

MVE 136 L12

Ex) AR(1), $\sigma_e^2 = 1$

$$x(n) + 0,5 x(n-1) = e(n)$$

$$z: \quad \underline{X}(z) + 0,5 \cdot \bar{z}^{-1} \underline{X}(z) = E(z)$$

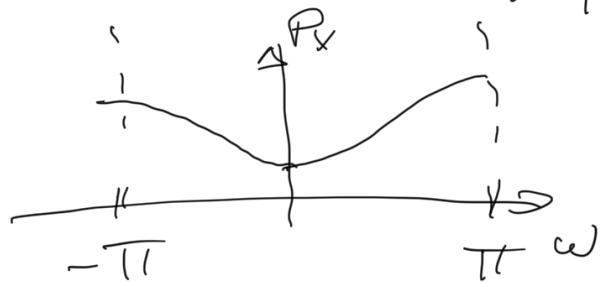
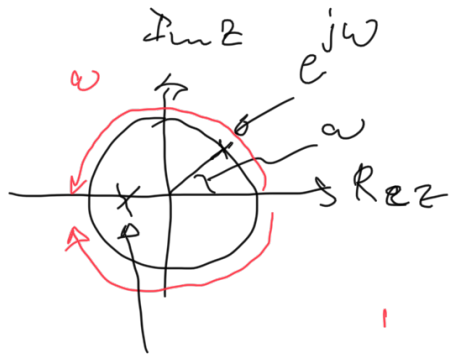
$$(1 + 0,5 \bar{z}^{-1}) \underline{X}(z) = E(z)$$

$$H(z) = \frac{\underline{X}(z)}{E(z)} = \frac{1}{1 + 0,5 \bar{z}^{-1}} = \frac{z}{z + 0,5}$$

$\Rightarrow$ pole at

$$z = -0,5$$

$$P_x(e^{j\omega}) = |H(z)|_{z=e^{j\omega}}^2 = \frac{1}{|e^{j\omega} + 0,5|^2}$$



Ex) MA(1), $\sigma_e^2 = 1$

$$x(n) = e(n) - 0,5 e(n-1)$$

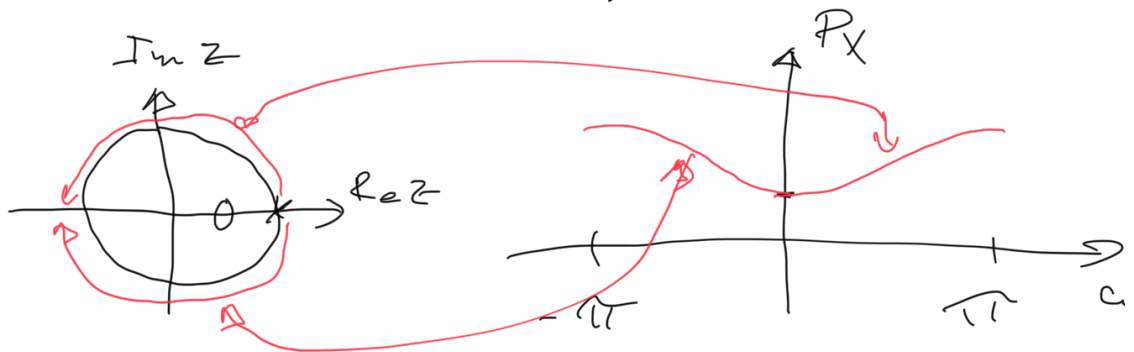
$$z: \quad \underline{X}(z) = E(z) - 0,5 E(z) \cdot \bar{z}^{-1}$$

$$\underline{X}(z) = (1 - 0,5 \bar{z}^{-1}) E(z)$$

$$\Rightarrow H(z) = \frac{Z(z)}{E(z)} = 1 - 0,5 z^{-1}$$

$$\Rightarrow \text{zero at } z = 0,5$$

$$P_x(e^{j\omega}) = |H(e^{j\omega})|^2 = |1 - 0,5 e^{-j\omega}|^2$$



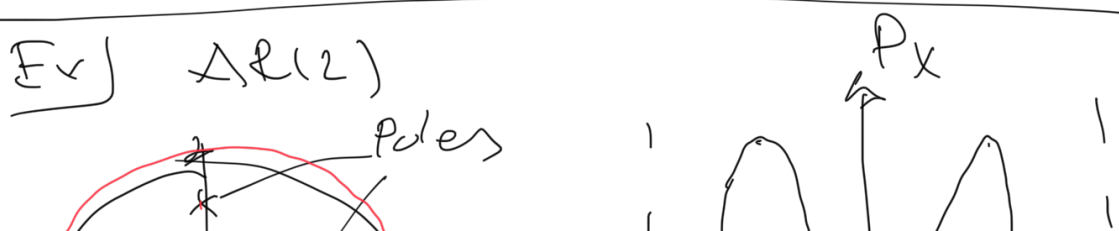
PSD for ARMA(p,q)

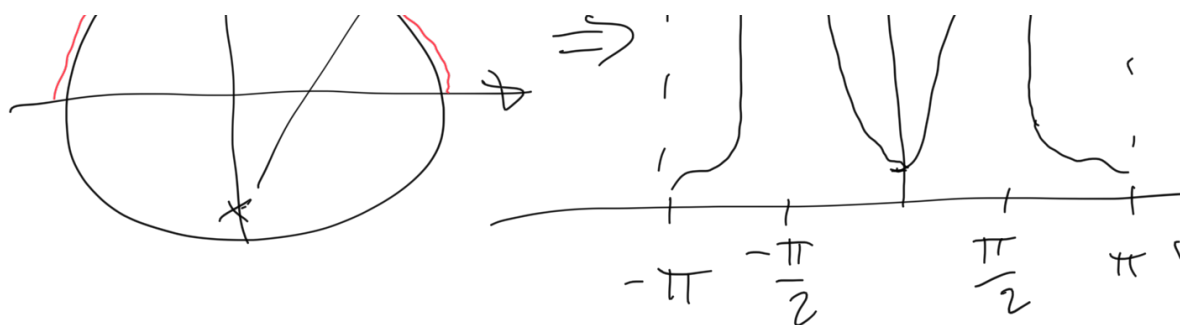
$$P_x(e^{j\omega}) = \sigma_e^2 |H(e^{j\omega})|^2 = \sigma_e^2 \left| \frac{B(e^{j\omega})}{A(e^{j\omega})} \right|^2$$

$$= \sigma_e^2 \frac{\prod_{i=1}^q |1 - z_i e^{-j\omega}|^2}{\prod_{i=1}^p |1 - p_i e^{-j\omega}|^2} \Big|_{z=e^{j\omega}}$$

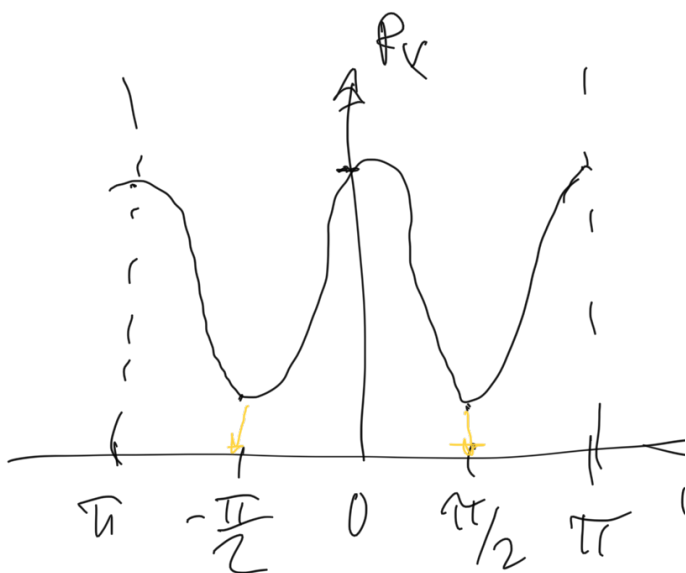
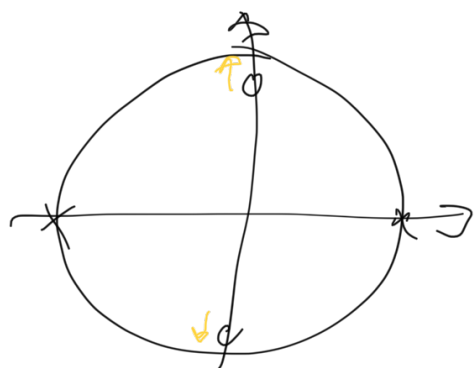
$\sigma_e^2 \frac{2\pi}{\pi}$ distances² between $e^{j\omega}$ & zero

π distance² between $e^{j\omega}$ and pole





Ex) $MA(2)$



Conclusions:

AR: good for peaks in $P_x(e^{j\omega})$

MA: good for notches in $P_x(e^{j\omega})$

ARMA combines both!

Back to AR(1) process

$$X(n) + 0.5X(n-1) = e(n)$$

Order 1 $n, n-1, n-2, \dots$

$$\left\{ \begin{array}{l} \vdots \\ r_x(k) + 0,5 r_y(k-1) = 0 \end{array} \right.$$

$$r_x(k) = -0,5 r_x(k-1)$$

$$|r_x(k)| > 0 \quad \forall k$$

$\Rightarrow r_x(k)$ has an infinite length if AR-process.

$$\begin{bmatrix} 1 & 0,5 \\ 0,5 & 1 \end{bmatrix} \begin{bmatrix} r_x(0) \\ r_x(1) \end{bmatrix} = \begin{bmatrix} \sigma_e^2 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} r_x(0) \\ r_x(1) \end{bmatrix} = \frac{1}{1-0,25} \begin{bmatrix} 1 & -0,5 \\ -0,5 & 1 \end{bmatrix} \begin{bmatrix} \sigma_e^2 \\ 0 \end{bmatrix}$$

Ex) $P=2 \quad \sigma_e^2 = 1$

$$r_x(0) + a_1 r_x(1) + a_2 r_x(2) = 1$$

$$\begin{cases} r_x(1) + a_1 r_x(0) + a_2 r_x(1) = 0 \\ r_x(2) + a_1 r_x(1) + a_2 r_x(0) = 0 \end{cases}$$

$\Leftrightarrow$

$$\begin{bmatrix} 1 & a_1 & a_2 \\ a_1 & 1+a_2 & 0 \\ a_2 & a_1 & 1 \end{bmatrix} \begin{bmatrix} r_x(0) \\ r_x(1) \\ r_x(2) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

MA(1), $\sigma_e^2 = 1$

$$x(n) = e(n) - 0.5e(n-1)$$

multiply $x(n-k)$ on both side

$$E\{x(n)x(n-k)\} = E\{x(n-k)(e(n) - 0.5e(n-1))\}$$

$r_x(k) \qquad (e(n-k) - 0.5e(n-k-1))$

if $k > 1 \Rightarrow r_x(k) = 0$

$$r_x(0) = E\{e^2(n) + 0.5^2 e^2(n-1)\} = 1 + 0.25 = 1.25$$

$$r_x(1) = 0.5 E\{e^2(n-1)\} = 0.5$$

for $k \geq 2 : r_x(k) = 0$

For general MA(q) models.

$$x(n) = e(n) + b_1 e(n-1) + \dots + b_q e(n-q)$$

$$\text{Def: } b(k) = \begin{cases} 1 & k=0 \\ b_k & k=1 \dots q \\ 0 & k > q, k < 0 \end{cases}$$

$$r_x[k] = \sum_{n=0}^q b(n) b(k-n) \quad -q \leq k \leq q$$

$\Rightarrow$ MA process has a finite length ACF

Estimate an AR(p) model from data. Given $\{x(n)\}_{n=0}^{N-1}$
estimate $a_1 \dots a_p$

1) Find $\hat{r}_x(k) = \frac{1}{N} \sum_{n=k}^{N-1} x(n)x(n-k)$

2) Use Y-W eqn. $k=1 \dots p$
 $r_x(k) + a_1 r_x(k-1) + \dots + a_p r_x(k-p) = 0$

Y-W $\Rightarrow$

$$\begin{bmatrix} r_x(0) & r_x(1) & \dots & r_x(p-1) \\ r_x(1) & r_x(0) & \dots & r_x(p-2) \\ \vdots & \vdots & \ddots & \vdots \\ r_x(p-1) & \dots & \dots & r_x(0) \end{bmatrix} \begin{bmatrix} a_1 \\ \vdots \\ a_p \end{bmatrix} = \begin{bmatrix} -r_x(1) \\ -r_x(2) \\ \vdots \\ -r_x(p) \end{bmatrix}$$

$$\underbrace{\quad}_{\underline{R}} \quad \underbrace{\quad}_{\underline{a}} \quad \underbrace{\quad}_{\underline{r}}$$

$$\hat{\underline{R}} \hat{\underline{a}} = \hat{\underline{r}}$$

$$\hat{\underline{a}} = \hat{\underline{R}}^{-1} \hat{\underline{r}}$$

$$\hat{\underline{a}} = \hat{\underline{R}} \setminus \hat{\underline{r}} \quad (\text{in Matlab})$$

~~$$\text{inv}(\hat{\underline{R}}) * \hat{\underline{r}}$$~~