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# A NOVEL NORMALIZED CROSS-CORRELATION BASED ECHO-PATH CHANGE DETECTOR

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**Abstract**—A double-talk detector is used to freeze acoustic echo canceller's (AEC) filter adaptation during periods of near-end speech. Increased sensitivity towards double-talk results in declaring echo-path changes as double-talk which adversely effects the performance of an AEC as we freeze adaptation when we really need to adapt. Thus, we need an efficient and simple echo-path change detector so as to differentiate any echo-path variations from double-talk condition. In this paper, we derive a novel test statistic for echo-path change detection. The proposed decision statistic detects any echo-path variations, is normalized properly and is computationally very efficient as compared to existing techniques. Simulations demonstrate the efficiency of the proposed algorithm.

## I. INTRODUCTION

Most teleconferencing conversations are conducted in the presence of acoustic echoes [1]; if the delay between the speech and its echo is more than a few tens of milliseconds, the echo is distinctly noticeable. An acoustic echo canceller (AEC) is used to remove the echo created due to the loudspeaker-microphone environment ( $\mathbf{h}$ ) [3]. In an AEC the echo-path (loudspeaker-microphone path  $\mathbf{h}$ ) is adaptively modelled using a filter ( $\hat{\mathbf{h}}$ ), which is then used to synthesize a replica of the echo ( $\hat{y}$ ). This synthesized replica of the echo is subtracted from the echo-corrupted microphone signal ( $m$ ) to get an echo-free signal ( $e$ ). When the near-end talker ( $v$ ) is active or when the speech comes from both the far-end ( $\mathbf{x}$ ) and near-end ( $v$ ), the adaptive filter coefficients diverge from the true echo path impulse response if the adaptation is not halted. A double-talk detector is used to stop the AEC's filter adaptation during periods of near-end speech [3]. A double-talk detector should be able to detect a double-talk condition quickly and accurately so as to freeze adaptation as soon as possible; at the same time it should be able to track any echo-path changes and should be able to distinguish the double-talk from the echo-path variations [5]. Typically, better immunity towards double-talk results in declaring echo-path changes as double-talk which adversely effects the performance of an AEC as we freeze adaptation when we really need to adapt. Thus, we need an efficient and simple echo-path change detector so as to differentiate

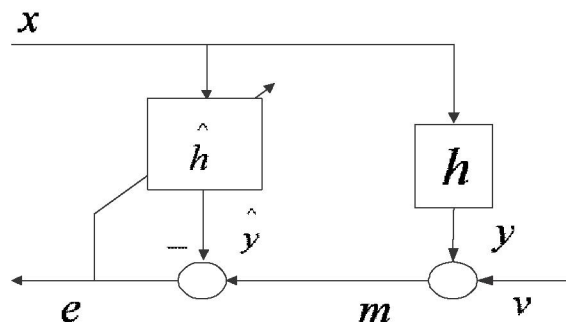


Fig. 1. Basic AEC Model

any echo-path variations from double-talk. An optimum decision variable for echo-path change detection should behave as follows:

- 1) If no echo-path variations i.e. when the adaptive filter is converged  $\xi_{EP} < T_{EP}$ .
- 2) During echo-path variations i.e. when the adaptive filter is not converged  $\xi_{EP} \geq T_{EP}$  and
- 3)  $\xi_{EP}$  is insensitive to near-end speech  $v$ .

Figure 1 shows the basic structure of the adaptive acoustic echo canceller. The far-end signal  $\mathbf{x}$  is filtered through the room impulse response  $\mathbf{h}$  to get the echo signal

$$y(n) = \mathbf{h}^T \mathbf{x} \quad (1)$$

where

$$\mathbf{h} = [h_0 \ h_1 \ \dots \ , \ h_{L-1}]^T,$$

$$\mathbf{x} = [x(n) \ x(n-1) \ \dots \ , \ x(n-L+1)]^T,$$

and  $L$  is the length of the echo-path. This echo signal is added to the near-end speech signal  $v$  to get the microphone signal

$$m(n) = y(n) + v(n) \quad (2)$$

We define the error signal at time  $n$  as

$$e(n) = m(n) - \hat{\mathbf{h}}^T \mathbf{x} \quad (3)$$

This error signal is used to adapt the  $L$  taps of the adaptive AEC filter  $\hat{\mathbf{h}}$ .

This paper is structured as follows: In section 2 the proposed echo-path change detection statistic is derived. We do a comprehensive study on the proposed algorithm for echo-path change detection in section 3 which is followed by a summary and conclusion in section 4.

## II. ECHO-PATH CHANGE DETECTION ALGORITHM

In this section, we derive a novel normalized cross-correlation based echo-path change detector which detects any echo-path variations. Referring to figure 1, the cross-correlation between the microphone signal  $m$ , and the cancellation error  $e$  is given by:

$$\begin{aligned}
 r_{em} &= E[em^T] \\
 &= E[(y+v - \hat{\mathbf{h}}^T \mathbf{x})(y+v)^T] \\
 &= E[(\mathbf{h}^T \mathbf{x} - \hat{\mathbf{h}}^T \mathbf{x} + v)(\mathbf{h}^T \mathbf{x} + v)^T] \\
 &= (\mathbf{h} - \hat{\mathbf{h}})^T E[\mathbf{x}\mathbf{x}^T] \mathbf{h} + E[vv^T] \\
 &= (\mathbf{h} - \hat{\mathbf{h}})^T R_{\mathbf{x}\mathbf{x}} \mathbf{h} + \sigma_v^2
 \end{aligned} \quad (4)$$

where  $\sigma_v^2$  is the variance of the near-end speech, the far-end speech vector  $\mathbf{x}$  the near-end signal  $v$  are independent and are assumed to be of zero mean. Variance of the microphone signal is given by:

$$\begin{aligned}
 \sigma_m^2 &= E[mm^T] = E[(y+v)(y+v)^T] \\
 &= E[yy^T] + E[vv^T] = E[\mathbf{h}^T \mathbf{x}(\mathbf{h}^T \mathbf{x})^T] + \sigma_v^2 \\
 &= \mathbf{h}^T R_{\mathbf{x}\mathbf{x}} \mathbf{h} + \sigma_v^2.
 \end{aligned} \quad (5)$$

and finally the variance of the cancellation error  $e$  is given by:

$$\begin{aligned}
 \sigma_e^2 &= E[ee^T] \\
 &= E[((\mathbf{h} - \hat{\mathbf{h}})^T \mathbf{x} + v)((\mathbf{h} - \hat{\mathbf{h}})^T \mathbf{x} + v)^T] \\
 &= (\mathbf{h} - \hat{\mathbf{h}})^T E[\mathbf{x}\mathbf{x}^T] (\mathbf{h} - \hat{\mathbf{h}}) + E[vv^T] \\
 &= (\mathbf{h} - \hat{\mathbf{h}})^T R_{\mathbf{x}\mathbf{x}} (\mathbf{h} - \hat{\mathbf{h}}) + \sigma_v^2
 \end{aligned} \quad (6)$$

We define our new normalized decision statistic to be

$$\xi_{AsifEPD} = \left| \frac{r_{em} - \sigma_e^2}{\sigma_m^2 - r_{em}} \right| \quad (7)$$

substituting equations 4, 5 and 6 in 7 we get:

$$\xi_{AsifEPD} = \left| \frac{(\mathbf{h} - \hat{\mathbf{h}})^T R_{\mathbf{x}\mathbf{x}} \hat{\mathbf{h}}}{\mathbf{h}^T R_{\mathbf{x}\mathbf{x}} \hat{\mathbf{h}}} \right| \quad (8)$$

we observe from equation 8, for  $\mathbf{h} \approx \hat{\mathbf{h}}$ ,  $\xi_{AsifEPD} \approx 0$  and for  $\mathbf{h} \neq \hat{\mathbf{h}}$ ,  $\xi_{AsifEPD} > 0$ . Thus the proposed echo-path change detector meets the needs of an optimal echo-path change detector.

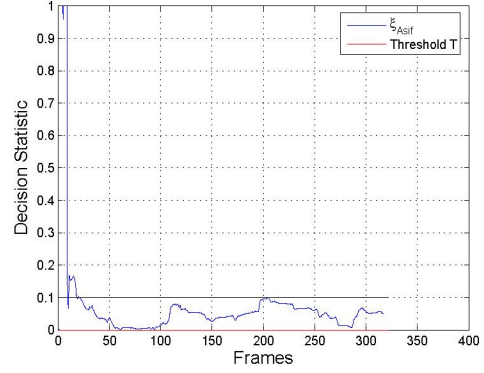


Fig. 2.  $\xi_{Asif}$  as function of time frames, Selecting detection threshold  $T$ .

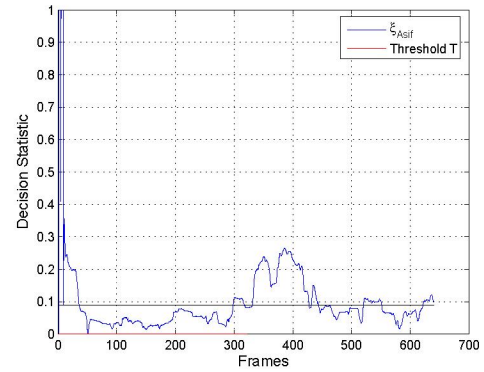


Fig. 3.  $\xi_{Asif}$  as function of time frames.

Our algorithm is computationally very efficient, we only need to do 3 multiplications, 3 additions, 2 subtractions and a division to compute the decision statistic at each sample (i.e. 9 operations per sample) as compared to  $2l + 2$  multiplications,  $l + 1$  divisions and  $3l + 1$  additions (i.e.  $6l + 4$  operations) per sample are required for the decision statistic proposed in [5]. Further, our decision statistic is normalized appropriately i.e. it is approximately zero in the absence of echo-path variations and is greater than zero during echo-path variations.

## III. EXPERIMENTS AND RESULTS

The values of  $r_{em}$ ,  $\sigma_m^2$  and  $\sigma_e^2$  in 7 are exact and not available in practice. As a result, the final decision statistic is given by:

$$\xi_{Asif} = \left| \frac{\hat{r}_{em} - \hat{\sigma}_e^2}{\hat{\sigma}_m^2 - \hat{r}_{em}} \right| \quad (9)$$

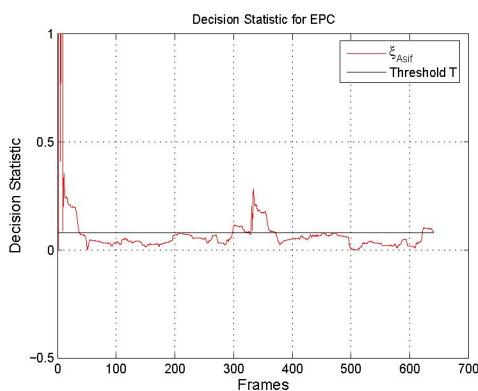


Fig. 4.  $\xi_{Asif}$  as function of time frames.

where the estimates denoted by a hat are obtained using the exponential recursive weighting algorithm, [4] [2]:

$$\begin{aligned}\hat{r}_{em}(t) &= \lambda \hat{r}_{em}(t-1) + (1-\lambda)e(t)m^T(t) \\ \hat{\sigma}_m^2(t) &= \lambda \hat{\sigma}_m^2(t-1) + (1-\lambda)m(t)m^T(t) \\ \hat{\sigma}_e^2(t) &= \lambda \hat{\sigma}_e^2(t-1) + (1-\lambda)e(t)e^T(t)\end{aligned}$$

The echo-path change detector works as follows:

- 1) When  $\xi_{Asif} > T$  ( $T$  is a properly chosen detection threshold), we declare that the echo-canceller has not converged i.e. the echo-path has changed, the adaptation is enabled even if the double-talk detector declares a double-talk.
- 2) Whenever  $\xi_{Asif} < T$ , the detector decides that the echo-canceller has converged i.e. there are no echo-path variations.

Detection threshold  $T$  is chosen to be slightly greater than the steady state value (the value of  $\xi_{Asif}$  in the absence of any echo-path variations) as shown in figure 2. The recorded digital speech sampled at 16 KHz is used as far-end speech  $\mathbf{x}$  and near-end speech  $\mathbf{v}$  and a measured  $L = 8000$  sample (500 ms) room impulse response of a  $10' \times 10' \times 8'$  room is used as the loudspeaker-microphone environment  $\mathbf{h}$ . The room response was collected using a stereo system.

To create echo-path variations we moved from the collected left channel impulse response to the right channel response after 320 frames, as can be seen in figure 3 these changes in echo-path were detected. The echo-path change statistic goes above the detection threshold  $T$  as observed, and hence the variations in echo-path are detected. Next, we increased the echo-path gain by 2 dB after 320 frames. Simulations demonstrate the efficiency of the algorithm, even variations in filter coefficients by 2dB are detected as shown in figure 4.

## IV. CONCLUSIONS

We have proposed a novel normalized sample by sample echo-path change detector. To summarize we list the major advantages of the proposed echo-path change detector:

- Detects any echo-path variations and is normalized appropriately i.e. the detection statistic is greater than zero only for echo-path variations.
- Low added complexity when implemented with any adaptive algorithm.
- Independent of the near-end speech/doubletalk.

The proposed echo-path change detector can also serve as a good download test for a two-path AEC, and when used with a good double-talk detector makes the complete system (AEC) very robust.

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