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

# Task 2 [Total points: 30]

```
In [1]:
```

```
import tensorflow as tf
device_name = tf.test.gpu_device_name()
if device_name != '/device:GPU:0':
    raise SystemError('GPU device not found')
print('Found GPU at: {}'.format(device_name))
```

Found GPU at: /device:GPU:0

In [2]:

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

Mounted at /content/drive

## 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: <a href="COVID-19">COVID-19</a> (https://drive.google.com/file/d/1Y88tggpQ1Piko 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] Multi-class Classification

```
In [55]:
```

```
import os
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from keras.applications.mobilenet import MobileNet
from keras.applications.vgg16 import VGG16
from keras import models
from keras import layers
from keras import optimizers
from keras.layers import BatchNormalization,GlobalAveragePooling2D
from keras.layers.core import Flatten, Dense
import keras.backend as K
from sklearn.manifold import TSNE
os.environ['OMP NUM THREADS'] = '1'
os.environ['CUDA VISIBLE DEVICES'] = '-1'
tf. version
```

```
'2.8.0'
```

Out[55]:

#### **Load Image Data**

#### In [56]:

```
DATA_LIST = os.listdir('/content/drive/MyDrive/CS 542 Machine Learning/Challeng e/Covid_Data_GradientCrescent/all/train')
DATASET_PATH = '/content/drive/MyDrive/CS 542 Machine Learning/Challenge/Covid_Data_GradientCrescent/all/train'
TEST_DIR = '/content/drive/MyDrive/CS 542 Machine Learning/Challenge/Covid_Data_GradientCrescent/all/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 memory
NUM_EPOCHS = 100
LEARNING_RATE = 0.0001 # start off with high rate first 0.001 and experiment with reducing it gradually
```

#### **Generate Training and Validation Batches**

#### In [57]:

```
train datagen = ImageDataGenerator(rescale=1./255,rotation range=50,featurewise
center = True,
                                    featurewise std normalization = True, width sh
ift range=0.2,
                                    height shift range=0.2, shear range=0.25, zoom
range=0.1,
                                    zca whitening = True, channel shift range = 20
                                    horizontal flip = True, vertical flip = True,
                                    validation split = 0.2,fill mode='constant')
train batches = train datagen.flow from directory(DATASET PATH, target size=IMAGE
_SIZE,
                                                   shuffle=True, batch size=BATCH
SIZE,
                                                   subset = "training", seed=42,
                                                   class mode="categorical")
valid batches = train datagen.flow from directory(DATASET PATH, target size=IMAGE
_SIZE,
                                                   shuffle=True, batch size=BATCH
SIZE,
                                                   subset = "validation",
                                                   seed=42,class mode="categorica
1")
```

/usr/local/lib/python3.7/dist-packages/keras\_preprocessing/image/image\_data\_generator.py:342: UserWarning: This ImageDataGenerator specifies `zca\_whitening` which overrides setting of`featurewise\_std\_normalization`.

warnings.warn('This ImageDataGenerator specifies '

Found 216 images belonging to 4 classes. Found 54 images belonging to 4 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 [58]:

```
#VGG16
VGG16 = VGG16(weights='imagenet',
                  include top=False,
                  input_shape=(224, 224, 3))
VGG16.trainable = False
vgg16 = models.Sequential(name='VGG16')
vgg16.add(VGG16)
vgg16.add(layers.Flatten())
vgg16.add(layers.Dense(256, name='feature dense', activation='relu'))
vgg16.add(layers.Dense(4, activation='softmax'))
vqq16.summary()
vgg16.compile(optimizer=tf.keras.optimizers.Adam(learning rate=0.0005),loss='cat
egorical_crossentropy',metrics=["acc"])
```

Model: "VGG16"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_12 (Flatten)	(None, 25088)	0
feature_dense (Dense)	(None, 256)	6422784
dense_11 (Dense)	(None, 4)	1028
Total params: 21,138,500 Trainable params: 6,423,812		

Non-trainable params: 14,714,688

#### In [48]:

```
# MobileNet
mobile = MobileNet(include_top=False, weights='imagenet', input_shape=(224, 224, 3))

mobilenet = models.Sequential(name='MobileNet')
mobilenet.add(mobile)
mobilenet.add(GlobalAveragePooling2D())
mobilenet.add(layers.Flatten())
mobilenet.add(layers.Dense(512, activation='relu'))
mobilenet.add(layers.Dense(256, name='feature_dense'))

mobilenet.add(layers.Dense(4, activation='softmax'))

mobilenet.summary()
mobilenet.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=0.0001),loss='categorical_crossentropy',metrics=["acc"])
```

#### Model: "MobileNet"

Layer (type)	Output Shape	Param #
mobilenet_1.00_224 (Functio nal)	(None, 7, 7, 1024)	3228864
<pre>global_average_pooling2d_5 (GlobalAveragePooling2D)</pre>	(None, 1024)	0
flatten_11 (Flatten)	(None, 1024)	0
dense_9 (Dense)	(None, 512)	524800
feature_dense (Dense)	(None, 256)	131328
dense_10 (Dense)	(None, 4)	1028
Total params: 3,886,020 Trainable params: 3,864,132 Non-trainable params: 21,888		======

#### [5 points] Train Model

#### In [59]:

```
#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
```

22

6

#### In [60]:

/usr/local/lib/python3.7/dist-packages/keras\_preprocessing/image/image\_data\_generator.py:720: UserWarning: This ImageDataGenerator specifies `featurewise\_center`, but it hasn't been fit on any training data. Fit it first by calling `.fit(numpy\_data)`.

warnings.warn('This ImageDataGenerator specifies '/usr/local/lib/python3.7/dist-packages/keras\_preprocessing/image/image\_data\_generator.py:739: UserWarning: This ImageDataGenerator specifies `zca\_whitening`, but it hasn't been fit on any training data. F it first by calling `.fit(numpy data)`.

warnings.warn('This ImageDataGenerator specifies '

```
Epoch 1/100
4 - acc: 0.3301 - val loss: 1.2102 - val acc: 0.4800
Epoch 2/100
21/21 [============== ] - 8s 390ms/step - loss: 1.213
7 - acc: 0.4757 - val loss: 0.9627 - val acc: 0.6400
Epoch 3/100
2 - acc: 0.4951 - val loss: 0.8244 - val acc: 0.7200
Epoch 4/100
21/21 [============= ] - 8s 390ms/step - loss: 1.027
6 - acc: 0.5728 - val loss: 0.9864 - val acc: 0.5200
Epoch 5/100
2 - acc: 0.6165 - val loss: 0.8828 - val acc: 0.6600
Epoch 6/100
4 - acc: 0.5680 - val loss: 0.9325 - val acc: 0.5200
Epoch 7/100
3 - acc: 0.5680 - val loss: 0.8394 - val acc: 0.5800
Epoch 8/100
21/21 [============ ] - 8s 377ms/step - loss: 0.867
2 - acc: 0.6311 - val loss: 0.8822 - val acc: 0.6200
Epoch 9/100
21/21 [============ ] - 8s 395ms/step - loss: 0.901
2 - acc: 0.6505 - val_loss: 0.7763 - val acc: 0.6200
Epoch 10/100
21/21 [============ ] - 8s 387ms/step - loss: 0.871
1 - acc: 0.6505 - val_loss: 0.7291 - val_acc: 0.6200
Epoch 11/100
1 - acc: 0.6650 - val loss: 0.8383 - val acc: 0.6200
Epoch 12/100
2 - acc: 0.6942 - val loss: 0.9880 - val acc: 0.5600
Epoch 13/100
21/21 [============= ] - 8s 395ms/step - loss: 0.795
9 - acc: 0.6505 - val loss: 0.7630 - val acc: 0.6400
Epoch 14/100
7 - acc: 0.6602 - val loss: 0.6895 - val acc: 0.6600
Epoch 15/100
6 - acc: 0.7039 - val loss: 0.6999 - val acc: 0.7200
Epoch 16/100
21/21 [============= ] - 8s 391ms/step - loss: 0.752
8 - acc: 0.7039 - val_loss: 0.7791 - val_acc: 0.6400
Epoch 17/100
9 - acc: 0.6553 - val loss: 0.7346 - val acc: 0.7000
Epoch 18/100
4 - acc: 0.6699 - val_loss: 0.9277 - val_acc: 0.5200
Epoch 19/100
21/21 [============ ] - 8s 385ms/step - loss: 0.781
7 - acc: 0.7087 - val loss: 0.7715 - val acc: 0.7000
Epoch 20/100
21/21 [============= ] - 8s 391ms/step - loss: 0.679
5 - acc: 0.7087 - val_loss: 0.9009 - val_acc: 0.6200
Epoch 21/100
```

```
21/21 [============== ] - 8s 387ms/step - loss: 0.796
5 - acc: 0.6602 - val loss: 0.7138 - val acc: 0.6600
Epoch 22/100
3 - acc: 0.7233 - val loss: 0.8226 - val acc: 0.5800
Epoch 23/100
21/21 [============ ] - 8s 384ms/step - loss: 0.694
2 - acc: 0.6942 - val loss: 0.8100 - val acc: 0.6400
Epoch 24/100
21/21 [============= ] - 8s 391ms/step - loss: 0.708
8 - acc: 0.7184 - val loss: 0.7186 - val acc: 0.6400
Epoch 25/100
4 - acc: 0.7184 - val loss: 0.7841 - val acc: 0.6800
Epoch 26/100
21/21 [============ ] - 8s 394ms/step - loss: 0.693
6 - acc: 0.6845 - val loss: 0.7016 - val acc: 0.7200
Epoch 27/100
21/21 [============ ] - 8s 400ms/step - loss: 0.669
6 - acc: 0.6942 - val loss: 0.6036 - val acc: 0.7200
Epoch 28/100
21/21 [============ ] - 8s 393ms/step - loss: 0.673
2 - acc: 0.7427 - val loss: 0.7567 - val acc: 0.6000
Epoch 29/100
21/21 [============= ] - 8s 388ms/step - loss: 0.626
0 - acc: 0.7670 - val loss: 0.7274 - val acc: 0.6600
Epoch 30/100
21/21 [============ ] - 8s 392ms/step - loss: 0.705
4 - acc: 0.6893 - val_loss: 0.9401 - val_acc: 0.5600
Epoch 31/100
4 - acc: 0.6845 - val loss: 0.5446 - val acc: 0.7000
Epoch 32/100
21/21 [============== ] - 8s 390ms/step - loss: 0.592
9 - acc: 0.7427 - val_loss: 0.7379 - val acc: 0.6800
Epoch 33/100
21/21 [============ ] - 8s 392ms/step - loss: 0.635
0 - acc: 0.7524 - val loss: 0.9842 - val acc: 0.5200
Epoch 34/100
8 - acc: 0.7476 - val loss: 0.6998 - val acc: 0.6600
Epoch 35/100
0 - acc: 0.7670 - val_loss: 0.8025 - val_acc: 0.6400
Epoch 36/100
9 - acc: 0.7190 - val loss: 0.8865 - val acc: 0.6600
Epoch 37/100
21/21 [============ ] - 8s 390ms/step - loss: 0.650
2 - acc: 0.6796 - val_loss: 0.7115 - val_acc: 0.6600
Epoch 38/100
21/21 [============ ] - 8s 395ms/step - loss: 0.631
3 - acc: 0.7184 - val_loss: 0.9736 - val_acc: 0.6400
Epoch 39/100
3 - acc: 0.7476 - val_loss: 0.7085 - val_acc: 0.6600
Epoch 40/100
21/21 [============ ] - 8s 393ms/step - loss: 0.626
0 - acc: 0.7379 - val_loss: 0.7374 - val_acc: 0.7000
Epoch 41/100
```

```
5 - acc: 0.7718 - val loss: 0.7911 - val acc: 0.6400
Epoch 42/100
21/21 [============= ] - 8s 385ms/step - loss: 0.660
2 - acc: 0.7233 - val loss: 0.6949 - val acc: 0.7000
Epoch 43/100
8 - acc: 0.7184 - val loss: 0.6801 - val acc: 0.6400
Epoch 44/100
21/21 [============ ] - 8s 392ms/step - loss: 0.593
7 - acc: 0.7330 - val loss: 0.5773 - val acc: 0.6800
Epoch 45/100
21/21 [============ ] - 8s 391ms/step - loss: 0.587
9 - acc: 0.7864 - val loss: 0.9509 - val acc: 0.5400
Epoch 46/100
0 - acc: 0.7330 - val loss: 0.9701 - val acc: 0.5600
Epoch 47/100
6 - acc: 0.7476 - val loss: 0.6868 - val acc: 0.7400
Epoch 48/100
21/21 [============ ] - 8s 397ms/step - loss: 0.626
0 - acc: 0.7379 - val loss: 0.8409 - val acc: 0.6000
Epoch 49/100
6 - acc: 0.7233 - val loss: 0.8180 - val acc: 0.6400
Epoch 50/100
21/21 [============= ] - 8s 395ms/step - loss: 0.596
4 - acc: 0.7524 - val loss: 0.7686 - val acc: 0.6400
Epoch 51/100
21/21 [============= ] - 8s 399ms/step - loss: 0.584
3 - acc: 0.7767 - val loss: 0.7334 - val acc: 0.7000
Epoch 52/100
21/21 [============ ] - 8s 388ms/step - loss: 0.602
4 - acc: 0.7524 - val_loss: 0.6984 - val_acc: 0.7200
Epoch 53/100
21/21 [============= ] - 8s 388ms/step - loss: 0.554
6 - acc: 0.7718 - val loss: 0.6975 - val acc: 0.7400
Epoch 54/100
21/21 [============ ] - 8s 391ms/step - loss: 0.532
2 - acc: 0.7718 - val loss: 0.7302 - val acc: 0.7000
Epoch 55/100
7 - acc: 0.8107 - val_loss: 0.6479 - val_acc: 0.6800
Epoch 56/100
21/21 [============== ] - 8s 392ms/step - loss: 0.553
0 - acc: 0.7718 - val loss: 0.6737 - val acc: 0.6200
Epoch 57/100
21/21 [============== ] - 8s 390ms/step - loss: 0.565
3 - acc: 0.7670 - val loss: 0.7575 - val acc: 0.6200
Epoch 58/100
9 - acc: 0.7670 - val loss: 0.6804 - val acc: 0.7200
Epoch 59/100
21/21 [============= ] - 8s 388ms/step - loss: 0.551
5 - acc: 0.7379 - val_loss: 0.7621 - val_acc: 0.7000
Epoch 60/100
7 - acc: 0.7718 - val loss: 0.7091 - val acc: 0.6000
Epoch 61/100
4 - acc: 0.7282 - val_loss: 0.7467 - val_acc: 0.6800
```

```
Epoch 62/100
3 - acc: 0.7767 - val loss: 0.5992 - val acc: 0.6800
Epoch 63/100
21/21 [============= ] - 8s 396ms/step - loss: 0.560
4 - acc: 0.7767 - val loss: 0.6629 - val acc: 0.6600
Epoch 64/100
21/21 [============= ] - 8s 389ms/step - loss: 0.501
4 - acc: 0.7913 - val loss: 0.7208 - val acc: 0.6600
Epoch 65/100
3 - acc: 0.7476 - val loss: 0.7954 - val acc: 0.6800
Epoch 66/100
21/21 [============ ] - 8s 393ms/step - loss: 0.508
9 - acc: 0.8048 - val loss: 0.7439 - val acc: 0.7200
Epoch 67/100
7 - acc: 0.7282 - val_loss: 0.7253 - val_acc: 0.6400
Epoch 68/100
4 - acc: 0.7573 - val loss: 0.8718 - val acc: 0.4800
Epoch 69/100
21/21 [============ ] - 8s 390ms/step - loss: 0.541
9 - acc: 0.7670 - val loss: 0.8435 - val acc: 0.5400
Epoch 70/100
4 - acc: 0.8252 - val loss: 0.6871 - val acc: 0.6600
Epoch 71/100
9 - acc: 0.7476 - val loss: 0.5000 - val acc: 0.7400
Epoch 72/100
21/21 [============ ] - 8s 395ms/step - loss: 0.587
4 - acc: 0.7767 - val_loss: 0.6767 - val_acc: 0.7000
Epoch 73/100
5 - acc: 0.7330 - val loss: 0.9293 - val acc: 0.5400
Epoch 74/100
9 - acc: 0.7427 - val loss: 0.6488 - val acc: 0.7000
Epoch 75/100
21/21 [============ ] - 8s 393ms/step - loss: 0.551
2 - acc: 0.7961 - val loss: 0.8361 - val acc: 0.6600
Epoch 76/100
6 - acc: 0.7816 - val loss: 0.5689 - val acc: 0.6800
Epoch 77/100
5 - acc: 0.8252 - val_loss: 0.5762 - val_acc: 0.7200
Epoch 78/100
5 - acc: 0.7718 - val_loss: 0.6611 - val_acc: 0.6600
Epoch 79/100
9 - acc: 0.7524 - val loss: 0.6583 - val acc: 0.7200
Epoch 80/100
21/21 [============= ] - 8s 391ms/step - loss: 0.588
6 - acc: 0.7379 - val_loss: 0.8412 - val_acc: 0.6600
Epoch 81/100
21/21 [============ ] - 8s 390ms/step - loss: 0.475
0 - acc: 0.8155 - val loss: 0.8145 - val acc: 0.6600
Epoch 82/100
```

```
21/21 [============ ] - 8s 386ms/step - loss: 0.474
5 - acc: 0.7961 - val loss: 0.6249 - val acc: 0.6600
Epoch 83/100
21/21 [============= ] - 8s 395ms/step - loss: 0.463
7 - acc: 0.8010 - val loss: 0.8369 - val acc: 0.6400
Epoch 84/100
21/21 [============ ] - 8s 385ms/step - loss: 0.520
1 - acc: 0.7718 - val loss: 0.6279 - val acc: 0.6400
Epoch 85/100
21/21 [============= ] - 8s 392ms/step - loss: 0.473
9 - acc: 0.7952 - val_loss: 0.7062 - val acc: 0.6000
Epoch 86/100
7 - acc: 0.7816 - val_loss: 1.0830 - val acc: 0.5400
Epoch 87/100
21/21 [============= ] - 8s 394ms/step - loss: 0.545
1 - acc: 0.7379 - val loss: 0.9031 - val acc: 0.6200
Epoch 88/100
21/21 [============ ] - 8s 386ms/step - loss: 0.450
9 - acc: 0.7864 - val loss: 0.5349 - val acc: 0.7000
Epoch 89/100
5 - acc: 0.7857 - val loss: 0.5987 - val acc: 0.6800
Epoch 90/100
21/21 [============ ] - 8s 395ms/step - loss: 0.599
6 - acc: 0.7621 - val loss: 0.6162 - val acc: 0.6800
Epoch 91/100
8 - acc: 0.7524 - val_loss: 0.6943 - val acc: 0.6800
Epoch 92/100
9 - acc: 0.7282 - val loss: 0.7491 - val acc: 0.6600
Epoch 93/100
21/21 [============ ] - 8s 381ms/step - loss: 0.478
2 - acc: 0.7816 - val loss: 0.6965 - val acc: 0.6800
Epoch 94/100
8 - acc: 0.8107 - val loss: 0.6054 - val acc: 0.6800
Epoch 95/100
21/21 [============ ] - 8s 389ms/step - loss: 0.572
7 - acc: 0.7476 - val loss: 1.0214 - val acc: 0.6000
Epoch 96/100
21/21 [============= ] - 9s 421ms/step - loss: 0.574
6 - acc: 0.7670 - val loss: 0.6953 - val acc: 0.6600
Epoch 97/100
21/21 [============ ] - 8s 383ms/step - loss: 0.462
0 - acc: 0.8398 - val loss: 0.5763 - val acc: 0.7000
Epoch 98/100
7 - acc: 0.8010 - val_loss: 0.6667 - val_acc: 0.7000
Epoch 99/100
21/21 [============= ] - 8s 400ms/step - loss: 0.508
5 - acc: 0.8010 - val loss: 0.6881 - val acc: 0.6600
Epoch 100/100
21/21 [============ ] - 8s 398ms/step - loss: 0.481
7 - acc: 0.7864 - val_loss: 0.5624 - val_acc: 0.7200
```

#### In [26]:

/usr/local/lib/python3.7/dist-packages/keras\_preprocessing/image/image\_data\_generator.py:720: UserWarning: This ImageDataGenerator specifies `featurewise\_center`, but it hasn't been fit on any training data. Fit it first by calling `.fit(numpy\_data)`.

warnings.warn('This ImageDataGenerator specifies '/usr/local/lib/python3.7/dist-packages/keras\_preprocessing/image/image\_data\_generator.py:739: UserWarning: This ImageDataGenerator specifies `zca\_whitening`, but it hasn't been fit on any training data. F it first by calling `.fit(numpy data)`.

warnings.warn('This ImageDataGenerator specifies '

```
Epoch 1/100
23 - acc: 0.4417 - val loss: 2.4747 - val acc: 0.2600
Epoch 2/100
21/21 [============== ] - 8s 401ms/step - loss: 0.705
6 - acc: 0.7136 - val loss: 2.9546 - val acc: 0.2000
Epoch 3/100
9 - acc: 0.7282 - val loss: 2.9787 - val acc: 0.3000
Epoch 4/100
21/21 [============= ] - 9s 405ms/step - loss: 0.561
6 - acc: 0.7913 - val loss: 3.0249 - val acc: 0.2400
Epoch 5/100
6 - acc: 0.7767 - val_loss: 2.3122 - val_acc: 0.3000
Epoch 6/100
3 - acc: 0.7621 - val loss: 2.3439 - val acc: 0.3000
Epoch 7/100
21/21 [============ ] - 8s 399ms/step - loss: 0.442
2 - acc: 0.8204 - val loss: 1.6809 - val acc: 0.4200
Epoch 8/100
21/21 [============ ] - 8s 399ms/step - loss: 0.494
4 - acc: 0.7816 - val loss: 2.5028 - val acc: 0.3600
Epoch 9/100
21/21 [============= ] - 8s 402ms/step - loss: 0.438
0 - acc: 0.8350 - val_loss: 1.7472 - val_acc: 0.3600
Epoch 10/100
21/21 [============ ] - 10s 460ms/step - loss: 0.37
02 - acc: 0.8738 - val_loss: 2.2474 - val_acc: 0.4200
Epoch 11/100
1 - acc: 0.8495 - val loss: 2.3004 - val acc: 0.3600
Epoch 12/100
6 - acc: 0.8204 - val loss: 1.5798 - val acc: 0.4400
Epoch 13/100
21/21 [============= ] - 8s 394ms/step - loss: 0.374
7 - acc: 0.8495 - val loss: 1.2463 - val acc: 0.5200
Epoch 14/100
4 - acc: 0.8398 - val loss: 1.6831 - val acc: 0.4400
Epoch 15/100
0 - acc: 0.8252 - val loss: 2.2206 - val acc: 0.3200
Epoch 16/100
21/21 [============= ] - 8s 401ms/step - loss: 0.294
2 - acc: 0.8786 - val_loss: 2.0693 - val_acc: 0.3600
Epoch 17/100
0 - acc: 0.8301 - val loss: 1.6325 - val acc: 0.4400
Epoch 18/100
21/21 [=============== ] - 8s 404ms/step - loss: 0.306
2 - acc: 0.8592 - val_loss: 1.5158 - val_acc: 0.4200
Epoch 19/100
21/21 [============ ] - 8s 399ms/step - loss: 0.285
2 - acc: 0.8738 - val loss: 0.8467 - val acc: 0.5600
Epoch 20/100
9 - acc: 0.8689 - val_loss: 1.1628 - val_acc: 0.5600
Epoch 21/100
```

```
21/21 [============ ] - 8s 394ms/step - loss: 0.235
8 - acc: 0.9078 - val loss: 1.6709 - val acc: 0.4200
Epoch 22/100
21/21 [============ ] - 8s 403ms/step - loss: 0.259
1 - acc: 0.9029 - val loss: 1.6148 - val acc: 0.5000
Epoch 23/100
21/21 [============ ] - 8s 402ms/step - loss: 0.273
4 - acc: 0.9029 - val loss: 1.2389 - val acc: 0.6000
Epoch 24/100
6 - acc: 0.8738 - val_loss: 0.9777 - val acc: 0.6200
Epoch 25/100
2 - acc: 0.9126 - val loss: 0.9395 - val acc: 0.6600
Epoch 26/100
3 - acc: 0.8883 - val loss: 0.6061 - val acc: 0.7200
Epoch 27/100
21/21 [============= ] - 8s 404ms/step - loss: 0.248
6 - acc: 0.8932 - val loss: 0.6149 - val acc: 0.7800
Epoch 28/100
21/21 [============= ] - 8s 403ms/step - loss: 0.164
2 - acc: 0.9320 - val loss: 0.4695 - val acc: 0.7800
Epoch 29/100
21/21 [============= ] - 8s 403ms/step - loss: 0.164
0 - acc: 0.9466 - val loss: 0.5033 - val acc: 0.8200
Epoch 30/100
21/21 [============ ] - 8s 395ms/step - loss: 0.254
6 - acc: 0.9126 - val_loss: 0.7416 - val_acc: 0.6600
Epoch 31/100
0 - acc: 0.8592 - val loss: 0.5365 - val acc: 0.7600
Epoch 32/100
21/21 [============== ] - 8s 397ms/step - loss: 0.169
2 - acc: 0.9078 - val loss: 0.9456 - val acc: 0.6800
Epoch 33/100
5 - acc: 0.9369 - val loss: 0.7911 - val acc: 0.7200
Epoch 34/100
5 - acc: 0.9854 - val loss: 0.5854 - val acc: 0.7800
Epoch 35/100
5 - acc: 0.9417 - val_loss: 1.0097 - val_acc: 0.7800
Epoch 36/100
1 - acc: 0.9143 - val loss: 0.8347 - val acc: 0.7800
Epoch 37/100
21/21 [============ ] - 8s 398ms/step - loss: 0.170
7 - acc: 0.9272 - val_loss: 0.8581 - val_acc: 0.6600
Epoch 38/100
21/21 [============ ] - 8s 398ms/step - loss: 0.156
2 - acc: 0.9320 - val_loss: 0.5998 - val_acc: 0.8000
Epoch 39/100
0 - acc: 0.9320 - val_loss: 0.7261 - val_acc: 0.7200
Epoch 40/100
21/21 [============ ] - 8s 399ms/step - loss: 0.127
1 - acc: 0.9417 - val_loss: 0.6601 - val_acc: 0.7600
Epoch 41/100
```

```
3 - acc: 0.9714 - val loss: 0.4811 - val acc: 0.8000
Epoch 42/100
21/21 [============= ] - 8s 397ms/step - loss: 0.068
2 - acc: 0.9757 - val loss: 0.7911 - val acc: 0.7400
Epoch 43/100
3 - acc: 0.9709 - val loss: 0.6227 - val acc: 0.8000
Epoch 44/100
21/21 [============ ] - 8s 393ms/step - loss: 0.109
1 - acc: 0.9476 - val loss: 1.1026 - val acc: 0.7200
Epoch 45/100
21/21 [============ ] - 8s 400ms/step - loss: 0.182
2 - acc: 0.9320 - val loss: 1.0055 - val acc: 0.6600
Epoch 46/100
6 - acc: 0.9320 - val loss: 0.7135 - val acc: 0.8000
Epoch 47/100
7 - acc: 0.9272 - val loss: 0.7175 - val acc: 0.7800
Epoch 48/100
21/21 [============= ] - 8s 401ms/step - loss: 0.118
9 - acc: 0.9563 - val loss: 0.6342 - val acc: 0.8000
Epoch 49/100
6 - acc: 0.9612 - val loss: 0.7755 - val acc: 0.7800
Epoch 50/100
21/21 [============= ] - 8s 408ms/step - loss: 0.166
5 - acc: 0.9272 - val loss: 0.6273 - val acc: 0.7800
Epoch 51/100
21/21 [============= ] - 8s 404ms/step - loss: 0.228
7 - acc: 0.9095 - val loss: 0.4424 - val acc: 0.8400
Epoch 52/100
21/21 [============ ] - 8s 398ms/step - loss: 0.162
6 - acc: 0.9320 - val_loss: 0.8378 - val_acc: 0.7400
Epoch 53/100
21/21 [============ ] - 8s 391ms/step - loss: 0.101
7 - acc: 0.9660 - val loss: 0.7555 - val acc: 0.8000
Epoch 54/100
21/21 [============ ] - 8s 405ms/step - loss: 0.125
2 - acc: 0.9515 - val loss: 1.1830 - val acc: 0.7000
Epoch 55/100
21/21 [============= ] - 8s 406ms/step - loss: 0.159
5 - acc: 0.9429 - val_loss: 0.8862 - val_acc: 0.7600
Epoch 56/100
21/21 [============== ] - 8s 403ms/step - loss: 0.128
2 - acc: 0.9417 - val loss: 1.0566 - val acc: 0.7800
Epoch 57/100
21/21 [============== ] - 8s 398ms/step - loss: 0.165
3 - acc: 0.9466 - val loss: 0.8299 - val acc: 0.6600
Epoch 58/100
3 - acc: 0.9320 - val loss: 1.2668 - val acc: 0.6000
Epoch 59/100
21/21 [============= ] - 8s 400ms/step - loss: 0.162
1 - acc: 0.9417 - val_loss: 0.6520 - val_acc: 0.7000
Epoch 60/100
9 - acc: 0.9515 - val loss: 0.5125 - val acc: 0.8200
Epoch 61/100
0 - acc: 0.9466 - val_loss: 0.6334 - val_acc: 0.7200
```

```
Epoch 62/100
3 - acc: 0.9524 - val loss: 1.3792 - val acc: 0.6400
Epoch 63/100
9 - acc: 0.9369 - val loss: 1.0849 - val acc: 0.7400
Epoch 64/100
21/21 [============= ] - 8s 401ms/step - loss: 0.122
9 - acc: 0.9466 - val loss: 1.2191 - val acc: 0.7000
Epoch 65/100
1 - acc: 0.9563 - val loss: 0.6663 - val acc: 0.7400
Epoch 66/100
21/21 [============ ] - 8s 402ms/step - loss: 0.098
3 - acc: 0.9709 - val loss: 0.8634 - val acc: 0.8200
Epoch 67/100
4 - acc: 0.9417 - val_loss: 0.5059 - val_acc: 0.8400
Epoch 68/100
2 - acc: 0.9806 - val loss: 0.6825 - val acc: 0.7200
Epoch 69/100
21/21 [=========== ] - 8s 402ms/step - loss: 0.056
0 - acc: 0.9709 - val loss: 0.5376 - val acc: 0.8000
Epoch 70/100
21/21 [============= ] - 8s 402ms/step - loss: 0.064
3 - acc: 0.9806 - val loss: 0.7946 - val acc: 0.7200
Epoch 71/100
2 - acc: 0.9757 - val loss: 0.6210 - val acc: 0.8200
Epoch 72/100
21/21 [============ ] - 8s 399ms/step - loss: 0.142
9 - acc: 0.9320 - val_loss: 0.9032 - val_acc: 0.7200
Epoch 73/100
2 - acc: 0.9709 - val loss: 0.9703 - val acc: 0.7800
Epoch 74/100
8 - acc: 0.9660 - val loss: 0.7244 - val acc: 0.7800
Epoch 75/100
21/21 [============ ] - 8s 401ms/step - loss: 0.091
2 - acc: 0.9709 - val loss: 0.9685 - val acc: 0.7600
Epoch 76/100
5 - acc: 0.9515 - val loss: 0.7558 - val acc: 0.7600
Epoch 77/100
21/21 [============== ] - 8s 404ms/step - loss: 0.099
8 - acc: 0.9660 - val_loss: 1.2212 - val_acc: 0.7400
Epoch 78/100
1 - acc: 0.9466 - val_loss: 1.1664 - val_acc: 0.7400
Epoch 79/100
2 - acc: 0.9563 - val loss: 1.1171 - val acc: 0.7000
Epoch 80/100
21/21 [============= ] - 8s 405ms/step - loss: 0.097
1 - acc: 0.9563 - val_loss: 0.8948 - val_acc: 0.7200
Epoch 81/100
21/21 [============ ] - 8s 404ms/step - loss: 0.083
7 - acc: 0.9757 - val loss: 1.1408 - val acc: 0.7200
Epoch 82/100
```

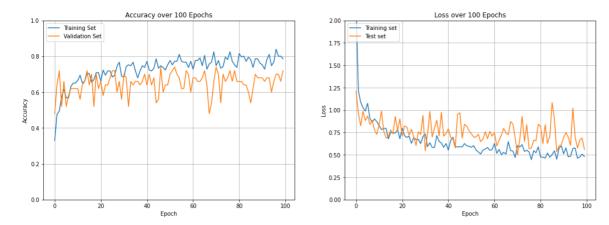
```
21/21 [============== ] - 8s 402ms/step - loss: 0.079
1 - acc: 0.9762 - val loss: 0.9764 - val acc: 0.7400
Epoch 83/100
6 - acc: 0.9709 - val loss: 0.7694 - val acc: 0.8000
Epoch 84/100
21/21 [============ ] - 8s 399ms/step - loss: 0.085
2 - acc: 0.9757 - val_loss: 0.7009 - val_acc: 0.8000
Epoch 85/100
21/21 [============= ] - 8s 392ms/step - loss: 0.040
8 - acc: 0.9903 - val_loss: 0.6142 - val acc: 0.8600
Epoch 86/100
4 - acc: 0.9709 - val loss: 0.8260 - val acc: 0.7600
Epoch 87/100
21/21 [============= ] - 8s 399ms/step - loss: 0.123
5 - acc: 0.9612 - val loss: 1.4495 - val acc: 0.6800
Epoch 88/100
21/21 [============= ] - 8s 401ms/step - loss: 0.150
0 - acc: 0.9515 - val loss: 0.9965 - val acc: 0.7400
Epoch 89/100
21/21 [============ ] - 8s 398ms/step - loss: 0.164
4 - acc: 0.9612 - val loss: 1.1376 - val acc: 0.7400
Epoch 90/100
21/21 [============ ] - 8s 392ms/step - loss: 0.088
5 - acc: 0.9660 - val loss: 0.9903 - val acc: 0.7200
Epoch 91/100
21/21 [============ ] - 8s 398ms/step - loss: 0.071
4 - acc: 0.9806 - val_loss: 0.8294 - val_acc: 0.7600
Epoch 92/100
1 - acc: 0.9563 - val loss: 0.8390 - val acc: 0.8000
Epoch 93/100
21/21 [============== ] - 8s 402ms/step - loss: 0.075
9 - acc: 0.9806 - val loss: 0.7582 - val acc: 0.8000
Epoch 94/100
6 - acc: 0.9806 - val loss: 0.9262 - val acc: 0.8400
Epoch 95/100
6 - acc: 0.9951 - val loss: 1.3100 - val acc: 0.8000
Epoch 96/100
0 - acc: 0.9806 - val loss: 0.7656 - val acc: 0.8400
Epoch 97/100
5 - acc: 0.9854 - val loss: 0.5082 - val acc: 0.8200
Epoch 98/100
21/21 [============ ] - 8s 399ms/step - loss: 0.054
4 - acc: 0.9854 - val_loss: 0.5319 - val_acc: 0.8400
Epoch 99/100
21/21 [============ ] - 8s 406ms/step - loss: 0.053
4 - acc: 0.9757 - val loss: 0.8367 - val acc: 0.7800
Epoch 100/100
21/21 [============ ] - 8s 399ms/step - loss: 0.137
7 - acc: 0.9466 - val_loss: 1.0237 - val_acc: 0.7800
```

## [5 points] Plot Accuracy and Loss During Training

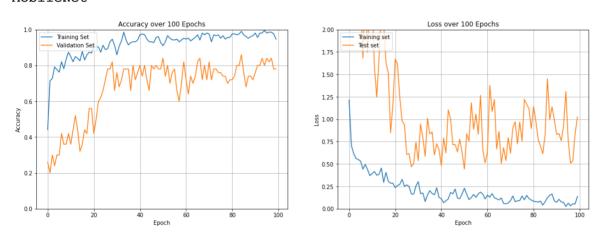
In [61]:

```
import matplotlib.pyplot as plt
%matplotlib inline
# #Accuracy
# plt.subplot(1,2,1)
def plot acc loss(result):
 plt.figure(figsize=(18,6))
 plt.subplot(1,2,1)
 plt.plot(result.history["acc"])
 plt.plot(result.history["val_acc"])
 plt.title("Accuracy over 100 Epochs")
 plt.ylim((0, 1))
 plt.xlabel("Epoch")
 plt.ylabel("Accuracy")
 plt.legend(["Training Set","Validation Set"],loc="upper left")
 plt.grid(True)
 # Loss
 plt.subplot(1,2,2)
 plt.ylim((0,2))
 plt.plot(result.history["loss"])
 plt.plot(result.history["val loss"])
 plt.title("Loss over 100 Epochs")
 plt.xlabel("Epoch")
 plt.ylabel("Loss")
 plt.legend(["Training set", "Test set"], loc="upper left")
 plt.grid(True)
 plt.show()
print("VGG16")
plot acc loss(vgg16 model)
print("MobileNet")
plot acc loss(mobilenet model)
```

## VGG16



## MobileNet



## **Testing Model**

```
In [51]:
```

```
test datagen = ImageDataGenerator(rescale=1. / 255)
eval generator = test datagen.flow from directory(TEST DIR, target size=IMAGE SIZ
Ε,
                                                   batch size=1, shuffle=True, seed
=42,class mode="categorical")
eval generator.reset()
print(len(eval generator))
def Test(model):
  x = model.evaluate generator(eval generator, steps = np.ceil(len(eval generator
)),
                            use multiprocessing = False, verbose = 1, workers=1)
  return x
```

