

# model\_training

May 1, 2019

## 1 Follow-Me Project

Congratulations on reaching the final project of the Robotics Nanodegree!

Previously, you worked on the Semantic Segmentation lab where you built a deep learning network that locates a particular human target within an image. For this project, you will utilize what you implemented and learned from that lab and extend it to train a deep learning model that will allow a simulated quadcopter to follow around the person that it detects!

Most of the code below is similar to the lab with some minor modifications. You can start with your existing solution, and modify and improve upon it to train the best possible model for this task.

You can click on any of the following to quickly jump to that part of this notebook: 1. Section ?? 2. Section ?? 3. Section ?? 4. Section [1.4](#) 5. Section [1.5](#) 6. Section [1.6](#)

### 1.1 Data Collection

We have provided you with a starting dataset for this project. Download instructions can be found in the README for this project's repo. Alternatively, you can collect additional data of your own to improve your model. Check out the "Collecting Data" section in the Project Lesson in the Classroom for more details!

```
In [36]: import os
import glob
import sys
import tensorflow as tf

from scipy import misc
import numpy as np

from tensorflow.contrib.keras.python import keras
from tensorflow.contrib.keras.python.keras import layers, models

from tensorflow import image

from utils import scoring_utils
from utils.separable_conv2d import SeparableConv2DKeras, BilinearUpSampling2D
from utils import data_iterator
from utils import plotting_tools
from utils import model_tools
```

## 1.2 FCN Layers

In the Classroom, we discussed the different layers that constitute a fully convolutional network (FCN). The following code will introduce you to the functions that you need to build your semantic segmentation model.

### 1.2.1 Separable Convolutions

The Encoder for your FCN will essentially require separable convolution layers, due to their advantages as explained in the classroom. The 1x1 convolution layer in the FCN, however, is a regular convolution. Implementations for both are provided below for your use. Each includes batch normalization with the ReLU activation function applied to the layers.

```
In [2]: def separable_conv2d_batchnorm(input_layer, filters, strides=1):
        output_layer = SeparableConv2DKeras(filters=filters, kernel_size=3, strides=strides,
                                             padding='same', activation='relu')(input_layer)

        output_layer = layers.BatchNormalization()(output_layer)
        return output_layer

    def conv2d_batchnorm(input_layer, filters, kernel_size=3, strides=1):
        output_layer = layers.Conv2D(filters=filters, kernel_size=kernel_size, strides=strides,
                                     padding='same', activation='relu')(input_layer)

        output_layer = layers.BatchNormalization()(output_layer)
        return output_layer
```

### 1.2.2 Bilinear Upsampling

The following helper function implements the bilinear upsampling layer. Upsampling by a factor of 2 is generally recommended, but you can try out different factors as well. Upsampling is used in the decoder block of the FCN.

```
In [3]: def bilinear_upsample(input_layer):
        output_layer = BilinearUpSampling2D((2,2))(input_layer)
        return output_layer
```

## 1.3 Build the Model

In the following cells, you will build an FCN to train a model to detect and locate the hero target within an image. The steps are: - Create an `encoder_block` - Create a `decoder_block` - Build the FCN consisting of encoder block(s), a 1x1 convolution, and decoder block(s). This step requires experimentation with different numbers of layers and filter sizes to build your model.

### 1.3.1 Encoder Block

Create an encoder block that includes a separable convolution layer using the `separable_conv2d_batchnorm()` function. The `filters` parameter defines the size or depth of the output layer. For example, 32 or 64.

```
In [115]: def encoder_block(input_layer, filters, strides=1):

    # TODO Create a separable convolution layer using the separable_conv2d_batchnorm()
    output_layer = separable_conv2d_batchnorm(input_layer, filters=filters, strides=strides)
    return output_layer

image_hw = 160
image_shape = (image_hw, image_hw, 3)
inputs = layers.Input(image_shape)
print(inputs.get_shape())
test_out_layer = encoder_block(inputs,32,)
print(test_out_layer.get_shape())

(?, 160, 160, 3)
(?, 23, 23, 32)
```

### 1.3.2 Decoder Block

The decoder block is comprised of three parts: - A bilinear upsampling layer using the `upsample_bilinear()` function. The current recommended factor for upsampling is set to 2. - A layer concatenation step. This step is similar to skip connections. You will concatenate the upsampled `small_ip_layer` and the `large_ip_layer`. - Some (one or two) additional separable convolution layers to extract some more spatial information from prior layers.

```
In [5]: def decoder_block(small_ip_layer, large_ip_layer, filters):

    # TODO Upsample the small input layer using the bilinear_upsample() function.
    Upsampled_small_ip_layer = bilinear_upsample(small_ip_layer)
    # TODO Concatenate the upsampled and large input layers using layers.concatenate
    concatenated_layers = layers.concatenate([Upsampled_small_ip_layer, large_ip_layer])
    # TODO Add some number of separable convolution layers
    output_layer = encoder_block(concatenated_layers, filters=filters, strides=1)
    return output_layer
```

### 1.3.3 Model

Now that you have the encoder and decoder blocks ready, go ahead and build your FCN architecture!

There are three steps: - Add encoder blocks to build the encoder layers. This is similar to how you added regular convolutional layers in your CNN lab. - Add a 1x1 Convolution layer using the `conv2d_batchnorm()` function. Remember that 1x1 Convolutions require a kernel and stride of 1. - Add decoder blocks for the decoder layers.

```
In [92]: def fcn_model(inputs, num_classes):

    # TODO Add Encoder Blocks.
    # Remember that with each encoder layer, the depth of your model (the number of filters)
    print(inputs.get_shape())
```

```

#Layer 1
layer_1 = encoder_block(inputs, filters=32, strides=3)
print(layer_1.get_shape())
#Layer 2
layer_2 = encoder_block(layer_1, filters=64, strides=2)
print(layer_2.get_shape())
layer_3 = encoder_block(layer_2, filters=128, strides=2)
print(layer_3.get_shape())
layer_4 = encoder_block(layer_3, filters=256, strides=2)
print(layer_4.get_shape())
# TODO Add 1x1 Convolution layer using conv2d_batchnorm().
layer_1x1 = conv2d_batchnorm(layer_4, filters=1024, kernel_size=1, strides=1)
print(layer_1x1.get_shape())
# TODO: Add the same number of Decoder Blocks as the number of Encoder Blocks
decoder_1 = decoder_block(layer_1x1, layer_3, filters=128)
print(decoder_1.get_shape())
decoder_2 = decoder_block(decoder_1, layer_2, filters=64)
print(decoder_2.get_shape())
decoder_3 = decoder_block(decoder_2, layer_1, filters=32)
print(decoder_3.get_shape())
x = decoder_block(decoder_3, inputs, filters=num_classes)
print(x.get_shape())
# The function returns the output layer of your model. "x" is the final layer obtained
return layers.Conv2D(num_classes, 3, activation='softmax', padding='same')(x)

```

## 1.4 Training

The following cells will use the FCN you created and define an output layer based on the size of the processed image and the number of classes recognized. You will define the hyperparameters to compile and train your model.

Please Note: For this project, the helper code in `data_iterator.py` will resize the copter images to 160x160x3 to speed up training.

```

In [116]: """
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""

image_hw = 160
image_shape = (image_hw, image_hw, 3)
inputs = layers.Input(image_shape)
num_classes = 3

# Call fcn_model()
output_layer = fcn_model(inputs, num_classes)
#print(output_layer.get_shape())

```

```
(?, 160, 160, 3)
```

```
(?, 80, 80, 32)
```

```
(?, 40, 40, 64)
(?, 20, 20, 128)
(?, 10, 10, 256)
(?, 10, 10, 1024)
(?, 20, 20, 128)
(?, 40, 40, 64)
(?, 80, 80, 32)
(?, 160, 160, 3)
```

### 1.4.1 Hyperparameters

Define and tune your hyperparameters. - **batch\_size**: number of training samples/images that get propagated through the network in a single pass. - **num\_epochs**: number of times the entire training dataset gets propagated through the network. - **steps\_per\_epoch**: number of batches of training images that go through the network in 1 epoch. We have provided you with a default value. One recommended value to try would be based on the total number of images in training dataset divided by the batch\_size. - **validation\_steps**: number of batches of validation images that go through the network in 1 epoch. This is similar to steps\_per\_epoch, except validation\_steps is for the validation dataset. We have provided you with a default value for this as well. - **workers**: maximum number of processes to spin up. This can affect your training speed and is dependent on your hardware. We have provided a recommended value to work with.

```
In [117]: learning_rate = 0.001
          batch_size = 50
          num_epochs = 200
          steps_per_epoch = 30
          validation_steps = 50
          workers = 2
```

```
In [118]: """
          DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
          """

          # Define the Keras model and compile it for training
          model = models.Model(inputs=inputs, outputs=output_layer)

          model.compile(optimizer=keras.optimizers.Adam(learning_rate), loss='categorical_crossentropy')

          # Data iterators for loading the training and validation data
          train_iter = data_iterator.BatchIteratorSimple(batch_size=batch_size,
                                                         data_folder=os.path.join('..', 'data', 'v
                                                         image_shape=image_shape,
                                                         shift_aug=True)

          val_iter = data_iterator.BatchIteratorSimple(batch_size=batch_size,
                                                         data_folder=os.path.join('..', 'data', 'v
                                                         image_shape=image_shape)
```

```

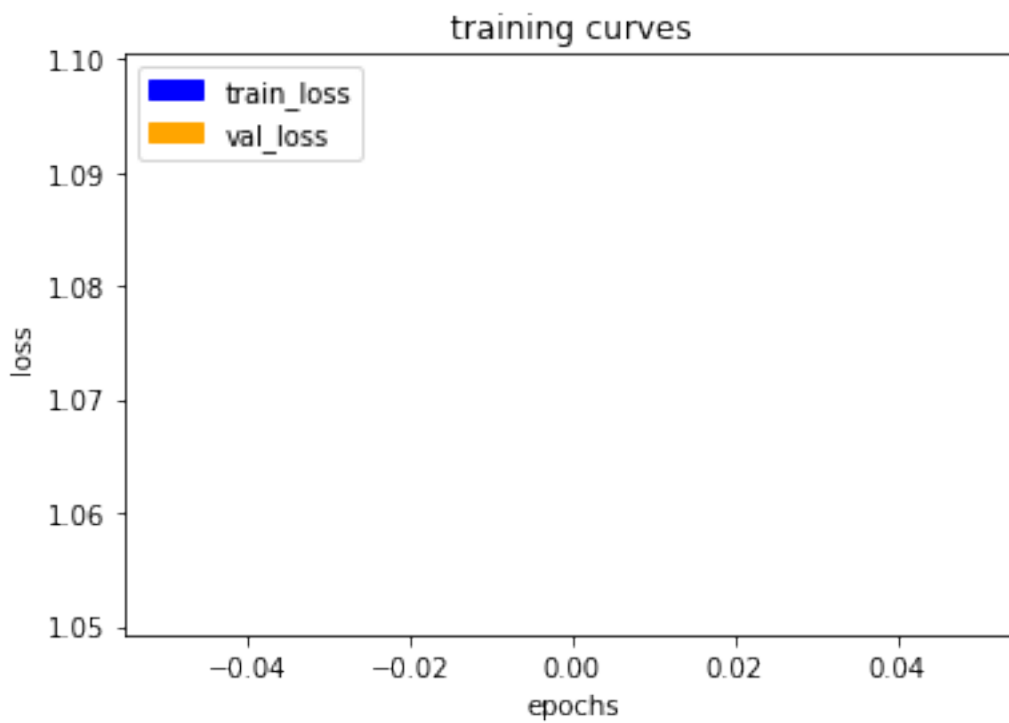
logger_cb = plotting_tools.LoggerPlotter()
callbacks = [logger_cb]

model.fit_generator(train_iter,
                    steps_per_epoch = steps_per_epoch, # the number of batches per epoch
                    epochs = num_epochs, # the number of epochs to train for,
                    validation_data = val_iter, # validation iterator
                    validation_steps = validation_steps, # the number of batches to validate
                    callbacks=callbacks,
                    workers = workers)

```

Epoch 1/200

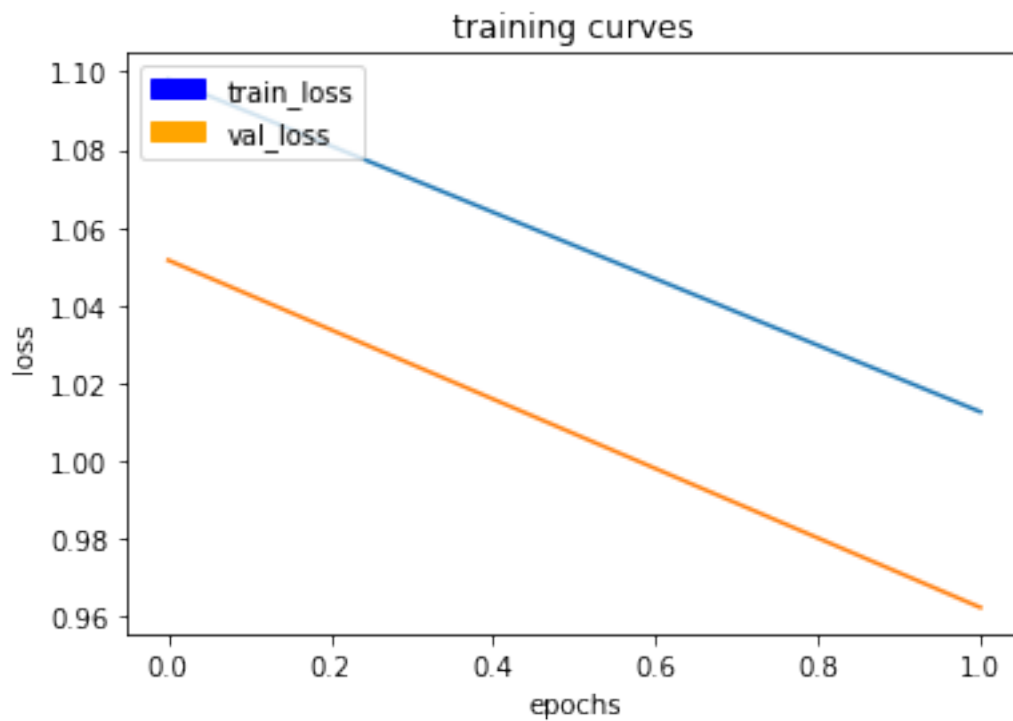
29/30 [=====>.] - ETA: 0s - loss: 1.0997



30/30 [=====] - 26s - loss: 1.0981 - val\_loss: 1.0516

Epoch 2/200

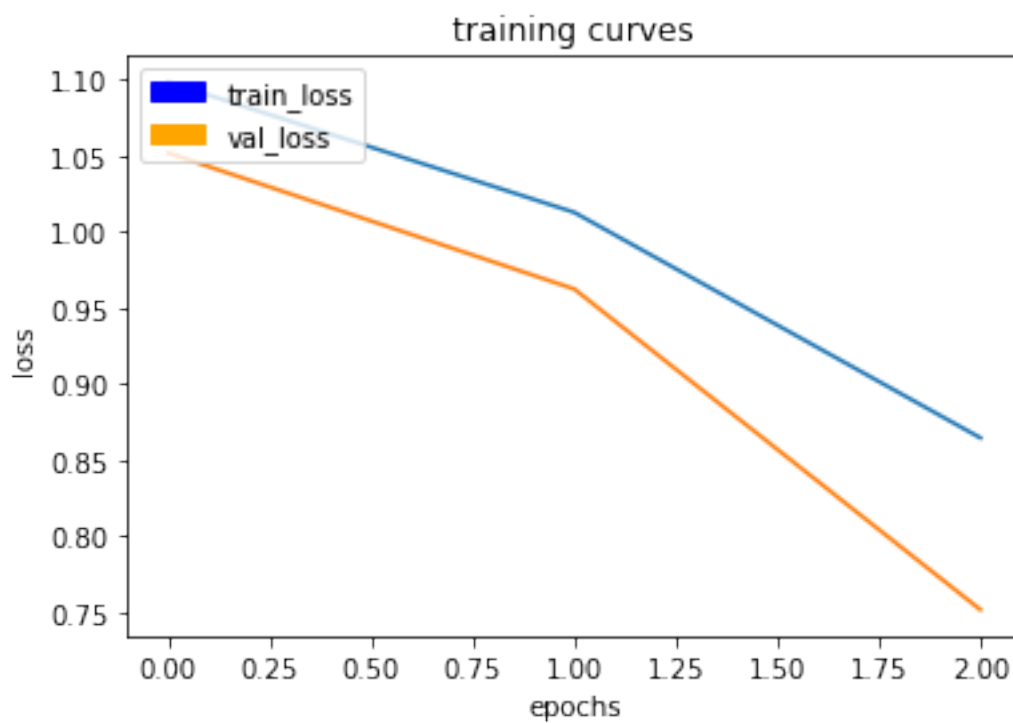
29/30 [=====>.] - ETA: 0s - loss: 1.0143



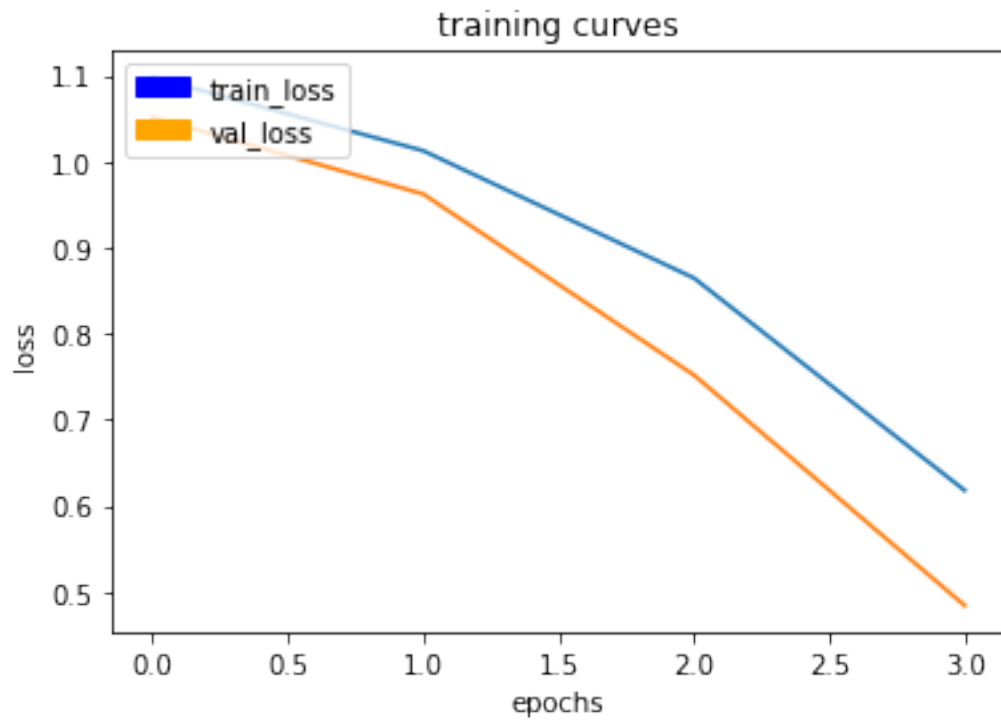
30/30 [=====] - 23s - loss: 1.0126 - val\_loss: 0.9622

Epoch 3/200

29/30 [=====>.] - ETA: 0s - loss: 0.8678

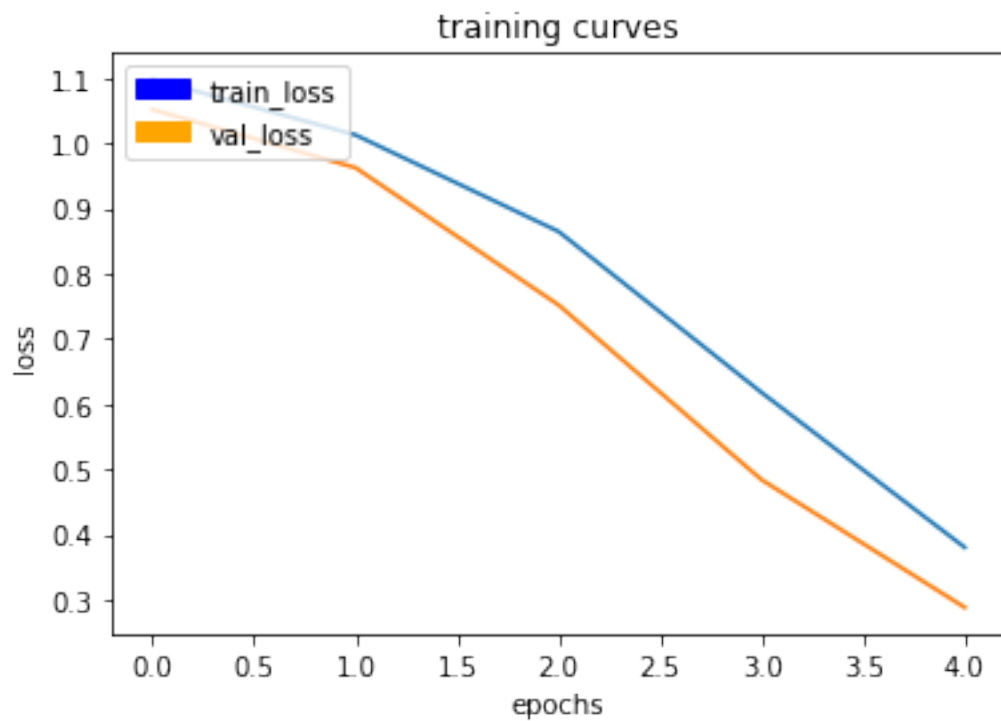


30/30 [=====] - 23s - loss: 0.8638 - val\_loss: 0.7517  
Epoch 4/200  
29/30 [=====>.] - ETA: 0s - loss: 0.6220



30/30 [=====] - 23s - loss: 0.6175 - val\_loss: 0.4837  
Epoch 5/200  
29/30 [=====>.] - ETA: 0s - loss: 0.3834

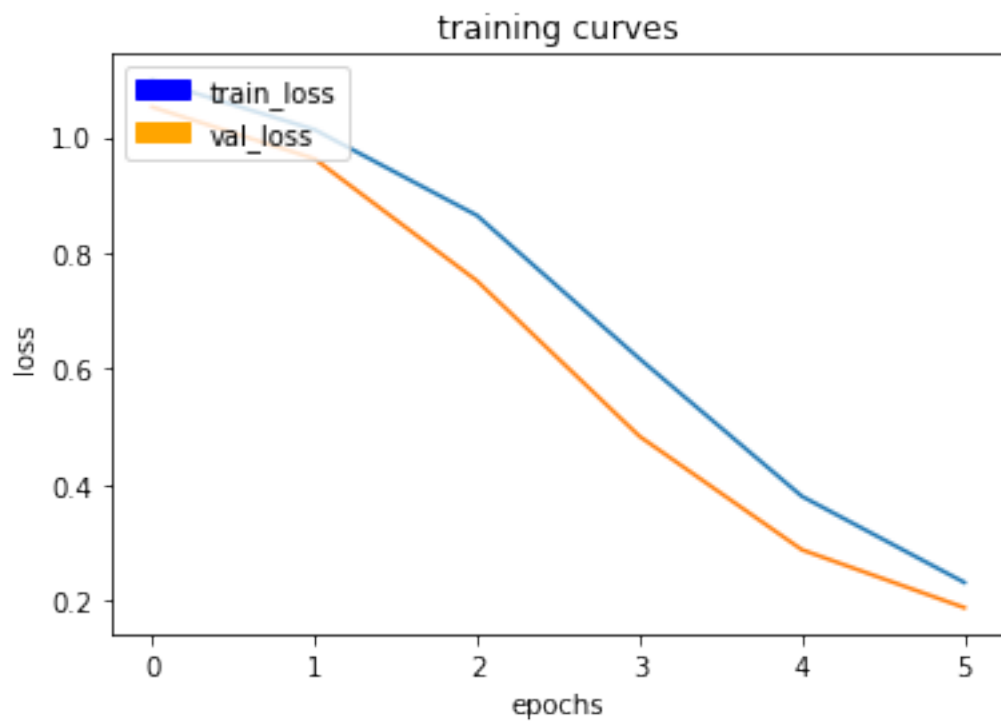




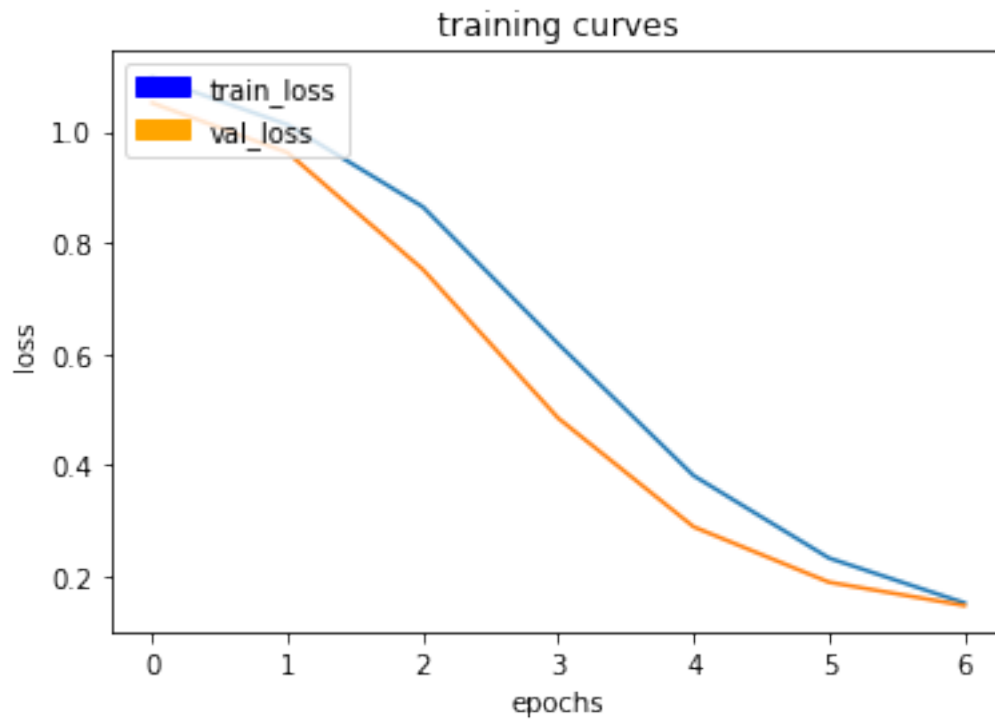
30/30 [=====] - 24s - loss: 0.3802 - val\_loss: 0.2880

Epoch 6/200

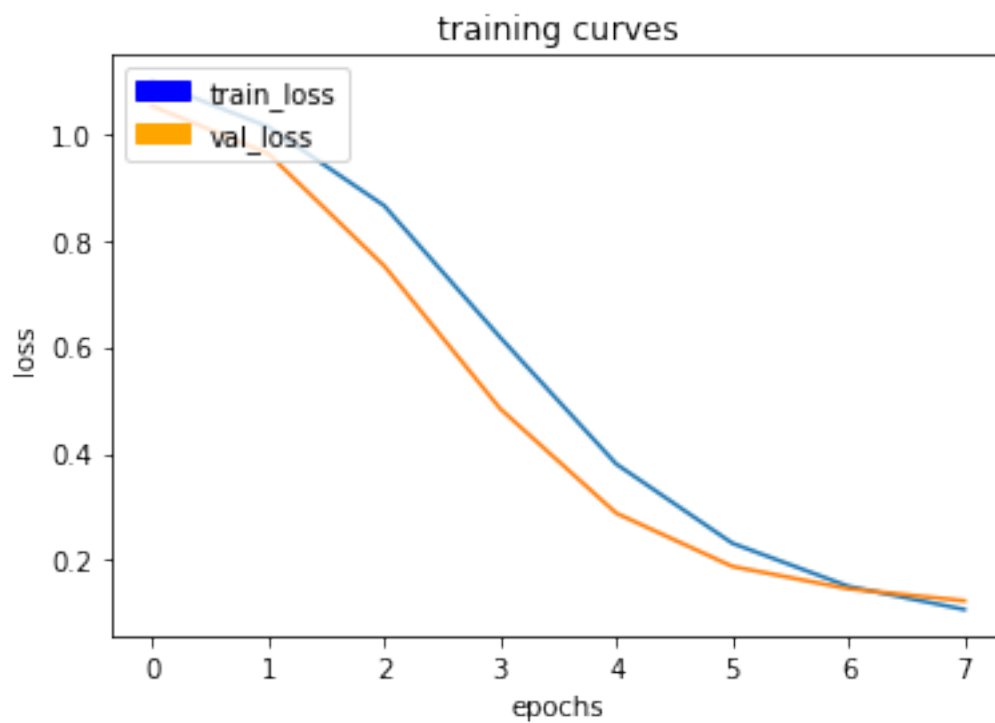
29/30 [=====>.] - ETA: 0s - loss: 0.2332



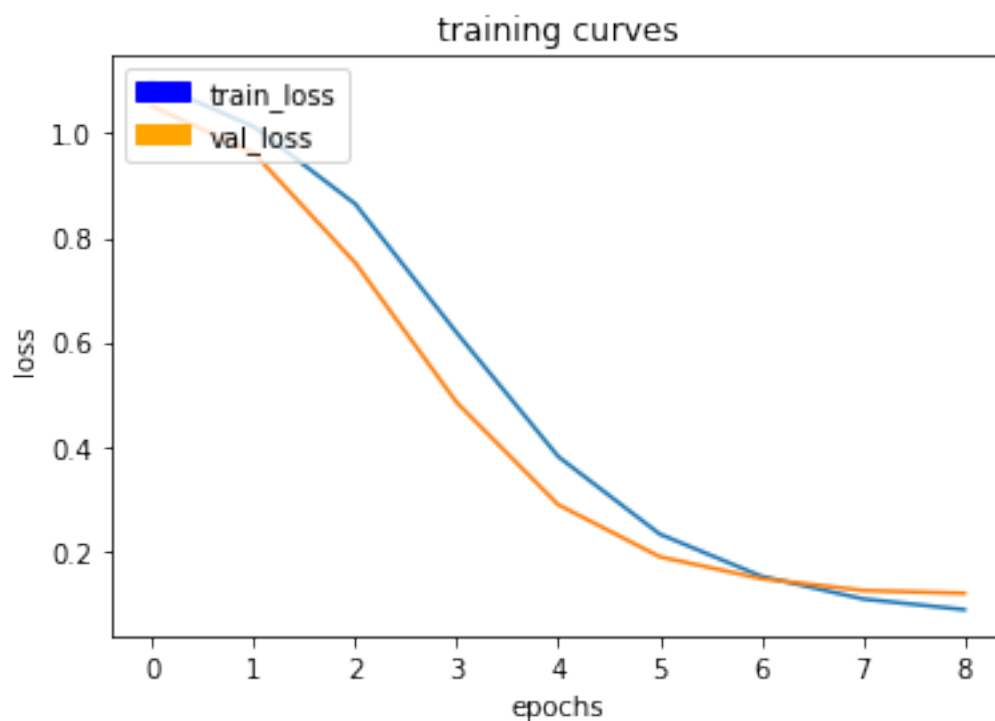
30/30 [=====] - 23s - loss: 0.2316 - val\_loss: 0.1882  
Epoch 7/200  
29/30 [=====>.] - ETA: 0s - loss: 0.1520



