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EE225D

Spring, 1999

LPC Analysis

Lecture 22

Hearing and Speech Engineering

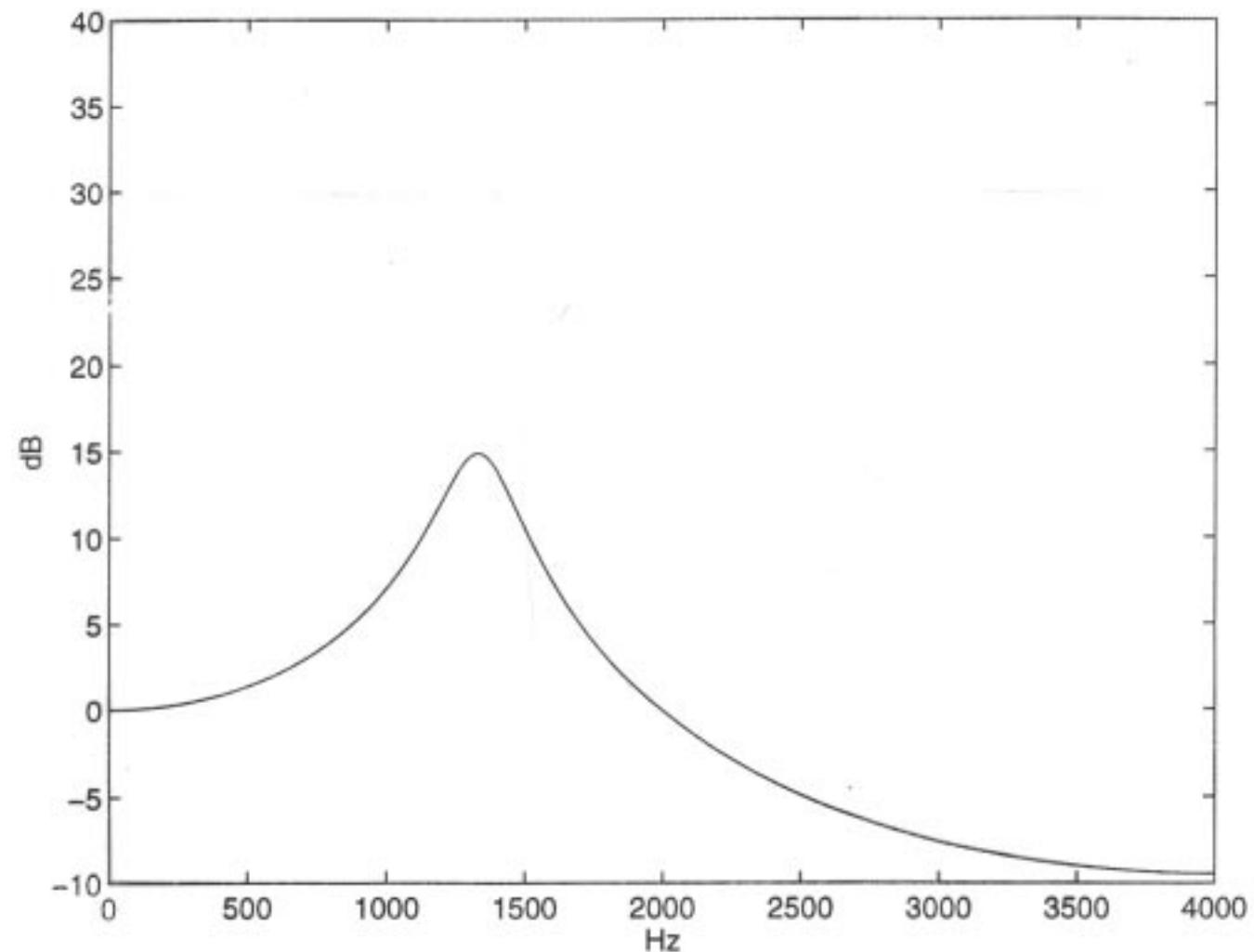
- Focus on power spectrum (not phase)
- Spectral envelope for phonetic discrimination
- Less accuracy required at high frequencies
- Emphasis on spectral peaks

Spectral Envelope Estimation

- Filter banks
- Cepstral Analysis
- Linear Predictive Coding (LPC)

Incorporate Production

- Assume simple excitation /vocal tract model
- Assume vocal tract like series resonators
- Find best spectrum based on resonators



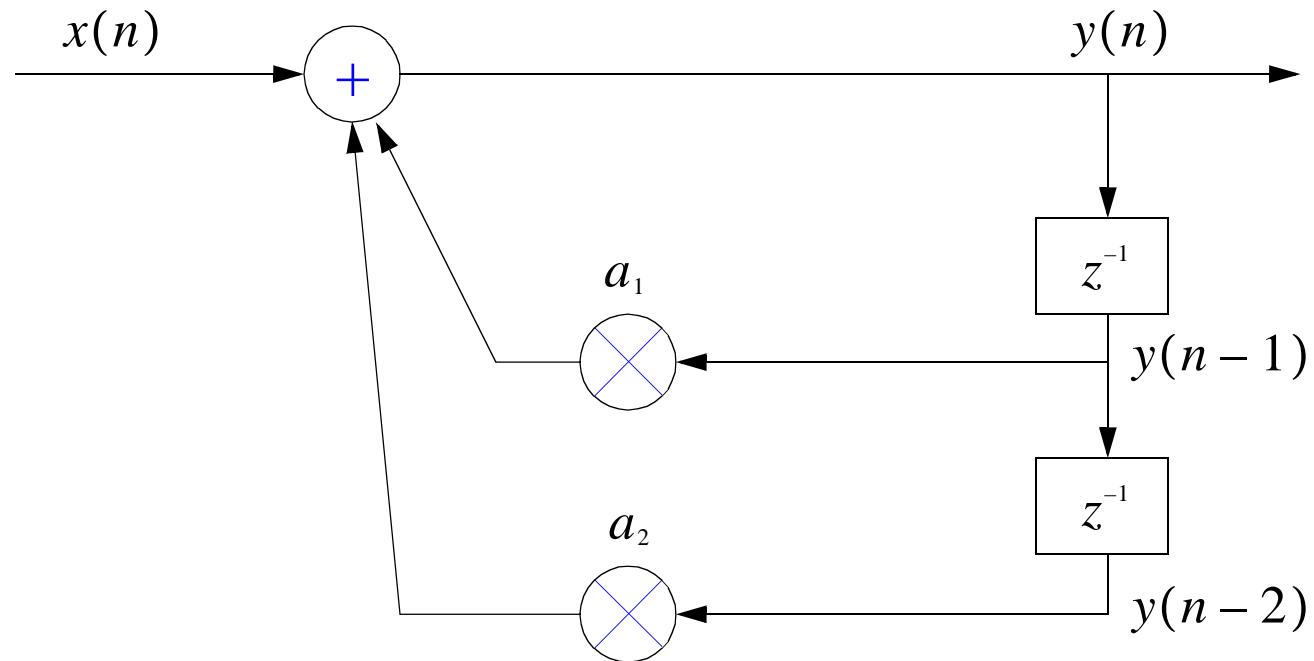
Pole-only resonator

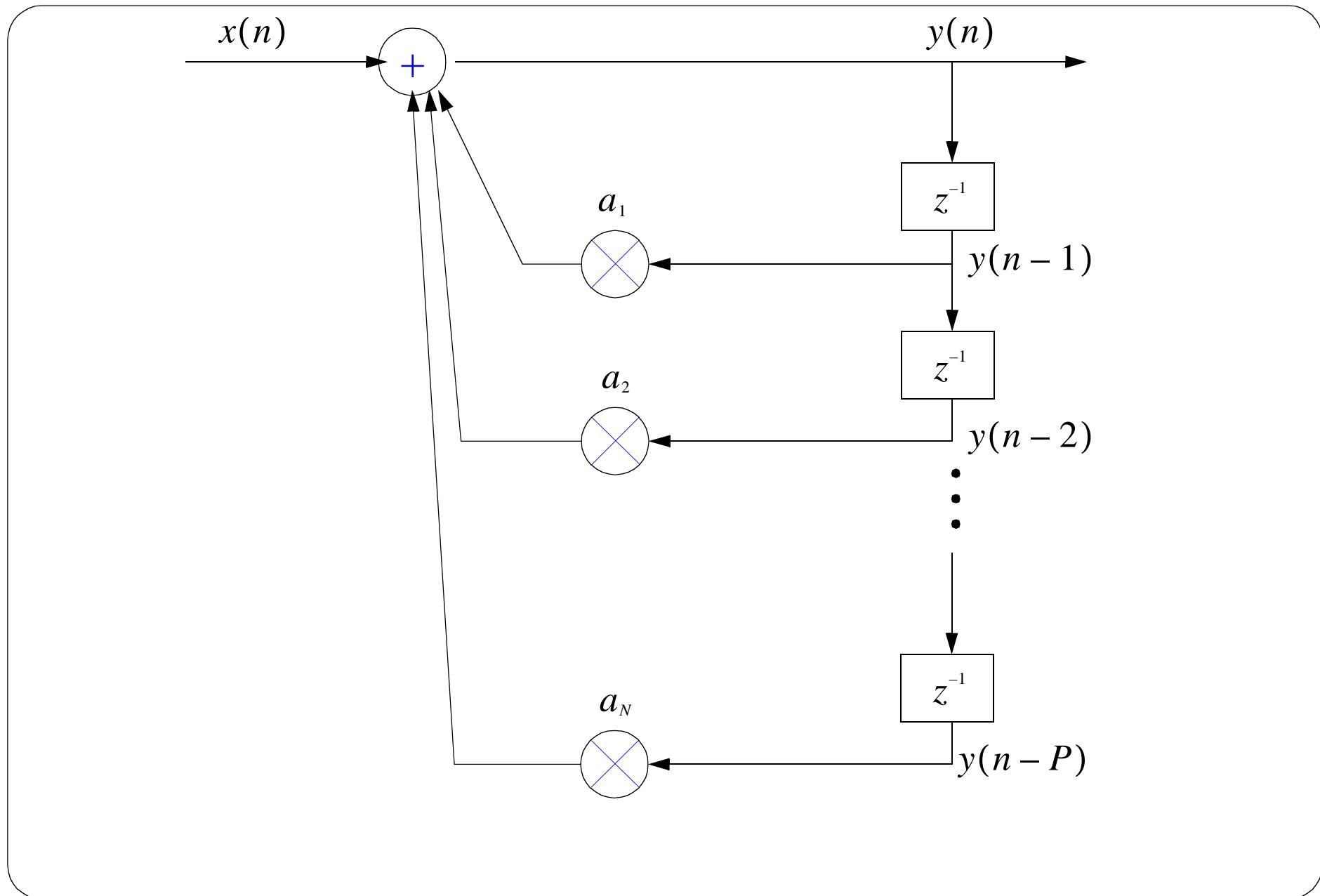
- $H_i(z) = \frac{1}{1 - b_i z^{-1} - c_i z^{-2}}$

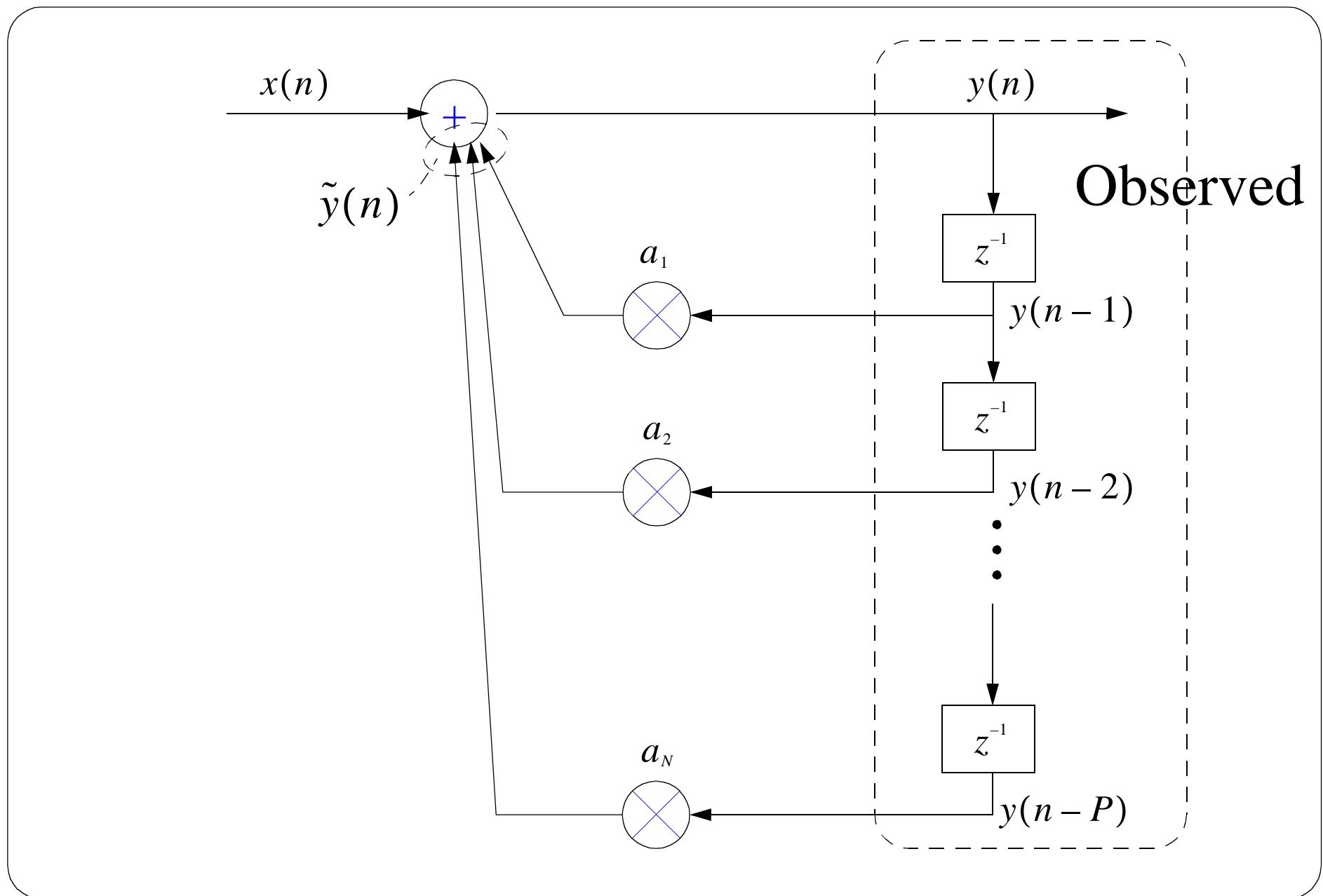
For complex pole pair,

- $b_i = 2r\cos\Theta, c_i = r^2$

where r is pole magnitude and Θ is pole angle.







Error Signal

$$e(n) = y(n) - \tilde{y}(n) = y(n) - \sum_{j=1}^P a_j y(n-j)$$

$$\begin{aligned} E(z) &= Y(z) - Y(z) = Y(z) - \sum_{j=1}^P a_j z^{-j} Y(z) \\ &= Y(z) \left(1 - \sum_{j=1}^P a_j z^{-j} \right) \end{aligned}$$

$\frac{1}{H(z)}$

or

$$E(z) = \frac{Y(z)}{H(z)}$$

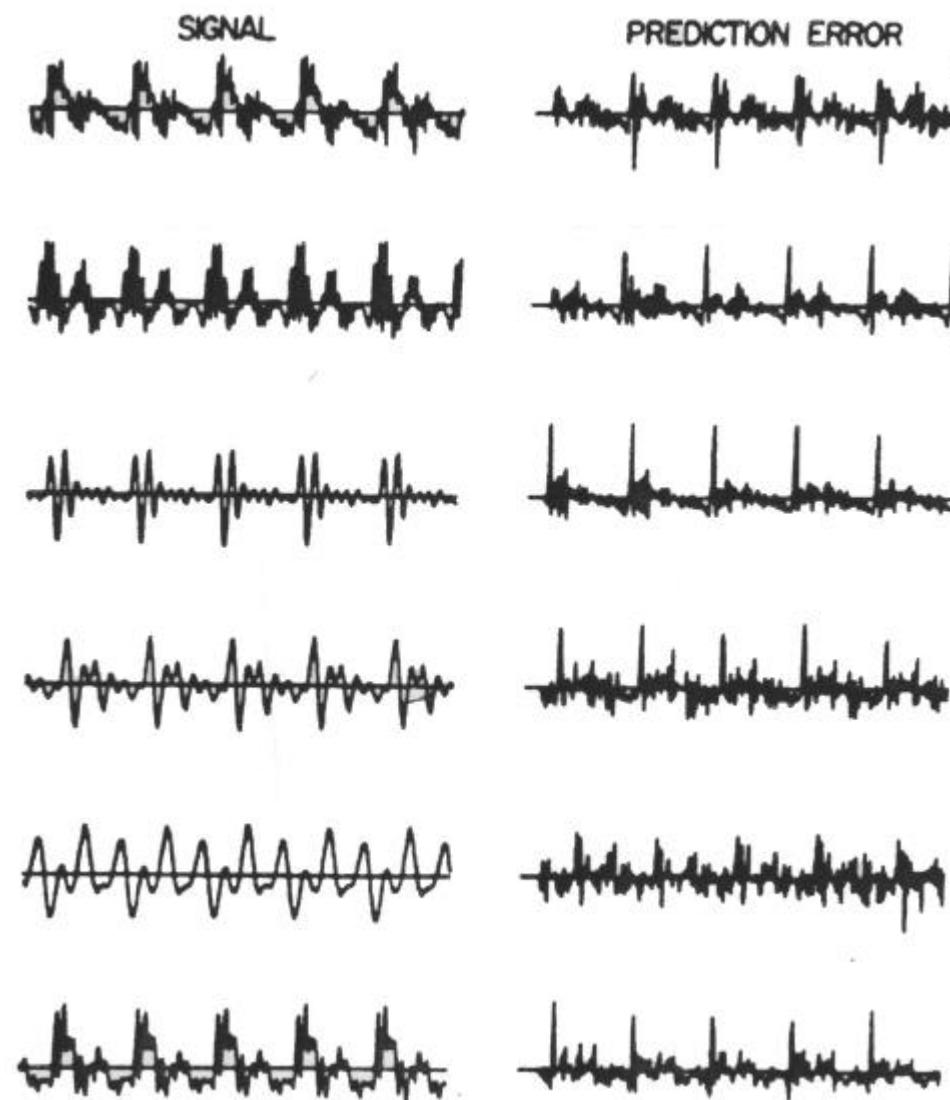


Figure 21.3 : Residual error waveforms for several vowels.

Some LPC Issues

- Error criterion for minimization
- Model order

Error Criterion

$$D = \sum_{n=0}^{N-1} e^2(n) = \int_{-\pi}^{\pi} |E(\omega)|^2 \frac{d\omega}{2\pi}$$

SO

$$D = \int_{-\pi}^{\pi} \frac{|Y(\omega)|^2}{|H(\omega)|^2} \frac{d\omega}{2\pi}$$

LPC peak modeling

- Total error constrained to be (at best) gain factor squared
- Error where model spectrum is larger contributes less
- Tends to “hug” peaks

