Class Challenge: Image Classification of COVID-19 X-rays

Task 1 [Total points: 30]

Setup

- · This assignment involves the following packages: 'matplotlib', 'numpy', and 'sklearn'.
- If you are using conda, use the following commands to install the above packages:

```
conda install matplotlib
conda install numpy
conda install -c anaconda scikit-learn
```

If you are using pip, use use the following commands to install the above packages:

```
pip install matplotlib
pip install numpy
pip install sklearn
```

Data

Please download the data using the following link: COVID-19 (https://drive.google.com/file/d/1Y88tggpQ1Pjko 7rntcPowOJs QNOrJ-/view).

 After downloading 'Covid_Data_GradientCrescent.zip', unzip the file and you should see the following data structure:

```
|--all
|-----train
|----test
|--two
|-----train
|-----test
```

• Put the 'all' folder, the 'two' folder and this python notebook in the **same directory** so that the following code can correctly locate the data.

[20 points] Binary Classification: COVID-19 vs. Normal

In [1]:

```
import os
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.preprocessing.image import ImageDataGenerator

os.environ['OMP_NUM_THREADS'] = '1'
os.environ['CUDA_VISIBLE_DEVICES'] = '-1'
tf.__version__
```

Out[1]:

'2.3.1'

Load Image Data

In [2]:

```
DATA_LIST = os.listdir('two/train')

DATASET_PATH = 'two/train'

TEST_DIR = 'two/test'

IMAGE_SIZE = (224, 224)

NUM_CLASSES = len(DATA_LIST)

BATCH_SIZE = 10 # try reducing batch size or freeze more layers if your GPU runs out of memo ry

NUM_EPOCHS = 40

LEARNING_RATE = 0.0005 # start off with high rate first 0.001 and experiment with reducing it gradually
```

Generate Training and Validation Batches

In [3]:

C:\Users\Soo Whan Park\Anaconda3\lib\site-packages\keras_preprocessing\image\image\image data_generator.py:342: User\understring: This ImageDataGenerator specifies `zca_whiten ing` which overrides setting of`featurewise_std_normalization`.

warnings.warn('This ImageDataGenerator specifies '

Found 104 images belonging to 2 classes. Found 26 images belonging to 2 classes.

[10 points] Build Model

Hint: Starting from a pre-trained model typically helps performance on a new task, e.g. starting with weights obtained by training on ImageNet.

In [8]:

```
vgg16 = tf.keras.applications.VGG16(weights='imagenet', include_top=False,input_shape = (224,22
4,3))
vgg19 = tf.keras.applications.VGG19(include_top=False, weights='imagenet',input_shape=(224,224,
3))
model = tf.keras.models.Sequential([
    vgg16,
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(256, activation='relu', name='dense_feature'),
    tf.keras.layers.Dropout(0.5),
    tf.keras.layers.Dense(1, activation='sigmoid')
])
vgg16.trainable=False
model.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_2 (Flatten)	(None, 25088)	0
dense_feature (Dense)	(None, 256)	6422784
dropout (Dropout)	(None, 256)	0
dense_2 (Dense)	(None, 1)	257

Total params: 21,137,729 Trainable params: 6,423,041 Non-trainable params: 14,714,688

[5 points] Train Model

In [9]:

```
#FIT MODEL
print(len(train_batches))
print(len(valid_batches))

STEP_SIZE_TRAIN=train_batches.n//train_batches.batch_size
STEP_SIZE_VALID=valid_batches.n//valid_batches.batch_size

opt = tf.keras.optimizers.Adam(learning_rate=LEARNING_RATE)
model.compile(optimizer=opt,loss=tf.keras.losses.BinaryCrossentropy(),metrics=['accuracy'])
history = model.fit(train_batches, epochs=NUM_EPOCHS, validation_data=valid_batches,batch_size=5
, steps_per_epoch=STEP_SIZE_TRAIN, validation_steps = STEP_SIZE_VALID)
```

```
11
3
Epoch 1/40
10/10 [=======] - 11s 1s/step - loss: 1.3668 - accuracy: 0.
5532 - val_loss: 0.2846 - val_accuracy: 0.9000
Epoch 2/40
10/10 [=======] - 11s 1s/step - loss: 0.8508 - accuracy: 0.
7447 - val_loss: 0.3527 - val_accuracy: 0.8000
Epoch 3/40
                 10/10 [=====
7128 - val_loss: 0.1087 - val_accuracy: 1.0000
Epoch 4/40
10/10 [===========] - 13s 1s/step - loss: 0.5328 - accuracy: 0.
8298 - val_loss: 0.2401 - val_accuracy: 0.9000
Epoch 5/40
10/10 [===========] - 13s 1s/step - loss: 0.3701 - accuracy: 0.
8617 - val_loss: 0.0977 - val_accuracy: 1.0000
Epoch 6/40
10/10 [===========] - 13s 1s/step - loss: 0.3843 - accuracy: 0.
8298 - val_loss: 0.1259 - val_accuracy: 0.9500
Epoch 7/40
                     =======] - 14s 1s/step - loss: 0.3414 - accuracy: 0.
10/10 [========
8830 - val_loss: 0.2249 - val_accuracy: 0.8500
Epoch 8/40
10/10 [=======] - 14s 1s/step - Ioss: 0.2388 - accuracy: 0.
8830 - val_loss: 0.1552 - val_accuracy: 0.9500
Epoch 9/40
10/10 [=========== ] - 15s 1s/step - loss: 0.2265 - accuracy: 0.
9000 - val_loss: 0.0786 - val_accuracy: 1.0000
Epoch 10/40
10/10 [=======] - 14s 1s/step - Ioss: 0.2184 - accuracy: 0.
9149 - val_loss: 0.1589 - val_accuracy: 0.9500
Epoch 11/40
10/10 [========] - 14s 1s/step - loss: 0.1867 - accuracy: 0.
9255 - val_loss: 0.2855 - val_accuracy: 0.8500
Epoch 12/40
10/10 [========] - 16s 2s/step - loss: 0.4140 - accuracy: 0.
8511 - val_loss: 0.0672 - val_accuracy: 1.0000
Epoch 13/40
10/10 [=======] - 16s 2s/step - loss: 0.3380 - accuracy: 0.
8936 - val_loss: 0.0815 - val_accuracy: 1.0000
Epoch 14/40
10/10 [==========] - 16s 2s/step - loss: 0.2272 - accuracy: 0.
9300 - val_loss: 0.1965 - val_accuracy: 0.9500
Epoch 15/40
10/10 [=======] - 14s 1s/step - loss: 0.2073 - accuracy: 0.
9149 - val_loss: 0.0741 - val_accuracy: 0.9500
Epoch 16/40
10/10 [========= ] - 14s 1s/step - loss: 0.2699 - accuracy: 0.
8900 - val_loss: 0.1353 - val_accuracy: 0.9500
Epoch 17/40
10/10 [=======] - 16s 2s/step - loss: 0.2177 - accuracy: 0.
9362 - val_loss: 0.0419 - val_accuracy: 1.0000
Epoch 18/40
10/10 [==========] - 17s 2s/step - loss: 0.1747 - accuracy: 0.
9255 - val_loss: 0.0615 - val_accuracy: 1.0000
Epoch 19/40
10/10 [========] - 13s 1s/step - loss: 0.2097 - accuracy: 0.
9200 - val_loss: 0.1721 - val_accuracy: 0.9000
Epoch 20/40
10/10 [=========== ] - 13s 1s/step - loss: 0.2036 - accuracy: 0.
```

