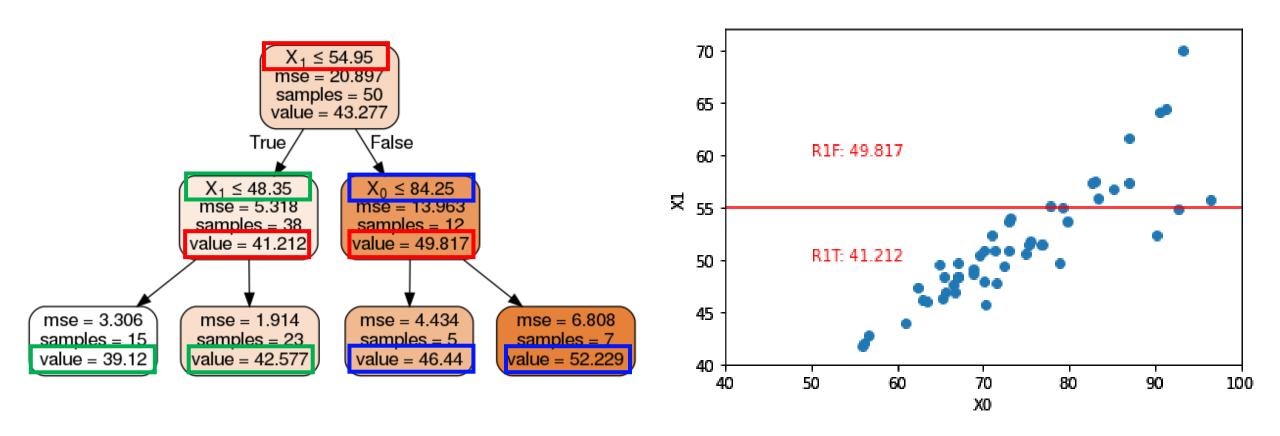
# Decision Trees Hyperparameters

# **Quick Review**



# Quick Review- metrics for splitting

### **Regression Tree**

### **MSE**

$$H(X_m) = \frac{1}{N_m} \sum_{i \in N_m} (y_i - \bar{y}_m)^2$$

### **MAE**

$$H(X_m) = \frac{1}{N_m} \sum_{i \in N_m} |y_i - \bar{y}_m|$$

### **Classification Tree**

### Gini

$$H(X_m) = \sum_{k} p_{mk} (1 - p_{mk})$$

### **Entropy**

$$H(X_m) = -\sum_{k} p_{mk} \log(p_{mk})$$

Information Gain = E(parent)-E(children)

# Outline

**Usage in sklearn** 

**Hyperparameters** 

```
from sklearn.tree import DecisionTreeRegressor, DecisionTreeClassifier
clf = DecisionTreeClassifier(random state=0)
clf.fit(X,y)
```

```
class sklearn.tree. DecisionTreeRegressor(*, criterion='mse', splitter='best', max depth=None, min samples split=2,
min\_samples\_leaf=1, min\_weight\_fraction\_leaf=0.0, max\_features=None, random\_state=None, max\_leaf\_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None, ccp_alpha=0.0)
```

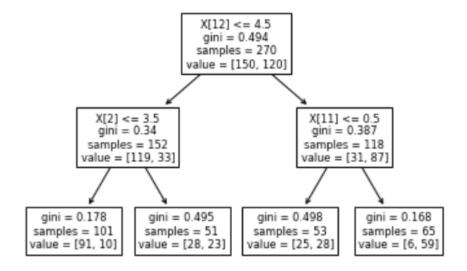
class sklearn.tree. DecisionTreeClassifier(\*, criterion='gini', splitter='best', max\_depth=None, min\_samples\_split=2, min\_samples\_leaf=1, min\_weight\_fraction\_leaf=0.0, max\_features=None, random\_state=None, max\_leaf\_nodes=None, min\_impurity\_decrease=0.0, min\_impurity\_split=None, class\_weight=None, ccp\_alpha=0.0)

> https://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeRegressor.html https://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeClassifier.html

