

# Assignment 3

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## Problem 2

```
# Set seed for reproducibility
set.seed(100)

# Define true parameter values
phi_true <- 0.9
theta_true <- 0.5
sigma2_true <- 1

# List of sample sizes
sample_sizes <- c(50, 200, 500)

# Number of simulations to run
num_simulations <- 1000

# Function to compute performance metrics
compute_metrics <- function(estimates, true_value, se_estimate = NULL) {
  mse <- mean((estimates - true_value)^2)
  mad <- mean(abs(estimates - true_value))

  if (!is.null(se_estimate)) {
    coverage <- mean(estimates - 1.96 * se_estimate < true_value &
                     true_value < estimates + 1.96 * se_estimate)
  } else {
    coverage <- NaN # Coverage is undefined for sigma^2
  }

  return(c(MSE = mse, MAD = mad, Coverage = coverage))
}
```

```

}

# Function to format results into a table
generate_table <- function(results, parameter_name) {
  kable(results,
    caption = paste("Performance Metrics for", parameter_name),
    col.names = c("MSE", "MAD", "Coverage"),
    row.names = TRUE,
    digits = 4)
}

# Initialize matrices for storing results
phi_results <- theta_results <- sigma2_results <- matrix(NA,
  nrow = length(sample_sizes), ncol = 3)
colnames(phi_results) <-
colnames(theta_results) <-
  colnames(sigma2_results) <- c("MSE", "MAD", "Coverage")
rownames(phi_results) <-
rownames(theta_results) <-
rownames(sigma2_results) <- paste0("n = ", sample_sizes)

# Run simulations for each sample size
for (i in 1:length(sample_sizes)) {
  n <- sample_sizes[i]

  # Initialize vectors to store estimates and standard errors
  phi_estimates <- theta_estimates <-
    sigma2_estimates <- phi_se_estimates <-
    theta_se_estimates <- numeric(num_simulations)

  for (j in 1:num_simulations) {
    # Simulate ARMA(1,1) process
    x <- arima.sim(n = n, list(ar = phi_true, ma = theta_true))
    fit <- arima(x, order = c(1, 0, 1))

    # Store parameter estimates and standard errors
    phi_estimates[j] <- fit$coef[1]
    theta_estimates[j] <- fit$coef[2]
    sigma2_estimates[j] <- fit$sigma2
    se <- sqrt(diag(fit$var.coef))
    phi_se_estimates[j] <- se[1]
    theta_se_estimates[j] <- se[2]
  }
}

```

```

}

# Compute performance metrics for phi, theta, and sigma^2
phi_results[i, ] <- compute_metrics(phi_estimates, phi_true,
                                   phi_se_estimates)
theta_results[i, ] <- compute_metrics(theta_estimates,
                                       theta_true, theta_se_estimates)
sigma2_results[i, ] <- compute_metrics(sigma2_estimates,
                                       sigma2_true)
}

# Display results for phi
generate_table(phi_results, "phi")

```

Table 1: Performance Metrics for phi

	MSE	MAD	Coverage
n = 50	0.0168	0.0917	0.909
n = 200	0.0019	0.0323	0.931
n = 500	0.0006	0.0180	0.939

```

# Display results for theta
generate_table(theta_results, "theta")

```

Table 2: Performance Metrics for theta

	MSE	MAD	Coverage
n = 50	0.0210	0.1123	0.915
n = 200	0.0042	0.0516	0.939
n = 500	0.0017	0.0325	0.946

```

# Display results for sigma^2
generate_table(sigma2_results, "sigma^2")

```

Table 3: Performance Metrics for  $\sigma^2$

	MSE	MAD	Coverage
n = 50	0.0445	0.1698	NaN
n = 200	0.0105	0.0825	NaN
n = 500	0.0043	0.0525	NaN

### Problem 3

```
# Load the cmort dataset
data(cmort)
# Fit AR(2) model using OLS
reg1 <- ar.ols(cmort, order = 2)

# Print model summary
cat("OLS Estimates:\n")
```

OLS Estimates:

```
print(reg1)
```

Call:

```
ar.ols(x = cmort, order.max = 2)
```

Coefficients:

```
      1      2
0.4286 0.4418
```

Intercept: -0.04672 (0.2527)

Order selected 2  $\sigma^2$  estimated as 32.32

```
# Calculate forecasts for 8 week horizon using OLS model
forecasts <- forecast(reg1, h = 8)

# Print forecasts
print(forecasts)
```

	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
1979.769	87.59986	80.31444	94.88529	76.45777	98.74196
1979.788	86.76349	78.83713	94.68985	74.64117	98.88581
1979.808	87.33714	78.19426	96.48002	73.35431	101.31997
1979.827	87.21350	77.48222	96.94478	72.33079	102.09621
1979.846	87.41394	77.09205	97.73583	71.62798	103.19990
1979.865	87.44522	76.71082	98.17963	71.02837	103.86208
1979.885	87.54719	76.45775	98.63662	70.58736	104.50701
1979.904	87.60471	76.23805	98.97136	70.22091	104.98850

```
# Extract point forecasts and prediction intervals
point_forecasts <- forecasts$mean
lower_pi <- forecasts$lower[, 2] # 95% lower prediction interval
upper_pi <- forecasts$upper[, 2] # 95% upper prediction interval

# Print forecasts and prediction intervals
cat("\nForecasts and 95% Prediction Intervals:\n")
```

Forecasts and 95% Prediction Intervals:

```
for (i in 1:8) {
  cat(sprintf("Week %d: %.4f (%.4f, %.4f)\n", i,
    point_forecasts[i], lower_pi[i], upper_pi[i]))
}
```

```
Week 1: 87.5999 (76.4578, 98.7420)
Week 2: 86.7635 (74.6412, 98.8858)
Week 3: 87.3371 (73.3543, 101.3200)
Week 4: 87.2135 (72.3308, 102.0962)
Week 5: 87.4139 (71.6280, 103.1999)
Week 6: 87.4452 (71.0284, 103.8621)
Week 7: 87.5472 (70.5874, 104.5070)
Week 8: 87.6047 (70.2209, 104.9885)
```

```
# Fit AR(2) model using Yule-Walker
reg2 <- ar.yw(cmort, order = 2)

cat("\nYule-Walker Estimates:\n")
```

Yule-Walker Estimates:

```
print(reg2)
```

Call:

```
ar.yw.default(x = cmort, order.max = 2)
```

Coefficients:

```
      1      2  
0.4339 0.4376
```

Order selected 2 sigma<sup>2</sup> estimated as 32.84

```
# Compare estimates and standard errors  
# Create a data frame for the comparison table  
comparison_df <- data.frame(  
  Parameter = c("AR1", "AR2"),  
  OLS_Estimate = c(reg1$ar[1], reg1$ar[2]),  
  OLS_SE = c(reg1$asy.se.coef$ar[1], reg1$asy.se.coef$ar[2]),  
  YW_Estimate = c(reg2$ar[1], reg2$ar[2]),  
  YW_SE = sqrt(diag(reg2$asy.var.coef))  
)  
  
# Display the comparison table  
kable(comparison_df,  
      col.names = c("Parameter",  
                    "OLS Estimate", "OLS SE", "Yule-Walker Estimate",  
                    "Yule-Walker SE"),  
      caption = "Comparison of Estimates and Standard Errors",  
      digits = 4)
```

Table 4: Comparison of Estimates and Standard Errors

Parameter	OLS Estimate	OLS SE	Yule-Walker Estimate	Yule-Walker SE
AR1	0.4286	0.0398	0.4339	0.04
AR2	0.4418	0.0398	0.4376	0.04

```
# Display variance estimates  
cat("Variance estimate (OLS):", sprintf("%.2f", reg1$var.pred), "\n")
```

Variance estimate (OLS): 32.32

```
cat("Variance estimate (Yule-Walker):", sprintf("%.2f", reg2$var.pred))
```

Variance estimate (Yule-Walker): 32.84

The results are consistent using 2 methods