Autocovariance generating function:

This is a function constructed by taking the j^{th} autocovariance, multiplying by some number z raised to the j^{th} power, and adding all the terms together:

$$g_y(z) = \sum_{-\infty}^{\infty} \gamma_j z^j$$

Ex. MA(1):
$$y_t = \varepsilon_t + \theta_1 \varepsilon_{t-1}$$

$$\gamma_0 = (1 + \theta_1^2)\sigma^2$$

$$\gamma_{-1} = \gamma_1 = \theta_1 \sigma^2$$

$$\gamma_2 = \gamma_{-2} = \gamma_3 = \gamma_{-3} = \dots = 0$$

$$g_y(z) = \gamma_0 z^0 + \gamma_{-1} z^{-1} + \gamma_1 z^1 = (1 + \theta_1^2) \sigma^2 + \theta_1^2 \sigma^2 z^{-1} + \theta_1^2 \sigma^2 z = ((1 + \theta_1^2) + \theta_1 z^{-1} + \theta_1 z) \sigma^2$$
$$= \sigma^2 (1 + \theta_1 z) (1 + \theta_1 z^{-1})$$

MA(q):

$$g_y z = \sigma^2 (1 + \theta_1 z^1 + \theta_2 z^2 + \dots + \theta_q z^q) (1 + \theta_1 z^{-1} + \theta_2 z^{-2} + \dots + \theta_q z^{-q})$$

$MA(\infty)$:

$$q_t(z) = \sigma^2 \theta(z) \theta(z^{-1}) = (\theta_1 + \theta_1 \theta_2 + \theta_2 \theta_3 + \dots) z^1 + (\theta_2 + \theta_1 \theta_3 + \theta_2 \theta_4 + \dots) z^2 + \dots + (\theta_j + \theta_1 \theta_{j+1} + \theta_2 \theta_{j+2} + \dots) z^j + \dots$$

$$\implies \gamma_1 = (\theta_1 + \theta_1\theta_2 + \theta_2\theta_3 + ...)$$

AR(1): Suppose stationarity

$$\phi(L)(y_t - \mu) = \varepsilon_t$$
 $(\phi(L) = (1 - \phi L))$

Stationarity $\implies \phi^{-1}(L)$ exists

$$\implies y_t - \mu = \phi^{-1}(L)\varepsilon_t = (\sum_{j=0}^{\infty} \phi^j L^j)\varepsilon_t$$

$$g_y(z) = \sigma^2 \phi^{-1}(z)\phi^{-1}(z^{-1}) = \sigma^2(\sum_{j=0}^{\infty} \phi^j z^j)(\sum_{j=0}^{\infty} \phi^j z^{-j})$$

$$= \sigma^2(1 + \phi^1 z^1 + \phi^2 z^2 + \dots)(1 + \phi^1 z^{-1} + \phi^2 z^{-2} + \dots)$$

$$\implies \gamma_j = \sigma^2(\phi^j + \phi^1\phi^{j+1} + \phi^2\phi^{j+2} + \dots)$$

$$= \sigma^2 \phi^j (1 + \phi^2 + \phi^4 + \phi^6 + \phi^8 + \ldots)$$

$$=\sigma^2 \phi^j \sum_{j=0}^{\infty} (\phi^2)^j = \frac{\sigma^2 \phi^j}{1-\phi^2}$$

Recall MA(q) process:

 $y_t = \mu + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \dots + \theta_q \varepsilon_{t-q}$ is stationary for all values $\{\theta_j\}_{j=1}^q$

- $\theta(L)$ can be factored as: $\theta(L) = (1 + \theta_1 L + ... + \theta_q L) = (1 \eta_1 L)(1 \eta_2 L)...(1 \eta_q L)$
- $\{\frac{1}{\eta_i}\}_{i=1}^q$ are the roots of $\theta(L)$
- Unlike AR process we don't need restrictions on $\theta(L)$ to ensure stationarity

Suppose $|\eta_i| < 1 \forall i$. Then $(1 - \eta_i L)^{-1} = \sum_{j=0}^{\infty} \eta_i^j L^j \forall i$.

Then:
$$\theta(L)^{-1} = (\sum_{j=0}^{\infty} \eta_1^j L^j)(\sum_{j=0}^{\infty} \eta_2^j L^j)...(\sum_{j=0}^{\infty} \eta_q^j L^j)$$

Begin with $y_t = \theta(L)\varepsilon_t$ if all roots of $\theta(L)$ lie outside the unit circle then $\theta^{-1}(L)$ exists and we can write: $\theta^{-1}(L)y_t = \varepsilon_t \implies \varepsilon_t = \theta^{-1}(L)y_t$

Motivation for Causality:

Suppose
$$y_t = \phi y_{t-1} + \varepsilon_t$$
 $\varepsilon_t \sim WN(0, \sigma^2)$

What if $\phi > 1$?

Run the recursion forward:

$$\begin{aligned} \phi y_{t-1} &= y_t - \varepsilon_t \implies y_{t-1} = \frac{1}{\theta} y_t - \frac{1}{\theta} \varepsilon_t \\ &= \frac{1}{\phi} (\frac{1}{\phi} y_{t+1} - \frac{1}{\phi} \varepsilon_{t+1}) - \frac{1}{\phi} \varepsilon_t = (\frac{1}{\phi})^2 y_{t+1} - (\frac{1}{\phi})^2 \varepsilon_{t+1} - \frac{1}{\phi} \varepsilon_t \\ &= \sum_{j=0}^{\infty} (\frac{1}{\phi})^j \varepsilon_{t+j} \end{aligned}$$

Stationary, but involves future $\varepsilon \implies$ but this isn't causal.

So need inside unit circle for causality (i.e. inly passed ε_s).

Invertability allows for causality.