### Visualization

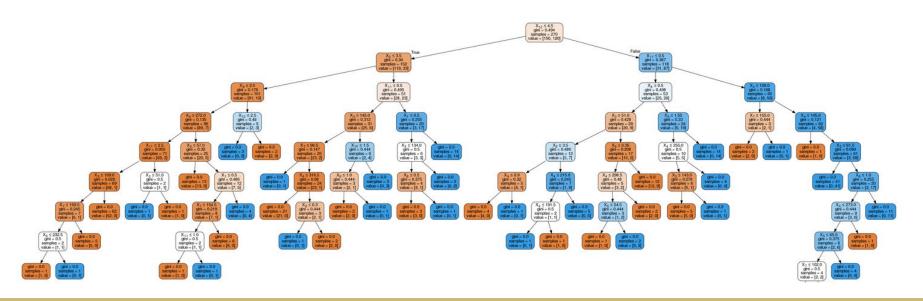
```
1 from sklearn import tree
2 tree.plot tree(clf.fit(X, y))
```

```
[Text(167.4, 181.2, 'X[12] \le 4.5 \cdot ] = 0.494 \cdot ] = 270 \cdot [150, 120]'),
Text(83.7, 108.72, 'X[2] \leq 3.5\ngini = 0.34\nsamples = 152\nvalue = [119, 33]'),
Text (251.100000000000000, 108.72, 'X[11] \le 0.5 \le 0.387 \le 118 \le [31, 87]'),
Text (292.95, 36.2399999999999, 'qini = 0.168\nsamples = 65\nvalue = [6, 59]')
```



### Visualization

```
from sklearn.tree import export graphviz
import graphviz
from six import StringIO
from IPython.display import Image
import pydotplus
dot data = StringIO()
export graphviz(clf, out file=dot data, filled=True, rounded=True, special characters=True)
graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
Image(graph.create png())
```



### Visualization

```
clf = DecisionTreeClassifier(max depth=2, random state=0)
2 clf.fit(X,y)
3 dot data = StringIO()
  export graphviz(clf, out file=dot data, filled=True, rounded=True, special characters=True)
5 graph = pydotplus.graph from dot data(dot data.getvalue())
6 Image(graph.create png())
                                   X_{12} \le 4.5
                                  gini = 0.494
                                 samples = 270
                               value = [150, 120]
                              True
                                              False
                       X_2 \le 3.5 gini = 0.34
                                               X_{11} \le 0.5
                                              gini = 0.387
            7
                      samples = 152
                                            samples = 118
                                           value = [31, 87]
                     value = [119, 33]
                       gini = 0.495
  gini = 0.178
                                             gini = 0.498
                                                                  gini = 0.168
samples = 101
                       samples = 51
                                            samples = 53
                                                                 samples = 65
value = [91, 10]
                      value = [28, 23]
                                           value = [25, 28]
                                                                 value = [6, 59]
```

# Ways to prevent overfitting

Decision trees are easy to overfit

- **Early Stopping**
- Pruning
- Ensembling

# Ways to prevent overfitting: Early stopping

```
max depth
             The maximum depth of the tree
min_samples_split The minimum number of samples required to split an internal node
min_samples_leaf The minimum number of samples required to be at a leaf node
                            The minimum weighted fraction of the sum total of
min weight fraction leaf
                            weights required to be at a leaf node
min_impurity_decrease A node will be split if this split induces a decrease of the
                         impurity greater than or equal to this value
```

max features The number of features to consider when looking for the best split

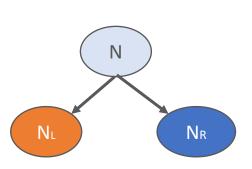
### Hyperparameters

max\_depth The maximum depth of the tree

min\_samples\_leaf The minimum number of samples required to be at a leaf node

min\_impurity\_decrease

A node will be split if this split induces a decrease of the impurity greater than or equal to this value



$$H_0 - \left(\frac{N_L}{N}, H_L + \frac{N_R}{N}, H_R\right)$$

# Other useful options

max\_features The number of features to consider when looking for the best split

sgrt lez

class\_weight None, balanced, custom

Complexity parameter for minimal cost-complexity pruning ccp\_alpha

### **Grid Search**

```
1 from sklearn.model selection import GridSearchCV
 1 rf = DecisionTreeClassifier()
 2 parameters = {'max_depth':[3,5,7,10], 'min_samples_leaf':[1,2,5,10]}
 3 clf = GridSearchCV(rf, parameters)
 1 clf.fit(X,y)
GridSearchCV(estimator=DecisionTreeClassifier(),
             param grid={'max depth': [3, 5, 7, 10],
                         'min samples leaf': [1, 2, 5, 10]})
    clf.best estimator
DecisionTreeClassifier(max depth=3, min samples leaf=5)
   clf.best_score_
0.8111111111111111
```