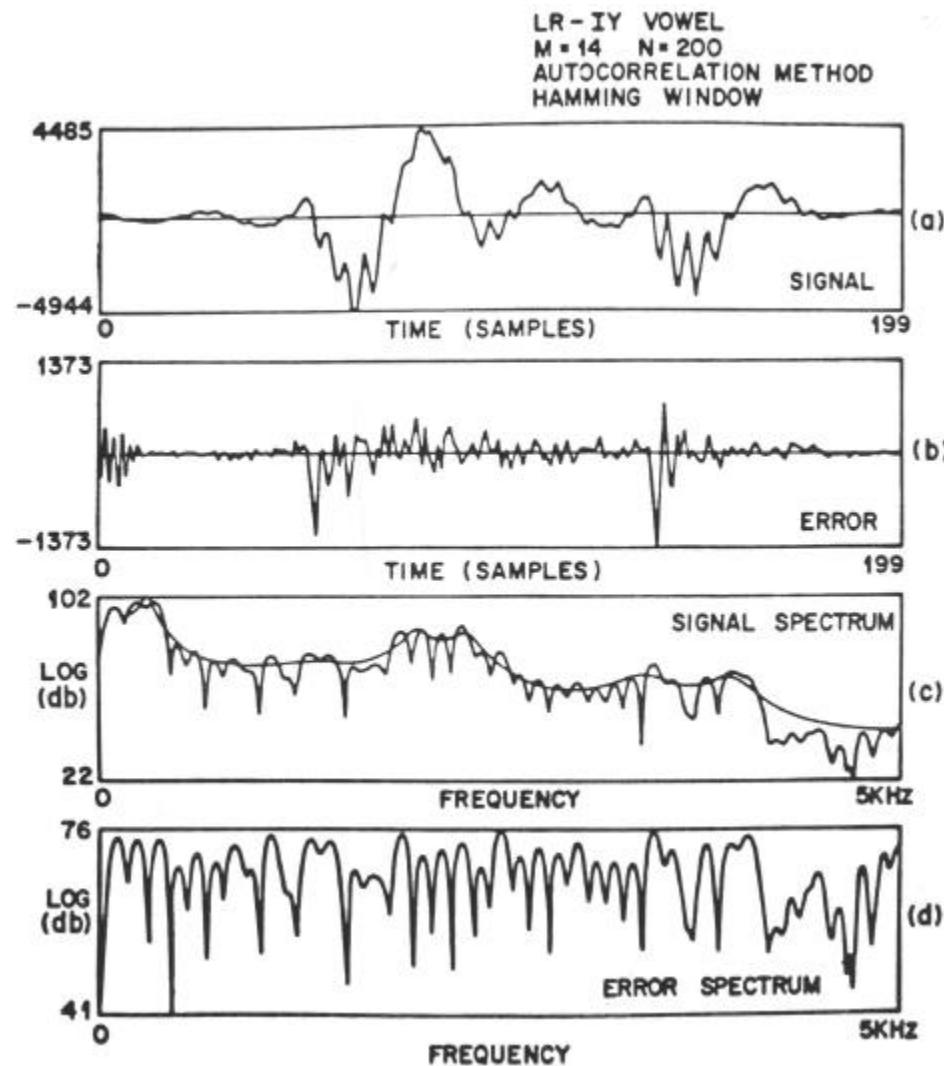


Figure 3.32 : Typical signals and spectra for LPC autocorrelation method for a segment of speech spoken by a male speaker (after Rabiner et al.)

More Effects of Error Criterion

- Globally tracks, but worse match in log spectrum for low values
- “Attempts” to model anti-aliasing filter
- Ill conditioned for wide range of values

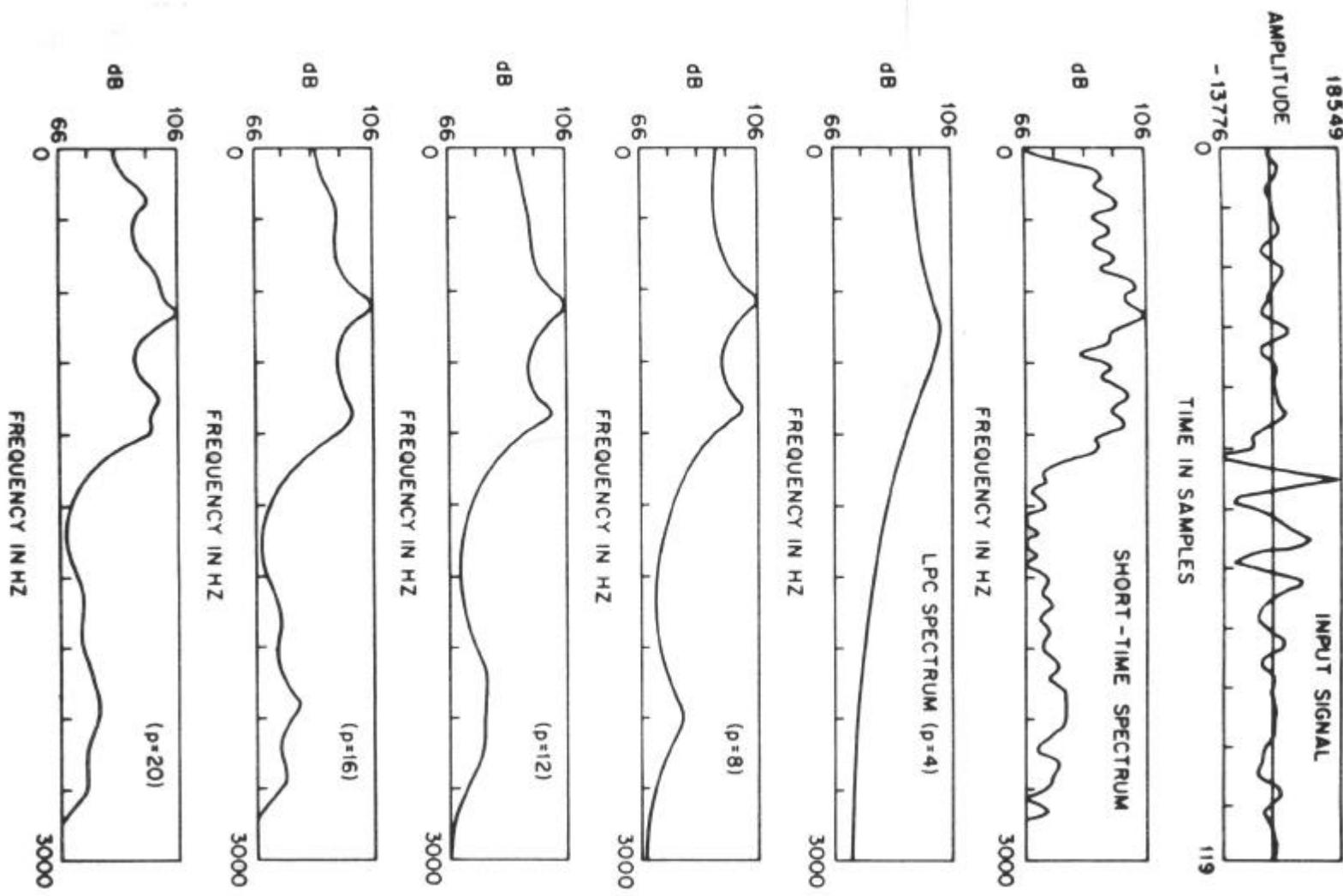
Other LPC properties

- Behavior in noise
- Sharpness of peaks
- Speaker dependence

Model Order

- Too few, can't represent formants
- Too many, model detail, e.g., harmonics
- Too many, low error, ill-conditioned matrices

Figure 3.36 : Spectra for a vowel sound for several values of predictor order, p.



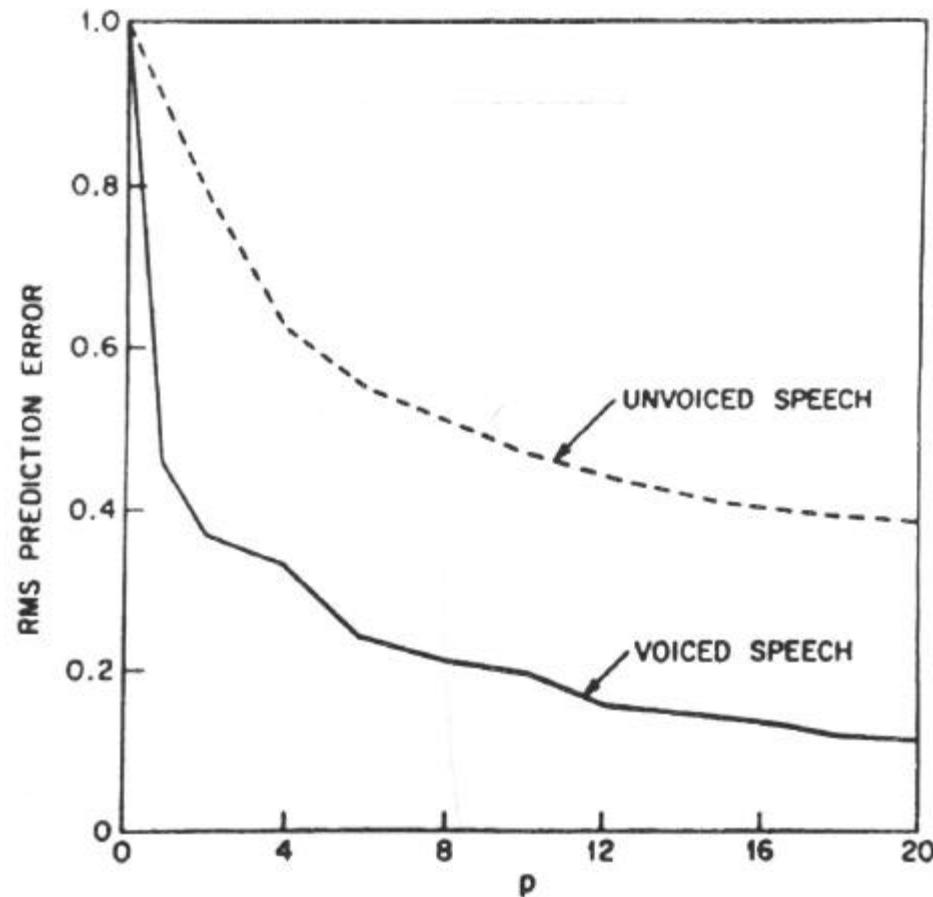


Figure 21.6 : RMS prediction error for different model orders.

Optimal Model Order

- Akaike Information Critetion (AR time series)
- Cross-validation (trial and error)

Coefficient Estimation

- Minimize squared error - set derivatives to zero
- Compute in blocks or on-line
- For blocks, use autocorrelation or covariance methods (windowing)

Minimizing the error

$$\bullet D = \sum_{n=0}^{N-1} e^2(n) = \sum_{n=0}^{N-1} \left(y(n) - \sum_{j=1}^P a_j y(n-j) \right)^2$$

If we take partial derivatives with respect to each a , we get P equations of the form.

$$\bullet \sum_{j=1}^P a_j \phi(i, j) = \phi(i, 0) \text{ for } i = 1, 2, \dots, P$$

Where $\phi(i, j)$ is a correlation sum between versions of the speech signal delayed by i and j points.

Solving the equations

- Autocorrelation method : Levinson or Durbin recursions ($O(P^2)$ operations) ; uses Toeplitz property, guaranteed stable
- Covariance method : Cholesky decomposition ($O(P^3)$ operations) - just uses symmetric property

LPC-based representations

- Predictor Polynomial
- Root pairs
- Reflection coefficients
- Log area ratios
- Cepstrum

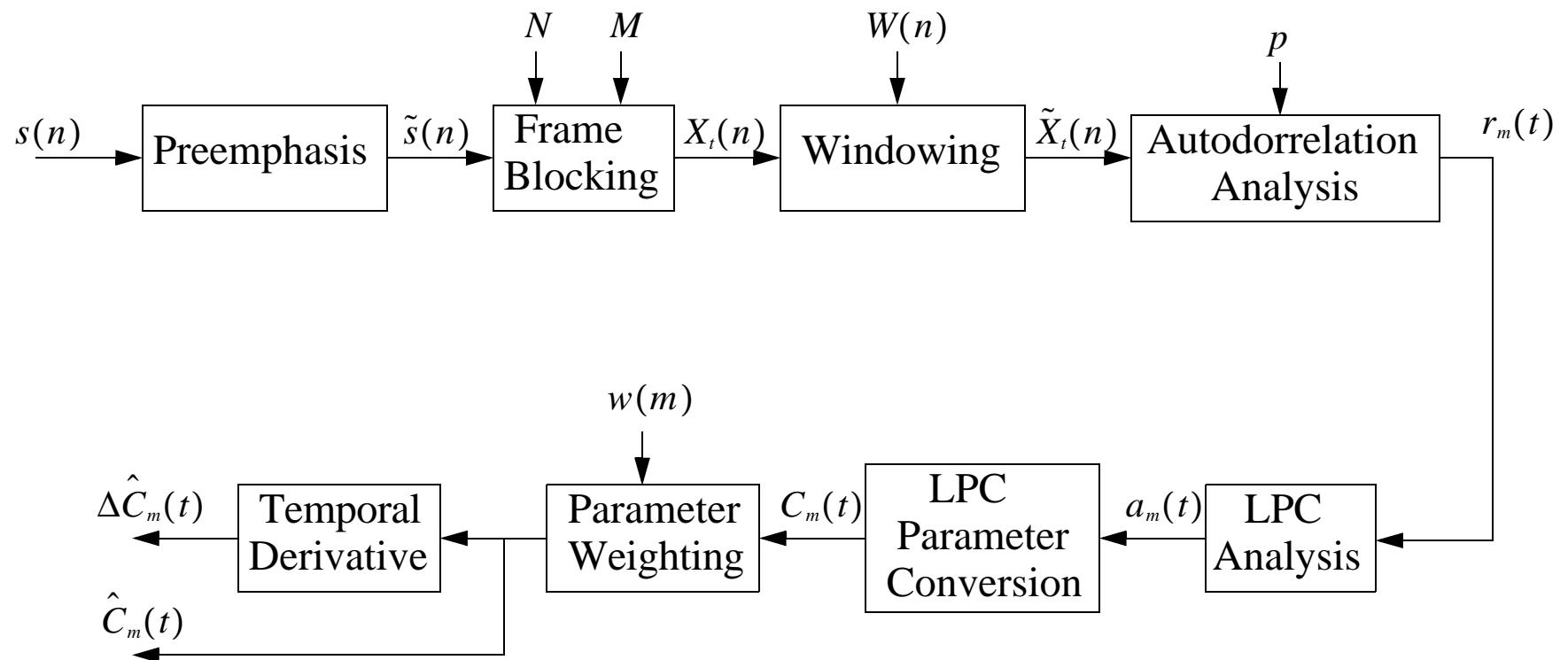


Figure 3.37 : Block diagram of LPC processor for speech recognition.

	Filter Banks	Cepstral Analysis	LPC
Reduced pitch effects	X	X	X
Excitation estimate		X	X
Direct access to spectra	X		
Less resolution at HF	X		
Orthogonal outputs		X	
Peak-hugging property			X
Reduced computation			X

Table 1 : Basic methods for spectral envelope estimation in speech.