Canceler for Hands-Free Teleconferencing," *Digital Signal Processing Journal*, Academic Press, vol. 6, no. 1, pp. 29-36, Jan.1996.

An Adaptive Acoustic Echo Canceler for Hands-Free Teleconferencing

Rajiv Porayath, John F. Doherty, and Steve F. Russell

Department of Electrical Engineering and Computer Engineering, Iowa State University, Ames, Iowa 50011

developed. The new echo canceler is based on an enhanced eliminating the acoustical echo is to use a discrete-
adaptive filtering algorithm and uses a spread-spectrumadaptive filtering algorithm and uses a spread-spectrumbased technique to mitigate the effects of double talk. The echo is predominant. This filter is used to quantitatest facility used to develop this new acoustic echo canceler tively characterize the acoustical link between the and the system design of echo canceler are described. Speaker and the microphone, thus generating an allocation of the acquisition of the acquisition

"information superhighway," which will provide the often marginally effective methods are used to alle-
capability for reasonable-cost two-way audio and viate this problem. One example is near-end speech capability for reasonable-cost two-way audio and viate this problem. One example is near-end speech video conferencing, is being rapidly implemented. detection followed by echo cancellation cessation.
With this capability, two geographically separate Another problem is the convergence rate of the adap-With this capability, two geographically separate Another problem is the convergence rate of the adapgroups of people can meet in a "virtual" way and tive filter. For a typical teleconference setup, the filgroups of people can meet in a "virtual" way and tive filter. For a typical teleconference setup, the fil-
avoid the expense of travel. Although technology is ter is on the order of 2000 taps and the input is making remote audio/video conferencing practical, correlated speech signals. This results in slow conthe meeting rooms used for these conferences are vergence of the adaptive filter. still frequently plagued by acoustic echoes that make

it difficult to carry on a meaningful conversation.
these problems by taking an entirely new approach it difficult to carry on a meaningful conversation. These problems by taking an entirely new approach
The efficient mitigation of these deleterious effects to eliminating acoustical echoes. This involves addin a hands-free telephony system is an active re- ing noise-like training signals to the speech before search and development area. The hands-free envi- propagation into the acoustical environment. A corronment is characterized by human communicators relation processor at the output of the microphone that are not ordinarily colocated with their respec- will determine the reverberation modes of the envitive telephones. The decoupling of the handset from ronment, allowing cancellation of the echoes. A mathe head introduces the room as another element in jor feature of the new technique is that it is nearly the voice system. The result is multiple reflections independent of the presence of near-end speech.

of the transmitted acoustic energy impinging on the Porayath, R., Doherty, J. F., and Russell, S. F., An Adap- listening apparatus, which are transmitted back as tive Acoustic Echo Canceler for Hands-Free Teleconferenc-
ing, *Digital Signal Processing* 6 (1996), 29–36.
of the acoustic echo patterns can change rapidly due of the acoustic echo patterns can change rapidly due A new technique for high quality acoustic echo cancella-
tion in the environment, for example, open-
tion in teleconference and speaker phone systems has been
dig and closing of doors. The standard approach to
developed. T electronically synthesized replica of the acoustical echo. This replica is then subtracted from the echo received by the microphone to decouple the speaker and the microphone. Figure 1 shows the standard I. INTRODUCTION approach used in acoustic echo cancellation. A major drawback in this technique is that the addition of near-end speech into the echo return path is not New telecommunications technology known as the accounted for by the adaptive filter. Complicated and ter is on the order of 2000 taps and the input is From the Control of the Co

to eliminating acoustical echoes. This involves add-

FIG. 1. A teleconference setup with an acoustic echo canceler.

Thus, the echoes are canceled and the near-end where the h_i are the coefficients of the transversal speech is transmitted undistorted. The speed of con- filter. These coefficients are estimated using either vergence of the adaptive algorithm is improved by an adaptive algorithm or a correlation. using a filter update that is orthogonal to the input One of the simplest adaptive algorithms is the nordata vector. A threshold function is used to switch malized least-mean-squares (NLMS) algorithm [2]. between the adaptive filter and the correlation esti- The implementation of such a system is not straightmate. These algorithms are described in the next forward. The problems in the model result in the section. **Following requirements on the adaptive echo can-**

The concept of an echo canceler is to consider the

echo path as a mapping function which is then repli-

cand near-end speech. An unprotected deaptive filter

echo prothesize the echo received by the micro-

exhibits unc

$$
\hat{d}(n) = \sum_{i=0}^{N-1} h_i x(n-i), \qquad (1)
$$

celer: (1) long impulse response lengths for the linear filter, (2) fast convergence characteristics for sig-
II. NEW ACOUSTIC ECHO CANCELER nal inputs such as speech, and (3) fast adaptability to variations in the echo path characteristics. An-

> nal to the signal vector $x(n - 1)$, where *n* is the time index. This orthogonal component is used in the update of the adaptive filter coefficient estima-

FIG. 2. Performance of the error square for the LMS and the new echo canceler algorithm.

tor. The procedure can be described as follows. The This is done by transmitting a pseudo-noise training

$$
W_N(n + 1) = W_N(n) + \mu e(n) Z_N(n), \qquad (2)
$$

nal estimated by the echo canceler. The equations

$$
Z_N(n) = X_N(n) - c(n)X_N(n-1)
$$
 (3)

$$
c(n) = \frac{X_N^T(n)X_N(n-1)}{X_N^T(n-1)X_N(n-1)} \tag{4}
$$

are defined, where $c(n)$ is the correlation coefficient for the current input and the delayed input. This algorithm provides fast convergence of the adaptive filter for speech signals as input, and a detailed where $\langle \rangle$ represents averaging, and is defined by study of the algorithm along with its performance the equation in acoustic echo cancellation is included in [5]. Performance improvements can be expected by requiring that the signal vector $Z_N(n)$ be orthogonal to the signal vector $x(n - 2)$ at the expense of increased computational complexity. Taking this notion of or-
thogonality further approaches the recursive leastthogonality further approaches the recursive least-
squares solution, which has a superior performance and *c* is the pseudo-noise training signal. Expanding squares solution, which has a superior performance and *c* is the pseudo-noise training signal. Expanding at increased computational complexity. The conver-
the signal representation and grouping terms gives gence of the new acoustic echo canceler is compared
to the LMS algorithm in Fig. 2.

A correlation technique is used to combat the ef*fect of double talk in the new acoustic echo canceler.*

update equation is given as signal. This signal is continuously transmitted into the room and the signal received by the microphone \mathbf{F} is then correlated to estimate the filter coefficients needed to model the room. This training signal is where *W* are the coefficients of the adaptive estima- independent of either the near-end or far-end speech tor, μ is the step size, and $e(n)$ is the error between and is used by the correlation algorithm to continu-
the signal received from the microphone and the sig-
ously and rapidly estimate the acoustic characteristhe signal received from the microphone and the sig-
nal estimate the acoustic characteris-
nal estimated by the echo canceler. The equations tics. Echo cancellation is again accomplished by subtracting the estimate of the echo signal from the $\overline{\text{microphone}}$ signal. Since the pseudo-noise training signal is uncorrelated with the other signals, it is not affected by the double talk. This is represented by the correlation process as

$$
\hat{h} = \langle s_m c \rangle, \tag{5}
$$

$$
\langle x \rangle = \frac{1}{N} \sum_{i=0}^{N-1} x_i, \quad N \to \infty,
$$
 (6)

the signal representation and grouping terms gives

$$
\hat{h} = \langle s_n c + (c \ast h) c + (s_f \ast h) c \rangle \tag{7}
$$

$$
\hat{h} = h + \langle s_n c \rangle + \langle c(s_f * h) \rangle. \tag{8}
$$

When the training signal correlation with the near-
tions and 1 division. It is seen that the new algo-

$$
\hat{h} = h. \tag{9}
$$

end speech signal, and *h* is the actual impulse retions of its performance in acoustic echo cancellation

method. The use of the switching technique was de-
cided upon because the adaptive algorithm perfor-
with acoustic echo cancellation. cided upon because the adaptive algorithm performance is superior to the correlation performance when double talk is not present. Likewise, the correlation method is superior when double talk is pres-
III. TEST FACILITY ent. The threshold function is obtained from the energy of the signal received by the microphone. A threshold level was determined from experiments A working prototype of the new acoustic echo can-
conducted to detect the presence of double talk [5]. Celer was developed in the laboratory. The test bed conducted to detect the presence of double talk [5].
The threshold is calculated using the relation

$$
t(n) = \mathrm{abs}\left[\frac{c^*e'}{e^*e'}\right].\tag{10}
$$

send into the room and *e* is a vector formed of the signal processor. The algorithms execute on the digierror samples after echo cancellation. These vectors tal signal processor, which also controls the data acwere determined experimentally to be 10 samples quisition board. The noise generator produces the long. The value of $t(n)$ is bounded to an upper value desired pseudo-noise training signal. The tape reof 1 and takes on values below 0.5 in the presence corder is used to provide the far-end speech. This of double talk. This threshold level is used to switch speech is mixed with the training signal. The level between the adaptive and correlation algorithms. $\qquad \qquad$ of the training signal is kept well below the near-

this algorithms is discussed for a digital signal proc- users. A second sample of the training signal is reessor which typically does a MAC (multiply and ac- quired to ensure that it is of the same level as that cumulate) in one instruction. The division operation injected into the room. The data acquisition board is processor dependent and is not included in the samples both these signals to be used by the digital MIPS (million instructions per second) specification. signal processor for processing. The data acquisition The update equation (2) requires $2N + 1$ instruc-
system mixes the speech and the training signal and tions, where *N* is the length of the echo canceler provides the composite signal to an audio amplifier taps. Equations (3) and (4), which form the update which drives the loudspeaker. The echo received by vector, require $6N + 1$ instructions and 1 division. the microphone is sampled by the data acquisition The correlation estimate requires *NM* instructions, system and provided to the digital signal processor. where *M* is the length of the pseudo-noise training The amplitude of the microphone signal is amplified signal. The threshold detector requires 20 instruc- and controlled to provide maximum dynamic range

end speech and the echo signal is zero (both good rithm has inherited the low computational complexapproximations) the estimate of the filter coeffi- ity $(O(N))$ operations of the gradient-type algorithcients becomes exact: mic schemes while providing faster convergence. Thus the new acoustic echo canceler can be imple*h* mented using 16 MIPS on a digital signal processor to obtain a 256-tap acoustic echo canceler with per-Here s_n is the near-end speech signal, s_f is the far-
formance levels that approach those used by re-
end speech signal, and h is the actual impulse re-
cursive least-squares algorithms which have a comsponse of the room. We see that the correlation tech-
nique provides good estimates of the room impulse computational complexity of this algorithm can be nique provides good estimates of the room impulse computational complexity of this algorithm can be
response in the presence of double talk. A detailed reduced by efficient implementation and practical response in the presence of double talk. A detailed reduced by efficient implementation and practical
study of the theory of this method along with simula-
approximations. Frequency domain, block adaptive study of the theory of this method along with simula- approximations. Frequency domain, block adaptive is included in [6].
A threshold function is used to switch between the lation provide complexity reductions. However, A threshold function is used to switch between the these methods introduce delays due to block pro-
ter coefficients obtained from the adaptive algo-cessing and their tracking ability is limited. The imfilter coefficients obtained from the adaptive algo- cessing and their tracking ability is limited. The imrithm and those obtained using the correlation portant note however is that these techniques can-

incorporates a 486-based personal computer (PC), a pseudo-random noise generator, a set of mixers, $t(n) = abs \left[\frac{c*e'}{e*e'} \right]$. (10) an audio amplifier, a data acquisition board, and a digital signal processor (DSP). This test facility is shown in Fig. 3. The PC is used to store the source of the algorithms used in the acoustic echo canceler. Here c is the vector formed of the training signal The PC downloads these algorithms to the digital The computational requirements of implementing end speech level so it remains unnoticeable to the the microphone is sampled by the data acquisition.

FIG. 3. Test facility for the new acoustic echo canceler.

in the analog-to-digital (A-to-D) converter. The digi- In the test facility, the memory and processing tal signal processor uses the three signals (training speed of the hardware and research software were signal, far-end speech plus training signal, and near- not sufficient to provide real-time processing. Inend microphone signal) to compute estimates of the stead, all signals were digitized, stored, and profilter coefficients and to do cancellation. The output cessed off-line, and the results were stored in memof the digital signal processor is the near-end speech ory. The stored results were then output to the Dwith the echo removed. This speech is then transmit-
to-A converter for demonstration and evaluation

written in the C language. The high-level flowchart of the main routine is shown in Fig. 4. This includes programs to: (1) implement the cancellation filter, (2) estimate the filter coefficients with an adaptive NLMS algorithm, (3) estimate the filter coefficients with the training signal, (4) calculate the threshold function value used to switch between the two coefficient estimates, and (5) control the entire system. These programs were designed, coded, and stored on the PC. For execution, the target code is downloaded to the digital signal processor. The data acquisition board consists of a four-channel analog-to-digital converter and four-channel digital-to-analog (D-to-A) converter. The sampling is done at a rate of 8000 samples per second with 16-bit resolution. The 16 bits are then transferred to the digital signal processor which carries out the processing in 32-bit floating point operations. The result is then converted to 16 bit words and transferred to the digital-to-analog **FIG. 4.** Flowchart of the main routine of the acoustic echo can-
celer algorithm.

ted to the far end. The farm of the farm of the farm of the farm of the purposes. Extensive prototype field tests were car-The software for the digital signal processor was ried out to determine the prototype performance

celer algorithm.

marized in the section on test results. The development of the prototype provided insight into the prob-
lems associated with the new method of acoustic

IV. SYSTEM DESIGN enhancement.

This section describes the design of the new acoustic echo canceler as a stand-alone unit which can be incorporated into a teleconferencing or speakerphone system. Figure 5 shows the system design.

Hardware

The analog audio includes: (1) the conditioning of analog signals, (2) the mixing of the training signal and speech, (3) the preconditioning of the signal, and (4) the audio power amplifier circuit. The analog-to-digital converter is three channels and the digital-to-analog converter is two channels. The DSP board has two fixed-point digital signal processors to carry out the computations and eight algorithm specific processors with associated memory. The control system includes the interface to the telephone lines and to the performance display. The training signal generator produces the pseudo-random noise and is driven by the clock. The clock provides synchronization to all the components. It is important to synchronize the entire system to make sure that **FIG. 6.** Hardware setup of the new acoustic echo canceler.

all sampling is done at the proper time. Clock offsets anywhere in the system will lead to degraded performance.

Figure 6 shows the hardware setup of the new acoustic echo canceler. A passive backplane is used to provide support, power supply, and transmission medium for communication between the various boards. The audio board is a full PC card which contains all the audio interface and signal conditioning circuits. The 4-channel analog I/O board contains the A-to-D and D-to-A and interfaces to the DSP board. The DSP board interfaces to the analog I/O and to the adaptive filter module used to carry out the filtering. The CPU board contains a Flash
EPROM and can be programmed through an RS232 FIG. 5. Block diagram of the new acoustic echo canceler. **Exam and can be programmed through an RS232** interface to the PC. The entire unit can operate as a stand-alone unit and has a telephone interface to specifications and compare them with existing facilitate its integration into existing teleconference
acoustic echo cancellation standards. These are sum-
or speaker phone systems.

lems associated with the new method of acoustic
echo cancellation. For example, the training signal
and the speech mixed with the training signal must
both be sampled at the same instant. Also, the dc
offset in the analogviding good flexibility for software maintenance and

V. TEST RESULTS VI. CONCLUSION

$$
ERL(k) = 10 \log \frac{E[x(k)x(k)]}{E[e(k)e(k)]}
$$
 (11)

$$
\text{ERLE}(k) = 10 \log \frac{E[y(k)y(k)]}{E[e(k)e(k)]}, \qquad (12)
$$

the expected value of the power of the uncancelled Engineering and Computer Engineering. echo, and the expected value of the power of the signal received by the microphone, respectively.
These terms are obtained by short-term averaging. REFERENCES

The implementation goal is to achieve 30 dB of ERLE because the ambient noise that is not created 1. Gritton, C. W. K., and Lin, D. W. Echo cancellation algoby the echo itself is typically measured at -30 -dB rithms. *IEEE ASSP* (Apr. 1984). level from the maximum received line signal level. 2. Haykin, S. *Adaptive Filter Theory.* Prentice–Hall, Englewood The prototype achieves an ERLE value of 30 to 35
dB and an ERL value of 35 to 40 dB with no double
talk. In the presence of double talk an ERLE value
of 25 to 30 dB is achieved.
When first used, the acoustic echo canceler

less than a 3-s initialization period (training) before 5. Porayath, R. Adaptive algorithm for acoustic echo cancella-
the audio system can work properly for full dupley tion, M.S. thesis, Iowa State University, Ames. Dec. the audio system can work properly for full duplex tion, M.S. thesis, Iowa State University, Ames. Dec. 1993.
Coneration, The exact value of initialization time will the Miller, J. W. "Correlation-based acoustic echo cance operation. The exact value of initialization time will the Miller, J. W. "Correlation-based acoustic echo canceler," M.S.
depend on the characteristics of a specific room. The prototype achieves 35-dB echo cancellation in after system initialization for teleconference applica- *Mobile Satellite Conference, Ottawa, Canada,* June 17–20, tions. Similarly there is an initialization time associ- 1990, pp. 188–193. ated with the double talk. The initialization time and Hillman, G. On acoustic echo cancellation im-

required by the correlation algorithm is less than 10 required by the correlation algorithm is less than 10
s for teleconference applications. For the conven-
s for teleconference applications. For the conven-
Glasgow, Scotland, May 1989, pp. 952–955. tional acoustic echo canceler the initialization time is indefinite since the canceler cannot adapt to double talk. A recovery time is defined as the time taken by the acoustic echo canceler to achieve a steady-
RAJIV PORAYATH was born in Palghat, India. He received state ERLE value of 35 dB following a period of dou- his B.E. degree in 1990 from Manipal Institute of Technology, ble talk. The prototype has a recovery time of less Manipal, India. His M.S. was received in 1993 from Iowa State
University, Ames. From 1992 to 1994, he was a research assis-

ization and the double talk algorithm provides 25- to scientist with the Center for Advanced Technology Development
30-dB echo cancellation during double-talk periods Brothers, Incorporated, Evanston, IL. He is a member of with an initialization time of less than 10 s . A recov-
Kappa Nu and the IEEE. ery time of 0.2 s is needed after a double-talk period to achieve a steady-state value of 35 dB. JOHN F. DOHERTY received the B.S. (Hon) degree from the

The two measures used to describe the perfor-
mance of the echo canceler are the echo return loss
(ERL) and the echo return loss enhancement
(ERLE), which are measures of the amount of echo
removed by the echo canceler [7,

^E[*e*(*k*)*e*(*k*)] (11) ACKNOWLEDGMENTS

^E[*e*(*k*)*e*(*k*)] , (12) This work was supported by a grant from the U.S. Department of Commerce, under USDOC Grant ITA87-02, through the Center where $E[x(k)x(k)]$, $E[e(k)e(k)]$, and $E[y(k)y(k)]$ for Advanced Technology Development (CATD) at Iowa State
are the expected values of the far-end speech power, Communication Research Group in the Department of Electrical

-
-
-
- properties. *Electron. Comm. Jpn.* **67-A** (1984), 19-22.
-
-
-
-

than 0.2 s.

The prototype acoustic echo canceler provides 35-

The prototype acoustic echo canceler provides 35-

dB echo cancellation within 1 s after system initial-

He worked on acoustic echo cancellation problems as He worked on acoustic echo cancellation problems as an assistant
scientist with the Center for Advanced Technology Development

Stevens Institute of Technology, Hoboken, NJ, in 1985, and the a B.S. degree in 1966 from Montana State University, Bozeman. Ph.D. degree from Rutgers University, New Brunswick, NJ, in His M.S. and Ph.D. in Electrical Engineering were received in 1990, all in Electrical Engineering. He worked as an integrated 1973 and 1978, respectively, from Iowa State University, Ames.
Circuit reliability engineer at IBM from 1982 to 1984. He was a From 1966 to 1970, he was in th circuit reliability engineer at IBM from 1982 to 1984. He was a From 1966 to 1970, he was in the Telecommunications Division member of the technical staff at AT&T Bell Laboratories from at Collins Radio Company. In 1976, he joined the Avionics Divi-
1985 to 1988, working in the area of underwater acquetical signal sion of Rockwell–Collins. From

City University of New York in 1982, the M.Eng degree from STEVE F. RUSSELL was born in Lewiston, Idaho. He received 1985 to 1988, working in the area of underwater acoustical signal
processing. Currently, he is an assistant professor and a Harpole-
Pentair Fellow at Iowa State University, Ames, in the Electrical
Pentair Fellow at Iowa S