

Lab 3 - MLR: Part 1

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```
library(readr)
```

```
## Warning: package 'readr' was built under R version 4.1.3
```

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 4.1.3
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.6      v dplyr   1.0.9
## v tibble  3.1.7      v stringr 1.4.0
## v tidyr   1.2.0      v forcats 0.5.1
## v purrr   0.3.4
```

```
## Warning: package 'ggplot2' was built under R version 4.1.3
```

```
## Warning: package 'tibble' was built under R version 4.1.3
```

```
## Warning: package 'tidyr' was built under R version 4.1.3
```

```
## Warning: package 'purrr' was built under R version 4.1.3
```

```
## Warning: package 'dplyr' was built under R version 4.1.3
```

```
## Warning: package 'stringr' was built under R version 4.1.3
```

```
## Warning: package 'forcats' was built under R version 4.1.3
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
teamperf <- read_csv("teamperf.csv")
```

```
## Rows: 23 Columns: 5
```

```
## -- Column specification -----
## Delimiter: ","
## dbl (5): team, intrapersonal_range, stress_range, mood_range, project_score
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

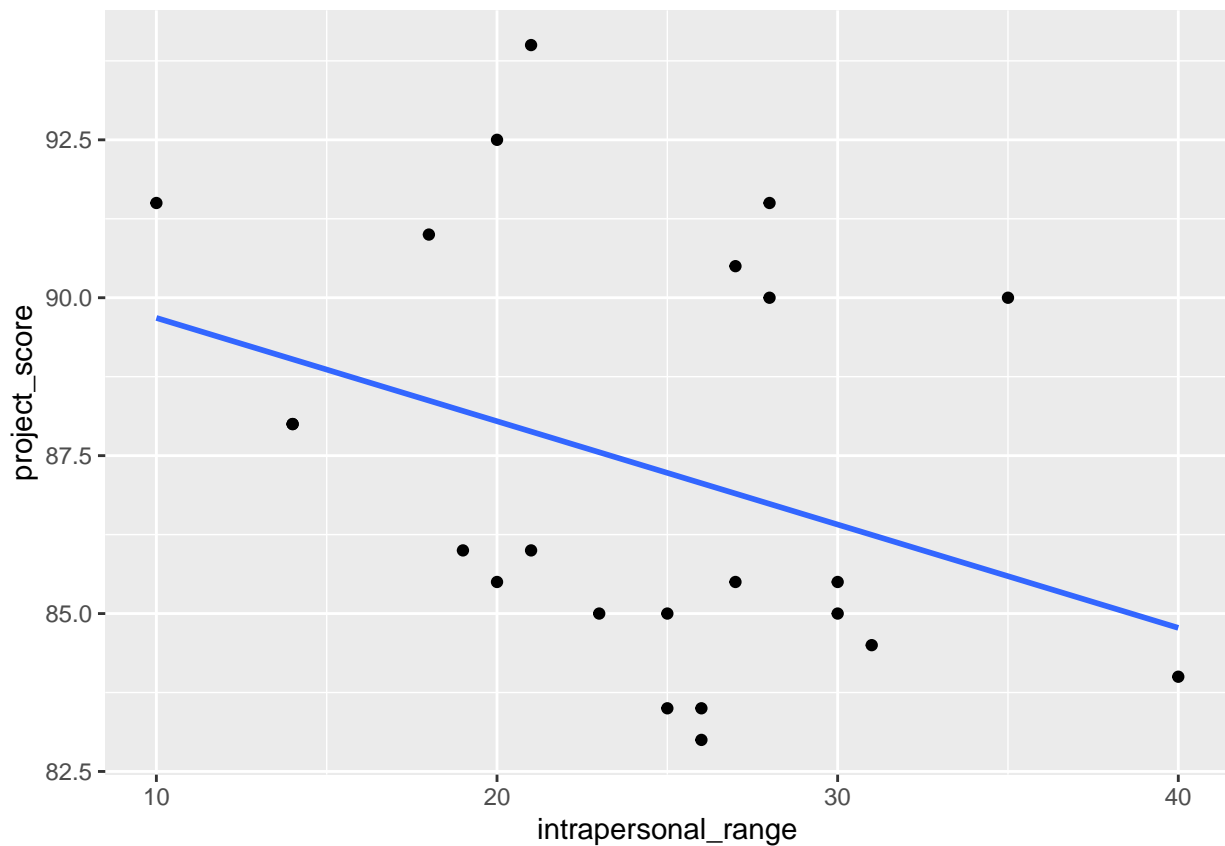
(a) Hypothesize a first-order model for project scores y as a function of x_1 , x_2 and x_3

Our dependent variable(y) is mean project score. Three independent variables for each team: range of intra-personal scores (x_1), range of stress management (x_2),and range of mood scores (x_3)

(b) Use a appropriate plots to verify a linear relationship

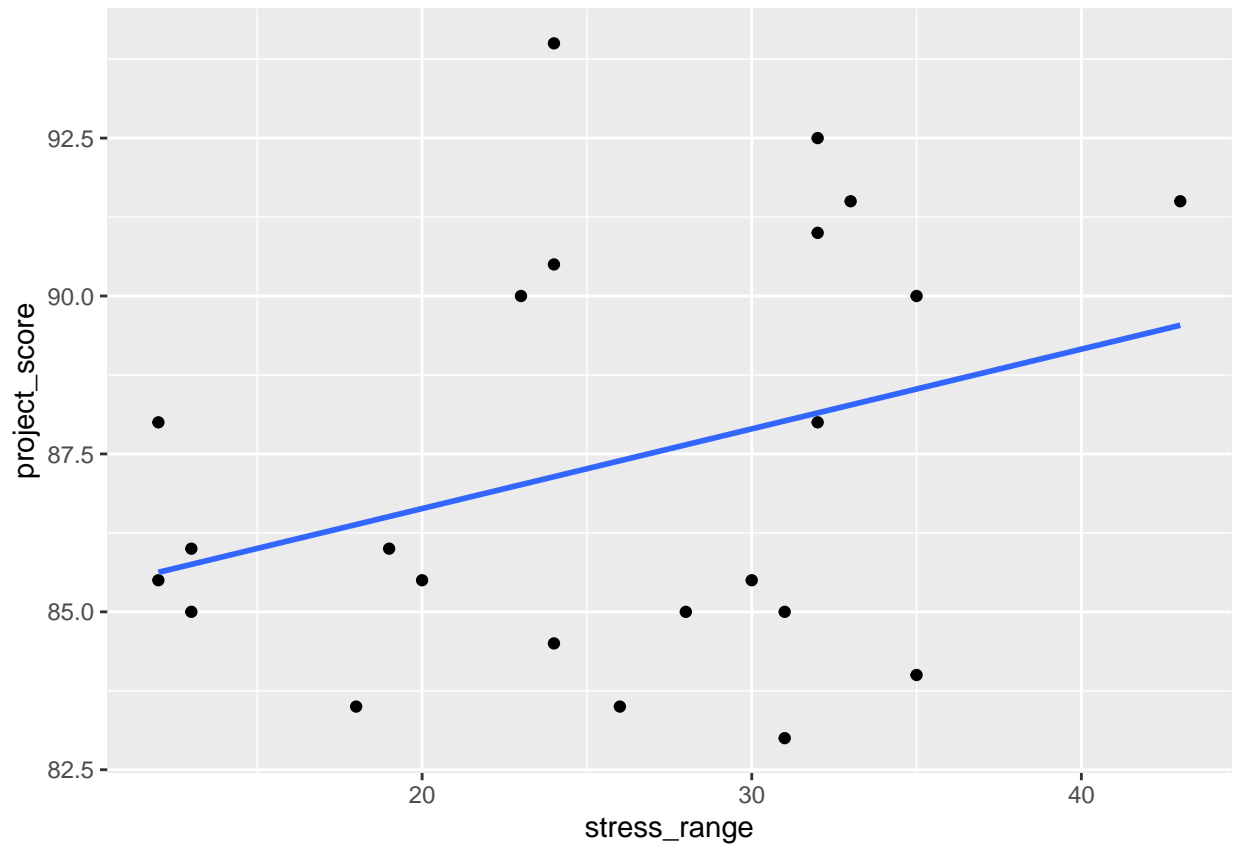
```
ggplot(data = teamperf, mapping = aes(x = intrapersonal_range, y = project_score))+  
  geom_point()+  
  geom_smooth(method='lm', se = FALSE)
```

'geom_smooth()' using formula 'y ~ x'



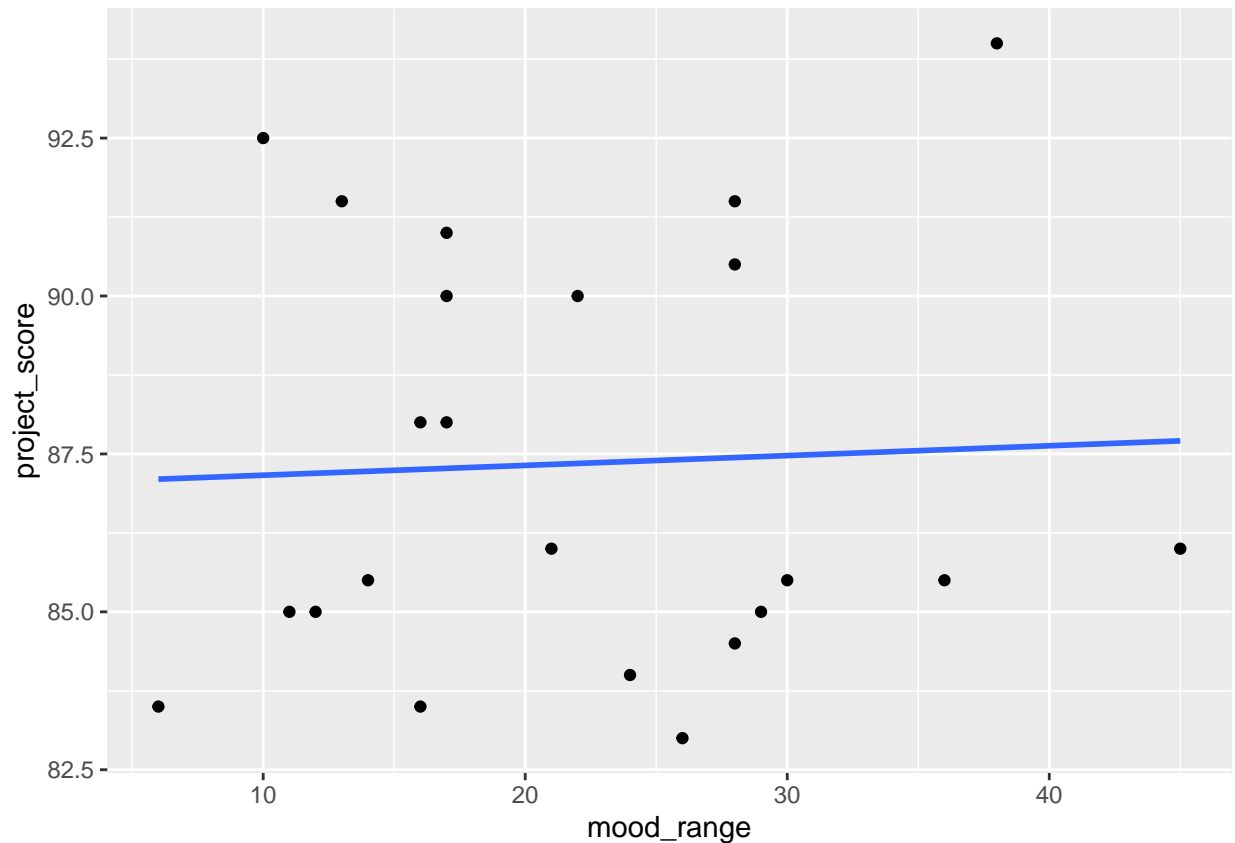
```
ggplot(data = teamperf, mapping = aes(x = stress_range, y = project_score))+  
  geom_point()+  
  geom_smooth(method='lm', se = FALSE)
```

'geom_smooth()' using formula 'y ~ x'



