# Lecture 14: LPC speech synthesis and autocorrelation-based pitch tracking

ECE 417, Multimedia Signal Processing
October 10, 2019

#### Outline

- The LPC-10 speech synthesis model
- Autocorrelation-based pitch tracking
- Inter-frame interpolation of pitch and energy contours
- The LPC-10 excitation model: white noise, pulse train
- Linear predictive coding: how to find the coefficients
- Linear predictive coding: how to make sure the coefficients are stable

# The LPC-10 speech synthesis model

Federal Standard 1015 has been redesignated as Federal Information Processing Standards Publication (FIPS PUB) 137. Issued by the National Institute of Standards and Technology pursuant FEDERAL STANDARD 1015 to Section 111(d) of the Federal Property and Administrative Services Act of 1949 as amended by the Computer Security Act of 1987, Public Law 100-235 ANALOG TO DIGITAL CONVERSION OF VOICE BY 2,400 BIT/SECOND LINEAR PREDICTIVE CODING Prepared By: **National Communications System** Office Of Technology & Standards **Published By:** General Services Administration

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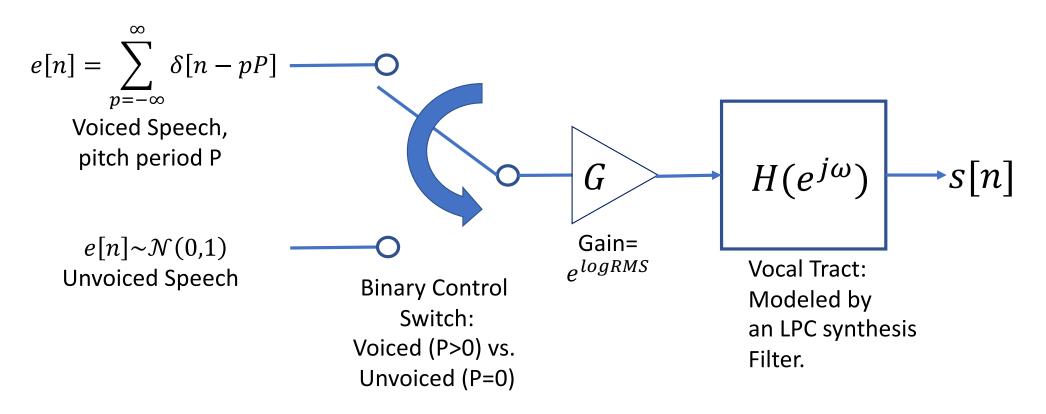
November 28, 1984

# The LPC-10 Speech Coder: Transmitted Parameters

Each frame is 54 bits, and is used to synthesize 22.5ms of speech. (54 bits/frame)/(0.0225 seconds/frame)=2400 bits/second

- Pitch: 7 bits/frame (127 distinguishable non-zero pitch periods)
- **Energy**: 5 bits/frame (32 levels, on a logRMS scale)
- 10 linear predictive coefficients (LPC): 41 bits/frame
- Synchronization: 1 bit/frame

# The LPC-10 speech synthesis model

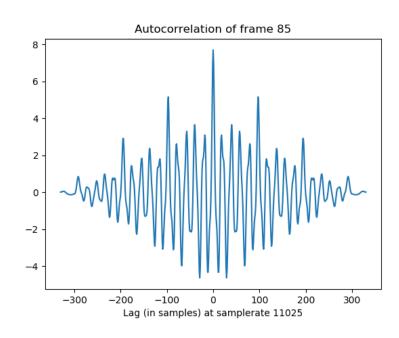


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#### Autocorrelation is maximum at n=0

$$r_{xx}[n] = \sum_{m=-\infty}^{\infty} x[m]x[m-n]$$



#### Autocorrelation is maximum at n=0

$$r_{xx}[n] = \sum_{m=-\infty}^{\infty} x[m]x[m-n] = x[n] * x[-n] = \mathcal{F}^{-1}\{|X(\omega)|^2\}$$
$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2 e^{j\omega n} d\omega$$

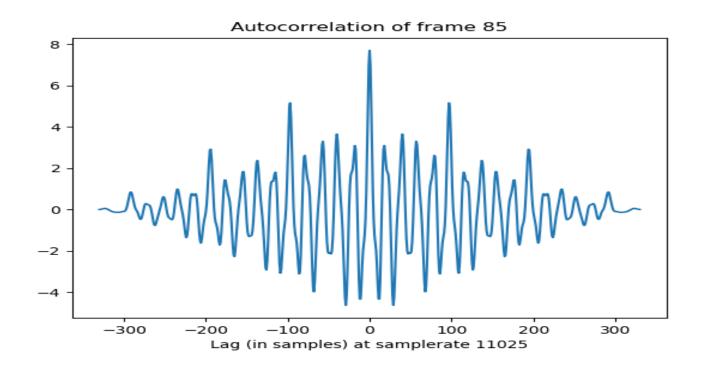
Notice that, for n=0, this becomes just Parseval's theorem:

$$r_{xx}[0] = \sum_{m=-\infty}^{\infty} x^2[m] = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2 d\omega$$

But since  $|X(\omega)|^2$  is positive and real, any value of  $e^{j\omega n}$  that is NOT positive and real will reduce the value of the integral!

$$r_{xx}[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2 e^{j\omega n} d\omega \le \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^2 d\omega = r_{xx}[0]$$

Example of an autocorrelation function computed from file0.wav, "Four score and seven years ago..."

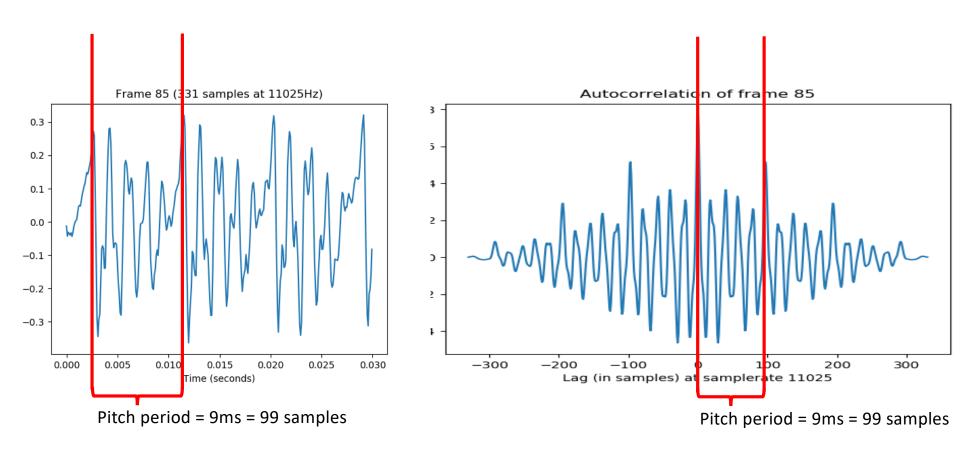


# Autocorrelation of a periodic signal

Suppose x[n] is periodic, x[n] = x[n-P]. Then the autocorrelation is also periodic:

$$r_{xx}[P] = \sum_{m=-\infty}^{\infty} x[m]x[m-P] = \sum_{m=-\infty}^{\infty} x^2[m] = r_{xx}[0]$$

