

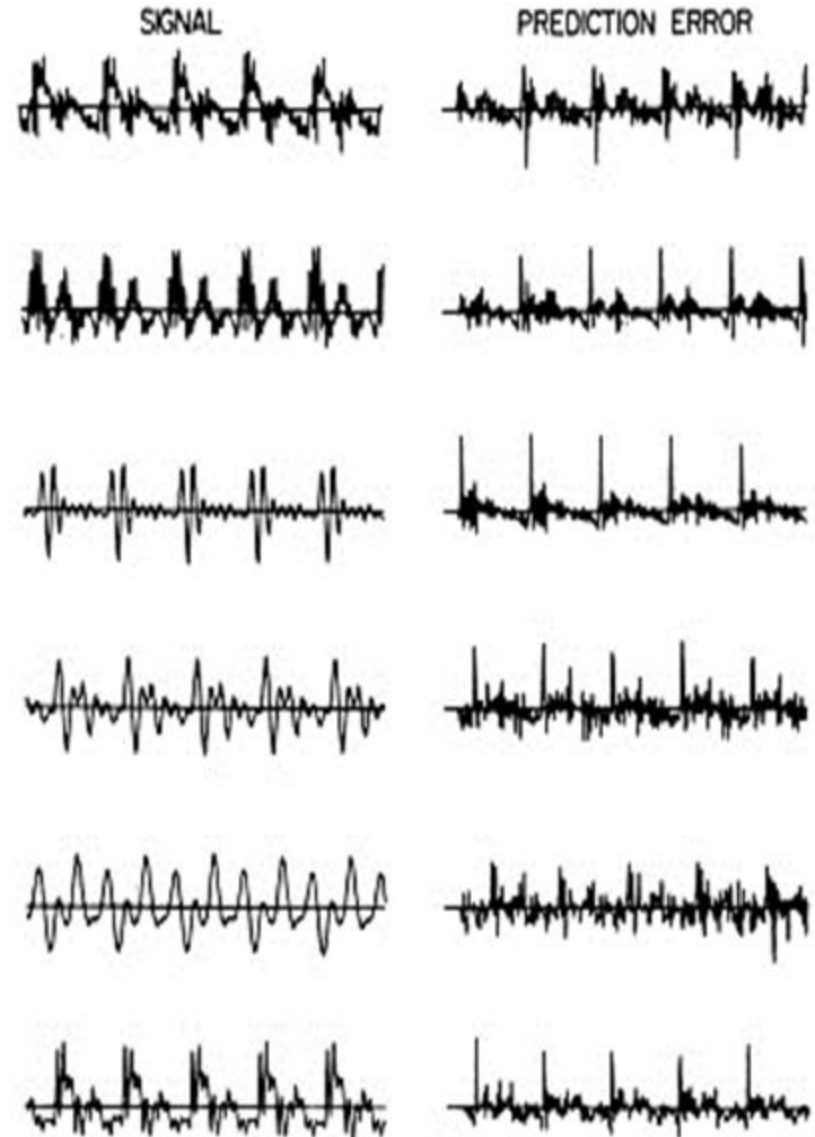
The Prediction Error Signal

Prediction Error Signal Behavior

- the prediction error signal is computed as

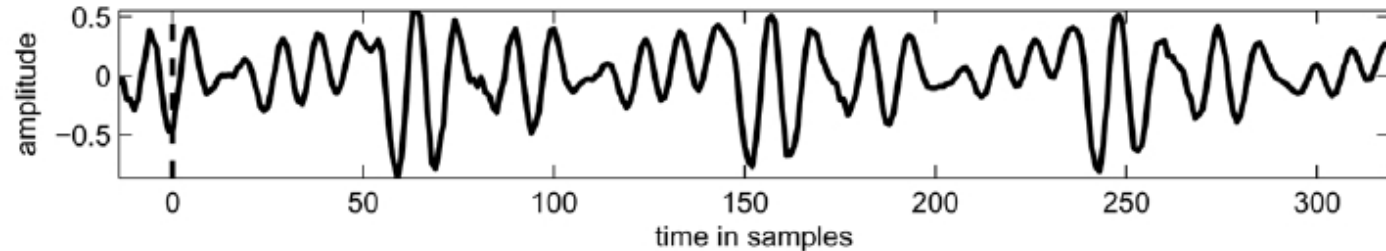
$$e(n) = s(n) - \sum_{k=1}^p \alpha_k s(n-k) = Gu(n)$$

- $e(n)$ should be large at the beginning of each pitch period (voiced speech) => good signal for pitch detection
- can perform autocorrelation on $e(n)$ and detect largest peak
- error spectrum is approximately flat-so effects of formants on pitch detection are minimized

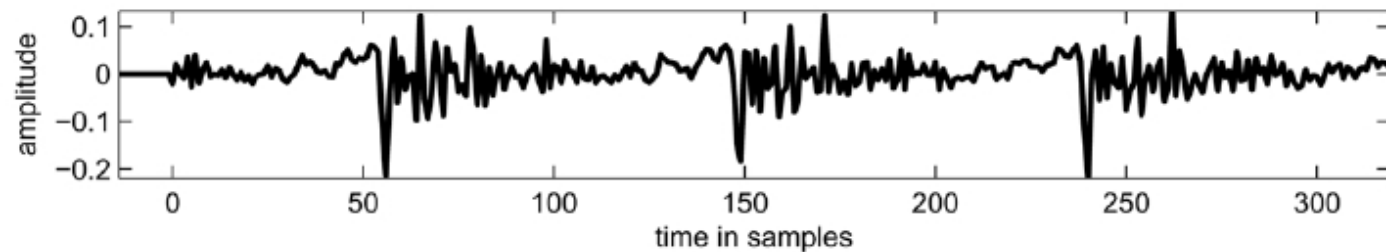


LP Speech Analysis

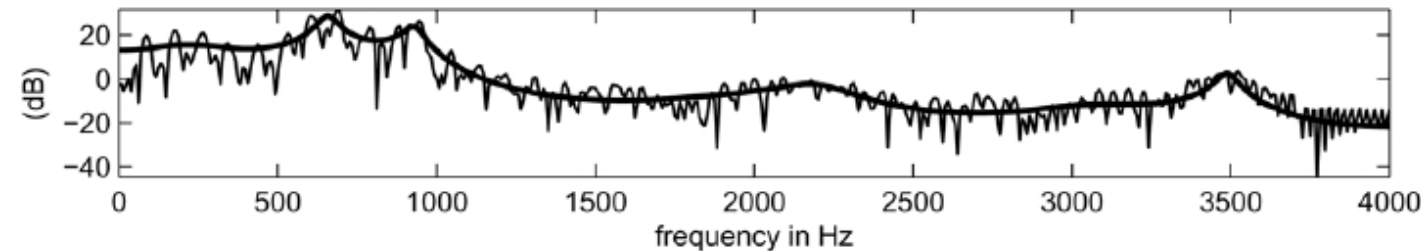
file:s5, ss:11000, frame size (L):320, lpc order (p):14, cov method



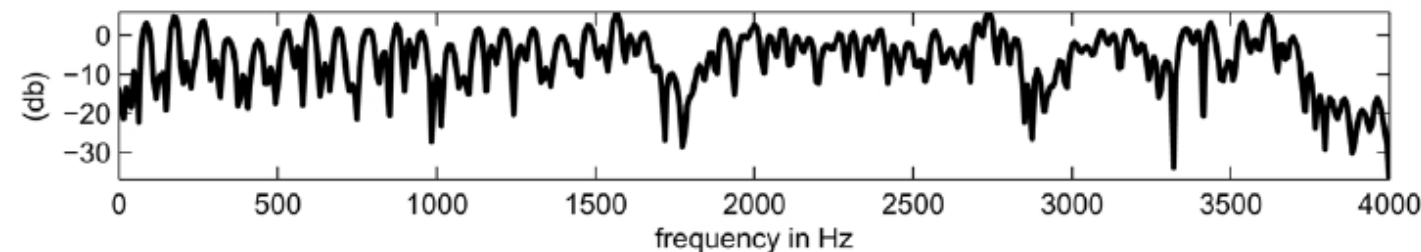
Top panel:
speech signal



Second panel:
error signal



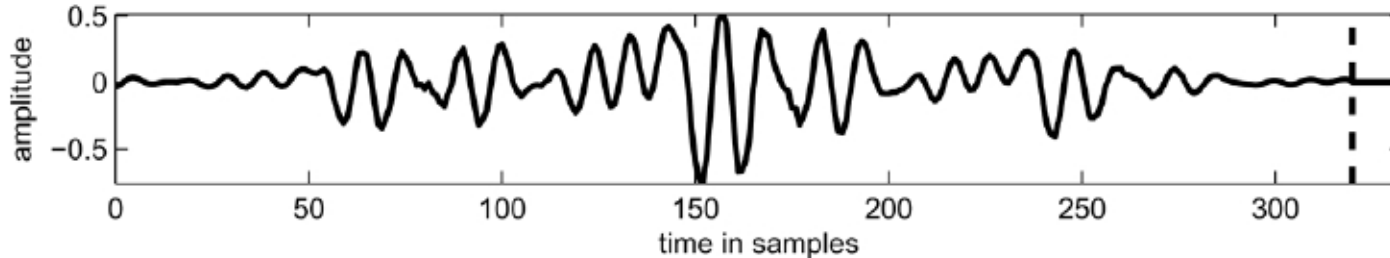
Third panel:
log magnitude
spectra of signal
and LP model



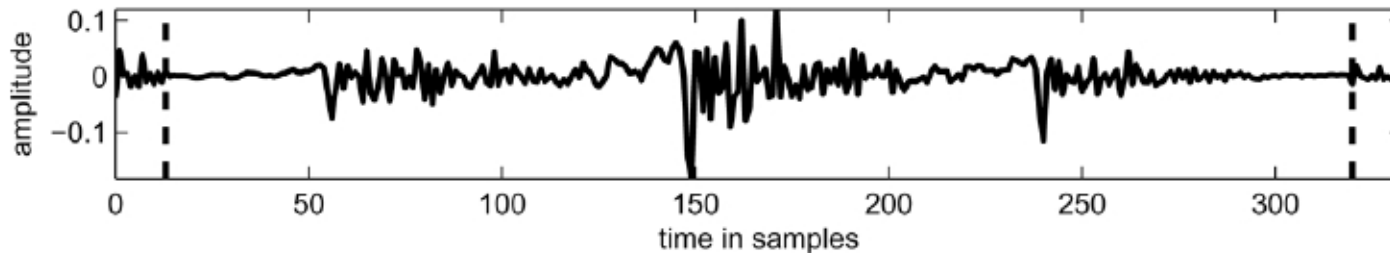
Fourth panel:
log magnitude
spectrum of error
signal

LP Speech Analysis

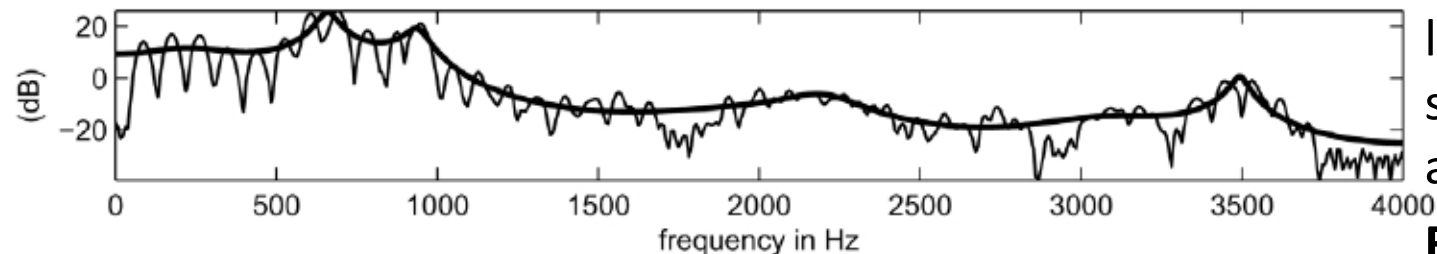
file:s5, ss:11000, frame size (L):320, lpc order (p):14, ac method



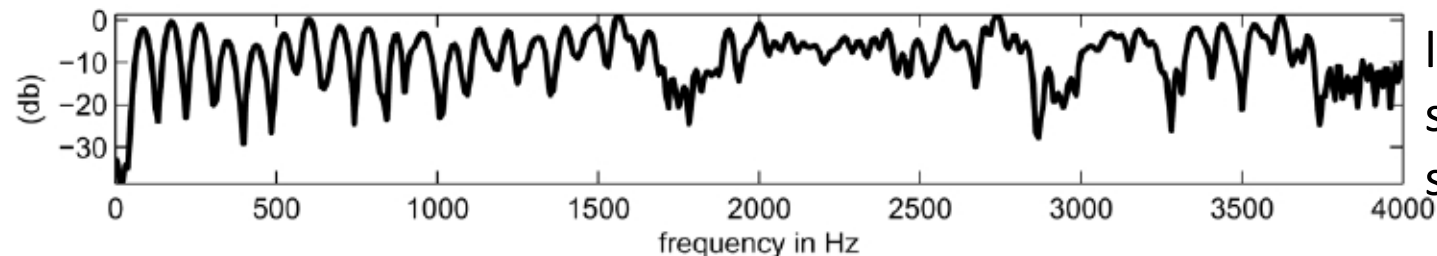
Top panel:
speech signal



Second panel:
error signal



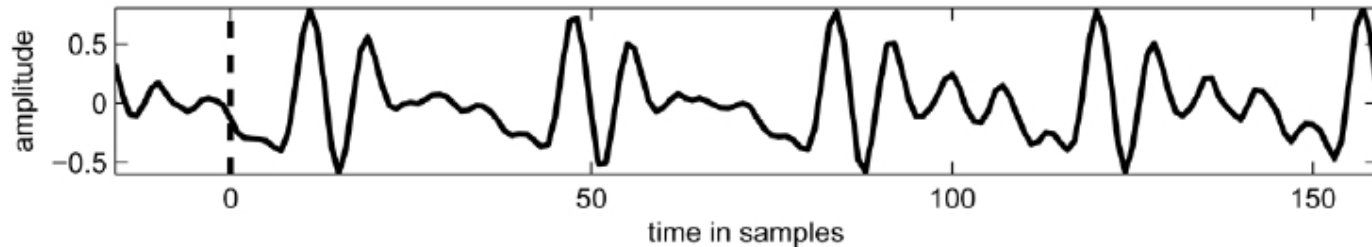
Third panel:
log magnitude
spectra of signal
and LP model



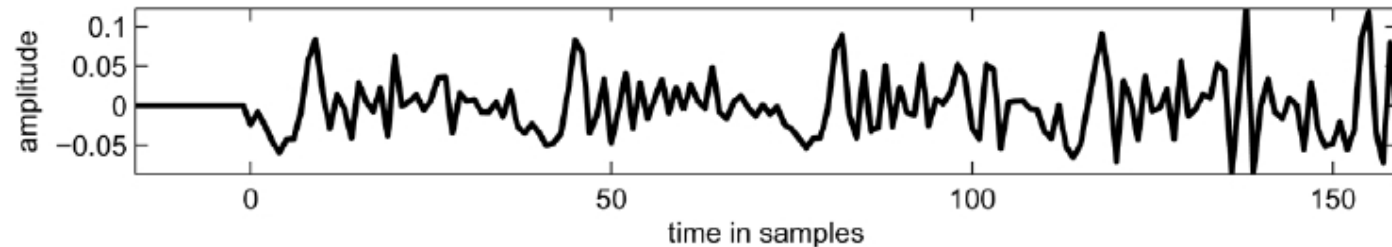
Fourth panel:
log magnitude
spectrum of error
signal

LP Speech Analysis

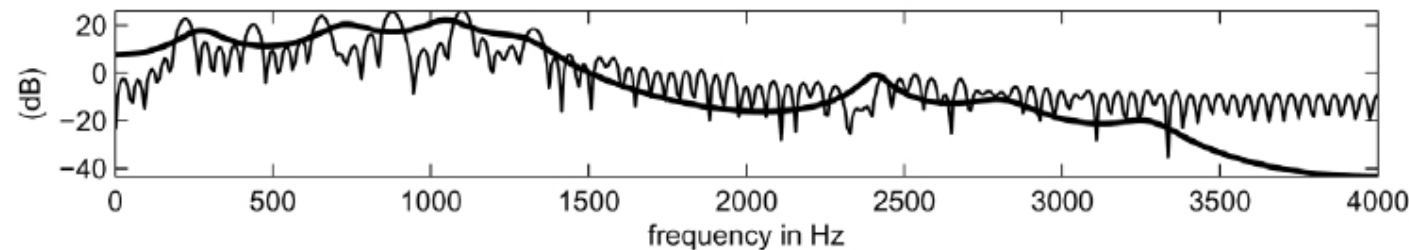
file:s3, ss:14000, frame size (L):160, lpc order (p):16, cov method



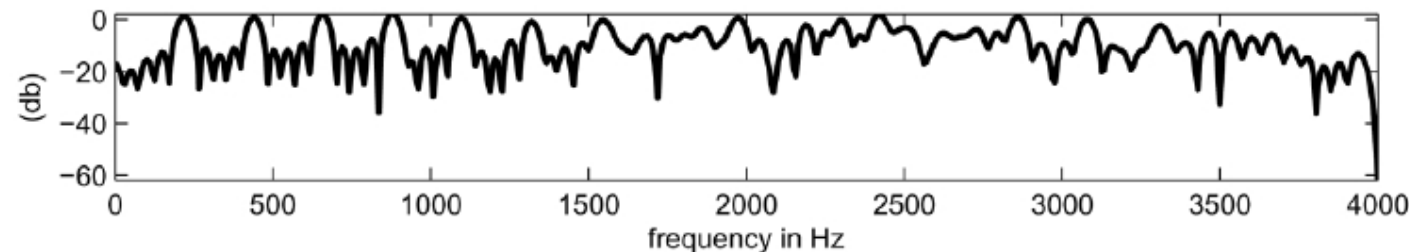
Top panel:
speech signal



Second panel:
error signal



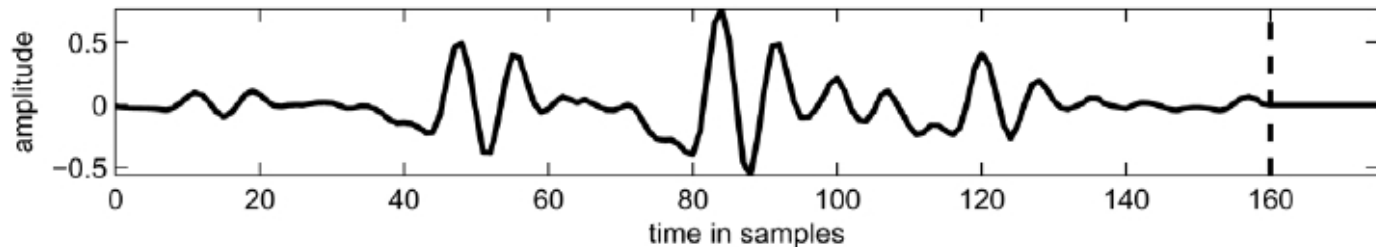
Third panel:
log magnitude
spectra of signal
and LP model



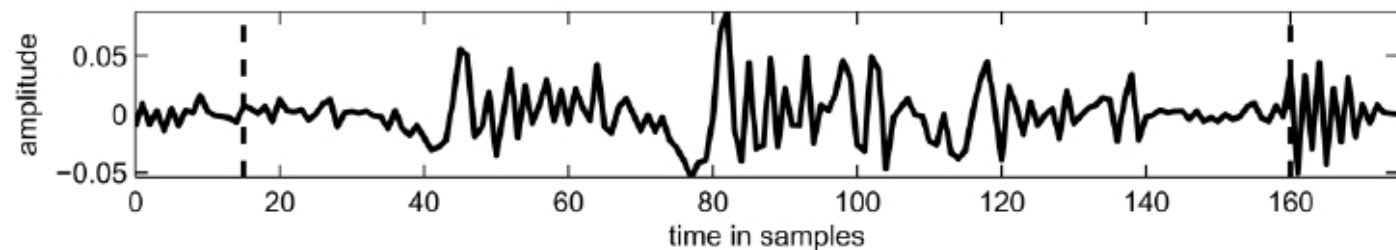
Fourth panel:
log magnitude
spectrum of error
signal

LP Speech Analysis

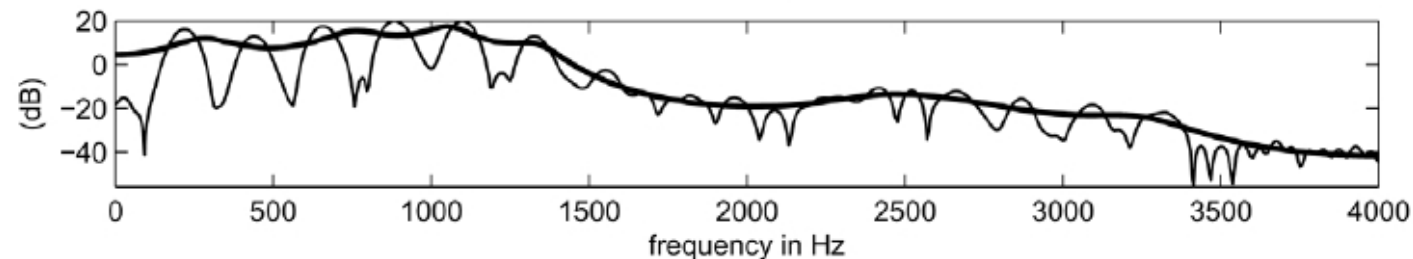
file:s3, ss:14000, frame size (L):160, lpc order (p):16, ac method



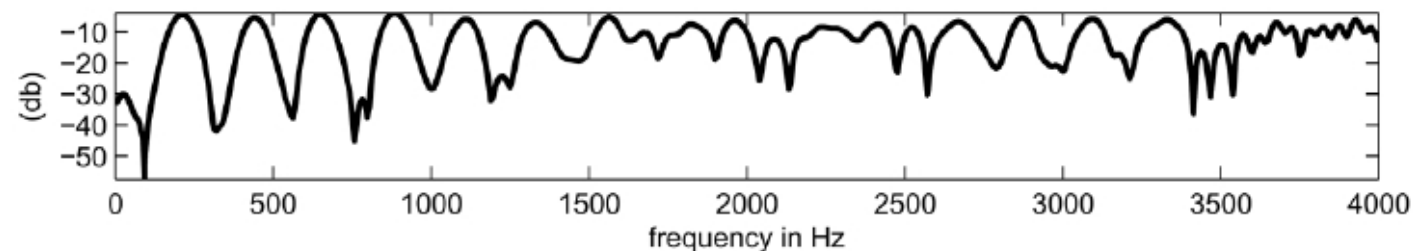
Top panel:
speech signal



Second panel:
error signal



Third panel:
log magnitude
spectra of signal
and LP model



Fourth panel:
log magnitude
spectrum of error
signal

Properties of the LPC Polynomial

Minimum-Phase Property of $A(z)$

$A(z)$ has all its zeros inside the unit circle

Proof: Assume that z_o ($|z_o|^2 > 1$) is a zero (root) of $A(z)$

$$A(z) = (1 - z_o z^{-1}) A'(z)$$

The minimum mean-squared error is

$$\begin{aligned} E_{\hat{n}} &= \sum_{m=-\infty}^{\infty} e_{\hat{n}}[m]^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |A(e^{j\omega})|^2 |S_{\hat{n}}(e^{j\omega})|^2 d\omega \quad \text{Parseval 定理} \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left| 1 - z_o e^{-j\omega} \right|^2 \left| A'(e^{j\omega}) \right|^2 \left| S_{\hat{n}}(e^{j\omega}) \right|^2 d\omega > 0 \\ \left| 1 - z_o e^{-j\omega} \right|^2 &= |z_o|^2 \left| 1 - (1/z_o^*) e^{-j\omega} \right|^2 \end{aligned}$$

Thus, $A(z)$ could not be the optimum filter because we could replace z_o by $(1/z_o^*)$ and decrease the error

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

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

PARCORs and Stability

- prove that $|k_i| \geq 1 \Rightarrow |z_j^{(i)}| \geq 1$ for some j

Proof: $A^{(i)}(z) = A^{(i-1)}(z) - k_i z^{-i} A^{(i-1)}(z^{-1}) = \prod_{j=1}^i (1 - z_j^{(i)} z^{-1})$

It is easily shown that $-k_i$ is the coefficient of z^{-i} in $A^{(i)}(z)$, i.e. $\alpha_i^{(i)} = k_i$.
Therefore

$$|k_i| = \prod_{j=1}^i |z_j^{(i)}|$$

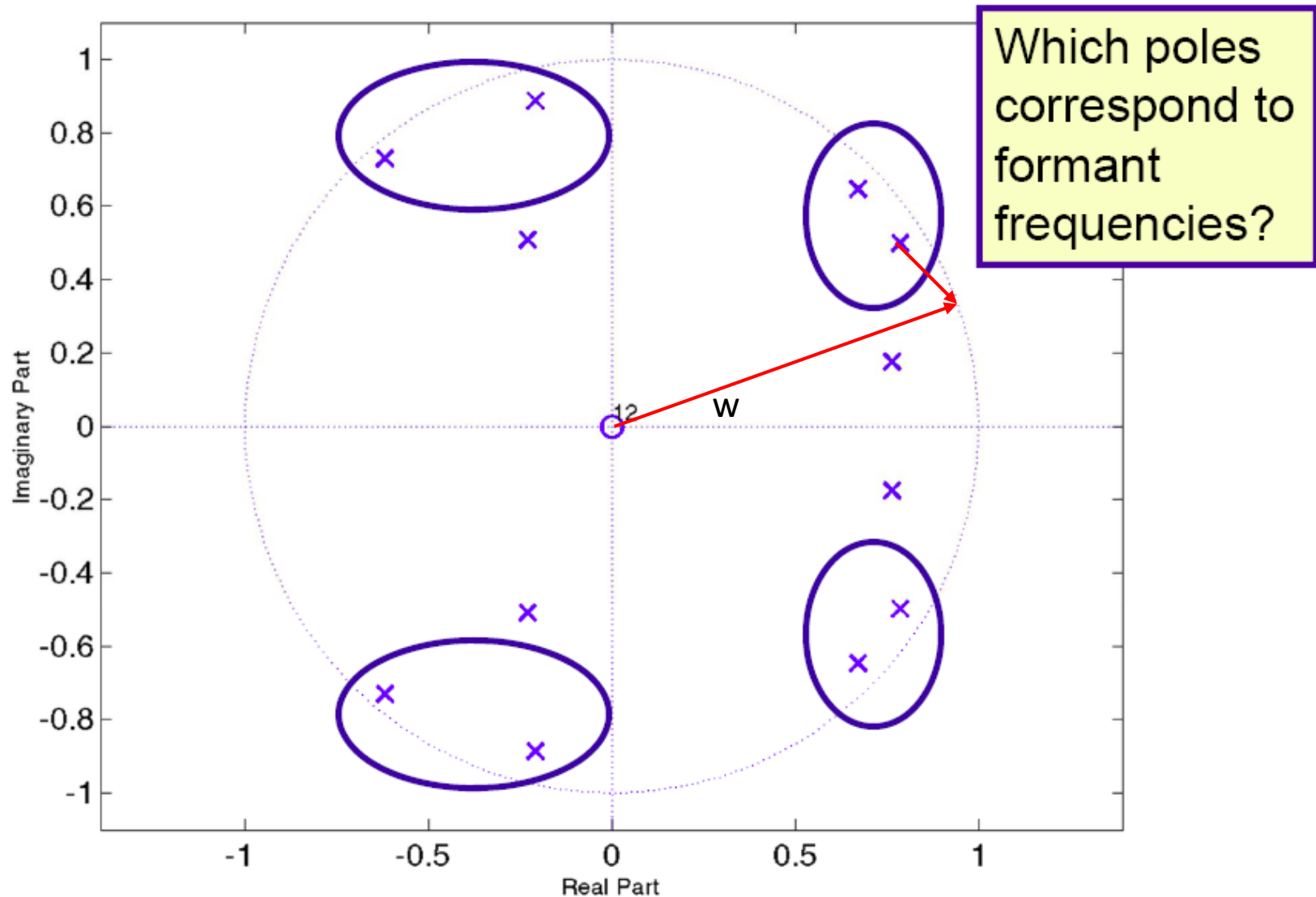
If $|k_i| \geq 1$, then either all the roots must be **on** the unit circle or at least one of them must be **outside** the unit circle

- $|k_i| < 1$ is a necessary and sufficient condition for $A(z)$ to be a minimum phase system and $1/A(z)$ to be a stable system