```
9468 - val_loss: 0.1535 - val_accuracy: 0.9000
Epoch 21/40
10/10 [==========] - 14s 1s/step - loss: 0.1357 - accuracy: 0.
9574 - val_loss: 0.0838 - val_accuracy: 0.9500
Epoch 22/40
10/10 [=========== ] - 13s 1s/step - loss: 0.1379 - accuracy: 0.
9468 - val_loss: 0.0242 - val_accuracy: 1.0000
Epoch 23/40
10/10 [=========== ] - 13s 1s/step - loss: 0.1019 - accuracy: 0.
9574 - val_loss: 0.0668 - val_accuracy: 1.0000
Epoch 24/40
                       =======] - 13s 1s/step - loss: 0.2107 - accuracy: 0.
10/10 [=====
9468 - val_loss: 0.0729 - val_accuracy: 0.9500
Epoch 25/40
10/10 [=======] - 13s 1s/step - loss: 0.2468 - accuracy: 0.
8936 - val_loss: 0.0348 - val_accuracy: 1.0000
Epoch 26/40
10/10 [=========== ] - 12s 1s/step - loss: 0.1046 - accuracy: 0.
9574 - val_loss: 0.0575 - val_accuracy: 1.0000
Epoch 27/40
                 -----] - 12s 1s/step - loss: 0.0842 - accuracy: 0.
10/10 [=======
9787 - val_loss: 0.0718 - val_accuracy: 0.9500
Epoch 28/40
10/10 [=======] - 12s 1s/step - Ioss: 0.1237 - accuracy: 0.
9468 - val_loss: 0.1208 - val_accuracy: 0.9500
Epoch 29/40
10/10 [=======] - 12s 1s/step - Ioss: 0.1880 - accuracy: 0.
9362 - val_loss: 0.0999 - val_accuracy: 0.9500
Epoch 30/40
10/10 [=========== ] - 13s 1s/step - loss: 0.2133 - accuracy: 0.
9149 - val_loss: 0.3189 - val_accuracy: 0.8500
Epoch 31/40
10/10 [======
                       =======] - 13s 1s/step - loss: 0.2183 - accuracy: 0.
8936 - val_loss: 0.0584 - val_accuracy: 1.0000
Epoch 32/40
10/10 [===========] - 13s 1s/step - loss: 0.1557 - accuracy: 0.
9468 - val_loss: 0.1706 - val_accuracy: 0.8500
Epoch 33/40
10/10 [=========== ] - 13s 1s/step - loss: 0.1601 - accuracy: 0.
9574 - val_loss: 0.0608 - val_accuracy: 1.0000
Epoch 34/40
10/10 [==========] - 12s 1s/step - loss: 0.1702 - accuracy: 0.
9468 - val_loss: 0.1405 - val_accuracy: 0.9500
Epoch 35/40
10/10 [==========] - 13s 1s/step - loss: 0.1395 - accuracy: 0.
9500 - val_loss: 0.0570 - val_accuracy: 1.0000
Epoch 36/40
10/10 [=======] - 13s 1s/step - Ioss: 0.1088 - accuracy: 0.
9681 - val_loss: 0.1375 - val_accuracy: 0.9000
Epoch 37/40
10/10 [========] - 12s 1s/step - loss: 0.1176 - accuracy: 0.
9574 - val_loss: 0.0493 - val_accuracy: 1.0000
Epoch 38/40
                       =======] - 13s 1s/step - loss: 0.0728 - accuracy: 0.
10/10 [======
9787 - val_loss: 0.0876 - val_accuracy: 0.9500
Epoch 39/40
10/10 [========] - 13s 1s/step - loss: 0.1640 - accuracy: 0.
9100 - val_loss: 0.0367 - val_accuracy: 1.0000
Epoch 40/40
10/10 [========== ] - 13s 1s/step - loss: 0.1594 - accuracy: 0.
9043 - val_loss: 0.1470 - val_accuracy: 0.9500
```

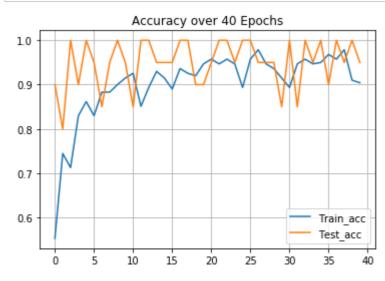
[5 points] Plot Accuracy and Loss During Training

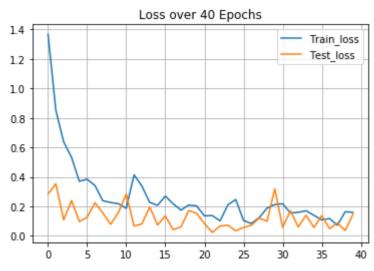
In [10]:

```
import matplotlib.pyplot as plt

plt.title('Accuracy over 40 Epochs')
plt.plot(history.history['accuracy'], label='Train_acc')
plt.plot(history.history['val_accuracy'], label = 'Test_acc')
plt.legend(loc='lower right')
plt.grid(True)
plt.show()

plt.title('Loss over 40 Epochs')
plt.plot(history.history['loss'], label='Train_loss')
plt.plot(history.history['val_loss'], label = 'Test_loss')
plt.legend(loc='upper right')
plt.grid(True)
plt.show()
```





Plot Test Results

In [11]:

