

Assignment 3

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Problem Statement :

Build the Image classification model

Importing the libraries

In [1]:

```
import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
```

a. Loading and preprocessing the image data

Grabbing CIFAR10 dataset

In [2]:

```
(train_images, train_labels), (test_images, test_labels) = datasets.cifar10.load_data
train_images, test_images = train_images / 255.0, test_images / 255.0
```

In [4]:

```
type(train_images)
```

Out[4]:

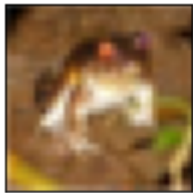
numpy.ndarray

Showing images of mentioned categories

In [5]:

```
class_names = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']

plt.figure(figsize=(10,10))
for i in range(10):
    plt.subplot(5,5,i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    plt.imshow(train_images[i])
    plt.xlabel(class_names[train_labels[i][0]])
plt.show()
```



frog



truck



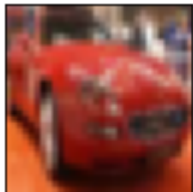
truck



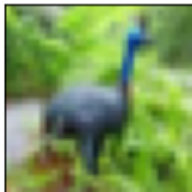
deer



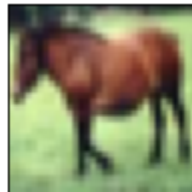
automobile



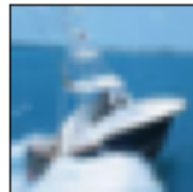
automobile



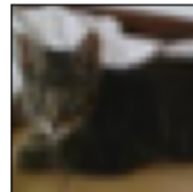
bird



horse



ship



cat

b. Defining the model's architecture

Building CNN model

In [6]:

```

model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10))

model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 30, 30, 32)	896
max_pooling2d (MaxPooling2D)	(None, 15, 15, 32)	0
conv2d_1 (Conv2D)	(None, 13, 13, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 6, 6, 64)	0
conv2d_2 (Conv2D)	(None, 4, 4, 64)	36928
flatten (Flatten)	(None, 1024)	0
dense (Dense)	(None, 64)	65600
dense_1 (Dense)	(None, 10)	650
Total params: 122,570		
Trainable params: 122,570		
Non-trainable params: 0		

c. Training the model

Model compilation

In [7]:

```

model.compile(optimizer='adam', loss=tf.keras.losses.SparseCategoricalCrossentropy(f

```

d. Estimating the model's performance

In [9]:

```
epochs = 10  
h = model.fit(train_images, train_labels, epochs=epochs, validation_data=(test_imag
```

```
Epoch 1/10  
1563/1563 [=====] - 66s 42ms/step - loss: 1.1  
581 - accuracy: 0.5890 - val_loss: 1.0580 - val_accuracy: 0.6277  
Epoch 2/10  
1563/1563 [=====] - 69s 44ms/step - loss: 1.0  
107 - accuracy: 0.6472 - val_loss: 0.9943 - val_accuracy: 0.6511  
Epoch 3/10  
1563/1563 [=====] - 66s 43ms/step - loss: 0.9  
166 - accuracy: 0.6772 - val_loss: 0.9544 - val_accuracy: 0.6693  
Epoch 4/10  
1563/1563 [=====] - 65s 41ms/step - loss: 0.8  
453 - accuracy: 0.7075 - val_loss: 0.9113 - val_accuracy: 0.6789  
Epoch 5/10  
1563/1563 [=====] - 66s 42ms/step - loss: 0.7  
916 - accuracy: 0.7241 - val_loss: 0.9141 - val_accuracy: 0.6884  
Epoch 6/10  
1563/1563 [=====] - 65s 42ms/step - loss: 0.7  
393 - accuracy: 0.7418 - val_loss: 0.8678 - val_accuracy: 0.7065  
Epoch 7/10  
1563/1563 [=====] - 66s 42ms/step - loss: 0.6  
958 - accuracy: 0.7573 - val_loss: 0.9028 - val_accuracy: 0.6945  
Epoch 8/10  
1563/1563 [=====] - 66s 42ms/step - loss: 0.6  
597 - accuracy: 0.7692 - val_loss: 0.8688 - val_accuracy: 0.7020  
Epoch 9/10  
1563/1563 [=====] - 65s 42ms/step - loss: 0.6  
243 - accuracy: 0.7818 - val_loss: 0.8780 - val_accuracy: 0.7071  
Epoch 10/10  
1563/1563 [=====] - 65s 42ms/step - loss: 0.5  
905 - accuracy: 0.7920 - val_loss: 0.8929 - val_accuracy: 0.7151
```

In []: