Create a classification model to predict the sentiment either (1 or 0) based on Disaster tweets

```
In [1]: import pandas as pd
In [2]: import nltk
In [3]: from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer
         from sklearn.model selection import train test split
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.naive bayes import MultinomialNB
         from sklearn.metrics import accuracy_score, classification_report
In [4]: | nltk.download('punkt')
         [nltk_data] Downloading package punkt to
         [nltk data]
                          C:\Users\HP\AppData\Roaming\nltk_data...
         [nltk data]
                        Package punkt is already up-to-date!
Out[4]: True
In [5]: | nltk.download('stopwords')
         [nltk data] Downloading package stopwords to
                          C:\Users\HP\AppData\Roaming\nltk_data...
         [nltk_data]
         [nltk data]
                        Package stopwords is already up-to-date!
Out[5]: True
        df=pd.read_csv('disaster_tweets_data(DS).csv')
In [6]:
In [7]: | df.head(5)
Out[7]:
                                              tweets target
            Our Deeds are the Reason of this #earthquake M...
          1
                     Forest fire near La Ronge Sask. Canada
                                                         1
          2
                 All residents asked to 'shelter in place' are ...
                                                         1
          3
               13,000 people receive #wildfires evacuation or...
                                                         1
              Just got sent this photo from Ruby #Alaska as ...
```

```
In [8]: df.tail(5)

Out[8]:

tweets target

7608 Two giant cranes holding a bridge collapse int... 1

7609 @aria_ahrary @TheTawniest The out of control w... 1

7610 M1.94 [01:04 UTC]?5km S of Volcano Hawaii. htt... 1

7611 Police investigating after an e-bike collided ... 1

7612 The Latest: More Homes Razed by Northern Calif... 1

In [9]: df.shape

Out[9]: (7613, 2)
```

1)Remove handle null values (if any).

```
In [10]: df.isnull().sum()
Out[10]: tweets  0
     target  0
     dtype: int64
```

Preprocess

```
In [11]: def preprocess_text(text):
    # Tokenizing words
    words = nltk.word_tokenize(text)

# Convert words to Lowercase
    words = [word.lower() for word in words]

# Removing punctuations
    words = [word for word in words if word.isalnum()]

# Removing stop words
    stop_words = set(stopwords.words('english'))
    words = [word for word in words if word not in stop_words]

# Stemming using Porter Stemmer (
    stemmer = PorterStemmer()
    words = [stemmer.stem(word) for word in words]
    return ' '.join(words)

df['preprocessed_tweets'] = df['tweets'].apply(preprocess_text)
```

split the dataset into training and testing sets

```
In [12]: X=df['preprocessed_tweets']
    y=df['target']
    X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.5,random_state=4

TF-IDF vectorization

In [13]: tfidf_vectorizer=TfidfVectorizer(max_features=5000)
    X_train_tfidf=tfidf_vectorizer.fit_transform(X_train)
    X_test_tfidf=tfidf_vectorizer.transform(X_test)
```

classification model(Multinomail naive bayes)

```
In [14]: | classifier=MultinomialNB()
In [15]: | classifier.fit(X_train_tfidf,y_train)
Out[15]:
          ▼ MultinomialNB
          MultinomialNB()
In [16]:
        y_pred=classifier.predict(X_test_tfidf)
In [17]: | accuracy=accuracy_score(y_test,y_pred)
In [18]: classification_rep=classification_report(y_test,y_pred)
In [19]: print("accuracy",accuracy,"\n classification_report:",classification_rep)
         accuracy 0.7972156553716837
          classification_report:
                                                 precision
                                                              recall f1-score
                                                                                  support
                                                  0.84
                     0
                             0.78
                                       0.90
                                                            2185
                     1
                             0.83
                                       0.66
                                                  0.74
                                                            1622
             accuracy
                                                  0.80
                                                            3807
            macro avg
                             0.80
                                       0.78
                                                  0.79
                                                            3807
         weighted avg
                             0.80
                                       0.80
                                                  0.79
                                                            3807
```

LogisticRegression

```
In [20]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix
```

```
In [21]: logistic_regression = LogisticRegression()
         logistic_regression.fit(X_train_tfidf,y_train)
In [22]:
Out[22]:
          ▼ LogisticRegression
          LogisticRegression()
In [23]: y pred logistic regression=logistic regression.predict(X test tfidf)
         confusion matrix logistic regression=confusion matrix(y test,y pred logistic re
In [24]:
         classification report logistic regression=classification report(y test,y pred ]
In [25]: |confusion_matrix_logistic_regression
Out[25]: array([[1954, 231],
                 [ 532, 1090]], dtype=int64)
In [26]:
        classification_report_logistic_regression
Out[26]:
                         precision
                                      recall f1-score
                                                          support\n\n
                                                                                0
         0.79
                   0.89
                                        2185\n
                                                                            0.67
                                                                                      0.
                              0.84
                                                          1
                                                                  0.83
         74
                                                                  0.80
                  1622\n\n
                              accuracy
                                                                            3807\n
                                                                                     mac
                                 0.78
                                           0.79
                                                     3807\nweighted avg
                                                                               0.80
         ro avg
                      0.81
         0.80
                   0.80
                              3807\n'
In [27]: | accuracy_lr=accuracy_score(y_test,y_pred_logistic_regression)
In [28]: |accuracy_lr
Out[28]: 0.7995797215655371
         KNN Classifier
In [29]: from sklearn.neighbors import KNeighborsClassifier
In [30]: knn_classifier=KNeighborsClassifier()
In [31]: knn_classifier.fit(X_train_tfidf,y_train)
Out[31]:
          ▼ KNeighborsClassifier
          KNeighborsClassifier()
In [32]: y_pred_knn_classifier=knn_classifier.predict(X_test_tfidf)
```

```
In [33]: confusion_matrix_knn_classifier=confusion_matrix(y_test,y_pred_knn_classifier)
         classification report knn classifier=classification report(y test,y pred knn c]
In [34]: | accuracy knn=accuracy score(y test,y pred knn classifier)
In [35]: confusion_matrix_knn_classifier
Out[35]: array([[2169,
                          16],
                        293]], dtype=int64)
                 [1329,
In [36]: classification report knn classifier
Out[36]:
                         precision
                                      recall f1-score
                                                                                0
                                                          support\n\n
         0.62
                   0.99
                              0.76
                                        2185\n
                                                                  0.95
                                                                            0.18
                                                                                      0.
                                                                  0.65
                                                                            3807\n
                 1622\n\n
                              accuracy
                                                                                     mac
         ro avg
                      0.78
                                 0.59
                                           0.53
                                                      3807\nweighted avg
                                                                               0.76
                              3807\n'
         0.65
                    0.57
In [37]: accuracy knn
Out[37]: 0.6467034410296821
In [38]:
         best_accuracy = 0
         best_model = ""
In [39]: | models = [
             ("Logistic Regression", y_pred_logistic_regression),
             ("K-Nearest Neighbors", y_pred_knn_classifier),
              ("Multinomail naive bayes",y_pred)]
In [40]: for model_name, predictions in models:
             if accuracy > best_accuracy:
                 best_accuracy = accuracy
                 best_model = model_name
```

best accuracy

```
In [41]: print("The model with the best accuracy is",best_model, "with an accuracy of",be
```

The model with the best accuracy is Logistic Regression with an accuracy of 0.7972156553716837