### Autocorrelation of a periodic signal is periodic

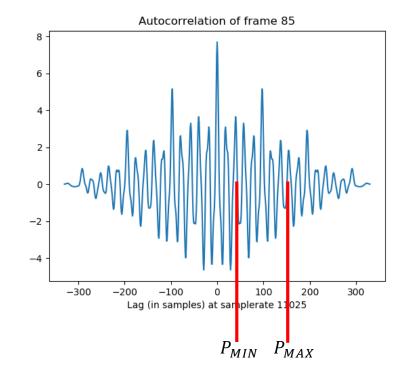


# Autocorrelation pitch tracking

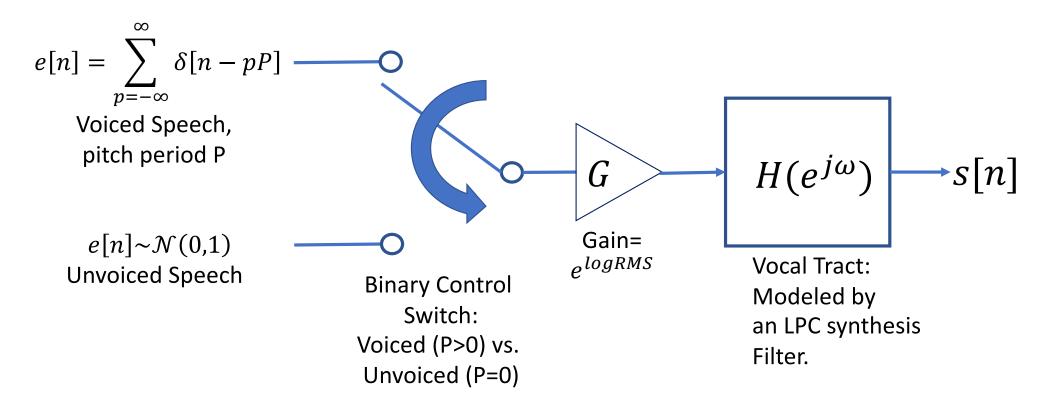
- Compute the autocorrelation
- Find the pitch period:

$$P = \underset{P_{MIN} \le m \le P_{MAX}}{\operatorname{argmax}} r_{xx}[m]$$

- The search limits,  $P_{MIN}$  and  $P_{MAX}$ , are important for good performance:
  - $P_{MIN}$  corresponds to a high woman's pitch, about  $F_S/P_{MIN} \approx 250~{\rm Hz}$
  - $P_{MAX}$  corresponds to a low man's pitch, about  $F_S/P_{MAX} \approx 80$  Hz



### The LPC-10 speech synthesis model



# The voiced/unvoiced decision

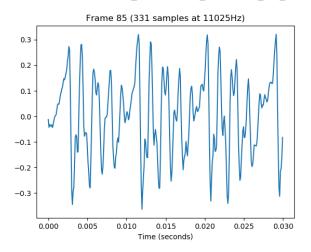
- x[n] voiced:  $r_{xx}[P] \approx r_{xx}[0]$
- x[n] unvoiced (white noise):  $r_{xx}[n] \approx \delta[n]$ , which means that  $r_{xx}[P] \ll r_{xx}[0]$

So a reasonable V/UV decision is:

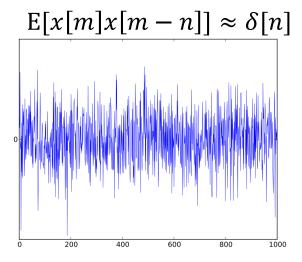
- $\frac{r_{\chi\chi}[P]}{r_{\chi\chi}[0]} \ge threshold$ : say the frame is voiced.
- $\frac{r_{\chi\chi}[P]}{r_{\chi\chi}[0]} < threshold$ : say the frame is unvoiced.

Setting threshold~0.25 works reasonably well.

#### voiced: $x[n + P] \approx x[n]$



#### unvoiced:

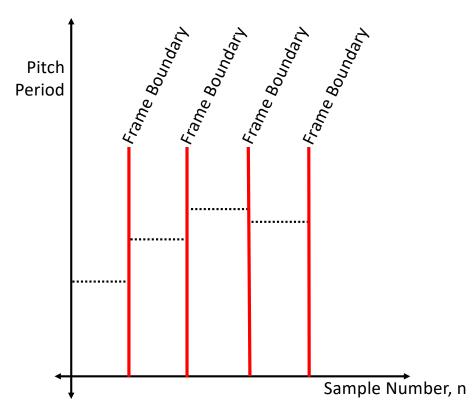


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### Inter-frame interpolation of pitch contours

We don't want the pitch period to change suddenly at frame boundaries; it sounds weird.



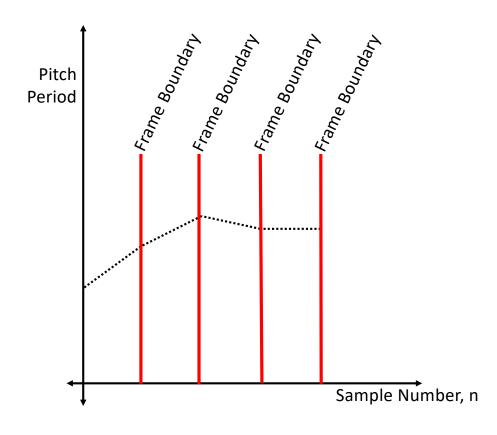
# Inter-frame interpolation of pitch contours

Linear interpolation sounds much better. We can accomplish linear interpolation using a formula like

$$P[n] = (1 - f)P_t + fP_{t+1}$$

#### Where

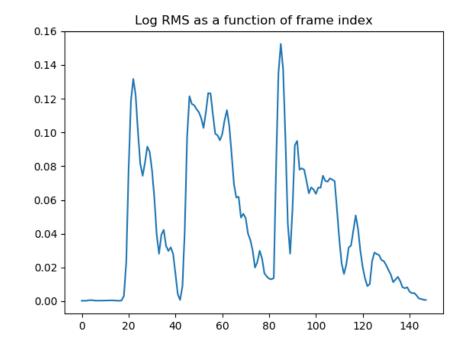
- $P_t$  is the pitch period in frame t
- $f = \frac{n-tS}{S}$  is how far sample n is from the beginning of frame t
- S is the frame-skip.



#### Inter-frame interpolation of energy

Linear interpolation is also useful for energy, EXCEPT: it sounds better if we interpolate log energy, not energy.

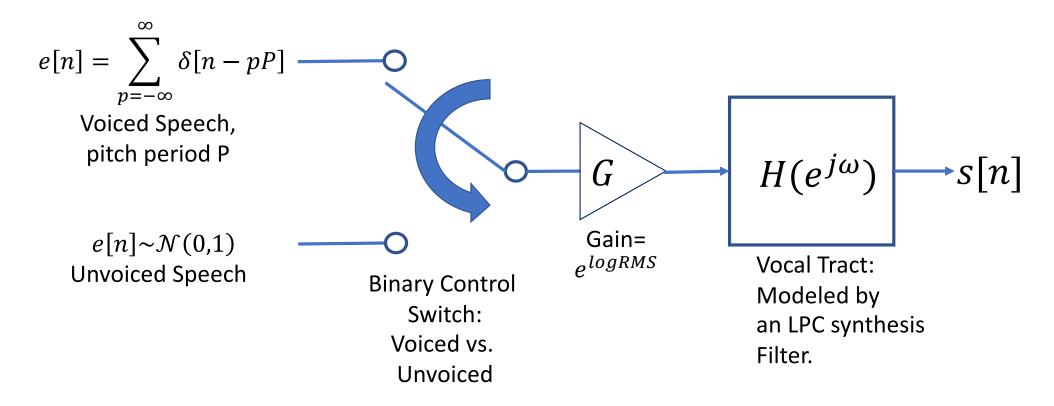
$$\log RMS_t = \log \sqrt{\frac{1}{L} \sum_{n=tS}^{tS+L-1} x^2[n]}$$



#### Outline

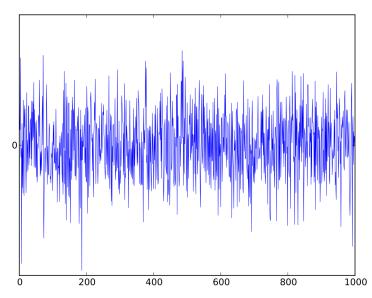
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# The LPC-10 speech synthesis model



### Unvoiced speech: e[n]=white noise

- Use zero-mean, unit-variance Gaussian white noise
- The choice, to use "unvoiced speech," is communicated by the special code word "P=0"



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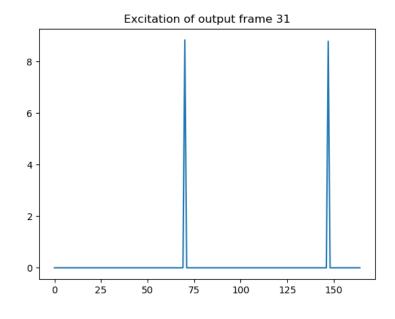
# Voiced speech: e[n]=pulse train

• The basic idea: $_{\infty}$ 

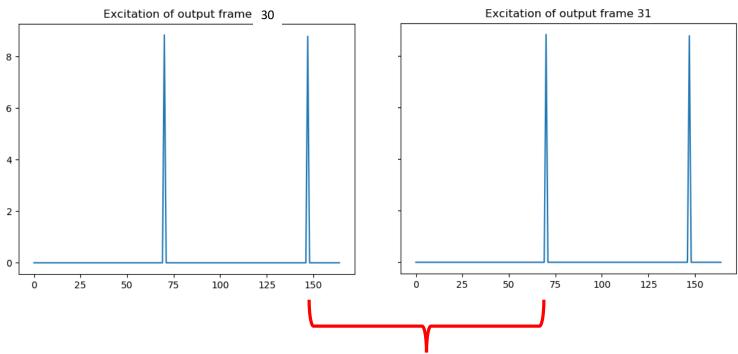
$$e[n] = \sum_{p = -\infty} \delta[n - pP]$$

• Modification #1: in order for the RMS to equal 1.0, we need to scale each pulse by  $\sqrt{P}$ :

$$e[n] = \sqrt{P} \sum_{p=-\infty}^{\infty} \delta[n - pP]$$



### Modification #2: the first pulse is not at n=0



Pitch period = 80 samples  $\Rightarrow$  first pulse in frame 31 can't occur until the 70<sup>th</sup> sample of the frame

# A mechanism for keeping track of pitch phase from one frame to the next

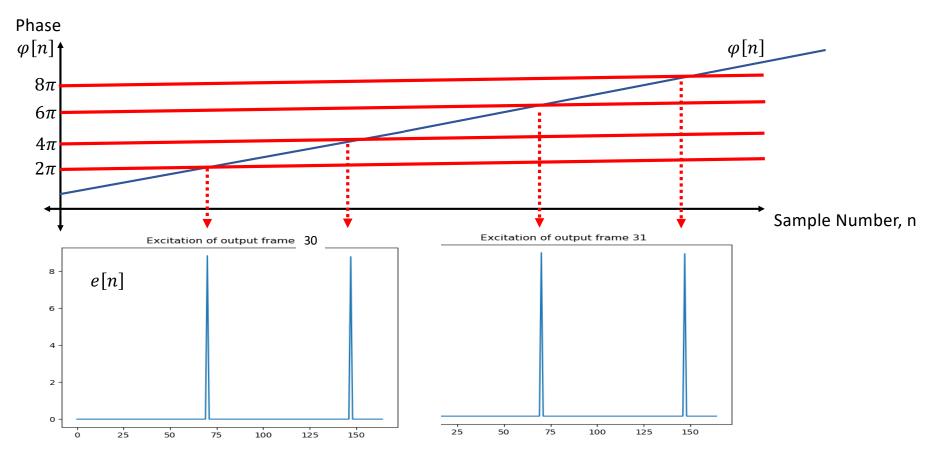
- Start out, at the beginning of the speech, with a pitch phase equal to zero,  $\varphi[0]=0$
- For every sample thereafter:
  - If the sample is unvoiced (P[n]=0), don't increment the pitch phase
  - If the sample is voiced (P[n]>0), then increment the pitch phase

$$\varphi[n] = \varphi[n-1] + \frac{2\pi}{P[n]}$$

• Every time the phase passes a multiple of  $2\pi$ , output a pitch pulse

$$e[n] = \begin{cases} \sqrt{P} & \left[ \frac{\varphi[n]}{2\pi} \right] - \left[ \frac{\varphi[n-1]}{2\pi} \right] > 0 \\ 0 & else \end{cases}$$

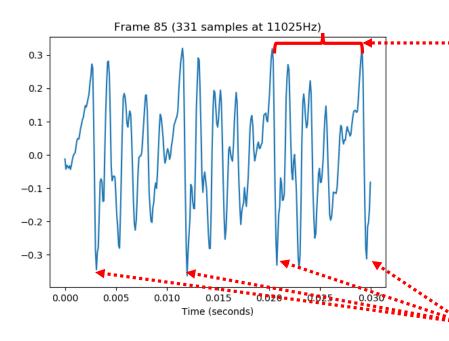
# The pitch phase method: generate an excitation pulse whenever pitch phase crosses a $2\pi$ -level



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# Speech is predictable

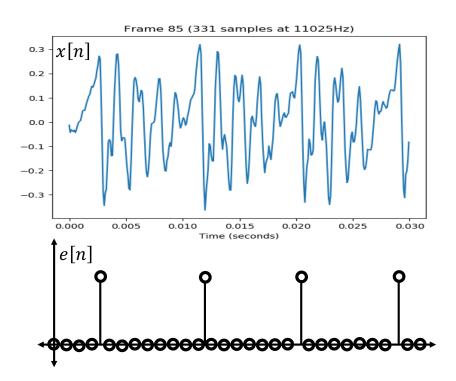


 Speech is not just white noise and pulse train. In fact, each sample is highly predictable from previous samples.

$$x[n] \approx \sum_{m=1}^{10} \alpha_m x[n-m]$$

 In fact, the pitch pulses are the only major exception to this
 predictability!

# Linear predictive coding (LPC)



#### The LPC idea:

1. Model the excitation as error

$$e[n] = x[n] - \sum_{m=1}^{10} \alpha_m x[n-m]$$

2. Force the coefficients  $\alpha_m$  to explain as much as they can, so that e[n] is as close to zero as possible.

### Linear predictive coding (LPC)

$$\varepsilon = E[e^{2}[n]] = E\left[\left(x[n] - \sum_{i=1}^{10} \alpha_{i}x[n-i]\right)^{2}\right]$$

$$\frac{\partial \varepsilon}{\partial \alpha_{j}} = -2E\left[x[n-j]\left(x[n] - \sum_{i=1}^{10} \alpha_{i}x[n-i]\right)\right]$$

Setting  $\frac{\partial \varepsilon}{\partial \alpha_i} = 0$  gives

$$E[x[n-j]x[n]] = \sum_{i=1}^{10} \alpha_i E[x[n-j]x[n-i]]$$

$$R_{xx}[|i-j|]$$

### Linear predictive coding (LPC)

So we have a set of linked equations, for  $1 \le j \le 10$ :

$$R_{xx}[j] = \sum_{i=1}^{10} \alpha_i R_{xx}[|i-j|]$$

- We can write these 10 equations as a 10x10 matrix equation:  $\vec{\gamma} = R\vec{\alpha}$
- ...which immediately gives the solution:  $\vec{\alpha} = R^{-1}\vec{\gamma}$
- ...where

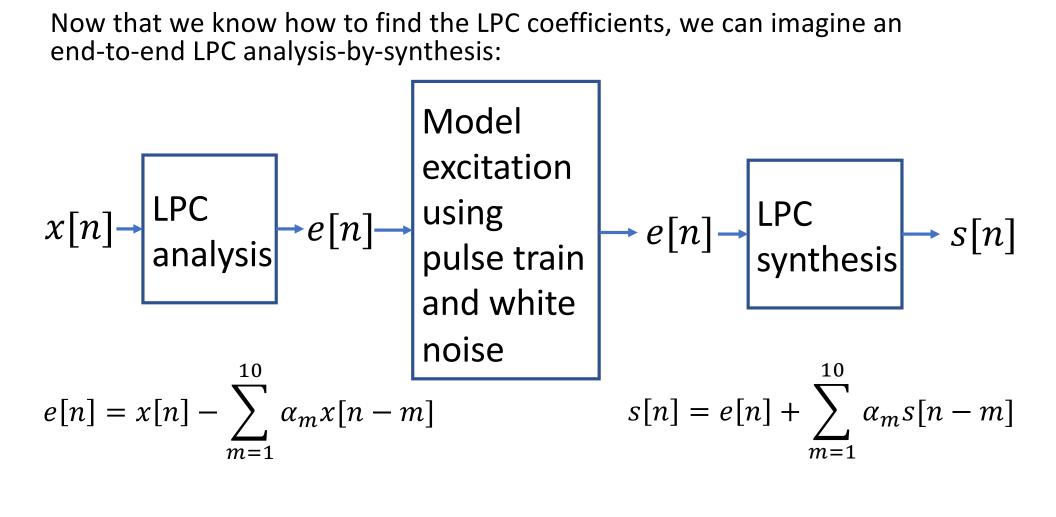
$$\vec{\gamma} = \begin{bmatrix} R_{xx}[1] \\ \vdots \\ R_{xx}[10] \end{bmatrix}, \qquad R = \begin{bmatrix} R_{xx}[0] & R_{xx}[1] & \cdots \\ R_{xx}[1] & R_{xx}[0] & \cdots \\ \vdots & \vdots & R_{xx}[0] \end{bmatrix}, \qquad \vec{\alpha} = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_{10} \end{bmatrix}$$

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#### Speech -> Excitation -> Speech

Now that we know how to find the LPC coefficients, we can imagine an



### The LPC Analysis Filter

The LPC Analysis Filter is an all-zeros filter (FIR = finite impulse response):

$$e[n] = x[n] - \sum_{m=1}^{10} \alpha_m x[n-m] \leftrightarrow E(z) = A(z)X(z)$$

...where...

$$A(z) = 1 - \sum_{m=1}^{10} \alpha_m z^{-m}$$

It's often useful to write this as a polynomial  $A(z)=a_0+a_1z^{-1}+\cdots$  where

$$a_j = \begin{cases} 1 & j = 0 \\ -\alpha_j & 1 \le j \le 10 \end{cases}$$

#### The LPC Synthesis Filter

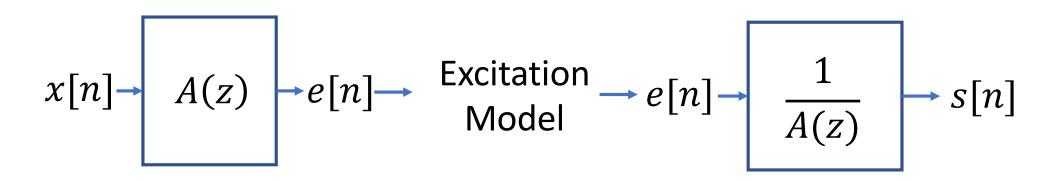
The LPC Synthesis Filter is an all-poles filter (IIR = infinite impulse response):

$$s[n] = e[n] + \sum_{m=1}^{10} \alpha_m s[n-m] \leftrightarrow S(z) = H(z)E(z)$$

...where...

$$H(z) = \frac{1}{A(z)} = \frac{1}{1 - \sum_{m=1}^{10} \alpha_m z^{-m}}$$

Speech -> Excitation -> Speech



### The Stability Problem

- The analysis filter is guaranteed to be stable, as long as the coefficients are finite. Suppose you know that  $|x[n]| \le X_{MAX}$ , and  $|\alpha_m| \le \alpha_{MAX}$ . Then, even in the worst possible case,  $|e[n]| \le 11\alpha_{MAX}X_{MAX}$ .
- The synthesis filter has no such guarantee. For example, suppose e[n] is just a delta function ( $e[n] = \delta[n]$ ), and suppose all of the  $\alpha_m = 0$  except the first one,  $\alpha_1 = 1.1$ . Then

$$s[n] = \delta[n] + 1.1s[n-1] = (1.1)^n$$

Which overflows your 16-bit sample buffer after only 110 samples. Your output will be full of NaNs, and you'll be saying "What happened...?"

#### How to Guarantee Stability

Fortunately, the LPC synthesis filter is causal, so it's easy to guarantee stability. We just need to make sure that all of the poles have magnitude less than 1:

$$|r_k| < 1$$

We find the poles like this:

$$H(z) = \frac{1}{A(z)} = \frac{1}{\sum_{m=0}^{10} a_m z^{-m}} = \frac{1}{\prod_{k=1}^{10} (1 - r_k z^{-1})}$$

in other words,

$$r_k = roots(A(z))$$

...which you can do using np.roots, if you define the polynomial correctly. Then you just truncate the magnitude,

$$r_k \leftarrow \min(|r_k|, 0.999)e^{j \not\preceq r_k}$$

...and then use np.poly to convert back from roots to polynomial.