

Assignment 8

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```
exer <- read.csv("https://stats.idre.ucla.edu/stat/data/exer.csv")
## Convert variables to factor
exer <- within(exer, {
  diet <- factor(diet)
  exertype <- factor(exertype)
  time <- factor(time)
  id <- factor(id)
})

# Load Pacakage
library(ggplot2)
library(lme4)

## Warning: package 'lme4' was built under R version 4.4.3

## Loading required package: Matrix

library(nlme)

## Warning: package 'nlme' was built under R version 4.4.3

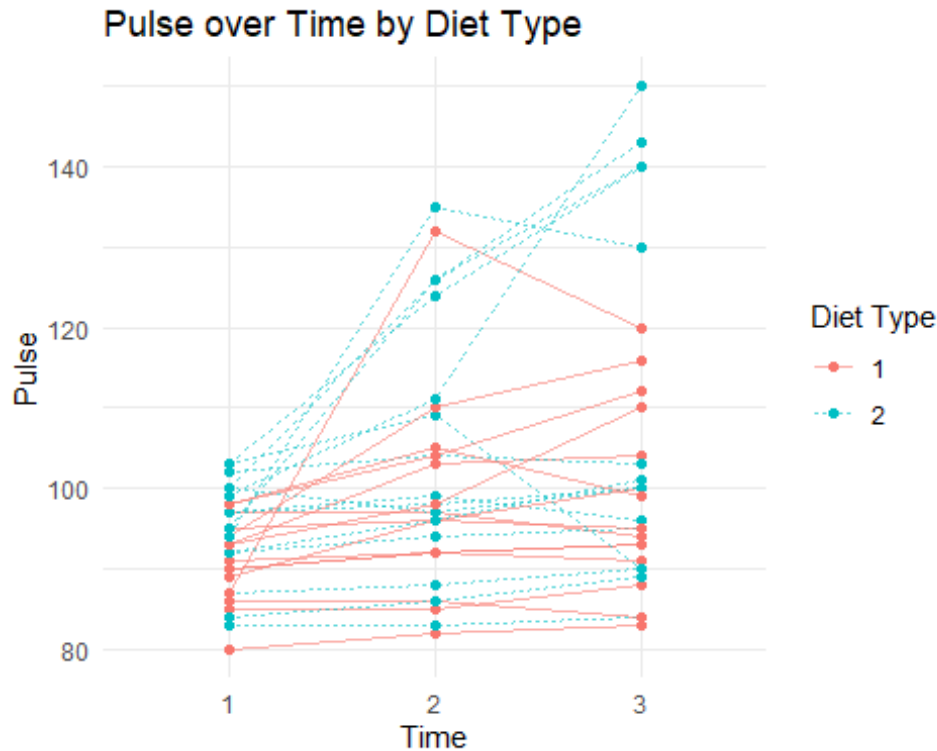
##
## Attaching package: 'nlme'

## The following object is masked from 'package:lme4':
##
##      lmList

library(Matrix)
```

Question 1:

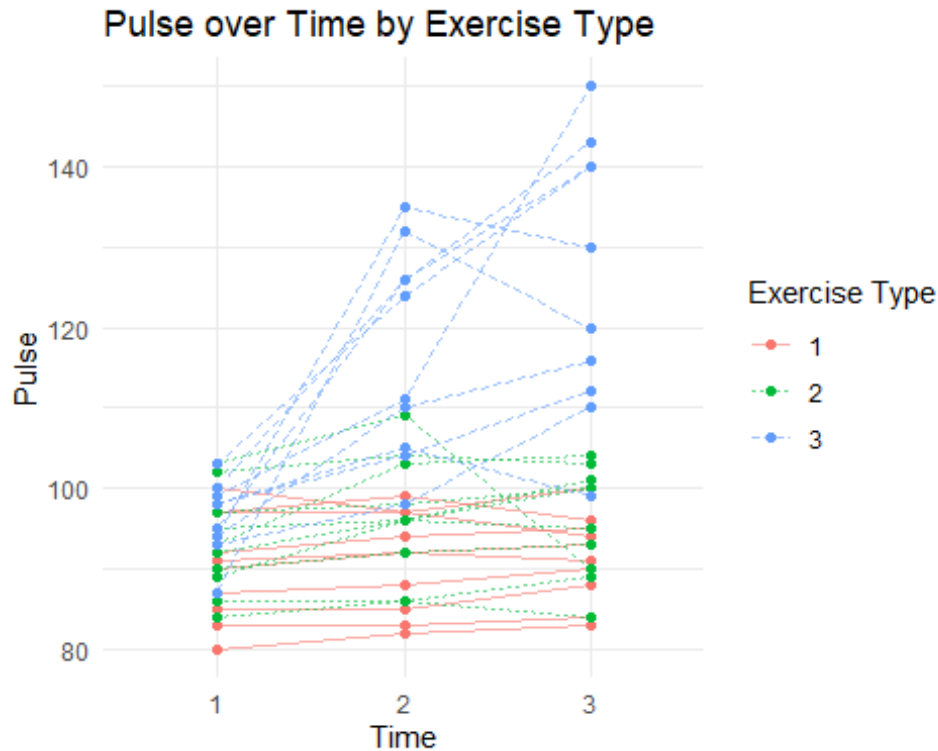
```
ggplot(exer, aes(x = time, y = pulse, group = id, color = diet, linetype = diet)) +
  geom_line(alpha = 0.6) +
  geom_point() +
  labs(title = "Pulse over Time by Diet Type",
       x = "Time",
       y = "Pulse",
       color = "Diet Type",
       linetype = "Diet Type") +
  theme_minimal()
```



Conclusion: Pulse tends to rise from Time 1 to Time 3, regardless of diet type. The blue dashed lines (Diet 2) show a more upward trend, especially between Time 2 and Time 3. There's greater spread in the pulse values for Diet 2, especially at Time 3, indicating more variation in individual responses. On the other hand, Diet Type 1 shows more clustered trends.

Question 2:

```
ggplot(exer, aes(x = time, y = pulse, group = id, color = exertype, linetype = exertype)) +
  geom_line(alpha = 0.6) +
  geom_point() +
  labs(title = "Pulse over Time by Exercise Type",
       x = "Time",
       y = "Pulse",
       color = "Exercise Type",
       linetype = "Exercise Type") +
  theme_minimal()
```



Conclusion: Participants in Exercise Type 3 show a sharp and consistent increase in pulse over time. Moreover, Exercise Type 2 shows a mild upward trend from Time 1 to Time 3. In addition, Exercise Type 1 shows minimal change in pulse over time.

Question 3:

```
# Fit linear regression model
lin <- lm(pulse ~ diet + exertype, data = exer)

# View summary of the model
summary(lin)

##
## Call:
## lm(formula = pulse ~ diet + exertype, data = exer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -22.811  -6.853   0.794   5.019  33.189
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   87.089     2.280   38.193 < 2e-16 ***
## diet2         7.489     2.280    3.284  0.00148 **
## exertype2      4.367     2.793    1.564  0.12158
## exertype3     22.233     2.793    7.961 6.34e-12 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.82 on 86 degrees of freedom
## Multiple R-squared:  0.488, Adjusted R-squared:  0.4701
## F-statistic: 27.32 on 3 and 86 DF, p-value: 1.668e-12
```

Question: 4

```
# Linear regression model (fixed effects only)
lm_model <- lm(pulse ~ diet + exertype, data = exer)
summary(lm_model)

##
## Call:
## lm(formula = pulse ~ diet + exertype, data = exer)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -22.811  -6.853   0.794   5.019  33.189
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   87.089      2.280  38.193 < 2e-16 ***
## diet2          7.489      2.280   3.284  0.00148 **
## exertype2      4.367      2.793   1.564  0.12158
## exertype3     22.233      2.793   7.961 6.34e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.82 on 86 degrees of freedom
## Multiple R-squared:  0.488, Adjusted R-squared:  0.4701
## F-statistic: 27.32 on 3 and 86 DF, p-value: 1.668e-12

# Linear mixed model with random intercept for id
lmer_model <- lmer(pulse ~ diet + exertype + (1 | id), data = exer)

## boundary (singular) fit: see help('isSingular')

summary(lmer_model)

## Linear mixed model fit by REML ['lmerMod']
## Formula: pulse ~ diet + exertype + (1 | id)
## Data: exer
##
## REML criterion at convergence: 666.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.10900 -0.63357  0.07345  0.46407  3.06847
##
```

```
## Random effects:
##   Groups   Name                Variance Std.Dev.
##   id       (Intercept)         0         0.00
##   Residual                    117        10.82
## Number of obs: 90, groups: id, 30
##
## Fixed effects:
##               Estimate Std. Error t value
## (Intercept)   87.089      2.280    38.193
## diet2          7.489      2.280     3.284
## exertype2      4.367      2.793     1.564
## exertype3     22.233      2.793     7.961
##
## Correlation of Fixed Effects:
##           (Intr) diet2  exrty2
## diet2      -0.500
## exertype2  -0.612  0.000
## exertype3  -0.612  0.000  0.500
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')
```

```
coef(lm_model)
```

```
## (Intercept)      diet2  exertype2  exertype3
##   87.088889    7.488889    4.366667    22.233333
```

```
coef(lmer_model)
```

```
## $id
##   (Intercept)      diet2 exertype2 exertype3
## 1    87.08889  7.488889  4.366667  22.23333
## 2    87.08889  7.488889  4.366667  22.23333
## 3    87.08889  7.488889  4.366667  22.23333
## 4    87.08889  7.488889  4.366667  22.23333
## 5    87.08889  7.488889  4.366667  22.23333
## 6    87.08889  7.488889  4.366667  22.23333
## 7    87.08889  7.488889  4.366667  22.23333
## 8    87.08889  7.488889  4.366667  22.23333
## 9    87.08889  7.488889  4.366667  22.23333
## 10   87.08889  7.488889  4.366667  22.23333
## 11   87.08889  7.488889  4.366667  22.23333
## 12   87.08889  7.488889  4.366667  22.23333
## 13   87.08889  7.488889  4.366667  22.23333
## 14   87.08889  7.488889  4.366667  22.23333
## 15   87.08889  7.488889  4.366667  22.23333
## 16   87.08889  7.488889  4.366667  22.23333
## 17   87.08889  7.488889  4.366667  22.23333
## 18   87.08889  7.488889  4.366667  22.23333
## 19   87.08889  7.488889  4.366667  22.23333
## 20   87.08889  7.488889  4.366667  22.23333
## 21   87.08889  7.488889  4.366667  22.23333
```

```
## 22      87.08889 7.488889 4.366667 22.23333
## 23      87.08889 7.488889 4.366667 22.23333
## 24      87.08889 7.488889 4.366667 22.23333
## 25      87.08889 7.488889 4.366667 22.23333
## 26      87.08889 7.488889 4.366667 22.23333
## 27      87.08889 7.488889 4.366667 22.23333
## 28      87.08889 7.488889 4.366667 22.23333
## 29      87.08889 7.488889 4.366667 22.23333
## 30      87.08889 7.488889 4.366667 22.23333
##
## attr(,"class")
## [1] "coef.mer"
```