```
In [3]:
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import tensorflow as tf
from keras.preprocessing.image import load img, ImageDataGenerator
import os
import random
In [4]:
data gen = ImageDataGenerator()
In [5]:
# read the images from directory and classify them
train images = data gen.flow from directory('Datasets/Assessment3 DL/train', classes=['dogs', 'cats'])
Found 40 images belonging to 2 classes.
In [32]:
# read the test images from directory and classify them
test images = data gen.flow from directory('Datasets/Assessment3 DL/test', classes=['dogs', 'cats'])
Found 20 images belonging to 2 classes.
In [6]:
train images
Out[6]:
<tensorflow.python.keras.preprocessing.image.DirectoryIterator at 0x1a8cd2f8588>
In [7]:
train_images.num_classes
Out[7]:
2
In [8]:
train images.classes
Out[8]:
In [9]:
train_images.image_shape
Out[9]:
(256, 256, 3)
In [10]:
```

list file names

fn_cats = os.listdir('Datasets/Assessment3_DL/train/cats')
fn_dogs = os.listdir('Datasets/Assessment3_DL/train/dogs')

In [11]:

In [12]:

```
# cat = 0
# dog = 1
```

In [13]:

```
df.head(n=40)
```

Out[13]:

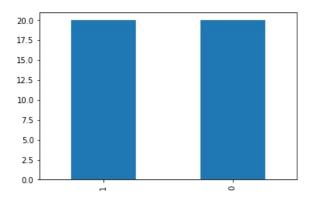
	filomomo.	
	filename	category
0	1.jpg	0
1	10.jpg	0
2	11.jpg	0
3	12.jpg	0
4	13.jpg	0
5	14.jpg	0
6	15.jpg	0
7	16.jpg	0
8	17.jpg	0
9	18.jpg	0
10	19.jpg	0
11	2.jpg	0
12	20.jpg	0
13	3.jpg	0
14	4.jpg	0
15	5.jpg	0
16	6.jpg	0
17	7.jpg	0
18	8.jpg	0
19	9.jpg	0
20	1.jpg	1
21	10.jpg	1
22	11.jpg	1
23	12.jpg	1
24	13.jpg	1
25	14.jpg	1
26	15.jpg	1
27	16.jpg	1
28	17.jpg	1
29	18.jpg	1
30		1
	19.jpg	
31	2.jpg	1
32	20.jpg	1
33	3.jpg	1
34	4.jpg	1
35	5.jpg	1
36	6.jpg	1
37	7.jpg	1
38	8.jpg	1
39	9.jpg	1

In [14]:

df.category.value_counts().plot.bar()

Out[14]:

<matplotlib.axes. subplots.AxesSubplot at 0x1a8cd451ec8>



In [15]:

```
# view the cat random sample
sample_cat = random.choice(fn_cats)
image_cat = load_img('Datasets\\Assessment3_DL\\train\\cats\\'+ sample_cat)
plt.imshow(image_cat)
```

Out[15]:

<matplotlib.image.AxesImage at 0x1a8cdc89b48>



In [16]:

```
#view the dog random sample
sample_dog = random.choice(fn_dogs)
image_dog = load_img('Datasets\\Assessment3_DL\\train\\dogs\\'+ sample_dog)
plt.imshow(image_dog)
```

Out[16]:

<matplotlib.image.AxesImage at 0x1a8cdd04b88>



In [18]:

```
model = tf.keras.models.Sequential()

model.add(tf.keras.layers.Conv2D(32, (5,5), activation='relu', input_shape=(256,256,3)))
model.add(tf.keras.layers.BatchNormalization())
model.add(tf.keras.layers.MaxPooling2D(pool_size=(2,2)))

model.add(tf.keras.layers.Conv2D(64, (5,5), activation='relu'))
model.add(tf.keras.layers.BatchNormalization())
model.add(tf.keras.layers.MaxPooling2D(pool_size=(2,2)))
```

In [19]:

```
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(32, activation='relu'))
model.add(tf.keras.layers.Dropout(0.4))
model.add(tf.keras.layers.BatchNormalization())
```

In [20]:

```
model.add(tf.keras.layers.Dense(2, activation='softmax'))
```

In [22]:

```
model.compile(loss='categorical_crossentropy', optimizer='rmsprop', metrics=['accuracy'])
```

In [23]:

model.summary()

Model: "sequential"

Layer (type)	0utput	Shape	Param #
conv2d (Conv2D)	(None,	252, 252, 32)	2432
batch_normalization (BatchNo	(None,	252, 252, 32)	128
max_pooling2d (MaxPooling2D)	(None,	126, 126, 32)	0
conv2d_1 (Conv2D)	(None,	122, 122, 64)	51264
batch_normalization_1 (Batch	(None,	122, 122, 64)	256
max_pooling2d_1 (MaxPooling2	(None,	61, 61, 64)	0
flatten (Flatten)	(None,	238144)	0
dense (Dense)	(None,	32)	7620640
dropout (Dropout)	(None,	32)	0
batch_normalization_2 (Batch	(None,	32)	128
dense_1 (Dense)	(None,	2)	66

Total params: 7,674,914 Trainable params: 7,674,658 Non-trainable params: 256

In [24]:

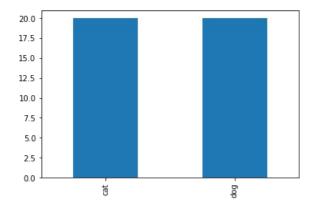
```
df['category'] = df['category'].replace({0:'cat', 1:'dog'})
```

In [25]:

```
df.category.value_counts().plot.bar()
```

Out[25]:

<matplotlib.axes. subplots.AxesSubplot at 0x1a8d4b55fc8>



In [27]:

```
# list file names
fn_test_cats = os.listdir('Datasets/Assessment3_DL/test/cats')
fn_test_dogs = os.listdir('Datasets/Assessment3_DL/test/dogs')
```

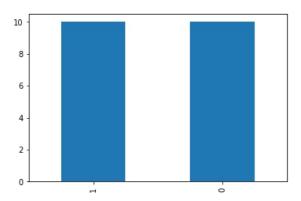
In [28]:

In [29]:

```
df_test.category.value_counts().plot.bar()
```

Out[29]:

<matplotlib.axes._subplots.AxesSubplot at 0x1a8d510da48>



In [30]:

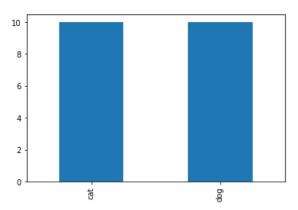
```
# converting into 0 = cat and 1 = dog
df_test.category = df_test.category.replace({0:'cat', 1:'dog'})
```

In [31]:

```
df_test.category.value_counts().plot.bar()
```

Out[31]:

<matplotlib.axes. subplots.AxesSubplot at 0x1a8d50f7f48>



In [33]:

```
# fit the model
model_fit = model.fit(train_images, epochs=100, batch_size=10, validation_data=test_images)
```

```
Epoch 1/100
5386 - val accuracy: 0.5000
Epoch 2/100
976 - val_accuracy: 0.4000
Epoch 3/100
2/2 [======
          ==============] - 3s 1s/step - loss: 0.2227 - accuracy: 0.9250 - val loss: 9.62
95 - val accuracy: 0.4000
Epoch 4/100
2/2 [=====
             34 - val accuracy: 0.5000
Epoch 5/100
41 - val accuracy: 0.3500
Epoch 6/100
61 - val_accuracy: 0.3500
Epoch 7/100
485 - val accuracy: 0.3500
Epoch 8/100
2/2 [======
            =========] - 2s 1s/step - loss: 0.1318 - accuracy: 0.9750 - val loss: 9.29
41 - val accuracy: 0.4000
Epoch 9/100
       2/2 [======
76 - val accuracy: 0.4000
Epoch 10/100
               ======] - 3s 1s/step - loss: 0.1498 - accuracy: 0.9500 - val loss: 10.7
2/2 [==
302 - val_accuracy: 0.4500
Epoch 11/100
14 - val accuracy: 0.4000
Epoch 12/100
08 - val_accuracy: 0.4500
Epoch 13/100
                ======] - 3s 1s/step - loss: 0.1375 - accuracy: 0.9750 - val_loss: 7.00
2/2 [==
05 - val accuracy: 0.4500
Epoch 14/100
2/2 [===
              =======] - 10s 5s/step - loss: 0.1111 - accuracy: 1.0000 - val loss: 5.5
046 - val accuracy: 0.4500
Epoch 15/100
            ========] - 3s 1s/step - loss: 0.0814 - accuracy: 1.0000 - val_loss: 4.67
2/2 [======
88 - val accuracy: 0.4000
Epoch 16/100
56 - val_accuracy: 0.4500
Epoch 17/100
57 - val accuracy: 0.4500
Epoch 18/100
2/2 [=====
          :===============] - 12s 6s/step - loss: 0.0688 - accuracy: 1.0000 - val loss: 3.7
379 - val accuracy: 0.4000
Epoch 19/100
```

```
15 - val accuracy: 0.4500
Epoch 20/100
2/2 [==
            ========] - 3s 1s/step - loss: 0.0865 - accuracy: 1.0000 - val loss: 2.25
23 - val accuracy: 0.4000
Epoch 21/100
24 - val accuracy: 0.4000
Epoch 22/100
77 - val accuracy: 0.4500
Epoch 23/100
2/2 [======
       42 - val accuracy: 0.4000
Epoch 24/100
78 - val accuracy: 0.3500
Epoch 25/100
2/2 [======
       23 - val_accuracy: 0.4500
Epoch 26/100
68 - val accuracy: 0.4000
Epoch 27/100
68 - val_accuracy: 0.4000
Epoch 28/100
36 - val accuracy: 0.4000
Epoch 29/100
           ========] - 3s 2s/step - loss: 0.0346 - accuracy: 1.0000 - val loss: 1.62
2/2 [====
10 - val accuracy: 0.4500
Epoch 30/100
2/2 [==
             ======] - 7s 4s/step - loss: 0.0343 - accuracy: 1.0000 - val loss: 2.03
10 - val accuracy: 0.3000
Epoch 31/100
37 - val_accuracy: 0.3500
Epoch 32/100
85 - val_accuracy: 0.4500
Epoch 33/100
2/2 [==
           :========] - 2s 1s/step - loss: 0.0456 - accuracy: 1.0000 - val loss: 1.85
69 - val accuracy: 0.5000
Epoch 34/100
             ======] - 2s 1s/step - loss: 0.0375 - accuracy: 1.0000 - val loss: 1.61
2/2 [===
96 - val accuracy: 0.4500
Epoch 35/100
2/2 [======
            ========] - 7s 3s/step - loss: 0.0754 - accuracy: 1.0000 - val loss: 2.10
04 - val_accuracy: 0.4000
Epoch 36/100
96 - val_accuracy: 0.5000
Epoch 37/100
79 - val accuracy: 0.4500
Epoch 38/100
99 - val accuracy: 0.4000
Epoch 39/100
2/2 [===
         :===============] - 3s 1s/step - loss: 0.0511 - accuracy: 0.9750 - val loss: 2.00
74 - val accuracy: 0.4500
Epoch 40/100
           ========] - 3s 1s/step - loss: 0.0462 - accuracy: 1.0000 - val loss: 3.61
2/2 [======
30 - val accuracy: 0.3500
Epoch 41/100
42 - val_accuracy: 0.5000
Epoch 42/100
35 - val_accuracy: 0.5000
Epoch 43/100
11 - val accuracy: 0.5000
Epoch 44/100
57 - val_accuracy: 0.5000
Epoch 45/100
2/2 [==
              ======] - 2s 1s/step - loss: 0.0305 - accuracy: 1.0000 - val_loss: 2.23
15 - val_accuracy: 0.5000
Epoch 46/100
58 - val_accuracy: 0.5500
```

```
Epoch 47/100
2/2 [==
           ======] - 7s 4s/step - loss: 0.0356 - accuracy: 1.0000 - val loss: 3.74
67 - val accuracy: 0.3500
Epoch 48/100
19 - val accuracy: 0.3500
Epoch 49/100
60 - val_accuracy: 0.5500
Epoch 50/100
72 - val_accuracy: 0.4000
Epoch 51/100
      2/2 [====
55 - val accuracy: 0.4500
Epoch 52/100
      2/2 [======
10 - val_accuracy: 0.4000
Epoch 53/100
41 - val_accuracy: 0.4500
Epoch 54/100
34 - val accuracy: 0.3000
Epoch 55/100
55 - val accuracy: 0.4000
Epoch 56/100
65 - val_accuracy: 0.3000
Epoch 57/100
           2/2 [==
66 - val accuracy: 0.3000
Epoch 58/100
79 - val_accuracy: 0.3000
Epoch 59/100
53 - val_accuracy: 0.4500
Epoch 60/100
12 - val accuracy: 0.3000
Epoch 61/100
        =========] - 2s 1s/step - loss: 0.0318 - accuracy: 1.0000 - val loss: 2.91
2/2 [==
36 - val accuracy: 0.4000
Epoch 62/100
2/2 [======
      91 - val accuracy: 0.3500
Epoch 63/100
24 - val_accuracy: 0.5000
Epoch 64/100
43 - val accuracy: 0.5000
Epoch 65/100
22 - val_accuracy: 0.3500
Epoch 66/100
2/2 [====
      22 - val accuracy: 0.3500
Epoch 67/100
          ========] - 8s 4s/step - loss: 0.0318 - accuracy: 1.0000 - val loss: 1.70
2/2 [======
59 - val_accuracy: 0.3500
Epoch 68/100
99 - val_accuracy: 0.4500
Epoch 69/100
43 - val_accuracy: 0.4000
Epoch 70/100
74 - val accuracy: 0.4500
Epoch 71/100
         :=========] - 2s 1s/step - loss: 0.0412 - accuracy: 1.0000 - val loss: 2.21
2/2 [======
21 - val accuracy: 0.4500
Epoch 72/100
      2/2 [======
46 - val accuracy: 0.5500
Epoch 73/100
          =======] - 2s 1s/step - loss: 0.0189 - accuracy: 1.0000 - val loss: 1.52
82 - val_accuracy: 0.5500
Epoch 74/100
```

```
83 - val accuracy: 0.5000
Epoch 75/100
2/2 [======
        ========] - 2s 1s/step - loss: 0.0204 - accuracy: 1.0000 - val loss: 1.81
86 - val accuracy: 0.5500
Epoch 76/100
48 - val accuracy: 0.4000
Epoch 77/100
86 - val_accuracy: 0.4000
Epoch 78/100
23 - val accuracy: 0.4000
Epoch 79/100
        2/2 [======
70 - val accuracy: 0.4000
Epoch 80/100
62 - val accuracy: 0.4000
Epoch 81/100
19 - val_accuracy: 0.5000
Epoch 82/100
04 - val accuracy: 0.5000
Epoch 83/100
31 - val_accuracy: 0.4500
Epoch 84/100
72 - val accuracy: 0.4500
Epoch 85/100
2/2 [===
       =========] - 2s 1s/step - loss: 0.0102 - accuracy: 1.0000 - val loss: 1.68
62 - val_accuracy: 0.4500
Epoch 86/100
31 - val accuracy: 0.4000
Epoch 87/100
18 - val_accuracy: 0.4500
Epoch 88/100
28 - val accuracy: 0.4500
Epoch 89/100
2/2 [======
     43 - val accuracy: 0.4000
Epoch 90/100
22 - val accuracy: 0.5000
Epoch 91/100
53 - val_accuracy: 0.5500
Epoch 92/100
31 - val accuracy: 0.4000
Epoch 93/100
80 - val accuracy: 0.5000
Epoch 94/100
00 - val accuracy: 0.5000
Epoch 95/100
37 - val accuracy: 0.5000
Epoch 96/100
81 - val_accuracy: 0.4500
Epoch 97/100
61 - val accuracy: 0.4500
Epoch 98/100
60 - val accuracy: 0.5500
Epoch 99/100
96 - val accuracy: 0.5500
Epoch 100/100
2/2 [======
      ==========] - 2s 1s/step - loss: 0.0131 - accuracy: 1.0000 - val loss: 1.63
75 - val_accuracy: 0.5500
```

```
In [48]:
```

```
loss, accuracy = model.evaluate(test_images)
```

In [49]:

fit the model - 200 iterations

```
model fit 200 = model.fit(train images, epochs=200, batch size=10, validation data=test images)
Epoch 1/200
           ========] - 8s 4s/step - loss: 0.0163 - accuracy: 1.0000 - val loss: 2.11
2/2 [===
00 - val accuracy: 0.4000
Epoch 2/200
             ======] - 2s 1s/step - loss: 0.0100 - accuracy: 1.0000 - val loss: 1.84
2/2 [=
74 - val accuracy: 0.4500
Fnoch 3/200
24 - val_accuracy: 0.5000
Epoch 4/200
13 - val_accuracy: 0.5500
Epoch 5/200
06 - val accuracy: 0.3500
Epoch 6/200
2/2 [===
           89 - val accuracy: 0.4000
Epoch 7/200
2/2 [=====
        58 - val accuracy: 0.4000
Epoch 8/200
45 - val accuracy: 0.4000
Epoch 9/200
55 - val accuracy: 0.4000
Epoch 10/200
41 - val_accuracy: 0.4000
Epoch 11/200
2/2 [====
            :======] - 3s 1s/step - loss: 0.0401 - accuracy: 0.9750 - val loss: 1.22
29 - val accuracy: 0.4500
Epoch 12/200
2/2 [======
          :========] - 2s 1s/step - loss: 0.0310 - accuracy: 1.0000 - val loss: 1.10
79 - val accuracy: 0.5000
Epoch 13/200
80 - val_accuracy: 0.4000
Epoch 14/200
02 - val_accuracy: 0.4000
Epoch 15/200
27 - val_accuracy: 0.4500
Epoch 16/200
33 - val_accuracy: 0.4500
Epoch 17/200
          ========] - 7s 4s/step - loss: 0.0120 - accuracy: 1.0000 - val_loss: 1.30
2/2 [==:
91 - val accuracy: 0.4500
Epoch 18/200
15 - val_accuracy: 0.4500
Epoch 19/200
88 - val_accuracy: 0.4500
Epoch 20/200
79 - val accuracy: 0.4500
Epoch 21/200
2/2 [==
             16 - val accuracy: 0.4500
Epoch 22/200
          -========] - 2s 1s/step - loss: 0.0052 - accuracy: 1.0000 - val loss: 1.60
2/2 [======
68 - val accuracy: 0.4500
Epoch 23/200
10 - val_accuracy: 0.4500
Epoch 24/200
71 - val_accuracy: 0.4500
```

```
Epoch 25/200
2/2 [==
           ======] - 6s 3s/step - loss: 0.0098 - accuracy: 1.0000 - val loss: 2.12
62 - val accuracy: 0.4500
Epoch 26/200
88 - val accuracy: 0.4000
Epoch 27/200
22 - val_accuracy: 0.5500
Epoch 28/200
15 - val_accuracy: 0.6000
Epoch 29/200
       2/2 [====
56 - val accuracy: 0.4500
Epoch 30/200
      2/2 [======
61 - val_accuracy: 0.4500
Epoch 31/200
18 - val_accuracy: 0.4500
Epoch 32/200
75 - val accuracy: 0.5000
Epoch 33/200
58 - val accuracy: 0.5500
Epoch 34/200
74 - val_accuracy: 0.5000
Epoch 35/200
2/2 [==
            :======] - 12s 6s/step - loss: 0.0073 - accuracy: 1.0000 - val loss: 1.3
243 - val accuracy: 0.4500
Fnoch 36/200
56 - val_accuracy: 0.5000
Epoch 37/200
53 - val_accuracy: 0.5500
Epoch 38/200
24 - val accuracy: 0.5000
Epoch 39/200
         =========] - 2s 1s/step - loss: 0.0180 - accuracy: 1.0000 - val loss: 1.09
2/2 [==:
73 - val accuracy: 0.5500
Epoch 40/200
2/2 [======
      33 - val accuracy: 0.3000
Epoch 41/200
33 - val_accuracy: 0.4500
Epoch 42/200
80 - val accuracy: 0.3000
Epoch 43/200
11 - val_accuracy: 0.4500
Epoch 44/200
2/2 [=====
       65 - val accuracy: 0.4500
Epoch 45/200
          ========] - 8s 4s/step - loss: 0.0116 - accuracy: 1.0000 - val loss: 2.01
2/2 [======
99 - val_accuracy: 0.5000
Epoch 46/200
47 - val_accuracy: 0.4500
Epoch 47/200
00 - val_accuracy: 0.4000
Epoch 48/200
59 - val_accuracy: 0.6000
Epoch 49/200
         :========] - 2s 1s/step - loss: 0.0086 - accuracy: 1.0000 - val loss: 1.40
2/2 [======
53 - val accuracy: 0.6000
Epoch 50/200
      2/2 [======
35 - val_accuracy: 0.5000
Epoch 51/200
           =======] - 2s 1s/step - loss: 0.0067 - accuracy: 1.0000 - val loss: 1.34
57 - val_accuracy: 0.5500
Epoch 52/200
```

```
56 - val accuracy: 0.5500
Epoch 53/200
            ======] - 2s 1s/step - loss: 0.0034 - accuracy: 1.0000 - val_loss: 1.48
2/2 [==
19 - val accuracy: 0.5500
Epoch 54/200
2/2 [==
          75 - val accuracy: 0.4000
Epoch 55/200
53 - val_accuracy: 0.4500
Epoch 56/200
78 - val_accuracy: 0.5000
Epoch 57/200
72 - val accuracy: 0.5000
Epoch 58/200
90 - val accuracy: 0.5000
Epoch 59/200
21 - val accuracy: 0.6000
Epoch 60/200
32 - val accuracy: 0.3500
Epoch 61/200
17 - val_accuracy: 0.3500
Epoch 62/200
39 - val accuracy: 0.5000
Epoch 63/200
2/2 [======
          55 - val accuracy: 0.4000
Epoch 64/200
23 - val accuracy: 0.4000
Epoch 65/200
95 - val_accuracy: 0.4000
Epoch 66/200
57 - val accuracy: 0.4000
Epoch 67/200
46 - val accuracy: 0.3000
Epoch 68/200
2/2 [=============] - 7s 3s/step - loss: 0.0061 - accuracy: 1.0000 - val_loss: 1.74
85 - val accuracy: 0.3000
Epoch 69/200
2/2 [===
          ========] - 2s 1s/step - loss: 0.0069 - accuracy: 1.0000 - val loss: 1.60
52 - val accuracy: 0.3000
Epoch 70/200
79 - val accuracy: 0.3000
Epoch 71/200
04 - val_accuracy: 0.4500
Epoch 72/200
81 - val accuracy: 0.4500
Epoch 73/200
          ========] - 2s 1s/step - loss: 0.0266 - accuracy: 1.0000 - val loss: 0.92
2/2 [===
43 - val accuracy: 0.4500
Epoch 74/200
2/2 [===
           :=======] - 2s 1s/step - loss: 0.0160 - accuracy: 1.0000 - val loss: 0.93
11 - val_accuracy: 0.5500
Epoch 75/200
89 - val accuracy: 0.5500
Epoch 76/200
04 - val accuracy: 0.4500
Epoch 77/200
52 - val_accuracy: 0.4500
Epoch 78/200
          =======] - 2s 1s/step - loss: 0.0146 - accuracy: 1.0000 - val_loss: 0.84
2/2 [=========
00 - val accuracy: 0.4500
Epoch 79/200
2/2 [===
            81 - val accuracy: 0.4000
Epoch 80/200
```

```
78 - val accuracy: 0.4000
Epoch 81/200
2/2 [==
            ======] - 6s 3s/step - loss: 0.0135 - accuracy: 1.0000 - val loss: 0.92
21 - val accuracy: 0.3500
Epoch 82/200
80 - val accuracy: 0.3500
Epoch 83/200
84 - val accuracy: 0.3500
Epoch 84/200
2/2 [======
       57 - val accuracy: 0.5500
Epoch 85/200
52 - val accuracy: 0.5500
Epoch 86/200
2/2 [======
       06 - val_accuracy: 0.5500
Epoch 87/200
87 - val_accuracy: 0.4500
Epoch 88/200
52 - val_accuracy: 0.5000
Epoch 89/200
43 - val accuracy: 0.4500
Epoch 90/200
            :======] - 2s 1s/step - loss: 0.0122 - accuracy: 1.0000 - val loss: 1.36
2/2 [===
48 - val accuracy: 0.5000
Epoch 91/200
2/2 [==
            ======] - 2s 1s/step - loss: 0.0089 - accuracy: 1.0000 - val loss: 1.29
43 - val accuracy: 0.5500
Epoch 92/200
50 - val_accuracy: 0.6000
Epoch 93/200
27 - val_accuracy: 0.4500
Epoch 94/200
2/2 [===
           70 - val accuracy: 0.6000
Epoch 95/200
            2/2 [==:
69 - val accuracy: 0.6000
Epoch 96/200
2/2 [======
           ========] - 2s 1s/step - loss: 0.0112 - accuracy: 1.0000 - val loss: 0.90
64 - val_accuracy: 0.6000
Epoch 97/200
48 - val_accuracy: 0.4000
Epoch 98/200
78 - val accuracy: 0.4500
Epoch 99/200
85 - val accuracy: 0.5000
Epoch 100/200
2/2 [===
        ==========] - 6s 3s/step - loss: 0.0056 - accuracy: 1.0000 - val loss: 1.51
12 - val accuracy: 0.6500
Epoch 101/200
          2/2 [======
06 - val accuracy: 0.6500
Epoch 102/200
02 - val_accuracy: 0.5500
Epoch 103/200
35 - val_accuracy: 0.5500
Epoch 104/200
         =========] - 8s 4s/step - loss: 0.0201 - accuracy: 1.0000 - val loss: 0.96
2/2 [======
59 - val accuracy: 0.5500
Epoch 105/200
11 - val_accuracy: 0.5500
Epoch 106/200
2/2 [==
             ======] - 3s 2s/step - loss: 0.0183 - accuracy: 1.0000 - val_loss: 1.23
87 - val_accuracy: 0.5500
Epoch 107/200
84 - val_accuracy: 0.5000
```

```
Epoch 108/200
             =======] - 2s 1s/step - loss: 0.0857 - accuracy: 0.9750 - val loss: 2.06
22 - val accuracy: 0.4000
Epoch 109/200
             2/2 [=
49 - val accuracy: 0.3500
Epoch 110/200
2/2 [======
           ========] - 7s 4s/step - loss: 0.0038 - accuracy: 1.0000 - val loss: 1.14
34 - val accuracy: 0.3500
Epoch 111/200
07 - val_accuracy: 0.5500
Epoch 112/200
29 - val accuracy: 0.5000
Epoch 113/200
2/2 [===
           ========] - 2s 1s/step - loss: 0.0278 - accuracy: 1.0000 - val loss: 1.04
14 - val accuracy: 0.5000
Epoch 114/200
2/2 [======
        08 - val accuracy: 0.5500
Epoch 115/200
            ========] - 7s 3s/step - loss: 0.0050 - accuracy: 1.0000 - val loss: 1.05
2/2 [======
38 - val_accuracy: 0.5000
Epoch 116/200
27 - val accuracy: 0.4500
Epoch 117/200
04 - val_accuracy: 0.4500
Epoch 118/200
        2/2 [====
10 - val accuracy: 0.5000
Epoch 119/200
           =======] - 2s 1s/step - loss: 0.0087 - accuracy: 1.0000 - val_loss: 0.91
2/2 [======
90 - val accuracy: 0.5000
Epoch 120/200
32 - val accuracy: 0.6000
Epoch 121/200
53 - val_accuracy: 0.6000
Epoch 122/200
34 - val accuracy: 0.6000
Epoch 123/200
93 - val_accuracy: 0.6000
Epoch 124/200
2/2 [==
             ======] - 2s 1s/step - loss: 0.0094 - accuracy: 1.0000 - val_loss: 0.96
11 - val accuracy: 0.6000
Epoch 125/200
           2/2 [==
45 - val_accuracy: 0.5500
Epoch 126/200
73 - val_accuracy: 0.5500
Epoch 127/200
83 - val_accuracy: 0.6000
Epoch 128/200
              ======] - 6s 3s/step - loss: 0.0210 - accuracy: 1.0000 - val loss: 1.00
2/2 [==
59 - val accuracy: 0.5500
Epoch 129/200
2/2 [===
          =========] - 7s 3s/step - loss: 0.0087 - accuracy: 1.0000 - val loss: 1.06
48 - val accuracy: 0.5500
Epoch 130/200
2/2 [======
        46 - val_accuracy: 0.5000
Epoch 131/200
72 - val_accuracy: 0.5000
Epoch 132/200
36 - val accuracy: 0.5000
Epoch 133/200
2/2 [======
        73 - val accuracy: 0.5000
Epoch 134/200
           2/2 [======
81 - val accuracy: 0.5000
Epoch 135/200
        2/2 [======
```

```
56 - val accuracy: 0.5000
Epoch 136/200
2/2 [======
          =======] - 7s 3s/step - loss: 0.0161 - accuracy: 1.0000 - val loss: 1.20
26 - val accuracy: 0.4000
Epoch 137/200
63 - val accuracy: 0.5000
Epoch 138/200
07 - val_accuracy: 0.5000
Epoch 139/200
78 - val accuracy: 0.5000
Epoch 140/200
2/2 [======
         ========] - 7s 3s/step - loss: 0.0100 - accuracy: 1.0000 - val loss: 1.41
79 - val accuracy: 0.4500
Epoch 141/200
70 - val accuracy: 0.4500
Epoch 142/200
64 - val_accuracy: 0.5500
Epoch 143/200
92 - val accuracy: 0.5500
Epoch 144/200
11 - val accuracy: 0.4500
Epoch 145/200
2/2 [===
         =======] - 3s 1s/step - loss: 0.0028 - accuracy: 1.0000 - val_loss: 1.57
95 - val accuracy: 0.4500
Epoch 146/200
2/2 [==
          =======] - 2s 1s/step - loss: 0.0120 - accuracy: 1.0000 - val loss: 1.86
54 - val accuracy: 0.5000
Epoch 147/200
54 - val accuracy: 0.5000
Epoch 148/200
82 - val_accuracy: 0.6000
Epoch 149/200
79 - val accuracy: 0.5000
Epoch 150/200
2/2 [===
       07 - val accuracy: 0.5500
Epoch 151/200
73 - val accuracy: 0.5500
Epoch 152/200
65 - val_accuracy: 0.7000
Epoch 153/200
63 - val accuracy: 0.6500
Epoch 154/200
35 - val accuracy: 0.6500
Epoch 155/200
21 - val accuracy: 0.6500
Epoch 156/200
      2/2 [======
76 - val accuracy: 0.6000
Epoch 157/200
77 - val_accuracy: 0.6500
Epoch 158/200
03 - val accuracy: 0.5500
Epoch 159/200
82 - val accuracy: 0.6000
Epoch 160/200
64 - val accuracy: 0.6000
Epoch 161/200
2/2 [======
