Modify the Bagging scratch code in our lecture such that:

- Calculate for oob evaluation for each bootstrapped dataset, and also the average score
- Change the code to "without replacement"
- Put everything into a class Bagging . It should have at least two methods, fit (X\_train, y\_train), and predict(X\_test)
- Modify the code from above to randomize features. Set the number of features to be used in each tree to be sqrt(n), and then select a subset of features for each tree. This can be easily done by setting our DecisionTreeClassifier max\_features to 'sqrt'

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In [1]: import numpy as np
          import matplotlib.pyplot as plt
In [95]:
          from sklearn.datasets import load iris
          from sklearn.model_selection import train test split
          from sklearn.metrics import classification report, accuracy score
          iris = load_iris()
          X = iris.data
          y = iris.target
          X_train, X_test, y_train, y_test = train_test_split(X, y,
                         test size=0.3, shuffle=True, random state=42)
In [162]: class Bagging:
              def init (self, boostrap ratio, without replacement = True, B = 5):
                  self.B = B
                  self.boostrap ratio = boostrap ratio
                  self.without_replacement = without_replacement
                  self.tree params = {'max_depth': 2, 'criterion':'gini', 'min_samples_split': 5}
                  self.models = [DecisionTreeClassifier(**self.tree params) for in range(B)]
              def fit(self, X_train, y_train):
                  m, n = X train.shape
                  #sample size for each tree
                  sample size = int(self.boostrap_ratio * len(X_train))
                  xsamples = np.zeros((B, sample size, n))
                  ysamples = np.zeros((B, sample size))
                  xoob = []
                  yoob = []
                  #subsamples for each model
                  for i in range(self.B):
                      idxes = []
                      oob idxes = []
                      ##sampling with replacement; i.e., sample can occur more than once
                      #for the same predictor
                      for j in range(sample_size):
                          idx = random.randrange(m) #<----with replacement #change so no repetition
                          # If there is replacement create indices of oob for future
                          if (self.without replacement):
                               # If idx is already in oob idxs, redo
                              while idx in oob idxes:
                                  idx = random.randrange(m)
                          idxes.append(idx)
                          oob idxes.append(idx)
                          xsamples[i, j, :] = X train[idx]
                          ysamples[i, j] = y train[idx]
                               #keep track of idx that i did not use for ith tree
                       \# if bootstrap is less than 0, we don't take the whole X
                       # Here I create an array the size of X with False values
                      mask = np.zeros((m), dtype=bool)
                      # oob_idx will not have all X's m values, so we are creating a mask that will select indice
          s we have
                      mask[oob idxes] = True # mask = [True, False, False, True ... etc]
                      # Here flip the incides
                      x_mask = X_train[~mask]
                      y_mask = y_train[~mask]
                      xoob.append(x mask)
                      yoob.append(y_mask)
                  #fitting each estimator
                  oob score = 0
                  total score = 0
                  for i, model in enumerate(self.models):
                      _X = xsamples[i, :]
                      _y = ysamples[i, :]
                      model.fit(_X, _y)
                      # calculating oob score here
                      # oob can now be used as a validation set! Because the model didn't see it
                      _X_test = np.asarray(xoob[i])
                      _y_test = np.asarray(yoob[i])
                      _yhat = model.predict(_X_test)
                      acc_score = accuracy_score(_y_test, _yhat)
                      total_score += acc_score
                      print(f"Accuracy score in three {i}: {acc_score}")
                   # calculate total oob score! finally!!
                  total_oob_score = total_score / len(models)
                  print(f"Total oob score is: {total_oob_score}")
              def predict(self, X test):
                  #make prediction and return the probabilities
                  predictions = np.zeros((self.B, X_test.shape[0]))
                  for i, model in enumerate(self.models):
                      yhat = model.predict(X test)
                      predictions[i, :] = yhat
                  yhat = stats.mode(predictions)[0][0]
                  return yhat
In [165]: model = Bagging(boostrap_ratio = 0.7, without_replacement=True, B = 5)
          model.fit(X train, y train)
          yhat = model.predict(X_test)
          Accuracy score in three 0: 0.9375
          Accuracy score in three 1: 0.84375
          Accuracy score in three 2: 0.9375
          Accuracy score in three 3: 0.75
          Accuracy score in three 4: 0.96875
          Total oob score is: 0.8875
In [166]: print(classification report(y test, yhat))
                        precision
                                    recall f1-score
                                                        support
                     0
                             1.00
                                      1.00
                                                 1.00
                                                             19
                                      1.00
                     1
                             1.00
                                                 1.00
                                                             13
                                      1.00
                                                 1.00
                             1.00
                                                             13
                                                 1.00
                                                             45
              accuracy
             macro avg
                             1.00
                                       1.00
                                                 1.00
                                                             45
          weighted avg
                             1.00
                                       1.00
                                                 1.00
                                                             45
In [85]: # Test
          aaa = np.array([1,2,3,4,5])
          bbb = np.array([1,2,3,4,5])
          print(~aaa)
```

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In [ ]:
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oob\_idx = [96, 67, 0, 91, 38, 26, 93, 48, 32, 6, 40, 14, 63, 78, 12, 36, 94, 4, 56, 53, 5, 41, 51, 72, 44, 19, 46, 34, 33, 43, 74, 92, 61, 85, 57, 22, 2, 30, 24, 82, 65, 70, 27, 95, 97, 76, 29, 39, 83, 80,

18, 73, 101, 103, 13, 47, 54, 3, 64, 99, 20, 81, 69, 7, 23, 89, 37, 98, 59, 45, 77, 11, 35]

[-2 -3 -4 -5 -6]

mask[oob\_idx] = True
xxx = X\_train[~mask]

print(xxx.shape)

(32, 4)

mask = np.zeros((m), dtype=bool)

In [97]: