

Neural network are individual nodes that form the layer in the network. Just like how neurons in our brain are connected to different areas.

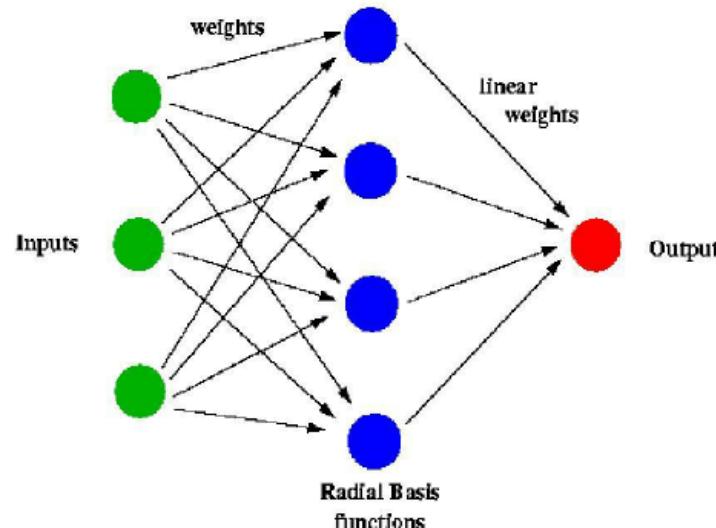
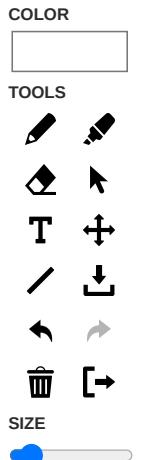
There are typically 3 layers in a Neural Network:

- 1) Input Layer
- 2) Hidden Layer
- 3) Output Layer

The input layer usually is a single layer, and it will have a weight that is assigned to them that changes the effect on the overall prediction result.

The neural network takes all the training in the input layer, then it passes the data through the hidden layer transforming the values based on the weights at each node.

Then it returns a value in the output layer.



In []: 1

CNN

A convolutional neural network is a specific kind of neural network with multiple layers. It processes data that has a grid like arrangement and then it extracts important features.

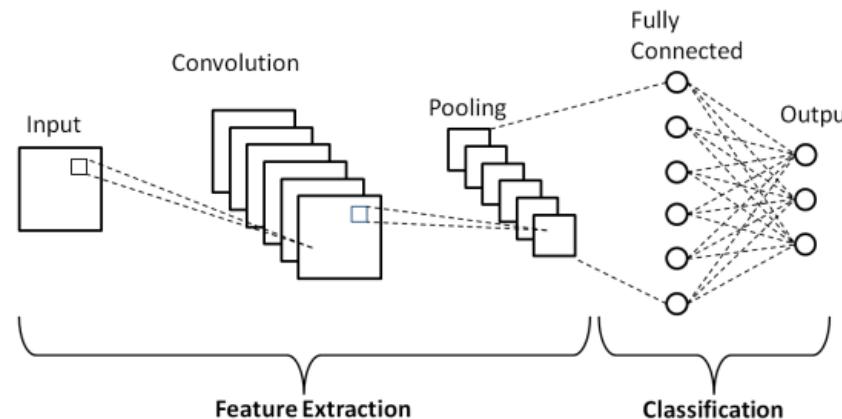
In this type of neural network they take image as an input & learn the various features on the image through filters.

This allows them to learn the important object present in the image.

For example: In CNN it will learn specific features of the cat that differentiate from the dogs. It can easily differentiate between the two.

One of the most important feature of CNN is that it sets apart from other Machine Learning Algorithm its ability to pre process the data by itself.

You dont have to spend lots of time in data pre-processing



In []: 1

example

In []: 1

```
In [1]: 1 import numpy as np
2 import pandas as pd
3 from keras.preprocessing.image import ImageDataGenerator, load_img
4 from tensorflow.keras.utils import to_categorical
5 from sklearn.model_selection import train_test_split
6 import matplotlib.pyplot as plt
7 import os
8 import random
```

In []: 1

```
In [2]: 1 # filenames[0].split('.')[0]
```

```
In [3]: 1 # reading the data and storing it into a Dataframe
```

```
In [4]: 1 filenames = os.listdir('/home/punit/Downloads/Python-Batch/PG-DS/train')
```

```
In [5]: 1 for val in filenames  
2     print(val)
```

cat.12371.jpg
dog.6189.jpg
cat.1208.jpg
dog.4534.jpg
cat.21.jpg
dog.11973.jpg
dog.10250.jpg
dog.8844.jpg
dog.6560.jpg
cat.2362.jpg
cat.1426.jpg
dog.11578.jpg
cat.8728.jpg
dog.763.jpg
cat.9261.jpg
dog.8855.jpg
cat.2495.jpg
cat.415.jpg
dog.3379.jpg
dog.1222.jpg

```
In [6]: 1 categories = []
2 for file_name in filenames:
3     category = file_name.split('.')[0]
4 #     print(category)
5     if category == 'dog':
6         categories.append(1)
7     else:
8         categories.append(0)
```

```
In [7]: 1 print(categories)
```

```
In [8]: 1 df = pd.DataFrame({  
2     'filename':filenames,  
3     'category':categories  
4 })
```

```
In [9]: 1 df.head()
```

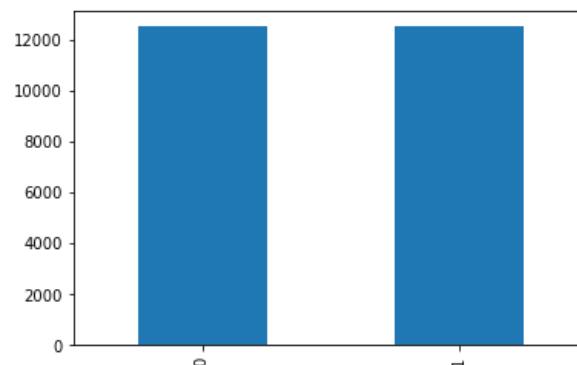
Out[9]:

	filename	category
0	cat.12371.jpg	0
1	dog.6189.jpg	1
2	cat.1208.jpg	0
3	dog.4534.jpg	1
4	cat.21.jpg	0

```
In [ ]: 1
```

```
In [10]: 1 df['category'].value_counts().plot.bar()
```

Out[10]: <AxesSubplot:>



```
In [ ]: 1
```

```
In [11]: 1 sample = random.choice(filenames)
```

```
In [12]: 1 sample
```

Out[12]: 'cat.1429.jpg'

```
In [13]: 1 img = load_img('train/'+sample)
```

```
In [14]: 1 plt.imshow(img)
```

```
Out[14]: <matplotlib.image.AxesImage at 0x7fe2d856da90>
```



```
In [15]: 1 plt.imshow(img)
```

```
Out[15]: <matplotlib.image.AxesImage at 0x7fe2d84d0be0>
```



```
In [ ]: 1
```

Building the Model

Input Layer: represent input image data. we will reshape image into single dimension array.

for instance: 64 * 64 (px) image will be converted into 4096 ----> which will be in array (4096,1) array

Convolutional Layer : This layer will extract features from the image.

Pooling Layer : Will take common feature in pool / also it will reduce spatial volume of input image after convolution.

Fully Connected Layer: This will connect the network from one layer to another.

Output Layer: It is the predicted value of the layer.

```
In [16]: 1 IMAGE_WIDTH = 128
          2 IMAGE_HEIGHT = 128
          3 IMAGE_SIZE = (IMAGE_WIDTH, IMAGE_HEIGHT)
          4 IMAGE_CHANNELS = 3
          5
          6 from keras.models import Sequential
          7 from keras.layers import Conv2D, MaxPooling2D, Dropout, Flatten, Dense, Activation, BatchNormalization
          8 model = Sequential()
          9
         10 model.add(Conv2D(32, (3,3), activation='relu', input_shape=(IMAGE_WIDTH, IMAGE_HEIGHT, IMAGE_CHANNELS)))
         11
         12 model.add(BatchNormalization())
         13
         14 model.add(MaxPooling2D(pool_size=(2,2)))
         15
         16 model.add(Dropout(0.25))
```

```
In [ ]: 1
```

Batch Normalization: It normalizes or applies a transformation that maintains the mean output close to 0 and the output of the standard deviation close to 1. It typically works during the training of the data.

Drop-out: it randomly sets the input to 0, which helps in preventing overfitting of the Data

```
In [ ]: 1
```

```
In [17]: 1 model.add(Conv2D(64, (3,3), activation='relu'))
          2 model.add(BatchNormalization())
          3 model.add(MaxPooling2D(pool_size=(2,2)))
          4 model.add(Dropout(0.25))
```

```
In [18]: 1 model.add(Conv2D(128, (3,3), activation='relu'))
          2 model.add(BatchNormalization())
          3 model.add(MaxPooling2D(pool_size=(2,2)))
          4 model.add(Dropout(0.25))
```

```
In [19]: 1 model.add(Flatten())
2 model.add(Dense(512,activation='relu'))
3 model.add(BatchNormalization())
4 model.add(Dropout(0.5))
5 model.add(Dense(2,activation='softmax')) #the output data should be in 0,1 (cat / dog)
6
7
8 model.compile(loss='categorical_crossentropy',optimizer='rmsprop',metrics=['accuracy'])
9
10
11 model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 126, 126, 32)	896
batch_normalization (BatchN ormalization)	(None, 126, 126, 32)	128
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0
dropout (Dropout)	(None, 63, 63, 32)	0
conv2d_1 (Conv2D)	(None, 61, 61, 64)	18496
batch_normalization_1 (BathchNormalization)	(None, 61, 61, 64)	256
max_pooling2d_1 (MaxPooling 2D)	(None, 30, 30, 64)	0
dropout_1 (Dropout)	(None, 30, 30, 64)	0
conv2d_2 (Conv2D)	(None, 28, 28, 128)	73856
batch_normalization_2 (BatchNormalization)	(None, 28, 28, 128)	512
max_pooling2d_2 (MaxPooling 2D)	(None, 14, 14, 128)	0
dropout_2 (Dropout)	(None, 14, 14, 128)	0
flatten (Flatten)	(None, 25088)	0
dense (Dense)	(None, 512)	12845568
batch_normalization_3 (BatchNormalization)	(None, 512)	2048
dropout_3 (Dropout)	(None, 512)	0

```
dense_1 (Dense)           (None, 2)          1026
```

```
=====
Total params: 12,942,786
Trainable params: 12,941,314
Non-trainable params: 1,472
```

