# FINALPROJECT\_jupyter

#### March 22, 2018

```
In [1]: import scipy.io as sio
        import matplotlib.pyplot as plt
        import numpy as np
        import seaborn as sns
        import pandas as pd
        from sklearn import svm
        from sklearn import tree
        from sklearn.model_selection import GridSearchCV
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.datasets import make_classification
In [2]: def convert_iris(label):
            Function to convert the labels from string to numeric
            if label == "Iris-virginica":
                return 1.0
            elif label == "Iris-setosa":
                return 0.0
            elif label == "Iris-versicolor":
                return 0.0
            else:
                return label
        def convert_breast(label):
            11 11 11
            Function to convert the labels from string to numeric
            if label == "M":
                return 1.0
            elif label == "B":
                return 0.0
            else:
                return label
        def simple_cross_validation(X_train_val, Y_train_val, k, fold):
```

```
11 11 11
    A simple cross-validation function for k-NN.
    X_train_val: Features for train and val set.
                 Shape: (num of data points, num of features)
    Y_train_val: Labels for train and val set.
                 Shape: (num of data points,)
    k:
                 Parameter k for k-NN.
                 The number of folds to do the cross-validation.
    fold:
    Return the average accuracy on validation set.
    val_acc_list = []
    train_acc_list = []
    for i in range(0,fold):
        size_split = int(X_train_val.shape[0]/fold)
        if i == 0:
            X_test_val_s = X_train_val[0:size_split,:]
            X_train_val_s = X_train_val[size_split:X_train_val.shape[0],:]
            Y_test_val_s = Y_train_val[0:size_split,]
            Y_train_val_s = Y_train_val[size_split:Y_train_val.shape[0],]
        else:
            X_test_val_s = X_train_val[i*size_split:i*size_split+size_split,:]
            train_beg_s = X_train_val[0:i*size_split,:]
            train_end s = X_train_val[i*size_split+size_split:X_train_val.shape[0],:]
            X_train_val_s = np.concatenate((train_beg_s,train_end_s))
        classifier1 = KNeighborsClassifier(algorithm='brute', n_neighbors=k)
        classifier2 = KNeighborsClassifier(algorithm='brute', n neighbors=k)
        classifier1.fit(X_test_val_s,Y_test_val_s.ravel())
        classifier2.fit(X_train_val_s,Y_train_val_s.ravel())
        val_acc_list.append(classifier1.score(X_test_val_s,Y_test_val_s))
        train_acc_list.append(classifier2.score(X_train_val_s,Y_train_val_s))
    return sum(val_acc_list) / len(val_acc_list), \
           sum(train_acc_list) / len(train_acc_list)
def simple_GridSearchCV_fit(X_train_val, Y_train_val, k_list, fold):
    A simple grid search function for k with cross-validation in k-NN.
    X_train_val: Features for train and val set.
                 Shape: (num of data points, num of features)
    Y_train_val: Labels for train and val set.
```

```
Shape: (num of data points,)
    k_list:
                 The list of k values to try.
    fold:
                 The number of folds to do the cross-validation.
    Return the val and train accuracy matrix of cross-validation.
    All combinations of k are included in the array.
    Shape: (len(k list), )
    11 11 11
    shape = len(k_list)
    val_acc_array = np.zeros(shape)
    train_acc_array = np.zeros(shape)
    for i in range(0,shape):
        val_acc_array[i], train_acc_array[i] = simple_cross_validation(
            X_train_val, Y_train_val, k_list[i], fold)
    return val_acc_array, train_acc_array
def draw_heatmap_knn(acc, acc_desc, k_list):
    Draw heatmap for k-nearest neighbors hyper parameters
   plt.figure(figsize = (2,4))
    ax = sns.heatmap(acc, annot=True, fmt='.3f', yticklabels=k_list, xticklabels=[])
    ax.collections[0].colorbar.set_label("accuracy")
    ax.set(ylabel='$k$')
   plt.title(acc_desc + ' w.r.t $k$')
    sns.set_style("whitegrid", {'axes.grid' : False})
    plt.show()
def draw_heatmap_linear(acc, acc_desc, depth_list):
    Draw heatmap for linear classifier hyper parameters
   plt.figure(figsize = (2,4))
    ax = sns.heatmap(acc, annot=True, fmt='.3f', yticklabels=depth_list, xticklabels=[
    ax.collections[0].colorbar.set_label("accuracy")
    ax.set(ylabel='depth')
   plt.title(acc_desc + ' w.r.t depth')
    sns.set_style("whitegrid", {'axes.grid' : False})
    plt.show()
```

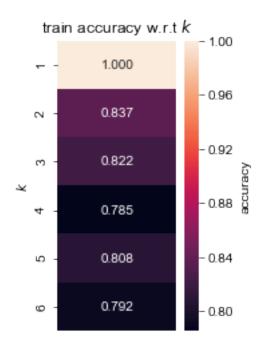
#### 1 Iris Dataset

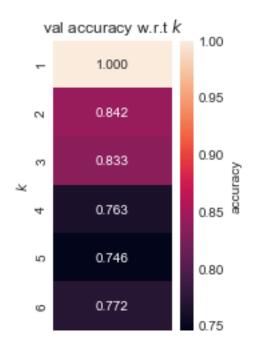
```
In [3]: iris = pd.read_csv('http://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris
    Clean and Split Dataset:
In [4]: iris['Iris-setosa'] = iris['Iris-setosa'].apply(convert_iris)
```

iris\_XY = iris.iloc[:,0:5].values

```
np.random.shuffle(iris_XY)
        iris_X = iris_XY[:,0:4]
        iris_Y = iris_XY[:,4:5]
        num_training = int(0.8*iris_X.shape[0])
        num_testing = int(0.2*iris_X.shape[0])
        iris_X_train = iris_X[:num_training]
        iris_Y_train = iris_Y[:num_training]
        iris_X_test = iris_X[num_training:]
        iris_Y_test = iris_Y[num_training:]
        print(iris_X_train.shape)
        print(iris_Y_train.shape)
        print(iris_X_test.shape)
        print(iris_Y_test.shape)
(119, 4)
(119, 1)
(30, 4)
(30, 1)
```

# 2 Performing KNN On Iris Data Set



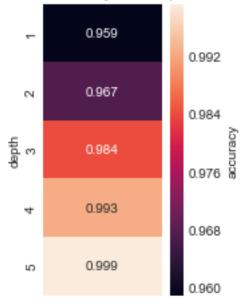


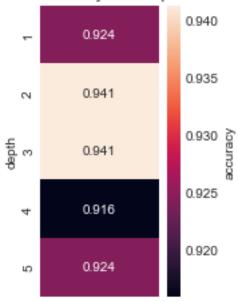
```
asserted = []
for i in range(0,preds.shape[0]):
    if preds[i] == iris_Y_test[i]:
        asserted.append(preds[i])
    test_accuracy = len(asserted)/preds.shape[0]
    print("Test Accuracy: {0:f}".format(test_accuracy))
Test Accuracy: 1.000000
```

# 3 Performing Decision Trees On Iris Data Set

```
In [8]: classifier = tree.DecisionTreeClassifier(
                            criterion='entropy')
        depth_list = [1, 2, 3, 4, 5]
        params = {"max_depth": depth_list}
        grid_dt_iris = GridSearchCV(classifier, params,
                                    return_train_score = True,
                                    cv = 10)
        grid_dt_iris.fit(iris_X_train, iris_Y_train)
Out[8]: GridSearchCV(cv=10, error_score='raise',
               estimator=DecisionTreeClassifier(class_weight=None, criterion='entropy', max_de
                    max_features=None, max_leaf_nodes=None,
                    min_impurity_decrease=0.0, min_impurity_split=None,
                    min_samples_leaf=1, min_samples_split=2,
                    min_weight_fraction_leaf=0.0, presort=False, random_state=None,
                    splitter='best'),
               fit_params=None, iid=True, n_jobs=1,
               param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
               refit=True, return_train_score=True, scoring=None, verbose=0)
In [9]: train_acc = grid_dt_iris.cv_results_['mean_train_score'].reshape(-1,1)
        draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
        val_acc = grid_dt_iris.cv_results_['mean_test_score'].reshape(-1,1)
        draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```

#### train accuracy w.r.t depth





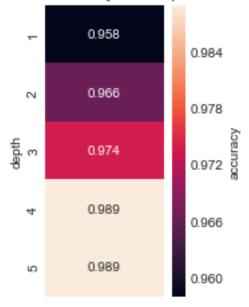
```
asserted.append(pred[i])
    test_accuracy = len(asserted)/pred.shape[0]
    print("Test Accuracy: {0:f}".format(test_accuracy))
    print(grid_dt_iris.best_params_)

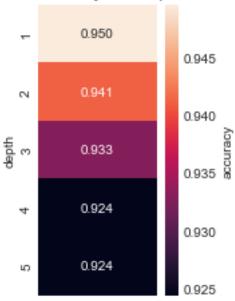
Test Accuracy: 0.933333
{'max_depth': 2}
```

# 4 Performing Random Forests On Iris Data Set

```
In [11]: depth_list = [1, 2, 3, 4, 5]
         params = {"max_depth": depth_list}
         classifier = RandomForestClassifier(max_depth=5,
                                             random_state=0)
         grid_rf_iris = GridSearchCV(classifier, params,
                                     return_train_score = True, cv = 10)
         grid_rf_iris.fit(iris_X_train, iris_Y_train.ravel())
Out[11]: GridSearchCV(cv=10, error_score='raise',
                estimator=RandomForestClassifier(bootstrap=True, class_weight=None, criterion=
                     max_depth=5, max_features='auto', max_leaf_nodes=None,
                     min_impurity_decrease=0.0, min_impurity_split=None,
                     min_samples_leaf=1, min_samples_split=2,
                     min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
                     oob_score=False, random_state=0, verbose=0, warm_start=False),
                fit_params=None, iid=True, n_jobs=1,
                param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
                refit=True, return_train_score=True, scoring=None, verbose=0)
In [12]: train_acc = grid_rf_iris.cv_results_['mean_train_score'].reshape(-1,1)
         draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
         val acc = grid rf_iris.cv_results_['mean_test_score'].reshape(-1,1)
         draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```

#### train accuracy w.r.t depth

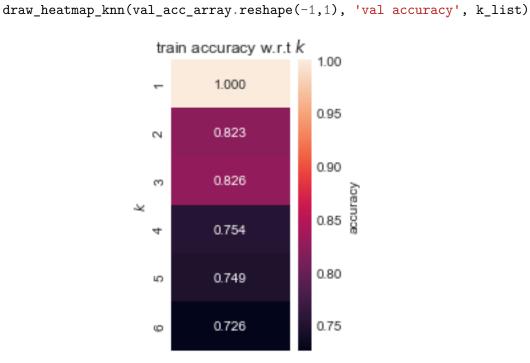


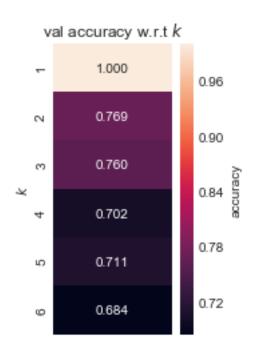


```
test_accuracy = len(asserted)/pred.shape[0]
         print("Test Accuracy: {0:f}".format(test_accuracy))
         print(grid_rf_iris.get_params())
Test Accuracy: 0.966667
{'cv': 10, 'error_score': 'raise', 'estimator__bootstrap': True, 'estimator__class_weight': No:
            max_depth=5, max_features='auto', max_leaf_nodes=None,
            min_impurity_decrease=0.0, min_impurity_split=None,
            min_samples_leaf=1, min_samples_split=2,
            min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
            oob_score=False, random_state=0, verbose=0, warm_start=False), 'fit_params': None,
  Breast Cancer Dataset
In [14]: breast = pd.read_csv('http://archive.ics.uci.edu/ml/machine-learning-databases/breast
  Clean and Split Dataset:
In [15]: breast.dropna(inplace = True)
         breast['M'] = breast['M'].apply(convert_breast)
         breast_XY = pd.concat([breast.iloc[:,0:1],breast.iloc[:,2:],
                                breast.iloc[:,1:2]],axis=1).values
         np.random.shuffle(breast_XY)
         breast_X = breast_XY[:,0:31]
         breast_Y = breast_XY[:,31:]
         num_training = int(0.8*breast_X.shape[0])
         num_testing = int(0.2*breast_X.shape[0])
         breast_X_train = breast_X[:num_training]
         breast_Y_train = breast_Y[:num_training]
         breast_X_test = breast_X[num_training:]
         breast_Y_test = breast_Y[num_training:]
         print(breast_X_train.shape)
         print(breast_Y_train.shape)
         print(breast_X_test.shape)
         print(breast_Y_test.shape)
(454, 31)
(454, 1)
(114, 31)
(114, 1)
```

asserted.append(pred[i])

# 6 Performing KNN On Breast Cancer Data Set

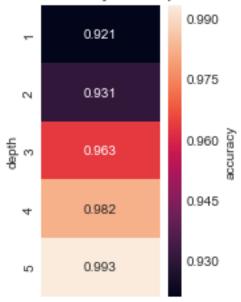


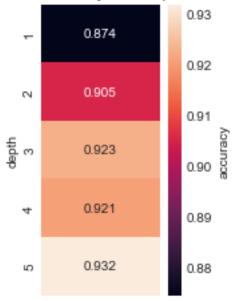


# 7 Performing Decision Trees On Breast Cancer Data Set

```
In [19]: classifier = tree.DecisionTreeClassifier(criterion='entropy')
                         depth_list = [1, 2, 3, 4, 5]
                         params = {"max_depth": depth_list}
                         grid_dt_breast = GridSearchCV(classifier, params, return_train_score = True, cv = 10)
                         grid_dt_breast.fit(breast_X_train, breast_Y_train)
Out[19]: GridSearchCV(cv=10, error_score='raise',
                                              estimator=DecisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClass_weight=None, criterion='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision='entropy', max_decision=
                                                             max_features=None, max_leaf_nodes=None,
                                                             min_impurity_decrease=0.0, min_impurity_split=None,
                                                             min_samples_leaf=1, min_samples_split=2,
                                                            min_weight_fraction_leaf=0.0, presort=False, random_state=None,
                                                             splitter='best'),
                                              fit_params=None, iid=True, n_jobs=1,
                                              param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
                                              refit=True, return_train_score=True, scoring=None, verbose=0)
In [20]: train_acc = grid_dt_breast.cv_results_['mean_train_score'].reshape(-1,1)
                         draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
                         val_acc = grid_dt_breast.cv_results_['mean_test_score'].reshape(-1,1)
                          draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```







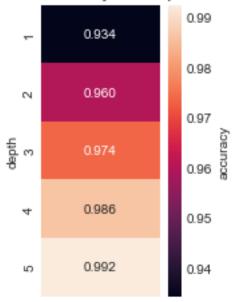
```
asserted.append(pred[i])
  test_accuracy = len(asserted)/pred.shape[0]
  print(test_accuracy)
  print(grid_dt_breast.best_params_)

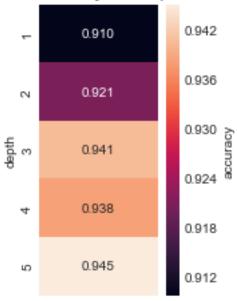
0.9385964912280702
{'max_depth': 5}
```

# 8 Performing Random Forests On Breast Cancer Data Set

```
In [22]: depth_list = [1, 2, 3, 4, 5]
         params = {"max_depth": depth_list}
         classifier = RandomForestClassifier(max_depth=5, random_state=0)
         grid_rf_breast = GridSearchCV(classifier, params, return_train_score = True, cv = 10)
         grid_rf_breast.fit(breast_X_train, breast_Y_train.ravel())
Out[22]: GridSearchCV(cv=10, error_score='raise',
                estimator=RandomForestClassifier(bootstrap=True, class_weight=None, criterion=
                     max_depth=5, max_features='auto', max_leaf_nodes=None,
                     min_impurity_decrease=0.0, min_impurity_split=None,
                     min_samples_leaf=1, min_samples_split=2,
                     min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
                     oob_score=False, random_state=0, verbose=0, warm_start=False),
                fit_params=None, iid=True, n_jobs=1,
                param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
                refit=True, return_train_score=True, scoring=None, verbose=0)
In [23]: train_acc = grid_rf_breast.cv_results_['mean_train_score'].reshape(-1,1)
         draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
         val_acc = grid_rf_breast.cv_results_['mean_test_score'].reshape(-1,1)
         draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```



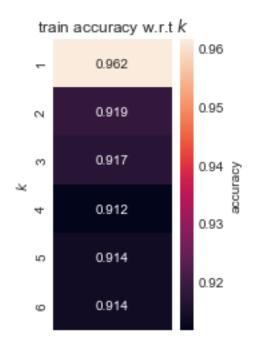


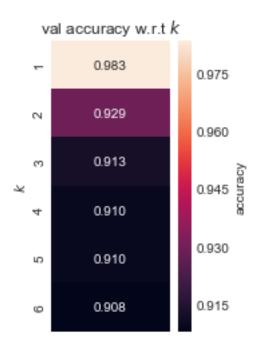


# 9 Mammographic Masses Dataset

```
In [25]: mammograph = pd.read_csv('http://archive.ics.uci.edu/ml/machine-learning-databases/ma
         mammograph = mammograph.replace({'?':np.nan}).dropna()
         mammograph_XY = mammograph.iloc[:,0:5].values
         np.random.shuffle(mammograph_XY)
         mammograph_X = mammograph_XY[:,0:4]
         mammograph_Y = mammograph_XY[:,4:5]
         num_training = int(0.8*mammograph_X.shape[0])
         num_testing = int(0.2*mammograph_X.shape[0])
         mam_X_train = mammograph_X[:num_training]
         mam_Y_train = mammograph_Y[:num_training]
         mam_X_test = mammograph_X[num_training:]
         mam_Y_test = mammograph_Y[num_training:]
         print(mam_X_train.shape)
         print(mam_Y_train.shape)
         print(mam_X_test.shape)
         print(mam_Y_test.shape)
(663, 4)
(663, 1)
(166, 4)
(166, 1)
```

# 10 Performing KNN On Mammograph Data Set

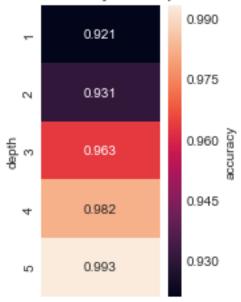


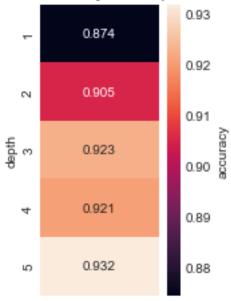


# 11 Performing Decision Trees On Mammograph Data Set

```
In [29]: classifier = tree.DecisionTreeClassifier(criterion='entropy')
                          depth_list = [1, 2, 3, 4, 5]
                          params = {"max_depth": depth_list}
                          grid_dt_mam = GridSearchCV(classifier, params, return_train_score = True, cv = 5)
                          grid_dt_mam.fit(mam_X_train, mam_Y_train)
Out[29]: GridSearchCV(cv=5, error_score='raise',
                                              estimator=DecisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClassifier(class_weight=None, criterion='entropy', max_decisionTreeClass_weight=None, criterion='entropy', max_decision='
                                                             max_features=None, max_leaf_nodes=None,
                                                             min_impurity_decrease=0.0, min_impurity_split=None,
                                                             min_samples_leaf=1, min_samples_split=2,
                                                             min_weight_fraction_leaf=0.0, presort=False, random_state=None,
                                                             splitter='best'),
                                              fit_params=None, iid=True, n_jobs=1,
                                              param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
                                              refit=True, return_train_score=True, scoring=None, verbose=0)
In [30]: train_acc = grid_dt_breast.cv_results_['mean_train_score'].reshape(-1,1)
                          draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
                          val_acc = grid_dt_breast.cv_results_['mean_test_score'].reshape(-1,1)
                          draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```







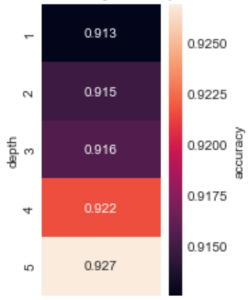
```
asserted.append(pred[i])
    test_accuracy = len(asserted)/pred.shape[0]
    print("Test Accuracy: {0:f}".format(test_accuracy))
    print(grid_dt_mam.best_params_)

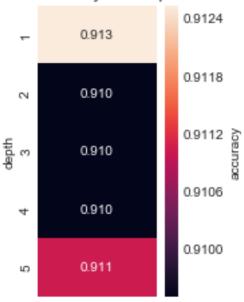
Test Accuracy: 0.897590
{'max_depth': 1}
```

# 12 Performing Random Forests On Mammograph Data Set

```
In [32]: depth_list = [1, 2, 3, 4, 5]
         params = {"max_depth": depth_list}
         classifier = RandomForestClassifier(max_depth=5, random_state=0)
         grid_rf_mam = GridSearchCV(classifier, params, return_train_score = True)
         grid_rf_mam.fit(mam_X_train, mam_Y_train.ravel())
Out[32]: GridSearchCV(cv=None, error_score='raise',
                estimator=RandomForestClassifier(bootstrap=True, class_weight=None, criterion=
                     max_depth=5, max_features='auto', max_leaf_nodes=None,
                     min_impurity_decrease=0.0, min_impurity_split=None,
                     min_samples_leaf=1, min_samples_split=2,
                     min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
                     oob_score=False, random_state=0, verbose=0, warm_start=False),
                fit params=None, iid=True, n jobs=1,
                param_grid={'max_depth': [1, 2, 3, 4, 5]}, pre_dispatch='2*n_jobs',
                refit=True, return_train_score=True, scoring=None, verbose=0)
In [33]: train_acc = grid_rf_mam.cv_results_['mean_train_score'].reshape(-1,1)
         draw_heatmap_linear(train_acc, 'train accuracy', depth_list)
         val_acc = grid_rf_mam.cv_results_['mean_test_score'].reshape(-1,1)
         draw_heatmap_linear(val_acc, 'val accuracy', depth_list)
```







```
asserted.append(pred[i])
    test_accuracy = len(asserted)/pred.shape[0]
    print("Test Accuracy: {0:f}".format(test_accuracy))
    print(grid_rf_mam.get_params())

Test Accuracy: 0.897590
{'cv': None, 'error_score': 'raise', 'estimator__bootstrap': True, 'estimator__class_weight': max_depth=5, max_features='auto', max_leaf_nodes=None,
    min_impurity_decrease=0.0, min_impurity_split=None,
    min_samples_leaf=1, min_samples_split=2,
    min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
    oob_score=False, random_state=0, verbose=0, warm_start=False), 'fit_params': None,
```