

```
In [1]: 1 import numpy as np
        2 import pandas as pd
        3 import matplotlib.pyplot as plt
        4 from keras.datasets import cifar10
```

```
In [2]: 1 (xtrain,ytrain),(xtest,ytest) = cifar10.load_data()
```

```
In [3]: 1 print("xtrain :- ",xtrain.shape)
        2 print("ytrain :- ",ytrain.shape)
        3 print("xtest :- ",xtest.shape)
        4 print("ytest :- ",ytest.shape)
```

```
xtrain :- (50000, 32, 32, 3)
ytrain :- (50000, 1)
xtest :- (10000, 32, 32, 3)
ytest :- (10000, 1)
```

```
In [4]: 1 from keras.utils import to_categorical
        2 ytrain = to_categorical(ytrain)
        3 ytest = to_categorical(ytest)
```

```
In [5]: 1 from keras.layers import AveragePooling2D,Conv2D,Dense,Activation,Flatten
        2 from keras.models import Sequential
```

```
In [6]: 1 xtrain.shape[-3:]
```

```
Out[6]: (32, 32, 3)
```

In [7]:

```

1 model = Sequential()
2 model.add(Conv2D(64,kernel_size=(3,3),strides = (2,2),input_shape=xtrain.shape[1:]))
3 model.add(AveragePooling2D((2,2)))
4 model.add(Conv2D(20,(3,3)))
5 model.add(AveragePooling2D((2,2)))
6 model.add(Flatten())
7 model.add(Dense(16))
8 model.add(Dense(12))
9 model.add(Dense(10,))
10 model.add(Activation('softmax'))
11 model.compile(loss = 'categorical_crossentropy', optimizer= 'adam', metrics=['accuracy'])
12 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 15, 15, 64)	1792
average_pooling2d (AveragePo	(None, 7, 7, 64)	0
conv2d_1 (Conv2D)	(None, 5, 5, 20)	11540
average_pooling2d_1 (Average	(None, 2, 2, 20)	0
flatten (Flatten)	(None, 80)	0
dense (Dense)	(None, 16)	1296
dense_1 (Dense)	(None, 12)	204
dense_2 (Dense)	(None, 10)	130
activation (Activation)	(None, 10)	0
=====		
Total params: 14,962		
Trainable params: 14,962		
Non-trainable params: 0		

```
In [8]: 1 history = model.fit(xtrain,ytrain,batch_size=256,epochs=250)

Epoch 1/250
196/196 [=====] - 4s 6ms/step - loss: 15.1255 - accuracy: 0.1631
Epoch 2/250
196/196 [=====] - 1s 5ms/step - loss: 2.4541 - accuracy: 0.2514
Epoch 3/250
196/196 [=====] - 1s 5ms/step - loss: 2.1407 - accuracy: 0.2757
Epoch 4/250
196/196 [=====] - 1s 5ms/step - loss: 2.0309 - accuracy: 0.2984
Epoch 5/250
196/196 [=====] - 1s 5ms/step - loss: 1.9733 - accuracy: 0.3106
Epoch 6/250
196/196 [=====] - 1s 5ms/step - loss: 1.9349 - accuracy: 0.3246
Epoch 7/250
196/196 [=====] - 1s 4ms/step - loss: 1.9110 - accuracy: 0.3341
```

