```
In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.model_selection import train_test_split
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
        from sklearn.preprocessing import LabelEncoder
        import tensorflow as tf
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dense, Dropout
        from tensorflow.keras.optimizers import Adam
In [5]: # Display the column names
        print(train_data.columns)
        Index(['uniq_id', 'crawl_timestamp', 'product_url', 'product_name',
                'product_category_tree', 'pid', 'retail_price', 'discounted_price',
                'image', 'is_FK_Advantage_product', 'description', 'product_rating',
               'overall_rating', 'brand', 'product_specifications'],
              dtype='object')
In [6]: import pandas as pd
        import numpy as np
        from sklearn.model_selection import train_test_split, GridSearchCV
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.preprocessing import LabelEncoder
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
        from keras.models import Sequential
        from keras.layers import Dense, Dropout
        from keras.optimizers import Adam
        import matplotlib.pyplot as plt
        import joblib
        # Load the dataset
        train_data = pd.read_csv("/Users/varunkumar/Downloads/hackathon/train_product_data.csv")
         test_data = pd.read_csv("/Users/varunkumar/Downloads/hackathon/test_data.csv")
        # Display basic information about the dataset
        print(train_data.head())
        print(train_data.columns)
         # Handle missing values
         train_data.dropna(subset=['description', 'product_category_tree'], inplace=True)
         # Extract category from product_category_tree
        train_data['category'] = train_data['product_category_tree'].apply(lambda x: x.split('>>')[0].strip())
        # Encode the labels
        label_encoder = LabelEncoder()
        train_data['category'] = label_encoder.fit_transform(train_data['category'])
        # TF-IDF Vectorization
        tfidf = TfidfVectorizer(max_features=5000, stop_words='english')
        X = tfidf.fit_transform(train_data['description']).toarray()
        y = train_data['category']
        # Split data into training and validation sets
        X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.2, random_state=42)
        # Logistic Regression Model
        log_reg = LogisticRegression(max_iter=1000)
        log_reg.fit(X_train, y_train)
        # Predictions and Evaluation
        y_pred = log_reg.predict(X_val)
        print("Accuracy:", accuracy_score(y_val, y_pred))
        print("Classification Report:\n", classification_report(y_val, y_pred))
        print("Confusion Matrix:\n", confusion_matrix(y_val, y_pred))
        # Neural Network Model
        model = Sequential([
            Dense(512, activation='relu', input_shape=(X_train.shape[1],)),
            Dense(256, activation='relu'),
            Dropout(0.3),
            Dense(len(np.unique(y)), activation='softmax')
        ])
         model.compile(optimizer=Adam(learning_rate=0.001), loss='sparse_categorical_crossentropy', metrics=['accuracy'])
        history = model.fit(X_train, y_train, epochs=10, batch_size=32, validation_data=(X_val, y_val))
        # Plot training history
        plt.plot(history.history['accuracy'], label='accuracy')
        plt.plot(history.history['val_accuracy'], label='val_accuracy')
        plt.xlabel('Epoch')
        plt.ylabel('Accuracy')
        plt.ylim([0, 1])
        plt.legend(loc='lower right')
        plt.show()
        # Evaluate on validation set
        val_loss, val_acc = model.evaluate(X_val, y_val, verbose=2)
        print("Validation Accuracy: ", val_acc)
        # Preprocess test data
         test_data['description'].fillna('', inplace=True)
         test_data_tfidf = tfidf.transform(test_data['description']).toarray()
        # Predict using the trained logistic regression model
         test_pred = log_reg.predict(test_data_tfidf)
         test_data['predicted_category'] = label_encoder.inverse_transform(test_pred)
        print(test_data.head())
         # Save the trained model
        joblib.dump(log_reg, 'logistic_regression_model.pkl')
         # Save the label encoder
        joblib.dump(label_encoder, 'label_encoder.pkl')
                                     uniq_id
                                                        crawl_timestamp \
        0 c2d766ca982eca8304150849735ffef9 2016-03-25 22:59:23 +0000
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        3 ce5a6818f7707e2cb61fdcdbba61f5ad 2016-03-25 22:59:23 +0000
           29c8d290caa451f97b1c32df64477a2c 2016-03-25 22:59:23 +0000
                                                  product_url \
        0 http://www.flipkart.com/alisha-solid-women-s-c...
           http://www.flipkart.com/aw-bellies/p/itmeh4grg...
           http://www.flipkart.com/alisha-solid-women-s-c...
           http://www.flipkart.com/alisha-solid-women-s-c...
           http://www.flipkart.com/dilli-bazaaar-bellies-...