        90 - val_accuracy: 0.5500
Epoch 162/200
89 - val accuracy: 0.5500
Epoch 163/200
```

```
66 - val accuracy: 0.5500
Epoch 164/200
            =======] - 8s 4s/step - loss: 0.0039 - accuracy: 1.0000 - val loss: 1.42
2/2 [=
02 - val accuracy: 0.6500
Epoch 165/200
2/2 [==
             ======] - 2s 1s/step - loss: 0.0112 - accuracy: 1.0000 - val loss: 1.21
32 - val accuracy: 0.4500
Epoch 166/200
64 - val_accuracy: 0.5000
Epoch 167/200
15 - val accuracy: 0.5000
Epoch 168/200
2/2 [====
          66 - val accuracy: 0.5000
Epoch 169/200
2/2 [======
          =========] - 2s 1s/step - loss: 0.0026 - accuracy: 1.0000 - val_loss: 1.33
24 - val accuracy: 0.5000
Epoch 170/200
2/2 [===
         63 - val accuracy: 0.5000
Epoch 171/200
79 - val accuracy: 0.5500
Epoch 172/200
16 - val_accuracy: 0.5500
Epoch 173/200
       2/2 [======
82 - val accuracy: 0.5500
Epoch 174/200
           ========] - 8s 4s/step - loss: 0.0013 - accuracy: 1.0000 - val loss: 1.17
2/2 [===
57 - val accuracy: 0.5500
Epoch 175/200
       2/2 [======
83 - val_accuracy: 0.5500
Epoch 176/200
38 - val_accuracy: 0.5500
Epoch 177/200
41 - val_accuracy: 0.5000
Epoch 178/200
49 - val accuracy: 0.5000
Epoch 179/200
42 - val_accuracy: 0.5000
Epoch 180/200
           =======] - 2s 1s/step - loss: 0.0037 - accuracy: 1.0000 - val_loss: 1.62
2/2 [==
00 - val accuracy: 0.4500
Epoch 181/200
68 - val_accuracy: 0.4500
Epoch 182/200
09 - val_accuracy: 0.5000
Epoch 183/200
2/2 [==:
             ======] - 8s 4s/step - loss: 8.4242e-04 - accuracy: 1.0000 - val loss:
1.5801 - val accuracy: 0.5500
Epoch 184/200
2/2 [==
           =======] - 9s 4s/step - loss: 0.0046 - accuracy: 1.0000 - val loss: 1.73
20 - val accuracy: 0.5000
Epoch 185/200
47 - val accuracy: 0.5000
Epoch 186/200
17 - val_accuracy: 0.5500
Epoch 187/200
69 - val accuracy: 0.4000
Epoch 188/200
70 - val_accuracy: 0.4500
Epoch 189/200
2/2 [====
       31 - val accuracy: 0.4500
Epoch 190/200
2/2 [======
            ========] - 3s 2s/step - loss: 0.0026 - accuracy: 1.0000 - val loss: 1.78
61 - val accuracy: 0.4500
```

```
Epoch 191/200
             :=======] - 2s 1s/step - loss: 0.0045 - accuracy: 1.0000 - val loss: 1.91
2/2 [===
35 - val accuracy: 0.4500
Epoch 192/200
1.9168 - val accuracy: 0.4500
Epoch 193/200
27 - val_accuracy: 0.5000
Epoch 194/200
06 - val_accuracy: 0.5500
Epoch 195/200
            2/2 [======
74 - val accuracy: 0.4500
Epoch 196/200
28 - val_accuracy: 0.4000
Epoch 197/200
58 - val_accuracy: 0.4000
Epoch 198/200
71 - val_accuracy: 0.4000
Epoch 19\overline{9}/200
27 - val accuracy: 0.4000
Epoch 200/200
73 - val_accuracy: 0.4000
In [50]:
# loss and accuracy after 200 iterations
loss, accuracy = model.evaluate(test_images)
1/1 [============= ] - 0s 2ms/step - loss: 1.0973 - accuracy: 0.4000
accuracy is reduced after 200 iterations. One reason could be less number of sample messages.
In [51]:
pred = model.predict(test images)
In [52]:
df_test['pred_category'] = np.argmax(pred, axis=1)
In [531:
```

df_test['pred_category'] = df_test['pred_category'].replace({0:'cat', 1: 'dog'})

In [54]:

df_test.head(20)

Out[54]:

	filename	category	pred_category
0	101.jpg	cat	cat
1	102.jpg	cat	dog
2	103.jpg	cat	cat
3	104.jpg	cat	dog
4	105.jpg	cat	cat
5	106.jpg	cat	cat
6	107.jpg	cat	dog
7	108.jpg	cat	dog
8	109.jpg	cat	dog
9	110.jpg	cat	dog
10	101.jpg	dog	cat
11	102.jpg	dog	cat
12	103.jpg	dog	dog
13	104.jpg	dog	dog
14	105.jpg	dog	dog
15	106.jpg	dog	cat
16	107.jpg	dog	cat
17	108.jpg	dog	dog
18	109.jpg	dog	dog
19	110.jpg	dog	dog

In []: