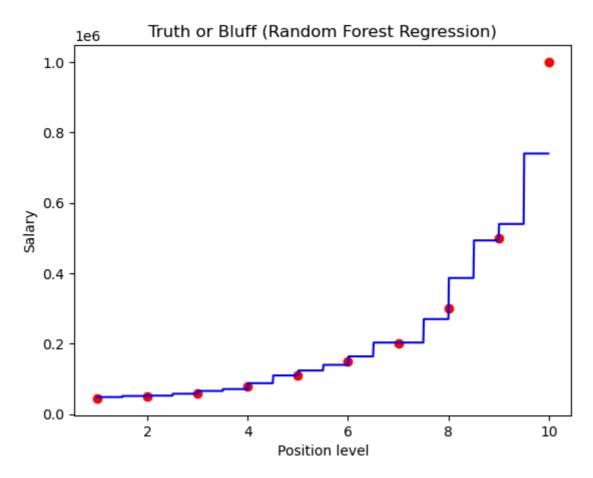
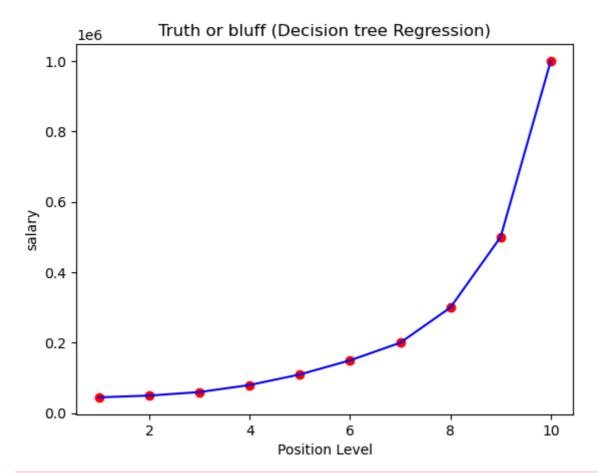
```
In [3]: # Random Forest Regression
        # Importing the libraries
        import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        # Importing the dataset
        #dataset = pd.read_csv(r"C:\Users\kdata\Desktop\KODI WORK\1. NARESH\1. MORNING B
        dataset = pd.read_csv(r'C:\Users\soham\OneDrive\Desktop\(OCT 02)\1.POLYNOMIAL RE
        X = dataset.iloc[:, 1:2].values
        y = dataset.iloc[:, 2].values
        # Splitting the dataset into the Training set and Test set
        """from sklearn.cross_validation import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
        # Feature Scaling
        from sklearn.preprocessing import StandardScaler
        sc_X = StandardScaler()
        X_train = sc_X.fit_transform(X_train)
        X_test = sc_X.transform(X_test)
        sc_y = StandardScaler()
        y_train = sc_y.fit_transform(y_train)
        # Fitting Random Forest Regression to the dataset
        #from sklearn.ensemble import RandomForestRegressor
        #regressor = RandomForestRegressor(n_estimators = 100, random_state = 0)
        #regressor.fit(X, y)
        from sklearn.ensemble import RandomForestRegressor
        regressor = RandomForestRegressor(n estimators=15)
        regressor.fit(X, y)
        # Predicting a new result
        y_pred = regressor.predict([[6.5]])
        # Visualising the Random Forest Regression results (higher resolution)one
        X \text{ grid} = \text{np.arange}(\text{min}(X), \text{max}(X), 0.01)
        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 (Random Forest Regression)')
        plt.xlabel('Position level')
        plt.ylabel('Salary')
        plt.show()
       C:\Users\soham\AppData\Local\Temp\ipykernel_8324\1760639366.py:41: DeprecationWar
```

C:\Users\soham\AppData\Local\Temp\ipykernel\_8324\1760639366.py:41: DeprecationWar
ning: Conversion of an array with ndim > 0 to a scalar is deprecated, and will er
ror in future. Ensure you extract a single element from your array before perform
ing this operation. (Deprecated NumPy 1.25.)
X grid = np.arange(min(X), max(X), 0.01)

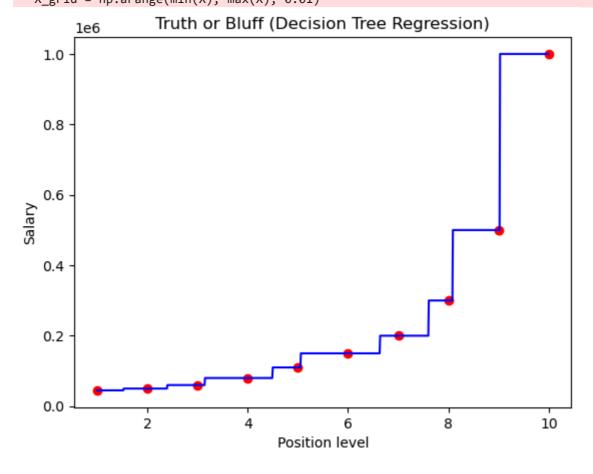


```
In [5]: # Decision Tree Regression
        # Importing the libraries
        import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        # Importing the dataset
        dataset = pd.read csv(r'C:\Users\soham\OneDrive\Desktop\(OCT 02)\1.POLYNOMIAL RE
        X = dataset.iloc[:, 1:2].values
        y = dataset.iloc[:, 2].values
        # Splitting the dataset into the Training set and Test set
        """from sklearn.cross_validation import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
        '''# Feature Scaling
        from sklearn.preprocessing import StandardScaler
        sc_X = StandardScaler()
        X_train = sc_X.fit_transform(X_train)
        X_test = sc_X.transform(X_test)
        sc y = StandardScaler()
        y_train = sc_y.fit_transform(y_train)'''
        #Fitting Decision Tree Regression to the datasetmae
        from sklearn.tree import DecisionTreeRegressor
        regressor = DecisionTreeRegressor(criterion = 'friedman_mse',splitter = 'random'
        regressor.fit(X, y)
        from sklearn.ensemble import RandomForestRegressor
        reg = RandomForestRegressor(n_estimators = 300, random_state = 0)
```

```
reg.fit(X,y)
# Predicting a new result
y_pred = reg.predict([[6.5]])
#now predict previous employee salary & visualize the result
#emplyoee said his salary was 161k but as per dt we got as 150 which was sama as
#what we got in decision tree as 10k less as from previous salary
plt.scatter(X, y, color = 'red')
plt.plot(X,regressor.predict(X), color = 'blue')
plt.title('Truth or bluff (Decision tree Regression)')
plt.xlabel('Position Level')
plt.ylabel('salary')
plt.show()
#first part is curve is very good & as i explained this is not a decision tree c
#algorithm of decission tree is by considering the entrophy and information gain
#as per our tutorial we have 2 independent variable diferent interval forms rect
#you have quastion if you taking average of each interval then how do you have a
#And you cannot find the average of independent variable & this is not a continu
#lets plot the higher resolution using tree models
#if you advance visualisation along with tree structure then you will get this r
# Visualising the Decision Tree Regression results (higher resolution)
X_{grid} = np.arange(min(X), max(X), 0.01)
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()
#if you check the plot you found the straight & verticle line hear and based on
#if you check the interval of 6 then you get the point of 150k & the range is 5.
#this is all about decission tree regression & for next session we will see the
```



C:\Users\soham\AppData\Local\Temp\ipykernel\_8324\1883255396.py:59: DeprecationWar
ning: Conversion of an array with ndim > 0 to a scalar is deprecated, and will er
ror in future. Ensure you extract a single element from your array before perform
ing this operation. (Deprecated NumPy 1.25.)
 X\_grid = np.arange(min(X), max(X), 0.01)



Tn Γ 1: