

```
In [1]: import numpy as np
import random
import matplotlib.pyplot as plt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dense, Flatten
```

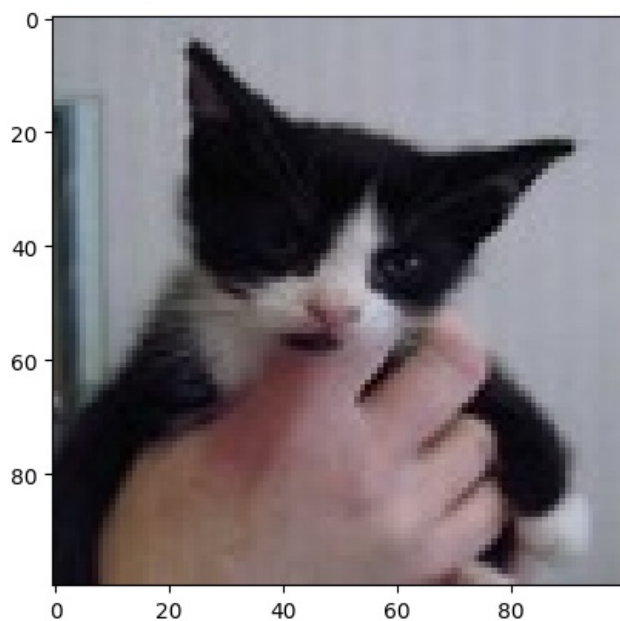
```
In [2]: X_train = np.loadtxt('input.csv', delimiter = ',')
Y_train = np.loadtxt('labels.csv', delimiter = ',')
X_test = np.loadtxt('input_test.csv', delimiter = ',')
Y_test = np.loadtxt('labels_test.csv', delimiter = ',')
```

```
In [3]: X_train = X_train.reshape(len(X_train), 100, 100, 3)
Y_train = Y_train.reshape(len(Y_train), 1)
X_test = X_test.reshape(len(X_test), 100, 100, 3)
Y_test = Y_test.reshape(len(Y_test), 1)
X_train = X_train/255.0
X_test = X_test/255.0
```

```
In [4]: print("Shape of X_train: ", X_train.shape)
print("Shape of Y_train: ", Y_train.shape)
print("Shape of X_test: ", X_test.shape)
print("Shape of Y_test: ", 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_test: (400, 1)
```

```
In [5]: idx = random.randint(0, len(X_train))
plt.imshow(X_train[idx, :])
plt.show()
```



```
In [ ]: # sequential model means the layers are going to be stacked up
model = Sequential([
    Conv2D(32, (3,3), activation = 'relu', input_shape = (100, 100, 3)),
    MaxPooling2D((2,2)),

    Conv2D(32, (3,3), activation = 'relu'),
    MaxPooling2D((2,2)),

    Flatten(),
    Dense(64, activation = 'relu'),
    Dense(1, activation = 'sigmoid')
])
```

```
In [7]: #way 2 of defining the model
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(1, activation = 'sigmoid'))
```

```
In [8]: model.compile(loss = 'binary_crossentropy', optimizer = 'adam', metrics = ['accuracy'])
```

```
In [10]: model.fit(X_train, Y_train, epochs = 5, batch_size = 64)
```

```
Epoch 1/5  
32/32 ————— 8s 263ms/step - accuracy: 0.7981 - loss: 0.4331  
Epoch 2/5  
32/32 ————— 8s 264ms/step - accuracy: 0.8327 - loss: 0.3758  
Epoch 3/5  
32/32 ————— 8s 235ms/step - accuracy: 0.8734 - loss: 0.3083  
Epoch 4/5  
32/32 ————— 8s 234ms/step - accuracy: 0.9136 - loss: 0.2407  
Epoch 5/5  
32/32 ————— 8s 244ms/step - accuracy: 0.9269 - loss: 0.2005
```

```
Out[10]: <keras.src.callbacks.history.History at 0x1c53e00b860>
```

```
In [11]: model.evaluate(X_test, Y_test)
```

```
13/13 ————— 1s 35ms/step - accuracy: 0.7086 - loss: 0.7385
```

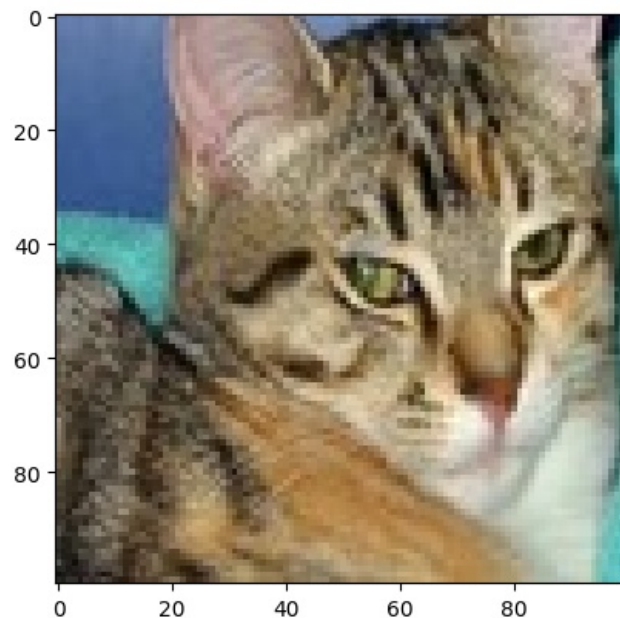
```
Out[11]: [0.8075342774391174, 0.6875]
```

```
In [12]: model.evaluate(X_test, Y_test)
```

```
13/13 ————— 1s 38ms/step - accuracy: 0.7086 - loss: 0.7385
```

```
Out[12]: [0.8075342774391174, 0.6875]
```

```
In [37]: idx2 = random.randint(0, len(Y_test))  
plt.imshow(X_test[idx2, :])  
plt.show()  
  
y_pred = model.predict(X_test[idx2, :].reshape(1, 100, 100, 3))  
y_pred = y_pred > 0.5  
  
if(y_pred == 0):  
    pred = 'dog'  
else:  
    pred = 'cat'  
  
print("Our model says it is a :", pred)
```



```
1/1 ————— 0s 34ms/step  
Our model says it is a : cat  
1/1 ————— 0s 34ms/step  
Our model says it is a : cat
```