

```
In [1]: 1 import pandas as pd
        2 import numpy as np
        3 import matplotlib.pyplot as plt
        4 import seaborn as sns
        5 import warnings
        6 from keras.datasets import fashion_mnist,cifar10
        7 from keras.models import Sequential
        8 from keras.layers import Dense, Activation, Dropout, Flatten, Conv2D, MaxPooling2D
        9 from keras.layers.normalization import BatchNormalization
       10 warnings.filterwarnings("ignore")
```

```
In [2]: 1 (xtrain,ytrain),(xtest,ytest) = cifar10.load_data()
        2 class_names=['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship',
```

```
In [3]: 1 print("Total Number of images :- ",xtrain.shape[0]+ytrain.shape[0])
```

Total Number of images :- 100000

```
In [4]: 1 print("No of classes present in the data :-",len(np.unique(ytrain)))
```

No of classes present in the data :- 10

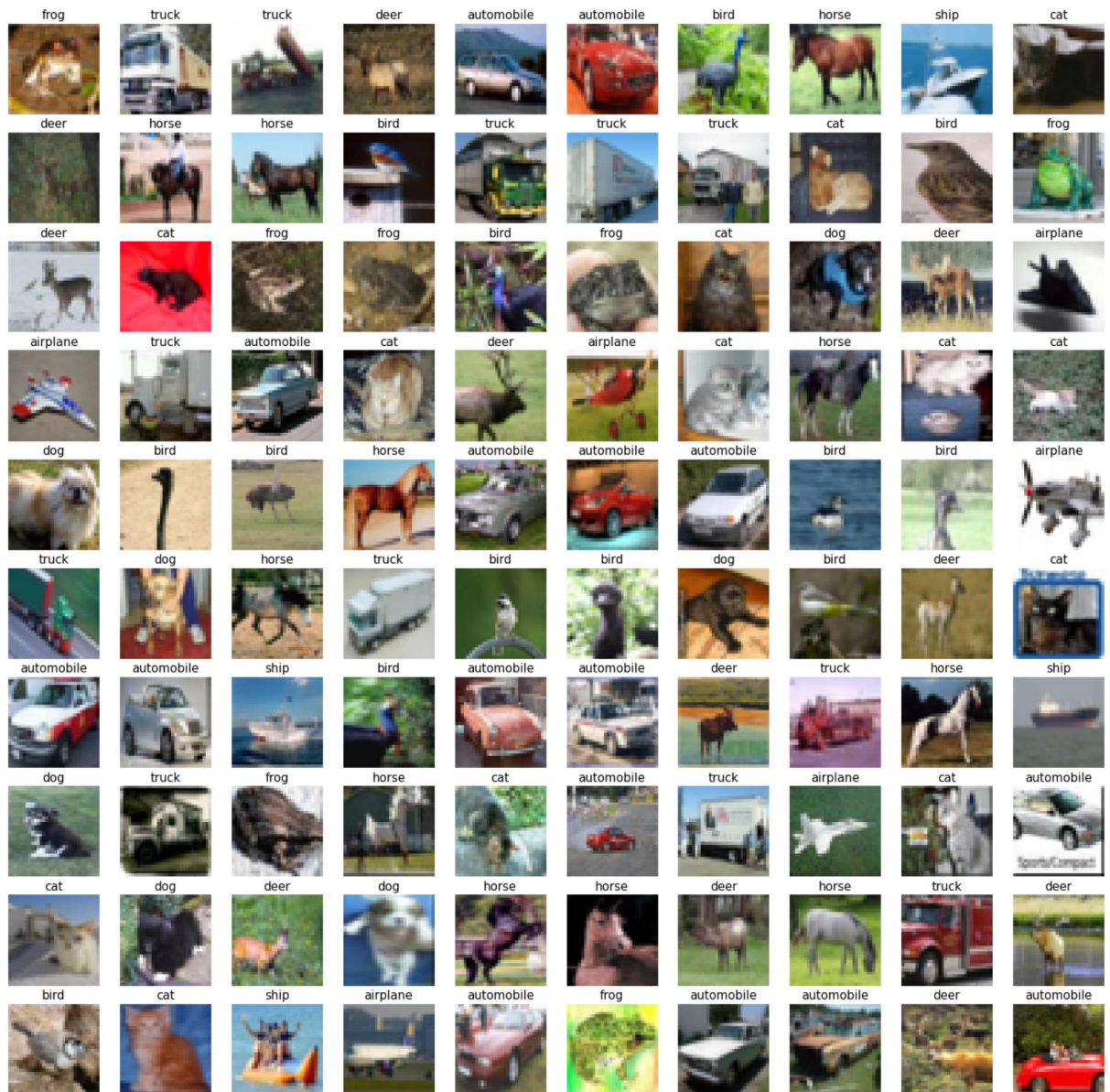
```
In [5]: 1 print("shape of images :- ",xtrain[0].shape)
```

shape of images :- (32, 32, 3)

```
In [6]: 1 fig = plt.figure()
2         _, axes = plt.subplots(10,10, figsize=(25,25))
3         axes = axes.flatten()
4         for img, ax, k in zip(xtrain, axes, ytrain):
5             ax.axis("off")
6             ax.set_title(class_names[k[0]], fontsize=15)
7             ax.imshow(img)
8         plt.suptitle('Training Data', fontsize=25)
9         plt.show()
```

<Figure size 432x288 with 0 Axes>

Training Data

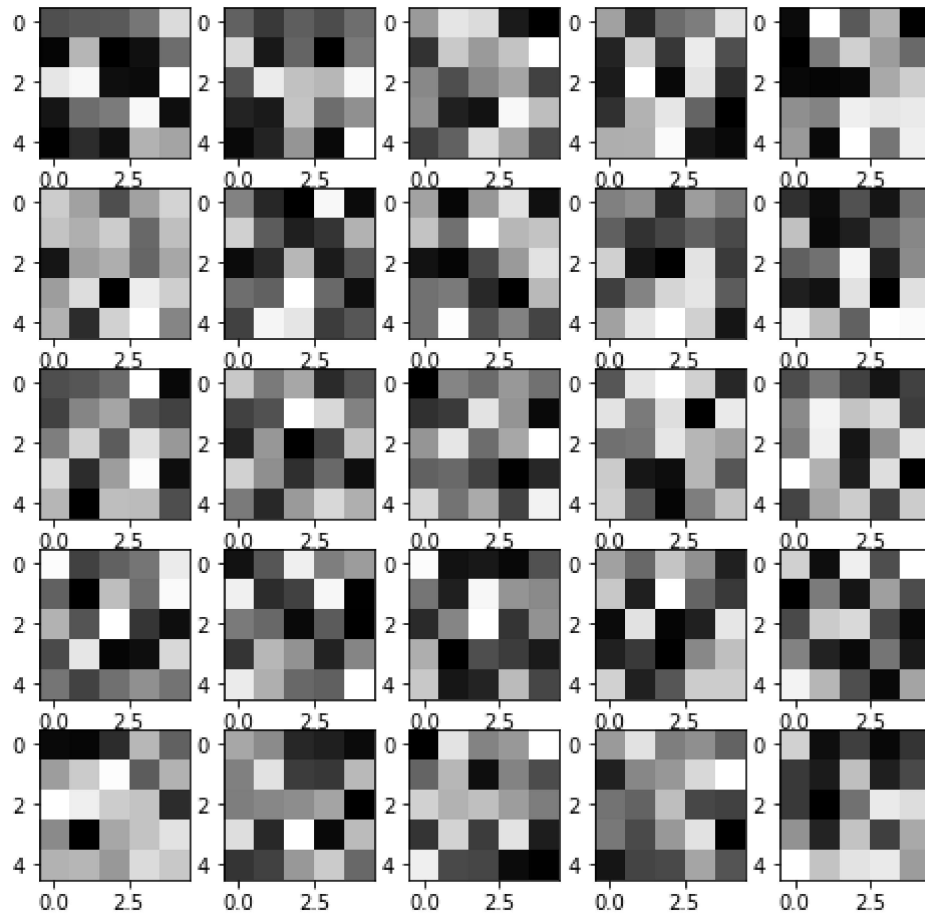


```
In [7]: 1 from keras.utils import to_categorical  
        2 ytrain=to_categorical(ytrain)  
        3 ytest=to_categorical(ytest)
```

```

In [8]: 1 plt.figure(figsize=(8,8))
        2 layers = Sequential()
        3 x = Conv2D(filters=64,kernel_size=(5,5),input_shape=(32,32,3))
        4 layers.add(x)
        5 x1w = x.get_weights()[0][:,:,0,:]
        6 for i in range(1,26):
        7     plt.subplot(5,5,i)
        8     plt.imshow(x1w[:, :, i], interpolation="nearest", cmap="gray")
        9 plt.show()

```



In [9]:

```
1 AlexNet = Sequential()
2
3 #1st Convolutional Layer
4 AlexNet.add(Conv2D(filters=96, input_shape=(32,32,3), kernel_size=(11,11), strides=(4
5 AlexNet.add(BatchNormalization())
6 AlexNet.add(Activation('relu'))
7
8 # 1st Maxpooling Layer
9 AlexNet.add(MaxPooling2D(pool_size=(2,2), strides=(2,2), padding='same'))
10
11 #2nd Convolutional Layer
12 AlexNet.add(Conv2D(filters=256, kernel_size=(5, 5), strides=(1,1), padding='same'))
13 AlexNet.add(BatchNormalization())
14 AlexNet.add(Activation('relu'))
15
16 # 2nd Maxpooling Layer
17 AlexNet.add(MaxPooling2D(pool_size=(2,2), strides=(2,2), padding='same'))
18
19 #3rd Convolutional Layer
20 AlexNet.add(Conv2D(filters=384, kernel_size=(3,3), strides=(1,1), padding='same'))
21 AlexNet.add(BatchNormalization())
22 AlexNet.add(Activation('relu'))
23
24 #4th Convolutional Layer
25 AlexNet.add(Conv2D(filters=384, kernel_size=(3,3), strides=(1,1), padding='same'))
26 AlexNet.add(BatchNormalization())
27 AlexNet.add(Activation('relu'))
28
29 #5th Convolutional Layer
30 AlexNet.add(Conv2D(filters=256, kernel_size=(3,3), strides=(1,1), padding='same'))
31 AlexNet.add(BatchNormalization())
32 AlexNet.add(Activation('relu'))
33
34 # 3rd Maxpooling Layer
35 AlexNet.add(MaxPooling2D(pool_size=(2,2), strides=(2,2), padding='same'))
36
37 #Passing it to a Fully Connected layer
38 AlexNet.add(Flatten())
39
40 # 1st Fully Connected Layer
41 AlexNet.add(Dense(4096, input_shape=(32,32,3,)))
42 AlexNet.add(BatchNormalization())
43 AlexNet.add(Activation('relu'))
44
45 # Add Dropout to prevent overfitting
46 AlexNet.add(Dropout(0.4))
47
48 #2nd Fully Connected Layer
49 AlexNet.add(Dense(4096))
50 AlexNet.add(BatchNormalization())
51 AlexNet.add(Activation('relu'))
52 #Add Dropout
53 AlexNet.add(Dropout(0.4))
54
55 #3rd Fully Connected Layer
56 AlexNet.add(Dense(1000))
```

```

57 AlexNet.add(BatchNormalization())
58 AlexNet.add(Activation('relu'))
59 #Add Dropout
60 AlexNet.add(Dropout(0.4))
61
62 #Output Layer
63 AlexNet.add(Dense(10))
64 AlexNet.add(BatchNormalization())
65 AlexNet.add(Activation('softmax'))
66
67 #Model Summary
68 AlexNet.summary()