```
ggplot(data = teamperf, mapping = aes(x = mood_range, y = project_score))+  
  geom_point()+  
  geom_smooth(method='lm', se = FALSE)
```

```
## 'geom_smooth()' using formula 'y ~ x'
```



(c) Fit the model in part (a), to the data using R

```
model = lm(project_score ~ intrapersonal_range + stress_range + mood_range, data = teamperf)
coef(model)
```

```
##          (Intercept) intrapersonal_range      stress_range      mood_range
##          86.64143382      -0.19937014          0.15605653          0.07028216
```

(d) Use matrices to verify the equations and produce the same coefficients.

```
n = nrow(teamperf)
p = length(coef(model))

X = cbind(rep(1, n), teamperf$intrapersonal_range, teamperf$stress_range, teamperf$mood_range)
y = teamperf$project_score
beta_hat = solve(t(X) %*% X) %*% t(X) %*% y
beta_hat
```

```
##          [,1]
## [1,] 86.64143382
## [2,] -0.19937014
```

```
## [3,] 0.15605653
## [4,] 0.07028216
```

(e) Is there sufficient evidence to indicate the overall model is statistically useful for predicting y ? Test using $\alpha = 0.10$.

```
summary(model)
```

```
##
## Call:
## lm(formula = project_score ~ intrapersonal_range + stress_range +
##     mood_range, data = teamperf)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1229 -1.8920 -0.8116  1.7606  5.1293
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    86.64143     3.26033   26.574  <2e-16 ***
## intrapersonal_range -0.19937     0.09537   -2.090   0.0503 .
## stress_range      0.15606     0.07835    1.992   0.0610 .
## mood_range       0.07028     0.06845    1.027   0.3174
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.023 on 19 degrees of freedom
## Multiple R-squared:  0.2816, Adjusted R-squared:  0.1681
## F-statistic: 2.482 on 3 and 19 DF,  p-value: 0.09204
```

The p-value for our model is 0.09204 which is less than 0.10, means that our model is statistically significant.

(f) Evaluate the model using statistics R^2 adjusted and 2s.

```
summary(model)
```

```
##
## Call:
## lm(formula = project_score ~ intrapersonal_range + stress_range +
##     mood_range, data = teamperf)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.1229 -1.8920 -0.8116  1.7606  5.1293
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    86.64143     3.26033   26.574  <2e-16 ***
## intrapersonal_range -0.19937     0.09537   -2.090   0.0503 .
```

```
## stress_range      0.15606    0.07835    1.992    0.0610 .
## mood_range        0.07028    0.06845    1.027    0.3174
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.023 on 19 degrees of freedom
## Multiple R-squared:  0.2816, Adjusted R-squared:  0.1681
## F-statistic: 2.482 on 3 and 19 DF,  p-value: 0.09204
```

```
2 * 3.023
```

```
## [1] 6.046
```

Adjusted R-squared for our model is 0.1681. Adjusted R² tells us the percentage of variation explained by only the independent variables that actually affect the dependent variable which is quite less in our case. Here, We can expect 95% of the observed values of y to lie within 2s (2 standard deviations away) of y, i.e.; 6.046

(g) Find and interpret a 95% confidence interval for y when $x_1 = 20$, $x_2 = 30$ and $x_3 = 25$.

```
new_team = data.frame(intrapersonal_range = 20, stress_range = 30, mood_range = 25)
new_team
```

```
##   intrapersonal_range stress_range mood_range
## 1                   20             30         25
```

```
predict(model, newdata = new_team, interval = "confidence", level = 0.95)
```

```
##           fit          lwr          upr
## 1 89.09278 87.18764 90.99792
```

(h) Find and interpret a 95% prediction interval for y when $x_1 = 20$, $x_2 = 30$ and $x_3 = 25$.

```
new_team1 = data.frame(intrapersonal_range = 20, stress_range = 30, mood_range = 25)
new_team1
```

```
##   intrapersonal_range stress_range mood_range
## 1                   20             30         25
```

```
predict(model, newdata = new_team1, interval = "prediction", level = 0.95)
```

```
##           fit          lwr          upr
## 1 89.09278 82.48456 95.701
```