Gradient Boosting - Lab

September 25, 2021

Modify the Gradient Boosting scratch code in our lecture such that: - Notice that we are still using max_depth = 1. Attempt to tweak min_samples_split, max_depth for the regression and see whether we can achieve better mse on our boston data - Notice that we only write scratch code for gradient boosting for regression, add some code so that it also works for binary classification. Load the breast cancer data from sklearn and see that it works. - Further change the code so that it works for multiclass classification. Load the digits data from sklearn and see that it works - Put everything into class

```
[1]: from sklearn.tree import DecisionTreeRegressor from sklearn.dummy import DummyRegressor from sklearn.model_selection import train_test_split import numpy as np
```

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[2]: class GradientBoosting:
         def __init__(self, n_estimators=5, learning_rate=1, max_depth = 1,_
     →min_samples_split = 2, regression=True):
             self.max_depth = max_depth
             self.min_samples_split = min_samples_split
             tree_params = {'max_depth': self.max_depth,
                     'min_samples_split': self.min_samples_split}
             self.n_estimators = n_estimators
             self.learning_rate = learning_rate
             self.regression = regression
             self.models = [DecisionTreeRegressor(**tree_params) for _ in_
      →range(n_estimators)]
             first_model = DummyRegressor(strategy='mean')
             self.models.insert(0, first_model)
         def grad(self, y, h):
             return y - h
         def fit(self, X, y):
             self.models[0].fit(X, y)
             for i in range(self.n_estimators):
                 y_pred = self.predict(X, self.models[:i+1], with_argmax=False)
                 residual = self.grad(y, y_pred)
```

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self.models[i+1].fit(X, residual)
   def predict(self, X, models=None, with_argmax=True):
       if models is None:
           models = self.models
       f0 = models[0].predict(X)
       boosting = sum(self.learning_rate * model.predict(X) for model in_
\rightarrowmodels[1:])
       y_pred = f0 + boosting
       # if the task is classification, apply softmax function to the
\rightarrow predicted value
       if not self.regression:
           y_pred = np.exp(y_pred) / np.sum(np.exp(y_pred), axis=1,__
→keepdims=True)
           if with_argmax:
               y_pred = np.argmax(y_pred, axis=1)
       return y_pred
```

[3]: # regression from sklearn.datasets import load_boston from sklearn.metrics import mean squared error from sklearn.ensemble import GradientBoostingRegressor X, y = load_boston(return_X_y=True) X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42) $n_{estimators} = 200$ learning_rate = 0.1 # setting max_depth to 3, min_samples_split to 2 $max_depth = 3$ min_samples_split = 2 model = GradientBoosting(n_estimators=n_estimators, learning rate=learning rate, max_depth = max_depth, min_samples_split = min_samples_split,) model.fit(X_train, y_train) yhat = model.predict(X_test) print("Our MSE: ", mean_squared_error(y_test, yhat))

Our MSE: 7.896061062860724 Sklearn MSE: 7.996273842366492

```
[4]: # binary classification
     from sklearn.datasets import load_breast_cancer
     from sklearn.metrics import accuracy_score
     from sklearn.ensemble import GradientBoostingClassifier
     X, y = load_breast_cancer(return_X_y=True)
     X_train, X_test, y_train, y_test = \
             train test split(X, y, test size=0.3, random state=42)
     y_train_encoded = np.zeros((y_train.shape[0], len(set(y))))
     for each_class in range(len(set(y))):
         cond = y_train==each_class
         y_train_encoded[np.where(cond), each_class] = 1
     model = GradientBoosting(n_estimators=200, learning_rate=0.1, max_depth = 3,
                      min_samples_split = 2,
                      regression=False)
     model.fit(X_train, y_train_encoded)
     yhat = model.predict(X_test)
     print("Our accuracy: ", accuracy_score(y_test, yhat))
     # sklearn
     sklearn_model = GradientBoostingClassifier(
         n_estimators=n_estimators,
         learning_rate = 0.1,
        max_depth=1
     yhat_sk = sklearn_model.fit(X_train, y_train).predict(X_test)
     print("Sklearn accuracy: ", accuracy_score(y_test, yhat_sk))
```

Our accuracy: 0.9649122807017544 Sklearn accuracy: 0.9649122807017544

```
[5]: # multi-class classification
     from sklearn.datasets import load_digits
     from sklearn.metrics import accuracy_score
     from sklearn.ensemble import GradientBoostingClassifier
     X, y = load_digits(return_X_y=True)
     X_train, X_test, y_train, y_test = \
             train_test_split(X, y, test_size=0.3, random_state=42)
     y_train_encoded = np.zeros((y_train.shape[0], len(set(y))))
     for each_class in range(len(set(y))):
         cond = y_train==each_class
         y_train_encoded[np.where(cond), each_class] = 1
     model = GradientBoosting(n_estimators=200, learning_rate=0.1, max_depth = 3,
                      min_samples_split = 2,
                      regression=False)
     model.fit(X_train, y_train_encoded)
     yhat = model.predict(X_test)
     print("Our accuracy: ", accuracy_score(y_test, yhat))
     # sklearn
     sklearn_model = GradientBoostingClassifier(
         n_estimators=n_estimators,
         learning_rate = 0.1,
        max_depth=1
     )
     yhat_sk = sklearn_model.fit(X_train, y_train).predict(X_test)
     print("Sklearn accuracy: ", accuracy_score(y_test, yhat_sk))
```

Our accuracy: 0.9314814814814815 Sklearn accuracy: 0.9481481481481482

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