Random Forest Provides a much more accurate result as compared to other machine Learning models

It has the below mentioned advantages:

- 1. It reduces overfitting in decision trees and helps to improve the accuracy
- 2. It is flexible to both classification and regression problems
- 3. It works well with both categorical and continuous values
- 4. It automates missing values present in the data
- 5. Normalising of data is not required as it uses a rule-based approach

```
from sklearn.linear_model import LinearRegression
lin_reg = evaluate_model(LinearRegression(), X_train, y_train)

LinearRegression
Accuracy: 65.52 %
Standard Deviation: 4.30 %

from sklearn.svm import SVR

svr = evaluate_model(SVR(), X_train, y_train)

SVR
Accuracy: 82.84 %
Standard Deviation: 2.31 %
```

```
from sklearn.ensemble import RandomForestRegressor
rf = evaluate_model(RandomForestRegressor(n_estimators = 100, random_state = 0), X_train, y_
{\tt RandomForestRegressor}
Accuracy: 85.58 %
Standard Deviation: 4.31 %
from sklearn.ensemble import AdaBoostRegressor
ada_boost = evaluate_model(AdaBoostRegressor(random_state=0, n_estimators=100), X_train, y_t
rain)
AdaBoostRegressor
Accuracy: 79.58 %
Standard Deviation: 2.79 %
import xgboost
xg_boost = evaluate_model(xgboost.XGBRegressor(), X_train, y_train)
XGBRegressor
Accuracy: 84.74 %
Standard Deviation: 5.43 %
```

When compared to other models it has yielded has much accurate result of 86% and above and on the other hand Linear regression model yields 66%~ accuracy,SVR yields 83%~ accuracy,AdaBoostRegressor yields 80%,XGBRegressor yields 85%