```
In [10]: import pandas as pd
         import numpy as np
         import urllib
         import scipy.optimize
         import random
         import urllib.request
         from sklearn import ensemble
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.model selection import cross val score
         from sklearn.model selection import GridSearchCV
         from sklearn.ensemble import AdaBoostClassifier
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import BaggingClassifier
         from sklearn.neighbors import KNeighborsClassifier
         #here we get the data
         df = pd.read_csv('iris.csv',
                          sep=r'\s*,\s*',engine = 'python', na_values = '?')
         df.dropna()
         X Y chart = pd.get dummies(df, drop first=False)
         X_Y_chart.drop(X_Y_chart.columns[len(X_Y_chart.columns)-1], axis=1, inpl
         ace=True)
         print(df.shape)
         print(X_Y_chart.shape)
         X and Y = X Y chart.values
         #np.random.shuffle(X and Y)
         X = X and Y[:, 0:-1]
         Y = X and Y[:, -1]
         print(X Y chart[:0])
         import matplotlib.pyplot as plt
         import seaborn as sns
         def draw heatmap(acc, acc desc, k list):
             plt.figure(figsize = (2,4))
             ax = sns.heatmap(acc, annot=True, fmt='.3f', yticklabels=k list, xti
         cklabels=[])
             ax.collections[0].colorbar.set label("accuracy")
             ax.set(ylabel='$params$')
             plt.title(acc desc)
             sns.set style("whitegrid", {'axes.grid': False})
             plt.show()
         (150, 5)
         (150, 5)
         Empty DataFrame
         Columns: [11, w1, 12, w2, versicolor Iris-versicolor]
         Index: []
```

(120, 4) (30, 4) (120,) (30,)

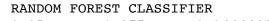
```
In [12]: random train acc = []
         adaboost train acc = []
         gradient_train_acc = []
         bagging_train_acc = []
          decision_train_acc = []
          random_test_acc = []
          adaboost_test_acc = []
         gradient test acc = []
         bagging_test_acc = []
         decision_test_acc = []
          i = 1;
         while i < 4:
              #here we split the data again
              np.random.shuffle(X and Y)
              X = X_and_Y[:, 0:-1]
              Y = X_and_Y[:, -1]
              X \text{ train val} = X[:int(0.8*len(X))]
              X_{\text{test}} = X[int(0.8*len(X)):]
              Y_train_val = Y[:int(0.8*len(Y))]
              Y \text{ test} = Y[\text{int}(0.8*\text{len}(Y)):]
              print(i)
              #then we run it on random forest
              random = RandomForestClassifier()
              print("\n RANDOM FOREST CLASSIFIER")
              #first grid search / cross validation for best parameter
              n_{list} = [1, 25, 50, 100, 150]
              parameters random = {'n estimators' : n list}
              clf random = GridSearchCV(random, parameters random, cv=3)
              clf random.fit(X train val, Y train val)
              #display result with heatmap
              print(clf_random.cv_results_['mean_test_score'])
              draw heatmap(clf random.cv results ['mean test score'].reshape(len(n
          list),1), "random training", n list)
              #choose the best parameter and train again
              random = RandomForestClassifier(n estimators = clf random.best param
         s_['n_estimators'])
              random = random.fit(X train val, Y train val)
              #calculate the test error
              test predict = random.predict(X test)
              test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
         st))])/len(Y_test)
              train_predict = random.predict(X train val)
              train acc = sum([train predict[i] == Y train val[i] for i in range(l
         en(Y train val))])/len(Y train val)
              print("random test accuracy with "+str(clf_random.best params ['n es
         timators'])+" is: " + str(test acc))
              random_test_acc.append(test_acc)
              random train acc.append(train acc)
              #run it with Adaboost
              dt = DecisionTreeClassifier()
              adaboost = AdaBoostClassifier(base estimator=dt)
              print("\n ADABOOST CLASSIFIER WITH DECISION TREE")
              #first grid search / cross validation for best parameter
              e list = [10,60,120,200,300]
              parameters adaboost = {'n estimators' : e list}
```

```
clf_adaboost = GridSearchCV(adaboost, parameters_adaboost, cv=3)
   clf_adaboost.fit(X_train_val,Y_train_val)
   #display result with heatmap
   print(clf_adaboost.cv_results_['mean_test_score'])
   draw_heatmap(clf_adaboost.cv_results_['mean_test_score'].reshape(len
(e_list),1), "adaboost_training", e_list)
    #choose the best parameter and train again
   adaboost = AdaBoostClassifier(base_estimator=dt, n_estimators=clf_ad
aboost.best_params_['n_estimators'])
   adaboost = adaboost.fit(X train val, Y train val)
   #calculate the test error
   test_predict = adaboost.predict(X_test)
   train_predict = adaboost.predict(X_train_val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
   print("Adaboost test accuracy with "+str(clf_adaboost.best_params_[
'n_estimators'])+" is: " + str(test_acc))
   adaboost_test_acc.append(test_acc)
   adaboost_train_acc.append(train_acc)
   #run it with gradient
   gradient = ensemble.GradientBoostingClassifier()
   print("\n GRADIENT BOOSTING CLASSIFIER")
   parameters_gradient = {'n_estimators': e_list}
   #first grid search / cross validation for best parameter
   clf gradient = GridSearchCV(gradient, parameters gradient, cv=3)
   clf gradient.fit(X train val, Y train val)
   #display result with heatmap
   print(clf_gradient.cv_results_['mean_test_score'])
   draw heatmap(clf gradient.cv results ['mean test score'].reshape(len
(e_list),1), "gradient_training", e_list)
   #choose the best parameter and train again
   gradient = ensemble.GradientBoostingClassifier(n estimators=clf grad
ient.best params ['n estimators'])
   gradient = gradient.fit(X_train_val,Y_train_val)
    #calculate the test error
   test predict = gradient.predict(X test)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y test)
   train predict = gradient.predict(X train val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   print("Gradient test accuracy with "+str(clf gradient.best params [
'n estimators'])+" is: " + str(test acc))
   gradient_test_acc.append(test_acc)
   gradient train acc.append(train acc)
   #run it with Bagging Family with KNeighborClassifer
   bagging = BaggingClassifier(KNeighborsClassifier())
   print("\n BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER")
   s_list = [0.15, 0.3, 0.45, 0.6, 0.75]
   parameters bagging = {'max samples' : s list}
   clf_bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
   clf_bagging.fit(X_train_val, Y_train_val)
   print("max_samples: " + str(clf_bagging.cv_results_['mean_test_scor
```

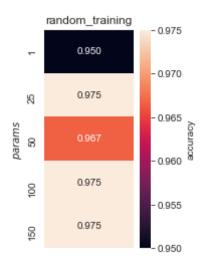
```
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
s_list),1),"Kneighbor_s_list_training", s_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=clf_
bagging.best_params_['max_samples'])
    best_max_samples = clf_bagging.best_params_['max_samples']
    f_{list} = [0.25, 0.4, 0.5, 0.6, 0.75]
    parameters_bagging = {'max_features': f_list}
    clf bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
    clf_bagging.fit(X_train_val, Y_train_val)
    print("max_features: " + str(clf_bagging.cv_results_['mean_test_scor
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
f_list),1),"Kneighbor_f_list_training", f_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=best
_max_samples,
                                max_features= clf_bagging.best_params_[
'max_features'])
    bagging = bagging.fit(X_train_val, Y_train_val)
    test_predict = bagging.predict(X_test)
    train_predict = bagging.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
    test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
    print("Kneighbor Bagging test accuracy with "+str(best_max_samples)+
          " and "+ str(clf bagging.best params ['max features'])+" is: "
 + str(test_acc))
    bagging_test_acc.append(test_acc)
    bagging train acc.append(train acc)
    #run it with Bagging Family with Decision tree
    decision = BaggingClassifier()
    parameters_decision = {'max_samples' : s_list}
    print("\n BAGGING CLASSIFIER WITH DECISION TREE")
    clf_decision = GridSearchCV(decision, parameters_decision, cv=3)
    clf decision.fit(X train val, Y train val)
   print("max samples: " + str(clf decision.cv results ['mean test scor
e']))
    draw_heatmap(clf_decision.cv_results_['mean_test_score'].reshape(len
(s list),1), "Decision s list training", s list)
    #choose the best parameter and train again for max features
    decision = BaggingClassifier(max_samples=clf_decision.best_params_[
'max_samples'])
    best max samples = clf decision.best params ['max samples']
    parameters_decision = {'max_features': f_list}
    clf_decision = GridSearchCV(decision, parameters_decision, cv=3)
    clf decision.fit(X train val, Y train val)
    print("max_features: " + str(clf_decision.cv_results_['mean_test_sco')
re']))
    draw heatmap(clf decision.cv results ['mean test score'].reshape(len
(f_list),1),"Decision_f_list_training", f_list)
    #choose the best parameter and train again for max_features
    decision = BaggingClassifier(max samples=best max samples,
```

```
max features= clf decision.best params [
'max features'])
    decision = decision.fit(X_train_val, Y_train_val)
    test predict = decision.predict(X test)
    train_predict = decision.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y train val)))/len(Y train val)
    test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
st))])/len(Y test)
    print("Decision Bagging test accuracy with "+str(best max samples)+
          " and "+ str(clf decision.best params ['max features'])+" is:
 " + str(test_acc))
    decision test acc.append(test acc)
    decision train acc.append(train acc)
    i = i+1;
    print("\n\n\n")
print("Random Forest Classifier 80/20 training accuracy:")
print(random train acc)
print("Adaboost Classifier 80/20 training accuracy:")
print(adaboost train acc)
print("Gradient Boost Classifier 80/20 training accuracy:")
print(gradient train acc)
print("Kneigher Bagging Classifier 80/20 training accuracy:")
print(bagging train acc)
print("Decision Tree Bagging Classifier 80/20 training accuracy:")
print(decision train acc)
print("Random Forest Classifier 80/20 test accuracy:
                                                             "+ str(sum(
random test acc)/len(random test acc)))
print("Adaboost Classifier 80/20 test accuracy:
                                                             "+ str(sum(
adaboost_test_acc)/len(adaboost_test_acc)))
print("Gradient Boost Classifier 80/20 test accuracy:
                                                             "+ str(sum(
gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 80/20 test accuracy:
                                                             "+ str(sum(
bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 80/20 test accuracy: "+ str(sum(
decision test acc)/len(decision test acc)))
print("Random Forest Classifier 80/20 test error:
                                                             "+ str(1-su
m(random test acc)/len(random test acc)))
print("Adaboost Classifier 80/20 test error:
                                                             "+ str(1-su
m(adaboost test acc)/len(adaboost test acc)))
print("Gradient Boost Classifier 80/20 test error:
                                                             "+ str(1-su
m(gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 80/20 test error:
                                                             "+ str(1-su
m(bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 80/20 test error:
                                                             "+ str(1-su
m(decision test acc)/len(decision test acc)))
```

1

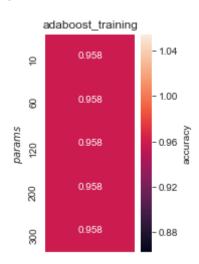


[0.95 0.975 0.96666667 0.975 0.975



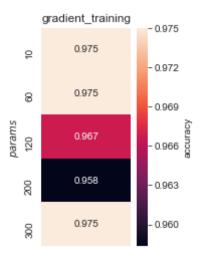
random test accuracy with 25 is: 0.9

ADABOOST CLASSIFIER WITH DECISION TREE
[0.95833333 0.95833333 0.95833333 0.95833333]

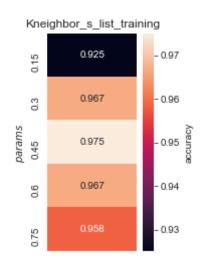


GRADIENT BOOSTING CLASSIFIER

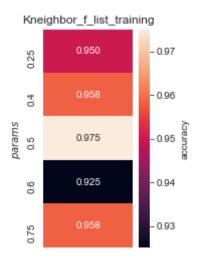
[0.975 0.96666667 0.95833333 0.975]



BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER

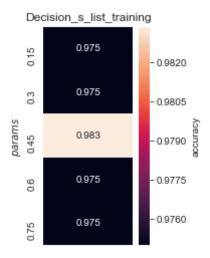


max features: [0.95 0.95833333 0.975 0.925 0.95833333]



Kneighbor Bagging test accuracy with 0.45 and 0.5 is: 0.9

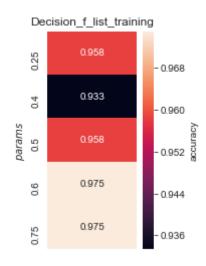
BAGGING CLASSIFIER WITH DECISION TREE



max_features: [0.95833333 0.93333333 0.95833333 0.975

0.975

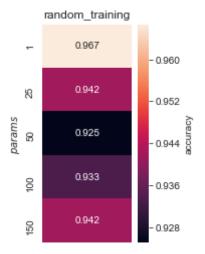
]



Decision Bagging test accuracy with 0.45 and 0.6 is: 0.9

2

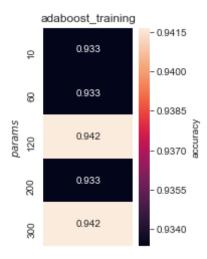
RANDOM FOREST CLASSIFIER [0.96666667 0.94166667 0.925 0.93333333 0.94166667]



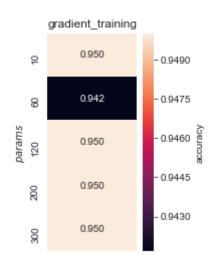
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random test accuracy with 1 is: 0.9666666666666667

ADABOOST CLASSIFIER WITH DECISION TREE
[0.93333333 0.93333333 0.94166667 0.93333333 0.94166667]

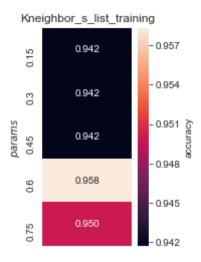


Adaboost test accuracy with 120 is: 0.9666666666666667



Gradient test accuracy with 10 is: 1.0

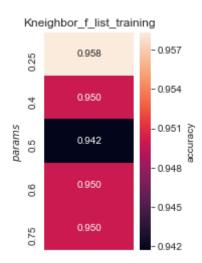
BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER
max_samples: [0.94166667 0.94166667 0.94166667 0.95833333 0.95]



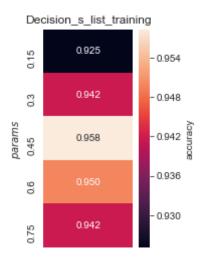
max_features: [0.95833333 0.95 0.94166667 0.95

0.95

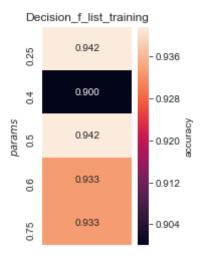
]



BAGGING CLASSIFIER WITH DECISION TREE max_samples: [0.925 0.94166667] 0.94166667 0.95833333 0.95

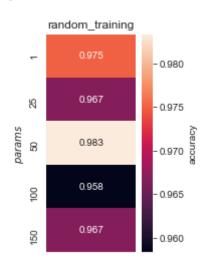


max_features: [0.94166667 0.9 0.94166667 0.93333333 0.93333333]



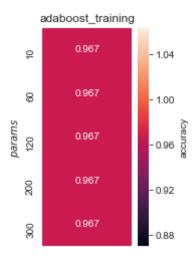
Decision Bagging test accuracy with 0.45 and 0.25 is: 1.0

3



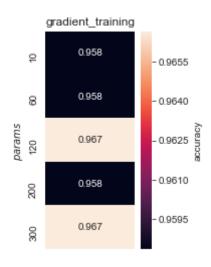
random test accuracy with 50 is: 0.9

ADABOOST CLASSIFIER WITH DECISION TREE
[0.96666667 0.96666667 0.96666667 0.96666667]



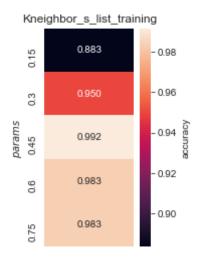
Adaboost test accuracy with 10 is: 0.9

GRADIENT BOOSTING CLASSIFIER
[0.95833333 0.95833333 0.96666667 0.95833333 0.96666667]

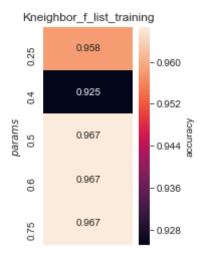


Gradient test accuracy with 120 is: 0.8666666666666667

BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER max_samples: [0.88333333 0.95 0.99166667 0.98333333 0.98333333]



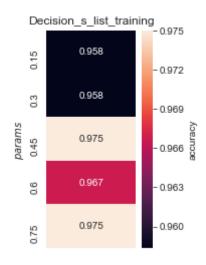
max_features: [0.95833333 0.925 0.96666667 0.96666667 0.96666667]



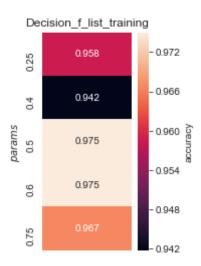
Kneighbor Bagging test accuracy with 0.45 and 0.5 is: 0.9

BAGGING CLASSIFIER WITH DECISION TREE max_samples: [0.95833333 0.95833333 0.975

0.96666667 0.975



max_features: [0.95833333 0.94166667 0.975 0.975 0.96666667]



Decision Bagging test accuracy with 0.45 and 0.5 is: 0.9

Random Forest Classifier 80/20 training accuracy: [1.0, 0.96666666666667, 1.0] Adaboost Classifier 80/20 training accuracy: [1.0, 1.0, 1.0] Gradient Boost Classifier 80/20 training accuracy: [1.0, 0.991666666666667, 1.0] Kneigher Bagging Classifier 80/20 training accuracy: [0.9833333333333333, 0.941666666666667, 0.983333333333333] Decision Tree Bagging Classifier 80/20 training accuracy: [0.991666666666667, 0.90833333333333, 0.983333333333333] Random Forest Classifier 80/20 test accuracy: 0.922222222222 Adaboost Classifier 80/20 test accuracy: 0.933333333333333 Gradient Boost Classifier 80/20 test accuracy: 0.933333333333333 Kneighbor Bagging Classifier 80/20 test accuracy: 0.9222222222222 Random Forest Classifier 80/20 test error: 0.077777777777777 Adaboost Classifier 80/20 test error: 0.06666666666666 76 Gradient Boost Classifier 80/20 test error: 0.06666666666666 Kneighbor Bagging Classifier 80/20 test error: 0.077777777777777 Decision Tree Bagging Classifier 80/20 test error: 0.06666666666666 76

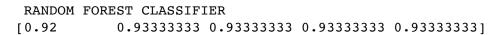
```
In [13]: random train acc = []
         adaboost train acc = []
         gradient_train_acc = []
         bagging_train_acc = []
          decision_train_acc = []
          random_test_acc = []
          adaboost_test_acc = []
         gradient test acc = []
         bagging_test_acc = []
         decision_test_acc = []
          i = 1;
         while i < 4:
              #here we split the data again
              np.random.shuffle(X and Y)
              X = X_and_Y[:, 0:-1]
              Y = X_and_Y[:, -1]
              X \text{ train val} = X[:int(0.5*len(X))]
              X_{\text{test}} = X[int(0.5*len(X)):]
              Y_{train\_val} = Y[:int(0.5*len(Y))]
              Y \text{ test} = Y[\text{int}(0.5*\text{len}(Y)):]
              print(i)
              #then we run it on random forest
              random = RandomForestClassifier()
              print("\n RANDOM FOREST CLASSIFIER")
              #first grid search / cross validation for best parameter
              n_{list} = [1, 25, 50, 100, 150]
              parameters random = {'n estimators' : n list}
              clf random = GridSearchCV(random, parameters random, cv=3)
              clf random.fit(X train val, Y train val)
              #display result with heatmap
              print(clf_random.cv_results_['mean_test_score'])
              draw heatmap(clf random.cv results ['mean test score'].reshape(len(n
          list),1), "random training", n list)
              #choose the best parameter and train again
              random = RandomForestClassifier(n estimators = clf random.best param
         s ['n estimators'])
              random = random.fit(X train val, Y train val)
              #calculate the test error
              test predict = random.predict(X test)
              test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
         st))])/len(Y_test)
              train_predict = random.predict(X train val)
              train acc = sum([train predict[i] == Y train val[i] for i in range(l
         en(Y train val))])/len(Y train val)
              print("random test accuracy with "+str(clf_random.best params ['n es
         timators'])+" is: " + str(test acc))
              random test acc.append(test acc)
              random train acc.append(train acc)
              #run it with Adaboost
              dt = DecisionTreeClassifier()
              adaboost = AdaBoostClassifier(base estimator=dt)
              print("\n ADABOOST CLASSIFIER WITH DECISION TREE")
              #first grid search / cross validation for best parameter
              e list = [10,60,120,200,300]
              parameters adaboost = {'n estimators' : e list}
```

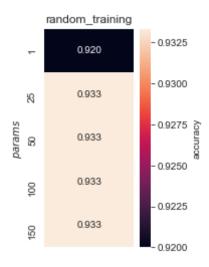
```
clf_adaboost = GridSearchCV(adaboost, parameters_adaboost, cv=3)
   clf_adaboost.fit(X_train_val,Y_train_val)
   #display result with heatmap
   print(clf adaboost.cv results ['mean test score'])
   draw heatmap(clf_adaboost.cv_results_['mean_test_score'].reshape(len
(e_list),1), "adaboost_training", e_list)
    #choose the best parameter and train again
   adaboost = AdaBoostClassifier(base_estimator=dt, n_estimators=clf_ad
aboost.best_params_['n_estimators'])
   adaboost = adaboost.fit(X train val, Y train val)
   #calculate the test error
   test_predict = adaboost.predict(X_test)
   train_predict = adaboost.predict(X_train_val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
   print("Adaboost test accuracy with "+str(clf_adaboost.best_params_[
'n_estimators'])+" is: " + str(test_acc))
   adaboost_test_acc.append(test_acc)
   adaboost_train_acc.append(train_acc)
   #run it with gradient
   gradient = ensemble.GradientBoostingClassifier()
   print("\n GRADIENT BOOSTING CLASSIFIER")
   parameters_gradient = {'n_estimators': e_list}
   #first grid search / cross validation for best parameter
   clf gradient = GridSearchCV(gradient, parameters gradient, cv=3)
   clf gradient.fit(X train val, Y train val)
   #display result with heatmap
   print(clf_gradient.cv_results_['mean_test_score'])
   draw heatmap(clf gradient.cv results ['mean test score'].reshape(len
(e_list),1), "gradient_training", e_list)
   #choose the best parameter and train again
   gradient = ensemble.GradientBoostingClassifier(n estimators=clf grad
ient.best params ['n estimators'])
   gradient = gradient.fit(X_train_val,Y_train_val)
    #calculate the test error
   test predict = gradient.predict(X test)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y test)
   train predict = gradient.predict(X train val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   print("Gradient test accuracy with "+str(clf gradient.best params [
'n estimators'])+" is: " + str(test acc))
   gradient_test_acc.append(test_acc)
   gradient train acc.append(train acc)
   #run it with Bagging Family with KNeighborClassifer
   bagging = BaggingClassifier(KNeighborsClassifier())
   print("\n BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER")
   s_list = [0.15, 0.3, 0.45, 0.6, 0.75]
   parameters bagging = {'max samples' : s list}
   clf_bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
   clf_bagging.fit(X_train_val, Y_train_val)
   print("max_samples: " + str(clf_bagging.cv_results_['mean_test_scor
```

```
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
s_list),1),"Kneighbor_s_list_training", s_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=clf_
bagging.best_params_['max_samples'])
    best_max_samples = clf_bagging.best_params_['max_samples']
    f_{list} = [0.25, 0.4, 0.5, 0.6, 0.75]
    parameters_bagging = {'max_features': f_list}
    clf bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
    clf_bagging.fit(X_train_val, Y_train_val)
    print("max_features: " + str(clf_bagging.cv_results_['mean_test_scor')
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
f_list),1),"Kneighbor_f_list_training", f_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=best
_max_samples,
                                max_features= clf_bagging.best_params_[
'max_features'])
    bagging = bagging.fit(X_train_val, Y_train_val)
    test_predict = bagging.predict(X_test)
    train_predict = bagging.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
    test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
    print("Kneighbor Bagging test accuracy with "+str(best_max_samples)+
          " and "+ str(clf bagging.best params ['max features'])+" is: "
 + str(test_acc))
    bagging_test_acc.append(test_acc)
    bagging train acc.append(train acc)
    #run it with Bagging Family with Decision tree
    decision = BaggingClassifier()
    parameters_decision = {'max_samples' : s_list}
    print("\n BAGGING CLASSIFIER WITH DECISION TREE")
    clf decision = GridSearchCV(decision, parameters decision, cv=3)
    clf decision.fit(X train val, Y train val)
   print("max samples: " + str(clf decision.cv results ['mean test scor
e']))
    draw_heatmap(clf_decision.cv_results_['mean_test_score'].reshape(len
(s_list),1),"Decision_s_list_training", s_list)
    #choose the best parameter and train again for max features
    decision = BaggingClassifier(max_samples=clf_decision.best_params_[
'max_samples'])
    best max samples = clf decision.best params ['max samples']
    parameters_decision = {'max_features': f_list}
    clf_decision = GridSearchCV(decision, parameters_decision, cv=3)
    clf decision.fit(X train val, Y train val)
    print("max_features: " + str(clf_decision.cv_results_['mean_test_sco')
re']))
    draw heatmap(clf decision.cv results ['mean test score'].reshape(len
(f_list),1),"Decision_f_list_training", f_list)
    #choose the best parameter and train again for max_features
    decision = BaggingClassifier(max samples=best max samples,
```

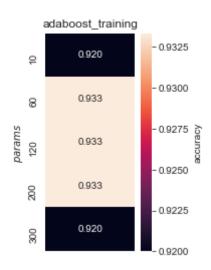
```
max features= clf decision.best params [
'max features'])
    decision = decision.fit(X_train_val, Y_train_val)
    test predict = decision.predict(X test)
    train_predict = decision.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y train val)))/len(Y train val)
    test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
st))])/len(Y test)
    print("Decision Bagging test accuracy with "+str(best max samples)+
          " and "+ str(clf decision.best params ['max features'])+" is:
 " + str(test_acc))
    decision test acc.append(test acc)
    decision train acc.append(train acc)
    i = i+1;
    print("\n\n\n")
print("Random Forest Classifier 50/50 training accuracy:")
print(random train acc)
print("Adaboost Classifier 50/50 training accuracy:")
print(adaboost train acc)
print("Gradient Boost Classifier 50/50 training accuracy:")
print(gradient train acc)
print("Kneigher Bagging Classifier 50/50 training accuracy:")
print(bagging train acc)
print("Decision Tree Bagging Classifier 50/50 training accuracy:")
print(decision train acc)
print("Random Forest Classifier 50/50 test accuracy:
                                                             "+ str(sum(
random test acc)/len(random test acc)))
print("Adaboost Classifier 50/50 test accuracy:
                                                             "+ str(sum(
adaboost_test_acc)/len(adaboost_test_acc)))
print("Gradient Boost Classifier 50/50 test accuracy:
                                                             "+ str(sum(
gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 50/50 test accuracy:
                                                             "+ str(sum(
bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 50/50 test accuracy: "+ str(sum(
decision test acc)/len(decision test acc)))
print("Random Forest Classifier 50/50 test error:
                                                             "+ str(1-su
m(random test acc)/len(random test acc)))
print("Adaboost Classifier 50/50 test error:
                                                             "+ str(1-su
m(adaboost test acc)/len(adaboost test acc)))
print("Gradient Boost Classifier 50/50 test error:
                                                             "+ str(1-su
m(gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 50/50 test error:
                                                             "+ str(1-su
m(bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 50/50 test error:
                                                             "+ str(1-su
m(decision test acc)/len(decision test acc)))
```

1



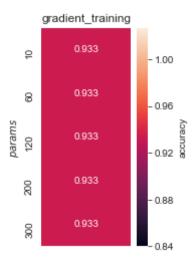


random test accuracy with 25 is: 0.96

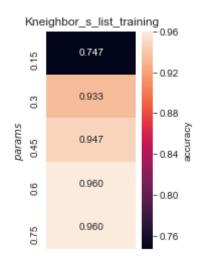


Adaboost test accuracy with 60 is: 0.92

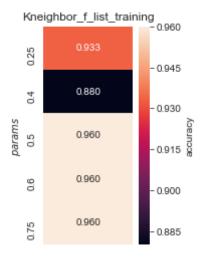
GRADIENT BOOSTING CLASSIFIER
[0.93333333 0.93333333 0.93333333 0.93333333]



Gradient test accuracy with 10 is: 0.92



max_features: [0.93333333 0.88 0.96 0.96 0.96]

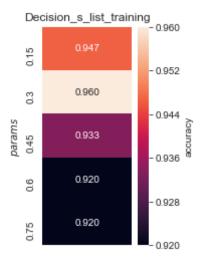


BAGGING CLASSIFIER WITH DECISION TREE max samples: [0.94666667 0.96 0.96

0.93333333 0.92 0.92

21/41

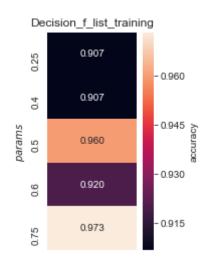
]



max_features: [0.90666667 0.90666667 0.96

0.92

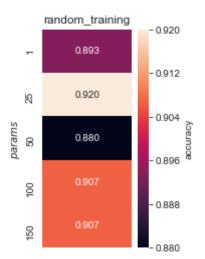
0.97333333]



2

RANDOM FOREST CLASSIFIER [0.89333333 0.92 0.88

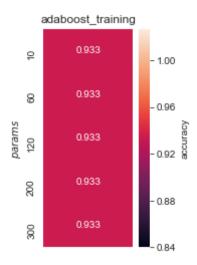
0.90666667 0.90666667]



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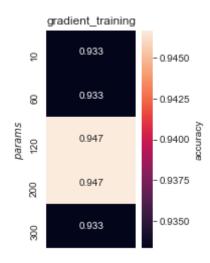
random test accuracy with 25 is: 0.9733333333333333

ADABOOST CLASSIFIER WITH DECISION TREE
[0.93333333 0.93333333 0.93333333 0.93333333]



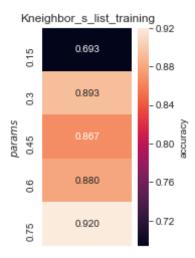
Adaboost test accuracy with 10 is: 0.96

GRADIENT BOOSTING CLASSIFIER
[0.93333333 0.93333333 0.94666667 0.94666667 0.93333333]

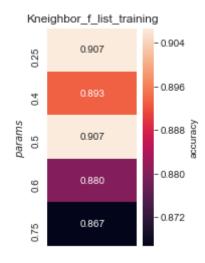


Gradient test accuracy with 120 is: 0.96

BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER
max_samples: [0.69333333 0.89333333 0.86666667 0.88 0.92]



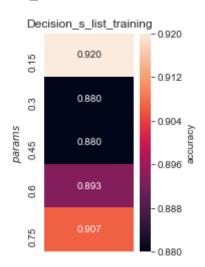
max_features: [0.90666667 0.89333333 0.90666667 0.88 0.86666667]



34

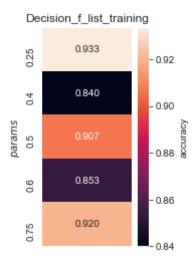
BAGGING CLASSIFIER WITH DECISION TREE 0.88 max_samples: [0.92 0.88

0.89333333 0.90666667]



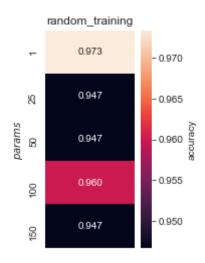
max_features: [0.93333333 0.84

0.90666667 0.85333333 0.92



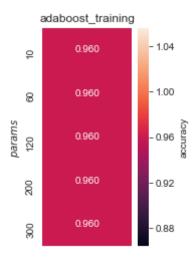
3

RANDOM FOREST CLASSIFIER
[0.97333333 0.94666667 0.94666667 0.96 0.94666667]

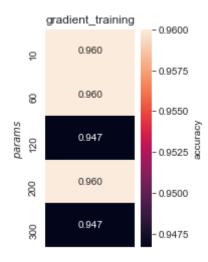


random test accuracy with 1 is: 0.96

ADABOOST CLASSIFIER WITH DECISION TREE [0.96 0.96 0.96 0.96 0.96]

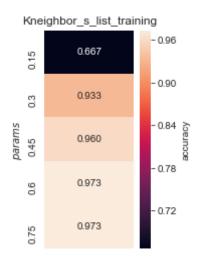


Adaboost test accuracy with 10 is: 0.96



Gradient test accuracy with 10 is: 0.9466666666666667

BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER max_samples: [0.66666667 0.93333333 0.96 0.97333333 0.97333333]

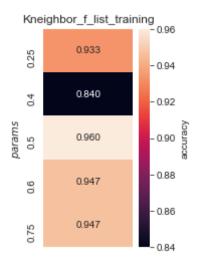


max_features: [0.93333333 0.84

0.96

0.94666667 0.94666667]

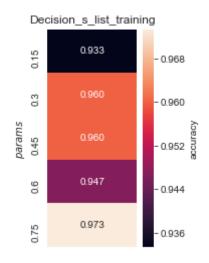
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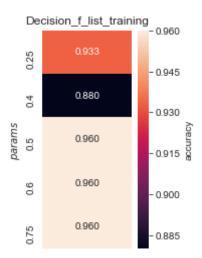
Kneighbor Bagging test accuracy with 0.6 and 0.5 is: 0.9466666666666667

BAGGING CLASSIFIER WITH DECISION TREE max_samples: [0.93333333 0.96 0.96

0.94666667 0.97333333]



max_features: [0.93333333 0.88 0.96 0.96 0.96]



```
Random Forest Classifier 50/50 training accuracy:
[1.0, 1.0, 0.946666666666667]
Adaboost Classifier 50/50 training accuracy:
[1.0, 1.0, 1.0]
Gradient Boost Classifier 50/50 training accuracy:
[1.0, 1.0, 1.0]
Kneigher Bagging Classifier 50/50 training accuracy:
[0.9733333333333334, 0.9466666666666667, 1.0]
Decision Tree Bagging Classifier 50/50 training accuracy:
[0.986666666666667, 0.92, 1.0]
Random Forest Classifier 50/50 test accuracy:
                                                  0.964444444444444
Adaboost Classifier 50/50 test accuracy:
                                                  0.94666666666666
Gradient Boost Classifier 50/50 test accuracy:
                                                  0.9422222222222
Kneighbor Bagging Classifier 50/50 test accuracy:
                                                  0.95555555555555
Random Forest Classifier 50/50 test error:
                                                  0.03555555555555
Adaboost Classifier 50/50 test error:
                                                  0.053333333333333
344
Gradient Boost Classifier 50/50 test error:
                                                  0.057777777777777
Kneighbor Bagging Classifier 50/50 test error:
                                                  0.044444444444444
Decision Tree Bagging Classifier 50/50 test error:
                                                  0.0711111111111111
12
```

```
In [16]: random train acc = []
         adaboost train acc = []
         gradient_train_acc = []
         bagging_train_acc = []
          decision_train_acc = []
          random_test_acc = []
          adaboost_test_acc = []
         gradient test acc = []
         bagging_test_acc = []
         decision_test_acc = []
          i = 1;
         while i < 4:
              #here we split the data again
              np.random.shuffle(X and Y)
              X = X_and_Y[:, 0:-1]
              Y = X_and_Y[:, -1]
              X \text{ train val} = X[:int(0.2*len(X))]
              X_{\text{test}} = X[int(0.2*len(X)):]
              Y_train_val = Y[:int(0.2*len(Y))]
              Y \text{ test} = Y[\text{int}(0.2*\text{len}(Y)):]
              print(i)
              #then we run it on random forest
              random = RandomForestClassifier()
              print("\n RANDOM FOREST CLASSIFIER")
              #first grid search / cross validation for best parameter
              n_{list} = [1, 25, 50, 100, 150]
              parameters random = {'n estimators' : n list}
              clf random = GridSearchCV(random, parameters random, cv=3)
              clf random.fit(X train val, Y train val)
              #display result with heatmap
              print(clf_random.cv_results_['mean_test_score'])
              draw heatmap(clf random.cv results ['mean test score'].reshape(len(n
          list),1), "random training", n list)
              #choose the best parameter and train again
              random = RandomForestClassifier(n estimators = clf random.best param
         s_['n_estimators'])
              random = random.fit(X train val, Y train val)
              #calculate the test error
              test predict = random.predict(X test)
              test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
         st))])/len(Y_test)
              train_predict = random.predict(X train val)
              train acc = sum([train predict[i] == Y train val[i] for i in range(l
         en(Y train val))])/len(Y train val)
              print("random test accuracy with "+str(clf_random.best params ['n es
         timators'])+" is: " + str(test acc))
              random test acc.append(test acc)
              random train acc.append(train acc)
              #run it with Adaboost
              dt = DecisionTreeClassifier()
              adaboost = AdaBoostClassifier(base estimator=dt)
              print("\n ADABOOST CLASSIFIER WITH DECISION TREE")
              #first grid search / cross validation for best parameter
              e list = [10,60,120,200,300]
              parameters adaboost = {'n estimators' : e list}
```

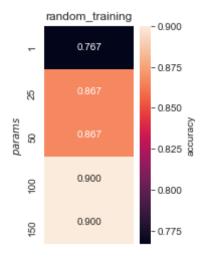
```
clf_adaboost = GridSearchCV(adaboost, parameters_adaboost, cv=3)
   clf_adaboost.fit(X_train_val,Y_train_val)
   #display result with heatmap
   print(clf_adaboost.cv_results_['mean_test_score'])
   draw_heatmap(clf_adaboost.cv_results_['mean_test_score'].reshape(len
(e_list),1), "adaboost_training", e_list)
    #choose the best parameter and train again
   adaboost = AdaBoostClassifier(base_estimator=dt, n_estimators=clf_ad
aboost.best_params_['n_estimators'])
   adaboost = adaboost.fit(X train val, Y train val)
   #calculate the test error
   test_predict = adaboost.predict(X_test)
   train_predict = adaboost.predict(X_train_val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
   print("Adaboost test accuracy with "+str(clf_adaboost.best_params_[
'n_estimators'])+" is: " + str(test_acc))
   adaboost_test_acc.append(test_acc)
   adaboost_train_acc.append(train_acc)
   #run it with gradient
   gradient = ensemble.GradientBoostingClassifier()
   print("\n GRADIENT BOOSTING CLASSIFIER")
   parameters_gradient = {'n_estimators': e_list}
   #first grid search / cross validation for best parameter
   clf gradient = GridSearchCV(gradient, parameters gradient, cv=3)
   clf gradient.fit(X train val, Y train val)
   #display result with heatmap
   print(clf_gradient.cv_results_['mean_test_score'])
   draw heatmap(clf gradient.cv results ['mean test score'].reshape(len
(e_list),1), "gradient_training", e_list)
   #choose the best parameter and train again
   gradient = ensemble.GradientBoostingClassifier(n estimators=clf grad
ient.best params ['n estimators'])
   gradient = gradient.fit(X_train_val,Y_train_val)
    #calculate the test error
   test predict = gradient.predict(X test)
   test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y test)
   train predict = gradient.predict(X train val)
   train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
   print("Gradient test accuracy with "+str(clf gradient.best params [
'n estimators'])+" is: " + str(test acc))
   gradient_test_acc.append(test_acc)
   gradient train acc.append(train acc)
   #run it with Bagging Family with KNeighborClassifer
   bagging = BaggingClassifier(KNeighborsClassifier())
   print("\n BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER")
   s_list = [0.3, 0.4, 0.5, 0.6, 0.7]
   parameters bagging = {'max samples' : s list}
   clf_bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
   clf_bagging.fit(X_train_val, Y_train_val)
   print("max_samples: " + str(clf_bagging.cv_results_['mean_test_scor
```

```
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
s_list),1),"Kneighbor_s_list_training", s_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=clf_
bagging.best_params_['max_samples'])
    best_max_samples = clf_bagging.best_params_['max_samples']
    f_{list} = [0.3, 0.4, 0.5, 0.6, 0.7]
    parameters_bagging = {'max_features': f_list}
    clf bagging = GridSearchCV(bagging, parameters_bagging, cv=3)
    clf_bagging.fit(X_train_val, Y_train_val)
    print("max_features: " + str(clf_bagging.cv_results_['mean_test_scor')
e']))
    draw_heatmap(clf_bagging.cv_results_['mean_test_score'].reshape(len(
f_list),1),"Kneighbor_f_list_training", f_list)
    #choose the best parameter and train again for max features
    bagging = BaggingClassifier(KNeighborsClassifier(), max_samples=best
_max_samples,
                                max_features= clf_bagging.best_params_[
'max_features'])
    bagging = bagging.fit(X_train_val, Y_train_val)
    test_predict = bagging.predict(X_test)
    train_predict = bagging.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y_train_val))])/len(Y_train_val)
    test_acc = sum([test_predict[i] == Y_test[i] for i in range(len(Y_te
st))])/len(Y_test)
    print("Kneighbor Bagging test accuracy with "+str(best_max_samples)+
          " and "+ str(clf bagging.best params ['max features'])+" is: "
 + str(test_acc))
    bagging_test_acc.append(test_acc)
    bagging train acc.append(train acc)
    #run it with Bagging Family with Decision tree
    decision = BaggingClassifier()
    parameters_decision = {'max_samples' : s_list}
    print("\n BAGGING CLASSIFIER WITH DECISION TREE")
    clf decision = GridSearchCV(decision, parameters decision, cv=3)
    clf decision.fit(X train val, Y train val)
   print("max samples: " + str(clf decision.cv results ['mean test scor
e']))
    draw_heatmap(clf_decision.cv_results_['mean_test_score'].reshape(len
(s_list),1),"Decision_s_list_training", s_list)
    #choose the best parameter and train again for max features
    decision = BaggingClassifier(max_samples=clf_decision.best_params_[
'max_samples'])
    best max samples = clf decision.best params ['max samples']
    parameters_decision = {'max_features': f_list}
    clf_decision = GridSearchCV(decision, parameters_decision, cv=3)
    clf decision.fit(X train val, Y train val)
    print("max_features: " + str(clf_decision.cv_results_['mean_test_sco
re']))
    draw_heatmap(clf_decision.cv_results_['mean_test_score'].reshape(len
(f_list),1),"Decision_f_list_training", f_list)
    #choose the best parameter and train again for max_features
    decision = BaggingClassifier(max samples=best max samples,
```

```
max features= clf decision.best params [
'max features'])
    decision = decision.fit(X_train_val, Y_train_val)
    test predict = decision.predict(X test)
    train_predict = decision.predict(X_train_val)
    train_acc = sum([train_predict[i] == Y_train_val[i] for i in range(l
en(Y train val)))/len(Y train val)
    test acc = sum([test predict[i] == Y test[i] for i in range(len(Y te
st))])/len(Y test)
    print("Decision Bagging test accuracy with "+str(best max samples)+
          " and "+ str(clf decision.best params ['max features'])+" is:
 " + str(test_acc))
    decision test acc.append(test acc)
    decision train acc.append(train acc)
    i = i+1;
    print("\n\n\n")
print("Random Forest Classifier 20/80 training accuracy:")
print(random train acc)
print("Adaboost Classifier 20/80 training accuracy:")
print(adaboost train acc)
print("Gradient Boost Classifier 20/80 training accuracy:")
print(gradient train acc)
print("Kneigher Bagging Classifier 20/80 training accuracy:")
print(bagging train acc)
print("Decision Tree Bagging Classifier 20/80 training accuracy:")
print(decision train acc)
print("Random Forest Classifier 20/80 test accuracy:
                                                             "+ str(sum(
random test acc)/len(random test acc)))
print("Adaboost Classifier 20/80 test accuracy:
                                                             "+ str(sum(
adaboost_test_acc)/len(adaboost_test_acc)))
print("Gradient Boost Classifier 20/80 test accuracy:
                                                             "+ str(sum(
gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 20/80 test accuracy:
                                                             "+ str(sum(
bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 20/80 test accuracy: "+ str(sum(
decision test acc)/len(decision test acc)))
print("Random Forest Classifier 20/80 test error:
                                                             "+ str(1-su
m(random test acc)/len(random test acc)))
print("Adaboost Classifier 20/80 test error:
                                                             "+ str(1-su
m(adaboost test acc)/len(adaboost test acc)))
print("Gradient Boost Classifier 20/80 test error:
                                                             "+ str(1-su
m(gradient test acc)/len(gradient test acc)))
print("Kneighbor Bagging Classifier 20/80 test error:
                                                             "+ str(1-su
m(bagging test acc)/len(bagging test acc)))
print("Decision Tree Bagging Classifier 20/80 test error:
                                                             "+ str(1-su
m(decision test acc)/len(decision test acc)))
```

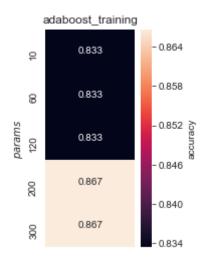
1

RANDOM FOREST CLASSIFIER
[0.76666667 0.86666667 0.9 0.9]

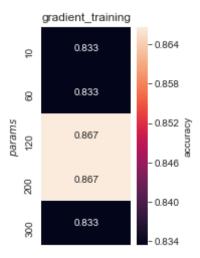


random test accuracy with 100 is: 0.925

ADABOOST CLASSIFIER WITH DECISION TREE
[0.83333333 0.83333333 0.86666667 0.86666667]



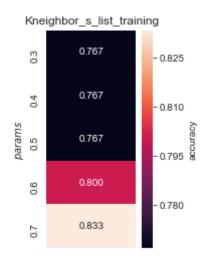
GRADIENT BOOSTING CLASSIFIER
[0.83333333 0.83333333 0.86666667 0.86666667 0.83333333]



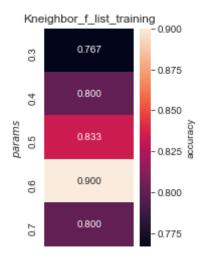
Gradient test accuracy with 120 is: 0.925

BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER max_samples: [0.76666667 0.76666667 0.76666667 0.8

0.83333333]

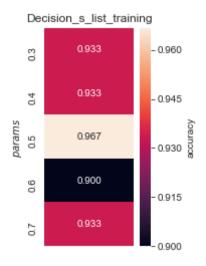


max features: [0.76666667 0.8 0.83333333 0.9 0.8]



BAGGING CLASSIFIER WITH DECISION TREE max_samples: [0.93333333 0.96666667 0.9

0.93333333]

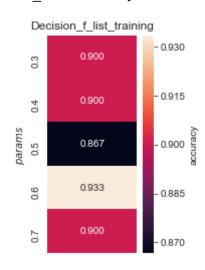


max_features: [0.9

0.9

0.86666667 0.93333333 0.9

]



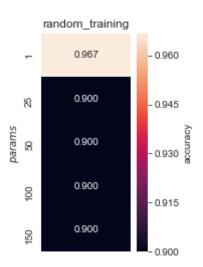
2

RANDOM FOREST CLASSIFIER [0.96666667 0.9 0.9

0.9

0.9

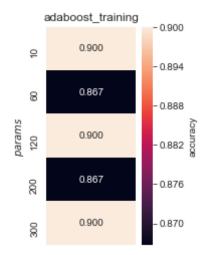
]



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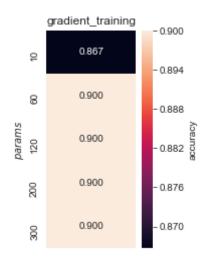
random test accuracy with 1 is: 0.891666666666667

1



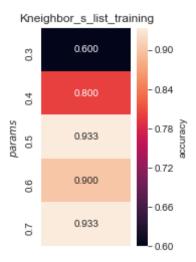
Adaboost test accuracy with 10 is: 0.875

GRADIENT BOOSTING CLASSIFIER
[0.86666667 0.9 0.9 0.9]

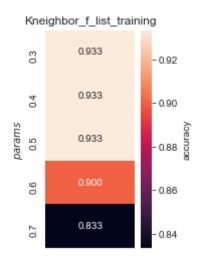


Gradient test accuracy with 60 is: 0.95

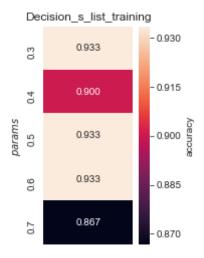
BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER
max samples: [0.6 0.8 0.93333333 0.9 0.93333333]



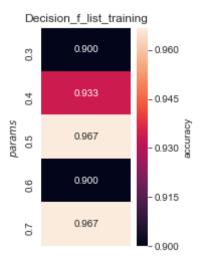
max_features: [0.93333333 0.93333333 0.93333333 0.9 0.83333333]



BAGGING CLASSIFIER WITH DECISION TREE
max_samples: [0.93333333 0.9 0.93333333 0.93333333 0.86666667]



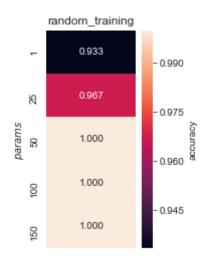
max features: [0.9 0.93333333 0.96666667 0.9 0.96666667]



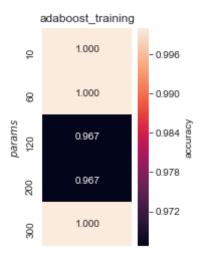
Decision Bagging test accuracy with 0.3 and 0.5 is: 0.875

3





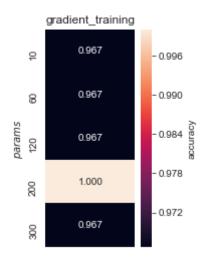
random test accuracy with 50 is: 0.925



Adaboost test accuracy with 10 is: 0.941666666666667

GRADIENT BOOSTING CLASSIFIER [0.96666667 0.96666667 1.

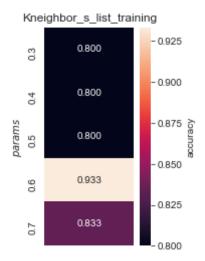
0.96666667]



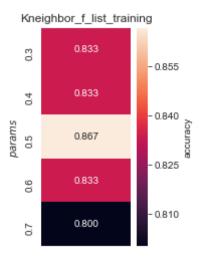
Gradient test accuracy with 200 is: 0.9416666666666667

BAGGING CLASSIFIER WITH KNEIGHBORS CLASSIFIER

max_samples: [0.8 0.8 0.93333333 0.83333333]

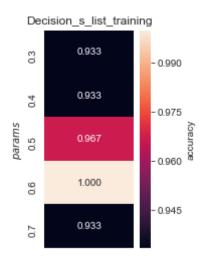


max features: [0.83333333 0.83333333 0.86666667 0.83333333 0.8

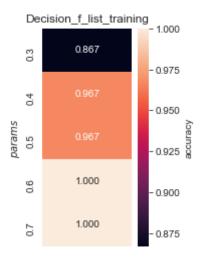


Kneighbor Bagging test accuracy with 0.6 and 0.5 is: 0.825

BAGGING CLASSIFIER WITH DECISION TREE
max_samples: [0.93333333 0.93333333 0.96666667 1. 0.93333333]



max_features: [0.86666667 0.96666667 0.96666667 1. 1.



```
Random Forest Classifier 20/80 training accuracy:
       [1.0, 1.0, 1.0]
       Adaboost Classifier 20/80 training accuracy:
       [1.0, 1.0, 1.0]
       Gradient Boost Classifier 20/80 training accuracy:
       [1.0, 1.0, 1.0]
       Kneigher Bagging Classifier 20/80 training accuracy:
       [0.966666666666667, 0.93333333333333, 0.933333333333333]
       Decision Tree Bagging Classifier 20/80 training accuracy:
       [0.9333333333333333, 0.9666666666666667, 1.0]
       Random Forest Classifier 20/80 test accuracy:
                                                          0.91388888888889
       Adaboost Classifier 20/80 test accuracy:
                                                          0.91666666666666
       Gradient Boost Classifier 20/80 test accuracy:
                                                          0.93888888888888
       Kneighbor Bagging Classifier 20/80 test accuracy:
                                                          0.875
       Random Forest Classifier 20/80 test error:
                                                          0.086111111111110
       Adaboost Classifier 20/80 test error:
                                                          0.083333333333333
       37
       Gradient Boost Classifier 20/80 test error:
                                                          0.0611111111111111
       Kneighbor Bagging Classifier 20/80 test error:
                                                          0.125
       Decision Tree Bagging Classifier 20/80 test error:
                                                          0.0861111111111111
       14
In [ ]:
In [ ]:
```