

## ➤ Medical Report Generation Case Study : Optimized Solution

```
!pip show tensorflow
!pip install plot_model
!pip install tensorboardcolab
%load_ext tensorboard
!rm -rf ./logs/
import warnings
warnings.filterwarnings("ignore")
import pandas as pd
import numpy as np
import re
import os
from nltk.tokenize import word_tokenize
import xml.etree.ElementTree as ET
from os import listdir
from os import path
import tensorflow as tf
from tensorflow.keras.preprocessing.image import load_img
from tensorflow.keras.preprocessing.image import img_to_array
from tensorflow.keras.applications.vgg16 import preprocess_input
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import LSTM
from tensorflow.keras.layers import Embedding
from tensorflow.keras.layers import Dropout
from tensorflow.keras.layers import add
from tensorflow.keras.callbacks import ModelCheckpoint
import pickle
from tqdm import tqdm
import random
from numpy import argmax
from tensorflow.keras.models import load_model
from nltk.translate.bleu_score import corpus_bleu
from tensorboardcolab import *
from tensorflow.keras.callbacks import TensorBoard
from datetime import datetime, timedelta
from tensorflow.keras.utils import plot_model
from tensorflow.keras.applications.xception import Xception
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from matplotlib import pyplot
from numpy import array
from prettytable import PrettyTable
```



Name: tensorflow  
 Version: 2.2.0rc4  
 Summary: TensorFlow is an open source machine learning framework for everyone.  
 Home-page: <https://www.tensorflow.org/>  
 Author: Google Inc.  
 Author-email: [packages@tensorflow.org](mailto:packages@tensorflow.org)  
 License: Apache 2.0  
 Location: /usr/local/lib/python3.6/dist-packages  
 Requires: scipy, tensorflow-estimator, absl-py, tensorboard, grpcio, keras-preprocessing, six, google-pasta,  
 Required-by: fancyimpute  
 Collecting plot\_model  
 Downloading <https://files.pythonhosted.org/packages/62/b8/0967e30391a7c07002c5e7bca868763dcfd2/>  
 Installing collected packages: plot-model  
 Successfully installed plot-model-0.20  
 Requirement already satisfied: tensorboardcolab in /usr/local/lib/python3.6/dist-packages (0.0.22)  
 Using TensorFlow backend.

```

from google.colab import drive
drive.mount('/content/drive')

```



Go to this URL in a browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=947318989803-6bn6qk](https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6qk)

Enter your authorization code:

.....

Mounted at /content/drive

## Here we Load all the images, preprocess it and extract feature vector using pretrained Xception n

```

# create image data augmentation generator
datagen = ImageDataGenerator(
    rotation_range = 15, # randomly rotate images in the range (degrees, 0 to 180)
    zoom_range = 0.2, # Randomly zoom image
    width_shift_range=0.1, # randomly shift images horizontally (fraction of total width)
    height_shift_range=0.1, # randomly shift images vertically (fraction of total height)
    horizontal_flip = True )

```

```

#load Xception model
model = Xception(weights='imagenet')

```



Downloading data from [https://storage.googleapis.com/tensorflow/keras-applications/xception/xception\\_91889664/91884032](https://storage.googleapis.com/tensorflow/keras-applications/xception/xception_91889664/91884032) [=====] - 1s 0us/step

```

#to get extracted image features
def extract_features(directory,model):
    # re-structure the model
    model = Model(inputs=model.inputs, outputs=model.layers[-2].output)
    model.summary()
    # extract features from each photo
    features = dict()
    for name in tqdm(listdir(directory)):
        # get image id
        image_id = name.split('.')[0]

```

```
image_id = name.split('.')[0]
# load an image from file
filename = path.join(directory, name)
image = load_img(filename, target_size=(299, 299))
# convert the image pixels to a numpy array
image = img_to_array(image)
# reshape data for the model
image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
image1 = preprocess_input(image)
feature = model.predict(image1, verbose=0)
features[image_id] = feature
it = datagen.flow(image, batch_size=1)
# generate samples
for i in range(1,3):
    # generate batch of images
    batch = it.next()
    # prepare the image for the xception model
    image = preprocess_input(batch)
    # get features
    feature = model.predict(image, verbose=0)
    # store feature
    id=image_id+str(i)
    features[id] = feature
return features

# extract features from all images
directory = '/content/drive/My Drive/images'
image_extracted_features = extract_features(directory,model)
```



0%| | 0/7470 [00:00<?, ?it/s]Model: "model"

Layer (type)	Output Shape	Param #	Connected to
=====:			
input_1 (InputLayer)	[(None, 299, 299, 3)]	0	
block1_conv1 (Conv2D)	(None, 149, 149, 32)	864	input_1[0][0]
block1_conv1_bn (BatchNormaliza	(None, 149, 149, 32)	128	block1_conv1[0][0]
block1_conv1_act (Activation)	(None, 149, 149, 32)	0	block1_conv1_bn[0][0]
block1_conv2 (Conv2D)	(None, 147, 147, 64)	18432	block1_conv1_act[0][0]
block1_conv2_bn (BatchNormaliza	(None, 147, 147, 64)	256	block1_conv2[0][0]
block1_conv2_act (Activation)	(None, 147, 147, 64)	0	block1_conv2_bn[0][0]
block2_sepconv1 (SeparableConv2	(None, 147, 147, 128)	8768	block1_conv2_act[0][0]
block2_sepconv1_bn (BatchNormal	(None, 147, 147, 128)	512	block2_sepconv1[0][0]
block2_sepconv2_act (Activation	(None, 147, 147, 128)	0	block2_sepconv1_bn[0][0]
block2_sepconv2 (SeparableConv2	(None, 147, 147, 128)	17536	block2_sepconv2_act[0][0]
block2_sepconv2_bn (BatchNormal	(None, 147, 147, 128)	512	block2_sepconv2[0][0]
conv2d (Conv2D)	(None, 74, 74, 128)	8192	block1_conv2_act[0][0]
block2_pool (MaxPooling2D)	(None, 74, 74, 128)	0	block2_sepconv2_bn[0][0]
batch_normalization (BatchNorma	(None, 74, 74, 128)	512	conv2d[0][0]
add (Add)	(None, 74, 74, 128)	0	block2_pool[0][0] batch_normalization[0][0]
block3_sepconv1_act (Activation	(None, 74, 74, 128)	0	add[0][0]
block3_sepconv1 (SeparableConv2	(None, 74, 74, 256)	33920	block3_sepconv1_act[0][0]
block3_sepconv1_bn (BatchNormal	(None, 74, 74, 256)	1024	block3_sepconv1[0][0]
block3_sepconv2_act (Activation	(None, 74, 74, 256)	0	block3_sepconv1_bn[0][0]
block3_sepconv2 (SeparableConv2	(None, 74, 74, 256)	67840	block3_sepconv2_act[0][0]
block3_sepconv2_bn (BatchNormal	(None, 74, 74, 256)	1024	block3_sepconv2[0][0]
conv2d_1 (Conv2D)	(None, 37, 37, 256)	32768	add[0][0]
block3_pool (MaxPooling2D)	(None, 37, 37, 256)	0	block3_sepconv2_bn[0][0]
batch_normalization_1 (BatchNor	(None, 37, 37, 256)	1024	conv2d_1[0][0]
add_1 (Add)	(None, 37, 37, 256)	0	block3_pool[0][0] batch_normalization_1[0][0]

block4_sepconv1_act (Activation (None, 37, 37, 256) 0	add_1[0][0]
block4_sepconv1 (SeparableConv2 (None, 37, 37, 728) 188672	block4_sepconv1_act[0][0]
block4_sepconv1_bn (BatchNormal (None, 37, 37, 728) 2912	block4_sepconv1[0][0]
block4_sepconv2_act (Activation (None, 37, 37, 728) 0	block4_sepconv1_bn[0][0]
block4_sepconv2 (SeparableConv2 (None, 37, 37, 728) 536536	block4_sepconv2_act[0][0]
block4_sepconv2_bn (BatchNormal (None, 37, 37, 728) 2912	block4_sepconv2[0][0]
conv2d_2 (Conv2D) (None, 19, 19, 728) 186368	add_1[0][0]
block4_pool (MaxPooling2D) (None, 19, 19, 728) 0	block4_sepconv2_bn[0][0]
batch_normalization_2 (BatchNor (None, 19, 19, 728) 2912	conv2d_2[0][0]
add_2 (Add) (None, 19, 19, 728) 0	block4_pool[0][0] batch_normalization_2[0][0]
block5_sepconv1_act (Activation (None, 19, 19, 728) 0	add_2[0][0]
block5_sepconv1 (SeparableConv2 (None, 19, 19, 728) 536536	block5_sepconv1_act[0][0]
block5_sepconv1_bn (BatchNormal (None, 19, 19, 728) 2912	block5_sepconv1[0][0]
block5_sepconv2_act (Activation (None, 19, 19, 728) 0	block5_sepconv1_bn[0][0]
block5_sepconv2 (SeparableConv2 (None, 19, 19, 728) 536536	block5_sepconv2_act[0][0]
block5_sepconv2_bn (BatchNormal (None, 19, 19, 728) 2912	block5_sepconv2[0][0]
block5_sepconv3_act (Activation (None, 19, 19, 728) 0	block5_sepconv2_bn[0][0]
block5_sepconv3 (SeparableConv2 (None, 19, 19, 728) 536536	block5_sepconv3_act[0][0]
block5_sepconv3_bn (BatchNormal (None, 19, 19, 728) 2912	block5_sepconv3[0][0]
add_3 (Add) (None, 19, 19, 728) 0	block5_sepconv3_bn[0][0] add_2[0][0]
block6_sepconv1_act (Activation (None, 19, 19, 728) 0	add_3[0][0]
block6_sepconv1 (SeparableConv2 (None, 19, 19, 728) 536536	block6_sepconv1_act[0][0]
block6_sepconv1_bn (BatchNormal (None, 19, 19, 728) 2912	block6_sepconv1[0][0]
block6_sepconv2_act (Activation (None, 19, 19, 728) 0	block6_sepconv1_bn[0][0]
block6_sepconv2 (SeparableConv2 (None, 19, 19, 728) 536536	block6_sepconv2_act[0][0]
block6_sepconv2_bn (BatchNormal (None, 19, 19, 728) 2912	block6_sepconv2[0][0]
block6_sepconv3_act (Activation (None, 19, 19, 728) 0	block6_sepconv2_bn[0][0]
block6_sepconv3 (SeparableConv2 (None, 19, 19, 728) 536536	block6_sepconv3_act[0][0]

block6_sepconv3_bn (BatchNormal (None, 19, 19, 728) 2912	block6_sepconv3[0][0]
add_4 (Add) (None, 19, 19, 728) 0	block6_sepconv3_bn[0][0] add_3[0][0]
block7_sepconv1_act (Activation (None, 19, 19, 728) 0	add_4[0][0]
block7_sepconv1 (SeparableConv2 (None, 19, 19, 728) 536536	block7_sepconv1_act[0][0]
block7_sepconv1_bn (BatchNormal (None, 19, 19, 728) 2912	block7_sepconv1[0][0]
block7_sepconv2_act (Activation (None, 19, 19, 728) 0	block7_sepconv1_bn[0][0]
block7_sepconv2 (SeparableConv2 (None, 19, 19, 728) 536536	block7_sepconv2_act[0][0]
block7_sepconv2_bn (BatchNormal (None, 19, 19, 728) 2912	block7_sepconv2[0][0]
block7_sepconv3_act (Activation (None, 19, 19, 728) 0	block7_sepconv2_bn[0][0]
block7_sepconv3 (SeparableConv2 (None, 19, 19, 728) 536536	block7_sepconv3_act[0][0]
block7_sepconv3_bn (BatchNormal (None, 19, 19, 728) 2912	block7_sepconv3[0][0]
add_5 (Add) (None, 19, 19, 728) 0	block7_sepconv3_bn[0][0] add_4[0][0]
block8_sepconv1_act (Activation (None, 19, 19, 728) 0	add_5[0][0]
block8_sepconv1 (SeparableConv2 (None, 19, 19, 728) 536536	block8_sepconv1_act[0][0]
block8_sepconv1_bn (BatchNormal (None, 19, 19, 728) 2912	block8_sepconv1[0][0]
block8_sepconv2_act (Activation (None, 19, 19, 728) 0	block8_sepconv1_bn[0][0]
block8_sepconv2 (SeparableConv2 (None, 19, 19, 728) 536536	block8_sepconv2_act[0][0]
block8_sepconv2_bn (BatchNormal (None, 19, 19, 728) 2912	block8_sepconv2[0][0]
block8_sepconv3_act (Activation (None, 19, 19, 728) 0	block8_sepconv2_bn[0][0]
block8_sepconv3 (SeparableConv2 (None, 19, 19, 728) 536536	block8_sepconv3_act[0][0]
block8_sepconv3_bn (BatchNormal (None, 19, 19, 728) 2912	block8_sepconv3[0][0]
add_6 (Add) (None, 19, 19, 728) 0	block8_sepconv3_bn[0][0] add_5[0][0]
block9_sepconv1_act (Activation (None, 19, 19, 728) 0	add_6[0][0]
block9_sepconv1 (SeparableConv2 (None, 19, 19, 728) 536536	block9_sepconv1_act[0][0]
block9_sepconv1_bn (BatchNormal (None, 19, 19, 728) 2912	block9_sepconv1[0][0]
block9_sepconv2_act (Activation (None, 19, 19, 728) 0	block9_sepconv1_bn[0][0]
block9_sepconv2 (SeparableConv2 (None, 19, 19, 728) 536536	block9_sepconv2_act[0][0]
block9_sepconv2_bn (BatchNormal (None, 19, 19, 728) 2912	block9_sepconv2[0][0]

block9_sepconv3_act (Activation (None, 19, 19, 728) 0	block9_sepconv2_bn[0][0]
block9_sepconv3 (SeparableConv2 (None, 19, 19, 728) 536536	block9_sepconv3_act[0][0]
block9_sepconv3_bn (BatchNormal (None, 19, 19, 728) 2912	block9_sepconv3[0][0]
add_7 (Add) (None, 19, 19, 728) 0	block9_sepconv3_bn[0][0] add_6[0][0]
block10_sepconv1_act (Activatio (None, 19, 19, 728) 0	add_7[0][0]
block10_sepconv1 (SeparableConv (None, 19, 19, 728) 536536	block10_sepconv1_act[0][0]
block10_sepconv1_bn (BatchNorma (None, 19, 19, 728) 2912	block10_sepconv1[0][0]
block10_sepconv2_act (Activatio (None, 19, 19, 728) 0	block10_sepconv1_bn[0][0]
block10_sepconv2 (SeparableConv (None, 19, 19, 728) 536536	block10_sepconv2_act[0][0]
block10_sepconv2_bn (BatchNorma (None, 19, 19, 728) 2912	block10_sepconv2[0][0]
block10_sepconv3_act (Activatio (None, 19, 19, 728) 0	block10_sepconv2_bn[0][0]
block10_sepconv3 (SeparableConv (None, 19, 19, 728) 536536	block10_sepconv3_act[0][0]
block10_sepconv3_bn (BatchNorma (None, 19, 19, 728) 2912	block10_sepconv3[0][0]
add_8 (Add) (None, 19, 19, 728) 0	block10_sepconv3_bn[0][0] add_7[0][0]
block11_sepconv1_act (Activatio (None, 19, 19, 728) 0	add_8[0][0]
block11_sepconv1 (SeparableConv (None, 19, 19, 728) 536536	block11_sepconv1_act[0][0]
block11_sepconv1_bn (BatchNorma (None, 19, 19, 728) 2912	block11_sepconv1[0][0]
block11_sepconv2_act (Activatio (None, 19, 19, 728) 0	block11_sepconv1_bn[0][0]
block11_sepconv2 (SeparableConv (None, 19, 19, 728) 536536	block11_sepconv2_act[0][0]
block11_sepconv2_bn (BatchNorma (None, 19, 19, 728) 2912	block11_sepconv2[0][0]
block11_sepconv3_act (Activatio (None, 19, 19, 728) 0	block11_sepconv2_bn[0][0]
block11_sepconv3 (SeparableConv (None, 19, 19, 728) 536536	block11_sepconv3_act[0][0]
block11_sepconv3_bn (BatchNorma (None, 19, 19, 728) 2912	block11_sepconv3[0][0]
add_9 (Add) (None, 19, 19, 728) 0	block11_sepconv3_bn[0][0] add_8[0][0]
block12_sepconv1_act (Activatio (None, 19, 19, 728) 0	add_9[0][0]
block12_sepconv1 (SeparableConv (None, 19, 19, 728) 536536	block12_sepconv1_act[0][0]
block12_sepconv1_bn (BatchNorma (None, 19, 19, 728) 2912	block12_sepconv1[0][0]

```

block12_sepconv2_act (Activatio (None, 19, 19, 728) 0      block12_sepconv1_bn[0][0]
-----
block12_sepconv2 (SeparableConv (None, 19, 19, 728) 536536   block12_sepconv2_act[0][0]
-----
block12_sepconv2_bn (BatchNorma (None, 19, 19, 728) 2912    block12_sepconv2[0][0]
-----
block12_sepconv3_act (Activatio (None, 19, 19, 728) 0      block12_sepconv2_bn[0][0]
-----
block12_sepconv3 (SeparableConv (None, 19, 19, 728) 536536   block12_sepconv3_act[0][0]
-----
block12_sepconv3_bn (BatchNorma (None, 19, 19, 728) 2912    block12_sepconv3[0][0]
-----
add_10 (Add)          (None, 19, 19, 728) 0      block12_sepconv3_bn[0][0]
                        add_9[0][0]
-----
block13_sepconv1_act (Activatio (None, 19, 19, 728) 0      add_10[0][0]
-----
block13_sepconv1 (SeparableConv (None, 19, 19, 728) 536536   block13_sepconv1_act[0][0]
-----
block13_sepconv1_bn (BatchNorma (None, 19, 19, 728) 2912    block13_sepconv1[0][0]
-----
block13_sepconv2_act (Activatio (None, 19, 19, 728) 0      block13_sepconv1_bn[0][0]
-----
block13_sepconv2 (SeparableConv (None, 19, 19, 1024) 752024   block13_sepconv2_act[0][0]
-----
block13_sepconv2_bn (BatchNorma (None, 19, 19, 1024) 4096    block13_sepconv2[0][0]
-----
conv2d_3 (Conv2D)      (None, 10, 10, 1024) 745472   add_10[0][0]
-----
block13_pool (MaxPooling2D) (None, 10, 10, 1024) 0      block13_sepconv2_bn[0][0]
-----
batch_normalization_3 (BatchNor (None, 10, 10, 1024) 4096    conv2d_3[0][0]
-----
add_11 (Add)          (None, 10, 10, 1024) 0      block13_pool[0][0]
                        batch_normalization_3[0][0]
-----
block14_sepconv1 (SeparableConv (None, 10, 10, 1536) 1582080   add_11[0][0]
-----
block14_sepconv1_bn (BatchNorma (None, 10, 10, 1536) 6144    block14_sepconv1[0][0]
-----
block14_sepconv1_act (Activatio (None, 10, 10, 1536) 0      block14_sepconv1_bn[0][0]
-----
block14_sepconv2 (SeparableConv (None, 10, 10, 2048) 3159552   block14_sepconv1_act[0][0]
-----
block14_sepconv2_bn (BatchNorma (None, 10, 10, 2048) 8192    block14_sepconv2[0][0]
-----
block14_sepconv2_act (Activatio (None, 10, 10, 2048) 0      block14_sepconv2_bn[0][0]
-----
avg_pool (GlobalAveragePooling2 (None, 2048) 0      block14_sepconv2_act[0][0]
=====
Total params: 20,861,480
Trainable params: 20,806,952
Non-trainable params: 54,528
=====
100%|████████████████████| 7470/7470 [1:47:55<00:00, 1.15it/s]

```

```
print('Number of total images: ' + len(image_extracted_features))
```



```
print('Number of total images: ', len(image_extracted_features),)
print('Extracted Feature vector size: ', len(image_extracted_features['CXR1_1_IM-0001-3001'][0]))
```



Number of total images: 22410  
Extracted Feature vector size: 2048

## ▼ 2) Load text data

```
# this code is for extracting data from xml files
# xml's sample code learned from https://www.youtube.com/watch?v=PNNg4xKbCtA
#https://docs.python.org/3.4/library/xml.etree.elementtree.html
id_impression=dict()
id_finding=dict()
directory = 'reports'
```

```
for filename in tqdm(listdir(directory)):
    if filename.endswith(".xml"):
        f=path.join(directory,filename)
        tree = ET.parse(f)
        root = tree.getroot()
        for child in root:
            if child.tag=='MedlineCitation':
                for attr in child:
                    if attr.tag=='Article':
                        for i in attr:
                            if i.tag=='Abstract':
                                for name in i:
                                    if name.get('Label')=='FINDINGS':
                                        finding=name.text
                                    elif name.get('Label')=='IMPRESSION':
                                        impression=name.text
```

```
for p_image in root.findall('parentImage'):
    idd = p_image.get('id')
    id_impression[idd]=impression
    id_finding[idd]=finding
    for i in range(1,3):
        id=idd+str(i)
        id_impression[id]=impression
        id_finding[id]=finding
```



100%

## ▼ Data Cleaning

### • 2.1 Check for None values

```
count1=0
#finding none values in impression
for k,v in id_impression.items():
    if id_impression[k] is None:
        count1=count1+1
```

```
count=count+1
print("Impression data contains",count1,"None Values")
```

 Impression data contains 156 None Values

```
count=0
#finding none values in finding
for k,v in id_finding.items():
    if id_finding[k] is None:
        count=count+1
print("Finding data contains",count,"None Values")
```

 Finding data contains 2991 None Values

```
count=0
#finding none values in findings and impressions
for k,v in id_finding.items():
    if (id_finding[k] is None) and (id_impression[k] is None) :
        count=count+1
print("There are",count,"datapoints whose Finding and impressions data are None")
```

 There are 120 datapoints whose Finding and impressions data are None

```
count=0
#finding none values in finding or impression
for k,v in id_finding.items():
    if (id_finding[k] is None) or (id_impression[k] is None) :
        count=count+1
print("There are",count,"datapoints whose Finding or impressions data are None")
```

 There are 3027 datapoints whose Finding or impressions data are None

## ▼ 2.2 Clean missing data

- We delete entries whose impressions are None

```
# removing none impressions
id_impression_none_removed = { k : v for k,v in id_impression.items() if v is not None}
print("After removing None contained impressions,number of final impressions: "+str(len(id_impression)) + " - "+" +"
```

 After removing None contained impressions,number of final impressions: 22410 - 156 = 22254

- We delete entries whose impressions or findings are None

```
#finding none values in finding
count=0
finding_for_2nd_model=dict()
#impression_for_2nd_model=dict()
for k,v in id_finding.items():
    if (id_finding[k] is None) or (id_impression[k] is None) :
        count=count+1
        continue
```

```
else:
```

```
    finding_for_2nd_model[k]=v
```

```
    #impression_for_2nd_model[k]=id_impression[k]
```

```
print("There are",count,"datapoints whose Finding or impressions data are None , so we have removed them.")
print("Number of final Finding datapoints: "+str(len(id_finding))+ " - ",count,"=",len(finding_for_2nd_model))
```



There are 3027 datapoints whose Finding or impressions data are None , so we have removed them.  
Number of final Finding datapoints: 22410-3027= 19383

## ▼ 2.3 Text Preprocessing

```
def clean_descriptions(descriptions,add_token):
```

```
    """this function cleans decription text"""
```

```
    descriptionss=dict()
```

```
    for key, desc in descriptions.items():
```

```
        sent = desc.replace('x-XXXX',' ')
```

```
        sent = sent.lower()
```

```
        sent = sent.replace('xxx',' ')
```

```
        sent = sent.replace('x-xxx',' ')
```

```
        sent = re.sub('[^A-Za-z]+',' ', sent)
```

```
        if add_token=='yes':
```

```
            sent = 'startseq ' +sent+ ' endseq'
```

```
        descriptionss[key]=sent.strip()
```

```
    return descriptionss
```

```
cleaned_id_impressions = clean_descriptions(id_impression_none_removed,'yes')
```

```
cleaned_id_findings = clean_descriptions(finding_for_2nd_model,'no')
```

## ▼ Train ,cv and test split

```
keys=list(cleaned_id_findings.keys())
```

```
random.shuffle(keys)
```

```
train_id,cv_id,test_id=keys[0:18883],keys[18883:19133],keys[19133:]
```

```
train_id = dict.fromkeys(train_id,1)
```

```
cv_id = dict.fromkeys(cv_id,1)
```

```
test_id = dict.fromkeys(test_id,1)
```

```
print("train size :\t",len(train_id))
```

```
print("cv size   :\t",len(cv_id))
```

```
print("test size :\t",len(test_id))
```



train size : 18883

cv size : 250

test size : 250

## ▼ Define Necessary Functions

```
def create_sequences(tokenizer, max_length, descriptions, image_features, finding_features, vocab_size):
    X1, X2, X3, y = list(), list(), list(), list()
    # walk through each image identifier
    for key, desc in descriptions.items():
        # encode the sequence
        seq = tokenizer.texts_to_sequences([desc])[0]
        # split one sequence into multiple X,y pairs
        for i in range(1, len(seq)):
            # split into input and output pair
            in_seq, out_seq = seq[:i], seq[i]
            # pad input sequence
            in_seq = pad_sequences([in_seq], maxlen=max_length)[0]
            # encode output sequence
            out_seq = to_categorical([out_seq], num_classes=vocab_size)[0]
            # store
            X1.append(image_features[key][0])
            #print(image_features[key][0])
            X2.append(finding_features[key])
            #print(finding_features[key])
            X3.append(in_seq)
            y.append(out_seq)
    return array(X1), array(X2), array(X3), array(y)
```

```
def finding_sequences(tokenizer, max_length, id_findings):
    finding_sequences = dict()
    for k, v in id_findings.items():
        seq = tokenizer.texts_to_sequences([v])[0]
        seq = pad_sequences([seq], maxlen=max_length)[0]
        finding_sequences[k] = seq
    return finding_sequences
```

```
# load clean respective set into memory
def load_respective_set(dictt, dataset):
    """ to load description of given dataset """
    descriptions = dict()
    for k, v in dataset.items():
        descriptions[k] = dictt[k]
    return descriptions
```

```
def load_image_features(dictt, dataset):
    """ to load image features of given dataset """
    features = {k: dictt[k] for k in dataset}
    return features
```

```
def create_tokenizer(descriptions):
    """fit a tokenizer for given descriptions """
    lines = list(descriptions.values())
    tokenizer = Tokenizer()
    tokenizer.fit_on_texts(lines)
    return tokenizer
```

## Prepare Train Data

```

train_image_features = load_image_features(image_extracted_features,train_id)
print('\nTotal train images    : ',len(train_image_features))

#-----
train_id_findings=load_respective_set(cleaned_id_findings, train_id)
print('\nTotal train findings   : ',len(train_id_findings))

findings_max_length = max(len(s.split()) for s in list(train_id_findings.values()))
print('\nMaximum Length of Findings: ',findings_max_length)

findings_tokenizer = create_tokenizer(train_id_findings)
train_id_finding_sequences = finding_sequences(findings_tokenizer,findings_max_length,train_id_findings)

findings_vocab_size = len(findings_tokenizer.word_index) + 1
print('\nVocab size of Findings: ',findings_vocab_size)

glove_words = pickle.load(open('/content/drive/My Drive/Xception/glove_vectors', 'rb'))
findings_embedding_matrix = np.zeros((findings_vocab_size, 300))
for word, i in findings_tokenizer.word_index.items():
    embedding_vector = glove_words.get(word)
    if embedding_vector is not None:
        findings_embedding_matrix[i] = embedding_vector

print("\n-----\n")

train_id_impressions=load_respective_set(cleaned_id_impressions, train_id)
print('\nTotal train impressions : ',len(train_id_impressions))

impressions_max_length = max(len(s.split()) for s in list(train_id_impressions.values()))
print('\nDescription maximum Length: ',impressions_max_length)

# prepare tokenizer
impressions_tokenizer = create_tokenizer(train_id_impressions)

impressions_vocab_size = len(impressions_tokenizer.word_index) + 1
print('\n Impressions Vocabulary Size: ', impressions_vocab_size)

impressions_embedding_matrix = np.zeros((impressions_vocab_size, 300))
for word, i in impressions_tokenizer.word_index.items():
    embedding_vector = glove_words.get(word)
    if embedding_vector is not None:
        impressions_embedding_matrix[i] = embedding_vector

# pad to fixed length
X1train, X2train, X3train, ytrain = create_sequences(impressions_tokenizer,impressions_max_length ,train_id_impr

```



Total train images : 18883

Total train findings : 18883

Maximum Length of Findings: 166

Vocab size of Findings: 1563

-----

Total train impressions : 18883

Description maximum Length: 114

Impressions Vocabulary Size: 1208

## Prepare CV Data

```
cv_image_features = load_image_features(image_extracted_features,cv_id)
print("\nTotal cv images : ',len(cv_image_features))
```

```
cv_id_findings=load_respective_set(cleaned_id_findings, cv_id)
print("\nTotal cv findings : ',len(cv_id_findings))
```

```
cv_id_finding_sequences = finding_sequences(findings_tokenizer,findings_max_length,cv_id_findings)
```

```
print("\n-----\n")
```

```
cv_id_impressions=load_respective_set(cleaned_id_impressions, cv_id)
print('Total cv impressions : ',len(cv_id_impressions))
```

```
X1cv, X2cv, X3cv,ycv = create_sequences(impressions_tokenizer, impressions_max_length,cv_id_impressions,cv_id_impressions)
```