```
import matplotlib.image as mpimg
test_datagen = ImageDataGenerator(rescale=1. / 255)
eval_generator = test_datagen.flow_from_directory(TEST_DIR,target_size=IMAGE_SIZE,
                                                  batch size=1.shuffle=True.seed=42.class mode=
"binary")
eval_generator.reset()
pred = model.predict_generator(eval_generator, 18, verbose=1)
for index, probability in enumerate(pred):
    image_path = TEST_DIR + "/" +eval_generator.filenames[index]
    image = mpimg.imread(image_path)
    if image.ndim < 3:</pre>
        image = np.reshape(image,(image.shape[0],image.shape[1],1))
        image = np.concatenate([image, image, image], 2)
#
          print(image.shape)
    pixels = np.array(image)
    plt.imshow(pixels)
    print(eval_generator.filenames[index])
    if probability > 0.5:
       plt.title("%.2f" % (probability[0]*100) + "% Normal")
    else:
        plt.title("%.2f" % ((1-probability[0])*100) + "% COVID19 Pneumonia")
    plt.show()
```

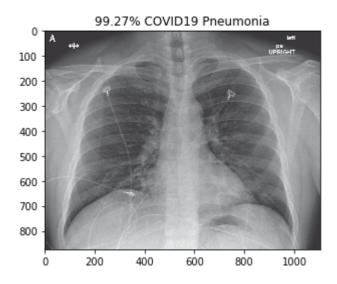
Found 18 images belonging to 2 classes.

WARNING:tensorflow:From <ipython-input-11-543347a5fba8>:7: Model.predict_generator (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.

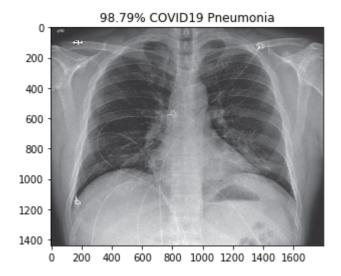
Instructions for updating:

Please use Model.predict, which supports generators.

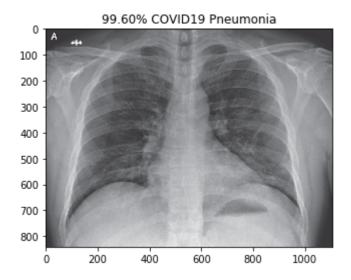
18/18 [======] - 2s 111ms/step covid\(\text{Wnejmoa} \) moa2001191_f3-PA.jpeg



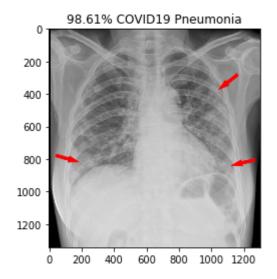
covid₩nejmoa2001191_f4.jpeg



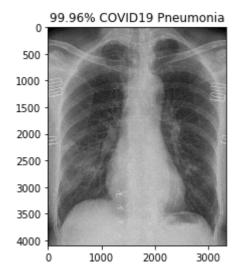
covid\mejmoa2001191_f5-PA.jpeg



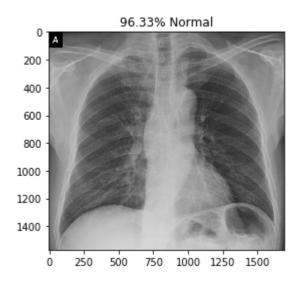
covid\madio1.2020200490.fig3.jpeg



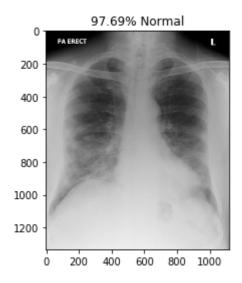
covidWryct.2020200028.fig1a.jpeg



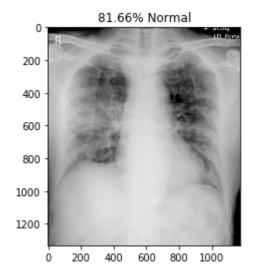
covid\footnote{\text{ryct.2020200034.fig2.jpeg}}



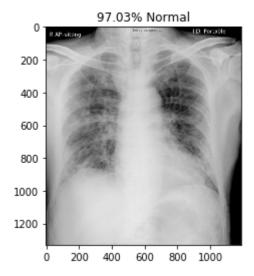
covid\footnote{wryct.2020200034.fig5-day0.jpeg



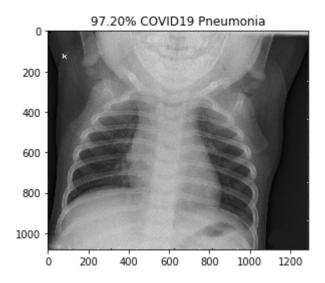
covidWryct.2020200034.fig5-day4.jpeg



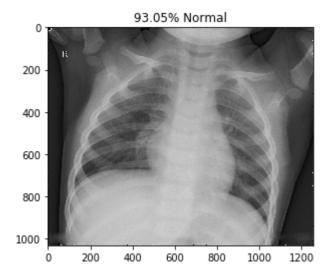
covid\footnote{wryct.2020200034.fig5-day7.jpeg



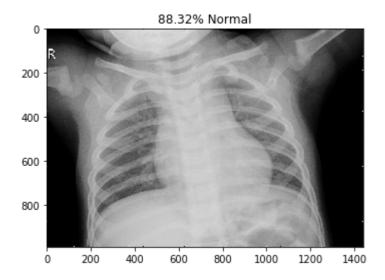
normal₩NORMAL2-IM-1385-0001.jpeg



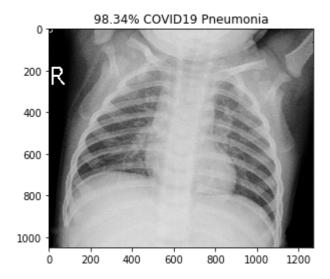
normal\WNORMAL2-IM-1396-0001.jpeg



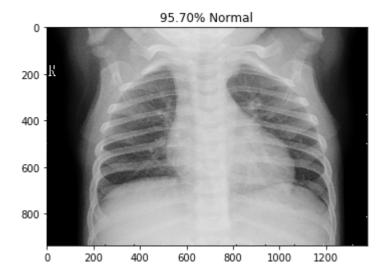
normal₩NORMAL2-IM-1400-0001.jpeg



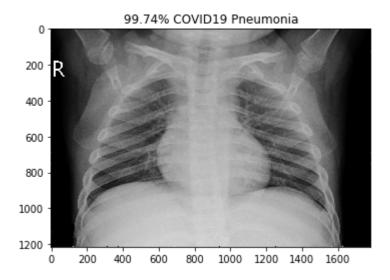
normal\NORMAL2-IM-1401-0001.jpeg



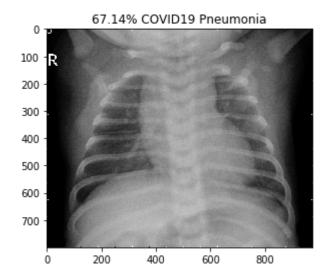
normal\NORMAL2-IM-1406-0001.jpeg



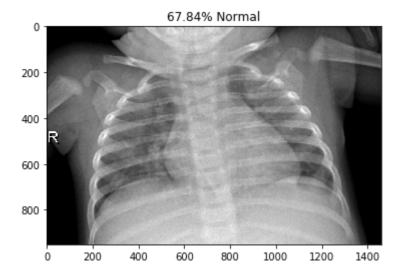
normal₩NORMAL2-IM-1412-0001.jpeg



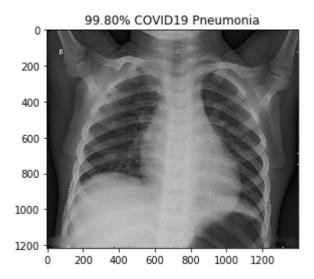
normal\NORMAL2-IM-1419-0001.jpeg



normal\NORMAL2-IM-1422-0001.jpeg



normal₩NORMAL2-IM-1423-0001.jpeg



[10 points] TSNE Plot

t-Distributed Stochastic Neighbor Embedding (t-SNE) is a widely used technique for dimensionality reduction that is particularly well suited for the visualization of high-dimensional datasets. After training is complete, extract features from a specific deep layer of your choice, use t-SNE to reduce the dimensionality of your extracted features to 2 dimensions and plot the resulting 2D features.

In [13]:

```
from sklearn.manifold import TSNE
intermediate_layer_model = models.Model(inputs=model.input,
                                        outputs=model.get_layer('dense_feature').output)
tsne_data_generator = test_datagen.flow_from_directory(DATASET_PATH,target_size=IMAGE_SIZE,
                                                  batch_size=1,shuffle=True,seed=42,class_mode=
"binary")
labels = []
num = len(tsne_data_generator)
for i in range(num):
    labels.extend(np.array(tsne_data_generator[i][1]))
#feature extraction
tsne_data_generator.reset()
features = intermediate_layer_model.predict_generator(tsne_data_generator)
#compress the dimensionality
tsne = TSNE(n_components=2)
tsne_result = tsne.fit_transform(features)
x=[i[0] for i in tsne_result]
y=[i[1] for i in tsne_result]
#plotting values
covid_x = []
covid_y = []
non\_covid\_x = []
non covid v = []
for i in range(len(labels)):
    if labels[i] == 0.0:
       non_covid_x.append(x[i])
        non_covid_y.append(y[i])
    else:
        covid_x.append(x[i])
        covid_y.append(y[i])
#plotting
plt.scatter(non_covid_x,non_covid_y, label='Normal')
plt.scatter(covid_x,covid_y, label='COVID-19')
plt.legend(loc='upper left')
plt.show()
```

Found 130 images belonging to 2 classes. 130

