▼ Import libraries

```
import subprocess
print((subprocess.check output("lscpu", shell=True).strip()).decode())
    CPU op-mode(s):
                                    32-bit, 64-bit
    Byte Order:
                                    Little Endian
    Address sizes:
                                    46 bits physical, 48 bits virtual
    On-line CPU(s) list:
                                    0.1
    Thread(s) per core:
    Core(s) per socket:
    Socket(s):
    NUMA node(s):
    Vendor ID:
                                    GenuineIntel
    CPU family:
                                    79
    Model:
    Model name:
                                    Intel(R) Xeon(R) CPU @ 2.20GHz
    Stepping:
    CPU MHz:
                                    2200.150
    BogoMIPS:
                                    4400.30
    Hypervisor vendor:
                                    KVM
    Virtualization type:
    Lld cache:
                                    32 KiB
    Lli cache:
                                    32 KiB
    L2 cache:
                                    256 KiB
    L3 cache:
                                    55 MiB
    NUMA node0 CPU(s):
                                    0,1
    Vulnerability Itlb multihit:
                                 Not affected
    Vulnerability Lltf:
                                    Mitigation; PTE Inversion
    Vulnerability Mds:
                                    Vulnerable; SMT Host state unknown
    Vulnerability Meltdown:
                                    Vulnerable
    Vulnerability Mmio stale data: Vulnerable
    Vulnerability Retbleed:
                                    Vulnerable
    Vulnerability Spec store bypass: Vulnerable
    Vulnerability Spectre v1:
                                   Vulnerable:
                                                 _user pointer sanitization and usercopy barriers only; no swapgs barriers
                                    Vulnerable, TBPB: disabled, STIBP: disabled, PBRSB-eIBRS: Not affected
    Vulnerability Spectre v2:
    Vulnerability Srbds:
                                    Not affected
    Vulnerability Tsx async abort: Vulnerable
    Flags:
                                   fpu vme de pse tsc msr pae mce cx8 apic sep mtrr pge mca cmov pat pse36 clflush mmx fxsr ss
!pip install transformers
### ----- Load libraries ----- ###
# Load Huggingface transformers
from transformers import TFBertModel, BertConfig, BertTokenizerFast
# Then what you need from tensorflow.keras
from tensorflow, keras, layers import Input, Dropout, Dense
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers.legacy import Adam
from tensorflow.keras.callbacks import EarlyStopping
from tensorflow.keras.initializers import TruncatedNormal
from tensorflow.keras.losses import CategoricalCrossentropy
from tensorflow.keras.metrics import CategoricalAccuracy
from tensorflow.keras.utils import to_categorical
# And pandas for data import + sklearn because you allways need sklearn
import pandas as pd
from sklearn.model selection import train test split
```

Loading the Data

```
train_raw = pd.read_csv('consumer_complaints.csv')
train_raw.head()
```

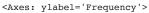
<ipython-input-52-c9bd59dabc04>:1: DtypeWarning: Columns (5,11) have mixed
train_raw = pd.read_csv('consumer_complaints.csv')

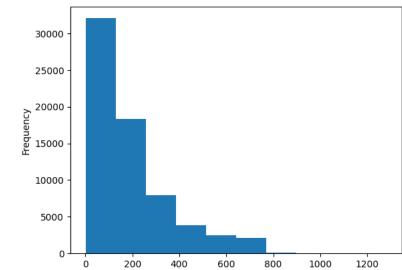
		${\tt date_received}$	product sub_product		issue	sub_issue	
	0	08/30/2013	Mortgage	Other mortgage	Loan modification,collection,foreclosure	NaN	
	1	08/30/2013	Mortgage	Other mortgage	Loan servicing, payments, escrow account	NaN	
	2	08/30/2013	Credit reporting	NaN	Incorrect information on credit report	Account status	
	3	08/30/2013	Student loan	Non-federal student loan	Repaying your loan	Repaying your loar	
train_1		00/00/0010 aw.shape	Debt	Ound's noud	False statements or	Attempted to collect	
		(555957, 18)					

→ Preprocessing Data

```
train_raw = train_raw[train_raw.consumer_complaint_narrative.notnull()]
train_raw.shape
(66806, 18)
```

train_raw.consumer_complaint_narrative.apply(lambda x: len(x.split())).plot(kind='hist')





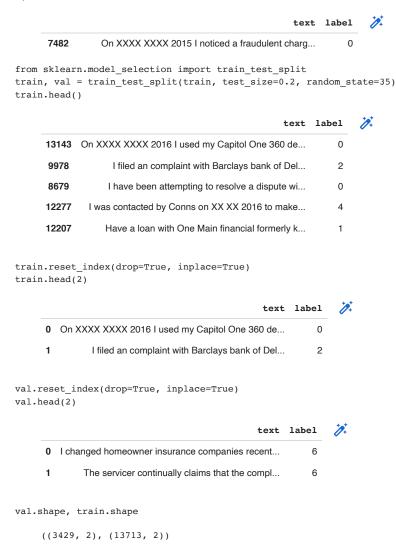
train_raw['len_txt'] =train_raw.consumer_complaint_narrative.apply(lambda x: len(x.split()))
train_raw.describe()

```
len_txt 🎢
             complaint_id
train_raw = train_raw[train_raw.len_txt >249]
train raw.shape
     (17142, 19)
train_raw = train_raw[['consumer_complaint_narrative', 'product']]
train_raw.reset_index(inplace=True, drop=True)
train_raw.head()
                    consumer_complaint_narrative
                                                           product
      0 In XX/XX/XXXX my wages that I earned at my job...
                                                           Mortgage
      1 XXXX was submitted XX/XX/XXXX. At the time I s...
                                                           Mortgage
           I spoke to XXXX of green tree representatives ...
                                                           Mortgage
         i opened XXXX Bank of America credit cards 15-...
                                                          Credit card
      4 I applied for a loan with XXXX XXXX and had pu... Consumer Loan
train raw.at[train raw['product'] == 'Credit reporting', 'product'] = 'Credit reporting, credit repair services, or other persona
train_raw.at[train_raw['product'] == 'Credit card', 'product'] = 'Credit card or prepaid card'
train_raw.at[train_raw['product'] == 'Prepaid card', 'product'] = 'Credit card or prepaid card'
train_raw.at[train_raw['product'] == 'Payday loan', 'product'] = 'Payday loan, title loan, or personal loan'
train_raw.at[train_raw['product'] == 'Virtual currency', 'product'] = 'Money transfer, virtual currency, or money service'
train_raw.head()
import numpy as np
for 1 in np.unique(train_raw['product']):
  print(1)
     Bank account or service
     Consumer Loan
     Credit card
     Credit reporting
     Debt collection
     Money transfers
     Mortgage
     Other financial service
     Payday loan
     Prepaid card
     Student loan
train raw['product'].value counts().sort values(ascending=False).plot(kind='bar')
```

```
<Axes: >
       6000
train_raw-train_raw.rename(columns = {'consumer_complaint_narrative':'text', 'product':'label'})
train_raw.head()
                                                                label
                                                 text
      0 In XX/XX/XXXX my wages that I earned at my job...
                                                             Mortgage
      1 XXXX was submitted XX/XX/XXXX. At the time I s...
                                                             Mortgage
      2
            I spoke to XXXX of green tree representatives ...
                                                             Mortgage
          i opened XXXX Bank of America credit cards 15-...
                                                            Credit card
          I applied for a loan with XXXX XXXX and had pu... Consumer Loan
from sklearn.preprocessing import LabelEncoder
LE = LabelEncoder()
train raw['label'] = LE.fit transform(train raw['label'])
train_raw.head()
                                                                 1
                                                 text label
      0 In XX/XX/XXXX my wages that I earned at my job...
                                                            6
      1 XXXX was submitted XX/XX/XXXX. At the time I s...
                                                            6
            I spoke to XXXX of green tree representatives ...
      2
                                                            6
          i opened XXXX Bank of America credit cards 15-...
      3
                                                            2
          I applied for a loan with XXXX XXXX and had pu...
                                                            1
len(np.unique(train_raw['label']))
     11
train = train_raw.copy()
train = train.reindex(np.random.permutation(train.index))
train.head()
                                                                          1
                                                         text label
       7482
                  On XXXX XXXX, 2015, I noticed a fraudulent cha...
       4894
                  We called Capiital One Bank at XXXX at XX/XX/X...
                                                                     2
      9282
                       A little over two years ago, I had a Macy 's d...
      11935
                  On XXXX/XXXX/2016 I received an email stating ...
             Dear CFPB, My name is XXXX XXXX and my husband...
Clean the text columns
import re
def clean_txt(text):
  text = re.sub("'", "",text)
  text=re.sub("(\\W)+"," ",text)
  return text
```

train['text'] = train.text.apply(clean_txt)

train.head()



▼ BERT: Use pretrained model to classify product and issue with Complaint Narrative

```
data = pd.read_csv('consumer_complaints.csv')
    <ipython-input-83-0fe41c053794>:1: DtypeWarning: Columns (5,11) have mixed types. Specify dtype option on import or set low_
      data = pd.read_csv('consumer_complaints.csv')
data.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 555957 entries, 0 to 555956
    Data columns (total 18 columns):
         Column
                                       Non-Null Count
                                                       Dtype
     0
         date_received
                                       555957 non-null object
         product
                                       555957 non-null object
                                       397635 non-null object
         sub product
         issue
                                       555957 non-null object
         sub_issue
                                       212622 non-null
         consumer complaint narrative 66806 non-null
         company_public_response
                                       85124 non-null
                                                       object
                                       555957 non-null
         company
         state
                                       551070 non-null object
     9
         zipcode
                                       551452 non-null object
     10
                                       77959 non-null
     11 consumer_consent_provided
                                      123458 non-null object
     12
         submitted via
                                       555957 non-null
                                                       object
     13 date_sent_to_company
                                       555957 non-null
     14 company_response_to_consumer 555957 non-null
     15
                                       555957 non-null
         timely response
                                                       object
                                       555957 non-null
         consumer_disputed?
     16
                                                       object
     17 complaint_id
                                       555957 non-null int64
```

```
dtypes: int64(1), object(17)
memory usage: 76.3+ MB
```

data.head()

```
date received product sub product
                                                                 issue sub issue con
                                      Other
                                                                  Loan
     0
             08/30/2013 Mortgage
                                                                             NaN
                                    mortgage
                                            modification, collection, foreclosure
                                      Other
                                                  Loan servicing, payments,
     1
             08/30/2013 Mortgage
                                                                             NaN
                                   mortgage
                                                          escrow account
                         Credit
                                               Incorrect information on credit
                                                                          Account
     2
            08/30/2013
                                       NaN
                       reporting
                                                                            status
                        Student
                                  Non-federal
                                                                         Repaying
     3
             08/30/2013
                                                       Repaying your loan
                                 student loan
                           loan
                                                                         your loan
                                                                         Attempted
                                                       False statements or
                          Debt
                                                                          to collect
             08/30/2013
                                  Credit card
                       collection
                                                           representation
                                                                            wrong
                                                                           amount
data['product'].nunique()
data['issue'].nunique()
### ----- Import data ---- ###
# Import data from csv
# Select required columns
data = data[['consumer_complaint_narrative', 'product', 'issue']]
# Remove a row if any of the three remaining columns are missing
data = data.dropna()
# Remove rows, where the label is present only ones (can't be split)
data = data.groupby('issue').filter(lambda x : len(x) > 1)
data = data.groupby('product').filter(lambda x : len(x) > 1)
# Set your model output as categorical and save in new label col
data['Issue label'] = pd.Categorical(data['issue'])
data['Product_label'] = pd.Categorical(data['product'])
# Transform your output to numeric
data['issue'] = data['Issue_label'].cat.codes
data['product'] = data['Product label'].cat.codes
# Split into train and test - stratify over Issue
data, data_test = train_test_split(data, test_size = 0.2, stratify = data[['issue']])
### ----- Setup RoBERTa ----- ###
# Name of the BERT model to use
model name = 'bert-base-uncased'
# Max length of tokens
max length = 100
# Load transformers config and set output_hidden_states to False
config = BertConfig.from_pretrained(model_name)
config.output hidden states = False
# Load BERT tokenizer
tokenizer = BertTokenizerFast.from pretrained(pretrained_model_name_or_path = model name, config = config)
# Load the Transformers BERT model
transformer_model = TFBertModel.from_pretrained(model_name, config = config)
    Some layers from the model checkpoint at bert-base-uncased were not used when initializing TFBertModel: ['nsp__cls', 'mlm_
```

- This IS expected if you are initializing TFBertModel from the checkpoint of a model trained on another task or with anothe - This IS NOT expected if you are initializing TFBertModel from the checkpoint of a model that you expect to be exactly iden

All the layers of TFBertModel were initialized from the model checkpoint at bert-base-uncased. If your task is similar to the task the model of the checkpoint was trained on, you can already use TFBertModel for predicti

```
#transformer model.layers[0]
```

```
### ----- Build the model ----- ###
# TF Keras documentation: https://www.tensorflow.org/api_docs/python/tf/keras/Model
# Load the MainLayer
bert = transformer model.layers[0]
# Build your model input
input_ids = Input(shape=(max_length,), name='input_ids', dtype='int32')
attention_mask = Input(shape=(max_length,), name='attention_mask', dtype='int32')
inputs = {'input_ids': input_ids, 'attention_mask': attention_mask}
# Load the Transformers BERT model as a layer in a Keras model
bert_model = bert(inputs)[1]
dropout = Dropout(config.hidden_dropout_prob, name='pooled_output')
pooled output = dropout(bert model, training=False)
# Then build your model output
issue = Dense(units=len(data.Issue_label.value_counts()), kernel_initializer=TruncatedNormal(stddev=config.initializer_range), national initializer_range = Dense(units=label.value_counts()), kernel_initializer_range = Dense(units=label.value_counts()),
product = Dense(units=len(data.Product_label.value_counts()), kernel_initializer=TruncatedNormal(stddev=config.initializer_range)
outputs = {'issue': issue, 'product': product}
# And combine it all in a model object
model = Model(inputs=inputs, outputs=outputs, name='BERT_MultiLabel_MultiClass')
# Take a look at the model
model.summary()
```

Model: "BERT_MultiLabel_MultiClass"

Layer (type)	Output Shape	Param #	Connected to
attention_mask (InputLayer)	[(None, 100)]	0	[]
input_ids (InputLayer)	[(None, 100)]	0	[]
bert (TFBertMainLayer)	TFBaseModelOutputWi thPoolingAndCrossAt tentions(last_hidde n_state=(None, 100, 768), pooler_output=(Non e, 768), past_key_values=No ne, hidden_states=N one, attentions=Non e, cross_attentions =None)	109482240	<pre>['attention_mask[0][0]', 'input_ids[0][0]']</pre>
ooled_output (Dropout)	(None, 768)	0	['bert[0][1]']
issue (Dense)	(None, 87)	66903	['pooled_output[0][0]']
product (Dense)	(None, 11)	8459	['pooled_output[0][0]']

Non-trainable params: 0

```
### ----- Train the model ----- ###
# Set an optimizer
optimizer = Adam(
    learning_rate=5e-05,
    epsilon=1e-08,
    decay=0.01,
    clipnorm=1.0)
# Set loss and metrics
loss = {'issue': CategoricalCrossentropy(from_logits = True), 'product': CategoricalCrossentropy(from_logits = True)}
metric = {'issue': CategoricalAccuracy('accuracy'), 'product': CategoricalAccuracy('accuracy')}
# Compile the model
model.compile(
    optimizer = optimizer,
    loss = loss,
   metrics = metric)
# Ready output data for the model
y_issue = to_categorical(data['issue'])
```

```
y_product = to_categorical(data['product'])
# Tokenize the input (takes some time)
x = tokenizer(
   text=data['consumer_complaint_narrative'].to_list(),
   add_special_tokens=True,
   max_length=max_length,
   truncation=True,
   padding=True,
   return tensors='tf',
   return_token_type_ids = False,
   return_attention_mask = True,
   verbose = True)
# Fit the model
history = model.fit(
   x={'input_ids': x['input_ids'] , 'attention_mask': x['attention_mask']},
   y={'issue': y_issue, 'product': y_product},
   validation split=0.2,
   batch_size=4,
   epochs=1)
     3787/10689 [=======>................] - ETA: 10:32:19 - loss: 3.4909 - issue loss: 2.6464 - product loss: 0.8445 - is
### ---- Evaluate the model ----- ###
# Ready test data
# test_y_issue = to_categorical(data_test['issue'])
test_y_product = to_categorical(val['label'])
test_x = tokenizer(
   text=val['text'].to list(),
   add_special_tokens=True,
   max_length=max_length,
   truncation=True,
   padding=True,
   return tensors='tf',
   return_token_type_ids = False,
   return_attention_mask = True,
   verbose = True)
# Run evaluation
model eval = model.evaluate(
   x={'input_ids': test_x['input_ids'],'attention_mask': test_x['attention_mask']},
   y={'product': test_y_product}
#model.predict('hello')
```

1 0s completed at 10:15 AM

×