

Naive Bayes predictor

```
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import StratifiedKFold
import numpy as np
import pickle
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
```

Load the Data Set

will be split to test and train

```
X = np.load("../X.npy")
y = np.load("../y.npy")

with open("../label_map.pkl", "rb") as f:
    label_map = pickle.load(f)
```

Split the data to n-fold

```
nfold = StratifiedKFold(n_splits=5, shuffle=True, random_state=777)
fold = 1
```

Separate them into 5 different modules

where the training data get scaled

```
import joblib
from sklearn.preprocessing import StandardScaler
for training_index, testing_index in nfold.split(X,
                                                    y): # the loop calls next()
    automatically to give the proper splitted sets for the fold
    # train_index: A NumPy array holds all indices of the training items in
    this fold, e.x. [0 4 7 ... 2887 2996 2999]
    print(f"\n--- Fold {fold} ---")

    # Get the training and testing data for this fold
    X_train, X_test = X[training_index], X[testing_index]
    y_train, y_test = y[training_index], y[testing_index]

    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    X_test = scaler.transform(X_test)

    # Normalize feature values
    """
    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
```

```

X_test = scaler.transform(X_test)
"""

# Train Naive Bayes model
model = GaussianNB()
model.fit(X_train, y_train)

# Predict and evaluate
y_pred = model.predict(X_test)
joblib.dump(model, f"NB_model{fold}.pkl")
joblib.dump(scaler, f"NB_scaler{fold}.pkl")

acc = accuracy_score(y_test, y_pred)
print(f"Accuracy: {acc:.4f}")
print("Classification Report:")
print(classification_report(y_test, y_pred,
target_names=label_map.values()))

# Confusion matrix (optional)
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cm)

fold += 1

```

--- Fold 1 ---

Accuracy: 0.7262

Classification Report:

	precision	recall	f1-score	support
cats	0.67	0.69	0.68	200
panda	0.69	0.79	0.74	199
spiders	0.84	0.69	0.76	200
accuracy			0.73	599
macro avg	0.74	0.73	0.73	599
weighted avg	0.74	0.73	0.73	599

Confusion Matrix:

```

[[139  40  21]
 [ 37 157   5]
 [ 31  30 139]]

```

--- Fold 2 ---

Accuracy: 0.7446

Classification Report:

	precision	recall	f1-score	support
cats	0.74	0.71	0.73	200
panda	0.70	0.86	0.77	199
spiders	0.81	0.66	0.73	200
accuracy			0.74	599
macro avg	0.75	0.74	0.74	599

weighted avg	0.75	0.74	0.74	599
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Confusion Matrix:

```
[[143  32  25]
 [ 22 171   6]
 [ 27  41 132]]
```

--- Fold 3 ---

Accuracy: 0.7262

Classification Report:

	precision	recall	f1-score	support
cats	0.73	0.65	0.69	200
panda	0.68	0.92	0.78	199
spiders	0.81	0.60	0.69	200
accuracy			0.73	599
macro avg	0.74	0.73	0.72	599
weighted avg	0.74	0.73	0.72	599

Confusion Matrix:

```
[[130  45  25]
 [ 11 184   4]
 [ 37  42 121]]
```

--- Fold 4 ---

Accuracy: 0.7007

Classification Report:

	precision	recall	f1-score	support
cats	0.68	0.67	0.67	200
panda	0.66	0.84	0.74	198
spiders	0.80	0.59	0.68	200
accuracy			0.70	598
macro avg	0.71	0.70	0.70	598
weighted avg	0.71	0.70	0.70	598

Confusion Matrix:

```
[[133  42  25]
 [ 26 167   5]
 [ 37  44 119]]
```

--- Fold 5 ---

Accuracy: 0.7441

Classification Report:

	precision	recall	f1-score	support
cats	0.68	0.69	0.68	200
panda	0.74	0.85	0.79	198
spiders	0.83	0.69	0.76	200
accuracy			0.74	598
macro avg	0.75	0.74	0.74	598
weighted avg	0.75	0.74	0.74	598

Confusion Matrix:

```
[[138  37  25]
 [ 27 168   3]
 [ 38  23 139]]
```