

LINEAR PREDICTIVE CODES FOR SPEECH RECOGNITION SYSTEM AT 121BPS

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Abstract:

This paper described the recognition of the phonetics related to numerals in Indian regional languages such as Marathi & Hindi by Nearest Neighbour rule. The segmentation is based on the location of the start and end points of the speech. The exact speech boundaries can be located and evaluated for linear predictive codes. The Linear Predictive Codes of the phonetics related to numerals in Indian regional languages such as Marathi & Hindi forms the codebook. The optimum distance between the test and the codebook linear predictive codes can be determined by the Dynamic Time Warping technique. Depending on the distance, the word is recognized by Nearest Neighbour rule. The accuracy of 88 % is achieved with highly reduction in the memory requirement & good SNR.

Keywords:

Pitch, Linear Predictive Codes, Dynamic Time Warping, Nearest Neighbour Rule

I. Introduction:

In nearly all speech recognition system developed for any language, speech signals are preprocessed to extract the features such as pitch [1]. Speech signal in time domain as raw data generated directly by the speech production system and contain all the acoustic information for recognition. The start and end points of the spoken speech is determined accurately. Two principle components of speech production system, the vocal track and

the excitation source, can be parameterized by time varying autoregressive filter to find the Linear Predictive codes (LPC). The Speech nonstationary is represented in a compact and parametric form on frame-by-frame basis. In order to implement optimization, the distance between frames of features is determined. The optimum distance between the LPC stored in the codebook and the currently spoken word is determined by dynamic time warping technique.

II. System Architecture:

Basic steps involved in the processing are:

1. End Point analysis
2. Pitch Analysis
3. LPC Analysis
4. Pattern classification
5. NN rule for decision of recognition.

Fig 1. Shows the Process diagram for phonetics related to numerals in Indian regional languages

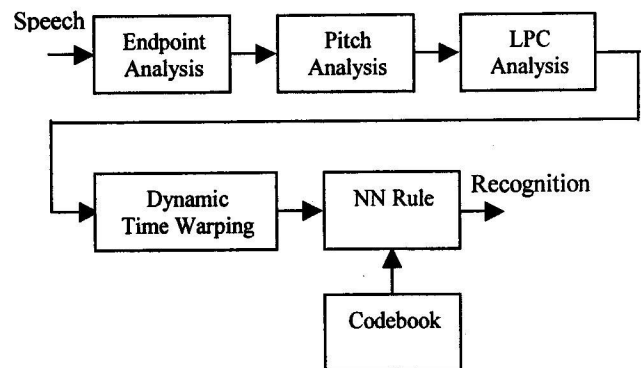


Fig 1. Process diagram for phonetics related to numerals in Indian regional languages

The acoustic waveform of phonetics related to numerals in Indian regional languages such as Marathi & Hindi is analyzed for correctly detecting the start point, end point and pitch of speech [2]-[4]. Linear Predictive coding analysis is performed to achieve the compressed form of large speech data. The optimum distance between the LPC's stored in the codebook and the spoken word under test respectively by the Dynamic time warping technique. The decision of recognition is performed by the Nearest-Neighbour rule.

III. Segmentation:

The phonetics related to numerals in Indian regional languages such as Marathi & Hindi are sampled and analyzed for Linear Predictive Codes. The LPC's are stored to form the codebook.

IV. Pitch Analysis:

Pitch detection is an important task to represent the speech for recognition. The autocorrelation technique estimates the pitch on frame basis by using sliding window technique to reduce the background noise [5]-[8]. The pitch correlation provided by the sliding window is defined as

$$R(T) = \max_{i=Ts}^{Ts-1} [\max_T Ri(T)]$$

where $Ri(T)$
 $= C(i, T+i) / \sqrt{C(i,i)C(T+i, T+i)}$
 Ts - Maximum sliding range
 $Ri(T)$ - Value of the normalized autocorrelation for delay i
 Autocorrelation function

$$C(k,l) = \sum_{n=0}^{N-1} s(n+k)s(n+l)$$

where $s(n)$ is the low pass speech signal, N is the frame size and k and l are the corresponding delays.

V. LPC Analysis:

The speech sample can be approximated as a linear combination of past samples. By minimizing the sum of the squared differences over the frames between the actual samples and the linearly predicted ones, a set of predictive coefficients is determined [9]-[13].

For LPC analysis, hamming windowed samples of speech frames is processed by fixed order digital system to flatten the speech signal.

In linear prediction, the unknown output is represented as a linear combination of past known samples, and the prediction coefficients are selected to minimize the mean square error.

If $x_1, x_2, x_3, \dots, x_n$ represent the data samples and $\hat{x}(n+1)$ represent the predictor for the next sample $x(n+1)$, then

$$x(n+1) = (a_1x_n + a_2x_{n-1} + \dots + a_nx_1)$$

$$= \sum_{k=1}^n a_k x_{n+1-k}$$

and the corresponding error

$$e(n+1) \triangleq x(n+1) - \hat{x}(n+1)$$

$$= \sum_{k=0}^n a_k x_{n+1-k}, a_0 \triangleq 1.$$

Minimization of the mean square error $E [e^2 (n+1)]$ with respect to the unknown qualities a_1, a_2, \dots, a_n gives rise to the standard set of linear equations

$$= \sum_{k=0}^n a_{n-k} r_{k-i+1} + 0, i = 1 \rightarrow n.$$

$$= \sum_{k=0}^n a_{n-k} r_{k-n} \triangleq r^2$$

The matrix can be represented as

$$\begin{bmatrix} r_0 & r_1 & \dots & r_n \\ r_1 & r_0 & \dots & r_{n-1} \\ \dots & \dots & \dots & \dots \\ r_n & r_{n-1} & \dots & r_0 \end{bmatrix} \begin{bmatrix} a_n \\ a_{n-1} \\ \dots \\ a_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \dots \\ r^2 \end{bmatrix}$$

The unique solution

$A_n(z) = 1 + a_1z^{-1} + \dots + a_nz^{-n}$ subject to $a_0=1$ represents the Levinsons polynomial of degree n . These polynomials have all zeros in $|z| < 1$, hence

$H(z) = r / A_n(z) = r / (1 + a_1z^{-1} + \dots + a_nz^{-n})$ that generates $x(n)$ representing stable system. The LPC vectors $[a_1, a_2, \dots, a_n]$ generated represents the speech.

VI. Dynamic Time Warping Framework:

Once the patterns have been determined, similarity between test and reference patterns is determined due to highly variable speaking rate; pattern similarity involves both time alignment and distance computation performed simultaneously [14]-[16].

The alignment function $w(t)$, which maps reference pattern R onto the corresponding parts of test pattern T is measured by calculating, optimized distance between the functions

$$d(T, R) = \|T - R\| = \sum_{i=0}^p (T_i - R_i)^2$$

where T_i and R_i are the i_{th} components of the vectors T and R respectively.

The Dynamic time warping (DTW) determines the optimum path, which minimizes the accumulated distance between test and the reference patterns. Subject to a set of path and end point constraints.

VII. Nearest Neighbour (NN) Rule:

The codebook contains R reference patterns $R^i, i = 1, 2, \dots, V$, and for each pattern the optimum distance is determined by DTW technique.

The NN rule is simply [17]

$i^* = \text{opt}[D^i]$ i.e. choose the pattern R^{i^*} with optimum distance as the recognized pattern.

VIII. Result:

The phonetics related to numerals in Indian regional languages such as Marathi & Hindi are sampled at 8KHz with 8 bits representing each sample. The data is processed in frame of 200 samples by hamming window technique with 100-sample step size is selected for overlapping to smoothen the data to determine the vector of pitch. The vector is obtained by pre-emphasizing speech data to flatten it using first order digital filter $H(z) = 1 - az^{-1}$ with coefficient $a = -0.97$. Table 1 shows the start and end frames of phonetics related to numerals in Indian regional Marathi language //ek//. The performance of speech recognition system using NN rule is presented in Table 2. The requirement of memory for phonetics related to numerals in Indian regional Marathi language is summarized in Table 3. Fig 2. shows the pitch boundaries for the phonetics related to numerical in Indian regional Marathi Language //ek// with system order 10.

Spoken Word= ek	Starting Frame	Ending Frame
Order=10	38	58
Order=11	48	68

Table 1. Showing the Start & End frames of //ek//

Speaker	Recognition success
5	60
26	76.92
50	88

Table 2. Performance of speech recognition System using NN rule

Numerical	Using Endpoint detection KBps	Using LPC Analysis KBps	SNR
//ek//	4.8	0.264	24
//don//	6.8	0.121	17
//teen//	5.2	0.286	26
//char//	4.4	0.033	22
//pach//	4.4	0.242	22
//saha//	3.8	0.209	19
//sat//	5.4	0.297	14
//aath//	6.4	0.352	16
//nau//	4.8	0.264	24
//shunya//	6.4	0.352	16

Table 3. Memory requirement

IX. Conclusion

The phonetics related to numerals in Indian regional languages such as Marathi & Hindi are recognized very effectively. The technique requires very less memory so as to increase the recognition speed with appreciable signal to noise ratio.

X. Figures

Figure 2 & 3 shows the recorded original signal, and the detection of pitch & end points for the phonetics related to numerical in Indian regional Marathi language //ek// respectively.

Figure 4 & 5 shows the extraction of the original signal and zero crossing rate of the speech signal respectively.

Fig 6 are the LPC spectrum.

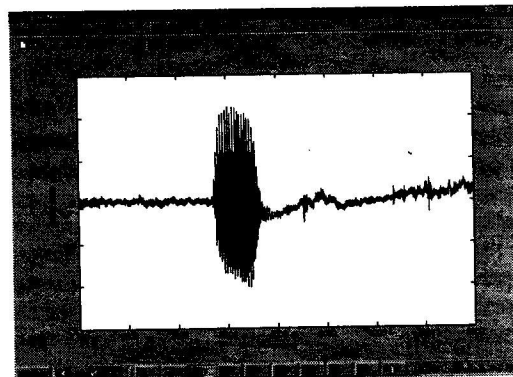


Figure2 shows the recorded original signal.

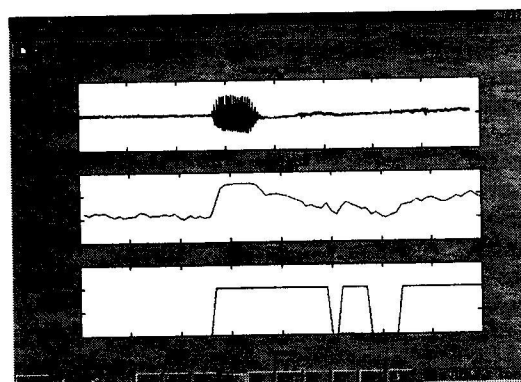


Figure3 the detection of pitch & end points for the phonetics.

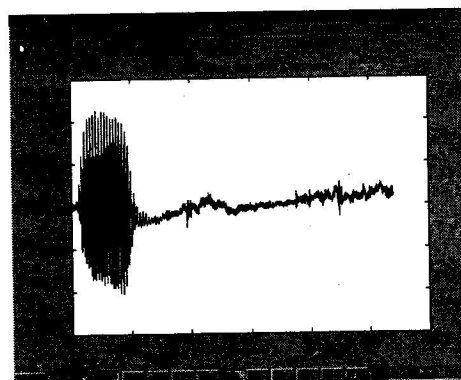


Figure 4 shows the extraction of the original signal.

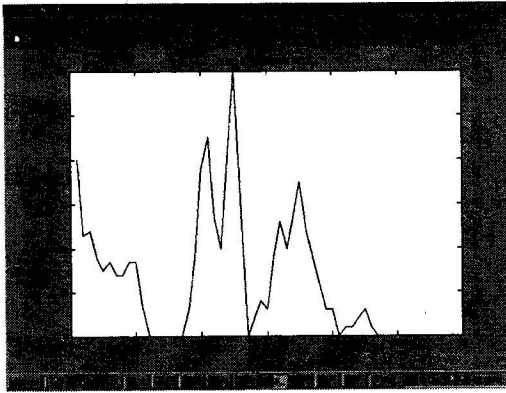


Figure 5 shows the zero crossing rate of the speech signal

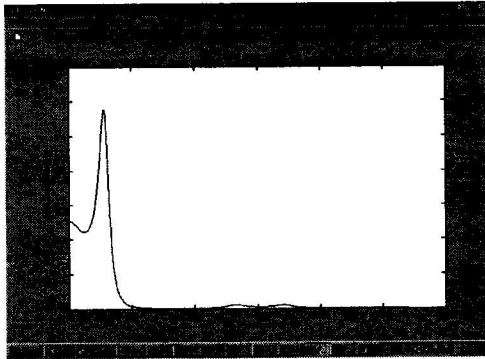


Figure 6 shows the LPC spectrum

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