

REFERENCE MANUAL for  
*Speech Signal Processing Toolkit Ver. 3.4.1*

April 28, 2011

The help message for every command can be obtained with the option “-h”. The help message brings explanation of the command, how to use, as well as its options.

Example: for the command `mcep` (% is the shell prompt)

```
> % mcep -h
>
> mcep - mel cepstral analysis
>
> usage:
> mcep [ options ] [ infile ] > stdout
> options:
> -a a : all-pass constant [0.35]
> -m m : order of mel cepstrum [25]
> -l l : frame length [256]
> -h : print this message
> (level 2)
> -i i : minimum iteration [2]
> -j j : maximum iteration [30]
> -d d : end condition [0.001]
> -e e : small value added to periodgram [0]
> infile:
> windowed sequences (float) [stdin]
> stdout:
> mel-cepstrum (float)
```

For more information related to this toolkit, please refer to <http://sourceforge.net/projects/sp-tk/>. In this site, the “Examples of Using Speech Signal Processing Toolkit” documentation file can be downloaded. If you have any bug reports, comments, or questions related this toolkit, please use the bug-tracker on SPTK website. We will try to answer every question, but we cannot guarantee it.

# Contents

acep	— adaptive cepstral analysis . . . . .	1
acorr	— obtain autocorrelation sequence . . . . .	3
agecep	— adaptive generalized cepstral analysis . . . . .	4
amcep	— adaptive mel-cepstral analysis . . . . .	6
average	— calculate mean for each block . . . . .	8
b2mc	— transform MLSA digital filter coefficients to mel-cepstrum . . . . .	9
bcp	— block copy . . . . .	10
bcut	— binary file cut . . . . .	12
bell	— ring a bell . . . . .	14
c2acr	— transform cepstrum to autocorrelation . . . . .	15
c2ir	— cepstrum to minimum phase impulse response . . . . .	16
c2sp	— transform cepstrum to spectrum . . . . .	17
cdist	— calculation of cepstral distance . . . . .	18
clip	— data clipping . . . . .	20
da	— play 16-bit linear PCM data . . . . .	21
dct	— DCT-II . . . . .	23
decimate	— decimation (data skipping) . . . . .	25
delay	— delay sequence . . . . .	26
delta	— delta calculation . . . . .	27
df2	— second order standard form digital filter . . . . .	29
dfs	— digital filter in standard form . . . . .	30
dmp	— binary file dump . . . . .	32
ds	— down-sampling . . . . .	34
echo2	— echo arguments to the standard error . . . . .	35
excite	— generate excitation . . . . .	36
extract	— extract vector . . . . .	37
fd	— file dump . . . . .	38
fdrw	— draw a graph . . . . .	39
fft	— FFT for complex sequence . . . . .	41
fft2	— 2-dimensional FFT for complex sequence . . . . .	42
fftcep	— FFT cepstral analysis . . . . .	45
fftr	— FFT for real sequence . . . . .	46
fftr2	— 2-dimensional FFT for real sequence . . . . .	47
fig	— plot a graph . . . . .	49
frame	— extract frame from data sequence . . . . .	56

freqt	— frequency transformation . . . . .	57
gc2gc	— generalized cepstral transformation . . . . .	58
gcep	— generalized cepstral analysis . . . . .	60
glogsp	— draw a log spectrum graph . . . . .	62
glsadf	— GLSA digital filter for speech synthesis . . . . .	64
gmm	— GMM parameter estimation . . . . .	66
gmmp	— calculation of GMM log-probability . . . . .	69
gnorm	— gain normalization . . . . .	71
grlogsp	— draw a running log spectrum graph . . . . .	72
grpdelay	— group delay of digital filter . . . . .	74
gwave	— draw a waveform . . . . .	75
histogram	— histogram . . . . .	77
idct	— Inverse DCT-II . . . . .	78
ifft	— inverse FFT for complex sequence . . . . .	80
ifft2	— 2-dimensional inverse FFT for complex sequence . . . . .	81
ignorm	— inverse gain normalization . . . . .	83
impulse	— generate impulse sequence . . . . .	84
imsvq	— decoder of multi stage vector quantization . . . . .	85
interpolate	— interpolation of data sequence . . . . .	86
ivq	— decoder of vector quantization . . . . .	87
lbg	— LBG algorithm for vector quantizer design . . . . .	88
levdur	— solve an autocorrelation normal equation using Levinson-Durbin method	91
linear_intpl	— linear interpolation of data . . . . .	93
lmadf	— LMA digital filter for speech synthesis . . . . .	95
lpc	— LPC analysis using Levinson-Durbin method . . . . .	98
lpc2c	— transform LPC to cepstrum . . . . .	99
lpc2lsp	— transform LPC to LSP . . . . .	101
lpc2par	— transform LPC to PARCOR . . . . .	104
lsp2lpc	— transform LSP to LPC . . . . .	106
lspcheck	— check stability and rearrange LSP . . . . .	107
lspdf	— LSP speech synthesis digital filter . . . . .	108
ltcdf	— all-pole lattice digital filter for speech synthesis . . . . .	109
mc2b	— transform mel-cepstrum to MLSA digital filter coefficients . . . . .	110
mcep	— mel cepstral analysis . . . . .	111
merge	— data merge . . . . .	113
mgc2mgc	— frequency and generalized cepstral transformation . . . . .	115
mgc2sp	— transform mel-generalized cepstrum to spectrum . . . . .	117
mgcep	— mel-generalized cepstral analysis . . . . .	119
mglsadf	— MGLSA digital filter for speech synthesis . . . . .	122
minmax	— find minimum and maximum values . . . . .	125
mlpg	— obtain parameter sequence from PDF sequence . . . . .	127
mlsadf	— MLSA digital filter for speech synthesis . . . . .	130
msvq	— multi stage vector quantization . . . . .	133
nan	— data check . . . . .	134
norm0	— normalize coefficients . . . . .	135

nrand	— generate normal distributed random value . . . . .	136
par2lpc	— transform PARCOR to LPC . . . . .	137
pca	— principal component analysis . . . . .	138
pcas	— calculate principal component scores . . . . .	139
phase	— transform real sequence to phase . . . . .	140
pitch	— pitch extraction . . . . .	142
poledf	— all pole digital filter for speech synthesis . . . . .	144
psgr	— XY-plotter simulator for EPSF . . . . .	145
ramp	— generate ramp sequence . . . . .	147
reverse	— reverse the order of data in each block . . . . .	148
rmse	— calculation of root mean squared error . . . . .	149
root_pol	— calculate roots of a polynomial equation . . . . .	150
sin	— generate sinusoidal sequence . . . . .	152
smcep	— mel-cepstral analysis using 2nd order all-pass filter . . . . .	153
snr	— evaluate SNR and segmental SNR . . . . .	155
sopr	— execute scalar operations . . . . .	157
spec	— transform real sequence to log spectrum . . . . .	160
step	— generate step sequence . . . . .	162
swab	— swap bytes . . . . .	163
train	— generate pulse sequence . . . . .	164
uels	— unbiased estimation of log spectrum . . . . .	165
ulaw	— $\mu$ -law compress/decompress . . . . .	167
us	— up-sampling . . . . .	168
us16	— up-sampling from 10 or 12 kHz to 16 kHz . . . . .	170
uscd	— up/down-sampling from 8, 10, 12, or 16 kHz to 11.025, 22.05, or 44.1 kHz	171
vopr	— execute vector operations . . . . .	172
vq	— vector quantization . . . . .	174
vstat	— vector statistics calculation . . . . .	175
vsum	— summation of vector . . . . .	177
wav2raw	— wav (RIFF) to raw . . . . .	179
window	— data windowing . . . . .	180
x2x	— data type transformation . . . . .	182
xgr	— XY-plotter simulator for X-window system . . . . .	184
zcross	— zero cross . . . . .	186
zerodf	— all zero digital filter for speech synthesis . . . . .	187
REFERENCESREFERENCES . . . . .		189
INDEX of TOPICS . . . . .		193



**NAME**

`acep` – adaptive cepstral analysis(4; 5)

**SYNOPSIS**

`acep` `[-m M]` `[-l L]` `[-t T]` `[-k K]` `[-p P]` `[-s]` `[-e E]` `[-P Pa]`  
`[ pefile ] < infile`

**DESCRIPTION**

`acep` uses adaptive cepstral analysis (4), (5), to calculate cepstral coefficients from unframed float data from standard input, sending the result to standard output. If `pefile` is given, `acep` writes the prediction error is written to that file.

The format for input and output data is float.

The algorithm to calculate recursively the adaptive cepstral coefficients is

$$\begin{aligned} \mathbf{c}^{(n+1)} &= \mathbf{c}^{(n)} - \mu^{(n)} \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_0^{(n)} &= -2e(n)\mathbf{e}^{(n)} \quad (\tau = 0) \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} &= -2(1 - \tau) \sum_{i=-\infty}^n \tau^{n-i} e(i)\mathbf{e}^{(i)} \quad (0 \leq \tau < 1) \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} &= \tau \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n-1)} - 2(1 - \tau)e(n)\mathbf{e}^{(n)} \\ \mu^{(n)} &= \frac{k}{M \boldsymbol{\varepsilon}^{(n)}} \\ \boldsymbol{\varepsilon}^{(n)} &= \lambda \boldsymbol{\varepsilon}^{(n-1)} + (1 - \lambda)e^2(n) \end{aligned}$$

where  $\mathbf{c} = [c(1), \dots, c(M)]^\top$ ,  $\mathbf{e}^{(n)} = [e(n-1), \dots, e(n-M)]^\top$ . Also, the gain is expressed by  $c(0)$  as follows:

$$c(0) = \frac{1}{2} \log \boldsymbol{\varepsilon}^{(n)}$$

In Figure 1, the system for adaptive cepstral analysis is shown.

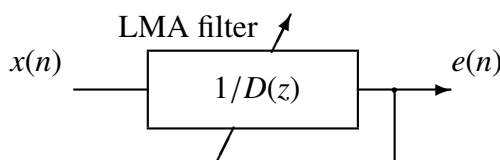


Figure 1: Adaptive cepstral analysis system

**OPTIONS**

<b>-m</b>	$M$	order of cepstrum	[25]
<b>-l</b>	$L$	leakage factor $\lambda$	[0.98]
<b>-t</b>	$T$	momentum constant $\tau$	[0.9]
<b>-k</b>	$K$	step size $k$	[0.1]
<b>-p</b>	$P$	output period of cepstrum	[1]
<b>-s</b>		output smoothed cepstrum	[FALSE]
<b>-e</b>	$E$	minimum value for $\varepsilon^{(n)}$	[0.0]
<b>-P</b>	$Pa$	number of coefficients of the LMA filter using the Padé approximation. $Pa$ should be 4 or 5.	[4]

**EXAMPLE**

In this example, the speech data is in the file *data.f* in float format, and the cepstral coefficients are written in the file *data.acep* for every block of 100 samples, and the prediction error can be found in *data.er*.

```
acep -m 15 -p 100 data.er < data.f > data.acep
```

**SEE ALSO**

uels, gcep, mcep, mgcep, amcep, agep, lmadf



**NAME**

`acorr` – obtain autocorrelation sequence

**SYNOPSIS**

`acorr` `[-m M]` `[-l L]` `[infile]`

**DESCRIPTION**

`acorr` calculates the  $m$ -th order autocorrelation function sequence for each frame of float data from `infile` (or standard input), sending the result to standard output. Namely, the input data is given by

$$x(0), x(1), \dots, x(L-1),$$

and the autocorrelation is evaluated as

$$r(k) = \sum_{m=0}^{L-1-k} x(m)x(m+k), \quad k = 0, 1, \dots, M,$$

and the output is the following autocorrelation function sequence,

$$r(0), r(1), \dots, r(M)$$

The format for input and output is float.

**OPTIONS**

`-m`  $M$  order of sequence [25]  
`-l`  $L$  frame length [256]

**EXAMPLE**

In the example below, the input file `data.f` is in float format. The frame length is 256, the frame period 100, every frame is passed through a Blackman window, and the autocorrelation function sequence is found in `data.acorr`.

```
frame +f -l 256 -p 100 < data.f | window | acorr -m 10 > data.acorr
```

**SEE ALSO**

`c2acr`, `levdur`

**NAME**

agcep – adaptive generalized cepstral analysis(9)

**SYNOPSIS**

**agcep** [ **-m** *M* ] [ **-c** *C* ] [ **-l** *L* ] [ **-t** *T* ] [ **-k** *K* ] [ **-p** *P* ]  
 [ **-s** ] [ **-n** ] [ **-e** *E* ] [ *pefile* ] < *infile*

**DESCRIPTION**

*agcep* uses adaptive generalized cepstral analysis (9) to calculate cepstral coefficients  $c_\gamma(m)$  from unframed float data from standard input, sending the result to standard output. If *pefile* is given, *agcep* writes the prediction error to that file.

The format for input and output data is float.

The algorithm to calculate recursively the adaptive generalized cepstral coefficients is

$$\begin{aligned} \mathbf{c}_\gamma^{(n+1)} &= \mathbf{c}_\gamma^{(n)} - \mu^{(n)} \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_0^{(n)} &= -2e_\gamma(n) \mathbf{e}_\gamma^{(n)} \quad (\tau = 0) \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} &= -2(1 - \tau) \sum_{i=-\infty}^n \tau^{n-i} e_\gamma(i) \mathbf{e}_\gamma^{(i)} \quad (0 \leq \tau < 1) \\ \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n)} &= \tau \hat{\mathbf{V}} \boldsymbol{\varepsilon}_\tau^{(n-1)} - 2(1 - \tau) e_\gamma(n) \mathbf{e}_\gamma^{(n)} \\ \mu^{(n)} &= \frac{k}{M \boldsymbol{\varepsilon}^{(n)}} \\ \boldsymbol{\varepsilon}^{(n)} &= \lambda \boldsymbol{\varepsilon}^{(n-1)} + (1 - \lambda) e_\gamma^2(n) \end{aligned}$$

where  $\mathbf{c}_\gamma = [c_\gamma(1), \dots, c_\gamma(M)]^\top$ ,  $\mathbf{e}_\gamma = [e_\gamma(n-1), \dots, e_\gamma(n-M)]^\top$ . The signal  $e_\gamma(n)$  is obtained passing the input signal  $x(n)$  through the filter  $(1 + \gamma F(z))^{-\frac{1}{\gamma}-1}$ , where

$$F(z) = \sum_{m=1}^M c_\gamma(m) z^{-m}.$$

In the case  $\gamma = -1/n$ , where  $n$  is a natural number, the adaptive generalized cepstral analysis system is shown in Figure 1. In the case  $n = 1$ , the adaptive generalized cepstral analysis is equivalent to the LMS linear predictor. In the case  $n \rightarrow \infty$ , the adaptive generalized cepstral analysis is equivalent to the adaptive cepstral analysis.

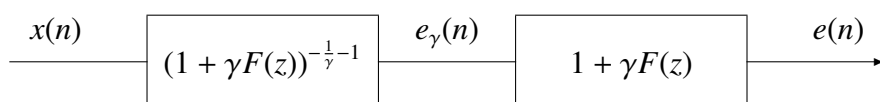
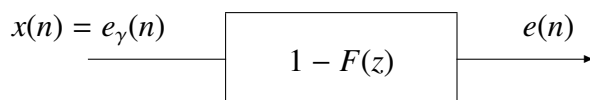
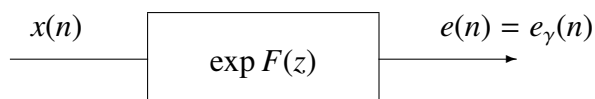
(a)  $-1 \leq \gamma \leq 0$ (b)  $\gamma = -1$ (c)  $\gamma = 0$ 

Figure 1: Adaptive generalized cepstral analysis system

**OPTIONS**

<b>-m</b>	$M$	order of generalized cepstrum	[25]
<b>-c</b>	$C$	power parameter $\gamma = -1/C$ for generalized cepstrum	[1]
<b>-l</b>	$L$	leakage factor $\lambda$	[0.98]
<b>-t</b>	$T$	momentum constant $\tau$	[0.9]
<b>-k</b>	$K$	step size $k$	[0.1]
<b>-p</b>	$P$	output period of generalized cepstrum	[1]
<b>-s</b>		output smoothed generalized cepstrum	[FALSE]
<b>-n</b>		output normalized generalized cepstrum	[FALSE]
<b>-e</b>	$E$	minimum value for $\varepsilon^{(n)}$	[0.0]

**EXAMPLE**

In this example, the speech data is in the file *data.f* in float format, the cepstral coefficients are written to the file *data.agcep*, and the prediction error can be found in *data.er*.

```
agcep -m 15 data.er < data.f > data.agcep
```

**SEE ALSO**

acep, amcep, glsadf

## NAME

amcep – adaptive mel-cepstral analysis(11; 12)

## SYNOPSIS

amcep [-m M] [-a A] [-l L] [-t T] [-k K] [-p P] [-s] [-e E]  
[-P Pa] [pfile] <infile

## DESCRIPTION

*amcep* uses adaptive mel-cepstral analysis to calculate mel-cepstral coefficients  $c_\alpha(m)$  from unframed float data from standard input, sending the result to standard output. If *pfile* is given, *amcep* writes the prediction error to that file.

The format for input and output data is float.

The algorithm to calculate recursively the adaptive mel-cepstral coefficients  $b(m)$  is

$$\begin{aligned} \mathbf{c}^{(n+1)} &= \mathbf{b}^{(n)} - \mu^{(n)} \hat{\nabla} \mathcal{E}_\tau^{(n)} \\ \hat{\nabla} \mathcal{E}_0^{(n)} &= -2e(n) \mathbf{e}_\Phi^{(n)} \quad (\tau = 0) \\ \hat{\nabla} \mathcal{E}_\tau^{(n)} &= -2(1 - \tau) \sum_{i=-\infty}^n \tau^{n-i} e(i) \mathbf{e}_\Phi^{(i)} \quad (0 \leq \tau < 1) \\ \hat{\nabla} \mathcal{E}_\tau^{(n)} &= \tau \hat{\nabla} \mathcal{E}_\tau^{(n-1)} - 2(1 - \tau) e(n) \mathbf{e}_\Phi^{(n)} \\ \mu^{(n)} &= \frac{k}{M \mathcal{E}^{(n)}} \\ \mathcal{E}^{(n)} &= \lambda \mathcal{E}^{(n-1)} + (1 - \lambda) e^2(n) \end{aligned}$$

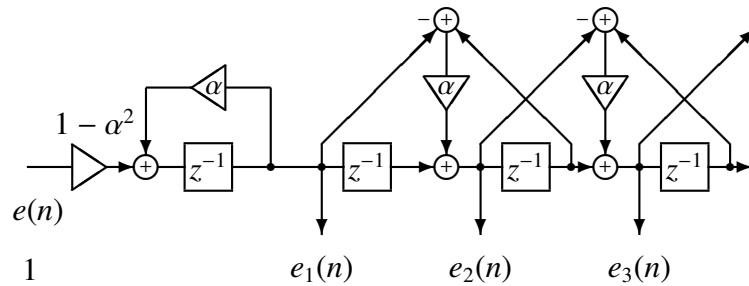


Figure 1: Filter  $\Phi_m(z)$

where  $\mathbf{b} = [b(1), b(2), \dots, b(M)]^\top$ ,  $\mathbf{e}_\Phi^{(n)} = [e_1(n), e_2(n), \dots, e_M(n)]^\top$ ,  $e_m(n)$  is the output of the inverse filter, which is obtained as shown in Figure 1, passing  $e(n)$  through the filter  $\Phi_m(z)$ .

The coefficients  $b(m)$  are equivalent to the coefficients of the MLSA filter. The mel-cepstral coefficients  $c_\alpha(m)$  can be obtained from  $b(m)$  through a linear transformation (refer to b2mc and mc2b).

In Figure 2, the adaptive mel-cepstral analysis system is shown.

The filter  $1/D(z)$  is realized by a MLSA filter.

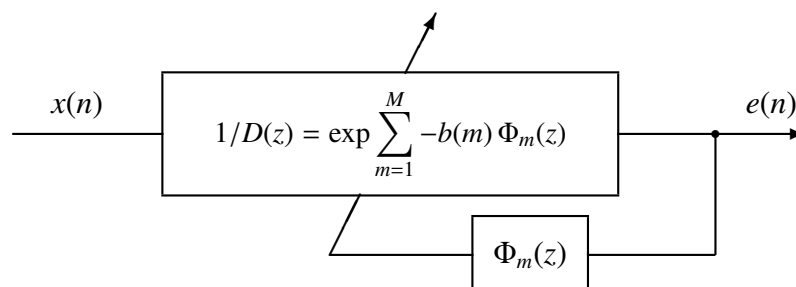


Figure 2: Adaptive mel-cepstral analysis system

## OPTIONS

<b>-m</b>	$M$	order of mel-cepstrum	[25]
<b>-a</b>	$A$	all-pass constant $\alpha$	[0.35]
<b>-l</b>	$L$	leakage factor $\lambda$	[0.98]
<b>-t</b>	$T$	momentum constant $\tau$	[0.9]
<b>-k</b>	$K$	step size $k$	[0.1]
<b>-p</b>	$P$	output period of mel-cepstrum	[1]
<b>-s</b>		output smoothed mel-cepstrum	[FALSE]
<b>-e</b>	$E$	minimum value for $\epsilon^{(n)}$	[0.0]
<b>-P</b>	$Pa$	number of coefficients of the MLSA filter using the Padé approximation. $Pa$ should be 4 or 5.	[4]

## EXAMPLE

In this example, the speech data is in the file *data.f* in float format, and the adaptive mel-cepstral coefficients are written to the file *data.amcep* for every block of 100 samples:

```
amcep -m 15 -p 100 < data.f > data.amcep
```

## SEE ALSO

acep, agcep, mc2b, b2mc, mlsadf

**NAME**

average – calculate mean for each block

**SYNOPSIS**

average [-l L] [-n N] [*infile*]

**DESCRIPTION**

*average* calculates the mean value for every  $L$ -length block from *infile* (or standard input), sending the result to standard output.

For the input data

$$x(0), x(1), \dots, x(L-1)$$

the output is calculated as follows:

$$\frac{x(0) + x(1) + \dots + x(L-1)}{L}$$

If  $L = 0$  then average of whole input data is calculated.

Input and output data are in float format.

**OPTIONS**

-l  $L$  number of items contained 1 frame [0]  
 -n  $N$  order of items contained 1 frame [L-1]

**EXAMPLE**

The output file *data.av* contains the mean taken from the whole data in *data.f* read in float format.

```
average < data.f > data.av
```

**SEE ALSO**

histogram, vsum, vstat

**NAME**

`b2mc` – transform MLSA digital filter coefficients to mel-cepstrum

**SYNOPSIS**

`b2mc` [ `-m M` ] [ `-a A` ] [ *infile* ]

**DESCRIPTION**

`b2mc` calculates mel-cepstral coefficients  $c_\alpha(m)$  from MLSA filter coefficients  $b(m)$  from *infile* (or standard input), sending the result to standard output.

Input and output data are in float format.

The transformation from  $b(m)$  coefficients to mel-cepstral coefficients  $c_\alpha(m)$  is as follows:

$$c_\alpha(m) = \begin{cases} b(M) & m = M \\ b(m) + \alpha b(m+1) & 0 \leq m < M \end{cases}$$

This transformation is the inverse transformation which is undertaken by the command `mc2b`.

**OPTIONS**

`-m` *M* order of mel cepstrum [25]  
`-a` *A* all-pass constant  $\alpha$  [0.35]

**EXAMPLE**

The example below converts the coefficients of an MLSA filter, which are in float format in file *data.b*, into mel-cepstral coefficients in file *data.mcep* with  $M = 15$  and  $\alpha = 0.35$ .

```
b2mc -m 15 < data.b > data.mcep
```

**SEE ALSO**

`mc2b`, `mcep`, `mlsadf`

**NAME**

bcp – block copy

**SYNOPSIS**

```
bcp [ -l l ] [ -L L ] [ -n n ] [ -N N ] [ -s s ] [ -S S ] [ -e e ] [ -f f ]
      [ +type ] [ infile ]
```

**DESCRIPTION**

*bcp* copies data blocks from *infile* (or standard input) to standard output, reformatting them according to command line options as shown below.

If the input format is ASCII, the basic input unit is a sequence of letters and the output block is partitioned with carriage returns.

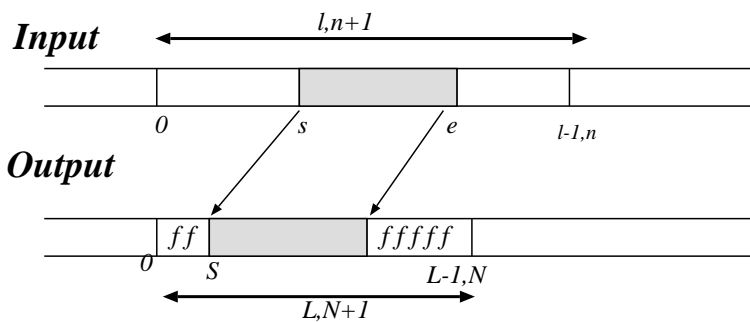


Figure 3: Example of the bcp command

**OPTIONS**

<b>-l</b>	<i>l</i>	number of items contained 1 block	[512]
<b>-L</b>	<i>L</i>	number of destination block size	[N/A]
<b>-n</b>	<i>n</i>	order of items contained 1 block	[1-1]
<b>-N</b>	<i>N</i>	order of destination block size	[N/A]
<b>-s</b>	<i>s</i>	start number	[0]
<b>-S</b>	<i>S</i>	start number in destination block	[0]
<b>-e</b>	<i>e</i>	end number	[EOF]
<b>-f</b>	<i>f</i>	fill into empty block	[0]



<i>+t</i>	data type		[f]
	c char (1 byte)	C	unsigned char (1 byte)
	s short (2 bytes)	S	unsigned short (2 bytes)
	i3 int (3 bytes)	I3	unsigned int (3 bytes)
	i int (4 bytes)	I	unsigned int (4 bytes)
	l long (4 bytes)	L	unsigned long (4 bytes)
	le long long (8 bytes)	LE	unsigned long long (8 bytes)
	f float (4 bytes)	d	double (8 bytes)
	a ASCII letter sequence		

**EXAMPLE**

Assume that data of the input file *data.f*  $a(0), a(1), a(2), \dots, a(20)$  is recursively written in float format.

If it is desired to copy the array  $a(1), a(2), \dots, a(10)$ , the following command can be used.

```
bcp +f -l 21 -s 1 -e 10 data.f > data.bcp
```

A different example with respect to the same input file *data.f* follows

```
bcp +f -l 21 -s 3 -e 5 -S 6 -L 10 data.f > data.bcp
```

In this example, the output block is

```
0, 0, 0, 0, 0, 0, a(3), a(4), a(5), 0
```

**SEE ALSO**

bcut, merge, reverse

**NAME**

`bcut` – binary file cut

**SYNOPSIS**

`bcut` [ `-s S` ] [ `-e E` ] [ `-l L` ] [ `-n N` ] [ `+type` ] [ *infile* ]

**DESCRIPTION**

*bcut* copies a selected portion of *infile* (or standard input) to standard output.

**OPTIONS**

<code>-s</code>	<i>S</i>	start number	[0]
<code>-e</code>	<i>E</i>	end number	[EOF]
<code>-l</code>	<i>L</i>	block length	[1]
<code>-n</code>	<i>N</i>	block order	[L-1]
<code>+t</code>		input data format	[f]
	<code>c</code>	char (1 byte)	<code>C</code> unsigned char (1 byte)
	<code>s</code>	short (2 bytes)	<code>S</code> unsigned short (2 bytes)
	<code>i3</code>	int (3 bytes)	<code>I3</code> unsigned int (3 bytes)
	<code>i</code>	int (4 bytes)	<code>I</code> unsigned int (4 bytes)
	<code>l</code>	long (4 bytes)	<code>L</code> unsigned long (4 bytes)
	<code>le</code>	long long (8 bytes)	<code>LE</code> unsigned long long (8 bytes)
	<code>f</code>	float (4 bytes)	<code>d</code> double (8 bytes)

**EXAMPLE**

In the example below, the input file *data.f* in float format is cut from the 3rd to the 5th float point:

```
bcut +f -s 3 -e 5 data.f > data.cut
```

For example, if the file *data.f* had the following data

1, 2, 3, 4, 5, 6, 7

the output file *data.cut* would be

4, 5, 6.

If the block length is assigned:

```
bcut +f -l 2 data.f -s 1 -e 2 > data.cut
```

then, the output file would contain the following data,

3,4,5,6

If the stationary part, say from the sample 100, of the output of a digital filter excited with pulse train is desired, then the following command can be used:

```
train -p 10 -l 256 | dfs -a 1 0.8 0.6 | bcut +f -s 100 > data.cut
```

In this case, the file *data.cut* will contain 156 points.

If we generate a *data.f* file passing a sinusoidal signal through a 256-length window as follows

```
sin -p 30 -l 2000 | window > data.f
```

and we want to take only the third window output, we could use the following command:

```
bcut +f -l 256 -s 3 -e 3 < data.f > data.cut
```

## SEE ALSO

bcp, merge, reverse



**NAME**

`c2acr` – transform cepstrum to autocorrelation

**SYNOPSIS**

`c2acr` [ **-m**  $M_1$  ] [ **-M**  $M_2$  ] [ **-l**  $L$  ] [ *infile* ]

**DESCRIPTION**

`c2acr` calculates  $M_2$ -th order autocorrelation coefficients from  $M_1$ -th order cepstral coefficients from *infile* (or standard input), writing the result to standard output. Give cepstral coefficients

$$c(0), c(1), \dots, c(M_1)$$

the corresponding autocorrelation coefficients are

$$r(0), r(1), \dots, r(M_2)$$

The format of input and output format is float.

The power spectrum is calculated from the logarithm spectrum, which is obtained from the Fourier transform of the  $M_1$ -th order cepstral analysis. The autocorrelation coefficients are obtained through the inverse Fourier transform of the power spectrum.

**OPTIONS**

<b>-m</b>	$M_1$	order of cepstrum	[25]
<b>-M</b>	$M_2$	order of autocorrelation	[25]
<b>-l</b>	$L$	FFT length	[256]

**EXAMPLE**

The output file *data.lpc* contains the 15-th order linear prediction coefficients calculated from the autocorrelation coefficients, which were obtained from the 30-th order cepstral coefficients file *data.cep*:

```
c2acr -m 30 -M 15 < data.cep | levdur -m 15 > data.lpc
```

**SEE ALSO**

uels, c2sp, c2ir, lpc2c

**NAME**

`c2ir` – cepstrum to minimum phase impulse response

**SYNOPSIS**

`c2ir` [-l L] [-m M<sub>1</sub>] [-M M<sub>2</sub>] [-i] [infile]

**DESCRIPTION**

`c2ir` calculates the minimum phase impulse response from minimum phase cepstral coefficients from *infile* (or standard input), sending the result to standard output. For example, if the input sequence is

$$c(0), c(1), c(2), \dots, c(M_1)$$

then the impulse response is calculated as

$$h(n) = \begin{cases} h(0) = \exp(c(0)) \\ h(n) = \sum_{k=1}^{M_1} \frac{k}{n} c(k) h(n-k) & n \geq 1 \end{cases}$$

and the output will contain

$$h(0), h(1), h(2), \dots, h(L-1)$$

The format of input and output format is float.

**OPTIONS**

**-m** M<sub>1</sub> order of cepstrum [25]  
**-M** M<sub>2</sub> length of impulse response [L-1]  
**-l** L order of impulse response [256]  
**-i** input minimum phase sequence [FALSE]

If the number of cepstral coefficients M<sub>1</sub> is not assigned and the order of the cepstral analysis is less than L, then the number of coefficients read is made equal to M<sub>1</sub>.

**EXAMPLE**

The output file *data.ir* contains the impulse response in the range  $n = 0 \sim 99$  obtained from the 30-th order cepstral coefficients file *data.cep*, in float format:

```
c2ir -l 100 -m 30 data.cep > data.ir
```

**SEE ALSO**

`c2sp`, `c2acr`

**NAME**

`c2sp` – transform cepstrum to spectrum

**SYNOPSIS**

`c2sp` [ `-m M` ] [ `-l L` ] [ `-p` ] [ `-o O` ] [ *infile* ]

**DESCRIPTION**

`c2sp` calculates the spectrum from the minimum phase cepstrum from *infile* (or standard input), sending the result to standard output. Input and output data are in float format.

**OPTIONS**

<code>-m</code>	<i>M</i>	order of cepstrum	[25]
<code>-l</code>	<i>L</i>	frame length	[256]
<code>-p</code>		output phase	[FALSE]
<code>-o</code>	<i>O</i>	output format	[0]

if the “`-p`” option is not assigned then

$O = 0$   $20 \times \log |H(z)|$

$O = 1$   $\ln |H(z)|$

$O = 2$   $|H(z)|$

if the “`-p`” option is assigned then

$O = 0$   $\arg |H(z)| \div \pi$  [ $\pi$  rad.]

$O = 1$   $\arg |H(z)|$  [rad.]

$O = 2$   $\arg |H(z)| \times 180 \div \pi$  [deg.]

**EXAMPLE**

The example below takes the 15-th order cepstrum from the file *data.cep* in float format, evaluates the running spectrum, and presents it in the screen:

```
c2sp -m 15 data.cep | grlogsp | xgr
```

**SEE ALSO**

`uels`, `mgc2sp`

**NAME**

`cdist` – calculation of cepstral distance

**SYNOPSIS**

`cdist` [ **-m** *M* ] [ **-o** *O* ] [ **-f** ] *cfile* [ *infile* ]

**DESCRIPTION**

*cdist* calculates the cepstral distance between the cepstral coefficients in *infile* (or standard input) and *cfile*, sending the result to standard output. For example, if the cepstral coefficients of the *infile* at frame *t* are

$$c_{1,t}(0), c_{1,t}(1), c_{1,t}(2), \dots, c_{1,t}(M)$$

and the cepstral coefficients of the *cfile* at frame *t* are

$$c_{2,t}(0), c_{2,t}(1), c_{2,t}(2), \dots, c_{2,t}(M)$$

then the squared cepstrum distance for every frame is

$$d(t) = \sum_{k=1}^M (c_{1,t}(k) - c_{2,t}(k))^2$$

and the total cepstral distance between both files is

$$d = \frac{1}{T} \sum_{t=0}^T d(t)$$

If the number of frames in *infile* or *cfile* is less than *T*, then evaluation is undertaken for the smallest number of frames.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of minimum-phase cepstrum	[25]
<b>-o</b>	<i>O</i>	output format	[0]
		<i>O</i> = 0 $\frac{10}{\ln 10} \sqrt{2d(t)}$ [db]	
		<i>O</i> = 1 $d(t)$	
		<i>O</i> = 2 $\sqrt{d(t)}$	
<b>-f</b>		output frame by frame	[FALSE]

**EXAMPLE**

In the example below, the squared spectral distance of the 15-th order cepstrum files *data1.cep* and *data2.cep*, both in float formats, is evaluated and displayed:

```
cdist -m 15 data1.cep data2.cep | dmp +f
```



**SEE ALSO**

acep, agcep, amcep, mcep

**NAME**

`clip` – data clipping

**SYNOPSIS**

**clip** [ **-y** *y<sub>min</sub>* *y<sub>max</sub>* ] [ **-ymin** *y<sub>min</sub>* ] [ **-ymax** *y<sub>max</sub>* ] [ *infile* ]

**DESCRIPTION**

*clip* clips the data from *infile* (or standard input) to minimum and maximum values specified on the command line, sending the result to standard output.

Input and output data are in float format.

**OPTIONS**

<b>-y</b>	<i>y<sub>min</sub></i> <i>y<sub>max</sub></i>	lower bound & upper bound	[−1.0 1.0]
<b>-ymin</b>	<i>y<sub>min</sub></i>	lower bound (ymax = inf)	[N/A]
<b>-ymax</b>	<i>y<sub>max</sub></i>	upper bound (ymin = -inf)	[N/A]

**EXAMPLE**

Suppose that the data of the file *data.f* is in float format with the following values,

1.0, 2.0, 3.0, 4.0, 5.0, 6.0

If we type the command

```
clip -y 2.5 5.5 < data.f > data.clip
```

the output *data.clip* will contain the values below.

2.5, 2.5, 3.0, 4.0, 5.0, 5.5

**NAME**

**da** – play 16-bit linear PCM data

**SYNOPSIS**

```
da [-s S] [-c C] [-g G] [-a A] [-o O] [-w] [-H H]
    [-v] [+type] [infile1] [infile2] ...
```

**DESCRIPTION**

*da* plays a series of input files (or standard input) on a system-dependent audio output device. If the system does not support the specified sampling frequency, *da* up-samples the data to a supported frequency. This command can be used under Linux (i386), FreeBSD (i386 newpcm driver), SunOS 4.1.x, SunOS 5.x (SPARC).

It is possible to change environment setting through following options

DA_GAIN	gain
DA_AMPGAIN	amplitude gain
DA_PORT	output port
DA_HDRSIZE	header size
DA_FLOAT	set the input data to float

**OPTIONS**

<b>-s</b>	<i>S</i>	sampling frequency, it can be used the following sampling frequencies 8, 10, 11.025, 12, 16, 20, 22.05, 32, 44.1, 48 (kHz).	[10]
<b>-g</b>	<i>G</i>	gain	[0]
<b>-a</b>	<i>A</i>	amplitude gain(0..100)	[N/A]
<b>-o</b>	<i>O</i>	output port(s : speaker, h : headphone)	[s]
<b>-w</b>		execute byte swap	[FALSE]
<b>-H</b>	<i>H</i>	header size in byte	[0]
<b>-v</b>		display filename	[FALSE]
<b>+type</b>		input data format	[f]

c	char (1 byte)	C	unsigned char (1 byte)
s	short (2 bytes)	S	unsigned short (2 bytes)
i3	int (3 bytes)	I3	unsigned int (3 bytes)
i	int (4 bytes)	I	unsigned int (4 bytes)
l	long (4 bytes)	L	unsigned long (4 bytes)
le	long long (8 bytes)	LE	unsigned long long (8 bytes)
f	float (4 bytes)	d	double (8 bytes)

**EXAMPLE**

In the following example, the speech data file *data.s* is played on the headphone. The sampling frequency is 8 kHz, and data is in short format.

```
da +s -s 8 -o h data.s
```

**BUGS**

In the Linux operating system, the output port can not be assigned.

## NAME

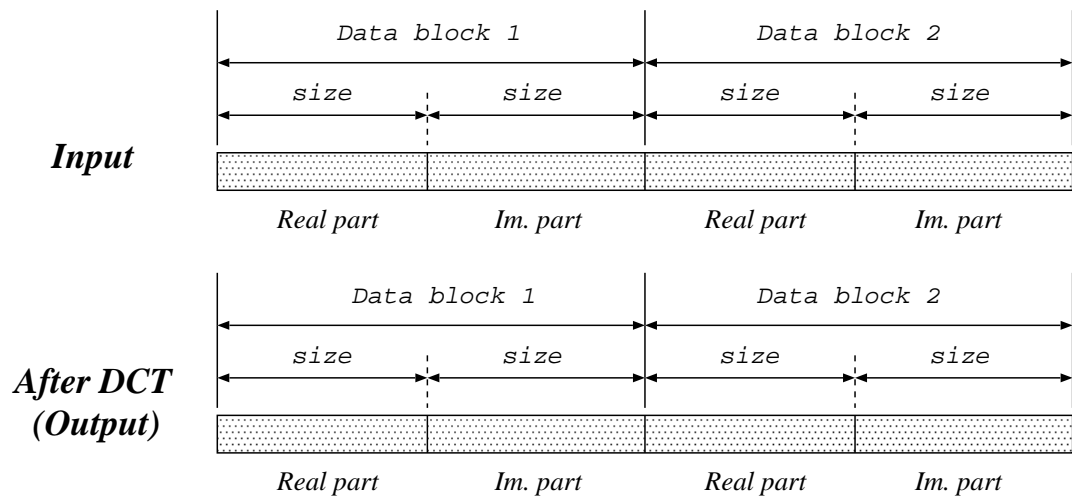
dct – DCT-II

## SYNOPSIS

dct [-l L] [-I] [-d] [infile]

## DESCRIPTION

*dct* calculates the Discrete Cosine Transformation II (DCT-II) of input data from *infile* (or standard input), sending the results to standard output. The input and output data is in float format, arranged as follows.



The Discrete Cosine Transformation II is

$$X_k = \sqrt{\frac{2}{L}} c_k \sum_{l=0}^{L-1} x_l \cos \left\{ \frac{\pi}{L} k \left( l + \frac{1}{2} \right) \right\}, \quad l = 0, 1, \dots, L$$

where

$$c_k = \begin{cases} 1 & (1 \leq k \leq L-1) \\ 1/\sqrt{2} & (k=0) \end{cases}$$

## OPTIONS

-l	L	DCT size	[256]
-I		use complex number	[FALSE]
-d		don't use FFT algorithm	[FALSE]

**EXAMPLE**

In this example, the DCT is evaluated from a complex-valued data file *data.f* in float format (real part: 256 points, imaginary part: 256 points), and the output is written to *data.dct*:

```
dct data.f -l 256 -I > data.dct
```

**SEE ALSO**

fft, idct

**NAME**

decimate – decimation (data skipping)

**SYNOPSIS**

**decimate** [ **-p** *P* ] [ **-s** *S* ] [ *infile* ]

**DESCRIPTION**

*decimate* picks up a sequence of input data from *infile* (or standard input) with interval *P* and start number *S*, sending the result to standard output.

If the input data is

$$x(0), x(1), x(2), \dots$$

then the output data is

$$x(S), x(S + P), x(S + 2P), x(S + 3P), \dots$$

Input and output data are in float format.

**OPTIONS**

<b>-p</b> <i>P</i>	decimation period	[10]
<b>-s</b> <i>S</i>	start sample	[0]

**EXAMPLE**

This example decimates input data from *data.f* file with interval 2, interpolates 0 with interval 2, and then outputs it to *data.di* file:

```
decimate -p 2 < data.f | interpolate -p 2 > data.di
```

**SEE ALSO**

interpolate

**NAME**

delay – delay sequence

**SYNOPSIS**

**delay** [ **-s** *S* ] [ **-f** ] [ *infile* ]

**DESCRIPTION**

*delay* delays the data from *infile* (or standard input) by inserting a specified number of zero samples at the beginning, sending the result to standard output. For example, if we want to delay the following data

$$x(0), x(1), \dots, x(T)$$

as

$$\underbrace{0, \dots, 0}_S, x(0), x(1), \dots, x(T).$$

We only need to set the “-s” option to *S*

$$\underbrace{0, \dots, 0}_S, x(0), x(1), \dots, x(T - S).$$

The format of input and output is float.

**OPTIONS**

<b>-s</b>	<i>S</i>	start sample	[0]
<b>-f</b>		keep file length	[FALSE]

**EXAMPLE**

If we have the following data in the input *data.f* file

1.0, 2.0, 3.0, 4.0, 5.0, 6.0

and we use the command below

```
delay -s 3 < data.f > data.delay
```

then the output file *data.delay* is

0.0, 0.0, 0.0, 1.0, 2.0, 3.0, 4.0, 5.0, 6.0

As another example, if we want to keep the same size of the input file, we can use the following command,

```
delay -s 3 -f < data.f > data.delay
```

and the output *data.delay* is

0.0, 0.0, 0.0, 1.0, 2.0, 3.0



**NAME**

delta – delta calculation

**SYNOPSIS**

**delta** [ **-m** *M* ] [ **-l** *L* ] [ **-t** *T* ] [ **-d** (*fn* | *d*<sub>0</sub> [*d*<sub>1</sub> ... ] ) ] [ **-r** *N<sub>R</sub>* *W*<sub>1</sub> [*W*<sub>2</sub>] ] [ *infile* ]

**DESCRIPTION**

*delta* calculates dynamic features from *infile* (or standard input), sending the result (static and dynamic features) to standard output. The input and output formats are

input ... ,  $x_t(0)$ , ... ,  $x_t(M)$ , ...

output ... ,  $x_t(0)$ , ... ,  $x_t(M)$ ,  $\Delta^{(1)}x_t(0)$ , ... ,  $\Delta^{(1)}x_t(M)$ , ... ,  $\Delta^{(n)}x_t(0)$ , ... ,  $\Delta^{(n)}x_t(M)$ , ...

Input and output data are in float format. The dynamic feature vector  $\Delta^{(n)}\mathbf{x}_t$  is obtained from the static feature vector as follows.

$$\Delta^{(n)}\mathbf{x}_t = \sum_{\tau=-L^{(n)}}^{L^{(n)}} w^{(n)}(\tau)\mathbf{x}_{t+\tau}$$

where  $n$  is the order of dynamic feature vector, for example, when we evaluate  $\Delta^2$  parameter,  $n=2$ .

**OPTIONS**

<b>-m</b>	<i>M</i>	order of vector	[25]
<b>-l</b>	<i>L</i>	length of vector	[ <i>M</i> + 1]
<b>-d</b>	( <i>fn</i>   <i>d</i> <sub>0</sub> [ <i>d</i> <sub>1</sub> ... ])	<i>fn</i> is the file name of the parameters $w^{(n)}(\tau)$ used when evaluating the dynamic feature vector. It is assume that the number of coefficients to the left and to the right have the same length, if this is not true than zeros are added to the short side. For example, if the coefficients are	[N/A]

$$w(-1), w(0), w(1), w(2), w(3)$$

then zeros are added to the left as follows.

$$0, 0, w(-1), w(0), w(1), w(2), w(3)$$

Instead of entering the filename *fn*, the coefficients(which compose the file *fn*) can be directly input in the command line. When the order of the dynamic feature vector is higher then one, then the sets of coefficients can be input one after the other as shown on the last example below. this option can not be used with the **-r** option.

**-r**  $N_R$   $W_1$  [ $W_2$ ] This option is used when  $N_R$ -th order dynamic parameters are used and the weighting coefficients  $w^{(n)}(\tau)$  are evaluated by regression.  $N_R$  can be made equal to 1 or 2. The variables  $W_1$  and  $W_2$  represent the widths of the first and second order regression coefficients, respectively. The first order regression coefficients for  $\Delta \mathbf{c}_t$  at frame  $t$  are evaluated as follows. [N/A]

$$\Delta \mathbf{c}_t = \frac{\sum_{\tau=-W_1}^{W_1} \tau \mathbf{c}_{t+\tau}}{\sum_{\tau=-W_1}^{W_1} \tau^2}$$

For the second order regression coefficients,  $a_2 = \sum_{\tau=-W_2}^{W_2} \tau^4$ ,  $a_1 = \sum_{\tau=-W_2}^{W_2} \tau^2$ ,  $a_0 = \sum_{\tau=-W_2}^{W_2} 1$  and

$$\Delta^2 \mathbf{c}_t = \frac{\sum_{\tau=-W_2}^{W_2} (a_0 \tau^2 - a_1) \mathbf{c}_{t+\tau}}{2(a_2 a_0 - a_1^2)}$$

this option can not be used with the **-d** option.

## EXAMPLE

In the example below, the first and second order dynamic features are calculated from 15-dimensional coefficient vectors from *data.f* using windows whose width are 1, and the resultant static and dynamic features are sent to *data.delta*:

```
delta -m 15 -r 2 1 1 data.static > data.delta
```

or

```
echo "-0.5 0 0.5" | x2x +af > delta
echo "0.25 -0.5 0.25" | x2x +af > accel
delta -m 15 -d delta -d accel data.pdf > data.par
```

## SEE ALSO

mlpg

**NAME**

df2 – second order standard form digital filter

**SYNOPSIS**

**df2** [ **-f**  $f_0$  ] [ **-p**  $f_1$   $b_1$  ] [ **-z**  $f_2$   $b_2$  ] [ *infile* ]

**DESCRIPTION**

*df2* filters data from *infile* (or standard output) with a second order standard form digital filter, sending the result to standard output. The central frequency and frequency band can be assigned through the options. The filter transfer function is

$$H(z) = \frac{1 - 2 \exp(-\pi b_2 / f_0) \cos(2\pi f_2 / f_0) z^{-1} + \exp(-2\pi b_2 / f_0) z^{-2}}{1 - 2 \exp(-\pi b_1 / f_0) \cos(2\pi f_1 / f_0) z^{-1} + \exp(-2\pi b_1 / f_0) z^{-2}}$$

If this command is used in cascade, an arbitrary filter can be designed using the options **-p** and **-z**. Input and output data are in float format.

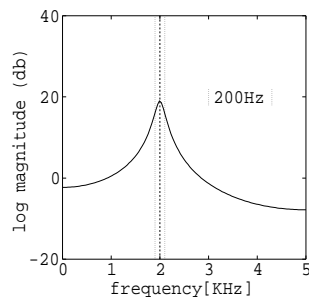
**OPTIONS**

<b>-f</b>	$f_0$	sampling frequency $f_0$ [Hz]	[10000]
<b>-p</b>	$f_1$ $b_1$	center frequency $f_1$ [Hz] and band width $b_1$ [Hz] of pole	[N/A]
<b>-z</b>	$f_2$ $b_2$	center frequency $f_2$ [Hz] and band width $b_2$ [Hz] of zero	[N/A]

**EXAMPLE**

The command below gives the impulse response of a filter with a pole at 2000 Hz and a frequency band of 200 Hz:

```
impulse | df2 -p 2000 200 | fdrw | xgr
```



**NAME**

dfs – digital filter in standard form

**SYNOPSIS**

**dfs** [ **-a** *K a(1) ... a(M)* ] [ **-b** *b(0) b(1) ... b(N)* ] [ **-p** *pfile* ] [ **-z** *zfile* ]  
[ *infile* ]

**DESCRIPTION**

*dfs* filters data from *infile* (or standard output) with a digital filter in standard form, sending the result to standard output. The filter transfer function is

$$H(z) = K \frac{\sum_{n=0}^N b(n)z^{-n}}{1 + \sum_{m=1}^M a(m)z^{-m}}$$

The format of input and output data is float.

**OPTIONS**

<b>-a</b> <i>K a(1) ... a(M)</i>	denominator coefficients, where <i>K</i> is the gain of the transfer function.	[N/A]
<b>-b</b> <i>b(0) b(1) ... b(N)</i>	numerator coefficients	[N/A]
<b>-p</b> <i>pfile</i>	denominator coefficients file in float format as follows <i>K, a(1), ..., a(M)</i>	[NULL]
<b>-z</b> <i>zfile</i>	numerator coefficients file in float format as follows <i>b(0), b(1), ..., b(N)</i>	[NULL]

If **-a**, **-p** options are not assigned, the denominator and *K* are made equal to 1. If **-b**, **-z** options are not assigned, the numerator is made equal to 1.

**EXAMPLE**

If we want to see the impulse response of the following transfer function

$$H(z) = \frac{1 + 2z^{-1} + z^{-2}}{1 + 0.9z^{-1}}$$

the command below can be used

```
impulse | dfs -a 1 0.9 -b 1 2 1 | dmp +f
```

If we want to see the frequency response plot of the digital filter whose coefficients are defined in float form by the files *data.p*, *data.z*, then we can use the following:

```
impulse | dfs -p data.p -z data.z | spec | fdw | xgr
```

The files *data.p* and *data.z* can be constructed using the command *x2x*.

**NAME**

`dmp` – binary file dump

**SYNOPSIS**

`dmp` [ `-n N` ] [ `-l L` ] [ `+type` ] [ `%form` ] [ `infile` ]

**DESCRIPTION**

`dmp` converts data from `infile` (or standard input) to human readable form, one sample per line with line numbers, sending the result to standard output.

**OPTIONS**

<code>-n</code>	<i>N</i>	block order (0,...,n)		[EOD]
<code>-l</code>	<i>L</i>	block length (1,...,l)		[EOD]
<code>+t</code>		input data format		[f]
	<code>c</code>	char (1 byte)	<code>C</code>	unsigned char (1 byte)
	<code>s</code>	short (2 bytes)	<code>S</code>	unsigned short (2 bytes)
	<code>i3</code>	int (3 bytes)	<code>I3</code>	unsigned int (3 bytes)
	<code>i</code>	int (4 bytes)	<code>I</code>	unsigned int (4 bytes)
	<code>l</code>	long (4 bytes)	<code>L</code>	unsigned long (4 bytes)
	<code>le</code>	long long (8 bytes)	<code>LE</code>	unsigned long long (8 bytes)
	<code>f</code>	float (4 bytes)	<code>d</code>	double (8 bytes)
<code>%form</code>		print format(printf style)		[N/A]

**EXAMPLE**

In this example, data is read from the input file `data.f` in float format, and the enumerated data is sent to the screen:

```
dmp +f data.f
```

For example, if the `data.f` file has the following values in float format

1, 2, 3, 4, 5, 6, 7

then the following output will be displayed on the screen:

```
0      1
1      2
2      3
3      4
4      5
5      6
6      7
```

In case we want to assign a block length:

```
dmp +f -n 2 data.f
```

Then the output would be

```
0      1
1      2
2      3
0      4
1      5
2      6
0      7
```

If we want to print on the screen the unit impulse response of a digital filter:

```
impulse | dfs -a 1 0.9 | dmp +f
```

If we want to print a sine wave then we can use the `%e` option of *printf* as follows:

```
sin -p 30 | dmp +f %e
```

If we want to represent the sine wave with three decimal points:

```
sin -p 30 | dmp +f %.3e
```

## SEE ALSO

x2x, fd

**NAME**

ds – down-sampling

**SYNOPSIS**

ds [ -s *S* ] [ *infile* ]

**DESCRIPTION**

*ds* down-samples data from *infile* (or standard input), sending the result to standard output.

The format of input and output data is float. The following filter coefficients can be used.

$S = 21$	<code>\$\$SPTK/lib/lpfcoef.2to1</code>
$S = 32$	<code>\$\$SPTK/lib/lpfcoef.3to2</code>
$S = 43$	<code>\$\$SPTK/lib/lpfcoef.4to3</code>
$S = 52, s = 54$	<code>\$\$SPTK/lib/lpfcoef.5to2up</code> <code>\$\$SPTK/lib/lpfcoef.5to2dn</code> ( <code>\$\$SPTK</code> is the directory where toolkit was installed.)

Filter coefficients are in ASCII format.

**OPTIONS**

-s *S* conversion type [21]

$S = 21$	down sampling by 2 : 1
$S = 32$	down sampling by 3 : 2
$S = 43$	down sampling by 4 : 3
$S = 52$	down sampling by 5 : 2
$S = 54$	down sampling by 5 : 4

**EXAMPLE**

In this example, the speech data in the input file *data.16*, which was sampled at 16 kHz in float format, is converted to an 8 kHz sampling rate:

```
ds data.16 > data.8
```



**NAME**

`echo2` – echo arguments to the standard error

**SYNOPSIS**

`echo2` [ `-n` ] [ `argument` ]

**DESCRIPTION**

`echo2` sends its command line arguments to standard error.

**OPTIONS**

`-n` no output newline

[FALSE]

**EXAMPLE**

This example prints "error!" in the standard error output:

```
echo2 -n "error!"
```

**NAME**

`excite` – generate excitation

**SYNOPSIS**

```
excite [ -p P ] [ -i I ] [ -n ] [ -s S ] [ infile ]
```

**DESCRIPTION**

`excite` generates an excitation sequence from pitch period information from *infile* (or standard input), sending the result to standard output. When the pitch period is nonzero (i.e. voiced), the excitation sequence consists of a pulse train at that pitch. When the pitch period is zero (i.e. unvoiced), the excitation sequence consists of Gaussian or M-sequence noise.

Input and output data are in float format.

**OPTIONS**

<code>-p</code>	<i>P</i>	frame period	[100]
<code>-i</code>	<i>I</i>	interpolation period	[1]
<code>-n</code>		gauss/M-sequence for unvoiced default is M-sequence	[FALSE]
<code>-s</code>	<i>S</i>	seed for nrand for Gaussian noise	[1]

**EXAMPLE**

In the example below, the excitation is generated from the *data.p* file and passed through a LPC synthesis filter whose coefficients are in the *data.lpc* file. The speech signal is output to the *data.syn* file.

```
excite < data.p | poledf data.lpc > data.syn
```

In case we use Gaussian noise to generate an unvoiced sound:

```
excite -n < data.p | poledf data.lpc > data.syn
```

**SEE ALSO**

`poledf`

**NAME**

extract – extract vector

**SYNOPSIS**

```
extract [-l L] [-i I] indexfile [ infile ]
```

**DESCRIPTION**

*extract* extracts selected vectors from *infile* (or standard input), sending the result to standard output. *indexfile* contains a previously-computed sequence of codebook indexes corresponding to the input vectors. Only those input vectors whose codebook index (from *indexfile*) matches the index given by the “-i” option are sent through to standard output.

**OPTIONS**

-l	<i>L</i>	order of vector	[10]
-i	<i>I</i>	codebook index	[0]

**EXAMPLE**

In the example below, a 10-th order vector file *data.v* in float format is quantized using a previously obtained codebook *data.idx* and writes to the output file *data.ex* the vectors which were quantized to the index 0 codeword.

```
extract -i 0 data.idx data.v > data.ex
```

**SEE ALSO**

ivq, vq

**NAME**

`fd` – file dump

**SYNOPSIS**

`fd` [ `-a A` ] [ `-n N` ] [ `-m M` ] [ `-ent` ] [ `+type` ] [ `%form` ] [ `infile` ]

**DESCRIPTION**

`fd` converts data from *infile* (or standard input) to human-readable multi-column format, sending the result to standard output.

**OPTIONS**

<code>-a</code>	<i>A</i>	address		[0]
<code>-n</code>	<i>N</i>	initial value for numbering		[0]
<code>-m</code>	<i>M</i>	modulo for numbering		[EOF]
<code>-ent</code>		number of data in each line		[0]
<code>+t</code>		data type		[c]
	<code>c</code>	char (1 byte)	<code>C</code>	unsigned char (1 byte)
	<code>s</code>	short (2 bytes)	<code>S</code>	unsigned short (2 bytes)
	<code>i3</code>	int (3 bytes)	<code>I3</code>	unsigned int (3 bytes)
	<code>i</code>	int (4 bytes)	<code>I</code>	unsigned int (4 bytes)
	<code>l</code>	long (4 bytes)	<code>L</code>	unsigned long (4 bytes)
	<code>le</code>	long long (8 bytes)	<code>LE</code>	unsigned long long (8 bytes)
	<code>f</code>	float (4 bytes)	<code>d</code>	double (8 bytes)
<code>%form</code>		print format(printf style)		[N/A]

**EXAMPLE**

This example displays the speech data in “sample.wav” with the corresponding addresses:

```
fd +c -a 0 sample.wav
```

Results:

```
000000 52 49 46 46 9a 15 00 00 57 41 56 45 66 6d 74 20 |RIFF....WAVEfmt |
000010 10 00 00 00 01 00 01 00 40 1f 00 00 40 1f 00 00 |.....@...@...|
000020 01 00 08 00 64 61 74 61 76 15 00 00 8a 8a 8f 99 |....datav.....|
```

⋮

**SEE ALSO**

`dmp`

**NAME**

`fdrw` – draw a graph

**SYNOPSIS**

```
fdrw [ -F F ] [ -R R ] [ -W W ] [ -H H ] [ -o xo yo ] [ -g G ] [ -m M ]
      [ -l L ] [ -p P ] [ -j J ] [ -n N ] [ -t T ] [ -y ymin ymax ] [ -z Z ] [ -b ]
      [ infile ]
```

**DESCRIPTION**

*fdrw* converts float data from *infile* (or standard input) to a plot formatted according to the FP5301 protocol, sending the result to standard output. Various options let you control the details of the plot layout.

**OPTIONS**

<b>-F</b>	<i>F</i>	factor	[1]
<b>-R</b>	<i>R</i>	rotation angle	[0]
<b>-W</b>	<i>W</i>	width of figure (×100 mm)	[1]
<b>-H</b>	<i>H</i>	height of figure (×100 mm)	[1]
<b>-o</b>	<i>xo yo</i>	origin in mm	[20 25]
<b>-g</b>	<i>G</i>	draw grid (0 ~ 2) (see also <code>fig</code> )	[1]
<b>-m</b>	<i>M</i>	line type (1 ~ 5) 1: solid 2: dotted 3: dot and dash 4: broken 5: dash	[0]
<b>-l</b>	<i>L</i>	line pitch	[0]
<b>-p</b>	<i>P</i>	pen number (1 ~ 10)	[1]
<b>-j</b>	<i>J</i>	join number (0 ~ 2)	[1]
<b>-n</b>	<i>N</i>	number of samples	[0]
<b>-t</b>	<i>T</i>	rotation of coordinate axis. When $T = -1$ , the reference point is on the top-left. When $T = 1$ the reference point is on the bottom-right.	[0]
<b>-y</b>	<i>ymin ymax</i>	scaling factor for y axis	[-1 1]
<b>-z</b>	<i>Z</i>	This option is used when data is written recursively in the y axis. The distance between two graphs in the y axis is given by <i>Z</i> .	[0]
<b>-b</b>		bar graph mode	[FALSE]

The *x* axis scaling is automatically done so that every point in the input file is plotted in equal interval for the assigned width. When **-n** option is omitted and the number of input samples is below 5000, then the block size is made equal to the number of samples. When the number of samples is above 5000, then the block size is made equal to 5000. When the **-y** option is omitted, the input data minimum value is made equal to *ymin* and the maximum value is made equal to *ymax*.

**EXAMPLE**

In the example below, the impulse response of a digital filter is drawn on the X window environment:

```
impulse | dfs -a 1 0.8 0.5 | fdrw -H 0.3 | xgr
```

The graph width is 10cm and its height is 3cm.

The next example draws on the X window environment the magnitude of the frequency response of a digital filter:

```
impulse | dfs -a 1 0.8 0.5 | spec | fdrw -y -60 40 | xgr
```

The y axis goes from -60 dB to 40 dB.

The running spectrum can be draw on the X window environment by:

```
fig -g 0 -W 0.4 << EOF
~~~~x 0 5
~~~~xscale 0 1 2 3 4 5
~~~~xname "FREQUENCY (kHz)"
EOF
spec < data |\
fdrw -W 0.4 -H 0.2 -g 0 -n 129 -y -30 30 -z 3 |\
xgr
```

The command *psgr* prints the output in a laser printer in the same way that it is printed on the screen. Since the *fdrw* command includes a sequence of commands for a plotter machine (FP5301 protocol) in the output file, its output can be directly sent to a printer.

**SEE ALSO**

fig, xgr, psgr

**NAME**

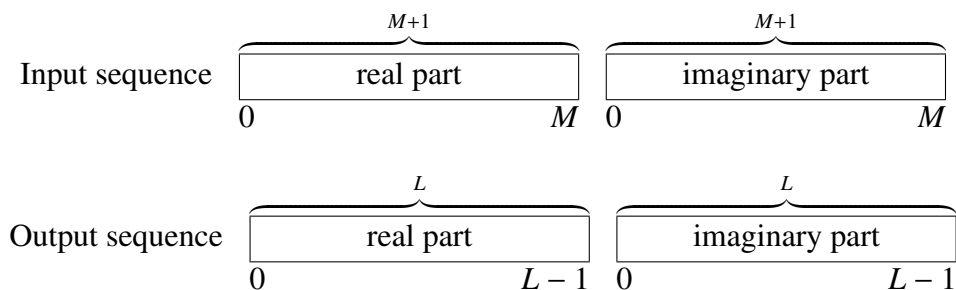
`fft` – FFT for complex sequence

**SYNOPSIS**

`fft [-l L] [-m M] [-{ A | R | I | P } ] [infile ]`

**DESCRIPTION**

`fft` uses the Fast Fourier Transform (FFT) algorithm to calculate the Discrete Fourier Transform (DFT) of complex-valued input data from `infile` (or standard input), sending the result to standard output. The input and output data is in float format, arranged as follows.

**OPTIONS**

<b>-l</b>	<i>L</i>	FFT size power of 2	[256]
<b>-m</b>	<i>M</i>	order of sequence	[L-1]
<b>-A</b>		amplitude	[FALSE]
<b>-R</b>		real part	[FALSE]
<b>-I</b>		imaginary part	[FALSE]
<b>-P</b>		output power spectrum	[FALSE]

**EXAMPLE**

This example reads a sequence of complex numbers in float format from `data.f` file (real part with 256 points and imaginary part with 256 points), evaluates its DFT and outputs it to `data.dft` file:

```
fft data.f -l 256 -A > data.dft
```

**SEE ALSO**

`fftr`, `spec`, `phase`

**NAME**

`fft2` – 2-dimensional FFT for complex sequence

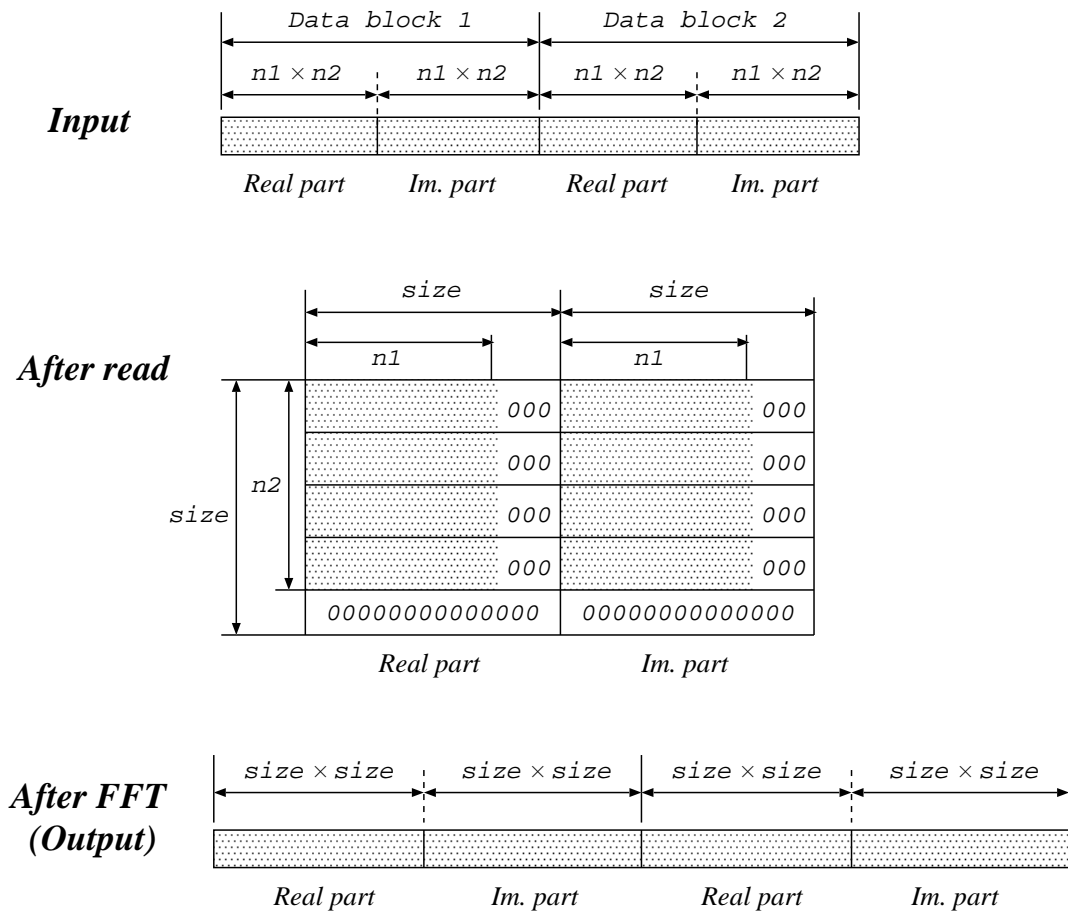
**SYNOPSIS**

`fft2 [-l L] [-m M1 M2] [-t] [-c] [-q] [-{A|R|I|P}]`

[ *infile* ]

**DESCRIPTION**

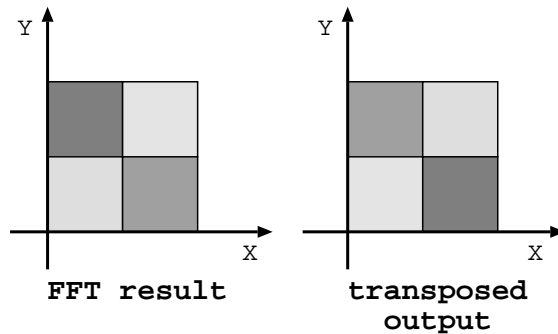
`fft2` uses the 2-dimensional Fast Fourier Transform (FFT) algorithm to calculate the 2-dimensional Discrete Fourier Transform (DFT) of complex-valued input data from *infile* (or standard input), sending the result to standard output. The input and output data is in float format, arranged as follows.



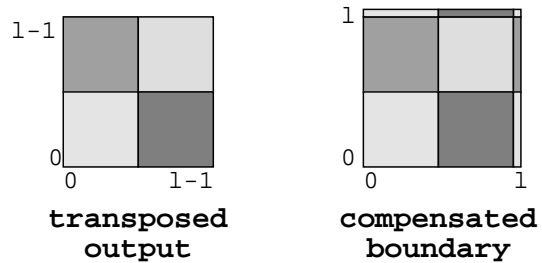


**OPTIONS**

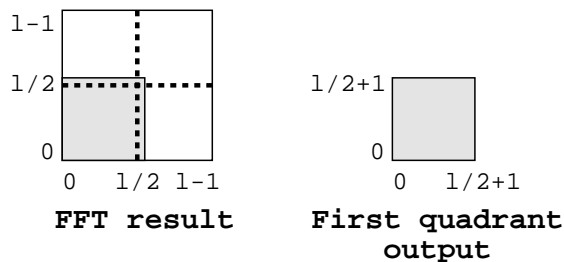
- l  $L$  FFT size power of 2 [64]
- m  $M_1 M_2$  order of sequence ( $M_1 \times M_2$ ). If file size  $k$  is smaller than  $64^2 \times 2$  and  $\sqrt{k \div 2}$  is integer value,  $M_1 = M_2 = \sqrt{k \div 2}$ . Otherwise output error message to standard error output and then terminate. [64,  $M_1$ ]
- t Output results in transposed form. [FALSE]



- c When results are transposed, 1 boundary data is copied from the opposite side, and then output  $(L + 1) \times (L + 1)$  data. [FALSE]



- q Output first 1/4 data of FFT results only. As in the above c option, boundary data is compensated and  $(\frac{L}{2} + 1) \times (\frac{L}{2} + 1)$  data are output. [FALSE]



- A amplitude [FALSE]

<b>-R</b>	real part	[FALSE]
<b>-I</b>	imaginary part	[FALSE]
<b>-P</b>	output power spectrum	[FALSE]

**EXAMPLE**

This example reads a sequence of 2-dimensional complex numbers in float format from *data.f* file, evaluates its 2-dimensional DFT and outputs it to *data.dft* file:

```
fft2 -A data.f > data.dft
```

**SEE ALSO**

fft, fft2, ifft

**NAME**

`fftcep` – FFT cepstral analysis

**SYNOPSIS**

`fftcep` [ **-m** *M* ] [ **-l** *L* ] [ **-j** *J* ] [ **-k** *K* ] [ **-e** *E* ] [ *infile* ]

**DESCRIPTION**

*fftcep* uses FFT cepstral analysis to calculate the cepstrum from windowed framed input data from *infile* (or standard input), sending the result to standard output. The windowed input time domain sequence of length *L* is

$$x(0), x(1), \dots, x(L-1)$$

Input and output data are in float format.

Assignment of the number of iterations *J* and the acceleration factor *K*, allows the use of the improved cepstral analysis method (1).

**OPTIONS**

<b>-m</b>	<i>M</i>	order of cepstrum	[25]
<b>-l</b>	<i>L</i>	frame length	[256]
<b>-j</b>	<i>J</i>	number of iteration	[0]
<b>-k</b>	<i>K</i>	acceleration factor	[0.0]
<b>-e</b>	<i>E</i>	epsilon	[0.0]

**EXAMPLE**

In the example below, speech data in float format is read from *data.f* and the cepstral coefficients are output to *data.cep*:

```
frame +f < data.f | window | fftcep > data.cep
```

**SEE ALSO**

uels



## NAME

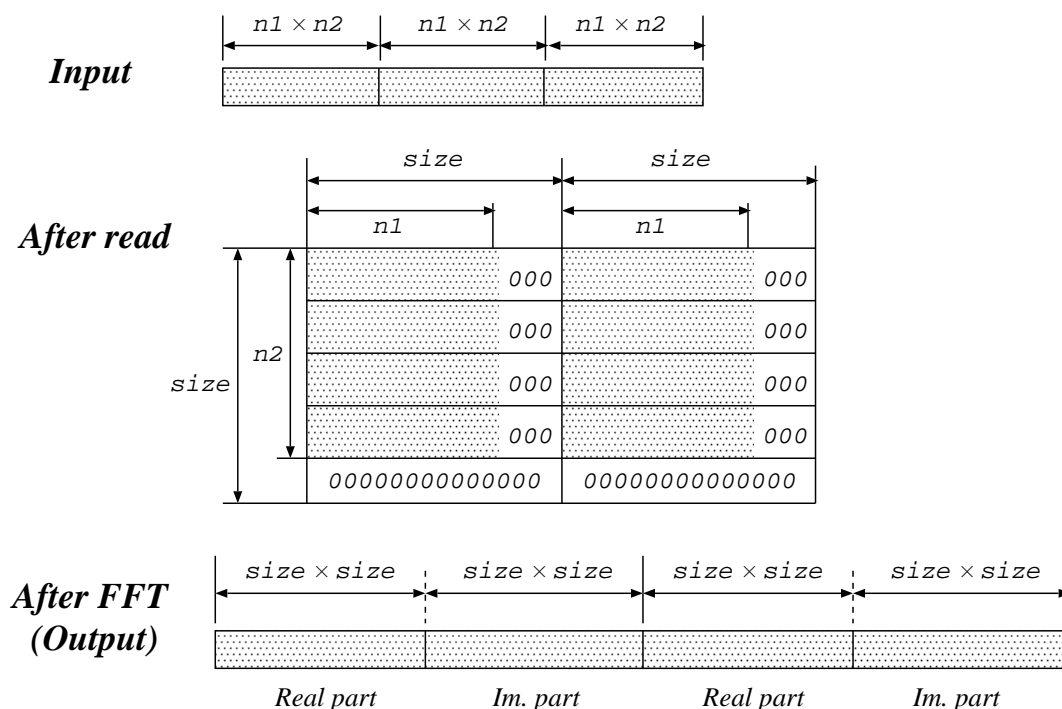
`fftr2` – 2-dimensional FFT for real sequence

## SYNOPSIS

`fftr2` **[-l L]** **[-m M<sub>1</sub> M<sub>2</sub>]** **[-t]** **[-c]** **[-q]** **[-{A|R|I|P}]** **[infile]**

## DESCRIPTION

`fftr2` uses the 2-dimensional Fast Fourier Transform (FFT) algorithm to calculate the 2-dimensional Discrete Fourier Transform (DFT) of real-valued input data from `infile` (or standard input), sending the result to standard output. The input and output data is in float format, arranged as follows.



## OPTIONS

- l**  $L$  FFT size power of 2 [64]
- m**  $M_1 M_2$  order of sequence ( $M_1 \times M_2$ ). If the file size  $k$  is smaller than  $64^2$  and  $\sqrt{k}$  is integer value,  $M_1 = M_2 = \sqrt{k}$ . Otherwise output error message to standard error output and then terminate. [64,  $M_1$ ]
- t** Output results in transposed form (see also `fft2`). [FALSE]
- c** When results are transposed, 1 boundary data is copied from the opposite side, and then output  $(L + 1) \times (L + 1)$  data (see also `fft2`). [FALSE]

<b>-q</b>	Output first 1/4 data of FFT results only. As in the above c option, boundary data is compensated and $(\frac{L}{2} + 1) \times (\frac{L}{2} + 1)$ data are output (see also fft2).	[FALSE]
<b>-A</b>	amplitude	[FALSE]
<b>-R</b>	real part	[FALSE]
<b>-I</b>	imaginary part	[FALSE]
<b>-P</b>	output power spectrum	[FALSE]

**EXAMPLE**

This example reads a sequence of 2-dimensional real numbers in float format from *data.f* file, evaluates its 2-dimensional DFT and outputs results to *data.dft* file:

```
fftr2 -A data.f > data.dft
```

**SEE ALSO**

fft, fft2, ifft

**NAME**

`fig` – plot a graph

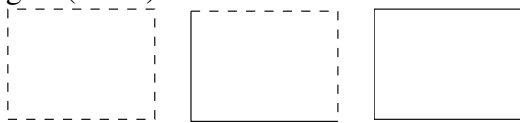
**SYNOPSIS**

```
fig [ -F F ] [ -R R ] [ -W W ] [ -H H ] [ -o xo yo ] [ -g G ] [ -p P ] [ -j J ]
[ -s S ] [ -f file ] [ -t ] [ infile ]
```

**DESCRIPTION**

*fig* draws a graph using information from *infile* (or standard input), sending the result in FP5301 plot format to standard output. This command is similar to the Unix command “graph” but includes some labeling functions. The output can be printed directly on a printer that supports the FP5301 protocol, displayed on an X11 display with the `xgr` command, or converted to PostScript with the `psgr` command.

**OPTIONS**

<b>-F</b>	<i>F</i>	factor	[1]
<b>-R</b>	<i>R</i>	rotation angle	[0]
<b>-W</b>	<i>W</i>	width of figure (×100mm)	[1]
<b>-H</b>	<i>H</i>	height of figure (×100mm)	[1]
<b>-o</b>	<i>xo yo</i>	origin in mm	[20 20]
<b>-g</b>	<i>G</i>	draw grid (0 ~ 2)	[2]
			
		<i>G</i> 0                      1                      2	
<b>-p</b>	<i>P</i>	pen number (1 ~ 10)	[1]
<b>-j</b>	<i>J</i>	join number (0 ~ 2)	[0]
<b>-s</b>	<i>S</i>	font size (1 ~ 4)	[1]
<b>-f</b>	<i>file</i>	The file assigned after this option is read before <i>infile</i> , that is, this option gives preference.	[NULL]
<b>-t</b>		transpose <i>x</i> and <i>y</i> axes	[FALSE]

**EXAMPLE**

Data in *data.fig* file is plotted in an X terminal in the example below:

```
fig data.fig |xgr
```

In this example, data in *data.fig* file is written in postscript, and seen with `ghostview`:

```
fig data.fig | psgr | ghostview -
```

**USAGE**

## COMMAND

The input data file can contain commands and data. Commands can be used for labeling, scaling, etc. Data is written in the  $(x\ y)$  coordinate pair form. Command values can be overwritten by entering new command values.

## COMMAND LINES

```
x [mel  $\alpha$ ] xmin xmax [xa]
y [mel  $\alpha$ ] ymin ymax [ya]
```

Assigns  $x$  and  $y$  scalings. Marks can be assigned in  $x$  and  $y$  axes through  $xa$  and  $ya$ . If no assigned of  $xa$  and  $ya$  is done, then  $xa = xmin$  and  $ya = ymin$ . If the optional “mel  $\alpha$ ”, where  $\alpha$  must be a number (for example, mel 0.35), is used, then labeling is undertaken as a frequency transformation of a minimum phase first order all-pass filter.

