Applied Machine Learning Coursework (001293509)

▼ Import libraries

```
!pip install -U pyforest hvplot
!pip install --upgrade spacy
!python -m spacy download en_core_web_sm
    Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
    Collecting pyforest
      Downloading pyforest-1.1.0.tar.gz (15 kB)
      Preparing metadata (setup.py) ... done
    Collecting hyplot
      Downloading hvplot-0.8.3-py2.py3-none-any.whl (3.2 MB)
                                                  - 3.2/3.2 MB 54.1 MB/s eta 0:00:00
    Requirement already satisfied: bokeh>=1.0.0 in /usr/local/lib/python3.9/dist-packages (from hvplot) (2.4.3)
    Requirement already satisfied: holoviews>=1.11.0 in /usr/local/lib/python3.9/dist-packages (from hvplot) (1.15.4)
    Requirement already satisfied: colorcet>=2 in /usr/local/lib/python3.9/dist-packages (from hvplot) (3.0.1)
    Requirement already satisfied: packaging in /usr/local/lib/python3.9/dist-packages (from hvplot) (23.0)
    Requirement already satisfied: numpy>=1.15 in /usr/local/lib/python3.9/dist-packages (from hvplot) (1.22.4)
    Requirement already satisfied: panel>=0.11.0 in /usr/local/lib/python3.9/dist-packages (from hvplot) (0.14.4)
    Requirement already satisfied: pandas in /usr/local/lib/python3.9/dist-packages (from hvplot) (1.5.3)
    Requirement already satisfied: param>=1.9.0 in /usr/local/lib/python3.9/dist-packages (from hyplot) (1.13.0)
    Requirement already satisfied: tornado>=5.1 in /usr/local/lib/python3.9/dist-packages (from bokeh>=1.0.0->hvplot) (6.2)
    Requirement already satisfied: pillow>=7.1.0 in /usr/local/lib/python3.9/dist-packages (from bokeh>=1.0.0->hvplot) (8.4.0
    Requirement already satisfied: PyYAML>=3.10 in /usr/local/lib/python3.9/dist-packages (from bokeh>=1.0.0->hvplot) (6.0)
    Requirement already satisfied: Jinja2>=2.9 in /usr/local/lib/python3.9/dist-packages (from bokeh>=1.0.0->hvplot) (3.1.2)
    Requirement already satisfied: typing-extensions>=3.10.0 in /usr/local/lib/python3.9/dist-packages (from bokeh>=1.0.0->hv
    Requirement already satisfied: pyct>=0.4.4 in /usr/local/lib/python3.9/dist-packages (from colorcet>=2->hvplot) (0.5.0)
    Requirement already satisfied: pyviz-comms>=0.7.4 in /usr/local/lib/python3.9/dist-packages (from holoviews>=1.11.0->hvpl
    Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.9/dist-packages (from pandas->hvplot) (2022.7.1)
    Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.9/dist-packages (from pandas->hvplot) (2.
    Requirement already satisfied: requests in /usr/local/lib/python3.9/dist-packages (from panel>=0.11.0->hvplot) (2.27.1)
    Requirement already satisfied: markdown in /usr/local/lib/python3.9/dist-packages (from panel>=0.11.0->hvplot) (3.4.3)
    Requirement already satisfied: tqdm>=4.48.0 in /usr/local/lib/python3.9/dist-packages (from panel>=0.11.0->hvplot) (4.65.
    Requirement already satisfied: setuptools>=42 in /usr/local/lib/python3.9/dist-packages (from panel>=0.11.0->hvplot) (67.
    Requirement already satisfied: bleach in /usr/local/lib/python3.9/dist-packages (from panel>=0.11.0->hvplot) (6.0.0)
    Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.9/dist-packages (from Jinja2>=2.9->bokeh>=1.0.0-
    Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.9/dist-packages (from python-dateutil>=2.8.1->pandas->h
    Requirement already satisfied: webencodings in /usr/local/lib/python3.9/dist-packages (from bleach->panel>=0.11.0->hvplot
    Requirement already satisfied: importlib-metadata>=4.4 in /usr/local/lib/python3.9/dist-packages (from markdown->panel>=(
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.9/dist-packages (from requests->panel>=0.11.6
    Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.9/dist-packages (from requests->panel>=0.1
    Requirement already satisfied: charset-normalizer~=2.0.0 in /usr/local/lib/python3.9/dist-packages (from requests->panel>
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.9/dist-packages (from requests->panel>=0.11.0->hvpl
    Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.9/dist-packages (from importlib-metadata>=4.4->markdov
    Building wheels for collected packages: pyforest
      Building wheel for pyforest (setup.py) \dots done
      \texttt{Created wheel for pyforest: filename=pyforest-1.1.0-py2.py3-none-any.whl size=14606 sha256=cle894106d0f6425194c2df6156c2}
      Stored in directory: /root/.cache/pip/wheels/d5/1a/3e/6193felc56168f5df4aef57d8411033ba4611881135d495727
    Successfully built pyforest
    Installing collected packages: pyforest, hvplot
    Successfully installed hvplot-0.8.3 pyforest-1.1.0
    Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
    Requirement already satisfied: spacy in /usr/local/lib/python3.9/dist-packages (3.5.1)
    Collecting spacy
      Downloading spacy-3.5.2-cp39-cp39-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (6.6 MB)
                                                  • 6.6/6.6 MB 36.2 MB/s eta 0:00:00
    Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.0.7)
    Requirement already satisfied: thinc<8.2.0,>=8.1.8 in /usr/local/lib/python3.9/dist-packages (from spacy) (8.1.9)
    Requirement already satisfied: srsly<3.0.0,>=2.4.3 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.4.6)
    Requirement already satisfied: typer<0.8.0,>=0.3.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (0.7.0)
    Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (23.0)
    Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.3.0)
    Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.0.8)
    Requirement already satisfied: setuptools in /usr/local/lib/python3.9/dist-packages (from spacy) (67.6.1)
import holoviews as hv
import hyplot.pandas
import itertools
import matplotlib.pyplot as plt
import numpy as np
import nltk
import pandas as pd
import re
import sklearn.model_selection
import spacy
import string
import seaborn as sns
import tensorflow as tf
import io
```

```
nltk.download('omw-1.4')
nltk.download('stopwords')
nltk.download('wordnet')
nltk.download('punkt')
    [nltk_data] Downloading package omw-1.4 to /root/nltk_data...
    [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk data] Unzipping corpora/stopwords.zip.
    [nltk data] Downloading package wordnet to /root/nltk data...
    [nltk_data] Downloading package punkt to /root/nltk_data...
    [nltk_data] Unzipping tokenizers/punkt.zip.
from google.colab import files
from imblearn.pipeline import Pipeline as ImbPipeline
from imblearn.over_sampling import RandomOverSampler, SMOTE
from keras import datasets, layers, models
from nltk import tokenize, MWETokenizer, TreebankWordTokenizer
from scipy import stats
from sklearn.base import TransformerMixin
from sklearn.ensemble import RandomForestClassifier
from sklearn.experimental import enable_halving_search_cv
from sklearn.feature_extraction.text import TfidfTransformer, TfidfVectorizer
from sklearn.impute import SimpleImputer
from sklearn.linear_model import SGDClassifier, LogisticRegression
from sklearn.metrics import confusion matrix, ConfusionMatrixDisplay, f1 score, accuracy score, roc curve, auc, roc auc score,
from sklearn.model_selection import GridSearchCV, RandomizedSearchCV, HalvingGridSearchCV, train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.neighbors import NearestNeighbors
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.svm import LinearSVC
from sklearn.tree import DecisionTreeClassifier
from pyforest import *
from tensorflow import keras
from time import time as tt
```

▼ Load Dataset

```
url = 'https://drive.google.com/file/d/10zV9L_LogAHwlaX0okgInnqAmW6QRQNn/view?usp=sharing'
data_file = 'https://drive.google.com/uc?export=download&id='+url.split('/')[-2]
data = pd.read_csv(data_file,encoding='latin-1')
data.head()
```

	text	label	query_index	1
0	Can I automatically top-up when traveling?	top_up_queries_or_issues	526cd7f17526	
1	What kind of fiat currency can I used for hold	other	f3cf7343067e	
2	I did not get the item I ordered. How should \dots	other	9a19501c3a3c	
3	Freeze my account it's been hacked.	needs_troubleshooting	d76b07db8cf8	
4	is there a reason that my payment didnt go thr	other	bd95ba09a18d	

data.describe()

	text	label	query_index	1
count	14195	13674	14195	
unique	13084	8	13672	
top	#	other	fc9b781a6b97	
freq	68	5036	2	

Data pre processing

```
data.isna().sum()

text 0
label 521
query_index 0
dtype: int64
```

```
data.count()
    text
                   14195
    label
                   13674
    query_index
                    14195
    dtype: int64
for i in list(data.columns):
 print(data[i].value_counts())
    I topped up but the app did not accept it.
    How do I know if my top up was unsuccessful?
    Oh my goodness, my card has been declined twice at ATM! I tried two different ATM, but each one declined my card! Can you
    I don't understand where this debit came from and want it removed.
    My statement shows different transaction times.
    What are the steps I need to take to cancel a transaction?
    after i got married i need to change my name
    I still have not received an answer as to why I was charged $1.00 in a transaction?
    Am I able to track the card that was just sent to me?
    Name: text, Length: 13084, dtype: int64
    other
                                 5036
    needs troubleshooting
                                 4305
                                 2598
    card queries or issues
    top_up_queries_or_issues
                                 1684
    Other
                                   21
    Card_queries_or_issues
                                   12
    Top_up_queries_or_issues
                                   11
    Needs_troubleshooting
    Name: label, dtype: int64
    fc9b781a6b97
    ea614b5c8b9a
    0c50afa79ab7
    a6e840dd13ce
    5fc6b0b669aa
                    2
    a0b6c96420d2
    1bb97b06d70c
    1ec16e8d1edd
    42d144e1974c
    f7e5a9b88449
    Name: query_index, Length: 13672, dtype: int64
```

