# Data Analysis and Lab\_Lab 3

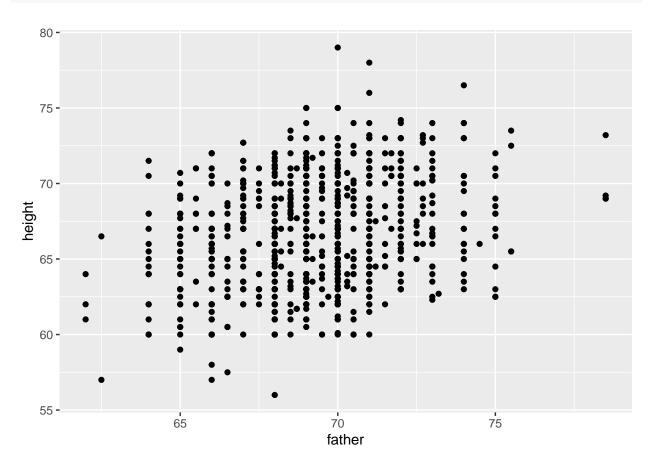
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# Exercise 1

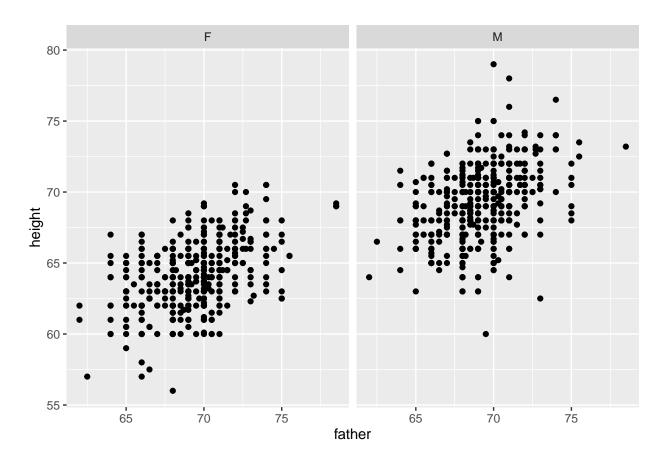
(a) Create a scatter plot of each person's height against their father's height

```
ggplot(data = Galton) +
geom_point(mapping = aes(x = father, y = height))
```



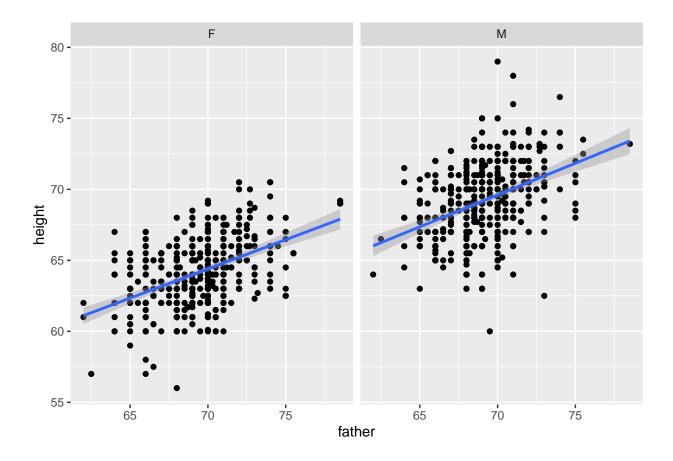
## (b) Separate your plot into facets by sex

```
ggplot(data = Galton) +
geom_point(mapping = aes(x = father, y = height)) +
facet_wrap(~sex)
```



(c) Add regression lines to all of your facets; use geom\_smooth(method = "lm").

```
ggplot(data = Galton) +
  geom_point(mapping = aes(x = father, y = height)) +
  facet_wrap(~sex) +
  geom_smooth(mapping = aes(x = father, y = height), method = "lm", formula = y ~ x)
```



(d) Analyze the results of (a), (b) and (c) in 2 ~ 3 sentences.

산점도상 자녀의 키와 아버지의 키는 그래프로 봤을 때 양의 상관관계를 가지는 것으로 보인다. 이와 같은 결과는 '남성과 여성의 키 차이를 고려'하여 자녀의 성별에 따라 서로 다른 facet에 각 그래프를 그려도 동일하게 나타난다. 실제 이에 따라 회귀직선을 mapping해 본 경우에도 비슷한 결과가 나온다.

#### Exercise 2

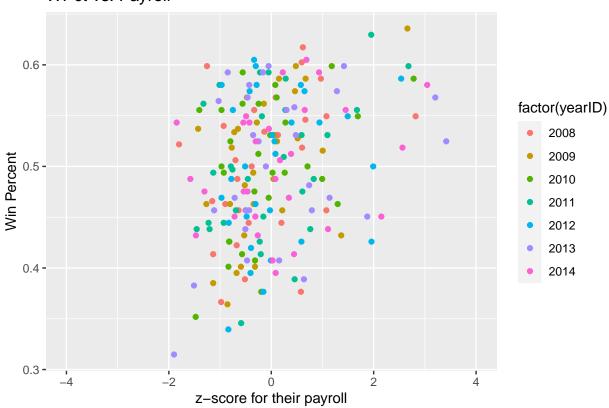
```
MLB_teams
```

(a) See how many variables you can illustrate on a single plot in R. The current record is 7. [Note: this is not good graphical practice—it is merely an exercise to help you understand how to use visual cues and aesthetics!]

```
MLB_teams %>%
  mutate(z_pay = (payroll-mean(payroll))/sd(payroll)) %>%
  ggplot() +
  geom_point(mapping = aes(x = z_pay, y = WPct, color = factor(yearID))) +
  xlim(-4,4) +
```

```
xlab('z-score for their payroll') +
ylab('Win Percent') +
labs(title = 'WPct vs. Payroll')
```

