

MTH 442 ASSIGNMENT 2

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1) Given MA(1) model: $X_t = w_t + \theta w_{t-1}$

ACF for lag 1:

$$\begin{aligned}\gamma(1) &= \text{Cov}(X_{t+1}, X_t) \\ &= \text{Cov}[(w_{t+1} + \theta w_t)(w_t + \theta w_{t-1})] \\ &= E[(w_{t+1} + \theta w_t)(w_t + \theta w_{t-1})] \quad [\because E[X_t] = 0] \\ &= \theta E[w_t^2] = \sigma_w^2 \theta\end{aligned}$$

$$\begin{aligned}\gamma(0) &= \text{Cov}(X_t, X_t) \\ &= E[X_t^2] \\ &= E[(w_t + \theta w_{t-1})(w_t + \theta w_{t-1})] \\ &= E[w_t^2] + \theta^2 E[w_{t-1}^2] \\ &= \sigma_w^2 (1 + \theta^2)\end{aligned}$$

$$\therefore \rho(1) = \frac{\gamma(1)}{\gamma(0)} = \frac{\theta}{1 + \theta^2}$$

We know that $AM \geq GM$,

$$\therefore \left[\frac{\theta + 1}{\theta} \right] \geq \sqrt{\theta \cdot 1}$$

$$\Rightarrow \left[\frac{\theta^2 + 1}{\theta} \right] \geq 2$$

$$\Rightarrow \left[\frac{\theta}{\theta^2 + 1} \right] \leq \frac{1}{2}$$

$$\Rightarrow \rho(1) \leq \frac{1}{2}$$

\therefore lag 1 ACF is bounded above by 0.5.

2) Given, $\{W_t; t=0, 1, \dots\}$: white noise process
 $|\phi| < 1$
 $X_0 = W_0$ and $X_t = \phi X_{t-1} + W_t ; t=1, 2, \dots$

(a) We can write iteratively as

$$\begin{aligned} X_t &= \phi X_{t-1} + W_t \\ &= \phi(\phi X_{t-2} + W_{t-1}) + W_t \\ &= \phi^2 X_{t-2} + (\phi W_{t-1} + W_t) \\ &\vdots \\ &= \phi^k X_{t-k} + \sum_{j=0}^{k-1} \phi^j W_{t-j} \end{aligned}$$

for $k=t$.

$$\begin{aligned} X_t &= \phi^t X_0 + \sum_{j=0}^{t-1} \phi^j W_{t-j} \\ \Rightarrow X_t &= \sum_{j=0}^{t-1} \phi^j W_{t-j} \quad [\because X_0 = W_0] \end{aligned}$$

$$\begin{aligned} (b) E(X_t) &= \sum_{j=0}^t E[\phi^j W_{t-j}] \\ &= \sum_{j=0}^t \phi^j E[W_{t-j}] \\ &= 0 \end{aligned}$$

$$\begin{aligned} (c) \text{Var}(X_t) &= \text{Var}\left(\sum_{j=0}^t \phi^j W_{t-j}\right) \\ &= \sum_{j=0}^t \phi^{2j} E\left[\left(\sum_{j=0}^t \phi^j W_{t-j}\right)\left(\sum_{j=0}^t \phi^j W_{t-j}\right)\right] \\ &= \sum_{j=0}^t \phi^{2j} E[W_{t-j}^2] \\ &= \sigma_w^2 \sum_{j=0}^t \phi^{2j} = \sigma_w^2 \left[\frac{1 - \phi^{2(t+1)}}{1 - \phi^2} \right] \end{aligned}$$

(d)

$$\begin{aligned}
 \text{Cov}(X_{t+h}, X_t) &= E(X_{t+h} X_t) \\
 &= E\left[\left(\sum_{j=0}^{t+h} \phi^j W_{t+h-j}\right)\left(\sum_{j=0}^t \phi^j W_{t+j}\right)\right] \quad [\because E[X_{t+h}] = E[X_t] = 0] \\
 &= E\left[\left(\sum_{j=0}^{h-1} \phi^j W_{t+h-j} + \sum_{j=h}^{t+h} \phi^j W_{t+h-j}\right)\left(\sum_{j=0}^t \phi^j W_{t+j}\right)\right] \\
 &= E\left[\left(\sum_{j=0}^{h-1} \phi^j W_{t+h-j}\right)\left(\sum_{j=0}^t \phi^j W_{t+j}\right)\right] + E\left[\left(\sum_{j=h}^{t+h} \phi^j W_{t+h-j}\right)\left(\sum_{j=0}^t \phi^j W_{t+j}\right)\right] \\
 &= 0 + E\left[\left(\sum_{j=h}^{t+h} \phi^j W_{t+h-j}\right)\left(\sum_{j=0}^t \phi^j W_{t+j}\right)\right] \\
 &= E\left[\left(\phi^h W_t + \phi^{h+1} W_{t+1} + \phi^{h+2} W_{t+2} + \dots + \phi^{t+h} W_0\right)\left(\phi^h W_t + \phi^{h+1} W_{t+1} + \dots + \phi^{t+h} W_0\right)\right] \\
 &= \phi^h E[W_t^2] + \phi^{h+2} E[W_{t+1}^2] + \dots + \phi^{t+2h} E[W_0^2] \\
 &= \phi^h \sigma_w^2 \left[\phi^h + \phi^{h+2} + \dots + \phi^{t+2h} \right] \\
 &= \phi^h \sigma_w^2 (1 + \phi^2 + \dots + \phi^{2h}) \\
 &= \phi^h \sigma_w^2 \cdot \frac{(1 - (\phi^2)^{t+1})}{1 - \phi^2} \\
 &= \phi^h \text{Var}(X_t)
 \end{aligned}$$

(e)

X_t is not stationary since $\text{Cov}(X_{t+h}, X_t)$ depends on t .

(f)

$$\begin{aligned}
 \text{As } t \rightarrow \infty \quad \text{Cov}(X_{t+h}, X_t) &= \phi^h \text{Var}(X_t) \\
 \text{it } \lim_{t \rightarrow \infty} \text{Cov}(X_{t+h}, X_t) &= h \phi^h \sigma_w^2 \frac{(1 - \phi^{2(t+1)})}{1 - \phi^2} \\
 &= \phi^h \sigma_w^2 \quad [\because |\phi| < 1]
 \end{aligned}$$

Now, the $\text{Cov}(X_{t+h}, X_t) = r(h)$ is independent of t and depends only on h .
 Further, $E[X_t] = 0$ and $\text{Var}(X_t)$ is finite

for every $t > 0$.

$\therefore X_t$ is asymptotically stationary.

(g) We can generate ~~n observations~~ more than n observations like $n+n_0$, where n_0 is fairly large and discard the first n_0 .

The remaining n observations can then be used to simulate asymptotically stationary AR(1) model

(h) Consider $X_0 = \frac{W_0}{\sqrt{1-\phi^2}}$

$$\text{Now, } X_t = \phi X_{t-1} + W_t \\ = \phi^t X_0 + \sum_{j=0}^{t-1} \phi^j W_{t-j}$$

$$\Rightarrow X_t = \frac{\phi^t W_0}{\sqrt{1-\phi^2}} + \sum_{j=0}^{t-1} \phi^j W_{t-j}$$

$$\Rightarrow E[X_t] = \frac{\phi^t}{\sqrt{1-\phi^2}} E[W_0] + \sum_{j=0}^{t-1} \phi^j E[W_{t-j}]$$

$$\Rightarrow E[X_t] = 0$$

$\therefore E[X_t]$ is constant.

$$\begin{aligned} r(h) &= \text{Cov}(X_{t+h}, X_t) = E[X_{t+h} X_t] \\ &= E\left[\left(\frac{\phi^{t+h}}{\sqrt{1-\phi^2}} W_0 + \sum_{j=0}^{t+h-1} \phi^j W_{t+h-j}\right) \left(\frac{\phi^t}{\sqrt{1-\phi^2}} W_0 + \sum_{j=0}^{t-1} \phi^j W_{t-j}\right)\right] \\ &= E\left[\left(\frac{\phi^{t+h}}{\sqrt{1-\phi^2}} W_0 + \sum_{j=h}^{t+h-1} \phi^j W_{t+h-j} + \sum_{j=0}^{t-1} \phi^j W_{t+h-j}\right) \left(\frac{\phi^t}{\sqrt{1-\phi^2}} W_0 + \sum_{j=0}^{t-1} \phi^j W_{t-j}\right)\right] \\ &= E\left[\left(\frac{\phi^{t+h}}{\sqrt{1-\phi^2}} W_0 + \sum_{j=h}^{t+h-1} \phi^j W_{t+h-j}\right) \left(\frac{\phi^t}{\sqrt{1-\phi^2}} W_0 + \sum_{j=0}^{t-1} \phi^j W_{t-j}\right)\right] \end{aligned}$$

$$\begin{aligned}
 &= \phi^h E[W_+^2] + \phi^{h+2} E[W_{+-}^2] + \dots + \phi^{h+2(t-1)} E[W_t^2] \\
 &\quad + \frac{\phi^{h+2t}}{1-\phi^2} E[W_0^2] \\
 &= \sigma_w^2 \phi^h \left(1 + \phi^2 + \dots + (\phi^2)^{t-1} + \frac{(\phi^2)^t}{1-\phi^2} \right) \\
 &= \sigma_w^2 \phi^h \cdot \left[\frac{(1-\phi^{2t})}{1-\phi^2} + \frac{\phi^{2t}}{1-\phi^2} \right] \\
 &= \sigma_w^2 \phi^h \cdot \frac{1}{1-\phi^2} = \frac{\phi^h \sigma_w^2}{1-\phi^2}
 \end{aligned}$$

$$\text{Var}(X_t) = \gamma(0) = \frac{\phi^h}{1-\phi^2} \sigma_w^2 < \infty$$

We can see that $\gamma(h)$ is independent of t and depends only on h . Further, $\text{Var}(X_t)$ is finite ~~for all t~~ .

$\therefore X_t$ is stationary when $X_0 = \frac{W_0}{\sqrt{1-\phi^2}}$

3) Given AR(2) model:

$$(1 - \phi_1 z - \phi_2 z^2) X_t = W_t$$

The AR polynomial will be:

$$\phi(z) = 1 - \phi_1 z - \phi_2 z^2$$

roots of $\phi(z)$ are:

$$\phi_1 \pm \frac{\sqrt{\phi_1^2 + 4\phi_2}}{-2\phi_1}$$

The given model is causal if and only if:

$$\left| \frac{\phi_1 \pm \sqrt{\phi_1^2 + 4\phi_2}}{-2\phi_1} \right| > 1$$

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The roots of $\phi(z)$ may be real and distinct, real and equal, or a complex conjugate pair.

Roots are real if $\phi_1^2 + 4\phi_2 \geq 0$ else roots are complex conjugates.

Consider z_1, z_2 as the roots of $\phi(z)$ such that $z_1, z_2 \in \mathbb{C}$. and $\bar{z}_1 = \bar{z}_2$.

Model is causal if and only if $|z_1| > 1$ and $|z_2| > 1$

~~We can write $\phi(z)$ as~~

$$\phi(z) = (1 - z_1^{-1}z)(1 - z_2^{-1}z)$$

$$\Rightarrow \phi(z) = 1 - (z_1^{-1} + z_2^{-1})z + z_1^{-1}z_2^{-1}z^2$$

Comparing the above expression with

$$\phi(z) = 1 - \phi_1 z - \phi_2 z^2$$

we get,

$$\phi_1 = z_1^{-1} + z_2^{-1}, \quad \phi_2 = -(z_1 z_2)^{-1}$$

Let $u_1 = z_1^{-1}$ and $u_2 = z_2^{-1}$

$$\Rightarrow \phi_1 = u_1 + u_2 \Rightarrow |u_1| = |z_1^{-1}| \text{ and } |u_2| = |z_2^{-1}|$$

$$\Rightarrow |u_1| = \frac{1}{|z_1|} < 1 \quad \text{and} \quad |u_2| = \frac{1}{|z_2|} < 1$$

Also,

$$\phi_2 = \frac{-1}{z_1 z_2} \Rightarrow |\phi_2| = \frac{1}{|z_1||z_2|} < 1$$

$$\phi_1 + \phi_2 = -1 = z_1^{-1} + z_2^{-1} - (z_1 z_2)^{-1} - 1$$

$$= u_1 + u_2 - u_1 u_2 - 1$$

$$= u_1(1 - u_2) + u_2 - 1$$

$$= -(1 - u_1)(1 - u_2)$$

$$\Rightarrow \phi_1 + \phi_2 < 1$$

$$\begin{aligned}
 \phi_2 - \phi_1 - 1 &= -z_1^{-1} - z_2^{-1} - (z_1 z_2)^{-1} - 1 \\
 &= -u_1 - u_2 - u_1 u_2 - 1 \\
 &= -(1+u_1)(1+u_2) \quad \text{←} \\
 &\quad \text{←}
 \end{aligned}$$

Consider the case when $z_1, z_2 \in \mathbb{R}$.

$$\therefore \phi_2 + \phi_1 - 1 = -(1-u_1)(1-u_2) < 0$$

$[\because |u_1| \text{ and } |u_2| < 1]$

$$\phi_2 - \phi_1 - 1 = -(1+u_1)(1+u_2) < 0$$

\therefore We have following conditions when roots are real,
 $|\phi_2| < 1, \phi_2 + \phi_1 < 1, \phi_2 - \phi_1 < 1$

Now, consider the case when $z_1, z_2 \in \mathbb{C}$ and
 $z_1 = \bar{z}_2$.

$$\begin{aligned}
 \phi_2 + \phi_1 - 1 &= -(1-u_1)(1-u_2) \\
 &= -(1-u_1)(1-\bar{u}_1) \quad [\because z_2 = \bar{z}_1 \Rightarrow u_2 = \bar{u}_1] \\
 &= -(1-u_1)(\bar{1-u_1}) \\
 &= -|(1-u_1)|^2 < 0
 \end{aligned}$$

$$\begin{aligned}
 \phi_2 - \phi_1 - 1 &= -(1+u_1)(1+u_2) \\
 &= -(1+u_1)(1+\bar{u}_1) \\
 &= -(1+u_1)(\bar{1+u_1}) \\
 &= -|(1+u_1)|^2 < 0
 \end{aligned}$$

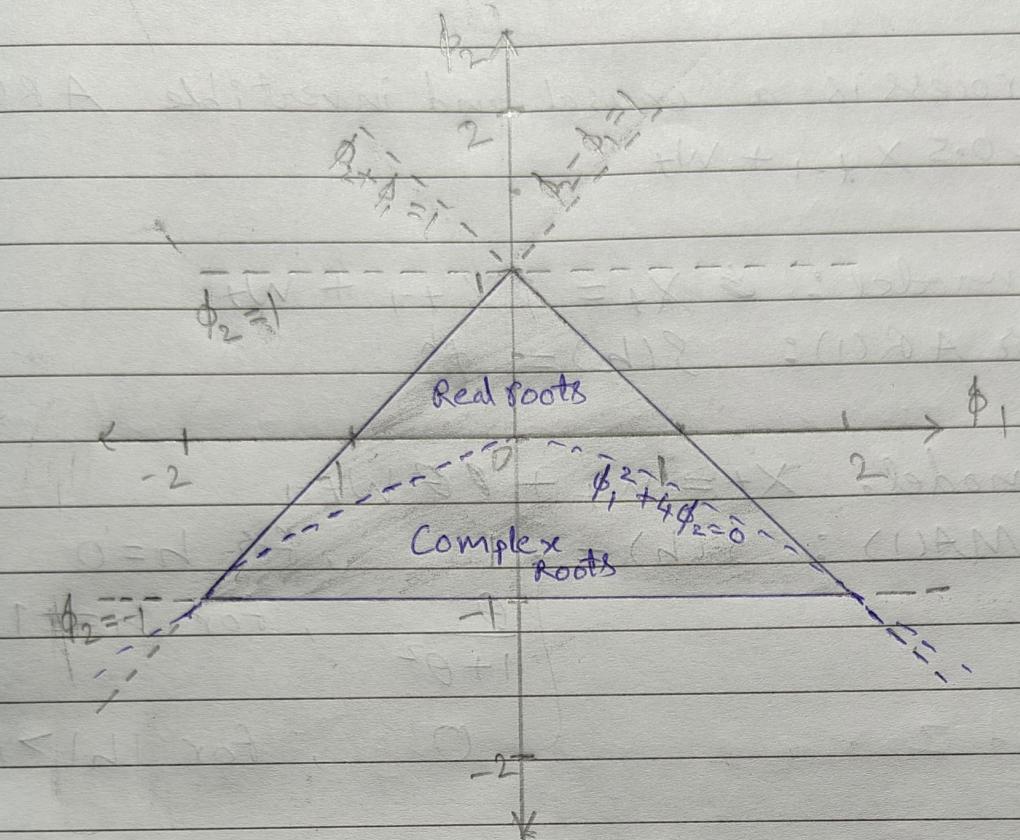
\therefore We again have following conditions when roots are complex conjugates,

$$|\phi_2| < 1, \phi_2 + \phi_1 < 1, \phi_2 - \phi_1 < 1$$

Thus, we have the following conditions in the general case to establish causality,
 $|\phi_2| < 1, \phi_2 + \phi_1 < 1, \phi_2 - \phi_1 < 1$

This causality condition specifies a triangular region in the parameter space.

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4) (i) Given, AR(2) model:

$$X_t = 0.25 X_{t-2} + W_t$$

AR polynomial is given by,
 $\phi(z) = 1 - 0.25 z^2$

Roots of $\phi(z)$:

$$\begin{aligned}\phi(z) &= 0 \\ \Rightarrow 1 - 0.25 z^2 &= 0 \\ \Rightarrow (z - \frac{1}{\sqrt{2}})(z + \frac{1}{\sqrt{2}}) &= 0 \\ \Rightarrow z &= \pm 2\end{aligned}$$

$$X_t = 0.25 X_{t-2} + W_t$$

Multiply each side by X_{t-h} for $h > 0$ and take expectation
 $E[X_t X_{t-h}] = 0.25 E[X_{t-2} X_{t-h}] + E[W_t X_{t-h}]$

$$\Rightarrow r(h) = 0.25 r(h-2), \quad h = 1, 2, \dots$$

Dividing by $r(0)$,

$$s(h) = 0.25 s(h-2)$$

$$\Rightarrow s(h) - 0.25 s(h-2) = 0$$

This is a difference equation with order 2.

\therefore Solution of $s(h)$ is

$$s(h) = c_1 z_1^{-h} + c_2 z_2^{-h}$$

Also, Initial conditions:

$$(i) s(0) = \frac{s(0)}{r(0)} = 1$$

$$(ii) s(1) = \frac{r(1)}{r(0)} = 0.25 s(-1) = 0.25 s(1-2)$$

$$\Rightarrow s(1) - 0.25 s(-1) = 0$$

$$\Rightarrow s(1)[1 - 0.25] = 0 \quad [\because s(-1) = s(1)]$$

$$\Rightarrow s(1) = 0$$

Using $\beta(0) = 1$,

$$\beta(0) = c_1 z_1^{-0} + c_2 z_2^{-0}$$

$$\Rightarrow c_1 + c_2 = 1$$

Using $\beta(1) = 0$,

~~$$\beta(1) = c_1 z_1^{-1} + c_2 z_2^{-1} = 0$$~~

$$\Rightarrow \frac{c_1}{2} - \frac{c_2}{2} = 0$$

$$\Rightarrow c_1 = c_2$$

\therefore We get, $c_1 = 1/2$ and $c_2 = 1/2$

\therefore ~~General~~ $\beta(h)$ becomes,

$$\beta(h) = \frac{(2)^{-h}}{2} + \frac{(-2)^{-h}}{2}$$

$$\Rightarrow \beta(h) = [1 + (-1)^h] \frac{2^{-(h+1)}}{2}$$

(ii)

Given AR(2) model:

$$X_t = -0.9 X_{t-2} + W_t$$

AR polynomial is given by:

$$\phi(z) = 1 + 0.9z^2$$

Roots of $\phi(z)$:

$$\phi(z) = 0$$

$$\Rightarrow 1 + 0.9z^2 = 0$$

$$\Rightarrow z^2 = -0.9^{-1}$$

$$\Rightarrow z = \pm \frac{i}{\sqrt{0.9}}$$

$$\arg(z) = \pi/2$$

$$\therefore z = \pm \frac{1}{\sqrt{0.9}} e^{i\pi/2}$$

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$$X_t = -0.9 X_{t-2} + W_t$$

Multiply each side by X_{t+h} for $h > 0$, and take expectation,

$$E[X_t X_{t+h}] = -0.9 E[X_{t-2} X_{t+h}] + E[W_t X_{t+h}]$$

$$\Rightarrow r(h) = -0.9 r(h-2) \quad ; \quad h=1, 2, \dots$$

Dividing by $r(0)$ gives,

$$\therefore s(h) = -0.9 s(h-2)$$

$$\Rightarrow s(h) + 0.9 s(h-2) = 0$$

This is a difference equation with order 2 and $\alpha_1 = 0, \alpha_2 = 0.9$.

\therefore Solution of $s(h)$ is

$$s(h) = C_1 z_1^{-h} + C_2 z_2^{-h}$$

$\because z_1$ and z_2 are a complex conjugate pair, then

$$C_2 = \bar{C}_1 \quad [\because s(h) \text{ is real}]$$

$$\Rightarrow s(h) = C_1 z_1^{-h} + \bar{C}_1 \bar{z}_1^{-h}$$

Writing $z_1 = |z_1| \exp(i\theta)$ in the polar representation,

$$s(h) = a |z_1|^h \cos(h\theta + b) = a (\sqrt{0.9})^h \cos\left(h\frac{\pi}{2} + b\right)$$

Initial conditions,

$$(i) \quad s(0) = \frac{s(0)}{s(0)} = 1$$

$$(ii) \quad s(1) + 0.9 s(1-2) = 0$$

$$\Rightarrow s(1) + 0.9 s(1) = 0$$

$$\Rightarrow s(1) = 0$$

$$[\because s(-1) = s(1)]$$

Using $s(0) = 1$,

$$s(0) = a (\sqrt{0.9})^0 \cos\left(\frac{b\pi}{2} + b\right) = a \cos b$$

$$\Rightarrow a \cos b = 1$$

Using $f(1) = 0$

$$f(1) = a \sqrt{0.9} \cos\left(\frac{\pi}{2} + b\right)$$

$$\Rightarrow \cos\left(\frac{\pi}{2} + b\right) = 0 \rightarrow b = 0$$

$$\therefore a \cos b = 1 \Rightarrow a = 1$$

$\therefore f(h)$ becomes

$$f(h) = (0.9)^{h/2} \cos\left(\frac{h\pi}{2}\right)$$

5) Given ARMA model:

$$X_t = 0.25 X_{t-2} + u$$

$$X_t = 0.80 X_{t-1} - 0.15 X_{t-2} + W_t - 0.30 W_{t-1}$$

$$\Rightarrow X_t - 0.8 X_{t-1} + 0.15 X_{t-2} = W_t - 0.30 W_{t-1}$$

AR polynomial:

$$\phi(z) = 1 - 0.8z + 0.15z^2 = (1 - 0.3z)(1 - 0.5z)$$

MA polynomial:

$$\theta(z) = 1 - 0.3z$$

We can see that the factor $(1 - 0.3z)$ is common to both the AR and MA polynomials. Therefore, yes, there is parameter redundancy.

The model reduces to an AR(1) process:

$$(1 - 0.5B) X_t = W_t$$

The root of AR polynomial is $z_\phi = 2 > 1$

\therefore The process is causal.

And there is no root such that $|Z_0| \leq 1$ in the MA polynomial, hence, the process is invertible.

\therefore The process is a causal and invertible AR(1):

$$X_t = 0.5 X_{t-1} + W_t$$

b) AR(1) model: $X_t = \phi X_{t-1} + W_t$

$$\text{ACF for AR(1)}: \rho(h) = \phi^h$$

MA(1) model: $X_t = W_t + \theta W_{t-1}$

$$\text{ACF for MA(1)}: \rho(h) = \begin{cases} 1 & , \text{ for } h=0 \\ \frac{\theta}{1+\theta^2} & , \text{ for } h=\pm 1 \\ 0 & , \text{ for } |h| > 1 \end{cases}$$

ARMA(1,1) model: $X_t = \phi X_{t-1} + W_t + \theta W_{t-1}$

$$\text{ACF for ARMA(1,1)}: \rho(h) = \frac{(1+\theta\phi)(\phi+\theta)}{1+2\theta\phi+\theta^2} \phi^{h-1}, \quad h \geq 1$$

We need to plot these functions for all three cases with $\phi=0.6$ and $\theta=0.9$.