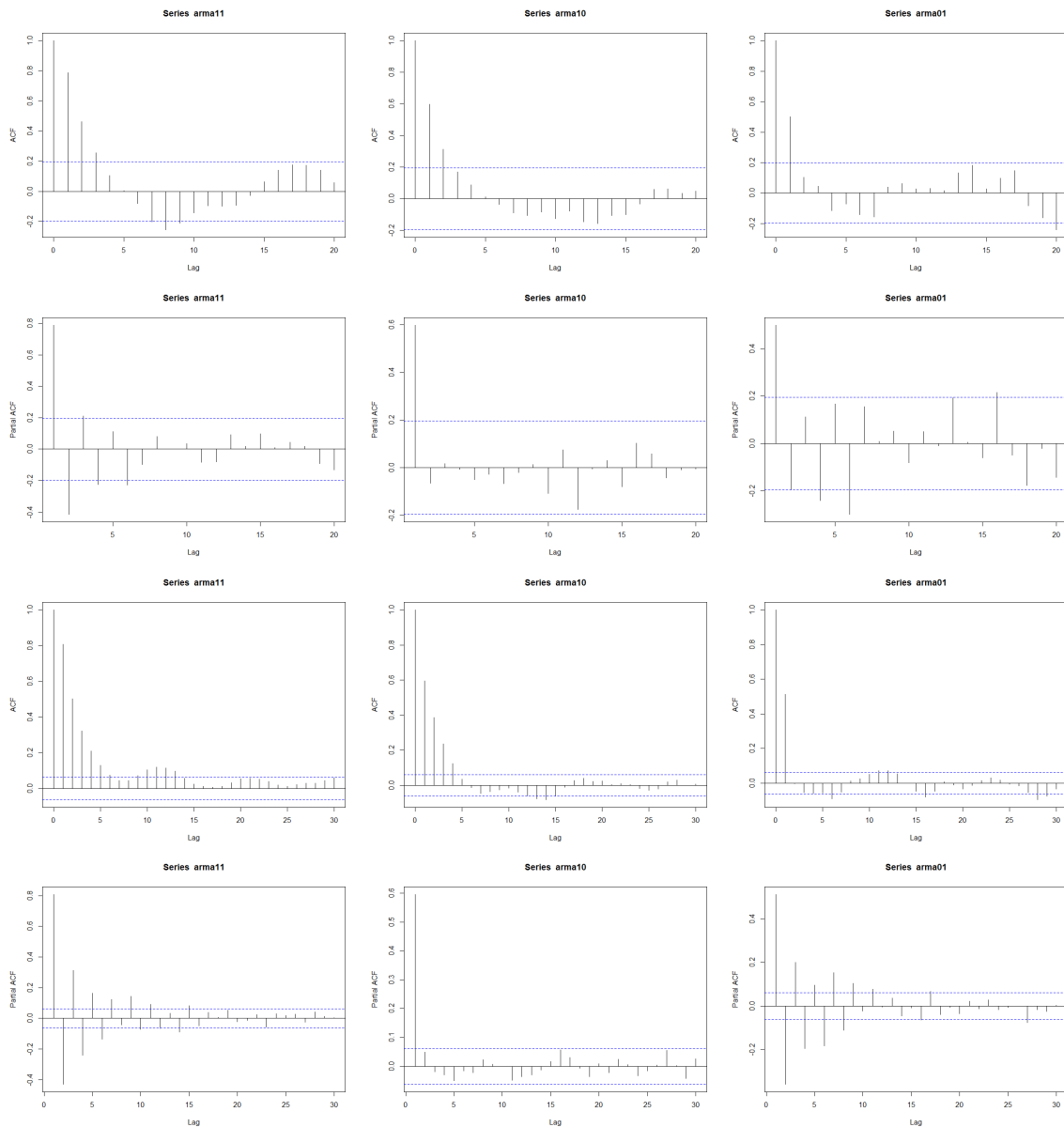


3.9 The plots is as follows. The plot is not so good because of white noise. If we sample more data, like  $n = 1000$ , which is shown below, it fits Table 3.1 better.



### #Problem 3.9

$n = 1000$

```
arma11 = arima.sim(n = n, list(ar = 0.6, ma = 0.9))
```

```
arma10 = arima.sim(n = n, list(ar = 0.6))
```

```
arma01 = arima.sim(n = n, list(ma = 0.9))
```

```
acf(arma11); acf(arma10); acf(arma01)
```

```
pacf(arma11); pacf(arma10); pacf(arma01)
```

3.12

$\forall h > 0$ , by Cauchy Inequality,

$$\begin{aligned} |\gamma(h)| &= |\mathbb{E}[(x_0 - \mu)(x_h - \mu)]| \\ &\leq \sqrt{\mathbb{E}[(x_0 - \mu)^2] \mathbb{E}[(x_h - \mu)^2]} \\ &= \gamma(0) \end{aligned}$$

The equation holds iff the distribution of  $x_0$  and  $x_h$  are equal. Since the pick of  $x_0$  is alternative, if  $\gamma(h) = \gamma(0)$ , we can prove that for any positive integer  $k$ ,  $x_0 = x_{kh}$ , thus,

$$\gamma(kh) = \gamma(0) > 0$$

Let  $k \rightarrow \infty$ , it is contradict to  $\gamma(h) \rightarrow 0$ . Thus,  $\forall h > 0$ ,  $|\gamma(h)| < \gamma(0)$ . It means that  $\Gamma$  is a strictly diagonally dominant matrix. So  $|\Gamma| \neq 0$ . By the proof of problem 1.25, the sample autocovariance is a non-negative definite function. So  $\Gamma$  is positive semidefinite. Since  $|\Gamma| \neq 0$ ,  $\Gamma$  is positive definite.

3.14 (a)

$$\begin{aligned} \arg \min_{g(x)} \text{MSE} &= \arg \min_{g(x)} \int \left[ \int (y - g(x))^2 f(y|x) \, dy \right] f(x) \, dx \\ &= \arg \min_{g(x)} \int (y - g(x))^2 f(y|x) \, dy \\ &= \arg \min_{g(x)} \left( \mathbb{E}[y^2|x] - 2\mathbb{E}[yg(x)|x] + \mathbb{E}[g^2(x)] \right) \\ &= \mathbb{E}[y|x] \end{aligned}$$

(b)

$$\begin{aligned} \text{MSE} &= \mathbb{E}[(x^2 + z - \mathbb{E}[x^2 + z|x])^2] \\ &= \mathbb{E}[z^2] \\ &= 1 \end{aligned}$$

(c) Apply  $g(x) = a + bx$  to the function of MSE,

$$\begin{aligned} \text{MSE} &= \int \left[ \int (y - a - bx)^2 f(y|x) \, dy \right] f(x) \, dx \\ &= \mathbb{E}[\mathbb{E}[(y - bx)^2|x] - 2a\mathbb{E}[y - bx|x] + a^2] \\ &= \mathbb{E}[(y - bx)^2] - 2a\mathbb{E}[y - bx] + a^2 \\ \arg \min_a \text{MSE} &= \mathbb{E}[y - bx] = 1 \end{aligned}$$

thus,

$$\begin{aligned}
 \text{MSE} &= \int \left[ \int (y - 1 - bx)^2 f(y|x) \, dy \right] f(x) \, dx \\
 &= \mathbb{E} \left[ \mathbb{E} \left[ (y - 1)^2 | x \right] - 2b \mathbb{E} [xy - x | x] + b^2 x^2 \right] \\
 &= b^2 \mathbb{E} [x^2] - 2b \mathbb{E} [xy - x] + \mathbb{E} [(y - 1)^2] \\
 \arg \min_b &= \frac{\mathbb{E} [xy - x]}{\mathbb{E} [x^2]} \\
 &= \frac{\mathbb{E} [xy]}{\mathbb{E} [x^2]} = 0 \\
 \text{MSE} &= \int \left[ \int (y - 1)^2 f(y|x) \, dy \right] f(x) \, dx \\
 &= \mathbb{E} [(x^2 + z - 1)^2] \\
 &= \mathbb{E} [x^4 + z^2 + 1 - 2x^2] \\
 &= 3 + 1 + 1 - 2 = 3
 \end{aligned}$$

Since  $x$  has zero mean, the slope is zero, and the intercept equals to mean of  $y$ . MSE measures the variance of  $y$ . In part (b), MSE measures the variance of  $z$