- 1. Linear Predictive Coding (LPC)
- 2. LPC is based on AR signal modeling
- 3. LPC is the basis of speech compression for cell phones, digital answering machines, etc.
- 4. LPC is a lossy compression scheme.
- 5. LPC is specifically tailored for speech. It does not work well for audio in general.

Human speech

- 1. Parts of speech: vowels, consonants, semivowels, and dipthongs.
- 2. voiced sounds: generated by vocal cords.
- 3. unvoiced sounds: do not involve vocal cords (uses mouth and nasal cavities).

Over short intervals (30 milliseconds), voiced speech:

- 1. resembles a quasi-periodic pulse train;
- 2. the interval between successive pulses are not exactly the same.
- 3. the amplitudes of successive pulse are not exactly the same.

Pitch frequency:

- 1. The pitch frequency is the reciprocal of the average period of the quasi-periodic pulse train.
- 2. Different people have different pitch frequencies. The pitch will vary slightly as a person speaks. A question often ends with a higher pitch (for some cultures).
- 3. Adult males: 80 120 Hz.
- 4. Adult females: 150 300 Hz.
- 5. Children: higher pitch than adults.

HUMAN SPEECH MODEL

The vocal tract is commonly modeled as a time-varying linear system.

vocal cords \longrightarrow vocal tract \longrightarrow speech

The vocal tract contains

- 1. larynx
- 2. pharynx
- 3. mouth cavity
- 4. nasal cavity

Generation of voiced speech:

- 1. The vocal cords generate the excitation signal and the vocal tract affects the sound of the speech.
- 2. As the mouth cavity changes shape, the sound of the speech changes.
- 3. The changing shape of the mouth cavity requires that it be modeled using a time-varying system.
- 4. However, over short intervals (10-30 milliseconds) it can be modelled as a time-invariant system.

The most commonly used model for the vocal tract is an all-pole LTI system.

$$
H(z) = \frac{1}{1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_p z^{-p}}
$$

 p is the model order.

Typical values of p are from 8 to 12.

$$
e(n) \longrightarrow \boxed{H(z)} \longrightarrow y(n)
$$

 $e(n)$ is called the excitation signal.

$$
y(n) = e(n) - a_1 y(n-1) - a_2 y(n-2) - \cdots - a_p y(n-p).
$$

- 1. This is an *autoregressive* (AR) model for the signal $y(n)$.
- 2. The problem: Given a signal (eg: a speech signal) $s(n)$, find coefficients a_k and a simple excitation signal $e(n)$ so that the output $y(n)$ is close to the given signal $s(n)$.
- 3. Then to represent $s(n)$ it is necessary only to save the coefficients a_k and the parameters of the excitation signal $e(n)$.

In spectral estimation, an AR system driven by a white noise is used to model a wide sense stationary random signal. In speech coding, the driving signal (excitation signal) is instead a quasi-periodic impulse train. However, we can still use the Yule-Walker equations to estimate the coefficients a_k for $1 \leq k \leq p$. LPC Speech compression consists of parts.

- 1. Segment the sampled speech signal into short intervals (10-30 milliseconds long). These segments are called frames and can be overlapping or nonoverlapping.
- 2. For each frame, compute the LPC parameters $(a_k$ for $1 \leq k \leq p)$ from the data. This can be done by solving the Yule-Walker equations, or by other related methods.
- 3. Compute the excitation signal $e(n)$.
- 4. Model the excitation with a small number of parameters (its pitch and amplitude during the frame). Sometimes a secondary excitation signal is used as well.
- 5. Quantize and code the parameters:
	- (a) the LPC coefficients $(a_k \text{ for } 1 \leq k \leq p)$
	- (b) the parameters of the excitation signal
	- (c) the parameters of the secondary excitation signal (if used)

To compute the LPC coefficients of a frame:

- 1. Let $b(n)$ denote the current frame.
- 2. Compute the autocorrelation of the frame $b(n)$

$$
r(n) = b(n) * b(-n)
$$

(We only need to compute the values $r(n)$ for $0 \le n \le p$.)

3. Solve the Yule-Walker equations. When $p = 4$ the Yule-Walker equations are:

For each frame there will be p LPC parameters.

The LPC parameters describe a different system $H(z)$ for each frame,

$$
H(z) = \frac{1}{1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_p z^{-p}}
$$

In the following Matlab code, each successive 160-point frame is extracted from the sampled speech signal, its autocorrelation function is computed, the Yule-Walker equations are solved, and the frequency response of the all-pole filter is plotted.

```
N = 160;p = 10;for k = 1:20n = (k-1)*N+[1:N];frame = s(n);
   corr = xcorr(frame)/N;r = corr(N:N+p);R = \text{toeplitz}(r(1:p));a = [1; -R\rr(2:p+1)'];
   [H,w] = \text{freqz}(1,a);plot(w/pi*Fs/2,20*log10(abs(H)))title('FREQUENCY RESPONSE OF H(z) in dB')
   xlabel('FREQUENCY (HERZ)')
end
```
This Matlab code can be shortened by using the Matlab function $a = 1pc(fname,p)$, which efficiently solves the Yule-Walker equations using the Levinson-Durbin algorithm. Other related Matlab functions are: levinson and aryule.

Example

- 1. Segment the signal 'why' into segments of 160 samples.
- 2. In this case, because $F_s = 11025$, the duration of each segment is about 14.5 milliseconds.
- 3. Use $p = 10$.

Frame 11

COMPUTING THE LPC COEFFICIENTS

Frame 16

The frequencies where peaks occur are called formants.

$$
y(n) = h(n) * e(n)
$$

$$
Y(z) = H(z) E(z)
$$

$$
E(z) = Y(z)/H(z)
$$

$$
y(n) \longrightarrow \boxed{1/H(z)} \longrightarrow e(n)
$$

 $1/H(z)$ is an FIR filter,

$$
\frac{1}{H(z)} = 1 + a_1 z^{-1} + a_2 z^{-2} + \dots + a_p z^{-p}
$$

To find an excitation signal that makes $y(n) = s(n)$, the excitation signal $e(n)$ can be found by filtering the $s(n)$ with the FIR filter $1/H(z)$.

The quantized excitation signal consists of a sequence of amplitudes and intervals.

THE QUANTIZED EXCITATION SIGNAL

