```
# This Python 3 environment comes with many helpful analytics
libraries installed
# It is defined by the kaggle/python Docker image:
https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
# Input data files are available in the read-only "../input/"
directory
# For example, running this (by clicking run or pressing Shift+Enter)
will list all files under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
# You can write up to 20GB to the current directory (/kaggle/working/)
that gets preserved as output when you create a version using "Save &
Run All"
# You can also write temporary files to /kaggle/temp/, but they won't
be saved outside of the current session
/kaggle/input/int3405-sentiment-analysis-problem/test.csv
/kaggle/input/int3405-sentiment-analysis-problem/full train.csv
 !pip install tensorflow==2.0
Requirement already satisfied: tensorflow==2.0 in
/opt/conda/lib/python3.7/site-packages (2.0.0)
Requirement already satisfied: astor>=0.6.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (0.8.1)
Requirement already satisfied: wrapt>=1.11.1 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.12.1)
Requirement already satisfied: google-pasta>=0.1.6 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (0.2.0)
Requirement already satisfied: numpy<2.0,>=1.16.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.19.5)
Requirement already satisfied: termcolor>=1.1.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.1.0)
Requirement already satisfied: absl-py>=0.7.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (0.15.0)
Requirement already satisfied: wheel>=0.26 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (0.37.1)
Requirement already satisfied: tensorboard<2.1.0,>=2.0.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (2.0.2)
Requirement already satisfied: tensorflow-estimator<2.1.0,>=2.0.0
in /opt/conda/lib/python3.7/site-packages (from tensorflow==2.0)
```

```
(2.0.1)
Requirement already satisfied: grpcio>=1.8.6 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.32.0)
Requirement already satisfied: protobuf>=3.6.1 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (3.19.4)
Requirement already satisfied: keras-applications>=1.0.8 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.0.8)
Requirement already satisfied: six>=1.10.0 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.15.0)
Requirement already satisfied: opt-einsum>=2.3.2 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (3.3.0)
Requirement already satisfied: gast==0.2.2 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (0.2.2)
Requirement already satisfied: keras-preprocessing>=1.0.5 in
/opt/conda/lib/python3.7/site-packages (from tensorflow==2.0) (1.1.2)
Requirement already satisfied: h5py in /opt/conda/lib/python3.7/site-
packages (from keras-applications>=1.0.8->tensorflow==2.0) (2.10.0)
Requirement already satisfied: setuptools>=41.0.0 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0.>=2.0.0->tensorflow==2.0) (59.8.0)
Requirement already satisfied: werkzeug>=0.11.15 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (2.2.2)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (0.4.6)
Requirement already satisfied: google-auth<2,>=1.6.3 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (1.35.0)
Requirement already satisfied: requests<3,>=2.21.0 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (2.28.1)
Requirement already satisfied: markdown>=2.6.8 in
/opt/conda/lib/python3.7/site-packages (from
tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (3.3.7)
Requirement already satisfied: cachetools<5.0,>=2.0.0 in
/opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (4.2.4)
Requirement already satisfied: rsa<5,>=3.1.4 in
/opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (4.8)
Requirement already satisfied: pyasn1-modules>=0.2.1 in
/opt/conda/lib/python3.7/site-packages (from google-auth<2,>=1.6.3-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (0.2.7)
Requirement already satisfied: requests-oauthlib>=0.7.0 in
/opt/conda/lib/python3.7/site-packages (from google-auth-
oauthlib<0.5,>=0.4.1->tensorboard<2.1.0,>=2.0.0->tensorflow==2.0)
(1.3.1)
Requirement already satisfied: importlib-metadata>=4.4 in
/opt/conda/lib/python3.7/site-packages (from markdown>=2.6.8-
```

```
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (4.13.0)
Requirement already satisfied: charset-normalizer<3,>=2 in
/opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0-
>tensorboard<2.1.0.>=2.0.0->tensorflow==2.0) (2.1.0)
Requirement already satisfied: certifi>=2017.4.17 in
/opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0-
>tensorboard<2.1.0.>=2.0.0->tensorflow==2.0) (2022.9.24)
Requirement already satisfied: idna<4,>=2.5 in
/opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (3.3)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in
/opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (1.26.12)
Requirement already satisfied: MarkupSafe>=2.1.1 in
/opt/conda/lib/python3.7/site-packages (from werkzeug>=0.11.15-
>tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (2.1.1)
Requirement already satisfied: typing-extensions>=3.6.4 in
/opt/conda/lib/python3.7/site-packages (from importlib-metadata>=4.4-
>markdown>=2.6.8->tensorboard<2.1.0,>=2.0.0->tensorflow==2.0)
(3.7.4.3)
Requirement already satisfied: zipp>=0.5 in
/opt/conda/lib/python3.7/site-packages (from importlib-metadata>=4.4-
>markdown>=2.6.8->tensorboard<2.1.0,>=2.0.0->tensorflow==2.0) (3.8.0)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in
/opt/conda/lib/python3.7/site-packages (from pyasn1-modules>=0.2.1-
>qoogle-auth<2,>=1.6.3->tensorboard<2.1.0,>=2.0.0->tensorflow==2.0)
(0.4.8)
Requirement already satisfied: oauthlib>=3.0.0 in
/opt/conda/lib/python3.7/site-packages (from requests-oauthlib>=0.7.0-
>google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.1.0,>=2.0.0-
>tensorflow==2.0) (3.2.0)
^C
ERROR: Operation cancelled by user
import os
!pip install pyvi
import numpy as np
with open('/kaggle/input/int3405-sentiment-analysis-problem/test.csv',
encoding="utf8") as f:
    lines = f.readlines()
    lines = "".join(lines)
^(
ERROR: Operation cancelled by user
#Ignoring the warnings
import warnings
warnings.filterwarnings('ignore')
import seaborn as sns
#Importing the required libraries
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import re, string, unicodedata
from tensorflow import keras
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import GlobalMaxPooling1D
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
from tensorflow.keras.models import load model
from tensorflow.keras.layers import *
from tensorflow.keras import backend
from tensorflow.keras import layers
from sklearn.metrics import fl score, confusion matrix
import tensorflow as tf
from pyvi import ViTokenizer
from pyvi import ViUtils
Tập dữ liệu gồm 2 trường:
     Key: "comment": các văn bản với nội dung đánh giá về quán ăn
     Key: "rating": là Label cho đoan văn bản đó có nôi dung tích cực hay tiêu cực.
df1 =
pd.read csv('/kaggle/input/int3405-sentiment-analysis-problem/full tra
in.csv')
df1 = df1.dropna()
df1 = df1.drop(['Unnamed: 0','RevId','UserId','image urls'], axis=1)
X train1 = list(df1['Comment'].values)
y train1 = list(df1['Rating'].values)
print(df1.shape)
df1.head()
(9070, 2)
                                              Comment
                                                        Rating
  Xôi dẻo, đô`ăn đâm vi. Hôp xôi được lót lá trô...
                                                           1.0
  Goi ship 1 xuất cari gà bánh naan và 3 miêng g...
                                                           0.0
  Thời tiết lanh như này, cả nhà rủ nhau đến leg...
                                                           1.0
                                                           0.0
  Em có đọc review thâý mng bảo trà sữa nướng đê\...
4 Đô`ăn rất ngon, nhà hàng cũng rất đẹp, tất cả ...
                                                           1.0
Check DF
# check for length train
X train = X train1 #+ X train2 + X train3
print(len(X train))
y_train = y_train1 #+ y_train2 + y_train3
print(len(y_train))
```

```
9070
9070
sum(y train)
print(sum(y train) / len(y train))
0.7878721058434399
# Check chart output data
sns.countplot(y train)
KeyboardInterrupt
                                           Traceback (most recent call
last)
/opt/conda/lib/python3.7/site-packages/tensorflow/python/keras/api/
_v1/keras/layers/__init__.py in <module>
      1 # Check chart output data
----> 2 sns.countplot(y train)
/opt/conda/lib/python3.7/site-packages/seaborn/ decorators.py in
inner_f(*args, **kwargs)
     44
     45
                kwargs.update({k: arg for k, arg in
zip(sig.parameters, args)})
                return f(**kwargs)
     47
            return inner f
     48
/opt/conda/lib/python3.7/site-packages/seaborn/categorical.py in
countplot(x, y, hue, data, order, hue order, orient, color, palette,
saturation, dodge, ax, **kwargs)
   3600
                estimator, ci, n boot, units, seed,
   3601
                orient, color, palette, saturation,
-> 3602
                errcolor, errwidth, capsize, dodge
   3603
            )
   3604
/opt/conda/lib/python3.7/site-packages/seaborn/categorical.py in
  init__(self, x, y, hue, data, order, hue_order, estimator, ci,
n boot, units, seed, orient, color, palette, saturation, errcolor,
errwidth, capsize, dodge)
                """Initialize the plotter."""
   1583
   1584
                self.establish_variables(x, y, hue, data, orient,
-> 1585
                                          order, hue order, units)
                self.establish colors(color, palette, saturation)
   1586
   1587
                self.estimate statistic(estimator, ci, n boot, seed)
/opt/conda/lib/python3.7/site-packages/seaborn/categorical.py in
establish variables(self, x, y, hue, data, orient, order, hue_order,
units)
```

```
155
                    # Figure out the plotting orientation
    156
                    orient = infer orient(
--> 157
                        x, y, orient,
require numeric=self.require numeric
    158
                    )
    159
/opt/conda/lib/python3.7/site-packages/seaborn/ core.py in
infer orient(x, y, orient, require numeric)
   13\overline{10}
   1311
            x type = None if x is None else variable type(x)
-> 1312
            y type = None if y is None else variable type(y)
   1313
   1314
            nonnumeric dv error = "{} orientation requires numeric
`{}` variable."
/opt/conda/lib/python3.7/site-packages/seaborn/ core.py in
variable type(vector, boolean type)
   1227
   1228
            # Special-case all-na data, which is always "numeric"
-> 1229
            if pd.isna(vector).all():
                return "numeric"
   1230
   1231
/opt/conda/lib/python3.7/site-packages/pandas/core/dtypes/missing.py
in isna(obj)
    136
            Name: 1, dtype: bool
    137
--> 138
            return isna(obi)
    139
    140
/opt/conda/lib/python3.7/site-packages/pandas/core/dtypes/missing.py
in _isna(obj, inf_as_na)
    181
                return obj.isna()
    182
            elif isinstance(obj, list):
--> 183
                return _isna_array(np.asarray(obj, dtype=object),
inf as na=inf as na)
            elif hasattr(obj, "__array__"):
    184
                return isna array(np.asarray(obj),
inf as na=inf as na)
KeyboardInterrupt:
Text Preprocessing
def clean text support(text):
    RE_EMOJI = re.compile('[\U00010000-\U0010ffff]', flags=re.UNICODE)
    text = re.sub(r"<.*?>", " ", text)
    text = re.sub(r"\n", " ", text)
```

```
text = re.sub(r"\s{2,}", " ", text)
text = RE_EMOJI.sub(r'', text)
    return text.strip().lower()
def clean text(X):
    processed = []
    for text in X:
        text = clean text support(text)
        text = ViTokenizer.tokenize(text)
        processed.append(text)
    return processed
text test = X train[100]
print(text test)
print('===')
print(X train[333])
X train final = X train
len(X train final)
print(X train final[:2])
print("===")
print(y train[:2])
# Attention Layer
from tensorflow.keras import initializers, regularizers, constraints
from tensorflow.keras import backend as K
class AttentionWithContext(tf.keras.layers.Layer):
    Attention operation, with a context/query vector, for temporal
data.
    Supports Masking.
    Follows the work of Yang et al.
[https://www.cs.cmu.edu/~diviv/docs/naacl16.pdf]
    "Hierarchical Attention Networks for Document Classification"
    by using a context vector to assist the attention
    # Input shape
        3D tensor with shape: `(samples, steps, features)`.
