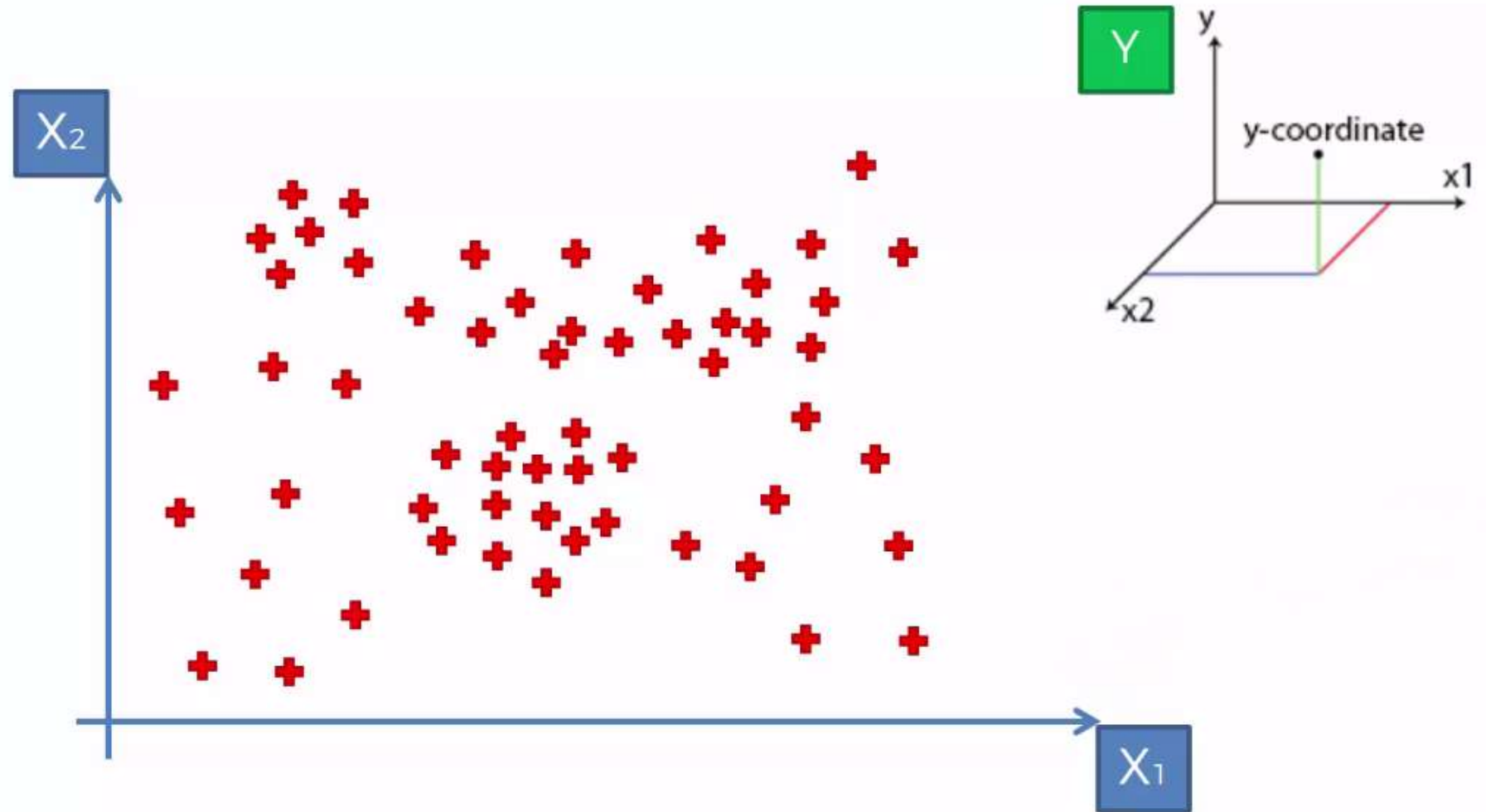




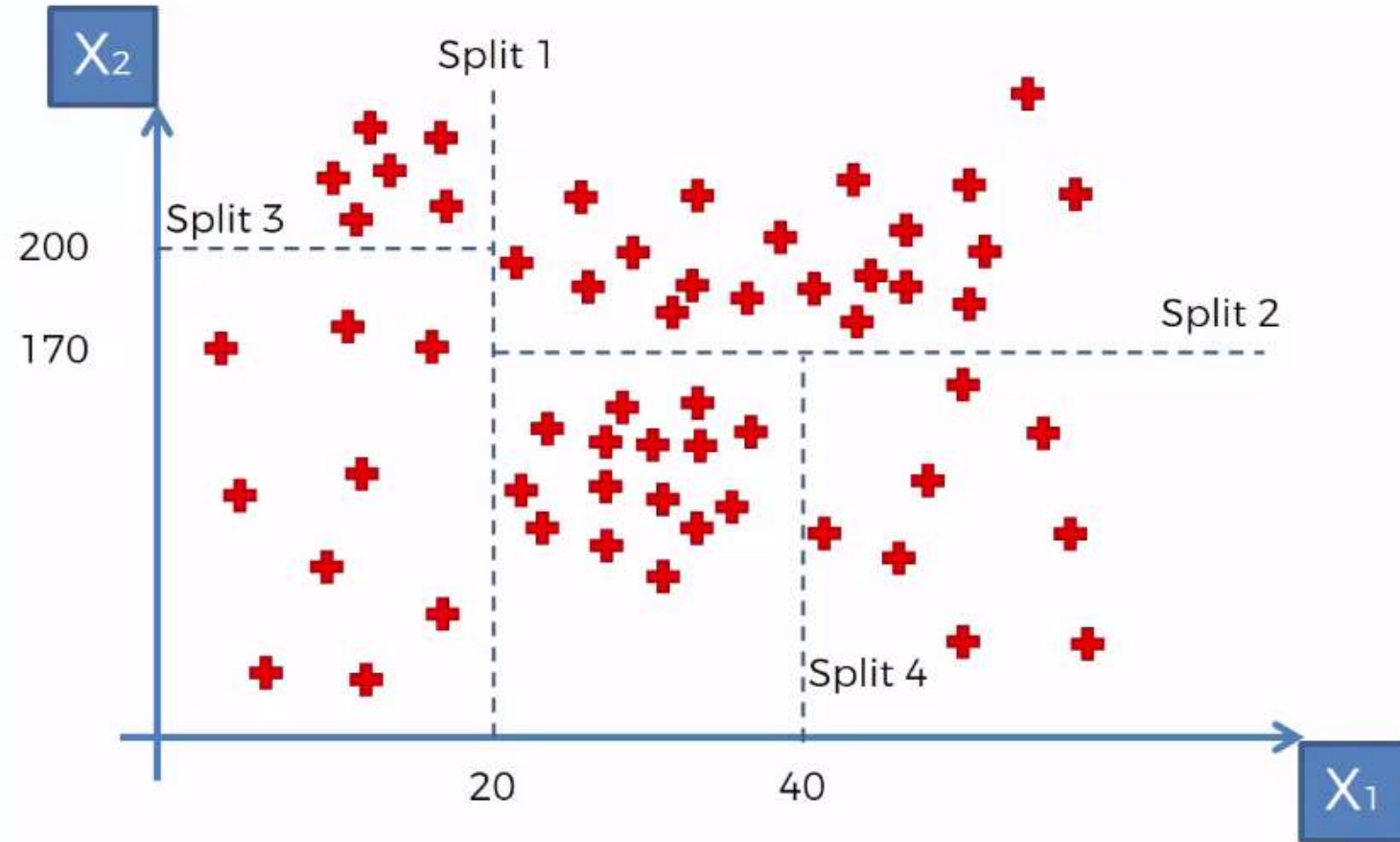
# DECISION TREE REGRESSION

Akhilesh Joshi

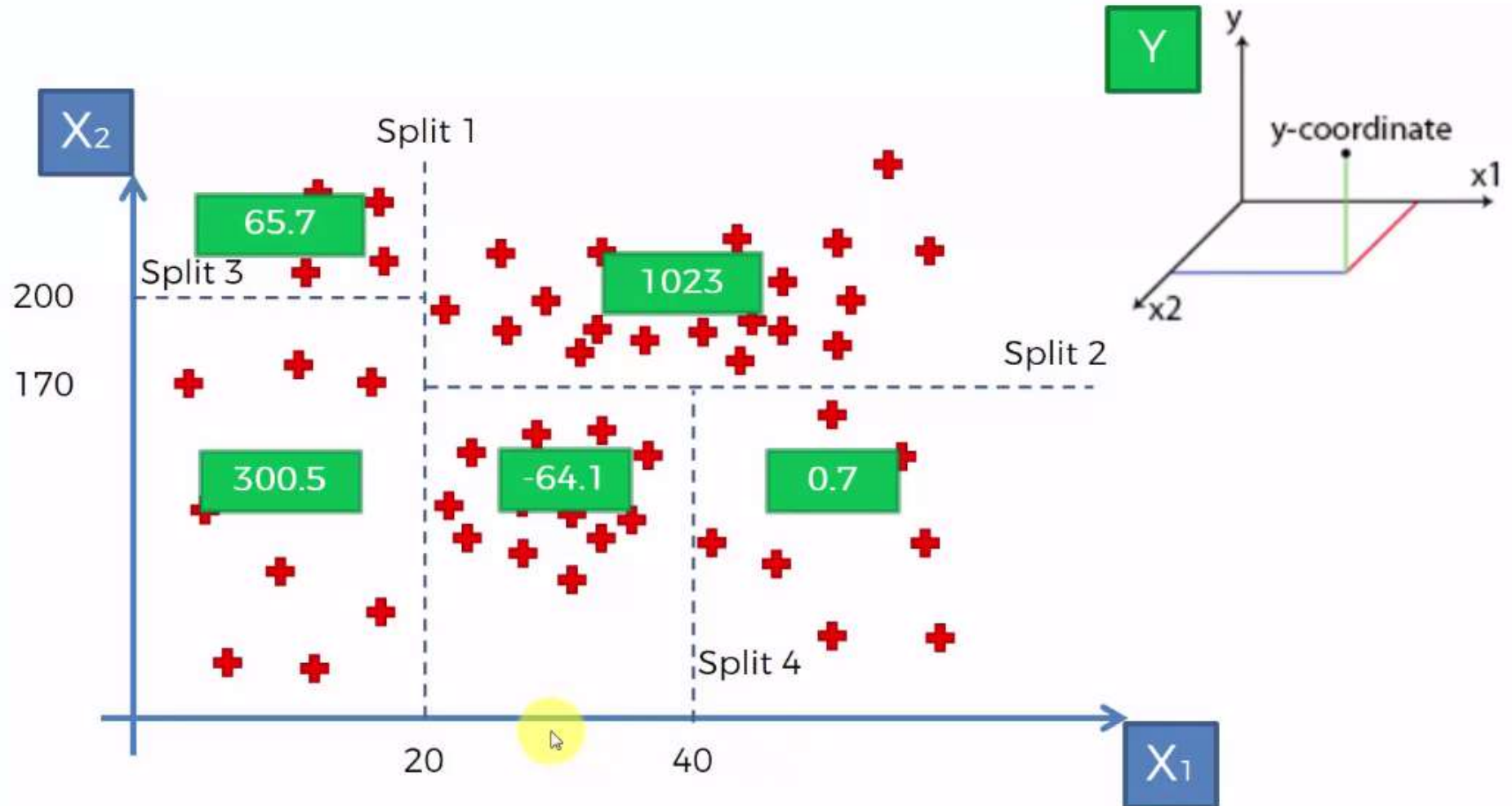
# Decision Tree Intuition



# Decision Tree Intuition

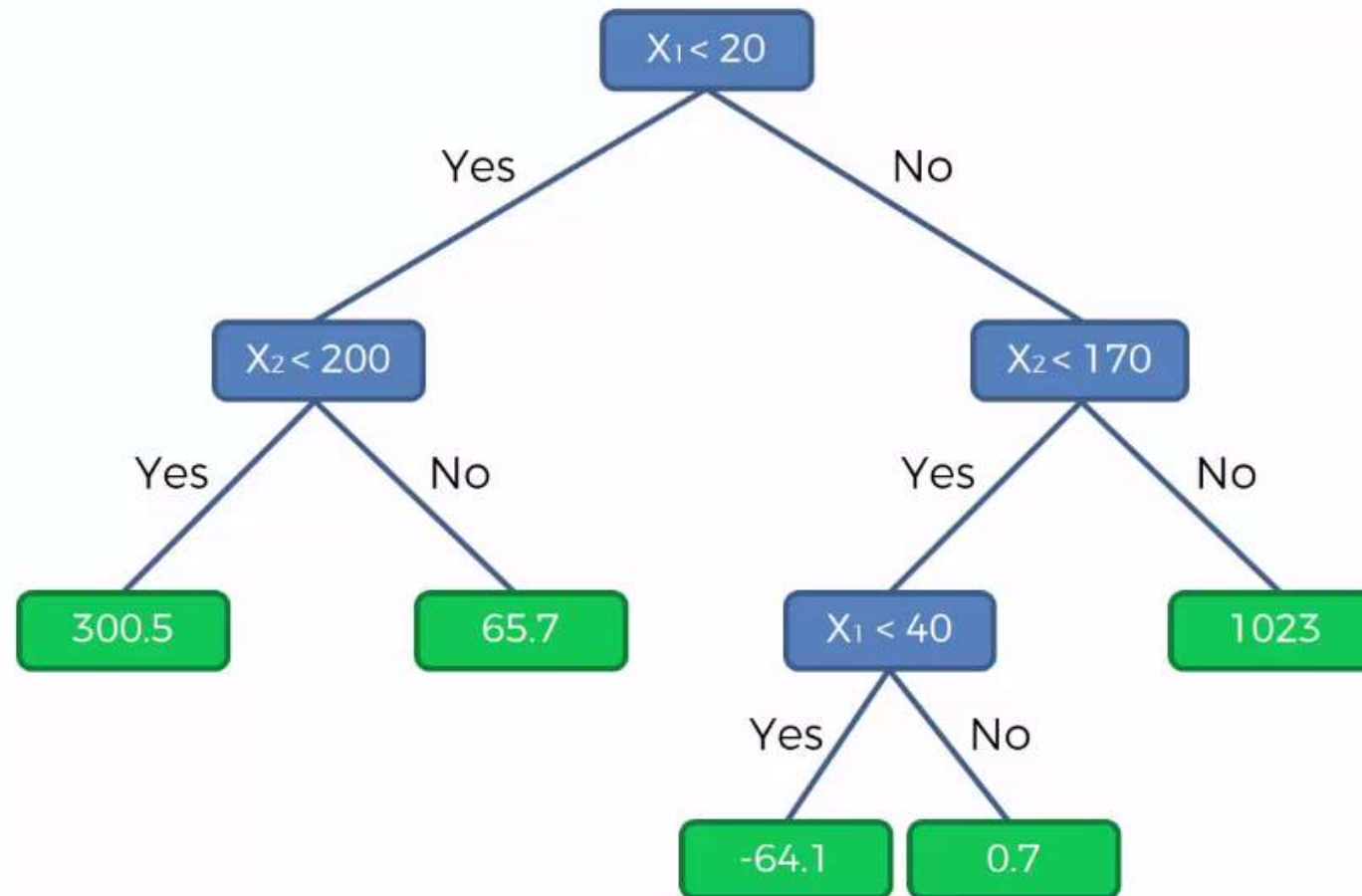


# Decision Tree Intuition

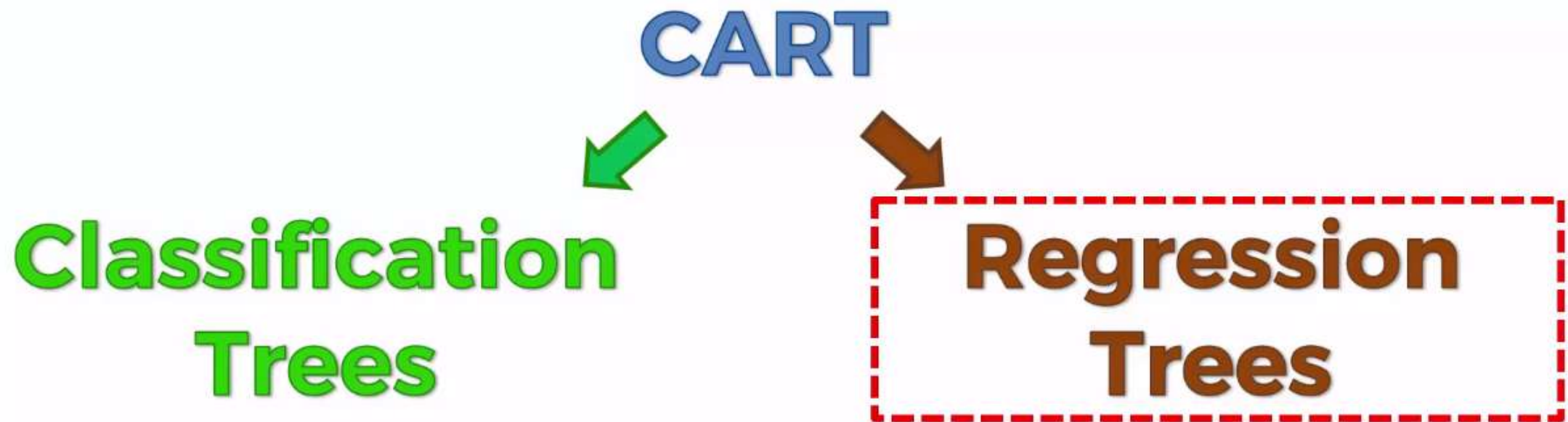


The value in green box represents the average of data points in that split

# Decision Tree Intuition



# Decision Tree Intuition





PYTHON



# READING FILE DYNAMICALLY

```
from tkinter import *  
from tkinter.filedialog import askopenfilename  
  
root = Tk()  
root.withdraw()  
root.update()  
file_path = askopenfilename()  
root.destroy()
```



# IMPORTING LIBRARIES

```
import pandas as pd
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

# IMPORTING DATASET

```
dataset = pd.read_csv(file_path)
```

```
X= dataset.iloc[:,1:2].values
```

```
y= dataset.iloc[:,2:3].values
```

# DECISION TREE REGRESSOR

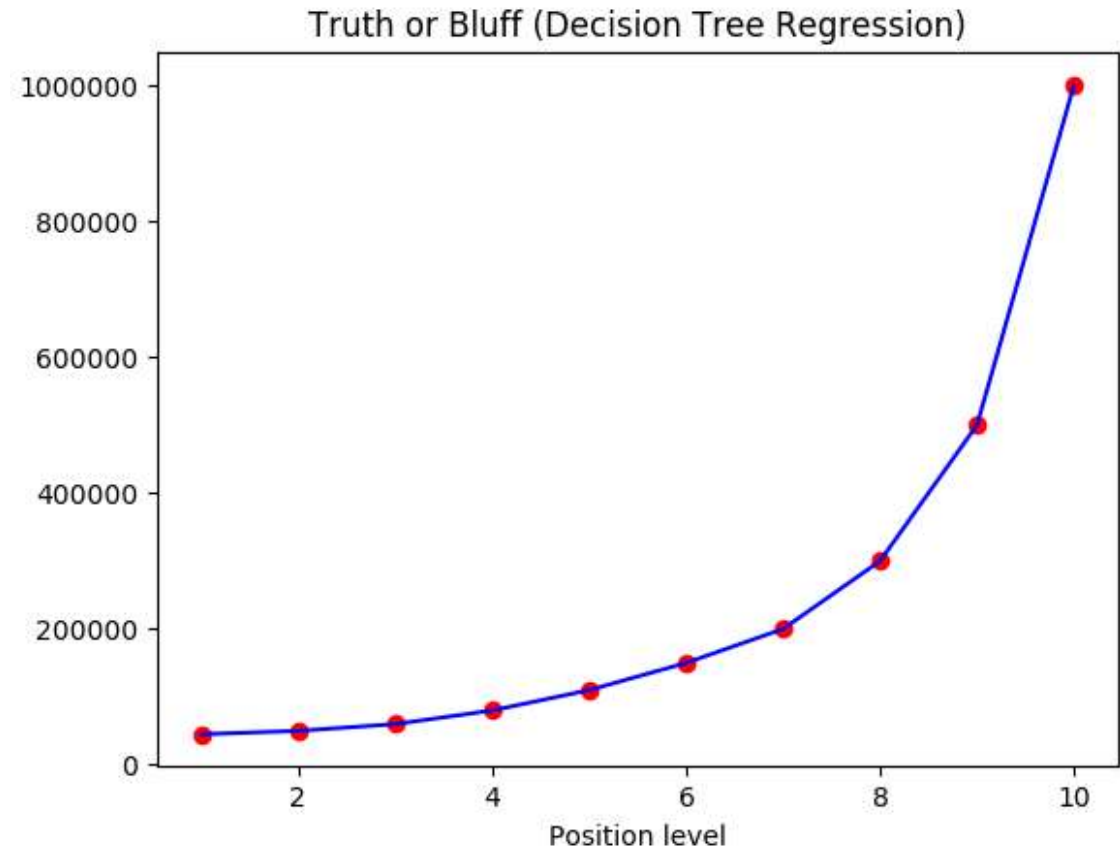
```
from sklearn.tree import DecisionTreeRegressor  
regressor = DecisionTreeRegressor(random_state=42)  
  
model = regressor.fit(X,y)
```

# PREDICTION

```
model.predict(6.5)
```

# SIMPLE PLOT

```
plt.scatter(X,y,color="red")  
plt.plot(X,model.predict(X),color="blue")  
plt.title('Truth or Bluff (Decision Tree  
Regression)')  
plt.xlabel('Position level')  
plt.ylabel('Salary')  
plt.show()
```



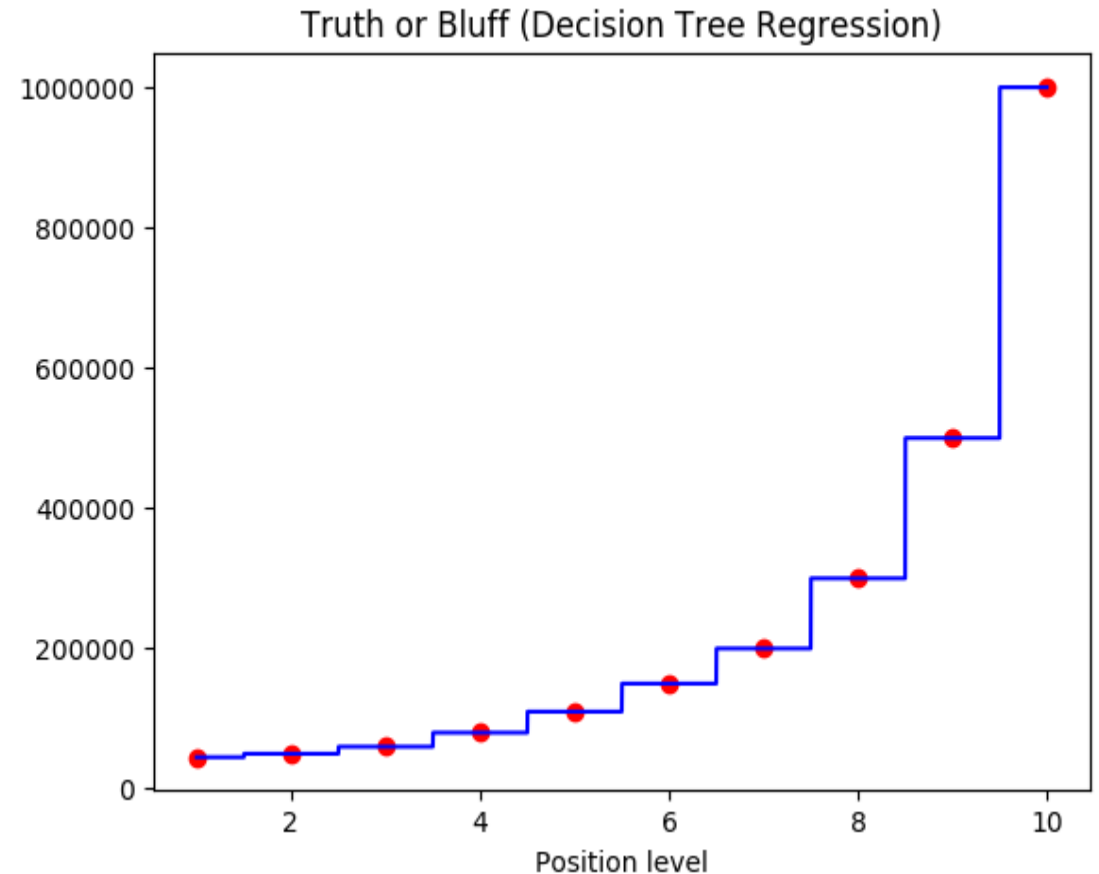
NOTE : Whats wrong here ? Well in the simple plot the Decision Tree Regressor model is treated as a continuous model. But it is not a continuous model. Decision Tree Regressor is a discrete model hence it should be treated as a discrete model.

FIX : plotting the same graph with grid with small step size say 0.01 will help us visualize better

# UPDATED PLOT

```
X_grid = np.arange(min(X), max(X), 0.001)
X_grid = X_grid.reshape((len(X_grid), 1))
plt.scatter(X, y, color = 'red')
plt.plot(X_grid, regressor.predict(X_grid),
color = 'blue')
plt.title('Truth or Bluff (Decision Tree
Regression)')
plt.xlabel('Position level')
plt.ylabel('Salary')
plt.show()
```

Note : Here the graph that is plotted gives us the clear discrete structure



R





# READ DATASET

```
library(readr)
```

```
dataset <- read_csv("D:/machine learning AZ/Machine Learning A-Z  
Template Folder/Part 2 – Regression/Section 7 – Support Vector  
Regression (SVR)/SVR/Position_Salaries.csv")
```

```
dataset= dataset[2:3]
```

# LIBRARY REQUIRED - RPART

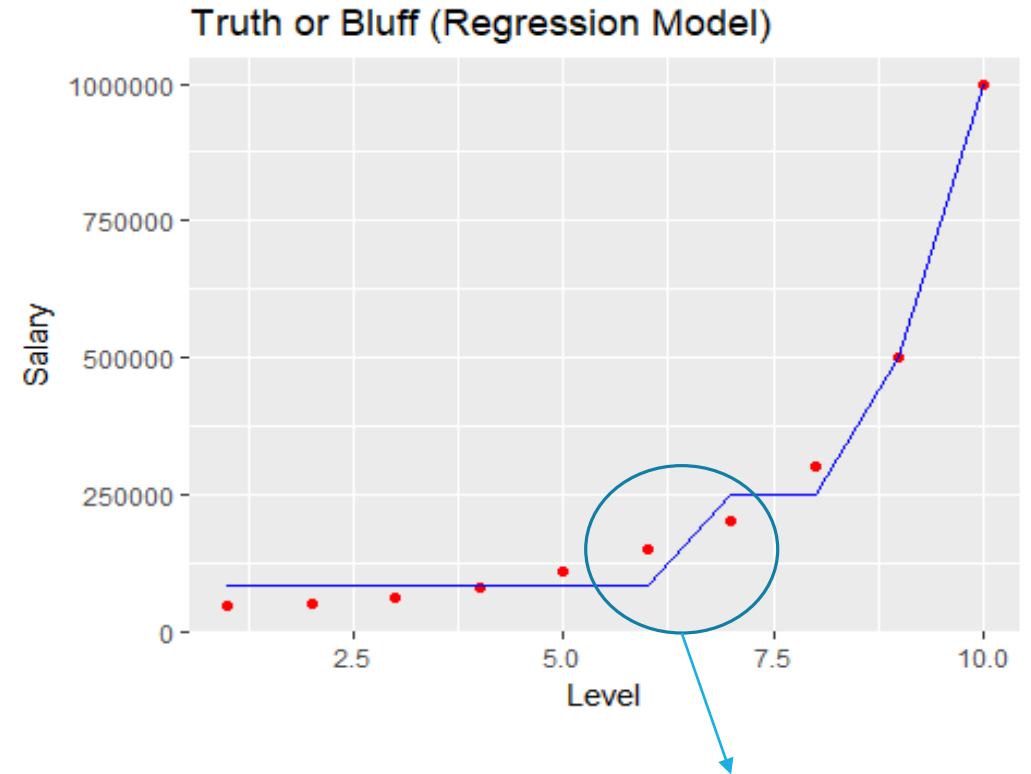
```
library('rpart')
```

```
regressor = rpart(Salary ~ . , data= dataset , control =  
rpart.control(minsplit = 1))
```

NOTE : do not forget to include control parameter as it decides the number of splits in your model.

# PLOT

```
# Visualising the Regression Model results
# install.packages('ggplot2')
library(ggplot2)
ggplot() +
  geom_point(aes(x = dataset$Level, y = dataset$Salary),
    colour = 'red') +
  geom_line(aes(x = dataset$Level, y = predict(regressor, newdata =
dataset)),
    colour = 'blue') +
  ggtitle('Truth or Bluff (Regression Model)') +
  xlab('Level') +
  ylab('Salary')
```

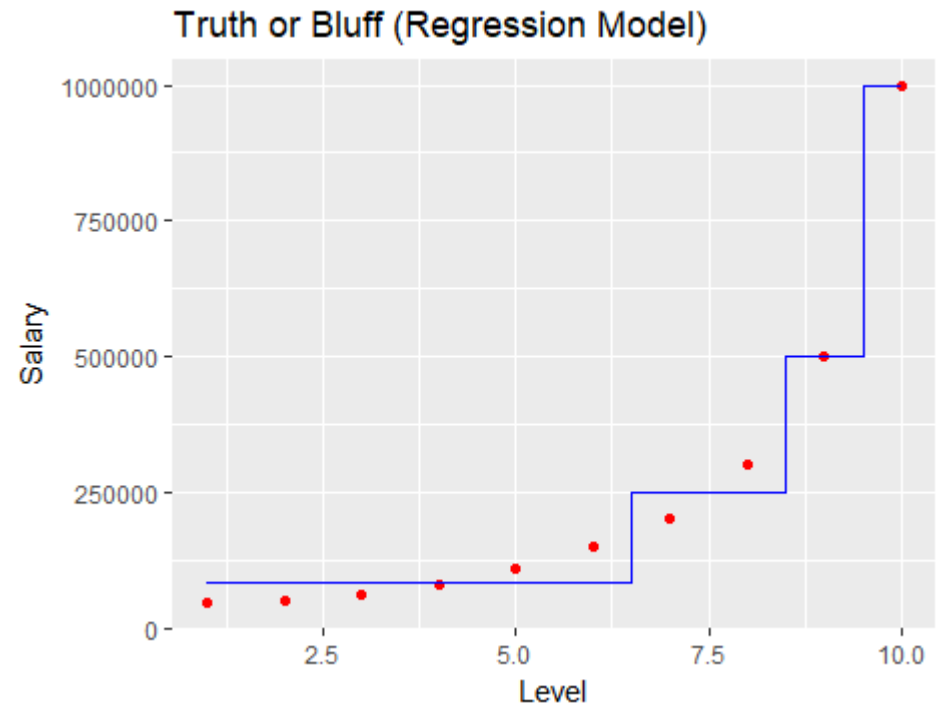


Here model is not treated as discrete hence plot simply joins the prediction points since we don't have any predictions for this interval.

Solution : plot Level as grid with step size as small as 0.01 or whatever you want it to be

# SMOOTHER PLOT

```
# install.packages('ggplot2')  
library(ggplot2)  
x_grid = seq(min(dataset$Level), max(dataset$Level), 0.001)  
ggplot() +  
  geom_point(aes(x = dataset$Level, y = dataset$Salary),  
    colour = 'red') +  
  geom_line(aes(x = x_grid, y = predict(regressor, newdata =  
    data.frame(Level = x_grid))),  
    colour = 'blue') +  
  ggtitle('Truth or Bluff (Regression Model)') +  
  xlab('Level') +  
  ylab('Salary')
```



**PERFECT PLOT**

# PREDICTIONS

```
prediction = predict(regressor,data.frame(Level=6.5))
```