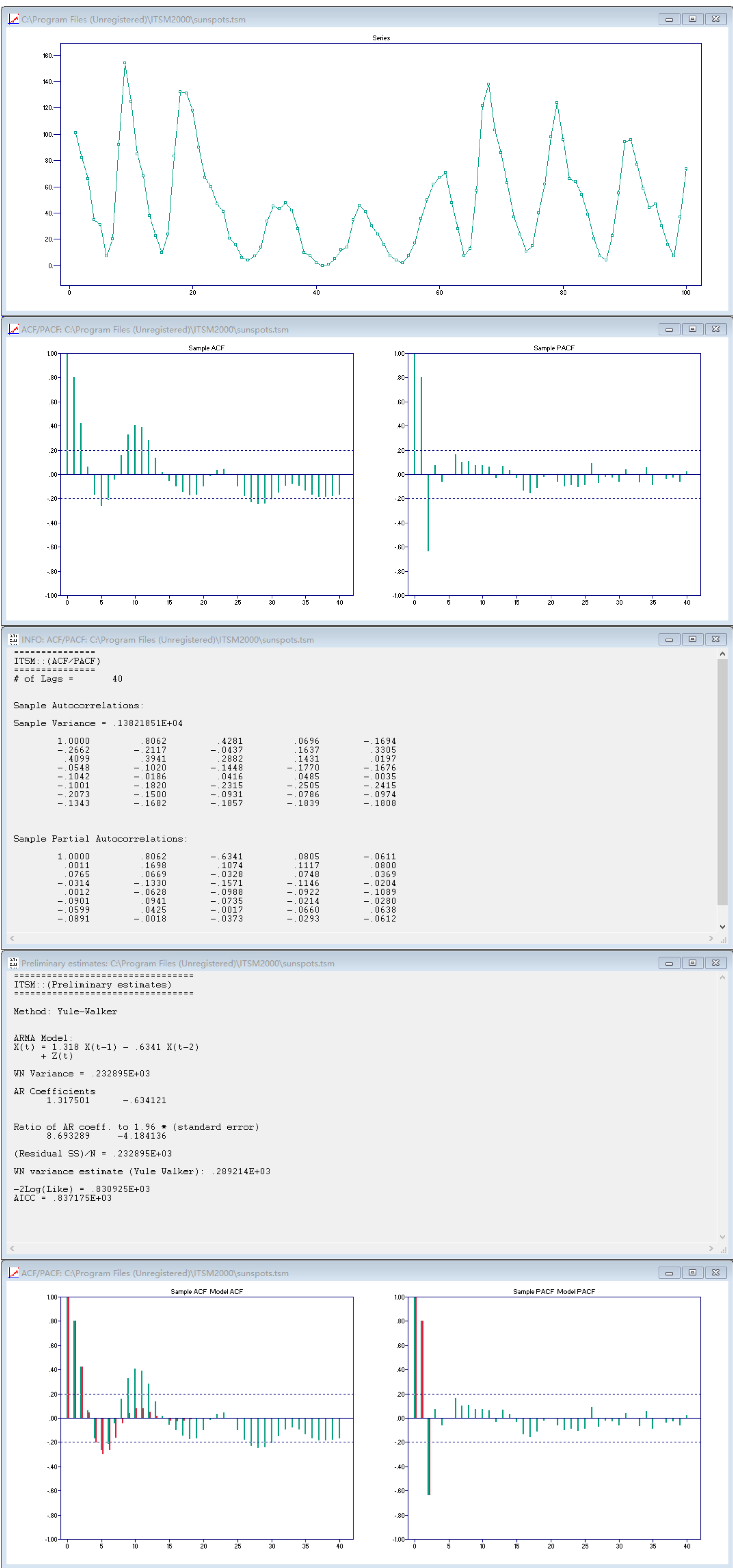


# Homework 5

```
library(itsmr)
```

2.16



3.2

(b)

$\psi(z) = 1 + 1.9z + 0.88z^2 = 0$  has roots  $z_1 = -\frac{10}{11}$ ,  $z_2 = -\frac{5}{4}$ , in which the first one is not outside the unit circle. So the process is not causal.

(d)

$\psi(z) = 1 + 1.8z + 0.81z^2 = 0$  has roots  $z_1 = z_2 = -\frac{10}{9}$ , which are both outside the unit circle. So the process is causal.

$\therefore \gamma(h) = (\frac{10}{9})^{-h}P(h)$ , where  $P(h)$  is a polynomial in  $h$  of degree 1. Thus  $\gamma(h) = (\frac{10}{9})^{-h}(ah + b)$ .

$\therefore \gamma(0) = b, \gamma(1) = \frac{9}{10}(a + b), \gamma(2) = \frac{81}{100}(2a + b)$ .

Combining with Y-W equations, which include

$$\gamma(0) - \phi_1\gamma(1) - \phi_1\gamma(2) = \sigma^2, \gamma(1) - \phi_1\gamma(0) - \phi_2\gamma(1) = 0, \gamma(2) - \phi_1\gamma(1) - \phi_2\gamma(0) = 0$$

We can get

$$a = -3.077\sigma^2, b = 1.462\sigma^2$$

$$\therefore \gamma(h) = (\frac{10}{9})^{-h}(-3.077h + 1.462)\sigma^2.$$

$$\therefore \rho(h) = (\frac{10}{9})^{-h} \frac{-3.077h + 1.462}{1.462}$$

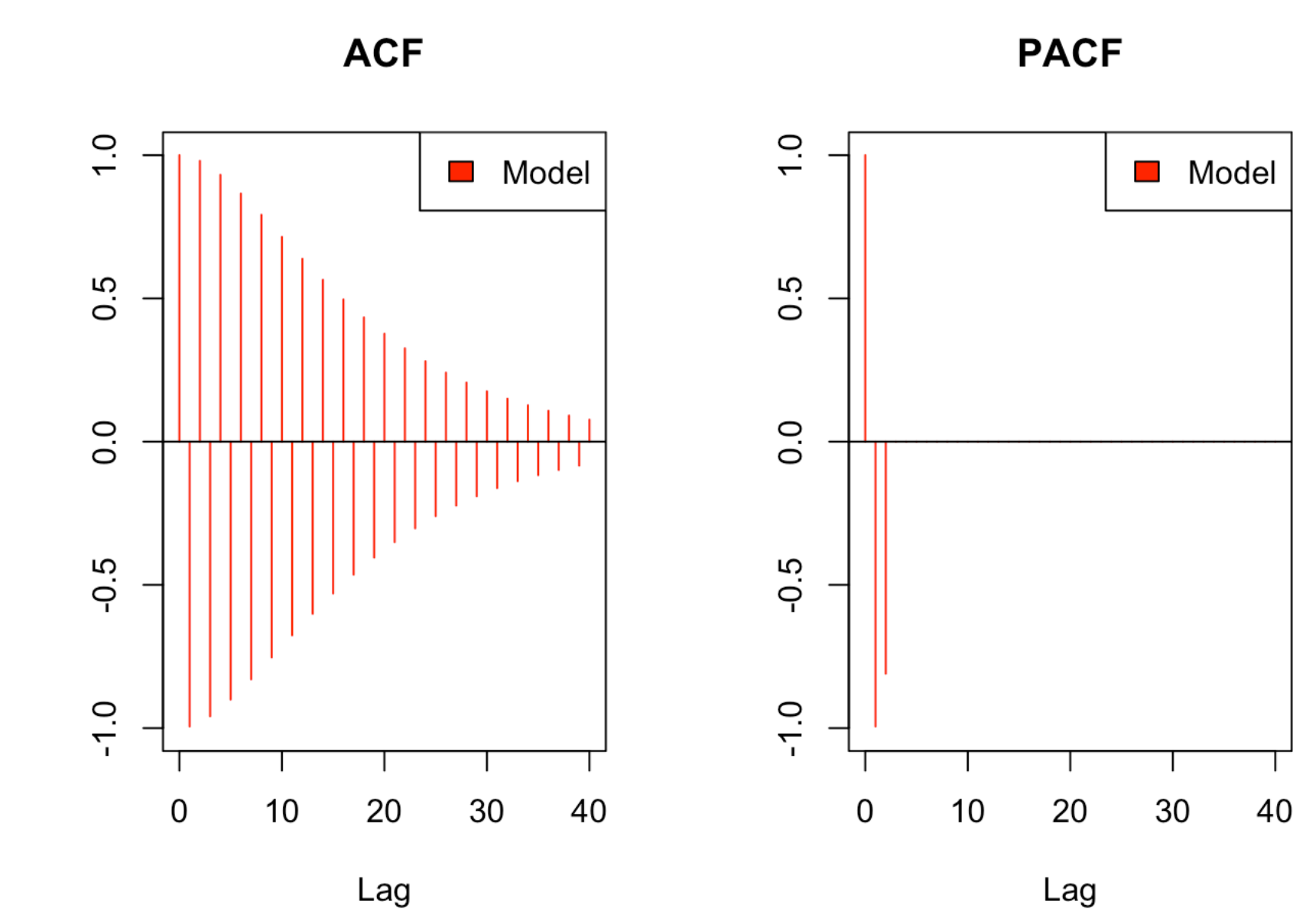
$$\therefore X_t = -1.8X_{t-1} - 0.81X_{t-2} + Z_t$$

