STAT2170/STAT6180 Assignment 2025

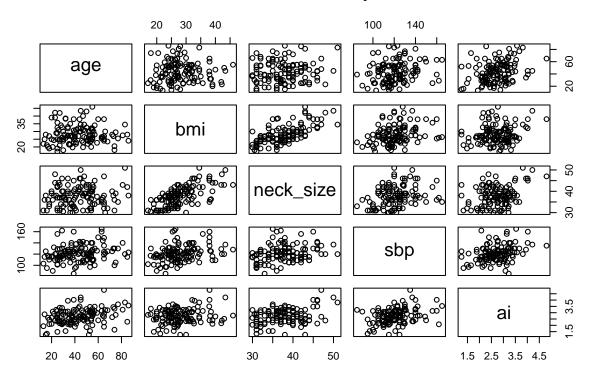
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Question 1

(a) Correlation and Plot

```
sleep <- read.csv("sleep.csv")
pairs(sleep, main="Pairwise Plots of Sleep Data")</pre>
```

Pairwise Plots of Sleep Data



round(cor(sleep), 2)

```
## age bmi neck_size sbp ai
## age 1.00 0.02 0.08 0.20 0.32
## bmi 0.02 1.00 0.67 0.31 0.19
```

We observe moderate correlation between neck size and arousal index, as well as some positive correlation with age and systolic blood pressure.

(b) Full Regression Model & Neck Size Impact

```
model_full <- lm(ai ~ age + bmi + neck_size + sbp, data=sleep)
summary(model_full)</pre>
```

```
##
## Call:
## lm(formula = ai ~ age + bmi + neck_size + sbp, data = sleep)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                            Max
## -1.67136 -0.32269 0.01491 0.35778 1.47595
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.159406
                          0.518207 -0.308 0.75893
                          0.002964
                                     2.965 0.00367 **
## age
               0.008789
               -0.009852
                          0.011312 -0.871 0.38557
## bmi
               0.040627
                          0.014208
                                    2.859 0.00503 **
## neck_size
               0.010218
                          0.003555
                                     2.875 0.00481 **
## sbp
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.5417 on 117 degrees of freedom
## Multiple R-squared: 0.2471, Adjusted R-squared: 0.2213
## F-statistic: 9.598 on 4 and 117 DF, p-value: 9.54e-07
confint(model_full, "neck_size", level = 0.95)
```

```
## 2.5 % 97.5 %
## neck_size 0.01248892 0.06876571
```

Each cm increase in neck size is associated with an estimated 0.041 increase in log-scaled arousal index (CI: $[0.012,\,0.069]$).

(c) ANOVA and Overall Model Test

```
Model: ai = \beta_0 + \beta_1 \cdot age + \beta_2 \cdot bmi + \beta_3 \cdot neck\_size + \beta_4 \cdot sbp + \varepsilon 
Hypotheses: H_0: All \beta = 0 (no relationship) H_1: At least one \beta \neq 0 anova(model_full) ## Analysis of Variance Table
```

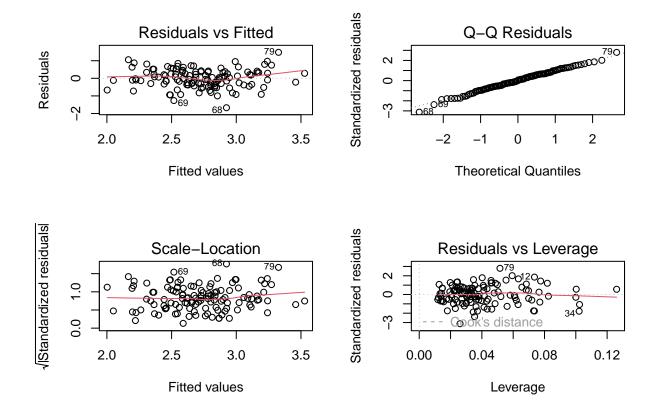
```
## Response: ai

## Property P
```

The F-statistic follows $F_{4,117}$ under H_0 and the p-value indicates strong evidence against H_0 . Thus, at least one predictor significantly explains variation in ai.

(d) Model Diagnostics

```
par(mfrow=c(2,2))
plot(model_full)
```



The residual plots suggest assumptions of linearity, normality, and homoscedasticity are reasonably met.

(e) R-squared

```
summary(model_full)$r.squared
```

[1] 0.2470586

About 24.7% of the variation in arousal index is explained by the predictors.

(f) Model Selection

```
step_model <- step(model_full)</pre>
```

```
## Start: AIC=-144.67
## ai ~ age + bmi + neck_size + sbp
```

```
##
##
               Df Sum of Sq
                                RSS
                                        AIC
## - bmi
                    0.22261 34.559 -145.88
                             34.337 -144.67
## <none>
## - neck size 1
                    2.39958 36.736 -138.43
## - sbp
                    2.42507 36.762 -138.35
                1
                    2.57977 36.916 -137.84
## - age
                1
##
## Step: AIC=-145.88
## ai ~ age + neck_size + sbp
##
               Df Sum of Sq
                                RSS
##
                                        AIC
## <none>
                             34.559 -145.88
                     2.2261 36.785 -140.27
## - sbp
                1
## - neck_size
                     2.6580 37.217 -138.84
                1
## - age
                1
                      2.7292 37.288 -138.61
```

summary(step_model)

```
##
## Call:
## lm(formula = ai ~ age + neck_size + sbp, data = sleep)
##
## Residuals:
##
       Min
                  1Q
                      Median
                                    3Q
                                       1.45627
## -1.65415 -0.35334 0.04008
                              0.37534
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.066166
                           0.506509 -0.131 0.89629
                0.009007
## age
                           0.002951
                                      3.053
                                            0.00280 **
## neck_size
                0.032630
                           0.010831
                                      3.013
                                            0.00317 **
                0.009579
                           0.003475
                                      2.757 0.00676 **
## sbp
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5412 on 118 degrees of freedom
## Multiple R-squared: 0.2422, Adjusted R-squared: 0.2229
## F-statistic: 12.57 on 3 and 118 DF, p-value: 3.452e-07
Final model: ai ~ age + neck_size + sbp
```

(g) Model Comparison

The adjusted R^2 of the reduced model is nearly identical to the full model, suggesting the excluded predictor (bmi) adds little value.

Question 2

(a) Balanced Design Check

```
energy <- read.csv("energy.csv")
table(energy$range, energy$menu)

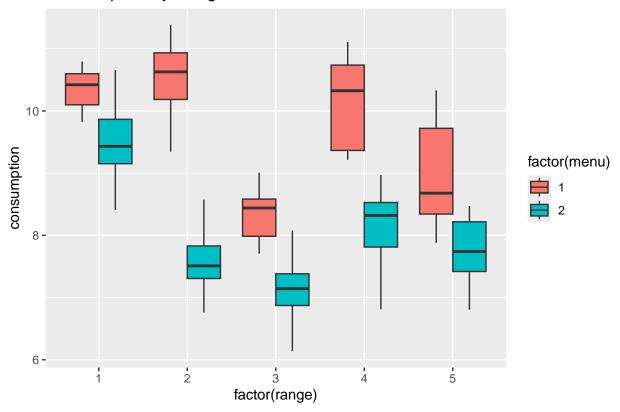
##
## 1 2
## 1 8 8
## 2 8 8
## 3 8 8
## 4 8 8
## 4 8 8
## 5 8 8</pre>
```

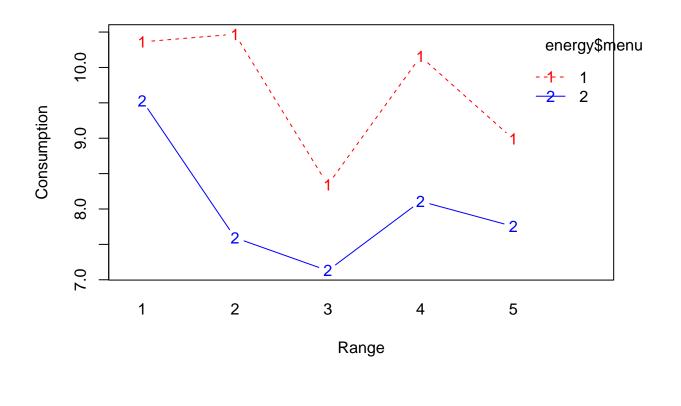
Each cell has equal observations \rightarrow the design is $\mathbf{balanced}.$

(b) Graphs

```
library(ggplot2)
ggplot(energy, aes(x=factor(range), y=consumption, fill=factor(menu))) +
    geom_boxplot(position=position_dodge()) +
    labs(title="Consumption by Range and Menu")
```

Consumption by Range and Menu





(c) Interaction Model

```
\begin{array}{l} consumption = \mu + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \varepsilon_{ijk} \\ \text{Where } \alpha_i = \text{range effect, } \beta_j = \text{menu effect, and } (\alpha\beta)_{ij} = \text{interaction.} \end{array}
```

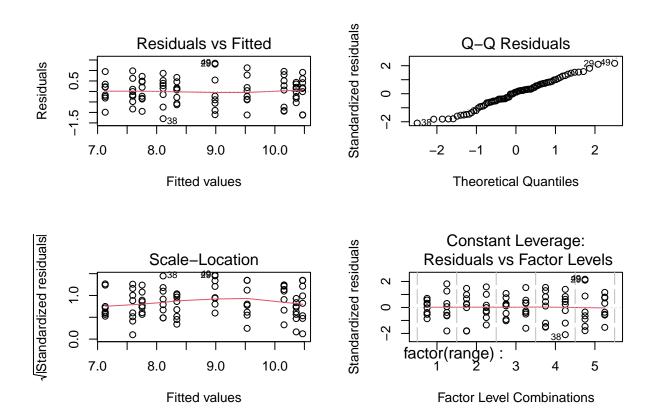
(d) ANOVA

```
interaction_model <- lm(consumption ~ factor(range)*factor(menu), data=energy)
anova(interaction_model)</pre>
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Significant interaction found (p < 0.001). Model diagnostics are below.

```
par(mfrow=c(2,2))
plot(interaction_model)
```



Conclusion: Both range, menu, and their interaction significantly affect energy consumption.