STAT 400 - Quiz 3

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Question 1

```
# Data
x <- c(26.8, 25.4, 28.9, 23.6, 27.7, 23.9, 24.7, 28.1, 26.9, 27.4, 22.6, 25.6)
y <- c(26.5, 27.3, 24.2, 27.1, 23.6, 25.9, 26.3, 22.5, 21.7, 21.4, 25.8, 24.9)

# Part (a): Regression Line
x_mean <- mean(x)
y_mean <- mean(y)
cov_xy <- cov(x, y)
var_x <- var(x)
beta_1 <- cov_xy / var_x
beta_0 <- y_mean - beta_1 * x_mean
cat("Regression Line: y =", beta_0, "+", beta_1, "* x\n")</pre>
```

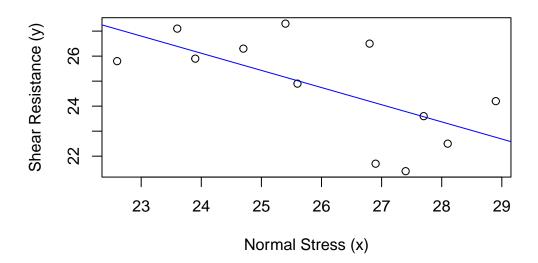
Regression Line: y = 42.5818 + -0.6860771 * x

```
# Part (b): Prediction
x_new <- 24.5
y_pred <- beta_0 + beta_1 * x_new
cat("Predicted Shear Resistance:", y_pred, "\n")</pre>
```

Predicted Shear Resistance: 25.77291

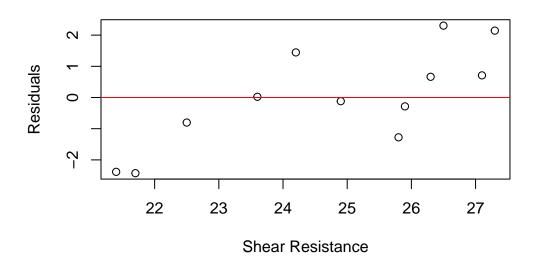
```
# Part (c): Scatter Plot
plot(x, y, main="Scatter Plot with Regression Line", xlab="Normal Stress (x)", ylab="Shear R
# Part (d): Graph Regression Line
abline(beta_0, beta_1, col="blue")
```

Scatter Plot with Regression Line



```
# Part (e): Residuals Plot
y_hat <- beta_0 + beta_1 * x
residuals <- y - y_hat
plot(y, residuals, main="Residuals vs. Shear Resistance", xlab="Shear Resistance", ylab="Res
abline(h=0, col='red')</pre>
```

Residuals vs. Shear Resistance



Question 2

```
# Part (a): Residual Variance (s^2)
SS_residual <- sum(residuals^2)
n <- length(x)
s2 <- SS_residual / (n - 2)
cat("Residual Variance (s^2):", s2, "\n")</pre>
```

Residual Variance (s^2): 2.688452

```
# Part (b): Hypothesis Test for Beta_1
SE_beta1 <- sqrt(s2 / sum((x - x_mean)^2))
t_value <- beta_1 / SE_beta1
cat("t-statistic for beta_1:", t_value, "\n")</pre>
```

t-statistic for beta_1: -2.745311

```
p_value <- 2 * (1 - pt(abs(t_value), df = n - 2))
cat("p-value for beta_1:", p_value, "\n")</pre>
```

p-value for beta_1: 0.02064371

```
# Part (c): Confidence Interval for Beta_0
alpha <- 0.05
t_critical <- qt(1 - alpha / 2, df = n - 2)
SE_beta0 <- sqrt(s2 * (1 / n + x_mean^2 / sum((x - x_mean)^2)))
CI_beta0 <- c(beta_0 - t_critical * SE_beta0, beta_0 + t_critical * SE_beta0)
cat("95% CI for beta_0:", CI_beta0, "\n")</pre>
```

95% CI for beta_0: 28.08434 57.07927

```
# Part (d): Confidence Interval for Beta_1
CI_beta1 <- c(beta_1 - t_critical * SE_beta1, beta_1 + t_critical * SE_beta1)
cat("95% CI for beta_1:", CI_beta1, "\n")</pre>
```

95% CI for beta_1: -1.242908 -0.1292458

Question 3

```
# Part (a): Coefficient of Determination (R^2)
SS_total <- sum((y - y_mean)^2)
R_squared <- 1 - (SS_residual / SS_total)
cat("R-squared:", R_squared, "\n")</pre>
```

R-squared: 0.4297683

```
# Part (b): Lack-of-Fit Test
SS_pure_error <- 0  # Replace with actual value if replicated x-values exist
SS_lack_of_fit <- SS_residual - SS_pure_error
df_pure_error <- 0  # Replace if replicated
df_lack_of_fit <- n - 2 - df_pure_error
MS_lack_of_fit <- SS_lack_of_fit / df_lack_of_fit
MS_pure_error <- ifelse(df_pure_error > 0, SS_pure_error / df_pure_error, NA)
F_lack_of_fit <- ifelse(!is.na(MS_pure_error), MS_lack_of_fit / MS_pure_error, NA)
cat("F-statistic for Lack-of-Fit:", F_lack_of_fit, "\n")</pre>
```

F-statistic for Lack-of-Fit: NA

```
# Part (c): Hypothesis Test Using F-statistic
SS_regression <- SS_total - SS_residual
df_regression <- 1
df_residual <- n - 2
MS_regression <- SS_regression / df_regression
MS_residual <- SS_residual / df_residual
F_statistic <- MS_regression / MS_residual
cat("F-statistic for regression:", F_statistic, "\n")</pre>
```

F-statistic for regression: 7.536732

Question 4

```
# Part (a): Correlation Coefficient
correlation <- cor(x, y)
cat("Correlation coefficient (r):", correlation, "\n")</pre>
```

Correlation coefficient (r): -0.6555672

```
# Part (b): Hypothesis Test for Rho
rho_0 <- -0.5
SE_r <- sqrt((1 - correlation^2) / (n - 2))
t_value_r <- (correlation - rho_0) / SE_r
cat("t-statistic for testing rho = -0.5:", t_value_r, "\n")</pre>
```

t-statistic for testing rho = -0.5: -0.6514669

```
p_value_r <- pt(t_value_r, df = n - 2)
cat("p-value for rho = -0.5:", p_value_r, "\n")</pre>
```

p-value for rho = -0.5: 0.2647157

```
# Part (c): Percentage of Variation Explained
variation_explained <- correlation^2 * 100
cat("Percentage of variation explained by X:", variation_explained, "%\n")</pre>
```

Percentage of variation explained by X: 42.97683 %

Question 5

```
# Given Value of X
x_{new} < -24.5
SE_CI \leftarrow sqrt(s2 * (1 / n + (x_new - x_mean)^2 / sum((x - x_mean)^2)))
SE_PI \leftarrow sqrt(s2 * (1 + 1 / n + (x_new - x_mean)^2 / sum((x - x_mean)^2)))
CI <- c(beta_0 + beta_1 * x_new - t_critical * SE_CI, beta_0 + beta_1 * x_new + t_critical *
PI <- c(beta_0 + beta_1 * x_new - t_critical * SE_PI, beta_0 + beta_1 * x_new + t_critical *
cat("95% Confidence Interval for mean response:", CI, "\n")
95% Confidence Interval for mean response: 24.43903 27.10679
cat("95% Prediction Interval for individual observation:", PI, "\n")
95% Prediction Interval for individual observation: 21.88365 29.66217
# Find Lowest Standard Error
SE_{values} \leftarrow sqrt(s2 * (1 + 1 / n + (x - x_mean)^2 / sum((x - x_mean)^2)))
min_SE <- min(SE_values)</pre>
min_SE_index <- which.min(SE_values)</pre>
cat("Observation with lowest SE (index):", min_SE_index, "\n")
Observation with lowest SE (index): 12
cat("Standard Error for this observation:", min_SE, "\n")
```

