

Random Forest Regression

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I will be building a random forest 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')

# Examine the Data
dim(positions)

## [1] 10  3

positions

##           Position Level  Salary
## 1 Business Analyst      1  45000
## 2 Junior Consultant      2  50000
## 3 Senior Consultant      3  60000
## 4           Manager      4  80000
## 5 Country Manager      5 110000
## 6   Region Manager      6 150000
## 7           Partner      7 200000
## 8   Senior Partner      8 300000
## 9           C-level      9 500000
## 10          CEO      10 1000000
```

From the table we can see there is some redundancy between the *Position* and *Level* column. Therefore 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]
```

Decision Tree Regressor

```
set.seed(1234)
regressor <- randomForest(x = positions[1],
                          y = positions$Salary,
                          ntree = 1000)
```

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

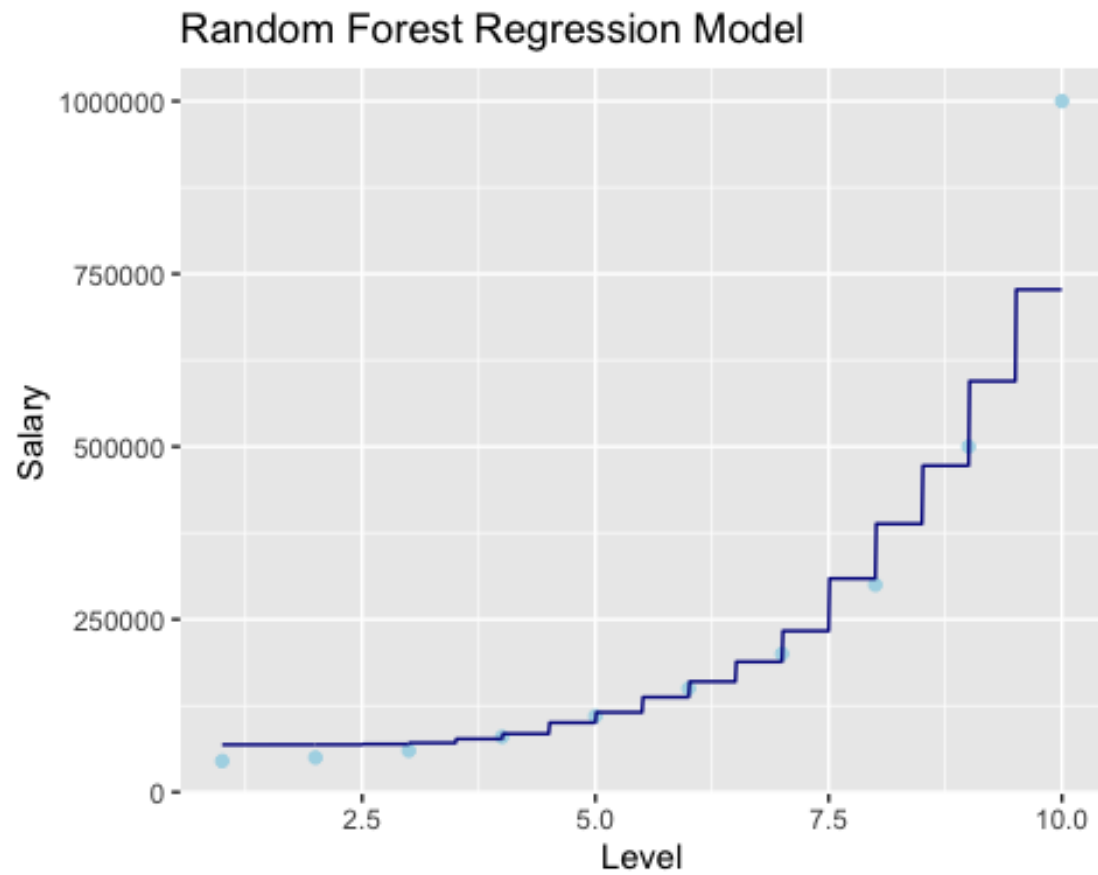
##          1
## 159894.2
```

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-continuous model.

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

# Visualizing the random forest 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('Random Forest Regression Model') +
  xlab('Level') +
  ylab('Salary')
```



Conclusion

The random forest regression model is considering the average in each of the split intervals. This model represents the data much better than the decision tree model.