Deep Learning for Pneumonia Detection from Chest X-Ray Images

Table of Contents

Table of Contents

- Import Module
 - Import the Libraries
 - Importing the Dataset (section 1 2)
- Data Visualization and Augmentation
 - Data Visualization
 - Data Augmentation (section 2 2)
- Build the Model Main
- Evaluate the Model
 - Confusion Matrix & Classification Report
 - Plotting Loss and Accuracy
- Hyperparameter Tuning with SGD Optimizer Model 2
- <u>Hyperparameter Tuning with Network Depth Model 3</u>
- Inferences and Comments

Import Module

Import the Libraries

```
In [9]: # importing the packages necessary for this assignment problem
        import numpy as np
        import pandas as pd
        import math
        import os
        import timeit
        from sklearn import preprocessing
        from sklearn import metrics
        from sklearn.metrics import roc auc score
        from sklearn.metrics import roc curve
        from sklearn.model_selection import train_test_split
        from sklearn.model_selection import RepeatedStratifiedKFold
        from sklearn.model_selection import GridSearchCV
        #for checking testing results
        from sklearn.metrics import classification_report, confusion_matrix
        import statsmodels.api as sm
        from sklearn.metrics import accuracy_score
        # Import visualization libraries
        from matplotlib import pyplot as plt
        plt.rc("font", size=14)
        import matplotlib.cm as cm
        from mpl toolkits.mplot3d import Axes3D
        from matplotlib.ticker import AutoLocator, MaxNLocator, LinearLocator, Multipl
        import seaborn as sns
        sns.set(style="white")
        sns.set(style="whitegrid", color_codes=True)
        # and we want to view the charts inline
        %matplotlib inline
        # Import deep learning libraries
        import tensorflow as tf
        print("TensorFlow version:", tf.__version__)
        from tensorflow.keras.layers import Dense, Flatten, Conv2D
        from tensorflow.keras import Model
        from numpy import expand_dims
        from keras.preprocessing.image import load_img
        from keras.preprocessing.image import img to array
        from keras.preprocessing.image import ImageDataGenerator
```

TensorFlow version: 2.8.2

Importing the Dataset

```
In [10]: # Mount BITS Google Drive
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
# All the train, test and val folders stored as keys in an array (arr)
In [11]:
         arr = next(os.walk('drive/MyDrive/Colab Notebooks/chest_xray'))[1]
         arr = arr[2:]
         # Create a dictionary using the aforementioned keys to form a dataframe
         d = \{\}
         for i in arr:
           d[i] = next(os.walk('drive/MyDrive/Colab Notebooks/chest_xray/chest_xray/'+i
         for i in arr:
             target = []
             images = []
             for j in d[i]:
                 files = next(os.walk('drive/MyDrive/Colab Notebooks/chest_xray/chest_x
                 images += files
                 target += [j]*len(files)
             d[i] = pd.DataFrame({'images':np.array(images).flatten(),'target':np.array
         # Appending absolute paths to the image url
         for i in arr:
             d[i]['image_paths'] = 'drive/MyDrive/Colab Notebooks/chest_xray/chest_xra
         train = d['train']
         val = d['val']
         test = d['test']
         # Display the dataframe
         train
```

Out[1	.1]:	
-------	------	--

	images	target	image_paths
0	person549_bacteria_2303.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
1	person533_virus_1055.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
2	person491_bacteria_2081.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
3	person504_bacteria_2127.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
4	person548_bacteria_2297.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
			
5213	IM-0512-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest
5214	IM-0523-0001-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest
5215	IM-0524-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest
5216	IM-0531-0001-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest
5217	IM-0520-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest

5218 rows × 3 columns

```
In [12]: # Shuffling the dataframe
    train = train.sample(frac=1).reset_index()
    train = train.drop(['index'],axis=1)
    train
```

Out[12]:		images	target	image_paths
_	0	person1814_bacteria_4669.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
	1	person359_bacteria_1646.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
	2	person420_bacteria_1847.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
	3	IM-0555-0001.jpeg	NORMAL	drive/MyDrive/Colab Notebooks/chest_xray/chest
	4	person1122_virus_1847.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
!	5213	person294_virus_611.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
!	5214	person827_bacteria_2738.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
!	5215	person1433_virus_2447.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
!	5216	person279_bacteria_1315.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest
!	5217	person448_bacteria_1933.jpeg	PNEUMONIA	drive/MyDrive/Colab Notebooks/chest_xray/chest

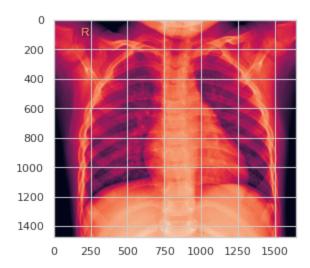
5218 rows × 3 columns

Data Visualization and Augmentation

Data Visualization

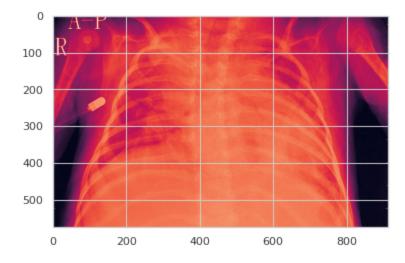
```
In [14]: # Display Normal image
    img_normal = plt.imread(train['image_paths'][3])
    plt.imshow(img_normal)
    print('target:' ,train['target'][3])
```

target: NORMAL



```
In [15]: # Display Pneumonia image
   img_pneumonia = plt.imread(train['image_paths'][0])
   plt.imshow(img_pneumonia)
   print('target:' ,train['target'][0])
```

target: PNEUMONIA



Data Augmentation

```
In [17]: #directory=train_path, #dataframe contains the full paths

ds_train = train_datagen.flow_from_dataframe(train, x_col = 'image_paths', y_colass_mode = 'binary', batch_size

ds_val = val_datagen.flow_from_dataframe(val, x_col = 'image_paths', y_col = 'class_mode = 'binary', batch_size = B

ds_test = test_datagen.flow_from_dataframe(test, x_col = 'image_paths', y_col class_mode = 'binary', batch_size = B
```

Found 5216 validated image filenames belonging to 2 classes. Found 16 validated image filenames belonging to 2 classes. Found 624 validated image filenames belonging to 2 classes.

/usr/local/lib/python3.7/dist-packages/keras_preprocessing/image/dataframe_it
erator.py:282: UserWarning: Found 2 invalid image filename(s) in x_col="image
_paths". These filename(s) will be ignored.
 .format(n invalid, x col)

Build the Model - Main

```
In [37]: model = tf.keras.Sequential()
         # Input layer
         model.add(tf.keras.layers.Conv2D(filters=16, kernel_size=3, activation='relu',
         model.add(tf.keras.layers.MaxPool2D(pool size=2, strides=2))
         # Hidden Layer 1
         model.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu')
         model.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 2
         model.add(tf.keras.layers.Conv2D(filters=64, kernel_size=3, activation='relu')
         model.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 3
         model.add(tf.keras.layers.Conv2D(filters=64, kernel_size=3, activation='relu')
         model.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 4 - Flattening
         model.add(tf.keras.layers.Flatten())
         model.add(tf.keras.layers.Dense(units=128, activation='relu'))
         # Output Layer
         model.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
         # Print Model Summary
         model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 16)	448
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 31, 31, 16)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	4640
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 14, 14, 32)	0
conv2d_2 (Conv2D)	(None, 12, 12, 64)	18496
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 6, 6, 64)	0
conv2d_3 (Conv2D)	(None, 4, 4, 64)	36928
<pre>max_pooling2d_3 (MaxPooling 2D)</pre>	(None, 2, 2, 64)	0
flatten (Flatten)	(None, 256)	0
dense (Dense)	(None, 128)	32896
dense_1 (Dense)	(None, 1)	129

Total params: 93,537 Trainable params: 93,537 Non-trainable params: 0

```
### CallBacks
In [38]:
         count = 0
         ### Save the best model
         saved_callbacks = tf.keras.callbacks.ModelCheckpoint('./saved_models/bestmodel
                                                               save weights only=False,
         ### Compile and Fit Model
         model.compile(optimizer = tf.keras.optimizers.Adam(), loss = tf.keras.losses.B
         history = model.fit(ds train, validation data = ds val, epochs=12, verbose=2,
         Epoch 1/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 1015s - loss: 0.4373 - accuracy: 0.8029 - val_loss: 0.5706 - val_ac
         curacy: 0.7500 - 1015s/epoch - 3s/step
         Epoch 2/12
         326/326 - 86s - loss: 0.2020 - accuracy: 0.9191 - val loss: 0.8160 - val accu
         racy: 0.6250 - 86s/epoch - 263ms/step
         Epoch 3/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 88s - loss: 0.1723 - accuracy: 0.9369 - val_loss: 0.4881 - val_accu
         racy: 0.8750 - 88s/epoch - 269ms/step
         Epoch 4/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 89s - loss: 0.1452 - accuracy: 0.9429 - val_loss: 0.3954 - val_accu
         racy: 0.8125 - 89s/epoch - 274ms/step
         Epoch 5/12
         326/326 - 86s - loss: 0.1386 - accuracy: 0.9440 - val_loss: 0.4305 - val_accu
         racy: 0.8125 - 86s/epoch - 264ms/step
         Epoch 6/12
         326/326 - 88s - loss: 0.1180 - accuracy: 0.9572 - val_loss: 0.6775 - val_accu
         racy: 0.6875 - 88s/epoch - 270ms/step
         Epoch 7/12
         326/326 - 88s - loss: 0.1259 - accuracy: 0.9519 - val_loss: 0.4522 - val_accu
         racy: 0.7500 - 88s/epoch - 271ms/step
         Epoch 8/12
         326/326 - 89s - loss: 0.1132 - accuracy: 0.9578 - val_loss: 0.5081 - val_accu
         racy: 0.6250 - 89s/epoch - 273ms/step
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 89s - loss: 0.1025 - accuracy: 0.9611 - val_loss: 0.3894 - val_accu
         racy: 0.8750 - 89s/epoch - 273ms/step
         Epoch 10/12
         326/326 - 87s - loss: 0.1160 - accuracy: 0.9559 - val_loss: 0.5935 - val_accu
         racy: 0.6875 - 87s/epoch - 268ms/step
         Epoch 11/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 89s - loss: 0.1115 - accuracy: 0.9590 - val_loss: 0.3646 - val accu
         racy: 0.8125 - 89s/epoch - 272ms/step
         Epoch 12/12
         INFO:tensorflow:Assets written to: ./saved models/bestmodel 0/assets
         326/326 - 88s - loss: 0.1022 - accuracy: 0.9615 - val_loss: 0.3255 - val_accu
         racy: 0.9375 - 88s/epoch - 270ms/step
```

Evaluate the Model

Confusion Matrix & Classification Report

```
In [40]:
         #Confusion Matrix and Classification Report
         Y_pred = model.predict(ds_test, 39)
         y_pred = np.argmax(Y_pred, axis=1)
         print('Confusion Matrix')
         print(confusion_matrix(ds_test.classes, y_pred))
         print('Classification Report')
         target_names = ['Normal', 'Pneumonia']
         print(classification_report(ds_test.classes, y_pred, target_names=target_names
         Confusion Matrix
         [[234
                 0]
          [390
                 0]]
         Classification Report
                       precision recall f1-score
                                                       support
               Normal
                            0.38
                                      1.00
                                                0.55
                                                           234
            Pneumonia
                            0.00
                                      0.00
                                                0.00
                                                           390
                                                0.38
                                                           624
             accuracy
                            0.19
                                      0.50
                                                0.27
                                                           624
            macro avg
         weighted avg
                            0.14
                                      0.38
                                                0.20
                                                           624
```

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131
8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se
t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t

o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131
8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

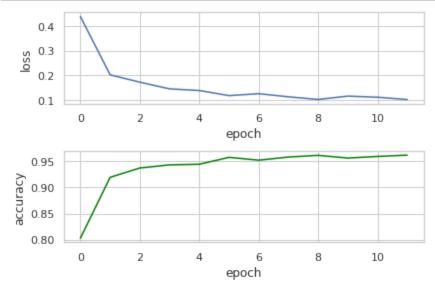
_warn_prf(average, modifier, msg_start, len(result))

Plotting Loss and Accuracy

```
In [41]: plt.subplot(2 ,1, 1)
    a = plt.plot(history.history['loss'])
    plt.ylabel('loss')
    plt.xlabel('epoch')

plt.subplot(2, 1 ,2)
    b = plt.plot(history.history['accuracy'],color="green")
    plt.ylabel('accuracy')
    plt.xlabel('epoch')

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



Hyperparameter Tuning with SGD Optimizer - Model 2

```
In [42]: model_2 = tf.keras.Sequential()
         # Input layer
         model_2.add(tf.keras.layers.Conv2D(filters=16, kernel_size=3, activation='relu
         model 2.add(tf.keras.layers.MaxPool2D(pool size=2, strides=2))
         # Hidden Layer 1
         model_2.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu
         model_2.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 2
         model 2.add(tf.keras.layers.Conv2D(filters=64, kernel size=3, activation='relu
         model_2.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 3
         model_2.add(tf.keras.layers.Conv2D(filters=64, kernel_size=3, activation='relu
         model_2.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 4 - Flattening
         model_2.add(tf.keras.layers.Flatten())
         model_2.add(tf.keras.layers.Dense(units=128, activation='relu'))
         # Output Layer
         model 2.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
         # Print Model Summary
         model 2.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_4 (Conv2D)	(None, 62, 62, 16)	448
<pre>max_pooling2d_4 (MaxPooling 2D)</pre>	(None, 31, 31, 16)	0
conv2d_5 (Conv2D)	(None, 29, 29, 32)	4640
<pre>max_pooling2d_5 (MaxPooling 2D)</pre>	(None, 14, 14, 32)	0
conv2d_6 (Conv2D)	(None, 12, 12, 64)	18496
<pre>max_pooling2d_6 (MaxPooling 2D)</pre>	(None, 6, 6, 64)	0
conv2d_7 (Conv2D)	(None, 4, 4, 64)	36928
<pre>max_pooling2d_7 (MaxPooling 2D)</pre>	(None, 2, 2, 64)	0
flatten_1 (Flatten)	(None, 256)	0
dense_2 (Dense)	(None, 128)	32896
dense_3 (Dense)	(None, 1)	129

Total params: 93,537 Trainable params: 93,537 Non-trainable params: 0

```
### CallBacks
In [43]:
         count = 0
         ### Save the best model
         saved_callbacks = tf.keras.callbacks.ModelCheckpoint('./saved_models/bestmodel
                           save weights only=False, monitor='val loss', save best only=
         ### Compile and Fit Model
         model_2.compile(optimizer = tf.keras.optimizers.SGD(), loss = tf.keras.losses.
         history 2 = model 2.fit(ds train, validation data = ds val, epochs=12, verbose
         Epoch 1/12
         INFO:tensorflow:Assets written to: ./saved models/bestmodel 0/assets
         326/326 - 91s - loss: 0.5813 - accuracy: 0.7421 - val_loss: 0.7957 - val_accu
         racy: 0.5000 - 91s/epoch - 280ms/step
         Epoch 2/12
         326/326 - 88s - loss: 0.5558 - accuracy: 0.7429 - val loss: 0.7961 - val accu
         racy: 0.5000 - 88s/epoch - 270ms/step
         Epoch 3/12
         326/326 - 90s - loss: 0.5014 - accuracy: 0.7590 - val_loss: 0.8956 - val accu
         racy: 0.5625 - 90s/epoch - 276ms/step
         Epoch 4/12
         INFO:tensorflow:Assets written to: ./saved models/bestmodel 0/assets
         326/326 - 89s - loss: 0.4146 - accuracy: 0.8066 - val_loss: 0.5913 - val_accu
         racy: 0.6875 - 89s/epoch - 274ms/step
         Epoch 5/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 90s - loss: 0.3224 - accuracy: 0.8589 - val_loss: 0.4584 - val_accu
         racy: 0.8125 - 90s/epoch - 276ms/step
         Epoch 6/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 91s - loss: 0.2773 - accuracy: 0.8854 - val_loss: 0.3833 - val_accu
         racy: 0.8125 - 91s/epoch - 279ms/step
         Epoch 7/12
         326/326 - 88s - loss: 0.2350 - accuracy: 0.9078 - val_loss: 0.4682 - val_accu
         racy: 0.6875 - 88s/epoch - 271ms/step
         Epoch 8/12
         326/326 - 88s - loss: 0.2229 - accuracy: 0.9107 - val_loss: 0.5048 - val accu
         racy: 0.6250 - 88s/epoch - 271ms/step
         Epoch 9/12
         326/326 - 88s - loss: 0.2025 - accuracy: 0.9199 - val_loss: 0.5802 - val_accu
         racy: 0.6875 - 88s/epoch - 270ms/step
         Epoch 10/12
         326/326 - 89s - loss: 0.1981 - accuracy: 0.9202 - val_loss: 0.3999 - val_accu
         racy: 0.7500 - 89s/epoch - 272ms/step
         326/326 - 88s - loss: 0.1812 - accuracy: 0.9321 - val_loss: 0.5754 - val_accu
         racy: 0.6875 - 88s/epoch - 270ms/step
         Epoch 12/12
         326/326 - 88s - loss: 0.1819 - accuracy: 0.9296 - val_loss: 0.8051 - val_accu
         racy: 0.6875 - 88s/epoch - 269ms/step
```

```
In [44]: model_2 = tf.keras.models.load_model('./saved_models/bestmodel_'+str(count))
model_2.evaluate(ds_test)
```

Out[44]: [0.3778236210346222, 0.8157051205635071]

```
In [45]: #Confusion Matrix and Classification Report
   Y_pred_2 = model_2.predict_generator(ds_test, 40)
   y_pred_2 = np.argmax(Y_pred_2, axis=1)
   print('Confusion Matrix')
   print(confusion_matrix(ds_test.classes, y_pred_2))
   print('Classification Report')
   target_names = ['Normal', 'Pneumonia']
   print(classification_report(ds_test.classes, y_pred_2, target_names=target_name)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: UserWarning: `Model.predict_generator` is deprecated and will be removed in a future versi on. Please use `Model.predict`, which supports generators.

WARNING:tensorflow:Your input ran out of data; interrupting training. Make su re that your dataset or generator can generate at least `steps_per_epoch * ep ochs` batches (in this case, 40 batches). You may need to use the repeat() function when building your dataset.

Confusion Matrix

[[234 0] [390 0]]

Classification Report

	precision	recall	f1-score	support
Normal	0.38	1.00	0.55	234
Pneumonia	0.00	0.00	0.00	390
accuracy			0.38	624
macro avg	0.19	0.50	0.27	624
weighted avg	0.14	0.38	0.20	624

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

Hyperparameter Tuning with Network Depth - Model 3

```
In [46]: model_3 = tf.keras.Sequential()
         # Input Layer
         model_3.add(tf.keras.layers.Conv2D(filters=16, kernel_size=3, activation='relu
         model 3.add(tf.keras.layers.MaxPool2D(pool size=2, strides=2))
         # Hidden Layer 1
         model_3.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu
         model_3.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 2
         model 3.add(tf.keras.layers.Conv2D(filters=64, kernel size=3, activation='relu
         model_3.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
         # Hidden Layer 3 - Flattening
         model_3.add(tf.keras.layers.Flatten())
         model_3.add(tf.keras.layers.Dense(units=128, activation='relu'))
         # Output Layer
         model_3.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
         # Print Model Summary
         model_3.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
conv2d_8 (Conv2D)	(None, 62, 62, 16)	448
<pre>max_pooling2d_8 (MaxPooling 2D)</pre>	(None, 31, 31, 16)	0
conv2d_9 (Conv2D)	(None, 29, 29, 32)	4640
<pre>max_pooling2d_9 (MaxPooling 2D)</pre>	(None, 14, 14, 32)	0
conv2d_10 (Conv2D)	(None, 12, 12, 64)	18496
<pre>max_pooling2d_10 (MaxPoolin g2D)</pre>	(None, 6, 6, 64)	0
<pre>flatten_2 (Flatten)</pre>	(None, 2304)	0
dense_4 (Dense)	(None, 128)	295040
dense_5 (Dense)	(None, 1)	129

Total params: 318,753 Trainable params: 318,753 Non-trainable params: 0

```
### CallBacks
In [47]:
         count = 0
         ### Save the best model
         saved_callbacks = tf.keras.callbacks.ModelCheckpoint('./saved_models/bestmodel
                           save weights only=False, monitor='val loss', save best only=
         ### Compile and Fit Model
         model_3.compile(optimizer = tf.keras.optimizers.Adam(), loss = tf.keras.losses
         history 3 = model 3.fit(ds train, validation data = ds val, epochs=12, verbose
         Epoch 1/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 91s - loss: 0.4191 - accuracy: 0.8129 - val_loss: 0.3850 - val_accu
         racy: 0.8125 - 91s/epoch - 280ms/step
         Epoch 2/12
         INFO:tensorflow:Assets written to: ./saved models/bestmodel 0/assets
         326/326 - 88s - loss: 0.2493 - accuracy: 0.8984 - val_loss: 0.3273 - val_accu
         racy: 0.8125 - 88s/epoch - 270ms/step
         Epoch 3/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 88s - loss: 0.1883 - accuracy: 0.9287 - val_loss: 0.3138 - val_accu
         racy: 0.8750 - 88s/epoch - 271ms/step
         Epoch 4/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 88s - loss: 0.1780 - accuracy: 0.9302 - val_loss: 0.2996 - val_accu
         racy: 0.9375 - 88s/epoch - 271ms/step
         Epoch 5/12
         326/326 - 89s - loss: 0.1569 - accuracy: 0.9396 - val loss: 0.3946 - val accu
         racy: 0.8750 - 89s/epoch - 273ms/step
         Epoch 6/12
         326/326 - 88s - loss: 0.1425 - accuracy: 0.9484 - val_loss: 0.3055 - val_accu
         racy: 0.9375 - 88s/epoch - 269ms/step
         326/326 - 88s - loss: 0.1343 - accuracy: 0.9473 - val_loss: 0.3967 - val_accu
         racy: 0.8125 - 88s/epoch - 269ms/step
         Epoch 8/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 91s - loss: 0.1349 - accuracy: 0.9479 - val_loss: 0.2722 - val_accu
         racy: 0.8125 - 91s/epoch - 280ms/step
         326/326 - 89s - loss: 0.1247 - accuracy: 0.9503 - val_loss: 0.3647 - val_accu
         racy: 0.8125 - 89s/epoch - 274ms/step
         Epoch 10/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 91s - loss: 0.1099 - accuracy: 0.9551 - val_loss: 0.2673 - val_accu
         racy: 0.8750 - 91s/epoch - 280ms/step
         Epoch 11/12
         INFO:tensorflow:Assets written to: ./saved_models/bestmodel_0/assets
         326/326 - 90s - loss: 0.1291 - accuracy: 0.9519 - val_loss: 0.2562 - val_accu
         racy: 0.9375 - 90s/epoch - 276ms/step
         Epoch 12/12
         326/326 - 90s - loss: 0.1182 - accuracy: 0.9549 - val_loss: 0.3900 - val_accu
         racy: 0.8750 - 90s/epoch - 277ms/step
```

```
In [48]: model_3 = tf.keras.models.load_model('./saved_models/bestmodel_'+str(count))
model_3.evaluate(ds_test)
```

Out[48]: [0.4252314567565918, 0.8669871687889099]

```
In [50]: #Confusion Matrix and Classification Report
    Y_pred_3 = model_3.predict_generator(ds_test, 40)
    y_pred_3 = np.argmax(Y_pred_3, axis=1)
    print('Confusion Matrix')
    print(confusion_matrix(ds_test.classes, y_pred_3))
    print('Classification Report')
    target_names = ['Normal', 'Pneumonia']
    print(classification_report(ds_test.classes, y_pred_3, target_names=target_name)
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: UserWarning: `Model.predict_generator` is deprecated and will be removed in a future versi on. Please use `Model.predict`, which supports generators.

WARNING:tensorflow:Your input ran out of data; interrupting training. Make su re that your dataset or generator can generate at least `steps_per_epoch * ep ochs` batches (in this case, 40 batches). You may need to use the repeat() function when building your dataset.

Confusion Matrix

[[234 0] [390 0]]

Classification Report

	precision	recall	f1-score	support
Normal	0.38	1.00	0.55	234
Pneumonia	0.00	0.00	0.00	390
accuracy			0.38	624
macro avg	0.19	0.50	0.27	624
weighted avg	0.14	0.38	0.20	624

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:131 8: UndefinedMetricWarning: Precision and F-score are ill-defined and being se t to 0.0 in labels with no predicted samples. Use `zero_division` parameter t o control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

Inferences and Comments

Explanations

- The results of the Adam optimizer are generally better than every other optimization algorithms, have faster computation time, and require fewer parameters for tuning. Because of all that, Adam is recommended as the default optimizer for most of the applications.
- The main model that used Adam optimizer and 4 hidden layers gave ~90% accuracy whereas the other 2 models gave a lower accuracy. The model that used SGD and 4 hidden layers gave ~82% accuracy. The model that used Adam optimizer and 3 hidden layers gave ~87% accuracy.