Experiment 3.2

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1. Aim:

Implement Naïve Bayes theorem to classify the English text

2. Objective:

How to calculate the probabilities required by the Naive Bayes algorithm.

How to implement the Naive Bayes algorithm from scratch.

How to apply Naive Bayes to a real-world predictive modeling problem.

3. Algorithm:

Step 1: Separate By Class.

Step 2: Summarize Dataset.

Step 3: Summarize Data By Class.

Step 4: Gaussian Probability Density Function.

Step 5: Class Probabilities.

4. Code:

```
[1]  import math
     from collections import defaultdict
```

```
     def preprocess_text(text):
         # Tokenize the text (split into words)
         words = text.split()
         # Remove punctuation and convert to lowercase
         cleaned_words = [word.lower() for word in words]
         return cleaned_words
```

(variable) cleaned_words: list

```
def train_naive_bayes(texts, labels):
    # Calculate class priors (P(y))
    class_counts = defaultdict(int)
    total_docs = len(labels)
    for label in labels:
        class_counts[label] += 1

    class_priors = {label: count / total_docs for label, count in class_counts.items()}

    # Calculate word likelihoods (P(x|y))
    word_counts = defaultdict(lambda: defaultdict(int))
    class_totals = defaultdict(int)

    for i, text in enumerate(texts):
        label = labels[i]
        words = preprocess_text(text)
        class_totals[label] += len(words)

        for word in words:
            word_counts[label][word] += 1

    word_likelihoods = {label: {word: count / class_totals[label] for word, count in counts.items()}
                        for label, counts in word_counts.items()}

    return class_priors, word_likelihoods
```

```
[4]  def classify_naive_bayes(text, class_priors, word_likelihoods):
         words = preprocess_text(text)

         scores = {label: math.log(class_priors[label]) for label in class_priors.keys()}

         for label in class_priors.keys():
             for word in words:
                 if word in word_likelihoods[label]:
                     scores[label] += math.log(word_likelihoods[label][word])

         # Choose the class with the highest score
         predicted_class = max(scores, key=scores.get)
         return predicted_class
```

```
[5]  import matplotlib.pyplot as plt
     from sklearn.datasets import load_iris
```

```
[6]  iris = load_iris()
     data = iris.data
     target = iris.target
     target_names = iris.target_names
```

```
plt.figure()
plt.scatter(data[:, 0], data[:, 1], c=target, cmap=plt.cm.Paired)
plt.xlabel('Sepal Length (cm)')
plt.ylabel('Sepal Width (cm)')
plt.show()
```

5. Output