```
xscale  $x_1\ x_2\ x_3\ \dots$ 
yscale  $y_1\ y_2\ y_3\ \dots$ 
```

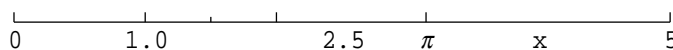
Assigns the points  $x_1, x_2, x_3, \dots$  and  $y_1, y_2, y_3, \dots$  in  $x$  and  $y$  axes. These points can be assigned with numbers or marks, Also, when a mixture of not-a-number + number (for example, '2,\*3.14) is needed, the following function should be used:

s	draws marks with half size.
\	only writes number.
@	does not write anything but assigns positions of marks.
none of the above	only marks are written.

Whenever the character is inside quotes, it appears in the position assigned by the string that precedes it. Please refer to the commands  $x/ynname$  for information on special characters.

(Example)

```
x 0 5
xscale 0 1.0 s1.5 '2 \2.5 '3.14 "\pi" @4 "x" 5
```

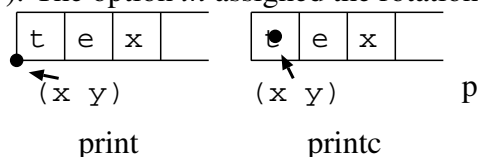


```
xname "text"
yname "text"
```

Labels  $x$  and  $y$  axes.  $text$  should be inside quote. Inside  $text$ ,  $\text{T}_{\text{E}}\text{X}$  commands can be used. Also, characters such as those that can be obtained with  $\text{T}_{\text{E}}\text{X}$  can be written with this command.

```
print x y "text" [th]
printc x y "text" [th]
```

This command writes  $text$  in the assigned position  $(x\ y)$ . The option  $th$  assigned the rotation degree.





title *x y "text" [th]*  
 titlec *x y "text" [th]*

This command is same as print(c). However, the basic unit is expressed in absolute value mm. The reference point is on the bottom-left.

csize *h [w]*

This command assigns in mm the character width and height, to be used in the following commands: *x/yscale, x/yname, print/c, title/c*

When the value of *w* is omitted, *w* is made equal to *h*. The default values for the option *-s* follows:

<i>-s</i>	<i>w</i>	<i>h</i>
1	2.5	2.2
2	5	2.6
3	2.5	4.4
4	5	4.4

pen *penno*

This command chooses the variable *penno*.  $1 \leq penno \leq 10$  Please refer to appendix.

join *joinno*

This command chooses the variable *joinno*.  $0 \leq joinno \leq 2$  Please refer to the appendix.

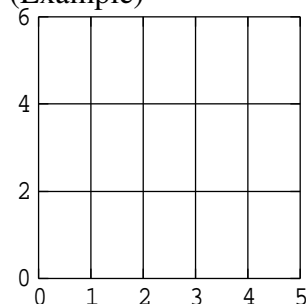
line *ltype [lpt]*

This command assigns the type *ltype* of the line which will connect data as well as the *lpt* pace. *lpt* is on mm unit. When *ltype=0*: no line is used to connect coordinate points. 1: solid 2: dotted 3: dot and dash 4: broken 5: dash Please refer to the appendix.

xgrid *x<sub>1</sub> x<sub>2</sub> ...*  
 ygrid *y<sub>1</sub> y<sub>2</sub> ...*

This command makes grids in the positions *x<sub>1</sub> x<sub>2</sub> ...*, *y<sub>1</sub> y<sub>2</sub> ...*.

(Example)



```
x 0 5
y 0 6
xscale 0 1 2 3 4 5
yscale 0 2 4 6
xgrid 1 2 3 4
ygrid 2 4
```

mark *label [th]*

This command draws a mark in the assigned coordinate position. The option *th* assigns angle(degree) that the string will be draw. If *label* is assigned  $\backslash 0$ , the mark is released. The way to write marks and special characters can be seen at the *label* section of data explanation.

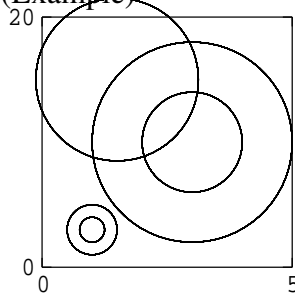
hight *h* [*w*]  
 italic *th*

This command defines the size of the label through its height *h*(mm) and width *w*(mm) whenever it is assigned and allows for writing a label in italic.

circle *x y r<sub>1</sub> r<sub>2</sub> ...*  
 xcircle *x y r<sub>1</sub> r<sub>2</sub> ...*  
 ycircle *x y r<sub>1</sub> r<sub>2</sub> ...*

This command writes a circle with radius *r<sub>1</sub> r<sub>2</sub> ...* with center on the coordinate (*x, y*). Also, the unit for the radius *r<sub>x</sub>* in the circle command is mm, for the xcircle command is the scale of the *x* axis and for the ycircle command is the scale of the *y* axis, as can be seen in the figure below.

(Example)



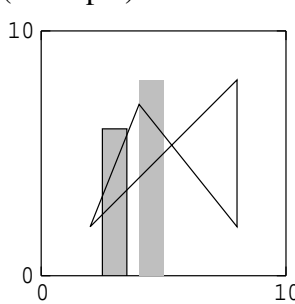
```
x 0 5
y 0 20
xscale 0 5
yscale 0 20

xcircle 3 10 1 2
ycircle 1 3 1 2
circle 1.5 15 13
```

box *x<sub>0</sub> y<sub>0</sub> x<sub>1</sub> y<sub>1</sub> [ x<sub>2</sub> y<sub>2</sub> ... ]*  
 paint *type*

This command draws a rectangle with paint *type* connecting (*x<sub>0</sub> y<sub>0</sub>*) and (*x<sub>1</sub> y<sub>1</sub>*) through a solid line. The line which connects (*x<sub>0</sub> y<sub>0</sub>*) and (*x<sub>1</sub> y<sub>1</sub>*) forms a diagonal of the rectangle. Also, if *x<sub>2</sub> y<sub>2</sub> ...* are assigned, a polygon is draw connecting (*x<sub>0</sub> y<sub>0</sub>*),(*x<sub>1</sub> y<sub>1</sub>*),(*x<sub>2</sub> y<sub>2</sub>*),... . In this case, Please do not use paint *type* different from 1. The default of paint is 1.

(Example)



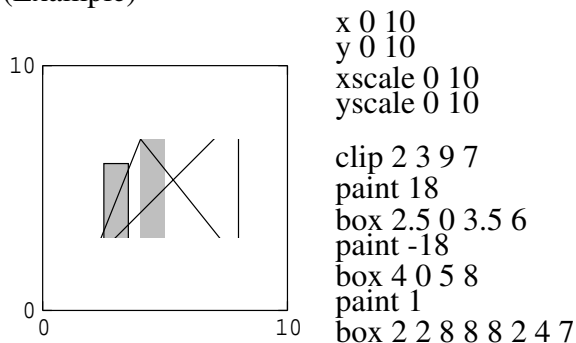
```
x 0 10
y 0 10
xscale 0 10
yscale 0 10

paint 18
box 2.5 0 3.5 6
paint -18
box 4 0 5 8
paint 1
box 2 2 8 8 2 4 7
```

```
clip  $x_0$   $y_0$   $x_1$   $y_1$ 
```

This command allows for drawing only inside the box defined by  $(x_0, y_0)$ ,  $(x_1, y_1)$ . When the coordinates  $(x_0, y_0)$ ,  $(x_1, y_1)$  are omitted, then the clip command is released.

(Example)



```
# any comment
```

This is used for comment lines. Whatever is written after this symbol is ignored by the fig command.

## DATA LINES

```
x y [label [th]]
```

The coordinates  $(x, y)$  are scaled by the values assigned in the command lines. If a string is written on *label*, then it will be written in the  $(x, y)$  position. No empty character (e.g., space character) should be left in *label*. When *label* is assigned in mark command, *label* replacement will take place only for this coordinate. The option *th* assigns the angle.

If  $\backslash n$ , where  $0 \leq n \leq 15$ , is assigned to *label*, the corresponding mark is drawn (refer to the appendix for the types of marks). When a minus sign is written before mark number, then the connecting line between marks passes through the center of each mark.

If a minus sign is not included, then connecting lines do not pass through the center of each mark. When  $n = 16$  ( $\backslash 16$ ), a small circle is written with diameter defined by the *hight* command. Also, special character and ASCII character can be written through code number when  $n > 32$ .

```
eod
EOD
```

This is the end of data sign. Coordinates before and after the eod sign are not connected.

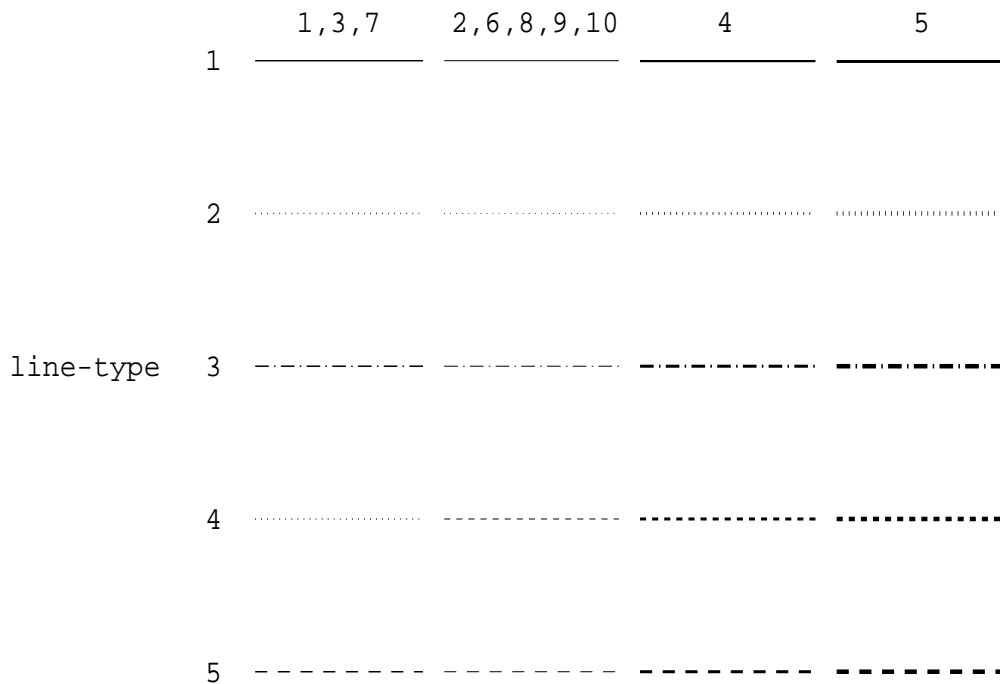
APPENDIX

- The following type of marks can be defined through *label*:

0	1	2	3	4	5	6	7
	•	×	□	△	○	◇	×
8	9	10	11	12	13	14	15
+	⊗	⊕	■	▲	●	◆	*

- The following types of pen and line can be defined:

[When output is obtained through the command `psgr`]  
`pen`






(Attention) The types of output generate through the pen command depends on the printer (Please try printing this page).









































[When output is obtained through the command xgr]  
 The following colors can be used.

pen type	1	2	3	4	5	6	7	8	9	10
color	black	blue	red	green	pink	orange	emerald	gray	brown	dark blue

- The following types of joins can be defined:

join type	0 Miter join	1 Round join	2 Bevel join
example			

- paint type:

0	1	2	3	4	5	6	7	8	9
									
10	11	12	13	14	15	16	17	18	19
									
-0	-1	-2	-3	-4	-5	-6	-7	-8	-9
									
-10	-11	-12	-13	-14	-15	-16	-17	-18	-19
									

(Attention) From 1 ~ 3 only a frame is draw, and for -9 and -19 the center is white and no frame is draw.

**NAME**

frame – extract frame from data sequence

**SYNOPSIS**

frame [ **-l** *L* ] [ **-n** ] [ **-p** *P* ] [ *infile* ]

**DESCRIPTION**

*frame* converts a sequence of input data from *infile* (or standard input) to a series of possibly-overlapping frames with period *P* and length *L*, sending the result to standard output. If the input data is  $x(0), x(1), \dots, x(T)$ , then the output data is

$$\begin{array}{ccccccc} 0 & , & 0 & , & \dots & , & x(L/2) \\ x(P - L/2) & , & x(P - L/2 + 1) & , & \dots & , & x(P + L/2) \\ x(2P - L/2) & , & x(2P - L/2 + 1) & , & \dots & , & x(2P + L/2) \\ & & & & & & \vdots \end{array}$$

**OPTIONS**

**-l** *L* frame length [256]  
**-p** *P* frame period [100]  
**-n** This option is used when instead of having  $x(0)$  is center point in the first frame we want to make  $x(0)$  as the first point of the first frame [FALSE]

**EXAMPLE**

In the example below, data is read from *data.f* file, frames of period 80 are applied, a Blackman window is passed, and a linear prediction analysis is undertaken. The output is written in *data.lpc* file:

```
frame -p 80 < data.f | window | lpc > data.lpc
```

**SEE ALSO**

bcp, x2x, bcut, window

**NAME**

freqt – frequency transformation

**SYNOPSIS**

**freqt** [ **-m**  $M_1$  ] [ **-M**  $M_2$  ] [ **-a**  $A_1$  ] [ **-A**  $A_2$  ] [ *infile* ]

**DESCRIPTION**

*freqt* converts an  $M_1$ -th order minimum phase sequence from *infile* (or standard input) into a frequency-transformed  $M_2$ -th order sequence, sending the result to standard output.

Given the input sequence

$$c_{\alpha_1}(0), c_{\alpha_1}(1), \dots, c_{\alpha_1}(M_1)$$

the frequency transform is

$$\alpha = (\alpha_1 - \alpha_2)/(1 - \alpha_1\alpha_2)$$

$$c_{\alpha_2}^{(i)}(m) = \begin{cases} c_{\alpha_1}(-i) + \alpha c_{\alpha_2}^{(i-1)}(0) & m = 0 \\ (1 - \alpha^2) c_{\alpha_2}^{(i-1)}(0) + \alpha c_{\alpha_2}^{(i-1)}(1) & m = 1 \\ c_{\alpha_2}^{(i-1)}(m-1) + \alpha (c_{\alpha_2}^{(i-1)}(m) - c_{\alpha_2}^{(i)}(m-1)) & m = 2, \dots, M_2 \end{cases} \quad (1)$$

$$i = -M_1, \dots, -1, 0$$

The  $M_2$ -th order frequency transformed output sequence is

$$c_{\alpha_2}^{(0)}(0), c_{\alpha_2}^{(0)}(1), \dots, c_{\alpha_2}^{(0)}(M_2)$$

Input and output data are in float format.

**OPTIONS**

<b>-m</b>	$M_1$	order of minimum phase sequence	[25]
<b>-M</b>	$M_2$	order of warped sequence	[25]
<b>-a</b>	$A_1$	all-pass constant of input sequence $\alpha_1$	[0]
<b>-A</b>	$A_2$	all-pass constant of output sequence $\alpha_2$	[0.35]

**EXAMPLE**

In the following example, the linear prediction coefficients in float format are read from *data.lpc* file, transformed in 30-th order LPC mel-cepstral coefficients, and written in *data.lpcmc* file:

```
lpc2c < data.lpc | freqt -m 30 > data.lpcmc
```

**SEE ALSO**

mgc2mgc

**NAME**

`gc2gc` – generalized cepstral transformation

**SYNOPSIS**

```
gc2gc [-m M1] [-g G1] [-c C1] [-n] [-u]
      [-M M2] [-G G2] [-C C2] [-N] [-U] [infile]
```

**DESCRIPTION**

`gc2gc` uses a regressive equation to transform a sequence of generalized cepstral coefficients with power parameter  $\gamma_1$  from *infile* (or standard input) into generalized cepstral coefficients with power parameter  $\gamma_2$ , sending the result to standard output.

Input and output data are in float format.

The regressive equation for the generalized cepstral coefficients follows.

$$c_{\gamma_2}(m) = c_{\gamma_1}(m) + \sum_{k=1}^{m-1} \frac{k}{m} (\gamma_2 c_{\gamma_1}(k) c_{\gamma_2}(m-k) - \gamma_1 c_{\gamma_2}(k) c_{\gamma_1}(m-k)), \quad m > 0.$$

For the above equation, in case  $\gamma_1 = -1, \gamma_2 = 0$ , then LPC cepstral coefficients are obtained from the LPC coefficients, in case  $\gamma_1 = 0, \gamma_2 = 1$ , minimum phase impulse response is obtained from cepstral coefficients.

If the coefficients  $c_{\gamma}(m)$  have not been normalized, then the input and output have following form.

$$1 + \gamma c_{\gamma}(0), \gamma c_{\gamma}(1), \dots, \gamma c_{\gamma}(M)$$

The following applies to the case the coefficients are normalized,

$$K_{\alpha}, \gamma c'_{\gamma}(1), \dots, \gamma c'_{\gamma}(M)$$

**OPTIONS**

<b>-m</b>	$M_1$	order of generalized cepstrum (input)	[25]
<b>-g</b>	$G_1$	gamma of generalized cepstrum (input)	[0]
		$\gamma_1 = G_1$	
<b>-c</b>	$C_1$	gamma of generalized cepstrum (input)	
		$\gamma_1 = -1/(\text{int})C_1$	
		$C_1$ must be $C_1 \geq 1$	
<b>-n</b>		regard input as normalized cepstrum	[FALSE]
<b>-u</b>		regard input as multiplied by $\gamma_1$	[FALSE]
<b>-M</b>	$M_2$	order of generalized cepstrum (output)	[25]
<b>-G</b>	$G_2$	gamma of generalized cepstrum (output)	[1]
		$\gamma_2 = G_2$	
<b>-C</b>	$C_2$	gamma of mel-generalized cepstrum (output)	
		$\gamma_2 = -1/(\text{int})G_2$	
		$C_2$ must be $C_2 \geq 1$	



<b>-N</b>	regard output as normalized cepstrum	[FALSE]
<b>-U</b>	regard output as multiplied by $\gamma_1$	[FALSE]

**EXAMPLE**

In the following example, generalized cepstral coefficients with  $M = 10$  and  $\gamma_1 = -0.5$  are read in float format from *data.gcep* file, transformed into 30-th order cepstral coefficients, and written to *data.cep*:

```
gc2gc -m 10 -c 2 -M 30 -G 0 < data.gcep > data.cep
```

**SEE ALSO**

gcep, mgcep, freqt, mgc2mgc, lpc2c

**NAME**

`gcep` – generalized cepstral analysis(6; 7; 8)

**SYNOPSIS**

`gcep` [ **-m** *M* ] [ **-g** *G* ] [ **-c** *C* ] [ **-l** *L* ] [ **-q** *Q* ] [ **-n** ] [ **-i** *I* ] [ **-j** *J* ] [ **-d** *D* ] [ **-e** *E* ]  
 [ **-f** *F* ] [ *infile* ]

**DESCRIPTION**

`gcep` uses generalized cepstral analysis to calculate normalized cepstral coefficients  $c'_\gamma(m)$  from  $L$ -length framed windowed input data from *infile* (or standard input), sending the result to standard output. The windowed input sequence of length  $L$  is

$$x(0), x(1), \dots, x(L-1)$$

Input and output data are in float format.

In the generalized cepstral analysis, the speech spectrum is estimated by the  $M$ -th order generalized cepstrum  $c_\gamma(m)$  or by normalized generalized cepstrum  $c'_\gamma(m)$  using the log spectrum through the unbiased estimation method.

$$\begin{aligned} H(z) &= s_\gamma^{-1} \left( \sum_{m=0}^M c_\gamma(m) z^{-m} \right) \\ &= K \cdot s_\gamma^{-1} \left( \sum_{m=1}^M c'_\gamma(m) z^{-m} \right) \\ &= \begin{cases} K \cdot \left( 1 + \gamma \sum_{m=1}^M c'_\gamma(m) z^{-m} \right)^{1/\gamma}, & -1 \leq \gamma < 0 \\ K \cdot \exp \sum_{m=1}^M c'_\gamma(m) z^{-m}, & \gamma = 0 \end{cases} \end{aligned}$$

To find the minimum value of cost function, if  $\gamma = -1$  then the linear prediction method is used, while if  $\gamma$  is different from  $-1$ , then Newton–Raphson method applied.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of generalized cepstrum	[25]
<b>-g</b>	<i>G</i>	gamma of generalized cepstrum $\gamma = G$	[0]
<b>-c</b>	<i>C</i>	gamma of generalized cepstrum $\gamma = -1/(\text{int})C$ <i>C</i> must be $C \geq 1$	
<b>-l</b>	<i>L</i>	frame length	[256]
<b>-n</b>		output normalized cepstrum	[FALSE]

- q**  $Q$  input data style [0]
- $Q = 0$  windowed data sequence
  - $Q = 1$   $20 \times \log |f(w)|$
  - $Q = 2$   $\ln |f(w)|$
  - $Q = 3$   $|f(w)|$
  - $Q = 4$   $|f(w)|^2$

Usually, the options below do not need to be assigned.

- i**  $I$  minimum iteration [2]
- j**  $J$  maximum iteration [30]
- d**  $D$  Newton-Raphson method end condition. The default value is  $D = 0.001$ . In this case the end point is achieved when the evaluation rate of  $\varepsilon^{(i)}$  is 0.001, that is when its value changes less than 0.1%. [0.001]
- e**  $E$  small value added to periodgram [0]
- f**  $F$  minimum value of the determinant of the normal matrix [0.000001]

### EXAMPLE

The following example read speech data in float format from *data.f* file, undertakes the 15-th order generalized cepstral analysis, and writes the results in *data.gcep*:

```
frame +f < data.f | window | gcep -m 15 > data.gcep
```

### SEE ALSO

uels, mcep, mgcep, glsadf

**NAME**

`glogsp` – draw a log spectrum graph

**SYNOPSIS**

`glogsp` [ **-F** *F* ] [ **-O** *O* ] [ **-x** *X* ] [ **-y** *ymin ymax* ] [ **-ys** *YS* ] [ **-p** *P* ] [ **-ln** *LN* ]  
 [ **-s** *S* ] [ **-l** *L* ] [ **-c** *comment* ] [ *infile* ]

**DESCRIPTION**

`glogsp` converts float-format log spectral data from *infile* (or standard input) to FP5301 plot format, sending the result to standard output. The output can viewed with `xgr`.

`glogsp` is implemented as a shell script that uses the `fig` and `fdrw` commands.

**OPTIONS**

<b>-F</b>	<i>F</i>	factor	[1]						
<b>-O</b>	<i>O</i>	origin of graph	[1]						
		1 ( 40,205) [mm]							
		2 (125,205) [mm]							
		3 ( 40,120) [mm]							
		4 (125,120) [mm]							
		5 ( 40, 35) [mm]							
		6 (125, 35) [mm]							
		<table style="border-collapse: collapse; text-align: center;"> <tr><td><b>1</b></td><td><b>2</b></td></tr> <tr><td><b>3</b></td><td><b>4</b></td></tr> <tr><td><b>5</b></td><td><b>6</b></td></tr> </table>	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	<b>6</b>	
<b>1</b>	<b>2</b>								
<b>3</b>	<b>4</b>								
<b>5</b>	<b>6</b>								
<b>-x</b>	<i>X</i>	<i>x</i> scale	[1]						
		1 normalized frequency (0 ~ 0.5)							
		2 normalized frequency (0 ~ $\pi$ )							
		4 frequency (0 ~ 4 kHz)							
		5 frequency (0 ~ 5 kHz)							
		8 frequency (0 ~ 8 kHz)							
		10 frequency (0 ~ 10 kHz)							
<b>-y</b>	<i>ymin ymax</i>	<i>y</i> scale[dB]	[0 100]						
<b>-ys</b>	<i>YS</i>	Y-axis scaling factor	[20]						
<b>-p</b>	<i>P</i>	pen number(1 ~ 10)	[1]						
<b>-ln</b>	<i>LN</i>	kind of line style(0 ~ 5) (see also <code>fig</code> )	[1]						
<b>-s</b>	<i>S</i>	start frame number	[0]						
<b>-l</b>	<i>L</i>	frame length	[256]						

**-c** comment comment for the graph [N/A]

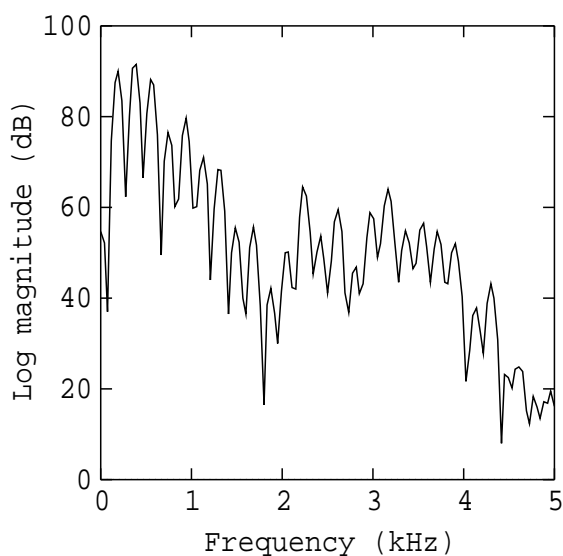
Usually, the options below do not need to be assigned.

**-W** *W* width of the graph ( mm) [0.6]  
**-H** *H* height of the graph ( mm) [0.6]  
**-v** over write mode [FALSE]  
**-o** *xo yo* origin of the graph. if -o option exists, -O is not effective [40 205]  
**-g** *G* type of frame of the graph (0 ~ 2) (see also fig) [2]  
**-f** *file* additional data file for fig [NULL]  
**-help** print help in detail

### EXAMPLE

In the example below, speech data sampled at 10 kHz is read in short format from *data.s* file, the magnitude of its log spectrum is evaluated and plotted on the screen:

```
x2x +sf data.s | bcut +f -s 4000 -e 4255 | window -n 2 | spec |\
glogsp -x 5 | xgr
```



### SEE ALSO

fig, fdrw, xgr, psgr, grlogsp, gwave

**NAME**

`glsadf` – GLSA digital filter for speech synthesis(18)

**SYNOPSIS**

`glsadf` [ **-m** *M* ] [ **-c** *C* ] [ **-p** *P* ] [ **-i** *I* ] [ **-v** ] [ **-t** ] [ **-n** ] [ **-k** ] [ **-P** *Pa* ] *gcfile*  
 [ *infile* ]

**DESCRIPTION**

*glsadf* derives a Generalized Log Spectral Approximation digital filter from normalized generalized cepstral coefficients in *gcfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output. The cepstral coefficients can be represented as  $K, c'_\gamma(1), \dots, c'_\gamma(M)$ .

Input and output data are in float format.

The transfer function  $H(z)$  are synthesis filter based on an  $M$  order normalized generalized cepstral coefficients  $c'_\gamma(m)$  is

$$H(z) = K \cdot D(z) = \begin{cases} K \cdot \left( 1 + \gamma \sum_{m=1}^M c'_\gamma(m) z^{-m} \right)^{1/\gamma}, & 0 < \gamma \leq -1 \\ K \cdot \exp \sum_{m=1}^M c'_\gamma(m) z^{-m}, & \gamma = 0 \end{cases}$$

In this case, we are considering only values for the power parameter  $\gamma = -1/C$ , where  $C$  is a natural number. The filter  $D(z)$  can be realized through a  $C$  level cascade as shown in figure1, where

$$\frac{1}{C(z)} = \frac{1}{1 + \gamma \sum_{m=1}^M c'_\gamma(m) z^{-m}}$$

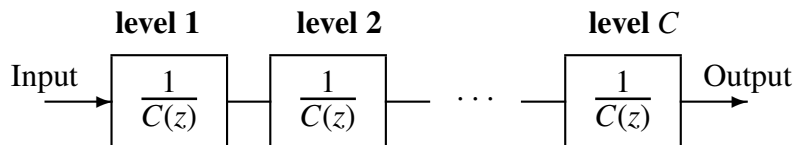


Figure 1: Structure of filter  $D(z)$

**OPTIONS**

<b>-m</b>	<i>M</i>	order of generalized cepstrum	[25]
<b>-c</b>	<i>C</i>	power parameter $\gamma = -1/C$ for generalized cepstrum if $C == 0$ then the LMA filter is used	[1]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-n</b>		regard input as normalized generalized cepstrum	[FALSE]
<b>-v</b>		inverse filter	[FALSE]
<b>-t</b>		transpose filter	[FALSE]
<b>-k</b>		filtering without gain	[FALSE]

The option below only works if  $C == 0$ .

<b>-P</b>	<i>Pa</i>	order of the Padé approximation <i>Pa</i> should be 4 or 5	[4]
-----------	-----------	---	-----

**EXAMPLE**

In this example, excitation is generated through the pitch data in the file *data.pitch* in float format, passed through a GLSA filter based on generalized cepstral coefficient file *data.gcep*, and the synthesized speech is output to *data.syn*:

```
excite < data.pitch | glsadf data.gcep > data.syn
```

**SEE ALSO**

ltdcf, lmadf, lspdf, mlsadf, mglsadf

**NAME**

`gmm` – GMM parameter estimation

**SYNOPSIS**

`gmm [-l L] [-m M] [-t T] [-s S] [-a A] [-b B] [-e E] [-v V] [-w W] [-f]`  
`[ infile ]`

**DESCRIPTION**

`gmm` uses expectation maximization (EM) algorithm to estimate Gaussian mixture model (GMM) parameters with diagonal covariance matrices from a sequence of vectors from `infile` (or standard input), sending the result to standard output.

The input sequence  $X$  consists of  $T$  float vectors  $\mathbf{x}$ , each of size  $L$ :

$$\begin{aligned} X &= [\mathbf{x}(0), \mathbf{x}(1), \dots, \mathbf{x}(T-1)], \\ \mathbf{x}(t) &= [x_t(0), x_t(1), \dots, x_t(L-1)]. \end{aligned}$$

The result is GMM parameters  $\lambda$  consisting of  $M$  mixture weights  $\mathbf{w}$  and  $M$  Gaussians with mean vector  $\boldsymbol{\mu}$  and variance vector  $\mathbf{v}$ , each of length  $L$ :

$$\begin{aligned} \lambda &= [\mathbf{w}, \boldsymbol{\mu}(0), \mathbf{v}(0), \boldsymbol{\mu}(1), \mathbf{v}(1), \dots, \boldsymbol{\mu}(M-1), \mathbf{v}(M-1)], \\ \mathbf{w} &= [w(0), w(1), \dots, w(M-1)], \\ \boldsymbol{\mu}(m) &= [\mu_m(0), \mu_m(1), \dots, \mu_m(L-1)], \\ \mathbf{v}(m) &= [\sigma_m^2(0), \sigma_m^2(1), \dots, \sigma_m^2(L-1)], \end{aligned}$$

where

$$\sum_{m=0}^{M-1} w(m) = 1.$$

GMM parameter set  $\lambda$  is initialized with an LBG algorithm, then the following EM steps are used iteratively to obtain new parameter set  $\hat{\lambda}$ :

$$\begin{aligned} \hat{w}(m) &= \frac{1}{T} \sum_{t=0}^{T-1} p(m | \mathbf{x}(t), \lambda), \\ \hat{\boldsymbol{\mu}}(m) &= \frac{\sum_{t=0}^{T-1} p(m | \mathbf{x}(t), \lambda) \mathbf{x}(t)}{\sum_{t=0}^{T-1} p(m | \mathbf{x}(t), \lambda)}, \\ \hat{\sigma}_m^2(l) &= \frac{\sum_{t=0}^{T-1} p(m | \mathbf{x}(t), \lambda) x_t^2(l)}{\sum_{t=0}^{T-1} p(m | \mathbf{x}(t), \lambda)} - \hat{\boldsymbol{\mu}}_m^2(l), \end{aligned}$$

where  $p(m | \mathbf{x}(t), \lambda)$  is a posterior probability of being in the  $m$ -th component at time  $t$ :

$$p(m | \mathbf{x}(t), \lambda) = \frac{w(m) \mathcal{N}(\mathbf{x}(t) | \boldsymbol{\mu}(m), \mathbf{v}(m))}{\sum_{k=0}^{M-1} w(k) \mathcal{N}(\mathbf{x}(t) | \boldsymbol{\mu}(k), \mathbf{v}(k))},$$



and

$$\begin{aligned} \mathcal{N}(\mathbf{x}(t) | \boldsymbol{\mu}(m), \mathbf{v}(m)) &= \frac{1}{(2\pi)^{L/2} |\boldsymbol{\Sigma}(m)|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{x}(t) - \boldsymbol{\mu}(m))' \boldsymbol{\Sigma}(m)^{-1} (\mathbf{x}(t) - \boldsymbol{\mu}(m)) \right\} \\ &= \frac{1}{(2\pi)^{L/2} \prod_{l=0}^{L-1} \sigma_m(l)} \exp \left\{ -\frac{1}{2} \sum_{l=0}^{L-1} \frac{(x_t(l) - \mu_m(l))^2}{\sigma_m^2(l)} \right\}, \end{aligned}$$

where  $\boldsymbol{\Sigma}(m)$  is a diagonal matrix with diagonal elements  $\mathbf{v}(m)$ :

$$\boldsymbol{\Sigma}(m) = \begin{bmatrix} \sigma_m^2(0) & & & 0 \\ & \sigma_m^2(1) & & \\ & & \ddots & \\ 0 & & & \sigma_m^2(L-1) \end{bmatrix}.$$

Average log-probability for training data  $X$

$$\log P(X) = \frac{1}{T} \sum_{t=0}^{T-1} \log \sum_{m=0}^{M-1} w(m) \mathcal{N}(\mathbf{x}(t) | \boldsymbol{\mu}(m), \mathbf{v}(m))$$

is increased by iterating the above steps. The average log-probability  $\log P(X)$  at each iterative step is printed on the standard error output. The EM steps are iterated at least  $A$  times and stopped at the  $B$ -th iteration or when there is a small absolute change in  $\log P(X)$  ( $\leq E$ ).

## OPTIONS

<b>-l</b>	$L$	length of vector	[26]
<b>-m</b>	$M$	number of Gaussian components	[16]
<b>-t</b>	$T$	number of training vectors	[N/A]
<b>-s</b>	$S$	seed of random variable for LBG algorithm	[1]
<b>-a</b>	$A$	minimum number of EM iterations	[0]
<b>-b</b>	$B$	maximum number of EM iterations ( $A \leq B$ )	[20]
<b>-e</b>	$E$	end condition for EM iteration	[0.00001]
<b>-v</b>	$V$	flooring value for variances	[0.001]
<b>-w</b>	$W$	flooring value for weights $(1/M)*W$	[0.001]
<b>-f</b>		full covariance	[FALSE]

## EXAMPLE

In the following example, a GMM with 8 Gaussian components is generated from training vectors *data.f* in float format, and GMM parameters are written to *gmm.f*.

```
gmm -m 8 data.f > gmm.f
```

If you want to model GMM with full covariance, add -f option.

```
gmm -m 8 -f data.f > gmm.f
```

**SEE ALSO**

gmmp, lbg

**NAME**

gmmp – calculation of GMM log-probability

**SYNOPSIS**

**gmmp** [ **-l** *L* ] [ **-m** *M* ] [ **-a** ] *gmmfile* [ *infile* ]

**DESCRIPTION**

*gmmp* calculates GMM log-probabilities of input vectors from *infile* (or standard input). *gmmfile* has the same file format as generated with *gmm* command, i.e., *gmmfile* consists of *M* mixture weights  $\mathbf{w}$  and *M* Gaussians with mean vector  $\boldsymbol{\mu}$  and diagonal variance vector  $\mathbf{v}$ , each of length *L*:

$$\begin{aligned}\lambda &= [\mathbf{w}, \boldsymbol{\mu}(0), \mathbf{v}(0), \boldsymbol{\mu}(1), \mathbf{v}(1), \dots, \boldsymbol{\mu}(M-1), \mathbf{v}(M-1)], \\ \mathbf{w} &= [w(0), w(1), \dots, w(M-1)], \\ \boldsymbol{\mu}(m) &= [\mu_m(0), \mu_m(1), \dots, \mu_m(L-1)], \\ \mathbf{v}(m) &= [\sigma_m^2(0), \sigma_m^2(1), \dots, \sigma_m^2(L-1)].\end{aligned}$$

The input sequence consists of *T* float vectors  $\mathbf{x}$ , each of size *L*:

$$\mathbf{x}(0), \mathbf{x}(1), \dots, \mathbf{x}(T-1).$$

The result is a sequence of log-probabilities of input vectors:

$$\log b(\mathbf{x}(0)), \log b(\mathbf{x}(1)), \dots, \log b(\mathbf{x}(T-1)),$$

or an average log-probability (if **-a** option is used):

$$\log P(\mathbf{X}) = \frac{1}{T} \sum_{t=0}^{T-1} \log b(\mathbf{x}(t)),$$

where

$$\begin{aligned}b(\mathbf{x}(t)) &= \sum_{m=0}^{M-1} w(m) \mathcal{N}(\mathbf{x}(t); \boldsymbol{\mu}(m), \mathbf{v}(m)), \\ \mathcal{N}(\mathbf{x}(t); \boldsymbol{\mu}(m), \mathbf{v}(m)) &= \frac{1}{(2\pi)^{L/2} \prod_{l=0}^{L-1} \sigma_m(l)} \exp \left\{ -\frac{1}{2} \sum_{l=0}^{L-1} \frac{(x_t(l) - \mu_m(l))^2}{\sigma_m^2(l)} \right\}.\end{aligned}$$

**OPTIONS**

<b>-l</b>	<i>L</i>	length of vector	[26]
<b>-m</b>	<i>M</i>	number of Gaussian components	[16]
<b>-a</b>		print average log-probability	[FALSE]

**EXAMPLE**

In the following example, frame log-probabilities of input data *data.f* for GMM with 8 Gaussians *gmm.f* are written to *probs.f*.

```
gmmp -m 8 gmm.f data.f > probs.f
```

**SEE ALSO**

`gmm`

**NAME**

gnorm – gain normalization

**SYNOPSIS**

**gnorm** [ **-m** *M* ] [ **-g** *G* ] [ **-c** *C* ] [ *infile* ]

**DESCRIPTION**

*gnorm* normalizes generalized cepstral coefficients  $c_\gamma(m)$  from *infile* (or standard input), sending the normalized generalized cepstral coefficients to standard output.

Input and output data are in float format.

The normalized generalized cepstral coefficients  $c'_\gamma(m)$  can be written as

$$c'_\gamma(m) = \frac{c_\gamma(m)}{1 + \gamma c_\gamma(0)}, \quad m > 0$$

Also, the gain  $K = c'_\gamma(0)$  is

$$K = \begin{cases} \left( \frac{1}{1 + \gamma c_\gamma(0)} \right)^{1/\gamma}, & 0 < |\gamma| \leq 1 \\ \exp c_\gamma(0), & \gamma = 0 \end{cases}$$

**OPTIONS**

- m** *M* order of generalized cepstrum [25]
- g** *G* power parameter  $\gamma$  of generalized cepstrum, [0]  
 $\gamma = G$
- c** *C* power parameter  $\gamma$  of generalized cepstrum,  
 $\gamma = -1/(\text{int})C$   
*C* must be  $C \geq 1$

**EXAMPLE**

In this example, generalized cepstral coefficients in float format are read from file *data.gcep* ( $M = 15, \gamma = -0.5$ ), normalized and output to *data.ngcep*:

```
gnorm -m 15 -c 2 < data.gcep > data.ngcep
```

**SEE ALSO**

ignorm, gcep, mgcep, gc2gc, mgc2mgc, freqt

**NAME**

`grlogsp` – draw a running log spectrum graph

**SYNOPSIS**

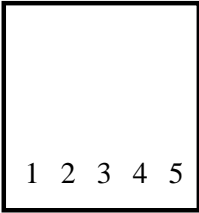
```
grlogsp [ -t ] [ -F F ] [ -O O ] [ -x X ] [ -y ymin ] [ -yy YY ] [ -yo YO ] [ -p P ]
[ -ln LN ] [ -s S ] [ -e E ] [ -n N ] [ -l L ]
[ -c comment1 ] [ -c2 comment2 ] [ -c3 comment3 ] [ infile ]
```

**DESCRIPTION**

`grlogsp` converts a sequence of float-format log spectra from *infile* (or standard input) to a running spectrum plot in FP5301 plot format, sending the result to standard output. The output can viewed with `xgr`.

`grlogsp` is implemented as a shell script that uses the `fig` and `fdw` commands.

**OPTIONS**

<b>-t</b>		transpose x and y axes	[FALSE]
<b>-F</b>	<i>F</i>	factor	[1]
<b>-O</b>	<i>O</i>	origin of graph	[1]
		1 ( 25, <i>YO</i> ) [mm]	
		2 ( 60, <i>YO</i> ) [mm]	
		3 ( 95, <i>YO</i> ) [mm]	
		4 (130, <i>YO</i> ) [mm]	
		5 (165, <i>YO</i> ) [mm]	
			
<b>-x</b>	<i>X</i>	( <i>YO</i> + 100, <i>X</i> ) [mm] if <code>-t</code> is specified. x scale	[1]
		1 normalized frequency (0 ~ 0.5)	
		2 normalized frequency (0 ~ $\pi$ )	
		4 frequency (0 ~ 4 kHz)	
		5 frequency (0 ~ 5 kHz)	
		8 frequency (0 ~ 8 kHz)	
		10 frequency (0 ~ 10 kHz)	
<b>-y</b>	<i>ymin</i>	y minimum	[-100]
<b>-yy</b>	<i>YY</i>	y scale [dB/10mm]	[100]
<b>-yo</b>	<i>YO</i>	y offset	[30]
<b>-p</b>	<i>p</i>	type of pen (1 ~ 10)	[2]

<b>-ln</b>	<i>LN</i>	style of line (0 ~ 5) (see also fig)	[1]
<b>-s</b>	<i>S</i>	start frame number	[0]
<b>-e</b>	<i>E</i>	end frame number	[EOF]
<b>-n</b>	<i>N</i>	number of frame	[EOF]
<b>-l</b>	<i>L</i>	frame length. Actually $\frac{L}{2}$ data are plotted.	[256]
<b>-c, c2, c3</b>	<i>comment1 ~ 3</i>	comment for the graph	[N/A]

Usually, the options below do not need to be assigned.

<b>-W</b>	<i>W</i>	width of the graph ( $\times 100$ mm)	[0.25]
<b>-H</b>	<i>H</i>	height of the graph ( $\times 100$ mm)	[1.5]
<b>-z</b>	<i>Z</i>	This option is used when data is written recursively in the y axis. the distance between two graphs in the y axis are given by Z. If Z is not given, Z is as same as F	
<b>-o</b>	<i>xo yo</i>	origin of the graph. if -o option exists, -O is not effective.	[95 30]
<b>-g</b>	<i>G</i>	type of frame of the graph (0 ~ 2) (see also fig)	[2]
<b>-cy</b>	<i>cy</i>	first comment position	[-8]
<b>-cy2</b>	<i>cy2</i>	second comment position	[-14]
<b>-cy3</b>	<i>cy3</i>	third comment position	[-20]
<b>-cs</b>	<i>cs</i>	font size of the comments	[1]
<b>-f</b>	<i>f</i>	additional data file for fig	[NULL]

## EXAMPLE

In this example, the magnitude of log spectrum is evaluated from data in *data.f* file in float format, and the graph with the running spectrum is sent in Postscript format to *data.ps* file:

```
frame +f < data.f | window |\
uels -m 15 | c2sp -m 15 |\
grlogsp | psgr > data.ps
```

## SEE ALSO

fig, fdrw, xgr, psgr, glogsp, gwave

**NAME**

`grpdelay` – group delay of digital filter

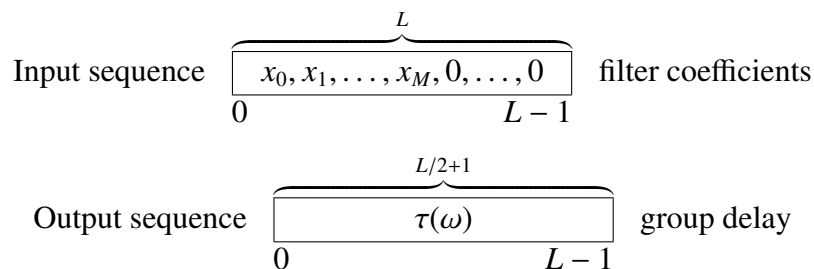
**SYNOPSIS**

`grpdelay` [ **-l** *L* ] [ **-m** *M* ] [ **-a** ] [ *infile* ]

**DESCRIPTION**

`grpdelay` computes the group delay of a sequence of filter coefficients from *infile* (or standard input), sending the result to standard output. Input and output data are in float format.

If the **-m** option is omitted and the length of an input data sequence is less than FFT size, the input file is padded with 0 and the FFT is evaluated as exemplified below. When the **-a** option is assigned, the gain is obtained from zero order input.

**OPTIONS**

<b>-l</b>	<i>L</i>	FFT size power of 2	[256]
<b>-m</b>	<i>M</i>	order of filter	[L-1]
<b>-a</b>		ARMA filter	[FALSE]

**EXAMPLE**

This example plots in the screen the group delay of impulse response of the filter with the following transfer function.

$$H(z) = \frac{1}{1 + 0.9z^{-1}}$$

```
impulse | dfs -a 1 0.9 | grpdelay | fdrw | xgr
```

**SEE ALSO**

delay, phase



**NAME**

`gwave` – draw a waveform

**SYNOPSIS**

```
gwave [ -F F ] [ -s S ] [ -e E ] [ -n N ] [ -i I ] [ -y ymax ] [ -y2 ymin ] [ -p P ]
      [ +ttype ] [ infile ]
```

**DESCRIPTION**

*gwave* converts speech waveform data from *infile* (or standard input) to FP5301 plot format, sending the result to standard output. The output can viewed with *xgr*.

*gwave* is implemented as a shell script that uses the *fig* and *fdrw* commands.

**OPTIONS**

<b>-F</b>	<i>F</i>	factor	[1]
<b>-s</b>	<i>S</i>	start point	[0]
<b>-e</b>	<i>E</i>	end point	[EOF]
<b>-n</b>	<i>N</i>	data number of one screen	[N/A]
		if this option is omitted, all of the data is plotted on one screen.	
<b>-i</b>	<i>I</i>	number of screen	[5]
<b>-y</b>	<i>y</i> <b>max</b>	maximum amplitude	[N/A]
		if this option is omitted, <i>y</i> <b>max</b> is maximum value of the input data.	
<b>-y2</b>	<i>y</i> <b>min</b>	minimum amplitude	[-YMAX]
<b>-p</b>	<i>P</i>	pen type(1 ~ 10)	[1]
<b>+t</b>		Input data format	[f]
	<i>c</i>	char (1 byte)	<i>C</i> unsigned char (1 byte)
	<i>s</i>	short (2 bytes)	<i>S</i> unsigned short (2 bytes)
	<i>i3</i>	int (3 bytes)	<i>I3</i> unsigned int (3 bytes)
	<i>i</i>	int (4 bytes)	<i>I</i> unsigned int (4 bytes)
	<i>l</i>	long (4 bytes)	<i>L</i> unsigned long (4 bytes)
	<i>le</i>	long long (8 bytes)	<i>LE</i> unsigned long long (8 bytes)
	<i>f</i>	float (4 bytes)	<i>d</i> double (8 bytes)
	<i>de</i>	long double (12 bytes)	

**EXAMPLE**

This example reads speech waveform file in float format from *data.f* and writes the output in Postscript format to *data.ps*.

```
gwave +f < data.f | psgr > data.ps
```

**SEE ALSO**

fig, fdrw, xgr, psgr, glogsp, grlogsp

**NAME**

histogram – histogram

**SYNOPSIS**

**histogram** [ **-l** *L* ] [ **-i** *I* ] [ **-j** *J* ] [ **-s** *S* ] [ **-n** ] [ *infile* ]

**DESCRIPTION**

*histogram* calculates histograms of frames of input data from *infile* (or standard input), sending the results to standard output.

Input and output data are in float format. The output can be graphed with *fdrw*.

If an input value is outside the specified interval, the exit status of *histogram* will be nonzero, but the output histogram will be generated nonetheless.

**OPTIONS**

<b>-l</b>	<i>L</i>	frame size	[0]
		<i>L</i> > 0	evaluate the histogram for every frame
		<i>L</i> = 0	evaluate the histogram for the whole file
<b>-i</b>	<i>I</i>	infimum	[0.0]
<b>-j</b>	<i>J</i>	supremum	[1.0]
<b>-s</b>	<i>S</i>	step size	[0.1]
<b>-n</b>		normalization	[FALSE]

**EXAMPLE**

The example below plots the histogram of the speech waveform file *data.f* in float format.

```
histogram -i -16000 -j 16000 -s 100 data.f | fdrw | xgr
```

**SEE ALSO**

average

**NAME**

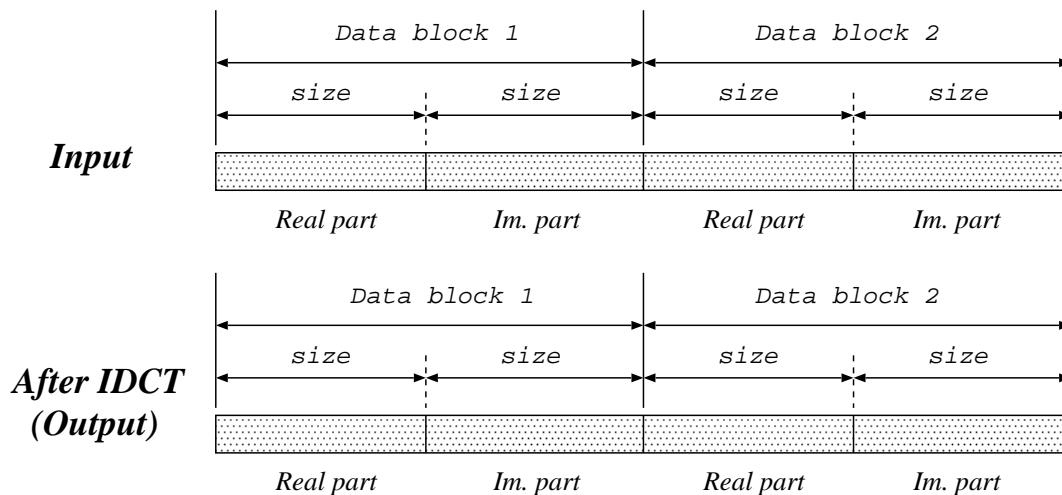
idct – Inverse DCT-II

**SYNOPSIS**

idct [-l L] [-c] [-d] [infile]

**DESCRIPTION**

*idct* calculates the Inverse Discrete Cosine Transformation II (IDCT-II) of input data from *infile* (or standard input), sending the results to standard output. The input and output data is in float format, arranged as follows.



The Inverse Discrete Cosine Transformation II is

$$x_l = \sqrt{\frac{2}{L}} c_l \sum_{k=0}^{L-1} X_k \cos \left\{ \frac{\pi}{L} \left( k + \frac{1}{2} \right) l \right\}, \quad l = 0, 1, \dots, L$$

where

$$c_l = \begin{cases} 1 & (1 \leq l \leq L-1) \\ 1/\sqrt{2} & (l = 0) \end{cases}$$

**OPTIONS**

-l	L	IDCT size	[256]
-c		use complex number	[FALSE]
-d		don't use FFT algorithm	[FALSE]

**EXAMPLE**

In this example, the IDCT is evaluated from a complex-valued data file *data.f* in float format (real part: 256 points, imaginary part: 256 points), and the output is written to *data.idct*:

```
idct data.f -l 256 -c > data.idct
```

**SEE ALSO**

fft, dct

**NAME**

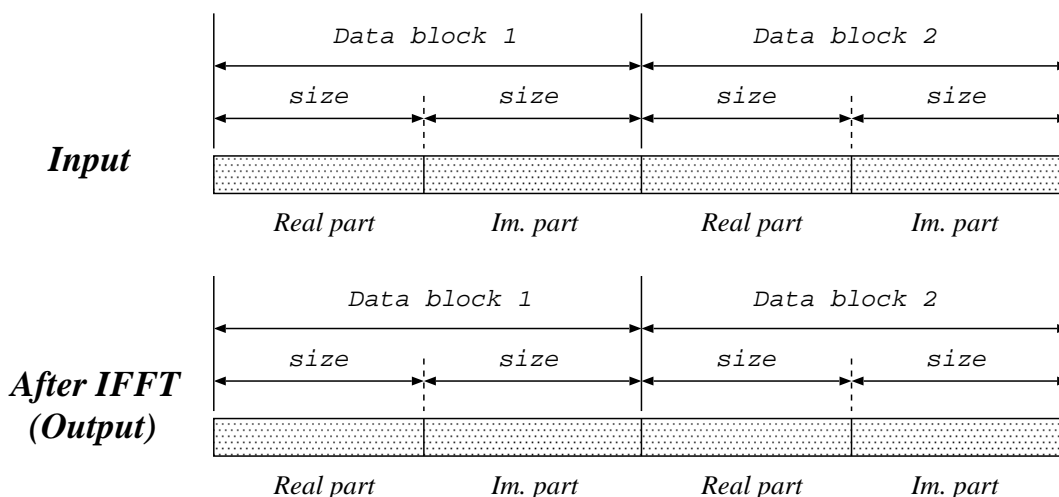
`ifft` – inverse FFT for complex sequence

**SYNOPSIS**

```
ifft [-l L] [-{R|I}] [infile ]
```

**DESCRIPTION**

`ifft` calculates the Inverse Discrete Fourier Transform (IDFT) of complex-valued data from *infile* (or standard input), sending the results to standard output. The input and output data is in float format, arranged as follows.

**OPTIONS**

```
-l L FFT size power of 2 [256]
-R output only real part [FALSE]
-I output only imaginary part [FALSE]
```

**EXAMPLE**

In this example, the inverse DFT is evaluated from a data file *data.f* in float format (real part: 256 points, imaginary part: 256 points), and the output is written to *data.iffi*:

```
ifft data.f -l 256 > data.iffi
```

**SEE ALSO**

`fft`, `fft2`, `fftr`, `fftr2`, `ifft2`

## NAME

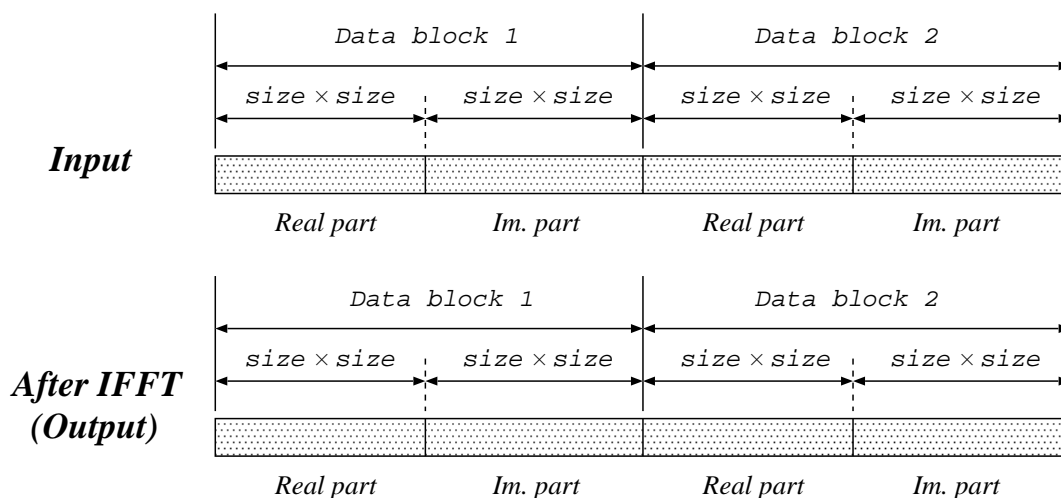
`ifft2` – 2-dimensional inverse FFT for complex sequence

## SYNOPSIS

`ifft2` [ `-l L` ] [ `+r` ] [ `-t` ] [ `-c` ] [ `-q` ] [ `-{ R | I }` ] [ *infile* ]

## DESCRIPTION

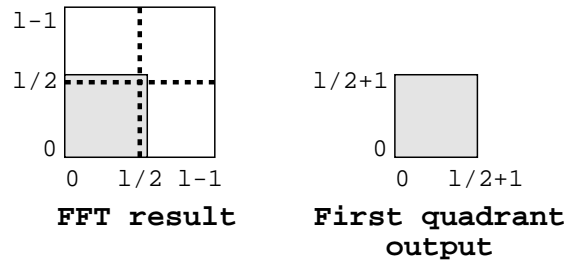
*ifft* calculates the 2-dimensional Inverse Discrete Fourier Transform (IDFT) of complex-valued data from *infile* (or standard input), sending the results to standard output. The input and output data is in float format, arranged as follows.



## OPTIONS

- `-l` *L* FFT size power of 2 [64]
- `+r` regard input as real values rather than complex values [FALSE]
- `-t` Output results in transposed form (see also `fft2`). [FALSE]
- `-c` When results are transposed, 1 boundary data is copied from the opposite side, and then output  $(L + 1) \times (L + 1)$  data (see also `fft2`). [FALSE]

- q** Output first 1/4 data of FFT results only. As in the above **c** option, boundary data is compensated and  $(\frac{L}{2} + 1) \times (\frac{L}{2} + 1)$  data are output. [FALSE]



- R** output only real part [FALSE]  
**-I** output only imaginary part [FALSE]

### EXAMPLE

This example reads a sequence of 2-dimensional complex numbers in float format from *data.f* file, evaluates its 2-dimensional IDFT and outputs it to *data.dft* file:

```
ifft2 < data.f > data.ifft2
```

### SEE ALSO

fft, fft2, ifft



**NAME**

`ignorm` – inverse gain normalization

**SYNOPSIS**

`ignorm [-m M] [-g G] [-c C] [infile]`

**DESCRIPTION**

`ignorm` unnormalizes normalized generalized cepstral coefficients  $c_\gamma(m)$  from `infile` (or standard input), sending the unnormalized generalized cepstral coefficients to standard output.

Input and output data are in float format.

To convert normalized generalized cepstral coefficients  $c'_\gamma(m)$  into not-normalized generalized cepstral coefficients  $c_\gamma(m)$ , the following equation can be used.

$$c_\gamma(m) = \left(c'_\gamma(0)\right)^\gamma c'_\gamma(m), \quad m > 0$$

Also, the gain  $K = c_\gamma(0)$  is

$$c_\gamma(0) = \begin{cases} \frac{\left(c'_\gamma(0)\right)^\gamma - 1.0}{\gamma}, & 0 < |\gamma| \leq 1 \\ \log c'_\gamma(0), & \gamma = 0 \end{cases}$$

**OPTIONS**

- |           |          |  |      |
|-----------|----------|--|------|
| <b>-m</b> | <i>M</i> | order of generalized cepstrum                    | [25] |
| <b>-g</b> | <i>G</i> | power parameter $\gamma$ of generalized cepstrum | [0]  |
|           |          | $\gamma = G$                                     |      |
| <b>-c</b> | <i>C</i> | power parameter $\gamma$ of generalized cepstrum |      |
|           |          | $\gamma = -1/(\text{int})C$                      |      |
|           |          | <i>C</i> must be $C \geq 1$                      |      |

**EXAMPLE**

In this example below, normalized generalized cepstral coefficients in float format are read from `data.ngcep` ( $M = 15, \gamma = -0.5$ ), and the not-normalized generalized cepstral coefficients are output to `data.gcep`.

```
ignorm -m 15 -c 2 < data.ngcep > data.gcep
```

**SEE ALSO**

`gcep`, `mgcep`, `gc2gc`, `mgc2mgc`, `freqt`

**NAME**

`impulse` – generate impulse sequence

**SYNOPSIS**

`impulse` `[-l L]` `[-n N]`

**DESCRIPTION**

*impulse* generates the unit impulse sequence of length  $L$ , sending the output to standard output. The output is in float format as follows.

$$\underbrace{1, 0, 0, \dots, 0}_L$$

If both `-l` and `-n` options are given, the last one is used.

**OPTIONS**

- |                 |  |       |
|-----------------|--|-------|
| <code>-l</code> | $L$ length of unit impulse<br>if $L < 0$ then endless sequence is generated. | [256] |
| <code>-n</code> | $N$ order of unit impulse  | [255] |

**EXAMPLE**

In the example below, an unit impulse sequence is passed through a digital filter and the results is presented on the screen.

```
impulse | dfs -a 1 0.9 -b 1 2 1 | dmp +f
```

**SEE ALSO**

`step`, `train`, `ramp`, `sin`, `nrand`

**NAME**

`imsvq` – decoder of multi stage vector quantization

**SYNOPSIS**

`imsvq` [ `-l L` ] [ `-n N` ] [ `-s S cbfile` ] [ `infile` ]

**DESCRIPTION**

`imsvq` decodes multi-stage vector-quantized data from a sequence of codebook indexes from `infile` (or standard input), using codebooks specified by multiple `-s` options, sending the result to standard output. The number of decoder stages is equal to the number of `-s` options.

Input data is in int format, and output data is in float format.

**OPTIONS**

<code>-l</code>	<code>L</code>	length of vector	[26]
<code>-n</code>	<code>N</code>	order of vector	[L-1]
<code>-s</code>	<code>S cbfile</code>	codebook	[N/A N/A]
	<code>S</code>	codebook size	
	<code>cbfile</code>	codebook file	

**EXAMPLE**

In the example below, the decoded vector `data.ivq` is obtained from the first stage codebook `cbfile1` and the second stage codebook `cbfile2`, both of size 256, as well as from the index file `data.vq`.

```
imsvq -s 256 cbfile1 -s 256 cbfile2 < data.vq > data.ivq
```

**SEE ALSO**

`msvq`, `ivq`, `vq`

**NAME**

interpolate – interpolation of data sequence

**SYNOPSIS**

**interpolate** [ **-p** *P* ] [ **-s** *S* ] [ **-d** ] [ *infile* ]

**DESCRIPTION**

*interpolate* supplements a sequence of input data from *infile* (or standard input) by 0 or input data with interval *P* and start number *S*, sending the result to standard output.

If the input data is

$$x(0), x(1), x(2), \dots$$

then the output data is

$$\underbrace{0, 0, \dots, 0}_{S-1}, \underbrace{x(0), 0, 0, \dots, 0}_P, \underbrace{x(1), 0, 0, \dots, 0}_P, x(2), \dots$$

If the **-d** option is given, the output data is

$$\underbrace{0, 0, \dots, 0}_{S-1}, \underbrace{x(0), x(0), x(0), \dots, x(0)}_P, \underbrace{x(1), x(1), x(1), \dots, x(1)}_P, x(2), \dots$$

Input and output data are in float format.

**OPTIONS**

<b>-p</b>	<i>P</i>	interpolation period	[10]
<b>-s</b>	<i>S</i>	start sample	[0]
<b>-d</b>		pad input data rather than 0	[FALSE]

**EXAMPLE**

This example decimates input data from *data.f* file with interval 2, interpolates 0 with interval 2, and then outputs it to *data.di* file:

```
decimate -p 2 < data.f | interpolate -p 2 > data.di
```

**SEE ALSO**

decimate

**NAME**

*ivq* – decoder of vector quantization

**SYNOPSIS**

***ivq*** [ **-l** *L* ] [ **-n** *N* ] *cbfile* [ *infile* ]

**DESCRIPTION**

*ivq* decodes vector-quantized data from a sequence of codebook indexes from *infile* (or standard input), using the codebook *cbfile*, sending the result to standard output. The decoded output vector format is

$$c_i(0), c_i(1), \dots, c_i(L-1).$$

Input data is in int format, and output data is in float format.

**OPTIONS**

<b>-l</b>	<i>L</i>	length of vector	[26]
<b>-n</b>	<i>N</i>	order of vector	[L-1]

**EXAMPLE**

In the following example, the decoded 25-th order output file *data.ivq* is obtained through the index file *data.vq* and codebook *cbfile*.

```
ivq cbfile data.vq > data.ivq
```

**SEE ALSO**

*vq*, *imsvq*, *msvq*

**NAME**

lbg – LBG algorithm for vector quantizer design

**SYNOPSIS**

**lbg** [-l *L*] [-n *N*] [-t *T*] [-s *S*] [-e *E*] [-f *F*] [-i *I*] [-m *M*] [-S *s*]  
[-c *C*] [-d *D*] [-r *R*] [*indexfile*] <*infile*

**DESCRIPTION**

*lbg* uses the LBG algorithm to train a codebook from a sequence of vectors from *infile* (or standard input), sending the result to standard output.

The input sequence consists of  $T$  float vectors  $\mathbf{x}$ , each of size  $L$

$$\mathbf{x}(0), \mathbf{x}(1), \dots, \mathbf{x}(T-1).$$

The result is a codebook consisting of  $E$  float vectors, each of length  $L$ ,

$$\mathbf{C}_E = \{\mathbf{c}_E(0), \mathbf{c}_E(1), \dots, \mathbf{c}_E(E-1)\},$$

generated by the following algorithm.

**step.0** When a initial codebook  $\mathbf{C}_S$  is not assigned, the initial codebook is obtained from the whole collection of training data as follows,

$$\mathbf{c}_1(0) = \frac{1}{T} \sum_{n=0}^{T-1} \mathbf{x}(n)$$

and the initial codebook with  $S = 1$  is  $\mathbf{C}_1 = \{\mathbf{c}_1(0)\}$ .

**step.1** From codebook  $\mathbf{C}_S$  obtain  $\mathbf{C}_{2S}$ . For this step, normalized random vector of size  $L$  and splitting factor  $R$  are used as follows,

$$\mathbf{c}_{2S}(n) = \begin{cases} \mathbf{c}_S(n) + R \cdot \mathbf{rnd} & (0 \leq n \leq S-1) \\ \mathbf{c}_S(n-S) - R \cdot \mathbf{rnd} & (S \leq n \leq 2S-1) \end{cases}$$

and we make  $D_0 = \infty$ ,  $k = 0$ .

**step.2** First, check that  $k \leq I$  where  $I$  is the maximum number of the iteration specified by  $-i$  option. If it is true, proceed the following steps. If not, then go to **step.4**. The present codebook  $\mathbf{C}_{2S}$  is now applied to the training vectors. After that the mean Euclidean distance  $D_k$  is evaluated from every training vector and the corresponding code vector. If the following condition

$$\left| \frac{D_{k-1} - D_k}{D_k} \right| < D$$

is valid then go to **step.4**. If it is not valid then go to **step.3**.

**step.3** Centroids are evaluated from the results obtained in **step.2**. The codebook  $C_{2S}$  is updated. Also, if a cell has training vectors less than  $M$ , then the corresponding code vector is erased from codebook, and a new code vector is generated from 1) the code vector  $c_{2S}(j)$  corresponding to the cell with more training vectors as follows.

$$c_{2S}(i) = c_{2S}(j) + R \cdot \mathbf{rnd}$$

Also,  $c_{2S}(j)$  is modified as follows.

$$c_{2S}(j) = c_{2S}(j) - R \cdot \mathbf{rnd}$$

2) the vector  $\mathbf{p}$  which internally divide two centroids in proportion to the number of training vectors for the cell. They were split from the same parent centroid. The vector  $\mathbf{p}$  can be written as follows,

$$\mathbf{p} = \frac{n_j c_{2S}(i) + n_i c_{2S}(j)}{n_i + n_j},$$

where  $n_i$  and  $n_j$  are the number of training vectors for the cell. The update method is as follows.

$$c_{2S}(i) = \mathbf{p} + R \cdot \mathbf{rnd},$$

$$c_{2S}(j) = \mathbf{p} - R \cdot \mathbf{rnd}.$$

The type of split can be specified by  $-c$  option. After that, we assign  $k = k + 1$  then go back to **step.2**

**step.4** If  $2S = E$  then end. If it is not then we make  $S = 2S$  and go back **step.1**.

## OPTIONS

<b>-l</b>	$L$	length of vector	[26]
<b>-n</b>	$N$	order of vector	[L-1]
<b>-t</b>	$T$	number of training vector	[N/A]
<b>-s</b>	$S$	initial codebook size	[1]
<b>-e</b>	$E$	final codebook size	[256]
<b>-f</b>	$F$	initial codebook filename	[NULL]
<b>-i</b>	$I$	maximum number of iteration for centroid update	[1000]
<b>-m</b>	$M$	minimum number of training vectors for each cell	[1]
<b>-S</b>	$s$	seed for normalized random vector	[1]
<b>-c</b>	$C$	type of exception procedure for centroid update when the number of training vectors for the cell is less than $M$	[1]
	$C = 1$	split the centroid with most training vectors	
	$C = 2$	split the vector which internally divide two centroids sharing the same parent centroid, in proportion to the number of training vectors for the cell.	

Usually, the options below do not need to be assigned.

<b>-d</b>	$D$	end condition	[0.0001]
<b>-r</b>	$R$	splitting factor	[0.0001]

**EXAMPLE**

In the following example, a codebook of size 1024 is generated from the 39-th order training vector *data.f* in float format. It is also specified that the iteration of centroid update is at most 100 times, each centroid contains at least 10 training vectors and the random vectors for the centroid update are generated with seed 5. The output is written to *cbfile*.

```
lbg -n 39 -e 1024 -i 100 -m 10 -S 5 < data.f > cbfile
```

**SEE ALSO**

vq, ivq, msvq



**NAME**

`levdur` – solve an autocorrelation normal equation using Levinson-Durbin method

**SYNOPSIS**

`levdur` [ `-m M` ] [ `-f F` ] [ `infile` ]

**DESCRIPTION**

`levdur` calculates linear prediction coefficients (LPC) from the autocorrelation matrix from `infile` (or standard input), sending the result to standard output.

The input is the  $M$ -th order autocorrelation matrix

$$r(0), r(1), \dots, r(M).$$

`levdur` uses the Levinson-Durbin algorithm to solve a system of linear equations obtained from the autocorrelation matrix.

Input and output data are in float format.

The linear prediction coefficients are the set of coefficients  $K, a(1), \dots, a(M)$  of the all-pole digital filter

$$H(z) = \frac{K}{1 + \sum_{i=1}^M a(i)z^{-i}}.$$

The linear prediction coefficients are evaluated by solving the following set of linear equations, which were obtained through the autocorrelation method,

$$\begin{pmatrix} r(0) & r(1) & \dots & r(M-1) \\ r(1) & r(0) & & \vdots \\ \vdots & & \ddots & \\ r(M-1) & \dots & & r(0) \end{pmatrix} \begin{pmatrix} a(1) \\ a(2) \\ \vdots \\ a(M) \end{pmatrix} = - \begin{pmatrix} r(1) \\ r(2) \\ \vdots \\ r(M) \end{pmatrix}$$

The Durbin iterative and efficient algorithm is used in the following taking advantage of the Toeplitz characteristic of the autocorrelation matrix:

$$E^{(0)} = r(0)$$

$$k(i) = \frac{-r(i) - \sum_{j=1}^i a^{(i-1)}(j)r(i-j)}{E^{(i-1)}}$$

$$a^{(i)}(i) = k(i)$$

$$a^{(i)}(j) = a^{(i-1)}(j) + k(i)a^{(i-1)}(i-j), \quad 1 \leq j \leq i-1 \quad (1)$$

$$E^{(i)} = (1 - k^2(i))E^{(i-1)} \quad (2)$$

Also, for  $i = 1, 2, \dots, M$ , equations (1) to (2) are applied recursively, and the gain  $K$  is calculated as follows.

$$K = \sqrt{E^{(M)}}$$

**OPTIONS**

**-m** *M* order of correlation [25]  
**-f** *F* minimum value of the determinant of the normal matrix [0.000001]

**EXAMPLE**

In this example, input data is read in float format from *data.f* and linear prediction coefficients are written to *data.lpc*:

```
frame +f < data.f | window | acorr -m 25 | levdur > data.lpc
```

**SEE ALSO**

acorr, lpc

**NAME**

`linear_intpl` – linear interpolation of data

**SYNOPSIS**

`linear_intpl` [ **-l** *L* ] [ **-m** *M* ] [ **-x** *x<sub>min</sub>* *x<sub>max</sub>* ] [ **-i** *x<sub>min</sub>* ] [ **-j** *x<sub>max</sub>* ] [ *infile* ]

**DESCRIPTION**

`linear_intpl` reads a 2-dimensional input data sequence from *infile* (or standard input) and outputs *y*-axis values when *x*-axis are linearly interpolated by equally-spaced  $L - 1$  points.

If the input data is

$$\begin{array}{c} x_0, y_0 \\ x_1, y_1 \\ \vdots \\ x_K, y_K \end{array}$$

then the output data is

$$y_0, y_1, \dots, y_{L-1}$$

Input and output data are in float format.

This command can interpolate data sequence whose *x*-axis is not equally-spaced, such as digital filter characteristics.

**OPTIONS**

<b>-l</b>	<i>L</i>	output length	[256]
<b>-m</b>	<i>M</i>	number of interpolation points	[ <i>L</i> -1]
<b>-x</b>	<i>x<sub>min</sub></i> <i>x<sub>max</sub></i>	minimum and maximum values of <i>x</i> -axis in input data	[0.0 0.5]
<b>-i</b>	<i>x<sub>min</sub></i>	minimum values of <i>x</i> -axis in input data	[0.0]
<b>-j</b>	<i>x<sub>max</sub></i>	maximum values of <i>x</i> -axis in input data	[0.5]

**EXAMPLE**

This example decimates input data from *data.f* file with interval 2, interpolates 0 with interval 2, and then outputs it to *data.di* file:

When input data *data.f* contains the following data,

$$\begin{array}{c} 0, 2 \\ 2, 2 \\ 3, 0 \\ 5, 1 \end{array}$$

this example linearly interpolates input data and outputs it to *data.intpl*

```
linear_intpl -m 10 -x 0 5 < data.f > data.intpl
```

The result becomes

```
2, 2, 2, 2, 2, 1, 0, 0.25, 0.5, 0.75, 1
```

**NAME**

`lmadf` – LMA digital filter for speech synthesis(5; 17)

**SYNOPSIS**

`lmadf` [ **-m** *M* ] [ **-p** *P* ] [ **-i** *I* ] [ **-P** *Pa* ] [ **-v** ] [ **-t** ] [ **-k** ] *cfile* [ *infile* ]

**DESCRIPTION**

`lmadf` derives a Log Magnitude Approximation filter from cepstral coefficients  $c(0), c(1), \dots, c(M)$  in *cfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

The LMA filter is an extremely precise approximation of the exponential transfer function obtained from  $M$ -th order cepstral coefficients  $c(m)$  as follows.

$$H(z) = \exp \sum_{m=0}^M c(m)z^{-m}$$

If we remove from the transfer function  $H(z)$  the gain  $K = \exp c(0)$ , then we obtain the following transfer function

$$D(z) = \exp \sum_{m=1}^M c(m)z^{-m},$$

which can be realized using the basic FIR filter

$$F(z) = \sum_{m=1}^M c(m)z^{-m}$$

as shown in figure 1(a). Also, as can be seen from figure 1(b), the basic filter  $F(z)$  can be decomposed as follows

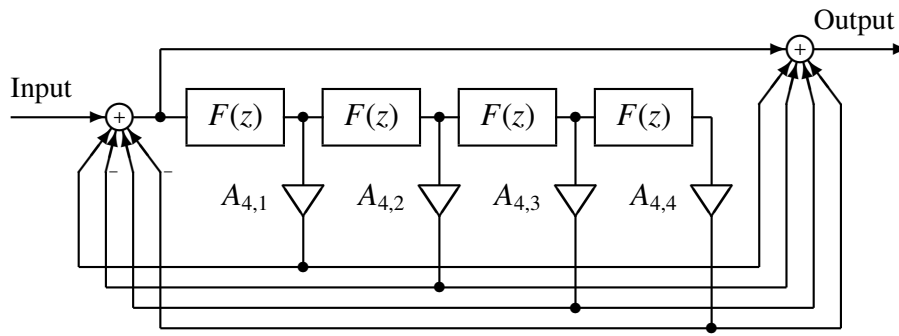
$$F(z) = F_1(z) + F_2(z)$$

where

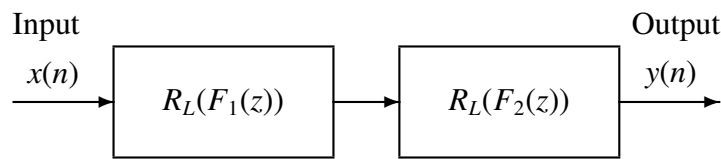
$$F_1(z) = c(1)z^{-1}$$

$$F_2(z) = \sum_{m=2}^M c(m)z^{-m}$$

By doing this decomposition, the accuracy of the approximation is improved. Also, the values of the coefficients  $A_{4,l}$  are given in table 1



(a)



(b)

Figure 1: (a)  $R_L(F(z)) \approx D(z)$   $L = 4$   
 (b) 2 level cascade realization  
 $R_L(F_1(z)) \cdot R_L(F_2(z)) \approx D(z)$

Table 1: The values for the coefficients  $A_{L,l}$

$l$	$A_{4,l}$	$A_{5,l}$
1	$4.999273 \times 10^{-1}$	$4.999391 \times 10^{-1}$
2	$1.067005 \times 10^{-1}$	$1.107098 \times 10^{-1}$
3	$1.170221 \times 10^{-2}$	$1.369984 \times 10^{-2}$
4	$5.656279 \times 10^{-4}$	$9.564853 \times 10^{-4}$
5		$3.041721 \times 10^{-5}$

**OPTIONS**

- m**  $M$  order of cepstrum [25]
- p**  $P$  frame period [100]
- i**  $I$  interpolation period [1]
- P**  $Pa$  order of the Padé approximation [4]  
 $Pa$  should be 4 or 5
- k** filtering without gain [FALSE]
- v** inverse filter [FALSE]

`-v` transpose filter [FALSE]

### EXAMPLE

In this example, the excitation is generated from the pitch data read in float format from *data.pitch*, passed through an LMA filter obtained from cepstrum file *data.cep*, and the synthesized speech is written to *data.syn*.

```
excite < data.pitch | lmadf data.cep > data.syn
```

### SEE ALSO

uels, acep, poledf, ltcdf, glsadf, mlsadf, mgladf

**NAME**

`lpc` – LPC analysis using Levinson-Durbin method

**SYNOPSIS**

`lpc [-l L] [-m M] [-f F] [infile]`

**DESCRIPTION**

`lpc` calculates linear prediction coefficients (LPC) from  $L$ -length framed windowed data from *infile* (or standard input), sending the result to standard output.

For each  $L$ -length input vector

$$x(0), x(1), \dots, x(L-1),$$

the autocorrelation function is calculated (see `acorr`), then the gain  $K$  and the linear prediction coefficients  $a(k)$  ( $1 \leq k \leq M$ )

$$K, a(1), \dots, a(M)$$

are calculated using the Levinson-Durbin algorithm (see `levdur`).

Input and output data are in float format.

**OPTIONS**

<code>-l</code>	$L$	frame length	[256]
<code>-m</code>	$M$	order of LPC	[25]
<code>-f</code>	$F$	minimum value of the determinant of the normal matrix	[0.000001]

**EXAMPLE**

In this example, the 20-th order linear prediction analysis is applied to input read from *data.f* in float format, and the linear prediction coefficients are written to *data.lpc*:

```
frame +f < data.f | window | lpc -m 20 > data.lpc
```

**SEE ALSO**

`acorr`, `levdur`, `lpc2par`, `par2lpc`, `lpc2c`, `lpc2lsp`, `lsp2lpc` `lpcdf`, `lspdf`



**NAME**

`lpc2c` – transform LPC to cepstrum

**SYNOPSIS**

`lpc2c` [ `-m`  $M_1$  ] [ `-M`  $M_2$  ] [ *infile* ]

**DESCRIPTION**

`lpc2c` calculates LPC cepstral coefficients from linear prediction (LPC) coefficients from *infile* (or standard input), sending the result to standard output. That is, when the input sequence is

$$\sigma, a(1), a(2), \dots, a(p)$$

where

$$H(z) = \frac{\sigma}{A(z)} = \frac{\sigma}{1 + \sum_{k=1}^P a(k)z^{-k}}$$

then the LPC cepstral coefficients are evaluated as follows.

$$c(n) = \begin{cases} \ln(h), & n = 0 \\ -a(n) = -\sum_{k=1}^{n-1} \frac{k}{n} c(k)a(n-k), & 1 \leq n \leq P \\ -\sum_{k=n-P}^{n-1} \frac{k}{n} c(k)a(n-k), & n > P \end{cases}$$

And the sequence of cepstral coefficients

$$c(0), c(1), \dots, c(M)$$

is output. Input and output data are in float format.

**OPTIONS**

`-m`  $M_1$  order of LPC [25]  
`-M`  $M_2$  order of cepstrum [25]

**EXAMPLE**

In the example below, a 10-th order LPC analysis is undertaken after passing the speech data *data.f* in float format through a window, 15-th order LPC cepstral coefficients are calculated, and the result is written in *data.cep*.

```
frame +f < data.f | window | lpc -m 10 |\
lpc2c -m 10 -M 15 > data.cep
```

**SEE ALSO**

lpc, gc2gc, mgc2mgc, freqt

**NAME**

`lpc2lsp` – transform LPC to LSP

**SYNOPSIS**

```
lpc2lsp [-m M] [-s S] [-k] [-l] [-o O] [-n N] [-p P] [-q Q] [-d D]
        [ infile ]
```

**DESCRIPTION**

`lpc2lsp` calculates line spectral pair (LSP) coefficients from  $M$ -th order linear prediction (LPC) coefficients from *infile* (or standard input), sending the result to standard output.

The gain  $K$  is included in the LPC input vectors

$$K, a(1), \dots, a(M)$$

but  $K$  is not used in the calculation of the LSP coefficients.

The  $M$ -th order polynomial linear prediction equation  $A(z)$  is

$$A_M(z) = 1 + \sum_{m=1}^M a(m)z^{-m}$$

The PARCOR coefficients satisfy the following equations.

$$\begin{aligned} A_m(z) &= A_{m-1}(z) - k(m)B_{m-1}(z) \\ B_m(z) &= z^{-1}(B_{m-1}(z) - k(m)A_{m-1}(z)) \end{aligned}$$

Also, the initial conditions are set as follows,

$$\begin{aligned} A_0(z) &= 1 \\ B_0(z) &= z^{-1}. \end{aligned} \tag{1}$$

When we are given the linear prediction polynomial equation of  $M$ -th order  $A_M(z)$ , and when the evaluation of  $A_{M+1}(z)$  is obtained with the value of  $k(M+1)$  equal to 1 or  $-1$ ,  $P(z)$  and  $Q(z)$  are defined as follow.

$$\begin{aligned} P(z) &= A_M(z) - B_M(z) \\ Q(z) &= A_M(z) + B_M(z) \end{aligned}$$

Making  $k(M+1)$  equal to  $\pm 1$  means that, with respect PARCOR coefficients, the boundary condition for the glottis of the fixed vocal tract model satisfies a perfect reflection characteristic. Also,  $A_M(z)$  can be expressed as

$$A_M(z) = \frac{P(z) + Q(z)}{2}.$$

When we express  $A_M(z)$  in this way,  $A_M(z)$  is stable. That is for the roots of  $A_M(z) = 0$  to be all inside the unit circle a necessary and sufficient condition is given in the following.

- All of the roots of  $P(z) = 0$  and  $Q(z) = 0$  are on the unit circle line.
- the roots of  $P(z) = 0$  and  $Q(z) = 0$  should be above the unit circle line and intercalate.

In other words, if the roots of  $P(z) = 0$  and  $Q(z) = 0$  satisfy the above condition, then  $A_M(z)$  is stable.

If we assume that  $M$  is a even number, then  $P(z)$  and  $Q(z)$  can be factorized as follows.

$$P(z) = (1 - z^{-1}) \prod_{i=2,4,\dots,M} (1 - 2z^{-1} \cos \omega_i + z^{-2})$$

$$Q(z) = (1 + z^{-1}) \prod_{i=1,3,\dots,M-1} (1 - 2z^{-1} \cos \omega_i + z^{-2})$$

Also, the values of  $\omega_i$  satisfy the following ordering condition.

$$0 < \omega_1 < \omega_2 < \dots < \omega_{M-1} < \omega_M < \pi$$

In the case,  $M$  is odd number solution can be found in a similar way. The coefficients  $\omega_i$  obtained through factorization are called LSP coefficients.

## OPTIONS

<b>-m</b>	$M$	order of LPC	[25]
<b>-s</b>	$S$	sampling frequency (kHz)	[10]
<b>-k</b>		output gain	[TRUE]
<b>-l</b>		output log gain instead of linear gain	[FALSE]
<b>-o</b>	$O$	output format	[0]
	0	normalized frequency ( $0 \dots \pi$ )	
	1	normalized frequency ( $0 \dots 0.5$ )	
	2	frequency (kHz)	
	3	frequency (Hz)	

Usually, the options below do not need to be assigned.

<b>-n</b>	$N$	split number of unit circle	[128]
<b>-p</b>	$P$	maximum number of interpolation	[4]
<b>-d</b>	$D$	end condition of interpolation	[1e-06]

## EXAMPLE

In the following example, speech data is read in float format from *data.f*, 10-th order LPC coefficients are calculated, and the LSP coefficients are evaluated and written to *data.lsp*:

```
frame +f < data.f | window | lpc -m 10 |\
lpc2lsp -m 10 > data.lsp
```

**SEE ALSO**

lpc, lsp2lpc, lspdf

**NAME**

`lpc2par` – transform LPC to PARCOR

**SYNOPSIS**

`lpc2par` [ `-m M` ] [ `-g G` ] [ `-c C` ] [ `-s` ] [ *infile* ]

**DESCRIPTION**

`lpc2par` calculates PARCOR coefficients from  $M$ -th order linear prediction (LPC) coefficients from *infile* (or standard input), sending the result to standard output.

The LPC input format is

$$K, a(1), \dots, a(M),$$

and the PARCOR output format is

$$K, k(1), \dots, k(M).$$

If the `-s` option is assigned, the stability of the filter is analyzed. If the filter is stable, then 0 is returned. If the filter is not stable, then 1 is returned to the standard output.

Input and output data are in float format.

The transformation from LPC coefficients to PARCOR coefficients is undertaken as follows:

$$k(m) = a^{(m)}(m)$$

$$a^{(m-1)}(i) = \frac{a^{(m)}(i) + a^{(m)}(m)a^{(m)}(m-i)}{1 - k^2(m)},$$

where  $1 \leq i \leq m-1$ ,  $m = p, p-1, \dots, 1$ . The initial condition is

$$a^{(M)}(m) = a(m), \quad 1 \leq m \leq M.$$

If we use the `-g` option, then the input contains normalized generalized cepstral coefficients with power parameter  $\gamma$  and the output contains the corresponding PARCOR coefficients. In other words, the input is

$$K, c'_\gamma(1), \dots, c'_\gamma(M)$$

and the initial condition is

$$a^{(M)}(m) = \gamma c'_\gamma(M), \quad 1 \leq m \leq M.$$

Also with respect to the stability analysis, the PARCOR coefficients are checked through the following equation.

$$-1 < k(m) < 1$$

If this condition satisfy then the filter is stable.

**OPTIONS**

- |           |          |                               |         |
|-----------|----------|-------------------------------|---------|
| <b>-m</b> | <i>M</i> | order of LPC                  | [25]    |
| <b>-g</b> | <i>G</i> | gamma of generalized cepstrum | [0]     |
|           |          | $\gamma = G$                  |         |
| <b>-c</b> | <i>C</i> | gamma of generalized cepstrum |         |
|           |          | $\gamma = -1/(\text{int})C$   |         |
|           |          | <i>C</i> must be $C \geq 1$   |         |
| <b>-s</b> |          | check stable or unstable      | [FALSE] |

**EXAMPLE**

In the example below, a linear prediction analysis is done in the input file *data.f* in float format, the LPC coefficients are then transformed into PARCOR coefficients, and the output is written to *data.rc*:

```
frame +f < data.f | window | lpc | lpc2par > data.rc
```

**SEE ALSO**

acorr, levdur, lpc, par2lpc, ltcdf

**NAME**

`lsp2lpc` – transform LSP to LPC

**SYNOPSIS**

`lsp2lpc` [ `-m` *M* ] [ `-s` *S* ] [ `-k` ] [ `-l` ] [ `-i` *I* ] [ *infile* ]

**DESCRIPTION**

`lsp2lpc` calculates linear prediction (LPC) coefficients from *M*-th order line spectral pair (LSP) coefficients from *infile* (or standard input), sending the result to standard output.

The LSP input format is

$$[K], l(1), \dots, l(M),$$

and the LPC output format is

$$K, a(1), \dots, a(M).$$

By default, `lsp2lpc` assumes that the LSP input vectors include the gain *K*, and it passes that gain value through to the LPC output vectors. However, if the `-k` option is present, `lsp2lpc` assumes that *K* is not present in the LSP input vectors, and it sets *K* to 1.0 in the LPC output vectors.

**OPTIONS**

<code>-m</code>	<i>M</i>	order of LPC	[25]
<code>-s</code>	<i>S</i>	sampling frequency(kHz)	[10]
<code>-k</code>		input & output gain	[TRUE]
<code>-l</code>		regard input as log gain and output linear gain	[FALSE]
<code>-i</code>	<i>I</i>	input format	[0]
	0	normalized frequency ( $0 \dots \pi$ )	
	1	normalized frequency ( $0 \dots 0.5$ )	
	2	frequency (kHz)	
	3	frequency (Hz)	

**EXAMPLE**

In the example below, 10-th order LSP coefficients in float format are read from file *data.lsp*, the linear prediction coefficients are evaluated, and written to *data.lpc*:

```
lsp2lpc -m 10 < data.lsp > data.lpc
```

**SEE ALSO**

`lpc`, `lpc2lsp`



**NAME**

lspcheck – check stability and rearrange LSP

**SYNOPSIS**

**lspcheck** [ **-m** *M* ] [ **-s** *S* ] [ **-k** ] [ **-i** *I* ] [ **-o** *O* ] [ **-r** *R* ] [ *infile* ]

**DESCRIPTION**

*lspcheck* tests the stability of the filter corresponding to the line spectral pair (LSP) coefficients from *infile* (or standard input), sending the result to standard output.

By default, the output is an ASCII report of the unstable frames. However, if the **-r** option is given, the output is frames of coefficients that have been rearranged so the filter is stable.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of LPC	[25]
<b>-s</b>	<i>S</i>	sampling frequency(kHz)	[10]
<b>-k</b>		input & output gain	[TRUE]
<b>-i</b>	<i>I</i>	input format	[0]
<b>-o</b>	<i>O</i>	output format	[I]
		0 normalized frequency ( $0 \dots \pi$ )	
		1 normalized frequency ( $0 \dots 0.5$ )	
		2 frequency (kHz)	
		3 frequency (Hz)	
<b>-r</b>	<i>R</i>	rearrange LSP	[FALSE]
		check the distance between two consecutive LSPs	
		and extend the distance (if it is smaller than $R \times \pi/M$ )	
		<i>s.t.</i> $0 < R < 1$	

**EXAMPLE**

In the following example, 10-th order LSP coefficients are read from *data.lsp* in float format, stability is checked, the unstable coefficients are rearranged so that they become stable, and the distance between two consecutive LSPs are extended to  $\pi/1000$  if it is smaller than  $\pi/1000$ , and the rearranged LSP coefficients are written to *data.lspr*:

```
lspcheck -m 10 -r 0.01 < data.lsp > data.lspr
```

**SEE ALSO**

lpc, lpc2lsp, lsp2lpc

**NAME**

`lspdf` – LSP speech synthesis digital filter

**SYNOPSIS**

`lspdf` [ **-m** *M* ] [ **-p** *P* ] [ **-i** *I* ] [ **-s** *S* ] [ **-o** *O* ] [ **-k** ] [ **-l** ] *lspfile* [ *infile* ]

**DESCRIPTION**

*lspdf* derives an LSP digital filter from line spectral pair (LSP) coefficients in *lspfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output are in float format.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of coefficients	[25]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-k</b>		filtering without gain	[FALSE]
<b>-l</b>		regard input as log gain	[FALSE]

**EXAMPLE**

In the example below, excitation is generated from pitch information given in *data.pitch* in float format, this excitation is passed through the LSP synthesis filter constructed from the LSP file *data.lsp*, and the synthesized speech is written to *data.syn*:

```
excite < data.pitch | lspdf data.lsp > data.syn
```

**SEE ALSO**

`lsp`, `lpc2lsp`

**NAME**

ltcdf – all-pole lattice digital filter for speech synthesis

**SYNOPSIS**

```
ltcdf [ -m M ] [ -p P ] [ -i I ] [ -k ] rcfile [ infile ]
```

**DESCRIPTION**

*lsp2df* derives an all-pole lattice digital filter from PARCOR coefficients in *rcfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of coefficients	[25]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-k</b>		filtering without gain	[FALSE]

**EXAMPLE**

In the example below, excitation is generated from pitch information given in *data.pitch* in float format, this excitation is passed through the lattice filter constructed from the LPC file *data.rc*, and the synthesized speech is written to *data.syn*:

```
excite < data.pitch | ltcdf data.k > data.syn
```

**SEE ALSO**

lpc, acorr, levdur, lpc2par, par2lpc, poledf, zerodf, lspdf

**NAME**

`mc2b` – transform mel-cepstrum to MLSA digital filter coefficients

**SYNOPSIS**

`mc2b` [ `-a A` ] [ `-m M` ] [ *infile* ]

**DESCRIPTION**

`mc2b` calculates MLSA filter coefficients  $b(m)$  from mel-cepstral coefficients  $c_\alpha(m)$  from *infile* (or standard input), sending the result to standard output.

Input and output data are in float format.

The equations are used for this transformation follows.

$$b(m) = \begin{cases} c_\alpha(M), & m = M \\ c_\alpha(m) - \alpha b(m+1), & 0 \leq m < M \end{cases}$$

These coefficients  $b(m)$  can be directly used in the implementation of a MLSA filter. This transformation is the inverse transformation undertaken by the command `b2mc`.

**OPTIONS**

`-a A` all-pass constant  $\alpha$  [0.35]  
`-m M` order of mel-cepstrum [25]

**EXAMPLE**

Speech data is read in float format from *data.f*, a 12-th order mel-cepstral analysis is undertaken, these mel-cepstral coefficients are transformed into MLSA filter coefficients, and these coefficients  $b(m)$  are written to *data.b*:

```
frame +f < data.f | window | mcep -m 12 |\
mc2b -m 12 > data.b
```

**SEE ALSO**

`mlsadf`, `mglisadf`, `b2mc`, `mcep`, `mgcep`, `amcep`

**NAME**

mcep – mel cepstral analysis(10; 12)

**SYNOPSIS**

**mcep** [ **-a** *A* ] [ **-m** *M* ] [ **-l** *L* ] [ **-q** *Q* ] [ **-i** *I* ] [ **-j** *J* ] [ **-d** *D* ] [ **-e** *E* ] [ **-f** *F* ]  
 [ *infile* ]

**DESCRIPTION**

*mcep* uses mel-cepstral analysis to calculate mel-cepstral coefficients  $c_\alpha(m)$  from  $L$ -length framed windowed data from *infile* (or standard input), sending the result to standard output.

Input and output data are in float format.

In the mel-cepstral analysis, the spectrum of the speech signal is modeled by  $M$ -th order mel-cepstral coefficients  $c_\alpha(m)$  as follows.

$$H(z) = \exp \sum_{m=0}^M c_\alpha(m) \tilde{z}^{-m}$$

For this command “mcep”, it is applied a cost function based on the unbiased estimation log spectrum method. The variable  $\tilde{z}^{-1}$  can be expressed as the following first order all-pass function

$$\tilde{z}^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}.$$

The phase characteristic is given by the variable  $\alpha$ . For a sampling rate 16 kHz,  $\alpha$  is made equal to 0.42. For a sampling rate 10 kHz,  $\alpha$  is made equal to 0.35. For a sampling rate 8 kHz,  $\alpha$  is made equal to 0.31. By making these choices for  $\alpha$ , the mel-scale becomes the good approximation to human sensitivity to the loudness speech sound.

The Newton-Raphson method is used to minimize the cost function when evaluating mel-cepstral coefficients.

**OPTIONS**

<b>-a</b>	<i>A</i>	all-pass constant $\alpha$	[0.35]
<b>-m</b>	<i>M</i>	order of mel cepstrum	[25]
<b>-l</b>	<i>L</i>	frame length	[256]
<b>-q</b>	<i>Q</i>	input data style	[0]
	$Q = 0$	windowed data sequence	
	$Q = 1$	$20 \times \log  f(w) $	
	$Q = 2$	$\ln  f(w) $	
	$Q = 3$	$ f(w) $	
	$Q = 4$	$ f(w) ^2$	

Usually, the options below do not need to be assigned.

<b>-i</b>	<i>I</i>	minimum iteration of Newton-Raphson method	[2]
<b>-j</b>	<i>J</i>	maximum iteration of Newton-Raphson method	[30]
<b>-d</b>	<i>D</i>	end condition of Newton-Raphson	[0.001]
<b>-e</b>	<i>E</i>	small value added to periodgram	[0.0]
<b>-f</b>	<i>F</i>	minimum value of the determinant of the normal matrix	[0.000001]

### EXAMPLE

Speech data is read in float format from *data.f* and analyzed, and mel-cepstral coefficients are written to *data.mcep*:

```
frame +f < data.f | window | mcep > data.mcep
```

```
frame +f < data.f | window | ffttr -A -H | mcep -q 3 > data.mcep
```

### SEE ALSO

uels, gcep, mgcep, mlsadf

## NAME

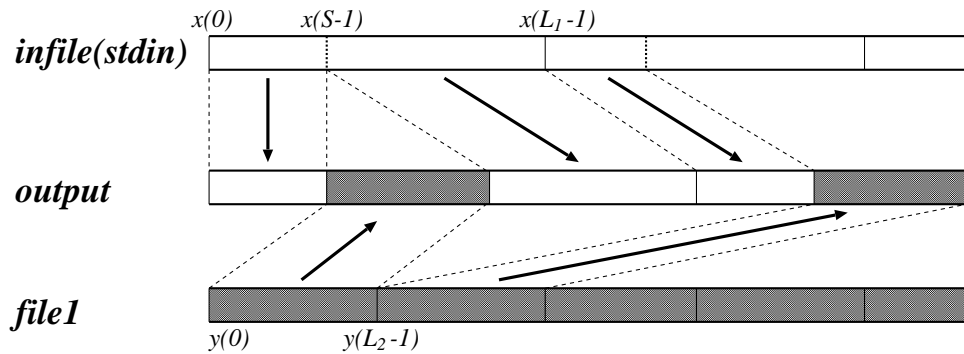
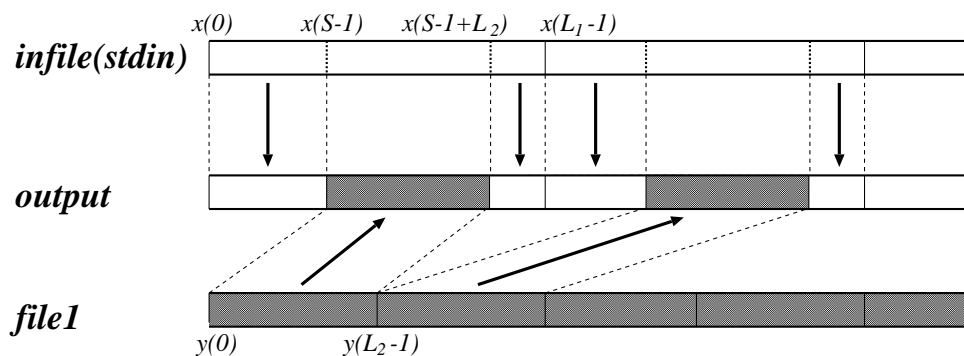
merge – data merge

## SYNOPSIS

```
merge [-s S ] [ -l L1 ] [ -n N1 ] [ -L L2 ] [ -N N2 ]
      [ -o ] [ +type ] file1 [ infile ]
```

## DESCRIPTION

*merge* merges, on a frame-by-frame basis, data from *file1* into the data from *infile* (or standard input), sending the result to standard output, as described below.

**Insert mode****Overwrite mode**

## OPTIONS

<b>-s</b>	$S$	insert point	[0]
<b>-l</b>	$L_1$	frame length of input data	[25]
<b>-n</b>	$N_1$	order of input data	[ $L_1 - 1$ ]
<b>-L</b>	$L_2$	frame length of insert data	[10]
<b>-N</b>	$N_2$	order of insert data	[ $L_2 - 1$ ]

<b>-o</b>	overwrite mode			[FALSE]
<b>+t</b>	input data format			[f]
	<b>c</b> char (1 byte)		<b>C</b> unsigned char (1 byte)	
	<b>s</b> short (2 bytes)		<b>S</b> unsigned short (2 bytes)	
	<b>i3</b> int (3 bytes)		<b>I3</b> unsigned int (3 bytes)	
	<b>i</b> int (4 bytes)		<b>I</b> unsigned int (4 bytes)	
	<b>l</b> long (4 bytes)		<b>L</b> unsigned long (4 bytes)	
	<b>le</b> long long (8 bytes)		<b>LE</b> unsigned long long (8 bytes)	
	<b>f</b> float (4 bytes)		<b>d</b> double (8 bytes)	

**EXAMPLE**

The following example inserts blocks of 2 samples from *data.f2* in short format into *data.f1*, also in short format, the frame length of the file *data.f1* is 3, and the blocks from *data.f2* will be inserted from the 3rd sample of every frame. The result is *data.merge*.

```
merge +f -s 2 -l 3 -L 2 +s data.f2 < data.f1 > data.merge
```

For example, if the *data.f1* file is

1, 1, 1, 2, 2, 2, ...

, and the *data.f2* file is

2, 3, 5, 6, ...

then the output *data.merge* will be

1, 1, 2, 3, 1, 2, 2, 5, 6, 2, ...

The next example overwrites blocks of 2 samples from *data.f2* in long format into *data.f1*, also in long format, the frame length of the file *data.f1* is 4, and the blocks from *data.f2* will be inserted from the 2nd sample of every frame. The result is *data.merge*.

```
merge +f -s 2 -l 4 -L 2 +l -o data.f2 < data.f1 > data.merge
```

For example, if the *data.f1* file is

1, 1, 1, 1, 2, 2, 2, 2, ...

, and the *data.f2* file is

3, 4, 5, 6, ...

then the output *data.merge* will be

1, 3, 4, 1, 2, 5, 6, 2, ...

**SEE ALSO**

bcp



**NAME**

`mgc2mgc` – frequency and generalized cepstral transformation

**SYNOPSIS**

```
mgc2mgc [ -m  $M_1$  ] [ -a  $A_1$  ] [ -g  $G_1$  ] [ -c  $C_1$  ] [ -n ] [ -u ]
[ -M  $M_2$  ] [ -A  $A_2$  ] [ -G  $G_2$  ] [ -C  $C_2$  ] [ -N ] [ -U ] [ infile ]
```

**DESCRIPTION**

`mgc2mgc` transforms mel-generalized cepstral coefficients  $c_{\alpha_1, \gamma_1}(0), \dots, c_{\alpha_1, \gamma_1}(M_1)$  from *infile* (or standard input) into a different set of mel-generalized cepstral coefficients  $c_{\alpha_2, \gamma_2}(0), \dots, c_{\alpha_2, \gamma_2}(M_2)$  sending the result to standard output.

$\alpha$  characterizes the frequency-warping transform, while  $\gamma$  characterizes the generalized log magnitude transform.

Input and output data are in float format.

Firstly, a frequency transformation ( $\alpha_1 \rightarrow \alpha_2$ ) is undertaken in the input mel-generalized cepstral coefficients  $c_{\alpha_1, \gamma_1}(m)$ , and  $c_{\alpha_2, \gamma_1}(m)$  is calculated as follows.

$$\alpha = (\alpha_2 - \alpha_1) / (1 - \alpha_1 \alpha_2)$$

$$c_{\alpha_2, \gamma_1}^{(i)}(m) = \begin{cases} c_{\alpha_1, \gamma_1}(-i) + \alpha c_{\alpha_2, \gamma_1}^{(i-1)}(0), & m = 0 \\ (1 - \alpha^2) c_{\alpha_2, \gamma_1}^{(i-1)}(0) + \alpha c_{\alpha_2, \gamma_1}^{(i-1)}(1), & m = 1 \\ c_{\alpha_2, \gamma_1}^{(i-1)}(m-1) + \alpha (c_{\alpha_2, \gamma_1}^{(i-1)}(m) - c_{\alpha_2, \gamma_1}^{(i-1)}(m-1)), & m = 2, \dots, M_2 \end{cases}$$

$$i = -M_1, \dots, -1, 0$$

Then the gain is normalized and  $c'_{\alpha_2, \gamma_1}(m)$  is evaluated.

$$K_{\alpha_2} = s_{\gamma_1}^{-1} (c_{\alpha_2, \gamma_1}^{(0)}(0)),$$

$$c'_{\alpha_2, \gamma_1}(m) = c_{\alpha_2, \gamma_1}^{(0)}(m) / (1 + \gamma_1 c_{\alpha_2, \gamma_1}^{(0)}(0)), \quad m = 1, 2, \dots, M_2$$

Afterwards,  $c'_{\alpha_2, \gamma_1}(m)$  is transformed into  $c'_{\alpha_2, \gamma_2}(m)$  through a generalized log transformation ( $\gamma_1 \rightarrow \gamma_2$ ).

$$c'_{\alpha_2, \gamma_2}(m) = c'_{\alpha_2, \gamma_1}(m) + \sum_{k=1}^{m-1} \frac{k}{m} \{ \gamma_2 c_{\alpha_2, \gamma_1}(k) c'_{\alpha_2, \gamma_2}(m-k) - \gamma_1 c_{\alpha_2, \gamma_2}(k) c'_{\alpha_2, \gamma_1}(m-k) \},$$

$$m = 1, 2, \dots, M_2$$

Finally, the gain is inverse normalized and  $c_{\alpha_2, \gamma_2}(m)$  is calculated.

$$c_{\alpha_2, \gamma_2}(0) = s_{\gamma_2} (K_{\alpha_2}),$$

$$c_{\alpha_2, \gamma_2}(m) = c'_{\alpha_2, \gamma_2}(m) (1 + \gamma_2 c_{\alpha_2, \gamma_2}(0)), \quad m = 1, 2, \dots, M_2$$

In case we represent input and output with  $\gamma$ , if the coefficients  $c_{\alpha,\gamma}(m)$  are not normalized, then the following representation is assumed

$$1 + \gamma c_{\alpha,\gamma}(0), \gamma c_{\alpha,\gamma}(1), \dots, \gamma c_{\alpha,\gamma}(M),$$

if they are normalized, then the following representation is assumed

$$K_{\alpha}, \gamma c'_{\alpha,\gamma}(1), \dots, \gamma c'_{\alpha,\gamma}(M).$$

## OPTIONS

<b>-m</b>	$M_1$	order of mel-generalized cepstrum (input)	[25]
<b>-a</b>	$A_1$	alpha of mel-generalized cepstrum (input)	[0]
<b>-g</b>	$G_1$	gamma of mel-generalized cepstrum (input)	[0]
		$\gamma_1 = G_1$	
<b>-c</b>	$C_1$	gamma of mel-generalized cepstrum (input)	
		$\gamma_1 = -1/(\text{int})C_1$	
		$C_1$ must be $C_1 \geq 1$	
<b>-n</b>		regard input as normalized mel-generalized cepstrum	[FALSE]
<b>-u</b>		regard input as multiplied by gamma	[FALSE]
<b>-M</b>	$M_2$	order of mel-generalized cepstrum (output)	[25]
<b>-A</b>	$A_2$	alpha of mel-generalized cepstrum (output)	[0]
<b>-G</b>	$G_2$	gamma of mel-generalized cepstrum (output)	[1]
		$\gamma_2 = G_2$	
<b>-C</b>	$C_2$	gamma of mel-generalized cepstrum (output)	
		$\gamma_2 = -1/(\text{int})G_2$	
		$C_2$ must be $C_2 \geq 1$	
<b>-N</b>		regard output as normalized mel-generalized cepstrum	[FALSE]
<b>-U</b>		regard input as multiplied by gamma	[FALSE]

## EXAMPLE

In the example below, 12-th order LPC coefficients are read in float format from *data.lpc*, 30-th order mel-cepstral coefficients are calculated and written to *data.mcep*:

```
mgc2mgc -m 12 -a 0 -g -1 -M 30 -A 0.31 -G 0
        < data.lpc > data.mcep
```

## SEE ALSO

uels, gcep, mcep, mgcep, gc2gc, freqt, lpc2c

**NAME**

`mgc2sp` – transform mel-generalized cepstrum to spectrum

**SYNOPSIS**

```
mgc2sp [-a A] [-g G] [-c C] [-m M] [-n] [-u] [-l L] [-p]
        [-o O] [infile]
```

**DESCRIPTION**

`mgc2sp` calculates the log magnitude spectrum from mel-generalized cepstral coefficients  $c_{\alpha,\gamma}(m)$  from *infile* (or standard input), sending the result to standard output.

Input and output data are in float format.

The mel-generalized cepstral coefficients  $c_{\alpha,\gamma}(m)$  are transformed into cepstral coefficients (refer to `mgc2mgc`) and then the log magnitude spectrum is calculated (refer to `spec`).

When the input data is normalized by the gain, then it can be represented as follows.

$$K_{\alpha} = s_{\gamma}^{-1} \left( c_{\alpha,\gamma}^{(0)}(0) \right),$$

$$c'_{\alpha,\gamma}(m) = c_{\alpha,\gamma}^{(0)}(m) / \left( 1 + \gamma c_{\alpha,\gamma}^{(0)}(0) \right), \quad m = 1, 2, \dots, M$$

In case we represent input with  $\gamma$ , if the coefficients  $c_{\alpha,\gamma}(m)$  are not normalized, then the following representation is assumed

$$1 + \gamma c_{\alpha,\gamma}(0), \gamma c_{\alpha,\gamma}(1), \dots, \gamma c_{\alpha,\gamma}(M)$$

if they are normalized, then the following representation is assumed

$$K_{\alpha}, \gamma c'_{\alpha,\gamma}(1), \dots, \gamma c'_{\alpha,\gamma}(M)$$

**OPTIONS**

<b>-a</b>	<i>A</i>	alpha $\alpha$	[0]
<b>-g</b>	<i>G</i>	power parameter $\gamma$ of mel-generalized cepstrum $\gamma = G$	[0]
<b>-c</b>	<i>C</i>	power parameter $\gamma$ of mel-generalized cepstrum $\gamma = -1/(\text{int})C$ <i>C</i> must be $C \geq 1$	
<b>-m</b>	<i>M</i>	order of mel-generalized cepstrum	[25]
<b>-n</b>		regard input as normalized cepstrum	[FALSE]
<b>-u</b>		regard input as multiplied by $\gamma$	[FALSE]
<b>-l</b>	<i>L</i>	FFT length	[256]

**-p** output phase [FALSE]  
**-o** *O* output format [0]

if the **-p** option is assigned, scale of output spectrum can be assigned.

$$O = 0 \quad 20 \times \log |H(z)|$$

$$O = 1 \quad \ln |H(z)|$$

$$O = 2 \quad |H(z)|$$

$$O = 3 \quad |H(z)|^2$$

if the **-p** option is not assigned, unit of output phase can be assigned.

$$O = 0 \quad \arg |H(z)| \div \pi \quad [\pi \text{ rad.}]$$

$$O = 1 \quad \arg |H(z)| \quad [\text{rad.}]$$

$$O = 2 \quad \arg |H(z)| \times 180 \div \pi \quad [\text{deg.}]$$

### EXAMPLE

In the following example, mel-generalized cepstral coefficients in float format are read from *data.mgcep* ( $M = 12, \alpha = 0.35, \gamma = -0.5$ ) and the log magnitude spectrum is evaluated and plotted:

```
mgc2sp -m 12 -a 0.35 -c 2 < data.mgcep | glogsp | xgr
```

### SEE ALSO

c2sp, mgc2mgc, gc2gc, freqt, gnorm, lpc2c

**NAME**

`mgcep` – mel-generalized cepstral analysis(13; 14)

**SYNOPSIS**

`mgcep` [-a A] [-g G] [-c C] [-m M] [-l L] [-q Q] [-o O]  
 [-i I] [-j J] [-d D] [-p P] [-e E] [-f F] [*infile*]

**DESCRIPTION**

`mgcep` uses mel-generalized cepstral analysis to calculate mel-generalized cepstral coefficients from  $L$ -length framed windowed input data from *infile* (or standard input), sending the result to standard output. There are several different output formats, controlled by the `-o` option.

When input signal has length  $L$ , then the time sequence is given by

$$x(0), x(1), \dots, x(L-1)$$

Input and output data are in float format.

In the mel-generalized cepstral analysis, the spectrum of the speech signal is modeled by  $M$ -th order mel-generalized cepstral coefficients  $c_{\alpha,\gamma}(m)$  as follows.

$$H(z) = s_{\gamma}^{-1} \left( \sum_{m=0}^M c_{\alpha,\gamma}(m) z^{-m} \right)$$

$$= \begin{cases} \left( 1 + \gamma \sum_{m=1}^M c_{\alpha,\gamma}(m) \tilde{z}^{-m} \right)^{1/\gamma}, & -1 \leq \gamma < 0 \\ \exp \sum_{m=1}^M c_{\alpha,\gamma}(m) \tilde{z}^{-m}, & \gamma = 0 \end{cases}$$

For this command “`mgcep`”, it is applied a cost function based on the unbiased estimation log spectrum method. The variable  $\tilde{z}^{-1}$  can be expressed as the following first order all-pass function

$$\tilde{z}^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}$$

The phase characteristic is given by the variable  $\alpha$ . For a sampling rate 10kHz,  $\alpha$  is made equal to 0.35. For a sampling rate 8kHz,  $\alpha$  is made equal to 0.31. By making these choices for  $\alpha$ , the mel-scale becomes the good approximation to human sensitivity to the loudness speech sound.

The Newton-Raphson method is used to minimize the cost function when evaluating mel-cepstral coefficients.

The mel-generalized cepstral analysis includes several other methods to analyze speech, depending on the values of  $\alpha$  and  $\gamma$  (refer to figure 1).

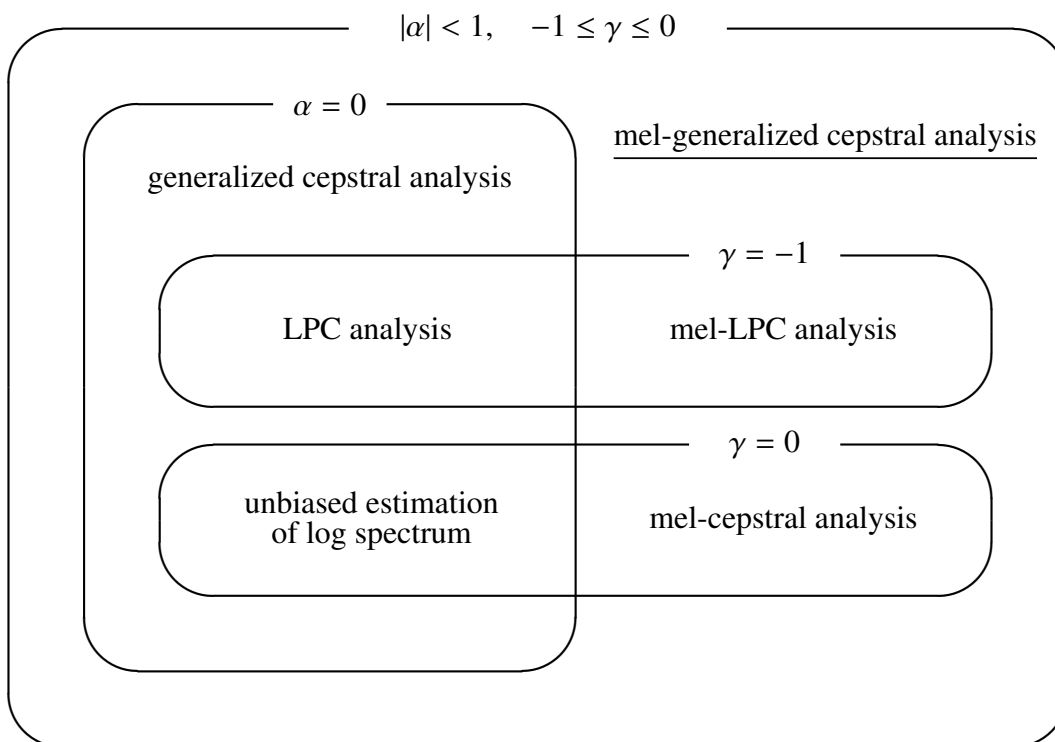


Figure 1: mel-generalized cepstral analysis and other method relations

## OPTIONS

<b>-a</b>	$A$	alpha $\alpha$	[0.35]
<b>-g</b>	$G$	power parameter of generalized cepstrum $\gamma$ $\gamma = G$	[0]
<b>-c</b>	$C$	power parameter of generalized cepstrum $\gamma$ $\gamma = -1/(\text{int})C$ $C$ must be $C \geq 1$	
<b>-m</b>	$M$	order of mel-generalized cepstrum	[25]
<b>-l</b>	$L$	frame length power of 2	[256]
<b>-q</b>	$Q$	input data style	[0]
		$Q = 0$ windowed data sequence	
		$Q = 1$ $20 \times \log  f(w) $	
		$Q = 2$ $\ln  f(w) $	
		$Q = 3$ $ f(w) $	
		$Q = 4$ $ f(w) ^2$	

<b>-o</b>	<i>O</i>	output format	[0]
	$O = 0$	$c_{\alpha,\gamma}(0), c_{\alpha,\gamma}(1), \dots, c_{\alpha,\gamma}(M)$	
	$O = 1$	$b_{\gamma}(0), b_{\gamma}(1), \dots, b_{\gamma}(M)$	
	$O = 2$	$K_{\alpha}, c'_{\alpha,\gamma}(1), \dots, c'_{\alpha,\gamma}(M)$	
	$O = 3$	$K, b'_{\gamma}(1), \dots, b'_{\gamma}(M)$	
	$O = 4$	$K_{\alpha}, \gamma c'_{\alpha,\gamma}(1), \dots, \gamma c'_{\alpha,\gamma}(M)$	
	$O = 5$	$K, \gamma b'_{\gamma}(1), \dots, \gamma b'_{\gamma}(M)$	

Usually, the options below do not need to be assigned.

<b>-i</b>	<i>I</i>	minimum iteration of Newton-Raphson method	[2]
<b>-j</b>	<i>J</i>	maximum iteration of Newton-Raphson method	[30]
<b>-d</b>	<i>D</i>	end condition of Newton-Raphson method	[0.001]
<b>-p</b>	<i>P</i>	order of recursions	[ <i>L</i> - 1]
<b>-e</b>	<i>E</i>	small value added to periodgram	[0]
<b>-f</b>	<i>F</i>	mimimum value of the determinant of the normal matrix	[0.000001]

### EXAMPLE

In the following speech data in float format is read from *data.f* and analyzed with  $\gamma = 0$ ,  $\alpha = 0$  (which correspond to UELS method for log spectrum estimation) and the resulting cepstral coefficients are written *data.cep*:

```
frame +f < data.f | window | mgcep > data.cep
```

In the same way if we want mel-cepstral coefficients:

```
frame +f < data.f | window | mgcep -a 0.35 > data.mcep
```

If we want linear prediction coefficients:

```
frame +f < data.f | window | mgcep -g -1 -o 5 > data.lpc
```

In this case the linear prediction coefficients are written in the following representation.

$$K, a(1), a(2), \dots, a(M)$$

In the following, speech data in float format is read from *data.f*, and analyzed with  $\gamma = 0$ ,  $\alpha = 0$  (which correspond to UELS method for log spectrum estimation) and the resulting cepstral coefficients are written *data.cep*:

```
frame +f < data.f | window | \  
fftr -A -H | mgcep -q 3 > data.cep
```

### SEE ALSO

uels, gcep, mcep, freqt, gc2gc, mgc2mgc, gnorm, mglsadf

**NAME**

`mglسادف` – MGLSA digital filter for speech synthesis(21; 22)

**SYNOPSIS**

`mglسادف` [ `-m M` ] [ `-a A` ] [ `-c C` ] [ `-p P` ] [ `-i I` ] [ `-v` ] [ `-t` ] [ `-k` ] [ `-P Pa` ]  
`mgcfile` [ `infile` ]

**DESCRIPTION**

`mglسادف` derives a Mel-Generalized Log Spectral Approximation digital filter from mel-generalized cepstral coefficients  $c_{\alpha,\gamma}(m)$  in `mgcfile` and uses it to filter an excitation sequence from `infile` (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

The transfer function  $H(z)$  related to the synthesis filter is obtained from the  $M$ -th order mel-generalized cepstral coefficients  $c_{\alpha,\gamma}(m)$  as follows.

$$H(z) = s_{\gamma}^{-1} \left( \sum_{m=0}^M c_{\alpha,\gamma}(m) \tilde{z}^{-m} \right) \quad (1)$$

$$= \begin{cases} \left( 1 + \gamma \sum_{m=0}^M c_{\alpha,\gamma}(m) \tilde{z}^{-m} \right)^{1/\gamma}, & 0 < \gamma \leq -1 \\ \exp \sum_{m=0}^M c_{\alpha,\gamma}(m) \tilde{z}^{-m}, & \gamma = 0 \end{cases}$$

where

$$\tilde{z}^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}$$

The transfer function  $H(z)$  can be rewritten as

$$H(z) = s_{\gamma}^{-1} \left( \sum_{m=0}^M b'_{\gamma}(m) \Phi_m(z) \right) = K \cdot D(z) \quad (2)$$

where

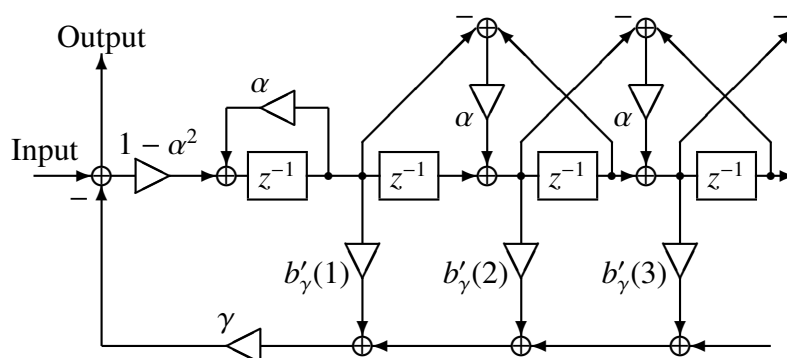
$$\Phi_m(z) = \begin{cases} 1, & m = 0 \\ \frac{(1 - \alpha^2)z^{-1}}{1 - \alpha z^{-1}} \tilde{z}^{-(m-1)}, & m \geq 1 \end{cases}$$

and

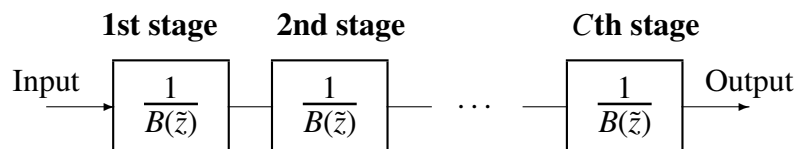
$$K = s_{\gamma}^{-1}(b_{\gamma}(0))$$

$$D(z) = s_{\gamma}^{-1} \left( \sum_{m=1}^M b_{\gamma}(m) \Phi_m(z) \right)$$





(a) Structure of filter  $1/B(z)$



(b)  $C$  level cascaded filter  $1/B(z)$

Figure 1: Realization synthesis filter  $D(z)$

Also, the coefficients  $b'_\gamma(m)$  are obtained from the coefficients  $c_{\alpha,\gamma}(m)$  by normalizing (refer to gnorm), and applying linear transformation (refer to mc2b and b2mc). Here we are considering only cases where power parameter is  $\gamma = -1/C$  ( $C$ :natural number). In this case the filter  $D(z)$  is realized as shown in figure (b), where each filter of the  $C$  level cascaded filter is realized as shown in figure (a), and can be expressed as

$$\frac{1}{B(\tilde{z})} = \frac{1}{1 + \gamma \sum_{m=1}^M b'_\gamma(m) \Phi_m(z)}$$

**OPTIONS**

<b>-m</b>	<i>M</i>	order of mel-generalized cepstrum	[25]
<b>-a</b>	<i>A</i>	alpha	[0.35]
<b>-c</b>	<i>C</i>	power parameter $\gamma = -1/C$ of generalized cepstrum if $C == 0$ , the MLSA filter is used	[1]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-v</b>		inverse filter	[FALSE]
<b>-t</b>		transpose filter	[FALSE]
<b>-k</b>		filtering without gain	[FALSE]

The option below only works if  $C == 0$ .

<b>-P</b>	<i>Pa</i>	order of the Padé approximation <i>Pa</i> should be 4 or 5	[4]
-----------	-----------	---	-----

**EXAMPLE**

In the following example, the excitation is constructed from pitch data read in float format from *data.pitch*, passed through an MGLSA filter built from the mel-generalized cepstrum in *data.mgcep*, and the synthesized speech is written to *data.syn*:

```
excite < data.pitch | mglسادf data.mgcep > data.syn
```

**SEE ALSO**

mgcep, poledf, zerodf, ltcdf, lmadf, mlsadf, glسادf

**NAME**

`minmax` – find minimum and maximum values

**SYNOPSIS**

`minmax` [ `-l L` ] [ `-n N` ] [ `-b B` ] [ `-d` ] [ *infile* ]

**DESCRIPTION**

*minmax* determines the *B* (default 1) minimum and maximum values, on a frame-by-frame basis, of the data from *infile* (or standard input), sending the result to standard output.

If the frame length *L* is 1, each input number is considered to be both the minimum and maximum value for its length-1 frame.

The input format is float. If the `-d` option is not given, the output format is float, consisting of the minimum and maximum values. If the `-d` option is give, the output format is ASCII, showing the positions within the frame where the minimum and maximum values occurred, as follows:

*value* : *position*<sub>0</sub>, *position*<sub>1</sub>, . . .

**OPTIONS**

<code>-l</code>	<i>L</i>	length of vector	[1]
<code>-n</code>	<i>N</i>	order of vector	[L-1]
<code>-b</code>	<i>B</i>	find n-best values	[1]
<code>-d</code>		output data number	[FALSE]

**EXAMPLE**

If, for example, the input data in *data.f* in float format is as follows

1, 1, 2, 3, 4, 5, 6, 7, 8, 9, 9, 10

, then the output of the following command

```
minmax data.f -l 6 > data.m
```

is written to *data.m* as follows.

1, 5, 6, 10

Also if the following command is applied

```
minmax -n 2 -d data.f
```

then the result is

1:0  
2:2  
3:0  
5:2  
6:0  
8:2  
9:0,1  
10:2

**NAME**

`mlpg` – obtain parameter sequence from PDF sequence(23)

**SYNOPSIS**

```
mlpg [-l L] [-m M] [-d (fn | d0 [d1 ...])] [-r NR W1 [W2]]
      [-i I] [-s S] [infile]
```

**DESCRIPTION**

`mlpg` calculates the maximum likelihood parameters from the means and diagonal covariances of Gaussian distributions from *infile* (or standard input), sending the result to standard output. The input format is

$$\dots, \mu_t(0), \dots, \mu_t(M), \mu_t^{(1)}(0), \dots, \mu_t^{(1)}(M), \dots, \mu_t^{(N)}(M), \\ \sigma_t^2(0), \dots, \sigma_t^2(M), \sigma_t^{(1)2}(0), \dots, \sigma_t^{(1)2}(M), \dots, \sigma_t^{(N)2}(M), \dots$$

Input and output data are in float format.

The speech parameter vector  $\mathbf{o}_t$  for every frame  $t$  is composed of the static feature vector  $\mathbf{c}_t$ , where

$$\mathbf{c}_t = [c_t(0), c_t(1), \dots, c_t(M)]^\top$$

and the dynamic feature vector  $\Delta^{(1)}\mathbf{c}_t, \dots, \Delta^{(N)}\mathbf{c}_t$ , that is,

$$\mathbf{o}_t = [\mathbf{c}'_t, \Delta^{(1)}\mathbf{c}'_t, \dots, \Delta^{(N)}\mathbf{c}'_t]^\top.$$

The dynamic feature vector  $\Delta^{(n)}\mathbf{c}_t$  is obtained from the static feature vector as follows.

$$\Delta^{(n)}\mathbf{c}_t = \sum_{\tau=-L^{(n)}}^{L^{(n)}} w^{(n)}(\tau)\mathbf{c}_{t+\tau}$$

where  $n$  is the order of dynamic feature vector, for example, when we evaluate  $\Delta^2$  parameter,  $n=2$ . The `mlpg` command reads probability density functions sequence

$$((\boldsymbol{\mu}_1, \boldsymbol{\Sigma}_1), (\boldsymbol{\mu}_2, \boldsymbol{\Sigma}_2), \dots, (\boldsymbol{\mu}_T, \boldsymbol{\Sigma}_T)),$$

where

$$\boldsymbol{\mu}_t = [\boldsymbol{\mu}'_t(0), \boldsymbol{\mu}'_t(1), \dots, \boldsymbol{\mu}'_t(N)]^\top \\ \boldsymbol{\Sigma}_t = \text{diag} [\boldsymbol{\Sigma}'_t(0), \boldsymbol{\Sigma}'_t(1), \dots, \boldsymbol{\Sigma}'_t(N)]$$

and evaluates the maximum likelihood parameter sequence  $(\mathbf{o}_1, \mathbf{o}_2, \dots, \mathbf{o}_T)$ , and sends the static feature vector sequence  $\mathbf{c}_t$ , that is  $(\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_T)$ , to the output. In the example above,  $\boldsymbol{\mu}^{(0)}, \boldsymbol{\Sigma}^{(0)}$  represent the static feature vector mean and covariance matrix, respectively, and  $\boldsymbol{\mu}^{(n)}, \boldsymbol{\Sigma}^{(n)}$  represent the  $n$ -th order dynamic feature vector mean and covariance matrix, respectively.

## OPTIONS

<b>-m</b>	$M$	order of vector	[25]
<b>-l</b>	$L$	length of vector	$[M + 1]$
<b>-d</b>	$(fn   d_0 [d_1 \dots])$	$fn$ is the file name of the parameters $w^{(n)}(\tau)$ used when evaluating the dynamic feature vector. It is assume that the number of coefficients to the left and to the right have the same length, if this is not true than zeros are added to the short side. For example, if the coefficients are	[N/A]

$$w(-1), w(0), w(1), w(2), w(3)$$

then zeros are added to the left as follows.

$$0, 0, w(-1), w(0), w(1), w(2), w(3)$$

Instead of entering the filename  $fn$ , the coefficients(which compose the file  $fn$ ) can be directly input in the command line. When the order of the dynamic feature vector is higher then one, then the sets of coefficients can be input one after the other as shown on the last example below. this option can not be used with the  $-r$  option.

<b>-r</b>	$N_R W_1 [W_2]$	This option is used when $N_R$ -th order dynamic parameters are used and the weighting coefficients $w^{(n)}(\tau)$ are evaluated by regression. $N_R$ can be made equal to 1 or 2. The variables $W_1$ and $W_2$ represent the widths of the first and second order regression coefficients, respectively. The first order regression coefficients for $\Delta \mathbf{c}_t$ at frame $t$ are evaluated as follows.	[N/A]
-----------	-----------------	--	-------

$$\Delta \mathbf{c}_t = \frac{\sum_{\tau=-W_1}^{W_1} \tau \mathbf{c}_{t+\tau}}{\sum_{\tau=-W_1}^{W_1} \tau^2}$$

For the second order regression coefficients,  $a_2 = \sum_{\tau=-W_2}^{W_2} \tau^4$ ,  $a_1 = \sum_{\tau=-W_2}^{W_2} \tau^2$ ,  $a_0 = \sum_{\tau=-W_2}^{W_2} 1$  and

$$\Delta^2 \mathbf{c}_t = \frac{\sum_{\tau=-W_2}^{W_2} (a_0 \tau^2 - a_1) \mathbf{c}_{t+\tau}}{2(a_2 a_0 - a_1^2)}$$

this option can not be used with the  $-d$  option.

**-i**  $I$  type of input PDFs [0]

$$\begin{aligned} I = 0 & \quad \mu, \quad \Sigma \\ I = 1 & \quad \mu, \quad \Sigma^{-1} \\ I = 2 & \quad \mu\Sigma^{-1}, \quad \Sigma^{-1} \end{aligned}$$

**-s**  $S$  range of influenced frames [30]

### EXAMPLE

In the example below, the number of parameters is 15, the width of the window for first or second order dynamic feature evaluation is 1, and the parameter sequence is evaluated from the probability density function.

```
mlpg -m 15 -r 2 1 1 data.pdf > data.par
```

or

```
echo "-0.5 0 0.5" | x2x +af > delta
echo "0.25 -0.5 0.25" | x2x +af > accel
mlpg -m 15 -d delta -d accel data.pdf > data.par
```

**NAME**

mlsadf – MLSA digital filter for speech synthesis(19; 20; 12)

**SYNOPSIS**

**mlsadf** [ -m *M* ] [ -a *A* ] [ -p *P* ] [ -i *I* ] [ -b ] [ -P *Pa* ] [ -v ] [ -t ] [ -k ]  
*mcfile* [ *infile* ]

**DESCRIPTION**

*mlsadf* derives a Mel Log Spectral Approximation digital filter from mel-cepstral coefficients  $c_\alpha(0), c_\alpha(1), \dots, c_\alpha(M)$  in *mcfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

The exponential transfer function  $H(z)$  related to the MLSA synthesis filter is obtained from the  $M$  order mel-cepstral coefficients  $c_\alpha(m)$  as follows.

$$H(z) = \exp \sum_{m=0}^M c_\alpha(m) \tilde{z}^{-m}$$

where

$$\tilde{z}^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}.$$

The highly accurate approximation method of the above transfer function is explained in the follow. Firstly, the transfer function  $H(z)$  is rewritten as

$$\begin{aligned} H(z) &= \exp \sum_{m=0}^M b(m) \Phi_m(z) \\ &= K \cdot D(z) \end{aligned}$$

where,

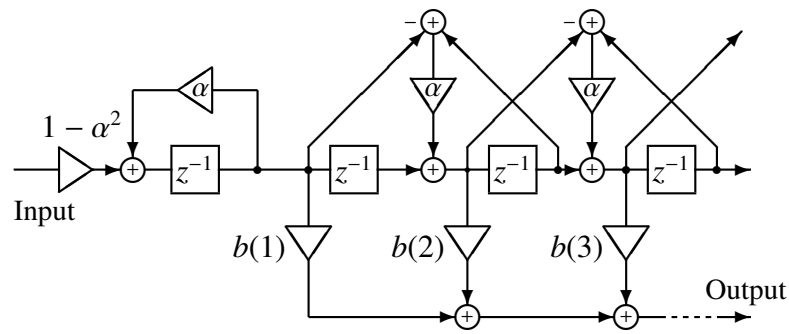
$$\Phi_m(z) = \begin{cases} 1, & m = 0 \\ \frac{(1 - \alpha^2)z^{-1}}{1 - \alpha z^{-1}} \tilde{z}^{-(m-1)}, & m \geq 1 \end{cases}$$

and

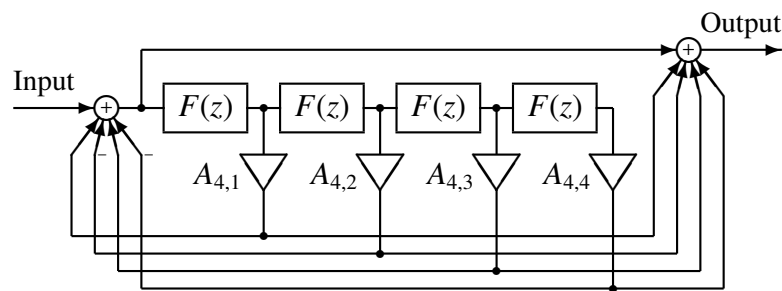
$$\begin{aligned} K &= \exp b(0) \\ D(z) &= \exp \sum_{m=1}^M b(m) \Phi_m(z) \end{aligned}$$

Also, the coefficients  $b(m)$  can be obtained through a linear transformation of  $c_\alpha(m)$  (refer to mc2b and b2mc).

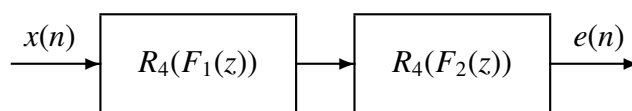




(a) Basic filter  $F(z)$



(b)  $R_L(F(z)) \approx D(z) \quad L = 4$



(c) Two-stage cascade structure  
 $R_4(F_1(z)) \cdot R_4(F_2(z)) \approx D(z)$

Figure 1: Realization of exponential transfer function  $1/D(z)$

The filter  $D(z)$  can be realized as shown in figure 1(b), where basic filter (figure 1(a)) is the following IIR filter.

$$F(z) = \sum_{m=1}^M b(m)\Phi_m(z)$$

If we want to improve the accuracy of the approximation, we can decompose the basic filter as shown in figure 1(c),

$$F(z) = F_1(z) + F_2(z)$$

where

$$F_1(z) = b(1)z^{-1}$$

$$F_2(z) = \sum_{m=2}^M b(m)\Phi_m(z)$$

Also, the coefficients  $A_{4,l}$  in figure 1(b) have same value as the LMA filter (refer to `lmadf`).

## OPTIONS

<b>-m</b>	<i>M</i>	order of mel-cepstrum	[25]
<b>-a</b>	<i>A</i>	all-pass constant $\alpha$	[0.35]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-b</b>		output filter coefficient $b(m)$ (coefficients which are linear transformed from mel-cepstrum)	[FALSE]
<b>-P</b>	<i>Pa</i>	order of the Padé approximation <i>Pa</i> should be 4 or 5	[4]
<b>-k</b>		filtering without gain	[FALSE]
<b>-v</b>		inverse filter	[FALSE]
<b>-t</b>		transpose filter	[FALSE]

## EXAMPLE

In the following example, the excitation is constructed from pitch data read in float format from `data.pitch`, passed through an MLSA filter built from the mel-cepstrum in `data.mcep`, and the synthesized speech is written to `data.syn`:

```
excite < data.pitch | mlsadf data.mcep > data.syn
```

## SEE ALSO

`mcep`, `amcep`, `poledf`, `zerodf`, `ltdcf`, `lmadf`, `glsadf`, `mgladf`

**NAME**

`msvq` – multi stage vector quantization

**SYNOPSIS**

```
msvq [ -l L ] [ -n N ] [ -s S cbfile ] [ -q ] [ infile ]
```

**DESCRIPTION**

`msvq` encodes the data from *infile* (or standard input) using multi-stage vector quantization with codebooks specified by multiple `-s` options, sending the result to standard output.

Input data is in float format and output data is in int format.

**OPTIONS**

<code>-l</code>	<i>L</i>	length of vector	[26]
<code>-n</code>	<i>N</i>	order of vector	[ <i>L</i> - 1]
<code>-s</code>	<i>S</i> <i>cbfile</i>	codebook	[N/A N/A]
	<i>S</i>	codebook size	
	<i>cbfile</i>	codebook file	
<code>-q</code>		output quantized vector	[FALSE]

**EXAMPLE**

In the example below, a two level vq is undertaken in input *data.f* file. the codebook sizes of *cbfile1* and *cbfile2* are 256 and the output is written to *data.vq*:

```
msvq -s 256 cbfile1 -s 256 cbfile2 < data.f > data.vq
```

**SEE ALSO**

`imsvq`, `vq`, `ivq`, `lbg`

**NAME**

`nan` – data check

**SYNOPSIS**

`nan [ infile ]`

**DESCRIPTION**

*nan* checks whether input data contains NaN (Not a Number) or Infinity, showing the positions where these values occurred.

**EXAMPLE**

This example reads input data *data.f* in float format and checks it:

```
nan data.f
```

**NAME**

norm0 – normalize coefficients

**SYNOPSIS**

**norm0** [ **-m** *M* ] [ *infile* ]

**DESCRIPTION**

*norm0* normalizes vectors from *infile* (or standard input) by dividing vector components by the zero-order component, sending the result to standard output.

For the input sequence

$$x(0), x(1), \dots, x(M),$$

the normalized output sequence is

$$1/x(0), x(1)/x(0), \dots, x(M)/x(0).$$

Input and output data are in float format.

**OPTIONS**

**-m** *M* order of input data [25]

**EXAMPLE**

Speech data is read from *data.f* in float format, the 15-th order autocorrelation coefficients are evaluated and normalized, and the results is written to *data.nacorr*:

```
frame +f < data.f | window | acorr -m 15 |\
norm0 -m 15 > data.nacorr
```

**SEE ALSO**

linear\_intpl

**NAME**

`nrnd` – generate normal distributed random value

**SYNOPSIS**

`nrnd` [ **-l** *L* ] [ **-s** *S* ] [ **-m** *M* ] [ **-v** *V* ] [ **-d** *D* ]

**DESCRIPTION**

*nrnd* generates a sequence of normally-distributed random values, sending the result to standard output.

Output data is in float format.

**OPTIONS**

<b>-l</b>	<i>L</i>	output length	[256]
		In the case $L \leq 0$ then random values will be generated indefinitely.	
<b>-s</b>	<i>S</i>	seed for nrnd	[1]
<b>-m</b>	<i>M</i>	mean of normal distribution	[0.0]
<b>-v</b>	<i>V</i>	variance of normal distribution	[1.0]
<b>-d</b>	<i>D</i>	standard deviation of normal distribution	[1.0]

**EXAMPLE**

Normal distributed random values of length 100 are generated and written to *data.rnd*:

```
nrnd -l 100 -s 3 > data.rnd
```

**NAME**

`par2lpc` – transform PARCOR to LPC

**SYNOPSIS**

`par2lpc` [ `-m M` ] [ *infile* ]

**DESCRIPTION**

`par2lpc` calculates linear prediction (LPC) coefficients from  $M$ -th order PARCOR coefficients from *infile* (or standard input), sending the result to standard output.

The PARCOR input format is

$$K, k(1), \dots, k(M),$$

and the LPC output format is

$$K, a(1), \dots, a(M).$$

Input and output data are in float format.

The transformation of PARCOR coefficients into linear prediction coefficients is undertaken by a part of Durbin algorithm as follows.

$$\begin{aligned} a^{(m)}(m) &= k(m) \\ a^{(m)}(i) &= a^{(m-1)}(i) + k(m)a^{(m-1)}(m-i), \quad 1 \leq i \leq m \end{aligned}$$

where  $m = 1, 2, \dots, p$ . The initial condition is

$$a^{(M)}(m) = a(m), \quad 1 \leq m \leq M.$$

**OPTIONS**

`-m M` order of LPC [25]

**EXAMPLE**

PARCOR coefficients are read in float format from *data.rc* and converted into the corresponding linear prediction coefficients. The output is written to *data.lpc*:

```
par2lpc < data.rc > data.lpc
```

**SEE ALSO**

`acorr`, `levdur`, `lpc`, `lpc2par`

**NAME**

*pca* – principal component analysis

**SYNOPSIS**

**pca** [ **-l** *L* ] [ **-n** *N* ] [ **-i** *I* ] [ **-e** *e* ] [ **-v** ] [ **-V** *fn* ] [ *infile* ]

**DESCRIPTION**

*pca* carries out principal component analysis from *infile* (or standard input) with jacobi method, sending the result to standard output. *pca* can also calculate contribution ratio with the eigen values.

In *infile*, the input training data set consists of *L*-dimension vectors:

$$\mathbf{x}(0), \mathbf{x}(1), \mathbf{x}(2), \mathbf{x}(3), \dots \quad \text{where } \mathbf{x}(i) = (x_i(1), x_i(2), \dots, x_i(L))$$

Input and output data are in float format.

**OPTIONS**

<b>-l</b>	<i>L</i>	dimensionality of vector	[3]
<b>-n</b>	<i>N</i>	number of output principal components	[2]
<b>-i</b>	<i>I</i>	limit of iteration on jacobi method	[10000]
<b>-e</b>	<i>e</i>	threshold of convergence on jacobi method	[0.000001]
<b>-v</b>		output eigen vectors and mean vector of the training data	[FALSE]
<b>-V</b>	<i>fn</i>	output eigen values and contribution rate (output filename = <i>fn</i> )	[FALSE]

**EXAMPLE**

In the example below, the eigen vectors and the eigen values are calculated from *data.f* which contains three-dimensional training vectors. The mean vectors and eigen vectors are sent to *pca.dat*, and the eigen values are sent to *eigen.dat*.

```
pca data.f -n 2 -l 3 -v -V eigen.dat > pca.dat
```

In the *pca.dat*, the mean vector is in the front of eigen vectors. In the *eigen.dat*, the eigen values and their contribution ratio are bound per the same principal component and ordered according to the magnitude of the eigen values.

**SEE ALSO**

*pcas*



**NAME**

`pcas` – calculate principal component scores

**SYNOPSIS**

`pcas` [ **-l** *L* ] [ **-n** *N* ] *pcafile* [ *infile* ]

**DESCRIPTION**

`pcas` calculates principal component scores from *infile* (or standard input) , sending the result to standard output.

The input data set must be composed of *L*-dimension, mean vector *m* and eigen vectors *e*(*i*):

$$\mathbf{m}, \mathbf{e}(0), \mathbf{e}(1), \mathbf{e}(2), \dots$$

$$\text{where } \mathbf{m} = (m(1), m(2), \dots, m(L)) \text{ and } \mathbf{e}(i) = (e_i(1), e_i(2), \dots, e_i(L))$$

Input and output data are in float format.

**OPTIONS**

**-l** *L* dimensionality of vector [3]  
**-n** *N* number principal components for output [2]

**EXAMPLE**

In the example below, the principal component scores are calculated from *test.dat* and sent to *score.dat*, using *pca.dat* in which the mean vectors and the eigen vectors are contained.

```
pcas pca.dat -l 3 -n 2 < test.dat > score.dat
```

In the *pca.dat*, the mean vector must be in the front of eigen vectors.

**SEE ALSO**

`pca`

**NAME**

phase – transform real sequence to phase

**SYNOPSIS**

phase [-l L] [-p *pfile*] [-z *zfile*] [-m M] [-n N] [*infile*]

**DESCRIPTION**

*phase* calculates the phase of the spectrum of a real sequence from *infile* (or standard input), sending the result to standard output. Assume that the input sequence is

$$x(0), x(1), \dots, x(L-1)$$

and the FFT is

$$\begin{aligned} X_k &= X(e^{j\omega}) \Big|_{\omega = \frac{2\pi k}{L}} \\ &= \sum_{m=0}^{L-1} x(m)e^{-j\omega m} \Big|_{\omega = \frac{2\pi k}{L}}, \quad k = 0, 1, \dots, L-1 \end{aligned}$$

Then the output is given by

$$Y_k = \arg X_k, \quad k = 0, 1, \dots, L/2$$

In this case the phase is written in continuous form. The output data angular frequency varies from  $0 \sim \pi$ . Input and output data are in float format.

If the **-p**, **-z** options are assigned then the phase of the corresponding filter related to the assigned coefficients is calculated <sup>1</sup>.

**OPTIONS**

- |           |              |  |        |
|-----------|--------------|--|--------|
| <b>-l</b> | <i>L</i>     | frame length power of 2  | [256]  |
| <b>-p</b> | <i>pfile</i> | numerator coefficients file  | [NULL] |
|           |              | The <i>pfile</i> should follow this structure in float format:   |        |
|           |              | $K, a(1), \dots, a(M)$   |        |
| <b>-z</b> | <i>zfile</i> | denominator coefficients file  | [NULL] |
|           |              | The <i>zfile</i> should follow this structure in float format:   |        |
|           |              | $b(0), b(1), \dots, b(N)$  |        |
|           |              | The contents of <i>pfile</i> and <i>zfile</i> should be in a similar form to that used in command <i>dfs</i> . When only the <b>-p</b> option is assigned then the denominator is made equal to 1. When only the <b>-z</b> option is assigned then the numerator and the gain <i>K</i> are made equal to 1. If neither <b>-p</b> nor <b>-z</b> are assigned, data is read from the standard input. |        |

<sup>1</sup> In this case the phase is not evaluated from the filter impulse response, the phase is evaluated from the difference between the numerator and denominator phases

<b>-m</b>	$M$	order of denominator polynomial In the case where the number of input data values is less than $M + 1$ , then $M$ is made equal to the number of input data values $-1$ . If the input data should not be analyzed in blocks of size $M + 1$ , then it is not necessary to assign a value to $M$ .	$[L - 1]$
<b>-n</b>	$N$	order of numerator polynomial Similarly to the $-m$ option, in the case where the number of input data values is less than $N + 1$ , then $N$ is made equal to the number of input data values $-1$ . If the input data should not be analyzed in blocks of size $N + 1$ , then it is not necessary to assign a value to $N$ .	$[L - 1]$
<b>-u</b>		unlapping	[TRUE]

**EXAMPLE**

In the example below, the phase characteristic of a digital filter with coefficients assigned by the files *data.p*, *data.z* in float format is displayed:

```
phase -p data.p -z data.z | fdrw | xgr
```

If the filter defined by *data.p*, *data.z* is stable then the following command gives rise to a similar result:

```
impulse | dfs -p data.p -z data.z | phase | fdrw | xgr
```

**SEE ALSO**

spec, fft, ffr, dfs

**BUGS**

When the sample interval between FFT points is large (the value assigned by the  $-l$  option is small), when the phase characteristic includes steep angles (when zeros and/or poles are close to the unit circle in the  $z$  domain), then sometimes phase is not properly drawn in continuous form.

**NAME**

*pitch* – pitch extraction

**SYNOPSIS**

```
pitch [ -s S ] [ -l L ] [ -t T ] [ -L Lo ] [ -H Hi ] [ -e E ]
      [ -i I ] [ -j J ] [ -d D ] [ infile ]
```

**DESCRIPTION**

*pitch* uses the cepstrum method to calculate the pitch period values corresponding to frames of input data of length *L* from *infile* (or standard input), sending the result to standard output. For unvoiced frames, the output value is 0.0. For voiced frames, the output value is proportional to the pitch period.

Input and output data are in float format.

To discriminate between voiced and unvoiced sounds, the unbiased estimation of log spectrum method is applied to evaluate ( $S/10 \times 25$ )-th order cepstrum. Then from these coefficients, the magnitude of log spectrum  $\hat{g}_i(\Omega_k)$  is evaluated. Finally the mean value  $v_i$  for every band is calculated.

$$v_i = \frac{1}{14n} \sum_{k=4n}^{17n} \hat{g}_i(\Omega_k), \quad (\Omega_k = \frac{2\pi k}{N}, n = N/256)$$

Here the FFT size *N* is square number greater than *L*.

If the speech sound is voiced ( $v_i > T$ ), then the FFT cepstral coefficients  $c(m)$  are transformed into  $c(m) \times m$ , and the peak frequency between *Lo* (Hz) and *Hi* (Hz) is the pitch. If the speech sound is unvoiced ( $v_i < T$ ) then 0 is output.

**OPTIONS**

<b>-s</b>	<i>S</i>	sampling frequency (kHz)	[10]
<b>-l</b>	<i>L</i>	frame length	[400]
<b>-t</b>	<i>T</i>	voiced/unvoiced threshold	[6.0]
<b>-L</b>	<i>Lo</i>	minimum fundamental frequency to search for (Hz)	[60]
<b>-H</b>	<i>Hi</i>	minimum fundamental frequency to search for (Hz)	[240]
<b>-e</b>	<i>E</i>	small value for calculate log-spectral envelope	[0.0]

Usually, the options below do not need to be assigned.

<b>-i</b>	<i>I</i>	minimum number of iteration	[2]
<b>-j</b>	<i>J</i>	maximum number of iteration	[30]
<b>-d</b>	<i>D</i>	end condition	[0.1]

**EXAMPLE**

Speech data with sampling rate 10kHz is read in float format from *data.f*, the pitch is evaluated, and the output is written to *data.pitch*:

```
frame +f -l 400 < data.f | window -l 400 |\
pitch -l 400 > data.pitch
```

**SEE ALSO**

excite

**NAME**

poledf – all pole digital filter for speech synthesis

**SYNOPSIS**

**poledf** [ **-m** *M* ] [ **-p** *P* ] [ **-i** *I* ] [ **-t** ] [ **-k** ] *afile* [ *infile* ]

**DESCRIPTION**

*poledf* derives an all pole standard form digital filter from the linear prediction (LPC) coefficients  $K, a(1), \dots, a(M)$  in *afile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

The transfer function  $H(z)$  of an all pole standard form filter is

$$H(z) = \frac{K}{1 + \sum_{m=1}^M a(m)z^{-m}}$$

**OPTIONS**

<b>-m</b>	<i>M</i>	order of coefficients	[25]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-t</b>		transpose filter	[FALSE]
<b>-k</b>		filtering without gain	[FALSE]

**EXAMPLE**

In the example below, the excitation is generated from pitch information read from *data.pitch* in float format, then it is passed through the standard form synthesis filter built from the linear prediction coefficients file *data.lpc*, and synthesized speech is output to *data.syn*:

```
excite < data.pitch | poledf data.lpc > data.syn
```

**SEE ALSO**

lpc, acorr, ltcdf, lmadf, zerodf

**NAME**

psgr – XY-plotter simulator for EPSF

**SYNOPSIS**

```
psgr [ -t title ] [ -s S ] [ -c C ] [ -x X ] [ -y Y ] [ -p P ] [ -r R ] [ -b ]
      [ -T T ] [ -B B ] [ -L L ] [ -R R ] [ -P ] [ infile ]
```

**DESCRIPTION**

*psgr* converts FP5301 plotter commands from *infile* (or standard input) to PostScript (EPSF or PS), sending the result to standard output.

**OPTIONS**

<b>-t</b>	<i>title</i>	title of figure	[NULL]
<b>-s</b>	<i>S</i>	shrink	[1.0]
<b>-c</b>	<i>C</i>	number of copy	[1]
<b>-x</b>	<i>X</i>	x offset (mm)	[0]
<b>-y</b>	<i>Y</i>	y offset (mm)	[0]
<b>-p</b>	<i>P</i>	paper (Letter, A0, A1, A2, A3, A4, A5, B0, B1, B2, B3, B4, B5)	[A4]
<b>-l</b>		landscape	[FALSE]
<b>-r</b>	<i>R</i>	resolution (dpi)	[600]
<b>-b</b>		bold font mode	[FALSE]
<b>-T</b>	<i>T</i>	top margin (mm)	[0]
<b>-B</b>	<i>B</i>	bottom margin (mm)	[0]
<b>-L</b>	<i>L</i>	left margin (mm)	[0]
<b>-R</b>	<i>R</i>	right margin (mm)	[0]
<b>-P</b>		output Postscript code	[FALSE]

**EXAMPLE**

This example sends to a printer a figure file *data.fig* written through the *fig* command:

```
fig data.fig | psgr | lpr
```

**BUGS**

- There is possibility that a part of the Y axis label is not properly output. In this case the user can change the margin to solve this problem.
- In the case that the size of the figure is modified, and included in a  $\text{T}_{\text{E}}\text{X}$ file, there is possibility that it dose not appear correctly. To solve this problem, please use  $\text{T}_{\text{E}}\text{X}$ options for including pictures and corresponding sizes.

**SEE ALSO**

fig, fdrw, xgr



**NAME**

ramp – generate ramp sequence

**SYNOPSIS**

**ramp** [ **-l** *L* ] [ **-n** *N* ] [ **-s** *S* ] [ **-e** *E* ] [ **-t** *T* ]

**DESCRIPTION**

*ramp* generates ramp sequences of length  $L$ , sending the result to standard output. The output is as follows.

$$\underbrace{S, S + T, S + 2T, \dots, S + (L - 1)T}_L$$

Output format is in float format. In the case the last value is assigned the generated sequence is,

$$\underbrace{S, S + T, S + 2T, \dots, E}_{(E-S)/T}$$

If the  $-l$  option,  $-e$  option and  $-n$  option are used in the same time, then only the last option are taken into account.

**OPTIONS**

<b>-l</b>	<i>L</i>	length of ramp sequence	[256]
		In the case $L \leq 0$ then ramp values will be generated indefinitely.	
<b>-n</b>	<i>N</i>	order of ramp sequence	[L-1]
<b>-s</b>	<i>S</i>	start value	[0]
<b>-e</b>	<i>E</i>	end value	[N/A]
<b>-t</b>	<i>T</i>	step size	[1]

**EXAMPLE**

The Following example output the sequence

$$y(n) = \exp(-n)$$

```
ramp | sopr -m -1 -E | dmp +f
```

**SEE ALSO**

impulse, step, train, sin

**NAME**

reverse – reverse the order of data in each block

**SYNOPSIS**

reverse [ **-l** *L* ] [ **-n** *N* ] [ *infile* ]

**DESCRIPTION**

*reverse* reverses the order of data within *L*-length blocks of input data from *infile* (or standard input), sending the result to standard output. The default value for *L* is the entire file. If *L* is given but the file length is not a multiple of *L*, leftover values are discarded as shown in the example below.

**OPTIONS**

<b>-l</b> <i>L</i>	length of block	[EOF]
<b>-n</b> <i>N</i>	order of block	[EOF-1]

**EXAMPLE**

Let's assume that the following data is read from *data.in* file in float format.

0.0, 1.0, 2.0, 3.0, 4.0, 5.0, 6.0, 7.0, 8.0, 9.0

The command

```
reverse -l 3 data.in > data.out
```

will write the output below to *data.out*.

2.0, 1.0, 0.0, 5.0, 4.0, 3.0, 8.0, 7.0, 6.0

**NAME**

`rmse` – calculation of root mean squared error

**SYNOPSIS**

`rmse [-l L] file1 [infile]`

**DESCRIPTION**

`rmse` calculates RMSE (Root Mean Square Error) of input data sequences from `infile` (or standard input) and `file1`, sending the results to standard output.

From given two files,  $L$ -length time series

$$\underbrace{x_1(0), x_1(1), \dots, x_1(L-1)}_{\text{series 1}}, \underbrace{x_2(0), x_2(1), \dots}_{\text{series 2}}$$

and

$$\underbrace{y_1(0), y_1(1), \dots, y_1(L-1)}_{\text{series 1}}, \underbrace{y_2(0), y_2(1), \dots}_{\text{series 2}}$$

are read, and then RMSE of these two series are calculated and output

$$\text{RMSE}_j = \sqrt{\sum_{m=0}^{L-1} (x_j(m) - y_j(m))^2 / L}$$

Input and output data are in float format.

**OPTIONS**

`-l L` data length to calculate RMSE. [0]  
If  $L = 0$ , RMSE of whole input data is output.

**EXAMPLE**

This example calculates the RMSE of input data files `data.f1` and `data.f2`, and output its maximum and minimum values:

```
rmse -l 26 data.f1 data.f2 | minmax | dmp +f
```

**SEE ALSO**

histogram, minmax

**NAME**

`root_pol` – calculate roots of a polynomial equation

**SYNOPSIS**

`root_pol` [ **-m** *M* ] [ **-n** *N* ] [ **-e** *E* ] [ **-i** ] [ **-s** ] [ **-r** ] [ *infile* ]

**DESCRIPTION**

`root_pol` finds root values of a polynomial equation from *infile* (or standard input), sending the result to standard output.

For given input file, read coefficients

$$a_0, a_1, \dots, a_n$$

of an  $n$ -th order polynomial equation

$$P(x) = a_0x^n + a_1x^{n-1} + \dots + a_{n-1}x + a_n$$

calculate root values by Durand-Kerner-Aberth method.

If roots of  $P(x)$  are  $z_i$ , the result is sent to standard output in complex form as

$$\begin{array}{ll} \text{Re}[z_0], & \text{Im}[z_0] \\ \text{Re}[z_1], & \text{Im}[z_1] \\ \vdots & \\ \text{Re}[z_{n-1}], & \text{Im}[z_{n-1}] \end{array}$$

or polar form as

$$\begin{array}{ll} |z_0|, & \arg[z_0] \\ |z_1|, & \arg[z_1] \\ \vdots & \\ |z_{n-1}|, & \arg[z_{n-1}] \end{array}$$

Both input and output data are in float format.

**OPTIONS**

<b>-m</b>	<i>M</i>	order of polynomial equation	[32]
<b>-n</b>	<i>N</i>	maximum iteration to search roots	[1000]
<b>-e</b>	<i>E</i>	error margin for roots $\varepsilon$	[ $10^{-14}$ ]
<b>-i</b>		set $a_0 = 1$	[FALSE]
<b>-s</b>		reverse order of coefficients	[FALSE]
<b>-r</b>		output results in polar form	[complex form]

**EXAMPLE**

This example calculates roots of a polynomial equation from file *data.z* and output its results in polar form:

```
root_pol -r < data.z | x2x +a 2
```

**NAME**

`sin` – generate sinusoidal sequence

**SYNOPSIS**

`sin [-l L] [-p P] [-m M]`

**DESCRIPTION**

`sin` generates a discrete sin wave sequence of period  $P$ , length  $L$  and magnitude  $M$ ,

$$x(n) = M \cdot \sin\left(\frac{2\pi}{P} \cdot n\right),$$

sending the result to standard output.

Output data is in float format.

**OPTIONS**

<code>-l</code>	$L$	length	[256]
		In the case $L \leq 0$ then sin values will be generated indefinitely.	
<code>-p</code>	$P$	period	[10.0]
<code>-m</code>	$M$	magnitude	[1.0]

**EXAMPLE**

The following example passes sin wave sequence through Blackman window and displays the results in the screen:

```
sin -p 12.3 | window | fdrw | xgr
```

**SEE ALSO**

impulse, step, train, ramp

**NAME**

smcep – mel-cepstral analysis using 2nd order all-pass filter(15; 16)

**SYNOPSIS**

```
smcep [-a A ][-t T ][-m M ][-l L ][-q Q ]
      [-i I ][-j J ][-d D ][-e E ][-f F ][infile ]
```

**DESCRIPTION**

*smcep* calculates the mel-cepstral coefficients from  $L$ -length framed windowed input data from *infile* (or standard input), sending the result to standard output. The analysis uses a second-order all-pass function raised to the  $1/2$  power:

$$A(z) = \left( \frac{z^{-2} - 2\alpha \cos \theta z^{-1} + \alpha^2}{1 - 2\alpha \cos \theta z^{-1} + \alpha^2 z^{-2}} \right)^{\frac{1}{2}},$$

$$\tilde{z}^{-1} = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}}.$$

Input and output data are in float format.

In the mel-cepstral analysis using a 2nd-order all pass function, the speech spectrum is modeled as  $m$ -th order cepstral coefficients  $c(m)$  as follows.

$$H(z) = \exp \sum_{m=0}^M c(m) B_m(e^{j\omega})$$

where

$$\operatorname{Re} [B_m(e^{j\omega})] = \frac{A^m(e^{j\omega}) + A^m(e^{-j\omega})}{2}$$

The Newton-Raphson method is applied to calculate the mel-cepstral coefficients through the minimization of the cost function.

**OPTIONS**

<b>-a</b>	$A$	all-pass constant $\alpha$	[0.35]
<b>-t</b>	$T$	emphasized frequency $\theta * \pi$ (rad)	[0]
<b>-m</b>	$M$	order of mel cepstrum	[25]
<b>-l</b>	$L_1$	frame length	[256]
<b>-L</b>	$L_2$	ifft size for making matrices	[1024]
<b>-q</b>	$Q$	input data style	[0]
		$Q = 0$ windowed data sequence	
		$Q = 1$ $20 \times \log  f(w) $	
		$Q = 2$ $\ln  f(w) $	
		$Q = 3$ $ f(w) $	
		$Q = 4$ $ f(w) ^2$	

Usually, the options below do not need to be assigned.

<b>-i</b>	$I$	minimum iteration of Newton-Raphson method	[2]
<b>-j</b>	$J$	maximum iteration of Newton-Raphson method	[30]
<b>-d</b>	$D$	end condition of Newton-Raphson	[0.001]
<b>-e</b>	$E$	small value added to periodgram	[0]
<b>-f</b>	$F$	minimum value of the determinant of the normal matrix	[0.000001]

**EXAMPLE**

In the example below, speech data is read in float format from *data.f*, analyzed, and resulting mel-cepstral coefficients are written to *data.mcep*:

```
frame +f < data.f | window | smcep > data.mcep
```

**SEE ALSO**

uels, gcep, mcep, mgcep, mlsadf



**NAME**

`snr` – evaluate SNR and segmental SNR

**SYNOPSIS**

`snr` [ `-l L` ] [ `-o O` ] *file1* [ *infile* ]

**DESCRIPTION**

`snr` calculates the SNR (Signal to Noise Ratio) and the  $\text{SNR}_{\text{seg}}$  (segmental SNR) between corresponding  $L$ -length frames of *file1* and *infile* (or standard input), sending the result to standard output. The output format is specified by the `-o` option.

The SNR and  $\text{SNR}_{\text{seg}}$  can be calculated through the following equation.

$$\text{SNR} = 10 \log \frac{\sum_n \{x(n)\}^2}{\sum_n \{e(n)\}^2} \quad [\text{dB}]$$

$$\text{SNR}_{\text{seg}} = \frac{1}{N_i} \sum_{i=1}^{N_i} \text{SNR}_i \quad [\text{dB}]$$

where

$$e(n) = x_1(n) - x_2(n)$$

The number of frame is represented by  $N_i$ . The segmental SNR has the characteristic that for signals with small amplitude such as consonant sounds it gives rise to a better subjective measure than the SNR.

**OPTIONS**

**-l** *L* frame length [256]  
**-o** *O* output data format [0]

- 0 SNR and SNRseg
- 1 SNR and SNRseg in detail
- 2 SNR
- 3 SNRseg

if 0 or 1 are assigned

then output data is written in ASCII format.

if 2 or 3 are assigned

then output data is written in float format

**EXAMPLE**

The following command reads the input files *data.f1* and *data.f2*, evaluates the SNR and segmental SNR, and sends the results to the standard output:

```
snr data.f1 data.f2
```

**SEE ALSO**

histogram, average, rmse

## NAME

sopr – execute scalar operations

## SYNOPSIS

```
sopr [-a A] [-s S] [-m M] [-d D] [-f F] [-c C] [-magic magic ]
      [-MAGIC MAGIC] [-ABS] [-INV] [-P] [-R] [-SQRT] [-LN]
      [-LOG2] [-LOG10] [-EXP] [-POW2] [-POW10] [-FIX] [-UNIT]
      [-CLIP] [-SIN] [-COS] [-TAN] [-ATAN] [-r mn] [-w mn] [infile]
```

## DESCRIPTION

*sopr* performs a sequence of scalar operations on float data from *infile* (or standard input), sending the float output data to standard output.

The sequence of operations is specified by command line options and is performed in the given order.

## OPTIONS

<b>-a</b>	<i>A</i>	addition $y = x + A$	[FALSE]
<b>-s</b>	<i>S</i>	subtraction $y = x - S$	[FALSE]
<b>-m</b>	<i>M</i>	multiplication $y = x * M$	[FALSE]
<b>-d</b>	<i>D</i>	division $y = x/D$	[FALSE]
<b>-f</b>	<i>F</i>	flooring $y = F$ if $x < F$	[FALSE]
<b>-c</b>	<i>C</i>	ceiling $y = C$ if $x > C$	[FALSE]
<b>-magic</b>	<i>magic</i>	remove magic number	[FALSE]
<b>-MAGIC</b>	<i>MAGIC</i>	replace magic number by <i>MAGIC</i>	[FALSE]
		if -magic option is not given, return error.	
		if -magic or -MAGIC option is given multiple times, also return error.	

If the argument of the above operation option is “*dB*”, “*cent*” or “*octave*” then the value  $20/\log_e 10$ ,  $1200/\log_e 2$  or  $1/\log_e 2$  is assigned respectively. Also similarly, if “*pi*” is written after the operation option, then its value will be used. Actuary expression such as “*ln2*”, “*exp10*”, “*sqrt30*” can also be used as arguments.

<b>-ABS</b>		absolute $y =  x $	[FALSE]
<b>-INV</b>		inverse $y = 1/x$	[FALSE]
<b>-P</b>		square $y = x^2$	[FALSE]
<b>-R</b>		square root $y = \sqrt{x}$	[FALSE]
<b>-SQRT</b>		square root $y = \sqrt{x}$	[FALSE]
<b>-LN</b>		logarithm $y = \log x$	[FALSE]
<b>-LOG2</b>		logarithm $y = \log_2 x$	[FALSE]
<b>-LOG10</b>		logarithm $y = \log_{10} x$	[FALSE]

<b>-EXP</b>		exponential $y = \exp x$	[FALSE]
<b>-POW2</b>		power of 2 $y = 2^x$	[FALSE]
<b>-POW10</b>		power of 10 $y = 10^x$	[FALSE]
<b>-FIX</b>		round $(int)x$	[FALSE]
<b>-UNIT</b>		unit step $u(x)$	[FALSE]
<b>-CLIP</b>		clipping $x * u(x)$	[FALSE]
<b>-SIN</b>		sin $y = \sin(x)$	[FALSE]
<b>-COS</b>		cos $y = \cos(x)$	[FALSE]
<b>-TAN</b>		tan $y = \tan(x)$	[FALSE]
<b>-ATAN</b>		atan $y = \text{atan}(x)$	[FALSE]
<b>-r</b>	<i>mn</i>	read from memory register <i>mn</i> ( $n = 0..9$ )	
<b>-w</b>	<i>mn</i>	write from memory register <i>mn</i> ( $n = 0..9$ )	

**EXAMPLE**

In the following example, a ramp function (0, 1, 2, ...) is multiplied by 2 (0, 2, 4, ...) and then 1 is added (1, 3, 5, ...):

```
ramp | sopr -m 2 -a 1 | dmp +f
```

The output file *data.avg* contains the average taken from data in files *data.f1* and *data.f2* read in float format:

```
vopr -a data.f1 data.f2 | sopr -d 2 > data.avg
```

Data is read in float format from *data.f*, and the results in dB is written to the output:

```
sopr data.f -LN -m dB | dmp +f
```

The following example gives the same results:

```
sopr data.f -LOG10 -m 20 | dmp +f
```

Also, the results in cent is written to the output:

```
sopr data.f -LN -m cent | dmp +f
```

The following example gives the same results:

```
sopr data.f -LOG2 -m 1200 | dmp +f
```

If we want to evaluate the following equation,

$$y = (1 + 3x + 4x^2)/(1 + 2x + 5x^2)$$

then memory registers can be used as follows.

```
sopr data.f -w m0 -m 5 -a 2 -m m0 -a 1 -w m1 \
-r m0 -m 4 -a 3 -m m0 -a 1 -d m1 | dmp +f
```

In the example above, m0 and m1 are memory registers. Registers from m0 to m9 can be used. The `-w` option is used to write into memory register, while the `-r` option is used to read from a register.

**SEE ALSO**

vopr, vsum

**NAME**

`spec` – transform real sequence to log spectrum

**SYNOPSIS**

`spec [-l L] [-m M] [-n N] [-z zfile] [-p pfile]  
[-e E] [-o O] [infile]`

**DESCRIPTION**

`spec` computes the log spectrum magnitude of framed windowed input data from `infile` (or standard input), sending the result to standard output.

Alternatively, given the poles (`-p pfile` option) and zeroes (`-z zfile` option) of a digital filter, `spec` computes the frequency response of that filter.

The output format is specified by the `-y` option.

If the input sequence is

$$x(0), x(1), \dots, x(L-1)$$

and the FFT algorithm is used to evaluate

$$\begin{aligned} X_k &= X(e^{j\omega}) \Big|_{\omega = \frac{2\pi k}{L}} \\ &= \sum_{m=0}^{L-1} x(m)e^{-j\omega m} \Big|_{\omega = \frac{2\pi k}{L}}, \quad k = 0, 1, \dots, L-1 \end{aligned}$$

then if the `-y` option is applied, then output is

$$Y_k = 20 \log_{10} |X_k|, \quad k = 0, 1, \dots, L/2$$

The output data corresponds to angular frequencies varying from  $0 \sim \pi$ . Input and output data are in float format.

If the `-p`, `-z` options are assigned then the phase of the corresponding filter related to the assigned coefficients is calculated <sup>2</sup>.

**OPTIONS**

<code>-l</code>	<code>L</code>	FFT window length <i>L</i> must be power of 2	[256]
<code>-m</code>	<code>M</code>	order of MA part In the case where the number of input data values is less than <i>M</i> + 1, then <i>M</i> is made equal to the number of input data values - 1. If the input data should not be analyzed in blocks of size <i>M</i> + 1, then it is not necessary to assign a value to <i>M</i> .	[0]

<sup>2</sup> In this case the phase is not evaluated from the filter impulse response, the phase is evaluated from the difference between the numerator and denominator phases

<b>-n</b>	<i>N</i>	order of AR part Similarly to the <b>-m</b> option, in the case where the number of input data values is less than $N + 1$ , then $N$ is made equal to the number of input data values $-1$ . If the input data should not be analyzed in blocks of size $N + 1$ , then it is not necessary to assign a value to $N$ .	[0]
<b>-z</b>	<i>zfile</i>	MA coefficients filename The <i>zfile</i> should follow this structure in float format: $b(0), b(1), \dots, b(N)$	[NULL]
<b>-p</b>	<i>pfile</i>	AR coefficients filename The <i>pfile</i> should follow this structure in float format: $K, a(1), \dots, a(M)$	[NULL]
<b>-e</b>	<i>E</i>	small value for calculating $\log()$	[0.0]
<b>-o</b>	<i>O</i>	output format	[0]
		$O = 0$ $20 \times \log  X_k $ $k = 0, 1, \dots, L/2$	
		$O = 1$ $\ln  X_k $ $k = 0, 1, \dots, L/2$	
		$O = 2$ $ X_k $ $k = 0, 1, \dots, L/2$	
		$O = 3$ $ X_k ^2$ $k = 0, 1, \dots, L/2$	

The contents of *pfile* and *zfile* should be in a similar form to that used in command *dfs*. When only the **-p** option is assigned then the denominator is made equal to 1. When only the **-z** option is assigned then the numerator and the gain  $K$  are made equal to 1. If neither **-p** nor **-z** are assigned, data is read from the standard input.

## EXAMPLE

In the example below, a pulse train excitation is passed through digital filter and Blackman window, then the log spectrum magnitude is evaluated and plotted on the screen:

```
train -p 50 | dfs -a 1 0.9 | window | spec | fdrw | xgr
```

This example evaluates the frequency response of digital filter with coefficients assigned by *data.p*, *data.z* in float format:

```
spec -p data.p -z data.z | fdrw | xgr
```

A similar results can be obtained with the following command when filter is stable:

```
impulse | dfs -p data.p -z data.z | spec | fdrw | xgr
```

## SEE ALSO

phase, fft, ftr, dfs

**NAME**

*step* – generate step sequence

**SYNOPSIS**

**step** [ **-l** *L* ] [ **-n** *N* ] [ **-v** *V* ]

**DESCRIPTION**

*step* generates a step sequence of length *L*, sending the result to standard output.

The output is in float format, as follows.

$$\underbrace{V, V, V, \dots, V}_L$$

**OPTIONS**

<b>-l</b>	<i>L</i>	length	[256]
		In the case $L \leq 0$ then values will be generated indefinitely.	
<b>-n</b>	<i>N</i>	order	[255]
<b>-v</b>	<i>V</i>	step value	[1.0]

**EXAMPLE**

In the following example, the unit step sequence passed through a digital filter and sent to the standard output:

```
step | dfs -a 1 -0.8 | dmp +f
```

**SEE ALSO**

impulse, train, ramp, sin



**NAME**

`swab` – swap bytes

**SYNOPSIS**

`swab` [ **-S**  $S_1$  ] [ **-s**  $S_2$  ] [ **-E**  $E_1$  ] [ **-e**  $E_2$  ] [ **+type** ] [ *infile* ]

**DESCRIPTION**

*swab* changes the byte order (from big-endian to little-endian or vice versa) of the input data from *infile* (or standard input), sending the result to standard output.

The range of input data that is changed can be restricted with the **-S**, **-E** or **-s**, **-e** options.

The **+type** option specifies the input and output data formats.

**OPTIONS**

<b>-S</b>	$S_1$	start address	[0]
<b>-s</b>	$S_2$	start offset number	[0]
<b>-E</b>	$E_1$	end address	[EOF]
<b>-e</b>	$E_2$	end offset number	[0]
<b>+type</b>		Input and output data format	[s]
	<b>s</b>	short (2 bytes)	<b>S</b> unsigned short (2 bytes)
	<b>i3</b>	int (3 bytes)	<b>I3</b> unsigned int (3 bytes)
	<b>i</b>	int (4 bytes)	<b>I</b> unsigned int (4 bytes)
	<b>l</b>	long (4 bytes)	<b>L</b> unsigned long (4 bytes)
	<b>le</b>	long long (8 bytes)	<b>LE</b> unsigned long long (8 bytes)
	<b>f</b>	float (4 bytes)	<b>d</b> double (8 bytes)

**EXAMPLE**

In the example below the byte order of the file *data.f* in float format is changed and written to *data.swab*:

```
swab +f data.f > data.swab
```

**NAME**

`train` – generate pulse sequence

**SYNOPSIS**

`train [-l L] [-p P]`

**DESCRIPTION**

`train` generates a normalized pulse train sequence or a sequence with values  $\pm 1$ , sending the result to standard output. Output data is in float format.

**OPTIONS**

- |           |          |  |       |
|-----------|----------|--|-------|
| <b>-l</b> | <i>L</i> | sequence length  | [256] |
| <b>-p</b> | <i>P</i> | frame period ( $P \geq 1.0$ )                                  | [0.0] |
|           |          | if $P = 0.0$ then a sequence with values $\pm 1$ is generated. |       |
| <b>-n</b> | <i>N</i> | type of normalization  | [1]   |
|           |          | When $x(n)$ is impulse sequence                                |       |
|           | 0        | no-normalization   |       |
|           | 1        | normalization as $\sum_{n=0}^{L-1} x^2(n) = 1$                 |       |
|           | 2        | normalization as $\sum_{n=0}^{L-1} x(n) = 1$                   |       |

**EXAMPLE**

The following example displays the spectrum of the signal obtained from passing a train pulse sequence through digital filter:

```
train | dfs -b 1 0.9 | window | spec | fdrw | xgr
```

**SEE ALSO**

impulse, sin, step, ramp

**NAME**

`uels` – unbiased estimation of log spectrum(2; 3)

**SYNOPSIS**

`uels [-m M] [-l L] [-q Q] [-i I] [-j J] [-d D] [-e E] [infile]`

**DESCRIPTION**

`uels` uses the unbiased estimation of log spectrum method to calculate cepstral coefficients  $c(m)$  from  $L$ -length framed windowed input data from `infile` (or standard input), sending the result to standard output.

Input and output data are in float format.

Until the proposition of the unbiased estimation of log spectrum method, the conventional methods had two main problems. Firstly the importance of smoothing of the log spectrum was not clear. Secondly it could not be guaranteed that the bias of the estimated value would be sufficiently small.

The evaluation procedure to obtain the unbiased estimation log spectrum values is similar to other improved methods to calculate cepstral coefficients. The main difference is that in UELS method a non-linear smoothing is used to guaranty that the estimation will be unbiased.

**OPTIONS**

<code>-m</code>	<i>M</i>	order of cepstrum	[25]
<code>-l</code>	<i>L</i>	frame length	[256]
<code>-q</code>	<i>Q</i>	input data style	[0]
		$Q = 0$	windowed data sequence
		$Q = 1$	$20 \times \log  f(w) $
		$Q = 2$	$\ln  f(w) $
		$Q = 3$	$ f(w) $
		$Q = 4$	$ f(w) ^2$

Usually, the options below do not need to be assigned.

<code>-i</code>	<i>I</i>	minimum iteration	[2]
<code>-j</code>	<i>J</i>	maximum iteration	[30]
<code>-d</code>	<i>D</i>	end condition	[0.001]
<code>-e</code>	<i>E</i>	small value added to periodgram	[0.0]

**EXAMPLE**

The example below reads data in float format, evaluates 15-th order log spectrum through UELS method, and sends spectrum coefficients to `data.cep`:

```
frame +f < data.f | window | uels -m 15 > data.cep
```

**SEE ALSO**

gcep, mcep, mgcep, lmadf

**NAME**

`ulaw` –  $\mu$ -law compress/decompress

**SYNOPSIS**

`ulaw` [ `-v` *V* ] [ `-u` *U* ] [ `-c` ] [ `-d` ] [ *infile* ]

**DESCRIPTION**

`ulaw` converts data between 8-bit  $\mu$ -law and 16-bit linear formats. The input data is `infile` (or standard input), and the output is sent to standard output.

If the input is  $x(n)$ , the output is  $y(n)$ , the largest value of input data is  $V$ , compression coefficient is  $U$ , then the compression is made through the following equation.

$$y(n) = \text{sgn}(x(n))V \frac{\log(1 + U \frac{|x(n)|}{V})}{\log(1 + U)}$$

And decompression is

$$y(n) = \text{sgn}(x(n))V \frac{(1 + u)^{|x(n)|/V} - 1}{U}$$

**OPTIONS**

<code>-v</code>	<i>V</i>	maximum of input	[32768]
<code>-u</code>	<i>U</i>	compression ratio	[256]
<code>-c</code>		coder mode	[TRUE]
<code>-d</code>		decoder mode	[FALSE]

**EXAMPLE**

In the following, 16-bit data read from `data.s` is compressed to 8-bit ulaw format, and written to `data.ulaw`

```
x2x +sf data.s | ulaw | sopr -d 256 | x2x +fc -r > data.ulaw
```

**NAME**

us – up-sampling

**SYNOPSIS**

**us** [ **-s** *S* ] [ **-c** *file* ] [ **-u** *U* ] [ **-d** *D* ] [ *infile* ]

**DESCRIPTION**

*us* up-samples data from *infile* (or standard input), sending the result to standard output. The format of input and output data is float. The following filter coefficients can be used.

*S* = 23*F*    \$SPTK/lib/lpfcoef.2to3f

*S* = 23*S*    \$SPTK/lib/lpfcoef.2to3s

*S* = 34      \$SPTK/lib/lpfcoef.3to4

*S* = 45      \$SPTK/lib/lpfcoef.4to5

*S* = 57      \$SPTK/lib/lpfcoef.5to7

*S* = 58      \$SPTK/lib/lpfcoef.5to8

(\$SPTK is the directory where toolkit was installed.)

The ratio between up-sampling and down-sampling can be modified by **-u**, **-d** options. If you want to specify filter coefficients, **-c** should also be given.

Filter coefficients are in ASCII format.

**OPTIONS**

<b>-s</b>	<i>S</i>	conversion type	[58]
	<i>S</i> = 23 <i>F</i>	up sampling by 2 : 3	
	<i>S</i> = 23 <i>S</i>	up sampling by 2 : 3	
	<i>S</i> = 34	up sampling by 3 : 4	
	<i>S</i> = 45	up sampling by 4 : 5	
	<i>S</i> = 57	up sampling by 5 : 7	
	<i>S</i> = 58	up sampling by 5 : 8	
<b>-c</b>	<i>file</i>	filename of low pass filter coefficients	[Default]
<b>-u</b>	<i>U</i>	up-sampling ratio	[N/A]
<b>-d</b>	<i>D</i>	down-sampling ratio	[N/A]

**EXAMPLE**

In this example, the speech data in the input file *data.16*, which was sampled at 16 kHz in short int format, is converted to an 44.1 kHz sampling rate:

```
x2x +sf data.16 | us -s 23F | us -s 23S | us -s 57 | \  
us -c /usr/local/SPTK/lib/lpfcoef.5to7 -u 7 -d 8 | \  
x2x +fs > data.44
```

Note:  $\frac{44100}{16000} = \frac{3 \times 3 \times 7 \times 7 \times 100}{2 \times 2 \times 5 \times 8 \times 100}$

**NAME**

us16 – up-sampling from 10 or 12 kHz to 16 kHz

**SYNOPSIS**

**us16** [ **-s** *S* ] [ *infile* ] [ *outfile* ]

**us16** [ **-s** *S* ] *infile1* ... [ *infileN* ] *outdir*

**DESCRIPTION**

*us16* up-samples data from 10 kHz or 12 kHz to 16 kHz. If *infile* and *outfile* arguments are not given, standard input and standard output are used. If several input files are given, the last argument is taken to be a directory name and multiple output files are created in that directory, with names similar to the input file names but the suffixes are changed to “.16”.

**OPTIONS**

**-s** *S* input sampling frequency 10—12 kHz [10]

**EXAMPLE**

In the example below, speech data sampled at 10 kHz is read from *data.10*, up sampling to 16 kHz is undertaken, and the results are written to *data.16*:

```
us16 -s 10 < data.10 > data.16
```



**NAME**

`uscd` – up/down-sampling from 8, 10, 12, or 16 kHz to 11.025, 22.05, or 44.1 kHz

**SYNOPSIS**

```
uscd [-s S S] [infile] [outfile]  
uscd [-s S S] infile1 ... [infileN] outdir
```

**DESCRIPTION**

`uscd` converts the sample rate from one of 8, 10, 12, or 16 kHz to one of 11.025, 22.04, or 44.1 kHz. If *infile* and *outfile* arguments are not given, standard input and standard output are used. If the last argument names a directory, each of the preceding argument files is re-sampled. The results are stored in multiple files in that directory, with base names the same as the input file base names, but with suffixes reflecting the new sample rate.

**OPTIONS**

```
-s S1 input sampling frequency 8|10|12|16 [10]  
-S S2 output sampling frequency 11.025|22.05|44.1 [11.025]  
S2 can be abbreviated as 11, 22, or 44.  
If the last command line argument is a directory name, the suffix  
for the output files is either “.11”, “.22”, or “.44.”
```

**EXAMPLE**

In the example below, speech data sampled at 16 kHz is read from *data.16*, up sampling to 22.05 kHz is undertaken, and the results are written to *data.22*:

```
uscd -s 16 22.05 < data.16 > data.22
```

**NAME**

vopr – execute vector operations

**SYNOPSIS**

```
vopr [ -l L ] [ -n N ] [ -i ] [ -a ] [ -s ] [ -m ] [ -d ]
      [ -ATAN2 ] [ file1 ] [ infile ]
```

**DESCRIPTION**

This command undertakes vector operations in input files. In other words

*file1* first vector file (if it is not assigned then stdin)

*infile* second vector file (if it is not assigned then stdin)

the first file gives the operation vectors **a** and the second file gives the operation vectors **b**. The assigned operation is undertaken and the results are sent to the standard output.

Input and output data are in float format.

The undertaken action depends on the number of assigned files as well as the vector lengths as exemplified in the following.

If two files are assigned (when only one file is assigned then it is assumed that it corresponds to *infile*) then depending on the values of vector sizes, the following actions are undertaken.

when  $L = 1$

<i>file1</i> (stdin)	$a_1$	$a_2$	...	$a_i$	...
<i>infile</i>	$b_1$	$b_2$	...	$b_i$	...
<i>Output</i> (stdout)	$y_1$	$y_2$	...	$y_i$	...

One data from one file corresponds to one data on the other file.

when  $L \geq 2$

<i>file1</i> (stdin)	$a_{11}, \dots, a_{1L}$	$a_{21}, \dots, a_{2L}$	$a_{31}, \dots, a_{3L}$	$a_{41}, \dots$
<i>infile</i>	$b_1, \dots, b_L$			
<i>Output</i> (stdout)	$y_{11}, \dots, y_{1L}$	$y_{21}, \dots, y_{2L}$	$y_{31}, \dots, y_{3L}$	$y_{41}, \dots$

In this case the operation vector is read only once *infile*, and the operations recursively undertaken in vectors from *file1*.

When the information related to **a** and **b** is contained in a single file, (that is, if only one file is assigned, or if no file assignment is made), then the *-i* option should be used and the action does not depend on the vector length.

when  $L \geq 1$

<i>file</i> (stdin)	$a_{11}, \dots, a_{1L}$	$b_{11}, \dots, b_{1L}$	$a_{21}, \dots, a_{2L}$	$b_{21}, \dots, b_{2L}$	
<i>Output</i> (stdout)	$y_{11}, \dots, y_{1L}$		$y_{21}, \dots, y_{2L}$		

Input vectors are read from a single file.

**OPTIONS**

<b>-l</b>	$L$	length of vector	[1]
<b>-n</b>	$N$	order of vector	[L-1]
<b>-i</b>		when only a file is specified, the file contains a and b.	[FALSE]
<b>-a</b>		addition $y_i = a_i + b_i$	[FALSE]
<b>-s</b>		subtraction $y_i = a_i - b_i$	[FALSE]
<b>-m</b>		multiplication $y_i = a_i * b_i$	[FALSE]
<b>-d</b>		division $y_i = a_i/b_i$	[FALSE]
<b>-ATAN2</b>		atan2 $y_i = \text{atan} 2(b_i, a_i)$	[FALSE]

**EXAMPLE**

The output file *data.c* contains addition of vectors in float format read from *data.a* and *data.b*:

```
vopr -a data.a data.b > data.c
```

In the following example, a sin wave is passed through a window with length 256 and coefficients given from *data.w*:

```
sin -p 30 -l 1000 | vopr data.w -l 256 -m | fdrw | xgr
```

The above example can be undertaken in similar way through the example below if the contents of *data.w* corresponds to Blackman window:

```
sin -p 30 -l 1000 | window | fdrw | xgr
```

**SEE ALSO**

sopr, vsum

**NAME**

vq – vector quantization

**SYNOPSIS**

**vq** [-l L] [-n N] [-q] *cbfile* [*infile*]

**DESCRIPTION**

*vq* uses vector quantization to compress vectors from *infile* (or standard input) according to the codebook *cbfile*, sending either codebook indexes or quantized vectors to standard output.

For each length  $L$  input vector

$$x(0), x(1), \dots, x(L-1)$$

*vq* finds the codebook vector  $c_i$  that minimizes the Euclidean distance

$$d_i = \frac{1}{L} \sum_{m=0}^{L-1} (x(m) - c_i(m))^2.$$

Input data is in float format. If the `-q` option is given, the output is the code vector  $[c_i(0), c_i(1), \dots, c_i(L-1)]$  in float format. If the `-q` option is not given, the output is codebook index  $i$  in int format.

**OPTIONS**

<code>-l</code>	$L$	length of vector	[26]
<code>-n</code>	$N$	order of vector	[L-1]
<code>-q</code>		output quantized vector	[FALSE]

**EXAMPLE**

In this example, a sequence of length 25 is read from *data.f* in float format. it is quantized using codebook *cbfile*, and the results are written to *data.vq*:

```
vq -q cbfile < data.f > data.vq
```

**SEE ALSO**

ivq, msvq, imsvq, lbg

**NAME**

vstat – vector statistics calculation

**SYNOPSIS**

vstat [-l L] [-n N] [-t T] [-c C] [-d] [-o O] [*infile*]

**DESCRIPTION**

*vstat* calculates the mean and covariance of groups of vectors from *infile* (or standard input), sending the result to standard output.

For each group of  $T$  input vectors of length  $L$ , *vstat* calculates the length  $L$  mean vector and the  $L \times L$  covariance matrix. That is, if the input data is

$$\overbrace{\underbrace{x_1(1), \dots, x_1(L)}_L, \underbrace{x_2(1), \dots, x_2(L)}_L, \dots, \underbrace{x_N(1), \dots, x_N(L)}_L}_{T \times L}, \dots$$

then the output will be

$$\underbrace{\mu(1), \dots, \mu(L)}_L, \overbrace{\underbrace{\sigma(11), \dots, \sigma(1L)}_L, \dots, \underbrace{\sigma(L1), \dots, \sigma(LL)}_L}_{L \times L}, \dots$$

evaluation of  $\boldsymbol{\mu}$ ,  $\boldsymbol{\Sigma}$  is undertaken by

$$\boldsymbol{\mu} = \frac{1}{N} \sum_{k=1}^N \mathbf{x}$$

$$\boldsymbol{\Sigma} = \frac{1}{N} \sum_{k=1}^N \mathbf{x}\mathbf{x}' - \boldsymbol{\mu}\boldsymbol{\mu}'$$

If the `-d` option is given, the length  $L$  diagonal of the covariance matrix is output instead of the entire  $L \times L$  matrix.

If the `-o 3` option is specified, *vstat* also calculates the confidence interval of the mean via Student's  $t$ -distribution for each dimension. That is, for each dimension, the confidence interval can be estimated at the confidence level  $\alpha$  (%) with satisfying the following condition;

$$t(\alpha, \phi) \geq \left| \frac{\mu(i) - m(i)}{\sqrt{\hat{\sigma}(i)^2 / L}} \right|, \quad i = 1, 2, \dots, L$$

where  $t(\alpha, \phi)$  is the upper  $0.5(100 - \alpha)$ th percentile of the  $t$ -distribution with the  $\phi$  degrees of freedom,  $m(i)$  is the population mean,  $\hat{\sigma}(i)^2$  is the unbiased variance. The confidence

level  $\alpha$  can be specified by `-c` option. The upper / lower bound  $u(i)$  and  $l(i)$  can be written as

$$u(i) = \mu(i) + t(\alpha, L - 1) \sqrt{\frac{\hat{\sigma}(i)^2}{L}},$$

$$l(i) = \mu(i) - t(\alpha, L - 1) \sqrt{\frac{\hat{\sigma}(i)^2}{L}}.$$

The order of the output is as follows.

$$\overbrace{\mu(1), \dots, \mu(L)}^L, \overbrace{u(1), \dots, u(L)}^L, \overbrace{l(1), \dots, l(L)}^L$$

Input and output data are in float format.

## OPTIONS

<code>-l</code>	$L$	length of vector	[1]
<code>-n</code>	$N$	order of vector	[L-1]
<code>-t</code>	$T$	number of vector	[N/A]
<code>-o</code>	$O$	output format	[0]
	$O = 0$	mean & covariance	
	$O = 1$	mean	
	$O = 2$	covariance	
	$O = 3$	mean & upper / lower bound of confidence interval via Student's t-distribution	
<code>-c</code>	$C$	confidence level of confidence interval (%)	[95.00]
<code>-d</code>		diagonal covariance	[FALSE]
<code>-i</code>		output inverse covariance instead of covariance	[FALSE]
<code>-r</code>		output correlation instead of covariance	[FALSE]

## EXAMPLE

The output file `data.stat` contains the mean and covariance matrix taken from the whole data in `data.f` read in float format.

```
vstat data.f > data.stat
```

In the example below, the mean of 15-th order coefficients vector is taken for every group of 3 frames and sent to `data.av`:

```
vstat -l 15 -t 3 -o 1 data.f > data.av
```

## SEE ALSO

average, vsum

**NAME**

`vsum` – summation of vector

**SYNOPSIS**

`vsum` [ **-l** *L* ] [ **-n** *N* ] [ *infile* ]

**DESCRIPTION**

`vsum` calculates the vector sum of groups of  $N$  input vectors of length  $L$  from *infile* (or standard input), sending the result to standard output. That is, if the input data is given by

$$\overbrace{\underbrace{a_1(1), \dots, a_1(L)}_L, \underbrace{a_2(1), \dots, a_2(L)}_L, \dots, \underbrace{a_N(1), \dots, a_N(L)}_L}_{N \cdot L}, \dots$$

then the output is

$$\underbrace{s(1), \dots, s(L)}_L, \dots$$

where  $s(n)$  is

$$s(n) = \sum_{k=1}^N a_k(n)$$

Input and output data are in float format.

**OPTIONS**

**-l** *L* order of vector [1]  
**-n** *N* number of vector [EOD]

**EXAMPLE**

The output file *data.sum* contains the summation for the whole data in file *data.f* read in float format:

```
vsum data.f > data.sum
```

In this example, the norm of 10-th order vectors are evaluated and written to *data.n*:

```
sopr data.f -P | vsum -n 10 | sopr -R > data.n
```

In the next example, 15-th order coefficients vectors are read from *data.f*, the average for every 3 frame is evaluated, and output to *data.av*:

```
vsum -l 15 -n 3 data.f | sopr -d 3 > data.av
```

**SEE ALSO**

sopr



**NAME**

wav2raw – wav (RIFF) to raw

**SYNOPSIS**

wav2raw [ **-swab** ] [ **-d** *D* ] [ **-n** ] [ **-N** ] [ **+type** ] [ *infile* ]

**DESCRIPTION**

*wav2raw* converts file format from wav to raw.

**OPTIONS**

<b>-swab</b>	change “Endian”	[FALSE]
<b>-d</b>	<i>D</i> destination directory	[N/A]
<b>-n</b>	normalization with the maximum value according to bit/sample of the wav file if max >= 255 (8bit), 32767 (16bit), 8388067 (24bit) or 2147483647 (32bit)	[FALSE]
<b>-N</b>	normalization with the maximum value	[FALSE]
<b>+type</b>	output data type	[f]
	<i>c</i> char (1 byte)	<i>C</i> unsigned char (1 byte)
	<i>s</i> short (2 bytes)	<i>S</i> unsigned short (2 bytes)
	<i>i3</i> int (3 bytes)	<i>I3</i> unsigned int (3 bytes)
	<i>i</i> int (4 bytes)	<i>I</i> unsigned int (4 bytes)
	<i>l</i> long (4 bytes)	<i>L</i> unsigned long (4 bytes)
	<i>le</i> long long (8 bytes)	<i>LE</i> unsigned long long (8 bytes)
	<i>f</i> float (4 bytes)	<i>d</i> double (8 bytes)
	<i>a</i> ascii	

**EXAMPLE**

The following example shows the wav file format read from *data.wav* is converted to the raw file format and normalized with the maximum value, and the output is *data.raw* in the same directory as *data.wav* unless *-d* option is given. :

```
wav2raw -N data.wav
```

**SEE ALSO**

raw2wav, swab

**NAME**

window – data windowing

**SYNOPSIS**

**window** [ **-l**  $L_1$  ] [ **-L**  $L_2$  ] [ **-n**  $N$  ] [ **-w**  $W$  ] [ *infile* ]

**DESCRIPTION**

*window* multiplies, on an element-by-element basis, length  $L$  input vectors from *infile* (or standard input) by a specified windowing function, sending the result to standard output.

For the input data

$$x(0), x(1), \dots, x(L_1 - 1)$$

and the windowing function

$$w(0), w(1), \dots, w(L_1 - 1),$$

the output is calculated as follows:

$$x(0) \cdot w(0), x(1) \cdot w(1), \dots, x(L_1 - 1) \cdot w(L_1 - 1).$$

If  $L_2$  is greater than  $L_1$ , then 0s are added to the output as follows.

$$\underbrace{x(0) \cdot w(0), x(1) \cdot w(1), \dots, x(L_1 - 1) \cdot w(L_1 - 1)}_{L_2}, 0, \dots, 0$$

Input and output data are in float format.

**OPTIONS**

- |           |       |   |           |
|-----------|-------|---|-----------|
| <b>-l</b> | $L_1$ | frame length of input ( $L \leq 2048$ )             | [256]     |
| <b>-L</b> | $L_2$ | frame length of output                              | [ $L_1$ ] |
| <b>-n</b> | $N$   | type of normalization                               | [1]       |
|           | 0     | no normalization                                    |           |
|           | 1     | normalization such as $\sum_{n=0}^{L-1} w^2(n) = 1$ |           |
|           | 2     | normalization such as $\sum_{n=0}^{L-1} w(n) = 1$   |           |
| <b>-w</b> | $W$   | type of window                                      | [0]       |
|           | 0     | Blackman  |           |
|           | 1     | Hamming   |           |
|           | 2     | Hanning   |           |
|           | 3     | Bartlett  |           |
|           | 4     | trapezoid   |           |
|           | 5     | rectangular   |           |

**EXAMPLE**

This example prints in the screen a sin wave function with period 20 after windowing it with a Blackman window:

```
sin -p 20 | window | fdrw | xgr
```

This example passes the excitation generated through a train pulse by a digital filter, applies to it a Blackman windowing function, evaluates the log magnitude spectrum through 512 points FFT, and plots the results in the screen:

```
train -p 50 | dfs -a 1 0.9 | window -l 50 -L 512 |\  
spec -l 512 | fdrw | xgr
```

**SEE ALSO**

fftr, spec

**NAME**

`x2x` – data type transformation

**SYNOPSIS**

`x2x` [ `+type1` ] [ `+type2` ] [ `%format` ] [ `+aN` ] [ `-r` ]

**DESCRIPTION**

`x2x` converts data from standard input to a different data type, sending the result to standard output.

The input and output data type are specified by command line options as described below.

**OPTIONS**

<code>+type1</code>	input data type	[f]
<code>+type2</code>	output data type	[type1]
	both options <code>type1, type2</code> can be assigned, one of the options below.	
	<code>c</code> char (1 byte)	<code>C</code> unsigned char (1 byte)
	<code>s</code> short (2 bytes)	<code>S</code> unsigned short (2 bytes)
	<code>i3</code> int (3 bytes)	<code>I3</code> unsigned int (3 bytes)
	<code>i</code> int (4 bytes)	<code>I</code> unsigned int (4 bytes)
	<code>l</code> long (4 bytes)	<code>L</code> unsigned long (4 bytes)
	<code>le</code> long long (8 bytes)	<code>LE</code> unsigned long long (8 bytes)
	<code>f</code> float (4 bytes)	<code>d</code> double (8 bytes)
	<code>de</code> long double (12 bytes)	<code>a</code> ASCII
	data type is converted from $t_1(type_1)$ to $t_2(type_2)$ . if $t_2$ is not assigned then no operation is undertaken, and the output file is equal to the input file.	
<code>+aN</code>	specify the column number $N$	[1]
<code>-r</code>	specify rounding off when a real number is substituted for a integer	[FALSE]
<code>-o</code>	clip by minimum and maximum of output data type if input data is over the range of output data type. if <code>-o</code> option is not given, process is aborted in the above case.	[FALSE]
<code>%format</code>	specify output format similar to 'printf()', if <code>type2</code> is ASCII.	[%g]

**EXAMPLE**

The following example converts data in ASCII format read from `data.asc`, converts to float format, and writes the output to `data.f`:

```
x2x +af < data.asc > data.f
```

This example reads data in float format from *data.f* converts to ASCII format, and sends the output to the screen:

```
x2x +fa < data.f
```

For example, if contents of *data.f* in float format are

```
1,2,3,4,5,6,7
```

then the following output is printed to the screen.

```
1
2
3
4
5
6
7
```

If for the same data in the example above the number of column is assigned: column

```
x2x +fa3 < data.f
```

the output is

```
1      2      3
4      5      6
7
```

The output uses the printf command *%e* format:

```
x2x +fa %9.4e < data.f
```

In this example the total number of characters for each number is 11, and the number of decimal points assigned to 4.

```
1.0000e+000
2.0000e+000
:
7.0000e+000
```

## SEE ALSO

dmp

**NAME**

`xgr` – XY-plotter simulator for X-window system

**SYNOPSIS**

```
xgr [ -s S ] [ -l ] [ -rv ] [ -m ] [ -bg BG ] [ -hl HL ] [ -bd BD ]
    [ -ms MS ] [ -g G ] [ -d D ] [ -t T ] [ infile ]
```

**DESCRIPTION**

`xgr` plots a graph from a sequence of FP5301 plotter commands, displaying the output on the screen in a new X window.

When the X window is created, the keyboard focus is initially assigned to that new window, which responds to a limited set of user interactions:

- Changing the window size truncates or expands the area in which the graph is displayed, but the graph stays the same size; it is not rescaled to fit the new window size.
- If the graph is larger than the window, the position within the window can be changed with “vi” cursor movement commands:
  - h: left scroll
  - j: down scroll
  - k: up scroll
  - l: right scroll
- To delete the window, type one of the following: “q”, “Ctrl-c”, “Ctrl-d”

**OPTIONS**

<code>-s</code>	<i>S</i>	shrink	[3.38667]
<code>-l</code>		landscape	[FALSE]
<code>-rv</code>		reverse mode	[FALSE]
<code>-m</code>		monochrome display mode	[FALSE]
<code>-bg</code>	<i>BG</i>	background color	[white]
<code>-hl</code>	<i>HL</i>	highlight color	[blue]
<code>-bd</code>	<i>BD</i>	border color	[blue]
<code>-ms</code>	<i>MS</i>	mouse color	[red]
<code>-g</code>	<i>G</i>	geometry	[NULL]
<code>-d</code>	<i>D</i>	display	[NULL]
<code>-t</code>	<i>T</i>	window title	[xgr]

**EXAMPLE**

The following example uses `fdrw` to draw a graph based on data read from `data.f`, and sends the output in a X-Window environment:

```
fdrw < data.f | xgr
```

## BUGS

- If the display server does not contain backing store function, then the hidden part of virtual screen is erased.
- To lessen the waiting time to display graphs, a image of virtual screen is copied to the memory. If the size assigned by the `-g` option is too small or if during the time the graph is being plotted an another window is put above the virtual screen, then a part of virtual screen will be erased. The `-s` option is suggested whenever the size of the virtual screen should be reduced.

## SEE ALSO

fig, fdrw

**NAME**

zcross – zero cross

**SYNOPSIS**

**zcross** [ **-l** *L* ] [ **-n** ] [ *infile* ]

**DESCRIPTION**

*zcross* determines the number of zero crossings within each length *L* input vector, sending the result to standard output as one float number for each input vector.

Input and output data are in float format.

**OPTIONS**

<b>-l</b>	<i>L</i> frame length	[256]
	if $L \leq 0$ then no data output.	
<b>-n</b>	normalized by frame length	[FALSE]

**EXAMPLE**

Data in float format is read from *data.f*, a zero crossing rate is computed, and the results is written to *data.zc*:

```
zcross < data.f > data.zc
```

**SEE ALSO**

frame, spec



**NAME**

`zerodf` – all zero digital filter for speech synthesis

**SYNOPSIS**

`zerodf` [ **-m** *M* ] [ **-p** *P* ] [ **-i** *I* ] [ **-t** ] [ **-k** ] *bfile* [ *infile* ]

**DESCRIPTION**

`zerodf` derives a standard-form FIR (all-zero) digital filter from the coefficients  $b(0), b(1), \dots, b(M)$  in *bfile* and uses it to filter an excitation sequence from *infile* (or standard input) to synthesize speech data, sending the result to standard output.

Input and output data are in float format.

The transfer function  $H(z)$  of an FIR filter in standard form is

$$H(z) = \sum_{m=0}^M b(m)z^{-m}$$

**OPTIONS**

<b>-m</b>	<i>M</i>	order of coefficients	[25]
<b>-p</b>	<i>P</i>	frame period	[100]
<b>-i</b>	<i>I</i>	interpolation period	[1]
<b>-t</b>		transpose filter	[FALSE]
<b>-k</b>		filtering without gain	[FALSE]

**EXAMPLE**

Excitation is generated from pitch information read in float format from *data.pitch*, passes through a FIR filter with coefficients read from *data.b*, and the synthesized speech is written to *data.syn*:

```
excite < data.pitch | zerodf data.b > data.syn
```

**SEE ALSO**

`poledf`, `lmadf`



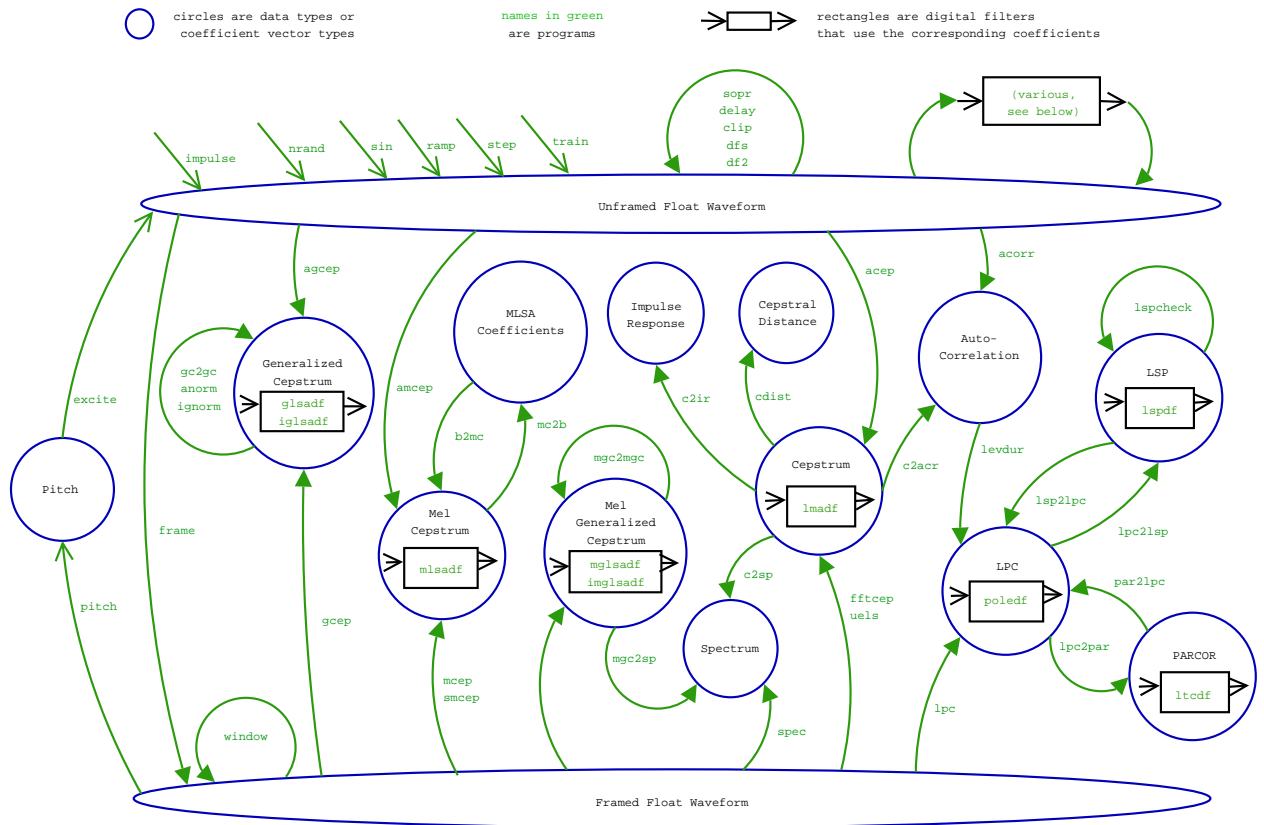
## REFERENCES

- [1] S. Imai and Y. Abe, "Spectral envelope extraction by improved cepstral method," *Journal of IEICE*, Vol.J62-A, No.4, pp.217–223, Apr. 1987. (*in Japanese*)
- [2] S. Imai and C. Furuichi, "Unbiased estimation of log spectrum," *Journal of IEICE*, Vol.J70-A, No.3, pp.471–480, Mar. 1987. (*in Japanese*)
- [3] S. Imai and C. Furuichi, "Unbiased estimator of log spectrum and its application to speech signal processing," *Signal Processing IV: Theory and Applications*, Vol.1, pp.203–206, Elsevier, North-Holland, 1988.
- [4] K. Tokuda, T. Kobayashi, S. Shiimoto, and S. Imai, "Adaptive cepstral analysis — Adaptive filtering based on cepstral representation —," *Journal of IEICE*, Vol.J73-A, No.7, pp.1207–1215, July 1990. (*in Japanese*)
- [5] K. Tokuda, T. Kobayashi, and S. Imai, "Adaptive cepstral analysis of speech," *IEEE Trans. Speech and Audio Process.*, Vol.3, No.6, pp.481–488, Nov. 1995.
- [6] K. Tokuda, T. Kobayashi, R. Yamamoto, and S. Imai, "Spectral estimation of speech based on generalized cepstral representation," *Journal of IEICE*, Vol.J72-A, No.3, pp.457–465, Mar. 1989. (*in Japanese*)
- [7] T. Kobayashi and S. Imai, "Spectral analysis using generalized cepstrum," *IEEE Trans. Acoust., Speech, Signal Process.*, Vol.ASSP-32, No.5, pp.1087–1089, Oct. 1984.
- [8] K. Tokuda, T. Kobayashi, and S. Imai, "Generalized cepstral analysis of speech — a unified approach to LPC and cepstral method," *Proc. ICSLP-90*, pp.37–40, Nov. 1990.
- [9] T. Fukada, K. Tokuda, T. Kobayashi, and S. Imai, "A study on adaptive generalized cepstral analysis," *IEICE Spring National Convention*, A-150, p.150, Mar. 1990. (*in Japanese*)
- [10] K. Tokuda, T. Kobayashi, T. Fukada, H. Saito, and S. Imai, "Spectral estimation of speech based on mel-cepstral representation," *Journal of IEICE*, Vol.J74-A, No.8, pp.1240–1248, Aug. 1991. (*in Japanese*)
- [11] K. Tokuda, T. Kobayashi, T. Fukada, and S. Imai, "Adaptive mel-cepstral analysis of speech," *Journal of IEICE*, Vol.J74-A, No.8, pp.1249–1256, Aug. 1991. (*in Japanese*)
- [12] T. Fukada, K. Tokuda, T. Kobayashi, and S. Imai, "An adaptive algorithm for mel-cepstral analysis of speech," *Proc. ICASSP-92*, pp.137–140, Mar. 1992.

- [13] K. Tokuda, T. Kobayashi, K. Chiba, and S. Imai, "Spectral estimation of speech by mel-generalized cepstral analysis," *Journal of IEICE*, Vol.J75-A, No.7, pp.1124–1134, July 1992. (in Japanese)
- [14] K. Tokuda, T. Kobayashi, T. Masuko, and S. Imai, "Mel-generalized cepstral analysis — a unified approach to speech spectral estimation," *Proc. ICSLP-94*, pp.1043–1046, Sep. 1994.
- [15] T. Wakako, K. Tokuda, T. Masuko, T. Kobayashi, and T. Kitamura, "Speech spectral estimation based on expansion of log spectrum by arbitrary basis functions," *Journal of IEICE*, Vol.J82-D-II, No.12, pp.2203–2211, Dec. 1999. (in Japanese)
- [16] C. Miyajima C, H. Watanabe, K. Tokuda, T. Kitamura, and S. Katagiri, "A new approach to designing a feature extractor in speaker identification based on discriminative feature extraction," *Speech Communication*, Vol.35, No.3, pp.203–218, Oct. 2001.
- [17] S. Imai, "Log magnitude approximation (LMA) filter," *Journal of IEICE*, Vol.J63-A, No.12, pp.886–893, Dec. 1987. (in Japanese)
- [18] T. Chiba, K. Tokuda, T. Kobayashi, and S. Imai, "Speech synthesis based on mel-generalized cepstral representation," *IEICE Spring National Convention*, A-243, p.243, Mar. 1988. (in Japanese)
- [19] S. Imai, "Cepstral analysis synthesis on the mel frequency scale," *Proc. ICASSP-83*, pp.93–96, Apr. 1983.
- [20] S. Imai, K. Sumita, and C. Furuichi, "Mel log spectrum approximation (MLSA) filter for speech synthesis," *Journal of IEICE*, Vol.J66-A, No.2, pp.122–129, Feb. 1983. (in Japanese)
- [21] T. Kobayashi, S. Imai, and Y. Fukuda, "Mel generalized-log spectrum approximation (MGLSA) filter," *Journal of IEICE*, Vol.J68-A, No.6, pp.610–611, June 1985. (in Japanese)
- [22] K. Koishida, G. Hirabayashi, K. Tokuda, and T. Kobayashi, "A 16kbit/s wideband CELP-based speech coder using mel-generalized cepstral analysis," *IEICE Trans. Inf. and Syst.*, vol.E83-D, no.4, pp.876–883, Apr. 2000.
- [23] K. Tokuda, T. Masuko, T. Yamada, T. Kobayashi, and S. Imai, "An algorithm for speech parameter generation from continuous mixture HMMs with dynamic features," *Proc. EUROSPEECH-95*, pp.757–760, Sep. 1995.

# Block diagram of SPTK commands

Mitch Bradley kindly provided us the following diagram to help users understand and remember the relationships between the SPTK commands and data representations.





# INDEX of TOPICS

## data operation

- bcp, 10
- bcut, 12
- dmp, 32
- fd, 38
- merge, 113
- minmax, 125
- reverse, 148
- swab, 163
- wav2raw, 179
- x2x, 182

## number operation

- sopr, 157
- vopr, 172

## data processing

- average, 8
- cdist, 18
- clip, 20
- delta, 27
- histogram, 77
- linear\_intpl, 93
- nan, 134
- pca, 138
- pcas, 139
- rmse, 149
- snr, 155
- vstat, 175
- vsum, 177

## sampling rate transformation

- ds, 34
- us, 168
- us16, 170
- uscd, 171

## DA transformation

- da, 21

## plotting graphs

- fdrw, 39
- fig, 49
- glogsp, 62
- grlogsp, 72
- gwave, 75
- psgr, 145
- xgr, 184

## signal generation

- excite, 36
- nrand, 136
- ramp, 147
- sin, 152
- step, 162
- train, 164

## digital filter

- df2, 29
- dfs, 30

## signal processing

- acorr, 3
- dct, 23
- decimate, 25
- delay, 26
- fft, 41
- fft2, 42
- fftcep, 45
- fftr, 46
- fftr2, 47
- frame, 56
- frequ, 57
- grpdelay, 74
- idct, 78
- ifft, 80
- ifft2, 81
- ignorm, 83

- impulse, 84
  - interpolate, 86
  - levdur, 91
  - lpc, 98
  - norm0, 135
  - phase, 140
  - pitch, 142
  - root\_pol, 150
  - spec, 160
  - ulaw, 167
  - window, 180
  - zcross, 186
- speech analysis and synthesis
- excite, 36
  - frame, 56
  - pitch, 142
  - window, 180
- speech analysis
- acep, 1
  - agecep, 4
  - amcep, 6
  - gcep, 60
  - mcep, 111
  - mgcep, 119
  - smcep, 153
  - uels, 165
- speech parameter transformation
- b2mc, 9
  - c2acr, 15
  - c2ir, 16
  - c2sp, 17
  - freqt, 57
  - gc2gc, 58
  - gnorm, 71
  - lpc2c, 99
  - lpc2lsp, 101
  - lpc2par, 104
  - lsp2lpc, 106
  - lspcheck, 107
  - mc2b, 110
  - mgc2mgc, 115
  - mgc2sp, 117
  - par2lpc, 137
- filters for speech synthesis
- glsadf, 64
  - lmadf, 95
  - lspdf, 108
  - ltcdf, 109
  - mglsadf, 122
  - mlsadf, 130
  - poledf, 144
  - zerodf, 187
- vector quantization
- extract, 37
  - imsvq, 85
  - ivq, 87
  - lbg, 88
  - msvq, 133
  - vq, 174
- parameter generation
- mlpg, 127
- others
- bell, 14
  - echo2, 35
- model training
- gmm, 66
- probability calculation
- gmmp, 69