```
In [0]:
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=947318989803-6bn6
qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect uri=urn%3aietf%3awg%3aoauth%3a2.0%
b&response type=code&scope=email%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdocs.test%20https%3a%2
www.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly
ttps%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly
Enter your authorization code:
Mounted at /content/drive
In [0]:
# ## https://stackoverflow.com/questions/32834731/how-to-delete-a-file-by-extension-in-python
# ## to delete all .npy files
# import os
# dir name = '/content/drive/My Drive/Colab Notebooks/Medical data case study'
# test = os.listdir(dir name)
# for item in test:
     if item.endswith(".npy"):
          os.remove(os.path.join(dir_name, item))
In [0]:
import tensorflow as tf
# You'll generate plots of attention in order to see which parts of an image
# our model focuses on during captioning
import matplotlib.pyplot as plt
from tqdm import tqdm
# Scikit-learn includes many helpful utilities
from sklearn.model selection import train test split
from sklearn.utils import shuffle
import os
# import tensorflow as tf
import xml.etree.ElementTree
import numpy as np
import pandas as pd
import re
import re
import numpy as np
import os
import time
import json
from glob import glob
from PIL import Image
import pickle
In [0]:
print(tf. version )
2.2.0-rc2
In [0]:
tf.test.gpu_device_name()
Out[0]:
'/device:GPU:0'
```

In [N] .

```
±11 [∪]•
```

```
!pip install -q tqdm
```

```
image_folder = '/content/drive/My Drive/Colab Notebooks/Medical data case study'
annotation_folder = '/content/drive/My Drive/Colab Notebooks/Medical data case study/ecgen-radiolo
gy'
```

```
## https://github.com/wisdal/diagnose-and-explain/blob/master/prepare dataset.py
def extract data():
   all_findings = []
   all_impressions = []
   all img names = []
   rids = []
   total count = 0 # Count of reports available in the dataset
   no image count = 0 # Count of reports having no associated chest image
   no impression count = 0 # Count of reports having an empty "Impression" section
   \verb"no_findings_count" = 0 \# \textit{Count of reports having an empty "Findings" section"}
    # Storing impressions, findings and the image names in vectors
   for file in tqdm(os.listdir(annotation_folder)):
    # for file in tqdm(annotation folder):
       total count += 1
       file = os.path.abspath(annotation_folder) + '/' + file
       e = xml.etree.ElementTree.parse(file).getroot()
       rid = e.find('pmcId').get('id') # Report Id
        # We choose to ignore reports having no associated image
       image id = e.find('parentImage')
       if image id is None:
            no image count += 1
            continue
       image_id = image_id.get('id')
         image_name = os.path.abspath('.') + '/' + image_id + '.png'
       image name = image folder + '/' + image id + '.png'
       findings = ''
       impression = ''
        # Parsing "Impression" and "Findings"
        for element in e.findall('MedlineCitation/Article/Abstract/AbstractText'):
            if element.get('Label') == 'FINDINGS':
                findings = element.text
            if element.get('Label') == 'IMPRESSION':
                impression = element.text
        # Sanity check: Skip this report if it has an empty "Impression" section
        if findings is None:
            no\_findings\_count += 1
            #findings = 'No finding'
            continue
        if impression is None:
            no_impression_count += 1
            continue
        # Transforming findings and impressions into lists of sentences
        # https://stackoverflow.com/questions/21840389/python-regular-expression-remove-period-
from-number-at-end-of-sentence
       findings = findings.replace("XXXX", "") #"XXXX" represents information anonymized
        findings=re.sub('((d+)[].])(?!([d]+))','g<2>',findings)
        findings = re.sub(" \d+", " ", findings)
         sentences = findings.split('.')
       sentences = findings
         del sentences[-1]
         sentences = ['<start> ' + sentence + ' <end>' for sentence in sentences]
        # sentences = ['<start> ' + sentences + ' <end>']
        findings = sentences
        impression = impression.replace("XXXX", "") #"XXXX" represents information anonymized
        impression=re.sub('((\d+)[\.])(?!([\d]+))','\g<2>',impression)
```

```
impression = re.sub(" \d+", " ", impression)
#https://www.tutorialspoint.com/How-to-remove-specific-characters-from-a-string-in-Python
        impression=impression.replace("1", "")
          sentences = impression.split('.')
         del sentences[-1]
        sentences = impression
        sentences = ['<start> ' + sentence + ' <end>' for sentence in sentences]
        sentences = ['<start> ' + sentences + ' <end>' ]
        impression = sentences
        #appending to vectors
        all_img_names.append(image_name)
        all_findings.append(findings)
        all_impressions.append(impression)
        rids.append(rid)
    print("Number of reports available:", total_count)
    print("Number of reports selected:", len(all_img_names))
    print("Number of reports not having images (skipped):", no image count)
    print("Number of reports with Impression section empty (skipped):", no impression count)
    print("Number of reports with Findings section empty:", no findings count)
    print("Total skipped:", no image count + no impression count + no findings count)
    return all findings, all impressions, all img names, rids
In [19]:
all findings, all impressions, all img names, report id=extract data()
100%| 3965/3965 [14:40<00:00, 4.50it/s]
Number of reports available: 3965
Number of reports selected: 3341
Number of reports not having images (skipped): 104
Number of reports with Impression section empty (skipped): 6
Number of reports with Findings section empty: 514
Total skipped: 624
In [0]:
# df=pd.read csv("/content/drive/My Drive/Colab Notebooks/x ray image with report (1).csv")
In [28]:
all impressions[0]
Out[28]:
['<start> No acute cardiopulmonary abnormalities. <end>']
In [0]:
def load_image(image_path):
    img = tf.io.read file(image path)
    img = tf.image.decode jpeg(img, channels=3)
    img = tf.image.resize(img, (512, 624))
    img = tf.keras.applications.inception v3.preprocess input(img)
    return img, image path
In [0]:
## Converting Lists to Dataframe
## https://stackoverflow.com/questions/30522724/take-multiple-lists-into-dataframe
# df = pd.DataFrame(
      { 'report_id':report_id,
          'findings': all findings,
       'impressions': all impressions,
```

'image names': all_img_names

df.to csv('x ray image with report.csv')

```
# df=pd.read csv('x ray image with report.csv')
In [49]:
impressions=[]
for i in all impressions:
 for j in i:
   impressions.append(j)
print(impressions[:1])
all impressions=impressions
['<start> No acute cardiopulmonary abnormalities. <end>']
In [23]:
image model = tf.keras.applications.InceptionV3(include top=False,
                                               weights='imagenet')
new_input = image_model.input
hidden layer = image model.layers[-1].output
image_features_extract_model = tf.keras.Model(new_input, hidden_layer)
Downloading data from https://storage.googleapis.com/tensorflow/keras-
applications/inception\_v3/inception\_v3\_weights\_tf\_dim\_ordering\_tf\_kernels\_notop.h5
In [0]:
# Get unique images
# all findings, all impressions, all img names, report id
encode train = sorted(set(all img names))
# Feel free to change batch_size according to your system configuration
image_dataset = tf.data.Dataset.from_tensor_slices(encode_train)
image_dataset = image_dataset.map(load_image, num_parallel_calls=tf.data.experimental.AUTOTUNE).bat
ch (16)
In [25]:
for img, path in tqdm(image dataset):
 batch features = image features extract model(img)
 batch features = tf.reshape(batch features,
                              (batch features.shape[0], -1, batch features.shape[3]))
  for bf, p in zip(batch features, path):
   path of feature = p.numpy().decode("utf-8")
   np.save(path of feature, bf.numpy())
209it [07:46, 2.23s/it]
In [0]:
# Find the maximum length of any caption in our dataset
def calc max length(tensor):
   return max(len(t) for t in tensor)
# Choose the top 5000 words from the vocabulary
top k = 5000
tokenizer = tf.keras.preprocessing.text.Tokenizer(num words=top k,
                                                 oov token="<unk>", split=' ', char level=False,
                                                 filters='!"#$%&()*+.,-/:;=?@[\]^_`{|}~ ')
tokenizer.fit on texts(all impressions)
train_seqs = tokenizer.texts_to_sequences(all_impressions)
# tokenizer.fit on texts(df['impressions'])
# train seqs = tokenizer.texts to sequences(df['impressions'])
```

tokenizer.word_index['<pad>'] = 0
tokenizer.index word[0] = '<pad>'

```
In [55]:
len(tokenizer.word_index)
Out[55]:
1225
In [62]:
print(all impressions[3])
print(tokenizer.index_word[10])
print(tokenizer.word index['<start>'])
print(tokenizer.index_word[1224])
print(tokenizer.word index['<start>'])
<start> No acute disease. <end>
the
contusion
In [0]:
# Create the tokenized vectors
# train_seqs = tokenizer.texts_to_sequences(all_impressions)
# Pad each vector to the max_length of the captions
# If you do not provide a max length value, pad sequences calculates it automatically
cap_vector = tf.keras.preprocessing.sequence.pad_sequences(train_seqs, padding='post')
# Calculates the max length, which is used to store the attention weights
max_length = calc_max_length(train_seqs)
In [64]:
max length
Out[64]:
114
In [67]:
train seqs[10]
Out[67]:
[2, 4, 5, 34, 6, 16, 3]
In [0]:
# Create training and validation sets using an 80-20 split
img_name_train, img_name_val, cap_train, cap_val = train_test_split(all_img_names,
                                                                     cap_vector,
                                                                      test size=0.2,
                                                                      random_state=0)
len(img_name_train), len(cap_train), len(img_name_val), len(cap_val)
Out[69]:
(2672, 2672, 669, 669)
In [0]:
# Feel free to change these parameters according to your system's configuration
```

```
BATCH_SIZE = 64
BUFFER_SIZE = 1000
embedding_dim = 256
units = 512
vocab_size = top_k + 1
num_steps = len(img_name_train) // BATCH_SIZE
# Shape of the vector extracted from InceptionV3 is (64, 2048)
# These two variables represent that vector shape
features_shape = 2048
attention_features_shape = 64
```

```
# Load the numpy files
def map_func(img_name, cap):
   img_tensor = np.load(img_name.decode('utf-8')+'.npy')
   return img_tensor, cap
```

In [0]:

```
class BahdanauAttention(tf.keras.Model):
  def init (self, units):
   super(BahdanauAttention, self). init ()
    self.W1 = tf.keras.layers.Dense(units)
    self.W2 = tf.keras.layers.Dense(units)
    self.V = tf.keras.layers.Dense(1)
  def call(self, features, hidden):
    # features(CNN encoder output) shape == (batch size, 64, embedding dim)
    # hidden shape == (batch size, hidden size)
    # hidden_with_time_axis shape == (batch_size, 1, hidden_size)
    hidden with time axis = tf.expand dims(hidden, 1)
    # score shape == (batch size, 64, hidden size)
    score = tf.nn.tanh(self.W1(features) + self.W2(hidden with time axis))
    # attention weights shape == (batch size, 64, 1)
    # you get 1 at the last axis because you are applying score to self.V
    attention weights = tf.nn.softmax(self.V(score), axis=1)
    # context vector shape after sum == (batch size, hidden size)
    context_vector = attention_weights * features
    context vector = tf.reduce sum(context vector, axis=1)
    return context vector, attention weights
class CNN Encoder(tf.keras.Model):
    # Since you have already extracted the features and dumped it using pickle
    \# This encoder passes those features through a Fully connected layer
    def init (self, embedding dim):
        super(CNN Encoder, self). init ()
        # shape after fc == (batch size, 64, embedding dim)
        self.fc = tf.keras.layers.Dense(embedding dim)
    def call(self, x):
       x = self.fc(x)
        x = tf.nn.relu(x)
        return x
```

```
class RNN Decoder(tf.keras.Model):
 def __init__ (self, embedding_dim, units, vocab_size):
   super(RNN_Decoder, self).__init__()
   self.units = units
    self.embedding = tf.keras.layers.Embedding(vocab size, embedding dim)
    self.gru = tf.keras.layers.GRU(self.units,
                                    return sequences=True,
                                    return state=True,
                                    recurrent initializer='glorot uniform')
    self.fc1 = tf.keras.layers.Dense(self.units)
    self.fc2 = tf.keras.layers.Dense(vocab size)
    self.attention = BahdanauAttention(self.units)
  def call(self, x, features, hidden):
    # defining attention as a separate model
    context vector, attention weights = self.attention(features, hidden)
    # x shape after passing through embedding == (batch_size, 1, embedding_dim)
    x = self.embedding(x)
    # x shape after concatenation == (batch_size, 1, embedding_dim + hidden_size)
    x = tf.concat([tf.expand_dims(context_vector, 1), x], axis=-1)
    # passing the concatenated vector to the GRU
    output, state = self.gru(x)
    # shape == (batch_size, max_length, hidden_size)
   x = self.fcl(output)
    # x shape == (batch size * max length, hidden_size)
   x = tf.reshape(x, (-1, x.shape[2]))
    # output shape == (batch_size * max_length, vocab)
    x = self.fc2(x)
   return x, state, attention weights
  def reset state(self, batch size):
    return tf.zeros((batch size, self.units))
```

```
encoder = CNN_Encoder(embedding_dim)
decoder = RNN_Decoder(embedding_dim, units, vocab_size)
```

In [0]:

```
optimizer = tf.keras.optimizers.Adam()
loss_object = tf.keras.losses.SparseCategoricalCrossentropy(
    from_logits=True, reduction='none')

def loss_function(real, pred):
    mask = tf.math.logical_not(tf.math.equal(real, 0))
    loss_ = loss_object(real, pred)

mask = tf.cast(mask, dtype=loss_.dtype)
loss_ *= mask

return tf.reduce_mean(loss_)
```

In [0]:

```
start epoch = 0
if ckpt manager.latest checkpoint:
 start_epoch = int(ckpt_manager.latest_checkpoint.split('-')[-1])
  # restoring the latest checkpoint in checkpoint path
  ckpt.restore(ckpt manager.latest checkpoint)
In [0]:
# adding this in a separate cell because if you run the training cell
# many times, the loss_plot array will be reset
loss_plot = []
In [79]:
len(tokenizer.word_index)
Out[79]:
1225
In [80]:
tokenizer.word_index['<pad>']
Out[80]:
In [81]:
tokenizer.index_word[1222]
Out[81]:
'thickness'
In [82]:
tokenizer.word_index['<start>']
Out[82]:
In [0]:
# tokenizer.index word[1223]='<start>'
# tokenizer.word index['<start>'] = 1223
# tokenizer.index word[1224]='<end>'
# tokenizer.word index['<end>'] = 1224
In [0]:
# tokenizer.word index['<start> ']
In [0]:
@tf.function
def train step(img tensor, target):
 loss = 0
  # initializing the hidden state for each batch
  # because the captions are not related from image to image
 hidden = decoder.reset state(batch size=target.shape[0])
  dec_input = tf.expand_dims([tokenizer.word_index['<start>']] * target.shape[0], 1)
  with tf.GradientTape() as tape:
     features = encoder(img tensor)
```

```
for i in range(1, target.shape[1]):
          # passing the features through the decoder
         predictions, hidden, _ = decoder(dec_input, features, hidden)
         loss += loss_function(target[:, i], predictions)
          # using teacher forcing
         dec input = tf.expand dims(target[:, i], 1)
  total loss = (loss / int(target.shape[1]))
  trainable_variables = encoder.trainable_variables + decoder.trainable_variables
  gradients = tape.gradient(loss, trainable variables)
  optimizer.apply_gradients(zip(gradients, trainable_variables))
  return loss, total_loss
In [0]:
# print(img tensor.shape)
# print(target.shape)
In [89]:
EPOCHS = 11
# with strategy.scope():
for epoch in range(start epoch, EPOCHS):
   start = time.time()
```

```
total loss = 0
    for (batch, (img_tensor, target)) in enumerate(dataset):
        batch loss, t loss = train step(img tensor, target)
        total_loss += t_loss
        if batch % 100 == 0:
            print ('Epoch {} Batch {} Loss {:.4f}'.format(
             epoch + 1, batch, batch_loss.numpy() / int(target.shape[1])))
    # storing the epoch end loss value to plot later
    loss_plot.append(total_loss / num steps)
    if epoch % 5 == 0:
      ckpt_manager.save()
    print ('Epoch {} Loss {:.6f}'.format(epoch + 1,
                                        total loss/num_steps))
    print ('Time taken for 1 epoch {} sec\n'.format(time.time() - start))
Epoch 1 Batch 0 Loss 0.5977
Epoch 1 Loss 0.396087
Time taken for 1 epoch 368.83429646492004 sec
Epoch 2 Batch 0 Loss 0.3455
Epoch 2 Loss 0.321712
Time taken for 1 epoch 157.97896194458008 sec
Epoch 3 Batch 0 Loss 0.3333
Epoch 3 Loss 0.274314
Time taken for 1 epoch 162.1726553440094 sec
Epoch 4 Batch 0 Loss 0.1958
Epoch 4 Loss 0.247043
Time taken for 1 epoch 163.10213565826416 sec
Epoch 5 Batch 0 Loss 0.2843
Epoch 5 Loss 0.231827
Time taken for 1 epoch 162.9009621143341 sec
Epoch 6 Batch 0 Loss 0.1763
Epoch 6 Loss 0.221931
Time taken for 1 epoch 162.6929795742035 sec
```

```
Epoch 7 Batch 0 Loss 0.1965
Epoch 7 Loss 0.213717
Time taken for 1 epoch 162.3735854625702 sec

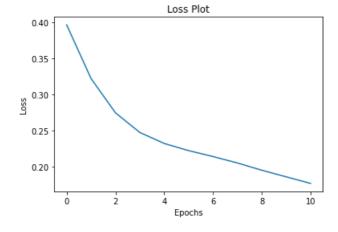
Epoch 8 Batch 0 Loss 0.1609
Epoch 8 Loss 0.204830
Time taken for 1 epoch 162.4220209121704 sec

Epoch 9 Batch 0 Loss 0.1808
Epoch 9 Loss 0.194778
Time taken for 1 epoch 162.6420075893402 sec

Epoch 10 Batch 0 Loss 0.2129
Epoch 10 Loss 0.185614
Time taken for 1 epoch 162.40754747390747 sec

Epoch 11 Batch 0 Loss 0.1838
Epoch 11 Loss 0.176504
Time taken for 1 epoch 162.49075555801392 sec
```

```
plt.plot(loss_plot)
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.title('Loss Plot')
plt.show()
```



```
def evaluate(image):
    attention_plot = np.zeros((max_length, attention_features_shape))
    hidden = decoder.reset state(batch size=1)
    temp_input = tf.expand_dims(load_image(image)[0], 0)
    img tensor val = image features extract model(temp input)
    img_tensor_val = tf.reshape(img_tensor_val, (img_tensor_val.shape[0], -1, img_tensor_val.shape[
31))
    features = encoder(img tensor val)
    dec input = tf.expand dims([tokenizer.word index['<start>']], 0)
    result = []
    for i in range(max length):
       predictions, hidden, attention_weights = decoder(dec_input, features, hidden)
        attention_plot[i] = tf.reshape(attention_weights, (-1, )).numpy()
        predicted_id = tf.random.categorical(predictions, 1)[0][0].numpy()
        result.append(tokenizer.index word[predicted id])
        if tokenizer.index_word[predicted_id] == ' <end>':
            return result, attention_plot
```

```
dec_input = tf.expand_dims([predicted_id], 0)

attention_plot = attention_plot[:len(result), :]
return result, attention_plot
```

```
def plot_attention(image, result, attention_plot):
    temp_image = np.array(Image.open(image))

fig = plt.figure(figsize=(10, 10))

len_result = len(result)

for l in range(len_result):
    temp_att = np.resize(attention_plot[l], (8, 8))
    ax = fig.add_subplot(len_result//2, len_result//2, l+1)
    ax.set_title(result[l])
    img = ax.imshow(temp_image)
    ax.imshow(temp_att, cmap='gray', alpha=0.6, extent=img.get_extent())

plt.tight_layout()
    plt.show()
```

In [104]:

```
# captions on the validation set
rid = np.random.randint(0, len(img_name_val))  #from live session code on URL Shortner
image = img_name_val[rid]
real_caption = ' '.join([tokenizer.index_word[i] for i in cap_val[rid] if i not in [0]])
result, attention_plot = evaluate(image)

print ('Real Caption:', real_caption)
print ('Prediction Caption:', ' '.join(result))
plot_attention(image, result, attention_plot)
```

ValueError: could not broadcast input array from shape (252) into shape (64)

In [106]:

```
from IPython.display import Image
Image(filename=image)
```

Out[106]:



