



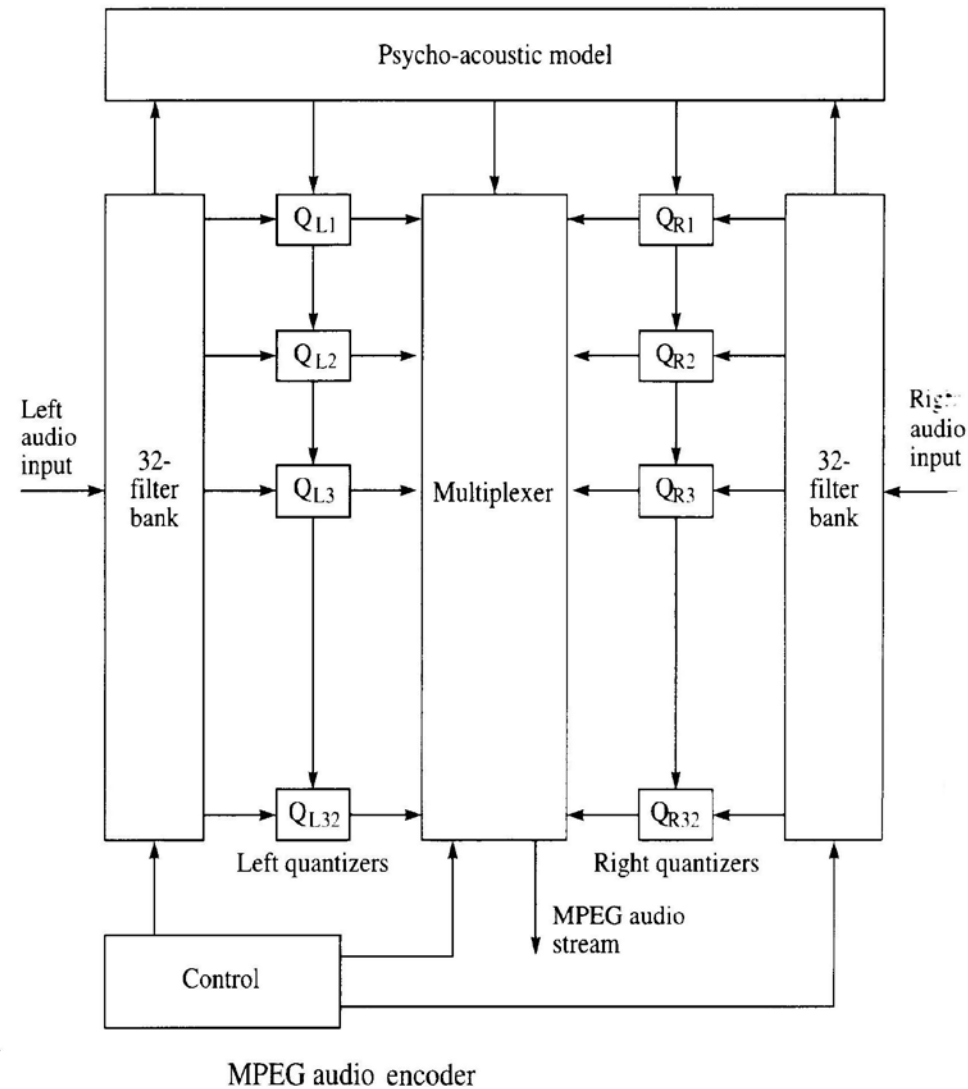
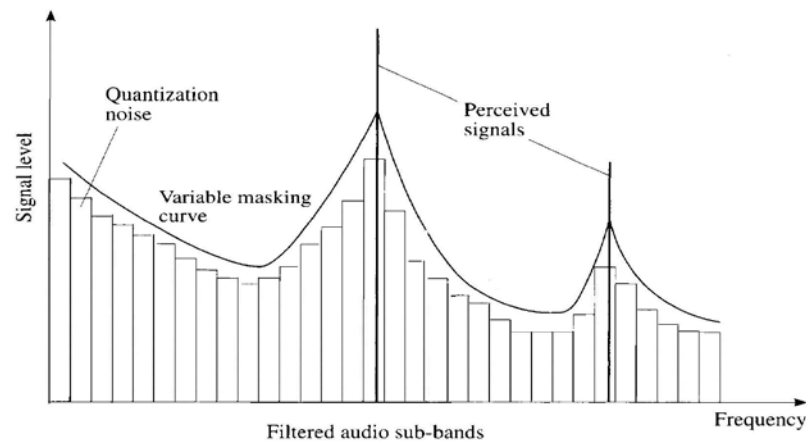
Applications

- Filter Banks
- Quadrature Mirror Filters and Implementation
- Subband Coding of Speech Signals

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Why Filter Banks?

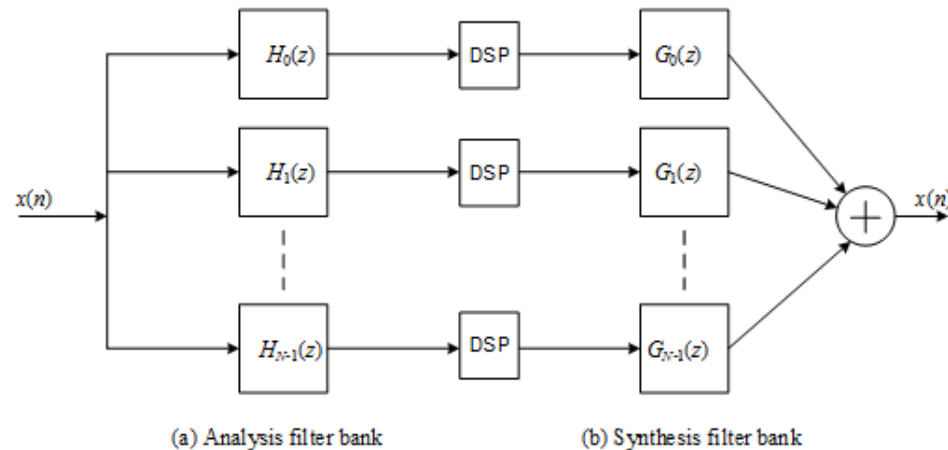
- Different characteristics are for different sub-bands (see figure below)
- Using suitable processing techniques (**i.e., code with no. of bits**) for different sub-bands
- See the encoder for 32 sub-bands



Digital Filter Banks

- An analysis filter bank consists of a set of parallel filters with system function $\{H_i(k)\}$ (Figure 9.1(a)) to split the signal into a number of subbands.

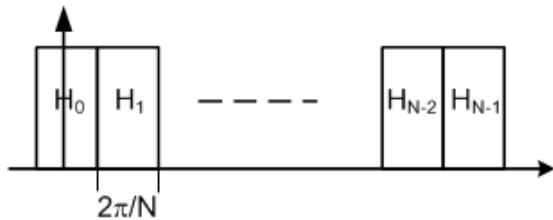
Figure 9.1 A digital filter bank



- A synthesis filter bank consists of a set of parallel filters with system function $\{G_i(k)\}$ (Figure 9.1(b))
- It is important to find an optimum method to divide the signal into different subbands so that they can be combined to restore the original signal **with distortions as small as possible**.
- One important issue is to minimize the effects of **non-ideal transitional bands** of filters.

Digital Filter Banks

- These filter banks are used for performing spectrum analysis and signal synthesis.
- The frequency response characteristics of filter k can be derived from



$$H_k(\omega) = H_0\left(\omega - \frac{2\pi k}{N}\right), \quad k = 0, 1, \dots, N-1 \quad (9.1)$$

where $H_0(\omega)$ is the spectrum of prototype filter required by the applications.

- The time domain impulse response of these filters are expressed by

$$h_k(n) = h_0(n)e^{j2\pi nk/N}, \quad k = 0, 1, \dots, N-1 \quad (9.2)$$

where $h_0(n)$ is the impulse response of prototype filter required by the applications

Digital Filter Banks

- The output of the lowpass filter is relatively narrow in bandwidth, the signal is decimated by a factor of $D \leq N$, which is expressed by

$$X_k(m) = \sum_n h_0(mD - n) e^{-j2\pi nk/N} x(n), \quad k = 0, 1, \dots, N-1; \quad m = 0, 1, \dots$$

where $\{X_k(m)\}$ are the DFT of k th band output.

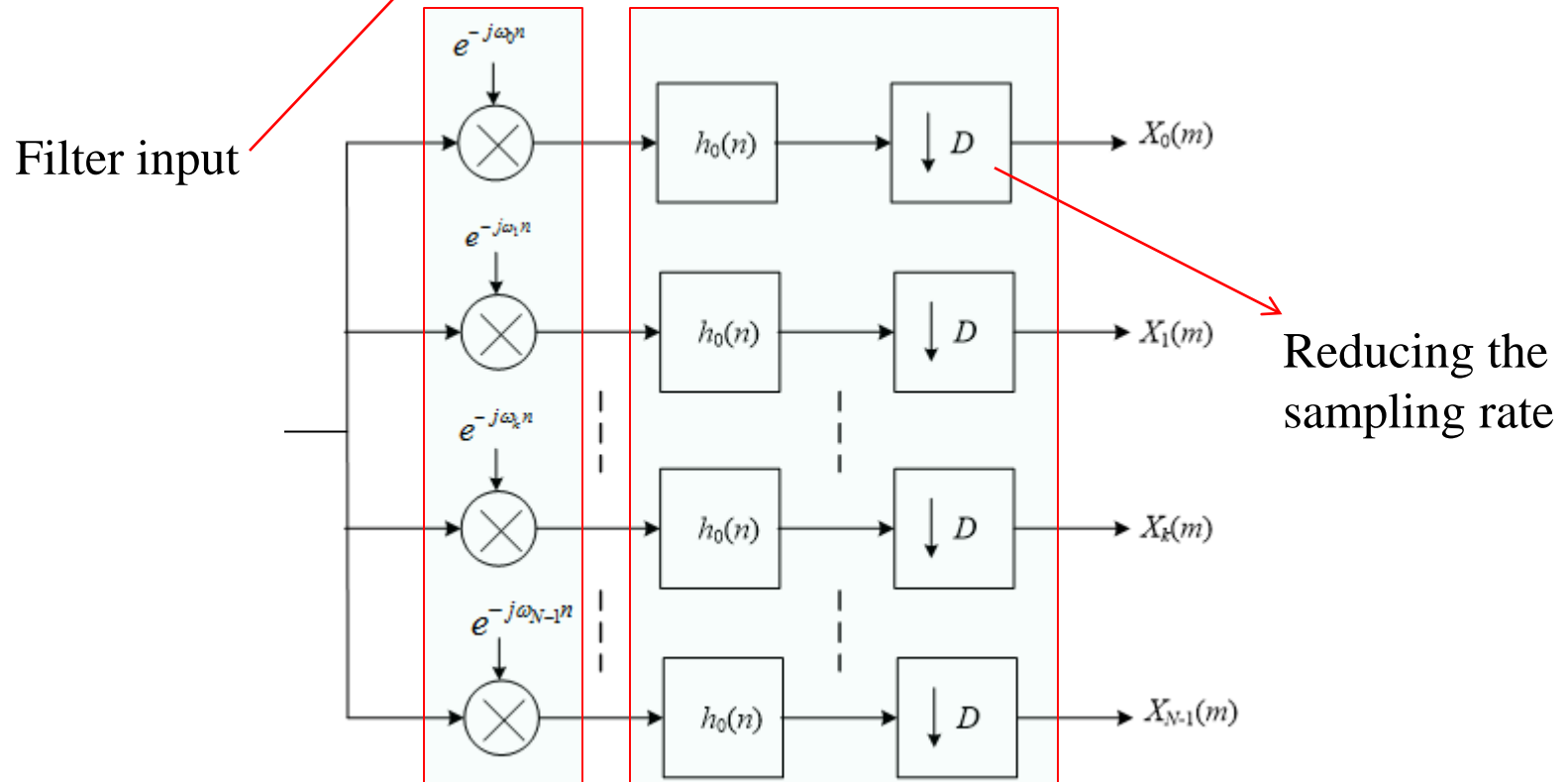


Figure 9.2 (a) Analysis part of a uniform DFT filter bank

Digital Filter Banks

- The synthesis process can be expressed into

$$v(n) = \frac{1}{N} \sum_{k=0}^{N-1} e^{j2\pi nk/N} \left[\sum_m Y_k(m) g_0(n-mD) \right]$$

$$= \sum_m g_0(n-mD) \left[\frac{1}{N} \sum_{k=0}^{N-1} Y_k(m) e^{j2\pi nk/N} \right] = \sum_m g_0(n-mD) y_n(m)$$

Increasing the
sampling rate

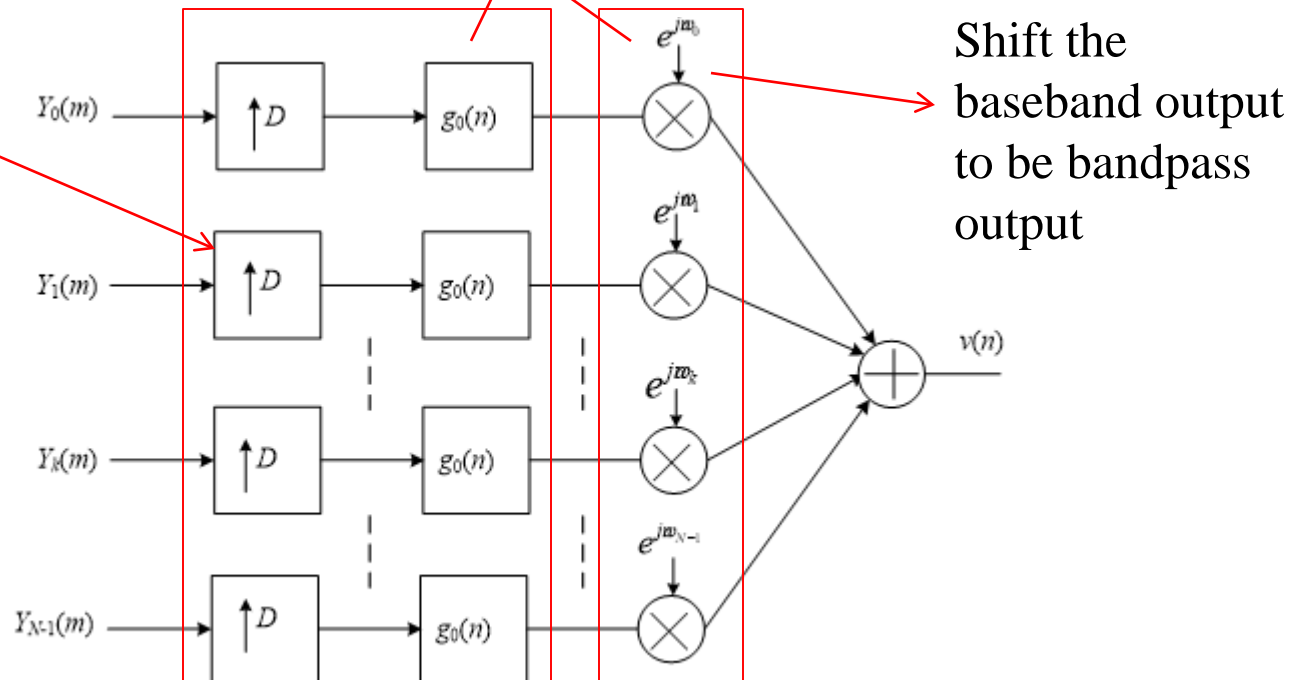
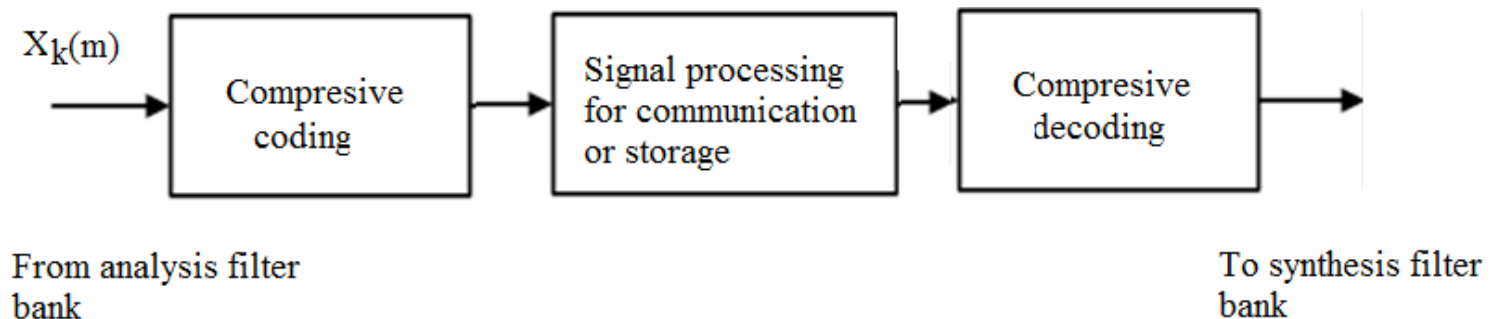


Figure 9.2 (b) Synthesis part of a uniform DFT filter bank, where $\omega_k = 2\pi k/N$

Digital Filter Banks

- In practical application, the outputs of these branches from the analysis process are further processed according to the specific requirement of the applications, for example audio or image coding/decoding and communications.



- You will learn in the subject of audio, image processing and digital communications. Some are done in your handphone and computer.
- The concept of **critically sampling** is that the output sampling frequency is the same as 2 times of the signal bandwidth.
- The filter bank is known as **oversampled** if the output sampling frequency of the filter bank is larger than 2 times of the signal bandwidth.



Digital Filter Banks

- In the above equation $1/N$ is a normalization factor, $\{y_n(m)\}$ represents samples of the inverse DFT sequence corresponding to $\{Y_k(m)\}$, $\{g_0(n)\}$ is the impulse response of the interpolation filter.
- $\{Y_k(m)\}$ is often the modified version of $\{X_k(m)\}$ in many applications
- **These arrangements in previous pages are not computational efficient**
- An alternative realization of the analysis filter bank is described by

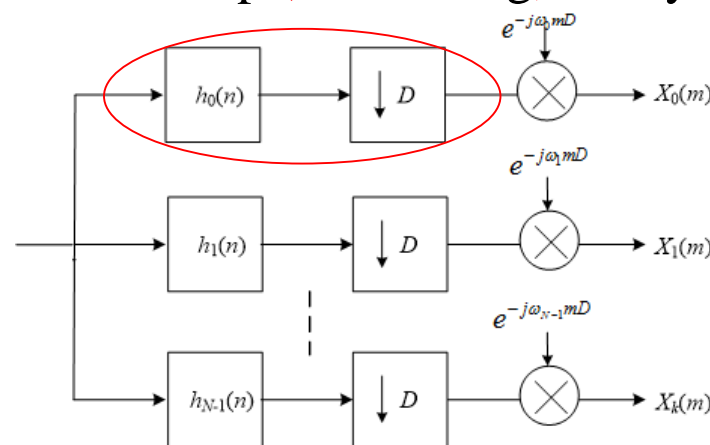
$$X_k(m) = \left[\sum_n x(n) h_0(mD - n) e^{j2\pi k(mD - n)/N} \right] e^{-j2\pi kmD/N} \quad (9.3)$$

where the impulse response of the bandpass filter is given by

$$h_k(n) = h_0(n) e^{j2\pi kn/N},$$

$$k = 0, 1, \dots, N-1$$

- Compared to the structure in Fig. 9.2 (a). here decimation before frequency shifting

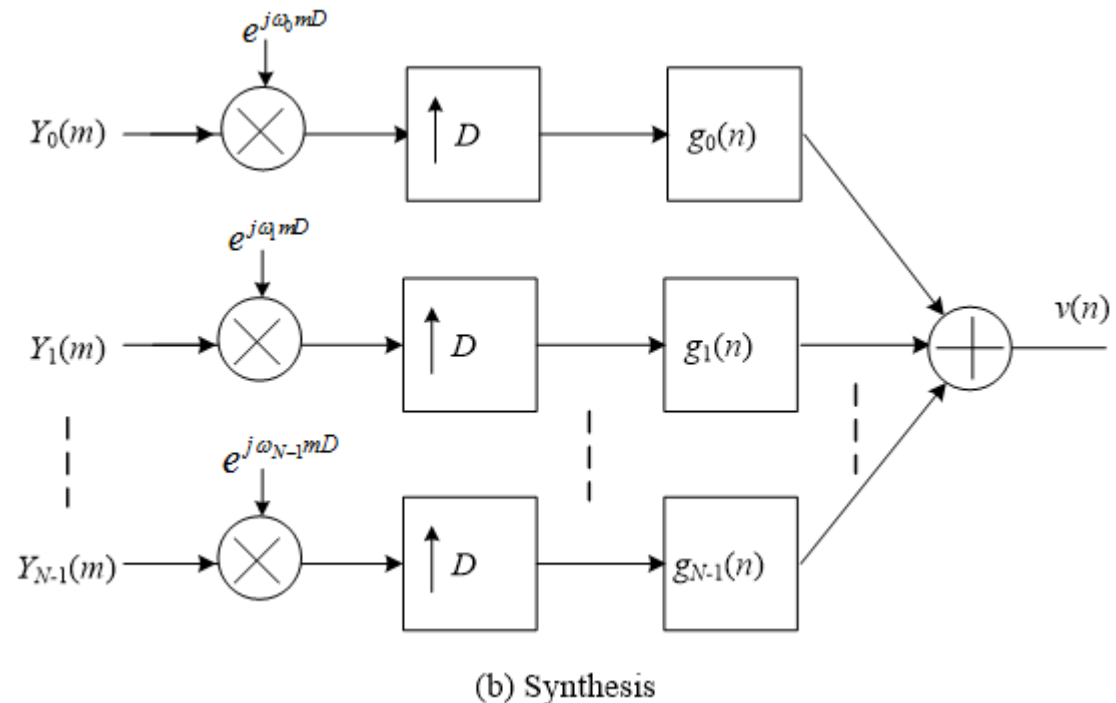


(a) Analysis

Digital Filter Banks

- Similarly, for synthesizing, we have

- Compared to the structure in Fig. 9.2(b), here frequency shifting before interpolation



- The output is

$$v(n) = \frac{1}{N} \sum_{k=0}^{N-1} \left\{ \sum_m [Y_k(m) e^{j2\pi k m D / N}] g_k(n - mD) \right\}$$

$$g_k(n) = g_0(n) e^{j2\pi n k / N}$$



Digital Filter Banks

- For analysis bank filter, the *critically sampled* filter bank is achieved by $N = D$ polyphase filters with impulse response

$$p_k(n) = h_0(nN - k), \quad k = 0, 1, \dots, N-1 \quad (9.4)$$

- The corresponding set of decimated input sequences are

$$x_k(n) = x(nN + k), \quad k = 0, 1, \dots, N-1 \quad (9.5)$$

- Putting (9.4) and (9.5) into (9.3), we have

$$X_k(m) = \left[\sum_n x(n) h_0(mD - n) e^{j2\pi k(mD - n)/N} \right] e^{-j2\pi kmD/N} \quad (9.3)$$

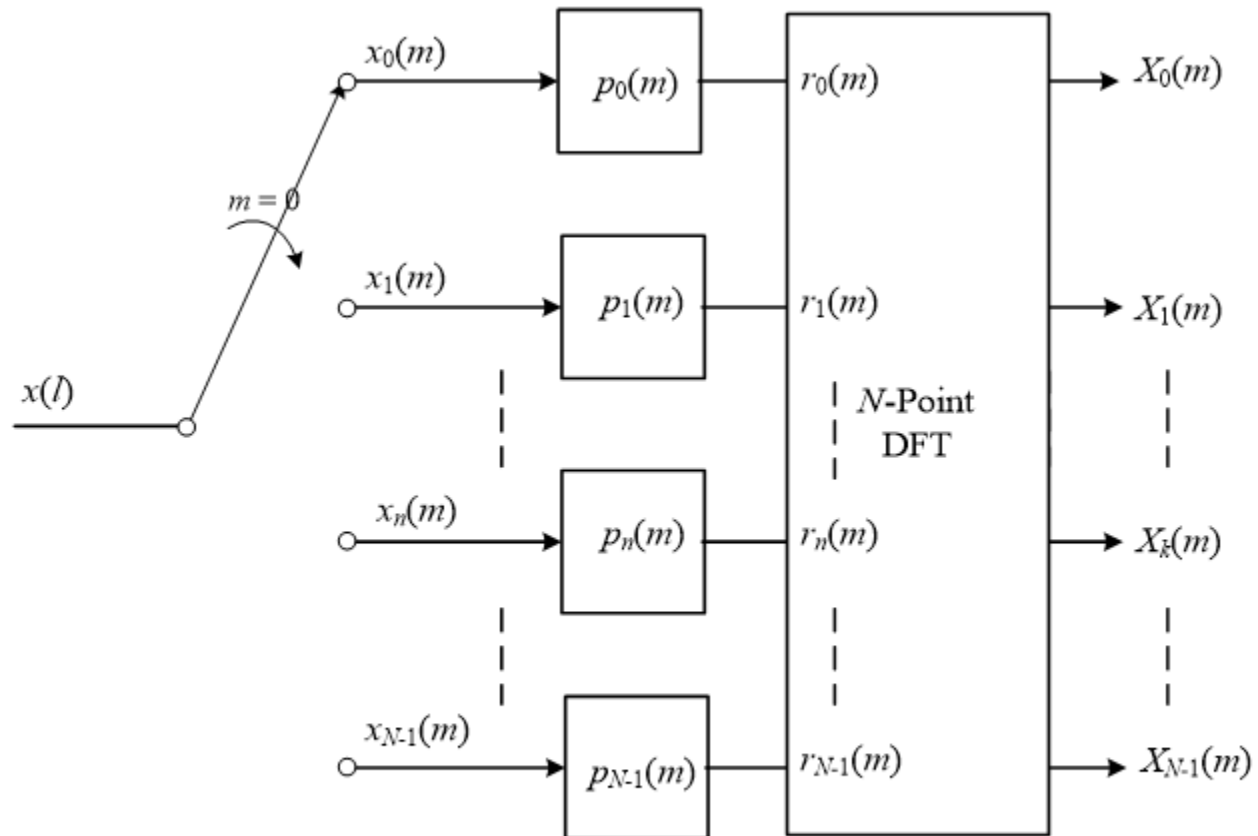
$$X_k(m) = \sum_{n=0}^{N-1} \left[\sum_l p_n(l) x_n(m-l) \right] e^{-j2\pi nk/N}, \quad k = 0, 1, \dots, D-1 \quad (9.6)$$

- The inner sum is the convolution of $\{p_n(l)\}$ with $x_n(l)$ and the outer sum is the N -point DFT of the filter outputs.

Digital Filter Banks

- The filter structure is in the figure below.

$$X_k(m) = \sum_{n=0}^{N-1} \left[\sum_l p_n(l) x_n(m-l) \right] e^{-j2\pi nk/N}$$

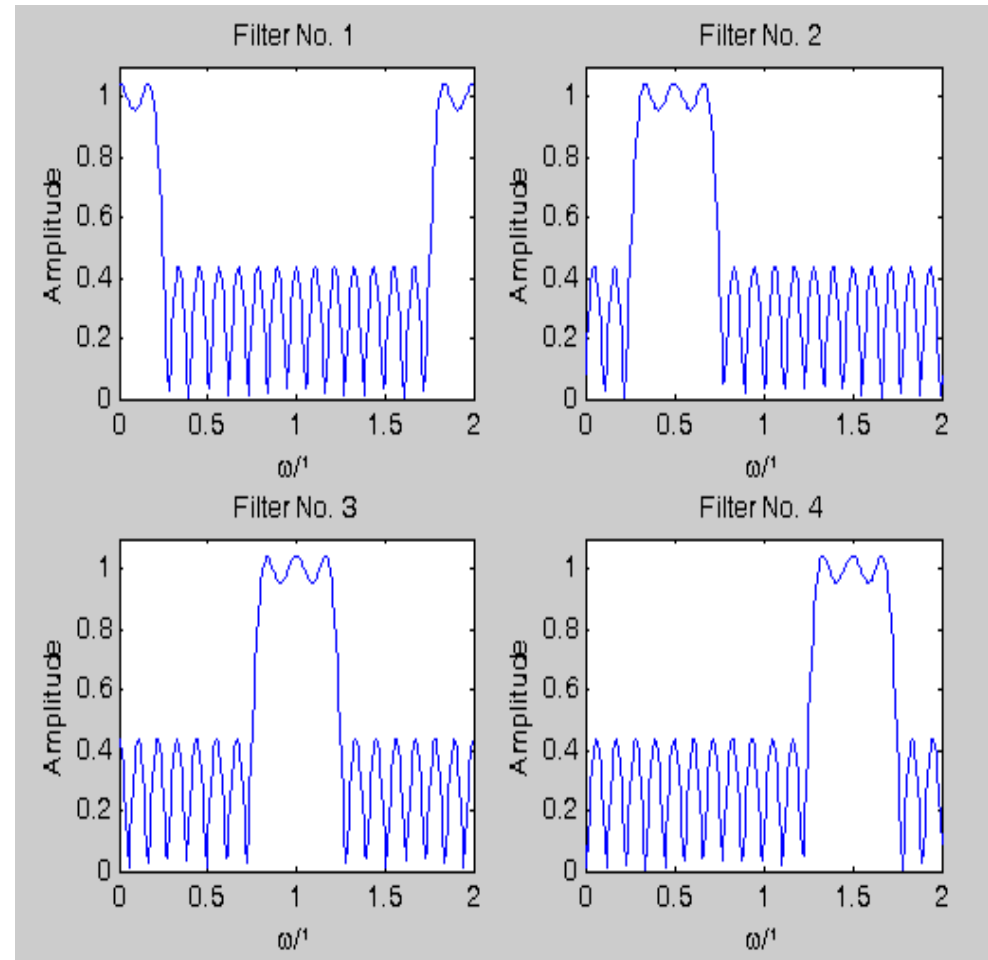


Digital filter bank structure for the computation of equation (9.6)

Digital Filter Banks

Example: MATLAB CODE

```
% Design the prototype LPF
b = remez(20, [0 0.2 0.25 1], [1 1
0 0], [10 1]);
% b contains the filter coefficients
w = 0:2*pi/255:2*pi;
n = 0:20;
for k = 1:4;
c = exp(2*pi*(k-1)*n*i/4) ;
% c contains the transform matrix
FB = b.*c;
% FB contains the DFT outputs
HB(k, :) = freqz(FB, 1, w);
end
```



Responses of polyphase filters for $D = 4$

Digital Filter Banks

- For synthesis bank filter, N polyphase filters are used for interpolation

- The corresponding set of output signals is

$$q_k(n) = g_0(nN + k), \quad k = 0, 1, \dots, N-1$$

- The output of the l th polyphase filter becomes

$$v_k(n) = v(nN + k), \quad k = 0, 1, \dots, N-1 \quad (9.7)$$

- The term in bracket is the N -point IDFT of $\{Y_k(m)\}$, which is denoted as $\{y_l(m)\}$, hence

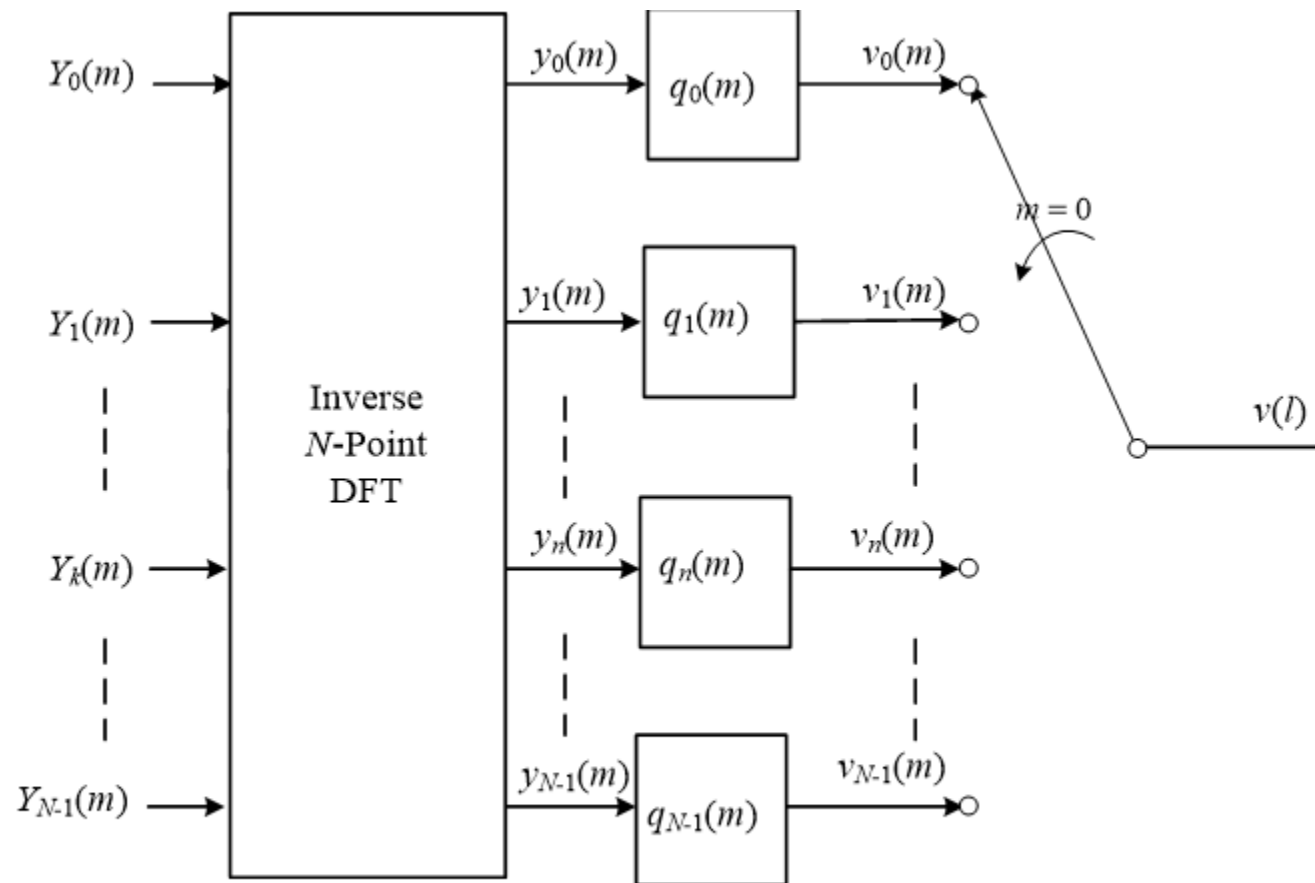
$$v_l(n) = \sum_m q_l(n-m) \left[\frac{1}{N} \sum_{k=0}^{N-1} Y_k(m) e^{j2\pi kl/N} \right], \quad l = 0, 1, \dots, D-1 \quad (9.8)$$

$$v_l(n) = \sum_m q_l(n-m) y_l(m)$$

Digital Filter Banks

- The structure is shown in the figure below

$$v_l(n) = \sum_m q_l(n-m) \left[\frac{1}{N} \sum_{k=0}^{N-1} Y_k(m) e^{j2\pi kl/N} \right]$$

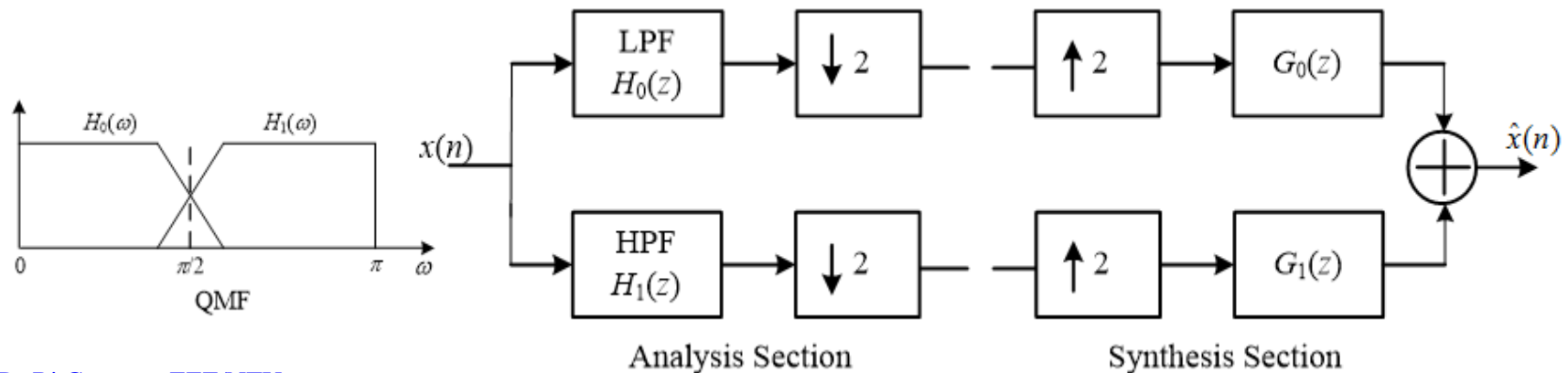




Quadrature Mirror Filters

Quadrature Mirror Filters

- Design of multirate filters is particularly important in achieving good performance. Aliasing and imaging resulting from decimation and interpolation process must be negligible.
- A practical solution to the aliasing problem is to use *quadrature mirror filters* (QMF) (Figure below). The basic building block is the two channel QMF bank and the frequency response of the filters are in the figure below.
- The lowpass and highpass filters in the analysis section have impulse response $h_0(n)$ and $h_1(n)$, respectively.
- Similarly, the lowpass and highpass filters in the synthesis section have impulse response $g_0(n)$ and $g_1(n)$.

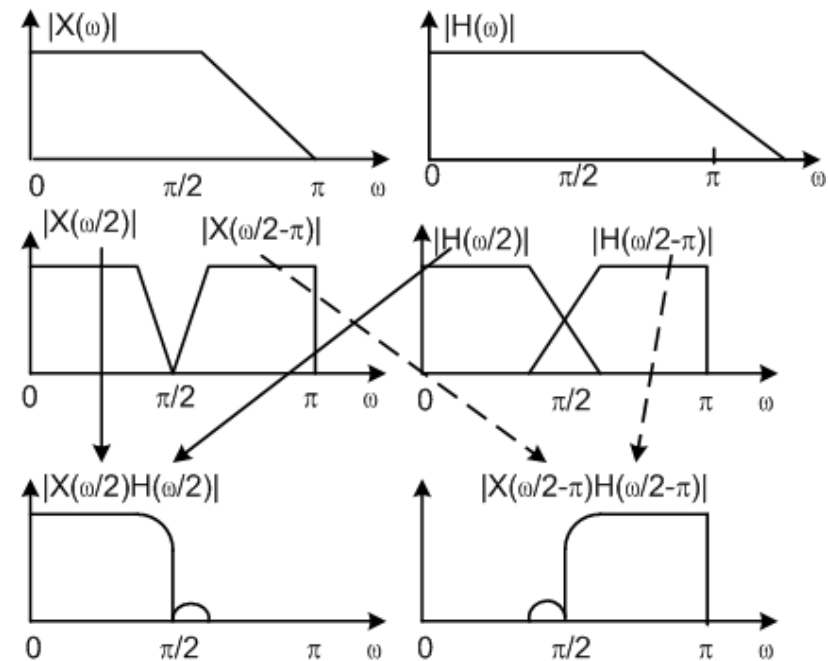


- The Fourier transforms of the signals at the outputs of two decimators are :

$$X_{a0}(\omega) = \frac{1}{2} \left[X\left(\frac{\omega}{2}\right) H_0\left(\frac{\omega}{2}\right) + X\left(\frac{\omega - 2\pi}{2}\right) H_0\left(\frac{\omega - 2\pi}{2}\right) \right]$$

$$X_{a1}(\omega) = \frac{1}{2} \left[X\left(\frac{\omega}{2}\right) H_1\left(\frac{\omega}{2}\right) + X\left(\frac{\omega - 2\pi}{2}\right) H_1\left(\frac{\omega - 2\pi}{2}\right) \right] \quad (10.1)$$

- Note that in the output of each branch, there is some aliasing components, which is the second term of the outputs.
- This is due to the consequence of filter design which has to use a non-zero transitional band.

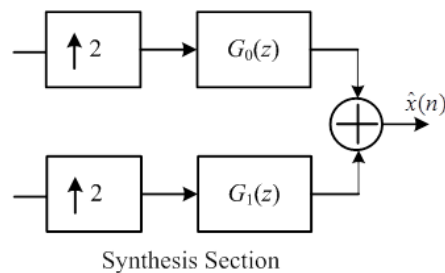




- If $X_{s0}(\omega)$ and $X_{s1}(\omega)$ represent two inputs to the synthesis section, the output is simply

$$\hat{X}(\omega) = X_{s0}(2\omega)G_0(\omega) + X_{s1}(2\omega)G_1(\omega)$$

- Let $X_{a0}(\omega) = X_{s0}(\omega)$ and $X_{a1}(\omega) = X_{s1}(\omega)$. By using (10.1), we have



$$\begin{aligned} \hat{X}(\omega) = & \frac{1}{2}[H_0(\omega)G_0(\omega) + H_1(\omega)G_1(\omega)]X(\omega) \\ & + \frac{1}{2}[H_0(\omega - \pi)G_0(\omega) + H_1(\omega - \pi)G_1(\omega)]X(\omega - \pi) \end{aligned}$$

- The first term is the desired output from the QMF bank. The second term represents the effect of aliasing. To eliminate the aliasing, we need

$$H_0(\omega - \pi)G_0(\omega) + H_1(\omega - \pi)G_1(\omega) = 0$$

Quadrature Mirror Filters

- The condition can be simply satisfied by selecting $G_0(\omega)$ and $G_1(\omega)$ as

$$G_0(\omega) = H_1(\omega - \pi) \quad G_1(\omega) = -H_0(\omega - \pi) \quad (10.5)$$

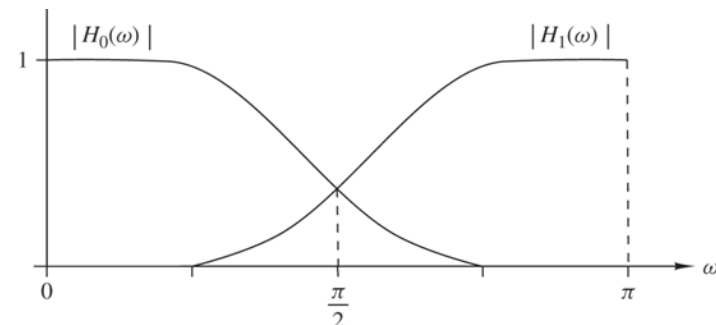
- If assuming that $H_0(\omega)$ is a lowpass filter and $H_1(\omega)$ is a mirror image highpass filter, we may express $H_0(\omega)$ and $H_1(\omega)$ as:

$$H_0(\omega) = H(\omega) \quad H_1(\omega) = H(\omega - \pi)$$

where $H(\omega)$ is the frequency response of a low pass filter. In the time domain, the corresponding relations are:

$$h_0(n) = h(n) \quad h_1(n) = (-1)^n h(n) \quad (10.6)$$

- As a consequence, $H_0(\omega)$ and $H_1(\omega)$ have mirror-image symmetry about the frequency $\omega = \pi/2$, as shown in the Figure.



Mirror image characteristics of the analysis filters $H_0(\omega)$ and $H_1(\omega)$.

Quadrature Mirror Filters

- To be consistent with the constraint (10.5), we select the lowpass filter $G_0(\omega)$ and $G_1(\omega)$ as

$$G_0(\omega) = 2H(\omega) \qquad G_1(\omega) = -2H(\omega - \pi)$$

- In the time domain, the relations become

$$g_0(n) = 2h(n) \qquad g_1(n) = -2(-1)^n h(n)$$

- The scaling factor 2 corresponds to the interpolation factor that used to normalize the overall frequency response of the QMF.
 - With the above arrangements, the component due to aliasing vanishes.
 - Thus the aliasing resulting from decimation in the analysis section is *perfectly canceled* by the image signal spectrum that arise due to interpolation.
 - The two channel QMF behaves as a linear, time-invariant system.



$$X_{a0}(\omega) = \frac{1}{2} \left[X\left(\frac{\omega}{2}\right) H_0\left(\frac{\omega}{2}\right) + X\left(\frac{\omega-2\pi}{2}\right) H_0\left(\frac{\omega-2\pi}{2}\right) \right] \quad 2019$$

Quadrature Mirror Filters

$$X_{a1}(\omega) = \frac{1}{2} \left[X\left(\frac{\omega}{2}\right) H_1\left(\frac{\omega}{2}\right) + X\left(\frac{\omega-2\pi}{2}\right) H_1\left(\frac{\omega-2\pi}{2}\right) \right]$$

- If we put $H_0(\omega)$, $H_1(\omega)$, $G_0(\omega)$ and $G_1(\omega)$ into the first term of (10.1)

$$\hat{X}(\omega) = [H^2(\omega) - H^2(\omega - \pi)]X(\omega)$$

where $H(\omega)$ is a frequency response of a lowpass filter and ideally we need

$$|H^2(\omega) - H^2(\omega - \pi)| = 1 \quad \text{for all } \omega$$

- If $H(\omega)$ has a linear phase, it can be expressed by

$$H(\omega) = H_r(\omega) e^{-j\omega(N-1)/2}$$

where N is the filter length, then

$$H^2(\omega) = H_r^2(\omega) e^{-j\omega(N-1)} = |H_r(\omega)|^2 e^{-j\omega(N-1)}$$

and

$$H^2(\omega - \pi) = H_r^2(\omega - \pi) e^{-j(\omega - \pi)(N-1)} = (-1)^{N-1} |H_r(\omega - \pi)|^2 e^{-j\omega(N-1)}$$

Quadrature Mirror Filters

- Therefore the overall transfer function of the two-channel QMF which uses linear phase FIR filter is

$$\frac{\hat{X}(\omega)}{X(\omega)} = \left[|H(\omega)|^2 - (-1)^{N-1} |H(\omega - \pi)|^2 \right] e^{-j\omega(N-1)} = A(\omega) e^{-j\omega(N-1)}$$

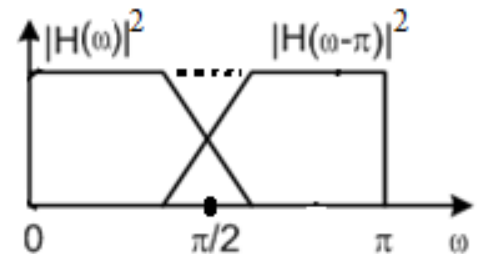
- It has a delay of $N-1$ samples and a magnitude characteristic

$$A(\omega) = |H(\omega)|^2 - (-1)^{N-1} |H(\omega - \pi)|^2$$

- When N is odd, $A(\pi/2) = 0$, which is an undesirable property for a QMF design. When N is even,

$$A(\omega) = |H(\omega)|^2 + |H(\omega - \pi)|^2$$

which avoids the problem of a zero at $\omega = \pi/2$



Quadrature Mirror Filters

- The condition $A(\omega) = 1$ cannot be met because any non-trivial linear-phase FIR filter $H(\omega)$ will introduce some amplitude distortion.
- The amount of distortion can be minimized by optimize the FIR filter coefficients, for example, one minimization method defines

$$J = w \int_{\omega_s}^{\pi} |H(\omega)|^2 d\omega + (1-w) \int_0^{\pi/2} [A(\omega) - 1]^2 d\omega$$

where $0 < w < 1$, is a weighting factor to control the tradeoff between stopband energy and the flatness of $A(\omega)$.

- For two channel QMFs, the use of halfband filters can eliminate completely both amplitude and phase as well as canceling aliasing distortion.

A Simple PR Quadrature Mirror Filter

Example:

- Consider a two-channel QMF bank with analysis/synthesis filters

$$H_0(z) = \frac{1}{\sqrt{2}}(1 + z^{-1}), \quad H_1(z) = \frac{1}{\sqrt{2}}(1 - z^{-1})$$

$$G_0(z) = \frac{1}{\sqrt{2}}(1 + z^{-1}), \quad G_1(z) = \frac{-1}{\sqrt{2}}(1 - z^{-1})$$

- It can be easily verify that

$$H_0(-z)G_0(z) + H_1(-z)G_1(z) = 0$$

$$H_0(z)G_0(z) + H_1(z)G_1(z) = 2z^{-1}$$

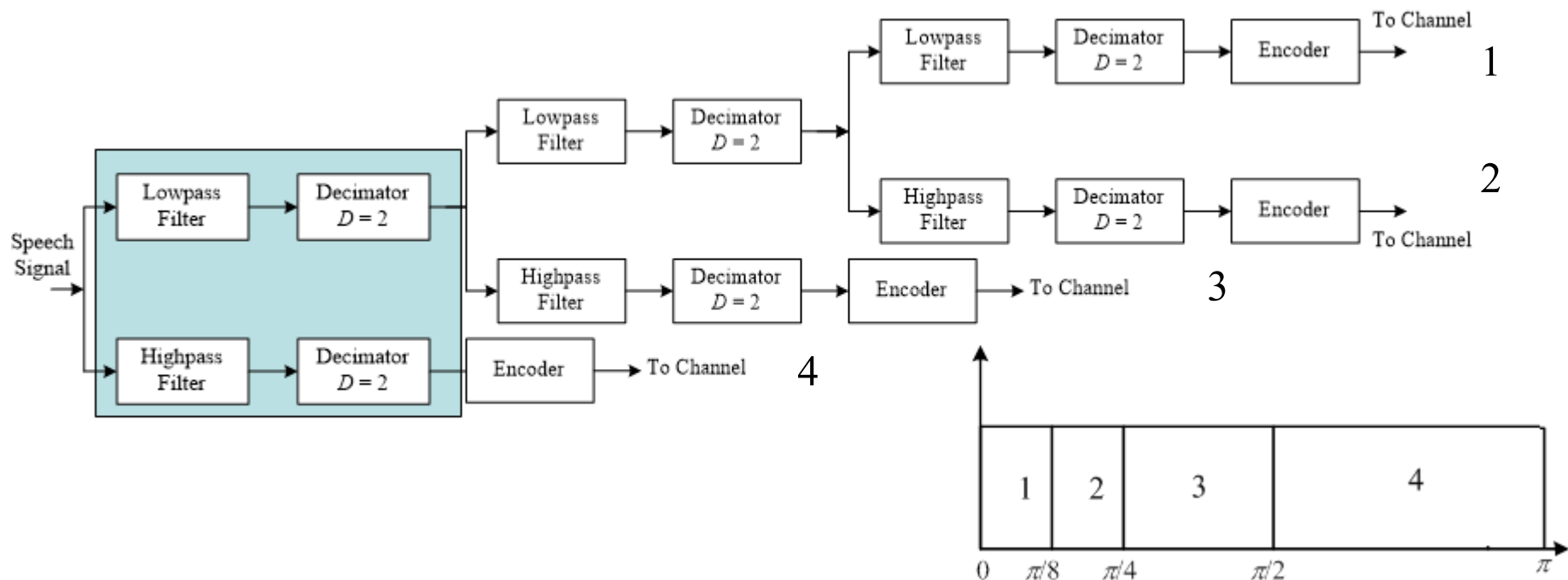
- In the time domain, the outputs of the analysis filters are

$$u_0[n] = \frac{1}{\sqrt{2}}(x[n] + x[n-1]), \quad u_1[n] = \frac{1}{\sqrt{2}}(x[n] - x[n-1])$$

Subband Coding of Speech Signals

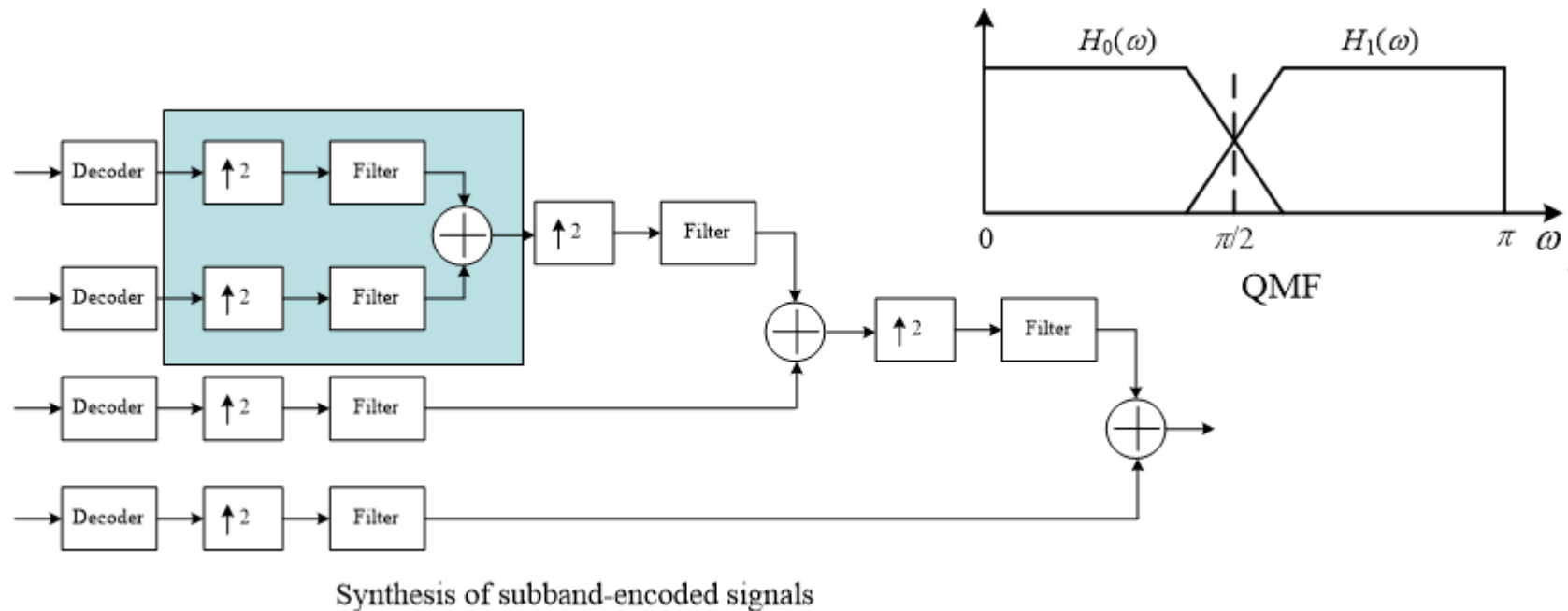
Subband Coding of Speech Signals

- Most of speech energy is contained in the lower frequencies so that better accuracy (using more bits for coding) is required
- The speech signal is divided into several bands and each is encoded separately with different accuracy.

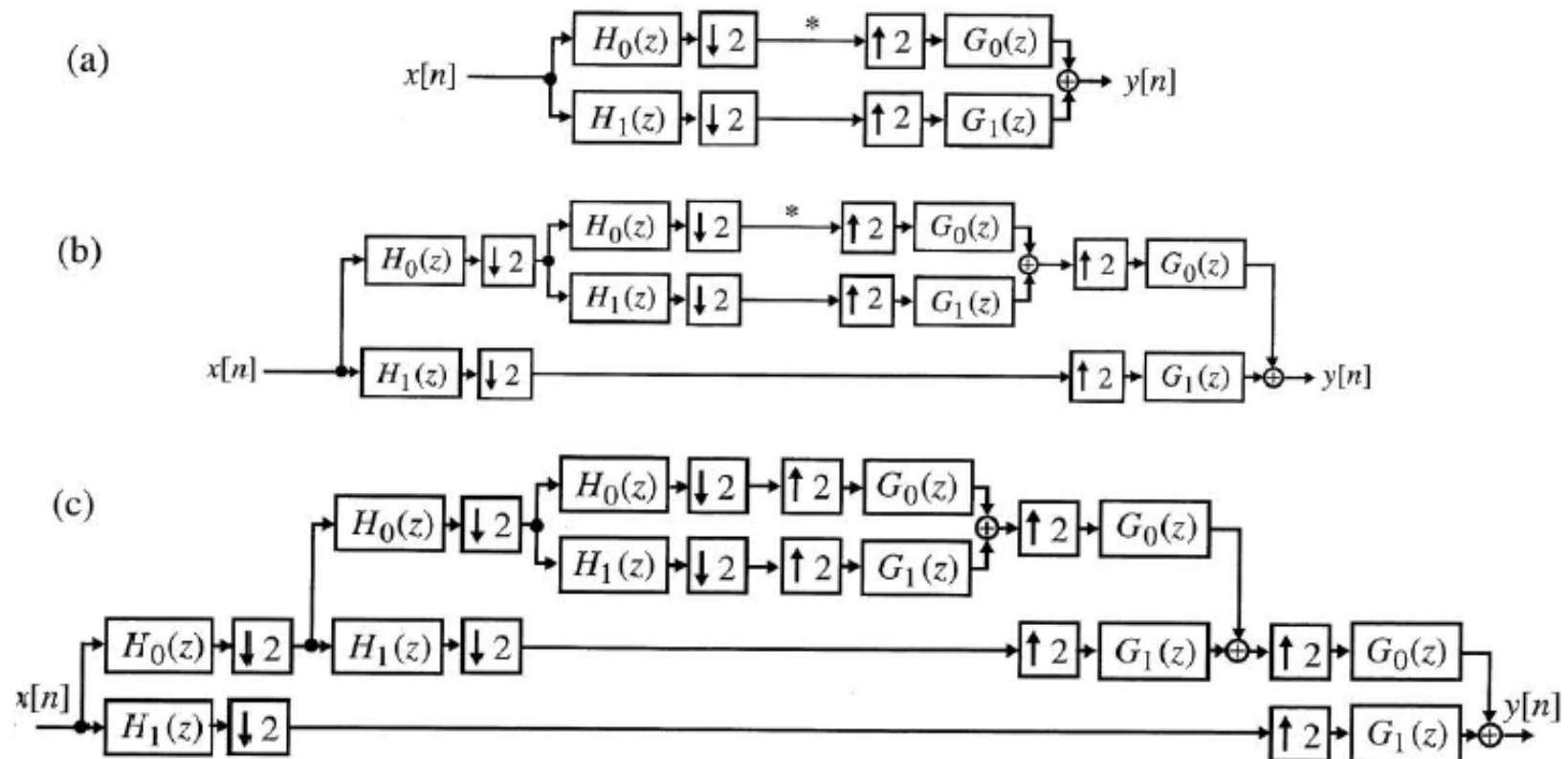


Subband Coding of Speech Signals

- A practical solution is to use quadrature mirror filters (QMF) with the filter characteristics shown in the Figure.



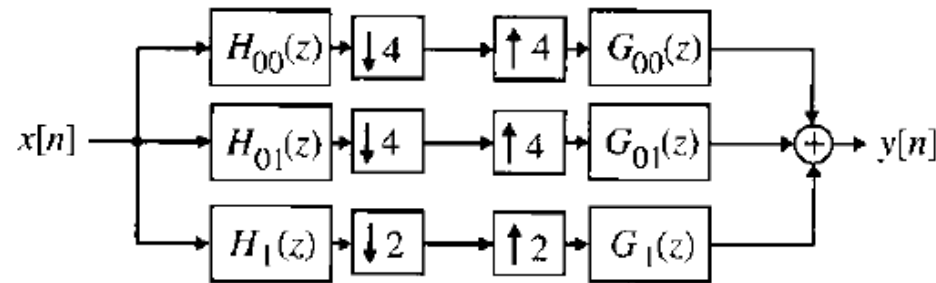
- The synthesis process for subband coding process is a reverse process of encoding process.
- We just need to design the shaded parts for analysis and synthesis process



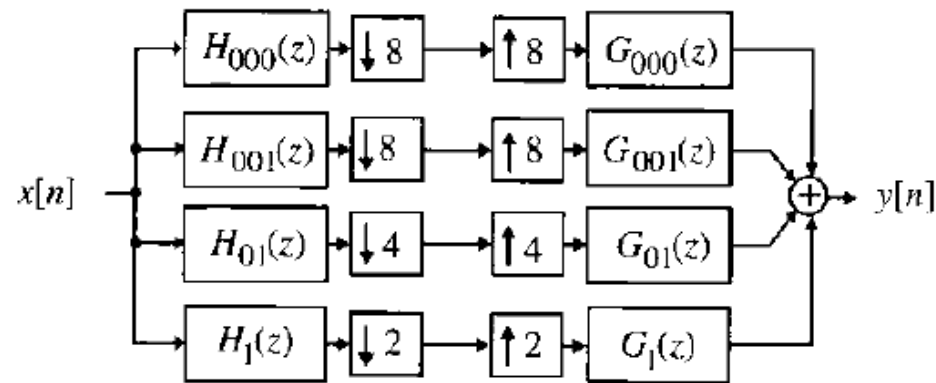
(a) A two-channel QMF bank, (b) a three-channel QMF bank derived from the two-channel QMF bank and (c) a four-channel QMF bank derived from the three-channel QMF bank.

Filter Banks with Unequal Passband Widths

- Inserting another two channel QMF bank in the top channel at the position marked by *;



(a)

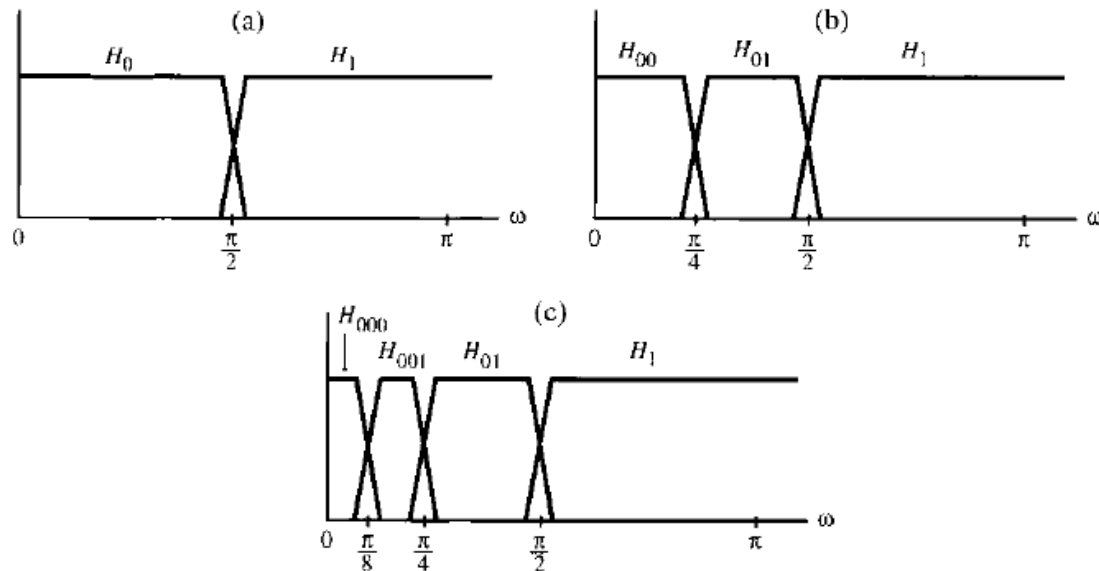


(b)

- Figure shows the development of iteratively deriving a three (a) and four channel (b) unequal passband width filter banks.

Filter Banks with Unequal Passband Widths

- The bandwidths of a two-channel, three-channel and four-channel are shown below



- The equivalent representation of the three-channel filter bank is

$$H_{00}(z) = H_0(z)H_0(z^2), \quad H_{01}(z) = H_0(z)H_1(z^2), \quad H_1(z)$$

$$G_{00}(z) = G_0(z)G_0(z^2), \quad G_{01}(z) = G_0(z)G_1(z^2), \quad G_1(z)$$

Filter Banks with Unequal Passband Widths

- Similarly, the four channel filter banks can be developed and its equivalent representation is

$$H_{000}(z) = H_0(z)H_0(z^2)H_0(z^4), \quad H_{001}(z) = H_0(z)H_0(z^2)H_1(z^4)$$

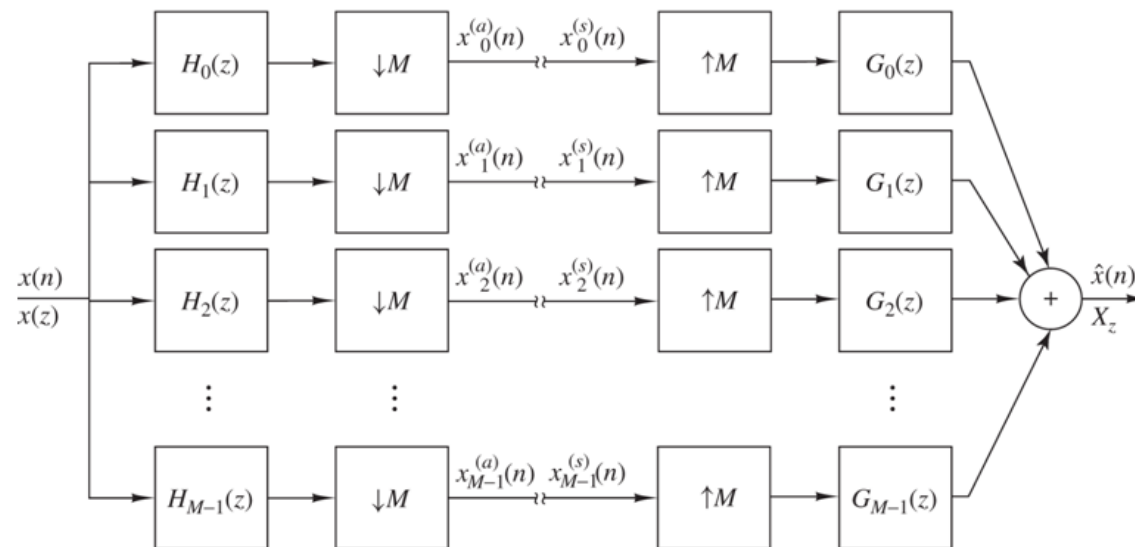
$$H_{01}(z) = H_0(z)H_1(z^2), \quad H_1(z),$$

$$G_{000}(z) = G_0(z)G_0(z^2)G_0(z^4), \quad G_{001}(z) = G_0(z)G_0(z^2)G_1(z^4),$$

$$G_{01}(z) = G_0(z)G_1(z^2), \quad G_1(z),$$

M -Channel QMF Bank

- We can have M -channel QMF bank shown in the Figure
- $x_k^{(a)}(n), 0 \leq k \leq M-1$ are the outputs of analysis filter
- $x_k^{(s)}(n), 0 \leq k \leq M-1$ are the inputs to the synthesis filter



An M -channel QMF bank.

- Although the math derivation is more complicated, the perfect reconstruction concept is still valid.
- Please refer to Proakis' book (4th edition) page 761 for more information.

Polyphase Form of M -Channel QMF Bank

- By using the polyphase filters, the k th filter $H_k(z)$ is represented by

$$H_k(z) = \sum_{m=0}^{M-1} z^{-m} P_{km}(z) \quad 0 \leq k \leq M-1$$

- The equations for the M polyphase filters in matrix form is

$$\mathbf{H}(z) = \mathbf{P}(z^M) \mathbf{a}(z) \quad (11.12.11)$$

where

$$\mathbf{H}(z) = [H_0(z) \ H_1(z) \ \cdots \ H_{M-1}(z)]^T$$

$$\mathbf{a}(z) = [1 \ z^{-1} \ z^{-2} \ \cdots \ z^{-(M-1)}]^T$$

and

$$\mathbf{P}(z) = \begin{bmatrix} P_{00}(z) & P_{01}(z) & \cdots & P_{0(M-1)}(z) \\ P_{10}(z) & P_{11}(z) & \cdots & P_{1(M-1)}(z) \\ \vdots & \vdots & \vdots & \vdots \\ P_{(M-1)0}(z) & P_{(M-1)1}(z) & \cdots & P_{(M-1)(M-1)}(z) \end{bmatrix}$$

Polyphase Form of M -Channel QMF Bank

- Figure 11.12.2 (a) below shows the analysis filter bank and (b) is obtained by applying the noble identity to (a).

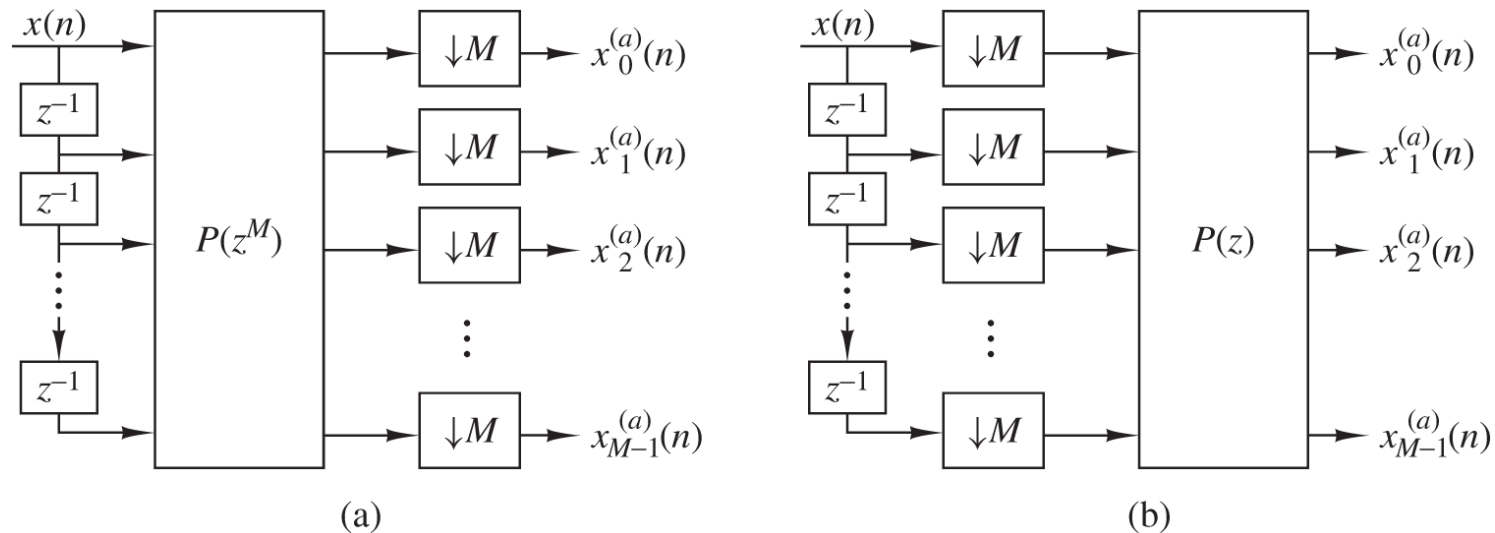


Figure 11.12.2 Polyphase structure of the analysis section of an M -channel QMF bank (a) before and (b) after applying the first noble identity.

- We use a transformed form for the polyphase representation of the synthesis section, which is given

$$G_k(z) = \sum_{m=0}^{M-1} z^{-(M-1-m)} Q_{km}(z^M) \quad 0 \leq k \leq M-1 \quad (11.12.14)$$

Polyphase Form of M -Channel QMF Bank

- In matrix form,

$$\mathbf{G}(z) = z^{-(M-1)} \mathbf{Q}(z^M) \mathbf{a}(z^{-1})$$

$$\mathbf{G}(z) = [G_0(z) \ G_1(z) \ \cdots \ G_{M-1}(z)]^T$$

$$\mathbf{a}(z^{-1}) = [1 \ z^1 \ z^2 \ \cdots \ z^{(M-1)}]^T$$

$$\mathbf{Q}(z) = \begin{bmatrix} Q_{00}(z) & Q_{01}(z) & \cdots & Q_{0M-1}(z) \\ Q_{10}(z) & Q_{11}(z) & \cdots & Q_{1M-1}(z) \\ \vdots & \vdots & \vdots & \vdots \\ Q_{M-10}(z) & Q_{M-11}(z) & \cdots & Q_{M-1M-1}(z) \end{bmatrix}$$

Polyphase Form of M -Channel QMF Bank

- The synthesis section is shown in Figure 11.12.3.

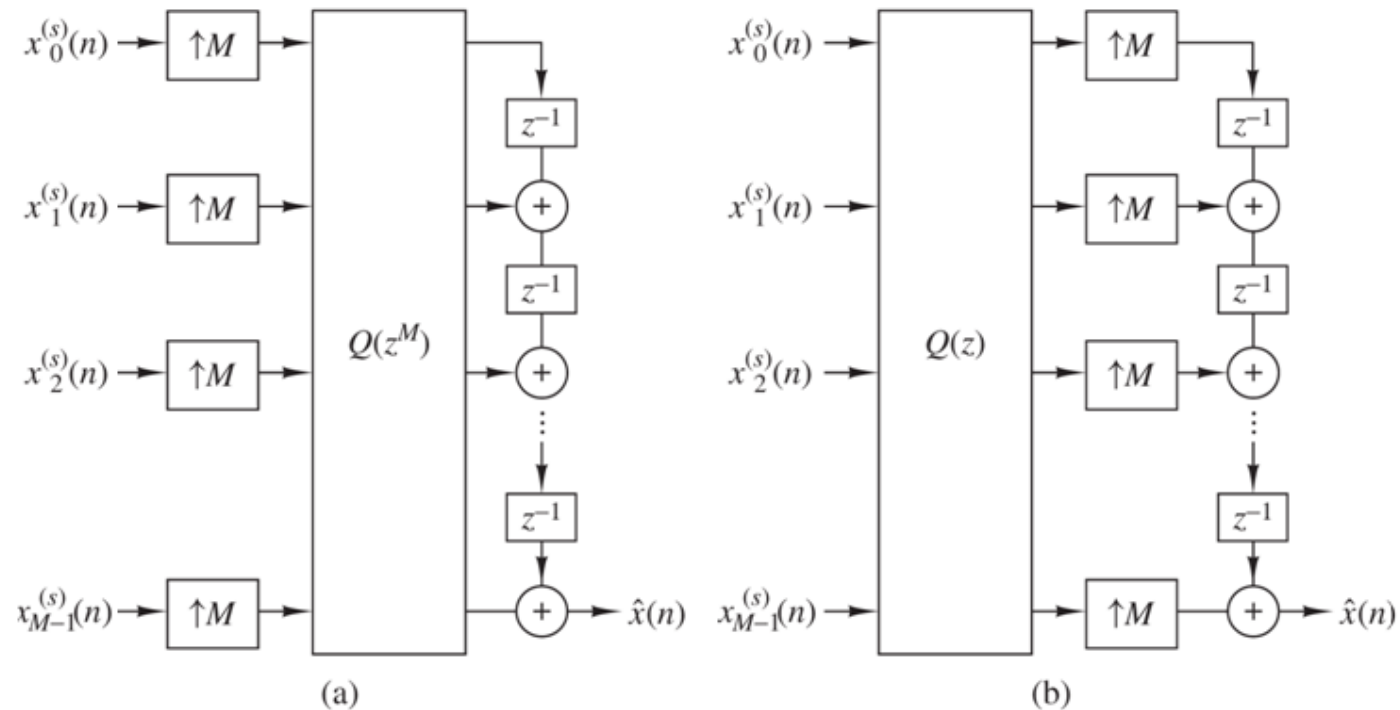


Figure 11.12.3 Polyphase structure of the synthesis section of an M -channel QMF bank (a) before and (b) after applying the first noble identity.

Polyphase Form of M -Channel QMF Bank

- The complete M channel QMF filter bank is given in 11.12.4.

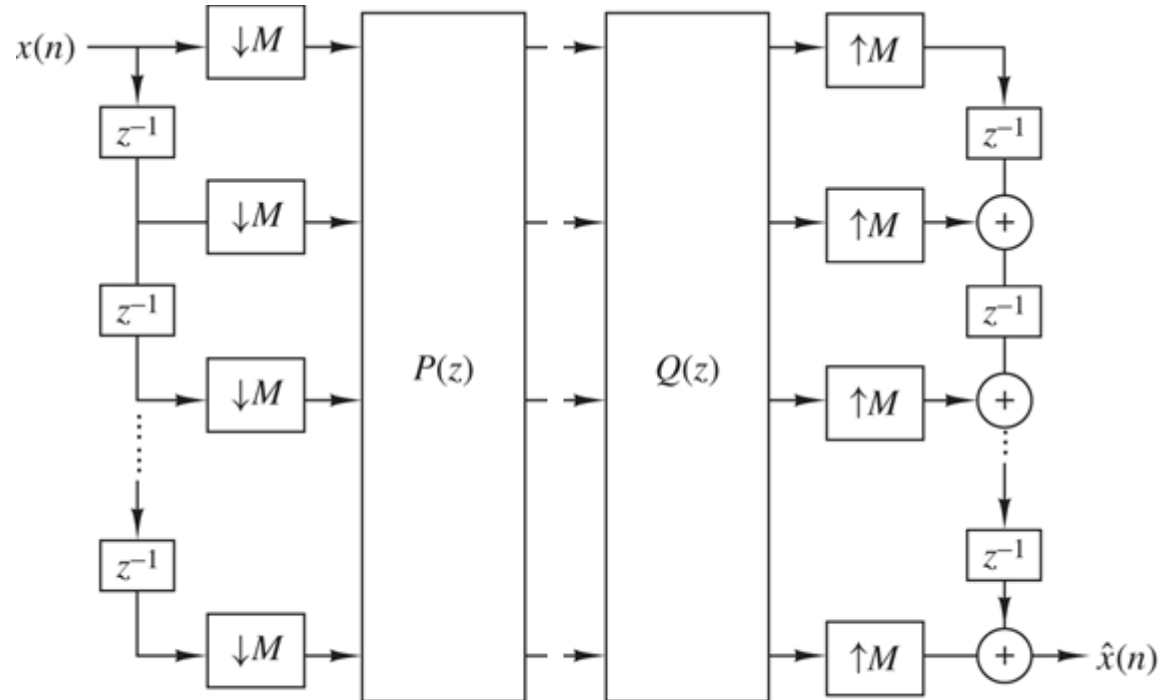


Figure 11.12.4 Polyphase realization of the M -channel QMF bank.

- The perfect reconstruction condition for the M channel QMF filter bank is

$$\mathbf{Q}(z)\mathbf{P}(z) = \mathbf{C}z^{-k}\mathbf{I}$$

- \mathbf{I} is an $M \times M$ identity matrix

Polyphase Form of M -Channel QMF Bank

- If $\mathbf{P}(z)$ is known, we have

$$\mathbf{Q}(z) = \mathbf{C}z^{-k} [\mathbf{P}(z)]^{-1}$$

EXAMPLE:

Suppose the polyphase matrix for a 3-channel perfect reconstruction FIR QMF bank is

$$\mathbf{P}(z^3) = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 3 & 1 \\ 1 & 2 & 1 \end{bmatrix}$$

Determine the analysis and synthesis filters in the QMF bank.

Solution:

The analysis filters are given by (11.12.11) as

$$\begin{bmatrix} H_0(z) \\ H_1(z) \\ H_2(z) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 3 & 1 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ z^{-1} \\ z^{-2} \end{bmatrix} = \begin{bmatrix} 1 + z^{-1} + 2z^{-2} \\ 2 + 3z^{-1} + z^{-2} \\ 1 + 2z^{-1} + z^{-2} \end{bmatrix}$$

Polyphase Form of M -Channel QMF Bank

- The inverse of $\mathbf{P}(z^3)$ is

$$\left[\mathbf{P}(z^3)\right]^{-1} = \frac{1}{2} \begin{bmatrix} 1 & 3 & -5 \\ -1 & -1 & 3 \\ 1 & -1 & 1 \end{bmatrix}$$

- With a scaling factor of 2, we have

$$\mathbf{Q}(z^3) = 2 \left[\mathbf{P}(z^3)\right]^{-1}$$

- By applying (11.12.15), the synthesis filter is obtained by

$$\begin{aligned} \begin{bmatrix} G_0(z) \\ G_1(z) \\ G_2(z) \end{bmatrix} &= z^{-(M-1)} \mathbf{Q}(z^M) \mathbf{a}(z^{-1}) \\ &= z^{-2} \begin{bmatrix} 1 & 3 & -5 \\ -1 & -1 & 3 \\ 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ z^1 \\ z^2 \end{bmatrix} = \begin{bmatrix} -5 + 3z^{-1} + z^{-2} \\ 3 - z^{-1} - z^{-2} \\ 1 - z^{-1} + z^{-2} \end{bmatrix} \end{aligned}$$

Demonstration



Summary

- Quadrature mirror filter in filter banks
- DFT filter bank structure
- M-Channel filter bank structure