Found 36 images belonging to 4 classes. 36

```
In [62]:
```

- acc: 0.7500

Test accuracy: 0.75

```
print('vgg16')
print('Test loss:' , Test(vgg16)[0])
print('Test accuracy:',Test(vgg16)[1])
vgg16
 5/36 [===>.....] - ETA: 0s - loss: 1.9903 - ac
c: 0.2000
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:9: User
Warning: `Model.evaluate_generator` is deprecated and will be remove
d in a future version. Please use `Model.evaluate`, which supports g
enerators.
  if name == ' main ':
36/36 [============ ] - 1s 25ms/step - loss: 0.7133
- acc: 0.7500
Test loss: 0.7132667899131775
```

36/36 [============== ] - 1s 24ms/step - loss: 0.7133

file:///Users/zj/Downloads/task2\_template.html

```
In [28]:
```

```
print('mobilenet')
print('Test loss:' , Test(mobilenet)[0])
print('Test accuracy:',Test(mobilenet)[1])
mobilenet
 4/36 [==>.....] - ETA: 0s - loss: 0.7772 - ac
c: 0.7500
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:9: User
Warning: `Model.evaluate_generator` is deprecated and will be remove
d in a future version. Please use `Model.evaluate`, which supports g
enerators.
  if __name__ == '__main__':
36/36 [============= ] - 1s 27ms/step - loss: 1.2211
- acc: 0.8333
Test loss: 1.2210861444473267
36/36 [============= ] - 1s 26ms/step - loss: 1.2211
- acc: 0.8333
Test accuracy: 0.8333333134651184
```

# [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 [66]:

```
intermediate layer model = models.Model(inputs=vgg16.input,
                                         outputs=vgg16.get layer('feature dense')
.output)
tsne eval generator = test datagen.flow from directory(DATASET PATH, target size=
IMAGE SIZE,
                                                   batch size=1, shuffle=False, see
d=42,class mode="categorical")
tsne eval generator.reset()
labels = tsne eval generator.classes
X = TSNE().fit transform(intermediate layer model.predict generator(tsne eval ge
nerator, verbose=1))
classes = ["COVID-19", "Normal", "pneumonia bac", "pneumonia vir"]
plt.figure(figsize=(18,6))
plt.subplot(1,2,1)
plt.title('VGG16')
for i in range(4):
    cluster = X[np.where(labels == i)]
    plt.scatter(cluster[:, 0], cluster[:, 1], label = classes[i])
plt.legend()
intermediate layer model = models.Model(inputs=mobilenet.input,
                                         outputs=mobilenet.get layer('feature den
se').output)
tsne eval generator = test datagen.flow from directory(DATASET PATH, target size=
IMAGE SIZE,
                                                   batch size=1, shuffle=False, see
d=42,class mode="categorical")
labels = tsne eval generator.classes
X = TSNE().fit_transform(intermediate_layer_model. predict_generator(tsne_eval_g
enerator, verbose=1))
classes = ["COVID-19", "Normal", "pneumonia bac", "pneumonia vir"]
plt.subplot(1,2,2)
plt.title('MobileNet')
for i in range(4):
    cluster = X[np.where(labels == i)]
    plt.scatter(cluster[:, 0], cluster[:, 1], label = classes[i])
plt.legend()
plt.show()
```

Found 270 images belonging to 4 classes.

/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:9: User Warning: `Model.predict\_generator` is deprecated and will be removed in a future version. Please use `Model.predict`, which supports gene rators.

/usr/local/lib/python3.7/dist-packages/sklearn/manifold/\_t\_sne.py:78 3: FutureWarning: The default initialization in TSNE will change fro m 'random' to 'pca' in 1.2.

FutureWarning,

/usr/local/lib/python3.7/dist-packages/sklearn/manifold/\_t\_sne.py:79 3: FutureWarning: The default learning rate in TSNE will change from 200.0 to 'auto' in 1.2.

FutureWarning,

Found 270 images belonging to 4 classes.

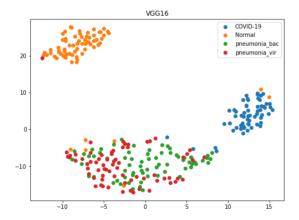
/usr/local/lib/python3.7/dist-packages/ipykernel\_launcher.py:28: Use rWarning: `Model.predict\_generator` is deprecated and will be remove d in a future version. Please use `Model.predict`, which supports ge nerators.

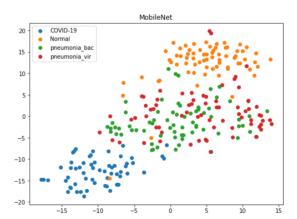
/usr/local/lib/python3.7/dist-packages/sklearn/manifold/\_t\_sne.py:78 3: FutureWarning: The default initialization in TSNE will change fro m 'random' to 'pca' in 1.2.

FutureWarning,

/usr/local/lib/python3.7/dist-packages/sklearn/manifold/\_t\_sne.py:79 3: FutureWarning: The default learning rate in TSNE will change from 200.0 to 'auto' in 1.2.

FutureWarning,





#### In [69]:

!jupyter nbconvert --to html '/content/drive/MyDrive/CS 542 Machine Learning/Cha llenge/Covid\_Data\_GradientCrescent/task2\_code.ipynb'

[NbConvertApp] Converting notebook /content/drive/MyDrive/CS 542 Mac hine Learning/Challenge/Covid\_Data\_GradientCrescent/task2\_code.ipynb to html

[NbConvertApp] Writing 620994 bytes to /content/drive/MyDrive/CS 542 Machine Learning/Challenge/Covid\_Data\_GradientCrescent/task2\_code.ht ml