Standard Error for this observation: 1.70906

```
# Data
x1 <- c(14.62, 15.63, 14.62, 15.00, 14.50, 15.25, 16.12, 15.13, 15.50, 15.13, 15.50, 16.12,
x2 <- c(226.0, 220.0, 217.4, 220.0, 226.5, 224.1, 220.5, 223.5, 217.6, 228.5, 230.2, 226.5, x3 <- c(7.000, 3.375, 6.375, 6.000, 7.625, 6.000, 3.375, 6.125, 5.000, 6.625, 5.750, 3.750, y <- c(128.40, 52.62, 113.90, 98.01, 139.90, 102.60, 48.14, 109.60, 82.68, 112.60, 97.52, 59
# Fit multiple linear regression model
model <- lm(y ~ x1 + x2 + x3)
summary(model)</pre>
```

```
Call:
lm(formula = y \sim x1 + x2 + x3)
Residuals:
            1Q Median
                            3Q
                                   Max
-6.9517 -2.3992 0.3168 1.8602 7.8955
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) -21.4696 51.6756 -0.415
                                         0.684
            -3.3243 3.5888 -0.926
                                          0.369
x1
x2
                        0.2747 0.897
             0.2465
                                          0.384
            20.3448 1.2576 16.178 6.65e-11 ***
xЗ
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
Residual standard error: 4.151 on 15 degrees of freedom
Multiple R-squared: 0.9916, Adjusted R-squared: 0.9899
F-statistic: 588.1 on 3 and 15 DF, p-value: 8.976e-16
# Part (b): Prediction for x1 = 14, x2 = 220, x3 = 5
new_data \leftarrow data.frame(x1 = 14, x2 = 220, x3 = 5)
y_pred <- predict(model, newdata = new_data)</pre>
cat("Predicted y:", y_pred, "\n")
```

Predicted y: 87.94123

```
# Full model
full_model <- lm(y ~ x1 + x2 + x3)

# Reduced model (excluding x1)
reduced_model <- lm(y ~ x2 + x3)

# Part (a): Compare Adjusted R^2
R2_adj_full <- summary(full_model)$adj.r.squared
R2_adj_reduced <- summary(reduced_model)$adj.r.squared
cat("Adjusted R^2 (Full Model):", R2_adj_full, "\n")</pre>
```

```
Adjusted R^2 (Full Model): 0.9898834
cat("Adjusted R^2 (Reduced Model):", R2_adj_reduced, "\n")
Adjusted R^2 (Reduced Model): 0.9899731
# Part (b): Compare Prediction Interval Widths
new_data \leftarrow data.frame(x2 = 220, x3 = 5)
# Full model prediction interval
PI_full <- predict(full_model, new_data = new_data, interval = "prediction")
Warning in predict.lm(full_model, new_data = new_data, interval = "prediction"): predictions
# Reduced model prediction interval
PI_reduced <- predict(reduced_model, new_data = new_data, interval = "prediction")
Warning in predict.lm(reduced_model, new_data = new_data, interval = "prediction"): prediction
# Calculate and print widths
PI_width_full <- PI_full[3] - PI_full[2]
PI_width_reduced <- PI_reduced[3] - PI_reduced[2]
cat("Prediction Interval Width (Full Model):", PI_width_full, "\n")
Prediction Interval Width (Full Model): 63.75113
cat("Prediction Interval Width (Reduced Model):", PI_width_reduced, "\n")
Prediction Interval Width (Reduced Model): 63.81373
Question 8
# ANOVA for the full model
anova_full <- anova(full_model)</pre>
print(anova_full)
```

Analysis of Variance Table Response: y Df Sum Sq Mean Sq F value 1 18252.3 18252.3 1059.42 2.490e-15 *** x1 x2 1 7634.1 7634.1 443.11 1.501e-12 *** xЗ 1 4509.2 4509.2 261.73 6.646e-11 *** Residuals 15 258.4 17.2 Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1 # Part (a): Sum of Squares SS_regression <- sum(anova_full\$`Sum Sq`[-length(anova_full\$`Sum Sq`)]) SS_residual <- anova_full\$`Sum Sq`[length(anova_full\$`Sum Sq`)] SS_total <- SS_regression + SS_residual # Degrees of Freedom df_regression <- sum(anova_full\$Df[-length(anova_full\$Df)])</pre> df_residual <- anova_full\$Df[length(anova_full\$Df)]</pre> df_total <- df_regression + df_residual</pre>

F-statistic: 588.0842

MS_regression <- SS_regression / df_regression

MS_residual <- SS_residual / df_residual

F_statistic <- MS_regression / MS_residual cat("F-statistic:", F_statistic, "\n")

Mean Squares

F-statistic

```
# p-value
p_value <- pf(F_statistic, df1 = df_regression, df2 = df_residual, lower.tail = FALSE)
cat("p-value:", p_value, "\n")</pre>
```

p-value: 8.975576e-16

```
# Data
profit <- c(157, -181, -253, 158, 75, 202, -451, 146, 89, -357, 522, 78, 5, -177, 123, 251,
income <- c(45000, 55000, 45800, 38000, 75000, 99750, 28000, 39000, 54350, 32500, 36750, 4250
gender <- c(1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0) # 1 = Male, 0 = Female
family_members <-c(1, 2, 4, 3, 4, 4, 1, 2, 1, 1, 1, 3, 2, 3, 2, 1, 1, 1, 1, 2)
# Fit model
credit_model <- lm(profit ~ income + gender + family_members)</pre>
summary(credit_model)
Call:
lm(formula = profit ~ income + gender + family_members)
Residuals:
   Min
            1Q Median
                            3Q
                                   Max
-347.24 -150.85
                  7.16 132.66 341.49
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
                                      0.237
(Intercept)
               3.008e+01 1.267e+02
                                              0.8153
income
               5.433e-03 2.741e-03
                                      1.982
                                              0.0649 .
              -2.367e+02 1.106e+02 -2.141
                                              0.0480 *
gender
family_members -4.924e+01 5.196e+01 -0.948
                                              0.3574
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 227.5 on 16 degrees of freedom
Multiple R-squared: 0.3075,
                               Adjusted R-squared:
F-statistic: 2.368 on 3 and 16 DF, p-value: 0.1091
```

```
# ------
# Question 10: Regression Model Selection

# Data
data <- data.frame(
    Y = c(11.2, 14.5, 17.2, 17.8, 19.3, 24.5, 21.2, 16.9, 14.8, 20.0, 13.2, 22.5),</pre>
```

```
X1 = c(56.5, 59.5, 69.2, 74.5, 81.2, 88.0, 78.2, 69.0, 58.1, 80.5, 58.3, 84.0),
  X2 = c(71.0, 72.5, 76.0, 79.5, 84.0, 86.2, 80.5, 72.0, 68.0, 85.0, 71.0, 87.2),
  X3 = c(38.5, 38.2, 42.5, 43.4, 47.5, 47.4, 44.5, 41.8, 42.1, 48.1, 37.5, 51.0),
  X4 = c(43.0, 44.8, 49.0, 56.3, 60.2, 62.0, 58.1, 48.1, 46.0, 60.3, 47.1, 65.2)
# Part (a): Forward Selection
# Null model (no predictors)
null_model <- lm(Y ~ 1, data = data)</pre>
# Full model (all predictors)
full_model <- lm(Y ~ ., data = data)</pre>
# Perform forward selection
forward_model <- step(null_model, scope = list(lower = null_model, upper = full_model), dire</pre>
Start: AIC=33.91
Y ~ 1
       Df Sum of Sq
                        RSS
                                AIC
+ X1
             158.41 12.978 4.940
+ X4
             145.29 26.100 13.324
+ X2
        1
             136.01 35.380 16.975
+ X3
             133.65 37.741 17.750
        1
<none>
                    171.389 33.908
Step: AIC=4.94
Y ~ X1
       Df Sum of Sq
                       RSS
                               AIC
                    12.978 4.9404
<none>
+ X2
        1
            1.92969 11.049 5.0088
+ X4
        1
            0.02886 12.949 6.9137
+ X3
        1
            0.01684 12.961 6.9249
# Output summary of the selected model
cat("Forward Selection Model Summary:\n")
```

Forward Selection Model Summary:

summary(forward_model)

```
Call:
lm(formula = Y ~ X1, data = data)
Residuals:
    Min
                  Median
              1Q
                              3Q
                                      Max
-1.75899 -0.86677 0.07325 0.85826 1.53439
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) -6.33592 2.20553 -2.873 0.0166 *
Х1
            ---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.139 on 10 degrees of freedom
Multiple R-squared: 0.9243, Adjusted R-squared: 0.9167
F-statistic: 122.1 on 1 and 10 DF, p-value: 6.331e-07
# Part (b): Backward Elimination
# Perform backward elimination
backward_model <- step(full_model, direction = "backward")</pre>
Start: AIC=8.34
Y \sim X1 + X2 + X3 + X4
      Df Sum of Sq
                   RSS
                         AIC
- X3
            0.0065 10.460 6.3515
- X4
            0.3963 10.850 6.7905
<none>
                  10.453 8.3440
- X2
      1 2.4315 12.885 8.8536
- X1
      1 15.0455 25.499 17.0446
Step: AIC=6.35
Y \sim X1 + X2 + X4
      Df Sum of Sq RSS
                             AIC
- X4
      1 0.5889 11.049 5.0088
```

```
<none>
                  10.460 6.3515
- X2
     1 2.4897 12.949 6.9137
- X1
      1 15.6378 26.098 15.3232
Step: AIC=5.01
Y \sim X1 + X2
      Df Sum of Sq RSS
- X2
      1 1.9297 12.978 4.9404
<none>
                  11.049 5.0088
- X1 1 24.3318 35.380 16.9750
Step: AIC=4.94
Y ~ X1
      Df Sum of Sq RSS
<none>
                   12.978 4.940
- X1
     1 158.41 171.389 33.908
# Output summary of the selected model
cat("Backward Elimination Model Summary:\n")
Backward Elimination Model Summary:
summary(backward_model)
Call:
lm(formula = Y ~ X1, data = data)
Residuals:
             1Q
                 Median
                               ЗQ
                                      Max
-1.75899 -0.86677 0.07325 0.85826 1.53439
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) -6.33592 2.20553 -2.873
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Х1

```
Multiple R-squared: 0.9243,
                             Adjusted R-squared: 0.9167
F-statistic: 122.1 on 1 and 10 DF, p-value: 6.331e-07
# Part (c): Stepwise Regression
# Perform stepwise regression
stepwise_model <- step(null_model, scope = list(lower = null_model, upper = full_model), directions
Start: AIC=33.91
Y ~ 1
      Df Sum of Sq
                       RSS
                              AIC
            158.41 12.978 4.940
+ X1
+ X4
            145.29 26.100 13.324
+ X2
        1
           136.01 35.380 16.975
+ X3
        1 133.65 37.741 17.750
<none>
                   171.389 33.908
Step: AIC=4.94
Y ~ X1
      Df Sum of Sq
                       RSS
                             AIC
<none>
                    12.978 4.940
+ X2
            1.930 11.049 5.009
        1
+ X4
            0.029 12.949 6.914
        1
+ X3
            0.017 12.961 6.925
        1
- X1
           158.411 171.389 33.908
# Output summary of the selected model
cat("Stepwise Regression Model Summary:\n")
Stepwise Regression Model Summary:
summary(stepwise_model)
Call:
lm(formula = Y ~ X1, data = data)
```

Residual standard error: 1.139 on 10 degrees of freedom

Residuals:

Min 1Q Median 3Q Max -1.75899 -0.86677 0.07325 0.85826 1.53439

Coefficients:

Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.139 on 10 degrees of freedom Multiple R-squared: 0.9243, Adjusted R-squared: 0.9167 F-statistic: 122.1 on 1 and 10 DF, p-value: 6.331e-07