Efficient Iterative Design Method for Cosine-Modulated QMF Banks

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Abstract—An iterative algorithm for the design of multichannel cosine-modulated quadrature mirror-image filter (QMF) banks with near-perfect reconstruction is proposed. The objective function is formulated as a quadratic function in each step whose minimum point can be obtained using a closed-form solution. This approach has high design efficiency and leads to filter banks with high stopband attenuation and low aliasing and amplitude distortions. The proposed approach is then extended to the design of multichannel cosine-modulated QMF banks with low reconstruction delays, which are often required, especially in real-time applications. Several design examples are included to demonstrate the proposed algorithms, and some comparisons are made with existing designs.

I. INTRODUCTION

MULTIRATE signal processing and filter bank systems have found high applications in areas such as speech and image coding [1]-[5] during the past 20 years, and extensive research has been carried out in the analysis and design of filter banks. The research effort was first focused on one-dimensional (1-D) two-channel filter banks [6]-[10], and later, it was extended to multichannel filter banks [11]-[19]. Among the various multichannel filter banks, the cosine-modulated QMF bank has two advantages:

1) The analysis and synthesis filters are all obtained from a prototype filter, and consequently, only the prototype filter needs to be designed. This results in high design efficiency.

2) There are polyphase structures that make the implementation of the filter bank very efficient.

Multichannel cosine-modulated QMF banks can be categorized into two groups: near-perfect and perfect-reconstruction banks.

To date, many approaches have been proposed for the design of multichannel cosine-modulated QMF banks. In [12]-[17], filter banks with near-perfect reconstruction are considered. In [18] and [19], methods for the design of perfect-reconstruction filter banks are proposed. With the exception of the method in [15], these approaches use standard constrained or unconstrained optimization techniques to obtain the design, which, in general, are quite time consuming.

In this paper, we propose a new optimization algorithm in which, instead of being minimized directly, the objective function is converted into a quadratic function of the filter coefficients whose minimum point can be obtained analytically. Then, a linear formula is adopted to update the coefficient vector, and the procedure is repeated until a termination criterion is met. The approach reduces the computational complexity significantly and leads to filter banks with high stopband attenuation and low aliasing and amplitude distortions.

Most cosine-modulated QMF banks investigated so far have reconstruction delays of $N - 1$, where $N$ is the filter length. When the filter length is high, which is very common in multichannel filter bank designs, the reconstruction delay will be long, which is undesirable for real-time applications. In [17], some preliminary results have been given through a design example on low-delay cosine-modulated QMF banks. In [20], a class of cosine-modulated filter banks was proposed whose reconstruction delay can take several values. In this paper, a general cosine-modulated QMF bank is considered in which no constraint is imposed on the filter length $N$ and the reconstruction delay $k_d$. By letting $k_d < N - 1$, a low-delay cosine-modulated QMF bank can be obtained. The iterative algorithm proposed is then applied for the design of low-delay filter banks with near-perfect reconstruction.

The organization of the paper is as follows: In Section II, the structure of multichannel cosine-modulated QMF banks is first reviewed, and the objective function is formed. Then, the iterative approach for minimizing the objective function is developed. In Section III, a general cosine-modulated QMF bank is proposed, and an iterative approach similar to that in Section II is applied to design low-delay cosine-modulated QMF banks. Section IV includes several design examples that demonstrate the proposed algorithms and provides performance comparisons with several other existing designs.

II. CONVENTIONAL COSINE-MODULATED QMF BANK

A. Review

In the $M$-channel filter bank shown in Fig. 1, the reconstructed signal is given by [14]

$$
\hat{X}(z) = \frac{1}{M} \sum_{k=0}^{M-1} F_k(z) \sum_{l=0}^{M-1} H_k(W_M^l z) X(W_M^l z)
= \sum_{l=0}^{M-1} T_l(z) X(W_M^l z)
$$

(1)
where
\[
W_M = e^{-2\pi j/M}
\]
\[
T_l(z) = \frac{1}{M} \sum_{k=0}^{M-1} F_k(z) H_k(W_{M}^l z).
\]

In order to cancel aliasing and achieve perfect reconstruction, it is required that
\[
T_l(z) = 0 \quad \text{for } l = 1, 2, \ldots, M-1 \quad (2a)
\]
\[
T_0(z) = z^{-k_d} \quad \text{where } k_d \text{ is an integer}. \quad (2b)
\]

In a cosine-modulated QMF bank, the transfer functions of the analysis and synthesis filters \(H_k(z)\) and \(F_k(z)\) for \(k = 0, 1, \ldots, M-1\), respectively, are obtained by modulating the transfer function \(P(z)\) of a prototype linear-phase lowpass FIR filter \(P\) with bandwidth \(\pi/(2M)\), i.e.

\[
H_k(z) = a_k c_k P(z W_{2M}^{(k+1)/2}) + a_k^* c_k^* P(z W_{2M}^{-(k+1)/2})
\]
\[
F_k(z) = a_k^* c_k P(z W_{2M}^{(k+1)/2}) + a_k c_k^* P(z W_{2M}^{-(k+1)/2})
\]

for \(0 \leq k \leq M-1\), where \(a_k = e^{j \psi_k}, c_k = W_{2M}^{(k+1)/2(N-1)/2}, W_{2M} = e^{-j \pi/M}\), and \(N\) is the length of filter \(P\). The superscript * denotes the complex conjugate operation. It follows that the impulse responses of the analysis and synthesis filters \(H_k\) and \(F_k\) are given by

\[
h_k(n) = 2p(n) \cos \left[ (2k + 1) \frac{\pi}{2M} \left( n - \frac{N-1}{2} \right) + \theta_k \right]
\]
\[
f_k(n) = 2p(n) \cos \left[ (2k + 1) \frac{\pi}{2M} \left( n - \frac{N-1}{2} \right) - \theta_k \right]
\]

for \(0 \leq n \leq N-1\) and \(0 \leq k \leq M-1\), respectively, and \(p(n)\) is the impulse response of the prototype filter \(P\). From (3) and (4), the analysis and synthesis filters are related by

\[
f_k(n) = h_k(N-1-n)
\]
\[
F_k(z) = z^{-(N-1)} \tilde{H}_k(z)
\]

where \(\tilde{H}_k(z) = H_k(z^{-1})\), for \(k = 0, \ldots, M-1\). As shown in [14], \(\theta_k\) can be chosen as

\[
\theta_{k+1} = \theta_k \pm \frac{\pi}{2} \quad \text{for } 0 \leq k < M-1
\]

in order to cancel significant aliasing terms. In order to assure a relatively flat overall amplitude response, it is required that

\[
\theta_0 = \pm \left( \frac{\pi}{4} + \frac{l \pi}{2} \right) \quad (6a)
\]
\[
\theta_{M-1} = \pm \left( \frac{\pi}{4} + \frac{m \pi}{2} \right) \quad (6b)
\]

where \(l\) and \(m\) are integers. A good choice of \(\theta_k\) that satisfies (5) and (6) is

\[
\theta_k = (2k + 1)\pi/4 \quad \text{for } 0 \leq k \leq M-1.
\]

Under these circumstances, the overall transfer function becomes

\[
T_0(z) = \frac{1}{M} \sum_{k=0}^{M-1} H_k(z) F_k(z)
\]
\[
= \frac{z^{-(N-1)}}{M} \sum_{k=0}^{M-1} H_k(z) \tilde{H}_k(z)
\]

and hence

\[
T_0(e^{j\omega}) = \frac{e^{-j\omega(N-1)}}{M} \sum_{k=0}^{M-1} |H_k(e^{j\omega})|^2.
\]

Note that (8) is derived on the basis of the assumption in (2a), and hence, it represents the overall frequency response of the filter bank exactly only if the aliasing components are exactly zero. From (8), it follows that the filter bank has a linear-phase response, which leads to a reconstructed signal with no phase distortion. If the amplitude distortion introduced by the filter bank is eliminated as well, i.e.,

\[
\sum_{k=0}^{M-1} |H_k(e^{j\omega})|^2 = 1
\]

for \(0 \leq \omega \leq \pi\), then perfect reconstruction will be achieved, and the reconstruction delay will be \(N-1\) sampling periods.

**B. New Iterative Method**

It can be shown that \(T_0(e^{j\omega})\) has a period of \(\pi/M\), and if the prototype filter \(P\) has a large stopband attenuation, (8) gives

\[
T_0(e^{j\omega}) \approx \frac{e^{-j\omega(N-1)}}{M} \left[ |P(e^{j(\omega-k\pi/M-\pi/2M)})|^2 + |P(e^{j(\omega-(k+1)\pi/M-\pi/2M)})|^2 \right]
\]

(10)
for \( \omega \) in the range of \((k + \frac{1}{2})\pi/M \leq \omega \leq (k + \frac{3}{2})\pi/M\) [14]. Note that the quantity in the square brackets of (10) is a frequency-shifted version of \(|P(e^{j\omega})|^2 + |P(e^{j\omega-\pi/M})|^2\), and, therefore, if this term can be made constant over \([0, \pi/M]\), then \(|T_0(e^{j\omega})|\) will be constant for all frequencies, that is, amplitude distortion is eliminated. Hence, the design problem can be reduced to the problem of minimizing the objective function

\[
E = E_1 + \alpha E_2
\]  

(11a)

where

\[
E_1 = \int_0^{\pi/M} \left[ |P(e^{j\omega})|^2 + |P(e^{j(\omega-\pi/M)})|^2 - 1 \right]^2 d\omega
\]  

(11b)

deals with the amplitude distortion of the overall transfer function \(T_0(\omega)\)

\[
E_2 = \int_{\omega_s}^{\pi} |P(e^{j\omega})|^2 d\omega
\]  

(11c)

deals with the stopband attenuation of the prototype filter \(P\), where \(\omega_s = \pi/(2M) + \varepsilon\), and \(\varepsilon\) is a positive constant that depends on the required transition width. Parameter \(\alpha\) is a positive weight. Since filter \(P\) is a linear-phase lowpass filter with a symmetrical impulse response, when \(N\) is even, its frequency response can be expressed as [21]

\[
P(e^{j\omega}) = M_p(\omega) e^{-j\omega(N-1)/2}
\]  

(12a)

\[
M_p(\omega) = 2p^T e(\omega)
\]  

(12b)

where \(e(\omega) = [\cos(N-1)/2 \cos(N/2-1)/2]^T\), and \(p = [p(0) p(1) \cdots p(N/2-1)]^T\). Therefore, \(E_1\) and \(E_2\) in (11) can be rewritten as

\[
E_1 = \int_0^{\pi/M} \left[ M_p^2(\omega) + M_p^2(\omega - \pi/M) - 1 \right]^2 d\omega
\]

\[
E_2 = \int_{\omega_s}^{\pi} M_p^2(\omega) d\omega.
\]

Minimizing \(E\) in (11a) using a standard optimization method tends to be time consuming since the objective function is highly nonlinear. Instead of minimizing function \(E\) directly, we adopt an iterative procedure in which the objective function in (11a) is modified to

\[
E' = E_1' + \alpha E_2'
\]  

(13a)

where

\[
E_1' = \sum_{0 \leq \omega \leq \pi/M} \left[ M_p(\omega) \cdot M_p(\omega - \pi/M) - 1 \right]^2
\]

\[
E_2' = \int_{\omega_s}^{\pi} M_p^2(\omega) d\omega
\]

\[
= 4q^T \left[ \int_{\omega_s}^{\pi} e(\omega) e^T(\omega) d\omega \right] q
\]

\[
= q^T U_s q
\]  

(13b)

and

\[
Q(e^{j\omega}) = M_q(\omega) e^{-j\omega(N-1)/2}
\]

\[
M_q(\omega) = 2q^T e(\omega)
\]

(14)

where \(q = [q(0) q(1) \cdots q(N/2-1)]^T\). \(Q(e^{j\omega})\) is the frequency response of a lowpass linear-phase FIR filter \(Q\) with a passband width \(\pi/(2M)\).

At the beginning of the iterative procedure, filter \(P\) is first designed using one of the conventional methods such as the window method, Remez exchange algorithm [21], etc. The summation in (13b) is carried out over a set of sampling points \(\Omega_p = \{\omega_{p1} = 0, \omega_{p2}, \cdots, \omega_{pk} = \pi/M\}\). Then, \(E'\) in (13a) can be formulated as a quadratic function of \(q\) given by

\[
E' = (Uq - d)^T (Uq - d) + \alpha (q^T U_s q)
\]

(14)

where \(d\) is a column vector with each entry being a 1, and

\[
U = H(\Omega_p)U(\Omega_p) + H\left(\Omega_p - \frac{\pi}{M}\right)U(\Omega_p - \frac{\pi}{M})
\]

(15a)

\[
H(\Omega_p) = \text{diag}(M_p(\omega_{p1}), \cdots, M_p(\omega_{pk}))
\]

(15b)

\[
U(\Omega_p) = \begin{bmatrix}
\cos(N-1)/2 & \cdots & \cos\omega_{p1}/2 \\
\cdots & \cdots & \cdots \\
\cos(N-1)/2 & \cdots & \cos\omega_{pk}/2
\end{bmatrix}
\]

(15c)

The \((i,j)\)th entry of \(U_s\) is given by

\[
u_{ij} = 4 \int_{\omega_s}^{\pi} \cos \left[ \left( i - 1 - \frac{N-1}{2} \right) \omega \right]
\]

\[
\cdot \cos \left[ \left( j - 1 - \frac{N-1}{2} \right) \omega \right]
\]

\[
\cdot \frac{\pi - \omega_s}{2} \cdot \frac{\sin \left( (2i - N - 1)\omega_s \right)}{2(2i - N - 1)}
\]

\[
\cdot \frac{\sin \left( (i - j)\omega_s \right)}{2(i - j)} \quad \text{for } i = j
\]

\[
\cdot \frac{\sin \left( (i + j - N - 1)\omega_s \right)}{2(i + j - N - 1)}
\]

\[
\cdot \frac{\sin \left( (i + j - N - 1)\omega_s \right)}{2(i + j - N - 1)}
\]

(15d)

for \(i, j = 1, 2, \cdots, N/2\). Since \(U^T U + \alpha U_s\) is positive definite, \(E'\) has a global minimum point at

\[
q = (U^T U + \alpha U_s)^{-1} (U^T d).
\]

(16)

Having obtained \(q\), a linear formula is used to update \(p\) as

\[
p := (1 - \tau)p + \tau q
\]

(17)

where \(\tau, 0 < \tau < 1\), is a smoothing parameter. The above process is repeated until \(\|p - q\|_2\) is less than a specified tolerance.

The iterative algorithm can now be summarized in terms of the following steps:
Algorithm 1

Step 1: Use a conventional method to design a linear-phase, lowpass, FIR filter of length $N$ with a bandwidth of $\pi/(2M)$. Then, use the coefficient vector of the filter obtained to initialize $p$.

Step 2: Use (15d) to compute matrix $U_r$.

Step 3: Form matrix $U$ and vector $q$ by using (15a) and (16), respectively.

Step 4: If $\|p - q\| < \epsilon$, where $\epsilon$ is a prescribed tolerance, output $p$ as the design result and stop; otherwise, update $p$ using (17) with $\tau$ in the range $0 < \tau < 1$, say, $\tau = 0.5$, and repeat from Step 3.

In the above analysis, $N$ is assumed to be even. If $N$ is odd, the frequency response of $P$ can still be expressed as in (12a) with

$$M_p(\omega) = p^T c(\omega)$$

where

$$c(\omega) = [\cos(N-1)\omega/2 \cos(N-3)\omega/2 \cdots \cos\omega]^T \quad p = [2p(0) 2p(1) \cdots 2p(N-3)/2 p(N-1/2)]^T$$

and Algorithm 1 is valid if (15c) and (15d) are modified as

$$U_t(\Omega_p) = \begin{bmatrix} \cos(N-1)\omega_p1/2 & \cdots & \cos\omega_p1 \\ \cos(N-1)\omega_p2/2 & \cdots & \cos\omega_p2 \\
\vdots & \ddots & \vdots \\ \cos(N-1)\omega_pk/2 & \cdots & \cos\omega_pk \end{bmatrix}$$

$$u_{i,j} = \begin{cases} \pi - \omega_a & \text{for } i = j = \frac{N+1}{2} \\ \pi - \omega_a & \text{for } i = j < \frac{N+1}{2} \\ \frac{\sin[(2i-N-1)\omega_a]}{2(2i-N-1)} & \text{for } i = j < \frac{N+1}{2} \\ \frac{\sin[(i-j-N-1)\omega_a]}{2(i-j-N-1)} & \text{for } i \neq j \end{cases}$$

for $i, j = 1, 2, \ldots, (N+1)/2$, respectively.

Although a rigorous proof of the algorithm’s convergence is not available at this time, the algorithm was found to always converge in a large set of designs when the initial point is obtained as in Step 1. Moreover, the algorithm converges faster if $\tau$ is chosen to be in the vicinity of 0.5.

III. LOW-DELAY COSINE-MODULATED QMF BANK

A. A General Cosine-Modulated QMF Bank

The reconstruction delay of the cosine-modulated QMF bank described in Section II (referred to as the conventional cosine-modulated QMF bank henceforth) is $N - 1$ sampling periods, where $N$ is the filter length. When the number of channels $M$ increases in a filter bank, the bandwidth of each filter decreases, and as a result, high-order filters are required to achieve the narrow passband. Consequently, the filter banks designed have long reconstruction delays, which are highly undesirable in some applications. In this section, we extend the method of Section II to the design of low-delay cosine-modulated QMF banks.

Consider a prototype FIR lowpass filter $P_L$ of length $N$ with an asymmetrical impulse response, a bandwidth of $\pi/(2M)$, and a phase response that is approximately linear. Its frequency response can be expressed as

$$P_L(e^{j\omega}) \approx |P_L(e^{j\omega})|e^{-j\omega k_d/2}$$

and hence, the group delay of the filter is $k_d/2$, where $k_d \leq N - 1$. The transfer functions of the analysis and synthesis filters $H_k(z)$ and $F_k(z)$ are related to the transfer function $P_L(z)$ of the prototype lowpass filter $P_L$ as

$$H_k(z) = e^{j\theta_k} P_L(W_{2M}^{(k+k_d)/2} z) + e^{-j\theta_k} P_L(W_{2M}^{-(k+k_d)/2} z)$$

$$F_k(z) = e^{j\psi_k} P_L(W_{2M}^{(k+k_d)/2} z) + e^{-j\psi_k} P_L(W_{2M}^{-(k+k_d)/2} z)$$

for $0 \leq k \leq M - 1$, where

$$\phi_k = (M - k_d)(2k + 1)\pi/(4M)$$

$$\psi_k = (-M + k_d)(2k + 1)\pi/(4M).$$

From (1) and (20), the aliasing terms in the reconstructed signal can be written as

$$T_l(z) = \frac{1}{M} \sum_{k=0}^{M-1} e^{j(\phi_k + \psi_k)} P_L(W_{2M}^{(2k+k_d)/2} z) P_L(W_{2M}^{-(k+k_d)/2} z)$$

$$+ e^{j(\psi_k - \psi_{k+1})} P_L(W_{2M}^{(2k+k_d)/2} z) P_L(W_{2M}^{-(k+k_d)/2} z)$$

$$+ e^{-j(\phi_k - \phi_{k+1})} P_L(W_{2M}^{(2k-k_d)/2} z) P_L(W_{2M}^{-(k-k_d)/2} z)$$

$$+ e^{-j(\psi_k + \psi_{k+1})} P_L(W_{2M}^{(2k-k_d)/2} z) P_L(W_{2M}^{-(k-k_d)/2} z)$$

for $1 \leq l \leq M - 1$. If we assume that the stopband attenuation of filter $P_L$ is high enough so that the overlap among nonadjacent channel filters can be neglected, the only nonzero terms in (21) occur when $k = M - l$, $k = M - l - 1$ in the second term, and $k = l, k = l - 1$ in the third term, and hence

$$T_l(z) = \frac{1}{M} \{ P_L(-W_{2M}^{(k+k_d)/2} z) P_L(-W_{2M}^{-(k+1)/2} z)$$

$$+ e^{j(\phi_{k+1} + \psi_{k+1})} P^2_L(W_{2M}^{(k+1)/2} z) e^{j\omega}$$

$$+ e^{-j(\phi_{k+1} + \psi_{k+1})} P^2_L(W_{2M}^{-(k+1)/2} z) e^{j\omega} \}.$$
Substituting (19) into (23) and considering $\omega$ in the range $[-\pi/(2M), \pi/(2M)]$, we have
\[
T_0(e^{j\omega}) = \frac{e^{-j\Delta \omega}}{M} \left[ |P_L(e^{j(\omega-\pi/2M)})|^2 + |P_L(e^{j(\omega+\pi/2M)})|^2 \right].
\] (24)

As in the design of conventional cosine-modulated QMF banks, if $|P_L(e^{j(\omega-\pi/2M)})|^2 + |P_L(e^{j(\omega+\pi/2M)})|^2$ is made sufficiently flat over $[-\pi/(2M), \pi/(2M)]$, then $T_0(e^{j\omega})$ will be flat for all frequencies. The perfect-reconstruction condition can be approximated by minimizing the error component
\[
E_{L1} = \int_{-\pi/(2M)}^{\pi/(2M)} |MTO_0(e^{j\omega}) - e^{-j\Delta \omega}|^2 d\omega
\]
\[
= \int_{-\pi/(2M)}^{\pi/(2M)} |e^{-j\Delta \omega}|^2 |P_L(e^{j(\omega-\pi/2M)})|^2
+ |P_L(e^{j(\omega+\pi/2M)})|^2 - e^{-j\Delta \omega}|^2 d\omega
\]
\[
= \int_{0}^{\pi/M} |P_L(e^{j\omega})|^2 e^{-j\Delta \omega}
+ e^{-j\Delta \omega} |P_L(e^{j(\omega-\pi/M)})|^2 - e^{-j\Delta \omega}|^2 d\omega
\]
\[
= \int_{0}^{\pi/M} |P_L(e^{j\omega})|^2 e^{-j\Delta \omega}
+ e^{-j\Delta \omega} |P_L(e^{j(\omega-\pi/M)})|^2 - e^{-j\Delta \omega}|^2 d\omega.
\] (25a)

Since the phase response of $P_L$ is only approximately linear, both the phase and amplitude responses must be optimized, and hence, the above objective function includes both phase and magnitude error components. By also considering the stopband attenuation of filter $P_L$, the error component
\[
E_{L2} = \int_{\omega_s}^{\pi} |P_L(e^{j\omega})|^2 d\omega.
\] (25b)

can be formed. A suitable objective function can now be constructed as
\[
E_L = E_{L1} + \alpha E_{L2}
\] (26)

where $\alpha$ is a positive constant. To achieve the desired design, the objective function in (26) is minimized with respect to the coefficients of $P_L$.

Having the coefficients of $P_L$, the impulse responses of the analysis filters $H_k$ and the synthesis filters $F_k$ can be obtained from (20) as
\[
h_k(n) = 2p_L(n) \cos \left( (2k + 1) \frac{\pi}{2M} \left( n - \frac{k+1}{2} \right) + \theta_k \right)
\] (27a)
\[
f_k(n) = 2p_L(n) \cos \left( (2k + 1) \frac{\pi}{2M} \left( n - \frac{k-1}{2} \right) - \theta_k \right)
\] (27b)

for $0 \leq n \leq N-1$ and $0 \leq k \leq M-1$, where $\theta_k = (2k + 1)\pi/4$ and $p_L(n)$ is the impulse response of the prototype filter $P_L$.

On comparing (27) with (4), we observe that the filter bank proposed in Section III is a general version of the filter bank of Section II, which includes the latter filter bank as the particular case when filter $P_L$ has a symmetrical impulse response, and $k_k = N-1$. It should also be emphasized that no constraint has been imposed on the filter length $N$ and the reconstruction delay $k_d$ of the cosine-modulated QMF bank in (27).

### B. Design Method

Instead of minimizing the highly nonlinear objective function in (26) directly, an iterative procedure similar to that of Section II-B can be used in which the objective function is modified as
\[
E_L' = E_{L1}' + \alpha E_{L2}'
\] (28a)

where
\[
E_{L1}' = \sum_{0 \leq \omega \leq \pi/M} |P_L(e^{j\omega})Q_L(e^{j\omega})
+ e^{-j\Delta \omega}|^2 d\omega.
\] (28b)
\[
E_{L2}' = \int_{\omega_s}^{\pi} |Q_L(e^{j\omega})|^2 d\omega.
\] (28c)

where $Q_L$ is a lowpass FIR filter with a bandwidth $\pi/(2M)$, and the frequency responses of $P_L$ and $Q_L$ can be expressed as
\[
P_L(e^{j\omega}) = \mathbf{p}_L e^{j\mathbf{c}_L(\omega)}
\]
\[
Q_L(e^{j\omega}) = \mathbf{q}_L e^{j\mathbf{c}_L(\omega)}
\]

where $\mathbf{c}_L(\omega) = \left[ 1, e^{-j\omega}, \ldots, e^{-j(N-1)\omega} \right]^T$, and $\mathbf{p}_L = [p(0), p(1), \ldots, p(N-1)]^T$, and $\mathbf{q}_L = [q(0), q(1), \ldots, q(N-1)]^T$ are the coefficient vectors of filters $P_L$ and $Q_L$, respectively.

We begin the iterative procedure by designing a filter $P_L$ with group delay $k_d/2 < (N-1)/2$ using a least-squares design approach similar to that in [22] and write $E_L'$ in (28a) as
\[
E'_L = (U \mathbf{q}_L - \mathbf{d}_L)^H (U \mathbf{q}_L - \mathbf{d}_L) + \alpha \mathbf{q}_L^H U \mathbf{s} \mathbf{q}_L
\] (29)

where $\mathbf{d}_L = [e^{-j\omega_1 k_d}, e^{-j\omega_2 k_d}, \ldots, e^{-j\omega_N k_d}]^T$, and superscript $H$ denotes complex conjugate transposition. In this case, we have

\[
U = H(\Omega_p)U(\Omega_p) + e^{-j\Delta \omega/\pi/M} H\left(\Omega_p - \frac{\pi}{M}\right)
\]
\[
\cdot U(\Omega_p - \frac{\pi}{M})
\] (30a)

\[
H(\Omega_p) = \text{diag}[P_L(e^{j\omega_1}), P_L(e^{j\omega_2}), \ldots, P_L(e^{j\omega_N})]
\] (30b)

\[
U(\Omega_p) = \begin{bmatrix}
1 & e^{-j\omega_1} & \cdots & e^{-j\omega_{N-1}} \\
1 & e^{-j\omega_2} & \cdots & e^{-j\omega_{N-1}} \\
\vdots & \vdots & \ddots & \vdots \\
1 & e^{-j\omega_N} & \cdots & e^{-j\omega_{N-1}}
\end{bmatrix}
\] (30c)

and $\mathbf{U}$ is given by (30d), which appears at the bottom of the next page. As $E'_L$ in (29) is a quadratic function in $\mathbf{q}_L$ with $[\text{Re}(U^H U)] > 0$, it has the unique global minimum point given by
\[
\mathbf{q}_L = (\text{Re}(U^H U) + \alpha \mathbf{U} \mathbf{s})^{-1} \cdot \text{Re}(U^H \mathbf{d}_L)
\] (31)
where $\text{Re}[]$ is the real part of $\cdot$. Having obtained $q_L$, a linear formula can be used to update $p_L$ as
\[ p_L := (1 - \tau)p_L + \tau q_L. \]  
(32)

The procedure is repeated until $\|p_L - q_L\|_2$ is less than a prescribed tolerance. The iterative algorithm for the design of cosine-modulated QMF banks with low reconstruction delays can now be summarized in terms of the following simple steps.

**Algorithm 2**

**Step 1:** Use a least-squares approach to design a lowpass FIR filter of length $N$ with stopband edge $\omega_s$ and group delay $k_d/2$, and use the coefficient vector of the filter obtained to initialize $p_L$.

**Step 2:** Use (30d) to compute matrix $U_s$.

**Step 3:** Form matrix $U$ and vector $q_L$ using (30a) and (31), respectively.

**Step 4:** If $\|p_L - q_L\|_2 < \epsilon$, where $\epsilon$ is a prescribed tolerance, output $p_L$ as the desired result and stop; otherwise, update $p_L$ using (32) with $\tau$ in the range $0 < \tau < 1$, say, $\tau = 0.5$, and repeat from Step 3.

In our experiments, we noticed that undesirable artifacts can sometimes occur in the amplitude responses of the analysis and synthesis filters when the reconstruction delay is low. Such artifacts have been observed as well by other researchers when designing low-delay two-channel filter banks [23], [24] and multichannel cosine-modulated filter banks [17], [20]. In the multichannel case, artifacts can occur in the transition band of the prototype filter $P_L$ when the group delay is low if there is no constraint on its transition band. These artifacts then lead to artifacts in the amplitude responses of the analysis and synthesis filters. An effective treatment of this problem is to modify the objective function in (28a) to include an additional term $\alpha_1 E_{L3}^p$, i.e.,
\[ E_L' = E_{L1}^p + \alpha E_{L2}^p + \alpha_1 E_{L3}^p \]  
(33)

where
\[ \alpha_1 E_{L3}^p = \alpha_1 \int_{\omega_{01}}^{\omega_{02}} |Q_L(e^{j\omega}) - e^{-j\omega_d/2}|^2 \, d\omega \]  
(34)

and $[\omega_{01}, \omega_{02}]$ is an interval in the transition region where the artifacts occur. It can be readily shown that the objective function in (33) can be written as
\[ E_L' = (Uq_L - d_L)^H(Uq_L - d_L) + \alpha q_L^H U_s q_L + \alpha_1 (q_L^H U_s q_L - 2b^T q_L + \omega_{02} - \omega_{01}) \]  
(35)

where $U$ and $U_s$ are given by (30a) and (30d), respectively, and $U_s$ is given by the expression at the bottom of the next page. The global minimum point of $E_L'$ in each step of the iterative algorithm can now be obtained as
\[ q_L = [\text{Re}(U^H U) + \alpha U_s + \alpha_1 U_s]^{-1} [\text{Re}(U^H d_L) + \alpha_1 b]. \]  
(36)

The proposed low-delay cosine-modulated QMF bank can be implemented by the efficient polyphase implementation using the discrete Fourier transform (DFT) as described in [17]. The numbers of multiplications and additions per output sample are comparable with those required by the cosine-modulated filter bank in [19], which is based on the extended lapped transformation.

**IV. DESIGN EXAMPLES**

In this section, we apply the iterative design algorithms described in Sections II and III to several examples. All the designs were carried out by running MATLAB programs on a
Sun SPARC workstation. The performance of the filter banks
designed are evaluated in terms of computational efficiency,
reconstruction error, and implementation considerations. The
performance criteria are as follows:

- iterations used in the design (IN)
- flops used in millions of floating point operations
  (MFLOPS)
- the maximum reconstruction error
  \[ E_r = \max_\omega \left| \left[ MT_0(e^{j\omega}) \right] - 1 \right| \]
- the aliasing error
  \[ E_a = \max_\omega E(\omega) \]

where
\[ E(\omega) = \frac{1}{M} \left( \sum_{t=1}^{M-1} T_t(e^{j\omega}) \right)^{1/2} \]
\[ T_t(e^{j\omega}) = \sum_{k=0}^{M-1} H_k(e^{j(\omega - 2\pi t/M)}) F_k(e^{j\omega}) \]

- the signal-to-noise ratio (SNR)
\[ \text{SNR} = 10 \log_{10} \left( \frac{\text{energy of input signal}}{\text{energy of reconstruction error}} \right) \]
\[ = 10 \log_{10} \left( \frac{\sum x^2(n)}{\sum [x(n) - \hat{x}(n + k_d)]^2} \right) \]

where \( k_d \) is the reconstruction delay

- the number of multiplications per output sample in the
  implementation (MPU) [17]
\[ \text{MPU} = 12 + 2N/M + 4 \log_2(2M) \]

- the number of additions per output sample in the
  implementation (APU) [17]
\[ \text{APU} = 4 + 2N/M + 4 \log_2(2M). \]

To calculate the SNR, the input signal \( x(n) \) was chosen to
be a random signal whose amplitude is uniformly distributed
between 0 and 1. The signal length was 1000.

A. Conventional Filter Banks

Algorithm 1 was used to design three cosine-modulated
QMF banks as follows:

Example 1: In this example, a four-band cosine-modulated
QMF bank satisfying the specifications
\[ M = 4, N = 112, \alpha = 200, \omega_s = 0.2109\pi, \tau = 0.5, \epsilon = 10^{-4} \]
was designed. The initial point \( p \) in Step 1 of Algorithm 1
was the coefficient vector of a lowpass FIR filter designed
by using the window method. The sampling grid contained
200 sampling points. The results obtained are summarized in
Table I. Fig. 2(a) and (b) show the amplitude responses of
the prototype filter and the analysis filters, respectively. To
demonstrate the design efficiency of the iterative algorithm,
the quasi-Newton method using the Broyden-Fletcher-Goldfarb-
Shanno (BFGS) formula for updating the approximation of
the inverse Hessian matrix of the objective function [25] was
used to design the same filter bank with the same initial point
and sampling points. The quasi-Newton method completed the
design with about 502 MFLOPS, which is almost 60 times that
used by the iterative algorithm.

Example 2: A 16-channel cosine-modulated QMF bank
was designed satisfying the specifications
\[ M = 16, N = 386, \alpha = 100, \]
\[ \omega_s = 0.0567\pi, \tau = 0.5, \epsilon = 10^{-4}. \]
The results obtained are listed in Table I. Graphical displays
for the filter bank in Example 2 are shown in Fig. 3(a)–(d).
Fig. 3(a) illustrates the amplitude responses of the analysis
filters. Fig. 3(b) shows the overall amplitude response of the
filter bank, i.e., \( |MT_0(e^{j\omega})| \), and Fig. 3(c) shows the aliasing
error \( E(\omega) \). Fig. 3(d) depicts the amplitude spectrums of
the input signal and the reconstruction error. From Table I and
the figures, it can be observed that the iterative algorithm
leads to analysis and synthesis filters with a high stopband
attenuation (≥ 120 dB) and low overall distortion and aliasing.
levels ($\approx -100$ dB). Such filter specifications were previously achieved with a constrained optimization method [16]. Using the algorithm in [16], the reconstruction error $E_r$ is of the order of $10^{-12}$, whereas in our approach, it is of the order of $10^{-6}$. However, since the aliasing errors $E_a$ achieved by both algorithms are of the order of $10^{-6}$ ($-120$ dB), in both cases the output signals $\hat{x}(n)$ approximate the input signals $x(n)$ with about $-100$ dB reconstruction error. This is evident from the large difference between the amplitude spectrums of the input and the reconstructed error signals as depicted in Fig. 3(d).

Example 3: A 32-band cosine-modulated QMF bank satisfying the specifications

$$M = 32, N = 513, \alpha = 100,$$
$$\omega_s = 0.0315\pi, \tau = 0.5, \epsilon = 10^{-4}$$

was designed. The results obtained are given in Table I. Fig. 4(a) shows the amplitude responses of the first three analysis filters obtained in the design. For comparison, the amplitude responses of the first three analysis filters adopted in the current Moving Picture Expert Group (MPEG) standard on audio compression [26] are depicted in Fig. 4(b). As can be seen from Table I, our design has an SNR of 97.1 dB as opposed to 84.3 dB for the MPEG filter bank with the same test signal. From Fig. 4, our design shows improved stopband attenuation with comparable transition bands with those in the MPEG filter bank.

B. Filter Banks with Low Reconstruction Delays

Algorithm 2 was used to design three low-delay cosine-modulated QMF banks.
Example 4: This example demonstrates that artifacts can occur when designing a low-delay cosine-modulated QMF bank and that it is possible to reduce them. First, a four-channel filter bank was designed with $N = 112, k_d = 55$, and no constraint was applied to the transition band of the prototype filter. Fig. 5(a) and (b) show the amplitude responses of the prototype filter $P_L$ obtained and the analysis filters, respectively. It can be observed that there is a 'bump' in the transition band of $P_L$, which, in turn, leads to the artifacts in the amplitude response of each analysis and synthesis filter. To reduce the artifacts, the modified objective function in (33) was used to design a filter bank satisfying the specifications

$$M = 4, N = 112, k_d = 55, \alpha = 10,$$

$$\alpha_1 = 10^{-3}, \omega_x = 0.2078\pi,$$

$$\tau = 0.5, \omega_{f1} = 0.1234\pi, \omega_{f2} = 0.1266\pi, \epsilon = 10^{-3}.$$  

The results obtained are summarized in Table II. Fig. 5(c) and (d) show the amplitude responses of the prototype filter $P_L$ and the analysis filters, respectively. Comparing them with Fig. 5(a) and (b), we note that the artifacts are reduced significantly. On comparing the filter bank designed in this example with that in Example 1, we observe that the reconstruction delay in the present design is only about 50% of that in Example 1. However, this low delay is achieved at the cost of reduced stopband attenuation and SNR ratio. This was found to be the case in a number of designs, and as a consequence, we believe that the designer can trade off stopband attenuation and reconstruction error for a lower reconstruction delay.

<table>
<thead>
<tr>
<th>IN</th>
<th>Example 1</th>
<th>Example 2</th>
<th>Example 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>MFLOPS</td>
<td>12</td>
<td>14</td>
<td>11</td>
</tr>
<tr>
<td>$E_r$</td>
<td>$3.2594 \times 10^{-6}$</td>
<td>$2.7563 \times 10^{-6}$</td>
<td>$2.9941 \times 10^{-6}$</td>
</tr>
<tr>
<td>$E_s$</td>
<td>$3.2178 \times 10^{-7}$</td>
<td>$2.5814 \times 10^{-7}$</td>
<td>$5.7532 \times 10^{-7}$</td>
</tr>
<tr>
<td>SNR (dB)</td>
<td>111.5</td>
<td>115.7</td>
<td>97.1</td>
</tr>
<tr>
<td>MPU</td>
<td>80</td>
<td>82</td>
<td>68</td>
</tr>
<tr>
<td>APU</td>
<td>72</td>
<td>74</td>
<td>60</td>
</tr>
</tbody>
</table>

Example 5: In this example, an eight-band low-delay cosine-modulated QMF bank satisfying the specifications

$$M = 8, N = 132, k_d = 65, \alpha = 20,$$

$$\alpha_1 = 10^{-3}, \omega_x = 0.1357\pi,$$

$$\tau = 0.5, \omega_{f1} = 0.0561\pi, \omega_{f2} = 0.0609\pi, \epsilon = 10^{-3}$$

was designed. The results are listed in Table II. Fig. 6(a) shows the amplitude response of the prototype filter, and Fig. 6(b) shows its group delay in the passband and transition band. Fig. 6(c) and (d) shows the amplitude response of the analysis filters obtained and the group delay of the filter bank. For comparison, Algorithm 1 was used to design a conventional cosine-modulated QMF bank with a filter length of 66 and a reconstruction delay of 65 sampling periods.
The SNR achieved by this filter bank is 62 dB, and the amplitude response of its prototype filter is also shown in Fig. 6(a). It is observed that with the same reconstruction delay, the low-delay cosine-modulated QMF bank shows improved performance over the conventional cosine-modulated QMF bank. From Fig. 6(b), it can be noticed that in the passband of the prototype filter, linear phase is well approximated, and the phase distortion in its stopband can be neglected due to very small amplitude response. As mentioned in Section III, in the low-delay QMF bank proposed, both the phase distortion and the amplitude distortion are minimized. Fig. 6(d) shows that the phase distortion of the filter bank obtained is small.

Example 6: A 32-band QMF bank satisfying the specifications

\[ M = 32, N = 513, k_d = 255, \alpha = 5000, \]
\[ \omega_s = 0.03417\pi, \tau = 0.5, \epsilon = 10^{-3} \]

was designed. The results are illustrated in Table II, and Fig. 7(a) shows the amplitude responses of the analysis filters. Compared with the current MPEG filter bank, the reconstruction delay of the filter bank designed is reduced by half. In addition, this low-delay filter bank can be implemented by a polyphase structure similar to that described in [26]. As a performance test, a ramp signal was used as the input. The input, the outputs from the low-delay QMF bank designed, and the current MPEG QMF bank are depicted in Fig. 7(b). The figure shows that, in agreement with the design specifications, our design only takes about 50% of the time required by the current MPEG filter bank to reconstruct the input signal.

### Table II

<table>
<thead>
<tr>
<th></th>
<th>Example 4</th>
<th>Example 5</th>
<th>Example 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>( k_d )</td>
<td>55</td>
<td>65</td>
<td>255</td>
</tr>
<tr>
<td>( \ln )</td>
<td>11</td>
<td>13</td>
<td>7</td>
</tr>
<tr>
<td>( E_c )</td>
<td>( 3.980 \times 10^{-8} )</td>
<td>( 1.8041 \times 10^{-8} )</td>
<td>( 7.1657 \times 10^{-5} )</td>
</tr>
<tr>
<td>( E_a )</td>
<td>( 5.1584 \times 10^{-6} )</td>
<td>( 5.8333 \times 10^{-8} )</td>
<td>( 4.6497 \times 10^{-6} )</td>
</tr>
<tr>
<td>SNR (dB)</td>
<td>88.3</td>
<td>82.8</td>
<td>90.4</td>
</tr>
<tr>
<td>MPU</td>
<td>80</td>
<td>62</td>
<td>68</td>
</tr>
<tr>
<td>APU</td>
<td>72</td>
<td>54</td>
<td>60</td>
</tr>
</tbody>
</table>

### V. Conclusion

An iterative algorithm for the design of conventional cosine-modulated QMF banks has been proposed. The algorithm is very efficient and can design filter banks with high stopband attenuation and low aliasing and amplitude distortions. Although a rigorous proof on the convergence of the algorithm is not available at this time, the algorithm converged in all the design examples attempted. A general version of the algorithm has then been developed, which can be used to design low-delay cosine-modulated QMF banks. This algorithm is based on a weighted objective function that depends on the error between the actual frequency response and that of a linear-phase ideal filter. Artifacts that can occur in the amplitude responses of the analysis and synthesis filters when designing low-delay filter banks can be reduced significantly by simply adding one more
error component to the objective function. The two algorithms have been illustrated by designing several filter banks, and the designs obtained have been compared with corresponding designs obtained by using some other known methods.

REFERENCES


Fig. 7. Example 6: (a) Analysis filters; (b) a ramp input and outputs from the low-delay QMF bank and the current MPEG QMF bank.


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