

# DEEP LEARNING LAB 1 REPORT

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importing the needed libraries

```
✓ [1] import numpy as np
4s
import tensorflow as tf

from tensorflow.keras.preprocessing.image import ImageDataGenerator

✓ [2] !git clone https://github.com/nickmccullum/cats-and-dogs.git
10s

Cloning into 'cats-and-dogs'...
remote: Enumerating objects: 10016, done.
remote: Counting objects: 100% (2/2), done.
remote: Compressing objects: 100% (2/2), done.
remote: Total 10016 (delta 0), reused 0 (delta 0), pack-reused 10014
Receiving objects: 100% (10016/10016), 216.39 MiB | 33.52 MiB/s, done.
Resolving deltas: 100% (3/3), done.
Updating files: 100% (10008/10008), done.
```

Preprocessing the Training Data

```
✓ [3] training_generator = ImageDataGenerator(
0s
    rescale = 1/255,

    shear_range = 0.2,

    zoom_range = 0.2,

    horizontal_flip = True)

✓ [4] training_set = training_generator.flow_from_directory('/content/cats-and-dogs/training_data',
1s
    target_size = (64, 64),

    batch_size = 32,

    class_mode = 'binary')

Found 8000 images belonging to 2 classes.

✓ [7] print('The training set is = ',training_set)
0s

The training set is = <keras.src.preprocessing.image.DirectoryIterator object at 0x7c525a208bb0>
```

Please print the output of `training_generator.flow_from_directory` as an image here [1 Mark]

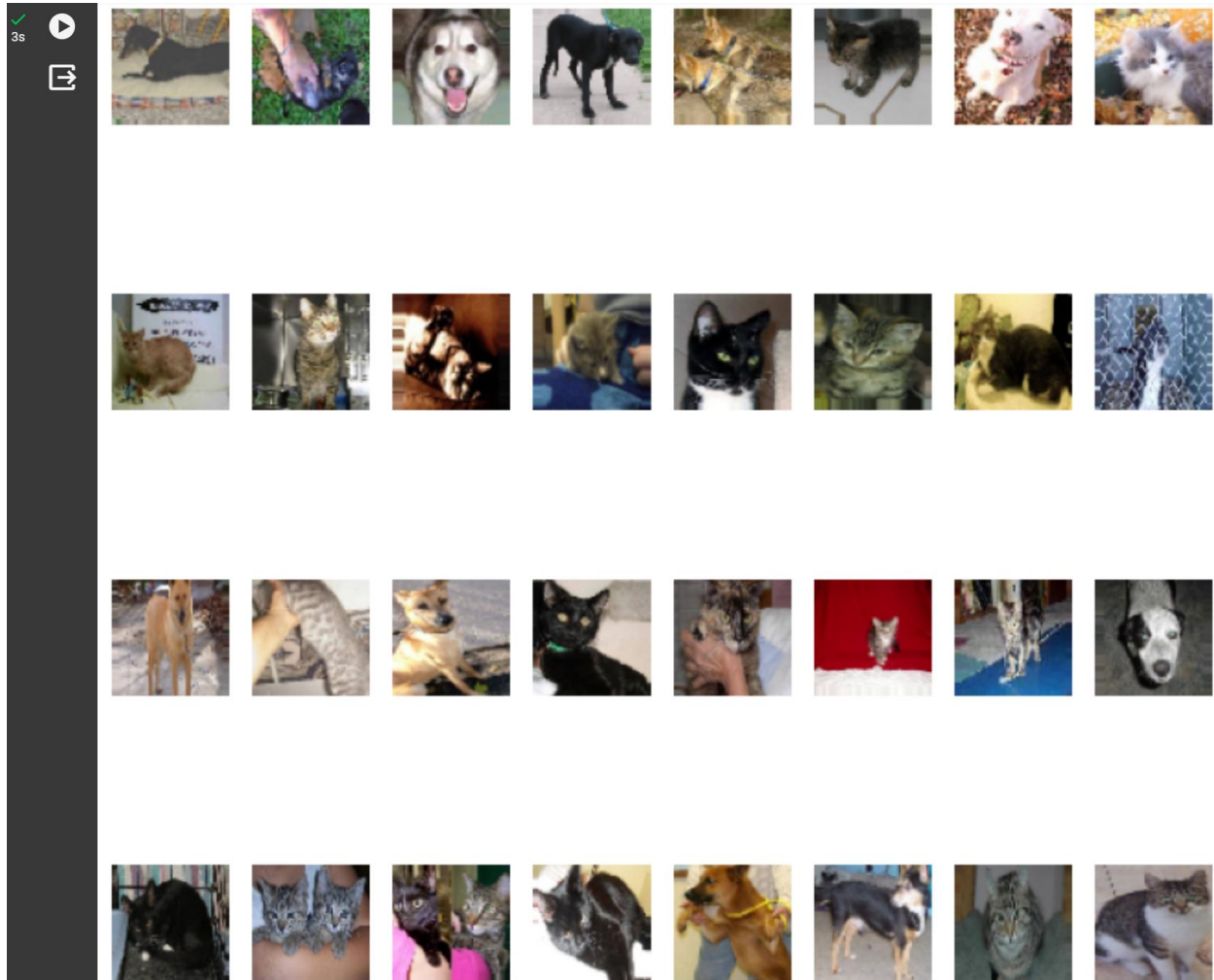
```
import matplotlib.pyplot as plt

# Retrieve a batch of images and labels from the generator
images, labels = next(training_set)

# Plot the images
fig, axes = plt.subplots(4, 8, figsize=(10, 10)) # will display 4*8 images = 32 images =batch size=32
axes = axes.ravel()

for i in range(len(images)):
    axes[i].imshow(images[i])
    axes[i].axis('off')

plt.show()
```

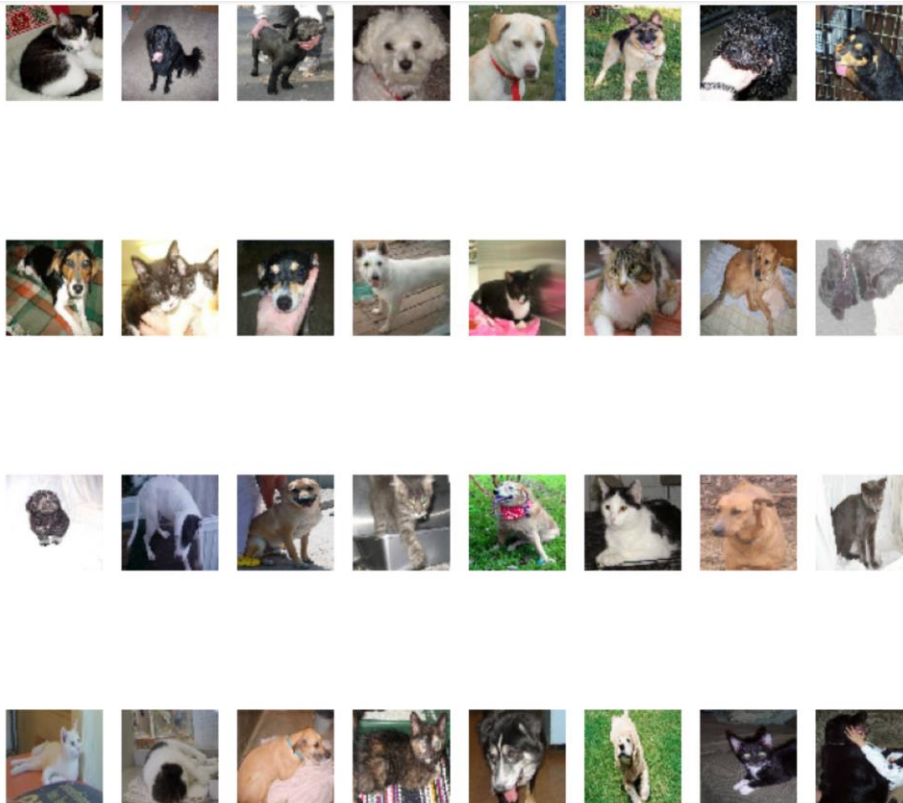


0s

```
Found 2000 images belonging to 2 classes.
```

✓  
4s

```
➡ Found 2000 images belonging to 2 classes.
```



preprocessing is complete^^

## Building Our Convolutional Neural Network

preprocessing is complete

Building Our Convolutional Neural Network

```
✓ [21] cnn = tf.keras.models.Sequential()  
0s
```

