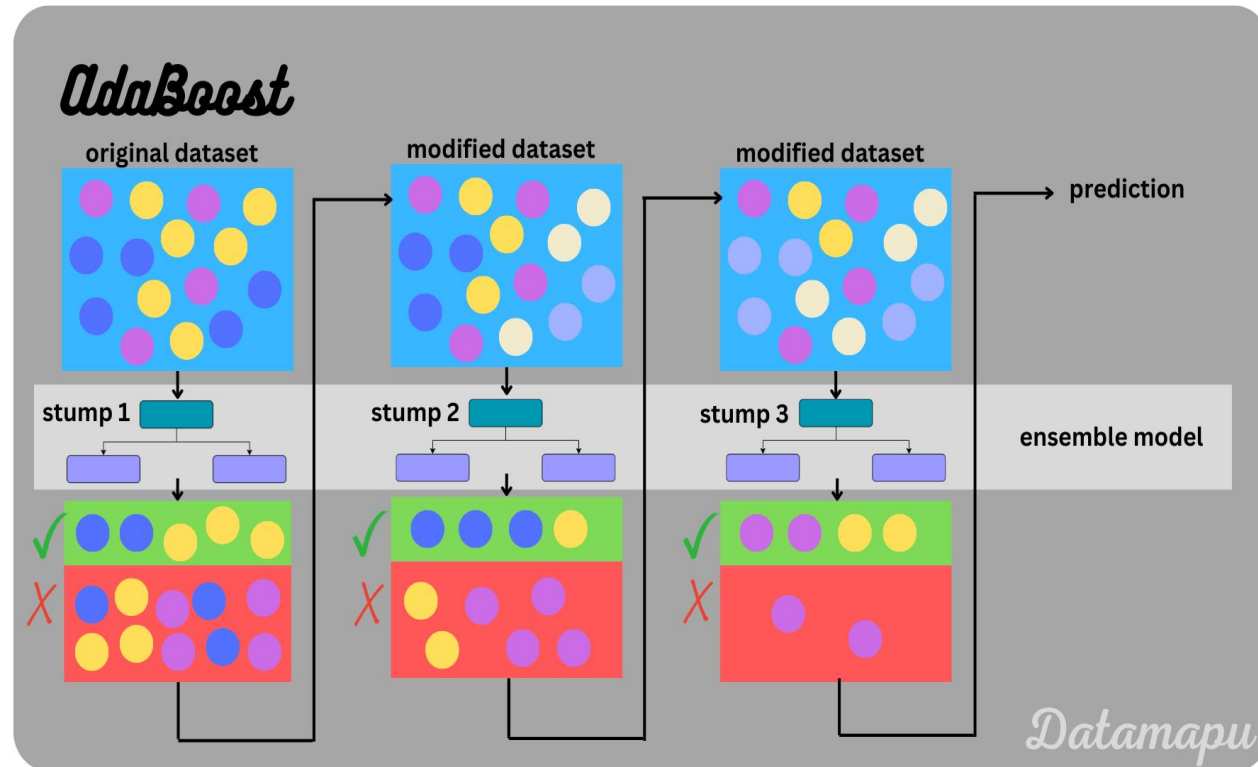



Boosting Algorithms

- 1) AdaBoost
- 2) Gradient Boosting
 - i) XG Boosting
 - ii) LG Boosting

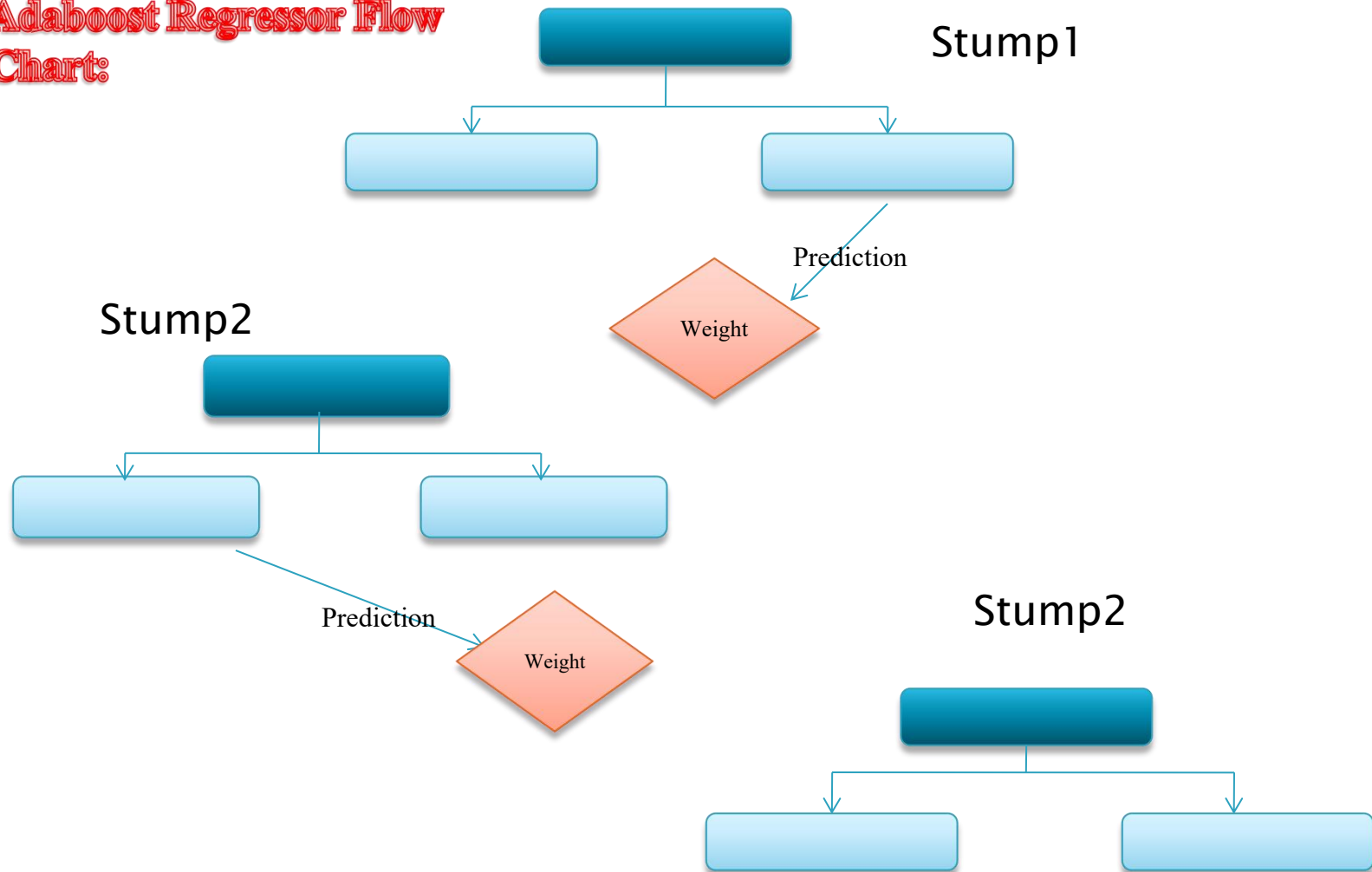
1) AdaBoost(Adaptive Boost)



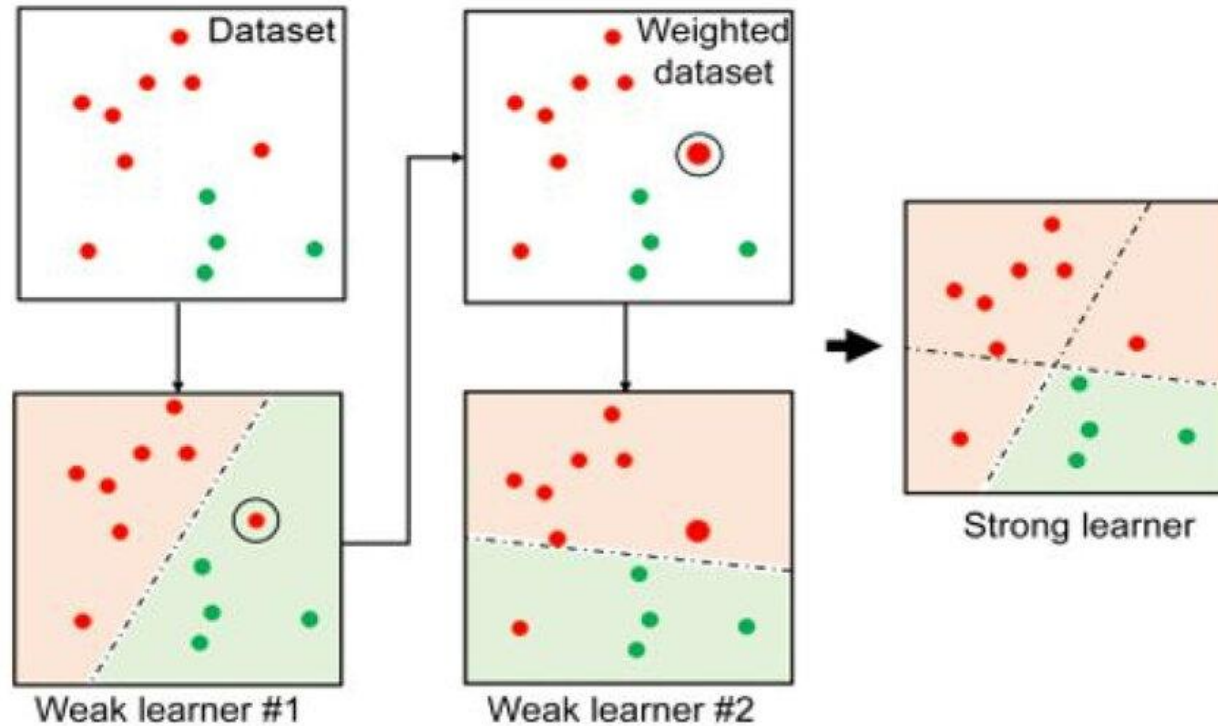
Working Method:

- } Adaboost Algorithm works same like boosting algorithm(ie. It do feature sampling in sequentially)
 - } It transforms weak learners to strong learners.
 - } AdaBoost is a Boosting algorithm, which means that the ensemble model is built sequentially and each new model builds on the results of the previous one, trying to improve its errors.
 - } The weights are determined in such a way that the wrongly predicted samples get higher weights than the correctly predicted samples.
- 

Adaboost Regressor Flow Chart:



Transformation of Weak learner into Strong learner:



AdaBoost algorithm

Before applying

```
[33]: from sklearn.preprocessing import StandardScaler  
sc=StandardScaler()  
X_train = sc.fit_transform(X_train)  
X_test = sc.transform(X_test)
```

```
[41]: from sklearn.svm import SVR  
regressor=SVR(kernel="sigmoid",C=10,coef0=1,epsilon=1)  
regressor.fit(X_train,Y_train)
```

C:\Users\Hxtreme\anaconda3\Lib\site-packages\sklearn\utils\validation.py:103: UserWarning: The shape of y is incompatible. Please change the shape of y to (n_samples,), for example use y = column_or_1d(y, warn=True)

```
[41]: SVR(C=10, coef0=1, epsilon=1, kernel='sigmoid')
```

```
[42]: Y_pred = regressor.predict(X_test)
```

```
[43]: from sklearn.metrics import r2_score  
r_score = r2_score(Y_test,Y_pred)  
r_score
```

```
[43]: -0.05571437556341796
```

After applying

```
81]: from sklearn.ensemble import AdaBoostRegressor  
regressor = AdaBoostRegressor(random_state=0, n_estimators=100)  
regressor.fit(X_train,Y_train)
```

C:\Users\Hxtreme\anaconda3\Lib\site-packages\sklearn\utils\validation.py:103: UserWarning: The shape of y is incompatible. Please change the shape of y to (n_samples,), for example use y = column_or_1d(y, warn=True)

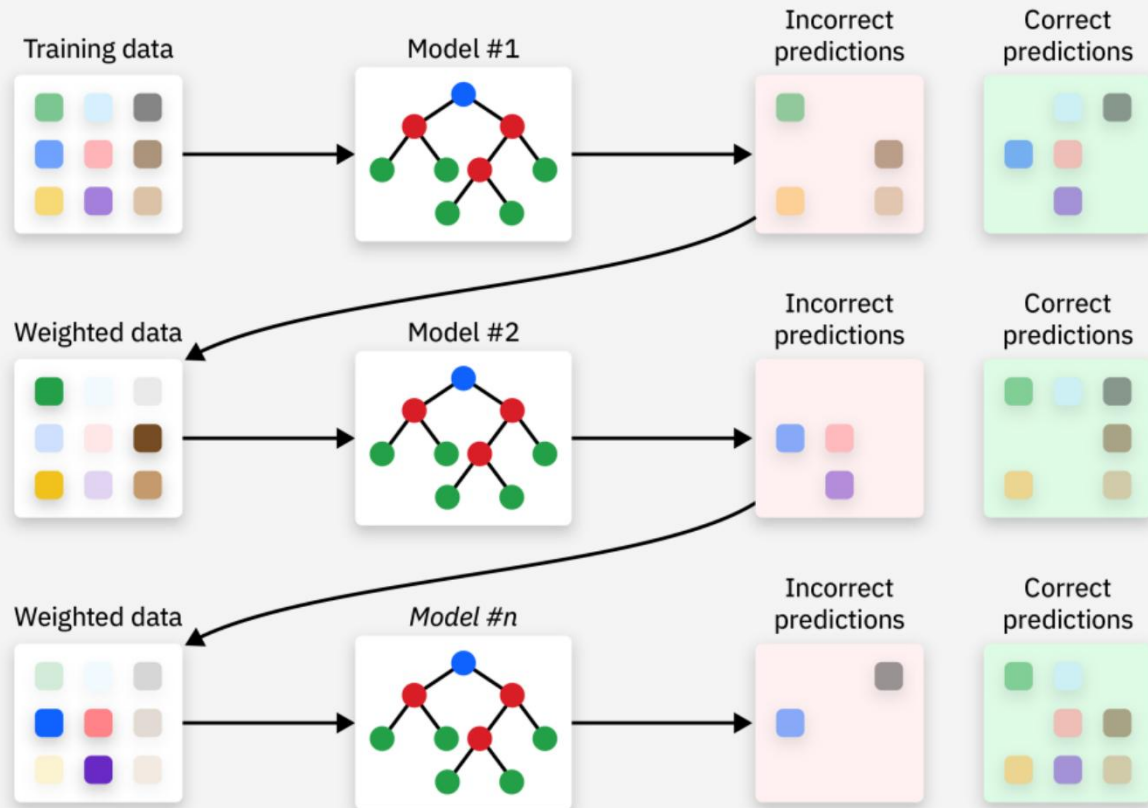
```
81]: AdaBoostRegressor  
AdaBoostRegressor(n_estimators=100, random_state=0)
```

```
82]: Y_pred = regressor.predict(X_test)
```

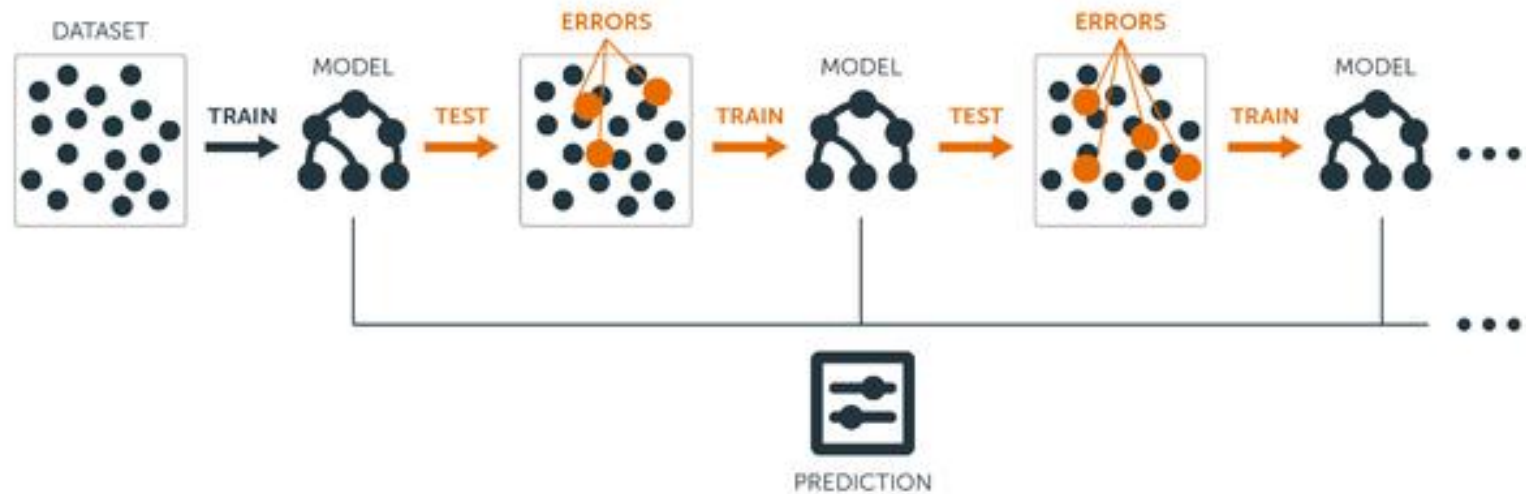
```
83]: from sklearn.metrics import r2_score  
r_score = r2_score(Y_test,Y_pred)  
r_score
```

```
83]: 0.9268799485154495
```



2) Gradient Boost

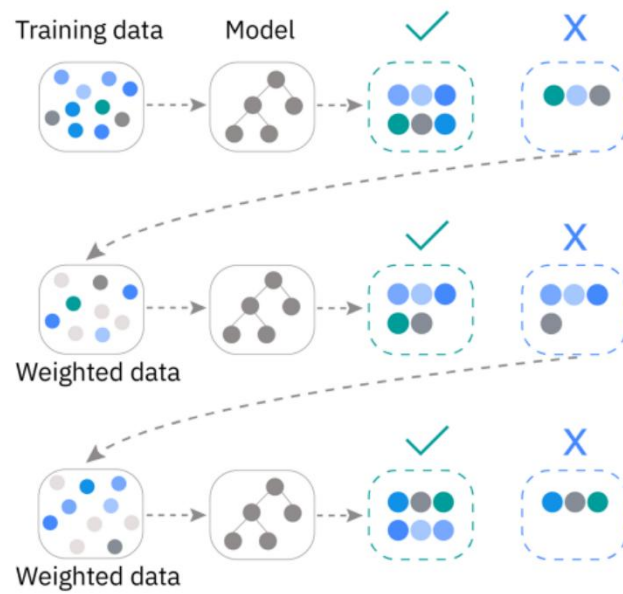


Flow Diagram for Gradient Boost:



About Gradient Boosting:

- } Gradient boosting is a machine learning technique that combines multiple weak prediction models into a single ensemble.
 - } The model improves sequentially but not incrementally weight.
 - } These weak models are typically decision trees, which are trained sequentially to minimize errors and improve accuracy.
 - } By combining multiple decision tree regressors or decision tree classifiers, gradient boosting can effectively capture complex relationships between features.
- 



Gradient Boosting Regressor

Before Boosting Algorithm

```
[33]: from sklearn.preprocessing import StandardScaler  
sc=StandardScaler()  
X_train = sc.fit_transform(X_train)  
X_test = sc.transform(X_test)
```

```
[41]: from sklearn.svm import SVR  
regressor=SVR(kernel="sigmoid",C=10,coef0=1,epsilon=1)  
regressor.fit(X_train,Y_train)
```

C:\Users\Hxtreme\anaconda3\Lib\site-packages\sklearn\utils\validation.py:103: FutureWarning: Please change the shape of y to (n_samples,), for example using y = column_or_1d(y, warn=True)

```
[41]: SVR  
SVR(C=10, coef0=1, epsilon=1, kernel='sigmoid')
```

```
[42]: Y_pred = regressor.predict(X_test)
```

```
[43]: from sklearn.metrics import r2_score  
r_score = r2_score(Y_test,Y_pred)  
r_score
```

```
[43]: -0.05571437556341796
```

After Boosting Algorithm

```
[5]: from sklearn.ensemble import GradientBoostingRegressor  
regressor = GradientBoostingRegressor(random_state=0)  
regressor.fit(X_train, Y_train)
```

C:\Users\Hxtreme\anaconda3\Lib\site-packages\sklearn\ensemble\gb.py:199: FutureWarning: Please change the shape of y to (n_samples,), for example using y = column_or_1d(y, warn=True) # TODO: Is this still required?

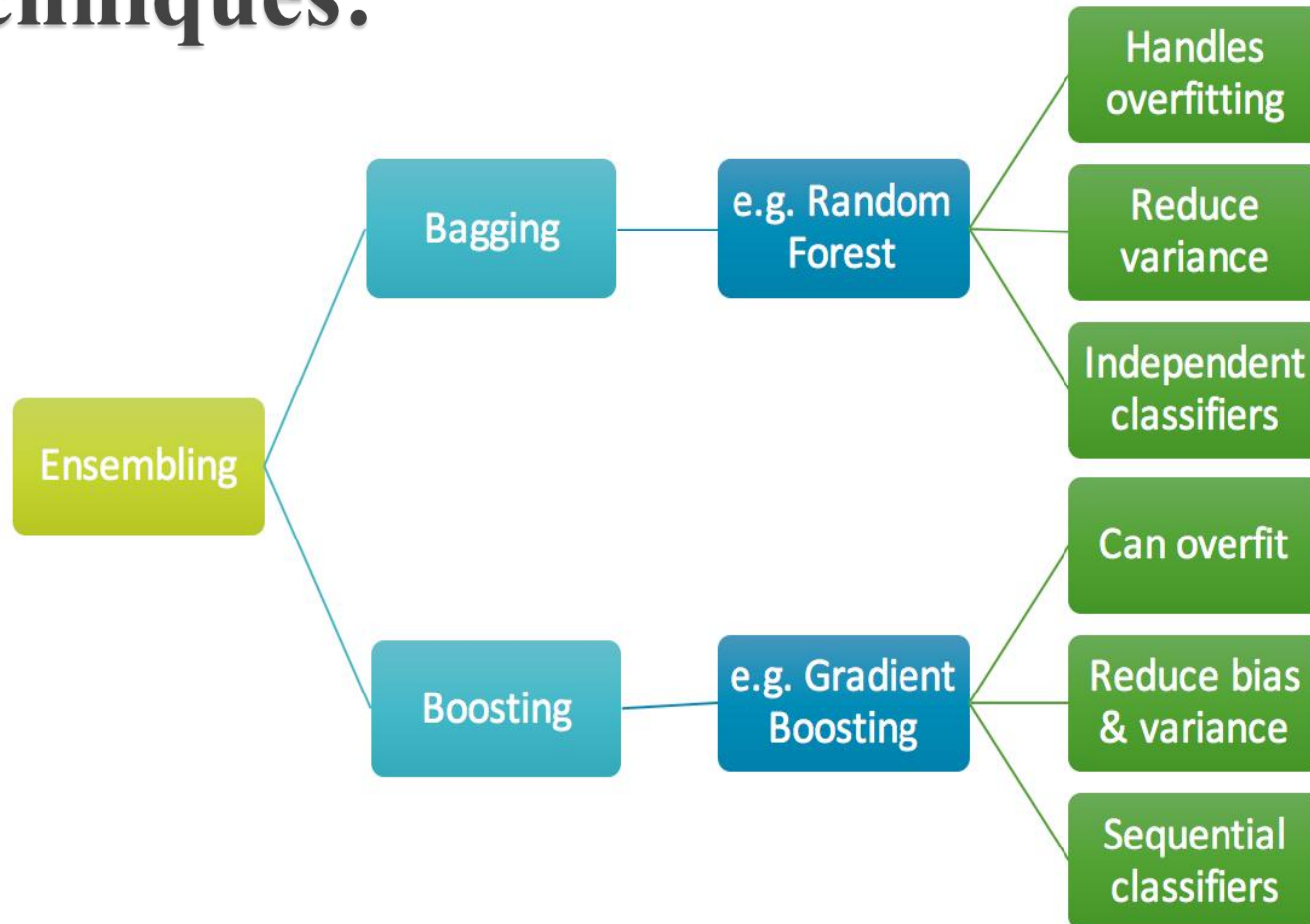
```
[5]: GradientBoostingRegressor  
GradientBoostingRegressor(random_state=0)
```

```
[9]: Y_pred = regressor.predict(X_test)
```

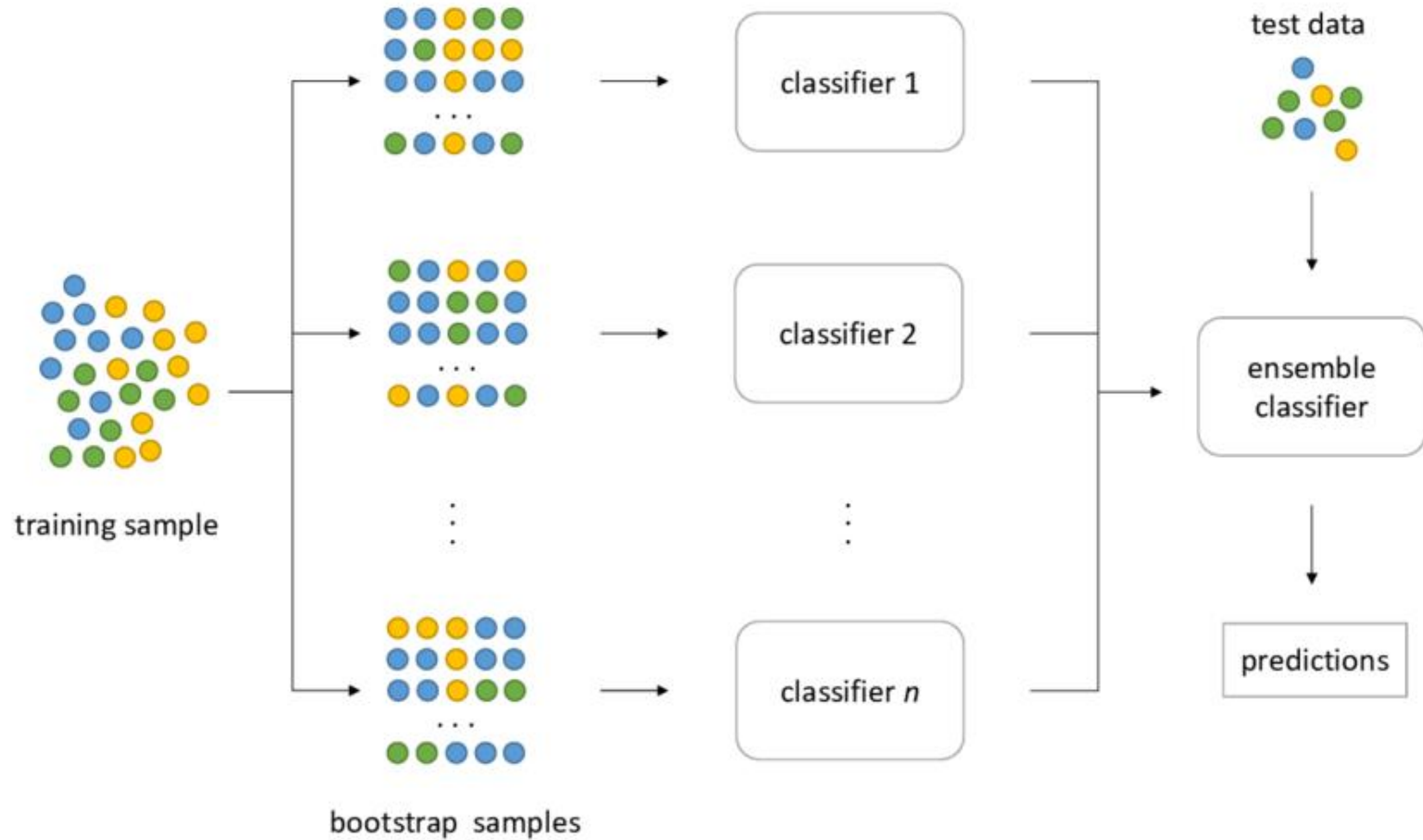
```
[10]: from sklearn.metrics import r2_score  
r_score = r2_score(Y_test,Y_pred)  
r_score
```

```
[10]: 0.9226242574216024
```


Comparison between two ensemble Techniques:



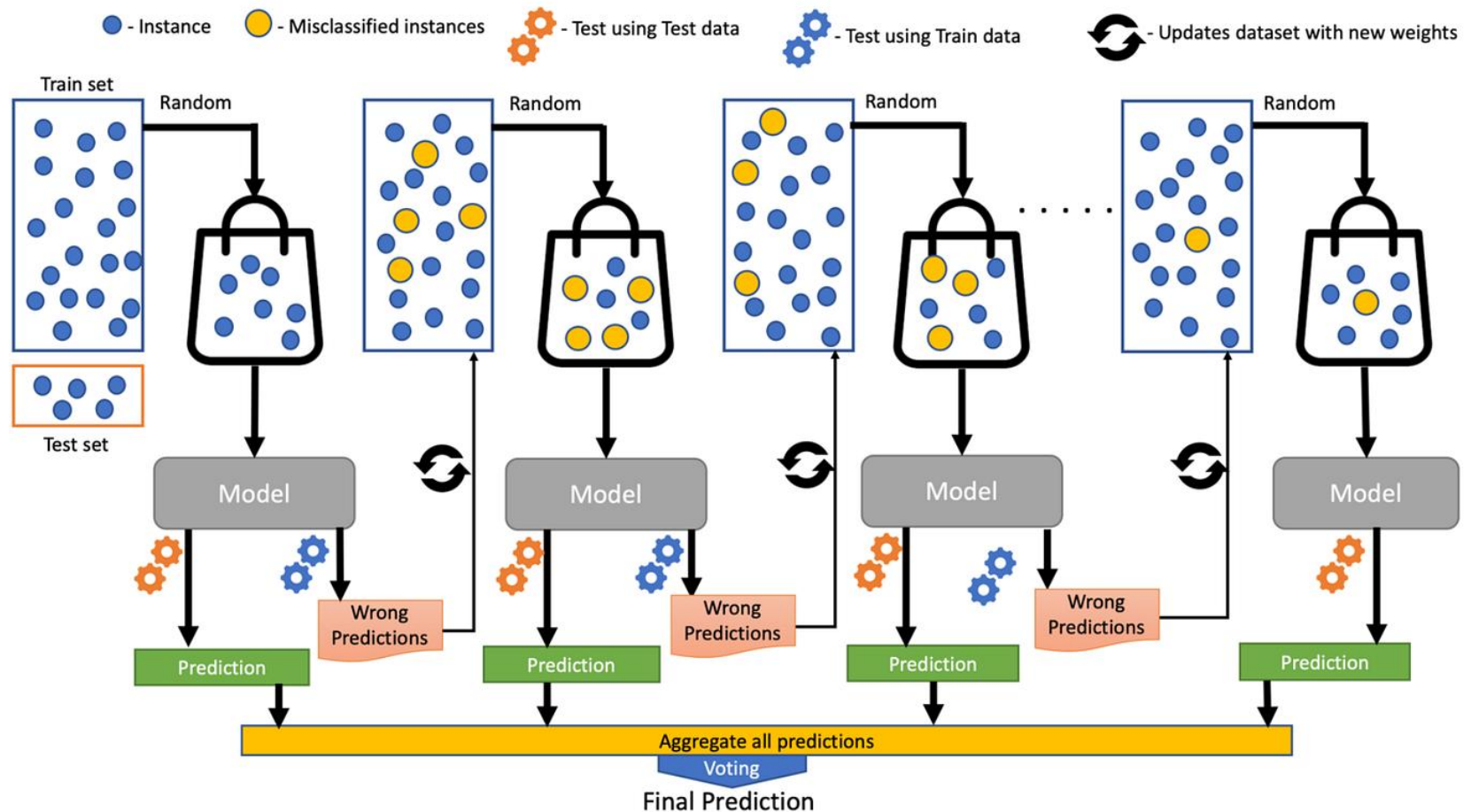
i) XG Boost



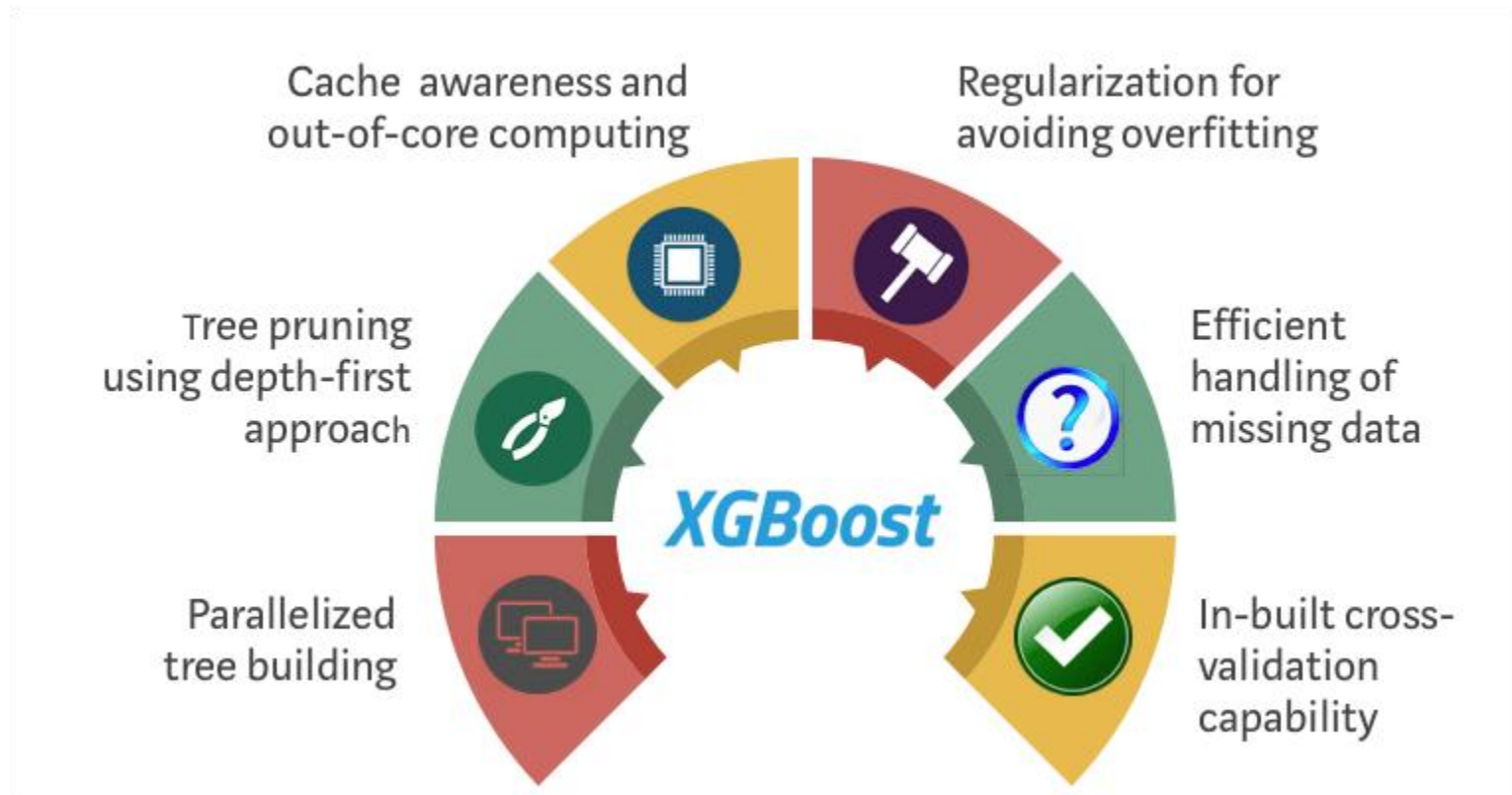
XG Boost(Extreme Gradient) working:

- } It is an implementation of gradient boosting to provide better speed and performance. It is decision tree based ensemble machine learning algorithm.
 - } It works same like gradient boosting but it has additional functionality.
 - } XGBoost predicts with greater accuracy and with less time complexity as compared to other machine learning algorithms.
 - } It is a distributed Machine Learning Process.
 - } It can handle large datasets.
- 

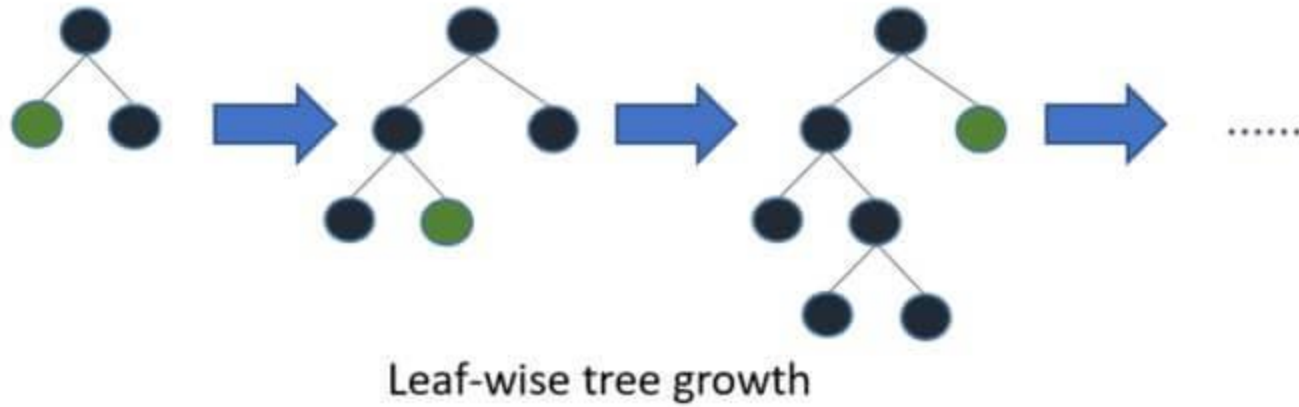
Flow Architecture:



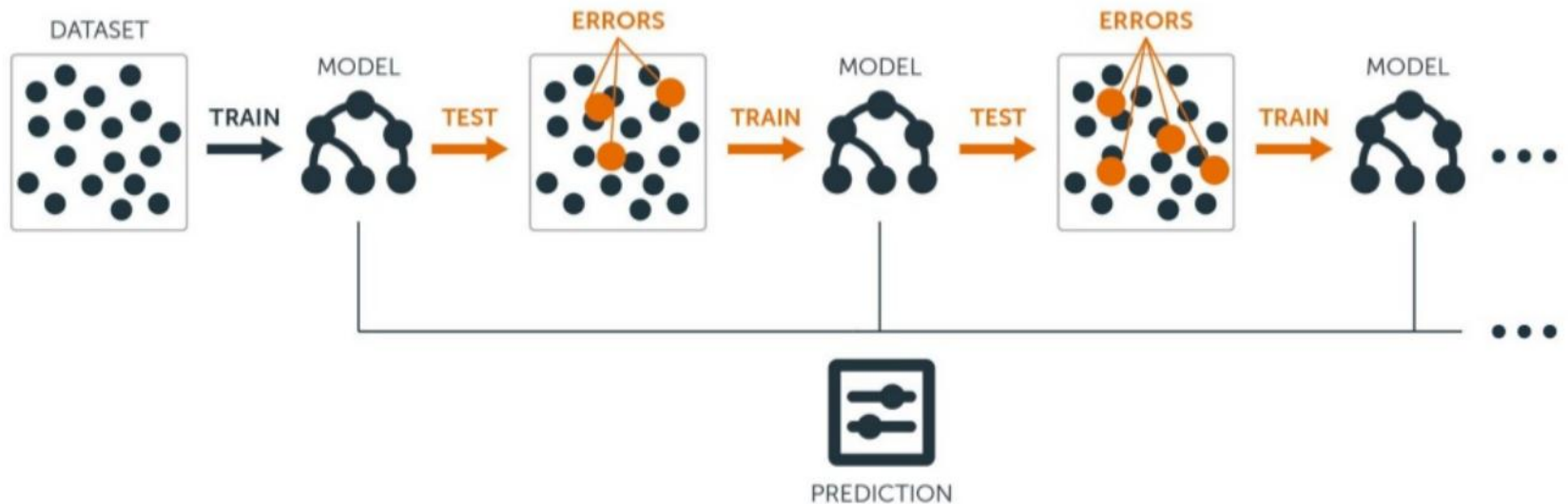
Advantages of XG Boosting:




ii) LG Boost Algorithm



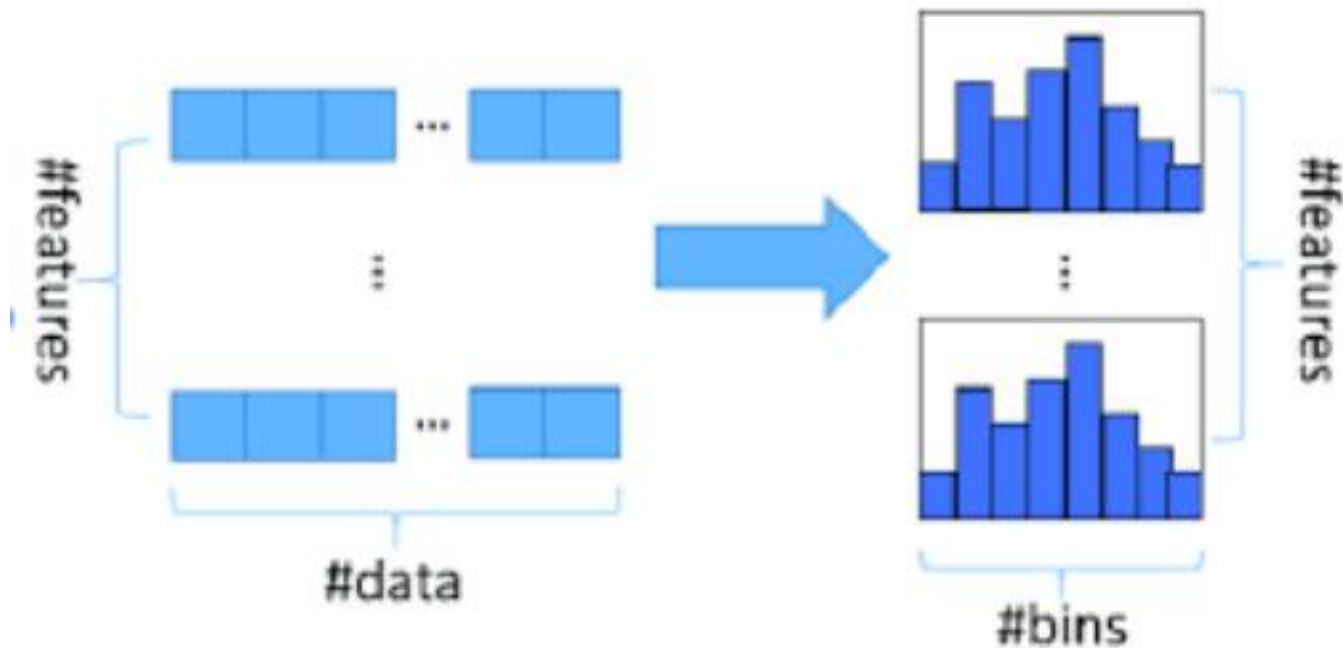
Flow Architecture for LG Boosting:



LG Boost(Light Gradient Boosting)

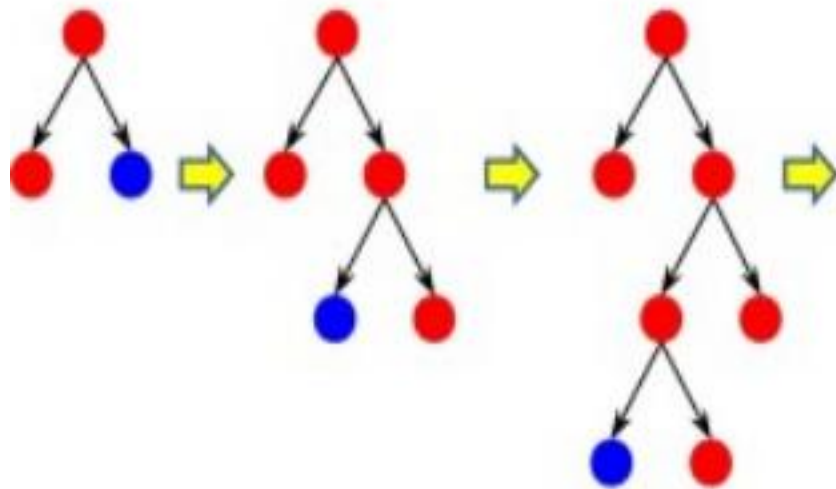
- } LightGBM is an open-source high-performance framework developed by Microsoft.
 - } It is an ensemble learning framework that uses gradient boosting method which constructs a strong learner by sequentially adding weak learners in a gradient descent manner.
 - } It uses histogram method for selecting the best fit.
 - } It can handle huge amount of data and poor to handle small dataset.
- 

LG Boosting uses histogram based method for continuous split into bins or buckets

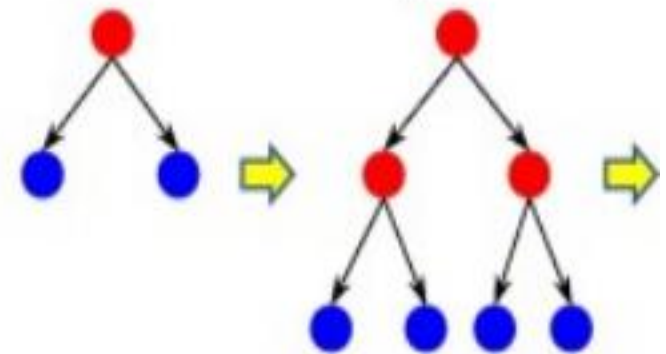


LG Boost Vs XG Boost

Leaf-wise tree growth in LightGBM



Level-wise tree growth



Difference between XG and LGBM

Aspect	XG Boost	Light GBM
Tree Growth	Level-wise	Leaf wise
Training Speed	Slower	Faster
Risk of Over fitting	Lower	Higher
CPU Performance	Strong	Moderate
Categorical Support Feature	Limited	Native

LG Boost algorithm

Before applying

```
[33]: from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

```
[41]: from sklearn.svm import SVR
regressor=SVR(kernel="sigmoid",C=10,coef0=1,epsilon=1)
regressor.fit(X_train,Y_train)
```

C:\Users\Hxtreme\anaconda3\Lib\site-packages\sklearn\utils\checked_array.py:113: UserWarning: Y has non-integer features. This is deprecated in version 0.22 and will raise an error in version 0.24. Please change the shape of y to (n_samples,), for y = column_or_1d(y, warn=True)

```
[41]: SVR
SVR(C=10, coef0=1, epsilon=1, kernel='sigmoid')
```

```
[42]: Y_pred = regressor.predict(X_test)
```

```
[43]: from sklearn.metrics import r2_score
r_score = r2_score(Y_test,Y_pred)
r_score
```

```
[43]: -0.05571437556341796
```

After applying

```
[106]: from sklearn import LGBMRegressor
regressor = LGBMRegressor(random_state=42)
regressor.fit(X_train, Y_train)
```

ImportError Traceback (most recent call last)
Cell In[106], line 1

```
----> 1 from sklearn import LGBMRegressor
      2 regressor = LGBMRegressor(random_state=42)
      3 regressor.fit(X_train, Y_train)
```

ImportError: cannot import name 'LGBMRegressor' from 'sklearn'

```
[107]: Y_pred = regressor.predict(X_test)
```

```
[108]: from sklearn.metrics import r2_score
r_score = r2_score(Y_test,Y_pred)
r_score
```

```
[108]: 0.9268799485154495
```