▼ Data quality assessment and Exploratory Data Analysis for cleaning data

Drop rows that have # symbol in text column

```
print(data[data["text"] == "#"])
          t.ext.
                                    label
                                           query_index
                   needs_troubleshooting 226t0c5be7cf
    106
             #
    139
             #
                                    other 10et93272e2e
    266
                                    other
                                           849t6f85a049
                                    other 73ft45a21d0f
    346
    530
             #
                  card_queries_or_issues
                                           ef6teff01216
                                      . . .
    13505
             #
                top_up_queries_or_issues
                                           33dte465441a
                   needs_troubleshooting
    13531
                                           451t8fa10c0d
    13799
             #
                                    other
                                           b25teb82e4a2
    13825
                top_up_queries_or_issues
                                           763tcd9275b1
    13884
                                    other 9eetccf1515a
    [68 rows x 3 columns]
data = data.drop(data[data.text == '#'].index)
Start all sentences with capital letter in text column
data['text'] = data['text'].apply(lambda x: x.capitalize())
print(data)
                  Can i automatically top-up when traveling?
           What kind of fiat currency can i used for hold...
           I did not get the item i ordered. how should ...
                         Freeze my account it's been hacked.
    3
           Is there a reason that my payment didnt go thr...
    4
          Can you tell me what the disposable cards are ...
    14191 The atm won't give me my card back. i need it ...
```

```
14192 Can you please tell me why my card payments ar...
    14193 The rate for a currency exchange was wrong whe...
    14194 Am i able to track the card that was just sent...
                               label
                                      query index
           top_up_queries_or_issues 526cd7f17526
    0
    1
                              other f3cf7343067e
    2
                              other 9a19501c3a3c
              needs troubleshooting d76b07db8cf8
    3
                              other bd95ba09a18d
    4
             card_queries_or_issues bd6df98cc746
    14190
    14191
             card_queries_or_issues e6197a1334b3
    14192
             card_queries_or_issues b922a2a5f687
    14193
              needs_troubleshooting cb1ed2c3ca95
    14194
             card_queries_or_issues f7e5a9b88449
    [14127 rows x 3 columns]
Fill empty values of label column with the most frequent value, in our case is other
print(data["label"].isnull().sum())
    521
'''# handling missing data
from sklearn.impute import SimpleImputer
# 1. Imputer
imptr_empl = SimpleImputer(missing_values = np.nan, strategy = 'most_frequent')
# 2. Fit the imputer object to the feature matrix
imptr_empl = imptr_empl.fit(data[['label']])
# 3. Call Transform to replace missing data in train_dataset (on specific columns) by the most frequent of the column to which
data[['label']] = imptr_empl.transform(data[['label']]) '''
     # handling missing data\nfrom sklearn.impute import SimpleImputer \n# 1. Imputer\nimptr empl = SimpleImputer(missing va)
    t_frequent') \n\n# 2. Fit the imputer object to the feature matrix\nimptr_empl = imptr_empl.fit(data[['label']])\n\n# 3.
    ssing data in train_dataset (on specific columns) by the most frequent of the column to which that missing data belongs t
    pl.transform(data[['label']])
Removing duplicates
1 =list(data.columns)
for i in 1:
 print(data[i].value_counts())
    I would like to exchange currencies
    How do i know if my top up was unsuccessful?
    Oh my goodness, my card has been declined twice at atm! i tried two different atm, but each one declined my card! can you
    I don't understand where this debit came from and want it removed.
    What do i do if the atm took my card?
    My statement shows different transaction times.
    What are the steps i need to take to cancel a transaction?
    After i got married i need to change my name
    I still have not received an answer as to why i was charged $1.00 in a transaction?
    Am i able to track the card that was just sent to me?
    Name: text, Length: 13083, dtype: int64
                                 5004
    needs troubleshooting
                                 4293
    card_queries_or_issues
                                 2586
    {\tt top\_up\_queries\_or\_issues}
                                 1672
    Other
                                   2.1
    Card_queries_or_issues
                                   12
    Top_up_queries_or_issues
                                   11
    Needs_troubleshooting
                                    7
    Name: label, dtype: int64
    bda7ab74290d
    dcd5dac08c5f
    d921fa76a483
    a862cbe02605
                    2
    529f3f67f869
                    2
```

Check if we have removed duplicates by renaming it

Name: query index, Length: 13604, dtype: int64

1

1

19fdb4300302

40b009669cb4

ba221ecd6e40 ec1b8fbbcc65 f7e5a9b88449

```
data.loc[data.label == 'Other', 'label'] = 'other'
data.loc[data.label == 'Needs_troubleshooting','label'] = 'needs_troubleshooting'
data.loc[data.label == 'Card_queries_or_issues', 'label'] = 'card_queries_or_issues'
data.loc[data.label == 'Top_up_queries_or_issues','label'] = 'top_up_queries_or_issues'
data=data.dropna()
data = data.drop_duplicates()
data['label'].value counts()
                                 4803
    needs_troubleshooting
                                 4153
    card queries or issues
                                 2495
    top_up_queries_or issues
                                 1632
    Name: label, dtype: int64
data.count()
    text
                   13083
    label
                   13083
    query_index
                    13083
    dtvpe: int64
```

▼ ENCODING CATEGORICAL DATA

3

```
# First check: what are the target categories?
data.value_counts()
    text
                                query_index
    \n\nwhat businesses accept this card?
    card queries or issues 532693cb5d73
    My withdrawal is pending, why?
    other
                               8b98db331107
    My wallet doesn't show my recent top up.
    top up queries or issues e78731b51985
    My wallet got stolen a couple hours ago and now i've seen there already is a withdrawal. help this is absolutely urgent
    i don't want to loose more money needs_troubleshooting My wallet is empty even though i topped it up an hour ago.
                                                                    9c262a198983
    top_up_queries_or_issues aa2c910b42e1
    I don't think the charges made when i had currency exchanged are right.
    needs troubleshooting
                              11ce0fdc214a
                                                1
    I don't think the exchange rate was right.
    needs troubleshooting
                               73b70bf28dfa
                                                 1
    I don't think the transaction has gone through, so can i cancel a transfer?
    other
                                60a77bb2ef8d
                                                1
    I don't understand how to top up my account, can you please explain the process?
                                6a0ad9821af4
    Ýýýl was in my statement as an extra fee.
    needs_troubleshooting
                               b03e2605e919
    Length: 13083, dtype: int64
temp = data.copy() #we make a copy and use a temporary variable name, since this is not the final transformation
# encode categorical data for the 'Label' column
# create an object of the LabelEncoder class
lblEncoder_X = LabelEncoder()
# apply LblEncoder object to our categorical variables (columns - 'Label') using the fit_transform method. This returns the co
temp['label'] = lblEncoder X.fit transform(temp['label']) # we can fit and transform all at once
temp.head()
                                                                     1
                                         text label query_index
            Can i automatically top-up when traveling?
                                                      526cd7f17526
                                                       f3cf7343067e
         What kind of fiat currency can i used for hold...
     2
          I did not get the item i ordered, how should ...
                                                   2 9a19501c3a3c
```

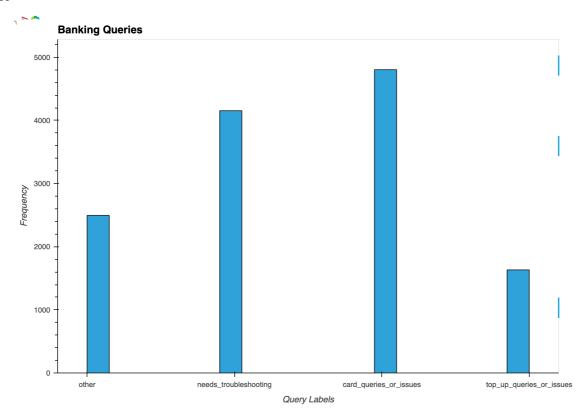
1 d76b07db8cf8

2 bd95ba09a18d

Freeze my account it's been hacked.

4 Is there a reason that my payment didnt go thr...

```
prot - temp[ raber ].nvprot.nrst(
    frame_height=500, frame_width=750,
    xlabel='Query Labels', ylabel='Frequency',
    title='Banking Queries',legend='bottom', xticks = [(0, labl[0]), (1, labl[1]), (2, labl[2]), (3, labl[3])],
    alpha=1, muted_alpha=0, muted_fill_alpha=0, muted_line_alpha=0.5,
    tools=['pan', 'box_zoom', 'wheel_zoom', 'undo', 'redo', 'hover', 'save', 'reset']
)
hv.extension('bokeh')
plot
```



▼ TEXT PRE-PROCESSING

```
!pip install ml-datasets
!pip install --upgrade spacy
!python -m spacy download en core web sm
!python -m spacy download en_core_web_md
import sklearn.model_selection
import spacy
import seaborn as sns
nlp = spacy.load('en core web sm')
    Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
    Collecting ml-datasets
       Downloading ml_datasets-0.2.0-py3-none-any.whl (15 kB)
    Requirement already satisfied: catalogue<3.0.0,>=0.2.0 in /usr/local/lib/python3.9/dist-packages (from ml-datasets) (2.0.
    Requirement already satisfied: numpy>=1.7.0 in /usr/local/lib/python3.9/dist-packages (from ml-datasets) (1.22.4)
    Requirement already satisfied: tqdm<5.0.0,>=4.10.0 in /usr/local/lib/python3.9/dist-packages (from ml-datasets) (4.65.0)
    Requirement already satisfied: srsly<3.0.0,>=1.0.1 in /usr/local/lib/python3.9/dist-packages (from ml-datasets) (2.4.6)
    Installing collected packages: ml-datasets
    Successfully installed ml-datasets-0.2.0
    Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>/ Requirement already satisfied: spacy in /usr/local/lib/python3.9/dist-packages (3.5.2)
    Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (23.0)
    Requirement already satisfied: preshed<3.1.0,>=3.0.2 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.0.8)
    Requirement already satisfied: setuptools in /usr/local/lib/python3.9/dist-packages (from spacy) (67.6.1)
    Requirement already satisfied: requests<3.0.0,>=2.13.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.27.1)
    Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.3.0)
    Requirement already satisfied: srsly<3.0.0,>=2.4.3 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.4.6)
    Requirement already satisfied: pydantic!=1.8,!=1.8.1,<1.11.0,>=1.7.4 in /usr/local/lib/python3.9/dist-packages (from space)
    Requirement already satisfied: pathy>=0.10.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (0.10.1)
    Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (1.0.4)
    Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in /usr/local/lib/python3.9/dist-packages (from spacy) (1.1.1)
Requirement already satisfied: thinc<8.2.0,>=8.1.8 in /usr/local/lib/python3.9/dist-packages (from spacy) (8.1.9)
    Requirement already satisfied: jinja2 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.1.2)
    Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (4.65.0)
    Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (1.0.9)
    Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.0.7)
    Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /usr/local/lib/python3.9/dist-packages (from spacy) (2.0.8)
    Requirement already satisfied: typer<0.8.0,>=0.3.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (0.7.0)
    Requirement already satisfied: smart-open<7.0.0,>=5.2.1 in /usr/local/lib/python3.9/dist-packages (from spacy) (6.3.0)
    Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in /usr/local/lib/python3.9/dist-packages (from spacy) (3.0.12
```

```
Requirement already satisfied: numpy>=1.15.0 in /usr/local/lib/python3.9/dist-packages (from spacy) (1.22.4)
Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/python3.9/dist-packages (from pydantic!=1.8,!=1)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.9/dist-packages (from requests<3.0.0,>=2.13.0->spackage (requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.9/dist-packages (from requests<3.0.0,>=2.13.0
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.9/dist-packages (from requests<3.0.0,>=2.13.0
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.9/dist-packages (from requests<3.0.0,>=2.1
Requirement already satisfied: blis<0.8.0,>=0.7.8 in /usr/local/lib/python3.9/dist-packages (from thinc<8.2.0,>=8.1
Requirement already satisfied: confection<1.0.0,>=0.0.1 in /usr/local/lib/python3.9/dist-packages (from typer<0.8.0,>=0.3.0->seequirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.9/dist-packages (from jinja2->spacy) (2.1.2)
2023-04-12 10:24:20.890012: W tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Could not find Tensor Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/Collecting en-core-web-sm==3.5.0

Downloading https://github.com/explosion/spacy-models/releases/download/en_core_web_sm=3.5.0/en_core_web_sm=3.5.0-py3-r
```

Requirement already satisfied: spacy<3.6.0,>=3.5.0 in /usr/local/lib/python3.9/dist-packages (from en-core-web-sm==3.5.0) Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0-> Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0-> Requirement already satisfied: pathy>=0.10.0 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core Requirement already satisfied: pydantic=1.8,!=1.8.1,<1.11.0,>=1.7.4 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core Requirement already satisfied: cymem<2.1.0,>=2.0.2 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core-web-sn Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core-web-sn already satisfied: typer<0.8.0,>=0.3.0 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core-web-sn already satisfied: packaging>=20.0 in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core-we Requirement already satisfied: setuptools in /usr/local/lib/python3.9/dist-packages (from spacy<3.6.0,>=3.5.0->en-core-we Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in /usr/local/lib/python3.9/dist-packa

▼ REMOVING STOP WORDS, PUNCTUATIONS and Lemmatisation

```
import string
#removing punctuations
def rem_pun(txt):
 txt nopunt = "".join([c for c in txt if c not in string.punctuation])
 return txt nopunt
data['no_punc'] = data['text'].apply(lambda x: rem_pun(x))
#tokenisation
def tokenize(column):
    tokenizer = TreebankWordTokenizer()
    result = tokenizer.tokenize(column)
   return result
data['tokens'] = data.apply(lambda x: tokenize(x['no_punc'].lower()), axis=1)
#removing stopwords
stopwords = nltk.corpus.stopwords.words('english')
def remove stop words(tokens):
 txt_clean = [word for word in tokens if word not in stopwords]
data['clean_text']=data['tokens'].apply(lambda x: remove_stop_words(x))
data['clean text']=data['clean text'].apply(lambda x: ' '.join([str(i) for i in x]))
data['clean_text'].head()
    0
                              automatically topup traveling
                  kind fiat currency used holding exchange
    1
         get item ordered go cancel order payment pleas...
    2
                                      freeze account hacked
                                    reason payment didnt go
    Name: clean_text, dtype: object
nltk.download('wordnet')
    [nltk_data] Downloading package wordnet to /root/nltk_data...
                  Package wordnet is already up-to-date!
    [nltk_data]
    True
#Lemmatization
wn = nltk.WordNetLemmatizer()
ps = nltk.PorterStemmer()
def lemmatization(tokens):
 text = [wn.lemmatize(word) for word in tokens]
 return text
\label{eq:dataset} \verb|data['lemmatised'] = \verb|data['clean_text'].apply(lambda x : lemmatization(x))| \\
data['lemmatised'] = data['lemmatised'].apply(lambda x: ''.join(map(str, x)))
data['lemmatised'].head()
```

0

1

```
automatically topup traveling
kind fiat currency used holding exchange
get item ordered go cancel order payment pleas...
freeze account hacke
```

#data after preprocessing
data.head()

clear	tokens	no_punc	query_index	label	text	
automatically topup t	[can, i, automatically, topup, when, traveling]	Can i automatically topup when traveling	526cd7f17526	top_up_queries_or_issues	Can i automatically top-up when traveling?	0
kind fiat curren holding ex	[what, kind, of, fiat, currency, can, i, used,	What kind of fiat currency can i used for hold	f3cf7343067e	other	What kind of fiat currency can i used for hold	1
get item ordered go order payment	[i, did, not, get, the, item, i, ordered, how,	I did not get the item i ordered how should i	9a19501c3a3c	other	I did not get the item i ordered. how should	2
freeze account	[freeze, my, account, its, been, hacked]	Freeze my account its been hacked	d76b07db8cf8	needs_troubleshooting	Freeze my account it's been hacked.	3
reason payment ([is, there, a, reason, that, my, payment, didn	Is there a reason that my payment didnt go thr	bd95ba09a18d	other	Is there a reason that my payment didnt go thr	4



Final data with clean text and labels

```
data_final = pd.DataFrame(columns=['text','clean','label'])
data_final['text'] = data['text']
data_final['clean'] = data['clean_text']
data_final['label'] = data['label']
data_final.head()
```

	text	clean	label	1
0	Can i automatically top-up when traveling?	automatically topup traveling	top_up_queries_or_issues	
1	What kind of fiat currency can i used for hold	kind fiat currency used holding exchange	other	
2	I did not get the item i ordered. how should	get item ordered go cancel order payment pleas	other	
3	Freeze my account it's been hacked.	freeze account hacked	needs_troubleshooting	
4	Is there a reason that my payment didnt go thr	reason payment didnt go	other	

```
lblEncoder_X = LabelEncoder()
# apply LblEncoder object to our categorical variables (columns - 'Label') using the fit_transform method. This returns the cc
data_final['label'] = lblEncoder_X.fit_transform(data_final['label'])
data_final.head()
```

0+	label	clean	text	
	3	automatically topup traveling	Can i automatically top-up when traveling?	0
	2	kind fiat currency used holding exchange	What kind of fiat currency can i used for hold	1
	2	get item ordered go cancel order payment pleas	I did not get the item i ordered. how should	2
	1	freeze account hacked	Freeze my account it's been hacked.	3
	2	reason payment didnt go	Is there a reason that my payment didnt go thr	4

Dataset split and Feature Scaling

```
X_train, X_test, y_train, y_test = train_test_split(data_final["clean"],data_final["label"],test_size=0.35,shuffle=True)
tfidf_vectorizer = TfidfVectorizer(use_idf=True)

X_train_vectors_tfidf = tfidf_vectorizer.fit_transform(X_train)

X_test_vectors_tfidf = tfidf_vectorizer.transform(X_test)
```

▼ Classification Task

```
#FITTING THE CLASSIFICATION MODEL using Naive Bayes(tf-idf)
nb_tfidf = MultinomialNB()
```

18

```
#nb tfidf.fit(X train vectors tfidf, y train)
nb_tfidf.fit(X_train_vectors_tfidf, y_train)
#Predict y value for test dataset
y predict = nb tfidf.predict(X test vectors tfidf)
y_prob = nb_tfidf.predict_proba(X_test_vectors_tfidf)[:,1]
print('Accuracy: {:.2f}'.format(accuracy_score(y_test, y_predict)))
print(classification_report(y_test,y_predict))
print('Confusion Matrix:\n',confusion_matrix(y_test, y_predict))
    Accuracy: 0.82
                  precision
                               recall f1-score
                                                  support
               0
                       0.90
                                 0.75
                                            0.82
                                                      887
               1
                       0.75
                                 0.89
                                            0.82
                                                      1412
               2
                       0.81
                                 0.82
                                            0.82
                                                     1712
               3
                       0.98
                                 0.74
                                            0.84
                                                       569
                                            0.82
                                                      4580
        accuracy
                       0.86
                                 0.80
                                                      4580
       macro avg
                                            0.82
    weighted avg
                       0.83
                                 0.82
                                            0.82
                                                      4580
    Confusion Matrix:
                         71
     [[ 668 115 97
     [ 30 1261 120
                        11
     [ 44 255 1411
                        21
        4 40 105 420]]
#Classifying the data Using Random Forest Classifier model
classifier = RandomForestClassifier(n_estimators=100, max_depth=5, random_state=100)
classifier.fit(X_train_vectors_tfidf, y_train)
# Predicting the Test set results
y pred = classifier.predict(X test vectors tfidf)
y_prob = classifier.predict_proba(X_test_vectors_tfidf)[:,1]
print('Accuracy: {:.2f}'.format(accuracy_score(y_test, y_pred)))
print(classification_report(y_test,y_pred))
print('Confusion Matrix: \n',confusion_matrix(y_test, y_pred))
    Accuracy: 0.53
                  precision
                               recall f1-score
                                                  support
               0
                       0.93
                                 0.09
                                            0.17
                                                      887
                       0.85
                                 0.51
                                            0.64
                                                      1412
               1
               2
                       0.45
                                 0.96
                                            0.61
                                                     1712
               3
                       0.00
                                 0.00
                                            0.00
                                                       569
                                            0.53
                                                     4580
        accuracy
       macro avg
                       0.56
                                 0.39
                                            0.35
                                                      4580
    weighted avg
                       0.61
                                 0.53
                                            0.46
                                                      4580
    Confusion Matrix:
     [[ 82 17 788
                         0 ]
         6 720 686
                        0]
         0
            73 1639
                        01
        0 36 533
                        0]]
    /usr/local/lib/python3.9/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-%
      _warn_prf(average, modifier, msg_start, len(result))
     /usr/local/lib/python3.9/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-%
      _warn_prf(average, modifier, msg_start, len(result))
    /usr/local/lib/python3.9/dist-packages/sklearn/metrics/ classification.py:1344: UndefinedMetricWarning: Precision and F-s
      warn prf(average, modifier, msg start, len(result))
#Classifying the data Using SGD Classifier model (Optional)
clf1 = SGDClassifier(loss="log loss", penalty="12")
clf1.fit(X_train_vectors_tfidf, y_train)
y pre = clf1.predict(X test vectors tfidf)
print('Accuracy: {:.2f}'.format(accuracy_score(y_test, y_pre)))
print(classification_report(y_test,y_pre))
print('Confusion Matrix: \n',confusion_matrix(y_test, y_pre))
    Accuracy: 0.87
                  precision
                               recall f1-score
                                                   support.
               0
                       0.88
                                 0.88
                                            0.88
                                                      887
               1
                       0.84
                                 0.88
                                            0.85
                                                      1412
                       0.87
                                 0.85
                                            0.86
                                                      1712
               2
                       0.95
                                 0.92
                                            0.93
               3
                                                       569
                                            0.87
                                                      4580
        accuracy
                        0.88
                                  0.88
                                            0.88
                                                      4580
       macro avg
                       0.87
                                 0.87
                                            0.87
                                                      4580
    weighted avg
```

```
Confusion Matrix:

[[ 781      42      48      16]

[ 37 1236      139      0]

[ 65      186 1449     12]

[ 9      16      23      521]]
```

Logistic Regression

```
# create custom class to add to the pipeline
from sklearn.base import TransformerMixin
class DenseTransformer(TransformerMixin):
    ''' Pipeline step to transform a sparse matrix into a dense one '''
    def fit(self, X, y=None, **fit_params):
        return self
   def transform(self, X, y=None, **fit_params):
       return X.toarray()
#FITTING THE CLASSIFICATION MODEL using Logistic Regression(tf-idf)
lr_tfidf=LogisticRegression(solver = 'liblinear', C=10, penalty = '12')
lr_tfidf.fit(X_train_vectors_tfidf, y_train)
#Predict y value for test dataset
y_predict = lr_tfidf.predict(X_test_vectors_tfidf)
y_prob = lr_tfidf.predict_proba(X_test_vectors_tfidf)[:,1]
print(classification_report(y_test,y_predict))
print('Confusion Matrix:',confusion_matrix(y_test, y_predict))
# fpr, tpr, thresholds = roc_curve(y_test, y_prob)
# roc_auc = auc(fpr, tpr)
# print('AUC:', roc_auc)
                  precision
                               recall f1-score
                       0.89
                                 0.89
                                           0.89
                       0.85
                                 0.89
                                           0.87
                                                      1412
                       0.88
                                 0.86
                                           0.87
                       0.94
               3
                                 0.93
                                           0.94
                                                      569
        accuracy
                                           0.88
                                                     4580
                       0.89
                                 0.89
       macro avg
                                           0.89
                                                     4580
    weighted avg
                       0.88
                                 0.88
                                           0.88
                                                     4580
    Confusion Matrix: [[ 788
                               37 47
                                       15]
     [ 27 1252 130
                        3 ]
                      15]
     [ 58 168 1471
            15
                 16 530]]
#Logistic Regression
lr tfidf=LogisticRegression(solver = 'liblinear', C=10, penalty = '12')
lr_tfidf.fit(X_train_vectors_tfidf, y_train)
#Predict y value for test dataset
y_predict = lr_tfidf.predict(X_test_vectors_tfidf)
y_prob = lr_tfidf.predict_proba(X_test_vectors_tfidf)[:,1]
print(classification_report(y_test,y_predict))
print('Confusion Matrix:',confusion matrix(y test, y predict))
                  precision
                               recall f1-score
                                                  support
               0
                       0.89
                                 0.89
                                           0.89
                                                      887
                       0.85
                                 0.89
                                           0.87
                                                      1412
               2
                       0.88
                                 0.86
                                           0.87
                                                     1712
                       0.94
                                 0.93
                                           0.94
                                                      569
                                           0.88
                                                     4580
        accuracy
                       0.89
                                 0.89
                                           0.89
                                                      4580
       macro avg
                                                      4580
    weighted avg
                       0.88
                                 0.88
                                           0.88
    Confusion Matrix: [[ 788
                               37 47
                                        151
     [ 27 1252 130
        58 168 1471
                      15]
         8
            15
                 16
                      530]]
```