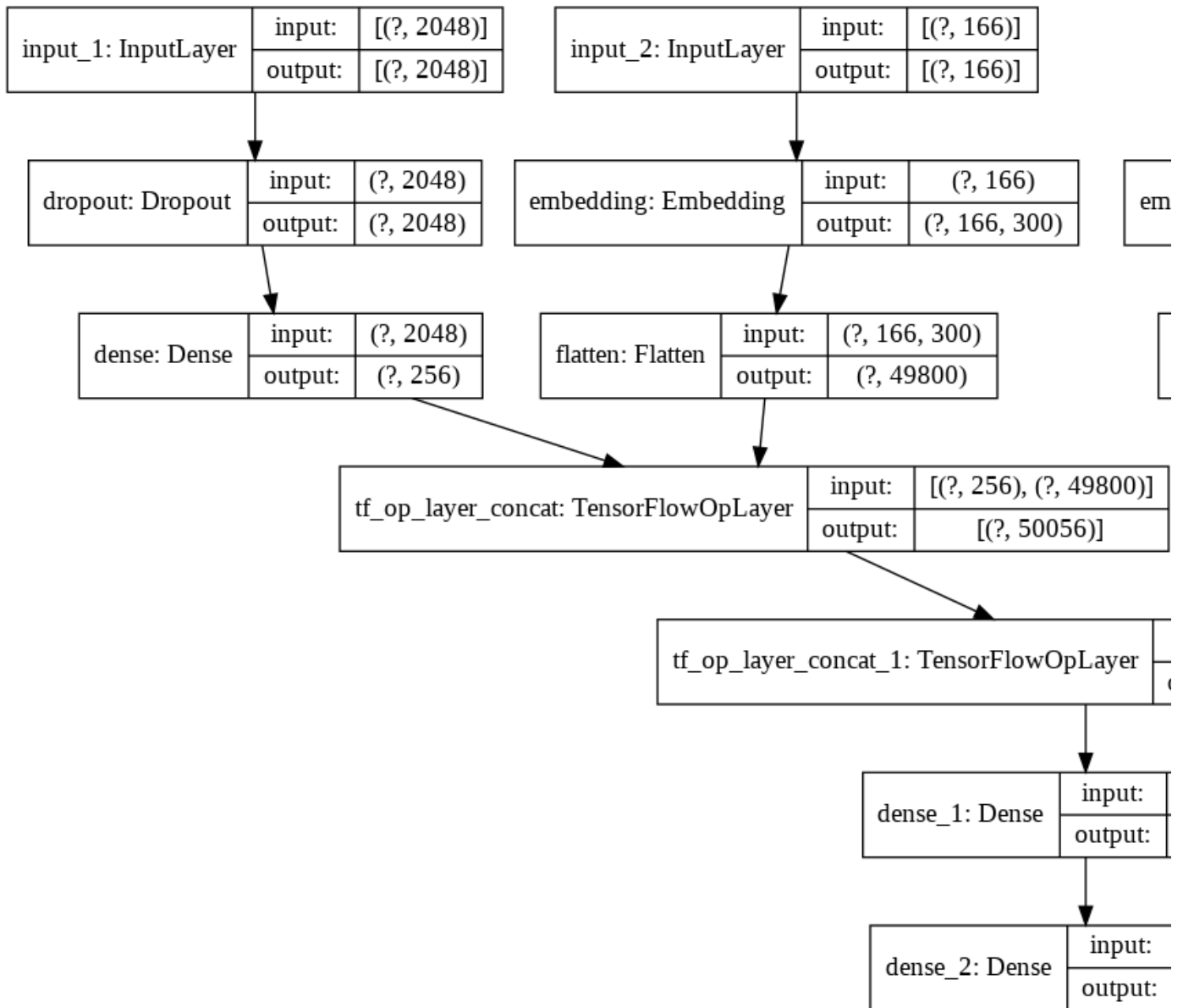
Total cv images : 250

Total cv findings : 250

-----

Total cv impressions : 250

## ➤ Build Deep Learning Model : ( [ image, findings ],impression )



## ▼ Define model

# define the captioning model

```
def define_model(findings_max_length, findings_vocab_size, impressions_max_length, impressions_vocab_size, findi
```

# feature extractor model

```
inputs1 = tf.keras.layers.Input(shape=(2048,))
```

```
image1 = tf.keras.layers.Dropout(0.5)(inputs1)
```

```
image2 = tf.keras.layers.Dense(256, activation='relu')(image1)
```

# finding model

```
inputs2 = tf.keras.layers.Input(shape=(findings_max_length,))
```

```
findings1 = tf.keras.layers.Embedding(findings_vocab_size, 300, weights=[findings_embedding_matrix], trainable
```

```
findings2 = tf.keras.layers.Flatten()(findings1)
```

```
im_f = tf.concat([image2, findings2], axis=1)
```

# sequence model

```
inputs3 = tf.keras.layers.Input(shape=(impressions_max_length,))
```

```
impressions1 = tf.keras.layers.Embedding(impressions_vocab_size, 300, weights=[impressions_embedding_matr
```

```
impressions1 = tf.keras.layers.Embedding(impressions_vocab_size,500,weights=[impressions_embedding_matrix])
impressions2 = tf.keras.layers.Dropout(0.5)(impressions1)
impressions3 = tf.keras.layers.LSTM(256)(impressions2)

# decoder model
decoder1= tf.concat([im_f, impressions3],axis=1)
decoder2 = tf.keras.layers.Dense(500, activation='relu')(decoder1)
outputs = tf.keras.layers.Dense(impressions_vocab_size, activation='softmax')(decoder2)
# tie it together [image, seq] [word]
model = tf.keras.models.Model(inputs=[inputs1, inputs2, inputs3], outputs=outputs)
# compile model
model.compile(loss='categorical_crossentropy', optimizer='adam')
# summarize model
model.summary()
plot_model(model, to_file='/content/drive/My Drive/Xception/Model2/finalmodel.png', show_shapes=True)
return model

# define the model
model = define_model(findings_max_length,findings_vocab_size,impressions_max_length,impressions_vocab_size
```





Model: "model"

## ▼ Run Model

```
xtrain=[X1train, X2train , X3train]
xcv=[X1cv, X2cv, X3cv]
# define checkpoint callback
checkpoint = tf.keras.callbacks.ModelCheckpoint('/content/drive/My Drive/Xception/Model2/finalcheck.hdf5', m

log_dir="/content/drive/My Drive/Xception/Model2/finaltensorboardlogs/logs/fit/" + datetime.now().strftime("%
tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir, histogram_freq=1, write_graph=True,write

# fit model
h=model.fit(xtrain, ytrain,batch_size=512, epochs=25, verbose=2,callbacks=[tensorboard_callback,checkpoint], va
```



WARNING:tensorflow:`write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback.  
Epoch 1/25

Epoch 00001: val\_loss improved from inf to 1.69536, saving model to /content/drive/My Drive/Xception/M  
314/314 - 161s - loss: 2.6970 - val\_loss: 1.6954  
Epoch 2/25

Epoch 00002: val\_loss improved from 1.69536 to 0.99211, saving model to /content/drive/My Drive/Xcepti  
314/314 - 160s - loss: 1.1903 - val\_loss: 0.9921  
Epoch 3/25

Epoch 00003: val\_loss improved from 0.99211 to 0.67842, saving model to /content/drive/My Drive/Xcepti  
314/314 - 160s - loss: 0.7286 - val\_loss: 0.6784  
Epoch 4/25

Epoch 00004: val\_loss improved from 0.67842 to 0.51504, saving model to /content/drive/My Drive/Xcepti  
314/314 - 159s - loss: 0.4776 - val\_loss: 0.5150  
Epoch 5/25

Epoch 00005: val\_loss improved from 0.51504 to 0.36081, saving model to /content/drive/My Drive/Xcepti  
314/314 - 158s - loss: 0.3292 - val\_loss: 0.3608  
Epoch 6/25

Epoch 00006: val\_loss improved from 0.36081 to 0.25499, saving model to /content/drive/My Drive/Xcepti  
314/314 - 158s - loss: 0.2474 - val\_loss: 0.2550  
Epoch 7/25

Epoch 00007: val\_loss improved from 0.25499 to 0.22702, saving model to /content/drive/My Drive/Xcepti  
314/314 - 158s - loss: 0.1933 - val\_loss: 0.2270  
Epoch 8/25

Epoch 00008: val\_loss improved from 0.22702 to 0.15800, saving model to /content/drive/My Drive/Xcepti  
314/314 - 157s - loss: 0.1587 - val\_loss: 0.1580  
Epoch 9/25

Epoch 00009: val\_loss improved from 0.15800 to 0.13293, saving model to /content/drive/My Drive/Xcepti  
314/314 - 158s - loss: 0.1382 - val\_loss: 0.1329  
Epoch 10/25

Epoch 00010: val\_loss improved from 0.13293 to 0.12608, saving model to /content/drive/My Drive/Xcepti  
314/314 - 157s - loss: 0.1309 - val\_loss: 0.1261  
Epoch 11/25

Epoch 00011: val\_loss improved from 0.12608 to 0.10120, saving model to /content/drive/My Drive/Xcepti  
314/314 - 155s - loss: 0.1142 - val\_loss: 0.1012  
Epoch 12/25

Epoch 00012: val\_loss improved from 0.10120 to 0.09612, saving model to /content/drive/My Drive/Xcepti  
314/314 - 156s - loss: 0.1052 - val\_loss: 0.0961  
Epoch 13/25

Epoch 00013: val\_loss did not improve from 0.09612  
314/314 - 155s - loss: 0.1107 - val\_loss: 0.1080  
Epoch 14/25

Epoch 00014: val\_loss improved from 0.09612 to 0.09071, saving model to /content/drive/My Drive/Xcepti  
314/314 - 155s - loss: 0.1010 - val\_loss: 0.0907  
Epoch 15/25

Epoch 00015: val\_loss did not improve from 0.09071  
314/314 - 155s - loss: 0.1040 - val\_loss: 0.1046  
Epoch 16/25

Epoch 00016: val\_loss improved from 0.09071 to 0.05937, saving model to /content/drive/My Drive/Xcepti  
314/314 - 154s - loss: 0.0959 - val\_loss: 0.0594  
Epoch 17/25

Epoch 00017: val\_loss improved from 0.05937 to 0.05246, saving model to /content/drive/My Drive/Xcepti  
314/314 - 155s - loss: 0.0865 - val\_loss: 0.0525  
Epoch 18/25

Epoch 00018: val\_loss did not improve from 0.05246  
314/314 - 154s - loss: 0.0862 - val\_loss: 0.0748  
Epoch 19/25

Epoch 00019: val\_loss improved from 0.05246 to 0.04458, saving model to /content/drive/My Drive/Xcepti  
314/314 - 154s - loss: 0.0872 - val\_loss: 0.0446  
Epoch 20/25

Epoch 00020: val\_loss did not improve from 0.04458  
314/314 - 153s - loss: 0.0801 - val\_loss: 0.0640  
Epoch 21/25

Epoch 00021: val\_loss did not improve from 0.04458  
314/314 - 153s - loss: 0.0790 - val\_loss: 0.0515  
Epoch 22/25

Epoch 00022: val\_loss did not improve from 0.04458  
314/314 - 154s - loss: 0.0731 - val\_loss: 0.0502  
Epoch 23/25

Epoch 00023: val\_loss did not improve from 0.04458  
314/314 - 153s - loss: 0.0759 - val\_loss: 0.0673  
Epoch 24/25

Epoch 00024: val\_loss improved from 0.04458 to 0.04004, saving model to /content/drive/My Drive/Xcepti  
314/314 - 154s - loss: 0.0770 - val\_loss: 0.0400  
Epoch 25/25

Epoch 00025: val\_loss did not improve from 0.04004  
314/314 - 153s - loss: 0.0852 - val\_loss: 0.0431

%tensorboard --logdir='/content/drive/My Drive/Xception/Model2/finaltensorboardlogs/logs/fit'



## TensorBoard

SCALARS

GRAPHS

DISTRIBUTIONS

HISTOGRAMS

☐ Show data download links☐ Ignore outliers in chart scalingTooltip sorting  
method: default

Smoothing



0.6

Horizontal Axis

STEP

RELATIVE

WALL

Runs

Write a regex to filter runs

☐ 20200501-104827/train☐ 20200501-104827/validation

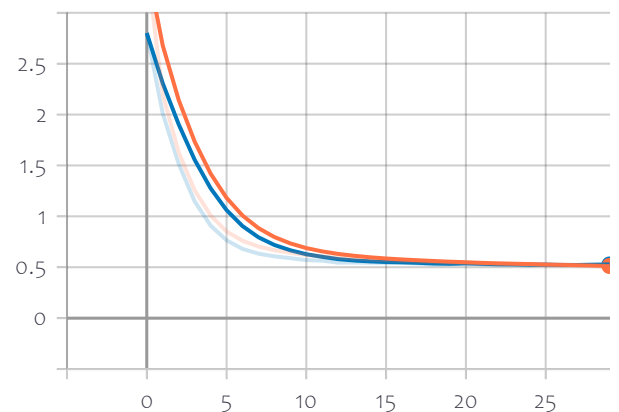
TOGGLE ALL RUNS

/content/drive/My Drive/Xception/Model1/  
checkmergetensorboardlogs1/logs/fit

Filter tags (regular expressions supported)

epoch\_loss

epoch\_loss



## ▼ Prepare test data

```
test_image_features = load_image_features(image_extracted_features, test_id)
print('\nTotal test images    : ', len(test_image_features))
```

```
test_id_findings = load_respective_set(cleaned_id_findings, test_id)
print('\nTotal test findings   : ', len(test_id_findings))
```

```
test_id_finding_sequences = finding_sequences(findings_tokenizer, findings_max_length, test_id_findings)

print("\n-----\n")

test_id_impressions = load_respective_set(cleaned_id_impressions, test_id)
print('Total test impressions : ', len(test_id_impressions))
```



Total test images : 250

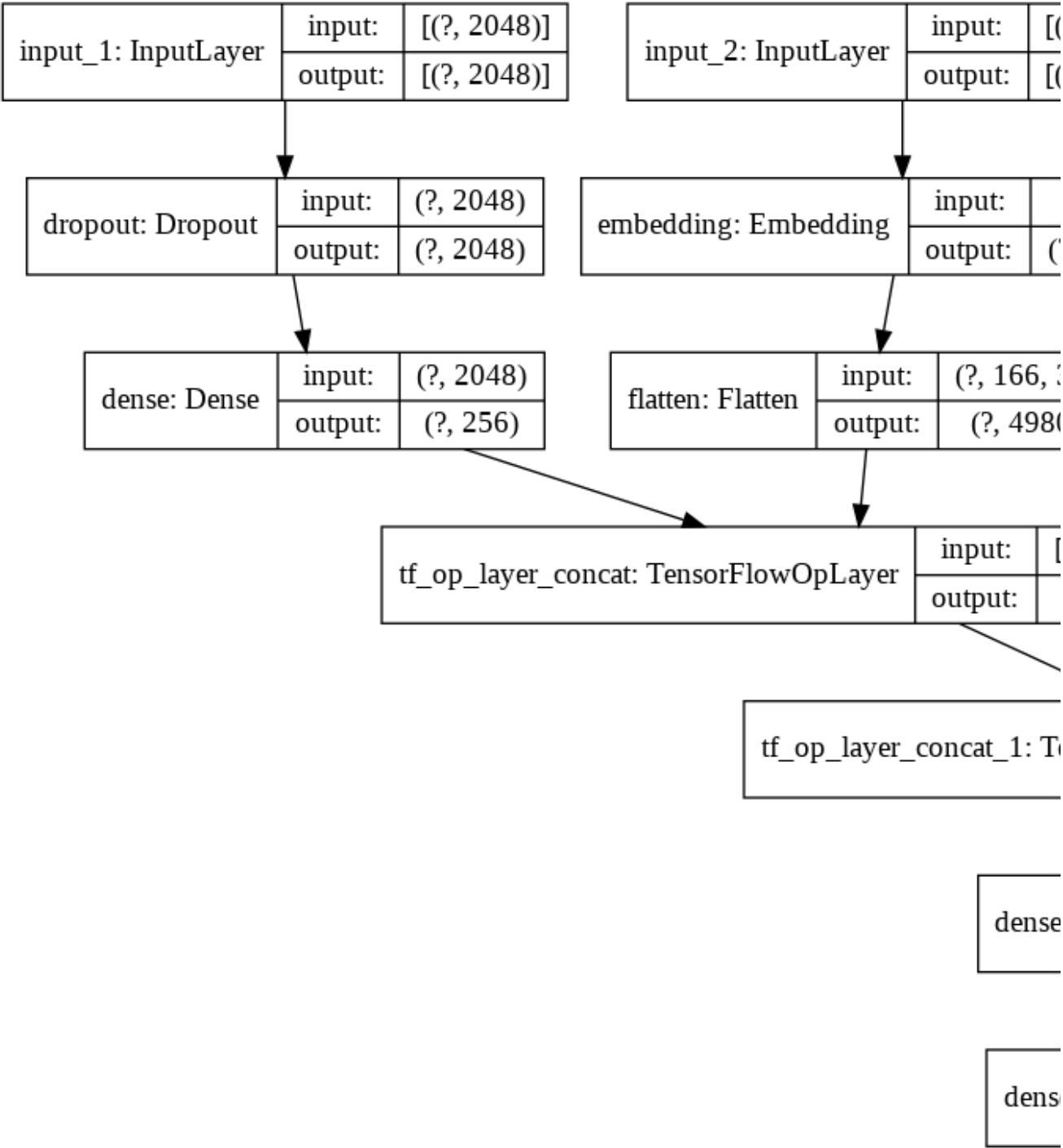
Total test findings : 250

-----

Total test impressions : 250

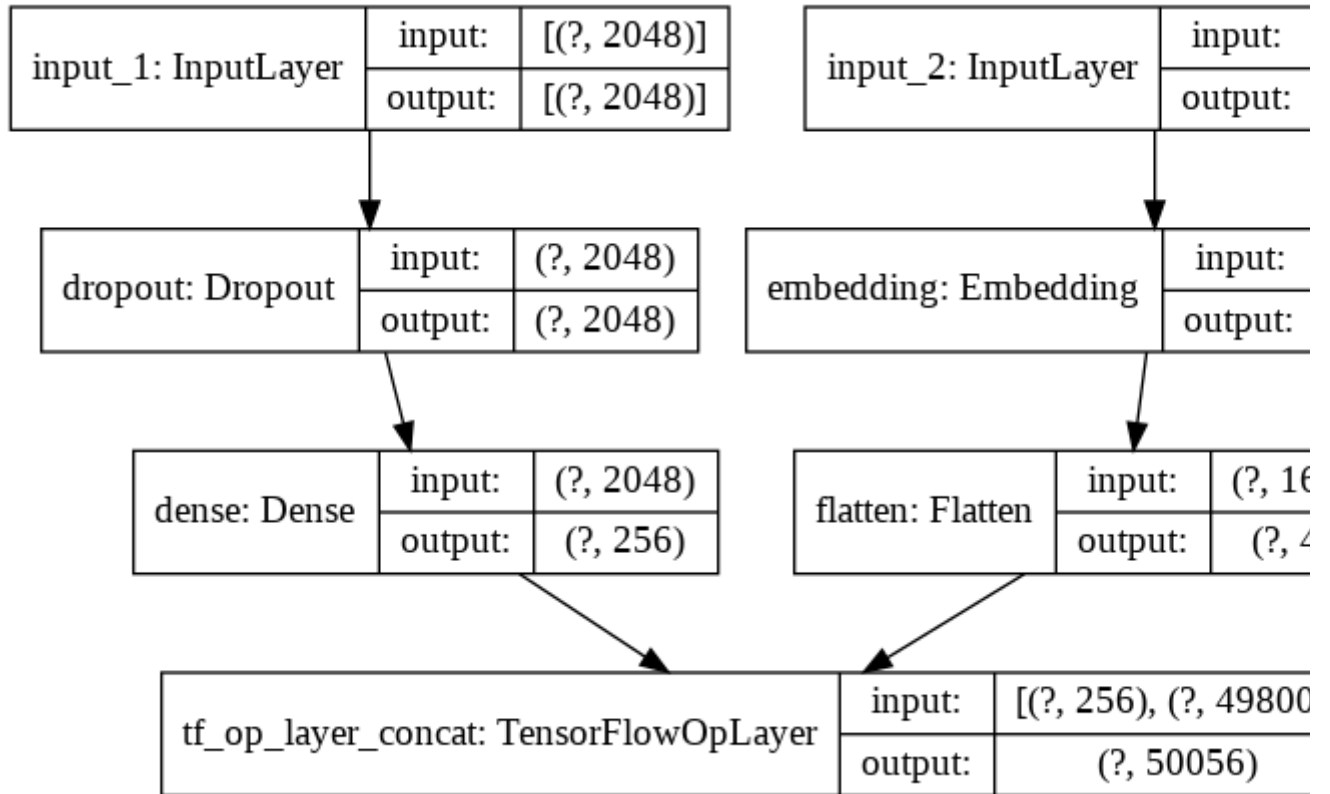
```
#here we display model architecture again just to understand further flow easily
# load the model
model1 = tf.keras.models.load_model('/content/drive/My Drive/Xception/Model2/finalcheck.hdf5')
tf.keras.utils.plot_model(model1, show_shapes=True)
```



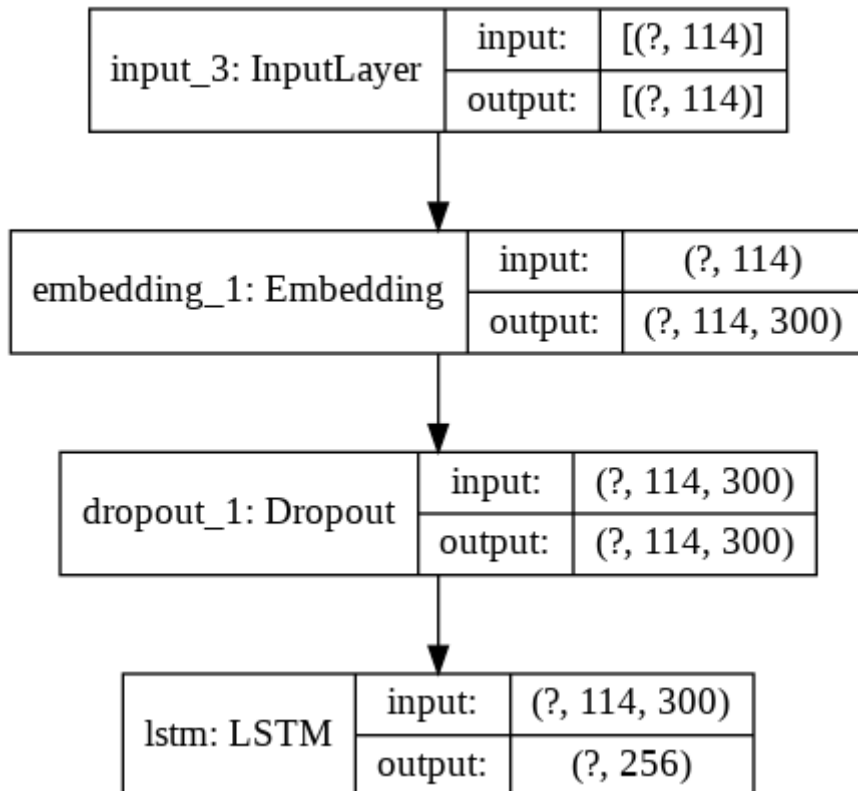


```
#we calling below architecture recurrently while predicting next word in the sequence so we trying to compute o
#https://stackoverflow.com/questions/41711190/keras-how-to-get-the-output-of-each-layer
#https://stackoverflow.com/questions/43452353/obtaining-output-of-an-intermediate-layer-in-tensorflow-keras
model1a = Model(inputs=model1.inputs[:2], outputs=model1.layers[-5].output)
tf.keras.utils.plot_model(model1a, show_shapes=True)
```





#we call below architecture for generating each word in the sequence  
 model1b = Model(inputs=model1.inputs[2], outputs=model1.layers[-4].output)  
 tf.keras.utils.plot\_model(model1b, show\_shapes=True)



# index of desired layer  
 idx = 11  
 # get the input shape of desired layer  
 input\_shape = model1.layers[idx].get\_input\_shape\_at(0)

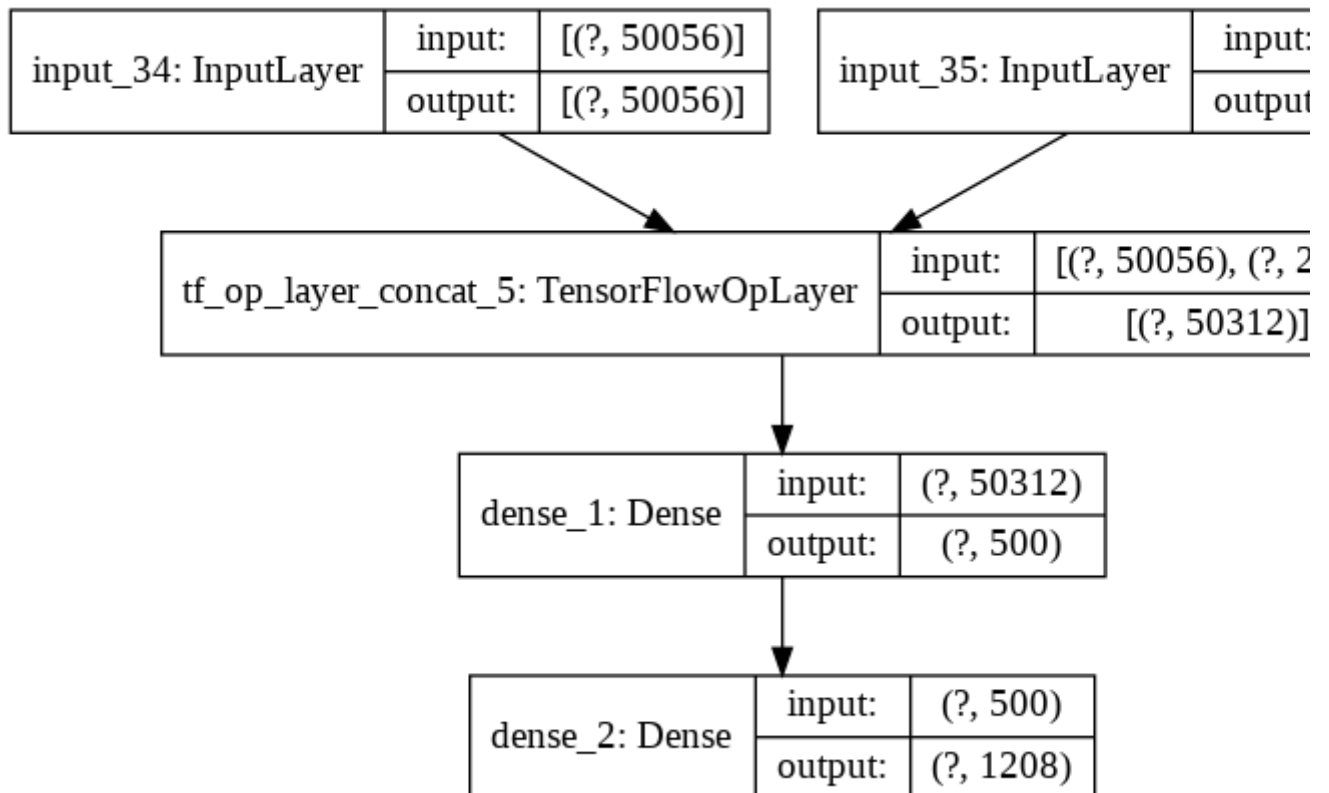
```

a,b=input_shape
layer_input1 = Input(shape=a[1])
layer_input2 = Input(shape=b[1])

merged_input= tf.concat([layer_input1, layer_input2],axis=1)
x = merged_input
for layer1 in model1.layers[12:]:
    x = layer1(x)

#https://keras.io/getting-started/functional-api-guide/
# create the model
new_model1 = Model(inputs=[layer_input1,layer_input2],outputs= x)
new_model1.load_weights("/content/drive/My Drive/Xception/Model2/finalcheck.hdf5", by_name=True)
tf.keras.utils.plot_model(new_model1, show_shapes=True)

```



## ▼ Evaluate Model

```

# map an integer to a word
def word_for_id(integer, tokenizer):
    for word, index in tokenizer.word_index.items():
        if index == integer:
            return word
    return None

# generate a description for an image(changed approach)
def generate_desc(tokenizer, photo, max_length, finding):
    # seed the generation process
    in_text = 'startseq'
    photo=photo.reshape((1,photo.shape[0]))
    finding=finding.reshape((1,finding.shape[0]))

```



```

finding=finding.reshape((1,finding.shape[0]))
output1=model1a.predict([photo,finding], verbose=0)
# iterate over the whole length of the sequence

for i in range(max_length):
    # integer encode input sequence
    sequence = tokenizer.texts_to_sequences([in_text])[0]
    # pad input
    sequence = pad_sequences([sequence], maxlen=max_length)
    output2 = model1b.predict(sequence, verbose=0)

    yhat = new_model1.predict([output1, output2], verbose=0)
    # convert probability to integer
    yhat = argmax(yhat)
    # map integer to word
    word = word_for_id(yhat, tokenizer)
    # stop if we cannot map the word
    if word is None:
        break
    # append as input for generating the next word
    in_text += ' ' + word
    # stop if we predict the end of the sequence
    if word == 'endseq':
        break
return in_text

```

# remove start/end sequence tokens from a summary

```

def cleanup_summary(summary):
    # remove start of sequence token
    index = summary.find('startseq ')
    if index > -1:
        summary = summary[len('startseq '):]
    # remove end of sequence token
    index = summary.find(' endseq')
    if index > -1:
        summary = summary[:index]
    return summary

```

# evaluate the skill of the model

```

def evaluate_model(descriptions, photos, tokenizer, max_length, findings):
    actual, predicted = list(), list()
    # step over the whole set
    for key, desc in descriptions.items():
        # generate description
        yhat = generate_desc(tokenizer, photos[key][0], max_length, findings[key])
        # clean up prediction
        yhat = cleanup_summary(yhat)
        # store actual and predicted
        references = [cleanup_summary(desc).split()]
        actual.append(references)
        predicted.append(yhat.split())
    # calculate BLEU score
    print('BLEU-1: %f' % corpus_bleu(actual, predicted, weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus_bleu(actual, predicted, weights=(0.5, 0.5, 0, 0)))

```

```
print('BLEU-2: %f' % corpus_bleu(actual, predicted, weights=(0.5, 0.5, 0, 0)))
print('BLEU-3: %f' % corpus_bleu(actual, predicted, weights=(0.3, 0.3, 0.3, 0)))
print('BLEU-4: %f' % corpus_bleu(actual, predicted, weights=(0.25, 0.25, 0.25, 0.25)))
```

```
# evaluate model
```

```
evaluate_model(test_id_impressions, test_image_features, impressions_tokenizer, impressions_max_length, test_id)
```



```
BLEU-1: 0.958754
```

```
BLEU-2: 0.950945
```

```
BLEU-3: 0.946297
```

```
BLEU-4: 0.927488
```

**Conclusion: This model is sensible and ready to predict on unseen data points.**

## ▼ Step by Step Procedure to solve this case study

- 1) Apply data augmentation technique on images to increase image data at the same time preprocess extracted feature vectors.
- 2) Now we done with image features ,lets move to the text data.As we know the text data we got is these files ,remove none values,preprocess it and store in dictionary format where key is a image id because it will be easier to get required data field in next steps.
- 3) we build our model on [image data+findings] + impressions data
  - First we split data into train,cv and test sets.
  - we define necessary functions after this we create input sequences i.e. we convert text data maximum length of impression's text data for further processing.
  - Training :Define Deep learning model for training and we save log files and give checkpoint path
  - Model Evaluation: we define necessary functions after this we load our best model to predict
- 5) Summarize results

