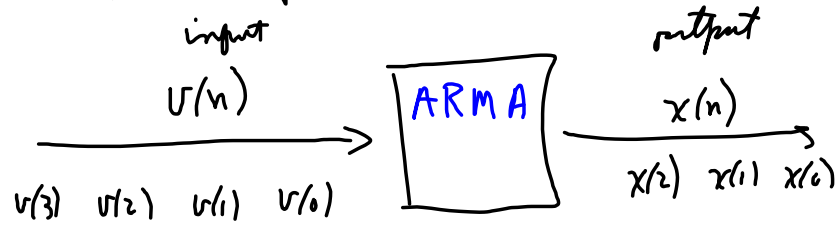


Discrete processes.



Difference Equation

$$x(n) = \underbrace{b(0)v(n) + b(1)v(n-1) + b(2)v(n-2) \dots}_{\text{Moving Average}} - \underbrace{a(1)x(n-1) - a(2)x(n-2) \dots}_{\text{Auto regression}}$$

$$\sum_{l=0}^P a(l) x(n-l) = \sum_{l=0}^q b(l) v(n-l)$$

$a(0) = 1$

In random process modeling, $v(n)$ are i.i.d. random variables: $\{v\}$ is white noise.

zero-mean

Let (1) $x(n) = \left(\frac{1}{2}\right)x(n-1) + v(n) \leftarrow \text{AR}$

Question 1. Given a measured random process ~~$x(t)$~~ $x(n)$, find an ARMA model which could have generated it.

First (use Bartlett's) extract the autocorrelation. $R_x(\tau)$.

Then find $\{a's, b's\}$ which give that autocorrelation.

Second question, give ARMA a 's & b 's,
what is $R_x(E)$ out of the output?

The "Yule-Walker" relates a 's, b 's, R_x 's.

Use them to answer question 1 or question 2,

Yule-Walker Equations.

$$a(0)x(n) + a(1)x(n-1) + a(2)x(n-2) + a(3)x(n-3)$$

$p=3$

$$= b v(n) + b(1)v(n-1) + b(2)v(n-2)$$

$q=2$

(3,2) ARMA process.

$$\sum_{l=0}^p a(l)x(n-l) = \sum_{l=0}^q b(l)v(n-l)$$

To get Y-W eqs, multiply by $x(n-k)$ & take $\{ \}$.

$$\sum_{l=0}^p a(l) \underbrace{\{ \{ x(n-l)x(n-k) \} \}}_{R_x(k-l)} = \sum_{l=0}^q b(l) \underbrace{\{ \{ v(n-l)x(n-k) \} \}}_{\text{"cross correlations"}}$$

$R_x(k-l)$

"cross correlations"

$R_{vx}(k-l)$

$$E \{ v(n-l) x(n-k) \}$$

Suppose $n-k < n-l$ (late noise, early output)

$$E \{ v(\text{late}) x(\text{early}) \} = 0$$

because $v(\text{late})$ is independent of $x(\text{early})$.

But $x(\text{late})$ depends on $v(\text{early})$, so not independent.

To get $\mathcal{E}\{x(\text{late})v(\text{early})\}$, we need
 to express $x(n)$ in terms of $v(n), v(n-1),$
 $v(n-2), \dots$

We need

$$x(n) = h(0)v(n) + h(1)v(n-1) + h(2)v(n-2) + h(3)v(n-3)$$

$$\Rightarrow \mathcal{E}\{x(n)v(n-r)\} = \mathcal{E}\{[h(0)v(n) + h(1)v(n-1) + \dots + h(r)v(n-r)]v(n-r)\}$$

$$= \begin{cases} h(r)\mathcal{E}\{v(n-r)^2\} & \text{if } r \geq 0 \\ 0 & \text{if } r < 0 \end{cases}$$

from \Rightarrow $n-k$ $n-l$
 mapping
 $r = l - k$
 $l - k \geq 0 \Leftrightarrow l \geq k$

$$\mathcal{E}\{v(i)v(j)\} = \begin{cases} 0, & \text{by independence + zero-mean,} \\ & \text{if } i \neq j \\ \sigma_v^2 & \text{if } i = j \end{cases}$$

$$\sum_{l=0}^p a(l) x(n-l) = \sum_{l=0}^q b(l) v(n-l)$$

• $x(n-k)$ & take $\{ \text{---} \}$.

$$\sum_{l=0}^p a(l) R_x(k-l) = \sum_{l=0}^q b(l) \left\{ x(n-k) v(n-l) \right\}$$

From previous page,
 $h(l-k) \sigma_v^2$ if $l \geq k$,

0 otherwise

Y-W

$$\sum_{l=0}^p a(l) R_x(k-l) = \sum_{l=k}^q b(l) h(l-k) \sigma_v^2$$

$$Y-W \sum_{l=0}^p a(l) R_x(k-l) = \sum_{l=k}^q b(l) h(l-k) \sigma_v^2$$

$$\begin{bmatrix} R_x(0) & R_x(-1) & R_x(-2) & R_x(-3) \\ R_x(1) & R_x(0) & R_x(-1) & R_x(-2) \end{bmatrix}$$

$$\begin{bmatrix} a(0) = 1 \\ a(1) \\ a(2) \\ a(3) \end{bmatrix}$$

$$= \begin{bmatrix} b(0) h(0) \sigma_v^2 & (k=q) \\ + b(1) h(1) \sigma_v^2 \\ + b(2) h(2) \sigma_v^2 \\ + b(3) \text{ No!} & \downarrow q=2 \\ \dots & (k=1) \\ b(1) h(0) \sigma_v^2 + \\ b(2) h(1) \sigma_v^2 \\ + b(3) \text{ No!} \end{bmatrix}$$

See notes for elaboration.

How do you find the h 's ?

$$x(n) = b(n) v(n) - a(1)x(n-1) - a(2)x(n-2) \quad p=2$$

$$x(0) = b(0)v(0) - a(1) \cdot 0 - a(2) \cdot 0$$

$$x(1) = b(1)v(1) - a(1)x(0) - a(2)x(-1)$$

$$= b(1)v(1) - a(1)b(0)v(0) - 0$$

$$x(2) = b(2)v(2) - a(1)[b(1)v(1) - a(1)b(0)v(0)] - a(2)b(0)v(0)$$

and so on.

Compare with

$$x(n) = h(0)v(n) + h(1)v(n-1) + h(2)v(n-2) + \dots$$

$h(0)$ $-a(1)h(0)$ $a(1)^2 b(0) - a(2)b(0)$
" " +