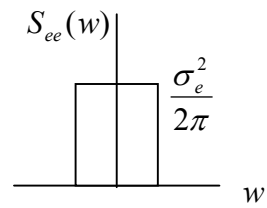
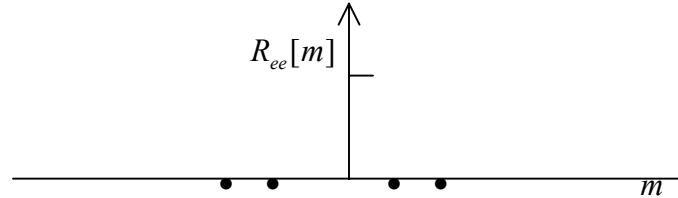


Random Signal Modelling

In statistics, random-process modeling using difference equations is known as time series analysis.

White Noise Sequence: The simplest model is the white noise $e[n]$. We shall assume that $e[n]$ is of 0-mean and variance σ_e^2 .



Moving Average model $MA(q)$ model.



The difference equation model is

$$X[n] = \sum_{i=0}^q b_i e[n-i]$$

$$\mu_e = 0 \Rightarrow \mu_Y = 0$$

and $e[n]$ is an uncorrelated sequence means

$$\sigma_Y^2 = \sum_{i=0}^q b_i^2$$

The autocorrelations are given by

$$\begin{aligned}
 R_{XX}[m] &= E X[n] X[n-m] \\
 &= \sum_{i=0}^q \sum_{j=0}^q b_i b_j E e[n-i] e[n-m-j] \\
 &= \sum_{i=0}^q \sum_{j=0}^q b_i b_j R_{ee}[m-i+j]
 \end{aligned}$$

Noting that $R_{ee}[m] = \sigma_e^2 \delta[m]$, we get

$$R_{ee}[m] = \sigma_e^2 \text{ when}$$

$$m-i+j=0$$

$$\Rightarrow i = m+j$$

The maximum value for $m+j$ is q so that

$$R_{XX}[m] = \sum_{j=0}^{q-m} b_j b_{j+m} \sigma_e^2 \quad m \geq 0$$

and

$$R_{XX}[-m] = R_{XX}[m]$$

Notice that $R_{XX}[m]$ is related by a nonlinear relationship with model parameters. Thus finding the model parameters is not simple.

The power spectral density is given by

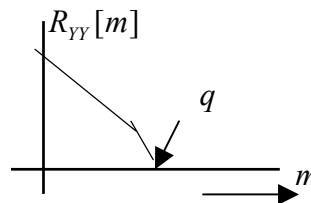
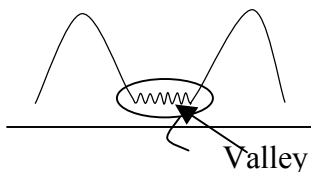
$$S_{XX}(w) = \frac{\sigma_e^2}{2\pi} |B(w)|^2$$

where

$$B(w) = b_0 + b_1 e^{-jw} + \dots + b_q e^{-jqw}$$

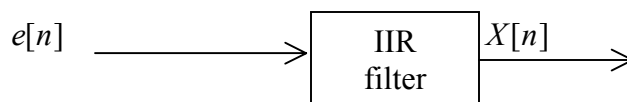
FIR system will give some zeros. So if the spectrum has some valleys then MA will fit well.

Test for MA process: $R_{YY}[m]$ becomes zero suddenly after some value of m .



Autoregressive Model

In time series analysis it is called AR(p) model.



The model is given by the difference equation

$$X[n] = -\sum_{i=1}^p a_i X[n-i] + e[n]$$

The transfer function $A(w)$ is given by

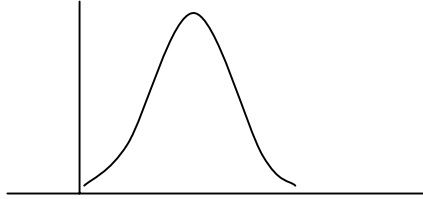
$$A(w) = \frac{1}{\sum_{i=0}^n a_i e^{-j\omega i}}$$

with $a_0 = 1$ (all poles model)

and

$$S_{yy}(\omega) = S_{xx}(\omega) = \frac{\sigma_e^2}{2\pi |A(\omega)|^2}$$

If there are sharp peaks in the spectrum, the AR(p) model may be suitable.



The autocorrelation function $R_{xx}[m]$ is given by

$$\begin{aligned} R_{xx}[m] &= E X[n] X[n-m] \\ &= -\sum_{i=1}^p a_i E X[n-i] X[n-m] + E e[n] X[n-m] \\ &= -\sum_{i=1}^p a_i R_{xx}[m-i] + \sigma_e^2 \delta[m] \end{aligned}$$

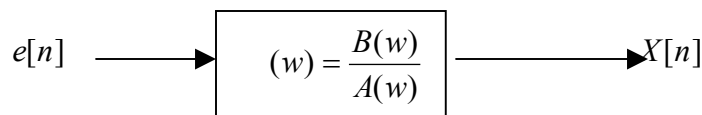
$$\therefore R_{xx}[m] = -\sum_{i=1}^p a_i R_{xx}[m-i] + \sigma_e^2 \delta[m] \quad \forall m \in I$$

The above relation gives a set of linear equations which can be solved to find a_i s.

This set of equations are known as Yule-Walker Equation.

ARMA – Autoregressive Moving Average Model.

Under the most practical situation, the process may be considered as an output of a filter that has both zeros and poles.



The model is given by

$$x[n] = -\sum_{i=1}^p a_i X[n-i] + \sum_{i=0}^q b_i v[n-i]$$

and is called the $ARMA(p, q)$ model.

The transfer function of the filter is given by

$$H(\omega) = \frac{B(\omega)}{A(\omega)}$$

$$S_{xx}(\omega) = \frac{|B(\omega)|^2 \sigma_e^2}{|A(\omega)|^2 2\pi}$$

How do we get the model parameters?

For $m \geq \max(p, q+1)$, there will be no contributions from b_i terms to $R_{xx}[m]$.

$$R_{yy}[m] = R_{xx}[m] = \sum_{i=1}^p a_i R_{xx}[m-i] \quad m \geq \max(p, q+1)$$

From a set of p Yule Walker equations, a_i parameters can be found out.

Then we can rewrite the equation

$$\tilde{X}[n] = X[n] + \sum_{i=1}^p a_i X[n-i]$$

$$\therefore \tilde{X}[n] = \sum_{i=0}^q b_i v[n-i]$$

From the above equation b_i s can be found out.

The $ARMA(p, q)$ is an economical model. Only $AR(p)$ only $MA(q)$ model may require a large number of model parameters to represent the process adequately. This concept in model building is known as the parsimony of parameters.

General Model building Steps

- Identification of p and q .
- Estimation of model parameters.
- Check the modeling error.
- If it is white noise then stop.
Else select new values for p and q
and repeat the process.

Other model:

ARIMA model: Here after differencing the data can be fed to an ARMA model.

SARMA model: Seasonal ARMA model etc.