# **Decision Tree Regression**

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I will be building a decision tree regression model based off the 'Position\_Salaries' data to determine 1.) determine if the model fits the data and 2.) to determine if a particular new hire's past salary was possibly \$160,000 as a region manager. In this dataset there are three columns *Position*, *Level*, and *Salary*. Salary is our dependent variable while the other two are our independent variables.

### **Preparing the Data**

```
# Set the seed
set.seed(1)
# Importing the data
positions <- read.csv('../../data/Position_Salaries.csv')</pre>
# Examine the Data
dim(positions)
## [1] 10 3
positions
##
              Position Level Salary
                               45000
## 1
      Business Analyst 1
## 2 Junior Consultant
                           2 50000
## 3 Senior Consultant
                           3 60000
## 4
               Manager
                           4 80000
                           5 110000
## 5
       Country Manager
## 6
        Region Manager
                           6 150000
                           7 200000
## 7
               Partner
## 8
        Senior Partner
                           8 300000
## 9
               C-level
                           9 500000
                          10 1000000
## 10
                   CEO
```

From the table we can see there is some redundancy between the *Position* and *Level* column. Therefor it would make sense to drop the *Position* column and just use the numeric *Level* and *Salary* columns. Since we only have 10 observations, it would not be useful to split the data into a training and test set.

```
# Saving the dataset with only the two necessary columns
positions <- positions[, 2:3]</pre>
```

#### **Decision Tree Regressor**

#### **Predicting a Result**

Is it likely that the new hire's past salary was actually \$160,000 as a level 6.5?

```
y.pred <- predict(regressor, data.frame(Level = 6.5))
y.pred
## 1
## 250000</pre>
```

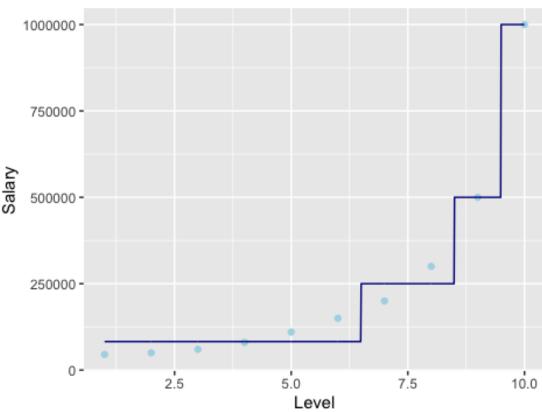
### **Visualizing the Decision Tree Model**

Visualizing the data will allow us to see if the decision tree regression is a good model for the data. It is important to note this is a non-continuos model.

```
# Increase the resolution
x_grid = seq(min(positions$Level), max(positions$Level), 0.01)

# Visualizing the Decision Tree Regression
ggplot() +
    geom_point(data = positions, aes(x = Level, y = Salary), col = 'lightblue')
+
    geom_line(aes(x = x_grid, y = predict(regressor, newdata = data.frame(Level
= x_grid))), col = 'darkblue') +
    ggtitle('Decision Tree Model') +
    xlab('Level') +
    ylab('Salary')
```





# Conclusion

The decision tree regression model is considering the average in each of the split intervals.