```
In [9]: 1 history = pd.DataFrame(history.history)
        2 history
```

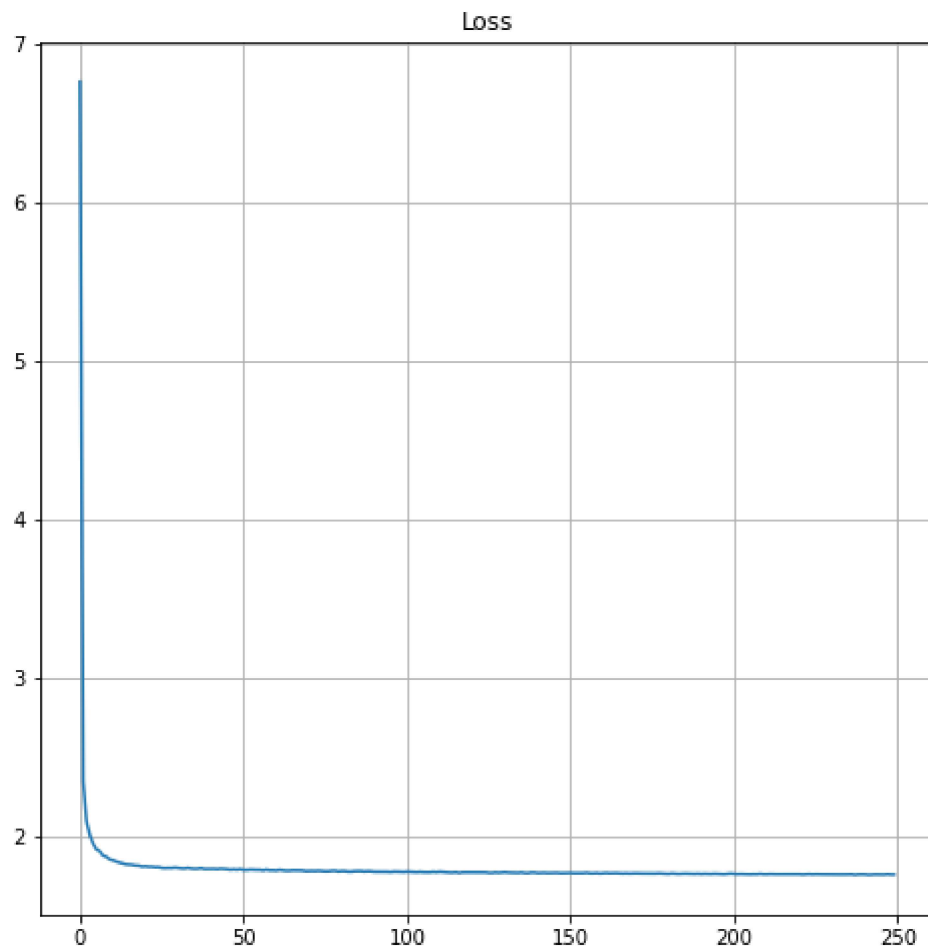
Out[9]:

	loss	accuracy
0	6.765108	0.19534
1	2.344730	0.25938
2	2.094036	0.28318
3	2.009014	0.30362
4	1.952580	0.31682
...
245	1.758080	0.39462
246	1.757231	0.39664
247	1.759657	0.39672
248	1.757356	0.39666
249	1.757499	0.39528

250 rows × 2 columns

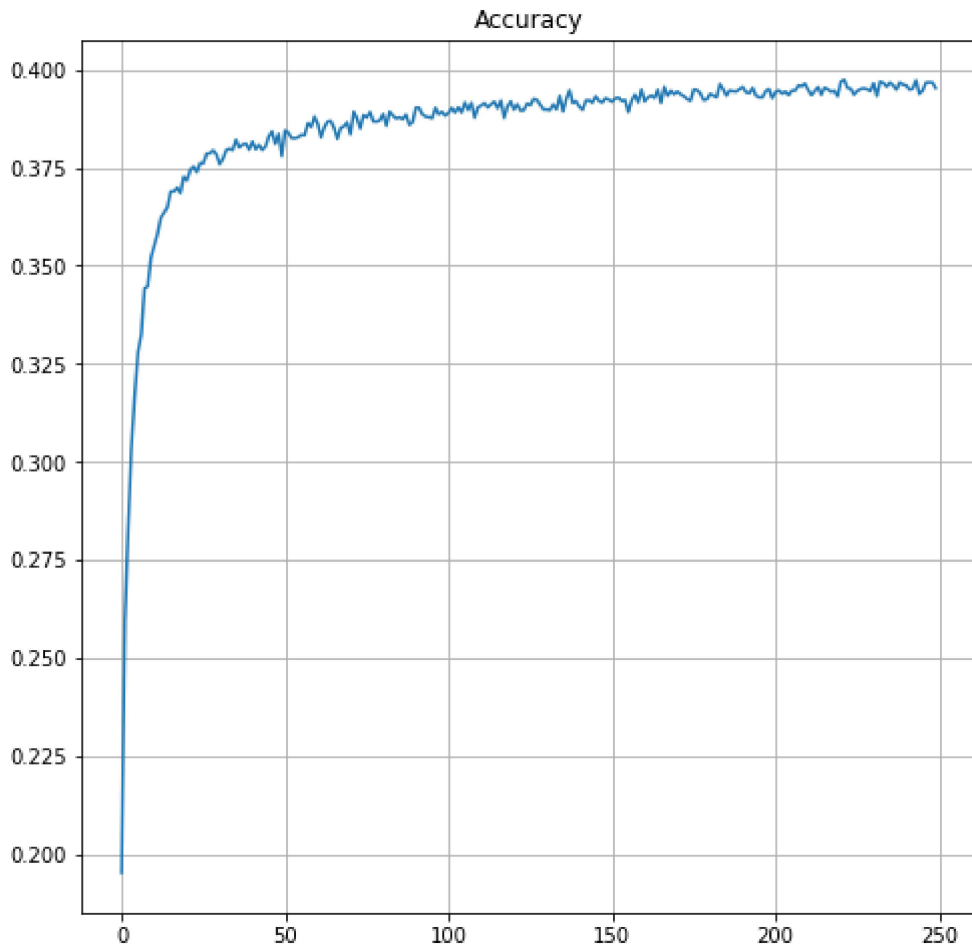
```
In [10]: 1 history.loss.plot.line(figsize=(8,8),title = "Loss",grid = "whitegrid",)
```

```
Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x7f0b660ae790>
```



```
In [11]: 1 history.accuracy.plot(figsize=(8,8),title = "Accuracy",grid = "whitegrid")
```

```
Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x7f0b5fff68d0>
```



```
In [12]: 1 predictions = np.argmax(model.predict(xtest),axis=1)
```

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

```
Out[13]: array([3, 8, 8, ..., 5, 5, 7])
```

```
In [14]: 1 ytest = np.argmax(ytest,axis=1)
```

```
In [15]: 1 from sklearn.metrics import accuracy_score, confusion_matrix  
        2 print(accuracy_score(predictions, ytest))
```

0.3779

```
In [16]: 1 print(confusion_matrix(predictions, ytest))
```

```
[[511 102 109 47 61 53 23 53 266 110]  
 [ 41 432 43 39 14 34 18 34 67 160]  
 [ 52 21 225 97 103 82 59 57 13 22]  
 [ 21 47 77 255 67 134 141 57 28 32]  
 [ 24 44 162 73 310 93 135 78 19 22]  
 [ 33 55 100 201 106 362 90 73 63 37]  
 [ 25 30 117 119 129 76 432 38 10 52]  
 [ 86 53 124 92 165 108 68 505 57 78]  
 [151 57 29 26 19 31 10 26 356 96]  
 [ 56 159 14 51 26 27 24 79 121 391]]
```