```
In [20]: 1 IMAGE_SIZE
```

```
Out[20]: (128, 128)
```

```
In [ ]: 1
```

```
In [21]: 1 df.head()
```

```
Out[21]:
```

	filename	category
0	cat.12371.jpg	0
1	dog.6189.jpg	1
2	cat.1208.jpg	0
3	dog.4534.jpg	1
4	cat.21.jpg	0

```
In [22]: 1 df['category'] = df['category'].replace({0:'cat',1:'dog'})
```

```
In [23]: 1 df
```

Out[23]:

	filename	category
0	cat.12371.jpg	cat
1	dog.6189.jpg	dog
2	cat.1208.jpg	cat
3	dog.4534.jpg	dog
4	cat.21.jpg	cat
...
24995	dog.9568.jpg	dog
24996	dog.1219.jpg	dog
24997	dog.12366.jpg	dog
24998	dog.7367.jpg	dog
24999	cat.3781.jpg	cat

25000 rows × 2 columns

```
In [24]: 1 train_df, validate_df = train_test_split(df,test_size=0.20,random_state=42)
2 train_df = train_df.reset_index(drop=True)
```

```
In [25]: 1 train_df.head()
```

Out[25]:

	filename	category
0	cat.8381.jpg	cat
1	cat.8034.jpg	cat
2	dog.6311.jpg	dog
3	cat.6820.jpg	cat
4	dog.8439.jpg	dog

```
In [26]: 1 validate_df = validate_df.reset_index(drop=True)
```

```
In [27]: 1 validate_df.head()
```

Out[27]:

	filename	category
0	cat.3451.jpg	cat
1	cat.6768.jpg	cat
2	dog.10764.jpg	dog
3	cat.4895.jpg	cat
4	dog.9377.jpg	dog

```
In [28]: 1 len(train_df)
```

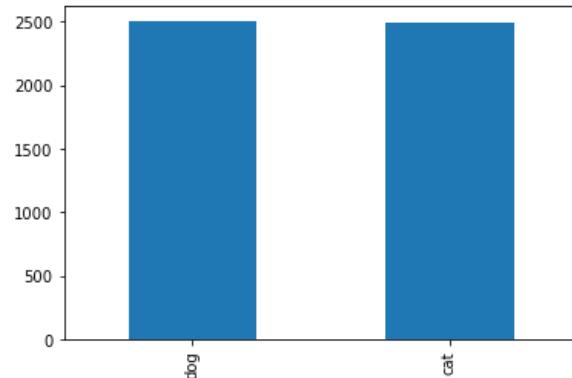
Out[28]: 20000

```
In [29]: 1 len(validate_df)
```

Out[29]: 5000

```
In [30]: 1 validate_df['category'].value_counts().plot.bar()
```

Out[30]: <AxesSubplot:>



```
In [31]: 1 validate_df['category'].value_counts()
```

Out[31]: dog 2505
cat 2495
Name: category, dtype: int64

```
In [32]: 1 train_df['category'].value_counts()
```

Out[32]: cat 10005
dog 9995
Name: category, dtype: int64

```
In [33]: 1 total_train = train_df.shape[0]
```

```
In [34]: 1 total_train
```

```
Out[34]: 20000
```

```
In [35]: 1 total_validate = validate_df.shape[0]
```

```
In [36]: 1 total_validate
```

```
Out[36]: 5000
```

```
In [37]: 1 batch_size = 15
```

```
In [ ]: 1
```

```
In [ ]: 1
```

Training generator

```
In [38]: 1 train_datagen = ImageDataGenerator(  
2     rotation_range=15,  
3     rescale=1./255,  
4     shear_range=0.1,  
5     zoom_range=0.1,  
6     horizontal_flip=True,  
7     width_shift_range=0.1,  
8     height_shift_range=0.1  
9 )
```

```
In [39]: 1 train_generator = train_datagen.flow_from_dataframe(train_df,  
2                                         'train/',  
3                                         x_col='filename',  
4                                         y_col='category',  
5                                         target_size=IMAGE_SIZE,  
6                                         class_mode='categorical',  
7                                         batch_size=batch_size  
8 )
```

Found 20000 validated image filenames belonging to 2 classes.

```
In [40]: 1 validation_datagen = ImageDataGenerator(  
2     rescale=1./255)
```

```
In [41]: 1 validation_generator = validation_datagen.flow_from_dataframe(validate_df,
2                                         'train/',
3                                         x_col='filename',
4                                         y_col='category',
5                                         target_size=IMAGE_SIZE,
6                                         class_mode='categorical',
7                                         batch_size=batch_size
8 )
```

Found 5000 validated image filenames belonging to 2 classes.

```
In [ ]: 1
```

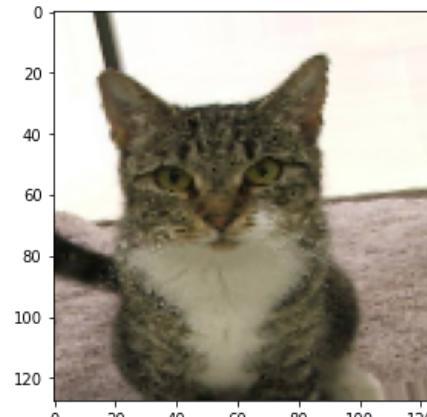
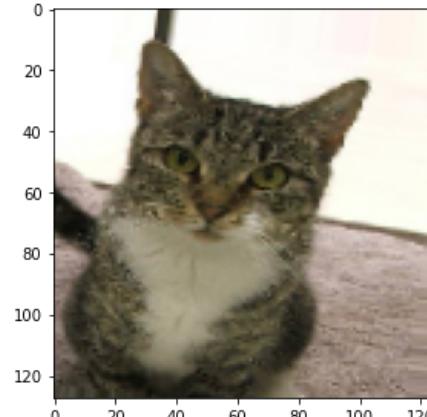
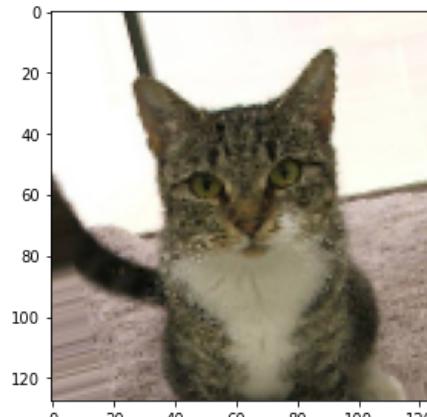
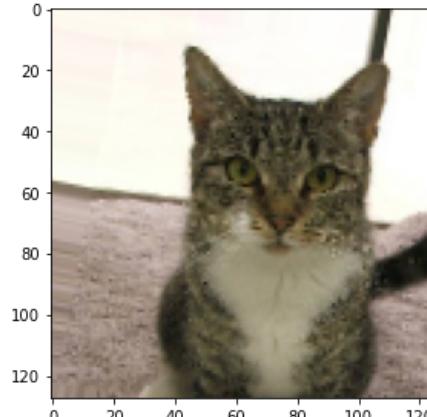
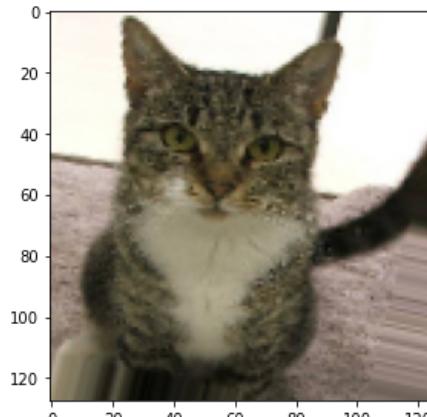
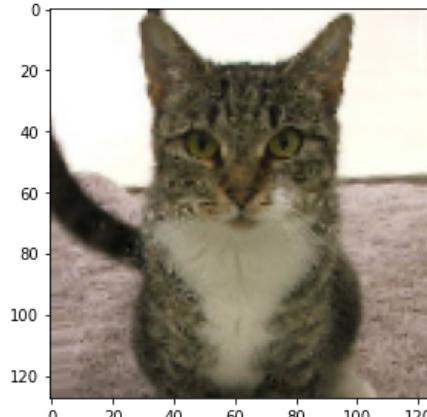
Example Generator

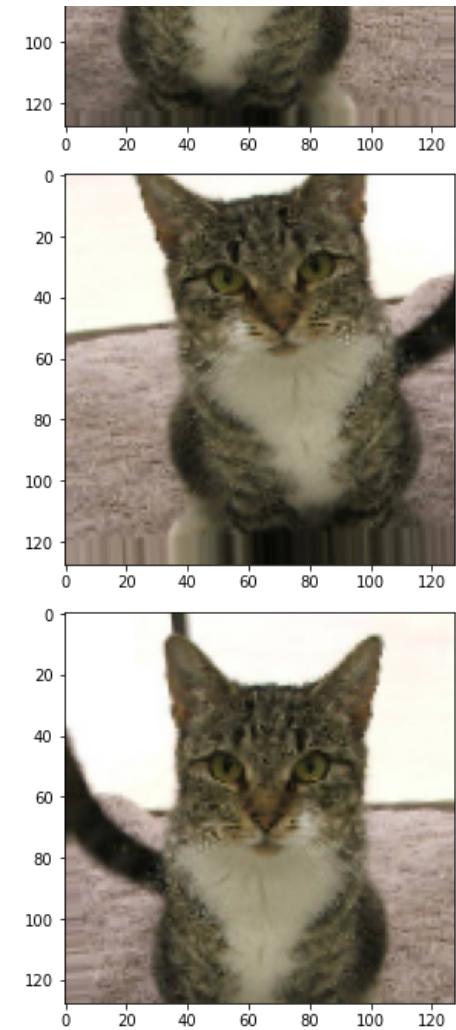
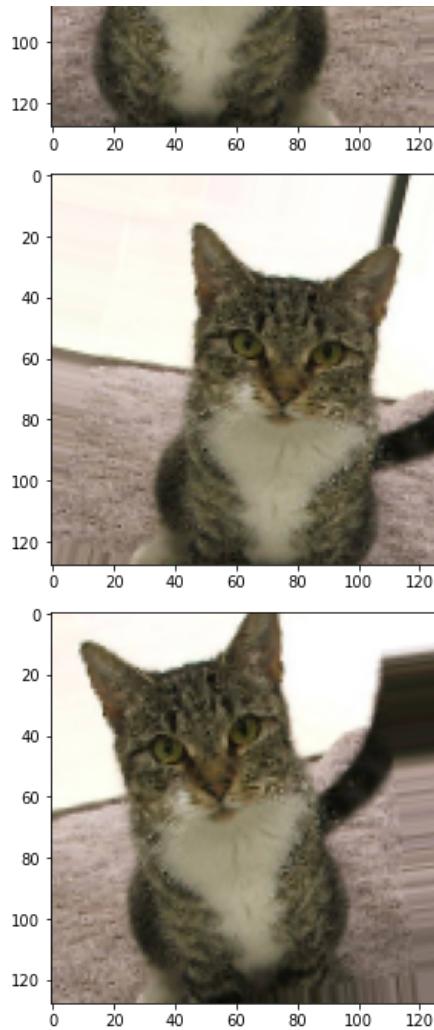
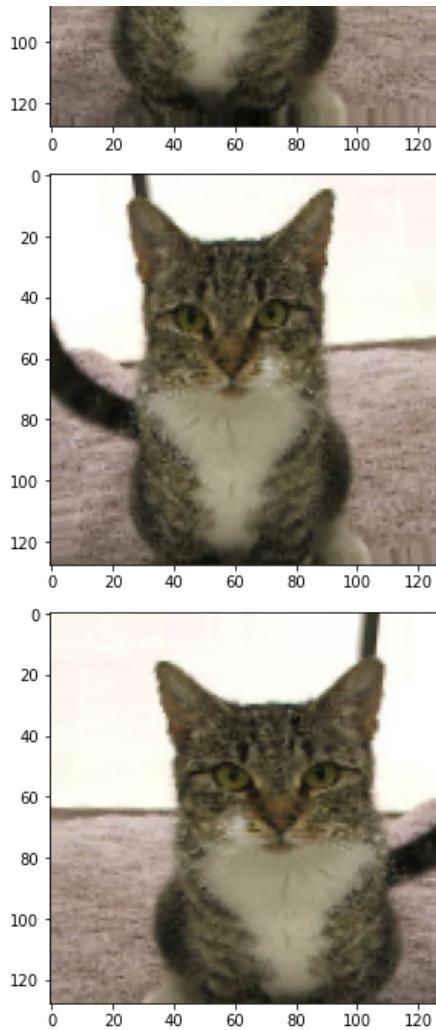
```
In [43]: 1 example_df = train_df.sample(n=1).reset_index(drop=True)
```

```
In [44]: 1 example_generator = train_datagen.flow_from_dataframe(example_df,'train/',x_col='filename',y_col='category',
2                                         target_size=IMAGE_SIZE,class_mode='categorical')
```

Found 1 validated image filenames belonging to 1 classes.

```
In [45]: 1 plt.figure(figsize=(20,20))
2 for i in range(0,15):
3     plt.subplot(5,3,i+1)
4     for X_batch,Y_batch in example_generator:
5         image = X_batch[0]
6         plt.imshow(image)
7         break
8 plt.tight_layout()
9 plt.show()
```





In [50]: 1 total_train // batch_size

Out[50]: 1333

```
In [47]: 1 history = model.fit_generator(  
2     train_generator,  
3     epochs=3,  
4     validation_data=validation_generator,  
5     validation_steps= total_validate // batch_size,  
6     steps_per_epoch = total_train // batch_size,  
7 )
```

```
<ipython-input-47-0e7355e3de92>:1: UserWarning: `Model.fit_generator` is deprecated and will be removed in a future version. Please use `Model.fit` which supports generators.  
    history = model.fit_generator()
```

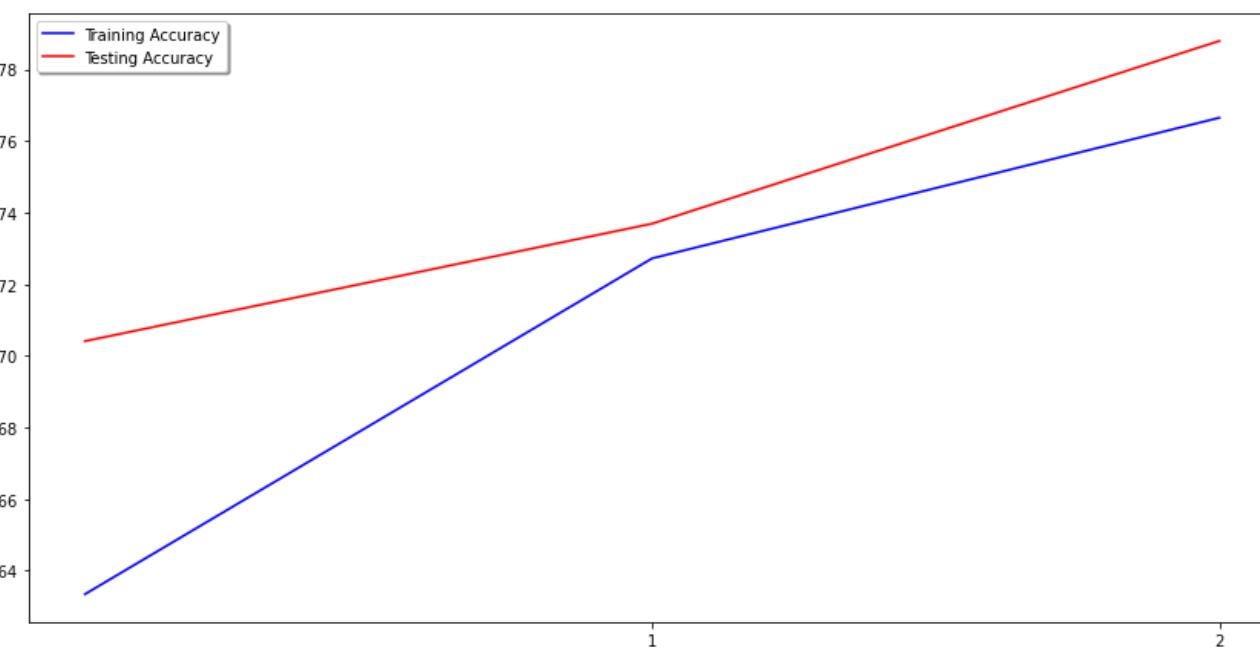
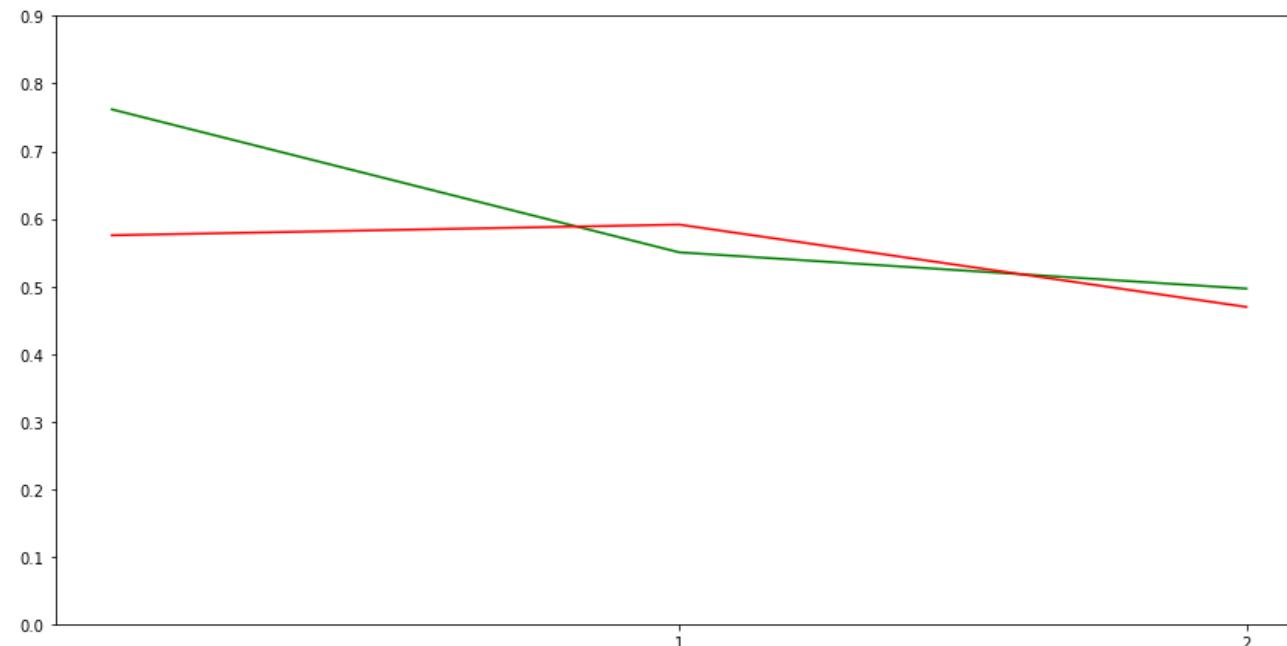
```
Epoch 1/3  
1333/1333 [=====] - 641s 480ms/step - loss: 0.7618 - accuracy: 0.6334 - val_loss: 0.5757 - val_accuracy: 0.7041  
Epoch 2/3  
1333/1333 [=====] - 628s 471ms/step - loss: 0.5505 - accuracy: 0.7272 - val_loss: 0.5915 - val_accuracy: 0.7369  
Epoch 3/3  
1333/1333 [=====] - 639s 479ms/step - loss: 0.4968 - accuracy: 0.7665 - val_loss: 0.4697 - val_accuracy: 0.7880
```

```
In [81]: 1
```

```
Out[81]: [0.6334250569343567, 0.7272454500198364, 0.7665249109268188]
```

Visualize Training

```
In [78]: 1 fig,(ax1,ax2) = plt.subplots(2,1,figsize=(12,12))
2
3 ax1.plot(history.history['loss'],color='g',label='Training Data Loss')
4 ax1.plot(history.history['val_loss'],color='r',label='Testing Data Loss')
5 ax1.set_xticks(np.arange(1,3,1))
6 ax1.set_yticks(np.arange(0,1,0.1))
7
8 ax2.plot(history.history['accuracy'],color='b',label='Training Accuracy')
9 ax2.plot(history.history['val_accuracy'],color='r',label='Testing Accuracy')
10 ax2.set_xticks(np.arange(1,3,1))
11
12 legend = plt.legend(loc='best',shadow=True)
13 plt.tight_layout()
14 plt.show()
```



In []:

1

```
In [55]: 1 test_file = os.listdir('test1/')
```

```
In [ ]: 1
```

```
In [57]: 1 test_df = pd.DataFrame({  
2     'filename':test_file  
3 })
```

```
In [61]: 1 test_df
```

Out[61]:

	filename
0	9895.jpg
1	11342.jpg
2	6449.jpg
3	9662.jpg
4	7947.jpg
...	...
12495	3796.jpg
12496	10089.jpg
12497	3565.jpg
12498	10366.jpg
12499	5025.jpg

12500 rows × 1 columns

```
In [58]: 1 nb_sample = test_df.shape[0]
```

```
In [59]: 1 nb_sample
```

Out[59]: 12500

```
In [84]: 1 round(nb_sample / batch_size)
```

Out[84]: 833

```
In [62]: 1 test_gen = ImageDataGenerator(rescale=1./255)
2 test_generator = test_gen.flow_from_dataframe(
3     'test1/',
4     x_col='filename',
5     y_col=None,
6     class_mode=None,
7     target_size=IMAGE_SIZE,
8     batch_size=batch_size,
9     shuffle=False)
```

Found 12500 validated image filenames.

```
In [65]: 1 predict = model.predict_generator(test_generator,steps=np.ceil(nb_sample/batch_size))
```

<ipython-input-65-46522c80dc03>:1: UserWarning: `Model.predict_generator` is deprecated and will be removed in a future version. Please use `Model.predict`, which supports generators.
predict = model.predict_generator(test_generator,steps=np.ceil(nb_sample/batch_size))

```
In [85]: 1 predict
```

```
Out[85]: array([[0.38720885, 0.6127911 ],
   [0.04973018, 0.9502699 ],
   [0.03544782, 0.96455216],
   ...,
   [0.9778145 , 0.0221855 ],
   [0.04795221, 0.95204777],
   [0.69287425, 0.30712578]], dtype=float32)
```

```
In [67]: 1 test_df['category'] = np.argmax(predict,axis=-1)
```

```
In [68]: 1 test_df
```

Out[68]:

	filename	category
0	9895.jpg	1
1	11342.jpg	1
2	6449.jpg	1
3	9662.jpg	0
4	7947.jpg	1
...
12495	3796.jpg	1
12496	10089.jpg	0
12497	3565.jpg	0
12498	10366.jpg	1
12499	5025.jpg	0

12500 rows × 2 columns

```
In [87]: 1 for k,v in train_generator.class_indices.items():
2     print(k,v)
```

cat 0
dog 1

```
In [69]: 1 label_map = dict((val,key) for key,val in train_generator.class_indices.items())
```

```
In [73]: 1 label_map
```

Out[73]: {0: 'cat', 1: 'dog'}

```
In [71]: 1 test_df['category'] = test_df['category'].replace(label_map)
```

In [72]: 1 test_df

Out[72]:

	filename	category
0	9895.jpg	dog
1	11342.jpg	dog
2	6449.jpg	dog
3	9662.jpg	cat
4	7947.jpg	dog
...
12495	3796.jpg	dog
12496	10089.jpg	cat
12497	3565.jpg	cat
12498	10366.jpg	dog
12499	5025.jpg	cat

12500 rows × 2 columns

In [74]: 1 test_df.head(20)

Out[74]:

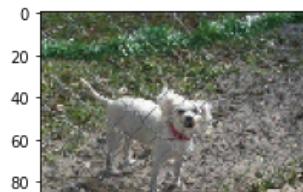
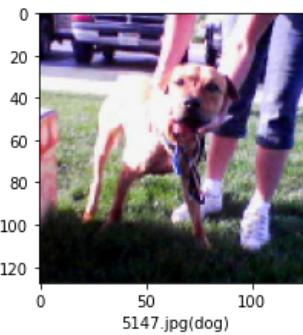
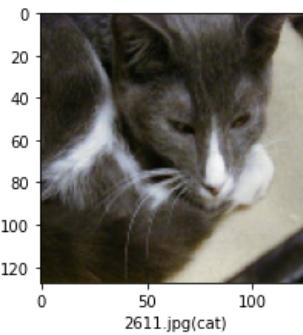
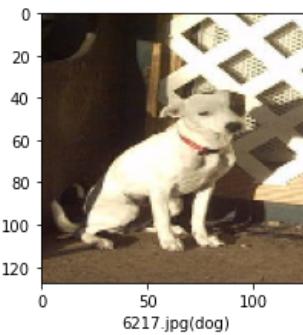
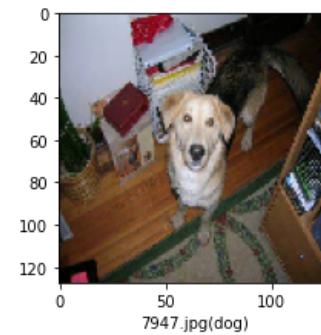
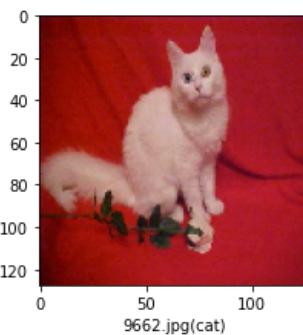
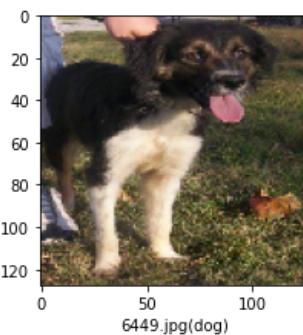
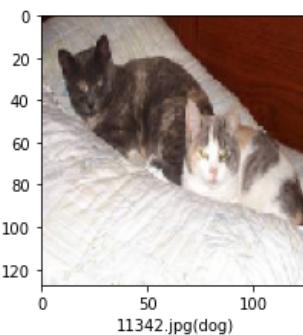
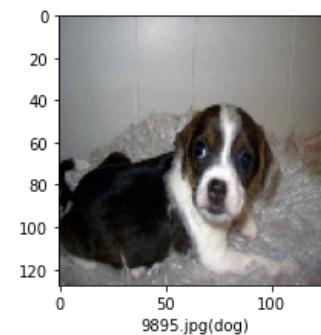
	filename	category
0	9895.jpg	dog
1	11342.jpg	dog
2	6449.jpg	dog
3	9662.jpg	cat
4	7947.jpg	dog
5	6217.jpg	dog
6	2611.jpg	cat
7	5147.jpg	dog
8	7145.jpg	dog
9	4591.jpg	cat
10	2661.jpg	dog
11	12120.jpg	dog
12	2328.jpg	cat
13	9788.jpg	dog
14	11114.jpg	dog
15	5964.jpg	dog
16	4201.jpg	cat
17	5911.jpg	dog
18	11524.jpg	dog
19	3518.jpg	cat

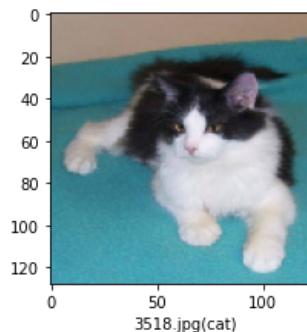
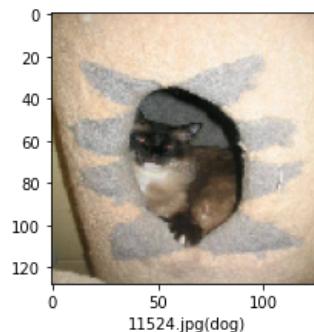
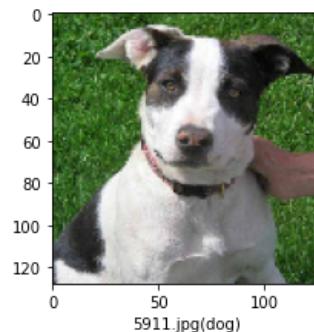
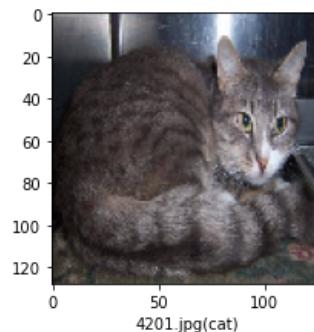
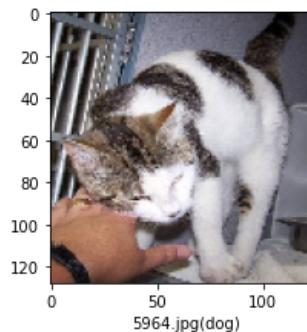
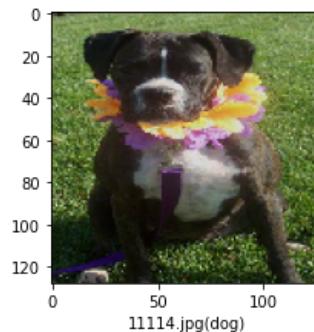
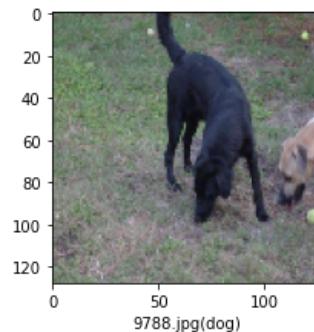
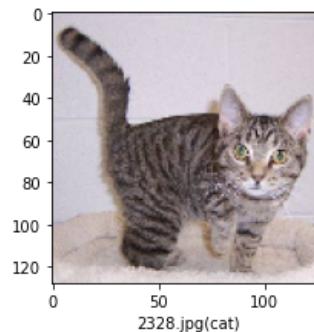
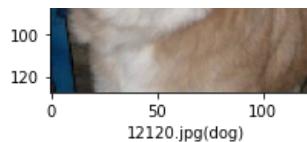
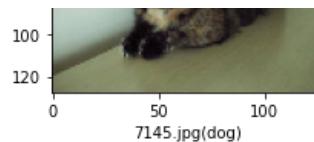
In [88]: 1 sample_data

Out[88]:

	filename	category
0	9895.jpg	dog
1	11342.jpg	dog
2	6449.jpg	dog
3	9662.jpg	cat
4	7947.jpg	dog
5	6217.jpg	dog
6	2611.jpg	cat
7	5147.jpg	dog
8	7145.jpg	dog
9	4591.jpg	cat
10	2661.jpg	dog
11	12120.jpg	dog
12	2328.jpg	cat
13	9788.jpg	dog
14	11114.jpg	dog
15	5964.jpg	dog
16	4201.jpg	cat
17	5911.jpg	dog
18	11524.jpg	dog
19	3518.jpg	cat

```
In [77]:  
1 sample_data = test_df.head(20)  
2 sample_data.head()  
3 plt.figure(figsize=(12,24))  
4 for index, row in sample_data.iterrows():  
5     filename = row['filename']  
6     category = row['category']  
7     img = load_img('test1/' + filename, target_size=IMAGE_SIZE)  
8     plt.subplot(6,4,index+1)  
9     plt.imshow(img)  
10    plt.xlabel(filename + ('+' + "{}".format(category) + ')')  
11    plt.tight_layout()  
12    plt.show()
```





In []:

1