



Analisis Deret Waktu

#11 Meeting

Model AR, MA, ARMA

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Model ARMA (p,q)

Ide Dasar

Model ARMA merupakan kombinasi dari $AR(p)$ dan $MA(q)$, sehingga biasa dituliskan $ARMA(p, q)$

$$Z_t = \phi_1 Z_{t-1} + \phi_2 Z_{t-2} + \dots + \phi_p Z_{t-p} + a_t - \theta_1 a_{t-1} - \theta_2 a_{t-2} - \dots - \theta_q a_{t-q}$$

Proses ARMA(1,1)

Model ARMA(1,1)

- Bila $p = 1$ dan $q = 1$ maka

$$Z_t = \phi Z_{t-1} + a_t - \theta a_{t-1}$$

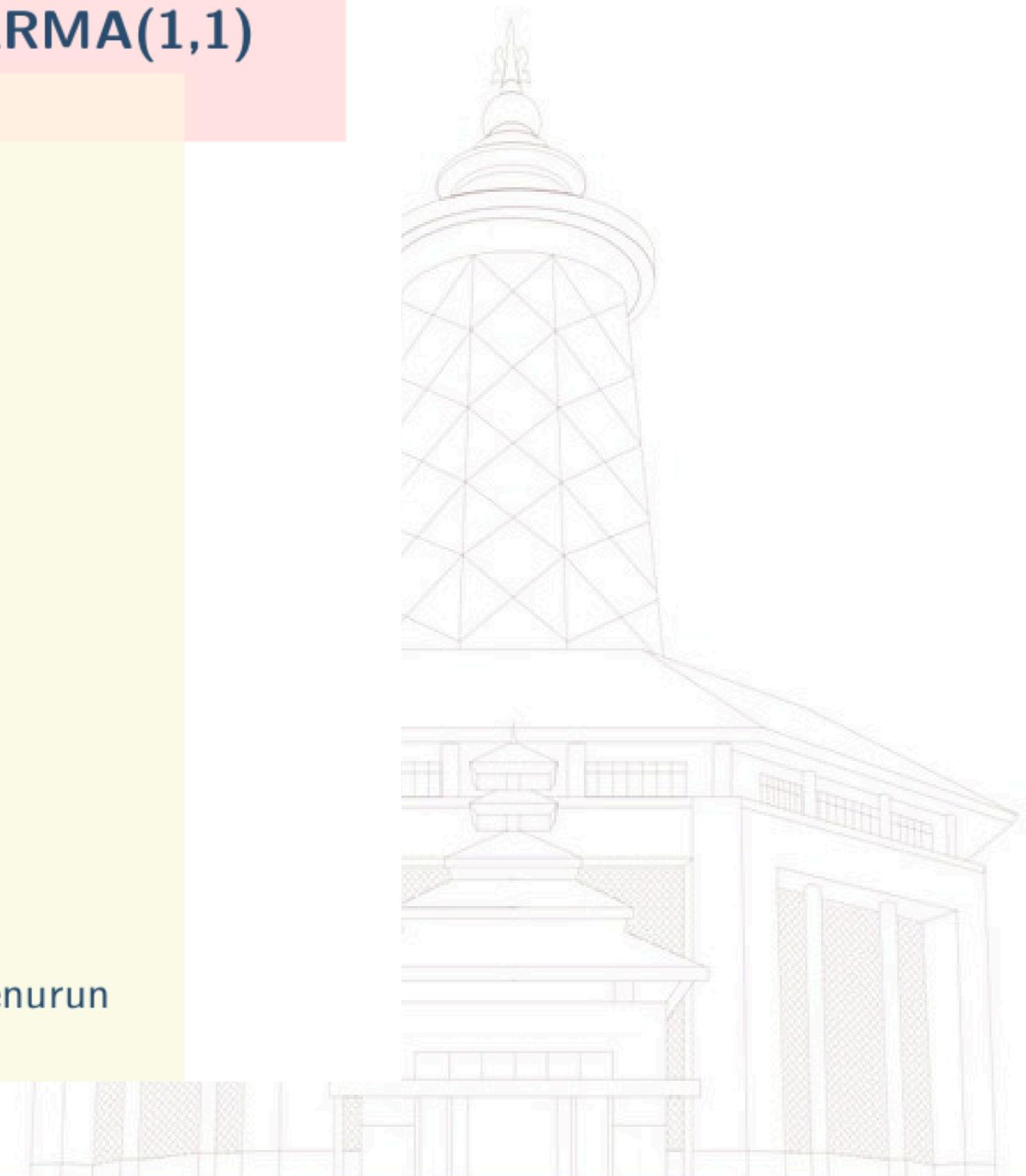
- Struktur koragam

$$\gamma_k = \frac{(1 - \theta\phi)(\phi - \theta)}{1 - \phi^2} \phi^{k-1} \sigma_a^2 \text{ untuk } k \geq 1$$

- Struktur korelasi

$$\rho_k = \frac{(1 - \theta\phi)(\phi - \theta)}{1 - 2\theta\phi + \theta^2} \phi^{k-1} \text{ untuk } k \geq 1$$

- Dengan demikian, autokorelasi Model ARMA juga menurun eksponensial





Proses AR(1)

Diketahui sebuah proses deret waktu stasioner $Z(t)$ mengikuti model AR(1) sebagai berikut:

$$Z(t) = 0.8 Z(t-1) + a(t)$$

Dimana $a(t)$ adalah white noise (WN) dengan rata-rata 0 dan varians 3.6.

Tugas:

Hitunglah:

1. Ragam (Varian) proses.
2. Koragam (Autokovarians).
3. Korelasi (ACF).
4. Korelasi Parsial (PACF).





Proses AR(1)

Kita memiliki parameter:

- $\phi_1 = 0.8$
- $\sigma_a^2 = 3.6$

1. Perhitungan Ragam (γ_0)

$$\gamma_0 = \frac{\sigma_a^2}{1 - \phi_1^2}$$

$$\gamma_0 = \frac{3.6}{1 - (0.8)^2}$$

$$\gamma_0 = \frac{3.6}{1 - 0.64}$$

$$\gamma_0 = \frac{3.6}{0.36} \quad \gamma_0 = 10$$

2. Perhitungan Koragam (γ_k)

gunakan rumus rekursif Yule-Walker: $\gamma_k = \phi_1 \gamma_{k-1}$

Untuk γ_1 (Lag 1)

$$\gamma_1 = \phi_1 \gamma_0$$

$$\gamma_1 = 0.8 \times 10$$

$$\gamma_1 = 8$$

Untuk γ_2 (Lag 2)

$$\gamma_2 = \phi_1 \gamma_1$$

$$\gamma_2 = 0.8 \times 8$$

$$\gamma_2 = 6.4$$

3. Perhitungan Korelasi / ACF (ρ_k)

gunakan rumus $\rho_k = \phi_1^k$

atau bisa juga $\rho_k = \gamma_k / \gamma_0$

Untuk ρ_1 (Lag 1)

$$\rho_1 = \phi_1^1 = 0.8 \text{ atau } \gamma_1 / \gamma_0 = 8/10 = 0.8$$

Untuk ρ_2 (Lag 2)

$$\rho_2 = \phi_1^2 = (0.8)^2 = 0.64$$

atau

$$\gamma_2 / \gamma_0 = 6.4/10 = 0.64$$

Untuk ρ_3 (Lag 3)

$$\rho_3 = \phi_1^3 = (0.8)^3 = 0.512$$

$$\rho_k = \{0.8, 0.64, 0.512, \dots\}$$

Nilainya meluruh (tails off)



Proses AR(1)

4. Perhitungan Korelasi Parsial / PACF (ϕ_{kk})

$$\phi_{kk} = \begin{cases} \phi_1 & \text{jika } k = 1 \\ 0 & \text{jika } k \geq 2 \end{cases}$$

Untuk ϕ_{11} (Lag 1)

$$\phi_{11} = \phi_1 = 0.8$$

Untuk ϕ_{22} (Lag 2)

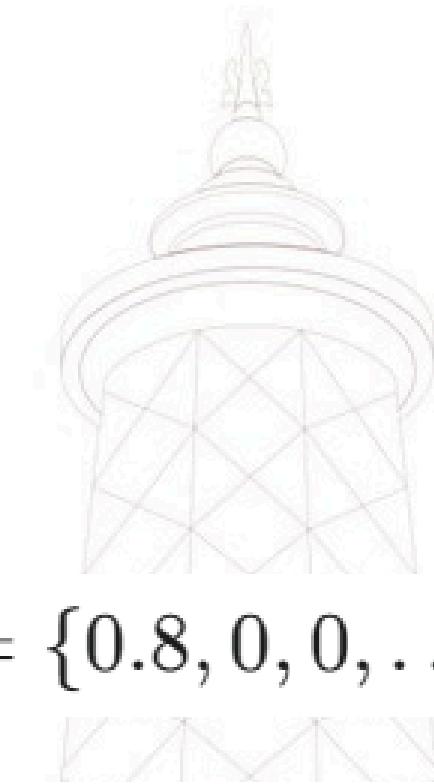
$$\phi_{22} = 0$$

Untuk ϕ_{33} (Lag 3)

$$\phi_{33} = 0$$

$$\phi_{kk} = \{0.8, 0, 0, \dots\}$$

Nilainya **terputus (cuts off)** setelah lag 1





Proses AR(1) dalam R

```
# -----  
# Verifikasi Perhitungan Teoretis AR(1)  
# Model:  $Z_t = 0.8 * Z_{t-1} + a_t$   
# Varians error:  $\sigma_a^2 = 3.6$   
# -----
```

```
# --- 1. Definisikan Parameter ---  
phi <- 0.8  
sigma_a_sq <- 3.6  
lag_maks <- 3 # Kita ingin hitung sampai lag 3
```

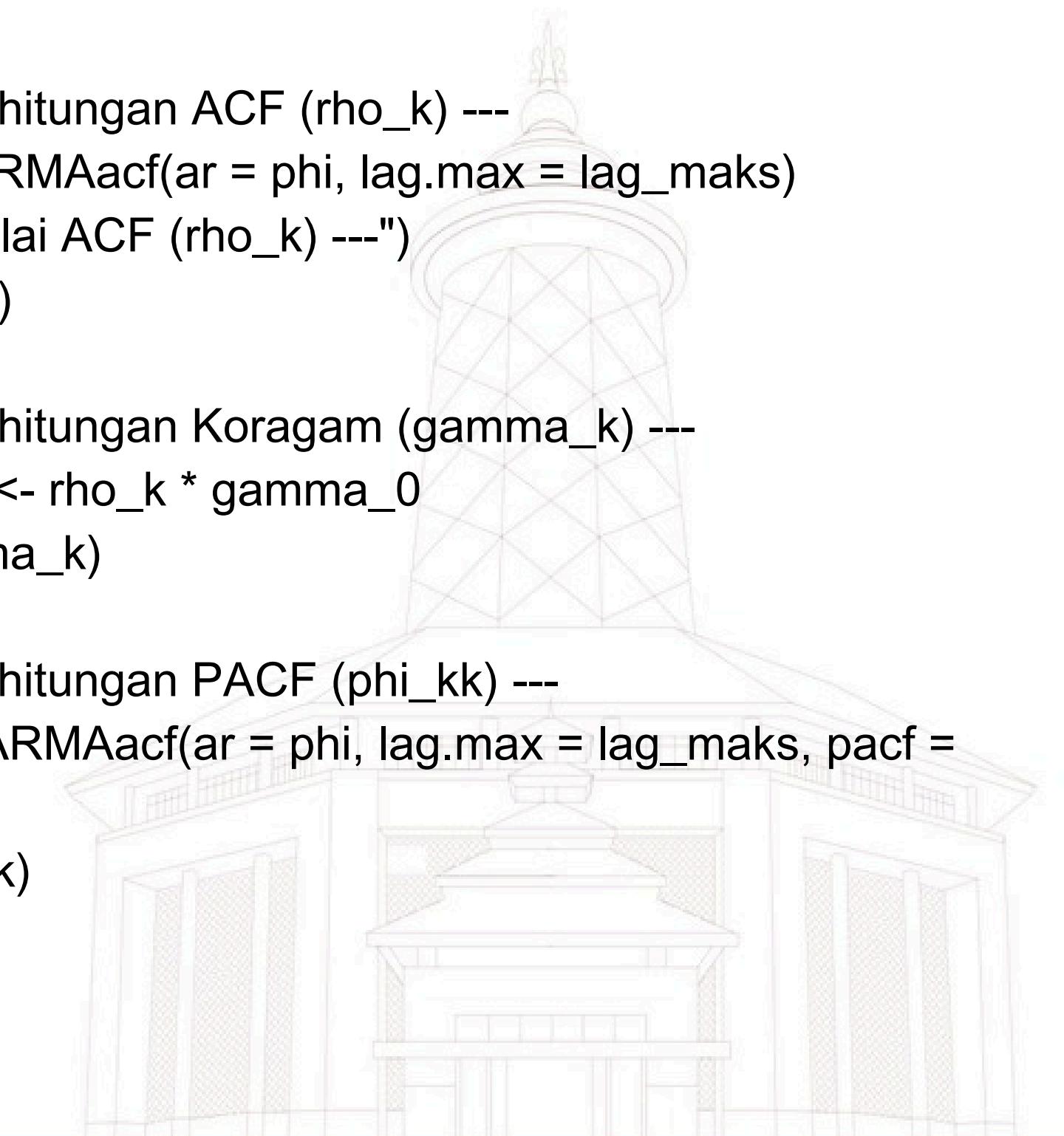
```
# --- 2. Perhitungan Ragam ( $\gamma_0$ ) ---  
# Rumus:  $\gamma_0 = \sigma_a^2 / (1 - \phi^2)$ 
```

```
gamma_0 <- sigma_a_sq / (1 - phi^2)  
print(paste("Ragam ( $\gamma_0$ ):", gamma_0))
```

```
# --- 3. Perhitungan ACF ( $\rho_k$ ) ---  
rho_k <- ARMAacf(ar = phi, lag.max = lag_maks)  
print("--- Nilai ACF ( $\rho_k$ ) ---")  
print(rho_k)
```

```
# --- 4. Perhitungan Koragam ( $\gamma_k$ ) ---  
gamma_k <- rho_k * gamma_0  
print(gamma_k)
```

```
# --- 5. Perhitungan PACF ( $\phi_{kk}$ ) ---  
phi_kk <- ARMAacf(ar = phi, lag.max = lag_maks, pacf =  
TRUE)  
print(phi_kk)
```





