# **Assignment 6**

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## **Problem Statement:**

Object detection using Transfer Learning of CNN architectures

## Importing the libraries

# In [1]:

```
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
tf.__version__
```

## Out[1]:

'2.8.0'

#### **Preprocessing for dataset**

#### In [2]:

#### In [3]:

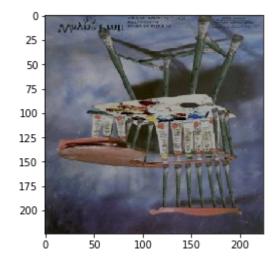
```
root_dir = '101_ObjectCategories'
img_generator_flow_train = img_generator.flow_from_directory(
    directory=root_dir,
    target_size=(224, 224),
    batch_size=32,
    shuffle=True,
    subset="training")

img_generator_flow_valid = img_generator.flow_from_directory(
    directory=root_dir,
    target_size=(224, 224),
    batch_size=32,
    shuffle=True,
    subset="validation")
```

Found 6444 images belonging to 102 classes. Found 2700 images belonging to 102 classes.

## In [4]:

```
imgs, labels = next(iter(img_generator_flow_train))
for img, label in zip(imgs, labels):
   plt.imshow(img)
   plt.show()
```





## a. Load in a pretrained model (InceptionV3)

### In [5]:

# b. Freeze parameters (weights) in model's lower convolutional layers

### In [6]:

```
base_model.trainable = False
```

## c. Add custom classifier with several layers of trainable parameters to model

## In [7]:

```
model = tf.keras.Sequential([
    base_model,
    tf.keras.layers.MaxPooling2D(),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(102, activation="softmax")
])
```

# In [8]:

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
inception_v3 (Functional)	(None, 5, 5, 2048)	21802784
<pre>max_pooling2d_4 (MaxPooling 2D)</pre>	(None, 2, 2, 2048)	0
flatten (Flatten)	(None, 8192)	0
dense (Dense)	(None, 102)	835686

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Total params: 22,638,470 Trainable params: 835,686

Non-trainable params: 21,802,784

# d. Train classifier layers on training data available for task

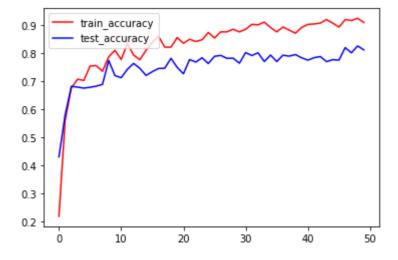
### In [9]:

#### In [10]:

```
model.fit(img generator flow train, validation data=img generator flow valid, steps
Epoch 1/50
0 - categorical accuracy: 0.2188 - val loss: 4.9710 - val categorica
l accuracy: 0.4307
Epoch 2/50
20/20 [============== ] - 281s 15s/step - loss: 3.323
5 - categorical accuracy: 0.5609 - val loss: 2.7433 - val categorica
l accuracy: 0.5800
Epoch 3/50
20/20 [============= ] - 277s 14s/step - loss: 2.033
5 - categorical accuracy: 0.6766 - val loss: 2.1982 - val categorica
l accuracy: 0.6822
Epoch 4/50
7 - categorical accuracy: 0.7078 - val loss: 2.1987 - val categorica
l accuracy: 0.6796
Epoch 5/50
20/20 [======
                   ==========] - 280s 15s/step - loss: 1.995
8 - categorical accuracy: 0.7031 - val loss: 2.3371 - val categorica
```

#### In [11]:

```
# Visualise train / Valid Accuracy
plt.plot(model.history.history["categorical_accuracy"], c="r", label="train_accuracy
plt.plot(model.history.history["val_categorical_accuracy"], c="b", label="test_accuracy"]
plt.legend(loc="upper left")
plt.show()
```



#### e. Fine-tune hyper parameters and unfreeze more layers as needed

#### In [12]:

#### In [ ]:

```
model.fit(img generator flow train, validation data=img generator flow valid, steps
Epoch 1/50
20/20 [============ ] - 474s 24s/step - loss: 3.9578
- categorical accuracy: 0.4359 - val loss: 126.2205 - val_categorical_
accuracy: 0.0163
Epoch 2/50
- categorical accuracy: 0.2875 - val loss: 13552.6338 - val categorica
l accuracy: 0.0481
Epoch 3/50
20/20 [============ ] - 462s 24s/step - loss: 3.5396
- categorical accuracy: 0.3250 - val loss: 65.9670 - val categorical a
ccuracy: 0.0663
Epoch 4/50
- categorical accuracy: 0.3812 - val loss: 62.4981 - val categorical a
ccuracy: 0.0722
Epoch 5/50
20/20 [============== ] - 463s 24s/step - loss: 2.8024
- categorical accuracy: 0.4109 - val loss: 4.9331 - val categorical ac
curacy: 0.2341
Epoch 6/50
- categorical_accuracy: 0.4328 - val_loss: 4.8107 - val categorical ac
curacy: 0.2370
Epoch 7/50
- categorical accuracy: 0.5250 - val loss: 3.3150 - val categorical ac
curacy: 0.4315
Epoch 8/50
20/20 [============ ] - 458s 24s/step - loss: 1.6358
- categorical accuracy: 0.6047 - val loss: 2.4610 - val categorical ac
curacy: 0.4459
Epoch 9/50
20/20 [============ ] - 460s 24s/step - loss: 1.6275
- categorical accuracy: 0.5953 - val loss: 2.1677 - val categorical ac
curacy: 0.5081
Epoch 10/50
gorical accuracy: 0.5938
In [ ]:
```