## **EXPERIMENT 7**

## **BAGGING**

```
In [2]: import numpy as np
        from sklearn.datasets import load iris
        from sklearn.model selection import train test split
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.metrics import accuracy score
        # Load data
        data = load iris()
        X = data.data
        y = data.target
        # Split data
        X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=42)
        # Bagging implementation
        class Bagging:
            def init (self, base model, n models=10):
                self.base_model = base_model
                self.n models = n models
                self.models = []
            def fit(self, X, y):
                for i in range(self.n models):
                    # Bootstrap sampling
                    indices = np.random.choice(len(X), len(X), replace=True)
                    X_sample, y_sample = X[indices], y[indices]
                    model = self.base_model()
                    model.fit(X sample, y_sample)
                    self.models.append(model)
                    print(f"Model {i+1} trained with accuracy: {accuracy_score(y_sample, model.predict(X_sample))}")
            def predict(self, X):
                # Aggregate predictions from each model
                predictions = np.zeros((X.shape[0], len(self.models)))
                for i, model in enumerate(self.models):
                    predictions[:, i] = model.predict(X)
                # Majority vote
                return np.round(np.mean(predictions, axis=1))
        # Initialize and train the bagging model
        bagging_model = Bagging(base_model=DecisionTreeClassifier, n_models=10)
        bagging model.fit(X train, y train)
        # Make predictions and evaluate
        y_pred = bagging_model.predict(X_test)
        print(f"Bagging Model Accuracy: {accuracy_score(y_test, y_pred)}")
       Model 1 trained with accuracy: 1.0
       Model 2 trained with accuracy: 1.0
       Model 3 trained with accuracy: 1.0
       Model 4 trained with accuracy: 1.0
       Model 5 trained with accuracy: 1.0
       Model 6 trained with accuracy: 1.0
       Model 7 trained with accuracy: 1.0
       Model 8 trained with accuracy: 1.0
       Model 9 trained with accuracy: 1.0
       Model 10 trained with accuracy: 1.0
       Bagging Model Accuracy: 1.0
```

## BOOSTING

```
In [5]: from sklearn.tree import DecisionTreeClassifier
from sklearn.utils import shuffle

class AdaBoost:
    def __init__(self, base_model, n_models=50):
        self.base_model = base_model
        self.n_models = n_models
        self.models = []
        self.alphas = []
```

```
def fit(self, X, y):
       n samples = X.shape[0]
       weights = np.ones(n_samples) / n_samples
       for i in range(self.n models):
           model = self.base_model()
           model.fit(X, y, sample_weight=weights)
            y_pred = model.predict(X)
            # Calculate error
           error = np.sum(weights * (y_pred != y)) / np.sum(weights)
            # Handle zero error
           if error == 0:
               alpha = 1.0
            else:
               alpha = 0.5 * np.log((1 - error) / error)
           # Update weights
           weights *= np.exp(alpha * (y_pred != y))
           weights /= np.sum(weights)
           # Check for NaN in weights
           if np.any(np.isnan(weights)):
                raise ValueError("Weights contain NaN values")
            self.models.append(model)
            self.alphas.append(alpha)
           print(f"Model {i+1}: Error = {error}, Alpha = {alpha}")
    def predict(self, X):
        predictions = np.zeros(X.shape[0])
        for model, alpha in zip(self.models, self.alphas):
           predictions += alpha * model.predict(X)
        return np.sign(predictions)
# Initialize and train the AdaBoost model
adaboost model = AdaBoost(base model=DecisionTreeClassifier, n models=50)
adaboost_model.fit(X_train, y_train)
# Make predictions and evaluate
y_pred = adaboost_model.predict(X_test)
print(f"AdaBoost Model Accuracy: {accuracy_score(y_test, y_pred)}")
```

```
Model 1: Error = 0.0, Alpha = 1.0
Model 2: Error = 0.0, Alpha = 1.0
Model 3: Error = 0.0, Alpha = 1.0
Model 4: Error = 0.0, Alpha = 1.0
Model 5: Error = 0.0, Alpha = 1.0
Model 6: Error = 0.0, Alpha = 1.0
Model 7: Error = 0.0, Alpha = 1.0
Model 8: Error = 0.0, Alpha = 1.0
Model 9: Error = 0.0, Alpha = 1.0
Model 10: Error = 0.0, Alpha = 1.0
Model 11: Error = 0.0, Alpha = 1.0
Model 12: Error = 0.0, Alpha = 1.0
Model 13: Error = 0.0, Alpha = 1.0
Model 14: Error = 0.0, Alpha = 1.0
Model 15: Error = 0.0, Alpha = 1.0
Model 16: Error = 0.0, Alpha = 1.0
Model 17: Error = 0.0, Alpha = 1.0
Model 18: Error = 0.0, Alpha = 1.0
Model 19: Error = 0.0, Alpha = 1.0
Model 20: Error = 0.0, Alpha = 1.0
Model 21: Error = 0.0, Alpha = 1.0
Model 22: Error = 0.0, Alpha = 1.0
Model 23: Error = 0.0, Alpha = 1.0
Model 24: Error = 0.0, Alpha = 1.0
Model 25: Error = 0.0, Alpha = 1.0
Model 26: Error = 0.0, Alpha = 1.0
Model 27: Error = 0.0, Alpha = 1.0
Model 28: Error = 0.0, Alpha = 1.0
Model 29: Error = 0.0, Alpha = 1.0
Model 30: Error = 0.0, Alpha = 1.0
Model 31: Error = 0.0, Alpha = 1.0
Model 32: Error = 0.0, Alpha = 1.0
Model 33: Error = 0.0, Alpha = 1.0
Model 34: Error = 0.0, Alpha = 1.0
Model 35: Error = 0.0, Alpha = 1.0
Model 36: Error = 0.0, Alpha = 1.0
Model 37: Error = 0.0, Alpha = 1.0
Model 38: Error = 0.0, Alpha = 1.0
Model 39: Error = 0.0, Alpha = 1.0
Model 40: Error = 0.0, Alpha = 1.0
Model 41: Error = 0.0, Alpha = 1.0
Model 42: Error = 0.0, Alpha = 1.0
Model 43: Error = 0.0, Alpha = 1.0
Model 44: Error = 0.0, Alpha = 1.0
Model 45: Error = 0.0, Alpha = 1.0
Model 46: Error = 0.0, Alpha = 1.0
Model 47: Error = 0.0, Alpha = 1.0
Model 48: Error = 0.0, Alpha = 1.0
Model 49: Error = 0.0, Alpha = 1.0
Model 50: Error = 0.0, Alpha = 1.0
AdaBoost Model Accuracy: 0.7111111111111111
```

## **STACKING**

```
In [4]: from sklearn.linear model import LogisticRegression
        class Stacking:
            def init (self, base models, meta model):
                self.base models = base models
                self.meta model = meta model
                self.meta X train = None
            def fit(self, X, y):
                meta_X = np.zeros((X.shape[0], len(self.base_models)))
                for i, model in enumerate(self.base_models):
                    model.fit(X, y)
                    meta_X[:, i] = model.predict(X)
                    print(f"Base Model {i+1} predictions: {meta X[:, i]}")
                self.meta model.fit(meta X, y)
            def predict(self, X):
                meta_X = np.zeros((X.shape[0], len(self.base_models)))
                for i, model in enumerate(self.base models):
                    meta_X[:, i] = model.predict(X)
                return self.meta_model.predict(meta_X)
        # Initialize base models and meta-model
        base_models = [DecisionTreeClassifier(), DecisionTreeClassifier(max_depth=3)]
        meta model = LogisticRegression()
        # Initialize and train the stacking model
```

```
stacking model = Stacking(base models=base models, meta model=meta model)
 stacking_model.fit(X_train, y_train)
 # Make predictions and evaluate
 y_pred = stacking_model.predict(X_test)
 print(f"Stacking Model Accuracy: {accuracy_score(y_test, y_pred)}")
Base Model 1 predictions: [1. 2. 2. 1. 2. 1. 2. 1. 0. 2. 1. 0. 0. 0. 1. 2. 0. 0. 0. 1. 2. 0.
 1. 2. 0. 2. 2. 1. 1. 2. 1. 0. 1. 2. 0. 0. 1. 1. 0. 2. 0. 0. 1. 1. 2. 1.
 2.\ \ 2.\ \ 1.\ \ 0.\ \ 0.\ \ 2.\ \ 2.\ \ 0.\ \ 1.\ \ 1.\ \ 2.\ \ 0.\ \ 2.\ \ 1.
2. 1. 1. 1. 0. 1. 1. 0. 1. 2. 2. 0. 1. 2. 2. 0. 2. 0. 1. 2. 2. 1. 2. 1.
1. 2. 2. 0. 1. 2. 0. 1. 2.]
Base Model 2 predictions: [1. 1. 2. 1. 2. 1. 2. 1. 0. 2. 1. 0. 0. 0. 1. 2. 0. 0. 0. 1. 2. 0. 0. 1. 2. 0.
1. 2. 0. 2. 2. 1. 1. 2. 1. 0. 1. 2. 0. 0. 1. 1. 0. 2. 0. 0. 2. 1. 2. 1.
 1. 2. 1. 0. 0. 1. 2. 0. 0. 1. 2. 0. 2. 2. 0. 1. 1. 2. 1. 2. 0. 2. 1.
2. \ 1. \ 1. \ 1. \ 0. \ 1. \ 1. \ 0. \ 1. \ 2. \ 2. \ 0. \ 1. \ 2. \ 1. \ 0. \ 2. \ 0. \ 1. \ 2. \ 2. \ 1. \ 2. \ 1.
1. 2. 2. 0. 1. 2. 0. 1. 2.]
Stacking Model Accuracy: 1.0
```

In [ ]:

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