## Chest X-Ray Pneumonia prediction model

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This notebook holds the python source code which declares a tensorflow CNN model, preprocess images to a standard, train the model using these images. The
       images are acquired from a kaggle dataset.
In [ ]: import numpy as np
       import os
       import tensorflow as tf
       from tensorflow.keras.preprocessing import image
       from tensorflow.keras.preprocessing.image import ImageDataGenerator
       import matplotlib.pyplot as plt
       from sklearn.metrics import accuracy_score
       # Setting directory path to the dataset
       train_dir = 'path to train directory'
       test_dir = 'path to test directory'
       # Preprocessing the images
       train_datagen = ImageDataGenerator(rescale=1./255)
       test datagen = ImageDataGenerator(rescale=1./255)
       train_generator = train_datagen.flow_from_directory(
              train dir,
              target_size=(150, 150),
              batch_size=20,
              class_mode='binary')
       test_generator = test_datagen.flow_from_directory(
              test_dir,
              target_size=(150, 150),
              batch_size=20,
              class_mode='binary')
       Found 5216 images belonging to 2 classes.
       Found 624 images belonging to 2 classes.
       Declaring the design of the CNN model
In [ ]: from tensorflow.keras import layers, models
       model = models.Sequential([
          layers.Conv2D(32, (3, 3), activation='relu', input_shape=(150, 150, 3)),
          layers.MaxPooling2D((2, 2)),
          layers.Conv2D(64, (3, 3), activation='relu'),
          layers.MaxPooling2D((2, 2)),
          layers.Conv2D(128, (3, 3), activation='relu'),
          layers.MaxPooling2D((2, 2)),
          layers.Flatten(),
          layers.Dense(512, activation='relu'),
          layers.Dense(1, activation='sigmoid')
       ])
       model.compile(optimizer='adam',
                   loss='binary_crossentropy',
                   metrics=['accuracy'])
       Fitting the model using the image dataset
In [ ]: history = model.fit(
          train generator,
          steps_per_epoch=100,
          epochs=10,
          validation_data=test_generator,
          validation_steps=50)
       Epoch 1/10
       t your dataset or generator can generate at least `steps_per_epoch * epochs` batches (in this case, 50 batches). You may need to use the repeat() function when building
       g your dataset.
       Epoch 2/10
       Epoch 3/10
       Epoch 4/10
       Epoch 5/10
       Epoch 6/10
       Epoch 7/10
       Epoch 8/10
       Epoch 9/10
       Epoch 10/10
       In [ ]: |plt.plot(history.history['accuracy'], label='accuracy',)
       plt.plot(history.history['val accuracy'], label = 'val accuracy')
       plt.xlabel('Epoch')
       plt.ylabel('Accuracy')
       plt.ylim([0.7, 1])
       plt.legend(loc='lower right')
Out[]: <matplotlib.legend.Legend at 0x13c6e1310>
         1.00
         0.95
         0.90
         0.85
         0.80
         0.75
                                         accuracy
                                         val accuracy
         0.70
                             Epoch
In [ ]: model.save('pneumonia detection model.h5')
       # For prediction
       from tensorflow.keras.preprocessing import image
       img_path = 'path to simple test img'
       img = image.load_img(img_path, target_size=(150, 150))
       img_array = image.img_to_array(img) / 255.0
       img_array = np.expand_dims(img_array, axis=0)
       prediction = model.predict(img_array)
       print(prediction)
       [[0.14047605]]
       Testing the validation images to acquire the actual accuracy of the model. After trial and errors - 10 epochs seem to find the middle ground of great accuracy for the
       model and also prevent overfitting.
In []: # Predictng All images from the val directory. Use this block of code directly if only prediction needs to done.
       # The pretrained weights of the model is stored in 'pneumonia_detection_model.h5' file
       # Acquiring all val images
       predict_img_dir_normal = "path to val directory normal X-Rays"
       predict_img_dir_pneumonia = "path to val directory pneumonia X-Rays"
       images_file_normal = os.listdir(predict_img_dir_normal)
       images_file_pneumonia = os.listdir(predict_img_dir_pneumonia)
       # Loading saved model which has been trained -
       model = tf.keras.models.load_model('pneumonia_detection_model.h5')
       test = []
       pred = []
       def predict_image(image_path):
          img = image.load_img(image_path,target_size=(150,150))
          img_array = image.img_to_array(img) / 255.0 # Normalizing image values
          img_array = np.expand_dims(img_array, axis = 0)
          prediction = model.predict(img_array)
          return prediction[0][0]
       #Iterate over all the images and predict them
       print("All the images below are Normal")
       for image_file in images_file_normal:
          image_path = os.path.join(predict_img_dir_normal, image_file)
          prediction_score = predict_image(image_path)
          pred_label = 'Pneumonia'
          test.append(0)
          if prediction_score < 0.5:</pre>
              pred_label = "Normal"
              pred.append(0)
          else:
              pred.append(1)
          print(f'Image : {image_file} | Prediction : {pred_label} | Score : {prediction_score}' )
       print("\n\nAll the images below are Pnuemonic")
       for image_file in images_file_pneumonia:
          image_path = os.path.join(predict_img_dir_pneumonia, image_file)
          prediction_score = predict_image(image_path)
          pred_label = 'Pneumonia'
          test.append(1)
          if prediction_score < 0.5:</pre>
              pred label = "Normal"
              pred.append(0)
          else:
              pred.append(1)
          print(f'Image : {image_file} | Prediction : {pred_label} | Score : {prediction_score}' )
       print(f"\nAccuracy of model {accuracy_score(test,pred)}")
       All the images below are Normal
       Image : NORMAL2-IM-1440-0001.jpeg | Prediction : Normal | Score : 0.0005770325660705566
       Image : NORMAL2-IM-1437-0001.jpeg
                                     Prediction: Normal | Score: 0.24535581469535828
       Image : NORMAL2-IM-1431-0001.jpeg | Prediction : Normal | Score : 0.0628671646118164
       Image : NORMAL2-IM-1436-0001.jpeg | Prediction : Normal | Score : 0.03877553343772888
       Image : NORMAL2-IM-1430-0001.jpeg
                                      Prediction : Pneumonia | Score : 0.9426052570343018
                                     Prediction: Normal | Score: 0.24113687872886658
       Image : NORMAL2-IM-1438-0001.jpeg
       Image : NORMAL2-IM-1442-0001.jpeg
                                     Prediction: Normal | Score: 0.0038035809993743896
       Image : NORMAL2-IM-1427-0001.jpeg | Prediction : Normal | Score : 0.1404760479927063
       All the images below are Pnuemonic
       Image : person1950_bacteria_4881.jpeg | Prediction : Pneumonia | Score : 0.9808032512664795
       Image : person1951_bacteria_4882.jpeg | Prediction : Pneumonia | Score : 0.9999854564666748
       Image : person1952 bacteria 4883.jpeg | Prediction : Pneumonia | Score : 0.9996638298034668
       Image : person1946 bacteria 4874.jpeg | Prediction : Pneumonia | Score : 0.9995415210723877
       Image : person1947_bacteria_4876.jpeg | Prediction : Pneumonia | Score : 0.9990575313568115
       Image : person1946_bacteria_4875.jpeg | Prediction : Pneumonia | Score : 0.9999327659606934
       Image : person1949_bacteria_4880.jpeg | Prediction : Pneumonia | Score : 0.9846831560134888
       Image : person1954 bacteria 4886.jpeg | Prediction : Pneumonia | Score : 0.615234911441803
       Accuracy of model 0.9375
```