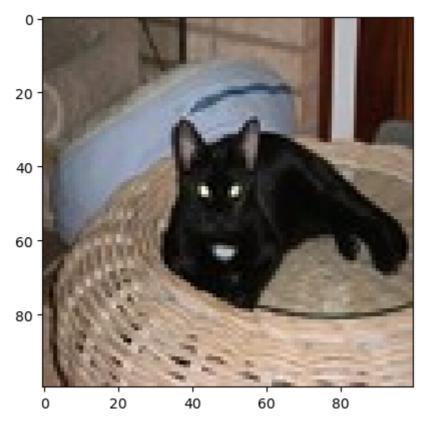
Import

```
In [1]:
           import numpy as np
           from tensorflow.keras.models import Sequential
           from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dense, Flatten
           import random
           import matplotlib.pyplot as plt
In [2]:
           import os
           print(os.getcwd())
           print(os.listdir('imageClassification'))
           /Users/rsn/ML:DL
           ['labels_test.csv', 'labels.csv', 'input.csv', 'input_test.csv']
           Adding Dataset
In [3]:
           X_train = np.loadtxt('imageClassification/input.csv', delimiter=',')
           Y_train = np.loadtxt('imageClassification/labels.csv', delimiter=',')
           X_test = np.loadtxt('imageClassification/input_test.csv', delimiter=',')
           Y test = np,loadtxt('imageClassification/labels test.csv', delimiter=',')
           Reshaping/shaping
           X_{train} = X_{train.reshape(len(X_{train}),100,100,3)}
In [4]:
           Y_train = Y_train.reshape(len(Y_train),1)
           X_{\text{test}} = X_{\text{test.reshape}}(\text{len}(X_{\text{test}}), 100, 100, 3)
           Y_test = Y_test.reshape(len(Y_test),1)
           print("shape of X_train", X_train.shape)
In [5]:
           print("shape of Y_train", Y_train.shape)
           print("shape of X_test", X_test.shape)
           print("shape of Y_train", Y_test.shape)
           shape of X_train (2000, 100, 100, 3)
           shape of Y_train (2000, 1)
           shape of X_test (400, 100, 100, 3)
           shape of Y_train (400, 1)
In [6]: #X_train[1,:]
           # <!--it shows values ranging upto 255 which is RGB values -->
In [7]:
           X train = X train/255.0
           X \text{ test} = X \text{ test/}255.0
            # here we divide it with 255 to rescale our value between 0 to 1
In [8]:
           # by using matplotlib we display the image
           idx = random.randint(0, len(X_train))
           plt.imshow(X_train[idx, :],cmap = "gray")
```

```
plt.figure(figsize = (15,2))
plt.show
```

Out [8]: <function matplotlib.pyplot.show(close=None, block=None)>



<Figure size 1500x200 with 0 Axes>

Model....

```
In [9]: model = Sequential()
# 1. first layer
model.add(Conv2D(32,(3,3),activation = 'relu',input_shape =(100,100,3)))
model.add(MaxPooling2D(2,2))

# 2. sec layer
model.add(Conv2D(32,(3,3),activation = 'relu'))
model.add(MaxPooling2D(2,2))

# 3.
model.add(Flatten())
# 4.
model.add(Dense(64,activation = 'relu'))
# 5.
model.add(Dense(1, activation = "sigmoid"))
```

/Users/rsn/anaconda3/lib/python3.11/site-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When using Sequential m odels, prefer using an `Input(shape)` object as the first layer in the model instead. super().__init__(activity_regularizer=activity_regularizer, **kwargs)

Compiling

Model Fit

```
model.fit(X_train, Y_train, epochs = 20, batch_size = 64)
In []:
           Epoch 1/20
           32/32 -
                                                             4s 106ms/step - accuracy: 0.4975 - loss: 0.7290
           Epoch 2/20
           32/32 -
                                                              4s 112ms/step - accuracy: 0.6191 - loss: 0.6602
           Epoch 3/20
                                                              3s 106ms/step - accuracy: 0.6787 - loss: 0.6033
           32/32
           Epoch 4/20
           32/32
                                                              3s 104ms/step - accuracy: 0.7351 - loss: 0.5374
           Epoch 5/20
                                                              3s 105ms/step - accuracy: 0.7584 - loss: 0.4771
           32/32 -
           Epoch 6/20
           32/32 -
                                                              3s 105ms/step - accuracy: 0.8023 - loss: 0.4206
           Epoch 7/20
           32/32 -
                                                              3s 105ms/step - accuracy: 0.8772 - loss: 0.3016
           Epoch 8/20
           32/32 -
                                                              3s 105ms/step - accuracy: 0.9236 - loss: 0.2429
           Epoch 9/20
           32/32
                                                              3s 107ms/step - accuracy: 0.9476 - loss: 0.1574
           Epoch 10/20
           32/32
                                                              3s 105ms/step - accuracy: 0.9627 - loss: 0.1110
           Epoch 11/20
           32/32 -
                                                              3s 106ms/step - accuracy: 0.9861 - loss: 0.0695
           Epoch 12/20
           32/32
                                                              3s 107ms/step - accuracy: 0.9929 - loss: 0.0437
           Epoch 13/20
           32/32
                                                              3s 108ms/step - accuracy: 0.9995 - loss: 0.0208
           Epoch 14/20
           32/32
                                                              3s 108ms/step - accuracy: 1.0000 - loss: 0.0141
           Epoch 15/20
           11/32 -
                                                              2s 116ms/step - accuracy: 1.0000 - loss: 0.0078
```

Test accuracy

```
In [ ]: model.evaluate(X_test, Y_test)
```

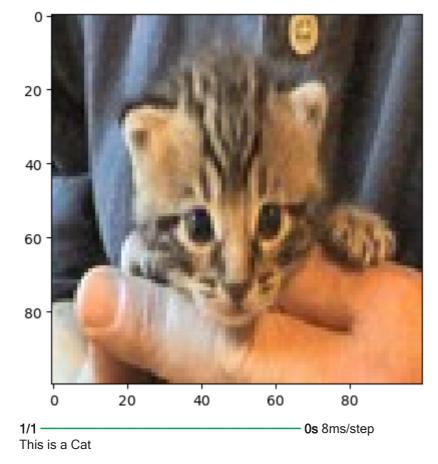
Prediction

```
In [19]: xyz = random.randint(0,len(Y_test))
plt.imshow(X_test[xyz, :])
plt.show()

y_pred = model.predict(X_test[xyz, :].reshape(1,100,100,3))

y_pred = y_pred > 0.5
if(y_pred == 0):
    pred = 'Dog'
else:
    pred = 'Cat'

print("This is a",pred)
```



In []:	
In []:	
TH []:	
In []:	