▼ Neural Network Model

Pre-process data

```
# create custom class to add to the pipeline
from sklearn.base import TransformerMixin
```

```
class DenseTransformer(TransformerMixin):
    ''' Pipeline step to transform a sparse matrix into a dense one '''
    def fit(self, X, y=None, **fit_params):
        return self
    def transform(self, X, y=None, **fit_params):
        return X.toarray()
#create pipeline
preprocessor = Pipeline(
       steps =[('tfidf', TfidfVectorizer(stop words='english', lowercase= True,
                                          max_features = 3000,
                                          ngram_range=(1,2))),
                ('to dense', DenseTransformer()),
      ]
    )
preprocessor.fit(data_final['clean'])
X_train_preprocessed = preprocessor.transform(X_train)
X_test_preprocessed = preprocessor.transform(X_test)
y_train
     2243
     9053
            3
     3711
            1
    7571
            0
    1613
            0
     6067
             0
     7613
            0
     5223
            2
     1067
     Name: label, Length: 8503, dtype: int64
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
# create a simple model with ONE hidden layer only
model = models.Sequential()
# we create a hidden layer with 20 nodes.
hidden layer nodes = 50
num_of_input_features = X_train_preprocessed.shape[1] #number of features = number of columns in the input matrix
model.add(layers.Input(shape=(num_of_input_features,)))
model.add(layers.Dense(hidden_layer_nodes, activation='relu'))
# let's add a dropout layer
dropout rate = 0.2
model.add(layers.Dropout(rate= dropout_rate))
num_categories = len(y_train.value_counts())
model.add(layers.Dense(num_categories, activation='softmax')) #is it clear why here we use "sigmoid" and use "softmax" for mul
learning_rate = 0.01
model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=learning_rate),
              loss=tf.keras.losses.SparseCategoricalCrossentropy(), # we use this function for MULTI-CLASS PROBLEMS. It expect
              metrics=['accuracy'])
# let's print a summary of the model to see what it's like
print(model.summary())
     Model: "sequential"
                                                            Param #
     Layer (type)
                                Output Shape
     dense (Dense)
                                 (None, 50)
                                                            150050
     dropout (Dropout)
                                  (None, 50)
                                                            0
     dense_1 (Dense)
                                  (None, 4)
                                                            204
     Total params: 150,254
     Trainable params: 150,254
```

https://colab.research.google.com/drive/10JFUYUCjM8fFv_Ahn159igqoMy_yBIQL#scrollTo=Li86JIQROaqs&printMode=true

Non-trainable params: 0

None

```
12/04/2023, 11:42
                                                         code_comp_1804.ipynb - Colaboratory
  y_train.dtype
       dtvpe('int64')
  batch size = 130 # The batch size indicates how many data points we use to compute each update to the parameters of the models
   epochs = 50 #how long we train the model for
   validation split = 0.2 #Fraction of the training data to be used as validation data.
   history = model.fit(X train preprocessed,
                      y_train.to_numpy(), #we add .to_numpy() because Keras doesn't like dataframes so we need to transform the
                      epochs=epochs,
                      batch_size=batch_size,
                      validation_split= validation_split)
       Epoch 1/50
       53/53 [====
                                 ========] - 1s 11ms/step - loss: 0.7526 - accuracy: 0.7146 - val_loss: 0.3634 - val_accuracy
       Epoch 2/50
       53/53 [====
                                     ====== 1 - 0s 6ms/step - loss: 0.2619 - accuracy: 0.9056 - val loss: 0.3285 - val accuracy:
       Epoch 3/50
       53/53 [====
                                    ======] - 0s 5ms/step - loss: 0.1704 - accuracy: 0.9394 - val loss: 0.3320 - val accuracy:
       Epoch 4/50
       53/53 [====
                                   ======= ] - 0s 6ms/step - loss: 0.1302 - accuracy: 0.9541 - val loss: 0.3502 - val accuracy:
       Epoch 5/50
       53/53 [====
                                ========] - 0s 5ms/step - loss: 0.1118 - accuracy: 0.9569 - val loss: 0.3722 - val accuracy:
       Epoch 6/50
       53/53 [===
                                            - 1s 11ms/step - loss: 0.1004 - accuracy: 0.9643 - val_loss: 0.3861 - val_accuracy
       Epoch 7/50
       53/53 [===
                                   ======] - 1s 11ms/step - loss: 0.0867 - accuracy: 0.9685 - val loss: 0.4050 - val accuracy
       Epoch 8/50
       53/53 [====
                                    ======1 - 1s 12ms/step - loss: 0.0807 - accuracy: 0.9702 - val loss: 0.4244 - val accuracy
       Epoch 9/50
       53/53 [====
                                     =====] - 1s 16ms/step - loss: 0.0722 - accuracy: 0.9728 - val loss: 0.4346 - val accuracy
       Epoch 10/50
       53/53 [====
                                            - 1s 12ms/step - loss: 0.0669 - accuracy: 0.9737 - val_loss: 0.4599 - val_accuracy
       Epoch 11/50
       53/53 [===
                                               0s 9ms/step - loss: 0.0632 - accuracy: 0.9774 - val_loss: 0.4729 - val_accuracy:
       Epoch 12/50
       53/53 [====
                                         ==] - 1s 12ms/step - loss: 0.0583 - accuracy: 0.9815 - val loss: 0.4819 - val accuracy
       Epoch 13/50
       53/53 [====
                                    ======] - 1s 11ms/step - loss: 0.0555 - accuracy: 0.9800 - val_loss: 0.5011 - val_accuracy
       Epoch 14/50
       53/53 [=====
                                ========1 - 1s 12ms/step - loss: 0.0552 - accuracy: 0.9790 - val loss: 0.5133 - val accuracy
       Epoch 15/50
       53/53 [=====
                                   ======] - 1s 15ms/step - loss: 0.0526 - accuracy: 0.9810 - val_loss: 0.5150 - val_accuracy
       Epoch 16/50
       53/53 [====
                                            - 1s 14ms/step - loss: 0.0522 - accuracy: 0.9791 - val_loss: 0.5289 - val_accuracy
       Epoch 17/50
       53/53 [==
                                            - 1s 12ms/step - loss: 0.0483 - accuracy: 0.9812 - val loss: 0.5440 - val accuracy
       Epoch 18/50
       53/53 [====
                               =======] - 1s 13ms/step - loss: 0.0493 - accuracy: 0.9819 - val_loss: 0.5496 - val_accuracy
       Epoch 19/50
                           =============== 1 - 1s 12ms/step - loss: 0.0479 - accuracy: 0.9816 - val loss: 0.5484 - val accuracy
       53/53 [=====
       Epoch 20/50
       53/53 [============] - 1s 13ms/step - loss: 0.0476 - accuracy: 0.9824 - val loss: 0.5841 - val accuracy
       Epoch 21/50
       53/53 [=====
                          =============== ] - 1s 15ms/step - loss: 0.0457 - accuracy: 0.9825 - val_loss: 0.5789 - val_accuracy
       Epoch 22/50
       53/53 [===
                                    ======] - 1s 11ms/step - loss: 0.0451 - accuracy: 0.9834 - val loss: 0.5763 - val accuracy
       Epoch 23/50
       53/53 [=============] - 1s 10ms/step - loss: 0.0437 - accuracy: 0.9840 - val loss: 0.5897 - val accuracy
       Epoch 24/50
       53/53 [=====
                             ========== 1 - 1s 14ms/step - loss: 0.0432 - accuracy: 0.9837 - val loss: 0.5945 - val accuracy
       Epoch 25/50
       53/53 [===========] - 1s 11ms/step - loss: 0.0439 - accuracy: 0.9832 - val loss: 0.6032 - val accuracy
       Epoch 26/50
       53/53 [=============] - 1s 16ms/step - loss: 0.0417 - accuracy: 0.9827 - val_loss: 0.6088 - val_accuracy
       Epoch 27/50
       53/53 [======
                         ========= ] - 1s 17ms/step - loss: 0.0408 - accuracy: 0.9835 - val loss: 0.6291 - val accuracy
       Epoch 28/50
       53/53 [=====
                            ========= ] - 1s 16ms/step - loss: 0.0416 - accuracy: 0.9838 - val loss: 0.6040 - val accuracy
       Epoch 29/50
       #Computing the Metrics of model by plotting confusion Matrix
   def plot confusion matrix(cm,target names,title='Confusion matrix',cmap=None,normalize=True):
      accuracy = np.trace(cm) / np.sum(cm).astype('float')
      misclass = 1 - accuracy
      if cmap is None:
          cmap = plt.get_cmap('Blues')
      plt.figure(figsize=(8, 6))
```

plt.imshow(cm, interpolation='nearest', cmap=cmap)

plt.title(title) plt.colorbar()

```
if target names is not None:
        tick_marks = np.arange(len(target_names))
        plt.xticks(tick_marks, target_names, rotation=45)
       plt.yticks(tick marks, target names)
    if normalize:
       cm = cm.astype('int')
    thresh = cm.max() / 1.5 if normalize else cm.max() / 2
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
       if normalize:
            plt.text(j, i, "{:,}".format(cm[i, j]),
                    horizontalalignment="center",
                     color="white" if cm[i, j] > thresh else "black")
        else:
           plt.text(j, i, "{:,}".format(cm[i, j]),
                     horizontalalignment="center",
                     color="white" if cm[i, j] > thresh else "black")
   plt.tight layout()
   plt.xticks(rotation=90, ha='right')
   plt.ylabel('True label')
   plt.xlabel('Predicted label\n\naccuracy={:0.4f}; misclass={:0.4f}'.format(accuracy, misclass))
   plt.show()
# Check performance on test data
test probabilities = model.predict(X test preprocessed)
# Since it's a multiclass problem, the output probabilities are given as one probability PER CLASS
# To get the final predicted label, we need to find the category with the HIGHEST probability
# We can get this by using the function np.argmax()
test_predictions = np.argmax(test_probabilities, axis=1)
# the result should be one integer number per data point that we can compare with the target labels
# let's show the classification report with all the metrics
# think about which metrics you think are the most important ones for this problem!
print(classification_report(y_test.to_numpy(),test_predictions,
                            target_names= labl)) # this is to give the real categories, not their encoded numbers
# let's also print the balanced accuracy score, since we know the dataset is not balanced
print(f'The balanced accuracy score is {balanced_accuracy_score(y_test.to_numpy(),test_predictions):.3f}\n')
# let's get all the numbers for the confusion matrix
cm = confusion matrix(y test.to numpy(),test predictions)
plot_confusion_matrix(cm=cm,target_names= labl)
```

144/144 [===========] - 1s 3ms/step					
	precision	recall	f1-score	support	
other	0.88	0.87	0.87	887	
needs_troubleshooting	0.84	0.89	0.86	1412	
card_queries_or_issues	0.88	0.86	0.87	1712	
top_up_queries_or_issues	0.89	0.83	0.86	569	
accuracy			0.87	4580	
macro avg	0.87	0.86	0.87	4580	
weighted avg	0.87	0.87	0.87	4580	

The balanced accuracy score is 0.863

