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

Task 2 [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.

 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

```
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[2]: '2.6.0'
```

Load Image Data

```
In [20]: DATA_LIST = os.listdir('all/train')
    DATASET_PATH = 'all/train'
    TEST_DIR = '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 ou
    NUM_EPOCHS = 100
    LEARNING_RATE = 0.0005 # start off with high rate first 0.001 and experiment with reduc
```

Generate Training and Validation Batches

Found 216 images belonging to 4 classes. Found 54 images belonging to 4 classes.

C:\ProgramData\Anaconda3\lib\site-packages\keras_preprocessing\image\image_data_generato r.py:342: UserWarning: This ImageDataGenerator specifies `zca_whitening` which overrides setting of`featurewise_std_normalization`.

warnings.warn('This ImageDataGenerator specifies '

Model 1: VGG16

[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.

```
vgg16 = tf.keras.applications.VGG16(weights='imagenet', include_top=False, input_shape=
vgg16.trainable = False
```

```
model = tf.keras.Sequential([
    vgg16,
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(4, activation='softmax')
])
model.compile(optimizer = 'adam', loss = 'categorical_crossentropy', metrics = ['categorical_crossentropy', metrics = ['categorical_crossentropy']]
```

In [6]:

In [8]:

model.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten (Flatten)	(None, 25088)	0
dense (Dense)	(None, 128)	3211392
dense_1 (Dense)	(None, 4)	516
=======================================	=======================================	========

Total params: 17,926,596 Trainable params: 3,211,908 Non-trainable params: 14,714,688

[5 points] Train Model

#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
res = model.fit(train batches, epochs=NUM EPOCHS, steps per epoch=STEP SIZE TRAIN, \
         validation data=valid batches, validation steps=STEP SIZE VALID)
22
C:\ProgramData\Anaconda3\lib\site-packages\keras preprocessing\image\image data generato
r.py:720: UserWarning: This ImageDataGenerator specifies `featurewise_center`, but it ha
sn't been fit on any training data. Fit it first by calling `.fit(numpy_data)`.
 warnings.warn('This ImageDataGenerator specifies '
C:\ProgramData\Anaconda3\lib\site-packages\keras preprocessing\image\image data generato
r.py:739: UserWarning: This ImageDataGenerator specifies `zca whitening`, but it hasn't
been fit on any training data. Fit it first by calling `.fit(numpy data)`.
 warnings.warn('This ImageDataGenerator specifies '
Epoch 1/100
uracy: 0.3641 - val_loss: 1.0474 - val_categorical_accuracy: 0.5200
Epoch 2/100
```

```
uracy: 0.5049 - val_loss: 0.9806 - val_categorical_accuracy: 0.6400
Epoch 3/100
uracy: 0.5388 - val_loss: 1.4163 - val_categorical_accuracy: 0.5600
Epoch 4/100
uracy: 0.5485 - val loss: 1.2122 - val categorical accuracy: 0.5800
Epoch 5/100
uracy: 0.5922 - val loss: 1.0407 - val categorical accuracy: 0.4800
Epoch 6/100
uracy: 0.5922 - val_loss: 0.9112 - val_categorical_accuracy: 0.6400
Epoch 7/100
uracy: 0.6796 - val_loss: 0.8962 - val_categorical_accuracy: 0.7200
Epoch 8/100
uracy: 0.6019 - val loss: 0.7324 - val categorical accuracy: 0.6600
Epoch 9/100
uracy: 0.6359 - val_loss: 0.8045 - val_categorical_accuracy: 0.6200
Epoch 10/100
uracy: 0.6748 - val_loss: 0.8133 - val_categorical_accuracy: 0.7000
Epoch 11/100
uracy: 0.6602 - val_loss: 0.8591 - val_categorical_accuracy: 0.6000
Epoch 12/100
uracy: 0.6796 - val_loss: 0.8321 - val_categorical_accuracy: 0.6400
Epoch 13/100
uracy: 0.7039 - val_loss: 0.8407 - val_categorical_accuracy: 0.6600
Epoch 14/100
uracy: 0.7233 - val_loss: 1.0960 - val_categorical_accuracy: 0.5200
Epoch 15/100
uracy: 0.6748 - val_loss: 0.6260 - val_categorical_accuracy: 0.7000
uracy: 0.6165 - val_loss: 1.0902 - val_categorical_accuracy: 0.5800
Epoch 17/100
uracy: 0.6117 - val_loss: 1.3743 - val_categorical_accuracy: 0.4600
Epoch 18/100
uracy: 0.6748 - val_loss: 0.6242 - val_categorical_accuracy: 0.7200
Epoch 19/100
uracy: 0.6505 - val loss: 0.7225 - val categorical accuracy: 0.7000
Epoch 20/100
uracy: 0.6990 - val_loss: 0.8855 - val_categorical_accuracy: 0.5000
Epoch 21/100
uracy: 0.7087 - val_loss: 0.5505 - val_categorical_accuracy: 0.8000
Epoch 22/100
```

```
uracy: 0.6796 - val_loss: 0.7556 - val_categorical_accuracy: 0.5600
Epoch 23/100
uracy: 0.6893 - val_loss: 0.7624 - val_categorical_accuracy: 0.6400
Epoch 24/100
uracy: 0.6845 - val loss: 0.5872 - val categorical accuracy: 0.8200
Epoch 25/100
uracy: 0.6990 - val loss: 0.6140 - val categorical accuracy: 0.7200
Epoch 26/100
uracy: 0.6650 - val_loss: 0.7307 - val_categorical_accuracy: 0.6400
Epoch 27/100
uracy: 0.6505 - val_loss: 0.6007 - val_categorical_accuracy: 0.6800
Epoch 28/100
uracy: 0.6650 - val loss: 0.6771 - val categorical accuracy: 0.6600
uracy: 0.7476 - val_loss: 0.5150 - val_categorical_accuracy: 0.7400
Epoch 30/100
uracy: 0.7282 - val_loss: 0.8201 - val_categorical_accuracy: 0.7000
Epoch 31/100
uracy: 0.7427 - val loss: 0.6378 - val categorical accuracy: 0.7200
Epoch 32/100
uracy: 0.7330 - val_loss: 0.5242 - val_categorical_accuracy: 0.7800
Epoch 33/100
uracy: 0.7573 - val_loss: 0.5628 - val_categorical_accuracy: 0.7400
Epoch 34/100
uracy: 0.6796 - val_loss: 0.6085 - val_categorical_accuracy: 0.6400
Epoch 35/100
uracy: 0.7573 - val_loss: 1.3283 - val_categorical_accuracy: 0.6000
uracy: 0.6602 - val_loss: 0.7400 - val_categorical_accuracy: 0.6400
Epoch 37/100
uracy: 0.6408 - val_loss: 0.6764 - val_categorical_accuracy: 0.6600
Epoch 38/100
uracy: 0.6990 - val_loss: 1.0589 - val_categorical_accuracy: 0.5600
Epoch 39/100
uracy: 0.7087 - val loss: 0.6743 - val categorical accuracy: 0.6200
Epoch 40/100
uracy: 0.7621 - val_loss: 0.7640 - val_categorical_accuracy: 0.6400
Epoch 41/100
uracy: 0.7233 - val_loss: 0.9113 - val_categorical_accuracy: 0.6000
Epoch 42/100
```

```
uracy: 0.7039 - val_loss: 0.7954 - val_categorical_accuracy: 0.6800
Epoch 43/100
uracy: 0.7476 - val_loss: 0.5490 - val_categorical_accuracy: 0.6800
Epoch 44/100
uracy: 0.7621 - val loss: 0.6132 - val categorical accuracy: 0.7200
Epoch 45/100
uracy: 0.7718 - val loss: 0.7525 - val categorical accuracy: 0.6600
Epoch 46/100
uracy: 0.7573 - val_loss: 0.5611 - val_categorical_accuracy: 0.7000
Epoch 47/100
uracy: 0.7524 - val_loss: 0.5471 - val_categorical_accuracy: 0.7800
Epoch 48/100
uracy: 0.7816 - val loss: 0.6566 - val categorical accuracy: 0.6600
Epoch 49/100
uracy: 0.7767 - val_loss: 0.5557 - val_categorical_accuracy: 0.7400
Epoch 50/100
uracy: 0.7913 - val_loss: 0.7965 - val_categorical_accuracy: 0.6400
Epoch 51/100
uracy: 0.7330 - val loss: 0.8610 - val categorical accuracy: 0.6400
Epoch 52/100
uracy: 0.7718 - val_loss: 0.9379 - val_categorical_accuracy: 0.6200
Epoch 53/100
uracy: 0.7476 - val_loss: 0.8440 - val_categorical_accuracy: 0.6000
Epoch 54/100
uracy: 0.7476 - val_loss: 0.6495 - val_categorical_accuracy: 0.7600
Epoch 55/100
uracy: 0.7864 - val_loss: 0.6188 - val_categorical_accuracy: 0.7000
uracy: 0.7427 - val_loss: 0.5679 - val_categorical_accuracy: 0.7400
Epoch 57/100
uracy: 0.7476 - val_loss: 0.6813 - val_categorical_accuracy: 0.6000
Epoch 58/100
uracy: 0.7476 - val_loss: 0.5886 - val_categorical_accuracy: 0.7200
Epoch 59/100
uracy: 0.7524 - val loss: 0.6768 - val categorical accuracy: 0.7200
Epoch 60/100
uracy: 0.8010 - val_loss: 0.6979 - val_categorical_accuracy: 0.7200
Epoch 61/100
uracy: 0.7233 - val_loss: 0.6975 - val_categorical_accuracy: 0.6000
Epoch 62/100
```

```
uracy: 0.7233 - val_loss: 0.7930 - val_categorical_accuracy: 0.6600
Epoch 63/100
uracy: 0.7767 - val_loss: 0.7554 - val_categorical_accuracy: 0.5800
Epoch 64/100
uracy: 0.7282 - val loss: 0.4864 - val categorical accuracy: 0.7800
Epoch 65/100
uracy: 0.7233 - val loss: 0.7780 - val categorical accuracy: 0.6200
Epoch 66/100
uracy: 0.7621 - val_loss: 0.9396 - val_categorical_accuracy: 0.5600
Epoch 67/100
uracy: 0.7282 - val_loss: 0.8330 - val_categorical_accuracy: 0.6200
Epoch 68/100
uracy: 0.7621 - val loss: 0.7627 - val categorical accuracy: 0.6400
Epoch 69/100
uracy: 0.8058 - val_loss: 0.8683 - val_categorical_accuracy: 0.6000
Epoch 70/100
uracy: 0.7184 - val_loss: 1.2533 - val_categorical_accuracy: 0.5400
Epoch 71/100
uracy: 0.6748 - val loss: 0.7456 - val categorical accuracy: 0.6400
Epoch 72/100
uracy: 0.7816 - val_loss: 0.8743 - val_categorical_accuracy: 0.6600
Epoch 73/100
uracy: 0.7427 - val_loss: 0.5851 - val_categorical_accuracy: 0.7000
Epoch 74/100
uracy: 0.7621 - val_loss: 0.5795 - val_categorical_accuracy: 0.7200
Epoch 75/100
uracy: 0.7961 - val_loss: 0.6762 - val_categorical_accuracy: 0.7000
uracy: 0.8107 - val_loss: 0.6786 - val_categorical_accuracy: 0.7200
Epoch 77/100
uracy: 0.7767 - val_loss: 0.8066 - val_categorical_accuracy: 0.6400
Epoch 78/100
uracy: 0.7864 - val loss: 1.1145 - val categorical accuracy: 0.6200
Epoch 79/100
uracy: 0.7621 - val loss: 0.4777 - val categorical accuracy: 0.8400
Epoch 80/100
uracy: 0.7864 - val_loss: 0.5568 - val_categorical_accuracy: 0.7600
Epoch 81/100
uracy: 0.7816 - val_loss: 0.6957 - val_categorical_accuracy: 0.6600
Epoch 82/100
```

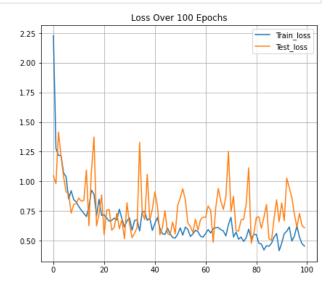
```
uracy: 0.8107 - val_loss: 0.7011 - val_categorical_accuracy: 0.6800
Epoch 83/100
uracy: 0.8155 - val_loss: 0.6044 - val_categorical_accuracy: 0.6800
Epoch 84/100
uracy: 0.8155 - val loss: 0.6969 - val categorical accuracy: 0.6600
Epoch 85/100
uracy: 0.7864 - val loss: 0.8051 - val categorical accuracy: 0.7000
Epoch 86/100
uracy: 0.8252 - val_loss: 0.5153 - val_categorical_accuracy: 0.7000
Epoch 87/100
uracy: 0.7961 - val_loss: 0.5039 - val_categorical_accuracy: 0.7400
Epoch 88/100
uracy: 0.7767 - val loss: 0.6858 - val categorical accuracy: 0.6600
uracy: 0.7913 - val_loss: 0.8444 - val_categorical_accuracy: 0.7000
Epoch 90/100
uracy: 0.8252 - val_loss: 0.6589 - val_categorical_accuracy: 0.7400
Epoch 91/100
uracy: 0.7816 - val_loss: 0.8178 - val_categorical_accuracy: 0.6000
Epoch 92/100
uracy: 0.7573 - val_loss: 0.6674 - val_categorical_accuracy: 0.6800
Epoch 93/100
uracy: 0.7379 - val_loss: 1.0272 - val_categorical_accuracy: 0.6600
Epoch 94/100
uracy: 0.7184 - val_loss: 0.9452 - val_categorical_accuracy: 0.6000
Epoch 95/100
uracy: 0.7670 - val_loss: 0.8646 - val_categorical_accuracy: 0.6200
uracy: 0.7864 - val_loss: 0.7234 - val_categorical_accuracy: 0.6400
Epoch 97/100
uracy: 0.7136 - val_loss: 0.6121 - val_categorical_accuracy: 0.6400
Epoch 98/100
uracy: 0.7573 - val_loss: 0.7294 - val_categorical_accuracy: 0.6000
Epoch 99/100
uracy: 0.8058 - val loss: 0.6275 - val categorical accuracy: 0.6800
Epoch 100/100
uracy: 0.8252 - val_loss: 0.6081 - val_categorical_accuracy: 0.7000
```

[5 points] Plot Accuracy and Loss During Training

```
In [9]: import matplotlib.pyplot as plt
```

```
fig, (ax1, ax2) = plt.subplots(1, 2)
fig.set_figheight(6)
fig.set_figwidth(15)
ax1.plot(res.history['categorical_accuracy'])
ax1.plot(res.history['val_categorical_accuracy'])
ax1.set_title('Accuracy Over ' + str(NUM_EPOCHS) + ' Epochs')
ax1.legend(['Train_acc', 'Test_acc'], loc='lower right')
ax1.grid(True)
ax2.set_title('Loss Over ' + str(NUM_EPOCHS) + ' Epochs')
ax2.plot(res.history['loss'])
ax2.plot(res.history['val_loss'])
ax2.legend(['Train_loss', 'Test_loss'], loc='upper right')
ax2.grid(True)
plt.show()
```





Testing Model

[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 [15]:
          from sklearn.manifold import TSNE
          intermediate layer model = models.Model(inputs=model.input,
                                                   outputs=model.get layer('feature dense').output
          tsne eval generator = test datagen.flow from directory(DATASET PATH,target size=IMAGE S
                                                             batch size=1, shuffle=True, seed=42, cla
          raise NotImplementedError("Extract features from the tsne data generator and fit a t-SN
                                     "and plot the resulting 2D features of the four classes.")
         Found 270 images belonging to 4 classes.
         {'covid': 0, 'normal': 1, 'pneumonia bac': 2, 'pneumonia vir': 3}
         Extracting features for 270 images.
         270/270 [=========== ] - 71s 265ms/step
         Training TSNE model.
           15
           10
            5
                                                 COVID-19
                                                 Normal
            0
                                                 Pneumonia bac
                                                 Pneumonia vir
          -10
```

Attempt 2: Revising the vgg16 model with additional layers

10

[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.

15

20

```
vgg16 = tf.keras.applications.VGG16(weights='imagenet', include_top=False, input_shape=
vgg16.trainable = False
model = tf.keras.Sequential([
    vgg16,
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(500, activation='relu'),
```

-15

-15

```
tf.keras.layers.Dropout(rate=0.2),
    tf.keras.layers.Dense(125, activation='relu'),
    tf.keras.layers.Dropout(rate=0.2),
    tf.keras.layers.Dense(4, activation='softmax')
])
model.compile(optimizer = tf.keras.optimizers.Adam(learning_rate=1e-5), loss = 'categor'
    metrics = ['categorical_accuracy'])
```

In [107...

model.summary()

Model: "sequential_25"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_25 (Flatten)	(None, 25088)	0
dense_76 (Dense)	(None, 500)	12544500
dropout_49 (Dropout)	(None, 500)	0
dense_77 (Dense)	(None, 125)	62625
dropout_50 (Dropout)	(None, 125)	0
dense_78 (Dense)	(None, 4)	504 ======

Total params: 27,322,317 Trainable params: 12,607,629 Non-trainable params: 14,714,688

[5 points] Train Model

```
In [108... #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

res = model.fit(train_batches, epochs=NUM_EPOCHS, steps_per_epoch=STEP_SIZE_TRAIN, \
    validation_data=valid_batches, validation_steps=STEP_SIZE_VALID)
```

```
uracy: 0.3058 - val_loss: 1.2964 - val_categorical_accuracy: 0.4400
Epoch 4/100
uracy: 0.3495 - val_loss: 1.2727 - val_categorical_accuracy: 0.4000
Epoch 5/100
uracy: 0.3952 - val loss: 1.2660 - val categorical accuracy: 0.4000
Epoch 6/100
uracy: 0.3738 - val loss: 1.1904 - val categorical accuracy: 0.5400
Epoch 7/100
uracy: 0.4466 - val_loss: 1.1790 - val_categorical_accuracy: 0.4400
Epoch 8/100
uracy: 0.4369 - val_loss: 1.1511 - val_categorical_accuracy: 0.4400
Epoch 9/100
uracy: 0.4903 - val loss: 1.1022 - val categorical accuracy: 0.5800
uracy: 0.5291 - val_loss: 1.0469 - val_categorical_accuracy: 0.6200
Epoch 11/100
uracy: 0.4757 - val_loss: 1.0814 - val_categorical_accuracy: 0.5200
Epoch 12/100
uracy: 0.5728 - val loss: 1.0509 - val categorical accuracy: 0.4800
Epoch 13/100
uracy: 0.5437 - val_loss: 1.0136 - val_categorical_accuracy: 0.5400
Epoch 14/100
uracy: 0.5680 - val_loss: 0.9926 - val_categorical_accuracy: 0.6200
Epoch 15/100
uracy: 0.5049 - val_loss: 0.9945 - val_categorical_accuracy: 0.5000
Epoch 16/100
uracy: 0.5049 - val_loss: 0.9241 - val_categorical_accuracy: 0.6200
uracy: 0.6068 - val_loss: 0.9012 - val_categorical_accuracy: 0.6400
Epoch 18/100
uracy: 0.5243 - val_loss: 1.0218 - val_categorical_accuracy: 0.4800
Epoch 19/100
uracy: 0.5922 - val_loss: 0.9368 - val_categorical_accuracy: 0.6000
Epoch 20/100
uracy: 0.5680 - val loss: 0.9529 - val categorical accuracy: 0.5200
Epoch 21/100
uracy: 0.5243 - val_loss: 0.9011 - val_categorical_accuracy: 0.5600
Epoch 22/100
uracy: 0.5922 - val_loss: 0.8851 - val_categorical_accuracy: 0.6800
Epoch 23/100
```

```
uracy: 0.6019 - val_loss: 0.8752 - val_categorical_accuracy: 0.6400
Epoch 24/100
uracy: 0.6165 - val_loss: 0.8295 - val_categorical_accuracy: 0.6600
Epoch 25/100
uracy: 0.5825 - val loss: 0.9095 - val categorical accuracy: 0.6000
Epoch 26/100
uracy: 0.5777 - val loss: 0.8205 - val categorical accuracy: 0.7600
Epoch 27/100
uracy: 0.5922 - val_loss: 0.9387 - val_categorical_accuracy: 0.5200
Epoch 28/100
uracy: 0.5810 - val_loss: 0.8439 - val_categorical_accuracy: 0.6800
Epoch 29/100
uracy: 0.6456 - val loss: 0.8396 - val categorical accuracy: 0.5800
Epoch 30/100
uracy: 0.5905 - val_loss: 0.8276 - val_categorical_accuracy: 0.6000
Epoch 31/100
uracy: 0.6408 - val_loss: 0.8018 - val_categorical_accuracy: 0.6800
Epoch 32/100
uracy: 0.6845 - val loss: 0.8046 - val categorical accuracy: 0.7000
Epoch 33/100
uracy: 0.6845 - val_loss: 0.8323 - val_categorical_accuracy: 0.6600
Epoch 34/100
uracy: 0.6214 - val_loss: 0.8489 - val_categorical_accuracy: 0.6400
Epoch 35/100
uracy: 0.6359 - val_loss: 0.8607 - val_categorical_accuracy: 0.5800
Epoch 36/100
uracy: 0.6408 - val_loss: 0.9271 - val_categorical_accuracy: 0.5400
uracy: 0.5777 - val_loss: 0.8072 - val_categorical_accuracy: 0.6000
Epoch 38/100
uracy: 0.6165 - val_loss: 0.7877 - val_categorical_accuracy: 0.6400
Epoch 39/100
uracy: 0.6019 - val_loss: 0.7711 - val_categorical_accuracy: 0.6200
Epoch 40/100
uracy: 0.6214 - val loss: 0.8392 - val categorical accuracy: 0.6200
Epoch 41/100
uracy: 0.6553 - val_loss: 0.8123 - val_categorical_accuracy: 0.6000
Epoch 42/100
uracy: 0.6699 - val_loss: 0.7008 - val_categorical_accuracy: 0.6800
Epoch 43/100
```

```
uracy: 0.6262 - val_loss: 0.7835 - val_categorical_accuracy: 0.5800
Epoch 44/100
uracy: 0.6762 - val_loss: 0.8003 - val_categorical_accuracy: 0.6400
Epoch 45/100
uracy: 0.6359 - val loss: 0.8018 - val categorical accuracy: 0.6000
Epoch 46/100
uracy: 0.6796 - val loss: 0.7967 - val categorical accuracy: 0.6000
Epoch 47/100
uracy: 0.6650 - val_loss: 0.7897 - val_categorical_accuracy: 0.6000
Epoch 48/100
uracy: 0.6650 - val_loss: 0.7925 - val_categorical_accuracy: 0.6000
Epoch 49/100
uracy: 0.6408 - val loss: 0.7188 - val categorical accuracy: 0.6600
Epoch 50/100
uracy: 0.6748 - val_loss: 0.9118 - val_categorical_accuracy: 0.5200
Epoch 51/100
uracy: 0.6796 - val_loss: 0.7118 - val_categorical_accuracy: 0.6600
Epoch 52/100
uracy: 0.6381 - val loss: 0.8619 - val categorical accuracy: 0.6000
Epoch 53/100
uracy: 0.6456 - val_loss: 0.8276 - val_categorical_accuracy: 0.6200
Epoch 54/100
uracy: 0.6952 - val_loss: 0.7786 - val_categorical_accuracy: 0.6400
Epoch 55/100
uracy: 0.6262 - val_loss: 0.7834 - val_categorical_accuracy: 0.5400
Epoch 56/100
uracy: 0.7039 - val_loss: 0.7350 - val_categorical_accuracy: 0.6600
uracy: 0.6990 - val_loss: 0.7322 - val_categorical_accuracy: 0.6400
Epoch 58/100
uracy: 0.7039 - val_loss: 0.7506 - val_categorical_accuracy: 0.7000
Epoch 59/100
uracy: 0.6650 - val_loss: 0.7285 - val_categorical_accuracy: 0.6400
Epoch 60/100
uracy: 0.6845 - val loss: 0.7820 - val categorical accuracy: 0.5600
Epoch 61/100
uracy: 0.7039 - val_loss: 0.7406 - val_categorical_accuracy: 0.6400
Epoch 62/100
uracy: 0.6553 - val_loss: 0.7649 - val_categorical_accuracy: 0.5800
Epoch 63/100
```

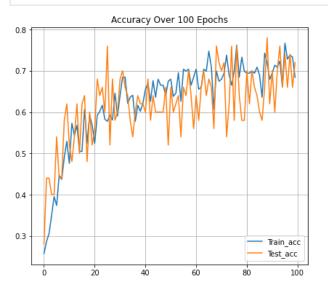
```
uracy: 0.6602 - val_loss: 0.7345 - val_categorical_accuracy: 0.6600
Epoch 64/100
uracy: 0.7039 - val_loss: 0.6959 - val_categorical_accuracy: 0.7000
Epoch 65/100
uracy: 0.6990 - val loss: 0.6780 - val categorical accuracy: 0.6400
Epoch 66/100
uracy: 0.7476 - val loss: 0.7135 - val categorical accuracy: 0.6800
Epoch 67/100
uracy: 0.7087 - val_loss: 0.6518 - val_categorical_accuracy: 0.6600
Epoch 68/100
uracy: 0.6068 - val_loss: 0.8039 - val_categorical_accuracy: 0.5600
Epoch 69/100
uracy: 0.6990 - val loss: 0.7028 - val categorical accuracy: 0.7600
uracy: 0.6748 - val_loss: 0.7585 - val_categorical_accuracy: 0.7200
Epoch 71/100
uracy: 0.6796 - val_loss: 0.7324 - val_categorical_accuracy: 0.7000
Epoch 72/100
uracy: 0.6942 - val_loss: 0.6031 - val_categorical_accuracy: 0.7200
Epoch 73/100
uracy: 0.7379 - val_loss: 0.7213 - val_categorical_accuracy: 0.5400
Epoch 74/100
uracy: 0.6893 - val_loss: 0.7330 - val_categorical_accuracy: 0.6200
Epoch 75/100
uracy: 0.6650 - val_loss: 0.6559 - val_categorical_accuracy: 0.7600
Epoch 76/100
uracy: 0.6990 - val_loss: 0.7027 - val_categorical_accuracy: 0.5800
uracy: 0.7621 - val_loss: 0.6649 - val_categorical_accuracy: 0.7600
Epoch 78/100
uracy: 0.6845 - val_loss: 0.6997 - val_categorical_accuracy: 0.6400
Epoch 79/100
uracy: 0.7330 - val_loss: 0.7243 - val_categorical_accuracy: 0.5800
Epoch 80/100
uracy: 0.6990 - val loss: 0.8181 - val categorical accuracy: 0.5800
Epoch 81/100
uracy: 0.6942 - val_loss: 0.6730 - val_categorical_accuracy: 0.7000
Epoch 82/100
uracy: 0.6942 - val_loss: 0.7864 - val_categorical_accuracy: 0.6200
Epoch 83/100
```

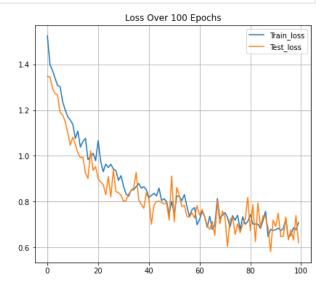
```
uracy: 0.6990 - val_loss: 0.6251 - val_categorical_accuracy: 0.7000
Epoch 84/100
uracy: 0.6942 - val_loss: 0.7933 - val_categorical_accuracy: 0.6600
Epoch 85/100
uracy: 0.7087 - val loss: 0.6870 - val categorical accuracy: 0.6400
Epoch 86/100
uracy: 0.6893 - val loss: 0.7381 - val categorical accuracy: 0.6000
Epoch 87/100
uracy: 0.6359 - val_loss: 0.7317 - val_categorical_accuracy: 0.5800
Epoch 88/100
uracy: 0.7427 - val_loss: 0.6850 - val_categorical_accuracy: 0.6600
Epoch 89/100
uracy: 0.7184 - val loss: 0.5812 - val categorical accuracy: 0.7800
Epoch 90/100
uracy: 0.6796 - val_loss: 0.7188 - val_categorical_accuracy: 0.6200
Epoch 91/100
uracy: 0.6942 - val_loss: 0.6916 - val_categorical_accuracy: 0.7000
Epoch 92/100
uracy: 0.7136 - val loss: 0.7494 - val categorical accuracy: 0.6000
Epoch 93/100
uracy: 0.7087 - val_loss: 0.6472 - val_categorical_accuracy: 0.7000
Epoch 94/100
uracy: 0.7233 - val_loss: 0.6472 - val_categorical_accuracy: 0.7600
Epoch 95/100
uracy: 0.6619 - val_loss: 0.7326 - val_categorical_accuracy: 0.6600
Epoch 96/100
uracy: 0.7670 - val_loss: 0.6320 - val_categorical_accuracy: 0.7400
uracy: 0.7282 - val_loss: 0.6764 - val_categorical_accuracy: 0.6600
Epoch 98/100
uracy: 0.7379 - val_loss: 0.6319 - val_categorical_accuracy: 0.7400
Epoch 99/100
uracy: 0.7330 - val loss: 0.7390 - val categorical accuracy: 0.6600
Epoch 100/100
uracy: 0.6845 - val loss: 0.6197 - val categorical accuracy: 0.7200
```

[5 points] Plot Accuracy and Loss During Training

```
import matplotlib.pyplot as plt
fig, (ax1, ax2) = plt.subplots(1, 2)
fig.set_figheight(6)
```

```
fig.set_figwidth(15)
ax1.plot(res.history['categorical_accuracy'])
ax1.plot(res.history['val_categorical_accuracy'])
ax1.set_title('Accuracy Over ' + str(NUM_EPOCHS) + ' Epochs')
ax1.legend(['Train_acc', 'Test_acc'], loc='lower right')
ax1.grid(True)
ax2.set_title('Loss Over ' + str(NUM_EPOCHS) + ' Epochs')
ax2.plot(res.history['loss'])
ax2.plot(res.history['val_loss'])
ax2.legend(['Train_loss', 'Test_loss'], loc='upper right')
ax2.grid(True)
plt.show()
```





Testing Model

Attempt 3: Try generating more fake data with data augmentation

Generate Training and Validation Batches

```
In [115... train_datagen = ImageDataGenerator(rescale=1./255,rotation_range=50,featurewise_center
```

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.

```
vgg16 = tf.keras.applications.VGG16(weights='imagenet', include_top=False, input_shape=
vgg16.trainable = False
model = tf.keras.Sequential([
    vgg16,
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(4, activation='softmax')
])
model.compile(optimizer = tf.keras.optimizers.Adam(learning_rate=1e-5),
    loss = 'categorical_crossentropy', metrics = ['categorical_accuracy'])
```

In [124...

model.summary()

Model: "sequential 29"

Layer (type)	Output Shape	Param #
vgg16 (Functional)	(None, 7, 7, 512)	14714688
flatten_29 (Flatten)	(None, 25088)	0
dense_85 (Dense)	(None, 128)	3211392
dense_86 (Dense)	(None, 4)	516
Total params: 17,926,596 Trainable params: 3,211,908		

[5 points] Train Model

Non-trainable params: 14,714,688

```
In [125... | #FIT MODEL
```

```
Epoch 1/500
uracy: 0.2961 - val_loss: 1.3821 - val_categorical_accuracy: 0.2200
Epoch 2/500
uracy: 0.3398 - val_loss: 1.3469 - val_categorical_accuracy: 0.3000
Epoch 3/500
uracy: 0.4029 - val_loss: 1.2987 - val_categorical_accuracy: 0.5200
Epoch 4/500
uracy: 0.4078 - val_loss: 1.2541 - val_categorical_accuracy: 0.4200
Epoch 5/500
uracy: 0.4126 - val_loss: 1.2271 - val_categorical_accuracy: 0.4800
uracy: 0.5291 - val_loss: 1.1538 - val_categorical_accuracy: 0.5400
Epoch 7/500
uracy: 0.5146 - val_loss: 1.1344 - val_categorical_accuracy: 0.5400
Epoch 8/500
uracy: 0.5485 - val_loss: 1.1357 - val_categorical_accuracy: 0.4800
Epoch 9/500
uracy: 0.5485 - val loss: 1.0950 - val categorical accuracy: 0.5400
Epoch 10/500
uracy: 0.5631 - val loss: 1.0731 - val categorical accuracy: 0.6200
Epoch 11/500
uracy: 0.5194 - val_loss: 1.0887 - val_categorical_accuracy: 0.5000
Epoch 12/500
uracy: 0.5485 - val_loss: 1.0651 - val_categorical_accuracy: 0.4800
Epoch 13/500
uracy: 0.6262 - val_loss: 1.0061 - val_categorical_accuracy: 0.7400
uracy: 0.5825 - val_loss: 1.0708 - val_categorical_accuracy: 0.4800
Epoch 15/500
uracy: 0.5194 - val_loss: 0.9748 - val_categorical_accuracy: 0.6600
Epoch 16/500
uracy: 0.6359 - val loss: 0.9666 - val categorical accuracy: 0.6200
Epoch 17/500
```

```
uracy: 0.6019 - val loss: 0.9611 - val categorical accuracy: 0.6000
Epoch 18/500
uracy: 0.6602 - val loss: 0.9859 - val categorical accuracy: 0.5000
Epoch 19/500
uracy: 0.5922 - val_loss: 0.9143 - val_categorical_accuracy: 0.6600
Epoch 20/500
uracy: 0.6796 - val loss: 0.9621 - val categorical accuracy: 0.5800
Epoch 21/500
uracy: 0.6650 - val loss: 0.9041 - val categorical accuracy: 0.6400
uracy: 0.7039 - val_loss: 0.8591 - val_categorical_accuracy: 0.5800
Epoch 23/500
uracy: 0.6408 - val loss: 0.9096 - val categorical accuracy: 0.6200
Epoch 24/500
uracy: 0.6429 - val loss: 0.8812 - val categorical accuracy: 0.6400
Epoch 25/500
uracy: 0.6699 - val_loss: 0.9165 - val_categorical_accuracy: 0.6000
uracy: 0.6650 - val_loss: 0.8861 - val_categorical_accuracy: 0.5600
Epoch 27/500
uracy: 0.6942 - val_loss: 0.8626 - val_categorical_accuracy: 0.6400
Epoch 28/500
uracy: 0.7087 - val_loss: 0.8623 - val_categorical_accuracy: 0.6000
uracy: 0.6990 - val_loss: 0.8460 - val_categorical_accuracy: 0.6400
Epoch 30/500
uracy: 0.6408 - val_loss: 0.9214 - val_categorical_accuracy: 0.6400
Epoch 31/500
uracy: 0.6650 - val_loss: 0.9201 - val_categorical_accuracy: 0.6000
Epoch 32/500
uracy: 0.6952 - val loss: 0.8307 - val categorical accuracy: 0.6400
uracy: 0.6796 - val_loss: 0.8540 - val_categorical_accuracy: 0.6400
Epoch 34/500
uracy: 0.6505 - val_loss: 0.9021 - val_categorical_accuracy: 0.5200
Epoch 35/500
uracy: 0.6990 - val_loss: 0.8136 - val_categorical_accuracy: 0.6600
Epoch 36/500
uracy: 0.6748 - val_loss: 0.8104 - val_categorical_accuracy: 0.6000
Epoch 37/500
```

```
uracy: 0.6165 - val loss: 0.8809 - val categorical accuracy: 0.6200
Epoch 38/500
uracy: 0.6845 - val loss: 0.7799 - val categorical accuracy: 0.6400
Epoch 39/500
uracy: 0.6990 - val_loss: 0.7913 - val_categorical_accuracy: 0.6400
Epoch 40/500
uracy: 0.6845 - val loss: 0.7951 - val categorical accuracy: 0.6600
Epoch 41/500
uracy: 0.7379 - val_loss: 0.7460 - val_categorical_accuracy: 0.6800
uracy: 0.7233 - val_loss: 0.7928 - val_categorical_accuracy: 0.7000
Epoch 43/500
uracy: 0.6893 - val loss: 0.8132 - val categorical accuracy: 0.6000
Epoch 44/500
uracy: 0.6845 - val loss: 0.8192 - val categorical accuracy: 0.6200
Epoch 45/500
uracy: 0.7184 - val_loss: 0.7871 - val_categorical_accuracy: 0.6000
Epoch 46/500
uracy: 0.6699 - val_loss: 0.7937 - val_categorical_accuracy: 0.6000
Epoch 47/500
uracy: 0.6699 - val_loss: 0.7656 - val_categorical_accuracy: 0.6600
Epoch 48/500
uracy: 0.6990 - val_loss: 0.8123 - val_categorical_accuracy: 0.6000
uracy: 0.6942 - val_loss: 0.8031 - val_categorical_accuracy: 0.6200
uracy: 0.7233 - val_loss: 0.7603 - val_categorical_accuracy: 0.6800
Epoch 51/500
uracy: 0.7136 - val_loss: 0.7727 - val_categorical_accuracy: 0.6200
Epoch 52/500
uracy: 0.6990 - val loss: 0.8109 - val categorical accuracy: 0.6400
uracy: 0.7087 - val_loss: 0.7407 - val_categorical_accuracy: 0.6600
Epoch 54/500
uracy: 0.7427 - val_loss: 0.7707 - val_categorical_accuracy: 0.6400
Epoch 55/500
uracy: 0.7379 - val_loss: 0.7603 - val_categorical_accuracy: 0.6600
Epoch 56/500
uracy: 0.7621 - val_loss: 0.7937 - val_categorical_accuracy: 0.6200
Epoch 57/500
```

```
uracy: 0.6990 - val loss: 0.7928 - val categorical accuracy: 0.6200
Epoch 58/500
uracy: 0.7136 - val loss: 0.7332 - val categorical accuracy: 0.6400
Epoch 59/500
uracy: 0.7136 - val_loss: 0.7883 - val_categorical_accuracy: 0.6200
Epoch 60/500
uracy: 0.7136 - val loss: 0.7645 - val categorical accuracy: 0.5400
Epoch 61/500
uracy: 0.7330 - val_loss: 0.6944 - val_categorical_accuracy: 0.7600
uracy: 0.7039 - val_loss: 0.6878 - val_categorical_accuracy: 0.7000
Epoch 63/500
uracy: 0.6845 - val loss: 0.7436 - val categorical accuracy: 0.6600
Epoch 64/500
uracy: 0.6990 - val loss: 0.7091 - val categorical accuracy: 0.6800
Epoch 65/500
uracy: 0.7190 - val_loss: 0.7491 - val_categorical_accuracy: 0.6400
uracy: 0.7330 - val_loss: 0.7089 - val_categorical_accuracy: 0.6600
Epoch 67/500
uracy: 0.6893 - val_loss: 0.7178 - val_categorical_accuracy: 0.7600
Epoch 68/500
uracy: 0.7330 - val_loss: 0.7202 - val_categorical_accuracy: 0.6800
uracy: 0.7524 - val_loss: 0.7328 - val_categorical_accuracy: 0.6800
Epoch 70/500
uracy: 0.7621 - val_loss: 0.7090 - val_categorical_accuracy: 0.6800
Epoch 71/500
uracy: 0.7039 - val_loss: 0.7922 - val_categorical_accuracy: 0.6200
Epoch 72/500
uracy: 0.6699 - val loss: 0.7654 - val categorical accuracy: 0.6800
Epoch 73/500
uracy: 0.7476 - val_loss: 0.7910 - val_categorical_accuracy: 0.6200
Epoch 74/500
uracy: 0.7670 - val_loss: 0.7384 - val_categorical_accuracy: 0.6600
Epoch 75/500
uracy: 0.7524 - val_loss: 0.7511 - val_categorical_accuracy: 0.6400
Epoch 76/500
uracy: 0.7039 - val_loss: 0.7950 - val_categorical_accuracy: 0.6600
Epoch 77/500
```

```
uracy: 0.7330 - val loss: 0.7377 - val categorical accuracy: 0.6800
Epoch 78/500
uracy: 0.7621 - val loss: 0.7289 - val categorical accuracy: 0.6400
Epoch 79/500
uracy: 0.7379 - val_loss: 0.7194 - val_categorical_accuracy: 0.7200
Epoch 80/500
uracy: 0.7621 - val loss: 0.7827 - val categorical accuracy: 0.6000
Epoch 81/500
uracy: 0.7330 - val_loss: 0.6818 - val_categorical_accuracy: 0.7000
uracy: 0.7427 - val_loss: 0.6890 - val_categorical_accuracy: 0.7400
Epoch 83/500
uracy: 0.7670 - val loss: 0.7012 - val categorical accuracy: 0.6800
Epoch 84/500
uracy: 0.7379 - val loss: 0.7140 - val categorical accuracy: 0.6400
Epoch 85/500
uracy: 0.7476 - val_loss: 0.7522 - val_categorical_accuracy: 0.6200
uracy: 0.7427 - val_loss: 0.7710 - val_categorical_accuracy: 0.5200
Epoch 87/500
uracy: 0.7427 - val_loss: 0.6979 - val_categorical_accuracy: 0.7000
Epoch 88/500
uracy: 0.7330 - val_loss: 0.6852 - val_categorical_accuracy: 0.6600
uracy: 0.7476 - val_loss: 0.7164 - val_categorical_accuracy: 0.6200
Epoch 90/500
uracy: 0.7330 - val_loss: 0.6902 - val_categorical_accuracy: 0.7000
Epoch 91/500
uracy: 0.7913 - val_loss: 0.7757 - val_categorical_accuracy: 0.6000
Epoch 92/500
uracy: 0.7621 - val loss: 0.6256 - val categorical accuracy: 0.7000
uracy: 0.7670 - val_loss: 0.8152 - val_categorical_accuracy: 0.6200
Epoch 94/500
uracy: 0.7379 - val_loss: 0.6905 - val_categorical_accuracy: 0.6800
Epoch 95/500
uracy: 0.7619 - val_loss: 0.6883 - val_categorical_accuracy: 0.6800
Epoch 96/500
uracy: 0.7282 - val_loss: 0.7228 - val_categorical_accuracy: 0.7000
Epoch 97/500
```

```
uracy: 0.7476 - val loss: 0.7072 - val categorical accuracy: 0.6600
Epoch 98/500
uracy: 0.7767 - val loss: 0.6805 - val categorical accuracy: 0.6600
Epoch 99/500
uracy: 0.7330 - val_loss: 0.6660 - val_categorical_accuracy: 0.7200
Epoch 100/500
uracy: 0.7233 - val_loss: 0.7528 - val_categorical_accuracy: 0.6400
Epoch 101/500
uracy: 0.7816 - val_loss: 0.7405 - val_categorical_accuracy: 0.6000
uracy: 0.7427 - val_loss: 0.6523 - val_categorical_accuracy: 0.6200
Epoch 103/500
uracy: 0.7864 - val loss: 0.6699 - val categorical accuracy: 0.7200
Epoch 104/500
uracy: 0.7816 - val loss: 0.6656 - val categorical accuracy: 0.6800
Epoch 105/500
uracy: 0.7282 - val_loss: 0.6369 - val_categorical_accuracy: 0.7000
uracy: 0.7961 - val_loss: 0.6747 - val_categorical_accuracy: 0.7000
Epoch 107/500
uracy: 0.7379 - val_loss: 0.6994 - val_categorical_accuracy: 0.7000
Epoch 108/500
uracy: 0.7379 - val_loss: 0.6638 - val_categorical_accuracy: 0.7600
uracy: 0.8204 - val_loss: 0.6461 - val_categorical_accuracy: 0.7800
Epoch 110/500
uracy: 0.7524 - val_loss: 0.7418 - val_categorical_accuracy: 0.6800
Epoch 111/500
uracy: 0.8010 - val_loss: 0.6703 - val_categorical_accuracy: 0.6600
Epoch 112/500
uracy: 0.7573 - val loss: 0.7378 - val categorical accuracy: 0.6600
Epoch 113/500
uracy: 0.7476 - val_loss: 0.7194 - val_categorical_accuracy: 0.6200
Epoch 114/500
uracy: 0.7427 - val_loss: 0.6743 - val_categorical_accuracy: 0.7000
Epoch 115/500
uracy: 0.7427 - val_loss: 0.6859 - val_categorical_accuracy: 0.7000
Epoch 116/500
uracy: 0.7476 - val_loss: 0.6546 - val_categorical_accuracy: 0.7200
Epoch 117/500
```

```
uracy: 0.7286 - val_loss: 0.7151 - val_categorical_accuracy: 0.7200
Epoch 118/500
uracy: 0.7427 - val loss: 0.6916 - val categorical accuracy: 0.6600
Epoch 119/500
uracy: 0.7476 - val_loss: 0.6935 - val_categorical_accuracy: 0.6800
Epoch 120/500
uracy: 0.7379 - val_loss: 0.6806 - val_categorical_accuracy: 0.6600
Epoch 121/500
uracy: 0.7282 - val_loss: 0.6946 - val_categorical_accuracy: 0.7200
uracy: 0.7718 - val_loss: 0.6838 - val_categorical_accuracy: 0.7000
Epoch 123/500
uracy: 0.7767 - val loss: 0.7039 - val categorical accuracy: 0.6600
Epoch 124/500
uracy: 0.7864 - val loss: 0.5464 - val categorical accuracy: 0.7200
Epoch 125/500
uracy: 0.7718 - val_loss: 0.7020 - val_categorical_accuracy: 0.6200
Epoch 126/500
uracy: 0.7767 - val_loss: 0.6127 - val_categorical_accuracy: 0.7200
Epoch 127/500
uracy: 0.7864 - val_loss: 0.7186 - val_categorical_accuracy: 0.6800
Epoch 128/500
uracy: 0.6942 - val_loss: 0.6821 - val_categorical_accuracy: 0.7200
uracy: 0.7427 - val_loss: 0.6464 - val_categorical_accuracy: 0.7000
Epoch 130/500
uracy: 0.7379 - val_loss: 0.6983 - val_categorical_accuracy: 0.7000
Epoch 131/500
uracy: 0.7913 - val_loss: 0.6611 - val_categorical_accuracy: 0.7200
Epoch 132/500
uracy: 0.7524 - val loss: 0.6338 - val categorical accuracy: 0.7200
Epoch 133/500
uracy: 0.7621 - val_loss: 0.7271 - val_categorical_accuracy: 0.6600
Epoch 134/500
uracy: 0.7718 - val_loss: 0.8153 - val_categorical_accuracy: 0.5800
Epoch 135/500
uracy: 0.7718 - val_loss: 0.6326 - val_categorical_accuracy: 0.6800
Epoch 136/500
uracy: 0.7670 - val_loss: 0.7190 - val_categorical_accuracy: 0.6200
Epoch 137/500
```

```
uracy: 0.7233 - val loss: 0.6729 - val categorical accuracy: 0.6400
Epoch 138/500
uracy: 0.7864 - val loss: 0.7383 - val categorical accuracy: 0.7000
Epoch 139/500
uracy: 0.7427 - val_loss: 0.6269 - val_categorical_accuracy: 0.7600
Epoch 140/500
uracy: 0.7816 - val loss: 0.6510 - val categorical accuracy: 0.7400
Epoch 141/500
uracy: 0.7670 - val_loss: 0.7577 - val_categorical_accuracy: 0.6600
uracy: 0.7670 - val_loss: 0.7339 - val_categorical_accuracy: 0.6800
Epoch 143/500
uracy: 0.7282 - val loss: 0.6982 - val categorical accuracy: 0.6800
Epoch 144/500
uracy: 0.7718 - val loss: 0.6339 - val categorical accuracy: 0.6600
Epoch 145/500
uracy: 0.8301 - val_loss: 0.6749 - val_categorical_accuracy: 0.6800
uracy: 0.7282 - val_loss: 0.6150 - val_categorical_accuracy: 0.7400
Epoch 147/500
uracy: 0.7961 - val_loss: 0.6319 - val_categorical_accuracy: 0.7400
Epoch 148/500
uracy: 0.7767 - val_loss: 0.7328 - val_categorical_accuracy: 0.6400
uracy: 0.7427 - val_loss: 0.7503 - val_categorical_accuracy: 0.5600
Epoch 150/500
uracy: 0.7864 - val_loss: 0.6920 - val_categorical_accuracy: 0.7200
Epoch 151/500
uracy: 0.8058 - val_loss: 0.7519 - val_categorical_accuracy: 0.6400
Epoch 152/500
uracy: 0.8204 - val loss: 0.7105 - val categorical accuracy: 0.6600
Epoch 153/500
uracy: 0.7767 - val_loss: 0.6403 - val_categorical_accuracy: 0.7200
Epoch 154/500
uracy: 0.7670 - val_loss: 0.7253 - val_categorical_accuracy: 0.5800
Epoch 155/500
uracy: 0.7767 - val_loss: 0.6426 - val_categorical_accuracy: 0.7000
Epoch 156/500
uracy: 0.7573 - val_loss: 0.6474 - val_categorical_accuracy: 0.6400
Epoch 157/500
```

```
uracy: 0.7816 - val loss: 0.7152 - val categorical accuracy: 0.7000
Epoch 158/500
uracy: 0.7621 - val_loss: 0.5751 - val_categorical_accuracy: 0.7600
Epoch 159/500
uracy: 0.7621 - val_loss: 0.5522 - val_categorical_accuracy: 0.7800
Epoch 160/500
uracy: 0.8107 - val_loss: 0.6632 - val_categorical_accuracy: 0.7400
Epoch 161/500
uracy: 0.7913 - val_loss: 0.7154 - val_categorical_accuracy: 0.7000
uracy: 0.7476 - val_loss: 0.6978 - val_categorical_accuracy: 0.6400
Epoch 163/500
uracy: 0.7816 - val loss: 0.6426 - val categorical accuracy: 0.6800
Epoch 164/500
uracy: 0.7524 - val loss: 0.6600 - val categorical accuracy: 0.7000
Epoch 165/500
uracy: 0.7718 - val_loss: 0.6410 - val_categorical_accuracy: 0.7600
uracy: 0.7524 - val_loss: 0.7462 - val_categorical_accuracy: 0.6400
Epoch 167/500
uracy: 0.7913 - val_loss: 0.7427 - val_categorical_accuracy: 0.6800
Epoch 168/500
uracy: 0.7767 - val_loss: 0.6857 - val_categorical_accuracy: 0.7200
uracy: 0.7816 - val_loss: 0.6365 - val_categorical_accuracy: 0.6800
Epoch 170/500
uracy: 0.7864 - val_loss: 0.7030 - val_categorical_accuracy: 0.6800
Epoch 171/500
uracy: 0.8155 - val_loss: 0.6679 - val_categorical_accuracy: 0.6600
Epoch 172/500
uracy: 0.7864 - val loss: 0.7113 - val categorical accuracy: 0.6000
Epoch 173/500
uracy: 0.7816 - val_loss: 0.6886 - val_categorical_accuracy: 0.7200
Epoch 174/500
uracy: 0.7718 - val_loss: 0.6664 - val_categorical_accuracy: 0.7200
Epoch 175/500
uracy: 0.8010 - val_loss: 0.7428 - val_categorical_accuracy: 0.7400
Epoch 176/500
uracy: 0.8058 - val_loss: 0.6113 - val_categorical_accuracy: 0.7200
Epoch 177/500
```

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uracy: 0.7621 - val_loss: 0.6056 - val_categorical_accuracy: 0.7200
Epoch 178/500
uracy: 0.7816 - val loss: 0.7411 - val categorical accuracy: 0.6600
Epoch 179/500
uracy: 0.7767 - val_loss: 0.7093 - val_categorical_accuracy: 0.6400
Epoch 180/500
uracy: 0.7816 - val_loss: 0.7101 - val_categorical_accuracy: 0.6400
Epoch 181/500
uracy: 0.7718 - val_loss: 0.6531 - val_categorical_accuracy: 0.6600
uracy: 0.7767 - val_loss: 0.7552 - val_categorical_accuracy: 0.6600
Epoch 183/500
uracy: 0.8000 - val loss: 0.6147 - val categorical accuracy: 0.6400
Epoch 184/500
uracy: 0.7816 - val loss: 0.6806 - val categorical accuracy: 0.6800
Epoch 185/500
uracy: 0.7621 - val_loss: 0.6649 - val_categorical_accuracy: 0.7400
Epoch 186/500
uracy: 0.7816 - val_loss: 0.6958 - val_categorical_accuracy: 0.6800
Epoch 187/500
uracy: 0.8058 - val_loss: 0.5609 - val_categorical_accuracy: 0.7400
Epoch 188/500
uracy: 0.7864 - val_loss: 0.7479 - val_categorical_accuracy: 0.6800
uracy: 0.8204 - val_loss: 0.5969 - val_categorical_accuracy: 0.7400
Epoch 190/500
uracy: 0.8010 - val_loss: 0.6383 - val_categorical_accuracy: 0.7800
Epoch 191/500
uracy: 0.7767 - val_loss: 0.6310 - val_categorical_accuracy: 0.7200
Epoch 192/500
uracy: 0.8058 - val loss: 0.6838 - val categorical accuracy: 0.6600
Epoch 193/500
uracy: 0.7621 - val_loss: 0.7220 - val_categorical_accuracy: 0.6400
Epoch 194/500
uracy: 0.8204 - val_loss: 0.7089 - val_categorical_accuracy: 0.7400
Epoch 195/500
uracy: 0.7864 - val_loss: 0.6131 - val_categorical_accuracy: 0.7000
Epoch 196/500
uracy: 0.7476 - val_loss: 0.6436 - val_categorical_accuracy: 0.7200
Epoch 197/500
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uracy: 0.7913 - val loss: 0.6058 - val categorical accuracy: 0.6800
Epoch 198/500
uracy: 0.8058 - val loss: 0.6343 - val categorical accuracy: 0.6800
Epoch 199/500
uracy: 0.7621 - val_loss: 0.6927 - val_categorical_accuracy: 0.7400
Epoch 200/500
uracy: 0.8204 - val loss: 0.7500 - val categorical accuracy: 0.6400
Epoch 201/500
uracy: 0.8010 - val_loss: 0.7077 - val_categorical_accuracy: 0.6400
uracy: 0.8107 - val_loss: 0.5865 - val_categorical_accuracy: 0.7600
Epoch 203/500
uracy: 0.7718 - val loss: 0.6647 - val categorical accuracy: 0.6800
Epoch 204/500
uracy: 0.8058 - val loss: 0.6556 - val categorical accuracy: 0.6600
Epoch 205/500
uracy: 0.7913 - val_loss: 0.6152 - val_categorical_accuracy: 0.7400
uracy: 0.7573 - val_loss: 0.5776 - val_categorical_accuracy: 0.7000
Epoch 207/500
uracy: 0.7905 - val_loss: 0.6833 - val_categorical_accuracy: 0.6400
Epoch 208/500
uracy: 0.7961 - val_loss: 0.6793 - val_categorical_accuracy: 0.7400
uracy: 0.7718 - val_loss: 0.6374 - val_categorical_accuracy: 0.7200
Epoch 210/500
uracy: 0.7767 - val_loss: 0.6513 - val_categorical_accuracy: 0.6600
Epoch 211/500
uracy: 0.7718 - val_loss: 0.6194 - val_categorical_accuracy: 0.7200
Epoch 212/500
uracy: 0.7913 - val loss: 0.6653 - val categorical accuracy: 0.6400
Epoch 213/500
uracy: 0.7913 - val_loss: 0.7259 - val_categorical_accuracy: 0.6400
Epoch 214/500
uracy: 0.7913 - val_loss: 0.6607 - val_categorical_accuracy: 0.6800
Epoch 215/500
uracy: 0.8058 - val_loss: 0.7039 - val_categorical_accuracy: 0.7200
Epoch 216/500
uracy: 0.8252 - val_loss: 0.6685 - val_categorical_accuracy: 0.6800
Epoch 217/500
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uracy: 0.7670 - val_loss: 0.6666 - val_categorical_accuracy: 0.7200
Epoch 218/500
uracy: 0.7913 - val loss: 0.6870 - val categorical accuracy: 0.6200
Epoch 219/500
uracy: 0.7816 - val_loss: 0.6686 - val_categorical_accuracy: 0.6800
Epoch 220/500
uracy: 0.8107 - val loss: 0.7399 - val categorical accuracy: 0.6400
Epoch 221/500
uracy: 0.7767 - val_loss: 0.6956 - val_categorical_accuracy: 0.6200
uracy: 0.7670 - val_loss: 0.5932 - val_categorical_accuracy: 0.7200
Epoch 223/500
uracy: 0.7913 - val loss: 0.7676 - val categorical accuracy: 0.6600
Epoch 224/500
uracy: 0.7718 - val loss: 0.6810 - val categorical accuracy: 0.7200
Epoch 225/500
uracy: 0.7864 - val_loss: 0.6906 - val_categorical_accuracy: 0.6800
Epoch 226/500
uracy: 0.8204 - val_loss: 0.7155 - val_categorical_accuracy: 0.6600
Epoch 227/500
uracy: 0.8350 - val_loss: 0.6732 - val_categorical_accuracy: 0.6400
Epoch 228/500
uracy: 0.7864 - val_loss: 0.6643 - val_categorical_accuracy: 0.6400
uracy: 0.7767 - val_loss: 0.6669 - val_categorical_accuracy: 0.6800
Epoch 230/500
uracy: 0.8301 - val_loss: 0.6519 - val_categorical_accuracy: 0.7000
Epoch 231/500
uracy: 0.8058 - val_loss: 0.5853 - val_categorical_accuracy: 0.6800
Epoch 232/500
uracy: 0.7816 - val loss: 0.6421 - val categorical accuracy: 0.7400
Epoch 233/500
uracy: 0.8107 - val_loss: 0.6296 - val_categorical_accuracy: 0.7000
Epoch 234/500
uracy: 0.8155 - val_loss: 0.5566 - val_categorical_accuracy: 0.7800
Epoch 235/500
uracy: 0.7864 - val_loss: 0.7274 - val_categorical_accuracy: 0.7200
Epoch 236/500
uracy: 0.7670 - val_loss: 0.7166 - val_categorical_accuracy: 0.6600
Epoch 237/500
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uracy: 0.7816 - val loss: 0.5799 - val categorical accuracy: 0.7600
Epoch 238/500
uracy: 0.7913 - val loss: 0.5992 - val categorical accuracy: 0.7200
Epoch 239/500
uracy: 0.7816 - val_loss: 0.6527 - val_categorical_accuracy: 0.6800
Epoch 240/500
uracy: 0.7767 - val_loss: 0.7102 - val_categorical_accuracy: 0.6600
Epoch 241/500
uracy: 0.8155 - val_loss: 0.6402 - val_categorical_accuracy: 0.6800
uracy: 0.7913 - val_loss: 0.6338 - val_categorical_accuracy: 0.6800
Epoch 243/500
uracy: 0.8155 - val loss: 0.6196 - val categorical accuracy: 0.7000
Epoch 244/500
uracy: 0.8252 - val loss: 0.6101 - val categorical accuracy: 0.6600
Epoch 245/500
uracy: 0.8095 - val_loss: 0.7619 - val_categorical_accuracy: 0.6400
uracy: 0.8155 - val_loss: 0.6658 - val_categorical_accuracy: 0.6600
Epoch 247/500
uracy: 0.8010 - val_loss: 0.6548 - val_categorical_accuracy: 0.6800
Epoch 248/500
uracy: 0.7767 - val_loss: 0.6949 - val_categorical_accuracy: 0.6600
uracy: 0.8048 - val_loss: 0.6564 - val_categorical_accuracy: 0.7200
Epoch 250/500
uracy: 0.7670 - val_loss: 0.6543 - val_categorical_accuracy: 0.6600
Epoch 251/500
uracy: 0.8058 - val_loss: 0.6005 - val_categorical_accuracy: 0.7000
Epoch 252/500
uracy: 0.8107 - val loss: 0.6071 - val categorical accuracy: 0.7400
Epoch 253/500
uracy: 0.7621 - val_loss: 0.6627 - val_categorical_accuracy: 0.6600
Epoch 254/500
uracy: 0.7816 - val_loss: 0.7201 - val_categorical_accuracy: 0.6600
Epoch 255/500
uracy: 0.8107 - val_loss: 0.6810 - val_categorical_accuracy: 0.6200
Epoch 256/500
uracy: 0.8204 - val_loss: 0.6815 - val_categorical_accuracy: 0.6600
Epoch 257/500
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uracy: 0.8058 - val loss: 0.6437 - val categorical accuracy: 0.6800
Epoch 258/500
uracy: 0.8058 - val_loss: 0.7287 - val_categorical_accuracy: 0.6000
Epoch 259/500
uracy: 0.7767 - val_loss: 0.6999 - val_categorical_accuracy: 0.5800
Epoch 260/500
uracy: 0.7816 - val_loss: 0.6622 - val_categorical_accuracy: 0.7200
Epoch 261/500
uracy: 0.7621 - val_loss: 0.5491 - val_categorical_accuracy: 0.7000
uracy: 0.8252 - val_loss: 0.6330 - val_categorical_accuracy: 0.7000
Epoch 263/500
uracy: 0.8010 - val loss: 0.7563 - val categorical accuracy: 0.6000
Epoch 264/500
uracy: 0.8058 - val loss: 0.6686 - val categorical accuracy: 0.6400
Epoch 265/500
uracy: 0.8107 - val_loss: 0.5992 - val_categorical_accuracy: 0.7200
uracy: 0.7961 - val_loss: 0.5786 - val_categorical_accuracy: 0.6800
Epoch 267/500
uracy: 0.8058 - val_loss: 0.6344 - val_categorical_accuracy: 0.7200
Epoch 268/500
uracy: 0.8333 - val_loss: 0.5789 - val_categorical_accuracy: 0.7000
uracy: 0.8058 - val_loss: 0.5819 - val_categorical_accuracy: 0.7200
Epoch 270/500
uracy: 0.7718 - val_loss: 0.6629 - val_categorical_accuracy: 0.6600
Epoch 271/500
uracy: 0.7961 - val_loss: 0.6421 - val_categorical_accuracy: 0.7000
Epoch 272/500
uracy: 0.8252 - val loss: 0.6905 - val categorical accuracy: 0.6600
Epoch 273/500
uracy: 0.8350 - val_loss: 0.7072 - val_categorical_accuracy: 0.6600
Epoch 274/500
uracy: 0.7961 - val_loss: 0.7368 - val_categorical_accuracy: 0.6400
Epoch 275/500
uracy: 0.7961 - val_loss: 0.6804 - val_categorical_accuracy: 0.6600
Epoch 276/500
uracy: 0.8155 - val_loss: 0.6917 - val_categorical_accuracy: 0.6200
Epoch 277/500
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uracy: 0.8010 - val_loss: 0.5501 - val_categorical_accuracy: 0.8000
Epoch 278/500
uracy: 0.7864 - val loss: 0.6174 - val categorical accuracy: 0.7200
Epoch 279/500
uracy: 0.8058 - val_loss: 0.7890 - val_categorical_accuracy: 0.6400
Epoch 280/500
uracy: 0.7864 - val_loss: 0.7243 - val_categorical_accuracy: 0.6800
Epoch 281/500
uracy: 0.8204 - val_loss: 0.6337 - val_categorical_accuracy: 0.6800
uracy: 0.8204 - val_loss: 0.7319 - val_categorical_accuracy: 0.6600
Epoch 283/500
uracy: 0.8301 - val loss: 0.6333 - val categorical accuracy: 0.7400
Epoch 284/500
uracy: 0.8048 - val loss: 0.7514 - val categorical accuracy: 0.6800
Epoch 285/500
uracy: 0.8107 - val_loss: 0.6579 - val_categorical_accuracy: 0.7400
uracy: 0.8107 - val_loss: 0.6662 - val_categorical_accuracy: 0.6600
Epoch 287/500
uracy: 0.8204 - val_loss: 0.6551 - val_categorical_accuracy: 0.7200
Epoch 288/500
uracy: 0.8301 - val_loss: 0.6703 - val_categorical_accuracy: 0.6800
uracy: 0.7816 - val_loss: 0.6878 - val_categorical_accuracy: 0.7200
Epoch 290/500
uracy: 0.8301 - val_loss: 0.5696 - val_categorical_accuracy: 0.7400
Epoch 291/500
uracy: 0.8155 - val_loss: 0.6663 - val_categorical_accuracy: 0.6400
Epoch 292/500
uracy: 0.7913 - val loss: 0.6057 - val categorical accuracy: 0.7200
uracy: 0.8204 - val_loss: 0.6093 - val_categorical_accuracy: 0.6400
Epoch 294/500
uracy: 0.8204 - val_loss: 0.6500 - val_categorical_accuracy: 0.6800
Epoch 295/500
uracy: 0.7961 - val_loss: 0.6964 - val_categorical_accuracy: 0.6000
Epoch 296/500
uracy: 0.8058 - val_loss: 0.6547 - val_categorical_accuracy: 0.6800
Epoch 297/500
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uracy: 0.8252 - val_loss: 0.6655 - val_categorical_accuracy: 0.7200
Epoch 298/500
uracy: 0.8107 - val loss: 0.6973 - val categorical accuracy: 0.6800
Epoch 299/500
uracy: 0.8592 - val_loss: 0.6049 - val_categorical_accuracy: 0.6400
Epoch 300/500
uracy: 0.8238 - val loss: 0.5941 - val categorical accuracy: 0.7800
Epoch 301/500
uracy: 0.7816 - val_loss: 0.7048 - val_categorical_accuracy: 0.7000
uracy: 0.8155 - val_loss: 0.7153 - val_categorical_accuracy: 0.6000
Epoch 303/500
uracy: 0.8058 - val loss: 0.7078 - val categorical accuracy: 0.6400
Epoch 304/500
uracy: 0.7961 - val loss: 0.6536 - val categorical accuracy: 0.7600
Epoch 305/500
uracy: 0.8204 - val_loss: 0.6584 - val_categorical_accuracy: 0.6400
uracy: 0.8058 - val_loss: 0.5631 - val_categorical_accuracy: 0.8000
Epoch 307/500
uracy: 0.7961 - val_loss: 0.6824 - val_categorical_accuracy: 0.6200
Epoch 308/500
uracy: 0.8398 - val_loss: 0.6174 - val_categorical_accuracy: 0.7000
uracy: 0.7961 - val_loss: 0.6556 - val_categorical_accuracy: 0.7000
uracy: 0.7864 - val_loss: 0.6409 - val_categorical_accuracy: 0.6600
Epoch 311/500
uracy: 0.8447 - val_loss: 0.5998 - val_categorical_accuracy: 0.7400
Epoch 312/500
uracy: 0.8058 - val loss: 0.6100 - val categorical accuracy: 0.7200
Epoch 313/500
uracy: 0.8107 - val_loss: 0.6194 - val_categorical_accuracy: 0.7400
Epoch 314/500
uracy: 0.8398 - val_loss: 0.6364 - val_categorical_accuracy: 0.7000
Epoch 315/500
uracy: 0.8155 - val_loss: 0.6138 - val_categorical_accuracy: 0.6800
Epoch 316/500
uracy: 0.8350 - val_loss: 0.6121 - val_categorical_accuracy: 0.7200
Epoch 317/500
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uracy: 0.8204 - val loss: 0.7032 - val categorical accuracy: 0.6600
Epoch 318/500
uracy: 0.8155 - val loss: 0.6822 - val categorical accuracy: 0.6600
Epoch 319/500
uracy: 0.8155 - val_loss: 0.8182 - val_categorical_accuracy: 0.6400
Epoch 320/500
uracy: 0.8544 - val_loss: 0.7251 - val_categorical_accuracy: 0.6400
Epoch 321/500
uracy: 0.8350 - val_loss: 0.6988 - val_categorical_accuracy: 0.7000
uracy: 0.7961 - val_loss: 0.8105 - val_categorical_accuracy: 0.6600
Epoch 323/500
uracy: 0.7816 - val loss: 0.6308 - val categorical accuracy: 0.7200
Epoch 324/500
uracy: 0.8301 - val loss: 0.4931 - val categorical accuracy: 0.7800
Epoch 325/500
uracy: 0.7767 - val_loss: 0.5362 - val_categorical_accuracy: 0.7800
Epoch 326/500
uracy: 0.8010 - val_loss: 0.6861 - val_categorical_accuracy: 0.6400
Epoch 327/500
uracy: 0.8107 - val_loss: 0.6720 - val_categorical_accuracy: 0.6600
Epoch 328/500
uracy: 0.8204 - val_loss: 0.6219 - val_categorical_accuracy: 0.7200
uracy: 0.8058 - val_loss: 0.6314 - val_categorical_accuracy: 0.6800
uracy: 0.8204 - val_loss: 0.6737 - val_categorical_accuracy: 0.7200
Epoch 331/500
uracy: 0.7670 - val_loss: 0.5700 - val_categorical_accuracy: 0.7800
Epoch 332/500
uracy: 0.7961 - val loss: 0.6157 - val categorical accuracy: 0.7200
Epoch 333/500
uracy: 0.8495 - val_loss: 0.7174 - val_categorical_accuracy: 0.5800
Epoch 334/500
uracy: 0.7670 - val_loss: 0.7013 - val_categorical_accuracy: 0.6600
Epoch 335/500
uracy: 0.8398 - val_loss: 0.5837 - val_categorical_accuracy: 0.7600
Epoch 336/500
uracy: 0.8010 - val_loss: 0.6561 - val_categorical_accuracy: 0.7200
Epoch 337/500
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uracy: 0.7762 - val loss: 0.6865 - val categorical accuracy: 0.7200
Epoch 338/500
uracy: 0.8689 - val loss: 0.5472 - val categorical accuracy: 0.7200
Epoch 339/500
uracy: 0.7816 - val_loss: 0.6250 - val_categorical_accuracy: 0.6600
Epoch 340/500
uracy: 0.8010 - val_loss: 0.6550 - val_categorical_accuracy: 0.6600
Epoch 341/500
uracy: 0.7864 - val_loss: 0.6339 - val_categorical_accuracy: 0.7800
uracy: 0.8301 - val_loss: 0.6384 - val_categorical_accuracy: 0.6400
Epoch 343/500
uracy: 0.8252 - val loss: 0.6400 - val categorical accuracy: 0.7000
Epoch 344/500
uracy: 0.8010 - val loss: 0.5456 - val categorical accuracy: 0.7200
Epoch 345/500
uracy: 0.8107 - val_loss: 0.6040 - val_categorical_accuracy: 0.7200
uracy: 0.8155 - val_loss: 0.5854 - val_categorical_accuracy: 0.7200
Epoch 347/500
uracy: 0.8155 - val_loss: 0.7586 - val_categorical_accuracy: 0.7000
Epoch 348/500
uracy: 0.8155 - val_loss: 0.6990 - val_categorical_accuracy: 0.6400
uracy: 0.8155 - val_loss: 0.6990 - val_categorical_accuracy: 0.7000
Epoch 350/500
uracy: 0.8301 - val_loss: 0.6735 - val_categorical_accuracy: 0.6800
Epoch 351/500
uracy: 0.8252 - val_loss: 0.7496 - val_categorical_accuracy: 0.6400
Epoch 352/500
uracy: 0.7961 - val loss: 0.7007 - val categorical accuracy: 0.6400
Epoch 353/500
uracy: 0.7961 - val_loss: 0.6018 - val_categorical_accuracy: 0.7800
Epoch 354/500
uracy: 0.8252 - val_loss: 0.7391 - val_categorical_accuracy: 0.6400
Epoch 355/500
uracy: 0.8058 - val_loss: 0.5911 - val_categorical_accuracy: 0.7000
Epoch 356/500
uracy: 0.8350 - val_loss: 0.6952 - val_categorical_accuracy: 0.6800
Epoch 357/500
```

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uracy: 0.8204 - val_loss: 0.6151 - val_categorical_accuracy: 0.6600
Epoch 358/500
uracy: 0.8301 - val loss: 0.6386 - val categorical accuracy: 0.6800
Epoch 359/500
uracy: 0.8301 - val_loss: 0.5848 - val_categorical_accuracy: 0.7800
Epoch 360/500
uracy: 0.8058 - val_loss: 0.6743 - val_categorical_accuracy: 0.6800
Epoch 361/500
uracy: 0.8204 - val_loss: 0.6475 - val_categorical_accuracy: 0.7400
uracy: 0.8155 - val_loss: 0.5579 - val_categorical_accuracy: 0.7200
Epoch 363/500
uracy: 0.8398 - val loss: 0.7069 - val categorical accuracy: 0.6000
Epoch 364/500
uracy: 0.7864 - val loss: 0.6783 - val categorical accuracy: 0.7000
Epoch 365/500
uracy: 0.8252 - val_loss: 0.7408 - val_categorical_accuracy: 0.6000
uracy: 0.8447 - val_loss: 0.8126 - val_categorical_accuracy: 0.7000
Epoch 367/500
uracy: 0.8301 - val_loss: 0.6383 - val_categorical_accuracy: 0.6800
Epoch 368/500
uracy: 0.8495 - val_loss: 0.7046 - val_categorical_accuracy: 0.7600
uracy: 0.8495 - val_loss: 0.6310 - val_categorical_accuracy: 0.7200
Epoch 370/500
uracy: 0.8495 - val_loss: 0.6672 - val_categorical_accuracy: 0.6800
Epoch 371/500
uracy: 0.7961 - val_loss: 0.5753 - val_categorical_accuracy: 0.7400
Epoch 372/500
uracy: 0.8495 - val loss: 0.6524 - val categorical accuracy: 0.7400
Epoch 373/500
uracy: 0.8010 - val_loss: 0.6069 - val_categorical_accuracy: 0.7200
Epoch 374/500
uracy: 0.8155 - val_loss: 0.6828 - val_categorical_accuracy: 0.6800
Epoch 375/500
uracy: 0.8252 - val_loss: 0.6710 - val_categorical_accuracy: 0.6800
Epoch 376/500
uracy: 0.7961 - val_loss: 0.5922 - val_categorical_accuracy: 0.7400
Epoch 377/500
```

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uracy: 0.8447 - val loss: 0.6358 - val categorical accuracy: 0.6800
Epoch 378/500
uracy: 0.8252 - val loss: 0.7140 - val categorical accuracy: 0.6600
Epoch 379/500
uracy: 0.8350 - val_loss: 0.5330 - val_categorical_accuracy: 0.7800
Epoch 380/500
uracy: 0.7718 - val_loss: 0.6027 - val_categorical_accuracy: 0.6400
Epoch 381/500
uracy: 0.7961 - val_loss: 0.6476 - val_categorical_accuracy: 0.6800
uracy: 0.8190 - val_loss: 0.7255 - val_categorical_accuracy: 0.6800
Epoch 383/500
uracy: 0.8301 - val loss: 0.6368 - val categorical accuracy: 0.6800
Epoch 384/500
uracy: 0.8252 - val loss: 0.5372 - val categorical accuracy: 0.7600
Epoch 385/500
uracy: 0.8333 - val_loss: 0.6953 - val_categorical_accuracy: 0.6800
uracy: 0.8495 - val_loss: 0.5890 - val_categorical_accuracy: 0.7800
Epoch 387/500
uracy: 0.8204 - val_loss: 0.6029 - val_categorical_accuracy: 0.7000
Epoch 388/500
uracy: 0.8398 - val_loss: 0.8315 - val_categorical_accuracy: 0.6600
uracy: 0.8204 - val_loss: 0.6658 - val_categorical_accuracy: 0.7000
Epoch 390/500
uracy: 0.8350 - val_loss: 0.6861 - val_categorical_accuracy: 0.6800
Epoch 391/500
uracy: 0.8204 - val_loss: 0.5046 - val_categorical_accuracy: 0.8000
Epoch 392/500
uracy: 0.8107 - val loss: 0.7161 - val categorical accuracy: 0.6200
Epoch 393/500
uracy: 0.7864 - val_loss: 0.6450 - val_categorical_accuracy: 0.7000
Epoch 394/500
uracy: 0.8204 - val_loss: 0.7090 - val_categorical_accuracy: 0.6600
Epoch 395/500
uracy: 0.8107 - val_loss: 0.6905 - val_categorical_accuracy: 0.7000
Epoch 396/500
uracy: 0.8641 - val_loss: 0.7451 - val_categorical_accuracy: 0.6600
Epoch 397/500
```

```
uracy: 0.8058 - val loss: 0.6899 - val categorical accuracy: 0.7400
Epoch 398/500
uracy: 0.8252 - val loss: 0.6134 - val categorical accuracy: 0.7400
Epoch 399/500
uracy: 0.8204 - val_loss: 0.6759 - val_categorical_accuracy: 0.7000
Epoch 400/500
uracy: 0.8544 - val loss: 0.6700 - val categorical accuracy: 0.6600
Epoch 401/500
uracy: 0.8350 - val_loss: 0.6249 - val_categorical_accuracy: 0.6800
uracy: 0.8447 - val_loss: 0.5967 - val_categorical_accuracy: 0.7600
Epoch 403/500
uracy: 0.8107 - val loss: 0.5873 - val categorical accuracy: 0.7200
Epoch 404/500
uracy: 0.8155 - val loss: 0.5678 - val categorical accuracy: 0.7200
Epoch 405/500
uracy: 0.8495 - val_loss: 0.5853 - val_categorical_accuracy: 0.7800
Epoch 406/500
uracy: 0.8204 - val_loss: 0.6781 - val_categorical_accuracy: 0.6600
Epoch 407/500
uracy: 0.8058 - val_loss: 0.7146 - val_categorical_accuracy: 0.6000
Epoch 408/500
uracy: 0.8107 - val_loss: 0.6188 - val_categorical_accuracy: 0.7800
uracy: 0.8447 - val_loss: 0.6158 - val_categorical_accuracy: 0.7400
Epoch 410/500
uracy: 0.8495 - val_loss: 0.5688 - val_categorical_accuracy: 0.7600
Epoch 411/500
uracy: 0.8155 - val_loss: 0.6952 - val_categorical_accuracy: 0.7200
Epoch 412/500
uracy: 0.8398 - val loss: 0.6529 - val categorical accuracy: 0.7000
Epoch 413/500
uracy: 0.8333 - val_loss: 0.5158 - val_categorical_accuracy: 0.7200
Epoch 414/500
uracy: 0.8155 - val_loss: 0.6059 - val_categorical_accuracy: 0.7200
Epoch 415/500
uracy: 0.8350 - val_loss: 0.6844 - val_categorical_accuracy: 0.7000
Epoch 416/500
uracy: 0.8252 - val_loss: 0.7333 - val_categorical_accuracy: 0.6800
Epoch 417/500
```

```
uracy: 0.8155 - val loss: 0.7434 - val categorical accuracy: 0.6200
Epoch 418/500
uracy: 0.8058 - val loss: 0.5129 - val categorical accuracy: 0.8000
Epoch 419/500
uracy: 0.8238 - val_loss: 0.5660 - val_categorical_accuracy: 0.7400
Epoch 420/500
uracy: 0.8398 - val_loss: 0.7318 - val_categorical_accuracy: 0.6600
Epoch 421/500
uracy: 0.8204 - val_loss: 0.6954 - val_categorical_accuracy: 0.7400
uracy: 0.8252 - val_loss: 0.7025 - val_categorical_accuracy: 0.6200
Epoch 423/500
uracy: 0.8204 - val loss: 0.6817 - val categorical accuracy: 0.6800
Epoch 424/500
uracy: 0.8738 - val loss: 0.6261 - val categorical accuracy: 0.7000
Epoch 425/500
uracy: 0.8301 - val_loss: 0.7493 - val_categorical_accuracy: 0.6800
Epoch 426/500
uracy: 0.8350 - val_loss: 0.5859 - val_categorical_accuracy: 0.7400
Epoch 427/500
uracy: 0.8058 - val_loss: 0.5629 - val_categorical_accuracy: 0.7200
Epoch 428/500
uracy: 0.8155 - val_loss: 0.5969 - val_categorical_accuracy: 0.7400
uracy: 0.8495 - val_loss: 0.5839 - val_categorical_accuracy: 0.7800
Epoch 430/500
uracy: 0.8350 - val_loss: 0.6880 - val_categorical_accuracy: 0.7000
Epoch 431/500
uracy: 0.8544 - val_loss: 0.7100 - val_categorical_accuracy: 0.7000
Epoch 432/500
uracy: 0.8155 - val loss: 0.7050 - val categorical accuracy: 0.6800
Epoch 433/500
uracy: 0.8252 - val_loss: 0.7191 - val_categorical_accuracy: 0.6800
Epoch 434/500
uracy: 0.8204 - val_loss: 0.6177 - val_categorical_accuracy: 0.7200
Epoch 435/500
uracy: 0.8058 - val_loss: 0.6694 - val_categorical_accuracy: 0.7000
Epoch 436/500
uracy: 0.8544 - val_loss: 0.6664 - val_categorical_accuracy: 0.7200
Epoch 437/500
```

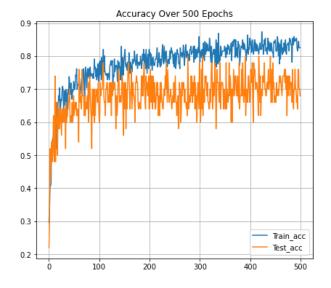
```
uracy: 0.8204 - val loss: 0.5099 - val categorical accuracy: 0.7600
Epoch 438/500
uracy: 0.8447 - val loss: 0.6506 - val categorical accuracy: 0.7000
Epoch 439/500
uracy: 0.8155 - val_loss: 0.5537 - val_categorical_accuracy: 0.7200
Epoch 440/500
uracy: 0.8155 - val_loss: 0.6082 - val_categorical_accuracy: 0.7200
Epoch 441/500
uracy: 0.8398 - val loss: 0.5801 - val categorical accuracy: 0.8000
uracy: 0.8155 - val_loss: 0.6662 - val_categorical_accuracy: 0.6600
Epoch 443/500
uracy: 0.8398 - val loss: 0.7844 - val categorical accuracy: 0.6600
Epoch 444/500
uracy: 0.8252 - val loss: 0.5982 - val categorical accuracy: 0.7600
Epoch 445/500
uracy: 0.8107 - val_loss: 0.7075 - val_categorical_accuracy: 0.6600
Epoch 446/500
uracy: 0.8107 - val_loss: 0.5402 - val_categorical_accuracy: 0.7800
Epoch 447/500
uracy: 0.7961 - val_loss: 0.6846 - val_categorical_accuracy: 0.7000
Epoch 448/500
uracy: 0.8447 - val_loss: 0.6282 - val_categorical_accuracy: 0.7200
uracy: 0.8398 - val_loss: 0.6233 - val_categorical_accuracy: 0.6600
Epoch 450/500
uracy: 0.8398 - val_loss: 0.6921 - val_categorical_accuracy: 0.6800
Epoch 451/500
uracy: 0.8495 - val_loss: 0.7043 - val_categorical_accuracy: 0.6600
Epoch 452/500
uracy: 0.8155 - val loss: 0.5662 - val categorical accuracy: 0.6600
Epoch 453/500
uracy: 0.8155 - val_loss: 0.6938 - val_categorical_accuracy: 0.7200
Epoch 454/500
uracy: 0.8155 - val_loss: 0.6686 - val_categorical_accuracy: 0.6800
Epoch 455/500
uracy: 0.8592 - val_loss: 0.5904 - val_categorical_accuracy: 0.7000
Epoch 456/500
uracy: 0.8447 - val_loss: 0.7660 - val_categorical_accuracy: 0.7000
Epoch 457/500
```

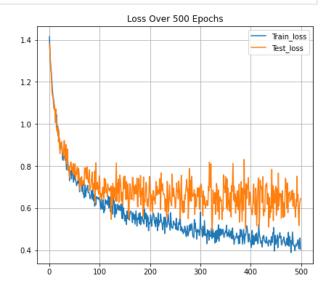
```
uracy: 0.8010 - val loss: 0.6104 - val categorical accuracy: 0.6800
Epoch 458/500
uracy: 0.8398 - val loss: 0.6432 - val categorical accuracy: 0.7000
Epoch 459/500
uracy: 0.7767 - val_loss: 0.6700 - val_categorical_accuracy: 0.7400
Epoch 460/500
uracy: 0.8350 - val_loss: 0.6537 - val_categorical_accuracy: 0.7200
Epoch 461/500
uracy: 0.7913 - val_loss: 0.6684 - val_categorical_accuracy: 0.7400
uracy: 0.8447 - val_loss: 0.5526 - val_categorical_accuracy: 0.7200
Epoch 463/500
uracy: 0.8495 - val loss: 0.6498 - val categorical accuracy: 0.7000
Epoch 464/500
uracy: 0.8107 - val loss: 0.6831 - val categorical accuracy: 0.6800
Epoch 465/500
uracy: 0.8048 - val_loss: 0.5975 - val_categorical_accuracy: 0.7200
Epoch 466/500
uracy: 0.8252 - val_loss: 0.5550 - val_categorical_accuracy: 0.7400
Epoch 467/500
uracy: 0.8107 - val_loss: 0.6268 - val_categorical_accuracy: 0.7400
Epoch 468/500
uracy: 0.8398 - val_loss: 0.5602 - val_categorical_accuracy: 0.7400
uracy: 0.8252 - val_loss: 0.6359 - val_categorical_accuracy: 0.7600
Epoch 470/500
uracy: 0.8155 - val_loss: 0.7417 - val_categorical_accuracy: 0.7200
Epoch 471/500
uracy: 0.8495 - val_loss: 0.7297 - val_categorical_accuracy: 0.7600
Epoch 472/500
uracy: 0.8107 - val loss: 0.6838 - val categorical accuracy: 0.7000
Epoch 473/500
uracy: 0.8058 - val_loss: 0.5992 - val_categorical_accuracy: 0.7400
Epoch 474/500
uracy: 0.8350 - val_loss: 0.7189 - val_categorical_accuracy: 0.6200
Epoch 475/500
uracy: 0.8350 - val_loss: 0.5782 - val_categorical_accuracy: 0.7400
Epoch 476/500
uracy: 0.8398 - val_loss: 0.6689 - val_categorical_accuracy: 0.6800
Epoch 477/500
```

```
uracy: 0.8398 - val loss: 0.6603 - val categorical accuracy: 0.7400
Epoch 478/500
uracy: 0.8155 - val loss: 0.6763 - val categorical accuracy: 0.7000
Epoch 479/500
uracy: 0.8544 - val_loss: 0.6302 - val_categorical_accuracy: 0.7400
Epoch 480/500
uracy: 0.8447 - val loss: 0.6427 - val categorical accuracy: 0.6800
Epoch 481/500
uracy: 0.8495 - val_loss: 0.6396 - val_categorical_accuracy: 0.7200
uracy: 0.8350 - val_loss: 0.6812 - val_categorical_accuracy: 0.7200
Epoch 483/500
uracy: 0.8350 - val loss: 0.6431 - val categorical accuracy: 0.7200
Epoch 484/500
uracy: 0.7816 - val loss: 0.6852 - val categorical accuracy: 0.6600
Epoch 485/500
uracy: 0.8544 - val_loss: 0.6087 - val_categorical_accuracy: 0.7600
Epoch 486/500
uracy: 0.8350 - val_loss: 0.6606 - val_categorical_accuracy: 0.6800
Epoch 487/500
uracy: 0.8350 - val_loss: 0.6702 - val_categorical_accuracy: 0.6400
Epoch 488/500
uracy: 0.8592 - val_loss: 0.5993 - val_categorical_accuracy: 0.7000
uracy: 0.8544 - val_loss: 0.6044 - val_categorical_accuracy: 0.7000
Epoch 490/500
uracy: 0.8447 - val_loss: 0.5570 - val_categorical_accuracy: 0.7600
Epoch 491/500
uracy: 0.8476 - val_loss: 0.6415 - val_categorical_accuracy: 0.7000
Epoch 492/500
uracy: 0.8495 - val loss: 0.6066 - val categorical accuracy: 0.7000
Epoch 493/500
uracy: 0.8544 - val_loss: 0.7514 - val_categorical_accuracy: 0.6600
Epoch 494/500
uracy: 0.8252 - val_loss: 0.6310 - val_categorical_accuracy: 0.6800
Epoch 495/500
uracy: 0.8155 - val_loss: 0.6661 - val_categorical_accuracy: 0.6400
Epoch 496/500
uracy: 0.8252 - val_loss: 0.5709 - val_categorical_accuracy: 0.7600
Epoch 497/500
```

[5 points] Plot Accuracy and Loss During Training

```
In [126...
          import matplotlib.pyplot as plt
          fig, (ax1, ax2) = plt.subplots(1, 2)
          fig.set_figheight(6)
          fig.set figwidth(15)
          ax1.plot(res.history['categorical_accuracy'])
          ax1.plot(res.history['val_categorical_accuracy'])
          ax1.set_title('Accuracy Over ' + str(NUM_EPOCHS) + ' Epochs')
          ax1.legend(['Train_acc', 'Test_acc'], loc='lower right')
          ax1.grid(True)
          ax2.set title('Loss Over ' + str(NUM EPOCHS) + ' Epochs')
          ax2.plot(res.history['loss'])
          ax2.plot(res.history['val loss'])
          ax2.legend(['Train loss', 'Test loss'], loc='upper right')
          ax2.grid(True)
          plt.show()
```





Testing Model

Model 2: AlexNet

[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 [128...
          # from https://towardsdatascience.com/implementing-alexnet-cnn-architecture-using-tenso
          model = tf.keras.models.Sequential([
              tf.keras.layers.Conv2D(filters=96, kernel_size=(11,11), strides=(4,4), activation='
              tf.keras.layers.BatchNormalization(),
              tf.keras.layers.MaxPool2D(pool size=(3,3), strides=(2,2)),
              tf.keras.layers.Conv2D(filters=256, kernel size=(5,5), strides=(1,1), activation='r
              tf.keras.layers.BatchNormalization(),
              tf.keras.layers.MaxPool2D(pool_size=(3,3), strides=(2,2)),
              tf.keras.layers.Conv2D(filters=384, kernel_size=(3,3), strides=(1,1), activation='r
              tf.keras.layers.BatchNormalization(),
              tf.keras.layers.Conv2D(filters=384, kernel_size=(3,3), strides=(1,1), activation='r
              tf.keras.layers.BatchNormalization(),
              tf.keras.layers.Conv2D(filters=256, kernel_size=(3,3), strides=(1,1), activation='r
              tf.keras.layers.BatchNormalization(),
              tf.keras.layers.MaxPool2D(pool size=(3,3), strides=(2,2)),
              tf.keras.layers.Flatten(),
              tf.keras.layers.Dense(4096, activation='relu'),
              tf.keras.layers.Dropout(0.5),
              tf.keras.layers.Dense(4096, activation='relu'),
              tf.keras.layers.Dropout(0.5),
              tf.keras.layers.Dense(4, activation='softmax')
          ])
          model.compile(loss='categorical crossentropy', optimizer=tf.optimizers.SGD(lr=1e-7), me
```

In [129...

```
model.summary()
```

Model: "sequential_30"

Layer (type)	Output	Shap	oe		Param #
conv2d_50 (Conv2D)	(None,	54,	54,	96)	34944
batch_normalization_50 (Batc	(None,	54,	54,	96)	384
max_pooling2d_30 (MaxPooling	(None,	26,	26,	96)	0
conv2d_51 (Conv2D)	(None,	26,	26,	256)	614656

batch_normalization_51 (Batc	(None,	26, 26, 256)	1024
max_pooling2d_31 (MaxPooling	(None,	12, 12, 256)	0
conv2d_52 (Conv2D)	(None,	12, 12, 384)	885120
batch_normalization_52 (Batc	(None,	12, 12, 384)	1536
conv2d_53 (Conv2D)	(None,	12, 12, 384)	1327488
batch_normalization_53 (Batc	(None,	12, 12, 384)	1536
conv2d_54 (Conv2D)	(None,	12, 12, 256)	884992
batch_normalization_54 (Batc	(None,	12, 12, 256)	1024
max_pooling2d_32 (MaxPooling	(None,	5, 5, 256)	0
flatten_30 (Flatten)	(None,	6400)	0
dense_87 (Dense)	(None,	4096)	26218496
dropout_51 (Dropout)	(None,	4096)	0
dense_88 (Dense)	(None,	4096)	16781312
dropout_52 (Dropout)	(None,	4096)	0
dense_89 (Dense)	(None,	4)	16388
Total naname: 46 769 000			

Total params: 46,768,900 Trainable params: 46,766,148 Non-trainable params: 2,752

[5 points] Train Model

racy: 0.2718 - val_loss: 1.4285 - val_categorical_accuracy: 0.2600

Epoch 2/100

```
Epoch 3/100
racy: 0.2476 - val_loss: 1.4257 - val_categorical_accuracy: 0.2600
Epoch 4/100
racy: 0.2718 - val_loss: 1.4125 - val_categorical_accuracy: 0.2800
Epoch 5/100
racy: 0.2573 - val_loss: 1.4218 - val_categorical_accuracy: 0.2800
racy: 0.2087 - val_loss: 1.4867 - val_categorical_accuracy: 0.2600
Epoch 7/100
racy: 0.2282 - val_loss: 1.4934 - val_categorical_accuracy: 0.2600
Epoch 8/100
racy: 0.2816 - val_loss: 1.5868 - val_categorical_accuracy: 0.2400
Epoch 9/100
racy: 0.1942 - val_loss: 1.5387 - val_categorical_accuracy: 0.2800
Epoch 10/100
racy: 0.3010 - val_loss: 1.5172 - val_categorical_accuracy: 0.3600
Epoch 11/100
racy: 0.2571 - val_loss: 1.5545 - val_categorical_accuracy: 0.3400
Epoch 12/100
racy: 0.1942 - val_loss: 1.5722 - val_categorical_accuracy: 0.3000
Epoch 13/100
racy: 0.2952 - val loss: 1.7581 - val categorical accuracy: 0.2600
Epoch 14/100
racy: 0.2524 - val_loss: 1.7039 - val_categorical_accuracy: 0.3000
Epoch 15/100
racy: 0.2621 - val loss: 1.7320 - val categorical accuracy: 0.3200
Epoch 16/100
racy: 0.2718 - val loss: 1.6356 - val categorical accuracy: 0.3000
Epoch 17/100
racy: 0.3107 - val_loss: 1.6825 - val_categorical_accuracy: 0.3000
racy: 0.3010 - val_loss: 1.6258 - val_categorical_accuracy: 0.3400
Epoch 19/100
racy: 0.3143 - val_loss: 1.7134 - val_categorical_accuracy: 0.2400
Epoch 20/100
racy: 0.2087 - val_loss: 1.6353 - val_categorical_accuracy: 0.2800
Epoch 21/100
racy: 0.2427 - val_loss: 1.6709 - val_categorical_accuracy: 0.2800
Epoch 22/100
racy: 0.3010 - val_loss: 1.6204 - val_categorical_accuracy: 0.3200
```

```
Epoch 23/100
racy: 0.2670 - val_loss: 1.5241 - val_categorical_accuracy: 0.3000
Epoch 24/100
racy: 0.2864 - val_loss: 1.6931 - val_categorical_accuracy: 0.2400
Epoch 25/100
racy: 0.2427 - val_loss: 1.4648 - val_categorical_accuracy: 0.4000
Epoch 26/100
racy: 0.2184 - val_loss: 1.4986 - val_categorical_accuracy: 0.3400
Epoch 27/100
racy: 0.2864 - val_loss: 1.6133 - val_categorical_accuracy: 0.2800
Epoch 28/100
racy: 0.2573 - val_loss: 1.5099 - val_categorical_accuracy: 0.3200
Epoch 29/100
racy: 0.2427 - val_loss: 1.5298 - val_categorical_accuracy: 0.3400
Epoch 30/100
racy: 0.1748 - val_loss: 1.5871 - val_categorical_accuracy: 0.3000
Epoch 31/100
racy: 0.2136 - val_loss: 1.6541 - val_categorical_accuracy: 0.2600
Epoch 32/100
racy: 0.3252 - val_loss: 1.7945 - val_categorical_accuracy: 0.1600
Epoch 33/100
racy: 0.2913 - val loss: 1.6282 - val categorical accuracy: 0.2400
Epoch 34/100
racy: 0.2816 - val_loss: 1.4641 - val_categorical_accuracy: 0.3600
Epoch 35/100
racy: 0.2379 - val loss: 1.5833 - val categorical accuracy: 0.2800
Epoch 36/100
racy: 0.2864 - val loss: 1.6305 - val categorical accuracy: 0.3400
Epoch 37/100
racy: 0.2767 - val_loss: 1.7364 - val_categorical_accuracy: 0.2400
racy: 0.2864 - val_loss: 1.4907 - val_categorical_accuracy: 0.3000
Epoch 39/100
racy: 0.2864 - val_loss: 1.5621 - val_categorical_accuracy: 0.2400
Epoch 40/100
racy: 0.2718 - val_loss: 1.4995 - val_categorical_accuracy: 0.2400
Epoch 41/100
racy: 0.3010 - val_loss: 1.7151 - val_categorical_accuracy: 0.2600
Epoch 42/100
racy: 0.2961 - val_loss: 1.4888 - val_categorical_accuracy: 0.3200
```

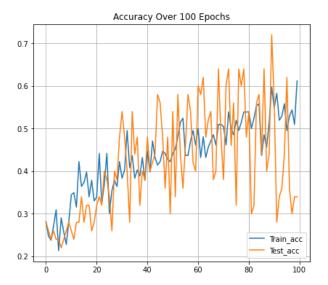
```
Epoch 43/100
racy: 0.2621 - val_loss: 1.6893 - val_categorical_accuracy: 0.2600
Epoch 44/100
racy: 0.2282 - val_loss: 1.5662 - val_categorical_accuracy: 0.3000
Epoch 45/100
racy: 0.2767 - val_loss: 1.5912 - val_categorical_accuracy: 0.3000
racy: 0.2718 - val_loss: 1.6003 - val_categorical_accuracy: 0.3000
Epoch 47/100
racy: 0.2670 - val_loss: 1.7495 - val_categorical_accuracy: 0.2800
Epoch 48/100
racy: 0.2476 - val_loss: 1.5963 - val_categorical_accuracy: 0.3200
Epoch 49/100
racy: 0.2767 - val_loss: 1.5506 - val_categorical_accuracy: 0.2400
Epoch 50/100
racy: 0.2427 - val_loss: 1.5835 - val_categorical_accuracy: 0.2800
Epoch 51/100
racy: 0.2427 - val_loss: 1.4519 - val_categorical_accuracy: 0.3400
Epoch 52/100
racy: 0.2379 - val_loss: 1.6366 - val_categorical_accuracy: 0.2800
Epoch 53/100
racy: 0.2330 - val loss: 1.5327 - val categorical accuracy: 0.3000
Epoch 54/100
racy: 0.2718 - val_loss: 1.6341 - val_categorical_accuracy: 0.3000
Epoch 55/100
racy: 0.1990 - val loss: 1.3422 - val categorical accuracy: 0.4400
Epoch 56/100
racy: 0.3447 - val loss: 1.6218 - val categorical accuracy: 0.2800
Epoch 57/100
racy: 0.2136 - val_loss: 1.5937 - val_categorical_accuracy: 0.3800
racy: 0.2476 - val_loss: 1.5012 - val_categorical_accuracy: 0.2800
Epoch 59/100
racy: 0.2767 - val_loss: 1.5073 - val_categorical_accuracy: 0.2800
Epoch 60/100
racy: 0.2330 - val_loss: 1.4351 - val_categorical_accuracy: 0.3800
Epoch 61/100
racy: 0.2621 - val_loss: 1.3189 - val_categorical_accuracy: 0.4400
Epoch 62/100
racy: 0.2670 - val_loss: 1.3600 - val_categorical_accuracy: 0.3400
```

Epoch 63/100

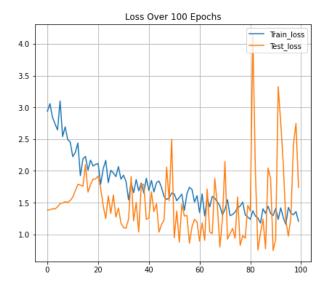
```
racy: 0.2476 - val_loss: 1.4634 - val_categorical_accuracy: 0.2800
Epoch 64/100
racy: 0.2330 - val loss: 1.3874 - val categorical accuracy: 0.3800
Epoch 65/100
racy: 0.2379 - val_loss: 1.5295 - val_categorical_accuracy: 0.2600
Epoch 66/100
racy: 0.2961 - val_loss: 1.5572 - val_categorical_accuracy: 0.3000
Epoch 67/100
racy: 0.2670 - val_loss: 1.3838 - val_categorical_accuracy: 0.4000
Epoch 68/100
racy: 0.3544 - val_loss: 1.5708 - val_categorical_accuracy: 0.3000
Epoch 69/100
racy: 0.3350 - val_loss: 1.4388 - val_categorical_accuracy: 0.3200
racy: 0.2816 - val loss: 1.4986 - val categorical accuracy: 0.2600
Epoch 71/100
racy: 0.3010 - val_loss: 1.6267 - val_categorical_accuracy: 0.3200
Epoch 72/100
0.2573
```

[5 points] Plot Accuracy and Loss During Training

```
In [70]:
          import matplotlib.pyplot as plt
          fig, (ax1, ax2) = plt.subplots(1, 2)
          fig.set figheight(6)
          fig.set figwidth(15)
          ax1.plot(res.history['categorical_accuracy'])
          ax1.plot(res.history['val_categorical_accuracy'])
          ax1.set title('Accuracy Over ' + str(NUM EPOCHS) + ' Epochs')
          ax1.legend(['Train acc', 'Test acc'], loc='lower right')
          ax1.grid(True)
          ax2.set_title('Loss Over ' + str(NUM_EPOCHS) + ' Epochs')
          ax2.plot(res.history['loss'])
          ax2.plot(res.history['val loss'])
          ax2.legend(['Train_loss', 'Test_loss'], loc='upper right')
          ax2.grid(True)
          plt.show()
```



Test accuracy: 0.5277777910232544



Testing Model