

1. Problem Statement :

Classification of ciFAR-10 images using a convolution neural network.

2. Proposed Solution :

3. Implementation details :

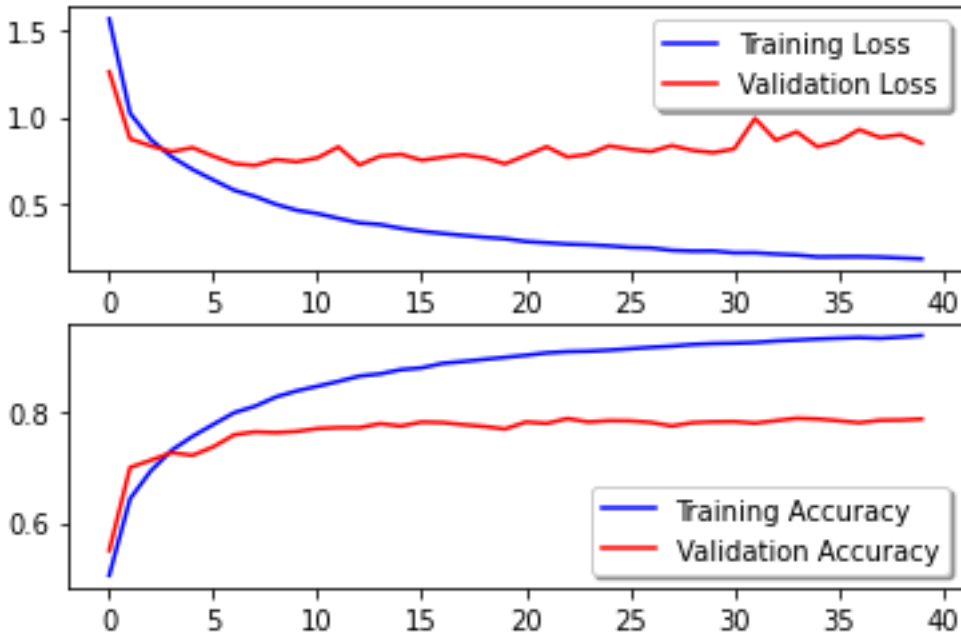
A. Download Data

B. Create Covnet Model

Model: "sequential_2"

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 32, 32, 32)	896
max_pooling2d_3 (MaxPooling2)	(None, 16, 16, 32)	0
batch_normalization_4 (Batch Normalization)	(None, 16, 16, 32)	128
conv2d_4 (Conv2D)	(None, 16, 16, 128)	36992
max_pooling2d_4 (MaxPooling2)	(None, 8, 8, 128)	0
batch_normalization_5 (Batch Normalization)	(None, 8, 8, 128)	512
dropout_3 (Dropout)	(None, 8, 8, 128)	0
flatten_2 (Flatten)	(None, 8192)	0
dense_3 (Dense)	(None, 512)	4194816
batch_normalization_6 (Batch Normalization)	(None, 512)	2048
dropout_4 (Dropout)	(None, 512)	0
dense_4 (Dense)	(None, 10)	5130
Total params: 4,240,522		
Trainable params: 4,239,178		
Non-trainable params: 1,344		

C. Perform Hyperparameter tuning



As you can after 17 epoch validation set starts to overfit so we choose epoch value as 17 and retrain the model with new hyperparamter (epoch = 18)

On Test set :
 Accuracy : 0.765900
 Loss : 0.843699

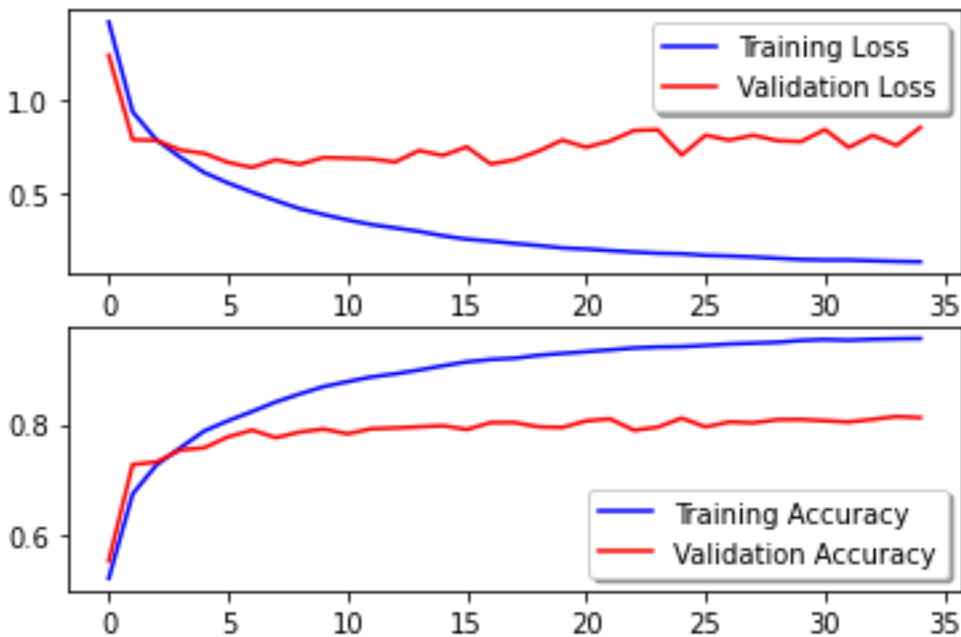
D: Adding Inception Block to our network block:

Model: "model_4"

Layer (type) to	Output Shape	Param #	Connected
=====			
input_4 (InputLayer)	(None, 32, 32, 3)	0	
conv2d_22 (Conv2D) input_4[0][0]	(None, 32, 32, 32)	896	
max_pooling2d_10 (MaxPooling2D) conv2d_22[0][0]	(None, 16, 16, 32)	0	

batch_normalization_10 (BatchNo	(None, 16, 16, 32)	128
max_pooling2d_10[0][0]		
conv2d_23 (Conv2D)	(None, 16, 16, 128)	36992
batch_normalization_10[0][0]		
max_pooling2d_11 (MaxPooling2D)	(None, 8, 8, 128)	0
conv2d_23[0][0]		
batch_normalization_11 (BatchNo	(None, 8, 8, 128)	512
max_pooling2d_11[0][0]		
conv2d_24 (Conv2D)	(None, 8, 8, 64)	8256
batch_normalization_11[0][0]		
conv2d_26 (Conv2D)	(None, 8, 8, 64)	8256
batch_normalization_11[0][0]		
max_pooling2d_12 (MaxPooling2D)	(None, 8, 8, 128)	0
batch_normalization_11[0][0]		
conv2d_25 (Conv2D)	(None, 8, 8, 64)	36928
conv2d_24[0][0]		
conv2d_27 (Conv2D)	(None, 8, 8, 64)	102464
conv2d_26[0][0]		
conv2d_28 (Conv2D)	(None, 8, 8, 64)	8256
max_pooling2d_12[0][0]		
concatenate_4 (Concatenate)	(None, 8, 8, 192)	0
conv2d_25[0][0]		
conv2d_27[0][0]		
conv2d_28[0][0]		
dropout_7 (Dropout)	(None, 8, 8, 192)	0
concatenate_4[0][0]		
flatten_4 (Flatten)	(None, 12288)	0
dropout_7[0][0]		

dense_7 (Dense) flatten_4[0][0]	(None, 512)	6291968
batch_normalization_12 (Batch Normalization) dense_7[0][0]	(None, 512)	2048
dropout_8 (Dropout) batch_normalization_12[0][0]	(None, 512)	0
dense_8 (Dense) dropout_8[0][0]	(None, 10)	5130
=====		
=====		
Total params: 6,501,834		
Trainable params: 6,500,490		
Non-trainable params: 1,344		



As you can after 13 epoch validation set starts to overfit so we choose epoch value as 13 and retrain the model with new hyperparamter (epoch = 13)

- On Test set :
Accuracy : 0.7936999797821045

Loss : 0.6680989986181259

E. Adding ResNet block to our network:

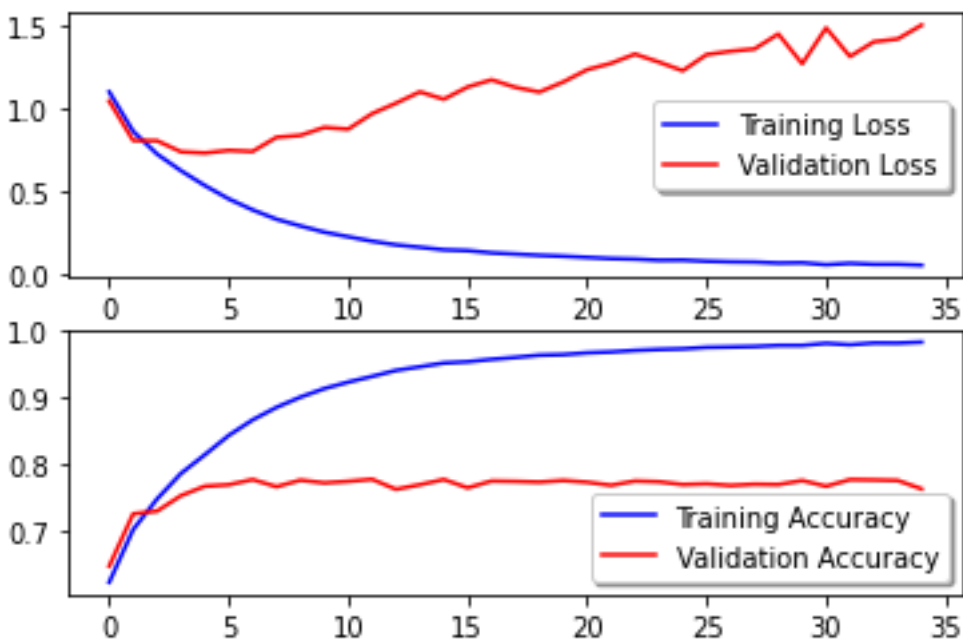
Model Architecture :

Model: "model_2"

Layer (type) to	Output Shape	Param #	Connected
=====			
input_2 (InputLayer)	(None, 32, 32, 3)	0	
<hr/>			
conv2d_7 (Conv2D) input_2[0][0]	(None, 32, 32, 32)	896	
<hr/>			
max_pooling2d_3 (MaxPooling2D) conv2d_7[0][0]	(None, 16, 16, 32)	0	
<hr/>			
batch_normalization_8 (BatchNor max_pooling2d_3[0][0]	(None, 16, 16, 32)	128	
<hr/>			
conv2d_8 (Conv2D) batch_normalization_8[0][0]	(None, 16, 16, 128)	36992	
<hr/>			
max_pooling2d_4 (MaxPooling2D) conv2d_8[0][0]	(None, 8, 8, 128)	0	
<hr/>			
batch_normalization_9 (BatchNor max_pooling2d_4[0][0]	(None, 8, 8, 128)	512	
<hr/>			
conv2d_9 (Conv2D) batch_normalization_9[0][0]	(None, 4, 4, 128)	147584	
<hr/>			
re_lu_5 (ReLU) conv2d_9[0][0]	(None, 4, 4, 128)	0	
<hr/>			
batch_normalization_10 (BatchNo re_lu_5[0][0]	(None, 4, 4, 128)	512	
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conv2d_10 (Conv2D)	(None, 4, 4, 128)	147584
batch_normalization_10[0][0]		
re_lu_6 (ReLU)	(None, 4, 4, 128)	0
conv2d_10[0][0]		
batch_normalization_11 (BatchNo	(None, 4, 4, 128)	512
re_lu_6[0][0]		
conv2d_11 (Conv2D)	(None, 4, 4, 128)	147584
batch_normalization_11[0][0]		
re_lu_7 (ReLU)	(None, 4, 4, 128)	0
conv2d_11[0][0]		
conv2d_12 (Conv2D)	(None, 4, 4, 128)	16512
batch_normalization_9[0][0]		
batch_normalization_12 (BatchNo	(None, 4, 4, 128)	512
re_lu_7[0][0]		
add_2 (Add)	(None, 4, 4, 128)	0
conv2d_12[0][0]		
batch_normalization_12[0][0]		
re_lu_8 (ReLU)	(None, 4, 4, 128)	0
add_2[0][0]		
batch_normalization_13 (BatchNo	(None, 4, 4, 128)	512
re_lu_8[0][0]		
dropout_3 (Dropout)	(None, 4, 4, 128)	0
batch_normalization_13[0][0]		
flatten_2 (Flatten)	(None, 2048)	0
dropout_3[0][0]		
dense_3 (Dense)	(None, 512)	1049088
flatten_2[0][0]		

batch_normalization_14 (BatchNo	(None, 512)	2048
dense_3[0][0]		
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dropout_4 (Dropout)	(None, 512)	0
batch_normalization_14[0][0]		
<hr/>		
dense_4 (Dense)	(None, 10)	5130
dropout_4[0][0]		
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=====		
Total params: 1,556,106		
Trainable params: 1,553,738		
Non-trainable params: 2,368		



As you can after 6 epoch validation set starts to overfit so we choose epoch value as 6 and retrain the model with new hyperparamter (epoch = 6)

On Test set :
 Accuracy : 0.7583000063896179
 Loss : 0.7259090369224548

Conclusion :
 Adding Inception block to our network helped us increase the accuracy. So adding Inception block is useful.