Adding Our Convolutional Layer

```
✓ [23] cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu', input_shape=[64, 64, 3]))  
0s
```

**What will be the add command if we wanted to have 64 feature detectors, with a kernel size of 5x5 [2 Marks]**

```
✓ [25] cnn.add(tf.keras.layers.Conv2D(filters=64, kernel_size=5, activation='relu')) #relu is the default activation function  
0s
```

## Adding Our Max Pooling Layer

Adding Our Max Pooling Layer

```
✓ [27] cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))  
0s
```

**What will be the add command if we wanted to have a max pooling layer which reduces the size of the image by a factor of 4 ? [2 Marks]**

What will be the add command if we wanted to have a max pooling layer which reduces the size of the image by a factor of 4 ?

```
✓ [28] cnn.add(tf.keras.layers.MaxPool2D(pool_size=4, strides=4))  
0s
```

## Adding Another Convolutional Layer and Pooling Layer

Adding Another Convolutional Layer and Pooling Layer

```
✓ [48] cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu'))  
0s  
cnn.add(tf.keras.layers.MaxPool2D(pool_size=2, strides=2))
```

## Adding The Flattening Layer To Our Convolutional Neural Network

### Adding The Flattening Layer To Our Convolutional Neural Network

```
✓ [49] cnn.add(tf.keras.layers.Flatten())  
0s
```

## Adding The Full Connection Layer To Our Convolutional Neural Network

### Adding The Full Connection Layer To Our Convolutional Neural Network

```
✓ [51] cnn.add(tf.keras.layers.Dense(units=128, activation='relu'))  
0s
```

## Adding The Output Layer To Our Convolutional Neural Network

### Adding The Output Layer To Our Convolutional Neural Network

```
✓ [52] cnn.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))  
0s
```

## Training the Convolutional Neural Network

### Training the Convolutional Neural Network

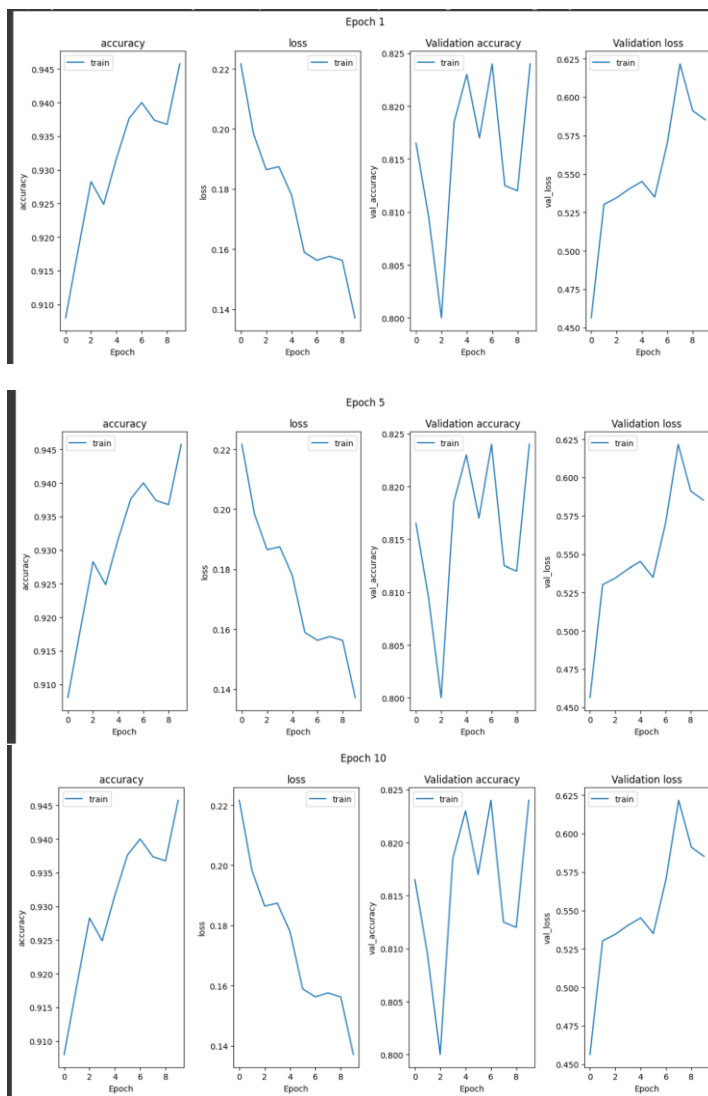
```
# Compile the model  
cnn.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])  
  
# Train the model  
history = cnn.fit(x=training_set, validation_data=test_set, epochs=10)  
  
# Plotting the training and validation metrics  
epoch_indices = [0, 4, 9] # Indices of the desired epochs  
  
for epoch_index in epoch_indices:  
    metrics = ['accuracy', 'loss', 'val_accuracy', 'val_loss']  
  
    plt.figure(figsize=(12, 6))  
  
    for i, metric in enumerate(metrics):  
        plt.subplot(1, 4, i+1)  
        plt.plot(history.history[metric])  
  
        # Check if metric starts with 'val_' to handle validation metrics correctly  
        if metric.startswith('val_'):  
            plt.title('Validation ' + metric[4:])  
        else:  
            plt.title(metric)  
  
        plt.xlabel('Epoch')  
        plt.ylabel(metric)  
        plt.legend(['train', 'val'])  
  
    plt.suptitle('Epoch {}'.format(epoch_index + 1))  
    plt.tight_layout()  
    plt.show()
```

```

Epoch 1/10
250/250 [=====] - 286s 1s/step - loss: 0.2216 - accuracy: 0.9080 - val_loss: 0.4563 - val_accuracy: 0.8165
Epoch 2/10
250/250 [=====] - 282s 1s/step - loss: 0.1983 - accuracy: 0.9183 - val_loss: 0.5302 - val_accuracy: 0.8095
Epoch 3/10
250/250 [=====] - 282s 1s/step - loss: 0.1865 - accuracy: 0.9283 - val_loss: 0.5345 - val_accuracy: 0.8000
Epoch 4/10
250/250 [=====] - 284s 1s/step - loss: 0.1874 - accuracy: 0.9249 - val_loss: 0.5403 - val_accuracy: 0.8185
Epoch 5/10
250/250 [=====] - 281s 1s/step - loss: 0.1780 - accuracy: 0.9316 - val_loss: 0.5451 - val_accuracy: 0.8230
Epoch 6/10
250/250 [=====] - 280s 1s/step - loss: 0.1590 - accuracy: 0.9376 - val_loss: 0.5350 - val_accuracy: 0.8170
Epoch 7/10
250/250 [=====] - 280s 1s/step - loss: 0.1563 - accuracy: 0.9400 - val_loss: 0.5702 - val_accuracy: 0.8240
Epoch 8/10
250/250 [=====] - 276s 1s/step - loss: 0.1576 - accuracy: 0.9374 - val_loss: 0.6217 - val_accuracy: 0.8125
Epoch 9/10
250/250 [=====] - 274s 1s/step - loss: 0.1563 - accuracy: 0.9367 - val_loss: 0.5912 - val_accuracy: 0.8120
Epoch 10/10
250/250 [=====] - 283s 1s/step - loss: 0.1372 - accuracy: 0.9457 - val_loss: 0.5853 - val_accuracy: 0.8240

```

Please share the output of the first, fifth and tenth epoch of training as an image here [2 marks]



## Making Predictions With Our Convolutional Neural Network

Please share the output probability that you have obtained for the two test images. Please mention the name of the image and the probability value obtained for it. [2 marks]

Image 1 is class 1 which is Dog

Image 2 is class 0 which is Cat

```
✓ [58] from tensorflow.keras.preprocessing import image
0s

✓ [61] test_image_1 = image.load_img('/content/cats-and-dogs/predictions/cat_or_dog_1.jpg', target_size = (64, 64))
0s
test_image_2 = image.load_img('/content/cats-and-dogs/predictions/cat_or_dog_2.jpg', target_size = (64, 64))

✓ [62] test_image_1 = image.img_to_array(test_image_1)
0s
test_image_2 = image.img_to_array(test_image_2)

✓ [63] test_image_1 = np.expand_dims(test_image_1, axis = 0)
0s
test_image_2 = np.expand_dims(test_image_2, axis = 0)

✓ [68] print(cnn.predict(test_image_1)) #Dog
0s
1/1 [=====] - 0s 27ms/step
[[1.]]

✓ [69] print(cnn.predict(test_image_2)) #Cat
0s
1/1 [=====] - 0s 27ms/step
[[0.]]
```