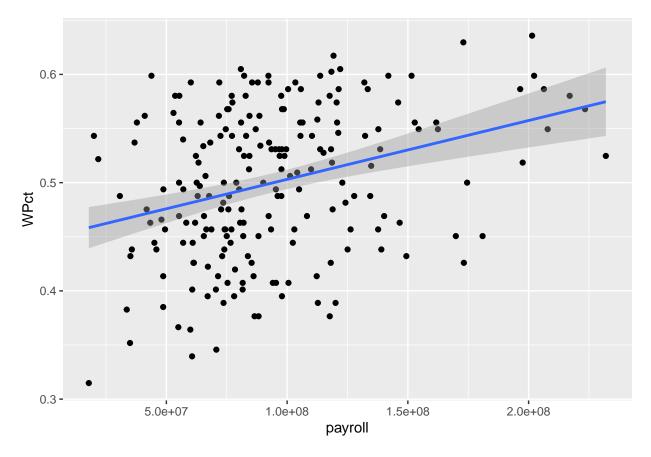
# WPct vs. Payroll



(b) Use the MLB\_teams data in the mdsr package to create an informative data graphic that illustrates the relationship between winning percentage WPct and payroll payroll in context.

```
ggplot(data = MLB_teams, mapping = aes(x = payroll, y = WPct)) +
  geom_point() +
  geom_smooth(method = "lm")
```

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



Overall there is a positive correlation between payroll and WPct. However, overall payroll tends to increase as time goes.

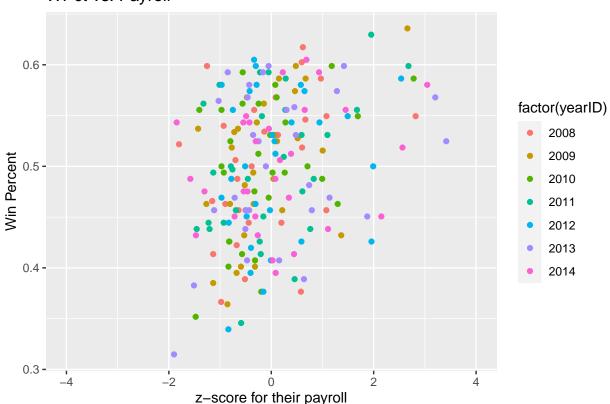
```
MLB_teams %>%
  group_by(yearID) %>%
  summarise(sum(payroll))
## # A tibble: 7 x 2
     yearID `sum(payroll)`
##
      <int>
                      <dbl>
##
       2008
                2684858670
## 1
## 2
       2009
                2664726994
## 3
       2010
                2721359865
## 4
       2011
                2784505291
## 5
       2012
                2932741192
## 6
       2013
                3034525648
## 7
       2014
                2994000466
```

So, we will use z-score. In z-scores, we can see teams below -2.0 did not exist, but teams with z-scores over 3.0 was 3 teams.

```
MLB_teams %>%
  mutate(z_pay = (payroll-mean(payroll))/sd(payroll)) %>%
  ggplot() +
  geom_point(mapping = aes(x = z_pay, y = WPct, color = factor(yearID))) +
```

```
xlim(-4,4) +
xlab('z-score for their payroll') +
ylab('Win Percent') +
labs(title = 'WPct vs. Payroll')
```

## WPct vs. Payroll



Exercises 3: Write a code to create a data object named Binary\_medv whose value is "rich" if the value of medv is greater than 25, "not so" if not. Use both with() and ifelse().

```
my_Boston <- Boston %>%
 mutate(Binary_medv = with(Boston, ifelse(medv > 25, 'rich', 'not so'))) %>%
  relocate(Binary_medv)
head(my_Boston)
##
    Binary_medv
                   crim zn indus chas
                                                        dis rad tax ptratio
                                      nox
                                             rm age
## 1
       not so 0.00632 18 2.31 0 0.538 6.575 65.2 4.0900
                                                             1 296
                                                                      15.3
         not so 0.02731 0 7.07
                                                             2 242
## 2
                                  0 0.469 6.421 78.9 4.9671
                                                                      17.8
## 3
           rich 0.02729 0 7.07
                                  0 0.469 7.185 61.1 4.9671
                                                             2 242
                                                                      17.8
          rich 0.03237 0 2.18
                                  0 0.458 6.998 45.8 6.0622 3 222
## 4
                                                                      18.7
## 5
          rich 0.06905 0 2.18 0 0.458 7.147 54.2 6.0622 3 222
                                                                      18.7
           rich 0.02985 0 2.18 0 0.458 6.430 58.7 6.0622 3 222
## 6
                                                                      18.7
```