

# STAT8111 Assignment1

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## Question 1

a.Examine first graphically and numerically correlation between variables, then comment :

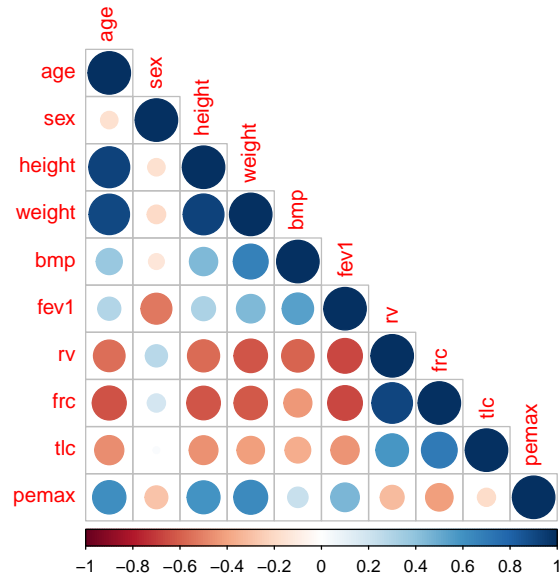
Numerical correlation

```
num_cor <- round(cor(data), 4)
num_cor
```

```
##          age      sex height weight      bmp      fev1      rv      frc      tlc
## age      1.0000 -0.1671  0.9261  0.9059  0.3778  0.2945 -0.5519 -0.6394 -0.4694
## sex     -0.1671  1.0000 -0.1675 -0.1904 -0.1376 -0.5283  0.2714  0.1836  0.0242
## height  0.9261 -0.1675  1.0000  0.9207  0.4408  0.3167 -0.5695 -0.6243 -0.4571
## weight  0.9059 -0.1904  0.9207  1.0000  0.6725  0.4488 -0.6215 -0.6173 -0.4185
## bmp      0.3778 -0.1376  0.4408  0.6725  1.0000  0.5455 -0.5824 -0.4344 -0.3649
## fev1     0.2945 -0.5283  0.3167  0.4488  0.5455  1.0000 -0.6659 -0.6651 -0.4430
## rv      -0.5519  0.2714 -0.5695 -0.6215 -0.5824 -0.6659  1.0000  0.9106  0.5891
## frc     -0.6394  0.1836 -0.6243 -0.6173 -0.4344 -0.6651  0.9106  1.0000  0.7044
## tlc     -0.4694  0.0242 -0.4571 -0.4185 -0.3649 -0.4430  0.5891  0.7044  1.0000
## pemax    0.6135 -0.2886  0.5992  0.6352  0.2295  0.4534 -0.3156 -0.4172 -0.1816
##          pemax
## age      0.6135
## sex     -0.2886
## height   0.5992
## weight   0.6352
## bmp      0.2295
## fev1     0.4534
## rv      -0.3156
## frc     -0.4172
## tlc     -0.1816
## pemax    1.0000
```

## Graphical correlation

```
corrplot(num_cor, type="lower")
```



Comment :

From the correlation plot, it can be seen that :

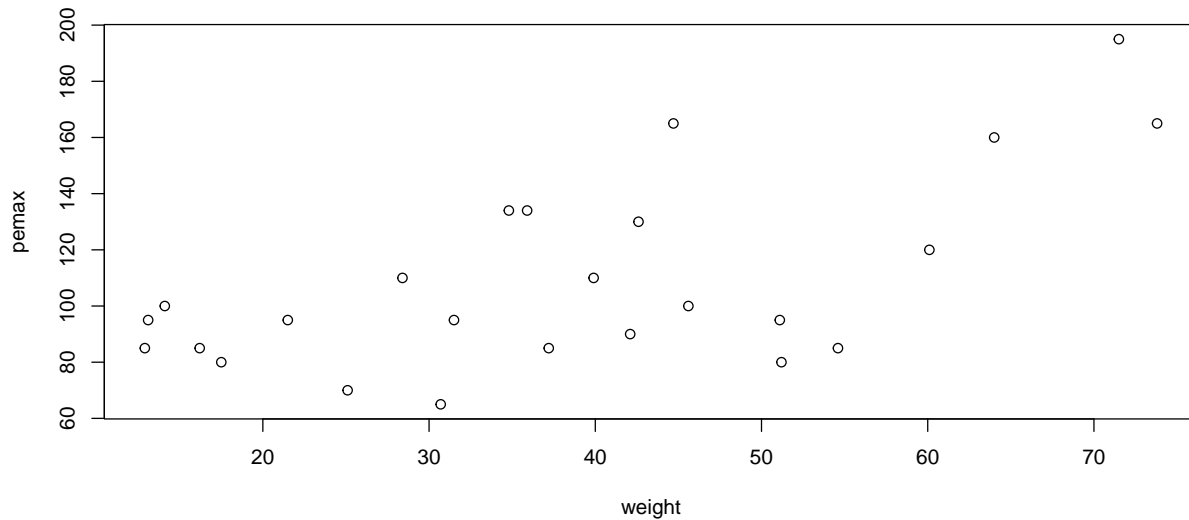
- There might be a strong positive linear relationship between `age` and `height`, `weight` and `pemax`. Beside that, `age` is negatively correlated with `frc`.
- `height` and `weight` are strongly correlated. In addition, `weight` has moderate negative correlation with `rv`, `frc`, `tlc` and positive correlation with `bmp`, `pemax`.
- Similarly, `fev1` has moderate negative correlation with `rv`, `frc`, `tlc`.
- Lastly, `rv` highly positively correlated with `frc`.

b. the relationship between `weight` and `pemax`

In this part, the relationship between **weight** and **pemax** will be examined. Specifically, **pemax** is the dependent variable (Y) and **weight** is the independent variable (X).

### Scatter Plot

```
plot(x = data$weight, y = data$pemax, xlab="weight", ylab = "pemax")
```



## Linear Model

```
model <- lm(pemax ~ weight, data = data)
summary(model)
```

```
##
## Call:
## lm(formula = pemax ~ weight, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44.30 -22.69   2.23  15.91  48.41
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  63.5456    12.7016   5.003 4.63e-05 ***
## weight       1.1867     0.3009   3.944 0.000646 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 26.38 on 23 degrees of freedom
## Multiple R-squared:  0.4035, Adjusted R-squared:  0.3776
## F-statistic: 15.56 on 1 and 23 DF, p-value: 0.0006457
```

The model equation  $\hat{pmax} = 63.5456 + 1.1867weight + \epsilon$

## Model Fit

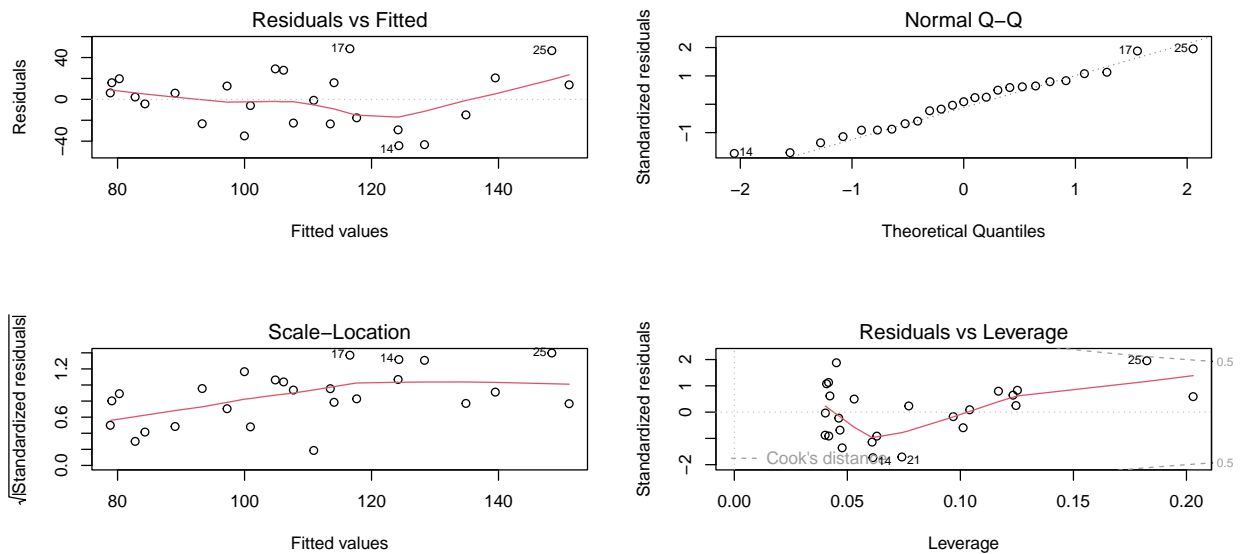
- The  $R^2 = 0.4035$  which means 40.35% of variation in **pemax** can be explained by **weight**. This show that the model is not fit well.

## Model interpretation

- According to the equation, for each unit increase in weight, the pemax will increase about 1.1867.
- **weight** is significant predictor since the p-value = 0.000646 ( $< 0.001$ ).

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

## Diagnostic



- The standardized Residuals versus Fitted values plot appears to be a random scatter about zero, so the model is adequate. This graph also shows some residuals which are 14, 17, 25 are low and high. This can be evidence of a small amount of heteroscedasticity.
- the Normal Q-Q plot is approximately linear, so it can be said that the normality assumption holds.

## c. Include in the previous model the sex variable

### Model 2

```
model_2 <- lm(pemax ~ weight + sex, data = data)
summary(model_2)
```

```
##
## Call:
## lm(formula = pemax ~ weight + sex, data = data)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -47.388 -16.850   0.073  13.168  43.748
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  70.9719    14.4644   4.907 6.61e-05 ***
## weight       1.1248     0.3056   3.681 0.00131 **
## sex         -11.4776    10.7963  -1.063 0.29926
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 26.31 on 22 degrees of freedom
## Multiple R-squared:  0.4327, Adjusted R-squared:  0.3811
## F-statistic: 8.388 on 2 and 22 DF,  p-value: 0.00196
```

### The model 2 equation

$$p\hat{m}ax = 70.9719 + 1.1248weight + (-11.4776)sex + \epsilon$$

### Model 3

```
model_3 <- lm(pemax ~ sex + weight, data = data)
summary(model_3)
```

```
##
## Call:
## lm(formula = pemax ~ sex + weight, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -47.388 -16.850   0.073  13.168  43.748
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  70.9719    14.4644   4.907 6.61e-05 ***
## sex         -11.4776    10.7963  -1.063 0.29926
## weight       1.1248     0.3056   3.681 0.00131 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 26.31 on 22 degrees of freedom
## Multiple R-squared:  0.4327, Adjusted R-squared:  0.3811
## F-statistic: 8.388 on 2 and 22 DF,  p-value: 0.00196
```

### The model 3 equation

$$p\hat{m}ax = 70.9719 + (-11.4776)sex + 1.1248weight + \epsilon$$

### Analyse the two proposed models

- Model\_2 and model\_3 are seem to be similar. The only difference is the order of **weight** and **sex**. The  $R^2$  of both model are 0.4327 which mean 43.27% of variant in **pmax** can be explained by **weight** and **sex**. On the one hand, in both two model, **sex** is insignificant predictor with p-value = 0.29926. On the other hand, **weight** is still significant predictor.
- For one unit increase in **weight** , **pmax** will increase 1.1248. The coefficient of **sex** represents the difference in **pmax** between females and males, while **weight** is same. In this case, models indicate that, on average, females have a **pmax** that is lower by 11.4776 units compared to males.

In conclusion, it appears that the order of variables (**weight** and **sex**) doesn't affect the results. Model\_2 and model\_3 are better than the first model on question b but these two still are not good model.

### Choose one model

In comparison of three models, model 2 and model 3 are appear to explain **pmax** better with higher  $R^2$  and adjusted  $R^2$ . Beside that, there are no multicollinearity issue between **weight** and **sex**. Therefore, I choose model 2.

**d. Construct a statistical model for the response variables pmax based on the normal response**

**distribution and the weight, bmp, fev1, rv, frc.**