$$\therefore P_n X_{n+1} = -1.8X_n - 0.81X_{n-1} \text{ for all } n \geq 2. \text{ So } \alpha(h) = 0 \text{ for all } h \geq 3, \alpha(2) = -0.81.$$

$$\text{By D-L algorithm, } \alpha(1) = \frac{1}{\gamma(0)}\gamma(1) = -\frac{171}{172}.$$

As convention  $\alpha(0) = 1$ .

```
model = list()
model$phi = c(-1.8, -0.81)
model$theta = c(0)
# model$sigma2 = c()
plota(model, h = 40)
```



3.3

(d)

Start with  $(1 - \phi_1z - \phi_2z^2)(\psi_0 + \psi_1z + \dots) = 1$ , where  $\phi_1 = -1.8, \phi_2 = -0.81$ .

Coefficient of  $z^0$ :  $\psi_0 = 1$ .

Coefficient of  $z^1$ :  $\psi_1 = \phi_1\psi_0 = \phi_1 = -1.8$ .

Coefficient of  $z^2$ :  $\psi_2 = \phi_1\psi_1 + \phi_2\psi_0 = -1.8\phi_1 + \phi_2 = 2.43$ .

Coefficient of  $z^3$ :  $\psi_3 = \phi_1\psi_2 + \phi_2\psi_1 = 2.43\phi_1 - 1.8\phi_2 = -2.916$ .

Coefficient of  $z^4$ :  $\psi_4 = \phi_1\psi_3 + \phi_2\psi_2 = -2.916\phi_1 + 2.43\phi_2 = 3.2805$ .

Coefficient of  $z^5$ :  $\psi_5 = \phi_1\psi_4 + \phi_2\psi_3 = 3.2805\phi_1 - 2.916\phi_2 = -3.54294$ .

3.4

$$X_t = Z_t + 0.8Z_{t-2} + 0.8^2Z_{t-4} + \dots = \sum_{i=0}^{\infty} 0.8^i Z_{t-2i}$$

$$\therefore \gamma(h) = \frac{0.8^2}{1-0.8^2} \text{ if } 2|h, \gamma(h) = 0 \text{ otherwise.}$$

$$\therefore \alpha(2) = 0.8, \alpha(h) = 0 \text{ otherwise.}$$

3.8

The stationary solution is  $X_t = \sum_{j=1}^{\infty} \phi^{-j} Z_{t+j}$  since  $|\phi| > 1$ . It follows that  $E(X_t) = 0, \gamma_X(h) = \frac{\sigma^2}{\phi^2 - 1} \phi^{-|h|}$ .

Now with  $W_t = X_t - \frac{1}{\phi} X_{t-1}, E(W_t) = 0$ .

$$\gamma_W(h) = \text{Cov}(X_{t+h} - \phi^{-1} X_{t+h-1}, X_t - \phi^{-1} X_{t-1}) = \gamma_X(h) - \phi^{-1} \gamma_X(h-1) - \phi^{-1} \gamma_X(h+1) + \phi^{-2} \gamma_X(h).$$

$$\therefore \gamma_W(0) = \frac{\sigma^2}{\phi^2}, \gamma_W(h) = 0 \text{ otherwise.}$$

Thus  $\{W_t\} \sim \text{WN}(0, \sigma_W^2)$  where  $\sigma_W^2 = \frac{\sigma^2}{\phi^2}$ .

Thus  $X_t$  is the unique stationary solution of  $X_t = \frac{1}{\phi} X_{t-1} + W_t$ .

3.9

(a)

$$\gamma(h) = \text{Cov}(\mu + Z_{t+h} + \theta_1 Z_{t+h-1} + \theta_{12} Z_{t+h-12}, \mu + Z_t + \theta_1 Z_{t-1} + \theta_{12} Z_{t-12}) = (1 + \theta_1^2 + \theta_{12}^2) \gamma_Z(h) + \theta_1 (\gamma_Z(h+1) + \gamma_Z(h-1)) + \theta_{12} (\gamma_Z(h+12) + \gamma_Z(h-12)) + \theta_1 \theta_{12} (\gamma_Z(h+11) + \gamma_Z(h-11))$$

$$\therefore \gamma(0) = (1 + \theta_1^2 + \theta_{12}^2) \sigma^2, \gamma(-1) = \gamma(1) = \theta_1 \sigma^2, \gamma(-12) = \gamma(12) = \theta_{12} \sigma^2, \gamma(-11) = \gamma(11) = \theta_1 \theta_{12} \sigma^2, \gamma(h) = 0 \text{ otherwise.}$$

(b)

$$\nabla \nabla_{12} X_t = \nabla (X_t - X_{t-12}) = X_t - X_{t-1} - X_{t-12} + X_{t-13}$$

```
library(itsmr)
nab = c()
for (x in 14:72){
  nab[x-13] = deaths[x]-deaths[x-1]-deaths[x-12]*deaths[x-13]
}
mean_s = mean(nab)
acvf_s = acvf(nab, 20)
paste0("The sample mean is ", mean_s, ',')
```

```
## [1] "The sample mean is 28.8305084745763."
```

```
paste0("The sample acvf with lag ", c(0:20), ' is ', acvf_s, ',')
```

```
## [1] "The sample acvf with lag 0 is 152669.632289572."
## [2] "The sample acvf with lag 1 is -54326.5279215499."
## [3] "The sample acvf with lag 2 is -15071.6823871964."
## [4] "The sample acvf with lag 3 is 14584.5679159018."
## [5] "The sample acvf with lag 4 is -17177.6942774091."
## [6] "The sample acvf with lag 5 is 6340.251282171."
## [7] "The sample acvf with lag 6 is 17420.9080237025."
## [8] "The sample acvf with lag 7 is -31164.4601736302."
## [9] "The sample acvf with lag 8 is -1087.51323163517."
## [10] "The sample acvf with lag 9 is 15277.1754512389."
## [11] "The sample acvf with lag 10 is -12434.6704823765."
## [12] "The sample acvf with lag 11 is 29801.9685021351."
## [13] "The sample acvf with lag 12 is -50866.8975406444."
## [14] "The sample acvf with lag 13 is 13767.9425355075."
## [15] "The sample acvf with lag 14 is 17757.6610948539."
## [16] "The sample acvf with lag 15 is -6199.56720015192."
## [17] "The sample acvf with lag 16 is -9656.4134210411."
## [18] "The sample acvf with lag 17 is 27981.3878731516."
## [19] "The sample acvf with lag 18 is -29455.8134181197."
## [20] "The sample acvf with lag 19 is 3692.78017713593."
## [21] "The sample acvf with lag 20 is 7569.42950350328."
```

(c)

Now we have

$$\theta_1 \sigma^2 = \hat{\gamma}(1), \theta_1 \theta_{12} \sigma^2 = \hat{\gamma}(11), \theta_{12} \sigma^2 = \hat{\gamma}(12)$$

```
theta_1 = acvf_s[12]/acvf_s[13]
theta_12 = acvf_s[12]/acvf_s[2]
sigma_2 = acvf_s[13]*acvf_s[2]/acvf_s[12]
paste0("theta_1 is ", theta_1, ',')
```

```
## [1] "theta_1 is -0.585881387366357."
```

```
paste0("theta_12 is ", theta_12, ',')
```

```
## [1] "theta_12 is -0.54857119794533."
```

```
paste0("sigma^2 is ", sigma_2, ',')
```

```
## [1] "sigma^2 is 92726.154291669."
```

So the model should be  $Y_t = 28.8305084745763 + Z_t - 0.585881387366357Z_{t-1} - 0.54857119794533Z_{t-12}$ , where  $\{Z_t\} \sim \text{WN}(0, 92726.154291669)$ .