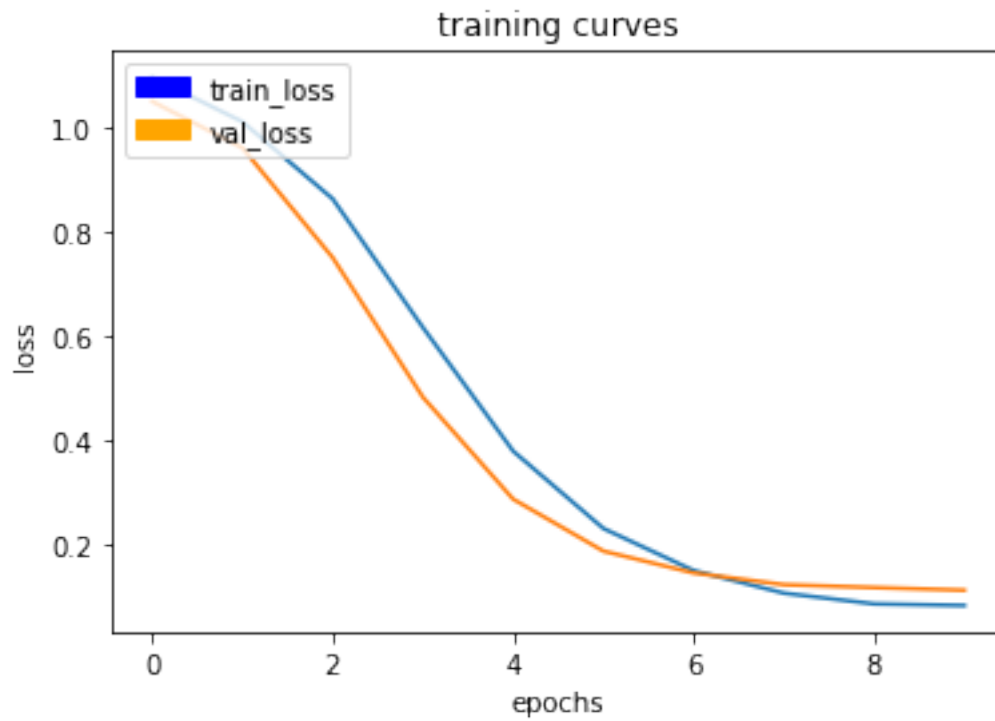
30/30 [=====] - 24s - loss: 0.1509 - val\_loss: 0.1463  
Epoch 8/200  
29/30 [=====>.] - ETA: 0s - loss: 0.1078



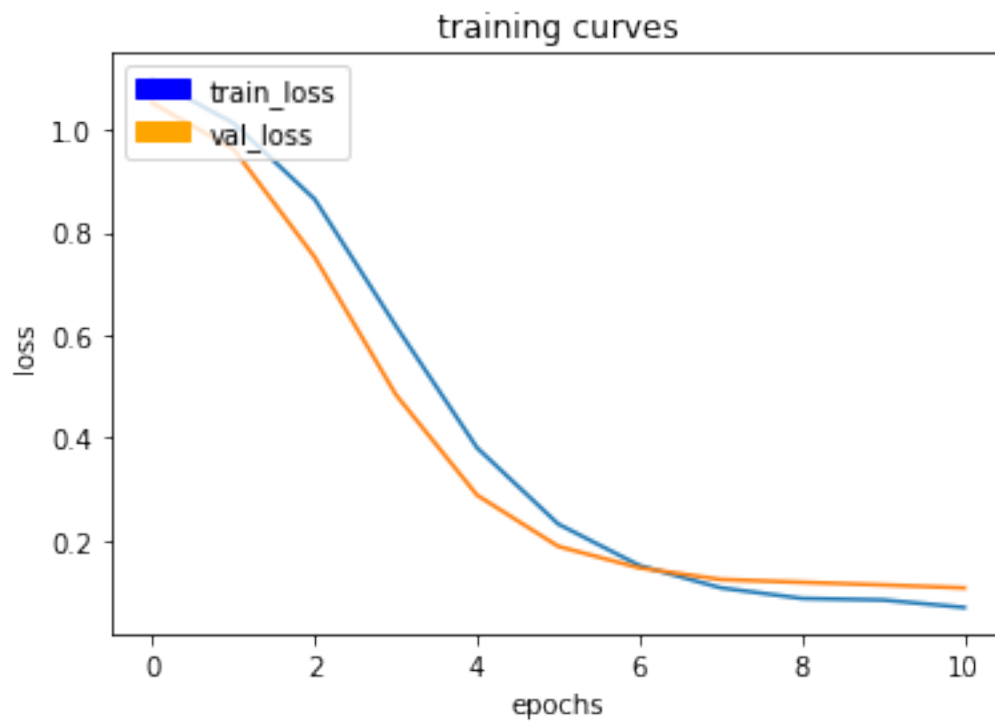
30/30 [======] - 23s - loss: 0.1075 - val\_loss: 0.1238  
 Epoch 9/200  
 29/30 [======>.] - ETA: 0s - loss: 0.0865



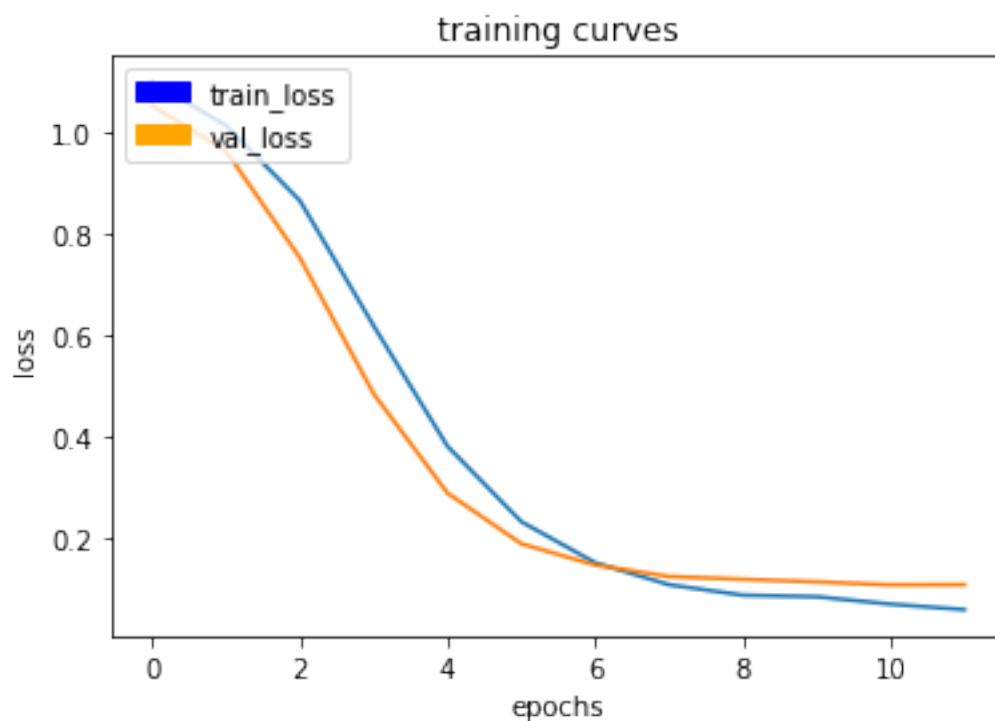
30/30 [=====] - 23s - loss: 0.0868 - val\_loss: 0.1183  
Epoch 10/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0844



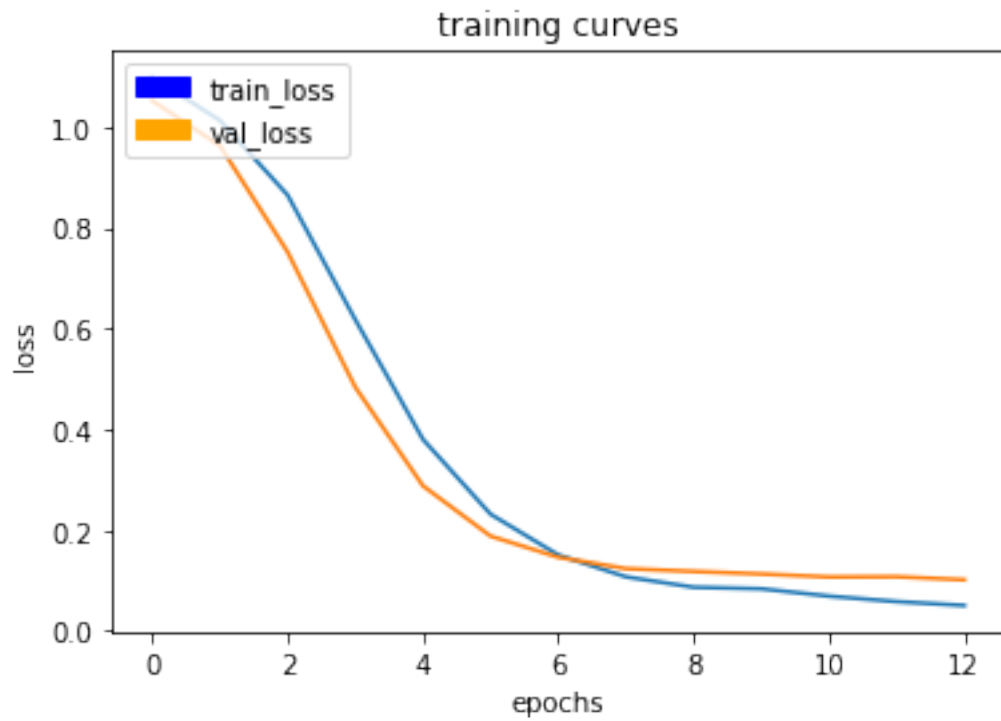
30/30 [=====] - 23s - loss: 0.0839 - val\_loss: 0.1132  
Epoch 11/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0698



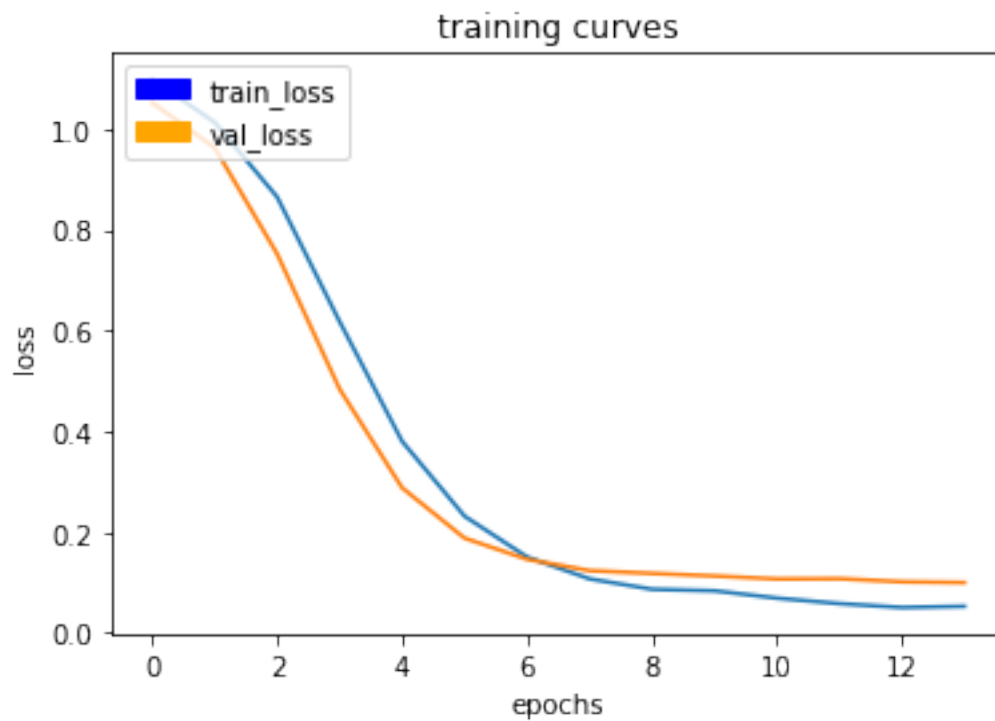
30/30 [=====] - 23s - loss: 0.0693 - val\_loss: 0.1073  
 Epoch 12/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0587



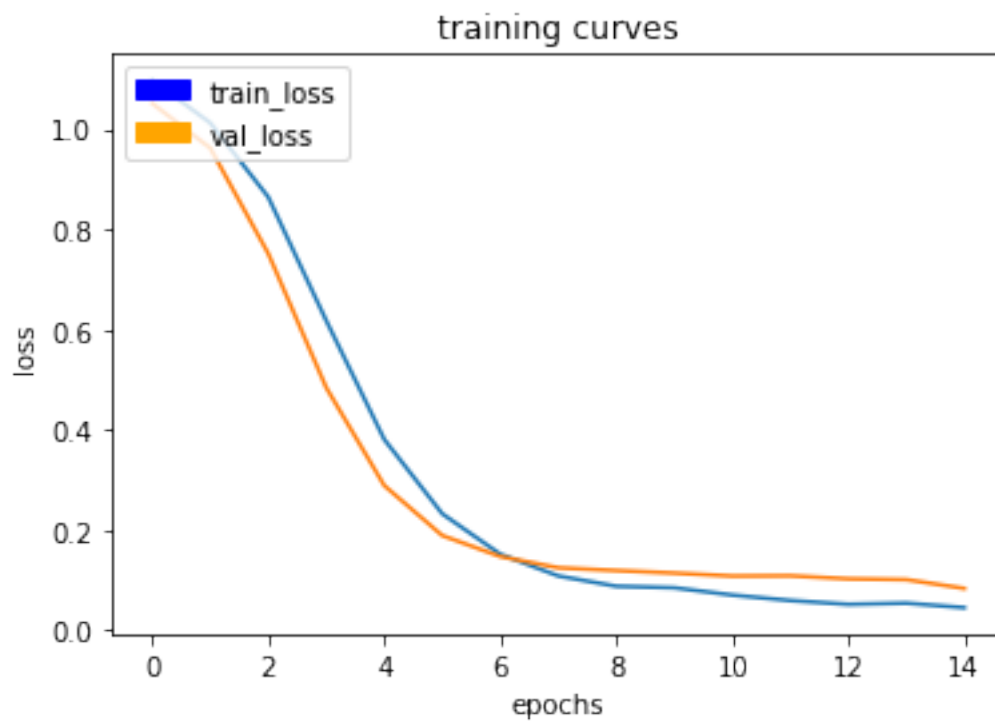
30/30 [=====] - 23s - loss: 0.0584 - val\_loss: 0.1078  
Epoch 13/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0507



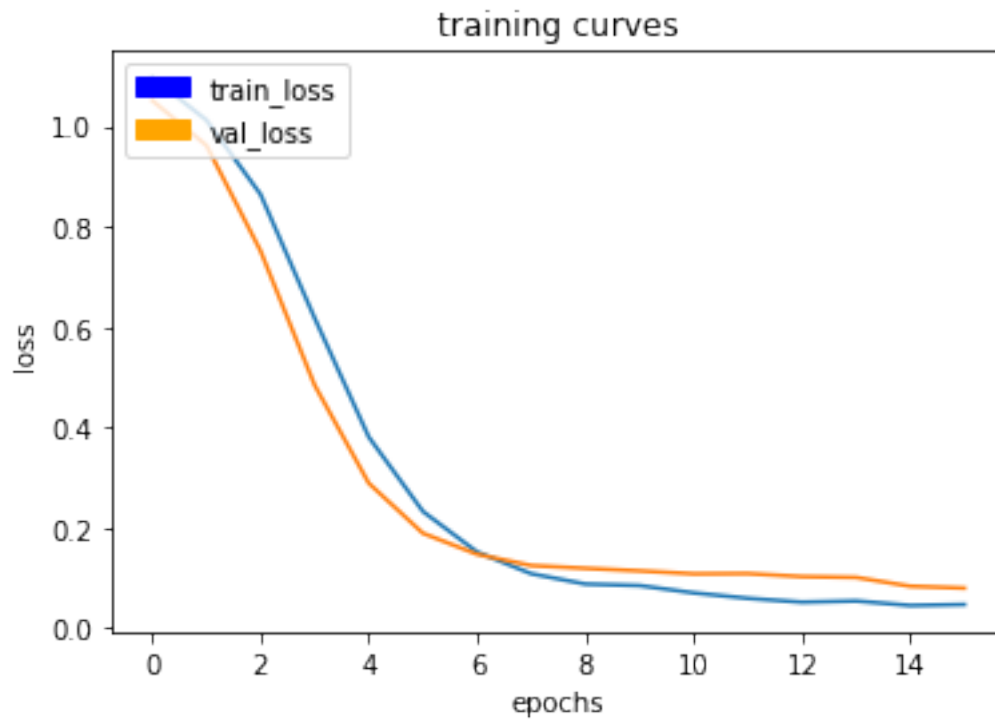
30/30 [=====] - 23s - loss: 0.0508 - val\_loss: 0.1017  
Epoch 14/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0531



30/30 [=====] - 23s - loss: 0.0529 - val\_loss: 0.1001  
 Epoch 15/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0441

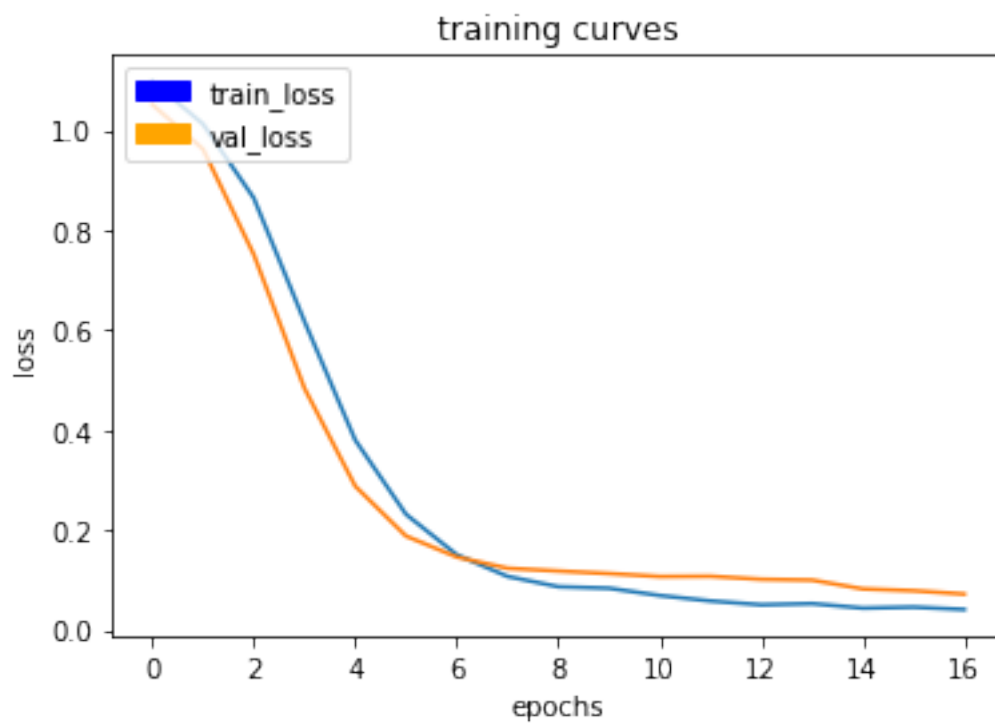


30/30 [=====] - 23s - loss: 0.0441 - val\_loss: 0.0825  
Epoch 16/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0449

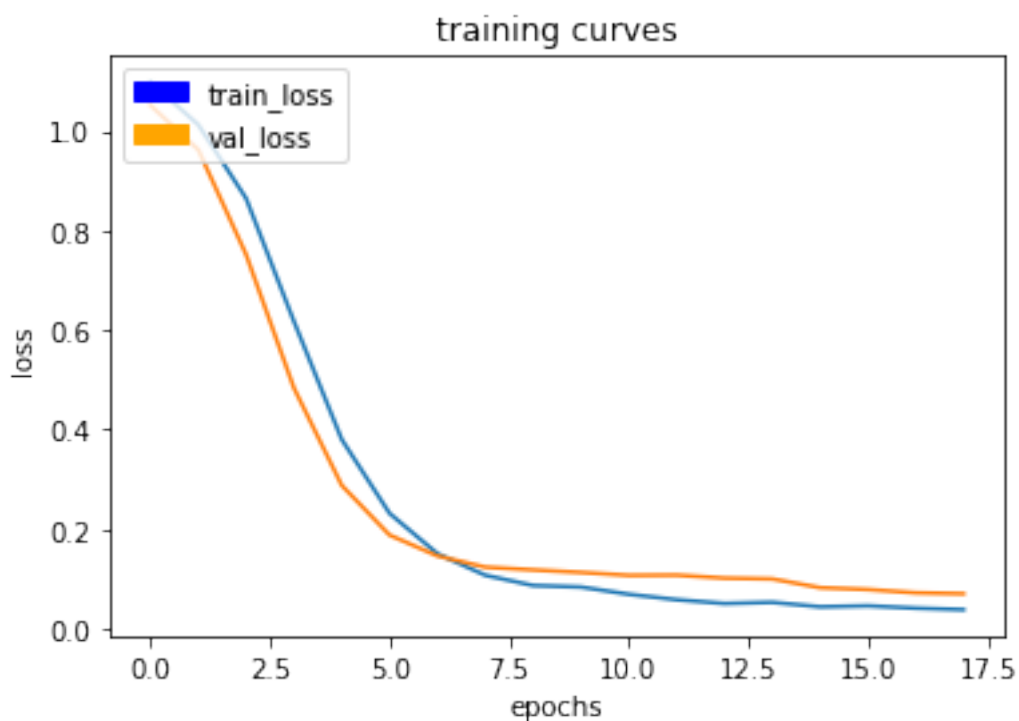


30/30 [=====] - 23s - loss: 0.0460 - val\_loss: 0.0789  
Epoch 17/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0420

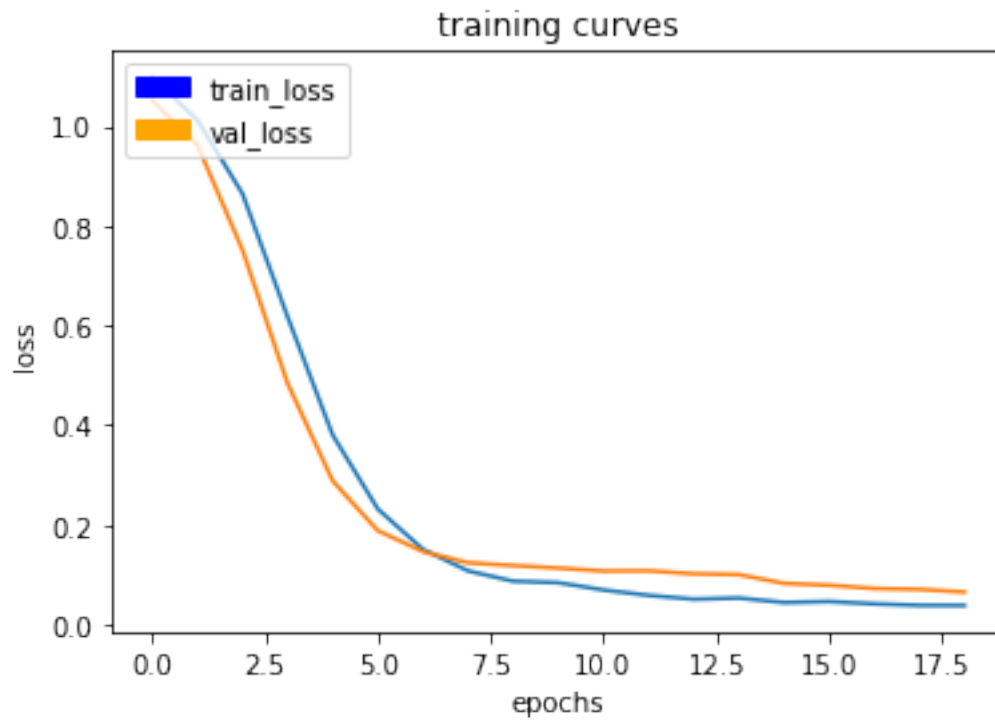




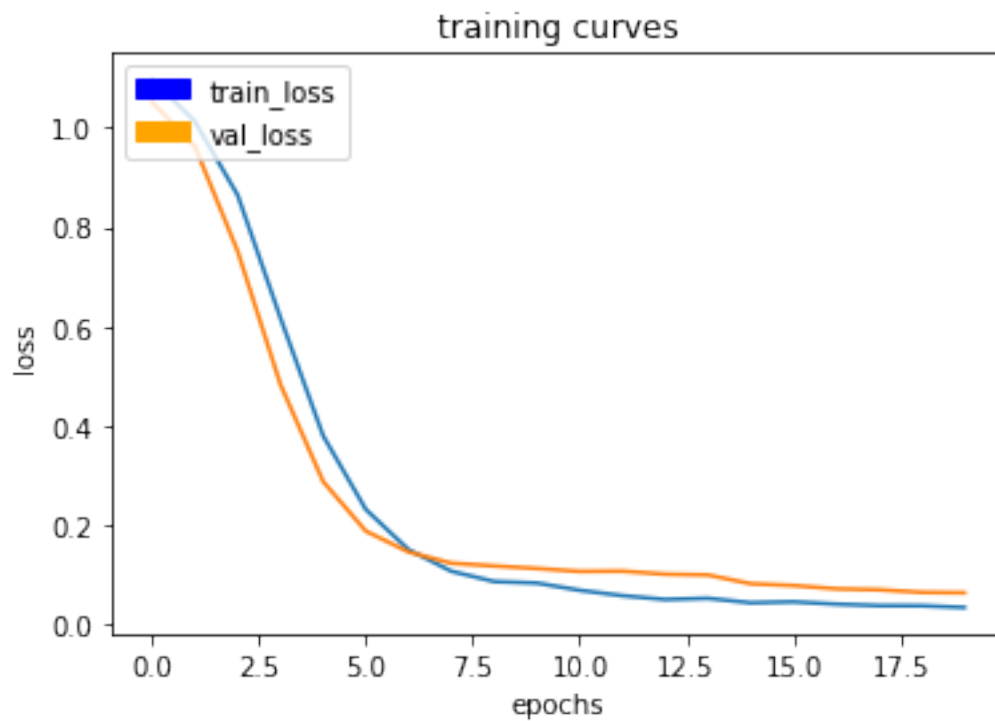
30/30 [=====] - 23s - loss: 0.0419 - val\_loss: 0.0722  
 Epoch 18/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0383



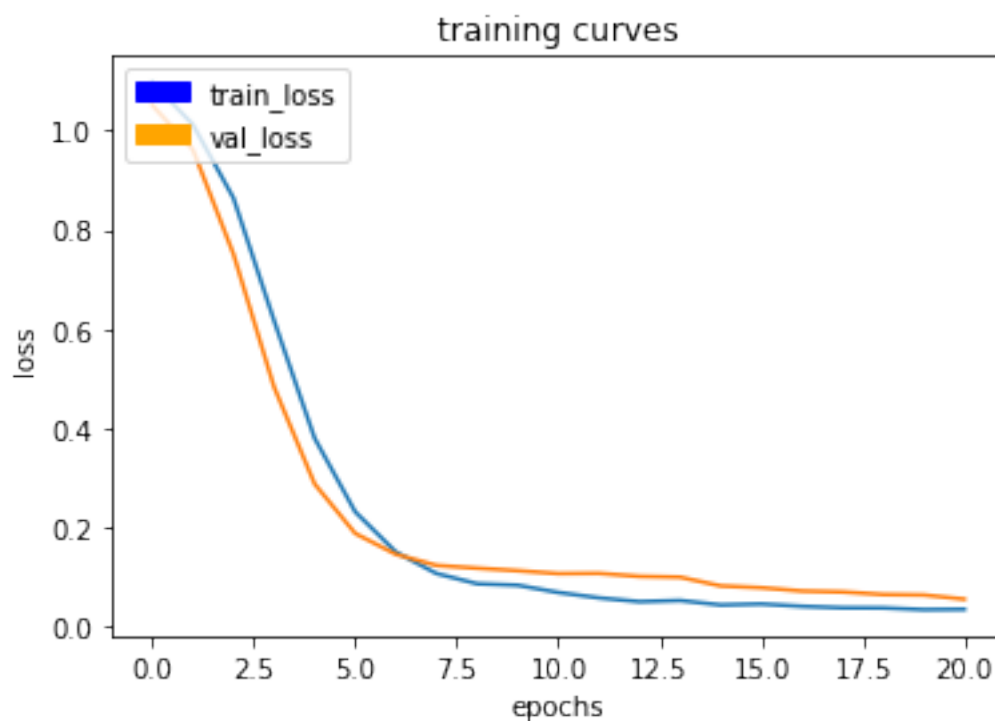
30/30 [=====] - 23s - loss: 0.0384 - val\_loss: 0.0704  
Epoch 19/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0383



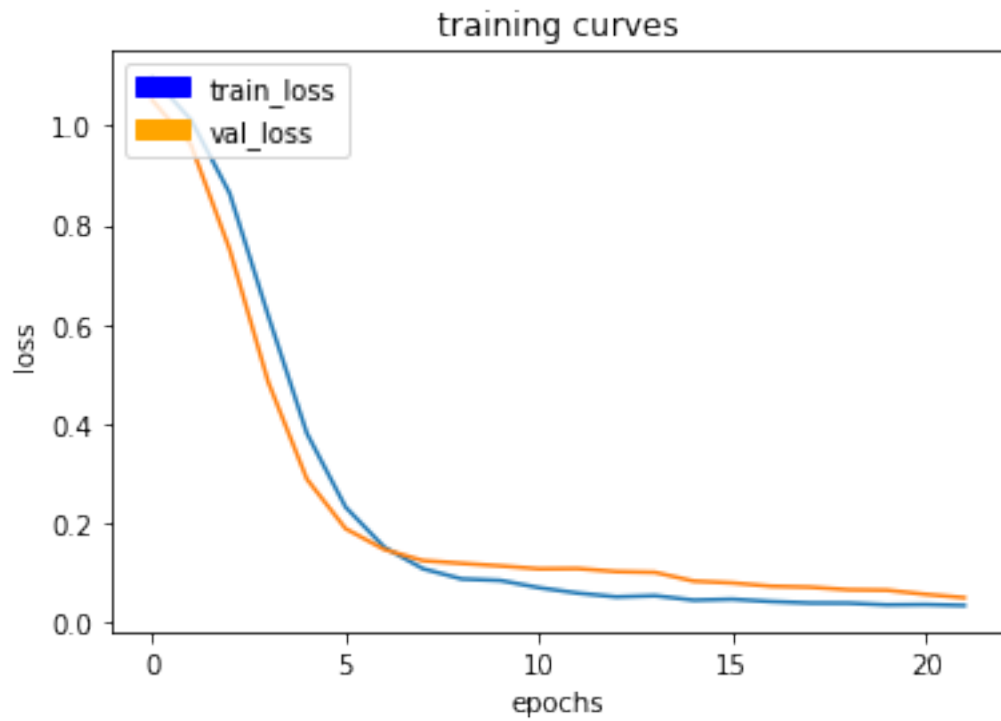
30/30 [=====] - 23s - loss: 0.0381 - val\_loss: 0.0652  
Epoch 20/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0348



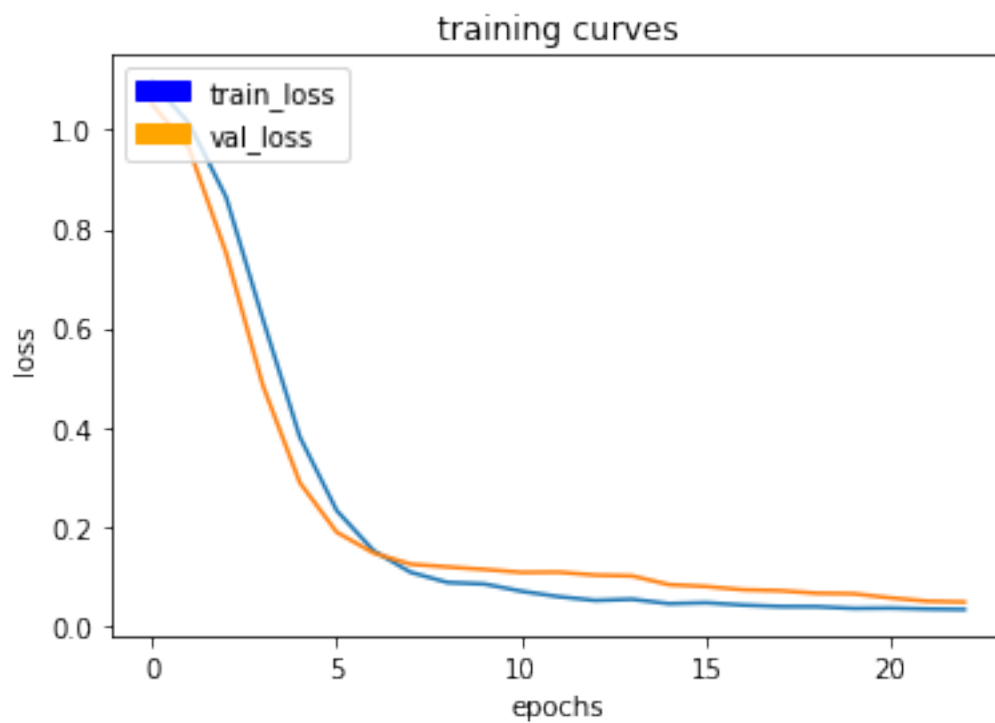
30/30 [=====] - 23s - loss: 0.0347 - val\_loss: 0.0642  
 Epoch 21/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0354



30/30 [=====] - 23s - loss: 0.0353 - val\_loss: 0.0558  
Epoch 22/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0335

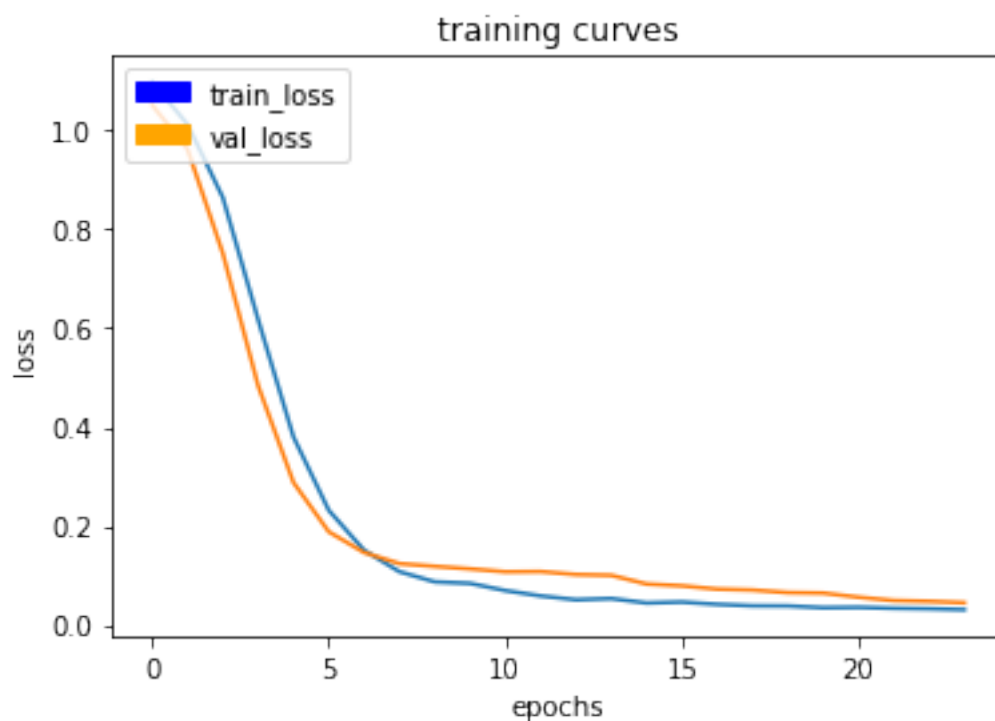


30/30 [=====] - 23s - loss: 0.0334 - val\_loss: 0.0491  
Epoch 23/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0331

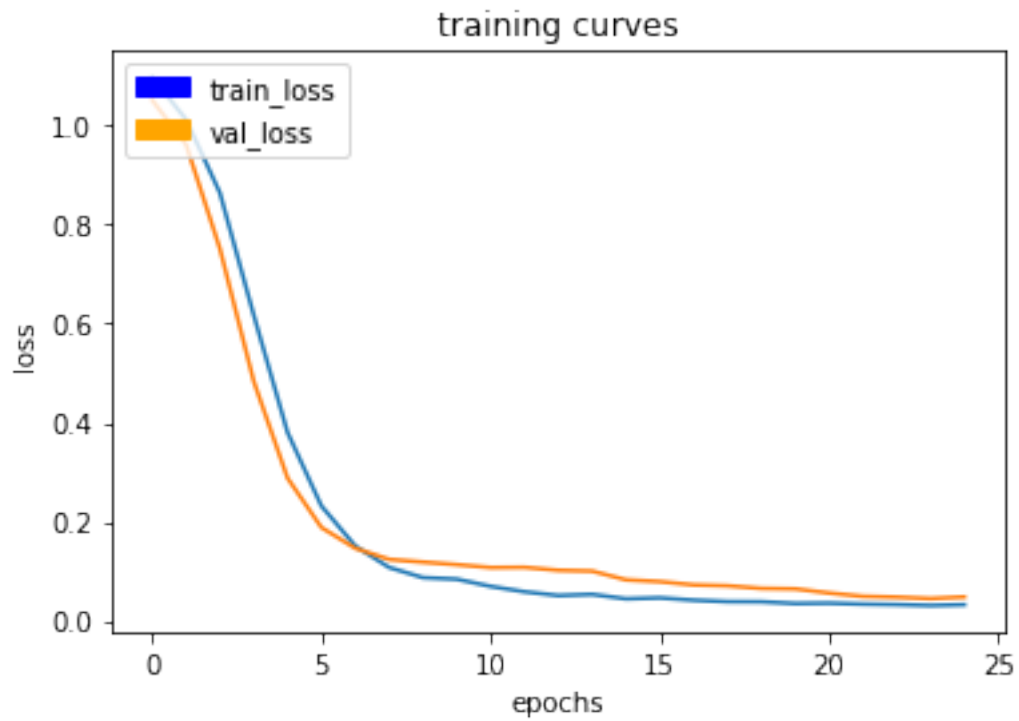


```

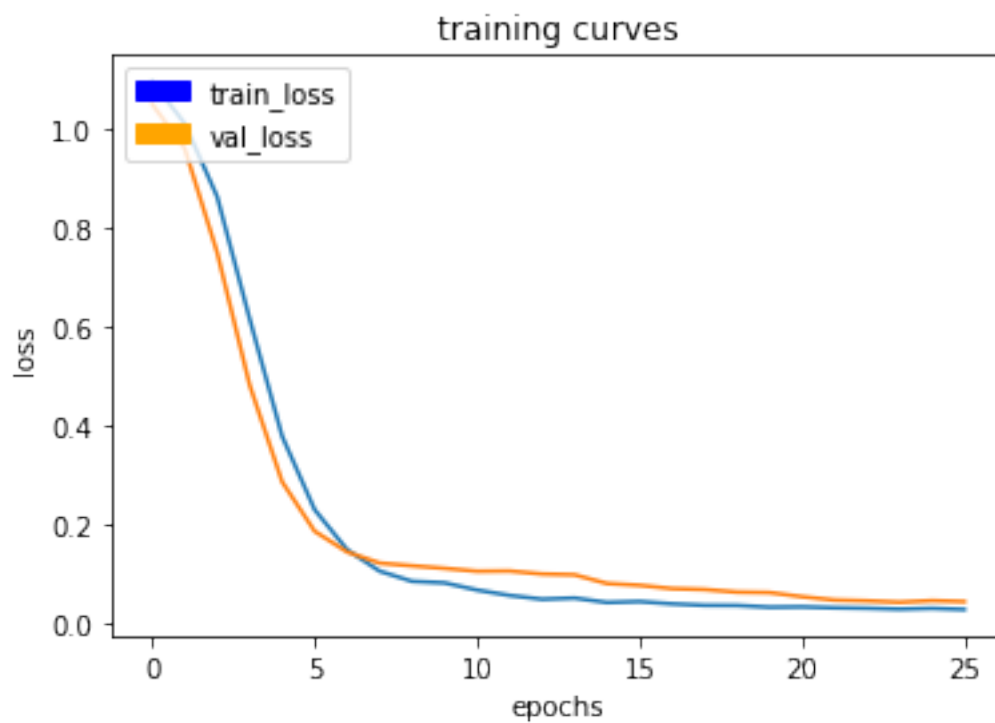
30/30 [=====] - 23s - loss: 0.0329 - val_loss: 0.0475
Epoch 24/200
29/30 [=====>.] - ETA: 0s - loss: 0.0308
  
```



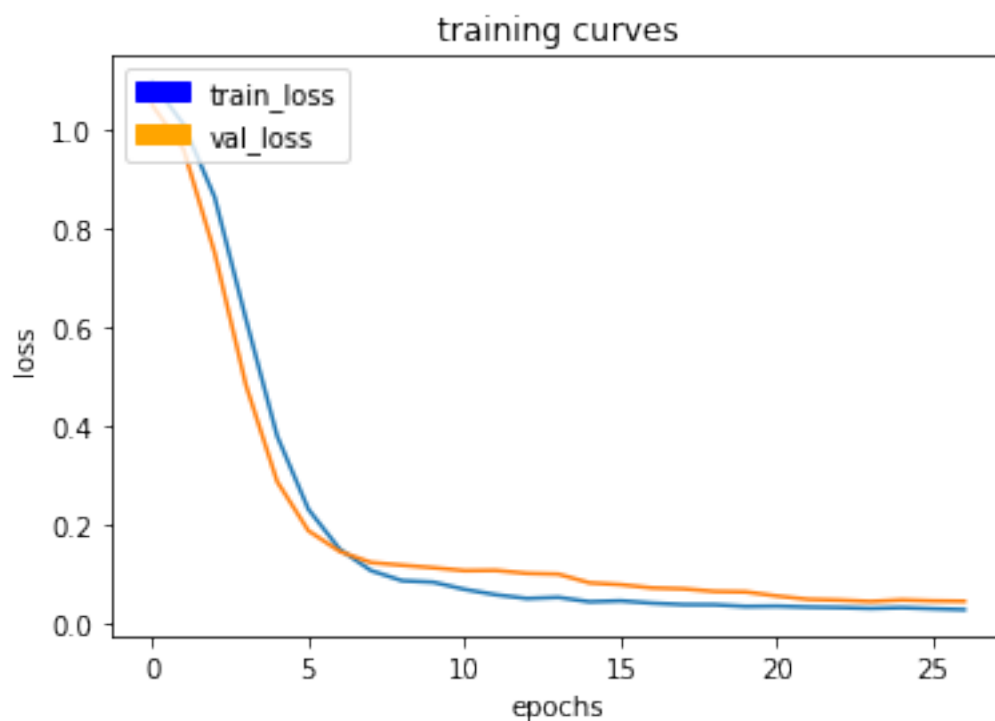
30/30 [=====] - 23s - loss: 0.0307 - val\_loss: 0.0446  
Epoch 25/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0322



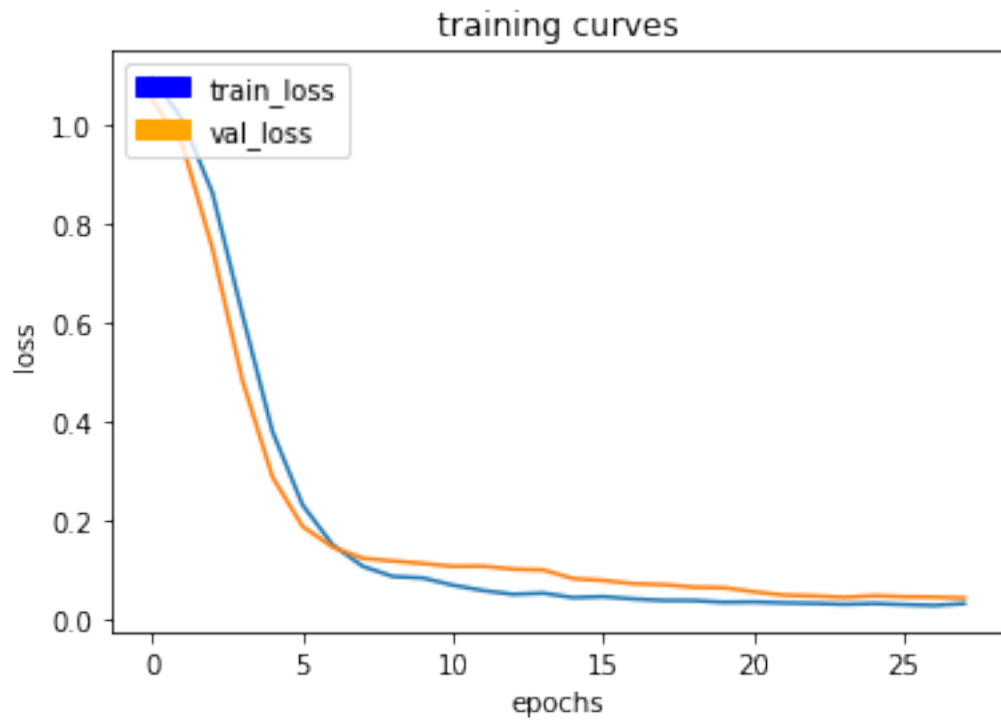
30/30 [=====] - 23s - loss: 0.0321 - val\_loss: 0.0479  
Epoch 26/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0302



30/30 [=====] - 23s - loss: 0.0300 - val\_loss: 0.0459  
 Epoch 27/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0283

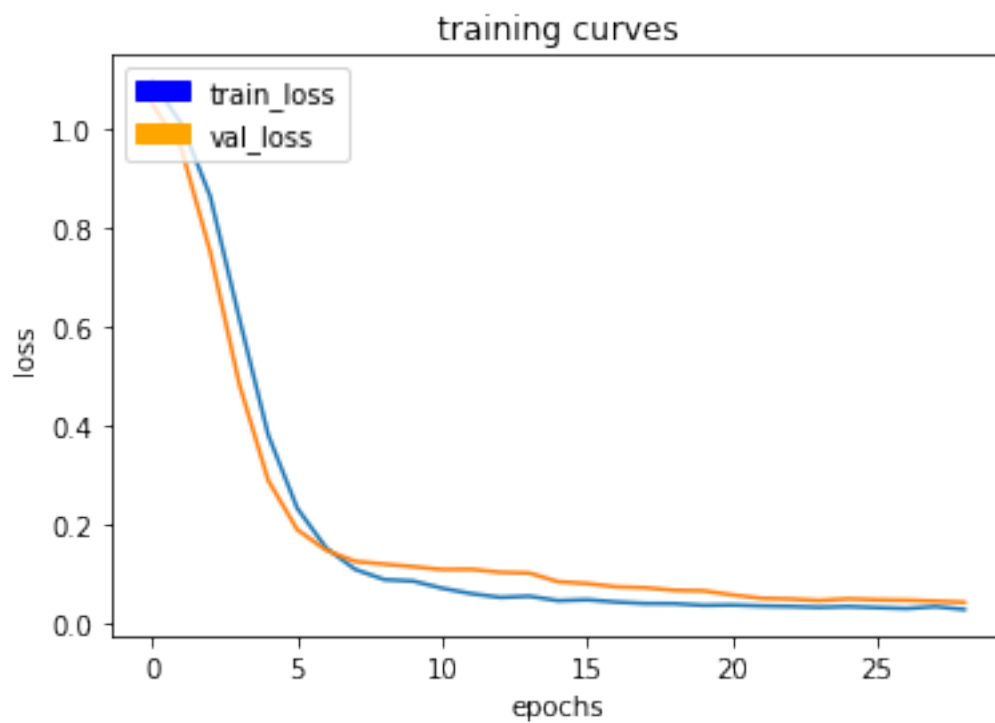


30/30 [=====] - 23s - loss: 0.0282 - val\_loss: 0.0452  
Epoch 28/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0323



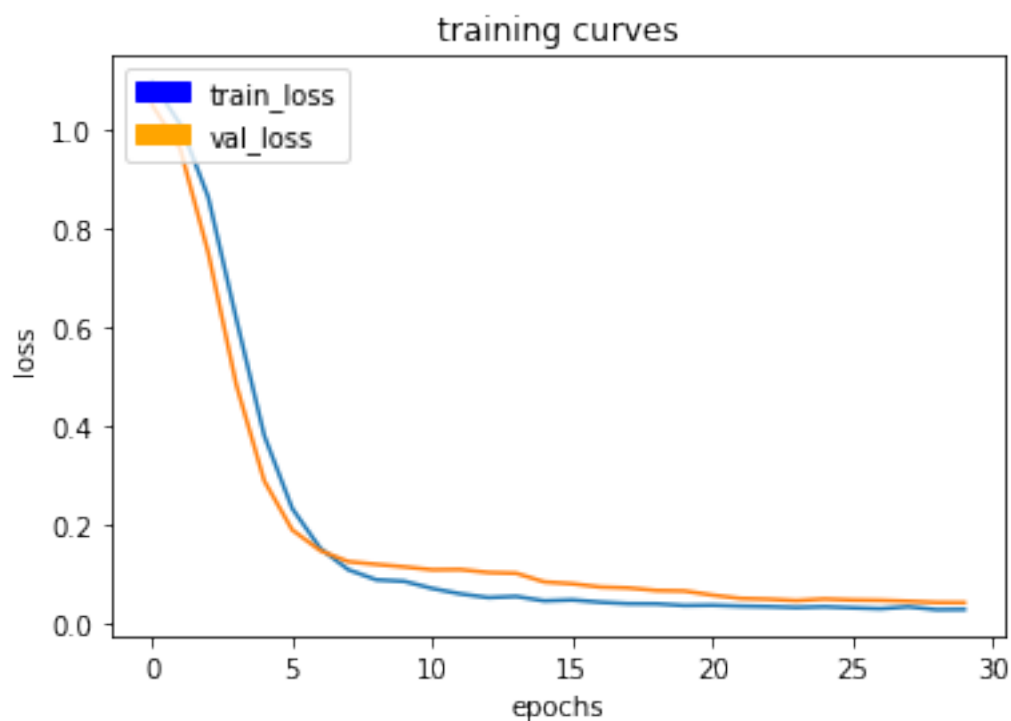
30/30 [=====] - 23s - loss: 0.0321 - val\_loss: 0.0433  
Epoch 29/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0264



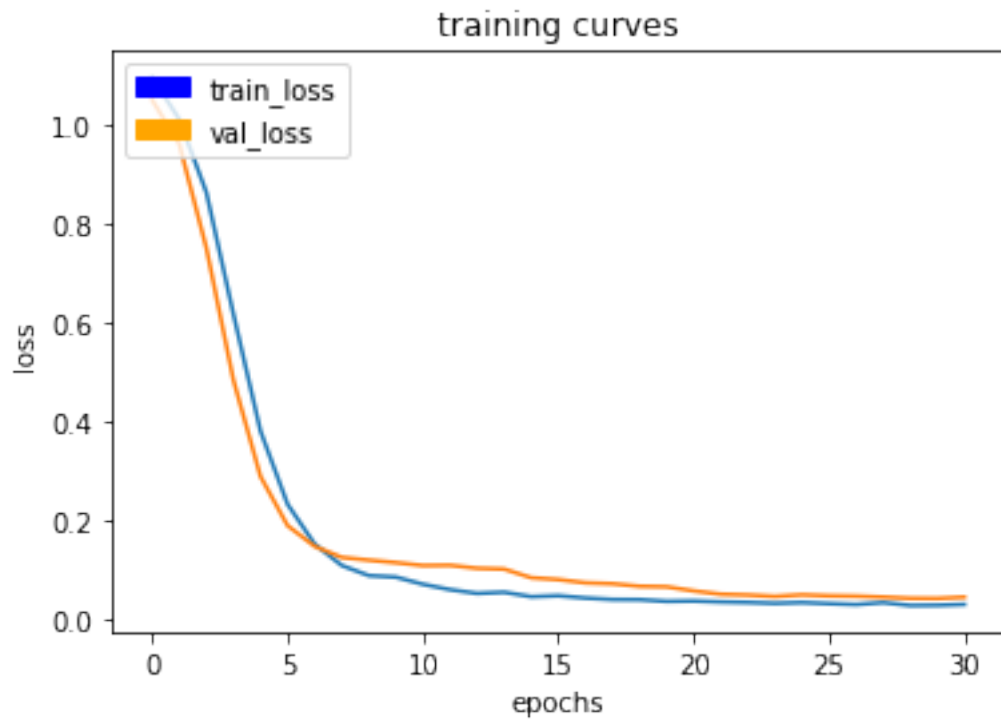


```

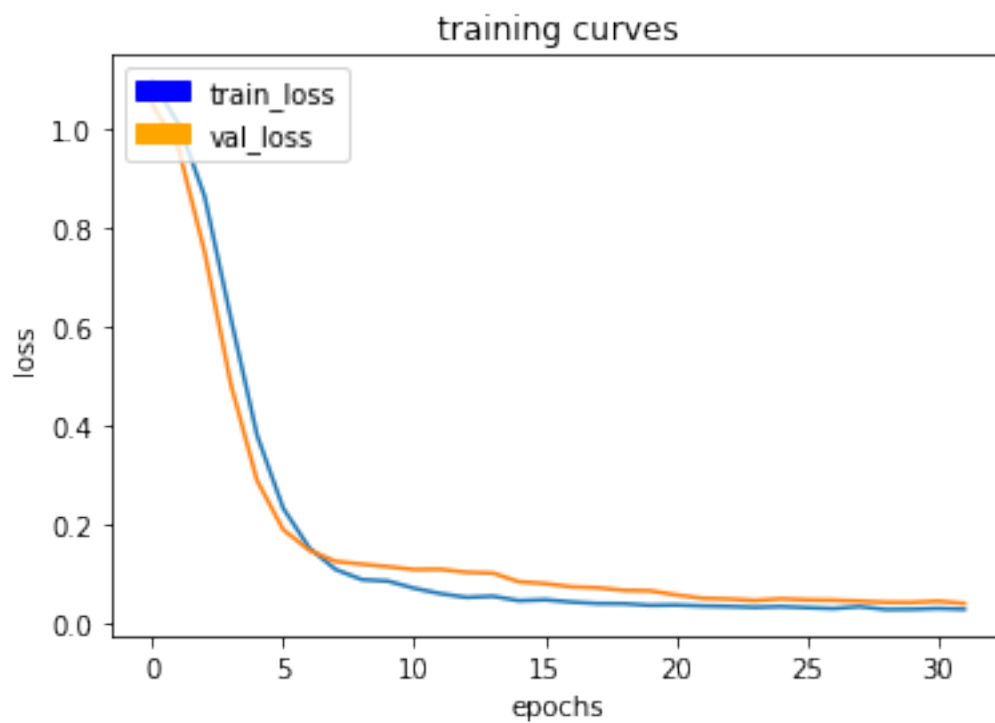
30/30 [=====] - 23s - loss: 0.0263 - val_loss: 0.0408
Epoch 30/200
29/30 [=====>.] - ETA: 0s - loss: 0.0268
  
```



30/30 [=====] - 23s - loss: 0.0266 - val\_loss: 0.0405  
Epoch 31/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0284

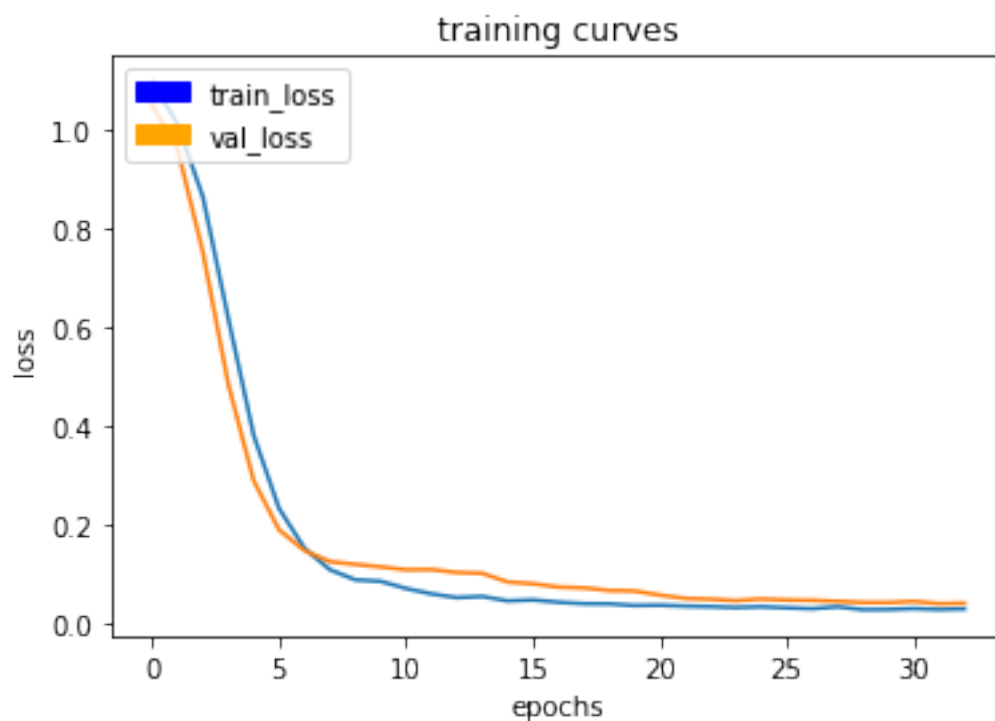


30/30 [=====] - 23s - loss: 0.0283 - val\_loss: 0.0431  
Epoch 32/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0269

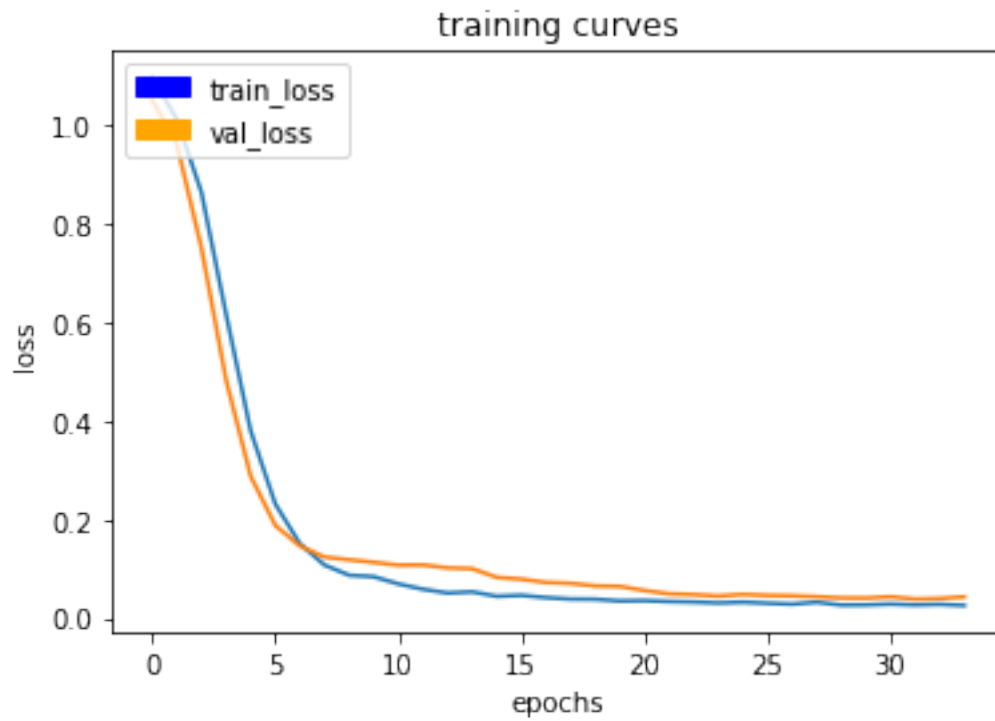


```

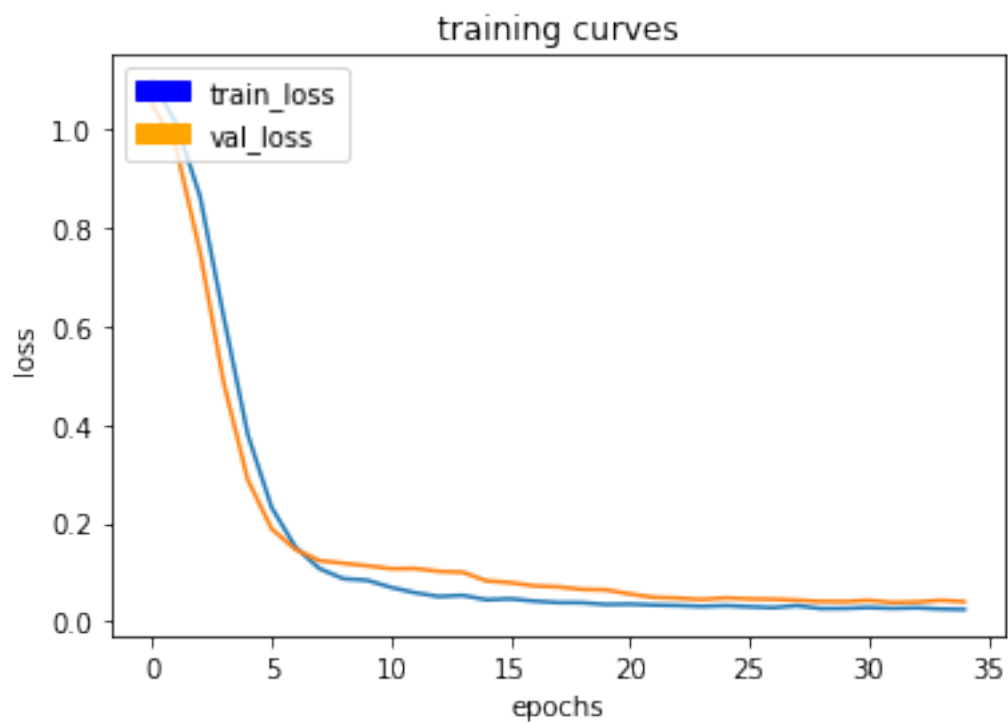
30/30 [=====] - 24s - loss: 0.0266 - val_loss: 0.0385
Epoch 33/200
29/30 [=====>.] - ETA: 0s - loss: 0.0280
  
```



30/30 [=====] - 23s - loss: 0.0278 - val\_loss: 0.0396  
Epoch 34/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0257



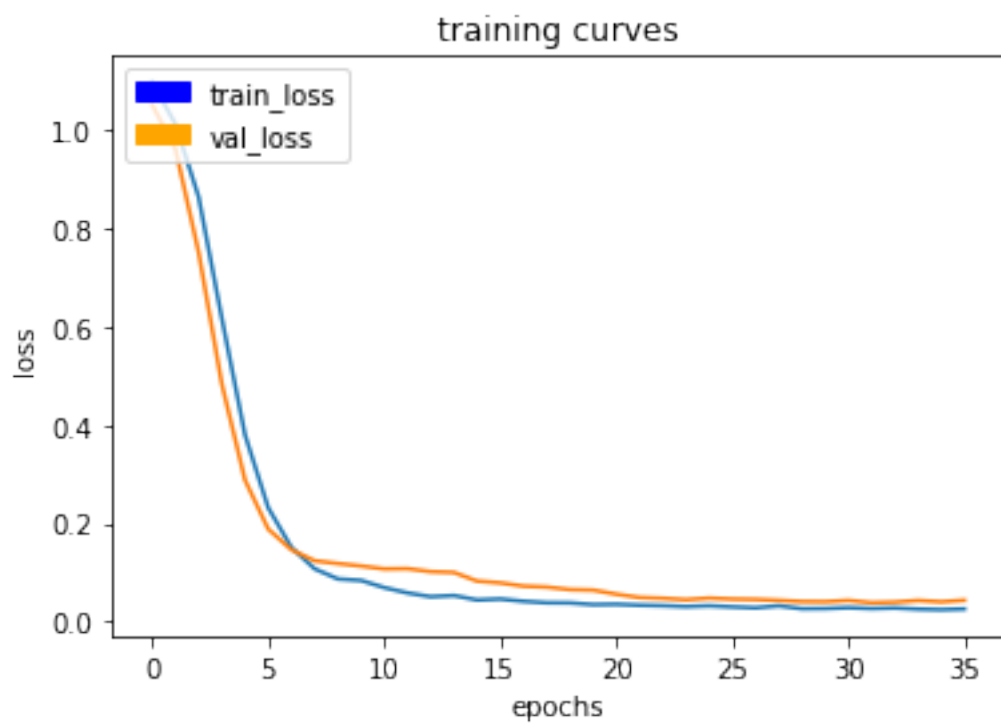
30/30 [=====] - 23s - loss: 0.0255 - val\_loss: 0.0430  
Epoch 35/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0237



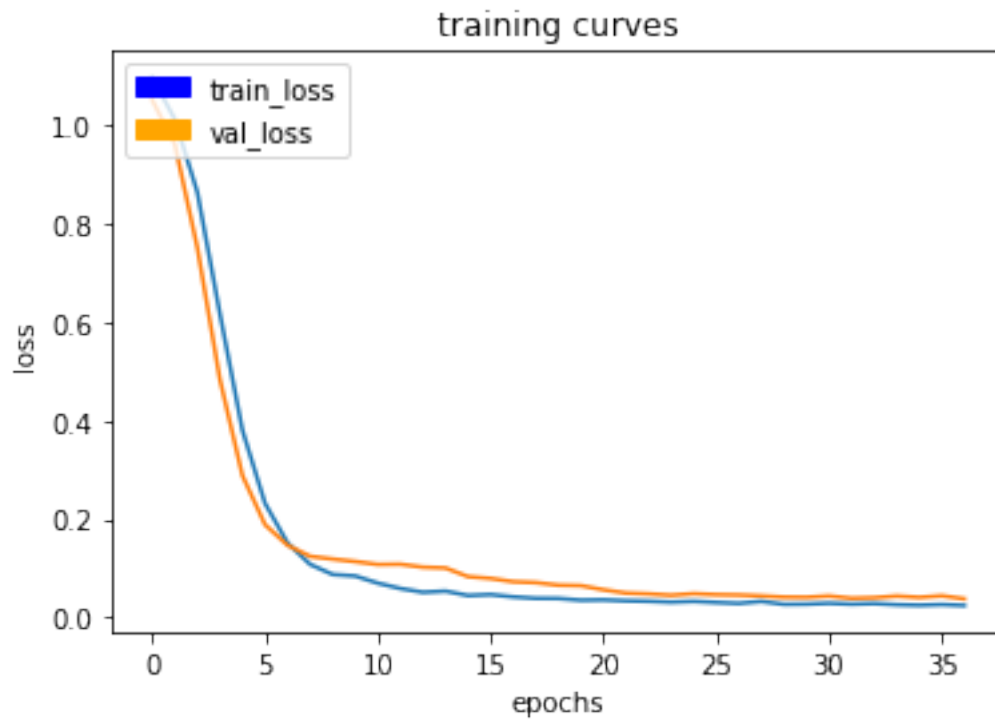
30/30 [=====] - 23s - loss: 0.0244 - val\_loss: 0.0403

Epoch 36/200

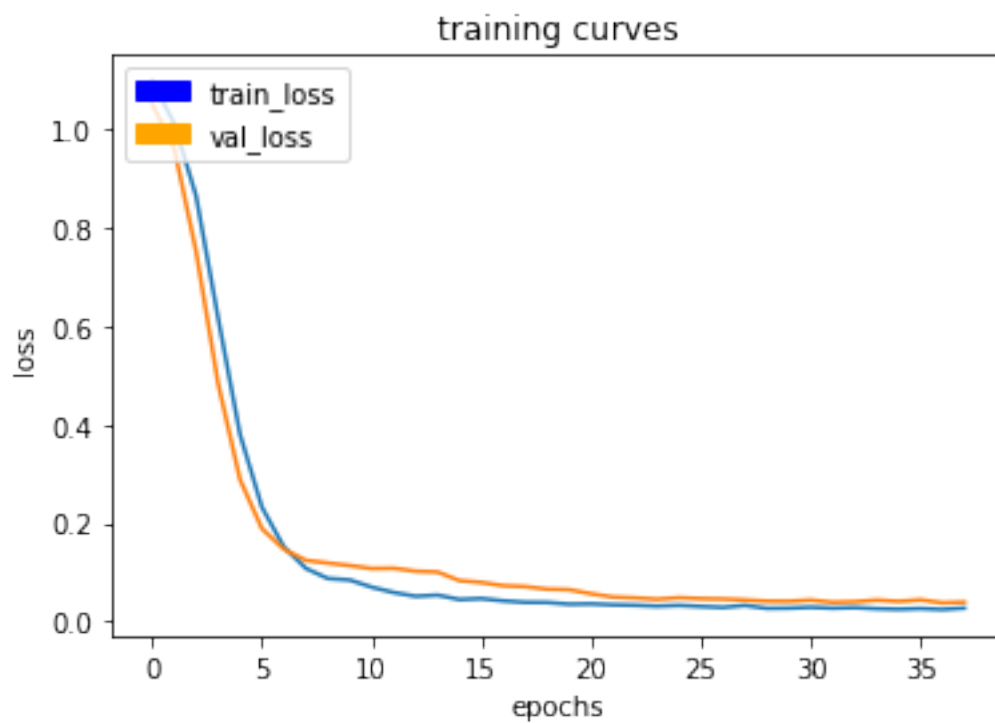
29/30 [=====>.] - ETA: 0s - loss: 0.0258



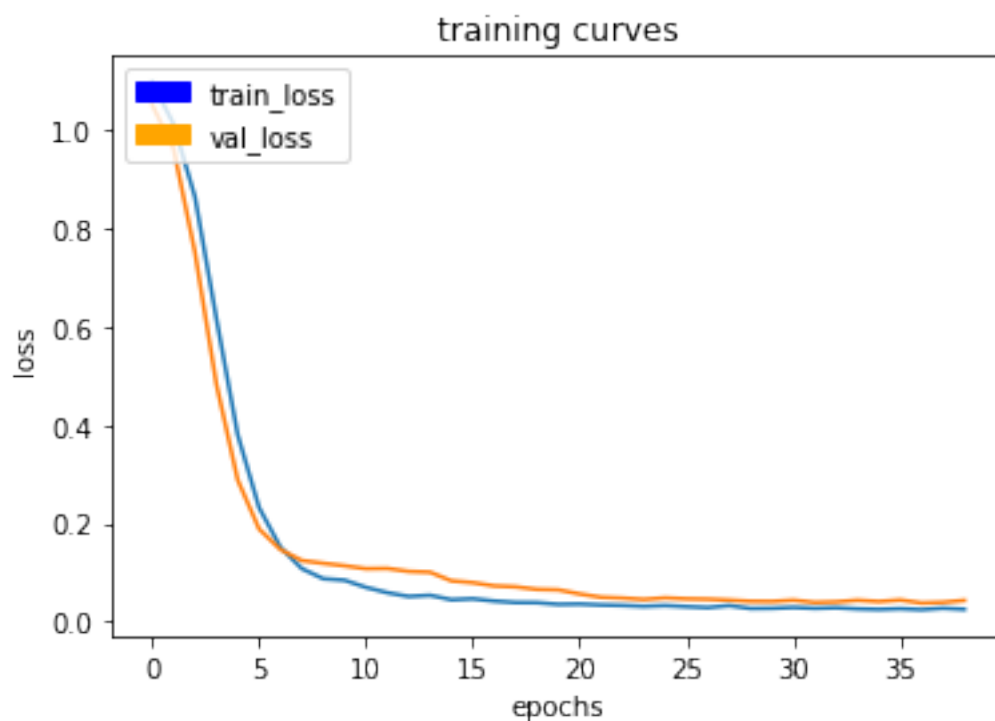
30/30 [=====] - 23s - loss: 0.0257 - val\_loss: 0.0435  
Epoch 37/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0241



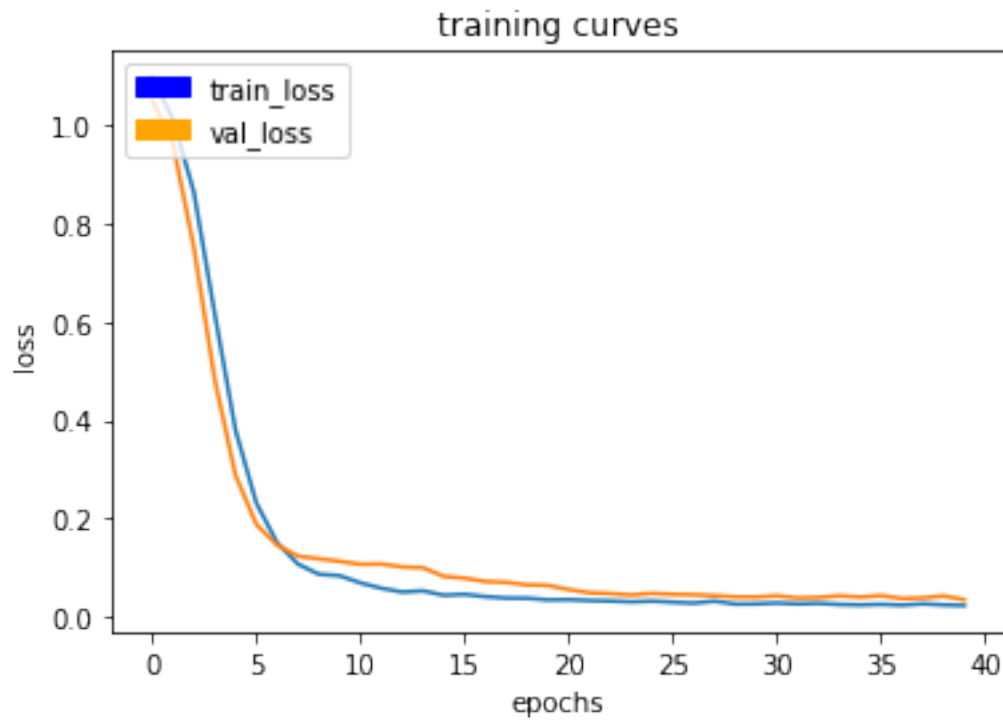
30/30 [=====] - 23s - loss: 0.0239 - val\_loss: 0.0374  
Epoch 38/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0270



30/30 [=====] - 23s - loss: 0.0267 - val\_loss: 0.0388  
 Epoch 39/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0237

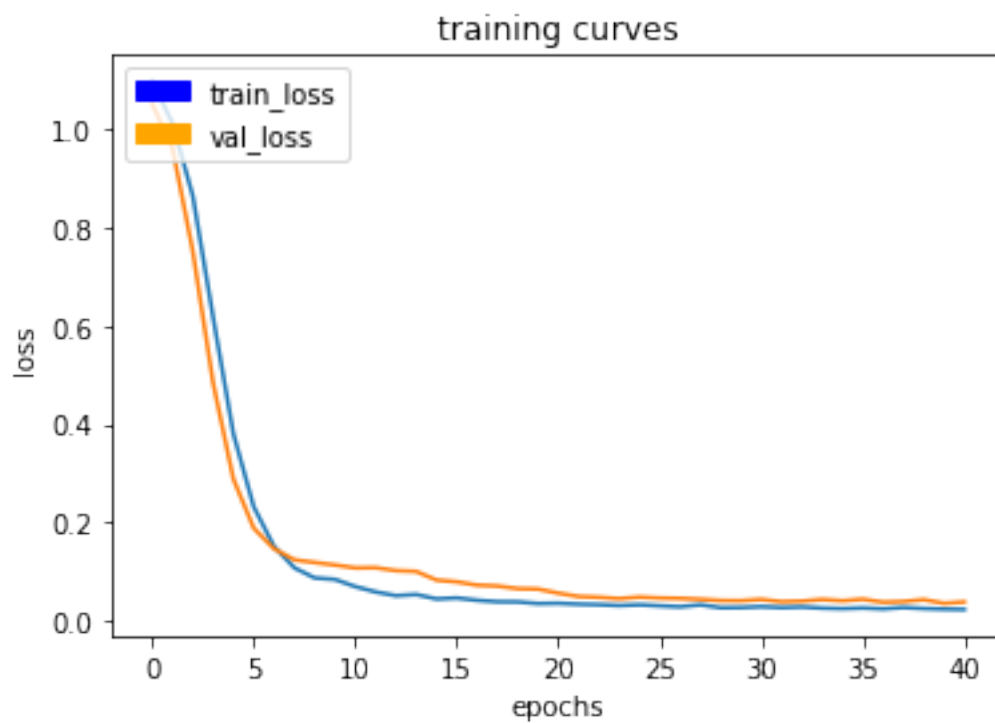


30/30 [=====] - 23s - loss: 0.0245 - val\_loss: 0.0427  
Epoch 40/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0234

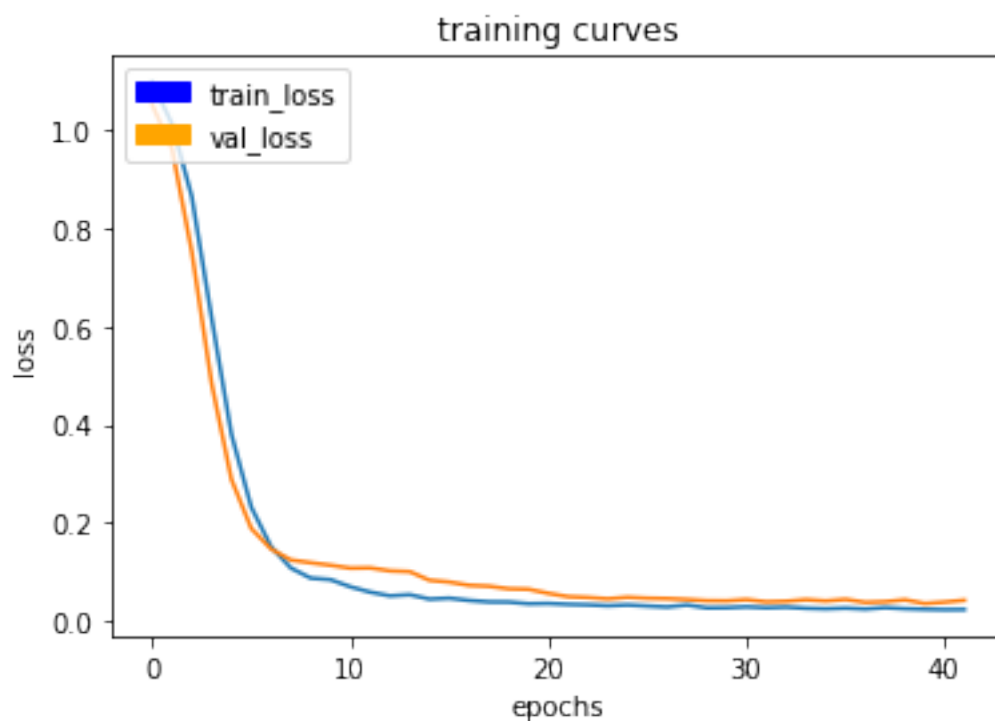


30/30 [=====] - 23s - loss: 0.0233 - val\_loss: 0.0351  
Epoch 41/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0225

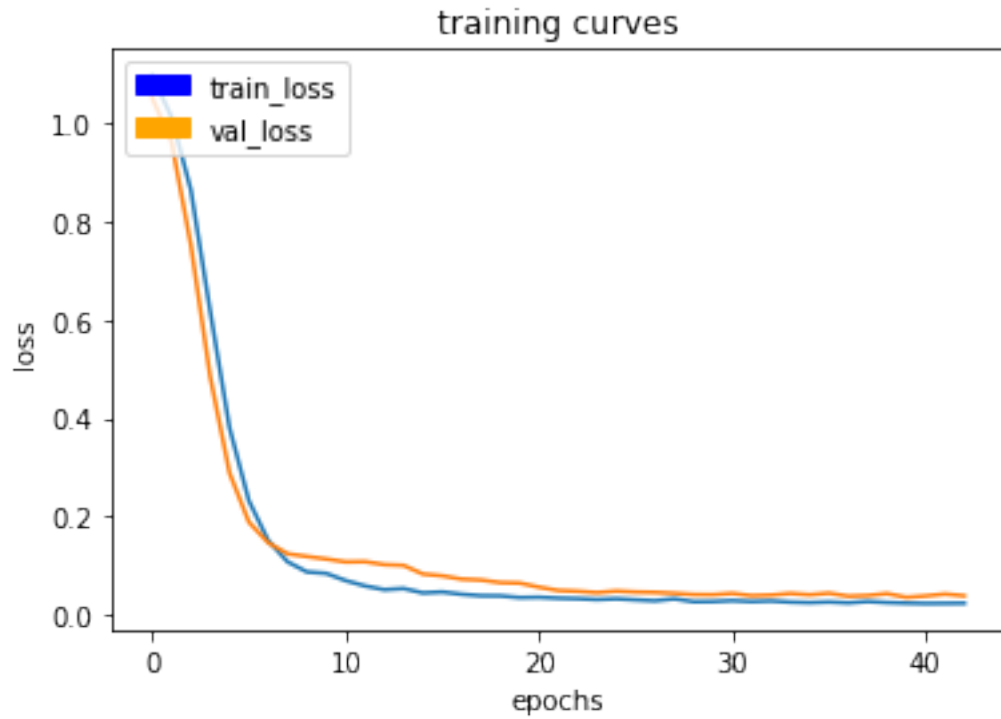




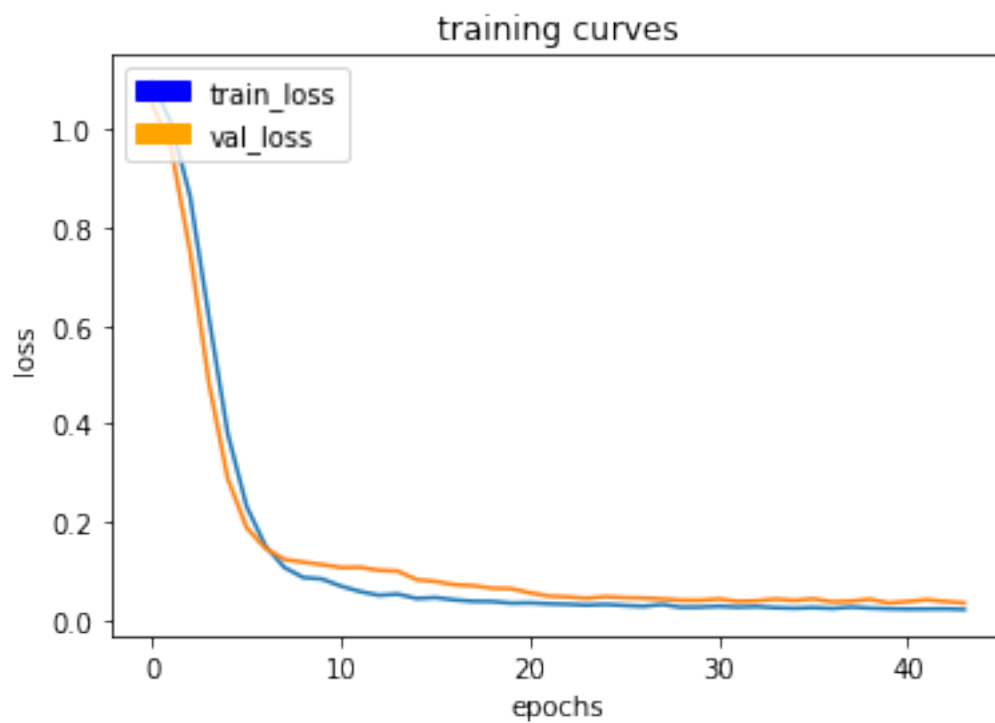
30/30 [=====] - 23s - loss: 0.0226 - val\_loss: 0.0380  
 Epoch 42/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0225



30/30 [=====] - 23s - loss: 0.0227 - val\_loss: 0.0418  
Epoch 43/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0230



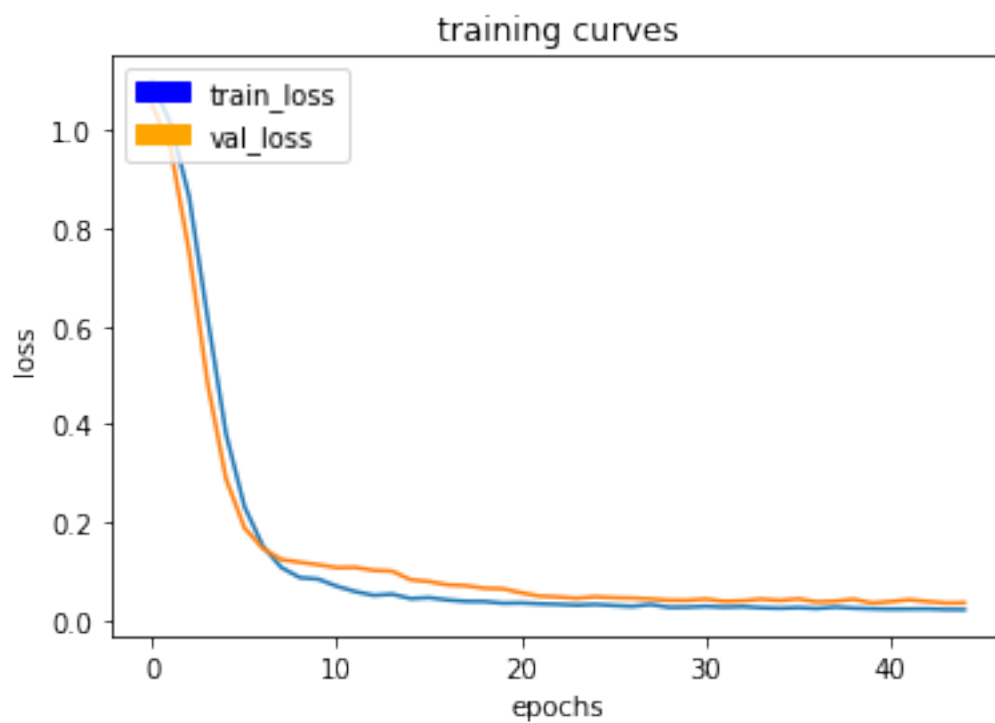
30/30 [=====] - 23s - loss: 0.0231 - val\_loss: 0.0382  
Epoch 44/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0219



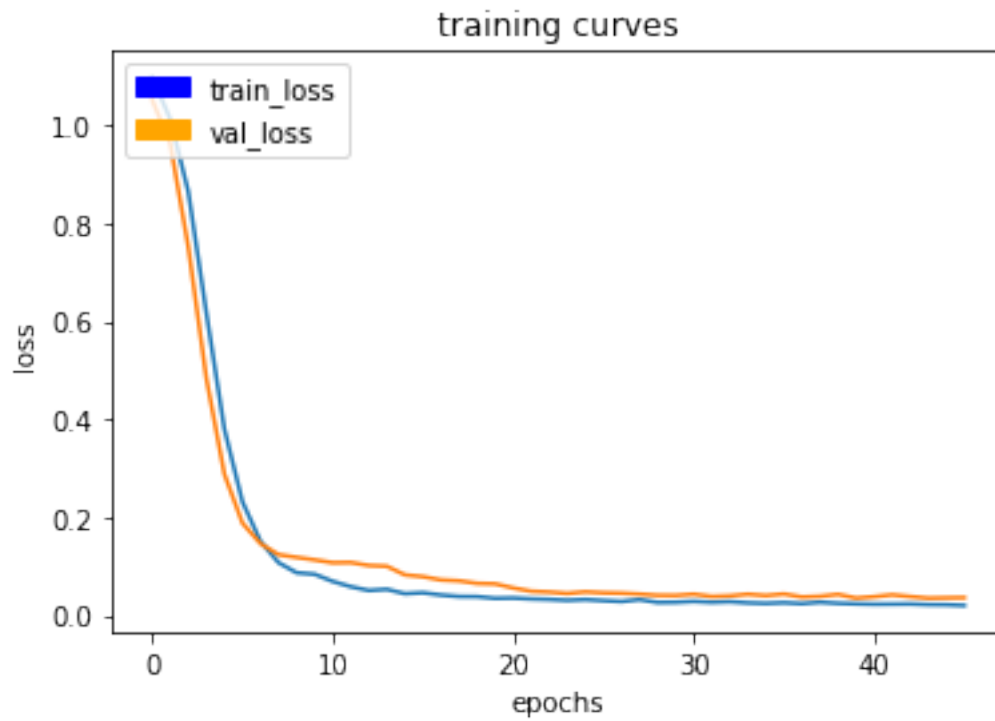
30/30 [=====] - 23s - loss: 0.0218 - val\_loss: 0.0350

Epoch 45/200

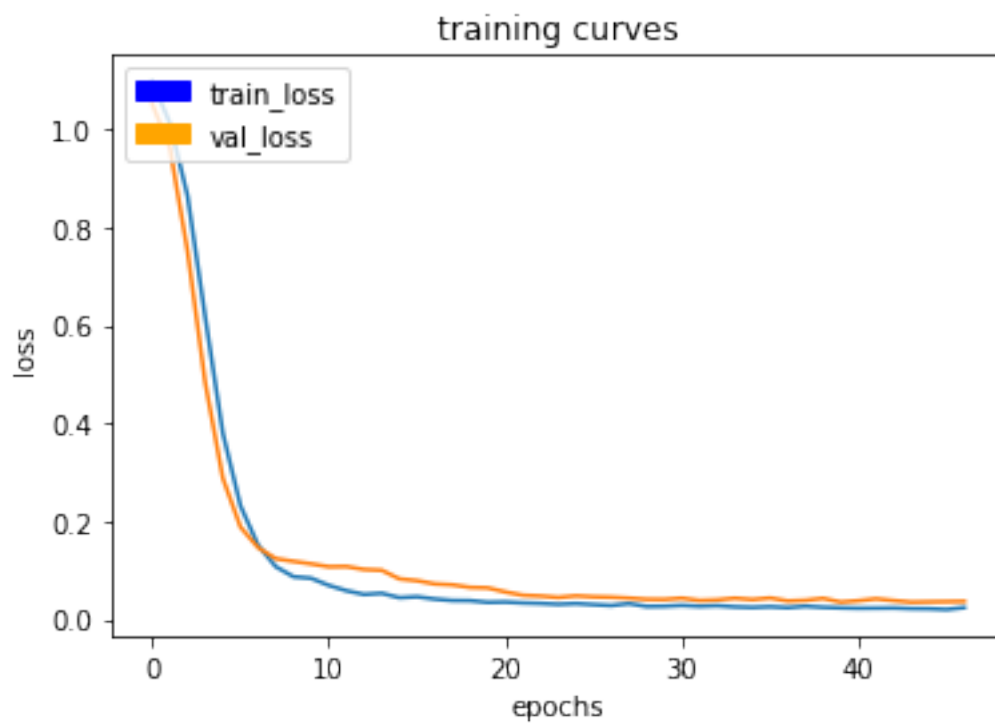
29/30 [=====>.] - ETA: 0s - loss: 0.0216



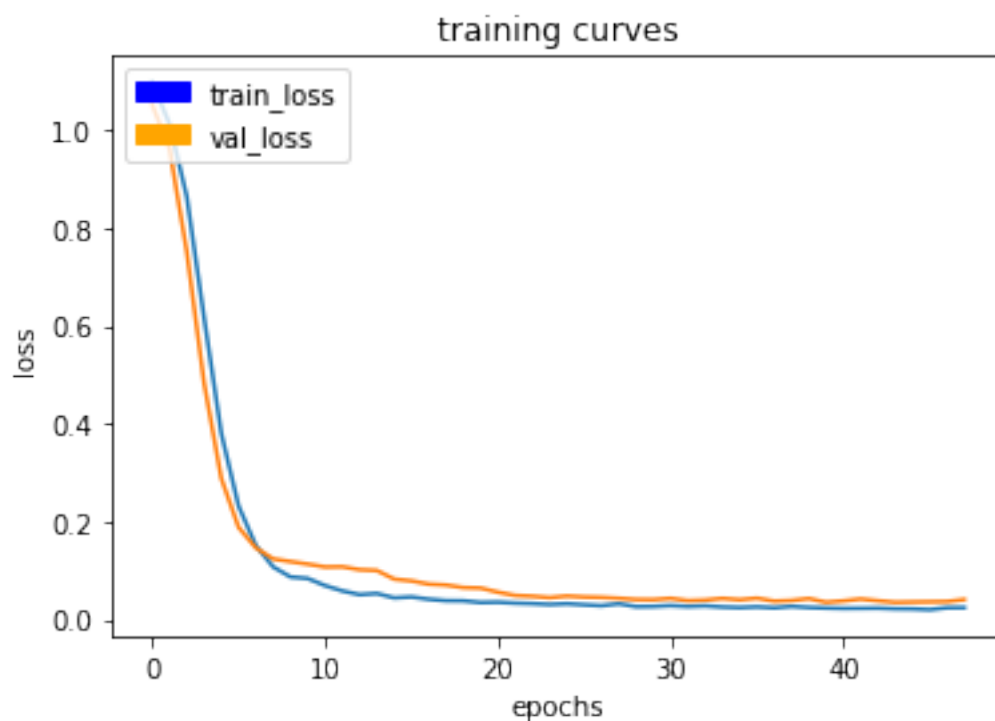
30/30 [=====] - 23s - loss: 0.0215 - val\_loss: 0.0356  
Epoch 46/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0197



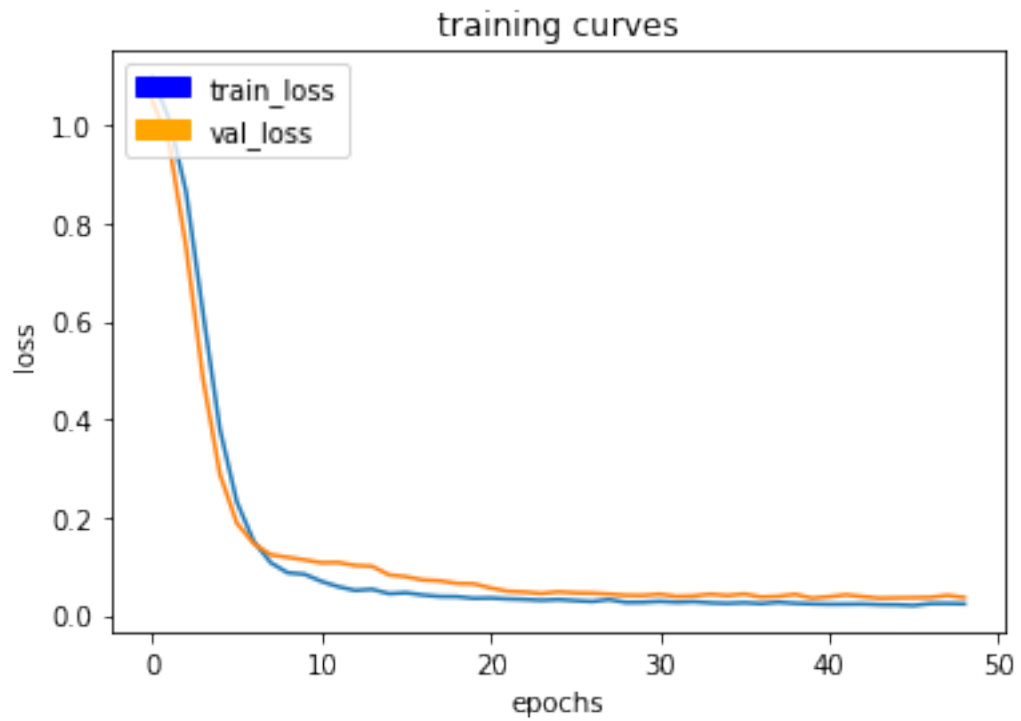
30/30 [=====] - 23s - loss: 0.0201 - val\_loss: 0.0361  
Epoch 47/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0240



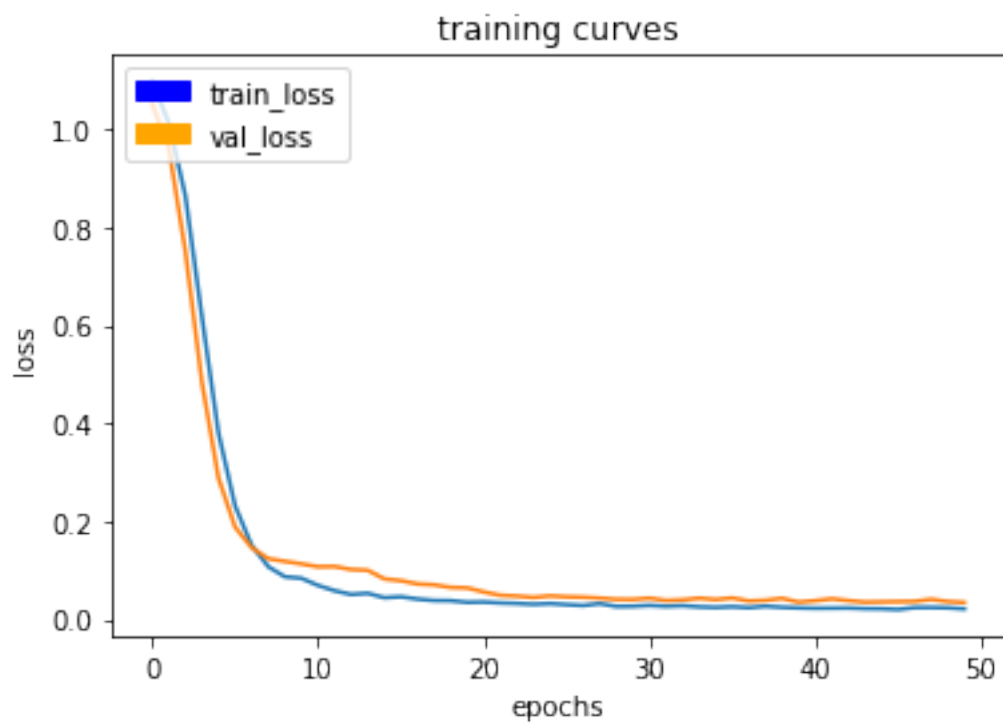
30/30 [=====] - 23s - loss: 0.0240 - val\_loss: 0.0366  
 Epoch 48/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0250



30/30 [=====] - 23s - loss: 0.0248 - val\_loss: 0.0408  
Epoch 49/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0237



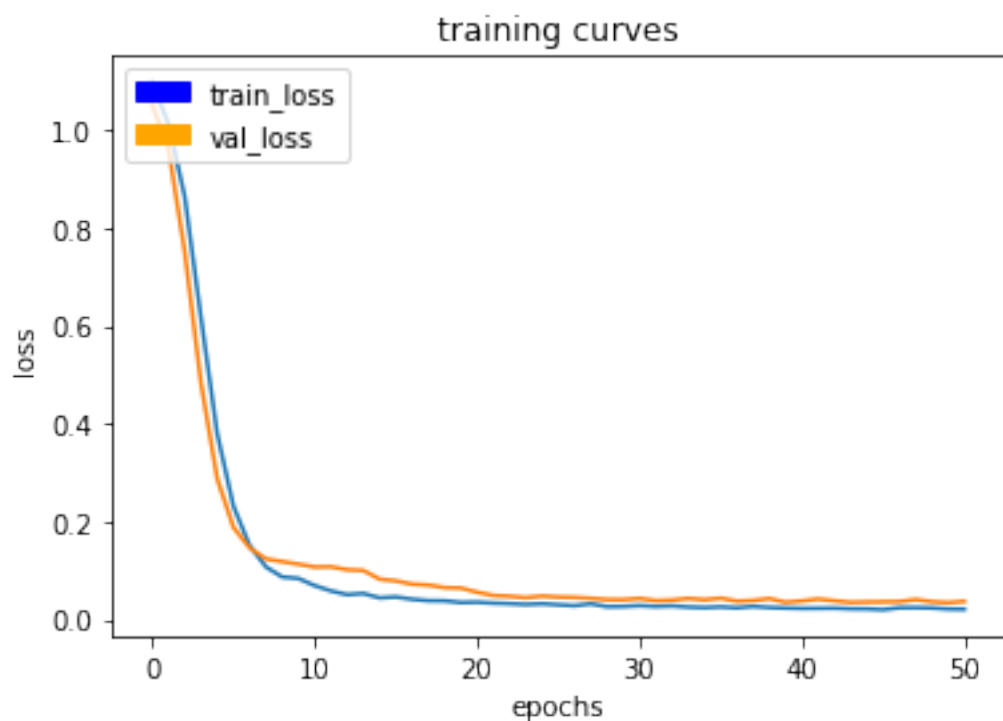
30/30 [=====] - 23s - loss: 0.0235 - val\_loss: 0.0363  
Epoch 50/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0210



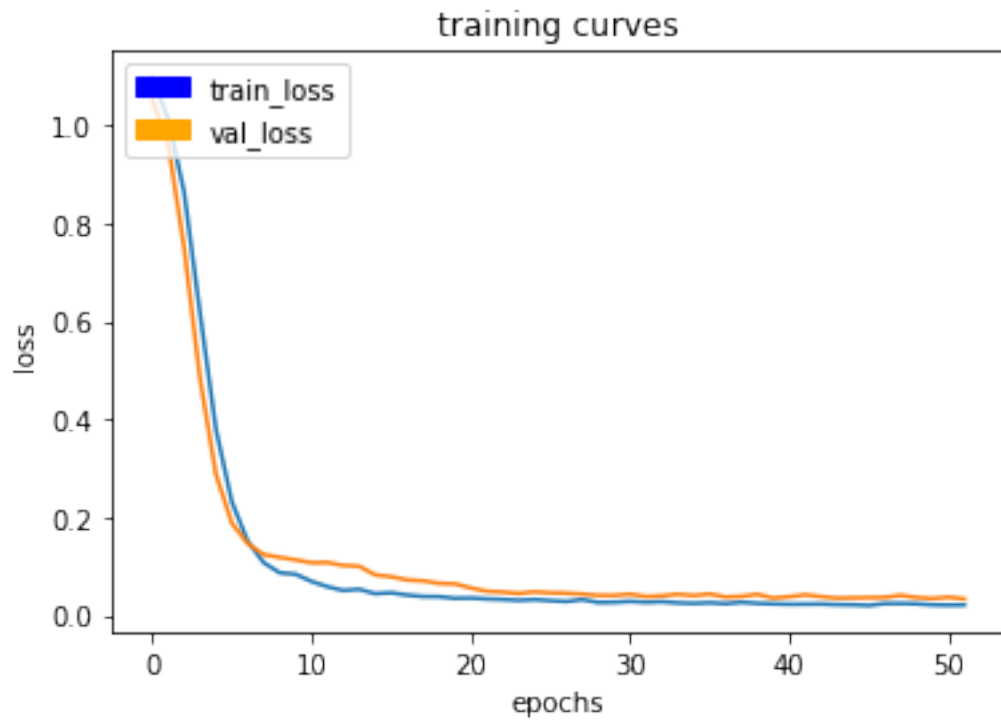
```

30/30 [=====] - 23s - loss: 0.0213 - val_loss: 0.0343
Epoch 51/200
29/30 [=====>.] - ETA: 0s - loss: 0.0208

```

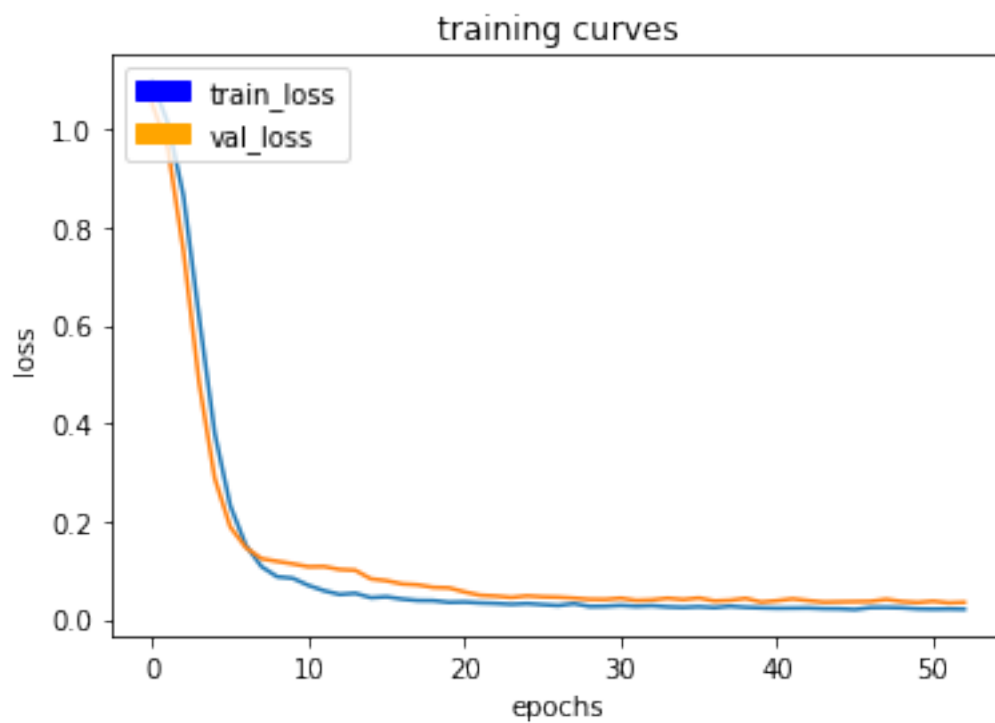


30/30 [=====] - 23s - loss: 0.0208 - val\_loss: 0.0369  
Epoch 52/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0212



30/30 [=====] - 23s - loss: 0.0212 - val\_loss: 0.0336  
Epoch 53/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0207

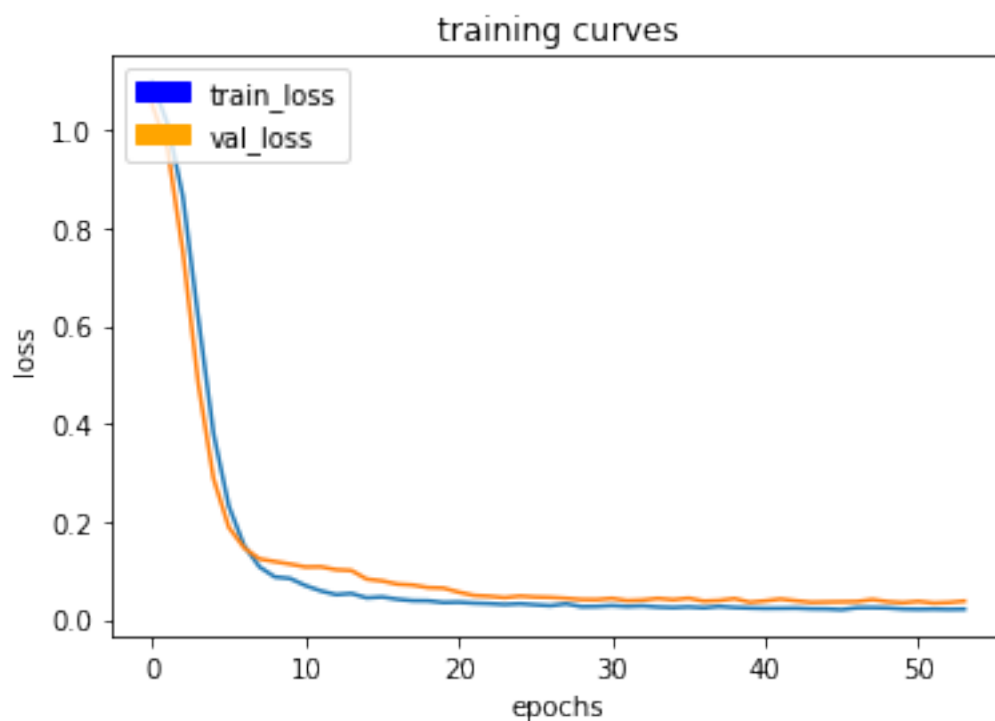




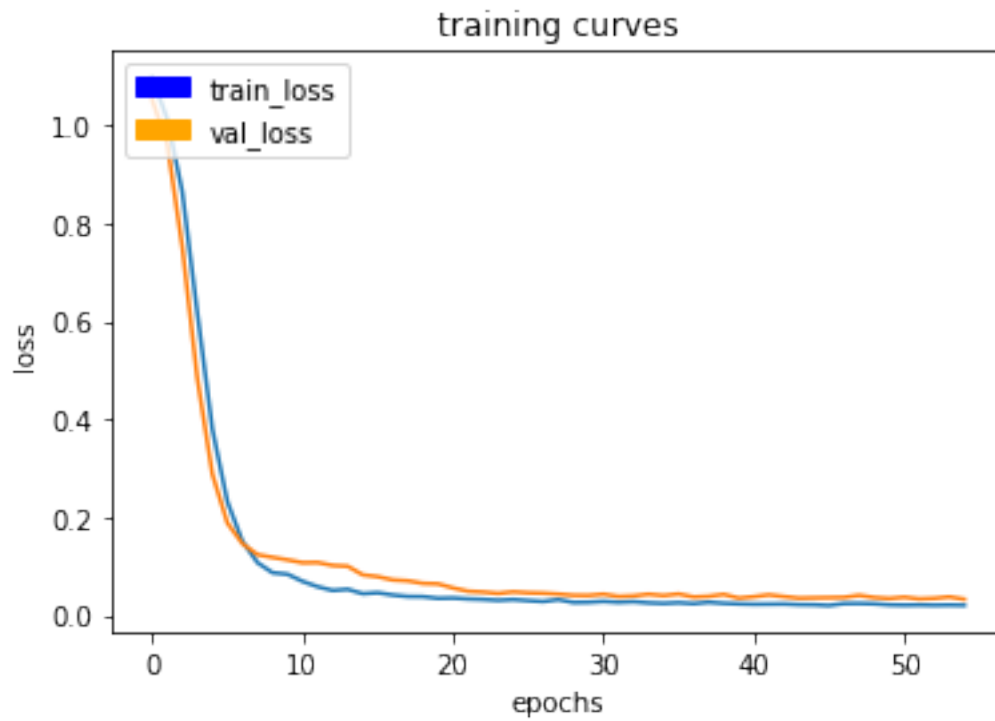
```

30/30 [=====] - 23s - loss: 0.0206 - val_loss: 0.0347
Epoch 54/200
29/30 [=====>.] - ETA: 0s - loss: 0.0212

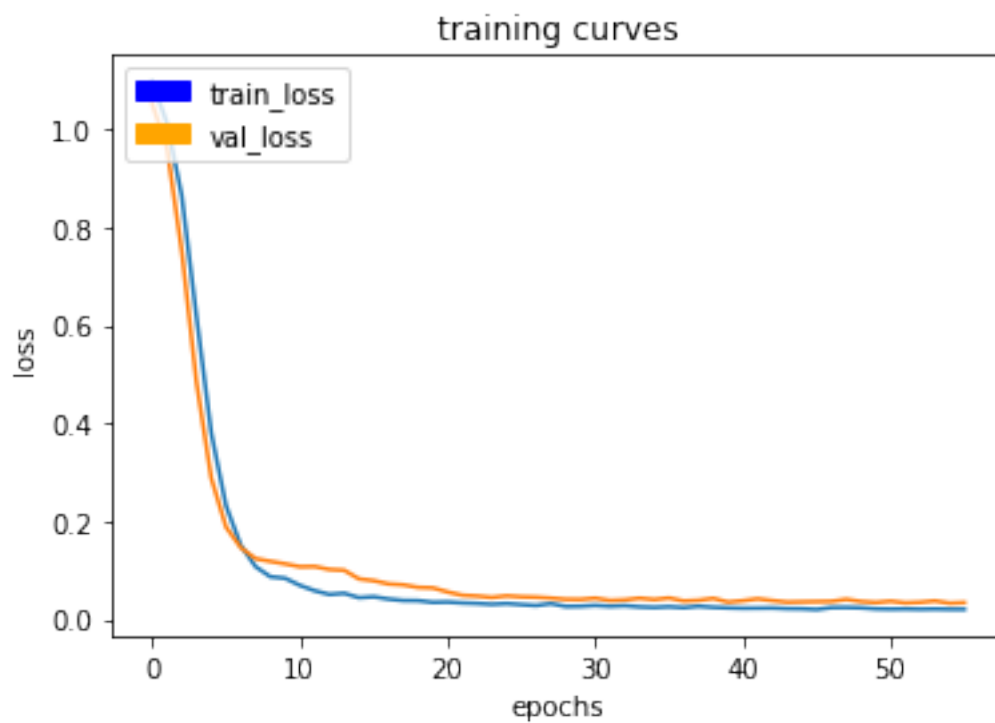
```



30/30 [=====] - 23s - loss: 0.0210 - val\_loss: 0.0376  
Epoch 55/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0206



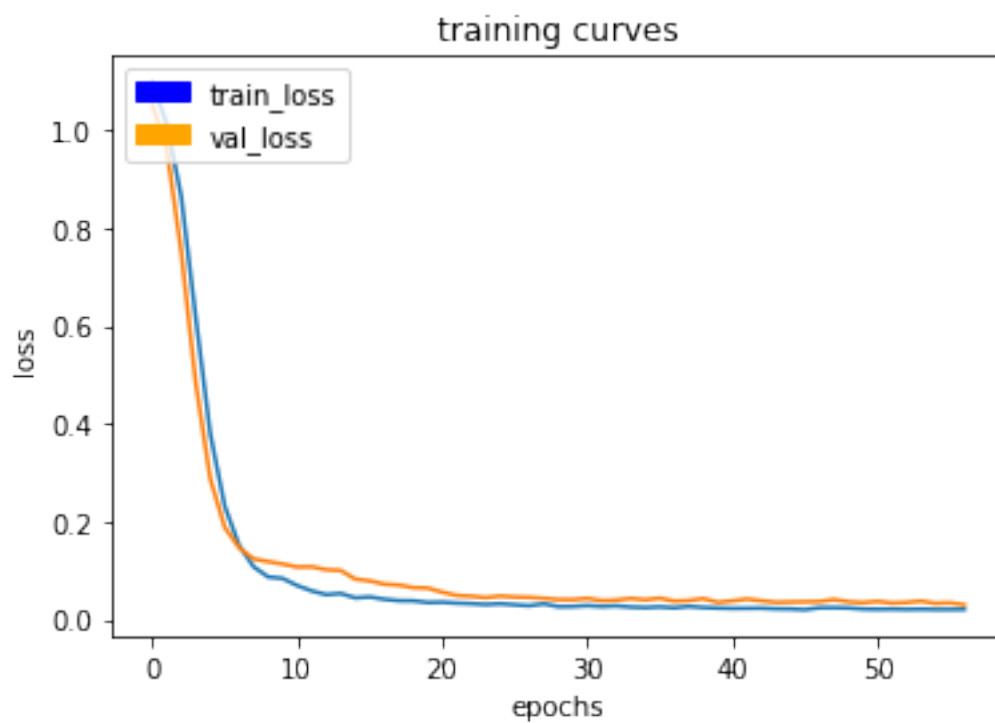
30/30 [=====] - 23s - loss: 0.0205 - val\_loss: 0.0329  
Epoch 56/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0204



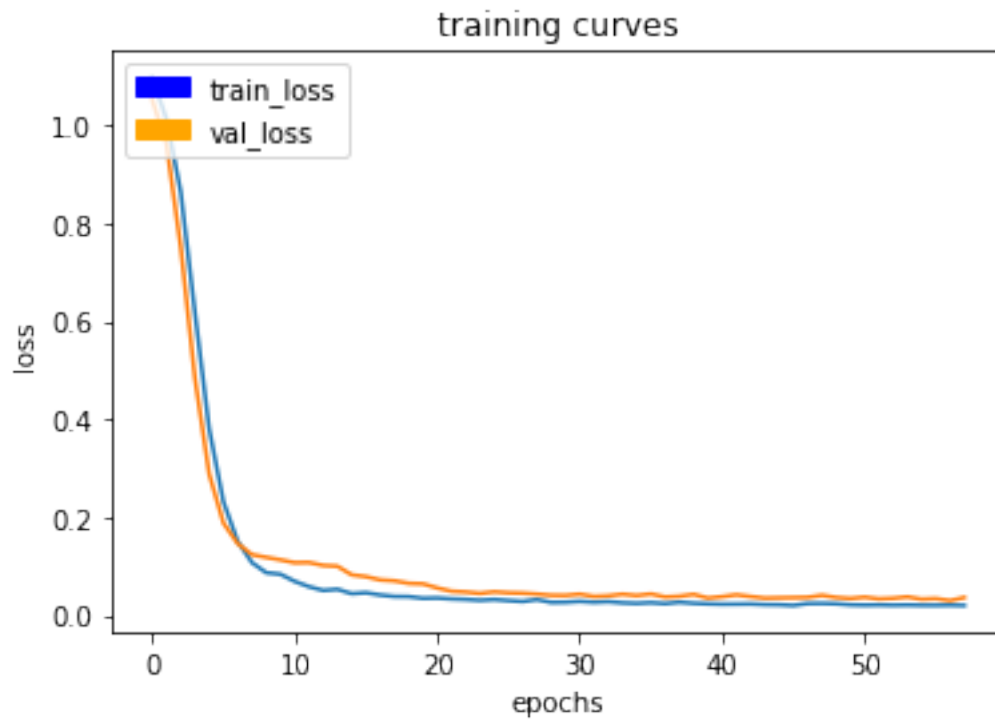
30/30 [=====] - 23s - loss: 0.0204 - val\_loss: 0.0341

Epoch 57/200

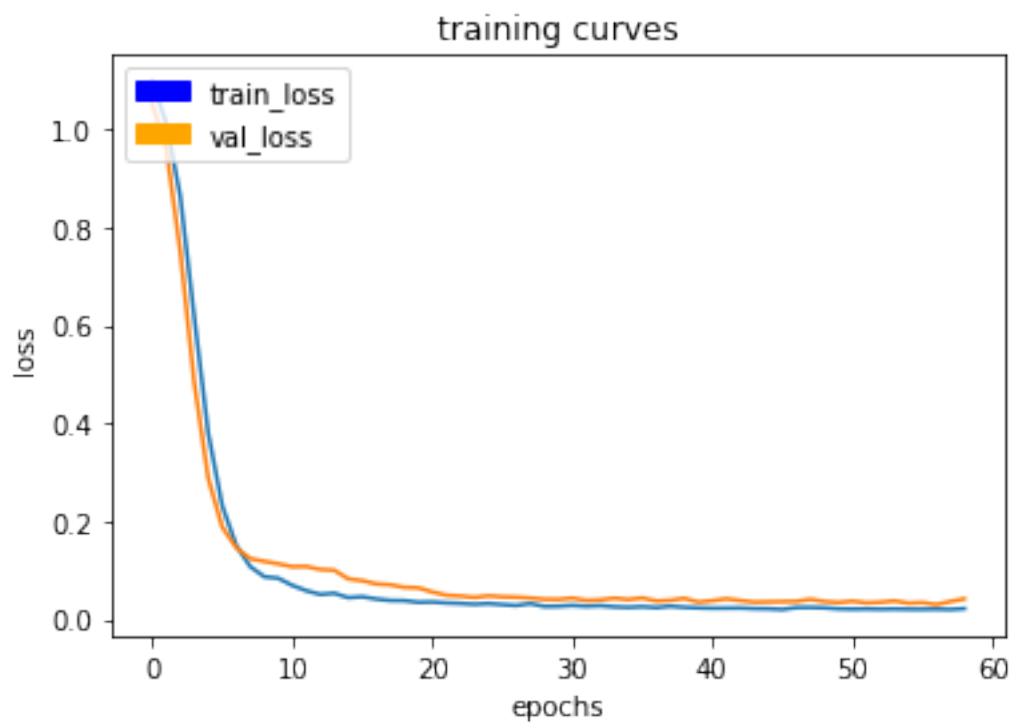
29/30 [=====>.] - ETA: 0s - loss: 0.0207



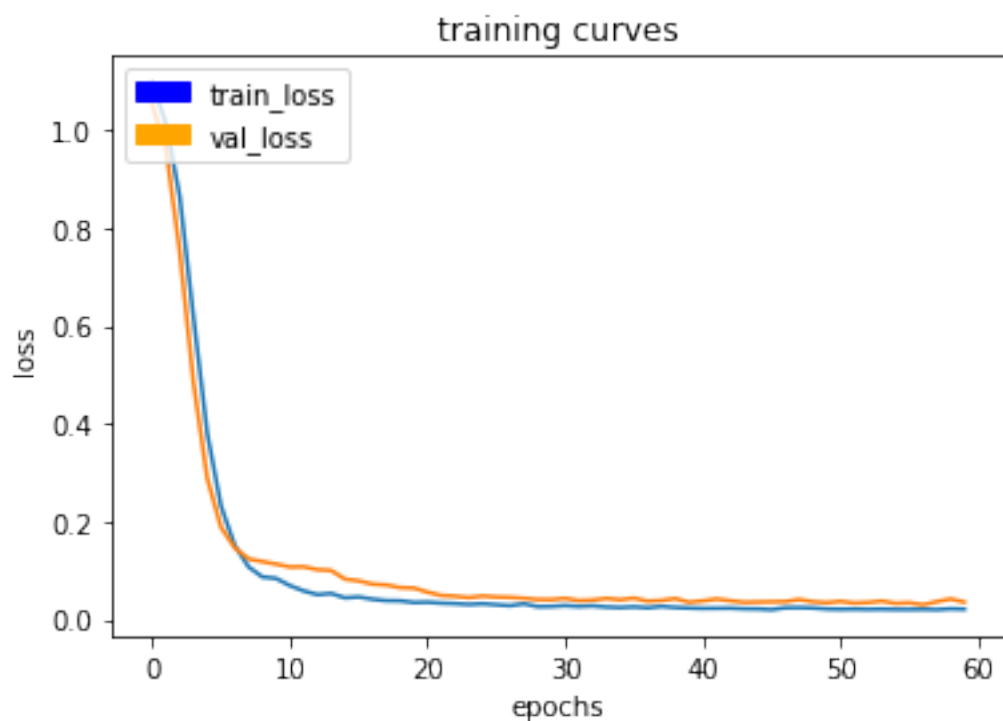
30/30 [=====] - 23s - loss: 0.0207 - val\_loss: 0.0300  
Epoch 58/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0201



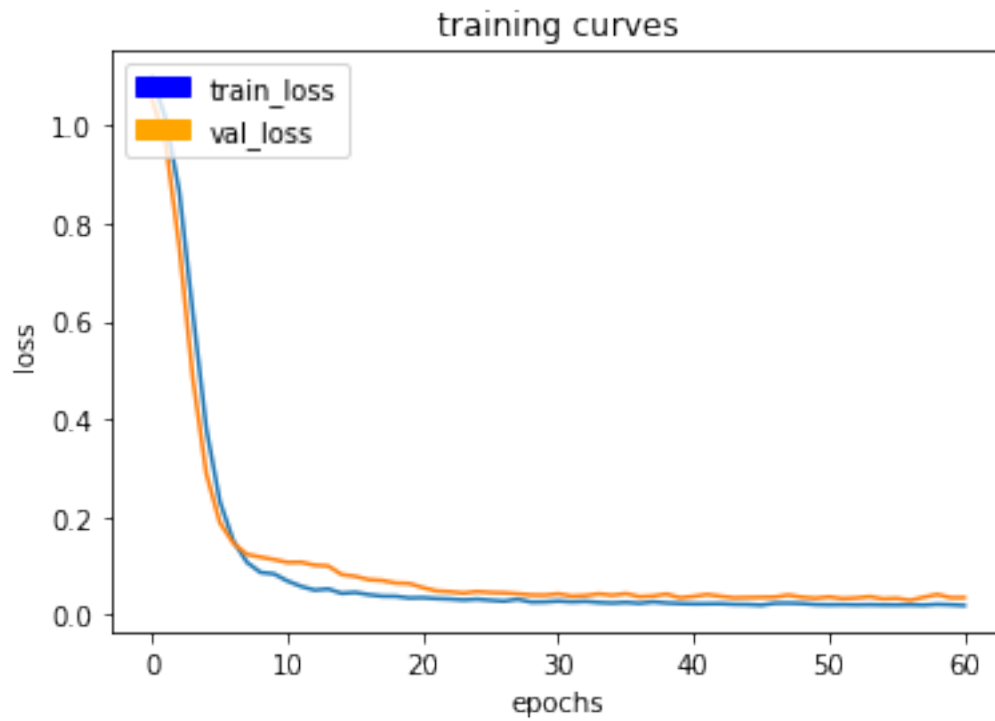
30/30 [=====] - 23s - loss: 0.0200 - val\_loss: 0.0363  
Epoch 59/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0216



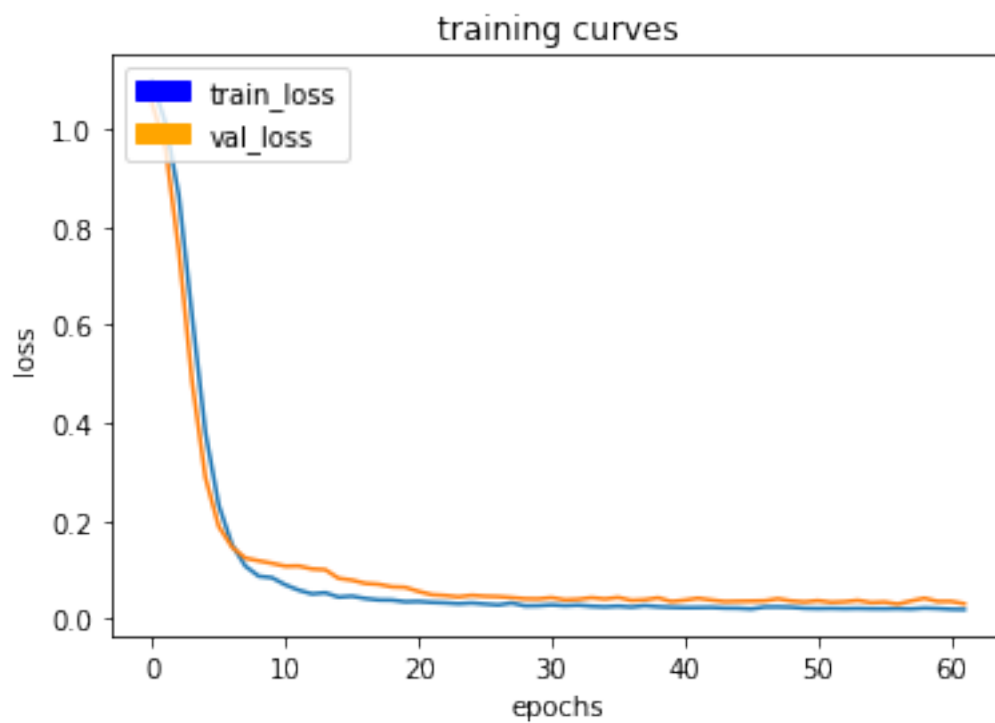
30/30 [=====] - 23s - loss: 0.0217 - val\_loss: 0.0419  
 Epoch 60/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0209



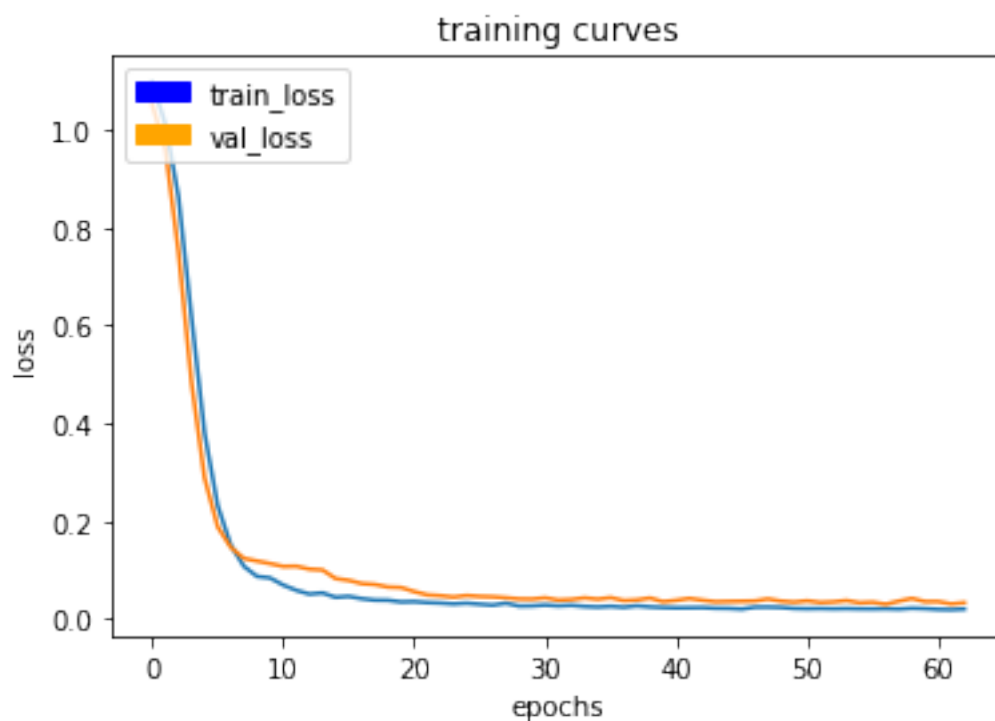
30/30 [=====] - 23s - loss: 0.0209 - val\_loss: 0.0354  
Epoch 61/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0195



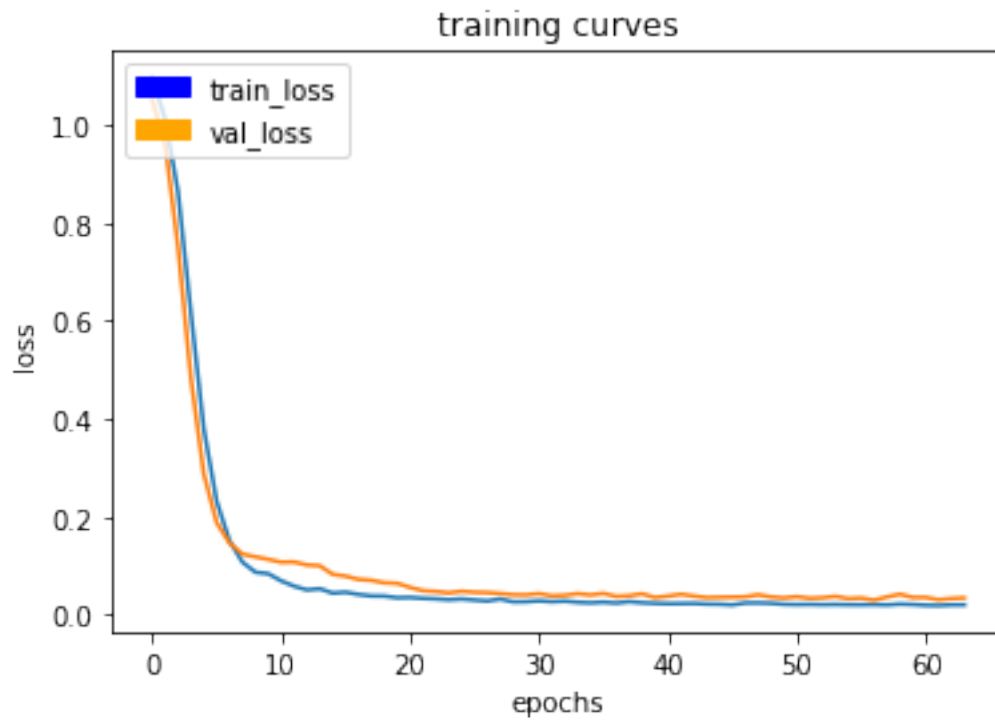
30/30 [=====] - 23s - loss: 0.0194 - val\_loss: 0.0354  
Epoch 62/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0188



30/30 [=====] - 23s - loss: 0.0189 - val\_loss: 0.0306  
 Epoch 63/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0203

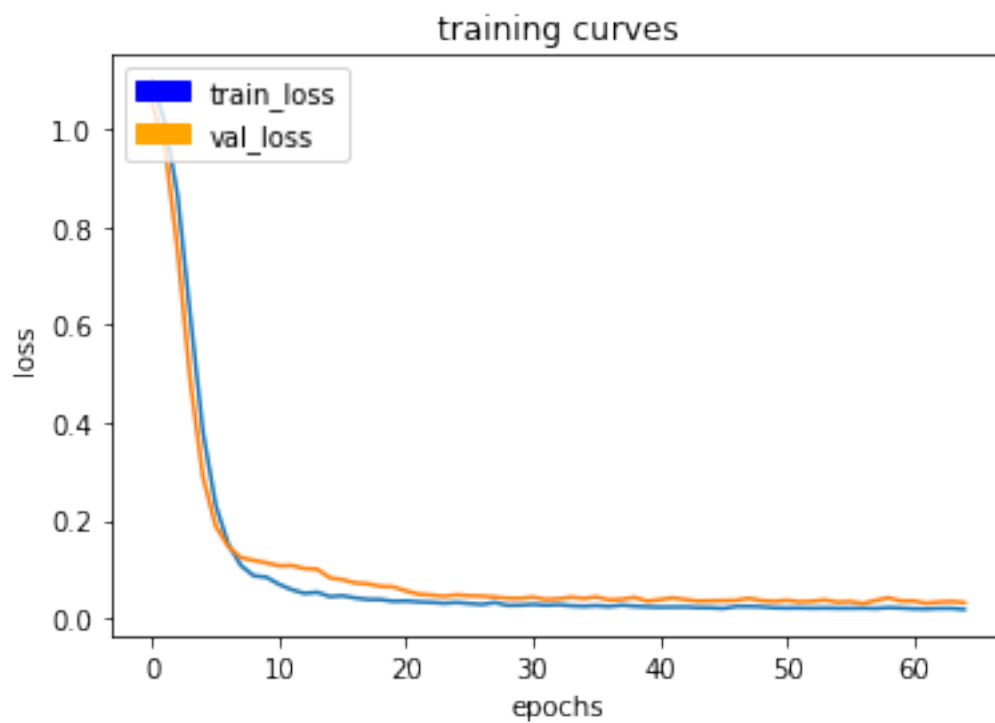


30/30 [=====] - 23s - loss: 0.0201 - val\_loss: 0.0331  
Epoch 64/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0203



30/30 [=====] - 23s - loss: 0.0201 - val\_loss: 0.0344  
Epoch 65/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0174

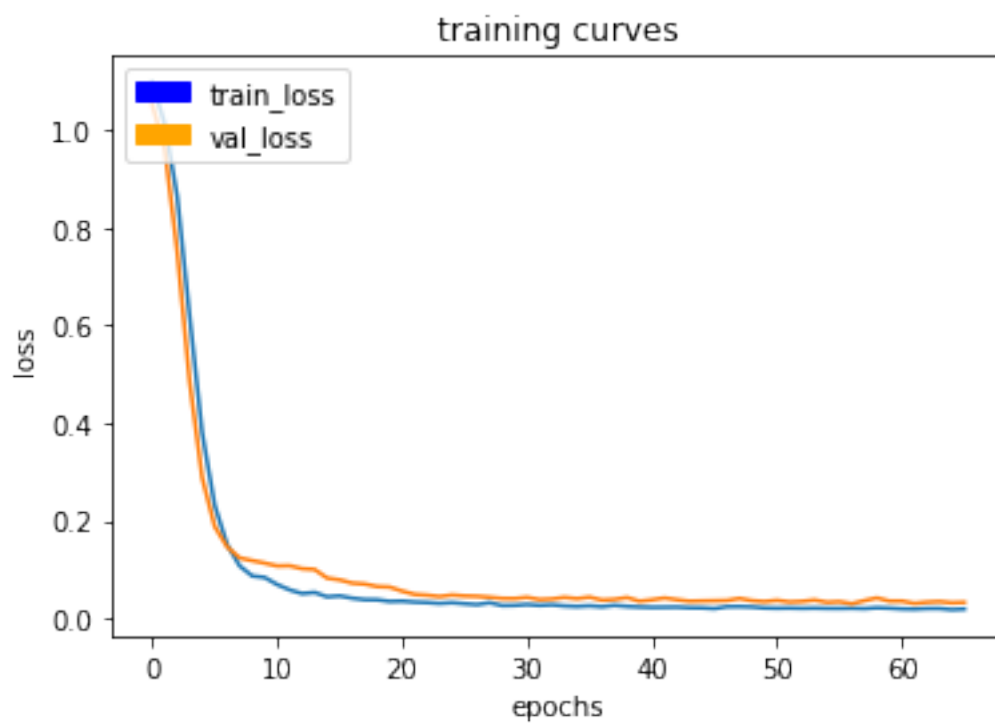




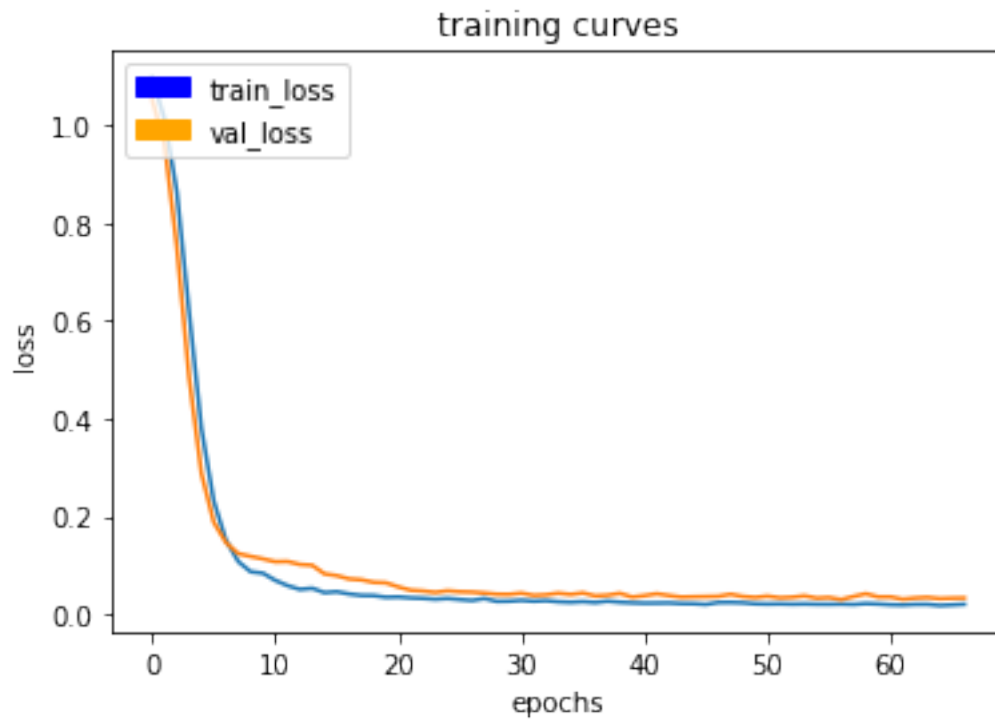
30/30 [======] - 23s - loss: 0.0179 - val\_loss: 0.0319

Epoch 66/200

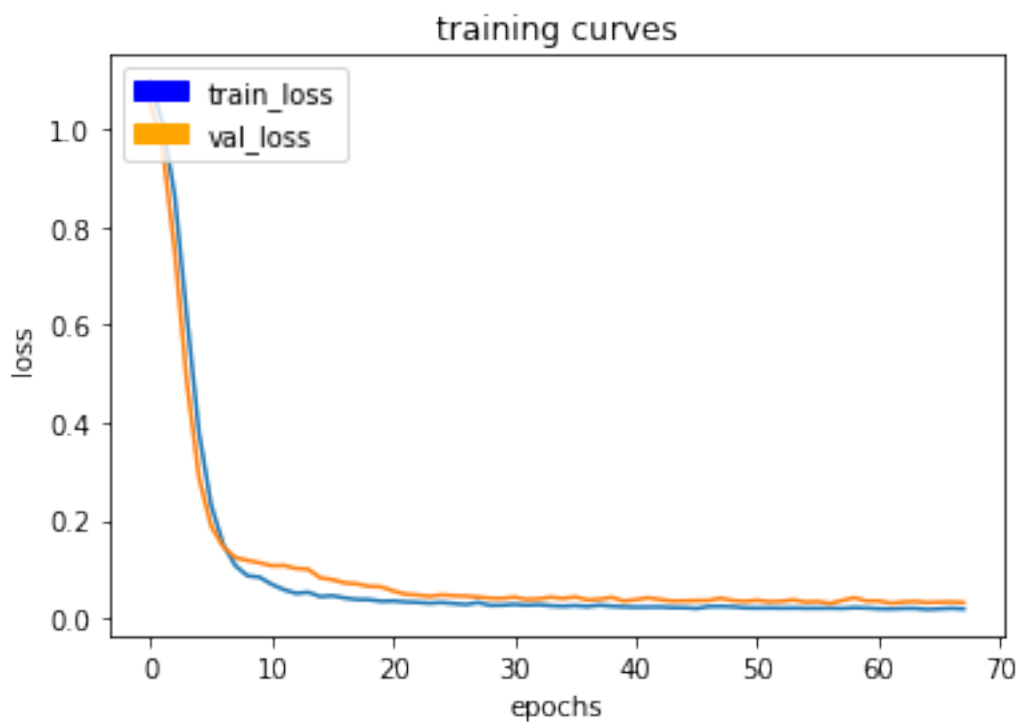
29/30 [======>.] - ETA: 0s - loss: 0.0190



30/30 [=====] - 23s - loss: 0.0190 - val\_loss: 0.0328  
Epoch 67/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0204

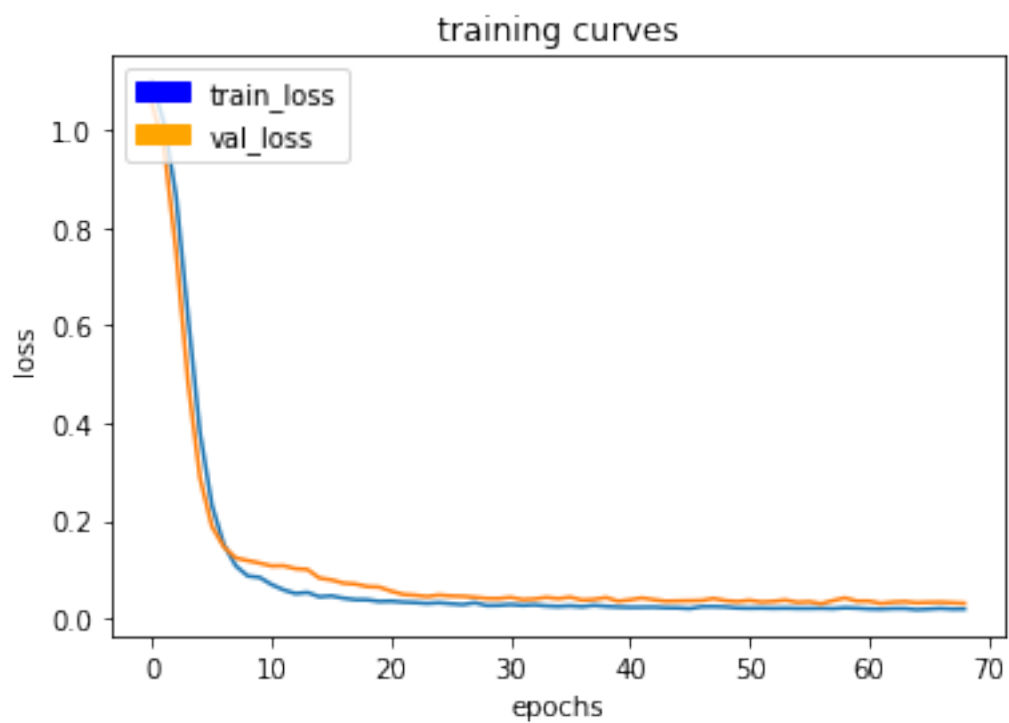


30/30 [=====] - 23s - loss: 0.0204 - val\_loss: 0.0332  
Epoch 68/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0191

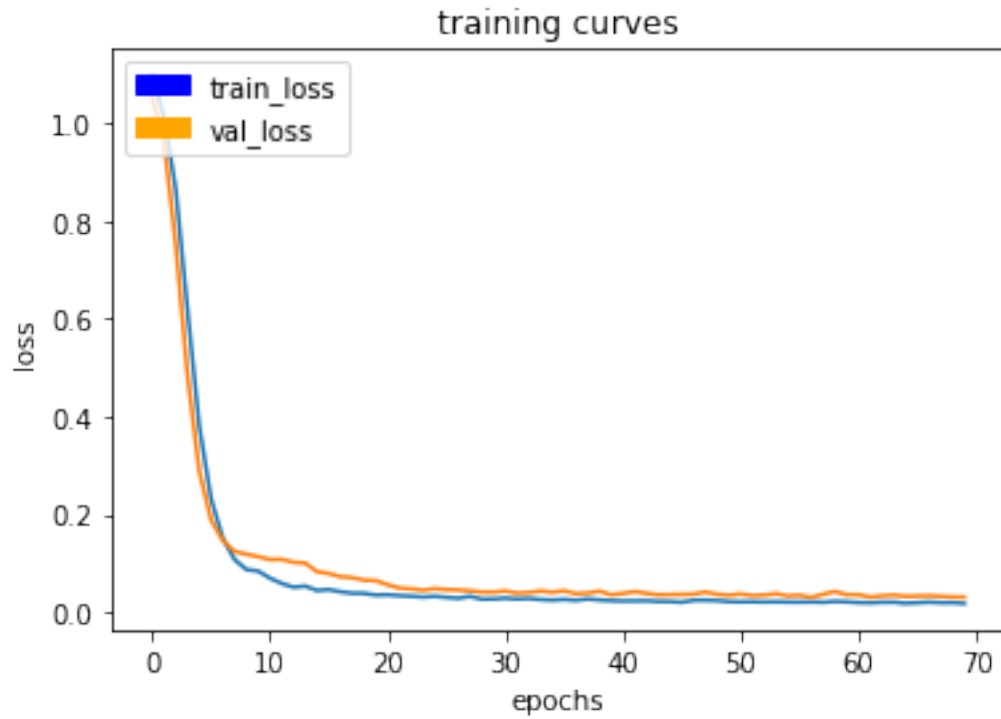


```

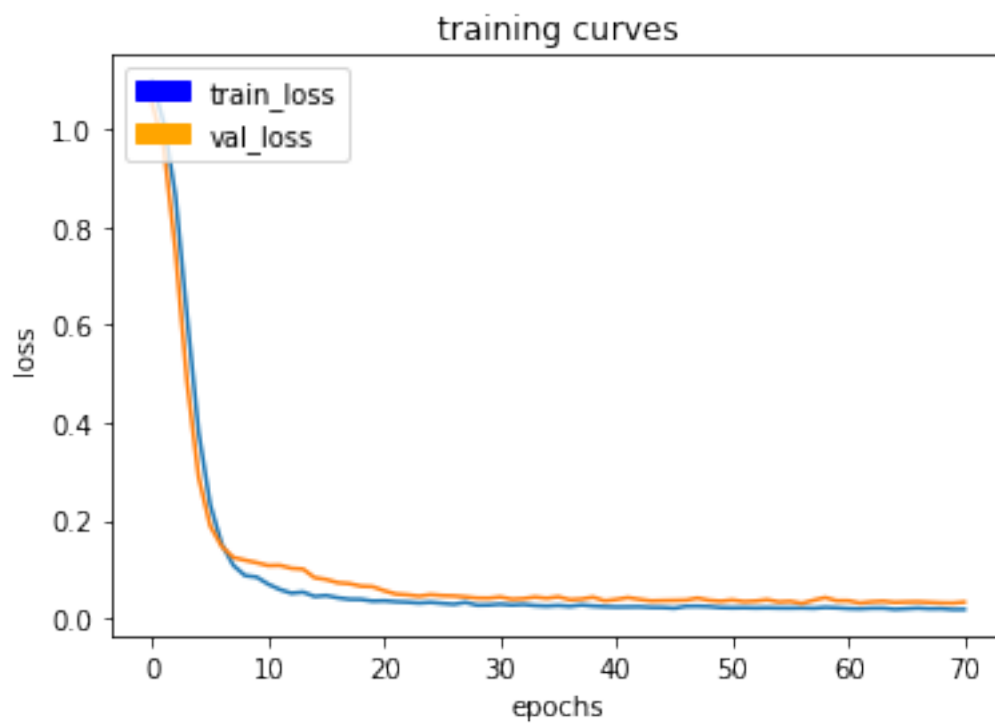
30/30 [=====] - 23s - loss: 0.0189 - val_loss: 0.0320
Epoch 69/200
29/30 [=====>.] - ETA: 0s - loss: 0.0193
  
```



30/30 [=====] - 23s - loss: 0.0192 - val\_loss: 0.0309  
Epoch 70/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0176

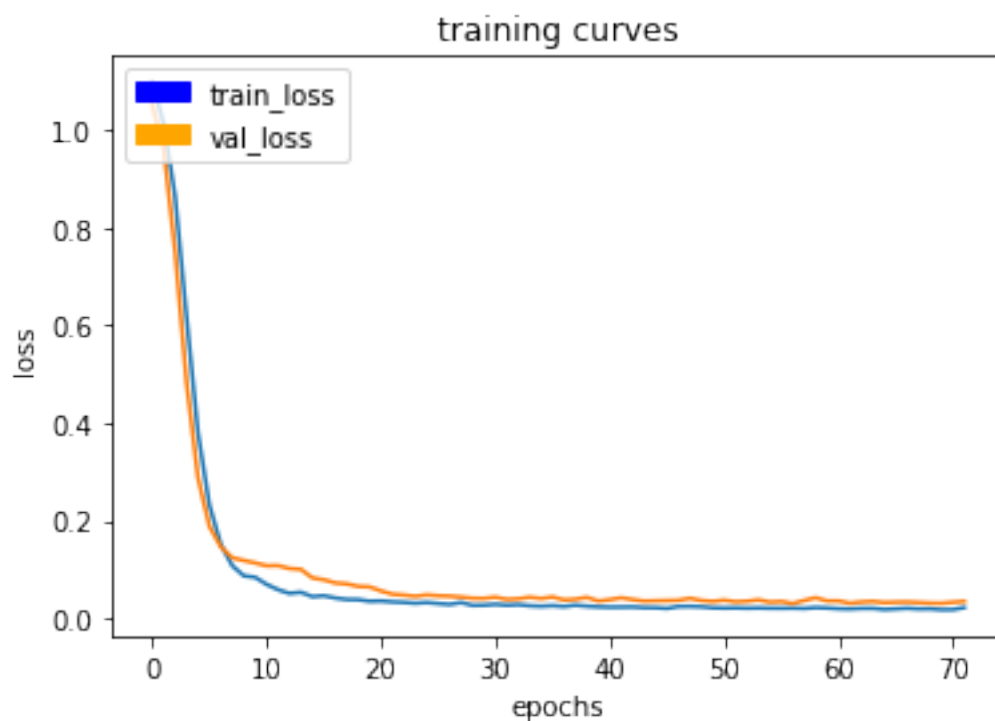


30/30 [=====] - 23s - loss: 0.0175 - val\_loss: 0.0306  
Epoch 71/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0176

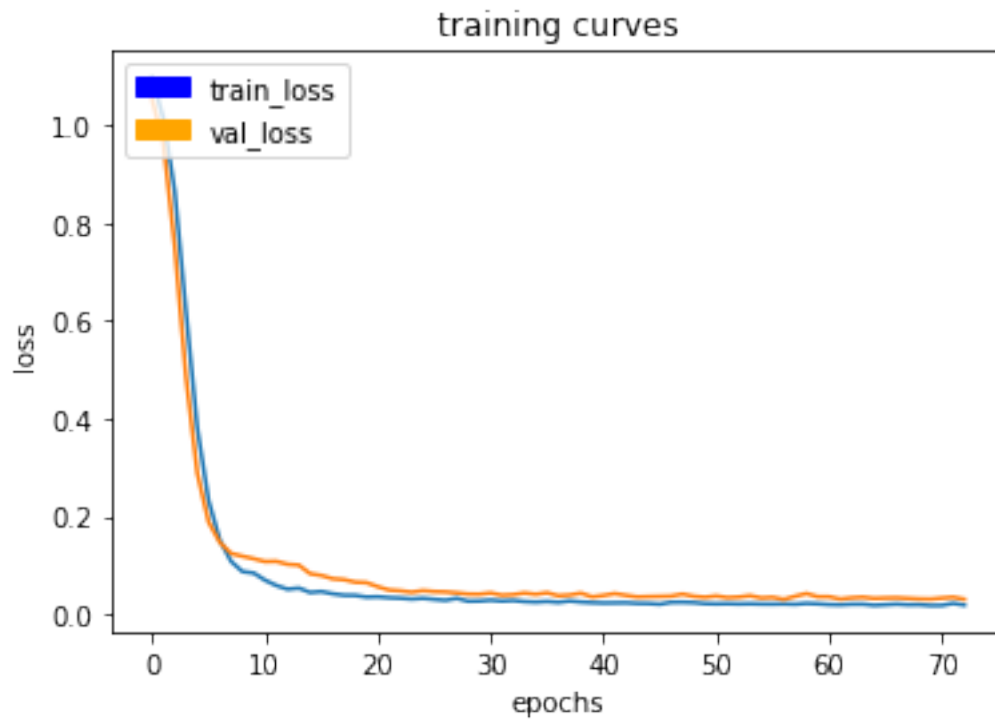


```

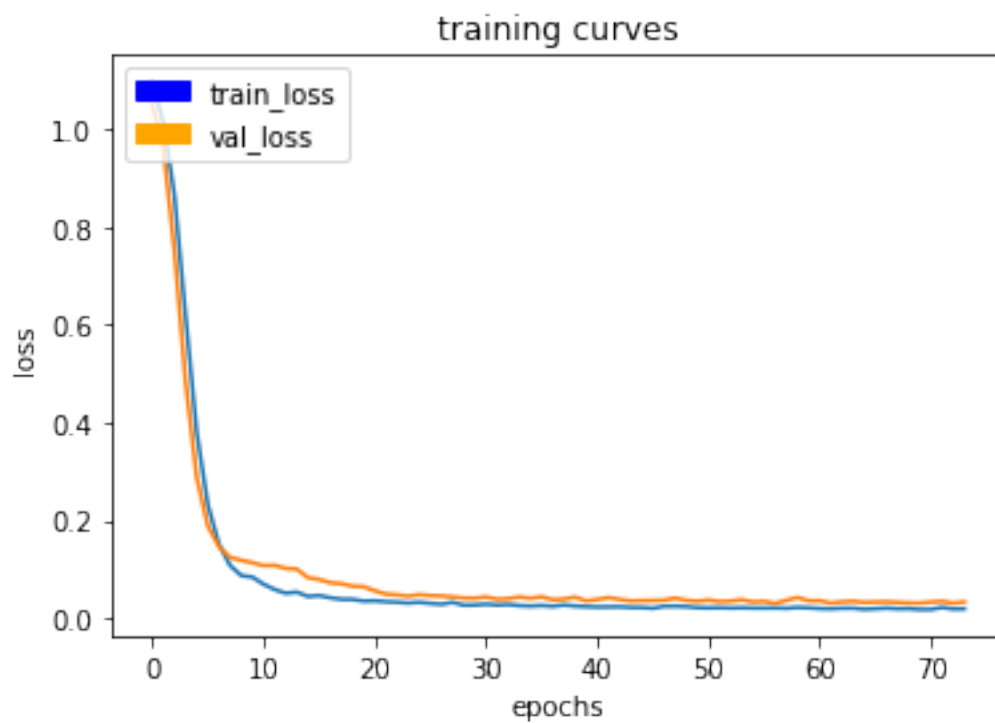
30/30 [=====] - 23s - loss: 0.0176 - val_loss: 0.0326
Epoch 72/200
29/30 [=====>.] - ETA: 0s - loss: 0.0219
  
```



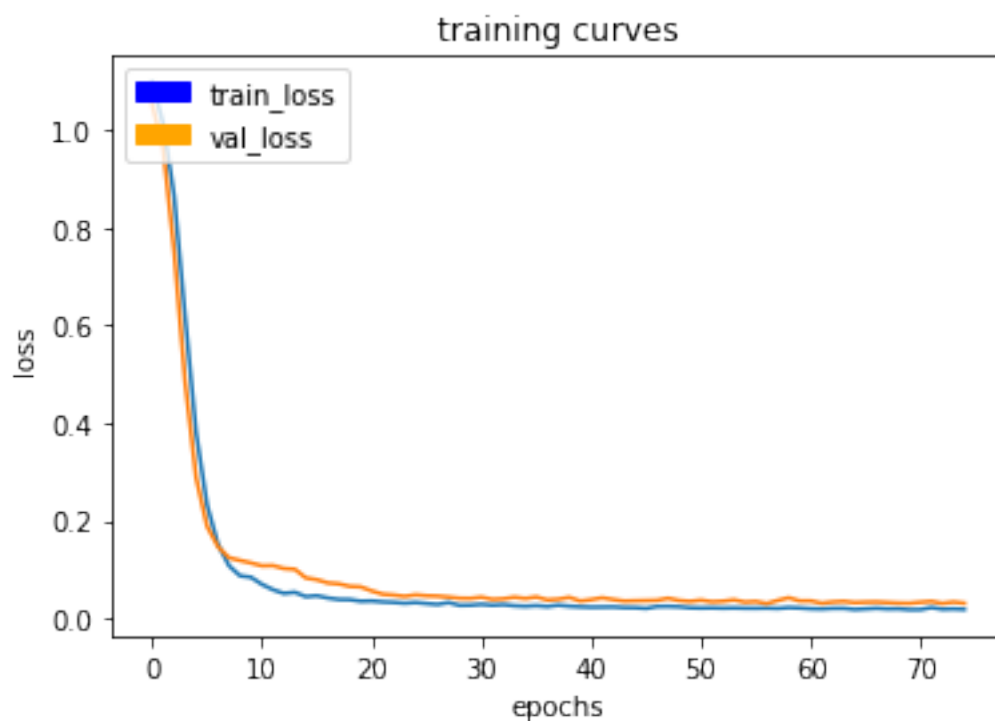
30/30 [=====] - 23s - loss: 0.0218 - val\_loss: 0.0343  
Epoch 73/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0184



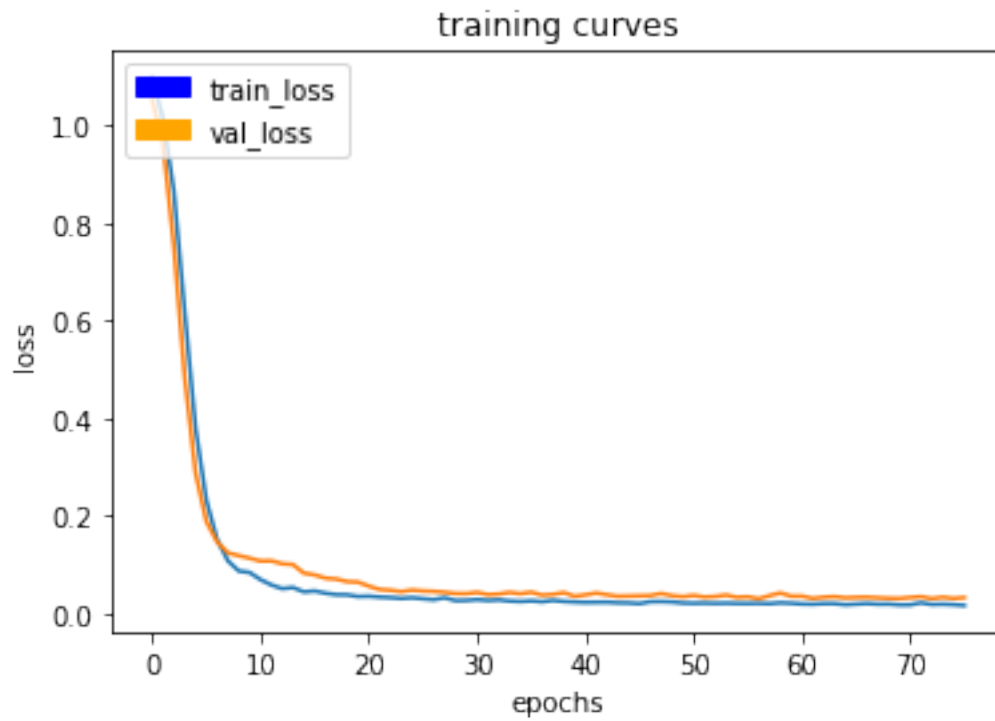
30/30 [=====] - 23s - loss: 0.0183 - val\_loss: 0.0303  
Epoch 74/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0189



30/30 [=====] - 23s - loss: 0.0189 - val\_loss: 0.0331  
 Epoch 75/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0180

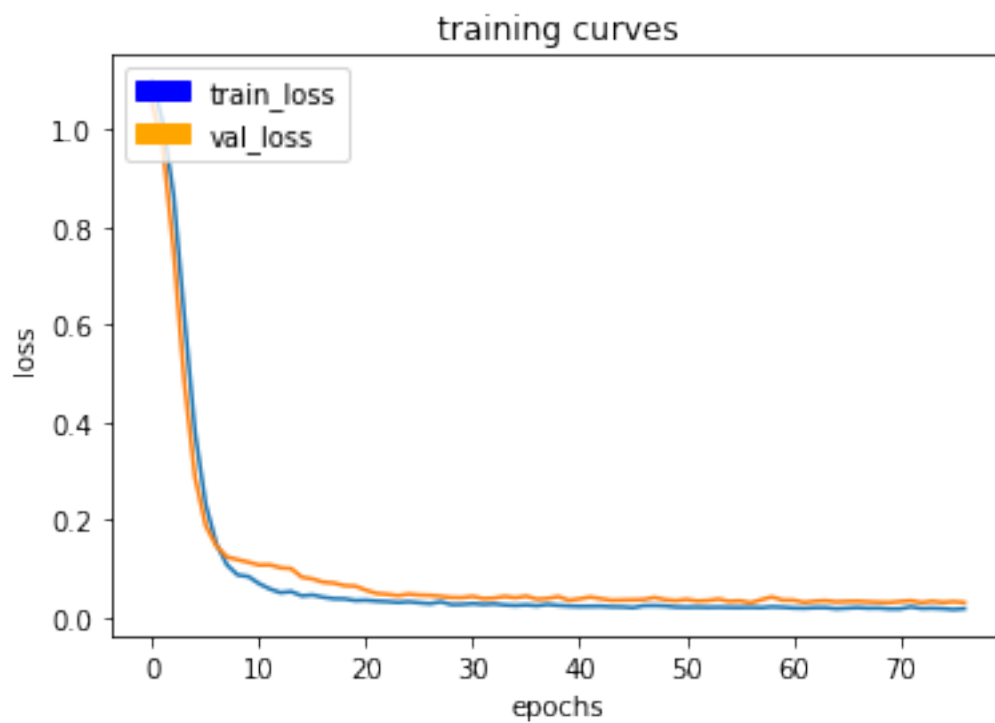


30/30 [=====] - 23s - loss: 0.0180 - val\_loss: 0.0307  
Epoch 76/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0163



30/30 [=====] - 23s - loss: 0.0163 - val\_loss: 0.0324  
Epoch 77/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0181

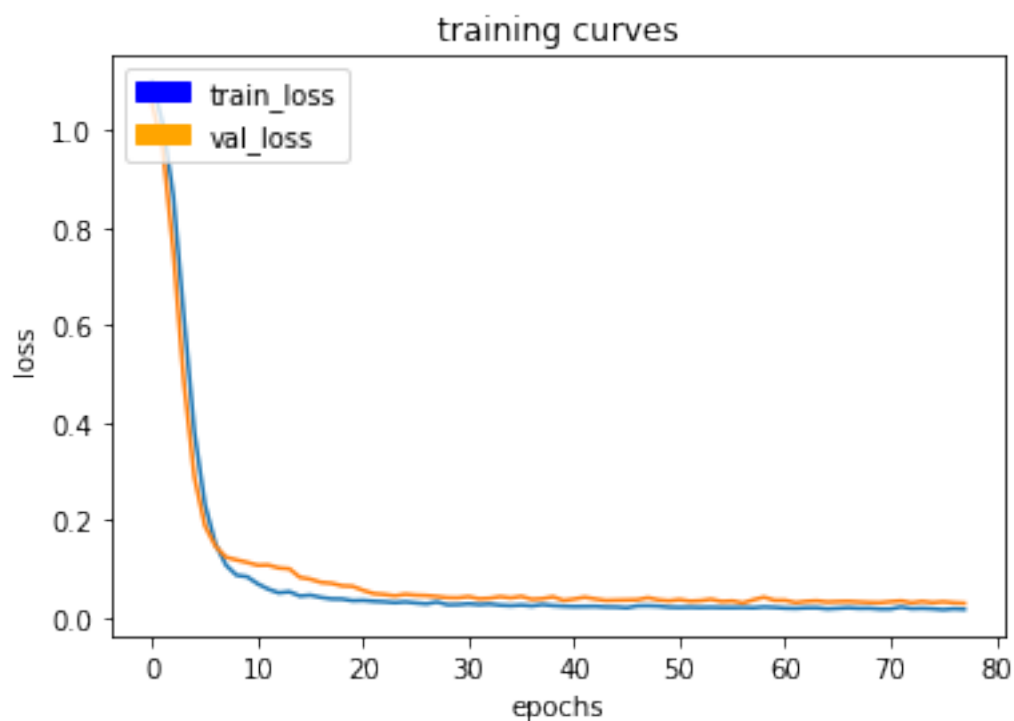




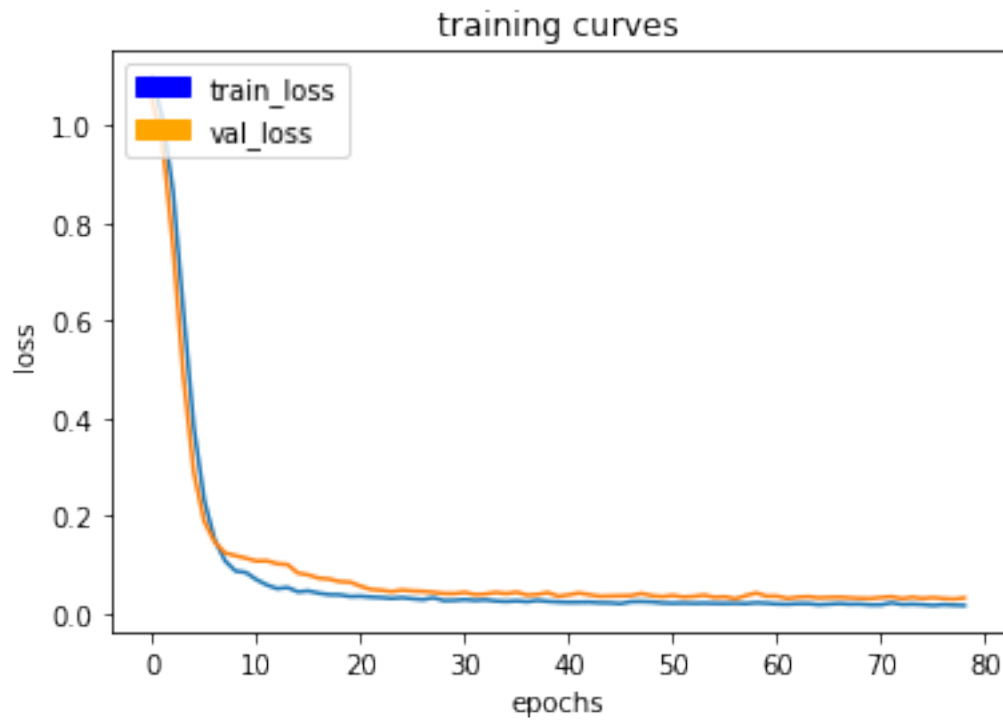
```

30/30 [=====] - 23s - loss: 0.0180 - val_loss: 0.0305
Epoch 78/200
29/30 [=====>.] - ETA: 0s - loss: 0.0171

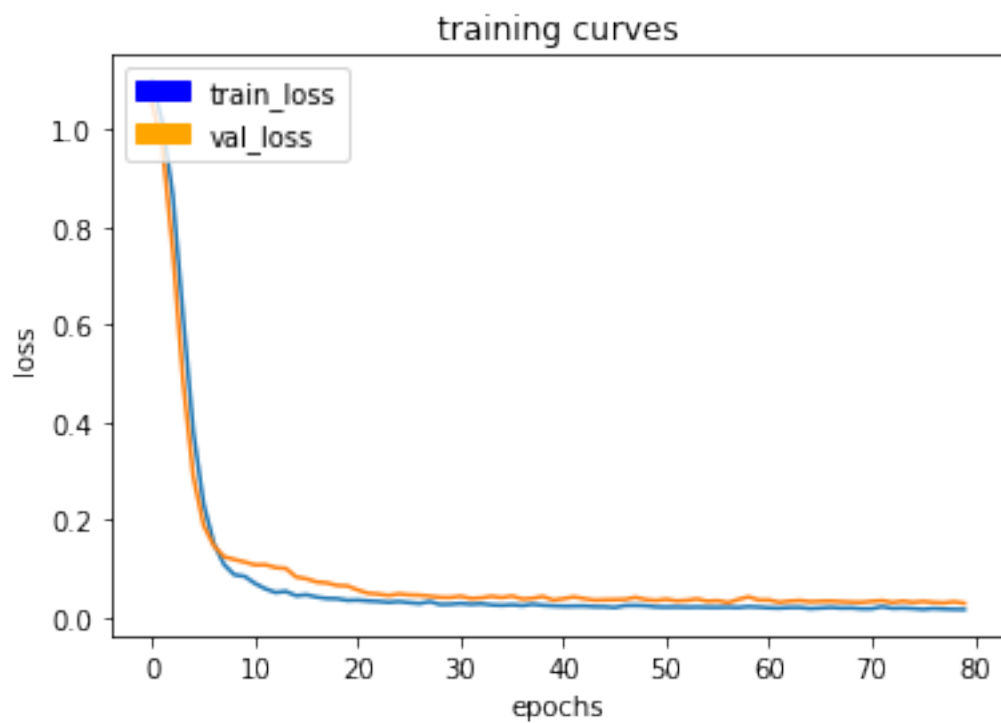
```



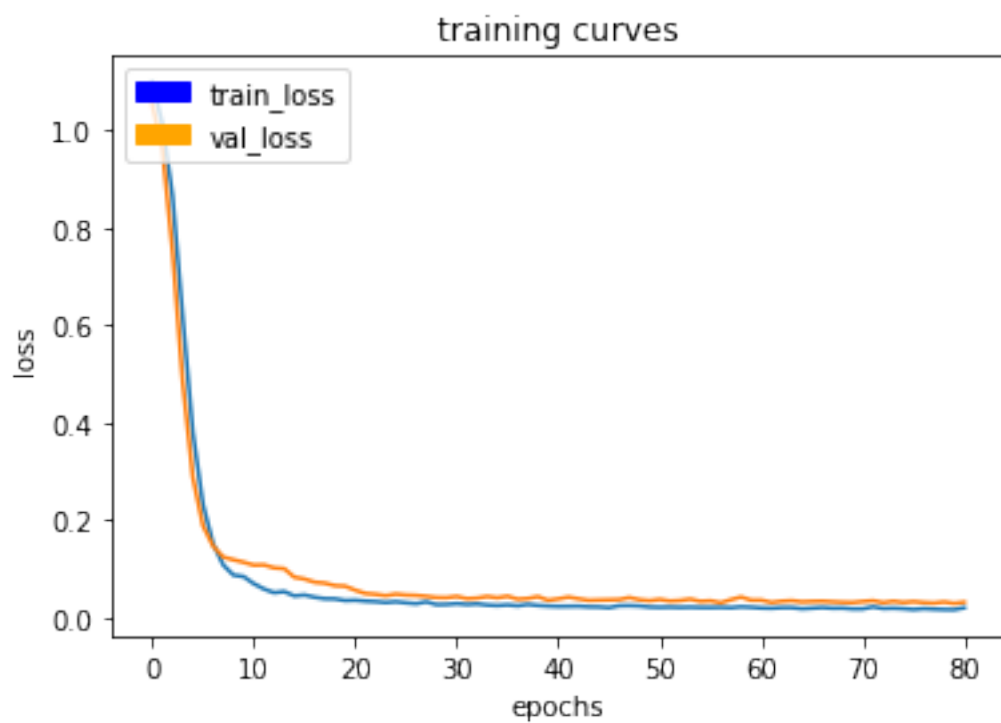
30/30 [=====] - 23s - loss: 0.0169 - val\_loss: 0.0294  
Epoch 79/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0163



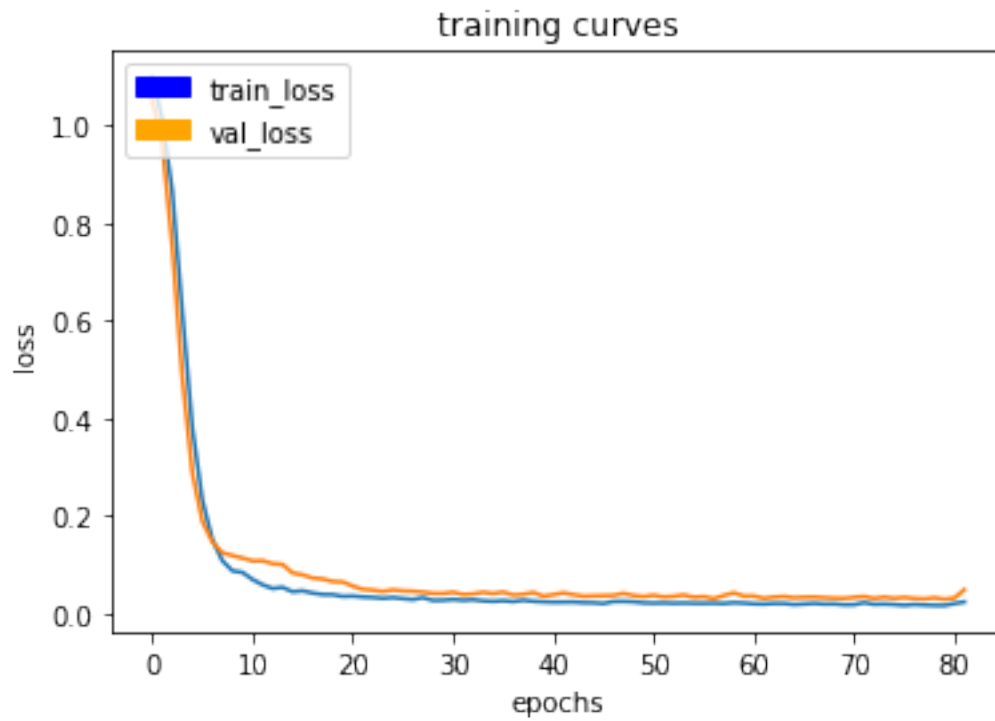
30/30 [=====] - 23s - loss: 0.0162 - val\_loss: 0.0317  
Epoch 80/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0160



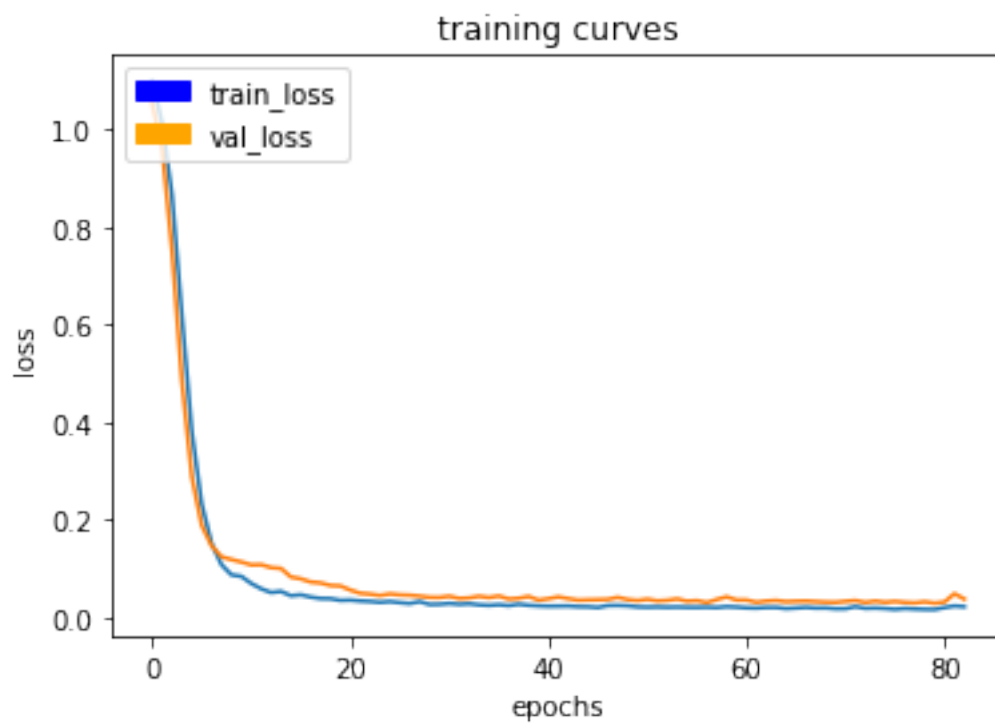
30/30 [=====] - 23s - loss: 0.0160 - val\_loss: 0.0288  
 Epoch 81/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0195



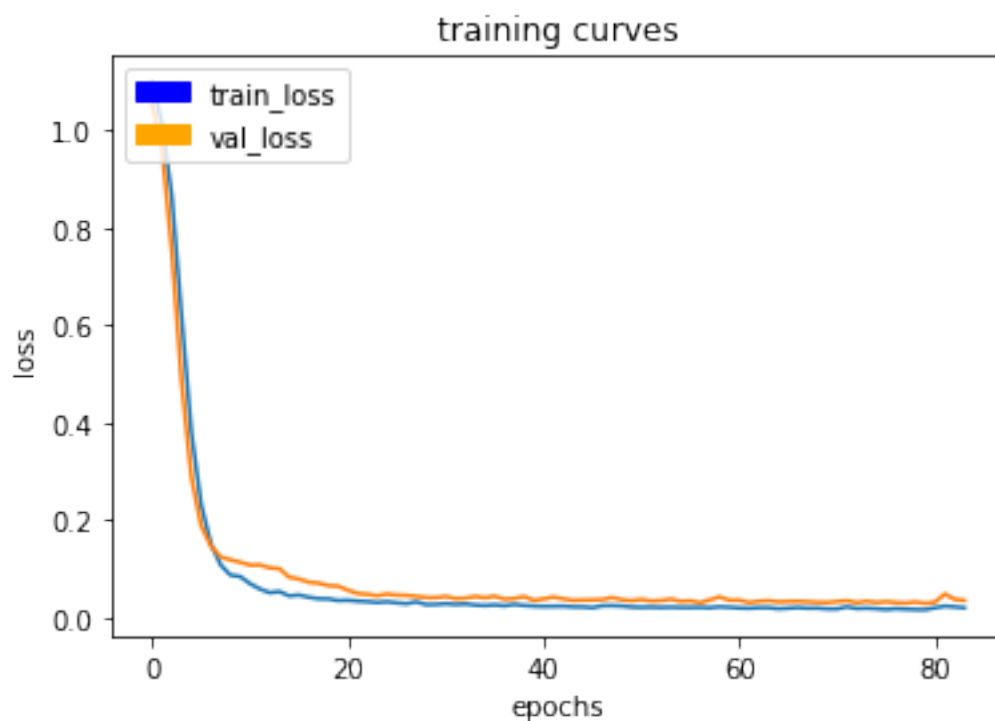
30/30 [=====] - 23s - loss: 0.0199 - val\_loss: 0.0313  
Epoch 82/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0227



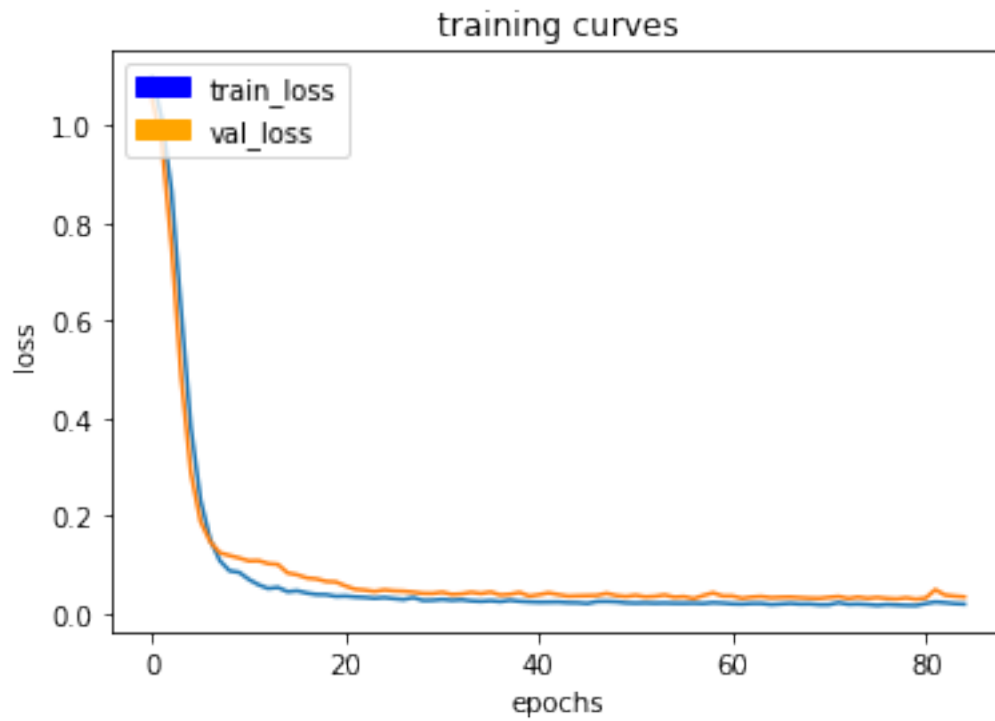
30/30 [=====] - 23s - loss: 0.0232 - val\_loss: 0.0483  
Epoch 83/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0217



