## **Simple Linear Regression**

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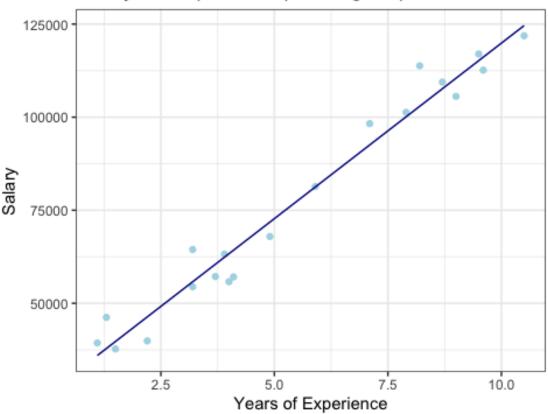
I will be building a simple linear regression model based off the 'Salary\_Data' data. In this dataset there are two columns, *Salary*, and *YearsExperience*. The YearsExperience is our dependent variable while the Salary is our independent variable. We want to determine if there are any correlations between profit and experience. Additionally, we want to determine if there is a linear dependency.

Simple Linear Regression Formula:

$$y = b_0 + b_1 x_1$$

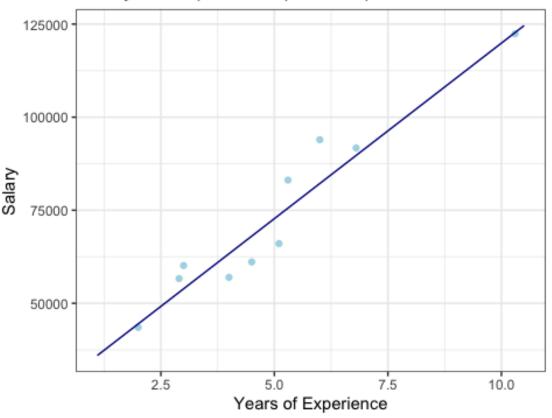
```
# Set seed
set.seed(1)
# Importing salary data
salary <- read.csv('.../.../data/Salary Data.csv')</pre>
# Splitting the dataset into the training set and test set
split <- sample.split(salary$Salary, SplitRatio = 2/3)</pre>
training.set <- subset(salary, split == TRUE)</pre>
test.set <- subset(salary, split == FALSE)</pre>
# Fitting simple linear regression to the training set
regressor <- lm(formula = Salary ~ YearsExperience,
                data = training.set)
# Predicting the test set results
y.pred <- predict(regressor, newdata = test.set)</pre>
# Visualizing the training set results
ggplot() +
  geom_point(aes(x = training.set$YearsExperience, y = training.set$Salary),
col = 'lightblue') +
  geom_line(aes(x = training.set$YearsExperience, y = predict(regressor,
newdata = training.set)), col = 'darkblue') +
  theme bw() +
  ggtitle('Salary VS Experience (Training Set)') +
  xlab('Years of Experience') +
 ylab('Salary')
```

## Salary VS Experience (Training Set)



```
# Visualizing the test set results
ggplot() +
    geom_point(aes(x = test.set$YearsExperience, y = test.set$Salary), col =
'lightblue') +
    geom_line(aes(x = training.set$YearsExperience, y = predict(regressor,
newdata = training.set)), col = 'darkblue') +
    theme_bw() +
    ggtitle('Salary VS Experience (Test Set)') +
    xlab('Years of Experience') +
    ylab('Salary')
```

## Salary VS Experience (Test Set)



## Conclusion

```
summary(regressor)
##
## Call:
## lm(formula = Salary ~ YearsExperience, data = training.set)
##
## Residuals:
               1Q Median
                               3Q
                                      Max
## -7522.2 -3584.8 -598.9 2187.8 10888.3
##
## Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   25594.7
                               2546.6
                                        10.05 8.26e-09 ***
## YearsExperience
                    9430.4
                                407.7
                                      23.13 7.71e-15 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 5483 on 18 degrees of freedom
## Multiple R-squared: 0.9675, Adjusted R-squared: 0.9656
## F-statistic: 535.1 on 1 and 18 DF, p-value: 7.707e-15
```

#