Efficient hearing aid algorithm using DCT with uniformly re-sampled and recursively modified audiogram values

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ABSTRACT

People with the hearing problems have different listening preferences and characteristics in hearing loss. So, hearing aids need algorithms that provide amplification based on frequency, so that the hearing-impaired persons can use hearing aids comfortably for a long duration. In this paper, a new algorithm is proposed for hearing aids in order to compensate for sensorineural and conductive hearing loss using discrete cosine transform (DCT). DCT coefficients of the input audio signal are multiplied with uniformly resampled and recursively modified audiogram values to compensate for hearing loss. This algorithm comprised of 4 stages namely precomputation to calculate gain values from audiogram, DCT, gain adjustment, and inverse DCT. In the above stated stages except precomputation, each stage requires only one matrix multiplication, which makes the proposed algorithm computational efficient. Performance of the proposed algorithm is compared with uniform filter banks, non-uniform filter banks, variable filter bank and reconfigurable filter banks. The algorithm is tested using audiograms with four different hearing loss cases. It is proved that the proposed algorithm provides less complexity, minimized delay and better matching with all types of audiograms, further, it also avoids degradation of audio signal due to sampling rate conversions in variable and reconfigurable filter banks.

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1. INTRODUCTION

Elderly people may not hear properly due to damaged nerve fibers and sensory cells of the inner ear [1]. Hearing aid may be used to compensate for this disability. The hearing aid is an electro acoustic device that amplifies sound signals to compensate for hearing loss. However, characteristics of hearing loss vary from person to person based on the hearing thresholds. It is considered that, normal hearing is between -10 to 20 dB, the mild hearing loss occurs between 20 to 40 dB, moderate is between 40 to 55 dB, moderately severe is between 55 to 70 dB and severe is 70 to 90 dB or profound greater than 90 dB [2], [3]. With a sensorineural hearing loss, one might lose only a certain band of frequency [4]. Thus, Normal hearing aid uniformly amplifies all frequencies in audio signals, but it needs to amplify only the sounds that can't be hear by hearing-impaired, if not the louder sounds become unbearable [5]. Therefore, in hearing aids, a particular band of frequencies of audio signals are subjected for suitable gain adjustment based on an audiogram to make the person understand the speech.

In the present scenario, the research is carried on in the implementation of a signal processing algorithm to compensate for different types of hearing losses. Current studies focus on filter bank structures with less computational complexity in order to reduce the hardware complexity and also to increase the speed of operation. Foremost, and mostly used filter banks are uniform filter bank [6]-[8] and nonuniform filter bank [9]-[11]. Over the past decade, the researchers implemented distinguished efficient variable filter bank structures [12], [13] and reconfigurable filter banks [14]-[20] to get better matching with an audiogram and to reduce the computational complexity of filter banks.

Considerably better match between the frequency response of the hearing aid and audiogram is achievable if a greater number of bands are assigned for a uniform and non-uniform filter bank. But a few drawbacks can be observed such as delay, power consumption, and the size of the hearing aid increase with the increase in the number of bands. In case of uniform and non-uniform filter banks one should compromise on either size and delay or matching error. To achieve better matching error with smaller delay and size filter bank structure should vary with type of the audiogram. Whereas, in uniform and non-unifrom filter banks the filter bank structure is fixed for all types of audiograms. To overcome these drawbacks, reconfigurable filter banks are introduced. In a reconfigurable filter bank the number of subbands in each band varies based on some parameters which gives different structures for different types of audiograms. Even in reconfigurable filter banks a shortcoming is observed, that it uses interpolation and decimation of filter coefficients and/or input signal to convert the sampling rate, which results in signal degradation and also, aliasing effect may occur due to sampling rate conversion. To overcome this hindering, the present research proposes a new technique in which the gains are adjusted in the frequency domain using DCT.

In this technique, DCT coefficients are multiplied with uniformly re-sampled and recursively modified audiogram values to adjust gain in frequency domain, after gain adjustment the frequency domain signal is converted back to time domain using inverse DCT. Audiogram values are re-sampled at uniform intervals of frequency and they are modified to get minimum matching error. This precomputation is performed for each audiogram before loading gain values into the hearing aid. The proposed DCT based algorithm is better when compared to fixed filter banks and reconfigurable filter banks. This is better in terms of complexity as it has only 3 matrix multiplications to perform gain adjustment in the frequency domain. It gives better matching error as it uses recursive modifications of audiogram values based on matching error. This DCT based technique need preprocessing to get gain values from audiogram which is not required in filter bank structures that is the only disadvantage of the proposed technique. Audiogram preprocessing is performed before loading the gain values into the processor. So, it won't affect the speed of the hearing aid system.

The proposed algorithm provides a simple solution to compensate for the hearing loss without any filter banks and sampling rate conversions. Totally 3 stages are needed for the whole process i) finding DCT for the input audio signal, ii) gain adjustment and iii) inverse DCT. The proposed algorithm is tested using audiograms with four different hearing loss cases such as mild hearing loss at high frequencies, mild to moderate hearing loss at low frequencies, moderate hearing loss at middle frequencies and mild conductive hearing loss. It is noted that the proposed algorithm provides less complexity, less delay and better matching with audiogram with all types of audiograms. It also avoids degradation of audio signal due to sampling rate conversions that are used in variable and reconfigurable filter banks.

The paper is organized as follows: Section 2 deals with the implementation of the proposed DCT based algorithm. Section 3 discusses the precomputation to find the gain values from audiogram. In section 4, design examples and performance evaluation proposed algorithm are tested with audiograms with four different hearing loss cases. Section 5 brings in experimental results and analysis. Finally, the conclusion is drawn in section 6.

2. PROPOSED DCT BASED ALGORITHM

In this proposed technique, the audio signal is transformed to the frequency domain using DCT to amplify the audio signal as per the above requirement. The commonly used transform domain approaches are based on DCT [21]. DCT is used to convert the data into a sum of cosine wave trays of different frequencies. As the DCT coefficients are arranged in ascending order with respect to their corresponding frequencies, it is very easy to adjust the gains. Uniformly re-sampled and recursively modified audiogram values are multiplied with the DCT coefficients of the audio signal to perform the gain adjustment. Audiogram re-sampling is discussed in the precomputation section of this paper. Figure 1 represents the block diagram of the proposed algorithm. As per the block diagram, input audio signal sensed by microphone is given to analog to digital converter (ADC) which converts the analog audio signal into a digital audio signal. Then, that digital signal is transformed into the frequency domain using DCT transform. Output of the DCT block is DCT coefficients. Further, in the gain adjustment block DCT coefficients are multiplied with uniformly

re-sampled and recursively modified audiogram values. Then, inverse DCT transforms the amplified DCT coefficients into time domain. Finally, digital to analog converter (DAC) is to convert the digital auditory compensated signal to analog and given to the audio speaker.

Figure 2 (a) explains the working procedure of the proposed algorithm. Consider the sampling frequency of the input audio signal Fs=16 kHz. In order to apply 80-point DCT on input signal 80 input samples need to be stored in the buffer. Then, apply 80-point DCT to the input samples stored in buffer to convert the time domain signal into frequency-domain. Later, in Figure 2 (b) the DCT coefficients are multiplied with uniformly re-sampled and recursively modified audiogram values to adjust gain values. Thus, to convert the amplified signals back to time-domain, apply 80-point inverse DCT. Now the output is time domain audio signal after gain adjustment.



Figure 1. Block diagram for proposed algorithm



Figure 2. These figures are; (a) working procedure of the proposed algorithm, and (b) modifying audiogram values recursively

3. PRECOMPUTATION

This section discusses how audiogram values are re-sampled and modified recursively in order to find the gain values that are needed to be adjusted in frequency domain.

3.1. Audiogram re-sampling

Graph in Figure 3 (a) represents the audiogram values for the hearing loss case with mild hearing loss at mid frequency. Graph in Figure 3 (b) is the uniformly re-samples audiogram values at 100 Hz that mean sampling interval in frequency domain is 100 Hz. From the audiogram in Figure 4 (a), it is observed that the audiogram is recorded at non-uniform frequencies. In order to get uniformly sampled audiogram values from the above audiogram, it is needed to be re-sampled at uniform intervals of frequency. This can be implemented by a simple computer application by the audiologist before programming the digital hearing aid. Interpolation is needed for re-sampling of an audiogram and interpolated samples should be an average of previous and next sample values. As the audiogram is recorded at non-uniform frequencies it may require non-uniform interpolation; for example, if the new sampling interval in frequency domain is 100 Hz that

means if the audiogram needs to be re-sampled at every 100 Hz then between 500 Hz and 1000 Hz 4 samples need to be interpolated, whereas between 4000 Hz and 8000 Hz 39 samples are to be interpolated. This non-uniform re-sampling of the audiogram is explained in the equation.



Figure 3. These figures are; (a) audiogram values, and (b) uniformly re-sampled audiogram values.

In (2) F(k) contains the frequencies at which the audiogram is recorded. In (1), G(k) contains the gain values in the audiogram corresponding to the frequencies defined in the F(k). N is the index of the current interpolating sample among the N number of samples. N is the number of samples to be interpolated between $(k-1)^{th}$ and k^{th} samples of the audiogram. F_d is the frequency difference between two int-erpolated samples (new uniform sampling interval in frequency domain) F_d =100. Elements in G_n(n,k) are expanded into a single vector of size 80 for F_d=100.

$$G_{n(k,n)} = G(k-1) + \frac{[G(k) - G(k-1)]n}{N}$$
(1)

where:

$$N = round \left[\frac{F(k) - F(k-1)}{F_d}\right]$$
⁽²⁾

$$F(0) = G(0) = 0 \tag{3}$$

3.2. Recursive modification of audiogram values

The proposed algorithm can be implemented directly but the matching error between audiogram and frequency response of the designed system is very high in some cases like moderate sensorineural hearing loss. To minimize this matching error, modification of gain values (audiogram re-sampled values) using a recursive algorithm is proposed. In this algorithm, the weighted matching error is recursively added to the uniformly re-sampled audiogram values. This process may be repeated till the matching error is reduced to minimum level. The number of iterations and the weight values depend on the type of audiogram. By trial and error, it is observed that the weight value is in between 0.1 and 1. Gain values are loaded into the hearing aid after re-sampling and modification. To generate modified gain values for the given audiogram this

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algorithm is implemented on PC. Changes in matching error concerning the number of iterations and weight is explained in detail in the section V experimental results and analysis.

The working procedure for recursively modifying audiogram gain values is shown in Figure 2 (b). In this process, gain values from audiogram are subjected to the uniformly re-sampling block to obtain uniformly re-sampled audiogram values. These gain values are recursively modified based on the matching error, if matching error is less than predefined threshold value then it undergoes one more iteration. Gain modification in this algorithm implies updating gain values with the weighted matching error

4. DESIGN EXAMPLES AND PERFORMANCE EVALUATION

The idea of the proposed algorithm for hearing aid is examined by using some examples. The performance of the proposed algorithm is evaluated using audiograms with four types of hearing loss cases. Based on levels of hearing thresholds, hearing losses are categorized as mild, moderate, moderately severe, severe and profound. Hearing loss cases like severe and profound may not be compensated using a hearing aid [22].

4.1. Example 1: Audiogram for mild hearing loss at high frequencies

Filter bank		Example 1		Example 2			E	xample 3		Example 4			
	Number	Maximum	Delay	Numbe	Maximum	Delay	Number	Maximu	De	Number	Maximum	Delay	
	of side	Matching	(ms)	r of	Matching	(ms)	of side	m	lay	of side	Matching	(ms)	
	bands	Error (dB)		side	Error (dB)		bands	Matchin	(m	bands	Error (dB)		
				bands				g Error	s)				
								(dB)					
Direct	8	6.39	4.3	-	-	-	-	-	-	8	5.86	4.3	
design													
[10]	10	9.61	15.7	8	3.2	5.7	8	9.2	15	10	3.67	5.7	
									.7				
[11]	16	2.10	12.8	-	-	-	-	-	-	-	-	-	
[15]	8	4.82	29	-	-	-	-	-	-	7	2.67	25	
[23]	7	5.63	12.1	-	-	-	-	-	-	7	1.84	2.1	
[17]	10	2.84	6.6	12	1.51	12	13	2.72		11	1.49	12	
									18				
[19]	6	2.84	15.7	12	1.49	12	13	2.72	18	7	1.36	1.09	
			5										
Proposed	-	1	7.4	-	0.88	7.4	-	1.87	7.	-	0.15	7.4	
algorithm									4				

Table 1. Comparison of the proposed algorithm with the filter bank structures in [10], [11], [15], [23], [17], [19]

4.2. Example 2: Audiogram for mild to moderate hearing loss at low frequencies

According to the audiogram shown in Figure 4 (b) the gain values are 45, 35, 20, 10, 5, 10 respectively. Figure 5 (b) represents the audiogram values and the frequency response of the hearing aid system. Figure 6 (b) represents the matching error. From Figures 5 (b), 6 (b), and Table 1 it is clear that the proposed algorithm performs better in terms of maximum matching error and delay, when compared with filter bank techniques. DCT based algorithm gives 0.88 dB matching error at 20 iterations and with weight 0.3.

4.3. Example 3: Audiogram for moderate hearing loss at middle frequencies

According to the audiogram shown in Figure 4 (c) the gain values are 10, 20, 40, 50, 20, and 10 respectively. Figure 5 (c) shows the audiogram values and the frequency response of the hearing aid system. Figure 6 (c) represents the matching error. From the above results, it is clear that the proposed DCT based algorithm gives 1.87 dB matching error at 20 iterations and with weight 0.4

4.4. Example 4: Audiogram for mild conductive hearing loss

According to the audiogram shown in Figure 4 (d) the gain values are 25, 25, 25, 35, 25, and 30 respectively. Figure 5 (d) shows the audiogram values and the frequency response of the hearing aid system. Figure 6 (d) represents the matching error. It is clear from the above stated Figures 4 (d), 5 (d), 6 (d), and Table 1 that the proposed algorithm performs better when compared with all filter bank techniques. From Table 2 the proposed DCT based algorithm gives 0.15 dB matching error at 20 iterations and with weight 1. From the above examples, it is evident that the DCT based auditory compensation is simple to implement and has only 80 multipliers in each of three stages namely DCT, gain adjustment and inverse DCT. The matching error is minimum when compared with the fixed filter bank and the reconfigurable filter bank



Figure 4. Audiogram for; (a) Mild hearing loss at high frequencies, (b) Mild to moderate hearing loss at low frequencies, (c) Moderate hearing loss at mid frequencies, and (d) Mild conductive hearing loss [10], [17], [19]



Figure 5. Audiogram values and the frequency response of hearing aid for; (a) mild hearing loss at highfrequencies, (b) mild to moderate hearing loss at low frequencies, (c) moderate hearing loss at midfrequencies, and (d) mild conductive hearing loss

Table 2. Maximum matching error for different hearing loss cases for given weights and the number of iterations

				0110	ciutions							
Audiogram	No. of iterat-ions	Maximum matching error for given weight										
Audiografii		0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1	
Example 1	0	-11.27	-11.27	-11.27	-11.27	-11.27	-11.27	-11.27	-11.27	-11.27	-11.27	
	5	-11.03	-10.67	-10	-9.2	-8.34	-7.4	-6.62	-5.82	-5.1	-4.44	
	10	-10.63	-9.18	-7.56	-6.05	-4.73	-3.59	-2.6	-1.74	-1.28	-0.95	
	15	-9.95	-7.59	-5.44	-3.7	-2.32	-1.36	-0.93	-0.65	-0.59	0.7	
	20	-9.18	-6.14	-3.76	-2.01	-1.07	-0.67	0.49	0.7	0.76	0.72	
Example 2	0	-9.03	-9.03	-9.03	-9.03	-9.03	-9.03	-9.03	-9.03	-9.03	-9.03	
	5	-6.43	-4.85	-3.63	-2.71	2.54	2.91	3.73	4.69	6.07	6.58	
	10	-4.88	-2.77	1.98	2.45	2.22	-1.74	-3	-3.67	2.13	6.17	
	15	-3.71	1.95	1.82	0.97	-1.68	-1.35	3.68	5.52	4.79	6.33	
	20	-2.8	1.68	0.88	-1.06	1.37	1.62	-3.53	-3.74	6.49	-3.64	
Example 3	0	-14.48	-14.48	-14.48	-14.48	-14.48	-14.48	-14.48	-14.48	-14.48	-14.48	
	5	-10.9	-8.88	-7.3	-6.14	-5.08	-4.22	-3.45	4.6	7.11	9.55	
	10	-8.87	-6.06	-4.08	2.75	4.53	5.43	5.19	-6.16	-9.53	-12.5	
	15	-7.34	-4.06	2.97	3.73	2.95	-5.31	-6.69	-6.38	16.17	-11.32	
	20	-6.03	-2.52	3.14	1.87	-4.53	4.12	12.53	15.9	20	-199	
Example 4	0	-2.89	-2.89	-2.89	-2.89	-2.89	-2.89	-2.89	-2.89	-2.89	-2.89	
	5	1.89	1.9	1.86	1.77	1.64	1.48	1.3	1.1	0.92	0.75	
	10	1.89	0.75	1.47	1.16	0.88	0.68	0.54	0.45	0.4	0.35	
	15	1.83	1.47	1.04	0.71	0.52	0.41	0.34	0.28	0.24	0.21	
	20	1.73	1.18	0.73	0.49	0.37	0.29	0.24	0.2	0.17	0.15	



Figure 6. Matching error for; (a) mild hearing loss at high frequencies, (b) mild to moderate hearing loss atlow frequencies, (c) moderate hearing loss at mid frequencies, and (d) mild conductive hearing loss.

5. EXPERIMENTAL RESULTS AND ANALYSIS

Matching errors for different types of audiograms with the different number of iterations and weights are shown in the Table 2. From the Table 2, it is clear that the matching error changes with the number of iterations and weight. In the case of mild hearing loss at high frequency (example 1) better matching can be observed at weight greater than or equals to 0.7. In the case of mild to moderate hearing loss at low frequency and moderate hearing loss at middle frequency (example 2 and 3) better matching can be observed at weight between 0.3 and 0.7. Whereas in conductive hearing loss case, better matching can be observed at weight equals to 1. In all cases matching error reduces with an increase in the number of iterations.

By analyzing Table 2 we can observe that for all cases matching error is better at 20 iterations. If hearing loss is at high frequency weight should be between 0.7 and 0.9. In case of low frequency hearing loss weight should be between 0.3 and 0.6. From example 3 in case of middle frequency weight should be between 0.4 and 0.7. For conductive hearing loss weight should be 1.

The proposed algorithm is tested using an audio signal for the audiogram with mild hearing loss in high frequency as shown in Figure 4. The frequency-domain representation of the input audio signal and the amplified signals are shown in Figure 7 (a). Thus, it is clear that the gain is maximum at the frequencies greater than 2000 Hz and the gain is changing with the frequency concerning the audiogram. From time-domain waveforms shown in Figure 7 (b), it is clear that signals with high frequency are amplified with high gain values and low-frequency components are amplified with smaller gain values.

5.1. Delay analysis

The proposed algorithm comprises three stages DCT, gain adjustment and inverse DCT. Before loading gain values into the hearing aid, Audiogram is re-sampled and modified. The proposed algorithm requires a buffer which is needed to compute the 80-point DCT. Total delay is the delay due to buffer plus delay of three matrix multiplications. According to [24], matrix multiplication with length 80 takes 0.8 ms, three such multiplications are needed for the proposed algorithm, so total delay due to matrix multiplications is $t_m=2.4$ ms. According to the (4) proposed algorithm takes 5 ms delay due to the buffer size of 80 at the

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input before applying DCT as shown in Figure 2 (a). From (5) the total delay between input and output signal is 7.4 ms.

$$T_b = \frac{N}{F_s} \tag{4}$$

where N is the size of the buffer in this case N=80, and F_s sampling frequency=16000 Hz. Therefore, T_b =5ms total delay.

$$T = T_b + T_m = 5 + 2.4 = 7.4 ms$$
(5)

5.2. Computational efficiency

To reduce the power consumption and delay hearing aid algorithm should be computationally efficient [25]. In the proposed algorithm DCT and IDCT takes more computations. One dimensional DCT and IDCT requires 2Nlog2(N) number of additions and multiplications, as shown Figure 7 [26]. So, 80-point DCT and IDCT takes 1120 number of multiplications and additions. To adjust the gain values in frequency domain 80 multiplication are needed. To perform the proposed algorithm on 80 samples, 1200 multiplications and 1120 additions are required. Total 14 additions and 15 multiplications are needed for one sample. In [19], the number of multipliers are 67 including all subbands, from this it is clear that multiplier complexity is reduced by 77.61%



Figure 7. Input and output of proposed hearing aid algorithm; (a) frequency domain representation and (b) time domain representation

6. CONCLUSION

In the present research, a DCT based auditory compensation using uniformly re-sampled and recursively modified audiogram values is implemented. The proposed algorithm provides a simple solution to compensate for the hearing loss without any filter banks and sampling rate conversions. Totally three stages are needed for the whole process: i) Finding DCT for the input audio signal, ii) Gain adjustment and iii) Inverse DCT. DCT coefficients of the audio signal are multiplied with uniformly re-sampled and recursively modified audiogram values to adjust the gains in frequency domain. The performance of the proposed algorithm is compared with different types of filter banks namely uniform filter bank nonuniform filter bank, variable filter bank and reconfigurable filter bank structures. The proposed algorithm is tested for different types of hearing loss cases like mild hearing loss at high frequencies, mild to moderate hearing loss at low frequencies, moderate hearing loss at middle frequencies and mild conductive hearing loss. From the above test, it is illustrated that the proposed DCT based algorithm provides better matching between the frequency response of hearing aid and audiogram. It is achieved with minimum delay and computational complexity.

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