                                                product_name product_category_tree \
        0
                         Alisha Solid Women's Cycling Shorts
                                                                          Clothing
                                                   AW Bellies
                                                                          Footwear
                                                                          Clothing
        2
                         Alisha Solid Women's Cycling Shorts
        3
                         Alisha Solid Women's Cycling Shorts
                                                                          Clothing
           dilli bazaaar Bellies, Corporate Casuals, Casuals
                                                                          Footwear
                        pid retail_price discounted_price \
        0 SRTEH2FF9KEDEFGF
                                    999.0
                                                       379.0
                                    999.0
                                                       499.0
           SH0EH4GRSUBJGZXE
           SRTEH2F6HUZMQ6SJ
                                    699.0
                                                       267.0
        3 SRTEH2FVVKRBAXHB
                                   1199.0
                                                       479.0
        4 SH0EH3DZBFR88SCK
                                    699.0
                                                       349.0
                                                        image is_FK_Advantage_product \
           ["http://img5a.flixcart.com/image/short/u/4/a/...
                                                                                 False
           ["http://img5a.flixcart.com/image/shoe/7/z/z/r...
                                                                                 False
           ["http://img5a.flixcart.com/image/short/6/2/h/...
                                                                                 False
           ["http://img6a.flixcart.com/image/short/p/j/z/...
                                                                                 False
        4 ["http://img6a.flixcart.com/image/shoe/b/p/n/p...
                                                                                 False
                                                  description
                                                                    product_rating \
        0 Key Features of Alisha Solid Women's Cycling S... No rating available
           Key Features of AW Bellies Sandals Wedges Heel... No rating available
           Key Features of Alisha Solid Women's Cycling S... No rating available
           Key Features of Alisha Solid Women's Cycling S... No rating available
           Key Features of dilli bazaaar Bellies, Corpora... No rating available
                overall_rating
                                        brand \
           No rating available
                                        Alisha
           No rating available
                                           ΑW
           No rating available
                                        Alisha
           No rating available
                                        Alisha
        3
           No rating available dilli bazaaar
                                      product_specifications
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        1
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        3
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                'image', 'is_FK_Advantage_product', 'description', 'product_rating',
               'overall_rating', 'brand', 'product_specifications'],
              dtype='object')
        Accuracy: 0.97033333333333334
        Classification Report:
                       precision
                                     recall f1-score
                                                       support
                           0.98
                                     0.96
                                                0.97
                                                           198
                                     0.75
                                                0.83
                   1
                           0.93
                                                            51
                   2
                           0.93
                                     0.84
                                                0.88
                                                            31
                           0.97
                                     1.00
                                                0.98
                                                          1083
                           0.92
                                     0.92
                                                0.92
                                                          107
                           1.00
                                     0.97
                                                0.98
                                                           220
                   6
                           0.95
                                     0.97
                                                0.96
                                                           154
                           0.99
                                     1.00
                                                1.00
                                                           608
                   8
                           0.98
                                     0.94
                                                0.96
                                                           114
                           0.93
                                     0.96
                                                0.95
                                                           171
                  10
                           0.96
                                     0.66
                                                0.78
                                                            41
                  11
                           1.00
                                     0.95
                                                0.97
                                                            79
                  12
                           0.78
                                     0.79
                                                0.78
                                                            48
                  13
                           1.00
                                     1.00
                                                1.00
                                                            95
                                                0.97
                                                          3000
            accuracy
                           0.95
                                     0.91
                                                0.93
                                                          3000
           macro avg
        weighted avg
                           0.97
                                     0.97
                                                0.97
                                                          3000
        Confusion Matrix:
         [[ 191
                   0
                        0
                             0
                                  2
                                       0
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             1
                 38
                       0
                            8
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                  0
                       1
                            2
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             0
                  0
                       0
                            5
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                                     214
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                       0
                                                                    75
                                                                               0]
             3
                  0
                       1
                                                                               0]
                  0
                       0
                                                                              95]]
        /Users/varunkumar/anaconda3/lib/python3.11/site-packages/keras/src/layers/core/dense.py:87: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. W
        hen using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.
          super().__init__(activity_regularizer=activity_regularizer, **kwargs)
        Epoch 1/10
        375/375
                                                   5s 12ms/step - accuracy: 0.7005 - loss: 1.0314 - val_accuracy: 0.9780 - val_loss: 0.0773
        Epoch 2/10
        375/375
                                                   4s 10ms/step - accuracy: 0.9884 - loss: 0.0427 - val_accuracy: 0.9807 - val_loss: 0.0585
        Epoch 3/10
        375/375
                                                   4s 11ms/step - accuracy: 0.9958 - loss: 0.0170 - val_accuracy: 0.9853 - val_loss: 0.0588
        Epoch 4/10
        375/375
                                                   4s 11ms/step - accuracy: 0.9966 - loss: 0.0117 - val_accuracy: 0.9850 - val_loss: 0.0612
        Epoch 5/10
        375/375
                                                   4s 11ms/step - accuracy: 0.9972 - loss: 0.0089 - val_accuracy: 0.9857 - val_loss: 0.0629
        Epoch 6/10
                                                   4s 11ms/step - accuracy: 0.9979 - loss: 0.0061 - val_accuracy: 0.9863 - val_loss: 0.0630
        375/375
        Epoch 7/10
        375/375
                                                   4s 10ms/step - accuracy: 0.9983 - loss: 0.0051 - val_accuracy: 0.9837 - val_loss: 0.0694
        Epoch 8/10
        375/375
                                                   4s 11ms/step - accuracy: 0.9980 - loss: 0.0057 - val_accuracy: 0.9847 - val_loss: 0.0712
        Epoch 9/10
        375/375
                                                   4s 10ms/step - accuracy: 0.9985 - loss: 0.0038 - val_accuracy: 0.9847 - val_loss: 0.0725
        Epoch 10/10
        375/375
                                                   4s 11ms/step - accuracy: 0.9985 - loss: 0.0060 - val_accuracy: 0.9817 - val_loss: 0.0810
            1.0
            0.8
            0.6
         Accuracy
           0.4
            0.2
                                                                  accuracy
                                                                  val_accuracy
            0.0
                               2
                  0
                                            4
                                                         6
                                                                      8
                                             Epoch
        94/94 - 0s - 2ms/step - accuracy: 0.9817 - loss: 0.0810
        Validation Accuracy: 0.9816666841506958
                                    uniq_id
                                                        crawl_timestamp \
        0 4fb99d98225f415e7ece96938e95628f 2015-12-20 08:26:17 +0000
           4ea284c8d38b2ea97a1c2a26f34e057c 2015-12-20 08:26:17 +0000
           ee6ce2c7045c54257e2a0b590e09c296 2015-12-20 08:26:17 +0000
           e797ba3b5f2e2d1fdc520e48486ab60e 2015-12-20 08:26:17 +0000
           f4d8d43858c8858c68d75ce07ac641c0 2015-12-20 08:26:17 +0000
                                                  product_url \
        0 http://www.flipkart.com/v-v-art-brass-bracelet...
           http://www.flipkart.com/kalpaveda-copper-cuff/...
           http://www.flipkart.com/thelostpuppy-book-cove...
        3 http://www.flipkart.com/riana-copper-bangle/p/...
           http://www.flipkart.com/inox-jewelry-stainless...
                                                                     pid retail_price \
                                         product_name
                               V&V ART Brass Bracelet BBAE6NYHCDTEZJTB
        0
                                                                                 470.0
                                                       BBAEDFFKZJTY7SZZ
                         Kalpaveda Copper Copper Cuff
                                                                                1200.0
           Thelostpuppy Book Cover for Apple iPad Air ACCEA4DZH6M5SFVH
        2
                                                                                2199.0
                           Riana Copper Copper Bangle BBAEAXFQHHMF3EYZ
                                                                                2499.0
        3
                    Inox Jewelry Stainless Steel Cuff BBAECH63WYDG6TE2
                                                                                1629.0
        4
           discounted_price
                              ["http://img6a.flixcart.com/image/bangle-brace...
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                      423.0
                              ["http://img6a.flixcart.com/image/bangle-brace...
        1
                              ["http://img5a.flixcart.com/image/cases-covers...
        2
        3
                      649.0
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                             ["http://img6a.flixcart.com/image/bangle-brace...
           is_FK_Advantage_product
                                                                           description \
        0
                              False V&V ART Brass Bracelet - Buy V&V ART Brass Bra...
                              False Kalpaveda Copper Copper Cuff\n
        1
                              False Thelostpuppy Book Cover for Apple iPad Air (Mu...
        3
                             False Riana Copper Copper Bangle - Buy Riana Copper ...
        4
                             False Inox Jewelry Stainless Steel Cuff\n
                product_rating
                                     overall_rating
                                                             brand \
        O No rating available No rating available
                                                           V&V ART
           No rating available No rating available
                                                         Kalpaveda
           No rating available No rating available Thelostpuppy
                             5
                                                             Riana
           No rating available No rating available Inox Jewelry
                                      product_specifications
                                                                  predicted_category
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                                                                           Jewellery
            {"product_specification"=>[{"key"=>"Stretchabl...
                                                                           Jewellerv
           {"product_specification"=>[{"key"=>"Brand", "v... Mobiles & Accessories
           {"product_specification"=>[{"key"=>"Collection...
                                                                           Jewellery
        4 {"product_specification"=>[{"key"=>"Stretchabl...
                                                                           Jewellery
```

In [

['label\_encoder.pkl']