P8106 Final Project: Predicting COVID-19 Recovery Time and Identifying Significant Risk Factors

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Background

[Check the report]

Data:

[Description check the report]

```
# For primary analysis:
# Dataset Loading:
load("data/recovery.Rdata")
set.seed(3521) # Runze Cui's uni(2183):
# Create a first random sample of 2000 participants:
dat1 <- dat[sample(1:10000, 2000),]</pre>
set.seed(3555) # Yuchen Hua's uni(3555)
# Create a second random sample of 2000 participants:
dat2 <- dat[sample(1:10000, 2000),]</pre>
# Merged the two datasets and remove repeated observations:
dat <- unique(rbind(dat1, dat2))</pre>
# Get rid of the id variable from the merged dataset and do the data cleaning:
dat <- dat %>%
 select(-id) %>%
 mutate(gender = as.factor(gender)) %>%
 mutate(race = as.factor(race)) %>%
 mutate(smoking = as.factor(smoking)) %>%
 mutate(hypertension = as.factor(hypertension)) %>%
 mutate(diabetes = as.factor(diabetes)) %>%
 mutate(vaccine = as.factor(vaccine)) %>%
 mutate(severity = as.factor(severity)) %>%
 mutate(study = as.factor(study)) %>%
 na.omit() %>%
 relocate(recovery_time)
head(dat)
       recovery_time age gender race smoking height weight bmi hypertension
## 8158
                 52 61 0 1 1 169.9 87.6 30.4
## 3387
                  24 60
                             1
                                  1
                                         2 173.4
                                                     70.6 23.5
                                                                         0
                  36 60
                            1
                                          1 178.2
                                                     79.9 25.1
                                                                         0
## 1709
                                  1
                            1
## 4051
                  23 70
                                  4
                                          0 167.4
                                                     77.7 27.7
                                                                         1
## 954
                  24 63
                                          0 175.4
                                                     88.7 28.8
                                                                         1
                 36 65
                            0
                                          0 160.4
                                                     74.4 28.9
## 531
                                  1
## diabetes SBP LDL vaccine severity study
          0 118 103 0
## 8158
                                   0
## 3387
             0 129 101
             0 130 107
## 1709
                             1
                                      0
                                            Α
## 4051
              0 145 128
                             1
                                            В
                              0
## 954
              0 131 100
                                      0
                                            Α
         0 137 153
## 531
```

```
# Separate the data as training and test data:
set.seed(3521)
# Specify rows of training data:
trRows <- createDataPartition(dat$recovery_time, p = 0.7, list = FALSE)
# Training data:
training <- dat[trRows, ]</pre>
## Covariates' matrix:
x <- model.matrix(recovery_time ~ ., dat)[trRows, -1]</pre>
## Response's vector:
y <- dat$recovery_time[trRows]</pre>
# Test data:
test <- dat[-trRows, ]</pre>
## Covariates' matrix:
x2 <- model.matrix(recovery_time ~ ., dat)[-trRows, -1]</pre>
## Response's vector:
y2 <- dat$recovery_time[-trRows]</pre>
# For secondary analysis:
dat2 <- dat %>%
  mutate(recovery_time = ifelse(recovery_time >= 30, 1, 0)) %>%
  mutate(recovery_time = as.factor(recovery_time))
# Training data:
training_sec <- dat2[trRows, ]</pre>
## Covariates' matrix:
x_sec <- model.matrix(recovery_time ~ ., dat2)[trRows, -1]</pre>
## Response's vector:
y_sec <- dat2$recovery_time[trRows]</pre>
# Test data:
test_sec <- dat2[-trRows, ]</pre>
## Covariates' matrix:
x2_sec <- model.matrix(recovery_time ~ ., dat2)[-trRows, -1]</pre>
## Response's vector:
y2_sec <- dat2$recovery_time[-trRows]</pre>
```

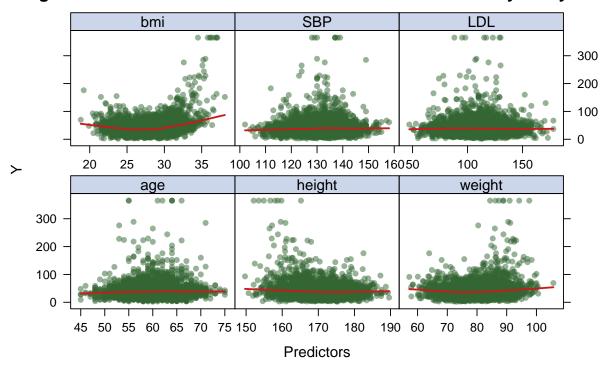
Exploratory Analysis and Data Visualization

[Description check the report]

```
# For primary analysis:
## For continuous variables:
theme = trellis.par.get()
theme$plot.symbol$col = rgb(.2, .4, .2, .5)
theme$plot.symbol$pch = 16
theme$plot.line$col = rgb(.8, .1, .1, 1)
theme$plot.line$lwd = 2
theme$strip.background$col = rgb(.0, .2, .6, .2)
trellis.par.set(theme)
```

```
featurePlot(x = dat %>% dplyr::select(age, height, weight, bmi, SBP, LDL),
    y = dat$recovery_time,
    plot = "scatter",
    span = .5,
    labels = c("Predictors","Y"),
    main = "Figure 1.1. Lattice Plots for Continuous Variables in Primary Analysis",
    type = c("p", "smooth"))
```

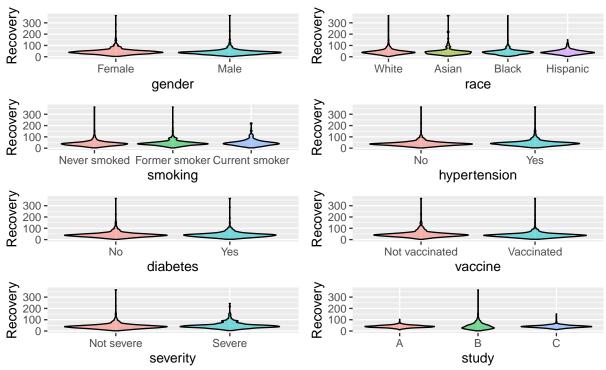
Figure 1.1. Lattice Plots for Continuous Variables in Primary Analysis

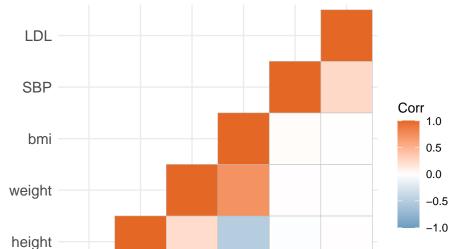


```
## For categorical variables:
gender plot = dat %>%
  ggplot(aes(x = gender, y = recovery_time, fill = gender)) +
  geom_violin(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Female','Male')) +
  vlab("Recovery") +
  theme(legend.position = "none")
race_plot = dat %>%
  ggplot(aes(x = race, y = recovery_time, fill = race)) +
  geom_violin(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('White','Asian','Black', 'Hispanic')) +
  vlab("Recovery") +
  theme(legend.position = "none")
smoking_plot = dat %>%
  ggplot(aes(x = smoking, y = recovery_time, fill = smoking)) +
 geom_violin(color = "black", alpha = .5) +
```

```
scale_x_discrete(labels = c('Never smoked', 'Former smoker', 'Current smoker')) +
  ylab("Recovery") +
  theme(legend.position = "none")
hyper_plot = dat %>%
  ggplot(aes(x = hypertension, y = recovery_time, fill = hypertension)) +
  geom_violin(color = "black", alpha = .5) +
 ylab("Recovery") +
  scale_x_discrete(labels = c('No','Yes')) +
  theme(legend.position = "none")
diabetes_plot = dat %>%
  ggplot(aes(x = diabetes, y = recovery_time, fill = diabetes)) +
  geom_violin(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('No','Yes')) +
  ylab("Recovery") +
  theme(legend.position = "none")
vac_plot = dat %>%
  ggplot(aes(x = vaccine, y = recovery_time, fill = vaccine)) +
  geom_violin(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Not vaccinated', 'Vaccinated')) +
 ylab("Recovery") +
 theme(legend.position = "none")
severity_plot = dat %>%
  ggplot(aes(x = severity, y = recovery_time, fill = severity)) +
  geom_violin(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Not severe','Severe')) +
 ylab("Recovery") +
 theme(legend.position = "none")
study_plot = dat %>%
  ggplot(aes(x = study, y = recovery_time, fill = study)) +
  geom_violin(color = "black", alpha = .5) +
  ylab("Recovery") +
 theme(legend.position = "none")
(gender_plot + race_plot + smoking_plot + hyper_plot) / (diabetes_plot + vac_plot + severity_plot + stu
  plot_layout(guides = "collect") +
 plot_annotation(title = "Figure 1.2. Violin Plots for Categorical Variables in Primary Analysis")
```







reight weight

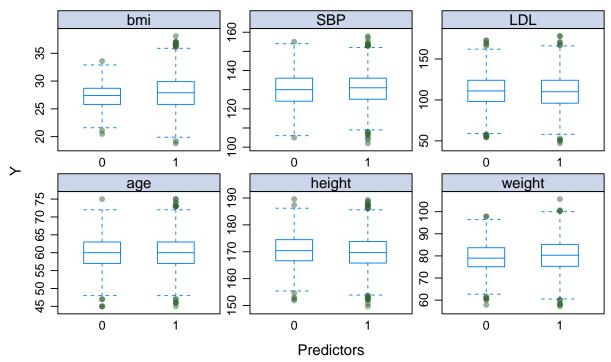
age

Figure 1.3. Correlation matrix for continuous datasat

```
# For secondary analysis:
## For continuous variables:
theme = trellis.par.get()
themeplot.symbol\\col = rgb(.2, .4, .2, .5)
themeplot.symbol\\pch = 16
themeplot.line col = rgb(.8, .1, .1, 1)
theme$plot.line$lwd = 2
theme$strip.background$col = rgb(.0, .2, .6, .2)
trellis.par.set(theme)
featurePlot(x = dat2 %>% dplyr::select(age, height, weight, bmi, SBP, LDL),
            y = dat2$recovery_time,
            plot = "box", pch = "|",
            scales = list(x = list(relation = "free"),
                         y = list(relation = "free")),
            labels = c("Predictors","Y"),
            main = "Figure 1.4. Lattice Plots for Continuous Variables in Secondary Analysis",
            auto.key = list(columns = 2))
```

D





```
## For categorical variables:
gender_plot_sec = dat2 %>%
  ggplot(aes(x = gender, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Female','Male')) +
  ylab("Recovery") +
  theme(legend.position = "none")
race_plot_sec = dat2 %>%
  ggplot(aes(x = race, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('White','Asian','Black', 'Hispanic')) +
  ylab("Recovery") +
  theme(legend.position = "none")
smoking_plot_sec = dat2 %>%
  ggplot(aes(x = smoking, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Never smoked', 'Former smoker', 'Current smoker')) +
  ylab("Recovery") +
  theme(legend.position = "none")
hyper_plot_sec = dat2 %>%
  ggplot(aes(x = hypertension, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  ylab("Recovery") +
  scale_x_discrete(labels = c('No','Yes')) +
```

```
theme(legend.position = "none")
diabetes_plot_sec = dat2 %>%
  ggplot(aes(x = diabetes, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('No','Yes')) +
 ylab("Recovery") +
 theme(legend.position = "none")
vac_plot_sec = dat2 %>%
  ggplot(aes(x = vaccine, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Not vaccinated','Vaccinated')) +
 ylab("Recovery") +
  theme(legend.position = "none")
severity_plot_sec = dat2 %>%
  ggplot(aes(x = severity, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
  scale_x_discrete(labels = c('Not severe','Severe')) +
 ylab("Recovery") +
 theme(legend.position = "none")
study_plot_sec = dat2 %>%
  ggplot(aes(x = study, fill = recovery_time)) +
  geom_bar(color = "black", alpha = .5) +
 ylab("Recovery") +
 theme(legend.position = "none")
(gender_plot_sec + race_plot_sec + smoking_plot_sec + hyper_plot_sec) / (diabetes_plot_sec + vac_plot_s
 plot_layout(guides = "collect") +
 plot_annotation(title = "Figure 1.5. Bar Plots for Categorical Variables in Secondary Analysis")
```

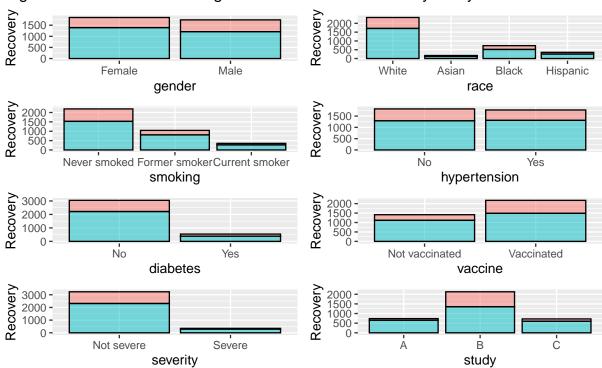


Figure 1.5. Bar Plots for Categorical Variables in Secondary Analysis

Note: red is 0 and blue is 1.

Primary Analysis

Recovery time as continuous variable.