    # Output shape
        2D tensor with shape: `(samples, features)`.
    How to use:
    Just put it on top of an RNN Layer (GRU/LSTM/SimpleRNN) with
return sequences=True.
    The dimensions are inferred based on the output shape of the RNN.
    Note: The layer has been tested with Keras 2.0.6
    Example:
        model.add(LSTM(64, return sequences=True))
        model.add(AttentionWithContext())
        # next add a Dense layer (for classification/regression) or
whatever...
```

 $\Pi_{i}\Pi_{j}\Pi_{j}\Pi_{j}$ 

```
def init (self, W regularizer=None, u regularizer=None,
b regularizer=None,
                 W constraint=None, u constraint=None,
b constraint=None,
                 bias=True, **kwargs):
        self.supports masking = True
        self.init = initializers.get('glorot uniform')
        self.W regularizer = regularizers.get(W regularizer)
        self.u regularizer = regularizers.get(u regularizer)
        self.b regularizer = regularizers.get(b regularizer)
        self.W constraint = constraints.get(W constraint)
        self.u constraint = constraints.get(u constraint)
        self.b constraint = constraints.get(b constraint)
        self.bias = bias
        super(AttentionWithContext, self).__init__(**kwargs)
    def build(self, input shape):
        assert len(input shape) == 3
        self.W = self.add weight(shape=(input shape[-1], input shape[-
1],),
                                 initializer=self.init,
                                 name='{} W'.format(self.name),
                                 regularizer=self.W regularizer,
                                 constraint=self.W constraint)
        if self.bias:
            self.b = self.add weight(shape=(input shape[-1],),
                                     initializer='zero',
                                     name='{}_b'.format(self.name),
                                     regularizer=self.b regularizer,
                                     constraint=self.b constraint)
        self.u = self.add weight(shape=(input shape[-1],),
                                 initializer=self.init,
                                 name='{} u'.format(self.name),
                                 regularizer=self.u regularizer,
                                 constraint=self.u constraint)
        super(AttentionWithContext, self).build(input shape)
    def compute mask(self, input, input mask=None):
        # do not pass the mask to the next layers
        return None
```

```
def get config(self):
        config = super().get config().copy()
        config.update({
             'W regularizer': self.W regularizer,
             'u regularizer': self.u regularizer,
            'b_regularizer': self.b_regularizer,
             'W constraint': self.W constraint,
            'u_constraint': self.u_constraint,
'b_constraint': self.b_constraint,
             'bias': self.bias.
            })
        return config
    def call(self, x, mask=None):
        uit = dot product(x, self.W)
        if self.bias:
            uit += self.b
        uit = K.tanh(uit)
        ait = dot product(uit, self.u)
        a = K.exp(ait)
        # apply mask after the exp. will be re-normalized next
        if mask is not None:
            # Cast the mask to floatX to avoid float64 upcasting in
theano
            a *= K.cast(mask, K.floatx())
        # in some cases especially in the early stages of training the
sum mav be almost zero
        # and this results in NaN's. A workaround is to add a very
small positive number \varepsilon to the sum.
        # a /= K.cast(K.sum(a, axis=1, keepdims=True), K.floatx())
        a /= K.cast(K.sum(a, axis=1, keepdims=True) + K.epsilon(),
K.floatx())
        a = K.expand dims(a)
        weighted input = x * a
        return K.sum(weighted input, axis=1)
    def compute output shape(self, input shape):
        return input shape[0], input shape[-1]
def dot product(x, kernel):
    Wrapper for dot product operation, in order to be compatible with
both
    Theano and Tensorflow
```

```
Args:
        x (): input
        kernel (): weights
    Returns:
    if K.backend() == 'tensorflow':
        return K.squeeze(K.dot(x, K.expand dims(kernel)), axis=-1)
    else:
        return K.dot(x, kernel)
# some properties
vocab size = 60000
maxlen = 250
encode dim = 20
tokenizer = Tokenizer()
tokenizer.fit on texts(X train final)
tokenized word list = tokenizer.texts to sequences(X train final)
X train padded = pad sequences(tokenized word list, maxlen = maxlen,
padding='post')
#EarlyStopping and ModelCheckpoint
es = EarlyStopping(monitor = 'val loss', mode = 'min', verbose = 1,
patience = 5)
mc = ModelCheckpoint('model best.h5', monitor = 'val loss', mode =
'min', verbose = 1, save best only = True)
Building model train
import tensorflow addons as tfa
# Build model
def create model():
    model = Sequential()
    embed = Embedding(input dim = vocab size, output dim = 20,
input length = X train padded.shape[1])
    model.add(embed)
    model.add(Dropout(0.4))
    model.add(Bidirectional(LSTM(200, return_sequences = True)))
    model.add(Dropout(0.3))
    model.add(AttentionWithContext())
    model.add(Dropout(0.3))
    model.add(Dense(512))
    model.add(LeakyReLU(alpha=0.2))
    model.add(Dense(256))
    model.add(LeakyReLU(alpha=0.2))
    model.add(Dense(1, activation = 'sigmoid'))
    model.summary()
    return model
```

Traning model

```
from sklearn.model selection import train test split
X train padded = np.asarray(X train padded)
y_train = np.asarray(y_train)
X train final2 = X train padded
y train final2 = y train
weight = sum(y_train_final2) / y_train_final2.shape[0]
#class weight
weight_for_0 = (1 / (1-(weight))) * 0.5
weight_for_1 = (1 / (weight)) * 0.5
class weight = {0: weight for 0, 1: weight for 1}
es = EarlyStopping(monitor = 'val loss', mode = 'min', verbose = 1,
patience = 5)
mc = ModelCheckpoint('model best.h5', monitor = 'f1 score', mode =
'min', verbose = 1, save best only = True)
batch size= 300
reduce lr = tf.keras.callbacks.ReduceLROnPlateau(monitor='f1 score',
factor=0.2,
                               patience=3, min lr=1e-5, verbose = 1)
tpu strategy = tf.distribute.MirroredStrategy()
# instantiating the model in the strategy scope creates the model on
the TPU
with tpu strategy.scope():
    model = create model() # define your model normally
    optim = tf.keras.optimizers.Adam(learning rate=5e-4)
    model.compile(optimizer='adam',
                      loss='binary crossentropy',
                     metrics=['accuracy',
                               tf.keras.metrics.Recall(),
                               tfa.metrics.F1Score(num_classes=1,
                                                   average='micro')
                              1
             )
len(X train final2)
train_x, train_y = np.array(X_train_final2), np.array(y_train_final2)
train data = tf.data.Dataset.from tensor slices((train x, train y))
batch size = 512 * tpu strategy.num replicas in sync
train data = train data.batch(batch size)
# val data = val data.batch(batch size)
# Disable AutoShard.
options = tf.data.Options()
options.experimental distribute.auto shard policy =
tf.data.experimental.AutoShardPolicy.OFF
```

```
train data = train data.with options(options)
# val data = val data.with options(options)
#Fit model
model.fit(train data ,
          epochs = 15, batch size = batch size, verbose = 1,
          callbacks = [reduce lr, es,mc],class weight=class weight)
df =
pd.read csv('/kaggle/input/int3405-sentiment-analysis-problem/test.csv
data test = pd.DataFrame({'input':df['Comment'],'id':df["RevId"]})
X test = list(data test['input'].values)
def clean text test(X):
    idx = 0
    y train = []
    processed = []
    for text in X:
        text = str(text)
        text = clean text support(text)
        input text pre accent = ViTokenizer.tokenize(text)
        processed.append(input_text_pre_accent)
    return processed
# X test = list()
X_test_final = clean_text_test(X_test)
my submission['Rating'].sum() / len(my_submission['Rating'])
my submission = pd.DataFrame({'RevId':
np.array(df["RevId"]).reshape(5103), 'Rating':
np.array(y pred).reshape(5103)})
my_submission.to_csv('submission.csv', index=False)
```