

Digital Signal Processing (review)

Sound = 1D signal - pressure as a function of time $p(t)$

• represent it as discrete-time samples $p[n]$ $n=0,1,2,\dots$



$T = nT$ sampling interval
 \rightarrow samp. freq. $f_s = \frac{1}{T}$
 (8kHz - 48kHz)

DT sigs

• computer \rightarrow finite precision $\Rightarrow p[n]$ is quantized

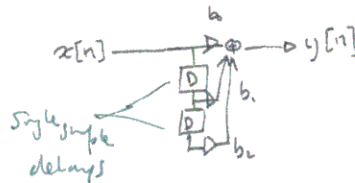
e.g. 16 bit linear $p_d[n] = K[-32768..32767]$

non-integer part thrown away \Rightarrow quantization noise.

Signal processing = signals + systems

Most useful class of systems = LTI/LCCDE

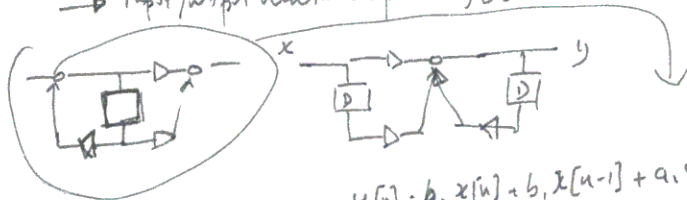
e.g. those as block diagrams:



Sys

\rightarrow input/output relationship: $y[n] = b_0 x[n] + b_1 x[n-1] + b_2 x[n-2]$

+ feedback



$y[n] = b_0 x[n] + b_1 x[n-1] + a_1 y[n-1]$

hence, LTI as etc don't change

$x = \alpha x_1 + \beta x_2 \Rightarrow y = \alpha y_1 + \beta y_2$

\Rightarrow impulse response $\delta[n] = \begin{cases} 1 & n=0 \\ 0 & \text{else} \end{cases}$

$x[n] = \delta[n] \Rightarrow y[n] = h[n]$ "impulse response"

but any $x[n] = \sum_m x_m \delta[n-m]$

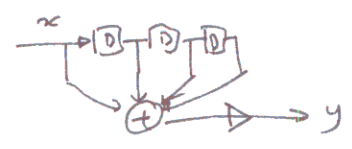
$\Rightarrow y[n] = \sum_m x_m h[n-m] \rightarrow$ CONVOLUTION $y[n] = x[n] * h[n]$
 $\Rightarrow h[n]$ "equivalent descr. of LTI system"

Conv

- examples
- moving average - effect on freq + phase \rightarrow MATLAB
 - simple feedback - INFINITE IR
 - 2nd order feedback \rightarrow SINUSOID $1 - 2r \cos \theta z^{-1} + r^2 z^{-2}$

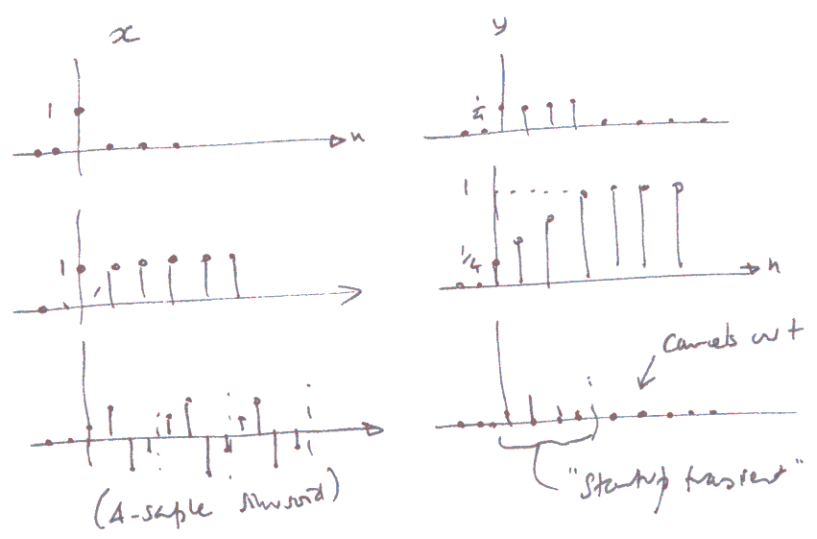
basic systems - examples

"Moving average" filter



$$y[n] = \frac{1}{4}[x[n] + x[n-1] + x[n-2] + x[n-3]]$$

$$= \left[\frac{1}{4} \frac{1}{4} \frac{1}{4} \frac{1}{4} \right] * x[n]$$



⇒ trade + period 4 sinusoid
 → LINEAR vs phase

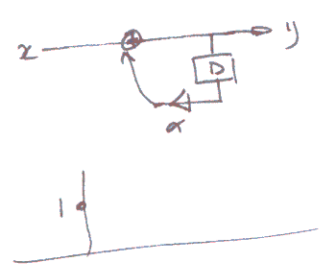
$$x = x_1 + x_2$$

$$\Rightarrow y = y_1 + y_2$$

⇒ effect of sinusoid REMOVED

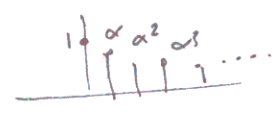
→ matlab

Feedback filter



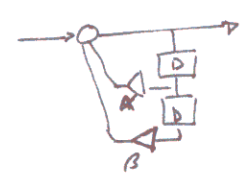
$$y[n] = x[n] + \alpha y[n-1]$$

α > 1 → unstable

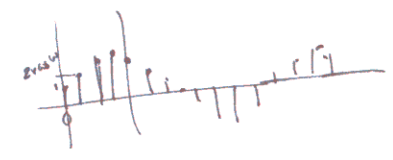


infinite IR
 (quite representative explicitly)

2nd order feedback



→ get oscillation (damped) (any freq)



stable/unstable.

$$\alpha = \cos 2\pi \omega_0; \beta = -r^2$$

Convolution & Correlation:

passing a signal through a system results in convolution of input and IR

$$y[n] = \sum_{m=0}^{n-1} h[m] x[n-m]$$

indices $\leq n$

$$= "h[n] * x[n]"$$

$$= x[n] * h[n] \rightarrow \text{pulse! HW}$$

- weighted sum of delayed x's
 - inner product between shifted $x \times h[-n]$

Correlation is explicitly inner product:

$$r_{xy}[l] = \sum x[n] y[n+l]$$

↑
lag

LOW

- peak in r_{xy} at $l_0 \Rightarrow x[n]$ "resembles" $y[n+l]$ (i.e. y is x delayed by l)
 (\Rightarrow matched filter)

- order matters: $r_{xy}[l] = r_{yx}[-l]$
- relates to conv: $r_{xy}[l] = x[l] * y[-l]$

Autocorrelation: $r_{xx}[l]$ finds shifts at which x resembles self i.e. PERIODICITY.

- ML example:
- picking out a pulse in noise
 - periodicity of voice (RT window)

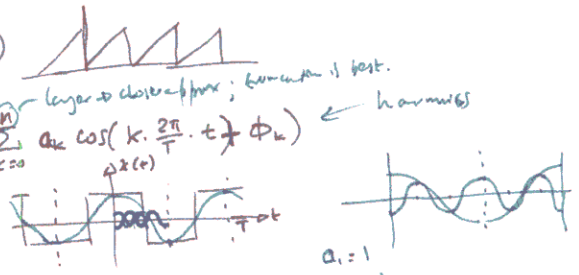
SOLA

FOURIER (frequency) domain:

Fourier series: periodic $x(t) = x(t+T)$

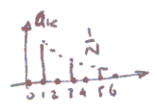
$$\Rightarrow x(t) \approx \sum_{k=0}^N a_k \cos(k \cdot \frac{2\pi}{T} \cdot t + \phi_k)$$

← harmonics



FS

→ Fourier domain



Complex sinusoids $\rightarrow x(t) = \sum_{k=-\infty}^{\infty} c_k e^{j \frac{2\pi k t}{T}}$
 complex; $c_k = c_{-k}^*$ for real x
 $c_k = a_k e^{j\phi_k}$ (?)

$$\Rightarrow \text{Fourier analysis } c_k = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) e^{-j \frac{2\pi k t}{T}} dt$$

convolution/matched filter/inner product

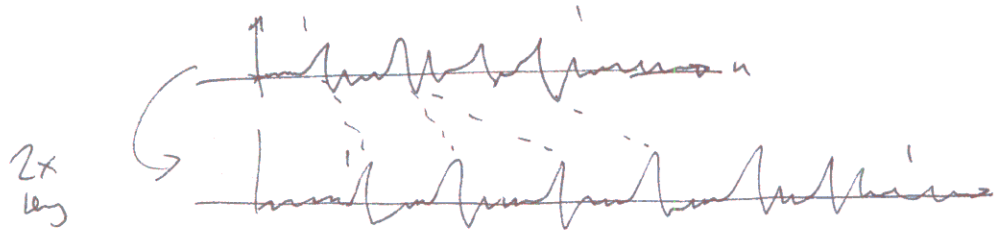
- works because sinusoids are ORTHOGONAL SET

$$x(t) = \sum c_k e^{j k \omega_0 t}$$

$$\Rightarrow \int x(t) e^{-j k_0 \omega_0 t} dt = \int \sum_k c_k e^{j(k-k_0)\omega_0 t} dt = 0 \text{ if } k \neq k_0$$

SOLA - Synchronous overlap Add

- timescale mod by duplicating fragments of signal



duplicate cycles?

but had to choose cycles.
 arbitrary → phase cancellation

→ choose best window (within limits) by
 max cross-correlation.

$$y'[rL+m] = \underbrace{\beta[m] \cdot y[rL+m]}_{\text{recursive ...}} + (1-\beta[m]) \cdot x[q \cdot rL+m + k_m]$$

$$k_m = \max_{0 \leq k \leq k_{max}} \frac{\sum_{m=0}^{N_{ov}} x[q \cdot rL+m+k] \cdot y[rL+m]}{\sqrt{\sum x^2[q \cdot rL+m+k] \cdot \sum y^2[rL+m]}}$$

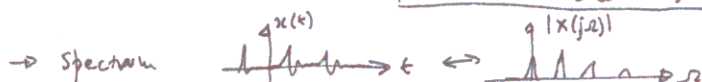
best alignment within window.

Fourier transform

- any $x(t)$ (not just periodic)
- continuous freq ω

$$FT \begin{cases} X(j\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt \\ x(t) = \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} dt \end{cases}$$

FT



(reduces to $\sum_k \delta(\omega - k\omega_0)$ if $x(t)$ is periodic)

Discrete time \rightarrow treat $x[n]$ as $x(t) = \sum_n x[n] \delta(t - nT)$ IMPULSE TRAIN

DTFT

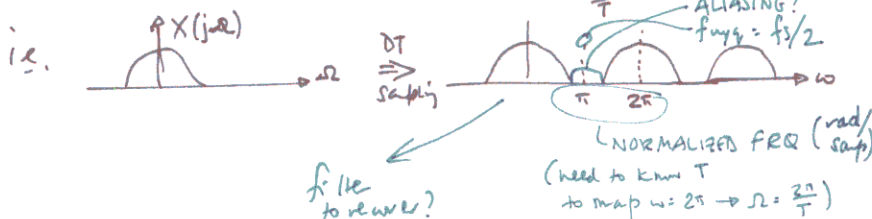
$$\rightarrow DTFT \begin{cases} X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\omega n} \\ x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega \end{cases}$$

Spectrum is periodic " $e^{j\omega n}$ " $\Rightarrow X(e^{j\omega}) = X(e^{j(\omega + 2\pi m)})$

• impulse train: multiplication in time domain by sample train

- modulation

? \equiv convolution in freq domain by



Examples - a little hard \therefore computer always ~~DTFT~~ discrete - dense sampling \rightarrow approx continuous $t \dots$?

- Fourier series - build up square wave
- pulse (2nd order system) \rightarrow large DTFT
- sample \rightarrow duplicated spectrum \rightarrow aliasing. \hookrightarrow if f_f too high / ~~spec~~ sampling too coarse

FT properties

- Linear
- time shift $g[n - n_0] \leftrightarrow e^{-j\omega n_0} G(e^{j\omega})$
- freq shift $e^{j\omega_0 n} g[n] \leftrightarrow G(e^{j(\omega - \omega_0)})$ e.g. modulated into a carrier (AM)
- parseval! $\sum_n g[n] h^*[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} G(e^{j\omega}) H^*(e^{j\omega}) d\omega$
e.g. $gg^* = |g|^2$

(from $X(e^{j\omega}) = \sum_n x[n] e^{-j\omega n}$)

Z-transform - mathematical convenience for DT analysis

Define $X(z) = \sum_{-\infty}^{\infty} x[n] z^{-n}$ \rightarrow represent $x[n]$ as polynomial in dummy complex z

Consider $G(z) \cdot H(z)$: polynomial multiplication, groups pairs where exponents add to a constant

$g[n] = \{g_0, g_1, g_2, \dots\}$

ZT

e.g. $(g_0 + g_1 z^{-1} + g_2 z^{-2}) \cdot (h_0 + h_1 z^{-1} + h_2 z^{-2})$

? $= g_0 h_0 + z^{-1}(g_1 h_0 + g_0 h_1) + z^{-2}(g_2 h_0 + g_1 h_1 + g_0 h_2) + z^{-3}(g_2 h_1 + g_1 h_2) + z^{-4} g_2 h_2$

conv

= CONVOLUTION! - subscripts add to n $c/w y[n] = \sum_{m} g[m] h[n-m]$
 $n = n + n - m$

i.e. $g[n] \rightarrow G(z)$

? $g[n] * h[n] \rightarrow G(z) \cdot H(z)$

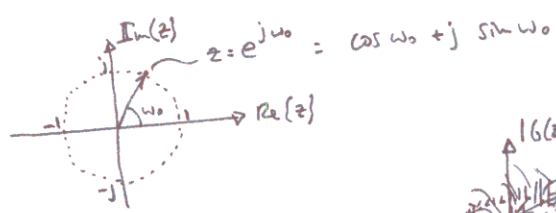
c/w DTFT

if $z = e^{j\omega} \Rightarrow G(z) = G(e^{j\omega}) \triangleq \sum_{-\infty}^{\infty} g[n] e^{-j\omega n} = \text{DTFT}$

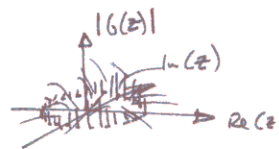
\Rightarrow EVALUATING z transform = unit circle ($z = e^{j\omega}, \omega: -\pi, \pi$) = freq. resp DTFT

DTFT

\rightarrow z-plane (complex)



\rightarrow can consider $G(z)$ as a surface (but complex)



+ DTFT as cylindrical slice on unit circle

ZT convolution property $\Rightarrow y[n] = x[n] * h[n] \Leftrightarrow Y(e^{j\omega}) = X(e^{j\omega}) \cdot H(e^{j\omega})$ \rightarrow FILTERING ($H(e^{j\omega})$ selects part of spectrum)

c/w system function

if have LCCDE system $x \rightarrow y$



from LCCDE

$y[n] = (-a_1) y[n-1] + (-a_2) y[n-2] + b_0 x[n] + b_1 x[n-1] + b_2 x[n-2]$

Z transform:

$Z\{x[n-1]\} = z^{-1} x(z)$

$\Rightarrow \sum_{j=0}^M a_j y[n-j] = \sum_{k=0}^N b_k x[n-k]$ ($a_0 \neq 1$)

ZT $\Rightarrow \sum_i a_i Y(z) z^{-i} = \sum_k b_k X(z) z^{-k}$

$\Rightarrow \frac{Y(z)}{X(z)} = H(z) = \frac{\sum b_k z^{-k}}{\sum a_i z^{-i}} = \frac{b_0 + b_1 z^{-1} + b_2 z^{-2}}{1 + a_1 z^{-1} + a_2 z^{-2}}$

ZT of IR: $H(z) \cdot X(z) = Y(z)$
 $h[n] * x[n] = y[n]$

Rational polynomial z transform.

$H(z)$ system function: rational polynomial in z , easily derived for actual system
 $\rightarrow H(e^{j\omega})$, frequency response - evaluate polynomial on u. circle

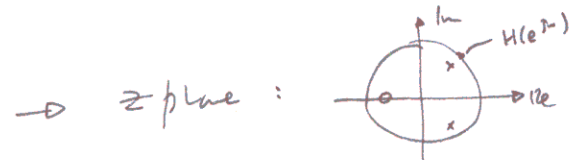
but the points on z plane of interest

$H(z) = 0$: zeros (roots of $\sum b_k z^{-k}$)
 $H(z) = \frac{1}{0} = \infty$: poles (roots of $\sum a_j z^{-j}$)

$$H(z) = \frac{b_0 \prod_k (1 - \xi_k z^{-k})}{\prod_j (1 - \lambda_j z^{-j})}$$

↑
num

Poles/zeros



\Rightarrow as $H(e^{j\omega})$ gets close to a pole, it gets large
zero, it gets small.

i.e. $H(e^{j\omega}) = \frac{b_0 (1 - \xi_1 z^{-1})(1 - \xi_2 z^{-1})}{(1 - \lambda_1 z^{-1})(1 - \lambda_2 z^{-1})}$

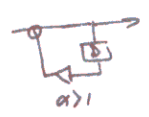
term dominates as $\lambda_2 z^{-1} \rightarrow 1$

eg. $y[n] = 2 \cos \omega_0 y[n-1] + r^2 y[n-2] + x[n]$
 $\Rightarrow H(z) \Rightarrow \lambda_i \Rightarrow z$ -plane \Rightarrow freq. resp.
 (and a zero)

(distance $(z - \lambda_2)$ (mult. top by z^h)
 by z^h)

STABILITY

Can make unstable filters:



Stability

VALIDITY of z transform assumes convergence

$$\sum x[n] z^{-n}$$

i.e. for "causal" ($n > 0$)

$\Rightarrow |z| > r_{\text{some}}$ will ensure convergence

- critical radius defined by poles of $H(z)$

- stable filter \Rightarrow converges for $H(e^{j\omega})$

\Rightarrow u. circle in ROC
 \Rightarrow all poles inside u. circle

(anti-causal would be reverse \rightarrow unv. stable filter "backward" in time?)

- \rightarrow DFT
- (FFT)
- ~~AFET~~
- Circconv
- OLA
- STFT
- pvoc.

DFT

periodic $x(t) \rightarrow$ Fourier Series $\sum_k c_k e^{j\frac{2\pi k}{T}t}$
 \hookrightarrow any $x(t) \rightarrow$ Fourier transform $x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) e^{j\omega t} d\omega$
 \hookrightarrow discrete $x[n] \rightarrow$ DTFT $x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega$
 \hookrightarrow periodic $x[n] \rightarrow$ DFT ...
 (ie. unique values only for $n=0..N-1$ (N deg. of freedom))
 \rightarrow can represent just N transform pts

DFT

$$X[k] = \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi}{N} \cdot k \cdot n}$$

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] W_N^{-nk}$$

$W_N = e^{-j\frac{2\pi}{N}}$

$x[k] = \begin{bmatrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{bmatrix}$

$\begin{bmatrix} 1 & & & \\ & W_N & & \\ & & \ddots & \\ & & & W_N^{-(N-1)} \end{bmatrix} \begin{bmatrix} x[0] \\ x[1] \\ \vdots \\ x[N-1] \end{bmatrix}$

matrix mul. Same shape I/O

"sample" exact values
 $X(e^{j\omega})$ for $x[n]$ zero outside $0..N-1$
 $\omega = \frac{2\pi k}{N}$
 or weights $\delta(\omega - \frac{2\pi k}{N})$
 for DFT of (unbounded) periodic $x[n]$

\rightarrow can actually calculate on a computer!
 (and fast - FFT!)

DFT & Circular Convolution

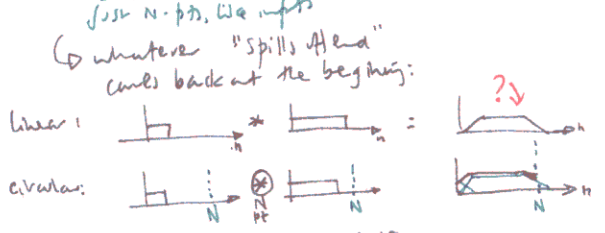
$\therefore x[n], X[k]$ imply periodic signal, multiplying DFTs corresponds to "circular convolution"

c/w linear convolution: $G(e^{j\omega}) \cdot H(e^{j\omega}) \longleftrightarrow \sum_{m=0}^{L+M-1} g[m] h[n-m]$

$G[k] \cdot H[k] \longleftrightarrow \sum_{m=0}^{N-1} g[m] h[(n-m)_N]$

(just N pts, we wrap)

Circ Conv.

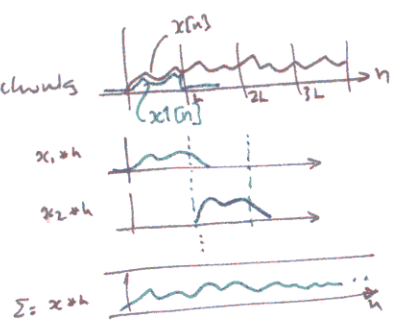


\Rightarrow must zero pad, ensure $N \geq L+M-1$, when using DFT for convolution.

Overlap-Add

by linearity of convolution, if $x[n] = x_1[n] + x_2[n] + x_3[n] + \dots$
 then $x[n] * h[n] = x_1[n] * h[n] + x_2[n] * h[n] + \dots$

e.g. break into short-time chunks



OLA

\Rightarrow can calculate one long convolution by set of shorter ones ("block convolution")
 \Rightarrow if $h[n]$ is short (Mpt), can do each block by DFT if zero pad: $N \geq L+M-1$
 \Rightarrow we FFT \rightarrow fast \rightarrow "BLOCK CONVOLUTION" (practical)

Short-time Fourier Transform

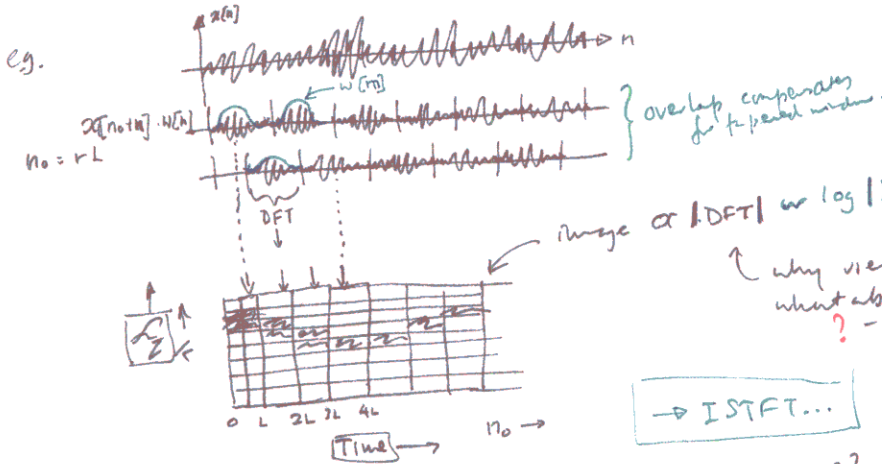
- full-up DTFT gives spectrum over all time
- time varying spectral content?
- break into pieces local to some time "short-time" segments
 FT \rightarrow spectral content "at that time" = STFT

STFT

$$\text{STFT (DFT-based)} \quad X[k, n_0] = \sum_{m=0}^{N-1} x[n_0+m] \cdot w[m] \cdot e^{-j\frac{2\pi k}{N}n}$$

fn of freq and time

- time-selection window
- zero for $n < 0, n \geq N$
- tapered? to avoid spectral ghosts?

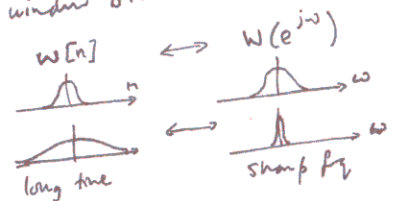


why vertical?
 what about ϕ - freq relative
 ? - EAR - relative ϕ insensitive

Effect of window

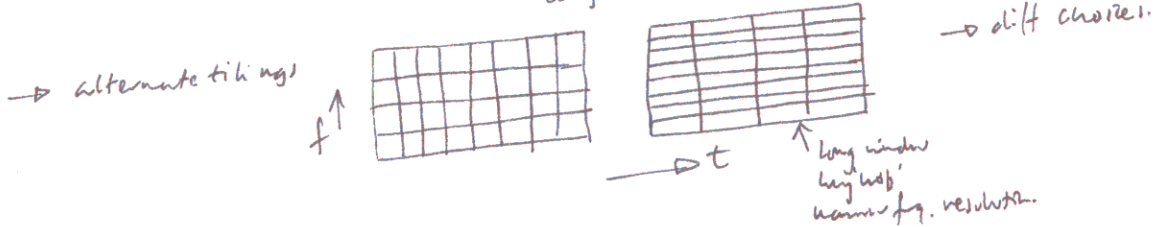
$$X[k, n_0] = \text{DFT} \{ x[n_0+m] \cdot w[m] \} = [e^{j\omega n_0} X(e^{j\omega})] \otimes W(e^{j\omega})$$

- ie. product
- is. Smoothed/blurred by window DFT
- \rightarrow T-F trade-off:



Windowing

plus long time \rightarrow larger hop



Phase Vocoder

- "stretch" a sound by "stretching" the spectrogram
 - time scale change w/o freq/spec change
 - (freq change w/o time scale)
 - (resampling = dilate t-f in inverse)
- easy to do by image - but doesn't specify phase, just |STFT|
 - need phase to reconstruct...
- duplicate ϕ ? → just duplicating little segments - \neq cancellation
- avoid \neq cancellation: maintain ϕ alignment
 - preserve ϕ - relativity constant → can interpolate.
- procedure: analyze to $|X|, \dot{\theta}(t, f)$
 - t scale: just interpolate
 - f scale: scale $\dot{\theta}$
 - then IDFT & overlap-add
 - (overlaps; rewind to avoid edges)

Flan & Golden:

reduce comms b/w i.e. speedup then slowdown
 or use $\frac{1}{2}$ b/w

or transmit $|X[k, n]|, \dot{\theta}[k, n]$ subsampled

what is intrinsic b/w? - analytically hard
 - limited by antennas? (somewhat)

signal → STFT → $|X[k, rL]|, \dot{\theta}_k[k, rL]$

→ interpolate eq by time factor q

$$\begin{aligned} \Rightarrow |Y[k, rL]| &\approx |X[k, q \cdot rL]| \\ &= (1-\alpha) |X[k, pL]| + \alpha |X[k, (p+1)L]| \\ p &= L \cdot q \cdot r \\ \alpha &= q \cdot r - p \end{aligned} \quad \left. \vphantom{\begin{aligned} \Rightarrow |Y[k, rL]| &\approx |X[k, q \cdot rL]| \\ &= (1-\alpha) |X[k, pL]| + \alpha |X[k, (p+1)L]| \\ p &= L \cdot q \cdot r \\ \alpha &= q \cdot r - p \end{aligned}} \right\} \text{linear interp}$$

$\dot{\theta}_y[k, rL] = \dots$

$$\rightarrow Y[k, rL] = |Y[k, rL]| \exp \left\{ j \left[q[k, (r-1)L] + L \cdot \dot{\theta}_y[k, rL] \right] \right\}$$

$$\rightarrow y[rL] = y[rL+m] = \sum_{k=0}^{N-1} Y[k, rL] \cdot e^{j \frac{2\pi k m}{N}} \quad \text{OLA} \quad \left(+ \sum Y[k, (r-1)L] e^{j \frac{2\pi k (m+L)}{N}} \right)$$