```

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
=====		
conv2d_1 (Conv2D)	(None, 8, 8, 96)	34944
batch_normalization (Batch Normalization)	(None, 8, 8, 96)	384
activation (Activation)	(None, 8, 8, 96)	0
max_pooling2d (MaxPooling2D)	(None, 4, 4, 96)	0
conv2d_2 (Conv2D)	(None, 4, 4, 256)	614656
batch_normalization_1 (Batch Normalization)	(None, 4, 4, 256)	1024
activation_1 (Activation)	(None, 4, 4, 256)	0
max_pooling2d_1 (MaxPooling2D)	(None, 2, 2, 256)	0
conv2d_3 (Conv2D)	(None, 2, 2, 384)	885120
batch_normalization_2 (Batch Normalization)	(None, 2, 2, 384)	1536
activation_2 (Activation)	(None, 2, 2, 384)	0
conv2d_4 (Conv2D)	(None, 2, 2, 384)	1327488
batch_normalization_3 (Batch Normalization)	(None, 2, 2, 384)	1536
activation_3 (Activation)	(None, 2, 2, 384)	0
conv2d_5 (Conv2D)	(None, 2, 2, 256)	884992
batch_normalization_4 (Batch Normalization)	(None, 2, 2, 256)	1024
activation_4 (Activation)	(None, 2, 2, 256)	0
max_pooling2d_2 (MaxPooling2D)	(None, 1, 1, 256)	0
flatten (Flatten)	(None, 256)	0
dense (Dense)	(None, 4096)	1052672
batch_normalization_5 (Batch Normalization)	(None, 4096)	16384

activation_5 (Activation)	(None, 4096)	0
dropout (Dropout)	(None, 4096)	0
dense_1 (Dense)	(None, 4096)	16781312
batch_normalization_6 (Batch Normalization)	(None, 4096)	16384
activation_6 (Activation)	(None, 4096)	0
dropout_1 (Dropout)	(None, 4096)	0
dense_2 (Dense)	(None, 1000)	4097000
batch_normalization_7 (Batch Normalization)	(None, 1000)	4000
activation_7 (Activation)	(None, 1000)	0
dropout_2 (Dropout)	(None, 1000)	0
dense_3 (Dense)	(None, 10)	10010
batch_normalization_8 (Batch Normalization)	(None, 10)	40
activation_8 (Activation)	(None, 10)	0
=====		
Total params: 25,730,506		
Trainable params: 25,709,350		
Non-trainable params: 21,156		

In [10]: 1 AlexNet.compile(loss = 'categorical\_crossentropy', optimizer= 'adam', metrics=['accuracy'])

In [11]: 1 batch\_size= 256  
2 epochs=250

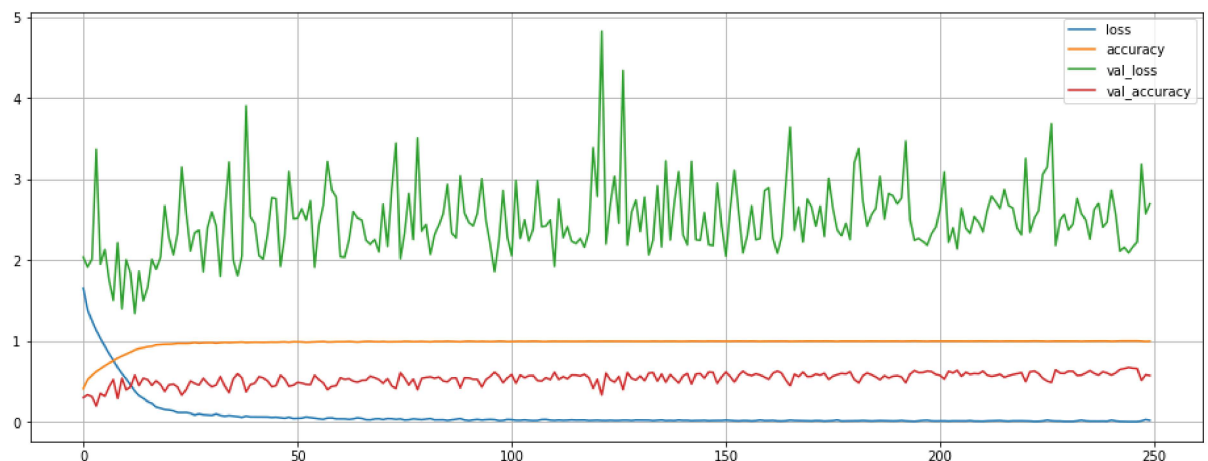
In [12]: 1 from sklearn.model\_selection import train\_test\_split  
2 xtrain,xval,ytrain,yval=train\_test\_split(xtrain,ytrain,test\_size=0.3,random\_state=0)



```
In [13]: 1 history = AlexNet.fit(xtrain,ytrain,validation_data=(xval,yval),batch_size=batch_size,epochs=250)
137/137 [=====] - 6s 43ms/step - loss: 0.0350 - accuracy: 0.9
829 - val_loss: 2.7593 - val_accuracy: 0.4448
Epoch 47/250
137/137 [=====] - 6s 42ms/step - loss: 0.0512 - accuracy: 0.9
844 - val_loss: 1.9196 - val_accuracy: 0.5747
Epoch 48/250
137/137 [=====] - 6s 43ms/step - loss: 0.0367 - accuracy: 0.9
888 - val_loss: 2.2962 - val_accuracy: 0.5295
Epoch 49/250
137/137 [=====] - 6s 43ms/step - loss: 0.0527 - accuracy: 0.9
836 - val_loss: 3.0946 - val_accuracy: 0.4366
Epoch 50/250
137/137 [=====] - 6s 43ms/step - loss: 0.0350 - accuracy: 0.9
893 - val_loss: 2.5101 - val_accuracy: 0.4505
Epoch 51/250
137/137 [=====] - 6s 42ms/step - loss: 0.0402 - accuracy: 0.9
874 - val_loss: 2.5161 - val_accuracy: 0.4858
Epoch 52/250
137/137 [=====] - 6s 42ms/step - loss: 0.0417 - accuracy: 0.9
870 - val_loss: 2.6322 - val_accuracy: 0.4776
```

```
In [14]: 1 history = pd.DataFrame(history.history)
```

```
In [15]: 1 plt.figure(figsize=(16,6))
2 plt.plot(history["loss"],label='loss')
3 plt.plot(history["accuracy"],label='accuracy')
4 plt.plot(history["val_loss"],label='val_loss')
5 plt.plot(history["val_accuracy"],label='val_accuracy')
6 plt.legend()
7 plt.grid("whitegrid")
8 plt.show()
```



```
In [16]: 1 ypredict = np.argmax(AlexNet.predict(xtest),axis=1)
```

```
In [17]: 1 ytest = np.argmax(ytest,axis=1)
```



In [18]: 1 ypredict

Out[18]: array([3, 8, 8, ..., 5, 1, 7])

In [19]: 1 from sklearn.metrics import confusion\_matrix, plot\_confusion\_matrix, accuracy\_score  
2 print(confusion\_matrix(ypredict, ytest))

```
[[534  8 45 21 12  6  2 14 16 21]
 [ 40 841 21 38 20 15 23 36 30 442]
 [112 17 689 164 256 146 97 155 16 46]
 [  3  9 24 331 39 101 29 31  7  9]
 [ 14  5 26 40 437 28 13 66  5  7]
 [  8  6 64 187 69 576 29 115  8 30]
 [ 22 12 65 113 91 65 766 26  4 30]
 [  9  6 10  9 22 20  6 513  1 15]
 [257 92 56 94 54 43 35 42 913 234]
 [  1  4  0  3  0  0  0  2  0 166]]
```

In [20]: 1 accuracy\_score(ypredict, ytest)

Out[20]: 0.5766