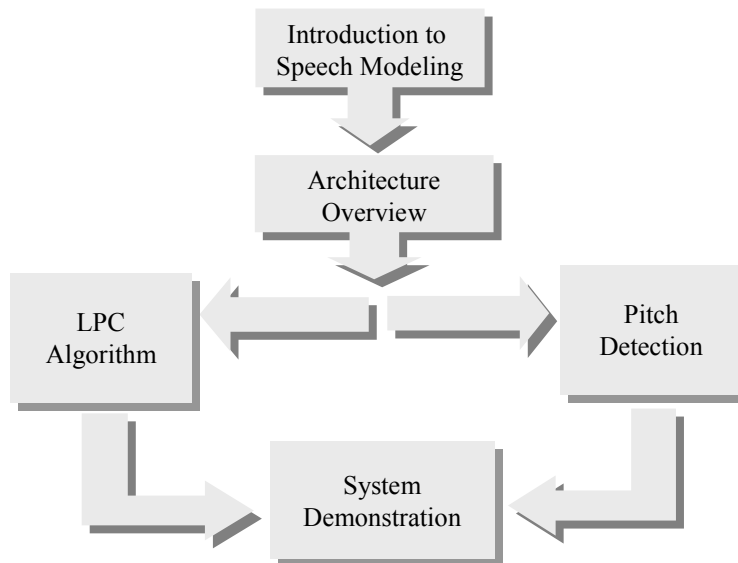


Implementation of Linear Predictive Coding (LPC) of Speech

213A class project
Spring 2000

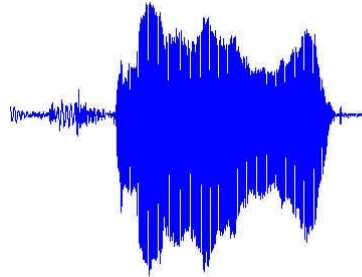
Jean François Frigon and Vladislav Teplitsky

Outline



Implementation of LPC

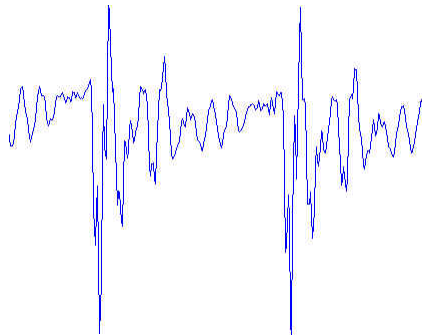
Speech Modeling – Non-stationary



- Speech is a highly non-stationary signal
- Dynamically changes over time
- Changes occur very quickly

Implementation of LPC

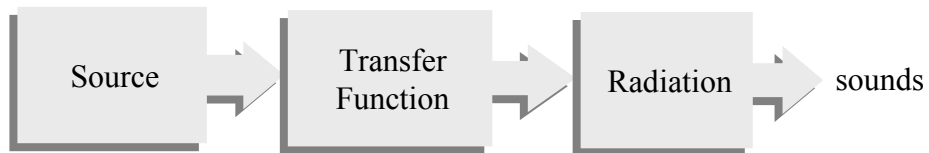
Speech Modeling – Frame Blocking



- Need to analyze the signal over many short segments, called frames
- Apply a short-duration (usually 20-30 msec) overlapping window (usually Hamming) to the speech signal in order to segment into frames
- A single frame of speech is stationary – perform analysis

Implementation of LPC

Speech Modeling – LTI Model



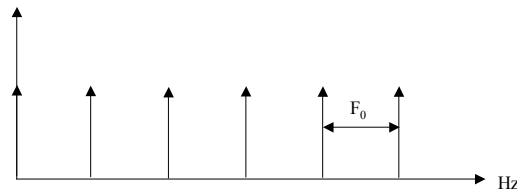
LTI Model is valid for

- moderately loud sounds
- short speech segments – frames (20 – 30 msec)

Implementation of LPC

Speech Modeling – Source (Voiced)

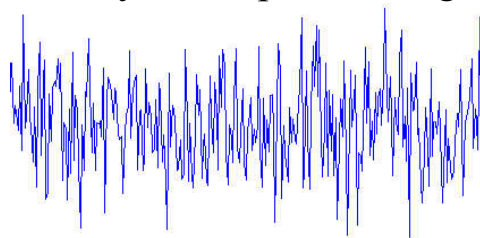
- Sounds are either **voiced** or **unvoiced**
- Voiced (e.g. all vowels) sounds are generated by vocal cords' vibrations
- These vibrations are periodic in time, thus are approximated by an impulse train
- Spacing between impulses is the **pitch**, F_0



Implementation of LPC

Speech Modeling – Source (Unvoiced)

- Unvoiced sounds (e.g. /sh/, /s/, /p/) are generated without vocal cords' vibrations
- The excitation is modeled by a White Gaussian Noise source
- Unvoiced sounds have no pitch since they are excited by a non-periodic signal



Implementation of LPC

Speech Modeling – Transfer Function

- Transfer function models the effects of the vocal tract on the source signal
- Transfer function is either all-pole (vowel model) or pole-and-zero (consonant model)
- Poles of the transfer function – resonances of the vocal tract - are called **formants**
- Human auditory system is much more sensitive to poles than to zeroes of the transfer function

Implementation of LPC

Speech Modeling – Transfer Function

- We will consider only an all-pole transfer function of the form:

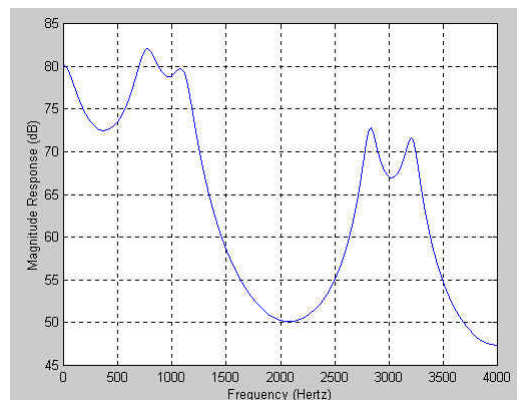
$$H(z) = \frac{G}{\prod_{i=1}^p (1 - a_i)(1 - a_i^*)}$$

- where G is the gain, p is the order (number of poles), and a_i is the pole.
- $p \approx 2 \times \text{Bandwidth of signal (in kHz)} + [2,3,4]$
- e.g. BW=4 kHz, then
 $p = 2 \times 4 + [2,3,4] \in [10,11,12]$

Implementation of LPC

Speech Modeling – Transfer Function

- Example: a 10th order transfer function model:



Implementation of LPC

Speech Modeling – Radiation

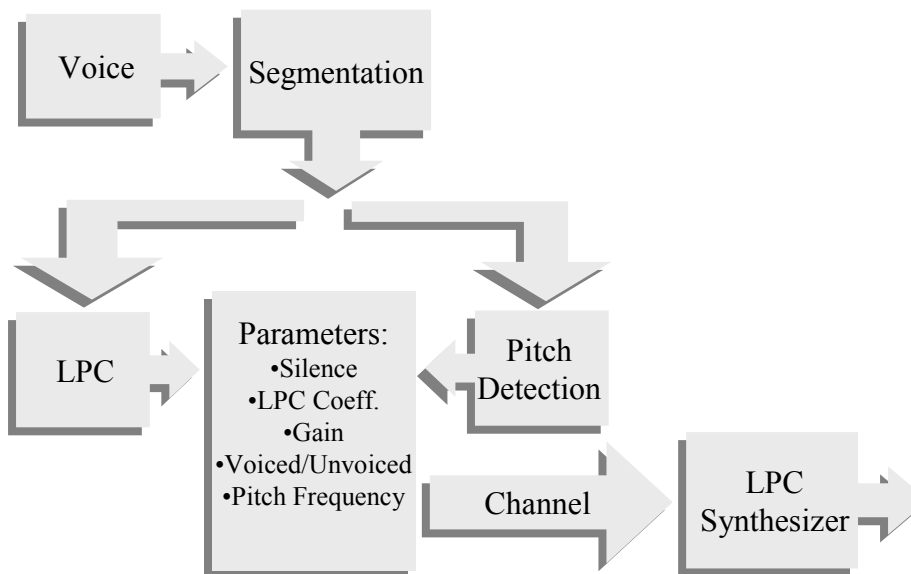
- Models how sound is radiated by the lips
- Usually approximated by a digital differentiator:

$$R(z) = 1 - z^{-1}$$

- Radiation is not important for classification of a sound
- Thus, we will omit it from our implementation

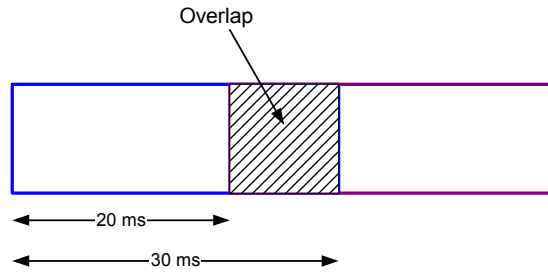
Implementation of LPC

Architecture Overview



Implementation of LPC

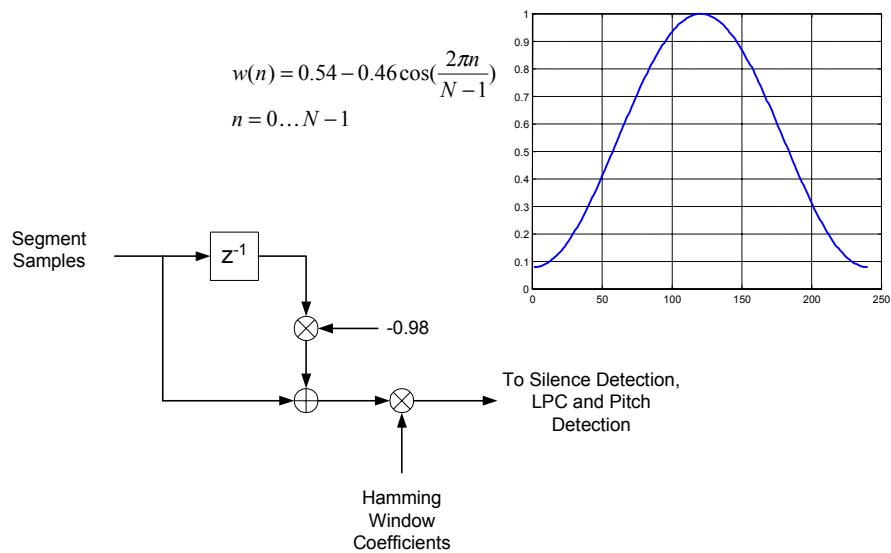
Voice Segmentation



- 8000 samples/sec
- 20 ms step size (160 samples)
- 30 ms window (240 samples)
- Process 240 samples in 20 ms

