Import Libraries

```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as pt
   import seaborn as sb
   import warnings
   warnings.filterwarnings('ignore')
   import cv2
```

loading data set

```
In [2]:
       Xtrain=np.loadtxt("input.csv",delimiter=",")
        Xtest=np.loadtxt("input_test.csv",delimiter=",")
        ytrain=np.loadtxt("labels.csv",delimiter=",")
        ytest=np.loadtxt("labels_test.csv",delimiter=",")
In [3]: Xtrain
Out[3]: array([[ 37., 39., 25., ..., 58., 54., 29.],
               [131., 128., 135., ..., 71., 96., 74.],
               [ 80., 92., 88., ..., 124., 119., 99.],
               . . . ,
               [231., 226., 230., ..., 62., 65., 72.],
               [ 61., 61., 63., ..., 135., 123., 123.],
               [ 64., 31., 12., ..., 61., 49., 35.]])
In [4]: len(Xtrain)
Out[4]: 2000
In [5]: len(Xtest)
Out[5]: 400
In [6]: len(ytrain)
Out[6]: 2000
```

Input Test Data

```
In [7]: import pandas as pd
In [8]: data=pd.read_csv("input_test.csv")
```

```
In [9]: data
 Out[9]:
               1,18000000000000000e+02 8,2000000000000000e+01 9,60000000000000000e+01 1,090000000
             0
                                 223.0
                                                         211.0
                                                                                 163.0
                                  73.0
                                                                                  43.0
             1
                                                          67.0
             2
                                   0.0
                                                           3.0
                                                                                  1.0
             3
                                  27.0
                                                          55.0
                                                                                  76.0
             4
                                 121.0
                                                         122.0
                                                                                 114.0
                                                         224.0
                                                                                 187.0
           394
                                 244.0
           395
                                 213.0
                                                         213.0
                                                                                 201.0
           396
                                 249.0
                                                         245.0
                                                                                 242.0
           397
                                  97.0
                                                          96.0
                                                                                 102.0
           398
                                  94.0
                                                          66.0
                                                                                  63.0
          399 rows × 30000 columns
In [10]: Xtrain
Out[10]: array([[ 37., 39., 25., ..., 58., 54., 29.],
                 [131., 128., 135., ..., 71., 96., 74.],
                 [ 80., 92., 88., ..., 124., 119., 99.],
                 [231., 226., 230., ..., 62., 65., 72.],
                 [ 61., 61., 63., ..., 135., 123., 123.],
                 [ 64., 31., 12., ..., 61., 49., 35.]])
In [11]: ytrain
Out[11]: array([0., 0., 0., ..., 1., 1., 1.])
```

```
In [12]: |sb.countplot(ytrain)
Out[12]: <AxesSubplot:ylabel='count'>
            1000
             800
             600
             400
             200
              0
                          0.0
                                                1.0
In [13]: Xtest
Out[13]: array([[118., 82.,
                              96., ..., 140.,
                                               79.,
                                                     16.],
                [223., 211., 163., ..., 70., 73., 78.],
                [ 73.,
                       67., 43., ..., 222., 211., 165.],
                [249., 245., 242., ..., 73., 72.,
```

78., 80.],

Here 0 represents Cat and 1 represents Dog

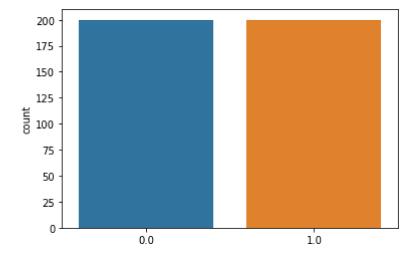
[94., 66., 63., ..., 119., 96., 80.]])

[97., 96., 102., ..., 84.,

```
In [14]: |ytest
1., 1., 1., 1., 1., 1., 1., 1., 1.
```



Out[15]: <AxesSubplot:ylabel='count'>



```
In [16]: Xtrain[0][0]
```

Out[16]: 37.0

```
In [17]: Xtrain[0,:]
Out[17]: array([37., 39., 25., ..., 58., 54., 29.])
In [18]: Xtrain[1,:]
Out[18]: array([131., 128., 135., ..., 71., 96., 74.])
In [19]: Xtrain[1,:].shape
Out[19]: (30000,)
In [20]: # above data is 1D array
#to process the image we require 2Dor3D array
```

Reshaping

```
In [21]: Xtrain=Xtrain.reshape(len(Xtrain),100,100,3)
    Xtest=Xtest.reshape(len(Xtest),100,100,3)
```

```
In [22]: Xtrain[1]
Out[22]: array([[[131., 128., 135.],
                  [160., 157., 164.],
                  [198., 192., 204.],
                  . . . ,
                  [250., 249., 247.],
                  [255., 255., 253.],
                  [250., 249., 245.]],
                 [[140., 137., 144.],
                  [127., 124., 131.],
                  [120., 114., 124.],
                  . . . ,
                  [251., 253., 252.],
                  [254., 255., 253.],
                  [254., 255., 251.]],
                 [[204., 202., 207.],
                  [187., 185., 190.],
                  [147., 142., 148.],
                  [249., 255., 255.],
                  [238., 247., 242.],
                  [232., 241., 236.]],
                 . . . ,
                 [[174., 182., 195.],
                  [172., 180., 193.],
                  [178., 186., 197.],
                  . . . ,
                  [ 87., 114.,
                  [ 75., 99.,
                                 83.],
                  [ 80., 105.,
                                 86.]],
                 [[166., 173., 189.],
                  [164., 172., 185.],
                  [172., 180., 193.],
                  . . . ,
                  [ 78., 106.,
                                 84.],
                  [ 72., 97.,
                                 76.],
                  [ 77., 102.,
                                 81.]],
                 [[173., 180., 196.],
                  [172., 179., 195.],
                  [174., 182., 195.],
                           91.,
                  [ 63.,
                                 69.],
                  [ 62., 87.,
                                 65.],
                  [ 71.,
                           96.,
                                 74.]]])
```

Reshaping

```
In [23]: ytrain=ytrain.reshape(len(ytrain),1)
ytest=ytest.reshape(len(ytest),1)

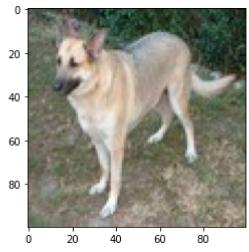
In [24]: ytrain.shape

Out[24]: (2000, 1)

In [25]: import matplotlib.pyplot as pt

In [26]: Xtrain=Xtrain/255 # dividing all the values by 255 will convert it to range from Xtest=Xtest/255

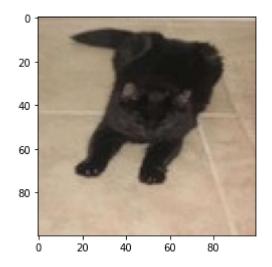
In [27]: pt.imshow(Xtrain[2,:])
Out[27]: <matplotlib.image.AxesImage at 0x13ae5375d30>
```



Here we use ramdom module

```
In [28]: import random
num=random.randint(0,len(Xtrain))
pt.imshow(Xtrain[num,:])
```

Out[28]: <matplotlib.image.AxesImage at 0x13ae5197400>



```
In [29]: | Xtrain[0,:]
Out[29]: array([[[0.14509804, 0.15294118, 0.09803922],
                  [0.10196078, 0.09411765, 0.03529412],
                  [0.13333333, 0.09803922, 0.03921569],
                  [0.22352941, 0.17254902, 0.1372549],
                  [0.23921569, 0.18431373, 0.14901961],
                  [0.25490196, 0.2
                                         , 0.16470588]],
                 [0.17647059, 0.16862745, 0.10980392],
                  [0.10980392, 0.09803922, 0.03137255],
                  [0.20392157, 0.15686275, 0.09411765],
                  [0.21176471, 0.16078431, 0.1254902],
                  [0.22352941, 0.16862745, 0.13333333],
                  [0.23921569, 0.18431373, 0.14901961]],
                 [[0.20392157, 0.17647059, 0.10196078],
                  [0.1254902, 0.09411765, 0.01960784],
                  [0.27058824, 0.21176471, 0.1372549],
                  [0.21176471, 0.15686275, 0.11372549],
                  [0.21960784, 0.16470588, 0.12156863],
                  [0.23137255, 0.17647059, 0.13333333]],
                 . . . ,
                 [[0.07843137, 0.15294118, 0.
                  [0.39607843, 0.49019608, 0.2627451],
                  [0.59607843, 0.71372549, 0.47058824],
                  [0.18039216, 0.16078431, 0.0745098],
                  [0.23529412, 0.21568627, 0.12941176],
                  [0.23529412, 0.21568627, 0.12941176]],
                 [[0.18039216, 0.25490196, 0.03529412],
                  [0.45490196, 0.54901961, 0.32156863],
                  [0.61176471, 0.72941176, 0.48627451],
                  [0.25098039, 0.23529412, 0.1372549],
                  [0.29411765, 0.27843137, 0.18039216],
                  [0.28235294, 0.26666667, 0.16862745]],
                 [[0.31764706, 0.39215686, 0.17254902],
                  [0.49411765, 0.58823529, 0.36078431],
                  [0.57254902, 0.69019608, 0.44705882],
                  [0.2627451, 0.24705882, 0.14901961],
                  [0.30588235, 0.29019608, 0.19215686],
                  [0.22745098, 0.21176471, 0.11372549]]])
```

CNN Building

```
In [30]: from tensorflow.keras.models import Sequential
    from tensorflow.keras.layers import Conv2D,MaxPooling2D,Dense,Flatten

In [31]: model=Sequential()
    model.add(Conv2D(32,(3,3),activation='relu',input_shape=(100,100,3)))
    model.add(MaxPooling2D(2,2))
    model.add(Conv2D(32,(3,3),activation='relu'))
    model.add(MaxPooling2D(2,2))
    model.add(Flatten())
    model.add(Dense(64,activation='relu'))
    model.add(Dense(64,activation='relu'))
```

now above data is sent to the model with shape 100x100x3

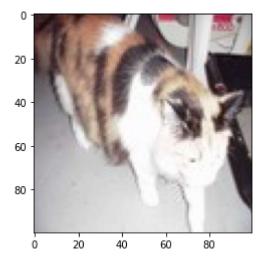
We are using adam as optimizer for better accuracy

```
In [32]:
    model.compile(loss="binary_crossentropy",optimizer="adam",metrics=['accuracy'])
```

We are fitting our data into model

```
In [34]: | model.fit(Xtrain,ytrain,batch size=64,epochs=5)
       Epoch 1/10
       32/32 [============= ] - 12s 385ms/step - loss: 0.5014 - accura
       cy: 0.7675
       Epoch 2/10
       32/32 [============= ] - 12s 359ms/step - loss: 0.4632 - accura
       cy: 0.7805
       Epoch 3/10
       32/32 [=============== ] - 12s 378ms/step - loss: 0.4097 - accura
       cy: 0.8035
       Epoch 4/10
       32/32 [============== ] - 11s 349ms/step - loss: 0.3714 - accura
       cy: 0.8250
       Epoch 5/10
       32/32 [=============== ] - 10s 313ms/step - loss: 0.3171 - accura
       cy: 0.8670
       Epoch 6/10
       32/32 [============== ] - 10s 315ms/step - loss: 0.2823 - accura
       cy: 0.8840
       Epoch 7/10
       32/32 [============== ] - 10s 309ms/step - loss: 0.2483 - accura
       cy: 0.9020
       Epoch 8/10
       cy: 0.9090
       Epoch 9/10
       32/32 [============== ] - 10s 315ms/step - loss: 0.1741 - accura
       cy: 0.9355
       Epoch 10/10
       32/32 [============= ] - 10s 316ms/step - loss: 0.1431 - accura
       cy: 0.9540
Out[34]: <keras.callbacks.History at 0x13a9e86ac10>
In [35]: model.evaluate(Xtest,ytest)
       13/13 [================ ] - 1s 38ms/step - loss: 0.7406 - accurac
       y: 0.7100
Out[35]: [0.7406277656555176, 0.7099999785423279]
```

```
In [38]: num=random.randint(0,len(ytest))
pt.imshow(Xtest[num,:])
pt.show()
```



```
In [40]: | ypr=ypr>0.5
    if ypr==1:
        print("dog")
    else:
        print("cat")
```

In []: