Multiclass classification:

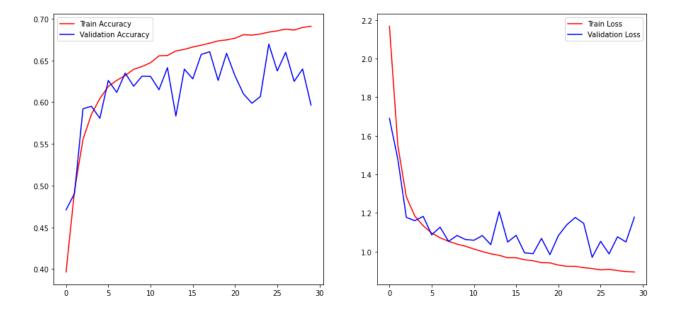
- (a) Download the CIFAR-10 and load the pickled data into your program.
- (b) Build a basic convolutional neural networks with several convolution, pooling, and normalization layers. Flatten the output of the convolution layers and pass it to a single dense layer that will produce the output using softmax activation.
- (c) Test the performance of the model you built and tune hyper parameters as needed.
- (d) Add one or two inception blocks and test performance.
- (e) Remove the inception blocks and add one or two residual blocks instead. Test performance and compare to previous results.

1. Hyperparameter tuning:

i. Initial model with 2 convolution layers, max pooling layers, Adding dropout rate 0.2: Model summary:

Layer (type)	Output	Shape	Param #
input_3 (InputLayer)	(None,	32, 32, 3)	0
conv2d_7 (Conv2D)	(None,	32, 32, 64)	256
leaky_re_lu_6 (LeakyReLU)	(None,	32, 32, 64)	0
max_pooling2d_6 (MaxPooling2	(None,	16, 16, 64)	0
dropout_7 (Dropout)	(None,	16, 16, 64)	0
batch_normalization_6 (Batch	(None,	16, 16, 64)	256
conv2d_8 (Conv2D)	(None,	16, 16, 128)	73856
leaky_re_lu_7 (LeakyReLU)	(None,	16, 16, 128)	0
max_pooling2d_7 (MaxPooling2	(None,	8, 8, 128)	0
dropout_8 (Dropout)	(None,	8, 8, 128)	0
<pre>batch_normalization_7 (Batch</pre>	(None,	8, 8, 128)	512
flatten_2 (Flatten)	(None,	8192)	0
dropout_9 (Dropout)	(None,	8192)	0
dense_2 (Dense)	(None,	10)	81930

Total params: 156,810 Trainable params: 156,426 Non-trainable params: 384



2. Adding inception blocks:

```
inception1 = Conv2D(32, kernel_size=1, strides=1, input_shape=self.img_shape, padding
= 'same')(img)
  hid1 = LeakyReLU()(inception1)
  hid1 = MaxPooling2D((2,2))(hid1)
  hid1 = Dropout(0.2)(hid1)
  hid1 = BatchNormalization()(hid1)
  print(hid1)
  inception2 = Conv2D(32, kernel_size=3, strides=1, input_shape=self.img_shape,
padding = 'same')(img)
  hid2 = LeakyReLU()(inception2)
  hid2 = MaxPooling2D((2,2))(hid2)
  hid2 = Dropout(0.2)(hid2)
  hid2 = BatchNormalization()(hid2)
  print(hid2)
  inception3 = Conv2D(32, kernel size=5, strides=1, input shape=self.img shape,
padding = 'same')(img)
  hid3 = LeakyReLU()(inception3)
  hid3 = MaxPooling2D((2,2))(hid3)
  hid3 = Dropout(0.2)(hid3)
  hid3 = BatchNormalization()(hid3)
  print(hid3)
```

Model: "model_3"

Layer (type)	Output	Shaj	pe ====		Param # =======	Connected to
<pre>input_4 (InputLayer)</pre>	(None,	32,	32,	3)	0	
conv2d_9 (Conv2D)	(None,	32,	32,	32)	128	input_4[0][0
conv2d_10 (Conv2D)	(None,	32,	32,	32)	896	input_4[0][0
conv2d_11 (Conv2D)	(None,	32,	32,	32)	2432	input_4[0][0
leaky_re_lu_8 (LeakyReLU)	(None,	32,	32,	32)	0	conv2d_9[0][
leaky_re_lu_9 (LeakyReLU) [0]	(None,	32,	32,	32)	0	conv2d_10[0]
leaky_re_lu_10 (LeakyReLU) [0]	(None,	32,	32,	32)	0	conv2d_11[0]
max_pooling2d_8 (MaxPooling2D) 8[0][0]	(None,	16,	16,	32)	0	leaky_re_lu_
max_pooling2d_9 (MaxPooling2D) 9[0][0]	(None,	16,	16,	32)	0	leaky_re_lu_
max_pooling2d_10 (MaxPooling2D) 10[0][0]	(None,	16,	16,	32)	0	leaky_re_lu_
dropout_10 (Dropout) d_8[0][0]	(None,	16,	16,	32)	0	max_pooling2
dropout_11 (Dropout) d_9[0][0]	(None,	16,	16,	32)	0	max_pooling2
dropout_12 (Dropout) d_10[0][0]	(None,	16,	16,	32)	0	max_pooling2

<pre>batch_normalization_8 (BatchNor][0]</pre>	(None,	16, 16, 32)	128	dropout_10[0
batch_normalization_9 (BatchNor][0]	(None,	16, 16, 32)	128	dropout_11[0
batch_normalization_10 (BatchNo][0]	(None,	16, 16, 32)	128	dropout_12[0
concatenate_2 (Concatenate) ization_8[0][0]	(None,	16, 16, 96)	0	batch_normal
<pre>ization_9[0][0] ization_10[0][0]</pre>				batch_normal
conv2d_12 (Conv2D) 2[0][0]	(None,	16, 16, 64)	6208	concatenate_
leaky_re_lu_11 (LeakyReLU) [0]	(None,	16, 16, 64)	0	conv2d_12[0]
max_pooling2d_11 (MaxPooling2D) 11[0][0]	(None,	8, 8, 64)	0	leaky_re_lu_
dropout_13 (Dropout) d_11[0][0]	(None,	8, 8, 64)	0	max_pooling2
batch_normalization_11 (BatchNo][0]	(None,	8, 8, 64)	256	dropout_13[0
conv2d_13 (Conv2D) ization_11[0][0]	(None,	8, 8, 128)	73856	batch_normal
leaky_re_lu_12 (LeakyReLU) [0]	(None,	8, 8, 128)	0	conv2d_13[0]
max_pooling2d_12 (MaxPooling2D) 12[0][0]	(None,	4, 4, 128)	0	leaky_re_lu_
dropout_14 (Dropout) d_12[0][0]	(None,	4, 4, 128)	0	max_pooling2

<pre>batch_normalization_12][0]</pre>	(BatchNo	(None,	4, 4, 128)	512	dropout_14[0
flatten_3 (Flatten) ization_12[0][0]		(None,	2048)	0	batch_normal
dropout_15 (Dropout) [0]		(None,	2048)	0	flatten_3[0]
dense_3 (Dense)][0]		(None,	10)	20490	dropout_15[0

Total params: 105,162
Trainable params: 104,586
Non-trainable params: 576

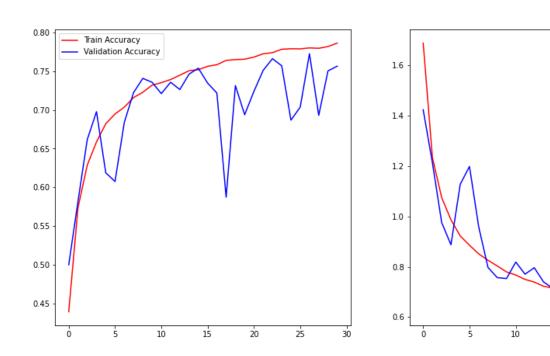
.....

Train Loss

15

20

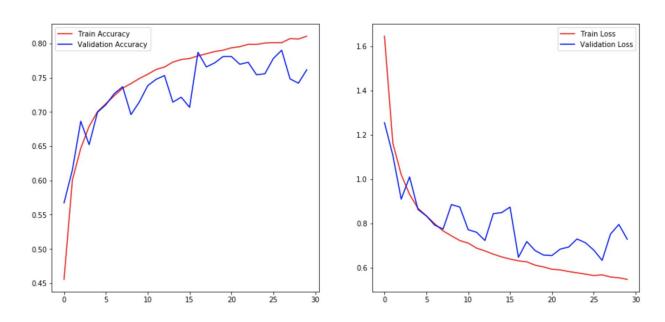
Validation Loss



Adding inception blocks has increased the performance to 75% on the validation set. This is because inception blocks increase the dimensionality of the features extracted by the convolution blocks.

3. Residual Block

Adding residual blocks has a better performance than the inception blocks.



Adding a second residual block

 A second residual block has not improved the performace, and is not necessary for our use case.