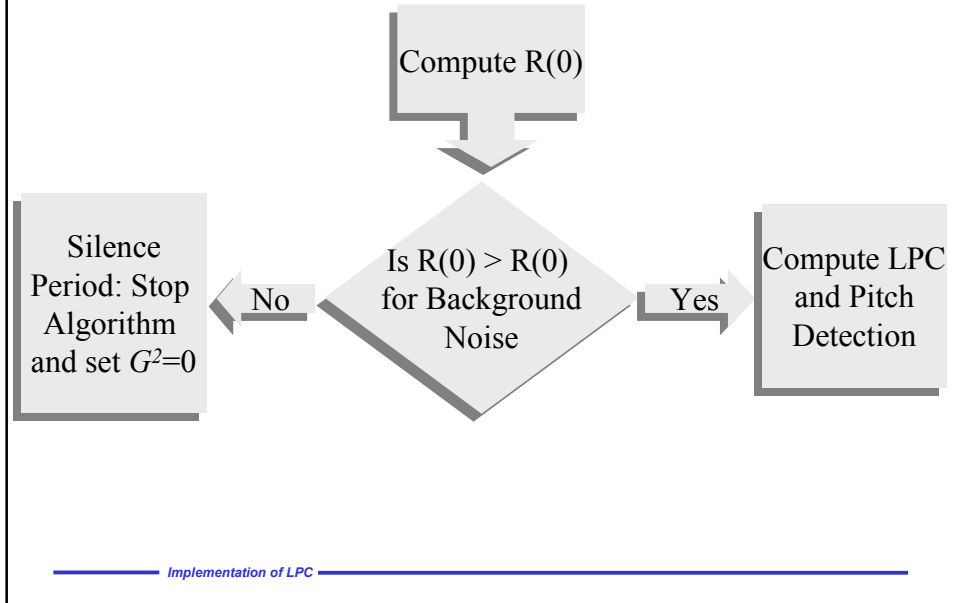
Implementation of LPC

Voice Segmentation - Filtering and Windowing



Implementation of LPC

Voice Segmentation - Silence Detection



LPC - Motivation

Speech Difference Equation for a p^{th} order filter:

$$s(n) = \sum_{k=1}^p a_k s(n-k) + Gu(n)$$

Want to minimize the mean-squared prediction error:

$$e(n) = s(n) - \sum_{k=1}^p \alpha_k s(n-k)$$

For a single input impulse or stationary white noise, the obtained coefficients are identical to the a_k 's

Implementation of LPC

LPC - Autocorrelation (1)

If we assume that $s(n)$ is zero outside the interval $0 \leq n \leq N-1$, we then need to solve the following set of linear equations:

$$\sum_{k=1}^p \alpha_k R(|i-k|) = R(i) \quad 1 \leq i \leq p$$

Where:

$$R(k) = \sum_{m=0}^{N-1-k} s(m)s(m+k)$$

Implementation of LPC

LPC - Autocorrelation (2)

In matrix form the set of linear equation can be expressed as:

$$\begin{bmatrix} R(0) & R(1) & R(2) & \dots & R(p-1) \\ R(1) & R(0) & R(1) & \dots & R(p-2) \\ R(2) & R(1) & R(0) & \dots & R(p-3) \\ \dots & \dots & \dots & \dots & \dots \\ R(p-1) & R(p-2) & R(p-3) & \dots & R(0) \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \dots \\ \alpha_p \end{bmatrix} = \begin{bmatrix} R(1) \\ R(2) \\ R(3) \\ \dots \\ R(p) \end{bmatrix}$$

Implementation of LPC

LPC - Levinson-Durbin Algorithm (1)

By exploiting

- Toeplitz structure of the matrix;
- Particular structure of the right-hand side of the linear system of equation

We can use the efficient Levinson-Durbin recursive procedure to solve this particular system of equations.

Implementation of LPC

LPC - Levinson-Durbin Algorithm (2)

The Levinson-Durbin recursive procedure is given by:

$$E^{(0)} = R(0)$$

for $1 \leq i \leq p$

$$k_i = \frac{\left[R(i) - \sum_{j=1}^{i-1} \alpha_j^{(i-1)} R(i-j) \right]}{E^{(i-1)}}$$

$$\alpha_i^{(i)} = k_i$$

for $1 \leq j \leq i-1$

$$\alpha_j^{(i)} = \alpha_j^{(i-1)} - k_i \alpha_{i-j}^{(i-1)}$$

$$E^{(i)} = (1 - k_i^2) E^{(i-1)}$$

The final solution is given by: $\alpha_j = \alpha_j^{(p)} \quad 1 \leq j \leq p$

Implementation of LPC

LPC - Gain Coefficient

It can be shown that the gain coefficient is given by:

$$G^2 = R(0) - \sum_{k=1}^p \alpha_k R(k) = E_n$$

Where E_n is the minimum mean squared error prediction and is given by $E^{(p)}$ from Levinson-Durbin's Algorithm.

We will transmit G^2 .

Implementation of LPC

LPC Algorithm

From Segmentation:
 $s(n)$ and $R(0)$

Compute $R(i)$ $1 \leq i \leq p$

Levinson-Durbin's Algorithm:
Find α_i $1 \leq i \leq p$
and G^2

Transmit
to decoder

Implementation of LPC

Pitch Detection - Motivation

- Recall that source can be either a periodic impulse train spaced by F_0 or random noise
- Autocorrelation function of a speech frame:

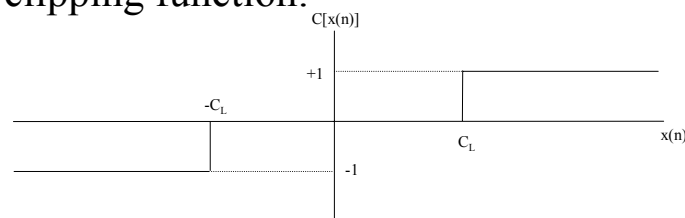
$$R(k) = \sum_{m=0}^{N-k-1} x(m)x(m+k)$$

- If $x(n)$ is periodic in N , then $R(k)$ is also periodic in N
- Thus, we can compute $R(k)$ and check if it's periodic

Implementation of LPC

Pitch Detection – Motivation

- First we clip the frame using 3-level center clipping function:



- That is:

$$C[x(n)] = \begin{cases} +1 & \text{if } x(n) > C_L \\ -1 & \text{if } x(n) < -C_L \\ 0 & \text{otherwise} \end{cases}$$

Implementation of LPC

Pitch Detection – Motivation

- Next we compute the modified autocorrelation function:

$$R_n(k) = \sum_{m=0}^{N-k-1} x(m)x(m+k)$$

- where $x(m)x(m+k)$ can have only 3 different values:

$$\begin{aligned} &+1 && \text{if } x(m) = x(m+k) \\ x(m)x(m+k) = &-1 && \text{if } x(m) \neq x(m+k) \\ &0 && \text{if } x(m) = 0 \text{ or } x(m+k) = 0 \end{aligned}$$

Implementation of LPC

Pitch Detection – Motivation

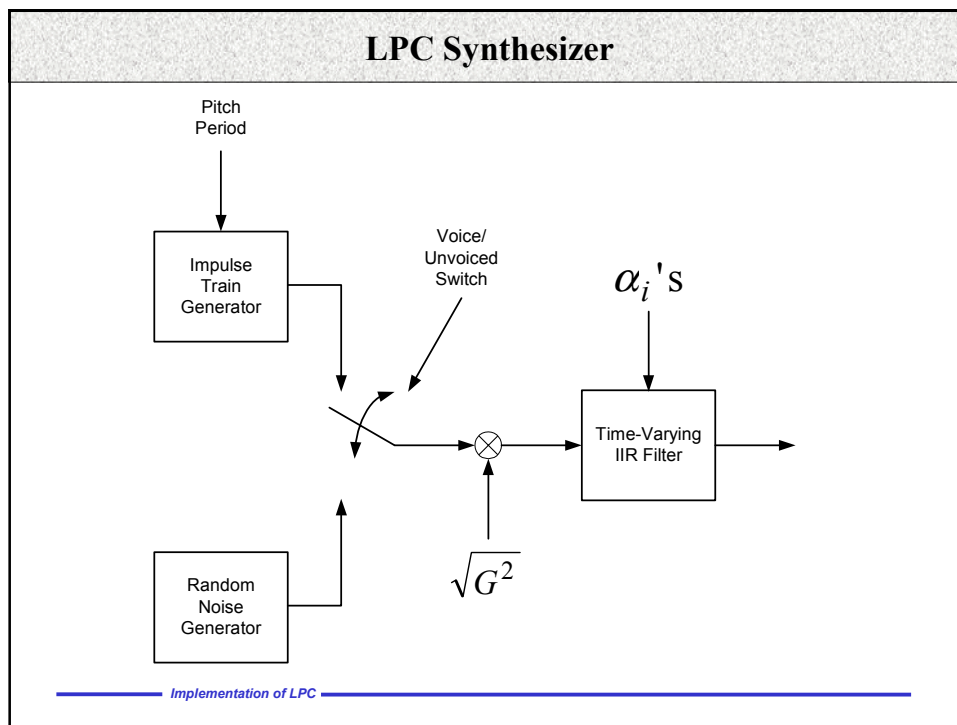
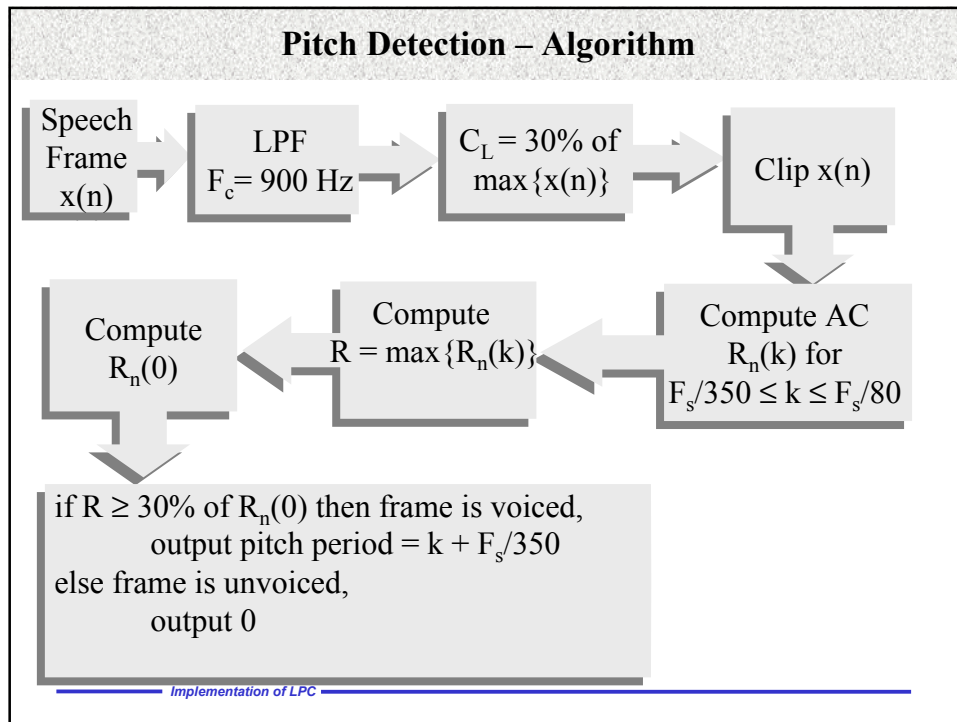
- We don't need to compute $R_n(k)$ for all values of k (i.e. $0 \leq k \leq N$)

	F_0 (Hz) min	F_0 (Hz) max
men	80	200
women	150	350

- Thus we only need to look in the range:

$$80 \text{ Hz} \leq F_0 \leq 350 \text{ Hz}$$

Implementation of LPC



References

- L. R. Rabiner and R. W. Schafer. *Digital Processing of Speech Signals*. Prentice Hall, Englewood Cliffs, New Jersey, 1978.
- Douglas O'Shaughnessy. *Speech Communication Human and Machine*. Addison Wesley Books, 1978.
- M. M. Sondhi. *New Methods of Pitch Extraction*. IEEE Trans. Audio and Electroacoustics, Vol. AU-16, No. 2, pp. 262-266, June 1968.