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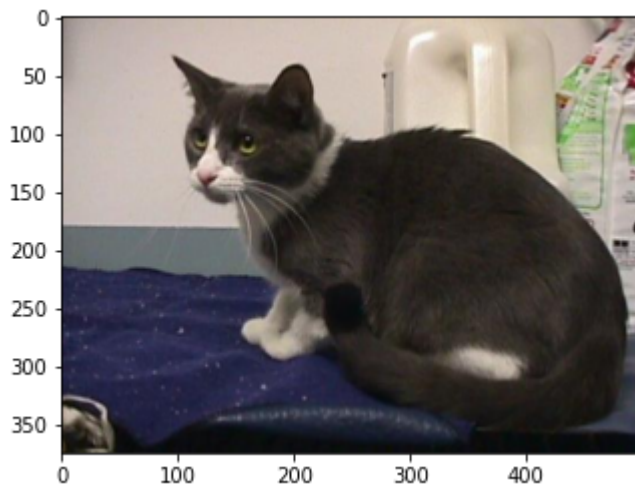
PROJECT : Cat & Dog Classification using Convolutional Neural Network

```
In [3]: 1 import warnings
        2 warnings.filterwarnings('ignore')
```

```
In [12]: 1 import matplotlib.pyplot as plt
        2 import cv2
        3 %matplotlib inline
```

```
In [57]: 1 #Upload image of cat number 4
        2 cat4 = cv2.imread('CATS_DOGS/train/CAT/4.jpg')
        3 cat4 = cv2.cvtColor(cat4,cv2.COLOR_BGR2RGB)
        4 plt.imshow(cat4)
```

Out[57]: <matplotlib.image.AxesImage at 0x25ce3bd4130>

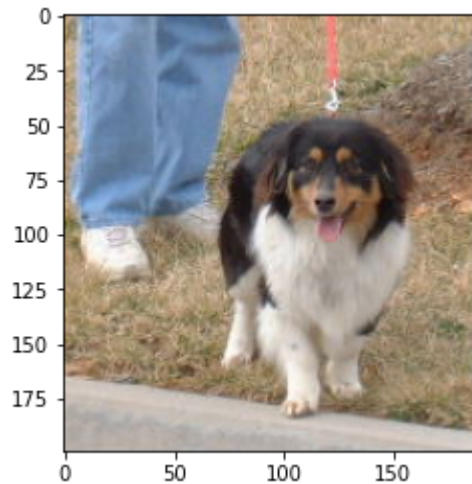


```
In [58]: 1 #image shape
        2 cat4.shape
```

Out[58]: (375, 500, 3)

```
In [59]: 1 #Upload image of dog number 4
2 dog = cv2.imread('CATS_DOGS/train/DOG/2.jpg')
3 dog = cv2.cvtColor(dog,cv2.COLOR_BGR2RGB)
4 plt.imshow(dog)
```

Out[59]: <matplotlib.image.AxesImage at 0x25ce5341c90>



```
In [60]: 1 #image shape
2 dog.shape
```

Out[60]: (199, 188, 3)

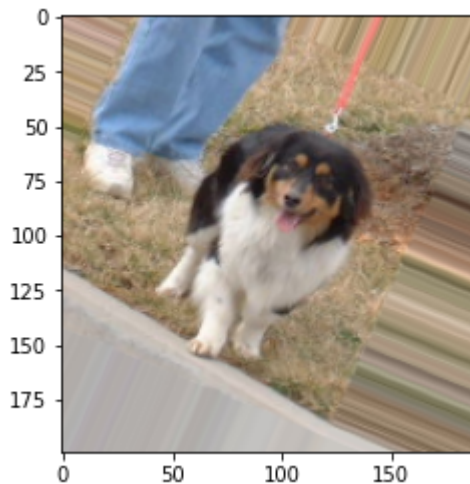
PREPROCESSING

```
In [6]: 1 from keras.preprocessing.image import ImageDataGenerator
```

```
In [62]: 1 #create an ImageDataGenerator object
2 image_gen = ImageDataGenerator(rotation_range=30,
3                                 width_shift_range=0.1,
4                                 height_shift_range=0.1,
5                                 shear_range=0.2,
6                                 zoom_range=0.2,
7                                 fill_mode='nearest',
8                                 horizontal_flip=True,
9                                 vertical_flip=False,
10                                rescale=1/255)
```

```
In [63]: 1 plt.imshow(image_gen.random_transform(dog))
```

```
Out[63]: <matplotlib.image.AxesImage at 0x25ce559f700>
```



```
In [64]: 1 image_gen.flow_from_directory('CATS_DOGS/train')
```

Found 18743 images belonging to 2 classes.

```
Out[64]: <keras.preprocessing.image.DirectoryIterator at 0x25ce3bdca00>
```

```
In [ ]: 1
```

CREATE MODEL

```
In [65]: 1 from keras.models import Sequential
2 from keras.layers import Activation, Dropout, Flatten, Conv2D, MaxPooling2D
```

```
In [66]: 1 model = Sequential()
2
3 model.add(Conv2D(filters=32, kernel_size=(3,3), input_shape=(150,150,3), a
4 model.add(MaxPooling2D(pool_size=(2,2)))
5
6 model.add(Conv2D(filters=64, kernel_size=(3,3), input_shape=(150,150,3), a
7 model.add(MaxPooling2D(pool_size=(2,2)))
8
9 model.add(Conv2D(filters=64, kernel_size=(3,3), input_shape=(150,150,3), a
10 model.add(MaxPooling2D(pool_size=(2,2)))
11
12 model.add(Flatten())
13
14 model.add(Dense(128, activation='relu'))
15
16 model.add(Dropout(0.5))
17
18 model.add(Dense(1, activation='sigmoid'))
19
20 model.compile(loss='binary_crossentropy',
21               optimizer='adam',
22               metrics=['accuracy'])
```



```
In [71]: 1 result = model.fit_generator(train_image_gen,epochs=100,steps_per_epoch
        2 validation_data=test_image_gen,validation_s
```

```
Epoch 1/100
150/150 [=====] - 116s 762ms/step - loss: 0.696
7 - accuracy: 0.5111 - val_loss: 0.6955 - val_accuracy: 0.5156
Epoch 2/100
150/150 [=====] - 108s 721ms/step - loss: 0.692
2 - accuracy: 0.5379 - val_loss: 0.6888 - val_accuracy: 0.6094
Epoch 3/100
150/150 [=====] - 115s 767ms/step - loss: 0.687
3 - accuracy: 0.5612 - val_loss: 0.6717 - val_accuracy: 0.5156
Epoch 4/100
150/150 [=====] - 93s 619ms/step - loss: 0.6718
- accuracy: 0.5958 - val_loss: 0.6809 - val_accuracy: 0.5573
Epoch 5/100
150/150 [=====] - 90s 600ms/step - loss: 0.6799
- accuracy: 0.5829 - val_loss: 0.6957 - val_accuracy: 0.5208
Epoch 6/100
150/150 [=====] - 92s 612ms/step - loss: 0.6750
- accuracy: 0.5858 - val_loss: 0.6544 - val_accuracy: 0.6042
Epoch 7/100
150/150 [=====] - 91s 605ms/step - loss: 0.6500
- accuracy: 0.6000 - val_loss: 0.6500 - val_accuracy: 0.6500
```

```
In [ ]: 1
```

EVALUATING THE MODEL

In [72]:

1	result.history['accuracy']
---	----------------------------

Out[72]: [0.5110832452774048,
0.5379166603088379,
0.5612499713897705,
0.5958333611488342,
0.5829166769981384,
0.5858333110809326,
0.6349999904632568,
0.636666555404663,
0.6604166626930237,
0.6479166746139526,
0.6616666913032532,
0.65625,
0.6766666769981384,
0.6858333349227905,
0.7030531167984009,
0.6937500238418579,
0.7172731161117554,
0.699999988079071,
0.7195833325386047,
0.7370833158493042,
0.7162500023841858,
0.6983333230018616,
0.7304166555404663,
0.7268925309181213,
0.7329166531562805,
0.7262499928474426,
0.731249988079071,
0.7395833134651184,
0.7391666769981384,
0.7524999976158142,
0.7433333396911621,
0.7450000047683716,
0.7641666531562805,
0.7662066221237183,
0.7570833563804626,
0.7462499737739563,
0.7754077911376953,
0.7637500166893005,
0.751666650772095,
0.766666507720947,
0.7554166913032532,
0.7712500095367432,
0.7595833539962769,
0.7674999833106995,
0.7716666460037231,
0.7917189598083496,
0.7891666889190674,
0.79708331823349,
0.8016666769981384,
0.7820995450019836,
0.7854453921318054,
0.7729166746139526,
0.7745833396911621,
0.7858333587646484,
0.8004166483879089,
0.7954166531562805,
0.7787500023841858,
0.7858333587646484,
0.8083333373069763,
0.800000011920929,
0.8125,

```

0.809166669845581,
0.8058333396911621,
0.8087499737739563,
0.8025000095367432,
0.7920833230018616,
0.8122124671936035,
0.8183333277702332,
0.8079166412353516,
0.8083333373069763,
0.8054166436195374,
0.8187500238418579,
0.815416693687439,
0.8324999809265137,
0.8183333277702332,
0.8204166889190674,
0.8170833587646484,
0.8133333325386047,
0.8125,
0.8038477897644043,
0.8402342200279236,
0.8162500262260437,
0.8287500143051147,
0.8070833086967468,
0.8070833086967468,
0.8147218823432922,
0.8195833563804626,
0.8362500071525574,
0.8423253893852234,
0.8149999976158142,
0.8137500286102295,
0.824999988079071,
0.8366666436195374,
0.840416669845581,
0.8450000286102295,
0.8423253893852234,
0.8333333134651184,
0.8352153897285461,
0.8366666436195374,
0.8424999713897705]

```

```

In [74]: 1 #save the model
          2 model.save('cat_dog_100epochs.h5')

```

PREDICTING ON NEW IMAGES

```

In [9]: 1 #importing the saved model
          2 from keras.models import load_model
          3 new_model = load_model('cat_dog_100epochs.h5')

```

```

In [10]: 1 test_data_gen = ImageDataGenerator(rescale=1/255)
          2 genrated_test_images = test_data_gen.flow_from_directory(
          3                                     'CATS_DOGS/test
          4                                     target_size=(15
          5                                     batch_size=batch
          6                                     class_mode='bin

```

Found 6251 images belonging to 2 classes.


```
In [11]: 1 genrated_test_images.class_indices
```

```
Out[11]: {'CAT': 0, 'DOG': 1}
```

```
In [12]: 1 predictions = new_model.predict(genrated_test_images)
391/391 [=====] - 294s 752ms/step
```

```
In [13]: 1 predictions
```

```
Out[13]: array([[0.18466547],
                [0.3501265 ],
                [0.99689215],
                ...,
                [0.77171296],
                [0.26374152],
                [0.04919095]], dtype=float32)
```

```
In [14]: 1 predictions.shape
```

```
Out[14]: (6251, 1)
```

```
In [15]: 1 class_prediction = (predictions >= 0.5).astype('int')
2 class_prediction
```

```
Out[15]: array([[0],
                [0],
                [1],
                ...,
                [1],
                [0],
                [0]])
```

```
In [16]: 1 class_prediction.shape
```

```
Out[16]: (6251, 1)
```

```
In [17]: 1 actual_classes = genrated_test_images.classes
2 actual_classes
```

```
Out[17]: array([0, 0, 0, ..., 1, 1, 1])
```

```
In [18]: 1 actual_classes.shape
```

```
Out[18]: (6251,)
```

```
In [19]: 1 from sklearn.metrics import classification_report, confusion_matrix
```

```
In [20]: 1 print(confusion_matrix(actual_classes, class_prediction))
```

```
[[1513 1613]
 [1540 1585]]
```

```
In [21]: 1 print(classification_report(actual_classes,class_prediction))
```

	precision	recall	f1-score	support
0	0.50	0.48	0.49	3126
1	0.50	0.51	0.50	3125
accuracy			0.50	6251
macro avg	0.50	0.50	0.50	6251
weighted avg	0.50	0.50	0.50	6251