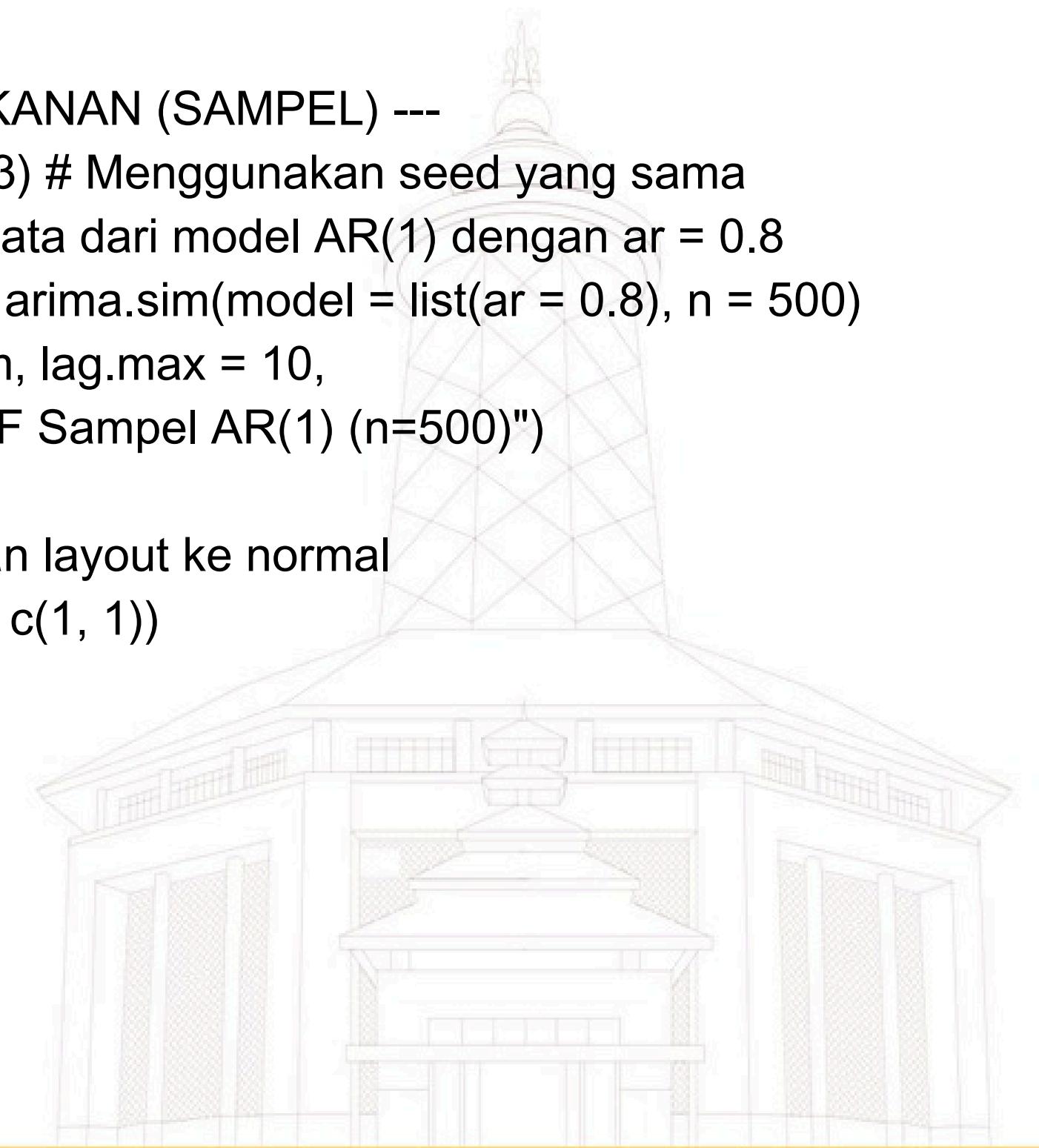
Plot ACF AR(1) dalam R

```
dev.new()
par(mfrow = c(1, 2))

# --- PLOT KIRI (TEORETIS) ---
# ACF akan meluruh secara eksponensial ( $0.8^k$ )
acf_teori <- ARMAacf(ar = 0.8, lag.max = 10, pacf =
FALSE)
plot(acf_teori, type = "h",
      main = "ACF Teoretis AR(1) (Meluruh)",
      xlab = "Lag (termasuk Lag 0)", ylab = "ACF",
      ylim = c(-1, 1),
      lwd = 3, col = "red")
abline(h = 0)
```

```
# --- PLOT KANAN (SAMPEL) ---
set.seed(123) # Menggunakan seed yang sama
# Simulasi data dari model AR(1) dengan ar = 0.8
data_sim <- arima.sim(model = list(ar = 0.8), n = 500)
acf(data_sim, lag.max = 10,
     main = "ACF Sampel AR(1) (n=500)")

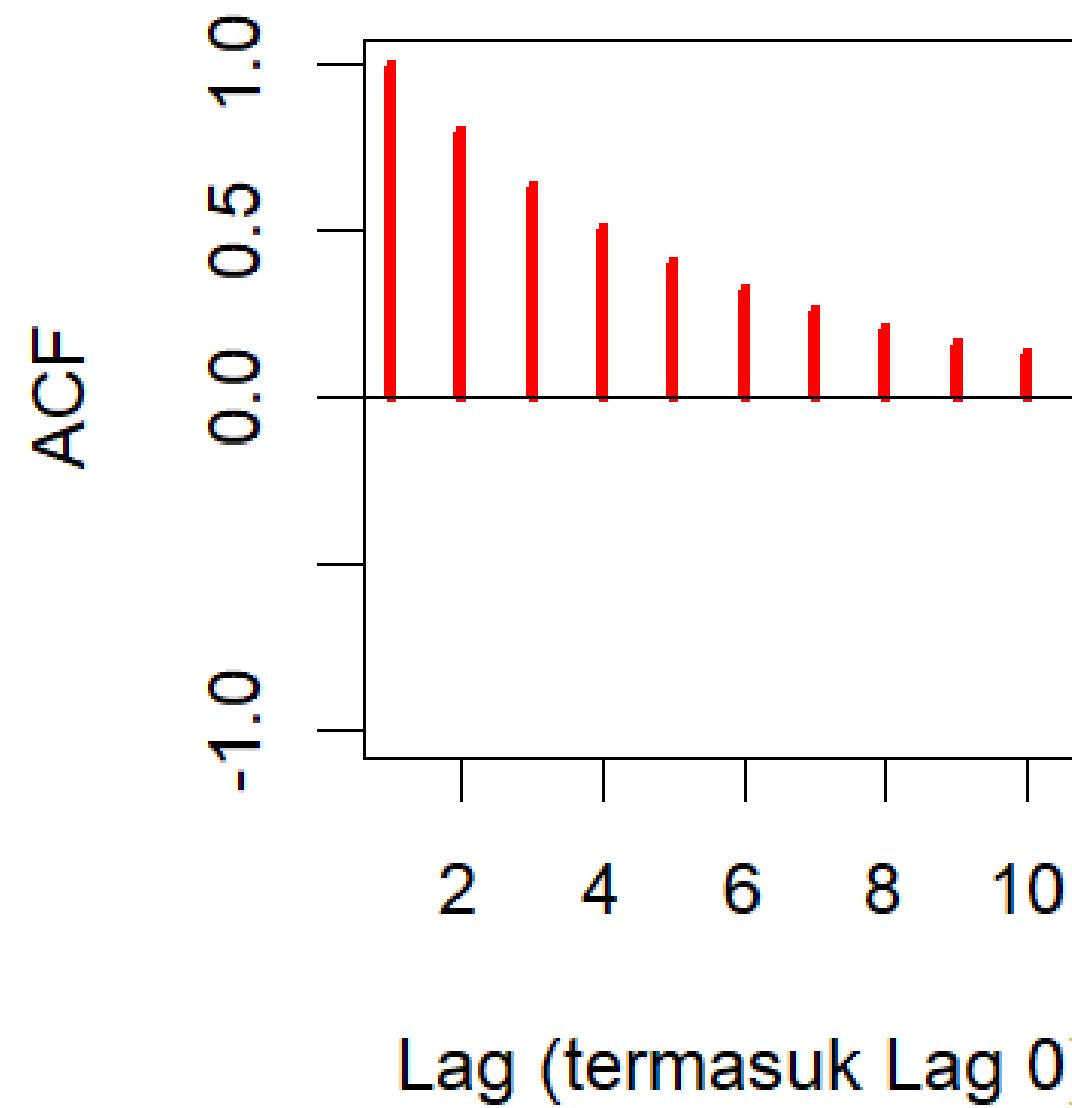
# Kembalikan layout ke normal
par(mfrow = c(1, 1))
```



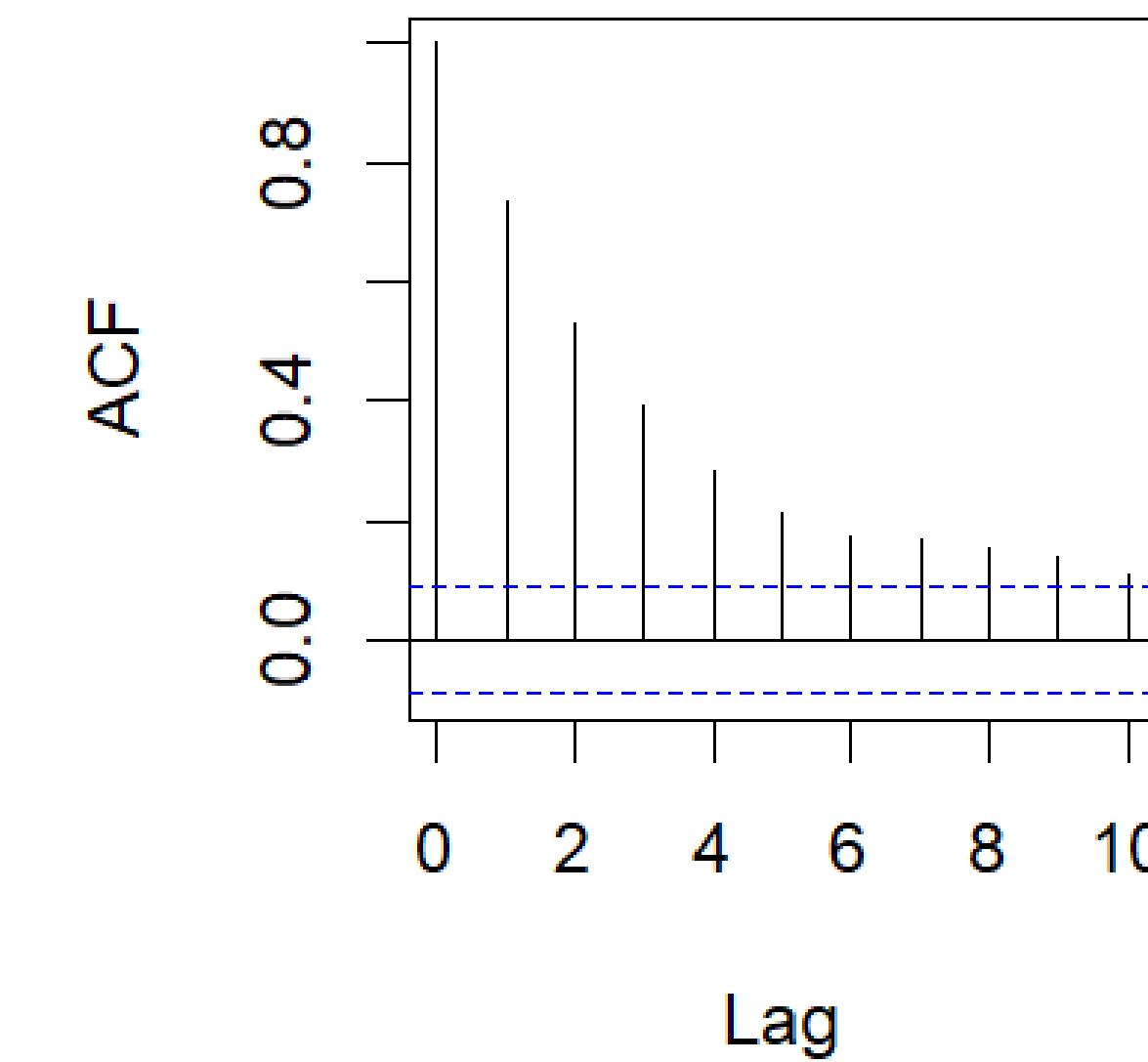


Plot ACF AR(1) dalam R

ACF Teoretis AR(1) (Meluruhan)



ACF Sampel AR(1) (n=500)

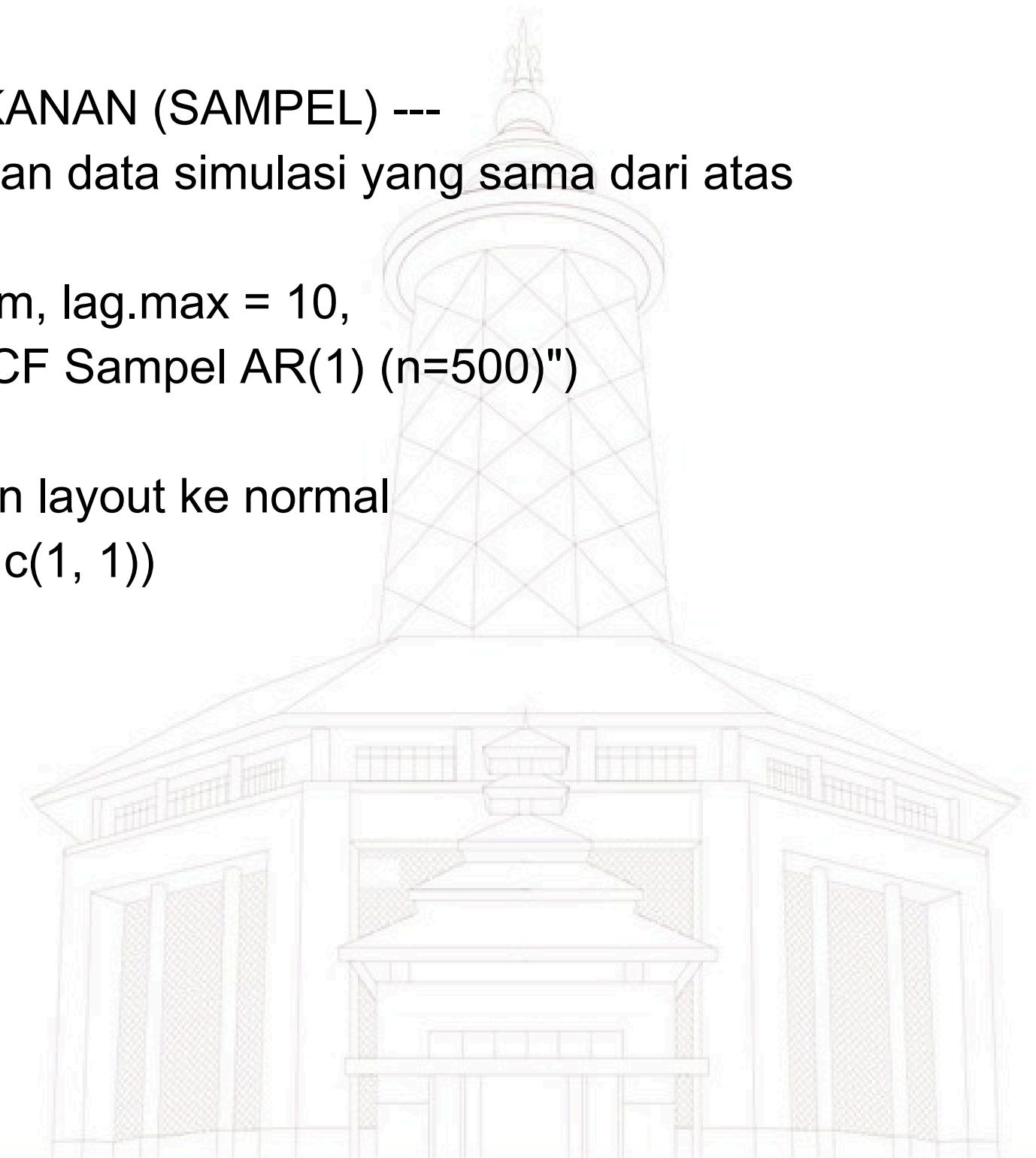




Plot PACF AR(1) dalam R

```
dev.new()  
par(mfrow = c(1, 2))  
  
# --- PLOT KIRI (TEORETIS) ---  
# PACF akan terputus setelah Lag 1  
pacf_teori <- ARMAacf(ar = 0.8, lag.max = 10, pacf =  
TRUE)  
plot(pacf_teori, type = "h",  
     main = "PACF Teoretis AR(1) (Terputus)",  
     xlab = "Lag", ylab = "PACF",  
     ylim = c(-1, 1),  
     lwd = 3, col = "red")  
abline(h = 0)
```

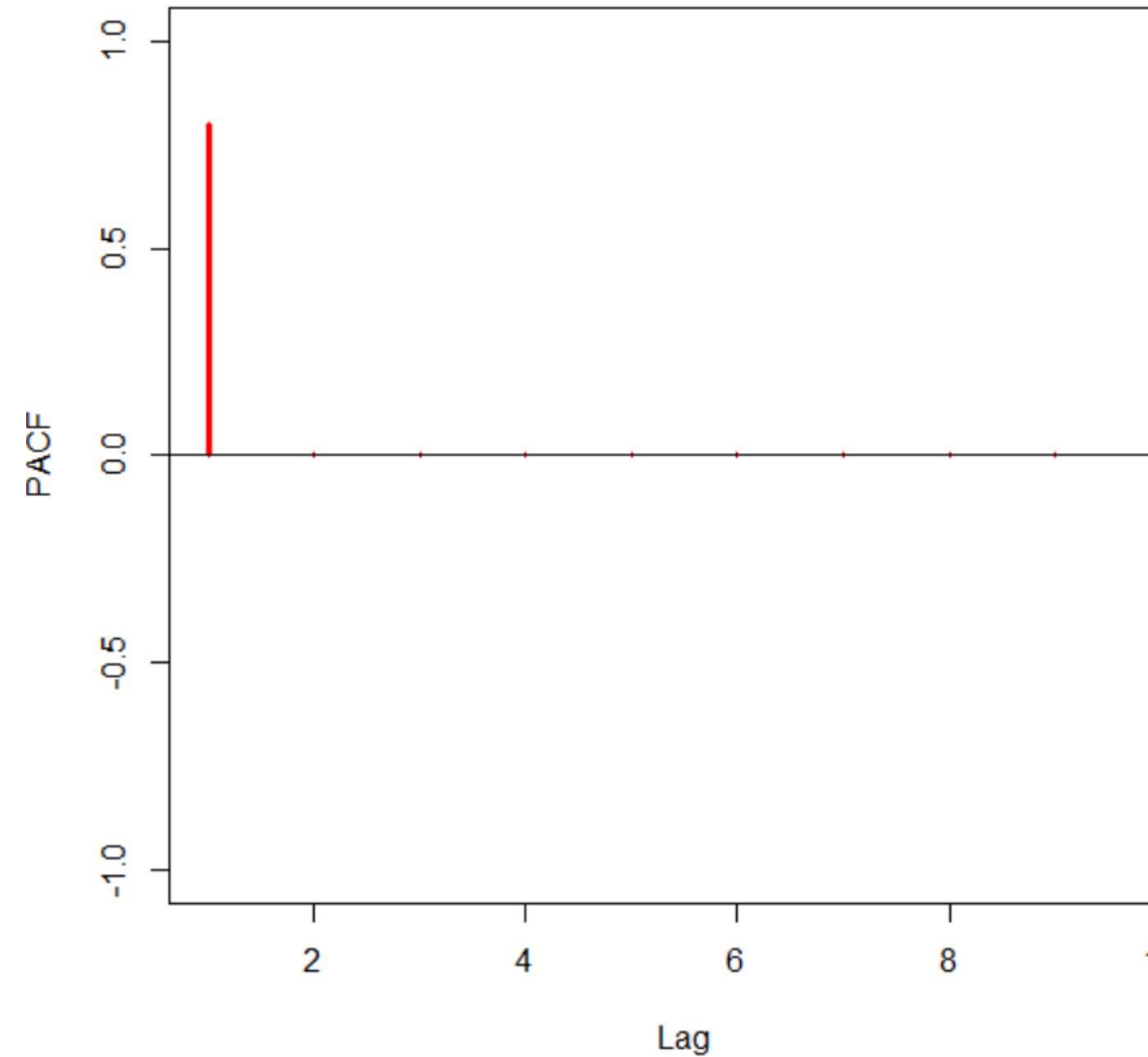
```
# --- PLOT KANAN (SAMPEL) ---  
# Kita gunakan data simulasi yang sama dari atas  
(data_sim)  
pacf(data_sim, lag.max = 10,  
      main = "PACF Sampel AR(1) (n=500)")  
  
# Kembalikan layout ke normal  
par(mfrow = c(1, 1))
```



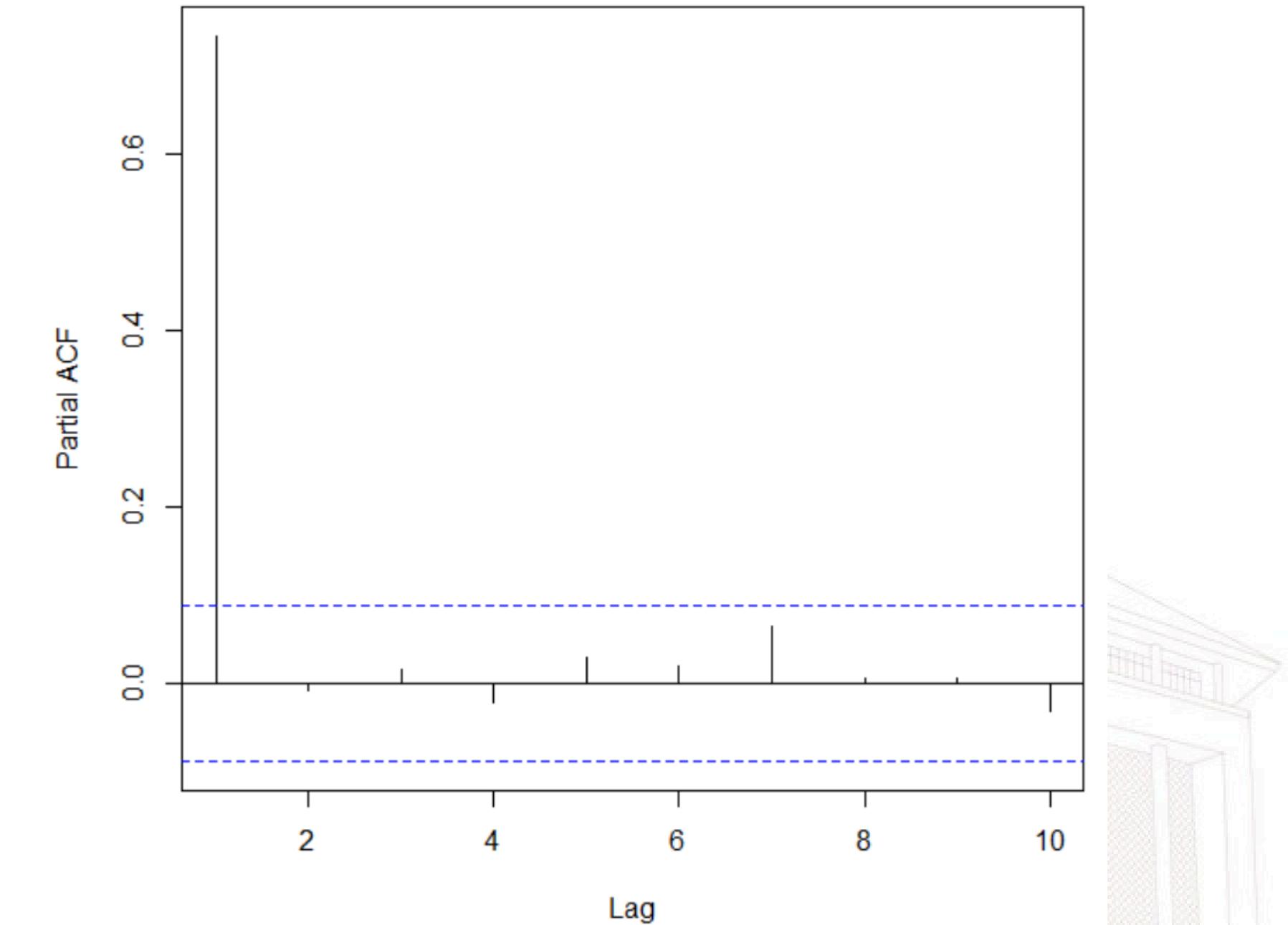


Plot PACF AR(1) dalam R

PACF Teoretis AR(1) (Terputus)



PACF Sampel AR(1) (n=500)





Proses MA(1)

Diketahui sebuah proses deret waktu stasioner $Z(t)$ mengikuti model MA(1) sebagai berikut:

$$Z(t) = a(t) - 0.5 a(t-1)$$

Dimana $a(t)$ adalah white noise (WN) dengan rata-rata 0 dan varians 8.

Tugas:

Hitunglah:

1. Ragam (Varian) proses.
2. Koragam (Autokovarians).
3. Korelasi (ACF).
4. Korelasi Parsial (PACF).





Proses MA(1)

Kita memiliki parameter:

- $\theta_1 = 0.5$
- $\sigma_a^2 = 8$

1. Perhitungan Ragam (γ_0)

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

$$\gamma_0 = 8 \times (1 + (0.5)^2)$$

$$\gamma_0 = 8 \times (1 + 0.25)$$

$$\gamma_0 = 8 \times (1.25)$$

$$\gamma_0 = 10$$

2. Perhitungan Koragam (γ_k)

Untuk γ_1 (Lag 1)

$$\gamma_1 = -\theta_1 \sigma_a^2$$

$$\gamma_1 = -0.5 \times 8$$

$$\gamma_1 = -4$$

Untuk γ_2 (Lag 2)

koragamnya nol untuk lag $k \geq 2$

$$\gamma_2 = 0$$

3. Perhitungan Korelasi / ACF (ρ_k)

Untuk ρ_1 (Lag 1)

$$\rho_1 = \frac{\gamma_1}{\gamma_0} = \frac{-4}{10} \quad \rho_1 = -0.4$$

atau

$$\frac{-\theta_1}{1+\theta_1^2} = \frac{-0.5}{1.25} = -0.4$$

Untuk ρ_2 (Lag 2)

$$\rho_2 = 0$$

Untuk ρ_3 (Lag 3)

$$\rho_3 = 0$$

$$\rho_k = \{-0.4, 0, 0, \dots\}$$

Nilainya **terputus (cuts off)** setelah lag 1



Proses MA(1)

4. Perhitungan Korelasi Parsial / PACF (ϕ_{kk})

$$\phi_{kk} = \frac{-\theta^k(1 - \theta^2)}{1 - \theta^{2(k+1)}}$$

Untuk ϕ_{11} (Lag 1)

$$\phi_{11} = \rho_1 = -0.4$$

Untuk ϕ_{22} (Lag 2)

$$\phi_{22} = \frac{-(0.5)^2(1 - 0.5^2)}{1 - 0.5^{2(2+1)}} = \frac{-(0.25)(1 - 0.25)}{1 - 0.5^6}$$

$$\phi_{22} = \frac{-0.25 \times 0.75}{1 - 0.015625} = \frac{-0.1875}{0.984375}$$

$$\phi_{22} \approx -0.1905$$

Untuk ϕ_{33} (Lag 3)

$$\phi_{33} = \frac{-(0.5)^3(1 - 0.5^2)}{1 - 0.5^{2(3+1)}} = \frac{-(0.125)(0.75)}{1 - 0.5^8}$$

$$\phi_{33} = \frac{-0.09375}{1 - 0.00390625}$$

$$\phi_{33} \approx -0.0941$$

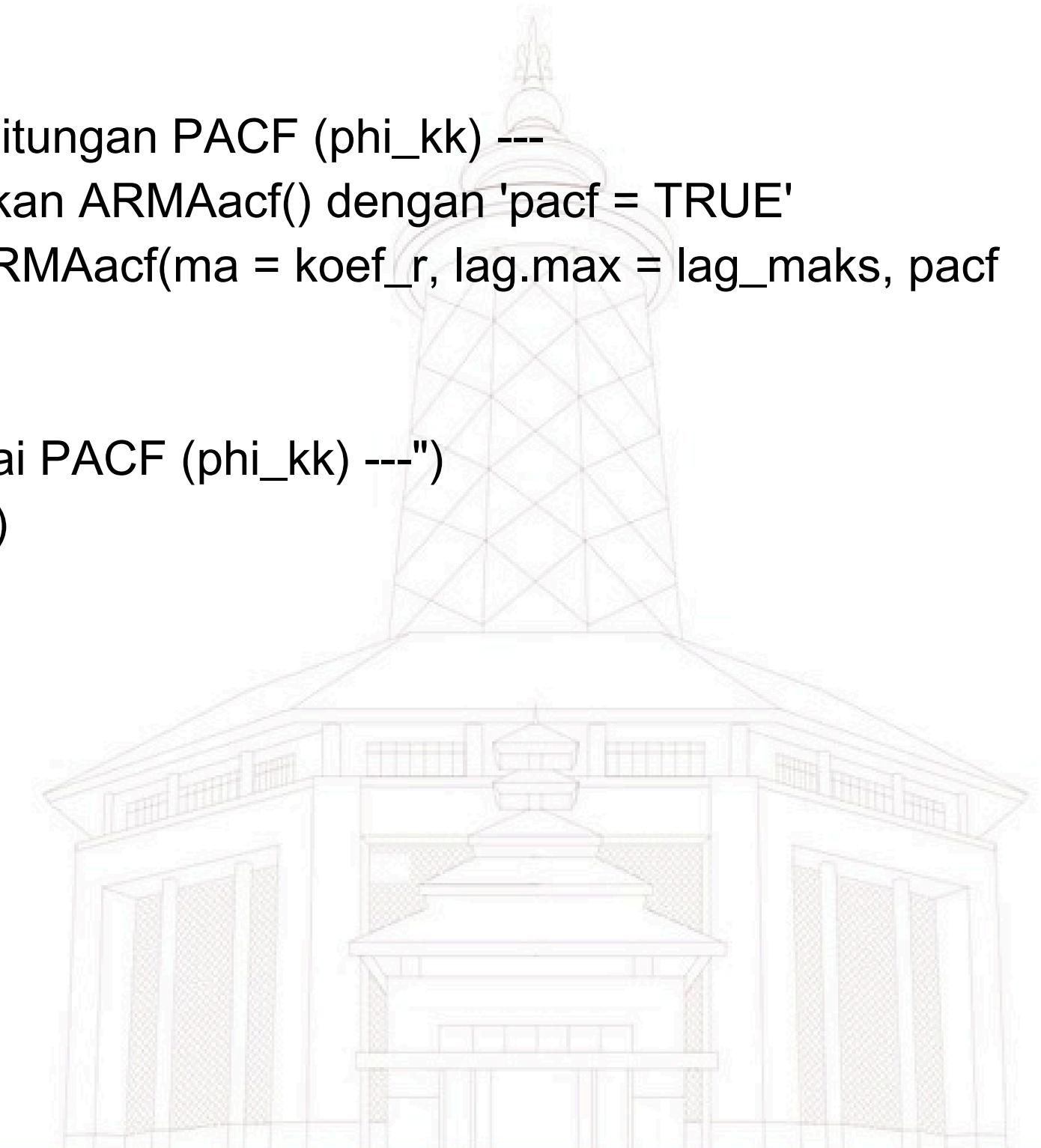
$$\phi_{kk} = \{-0.4, -0.19, -0.09, \dots\}$$

Nilainya meluruh (tails off)



Proses MA(1) dalam R

```
# -----  
# Verifikasi Perhitungan Teoretis MA(1)  
# Model:  $Z_t = a_t - 0.5 * a_{t-1}$   
# Konvensi R: ma = -0.5  
# -----  
  
# --- 1. Definisikan Parameter ---  
theta_bj <- 0.5  
koef_r <- -theta_bj # Koefisien untuk R  
lag_maks <- 3  
  
# --- 2. Perhitungan ACF (rho_k) ---  
# Kita gunakan fungsi ARMAacf()  
rho_k <- ARMAacf(ma = koef_r, lag.max = lag_maks)  
  
print(rho_k)  
  
# --- 3. Perhitungan PACF (phi_kk) ---  
# Kita gunakan ARMAacf() dengan 'pacf = TRUE'  
phi_kk <- ARMAacf(ma = koef_r, lag.max = lag_maks, pacf  
= TRUE)  
  
print("--- Nilai PACF (phi_kk) ---")  
print(phi_kk)
```





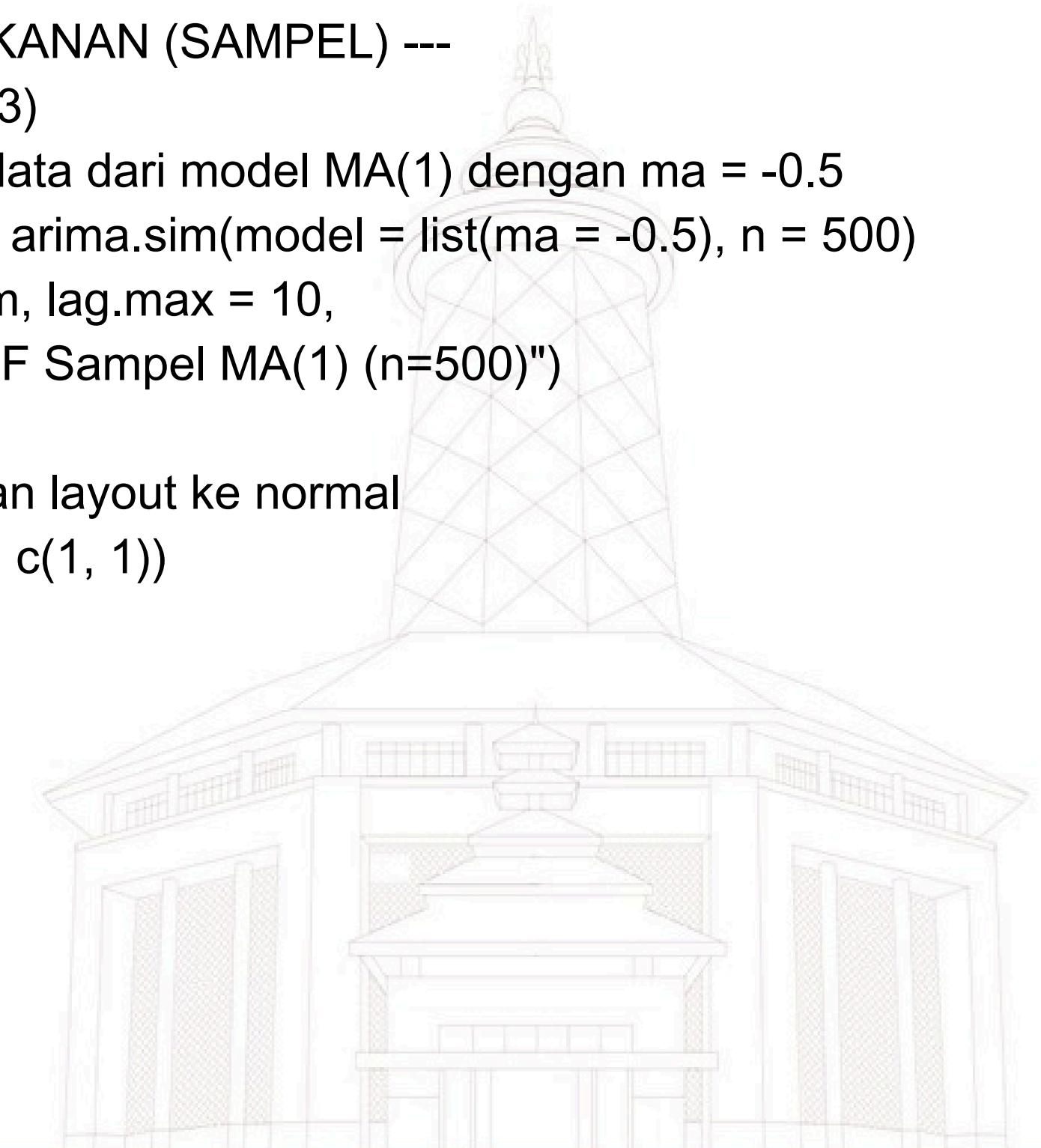
Plot ACF MA(1) dalam R

```
dev.new()
# Atur layout jadi 1 baris, 2 kolom
par(mfrow = c(1, 2))

# --- PLOT KIRI (TEORETIS) ---
# ACF akan terputus setelah Lag 1
acf_teori <- ARMAacf(ma = -0.5, lag.max = 10, pacf =
FALSE)
plot(acf_teori, type = "h",
     main = "ACF Teoretis MA(1) (Terputus)",
     xlab = "Lag (termasuk Lag 0)", ylab = "ACF",
     ylim = c(-1, 1),
     lwd = 3, col = "red")
abline(h = 0)
```

```
# --- PLOT KANAN (SAMPEL) ---
set.seed(123)
# Simulasi data dari model MA(1) dengan ma = -0.5
data_sim <- arima.sim(model = list(ma = -0.5), n = 500)
acf(data_sim, lag.max = 10,
     main = "ACF Sampel MA(1) (n=500)")

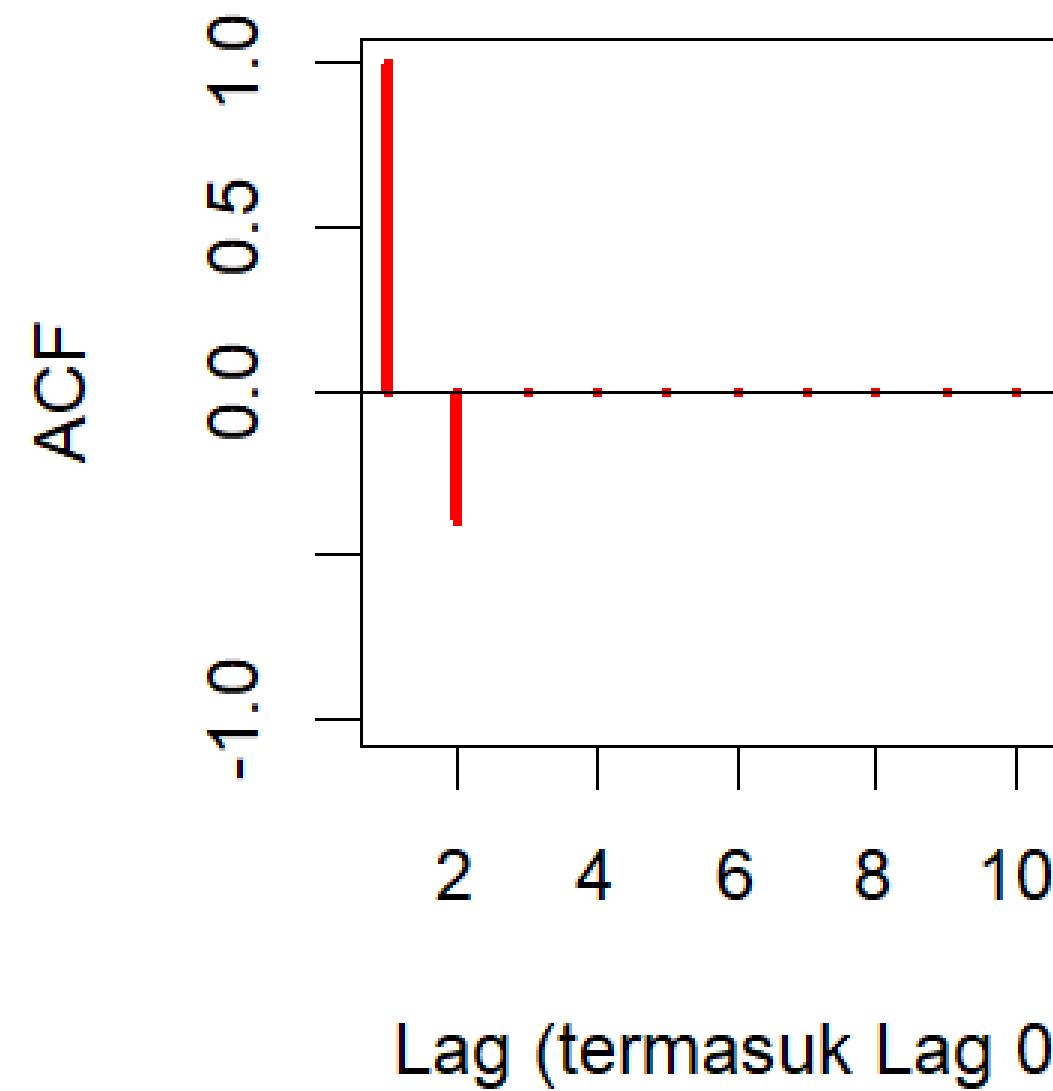
# Kembalikan layout ke normal
par(mfrow = c(1, 1))
```



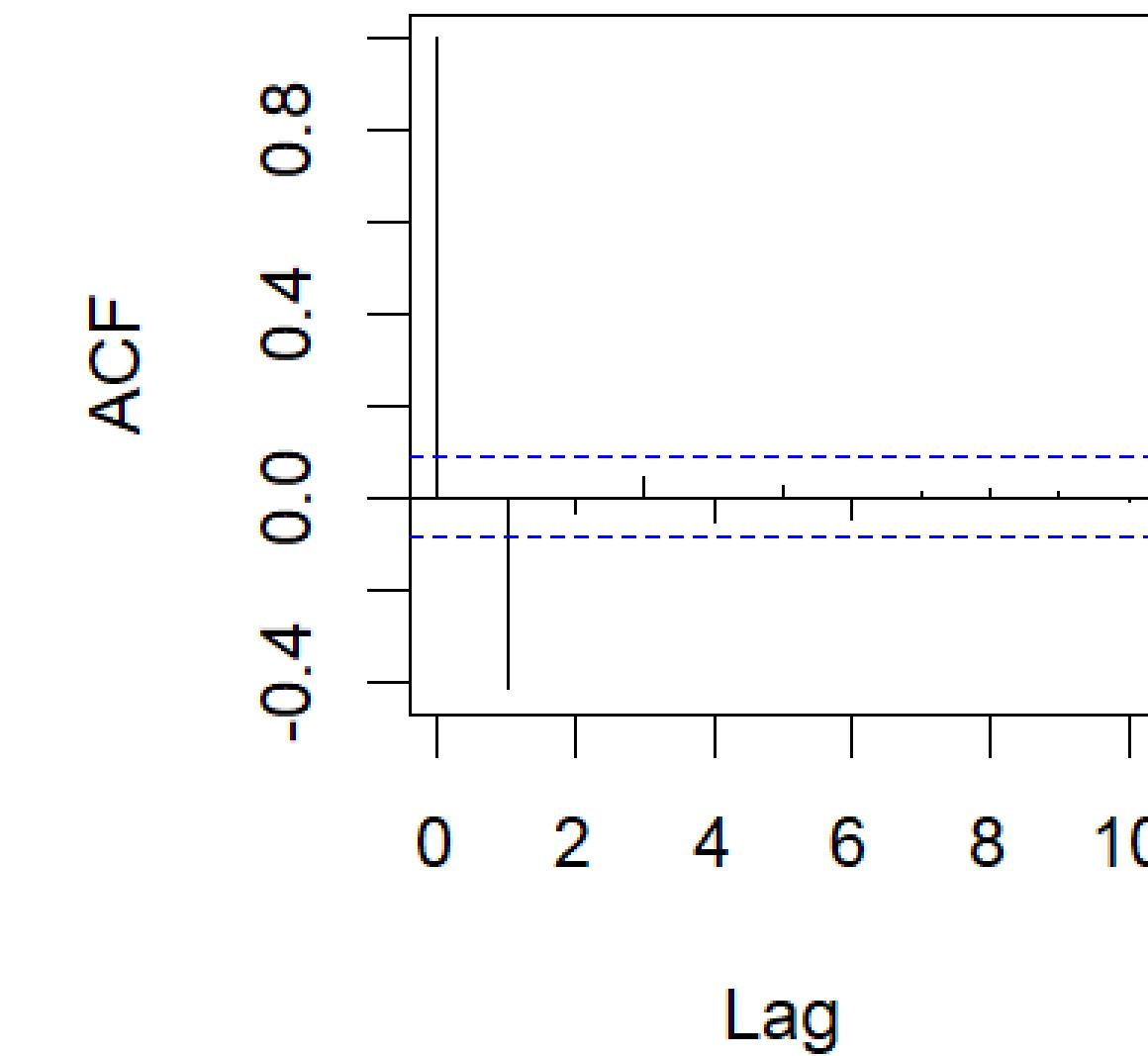


Plot ACF MA(1) dalam R

ACF Teoretis MA(1) (Terputus)



ACF Sampel MA(1) (n=500)

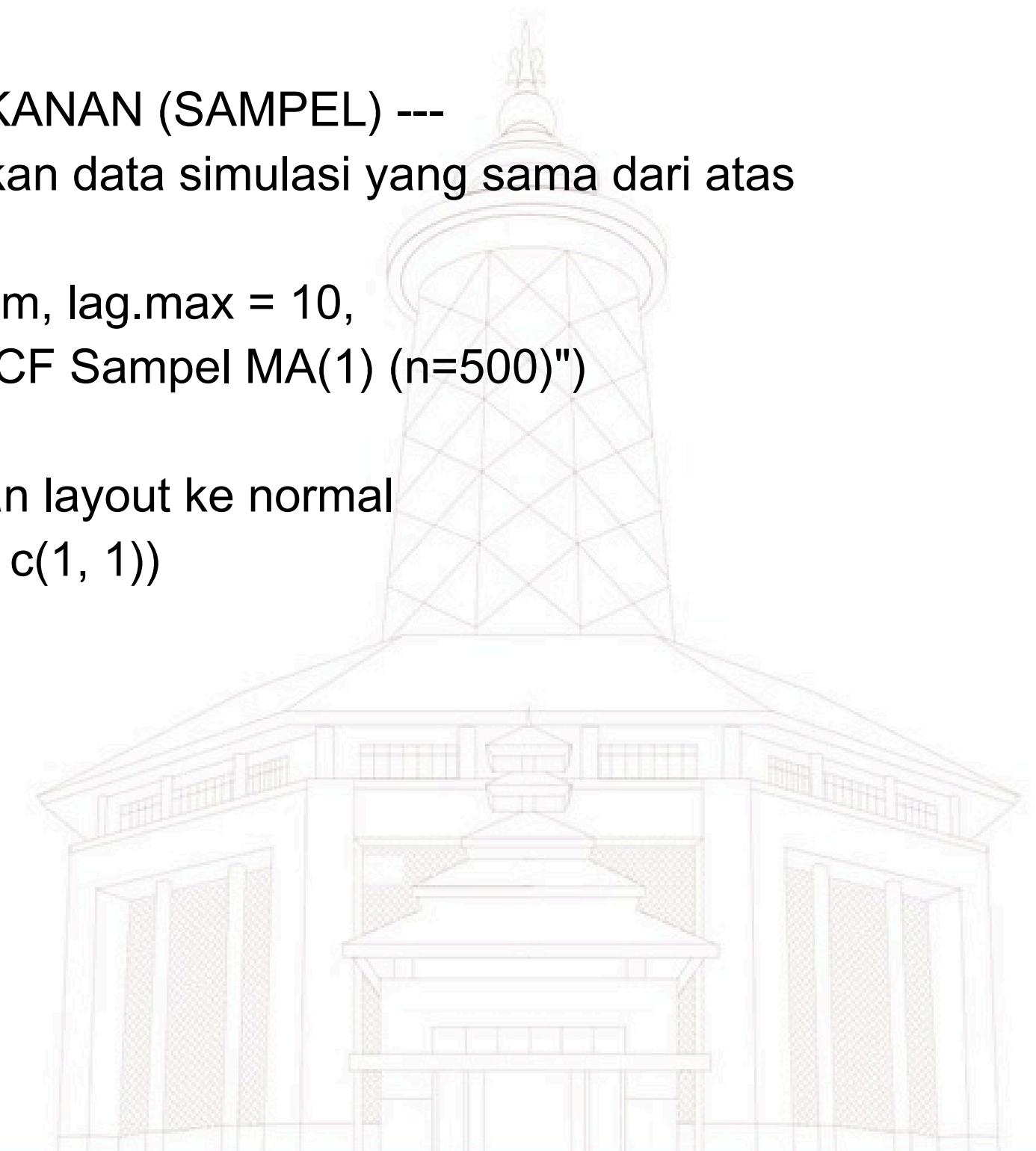




Plot PACF MA(1) dalam R

```
dev.new()  
# Atur layout jadi 1 baris, 2 kolom  
par(mfrow = c(1, 2))  
  
# --- PLOT KIRI (TEORETIS) ---  
# PACF akan meluruh  
pacf_teori <- ARMAacf(ma = -0.5, lag.max = 10, pacf =  
TRUE)  
plot(pacf_teori, type = "h",  
     main = "PACF Teoretis MA(1) (Meluruh)",  
     xlab = "Lag", ylab = "PACF",  
     ylim = c(-1, 1),  
     lwd = 3, col = "red")  
abline(h = 0)
```

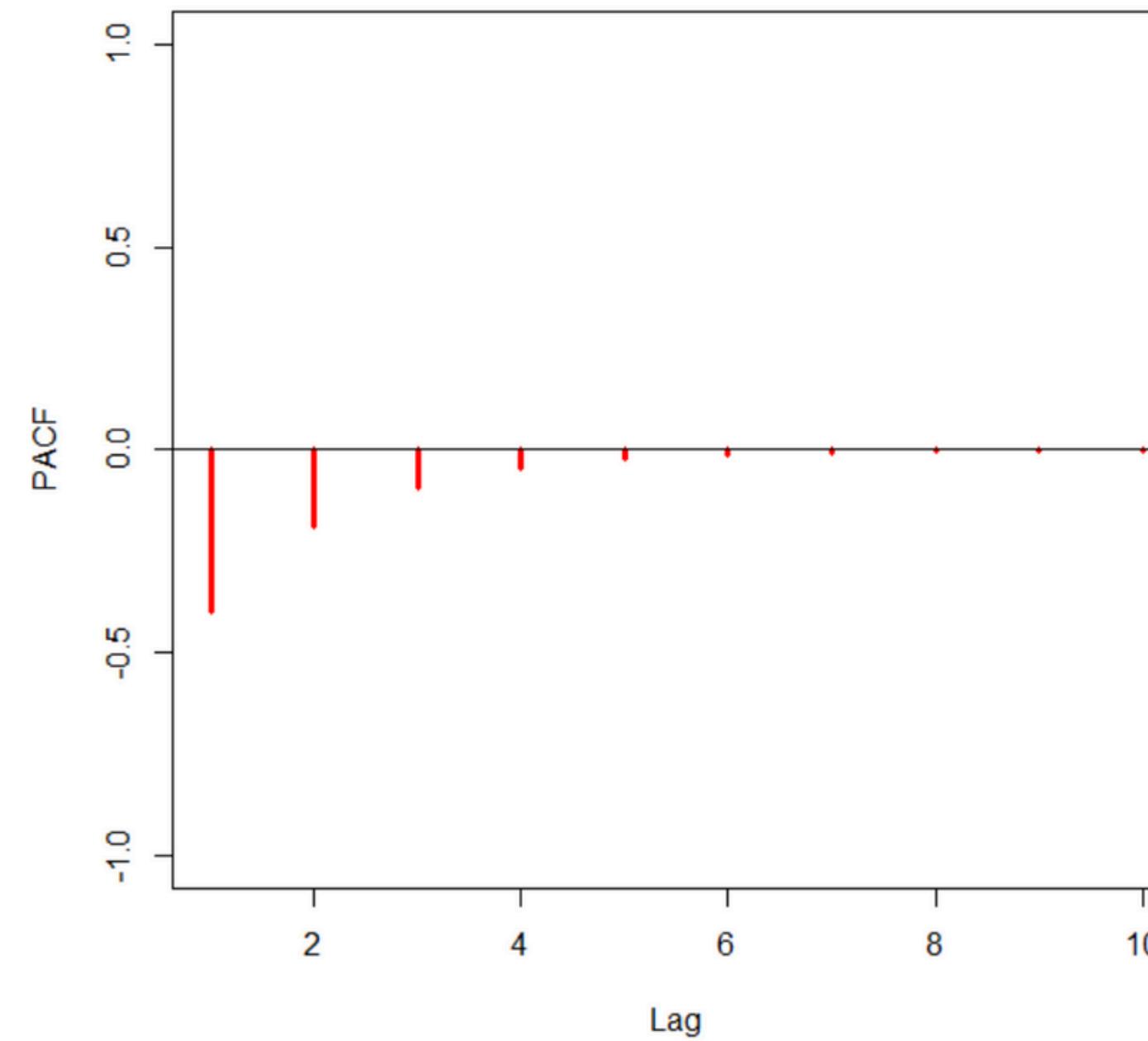
```
# --- PLOT KANAN (SAMPEL) ---  
# Kita gunakan data simulasi yang sama dari atas  
(data_sim)  
pacf(data_sim, lag.max = 10,  
      main = "PACF Sampel MA(1) (n=500)")  
  
# Kembalikan layout ke normal  
par(mfrow = c(1, 1))
```



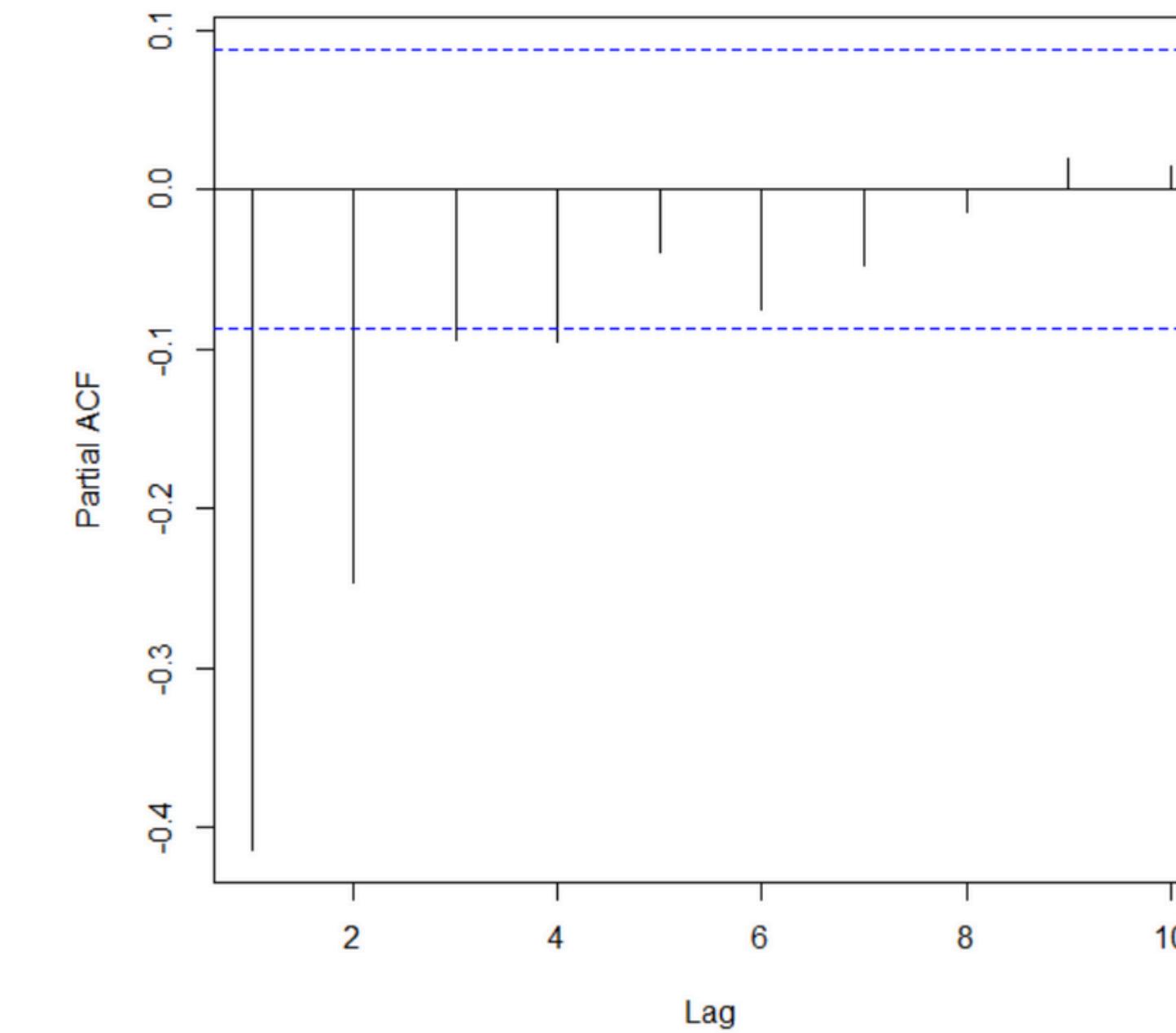


Plot PACF MA(1) dalam R

PACF Teoretis MA(1) (Meluruhan)



PACF Sampel MA(1) (n=500)





Proses ARMA(1,1)

Diketahui sebuah proses deret waktu stasioner $Z(t)$ mengikuti model MA(1) sebagai berikut:

$$Z(t) = 0.5 Z(t-1) + a(t) - 0.4 a(t-1)$$

Dimana $a(t)$ adalah white noise (WN) dengan rata-rata 0 dan varians 1.

Tugas:

Hitunglah:

1. Ragam (Varian) proses.
2. Koragam (Autokovarians).
3. Korelasi (ACF).
4. Korelasi Parsial (PACF).





Proses ARMA(1,1)

Kita memiliki parameter:

- $\phi_1 = 0.5$
- $\theta_1 = 0.4$
- $\sigma_a^2 = 1$

1. Perhitungan Ragam (γ_0)

Rumus γ_0 :

$$\begin{aligned}\gamma_0 &= \text{Var}(Z_t) \\ &= \text{Var}(\phi_1 Z_{t-1} + a_t - \theta_1 a_{t-1}) \\ &= \text{Var}(\phi_1 Z_{t-1}) + \text{Var}(a_t) + \text{Var}(\theta_1 a_{t-1}) - 2\text{Cov}(\phi_1 Z_{t-1}, \theta_1 a_{t-1}) \\ (\text{Catatan: } \text{Cov}(Z_{t-1}, a_{t-1}) &= \sigma_a^2)\end{aligned}$$

$$\begin{aligned}\gamma_0 &= \phi_1^2 \gamma_0 + \sigma_a^2 + \theta_1^2 \sigma_a^2 - 2\phi_1 \theta_1 \sigma_a^2 \\ \gamma_0 &= \phi_1^2 \gamma_0 + \sigma_a^2 (1 + \theta_1^2 - 2\phi_1 \theta_1) \\ \gamma_0 (1 - \phi_1^2) &= \sigma_a^2 (1 + \theta_1^2 - 2\phi_1 \theta_1) \\ \gamma_0 (1 - 0.5^2) &= 1 (1 + 0.4^2 - 2 \times 0.5 \times 0.4) \\ \gamma_0 (1 - 0.25) &= 1 (1 + 0.16 - 0.4) \\ \gamma_0 (0.75) &= 0.76 \\ \gamma_0 &= \frac{0.76}{0.75} \\ &\approx \mathbf{1.0133}\end{aligned}$$

Proses ARMA(1,1)

2. Perhitungan Koragam (γ_k)

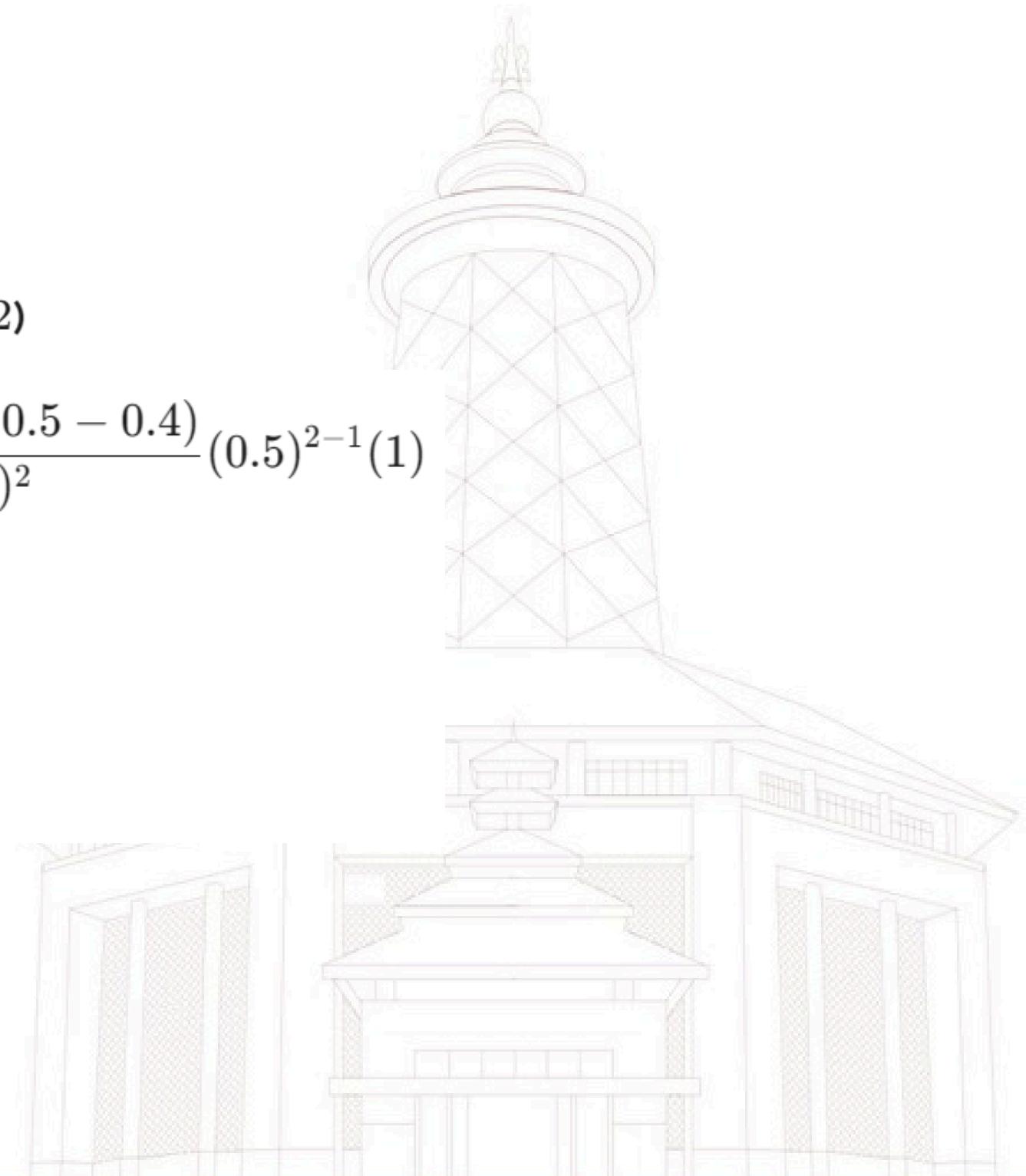
$$\gamma_k = \frac{(1 - \theta\phi)(\phi - \theta)}{1 - \phi^2} \phi^{k-1} \sigma_a^2$$

Untuk γ_1 (Lag 1, set $k = 1$)

$$\begin{aligned}\gamma_1 &= \frac{(1 - (0.4)(0.5))(0.5 - 0.4)}{1 - (0.5)^2} (0.5)^{1-1}(1) \\ &= \frac{(1 - 0.2)(0.1)}{1 - 0.25} (0.5)^0(1) \\ &= \frac{(0.8)(0.1)}{0.75}(1) \\ &= \frac{0.08}{0.75} \\ &\approx \mathbf{0.1067}\end{aligned}$$

Untuk γ_2 (Lag 2, set $k = 2$)

$$\begin{aligned}\gamma_2 &= \frac{(1 - (0.4)(0.5))(0.5 - 0.4)}{1 - (0.5)^2} (0.5)^{2-1}(1) \\ &= \left(\frac{0.08}{0.75}\right) \times (0.5)^1 \\ &= \frac{0.04}{0.75} \\ &\approx \mathbf{0.0533}\end{aligned}$$



Proses ARMA(1,1)

3. Perhitungan Korelasi / ACF (ρ_k) $\rho_k = \gamma_k / \gamma_0$

Untuk ρ_1 (Lag 1)

$$\begin{aligned}\rho_1 &= \frac{\gamma_1}{\gamma_0} \\ &= \frac{0.08/0.75}{0.76/0.75} \\ &= \frac{0.08}{0.76} \\ &\approx \mathbf{0.1053}\end{aligned}$$

Untuk ρ_2 (Lag 2)

$$\begin{aligned}\rho_2 &= \frac{\gamma_2}{\gamma_0} \\ &= \frac{0.04/0.75}{0.76/0.75} \\ &= \frac{0.04}{0.76} \\ &\approx \mathbf{0.0526}\end{aligned}$$

Perilaku ACF (seperti AR(1))

pola meluruh (tails off)

$$\rho_k = \phi_1 \rho_{k-1} \text{ (untuk } k \geq 2)$$

4. Perhitungan Korelasi Parsial / PACF (ϕ_{kk})

Untuk ϕ_{11} (Lag 1)

$$\begin{aligned}\phi_{11} &= \rho_1 \\ &\approx \mathbf{0.1053}\end{aligned}$$

Untuk ϕ_{22} (Lag 2)

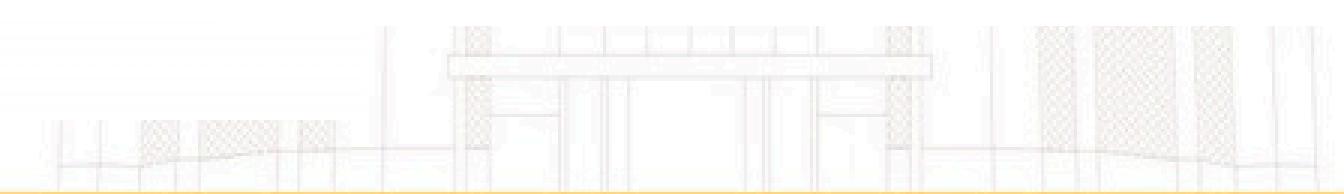
$$\begin{aligned}\phi_{22} &= \frac{\rho_2 - \rho_1^2}{1 - \rho_1^2} \\ &= \frac{0.0526 - (0.1053)^2}{1 - (0.1053)^2} \\ &= \frac{0.0526 - 0.0111}{1 - 0.0111} \\ &= \frac{0.0415}{0.9889} \\ &\approx \mathbf{0.042}\end{aligned}$$

$$\phi_{kk} = \frac{-\theta^k(1 - \theta^2)}{1 - \theta^{2(k+1)}} \quad \text{untuk } k \geq 1$$



Perilaku PACF (seperti MA(1))

pola meluruh (tails off)





Plot ACF ARMA(1,1) dalam R

```
dev.new()

# Atur layout jadi 1 baris, 2 kolom
par(mfrow = c(1, 2))

# --- PLOT KIRI (TEORETIS) ---
# ACF akan meluruh
acf_teori <- ARMAacf(ar = 0.5, ma = -0.4, lag.max = 10,
pacf = FALSE)

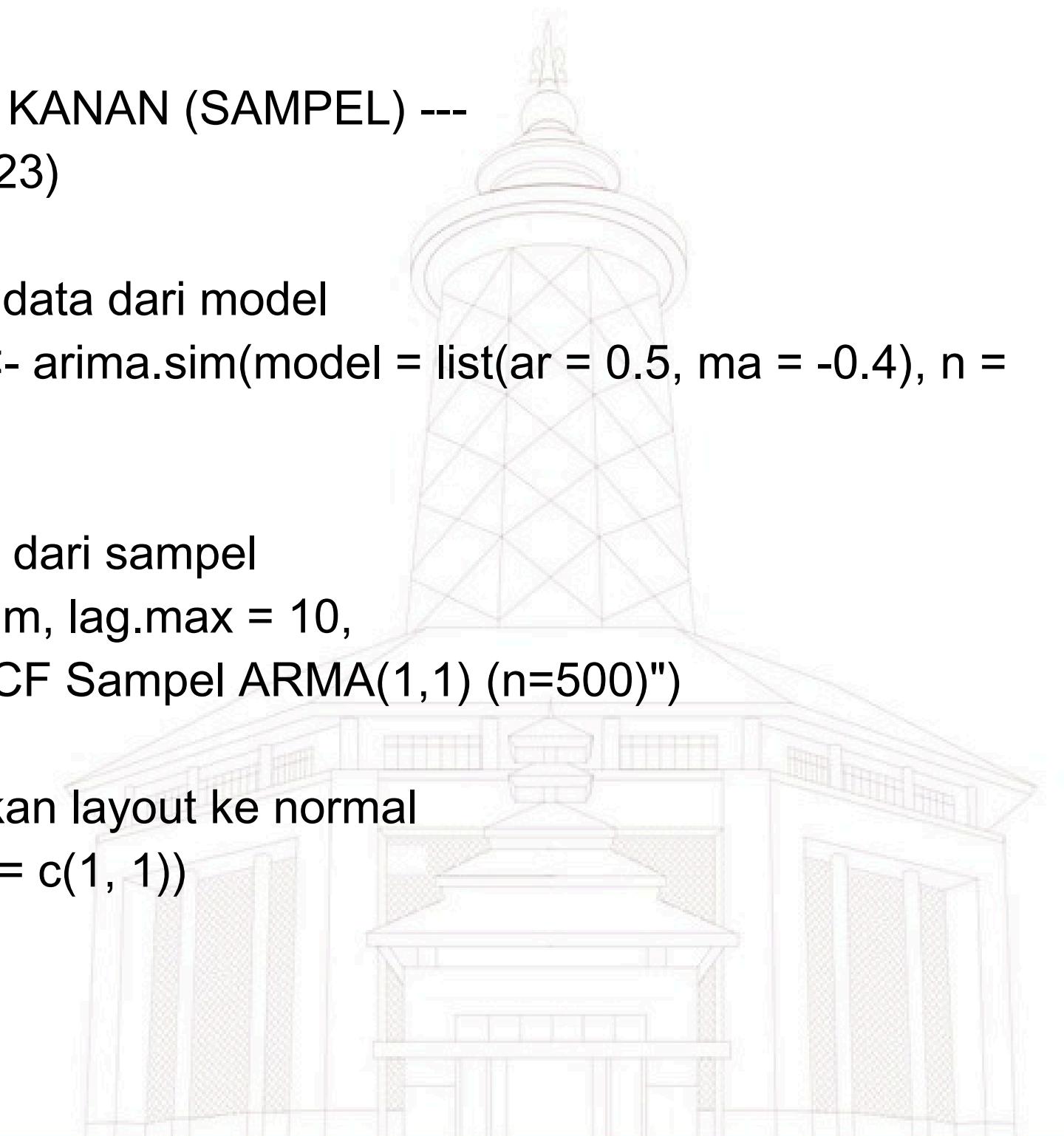
plot(acf_teori, type = "h",
      main = "ACF Teoretis ARMA(1,1) (Meluruh)",
      xlab = "Lag (termasuk Lag 0)", ylab = "ACF",
      ylim = c(-1, 1),
      lwd = 3, col = "red")
abline(h = 0)
```

```
# --- PLOT KANAN (SAMPEL) ---
set.seed(123)
```

```
# Simulasi data dari model
data_sim <- arima.sim(model = list(ar = 0.5, ma = -0.4), n =
500)

# Plot ACF dari sampel
acf(data_sim, lag.max = 10,
     main = "ACF Sampel ARMA(1,1) (n=500)")
```

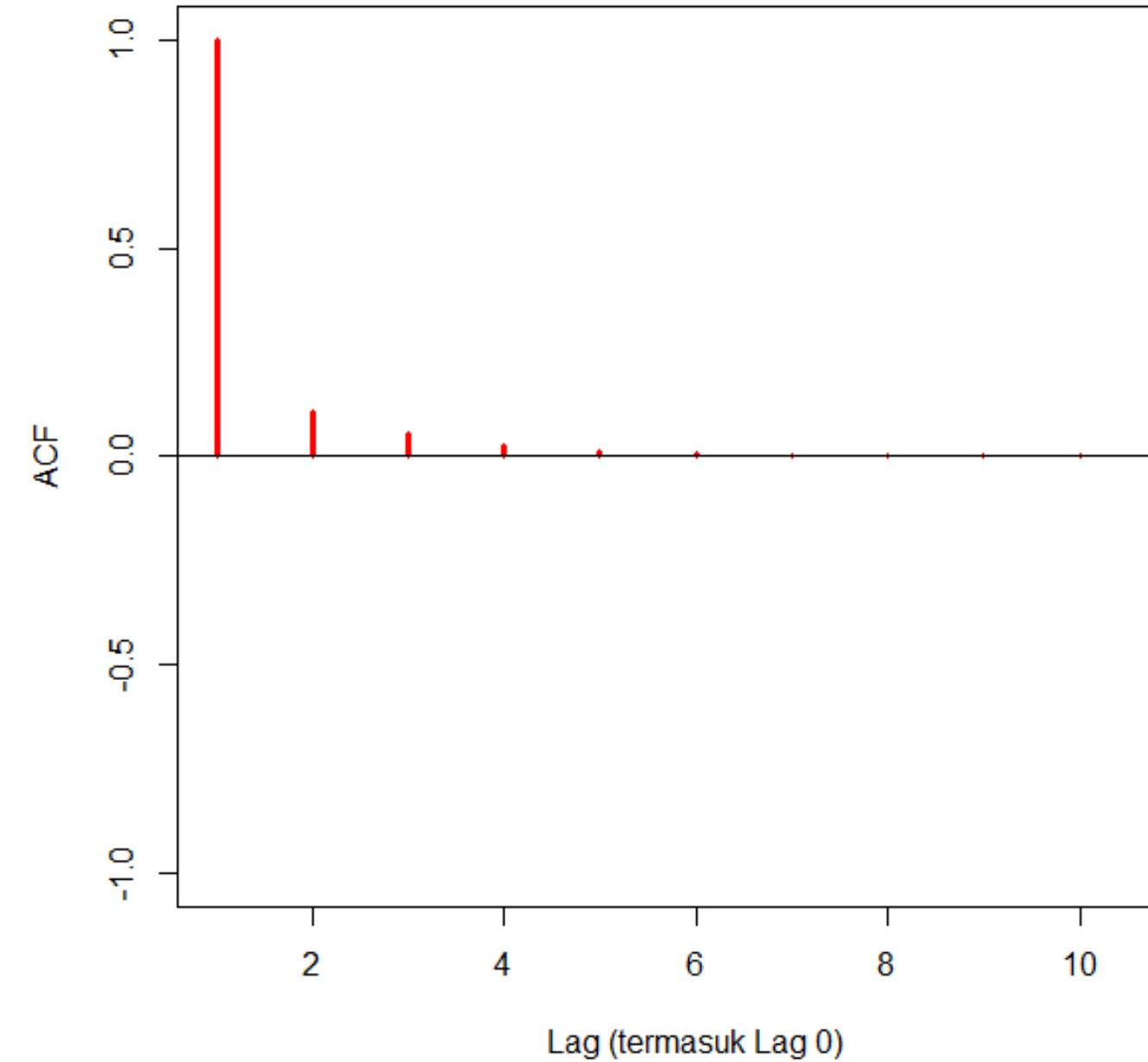
```
# Kembalikan layout ke normal
par(mfrow = c(1, 1))
```



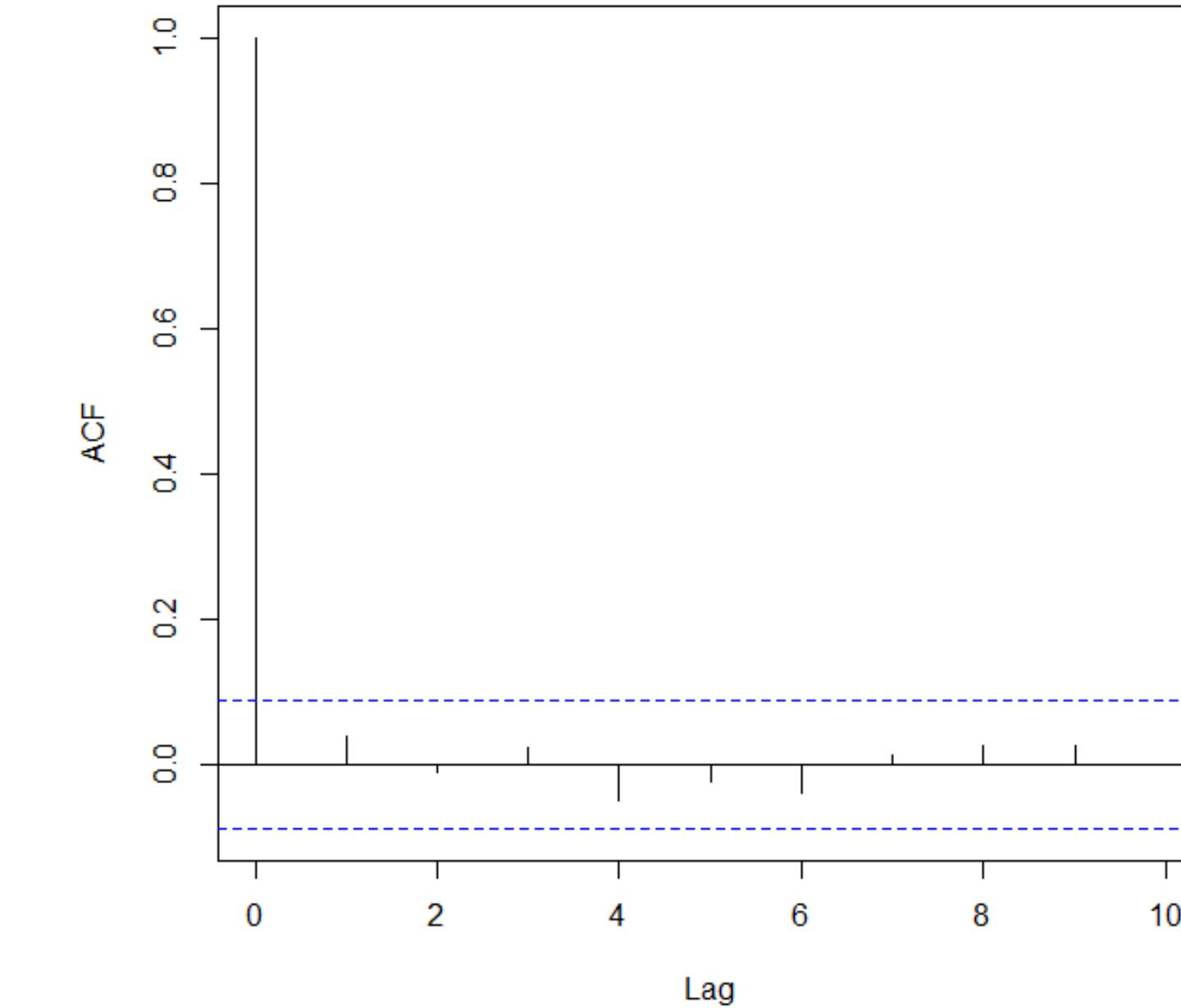


Plot ACF ARMA(1,1) dalam R

ACF Teoretis ARMA(1,1) (Meluruh)



ACF Sampel ARMA(1,1) (n=500)





Plot PACF ARMA(1,1) dalam R

```
dev.new()

# Atur layout jadi 1 baris, 2 kolom
par(mfrow = c(1, 2))

# --- PLOT KIRI (TEORETIS) ---
# PACF akan meluruh
pacf_teori <- ARMAacf(ar = 0.5, ma = -0.4, lag.max = 10,
pacf = TRUE)

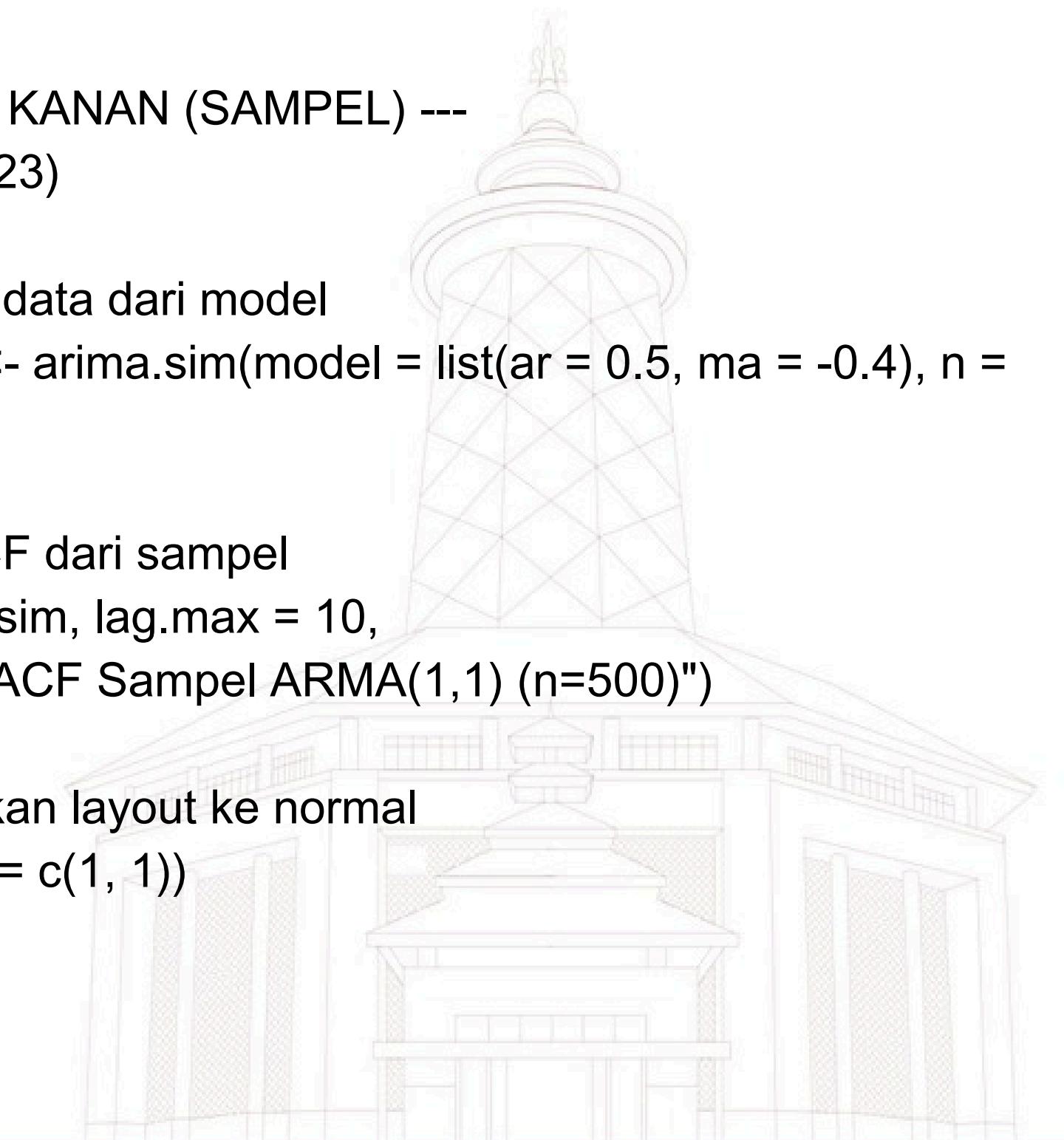
plot(pacf_teori, type = "h",
      main = "PACF Teoretis ARMA(1,1) (Meluruh)",
      xlab = "Lag", ylab = "PACF",
      ylim = c(-1, 1),
      lwd = 3, col = "red")
abline(h = 0)
```

```
# --- PLOT KANAN (SAMPEL) ---
set.seed(123)

# Simulasi data dari model
data_sim <- arima.sim(model = list(ar = 0.5, ma = -0.4), n =
500)

# Plot PACF dari sampel
pacf(data_sim, lag.max = 10,
      main = "PACF Sampel ARMA(1,1) (n=500)")

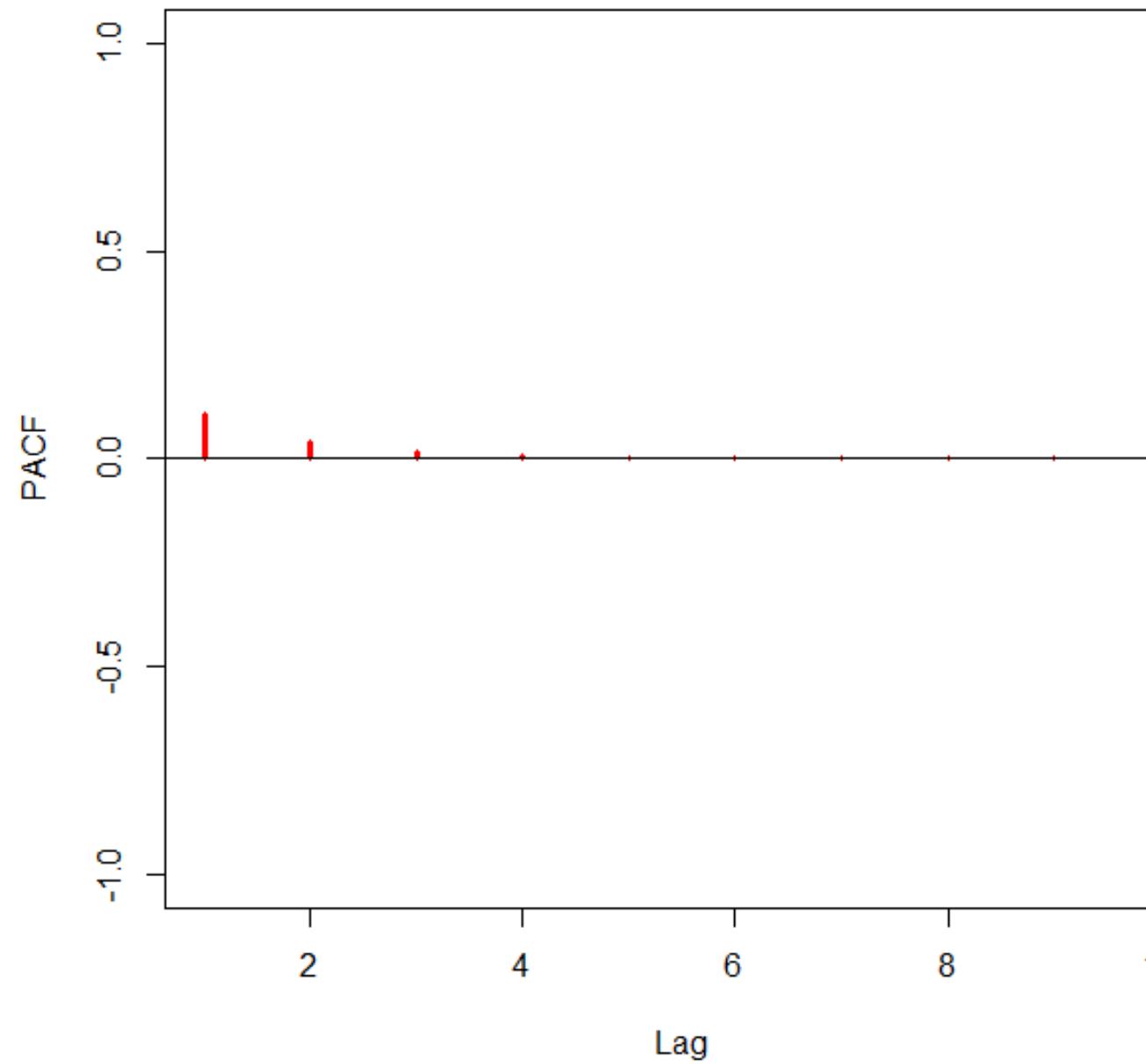
# Kembalikan layout ke normal
par(mfrow = c(1, 1))
```



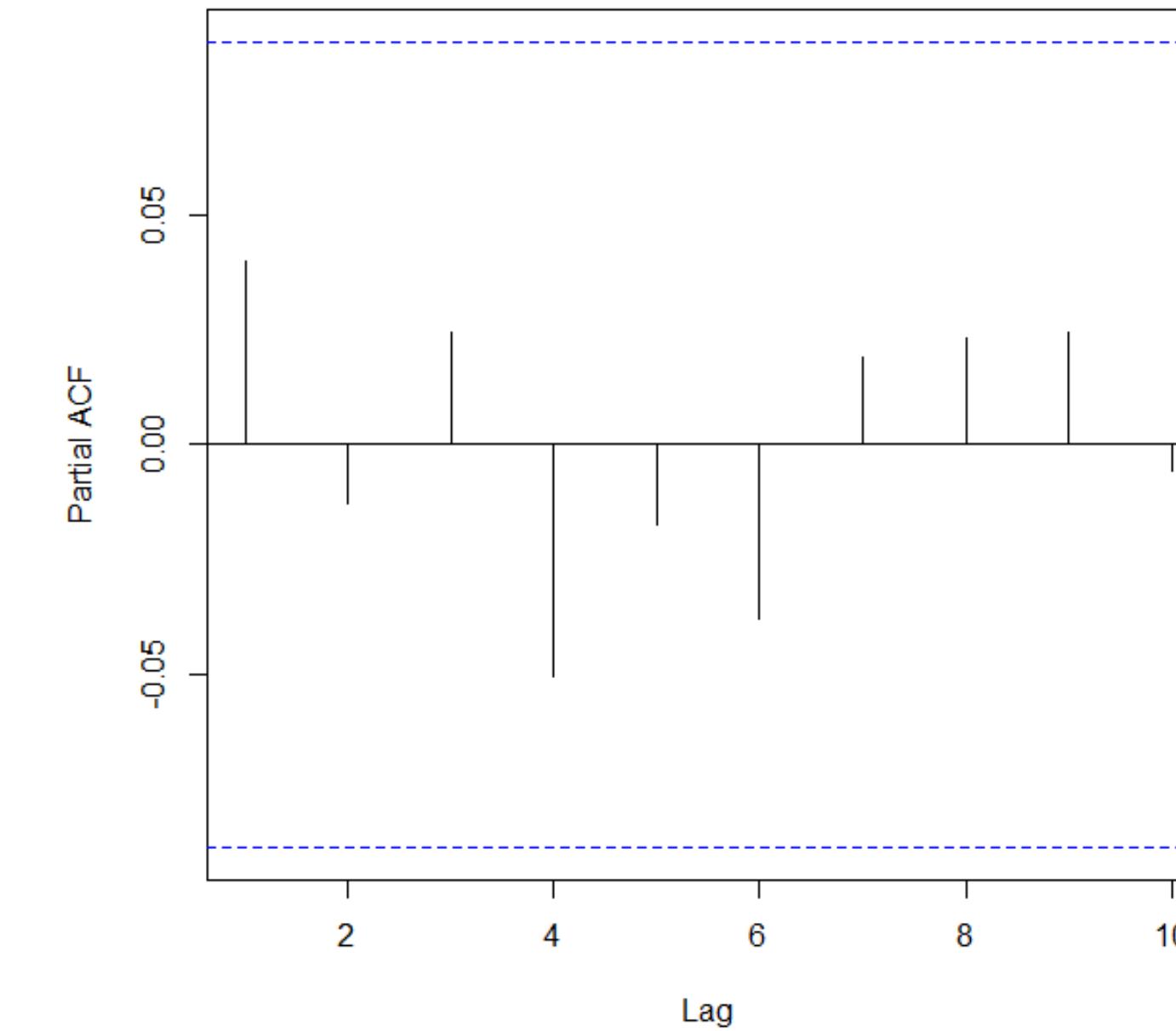


Plot PACF ARMA(1,1) dalam R

PACF Teoretis ARMA(1,1) (Meluruhan)



PACF Sampel ARMA(1,1) (n=500)





SEE YOU NEXT WEEK !

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