

Digital Speech Processing— Lecture 11

Modifications, Filter Bank Design Methods

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Modifications to STFT

- modifications to short-time spectrum can be fixed (non-time varying) or time-varying
- fixed modification (no variability with n)

$$\hat{X}_n(e^{j\omega_k}) = X_n(e^{j\omega_k}) \cdot P(e^{j\omega_k})$$

- assume inverse DFT of $P(e^{j\omega_k})$ exists, then

$$p[n] = \frac{1}{N} \sum_{k=0}^{N-1} P(e^{j\omega_k}) e^{j\omega_k n}$$

- where N is the number of frequencies at which $P(e^{j\omega_k})$ is evaluated

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Fixed Modifications using FBS

- using FBS methods we get

$$\begin{aligned} \hat{y}[n] &= \sum_{k=0}^{N-1} X_n(e^{j\omega_k}) P(e^{j\omega_k}) e^{j\omega_k n} \\ &= \sum_{k=0}^{N-1} \left[\sum_{m=-\infty}^{\infty} w[n-m] x[m] e^{-j\omega_k m} \right] P(e^{j\omega_k}) e^{j\omega_k n} \\ &= \left[\sum_{m=-\infty}^{\infty} w[n-m] x[m] \right] \sum_{k=0}^{N-1} P(e^{j\omega_k}) e^{j\omega_k (n-m)} \\ &= \sum_{m=-\infty}^{\infty} w[n-m] x[m] N p[n-m] \\ \hat{y}[n] &= Nx[n] * [w[n] \cdot p[n]] \end{aligned}$$

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Fixed Modifications using FBS

$$\hat{y}[n] = Nx[n] * [w[n] \cdot p[n]]$$

- $\hat{y}[n] = x[n]$ convolved with $[w[n] \cdot p[n]] \Rightarrow$ equivalent to linear filtering operation on $x[n]$
– ideally want $\hat{y}[n] = x[n] * p[n]$
 \Rightarrow need duration of $p[n] \ll$ duration of $w[n]$
- $p[n]$ is a periodic sequence of period N (sampled in frequency at N frequencies)
- if $w[n]$ is longer than $N \Rightarrow$ repetitive structure in $p[n] \cdot w[n] = h_p[n] \Rightarrow$ need to avoid long filters (IIR) but instead use RW so that
 $h_p[n] = p[n] \cdot w[n] \approx p[n], 0 \leq n \leq N-1$

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Time-Varying Modifications

- represent time-varying modification as

$$\hat{X}_n(e^{j\omega_k}) = X_n(e^{j\omega_k}) \cdot P_n(e^{j\omega_k})$$

- with time-varying IR, $p_n(m)$, defined as

$$p_n[m] = \frac{1}{N} \sum_{k=0}^{N-1} P_n(e^{j\omega_k}) e^{j\omega_k m}$$

- solve for $\hat{y}[n]$ as

$$\begin{aligned} \hat{y}[n] &= \sum_{k=0}^{N-1} X_n(e^{j\omega_k}) P_n(e^{j\omega_k}) e^{j\omega_k n} \\ &= \sum_{k=0}^{N-1} e^{-j\omega_k n} \sum_{m=-\infty}^{\infty} x[n-m] w[m] e^{j\omega_k m} P_n(e^{j\omega_k}) e^{j\omega_k n} \end{aligned}$$

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Time-Varying Modifications

$$\begin{aligned} \hat{y}[n] &= \sum_{k=0}^{N-1} e^{-j\omega_k n} \sum_{m=-\infty}^{\infty} x[n-m] w[m] e^{j\omega_k m} P_n(e^{j\omega_k}) e^{j\omega_k n} \\ &= \sum_{m=-\infty}^{\infty} x[n-m] w[m] \sum_{k=0}^{N-1} P_n(e^{j\omega_k}) e^{j\omega_k m} \\ &= \sum_{m=-\infty}^{\infty} x[n-m] w[m] N p_n[m] \end{aligned}$$

$$\hat{y}[n] = N \sum_{m=-\infty}^{\infty} x[n-m] w[m] p_n[m] = Nx[n] * [p_n[m] w[m]]$$

- modified output is the window $w[m]$ weighted by the modification $p_n[m]$ and convolved with the input $x[n] \Rightarrow$ window 'time limits' effects of modifications and prevents smearing in time
- for FBS spectral modifications lead to convolving the original signal with a time-limited, window weighted version of the time response due to the modification

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Fixed Modifications-OLA

- for a fixed modification we again have

$$\hat{X}_n(e^{j\omega_k}) = X_n(e^{j\omega_k})P(e^{j\omega_k})$$

- the basic OLA method gives

$$y[n] = \sum_{r=-\infty}^{\infty} \left[\frac{1}{N} \sum_{k=0}^{N-1} Y_r(e^{j\omega_k}) e^{j\omega_k n} \right]$$

$$\begin{aligned} \hat{y}[n] &= \sum_{r=-\infty}^{\infty} \frac{1}{N} \sum_{k=0}^{N-1} Y_r(e^{j\omega_k}) P(e^{j\omega_k}) e^{j\omega_k n} \\ &= \frac{1}{N} \sum_{r=-\infty}^{\infty} \sum_{k=0}^{N-1} \sum_{\ell=-\infty}^{\infty} x[\ell] w[rR - \ell] e^{-j\omega_k \ell} P(e^{j\omega_k}) e^{j\omega_k n} \\ &= \frac{1}{N} \sum_{\ell=-\infty}^{\infty} x[\ell] \left[\sum_{k=0}^{N-1} P(e^{j\omega_k}) e^{j\omega_k (n-\ell)} \right] \left[\sum_{r=-\infty}^{\infty} w[rR - \ell] \right] \end{aligned}$$

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Fixed Modifications-OLA

$$\hat{y}[n] = \frac{1}{N} \sum_{\ell=-\infty}^{\infty} x[\ell] \left[\sum_{k=0}^{N-1} P(e^{j\omega_k}) e^{j\omega_k (n-\ell)} \right] \left[\sum_{r=-\infty}^{\infty} w[rR - \ell] \right]$$

$$\hat{y}[n] = \sum_{\ell=-\infty}^{\infty} x[\ell] p[n - \ell] W(e^{j0}) / R = \left(\frac{1}{R} \right) W(e^{j0}) [x[n] * p[n]]$$

- $\hat{y}[n]$ is the convolution of $x[n]$ with the time response of the spectral modification ($p[n]$)—with no window modifications
- need to use larger FFT sizes to account for (prevent) aliasing due to duration of $p[n] \Rightarrow N + N_0 - 1$ where N is the window size, N_0 is the duration of $p[n]$

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Time Varying Modifications-OLA

- for the case of a time-varying modification, using OLA, we obtain

$$\hat{y}[n] = \sum_{r=-\infty}^{\infty} \frac{1}{N} \left[\sum_{k=0}^{N-1} Y_r(e^{j\omega_k}) P_r(e^{j\omega_k}) \right] e^{j\omega_k n}$$

- which after a great deal of manipulation can be put in the form

$$\begin{aligned} &= \frac{1}{N} \sum_{\ell=-\infty}^{\infty} x[\ell] \sum_{r=-\infty}^{\infty} w[rR - \ell] \left[\sum_{k=0}^{N-1} P_r(e^{j\omega_k}) e^{j\omega_k (n-\ell)} \right] \\ &= \sum_{\ell=-\infty}^{\infty} x[\ell] \sum_{r=-\infty}^{\infty} w[rR - \ell] p_r[n - \ell] \end{aligned}$$

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Time Varying Modifications-OLA

- if we make the substitutions $q = n - \ell$, or $\ell = n - q$, we get

$$\hat{y}[n] = \sum_{q=-\infty}^{\infty} x[n - q] \sum_{r=-\infty}^{\infty} p_r[q] w[rR - n + q]$$

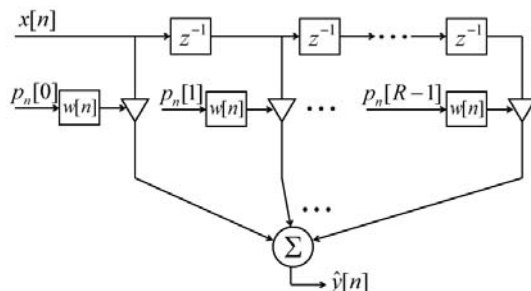
- if we let $\hat{p}[n - q, q] = \hat{p}[m, q] = \sum_{r=-\infty}^{\infty} p_r[q] w[rR - m]$

$$\hat{y}[n] = \sum_{q=-\infty}^{\infty} x[n - q] \hat{p}[n - q, q]$$

- $\hat{p}[m, q]$ is the convolution of $p_r[q]$ and $w(r)$ for all $q \Rightarrow$ each coefficient of the time response due to the time-varying modification is smoothed by the window

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Time Varying Modifications-OLA



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Modifications with FBS and OLA

- OLA—any modification is first “**bandlimited by the window**” then acts as a true convolution on the input signal
- FBS—any modification is first “**time limited by the window**” and could change instantaneously

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Additive Modifications-Quantization

- important to understand effects of additive, signal independent modifications to short-time spectrum as might occur for quantization

$$\tilde{X}_n(e^{j\omega_k}) = X_n(e^{j\omega_k}) + E(e^{j\omega_k})$$

where the additive sequence is

$$e[n] = \sum_{k=0}^{N-1} E(e^{j\omega_k}) e^{j\omega_k n}$$

- for FBS method we get

$$\begin{aligned} \hat{y}[n] &= \sum_{k=0}^{N-1} [X_n(e^{j\omega_k}) + E(e^{j\omega_k})] e^{j\omega_k n} \\ &= y[n] + e[n] \end{aligned}$$

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Additive Modifications-Quantization

- => additive spectral modification yields additive signal
- for OLA method we get

$$\begin{aligned} \hat{y}[n] &= \sum_{r=-\infty}^{\infty} \frac{1}{N} \sum_{k=0}^{N-1} (Y_r(e^{j\omega_k}) + E_r(e^{j\omega_k})) e^{j\omega_k n} \\ &= y[n] + \sum_{r=-\infty}^{\infty} \left[\frac{1}{N} \sum_{k=0}^{N-1} E_r(e^{j\omega_k}) e^{j\omega_k n} \right] \\ &= y[n] + \sum_{r=-\infty}^{\infty} \tilde{e}_r[n] \end{aligned}$$

- => larger additive signal because of overlap between analysis frames
- for HW there is about 4 times larger additive signal

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STFT Summary

- $X_n(e^{j\omega}) = \sum_{m=-\infty}^{\infty} w[n-m] x[m] e^{-j\omega m}$ -- STFT
- $w[n]$ is the analysis window
- $X_n(e^{j\omega})$ can be considered output of bandpass filter, translated to 0 frequency or normal Fourier Transform of the sequence $x[m]w(n-m)$
- use the sampling theorem to define sampling rate in time and frequency domain representations of the window => 2-4 times higher sampling rates than for stationary signal
- two synthesis procedures evolved

$$\text{FBS} \Rightarrow y[n] = \sum_{k=0}^{N-1} X_n(e^{j\omega_k}) e^{j\omega_k n} \text{ -- sum together the bandpass filter outputs}$$

$$\text{OLA} \Rightarrow y[n] = \sum_{r=-\infty}^{\infty} \frac{1}{N} \sum_{k=0}^{N-1} Y_r(e^{j\omega_k}) e^{j\omega_k n}$$

- with $Y_r(e^{j\omega_k}) = X_{n-r}(e^{j\omega_k})$ => windowed segments spaced by R samples in time are overlapped and added together
- FBS and OLA are dual methods

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Digital Filter Bank Designs

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Digital Filter Bank Designs

- composite frequency response approximates flat magnitude and linear phase
- showed ideal conditions for perfect reconstruction
 - choose same $w[n]$ for all channels of filter bank
 - $w[n]$ has zeros equally spaced at N sample intervals

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Uniform Filter Bank Designs

- simple design procedure for uniform bank of filters
 - choose number of filters (this is determined by the desired frequency spacing)
 - design LPF with desired frequency resolution and with zeros appropriately spaced in time
- problems with this ideal case:
 - may want non-uniform filter spacing (model ear characteristics)
 - certain frequency bands (e.g., 0-100 Hz) often omitted (almost no essential speech information)
 - no design procedure for LPF allows simultaneous constraints on both time and frequency response => it is very difficult to design the desired filters

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Modified Filter Bank Structure

- each channel in FB scaled by complex gain $P_k = |P_k| e^{j\phi_k}$

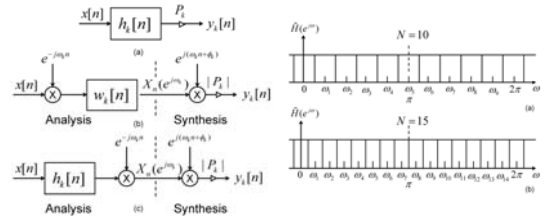
$$y[n] = \sum_{k=0}^{N-1} y_k[n] = \sum_{k=0}^{N-1} P_k X_n(e^{j\omega_k}) e^{j\omega_k n}$$

$$\tilde{h}[n] = \sum_{k=0}^{N-1} P_k h_k[n] = \sum_{k=0}^{N-1} |P_k| w_k[n] e^{j(\omega_k n + \phi_k)}$$

- P_k provides adjustment of gain and phase for each channel to make the overall composite filter bank better approximate the ideal filter bank

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Modified Filter Bank Structure



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FB Designs-Step 1

- Step 1—choose set of analysis frequencies, $\omega_k, 0 \leq k \leq N-1$
assume symmetry in frequency $\Rightarrow \omega_{N-k} = 2\pi - \omega_k$
assume symmetry in time $\Rightarrow w_k[n] = w_{N-k}[n]$
- giving $X_n(e^{j\omega_k}) = X_n^*(e^{j(2\pi-\omega_k)}) = X_n^*(e^{-j\omega_k})$
- can show that for N even we get

$$\tilde{h}[n] = P_0 w_0[n] + \sum_{k=1}^{N/2-1} 2|P_k| w_k[n] \cos(\omega_k n + \phi_k) + P_{N/2} w_{N/2}[n] (-1)^n$$

- and for N odd we get

$$\tilde{h}[n] = P_0 w_0[n] + \sum_{k=1}^{(N-1)/2} 2|P_k| w_k[n] \cos(\omega_k n + \phi_k)$$

- given set of frequencies $\{\omega_k\}$, $0 \leq k \leq N-1$, we need to design set of lowpass filters (or analysis windows) $\{w_k[n]\}$ with the desired frequency resolution and the desired composite response

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Effect of Omitted Channels

- channels are often omitted because they contain virtually no useful information (e.g., channel around $\omega=0$ or around $\omega=\pi$)
- to see the effect of omitting channels from the filter bank, assume uniform frequency spacing, $\omega_k = 2\pi k / N$, and identical analysis windows, $w_k[n] = w[n]$

$$\tilde{h}[n] = w[n] \sum_{k=0}^{N-1} P_k e^{j2\pi kn/N}$$

- defining $p[n] = \sum_{k=0}^{N-1} P_k e^{j2\pi kn/N}$, we get

$$\tilde{h}[n] = w[n] p[n] \text{ where } p[n] \text{ is periodic with period } N.$$

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Effect of Omitted Channels

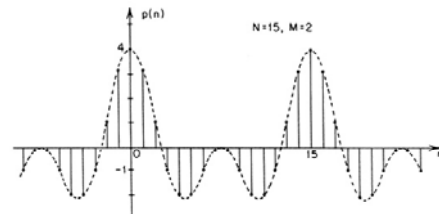
- to omit channels, we set $P_k = 0$ for those channels k that are omitted, e.g., omit channel 0 ($k=0$), and all channels above $\omega_M = 2\pi M / N$, $P_k = 0$ for $k > M$.
- in this case we get

$$p[n] = \sum_{k=1}^M e^{j2\pi kn/N} + \sum_{k=N-M}^{N-1} e^{j2\pi kn/N} = \sum_{k=1}^M (e^{j2\pi kn/N} + e^{-j2\pi kn/N})$$

$$\rho[n] = \frac{\sin\left[\frac{\pi}{N}(2M+1)n\right]}{\sin\left[\frac{\pi}{N}n\right]} - 1$$

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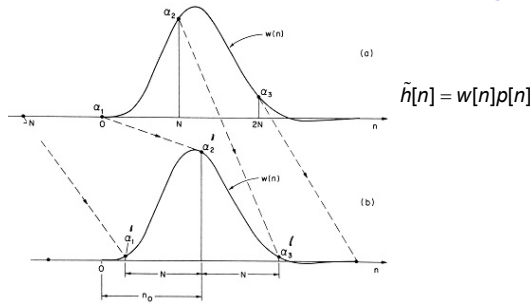
Effect of Omitted Channels



$$\rho[n] = \frac{\sin\left[\frac{\pi}{N}(2M+1)n\right]}{\sin\left[\frac{\pi}{N}n\right]} - 1$$

- $\rho[n]$ periodic with period $N=15$
- pulse amplitudes and widths depend on N and M
- for $M=(N-1)/2$ - all channels included - $\rho[n]$ is an impulse train spaced every $N=15$ samples

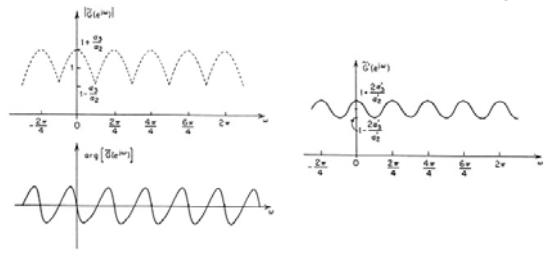
Effects of Imperfect LPF Design



- assume we can delay $w[n]$ by arbitrary amount n_0 giving $w[n - n_0]$
- Case 1: $\tilde{h}[n] = \alpha_2 \delta[n - N] + \alpha_3 \delta[n - 2N]$
- Case 2: $\tilde{h}[n] = \alpha_1 \delta[n - n_0] + \alpha_2 \delta[n - N - n_0] + \alpha_3 \delta[n - 2N - n_0]$

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Effects of Imperfect LPF Design



- to achieve a shift of n_0 samples, let

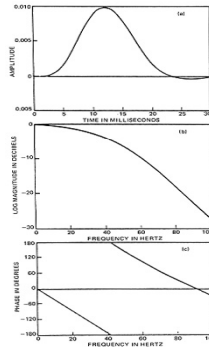
$$P_k = e^{-j2\pi k n_0 / N}, \quad 0 \leq k \leq N-1$$

$$p[n] = \sum_{k=0}^{N-1} e^{j\frac{2\pi}{N}k(n-n_0)} = N \sum_{r=-\infty}^{\infty} \delta[n - n_0 - rN]$$

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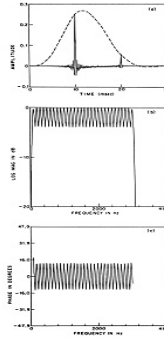
IIR Filter Banks

- assume $F_s = 10,000$ samples/sec
- assume uniform filter bank with spacing 100 Hz $\Rightarrow N = 10,000/100 = 100, \omega_k = 2\pi k/100, 0 \leq k \leq M$
- analysis range of 100 Hz to 3000 Hz $\Rightarrow M = 30$ channels
- Bessel (maximally flat delay) filters derived from 6th order analog filter
- 30 msec duration impulse response \Rightarrow 300 samples at 10,000 Hz SR
- 60 Hz filter bandwidth



IIR Filter Banks

- for $M=30, N=100$, we get $p[n] = \frac{\sin(0.61\pi n)}{\sin(0.01\pi n)} - 1$
- broadening of IR from 30 channels
- since IR duration $> 2N$ samples, there are at least 2 strong peaks in $\tilde{h}[n]$
- significant ripple in composite frequency response (4 dB, 25 degrees)
- using delay to equalize amplitudes of first and third pulses, get better frequency response (0.8 dB ripple, 0.6 degrees phase deviation)



Summary of FB Design

- determine **filter spacing** and **number of filters**
- design filter** to meet frequency selectivity for each channel
- evaluate $w(n)$** and choose delay to minimize ripple
- evaluate **composite response** and iterate solution if response does not meet specs

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FIR Filter Banks

- exact linear phase designs when $w(n) = w(L-1-n)$
- good design methods—windowing, optimal equiripple designs

Window Design Method

- design ideal LPF as

$$W_d(e^{j\omega}) = \begin{cases} e^{-j\omega n_0} & |\omega| \leq \omega_p \\ 0 & \text{otherwise} \end{cases}$$

- where $n_0 = (L-1)/2$ for L -point window

- can solve for ideal IR as

$$w_d[n] = \frac{1}{2\pi} \int_{-\omega_p}^{\omega_p} e^{-j\omega n_0} e^{j\omega n} d\omega = \frac{\sin(\omega_p(n-n_0))}{\pi(n-n_0)}$$

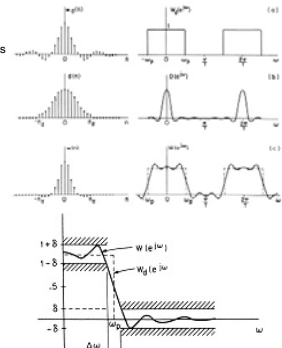
- use finite duration window to truncate $w_d(n)$ giving

$$w[n] = w_d[n] \Delta(n - n_0), \quad -N_0 \leq n \leq N_0$$

$$W(e^{j\omega}) = W_d(e^{j\omega}) \otimes \Delta(e^{j\omega})$$

$$= \frac{1}{2\pi} \int_{-\omega_p}^{\omega_p} W_d(e^{j\theta}) \Delta(e^{j(\omega-\theta)}) d\theta$$

- this leads to a non-zero transition region between the passband and stopband and ripples in both bands



Window Design Properties

1. transition region, $\Delta\omega$, inversely proportional to L
2. $W(e^{j\omega})$ antisymmetric around ω_p
3. peak errors in passband and stopband nearly equal
4. approximation error greatest near ω_p

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Kaiser Window Designs

- Kaiser window designs close to optimal

$$d[n] = \begin{cases} I_0 \left[\alpha \sqrt{1 - (n/n_d)^2} \right] / I_0(\alpha) & |n| \leq n_d \\ 0 & \text{otherwise} \end{cases}$$

- $I_0(\alpha)$ is the zeroth order Bessel function of the first kind
- α is a tradeoff between transition width and peak approximation error

$$L = \frac{-20 \log_{10} \delta - 7.95}{14.36 \Delta f} + 1$$

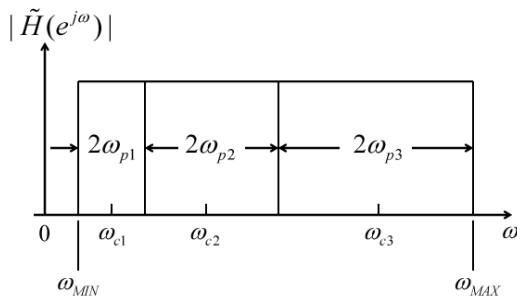
$$\Delta f = \Delta\omega / (2\pi)$$

Procedure for design:

1. δ and Δf chosen $\Rightarrow L$
2. α computed from formula

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Non-Uniform Filter Banks



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Ideal Non-Uniform Designs

- filter bank bandpass filter response of form

$$h_k[n] = P_k W_k[n] e^{j\omega_k n}, \quad 0 \leq k \leq N-1$$

- consider designs using a common window, $d[n]$, for all channels
- the composite frequency response is

$$\tilde{H}(e^{j\omega}) = \sum_{k=0}^{N-1} P_k W_k(e^{j(\omega-\omega_k)})$$

- using the same design window for each channel gives

$$W_k(e^{j(\omega-\omega_k)}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} W_{dk}(e^{j(\theta-\omega_k)}) D(e^{j(\omega-\theta)}) d\theta$$

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Ideal Non-Uniform Designs

- solving for $\tilde{H}(e^{j\omega})$ gives

$$\tilde{H}(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \sum_{k=0}^{N-1} P_k W_{dk}(e^{j(\theta-\omega_k)}) D(e^{j(\omega-\theta)}) d\theta$$

- letting

$$\tilde{H}_d(e^{j\omega}) = \sum_{k=0}^{N-1} P_k W_{dk}(e^{j(\omega-\omega_k)})$$

- gives

$$\tilde{H}(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \tilde{H}_d(e^{j\omega}) D(e^{j(\omega-\theta)}) d\theta$$

- where $\tilde{H}_d(e^{j\omega})$ is the desired composite frequency response

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Ideal Non-Uniform Designs

- assume $P_k = 1, 0 \leq k \leq N-1$, and that the bandwidths and center frequencies of $W_{dk}(e^{j(\omega-\omega_k)})$ are such that the entire frequency range $-\pi \leq \omega \leq \pi$ is covered, giving

$$\tilde{H}_d(e^{j\omega}) = e^{-j\omega n_d}, \quad -\pi \leq \omega \leq \pi$$

$$\tilde{H}(e^{j\omega}) = \frac{1}{2\pi} \int_{-\pi}^{\pi} e^{-j\omega n_d} D(e^{j(\omega-\theta)}) d\theta = d[n_d] e^{-j\omega n_d}$$

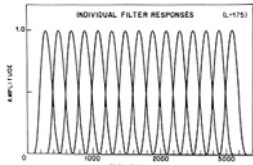
- giving a composite impulse response of

$$\tilde{h}[n] = d[n_d] \delta[n - n_d]$$

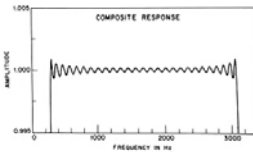
- this says that if the composite desired response is flat with linear phase then the actual composite response of the filter bank, using filters all designed with the same window, is also ideal— independent of how the center frequencies and bandwidths are distributed, and no matter what filter design window is used
- this says that perfect reconstruction is theoretically possible using FIR filters with an arbitrary distribution of center frequencies and bandwidths

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FIR Filter Bank Examples

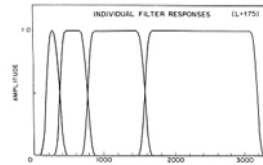


- $F_s = 9.6$ kHz
- want 15 filters to cover range from 200-3200 Hz
- lowpass cutoff frequency is $F_p = \frac{\omega_p}{2\pi T} = \frac{3200 - 200}{2(15)} = 100$ Hz
- center frequencies $F_k = \frac{\omega_k}{2\pi T} = 200k + 100, 1 \leq k \leq 15$
- desire 60 dB attenuation $\Rightarrow \alpha = 5.65$ with 200 Hz transition bands $\Rightarrow L = 175$

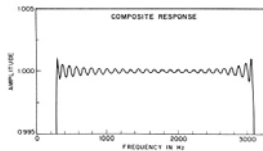


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FIR Filter Bank Examples

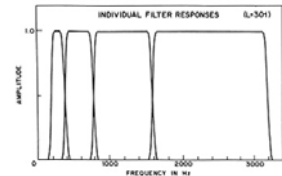


- $F_s = 9.6$ kHz
- want 4 octave band filters \Rightarrow bandwidth doubles with each successive filter \Rightarrow bandwidths of 200, 400, 800 and 1600 Hz, with center frequencies of 300, 600, 1200 and 2400 Hz
- use $L=175$

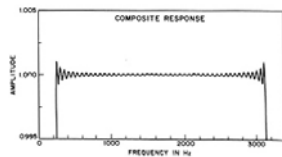


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FIR Filter Bank Examples



- want narrower transition bands
- use larger value of $L=501$



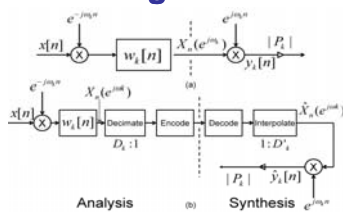
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Applications of STFT

- vocoders \Rightarrow voice coders, code speech at rates much lower than waveform coders
- removal of additive noise
- de-reverberation
- speed-up and slow-down of speech for speed learning, aids for the handicapped

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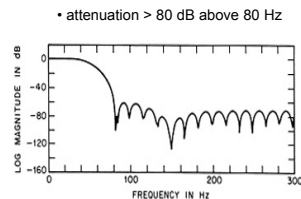
Coding of STFT



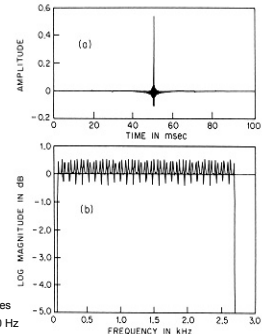
- elements of STFT
 1. set of $\{\omega_k\}$ chosen to cover frequency range of interest
 2. $W_k(n)$ -set of lowpass analysis windows
 3. P_k - set of complex gains to make composite frequency response as close to ideal as possible \Rightarrow goal is to sample STFT at rates lower than $x(n)$

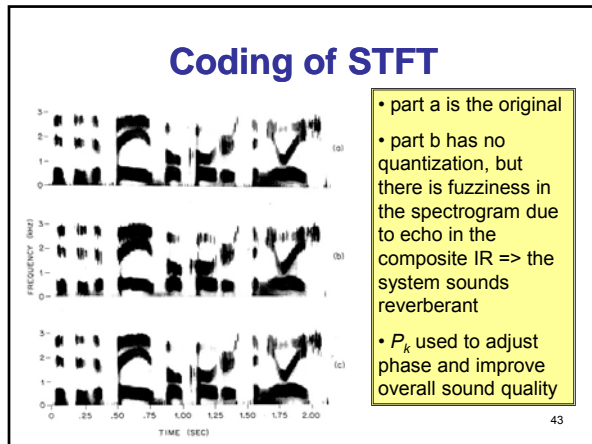
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Coding of STFT

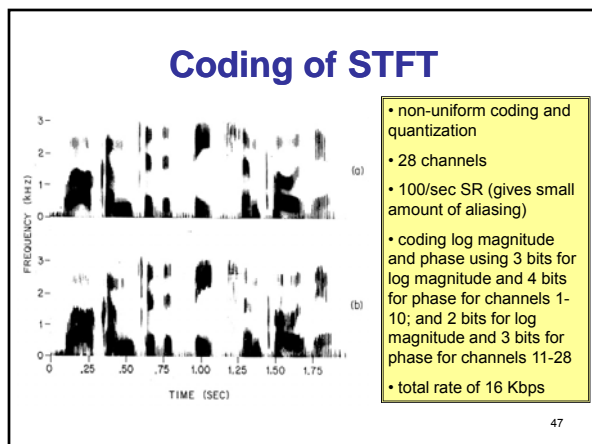
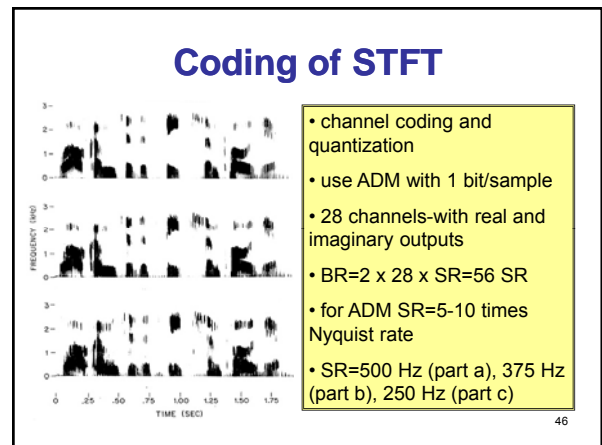
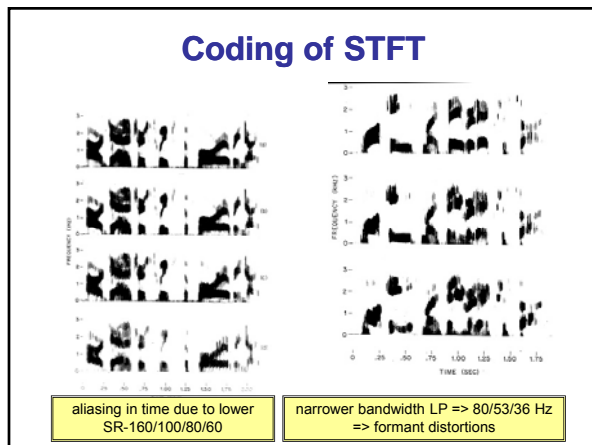


- System example
 - $F_s = 12195$ samples/sec
 - $N=128$
 - $\omega_k = 2\pi k / 128, F_k = 95.273$ kHz
 - $w_k[n] = w[n]$ - linear phase FIR filter, length 731 samples
 - $P_0 = 0, P_k = 0, 28 < k < 100$ - preserved band up to 2690 Hz



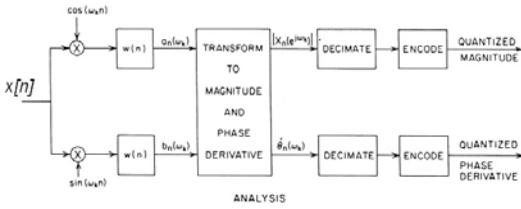


- ### Coding of STFT
- quantizing STFT's
 - BR=sampling rate x bits/sample
 - channel sampling rate depends on bandwidth of lowpass filter, e.g., 80 Hz BW => SR ≥ 160 /sec
 - using lower SR => aliasing in time
 - narrower bandwidth LP => reverberance because of "holes" in spectrum => need to increase number of channels
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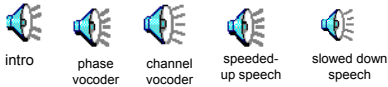


- ### STFT Coding Summary
- sample channels at sufficient rate, quantize to 12 bits/sample => perceptually perfect reproduction of $x(n)$ possible with bit rates of 100 Kbps
 - reduced bit rate => quantize more coarsely, reduce sampling rate => 16 Kbps rates possible
 - remove redundancy of the speech signal by higher order modeling prior to STFT operations
 - cannot use SNR to evaluate STFT since degradations are perceived as modifications of speech quality and intelligibility-not additive noise => spectrograms, subjective quality testing
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Phase Vocoder



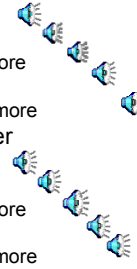
used for speed-up and slow-down of speech



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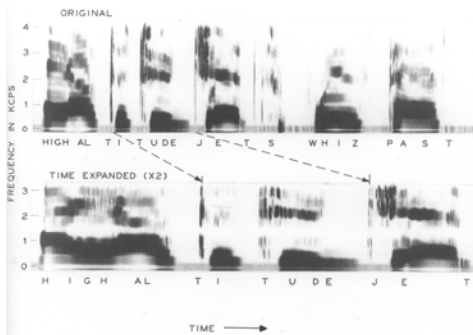
Additional Examples of Rate Changes in Speech

- Male Speaker
 - Original rate
 - Speeded up
 - Speeded up more
 - Slowed down
 - Slowed down more
- Female Speaker
 - Original rate
 - Speeded up
 - Speeded up more
 - Slowed down
 - Slowed down more



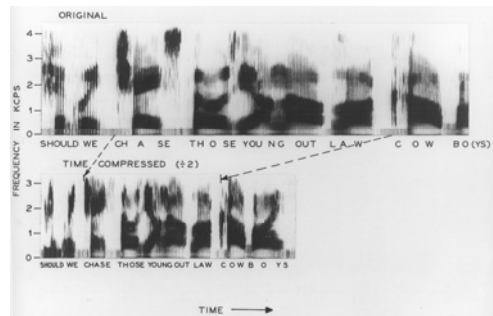
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Phase Vocoder Time Expanded



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Phase Vocoder Time Compressed



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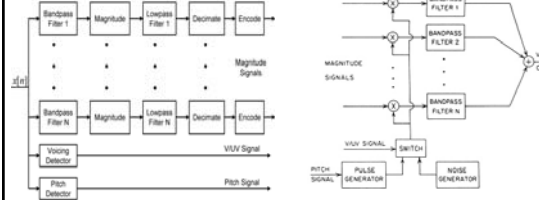
Channel Vocoder



- interpret STFT so that each channel can be thought of as a bandpass filter with center frequency ω_k
 - magnitude of STFT can be approximated by envelope detection on the BPF output (FWR and LPF)
 - **analyzer**-bank of channels; need excitation info (the phase component) => V/UV detector, pitch detector
 - **synthesizer**-channel signal control channel amplitude; excitation signals control detailed structure of output for a given channel; V/UV choice of excitation source
- => **highly reverberant speech because of total lack of control of composite filter bank response**

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Channel Vocoder



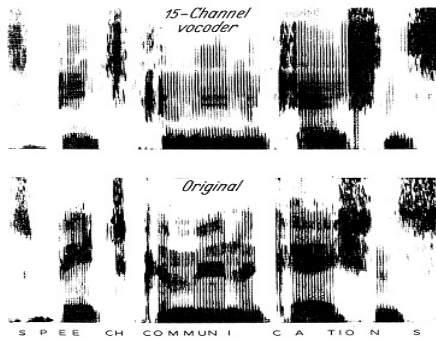
- 1200-9600 bps
- 600 bps for pitch and V/UV
- easy to modify pitch, timing



Channel Vocoder-
2.4 kbps

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Channel Vocoder



S P E E C H C O M M U N I C A T I O N S 55