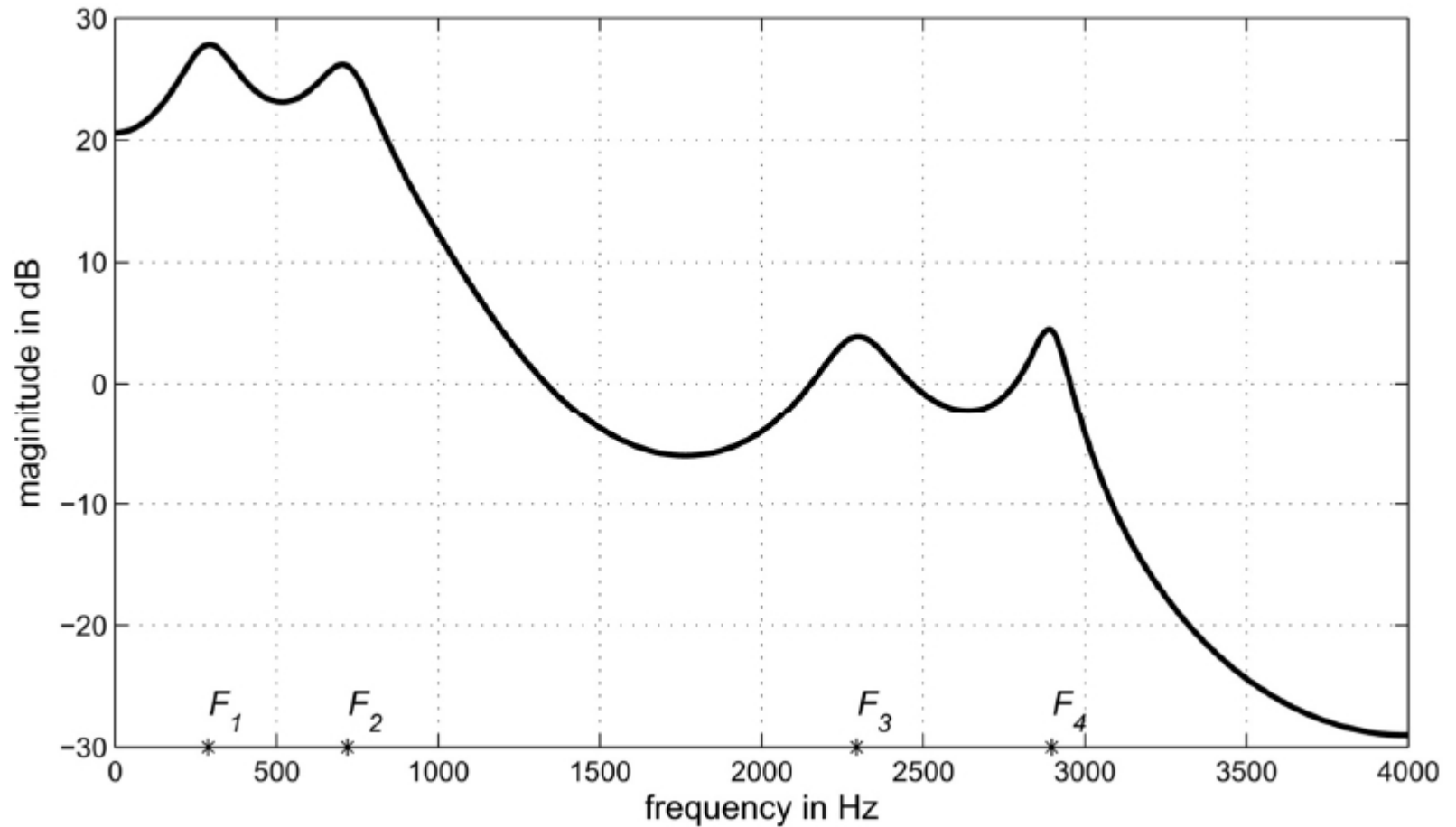
Root Locations for Optimum LP Model

$$\begin{aligned}\tilde{H}(z) &= \frac{G}{A(z)} = \frac{G}{1 - \sum_{i=1}^p \alpha_i z^{-i}} \\ &= \frac{G}{\prod_{i=1}^p (1 - z_i z^{-1})} = \frac{Gz^p}{\prod_{i=1}^p (z - z_i)}\end{aligned}$$

Pole-Zero Plot for Model



Pole Locations



Pole Locations ($F_s=10,000$ Hz)

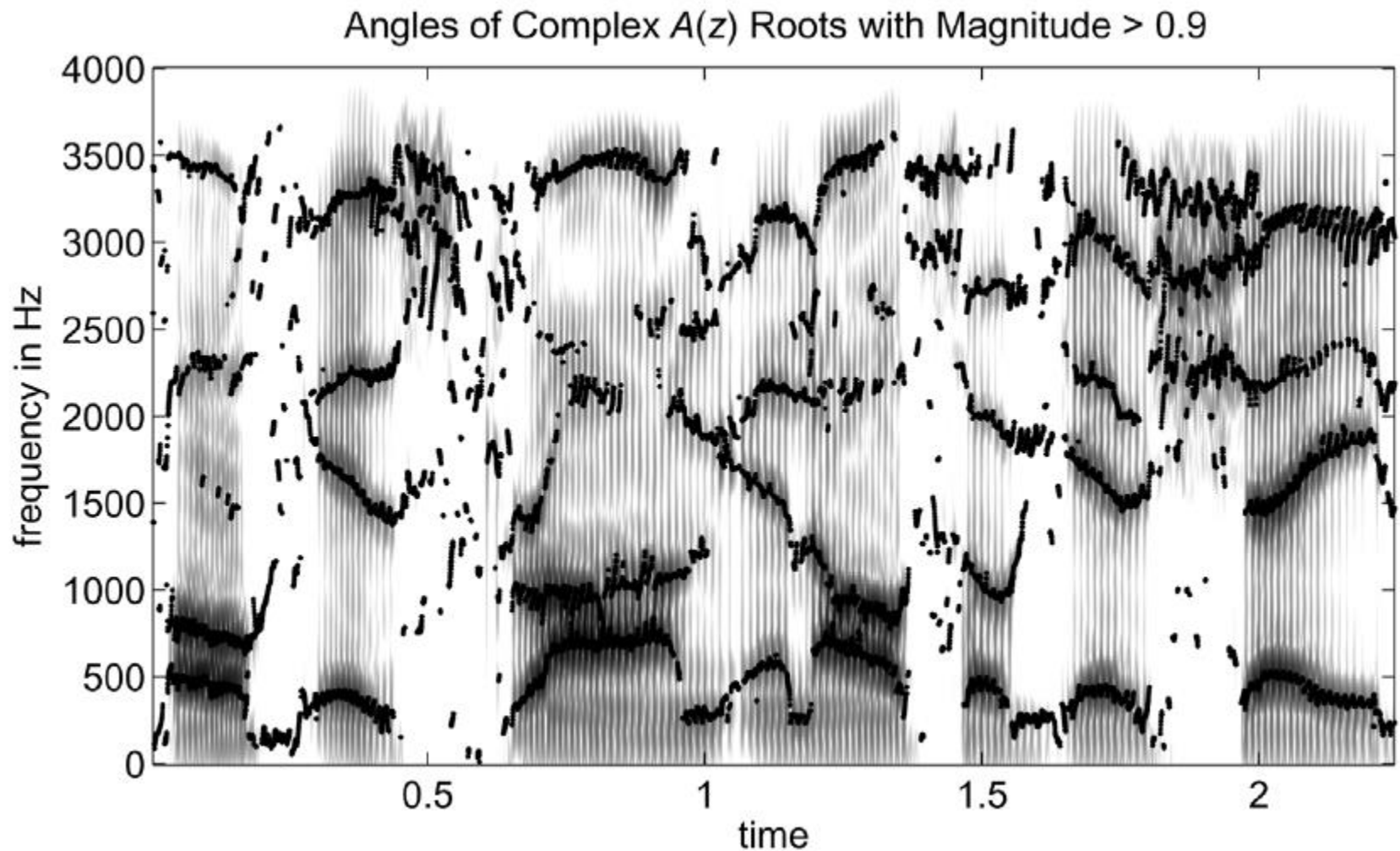
<i>root magnitude</i>	<i>θ root angle(degrees)</i>	<i>F root angle (Hz)</i>	<i>formant</i>
0.9308	10.36	288	F_1
0.9308	-10.36	-288	F_1
0.9317	25.88	719	F_2
0.9317	-25.88	-719	F_2
0.7837	35.13	976	
0.7837	-35.13	-976	
0.9109	82.58	2294	F_3
0.9109	-82.58	-2294	F_3
0.5579	91.44	2540	
0.5579	-91.44	-2540	
0.9571	104.29	2897	F_4
0.9571	-104.29	-2897	F_4

$$F = (\theta / 180) \cdot (F_s / 2)$$

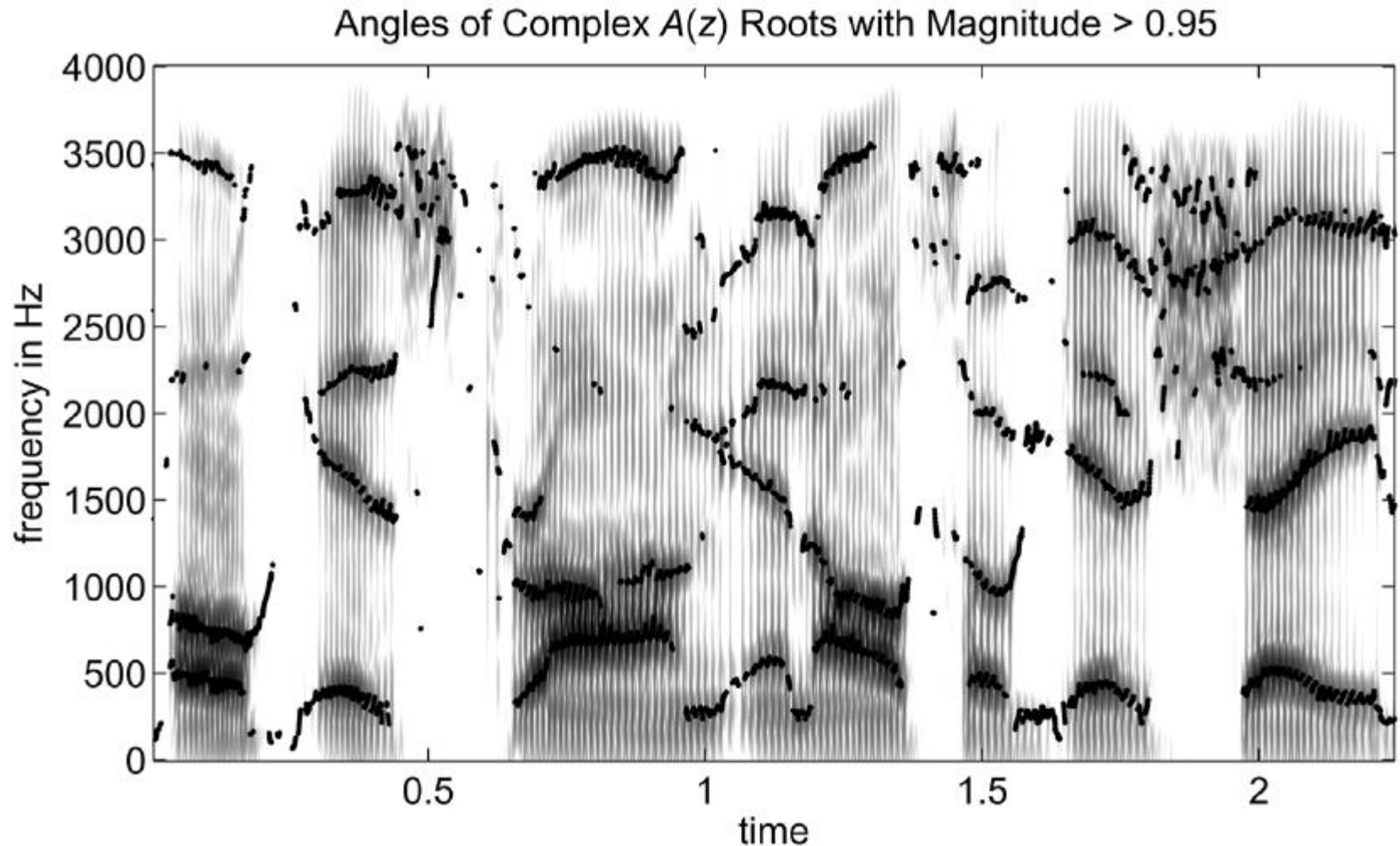
Estimating Formant Frequencies

- compute $A(z)$ and factor it
- find roots that are close to the unit circle.
- compute equivalent analog frequencies from the angles of the roots.
- plot formant frequencies as a function of time.

Spectrogram with LPC Roots



Spectrogram with LPC Roots



Alternative Representations of the LP Parameters

LP Parameter Sets

<i>Parameter Set</i>	<i>Representation</i>
LP Coefficients and Gain	$\{\alpha_k, 1 \leq k \leq p\}, G$
PARCOR Coefficients	$\{k_i, 1 \leq i \leq p\}$
Log Area Ratio Coefficients	$\{g_i, 1 \leq i \leq p\}$
Roots of Predictor Polynomial	$\{z_k, 1 \leq k \leq p\}$
Impulse Response of $H(z)$	$\{h[n], 0 \leq n \leq \infty\}$
LP Cepstrum	$\{\hat{h}[n], -\infty \leq n \leq \infty\}$
Autocorrelation of Impulse Response	$\{\tilde{R}(i), -\infty \leq i \leq \infty\}$
Autocorrelation of Predictor Polynomial	$\{R_a[i], -p \leq i \leq p\}$
Line Spectral Pair Parameters	$P(z), Q(z)$

PARCOR

- PARCORs to Prediction Coefficients
 - assume that $k_i, i=1,2, \dots, p$ are given. Then we can skip the computation of k_i in the Levinson recursion.

```
for  $i = 1, 2, \dots, p$   
     $\alpha_i^{(i)} = k_i$   
    if  $i > 1$ , then for  $j = 1, 2, \dots, i - 1$   
         $\alpha_j^{(i)} = \alpha_j^{(i-1)} - k_i \alpha_{i-j}^{(i-1)}$   
    end  
end  
 $\alpha_j = \alpha_j^{(p)} \quad j = 1, 2, \dots, p$ 
```

PARCOR

- Prediction Coefficients to PARCORs
 - assume that $\alpha_j, j=1,2, \dots, p$ are given. Then we can work backwards through the Levinson Recursion.

```

$$\alpha_j^{(p)} = \alpha_j \quad \text{for } j = 1, 2, \dots, p$$

$$k_p = \alpha_p^{(p)}$$

$$\text{for } i = p, p-1, \dots, 2$$

$$\quad \text{for } j = 1, 2, \dots, i-1$$

$$\quad \quad \alpha_j^{(i-1)} = \frac{\alpha_j^{(i)} + k_i \alpha_{i-j}^{(i)}}{1 - k_i^2}$$

$$\quad \text{end}$$

$$\quad \quad k_{i-1} = \alpha_{i-1}^{(i-1)}$$

$$\text{end}$$

```

Log Area Ratio

- log area ratio coefficients from PARCOR coefficients

$$g_i = \log \left[\frac{A_{i+1}}{A_i} \right] = \log \left[\frac{1 - k_i}{1 + k_i} \right] \quad 1 \leq i \leq p$$

with inverse relation

$$k_i = \frac{1 - e^{g_i}}{1 + e^{g_i}} \quad 1 \leq i \leq p$$

Roots of Predictor Polynomial

- roots of the predictor polynomial

$$A(z) = 1 - \sum_{k=1}^p \alpha_k z^{-k} = \prod_{k=1}^p (1 - z_k z^{-1})$$

where each root can be expressed as a z-plane i.e.,

$$z_k = z_{kr} + j z_{ki}$$

- important for formant estimation

Impulse Response of $H(z)$

- IR of all pole system

$$h(n) = \sum_{k=1}^p \alpha_k h(n-k) + G\delta(n) \quad 0 \leq n$$

LP Cepstrum

- cepstrum of IR of overall LP system from predictor coefficients

$$\hat{h}(n) = \alpha_n + \sum_{k=1}^{n-1} \left(\frac{k}{n} \right) \hat{h}(k) \alpha_{n-k} \quad 1 \leq n$$

- predictor coefficients from cepstrum of IR

$$\alpha_n = \hat{h}(n) - \sum_{k=1}^{n-1} \left(\frac{k}{n} \right) \hat{h}(k) \alpha_{n-k} \quad 1 \leq n$$

where

$$H(z) = \sum_{n=0}^{\infty} h(n) z^{-n} = \frac{G}{1 - \sum_{k=1}^p \alpha_k z^{-k}}$$

Autocorrelation of IR

- autocorrelation of IR

$$\tilde{R}(i) = \sum_{n=0}^{\infty} h(n)h(n-i) = \tilde{R}(-i)$$

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

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

Autocorrelation of Predictor Polynomial

- autocorrelation of the predictor polynomial

$$A(z) = 1 - \sum_{k=1}^p \alpha_k z^{-k}$$

with IR of the inverse filter

$$a(n) = \delta(n) - \sum_{k=1}^p \alpha_k \delta(n - k)$$

with autocorrelation

$$R_a(i) = \sum_{k=0}^{p-i} a(k)a(k+i) \quad 0 \leq i \leq p$$

Line Spectral Pairs

- Quantization of LP Parameters
- consider the magnitude-squared of the model frequency response

$$\left| H(e^{j\omega}) \right|^2 = \frac{1}{\left| A(e^{j\omega}) \right|^2} = P(\omega, g)$$

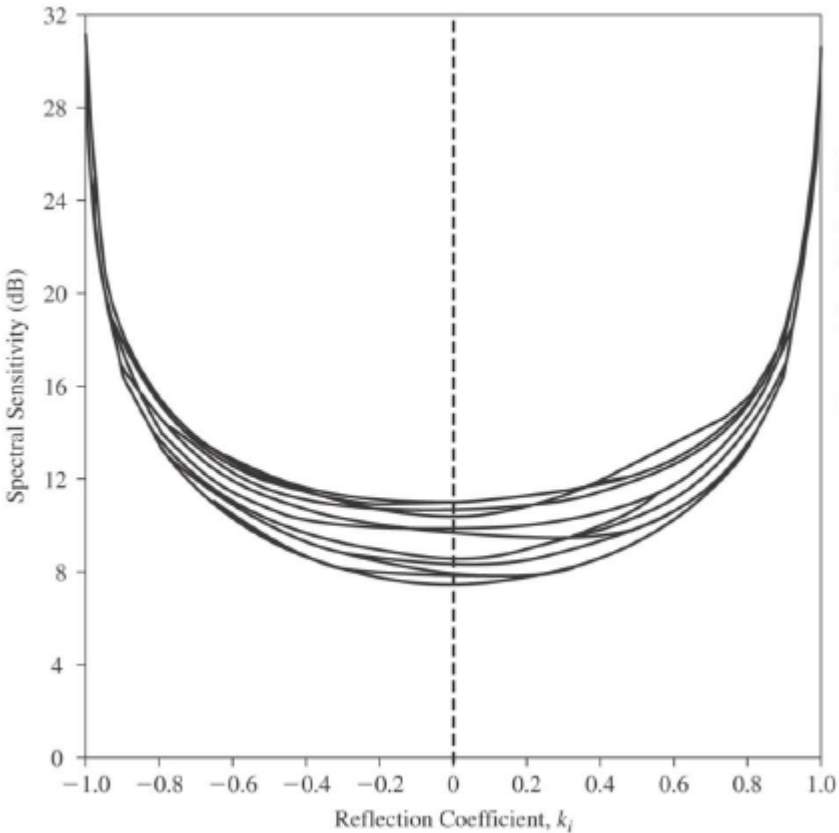
where g is a parameter that affects P .

- spectral sensitivity can be defined as

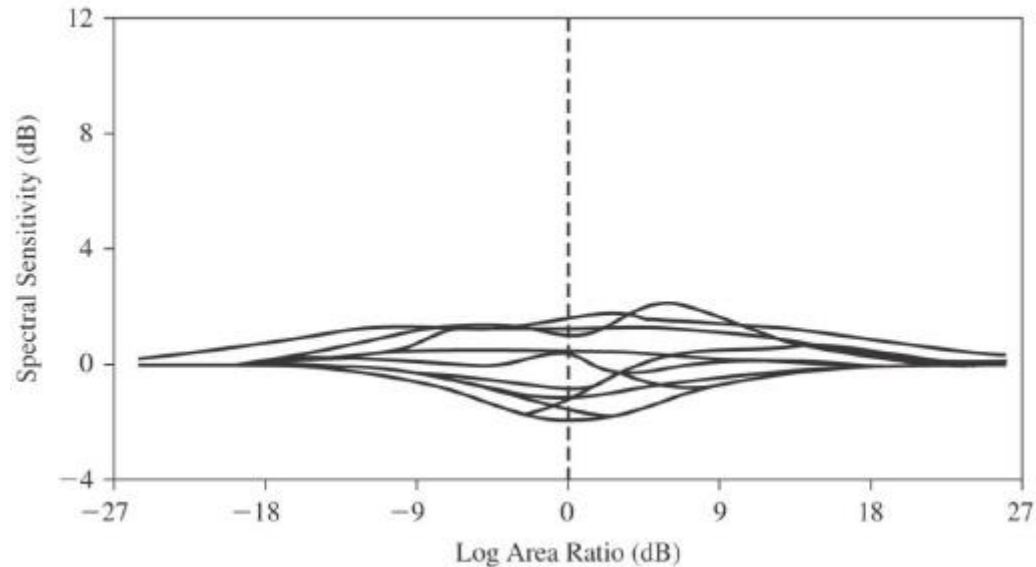
$$\frac{\partial S}{\partial g_i} = \lim_{\Delta g_i \rightarrow 0} \left| \frac{1}{\Delta g_i} \left[\frac{1}{2\pi} \int_{-\pi}^{\pi} \left| \log \frac{P(\omega, g_i)}{P(\omega, g_i + \Delta g_i)} \right| d\omega \right] \right|$$

which measures sensitivity to errors in the g_i parameters

Line Spectral Pairs



spectral sensitivity for k_i parameters;
low sensitivity around 0; high
sensitivity around 1



spectral sensitivity for log area ratio
parameters, g_i – low sensitivity for
virtually entire range is seen

Line Spectral Pairs

$$A(z) = 1 + \alpha_1 z^{-1} + \alpha_2 z^{-2} + \dots + \alpha_p z^{-p}$$

= all-zero prediction filter with all zeros, z_k , inside the unit circle

$$\tilde{A}(z) = z^{-(p+1)} A(z^{-1}) = \alpha_p z^{-1} + \dots + \alpha_2 z^{-p+1} + \alpha_1 z^{-p} + z^{-(p+1)}$$

= reciprocal polynomial with inverse zeros, $1/z_k$

- Consider the following

$$L(z) = \frac{\tilde{A}(z)}{A(z)} = \text{allpass system} \Rightarrow |L(e^{j\omega})| = 1, \text{ all } \omega$$

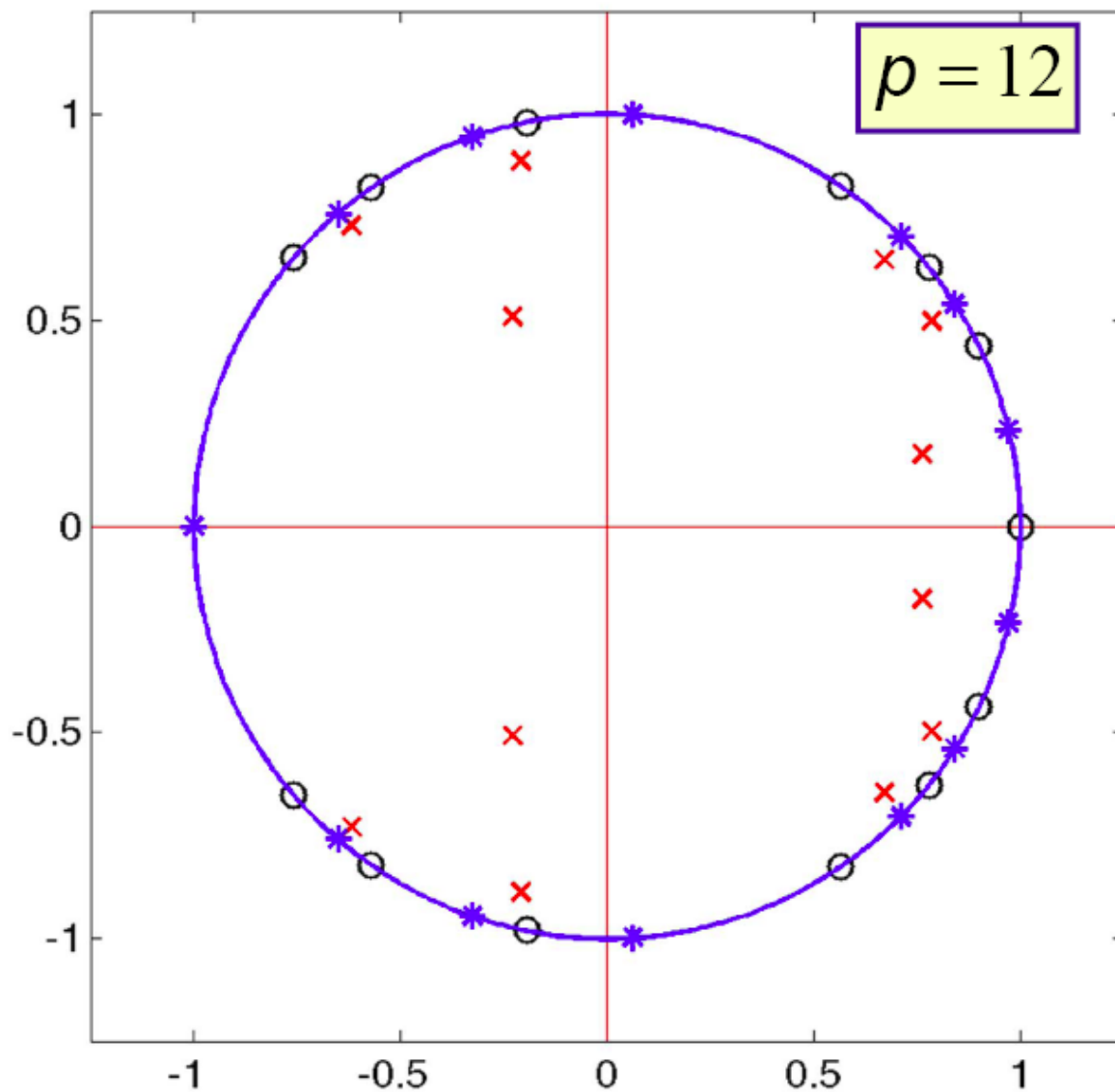
- Form the symmetric polynomial $P(z)$ as

$$P(z) = A(z) + \tilde{A}(z) = A(z) + z^{-(p+1)} A(z^{-1}) \Rightarrow P(z) \text{ has zeros for } L(z) = -1; (A(z) = -\tilde{A}(z)) \\ \Rightarrow \arg\{L(e^{j\omega_k})\} = (k + 1/2) \cdot 2\pi, k = 0, 1, \dots, p-1$$

- Form the anti-symmetric polynomial $Q(z)$ as

$$Q(z) = A(z) - \tilde{A}(z) = A(z) - z^{-(p+1)} A(z^{-1}) \Rightarrow Q(z) \text{ has zeros for } L(z) = +1; (A(z) = \tilde{A}(z)) \\ \Rightarrow \arg\{L(e^{j\omega_k})\} = k \cdot 2\pi, k = 0, 1, \dots, p-1$$

LSP Example



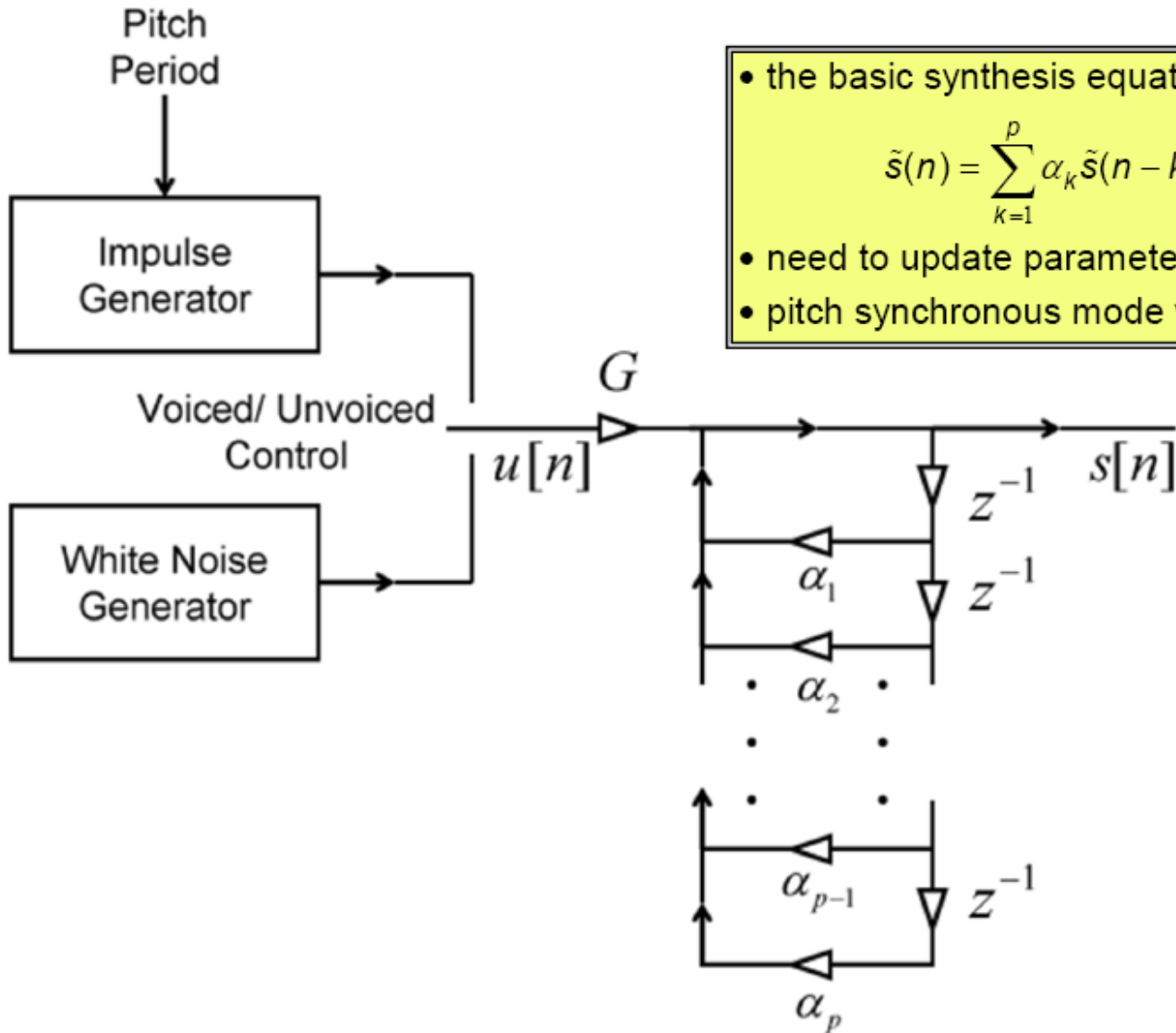
- * $P(z)$ roots
- o $Q(z)$ roots
- x $A(z)$ roots

Line Spectral Pairs

- properties of LSP parameters
 1. all the roots of $P(z)$ and $Q(z)$ are on the unit circle
 2. a necessary and sufficient condition for $|k_i| < 1, i = 1, 2, \dots, p$ is that the roots of $P(z)$ and $Q(z)$ alternate on the unit circle
 3. the LSP frequencies get close together when roots of $A(z)$ are close to the unit circle

Applications

Speech Synthesis



- the basic synthesis equation is

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

- need to update parameters every 10 msec or so
- pitch synchronous mode works best

Speech Coding

1. Extract α_k parameters properly
2. Quantize α_k parameters properly so that there is little quantization error
 - Small number of bits go into coding the α_k coefficients
3. Represent $e(n)$ via:
 - Pitch pulses and noise—LPC Coding
 - Multiple pulses per 10 msec interval—MPLPC Coding
 - Codebook vectors—CELP
 - Almost all of the coding bits go into coding of $e(n)$

LPC Vocoder