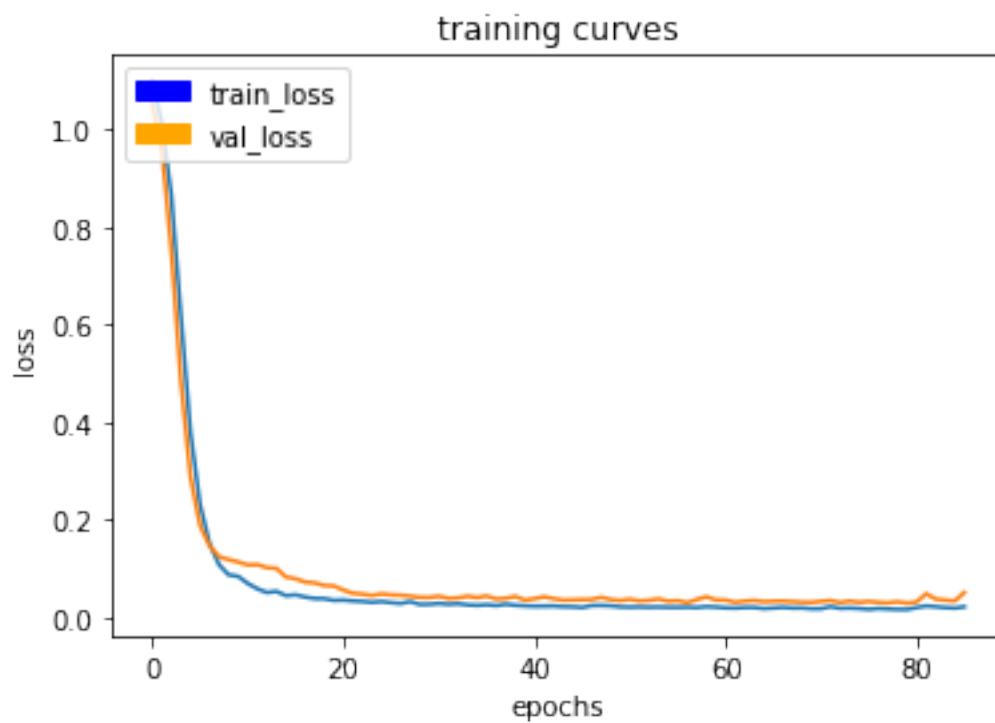
30/30 [=====] - 23s - loss: 0.0217 - val\_loss: 0.0372  
 Epoch 84/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0198



30/30 [=====] - 23s - loss: 0.0198 - val\_loss: 0.0351  
Epoch 85/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0189

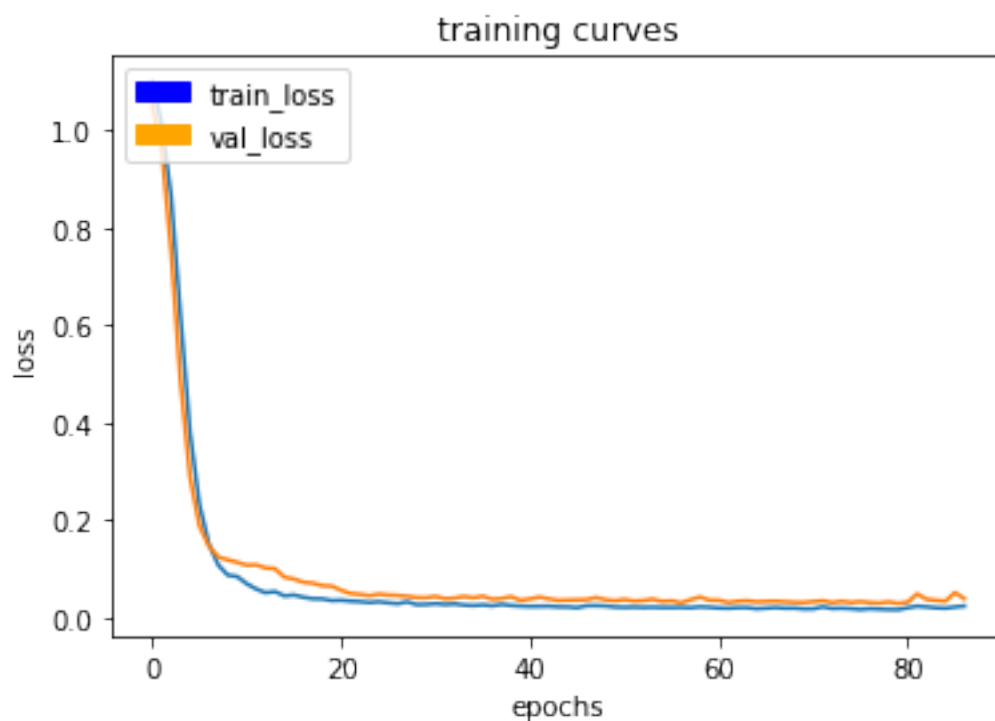


30/30 [=====] - 23s - loss: 0.0189 - val\_loss: 0.0335  
Epoch 86/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0208

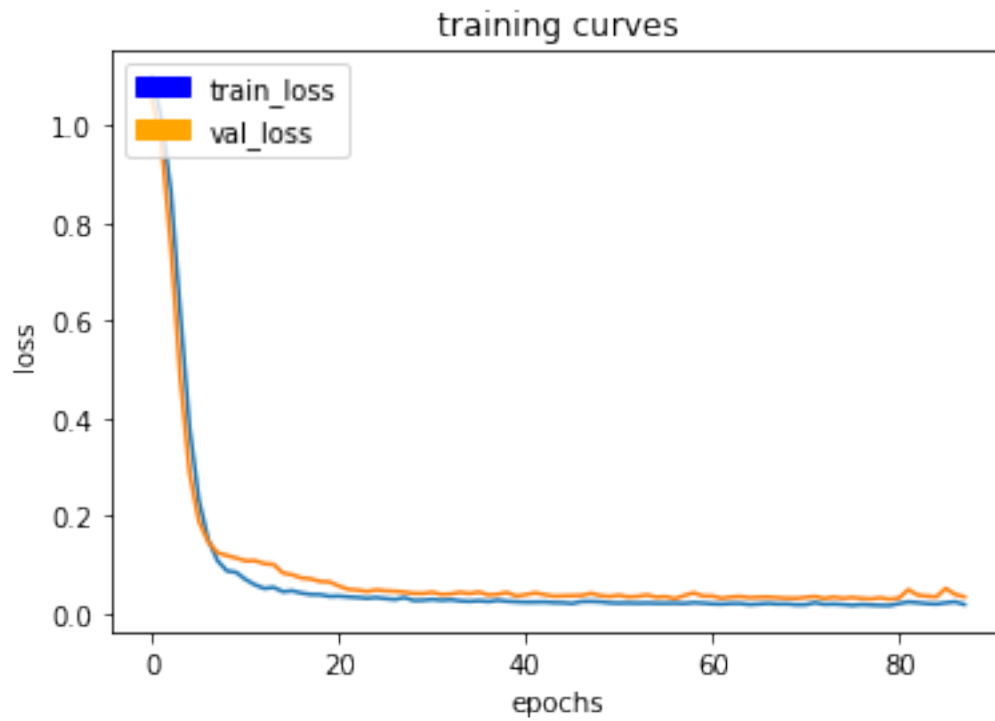


```

30/30 [=====] - 23s - loss: 0.0215 - val_loss: 0.0508
Epoch 87/200
29/30 [=====>.] - ETA: 0s - loss: 0.0236
  
```

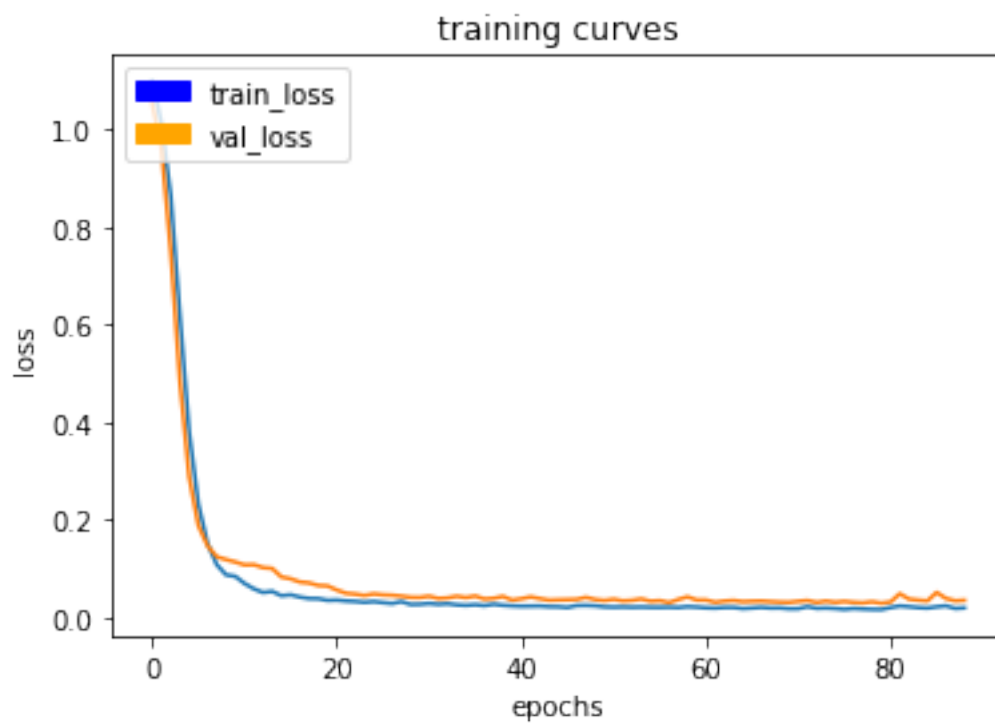


30/30 [=====] - 23s - loss: 0.0234 - val\_loss: 0.0386  
Epoch 88/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0180



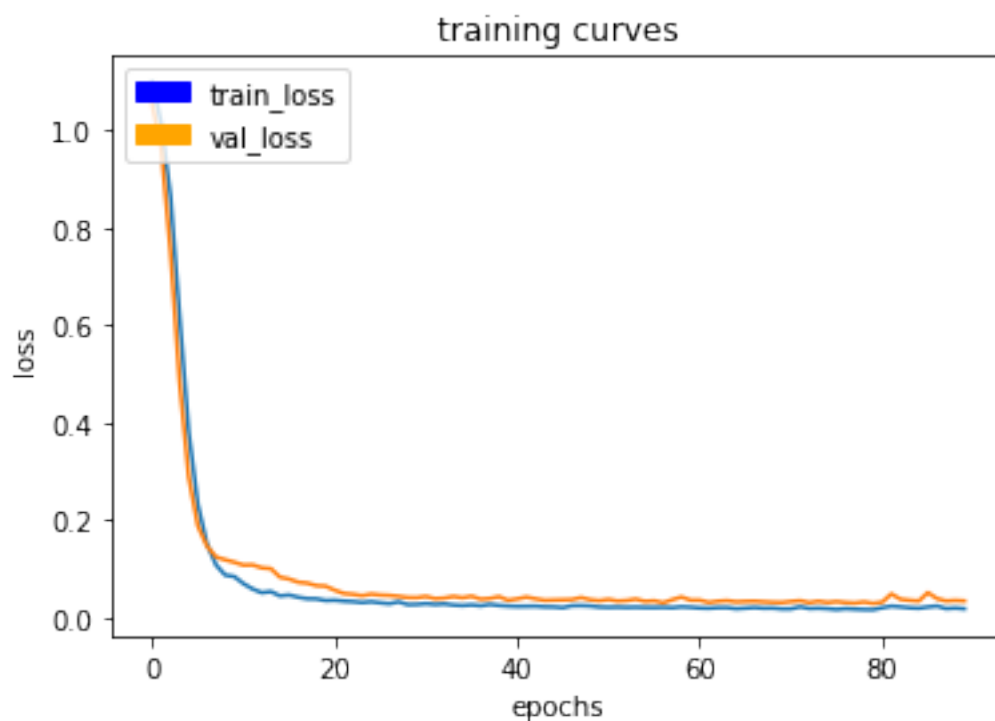
30/30 [=====] - 23s - loss: 0.0181 - val\_loss: 0.0337  
Epoch 89/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0194



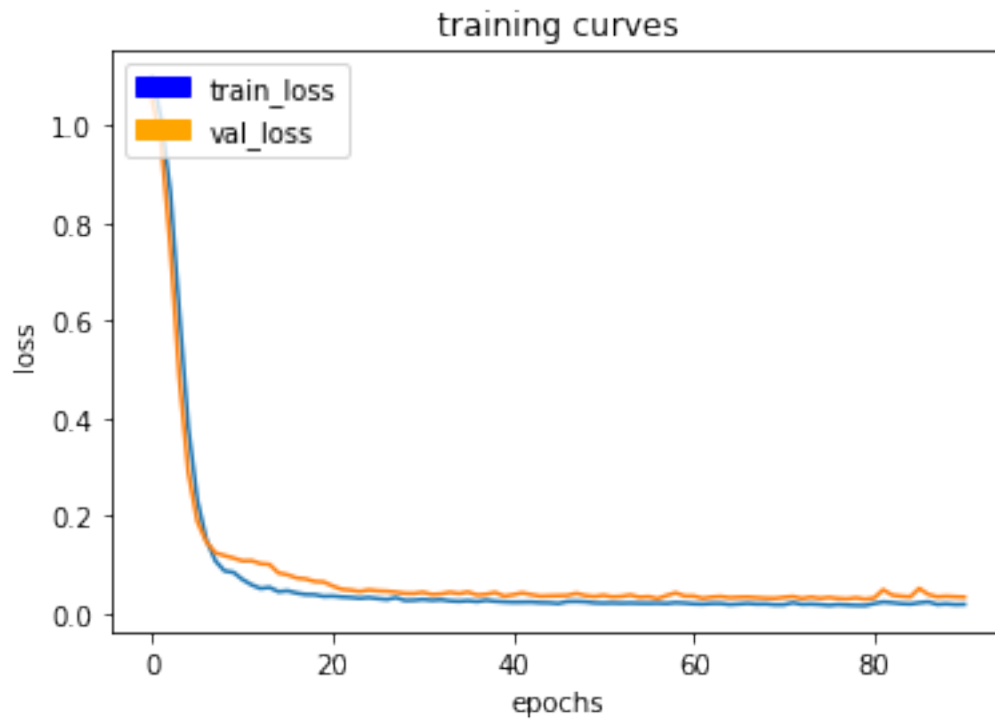


```

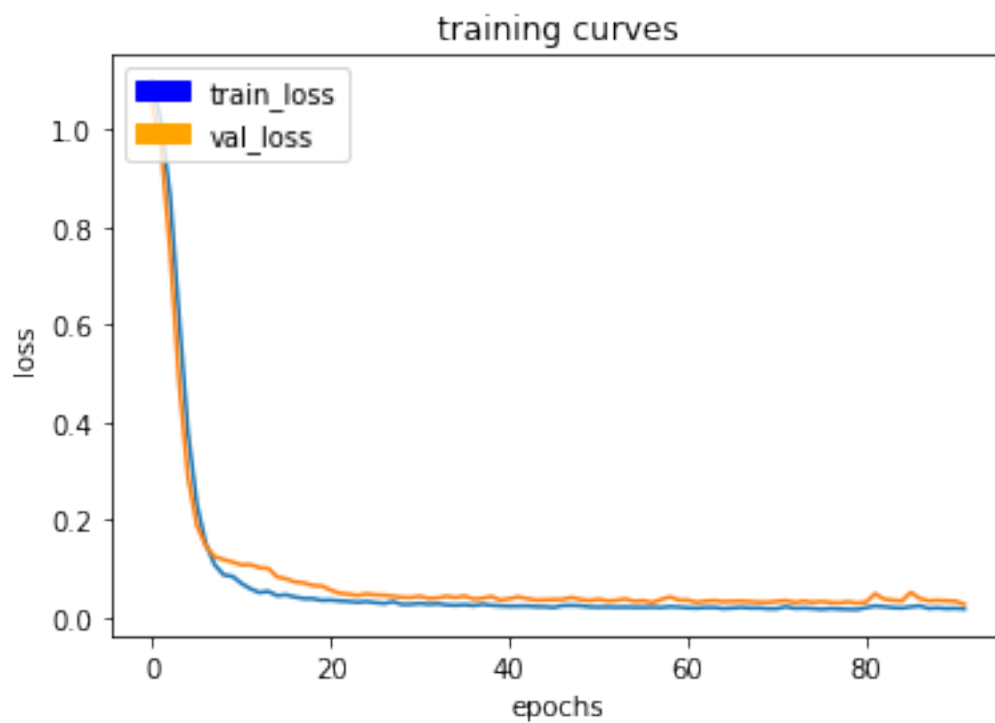
30/30 [=====] - 23s - loss: 0.0193 - val_loss: 0.0348
Epoch 90/200
29/30 [=====>.] - ETA: 0s - loss: 0.0178
  
```



30/30 [=====] - 23s - loss: 0.0178 - val\_loss: 0.0338  
Epoch 91/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0182

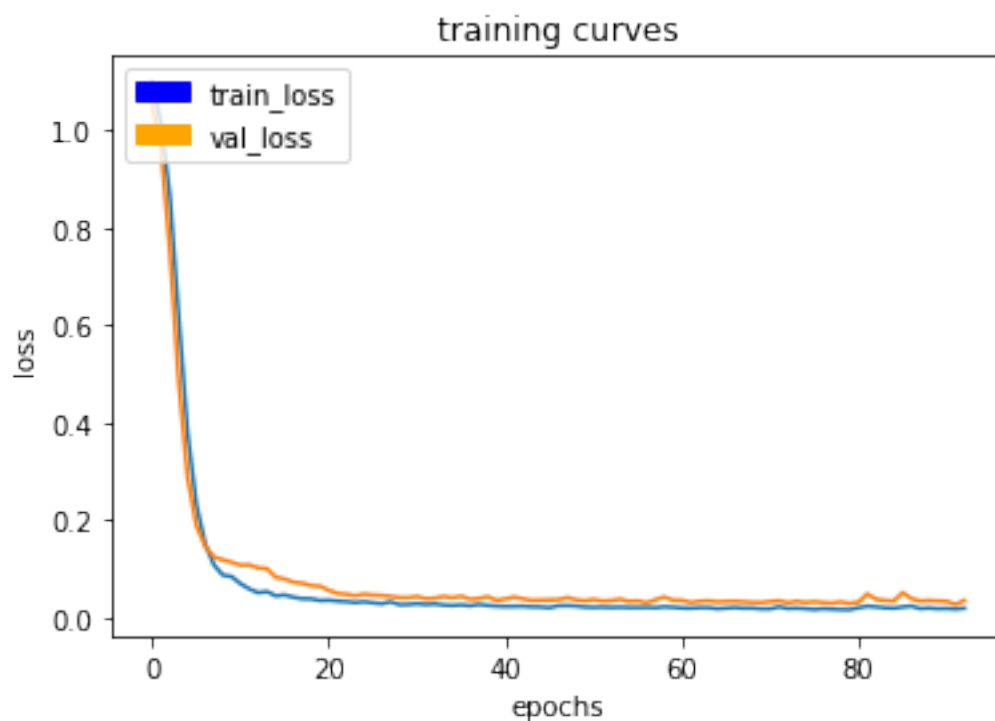


30/30 [=====] - 23s - loss: 0.0183 - val\_loss: 0.0331  
Epoch 92/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0172

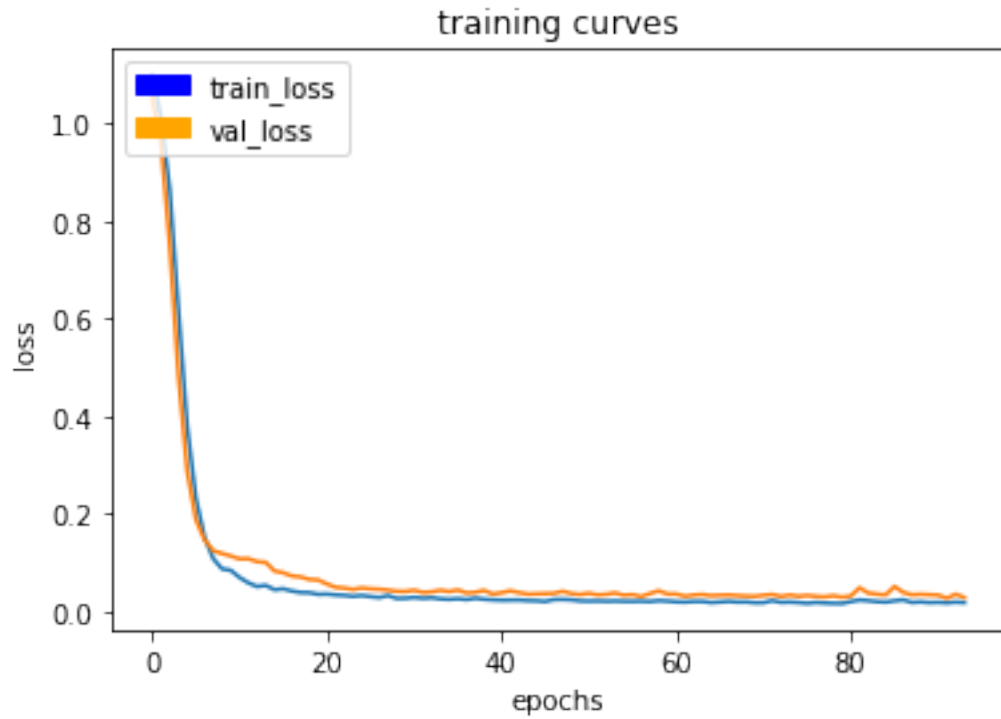


```

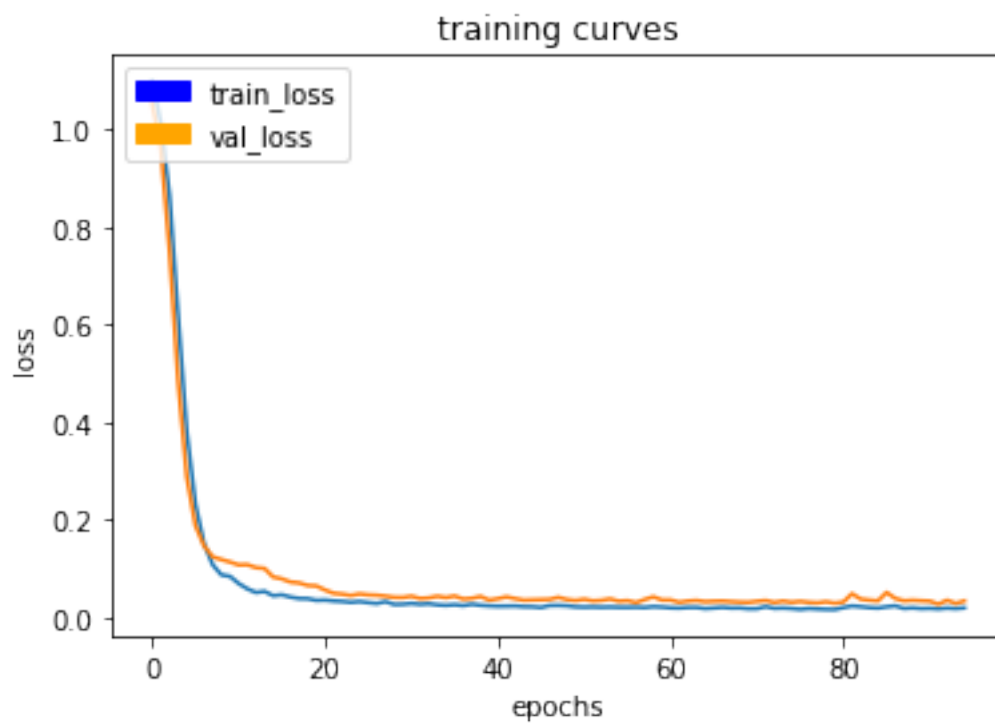
30/30 [=====] - 23s - loss: 0.0172 - val_loss: 0.0267
Epoch 93/200
29/30 [=====>.] - ETA: 0s - loss: 0.0190
  
```



30/30 [=====] - 23s - loss: 0.0189 - val\_loss: 0.0349  
Epoch 94/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0179



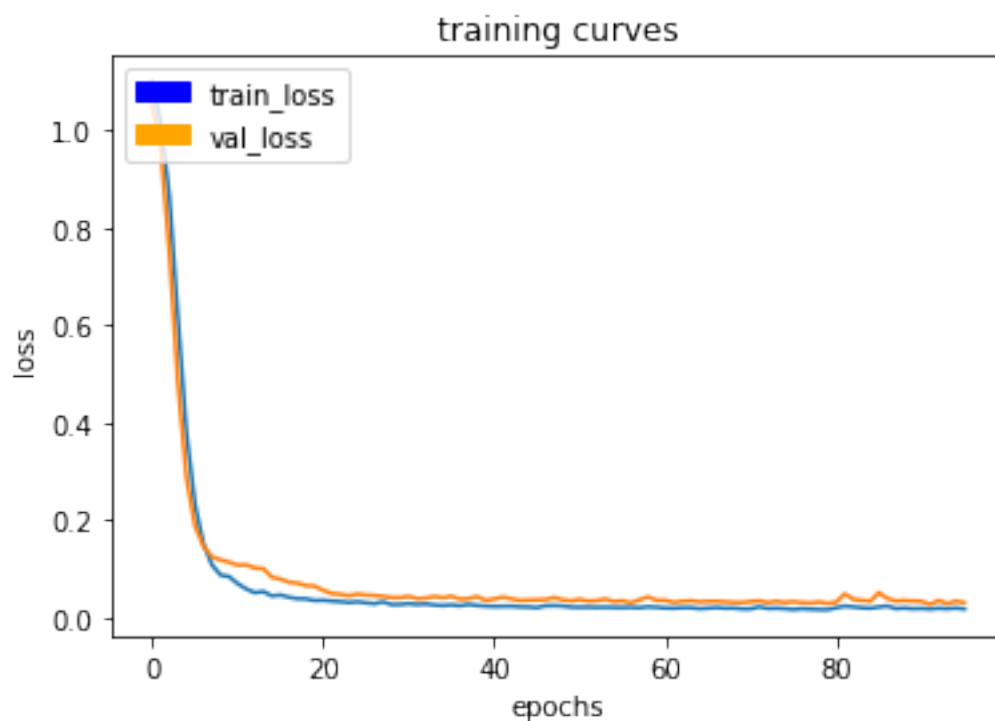
30/30 [=====] - 23s - loss: 0.0178 - val\_loss: 0.0282  
Epoch 95/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0192



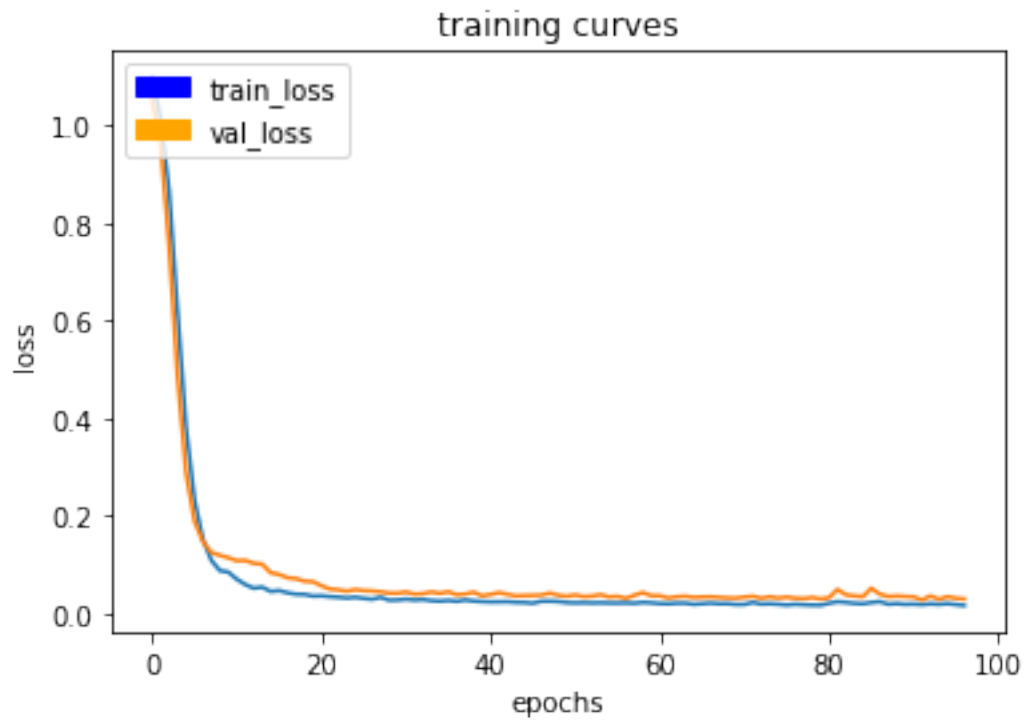
```

30/30 [=====] - 23s - loss: 0.0193 - val_loss: 0.0334
Epoch 96/200
29/30 [=====>.] - ETA: 0s - loss: 0.0173

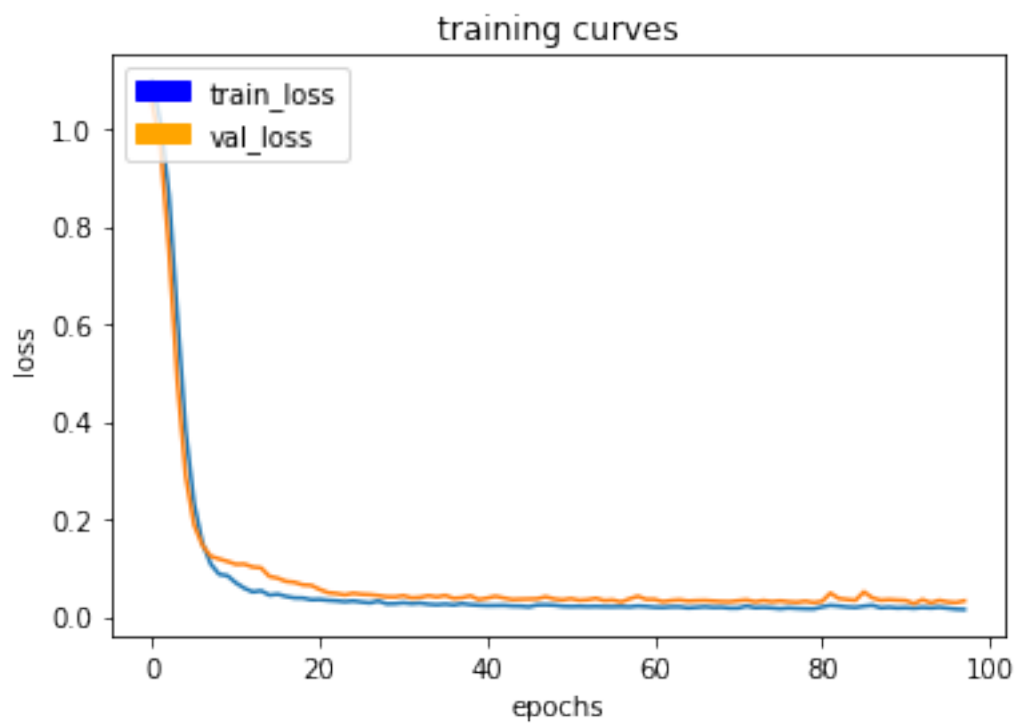
```



30/30 [=====] - 23s - loss: 0.0173 - val\_loss: 0.0302  
Epoch 97/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0154

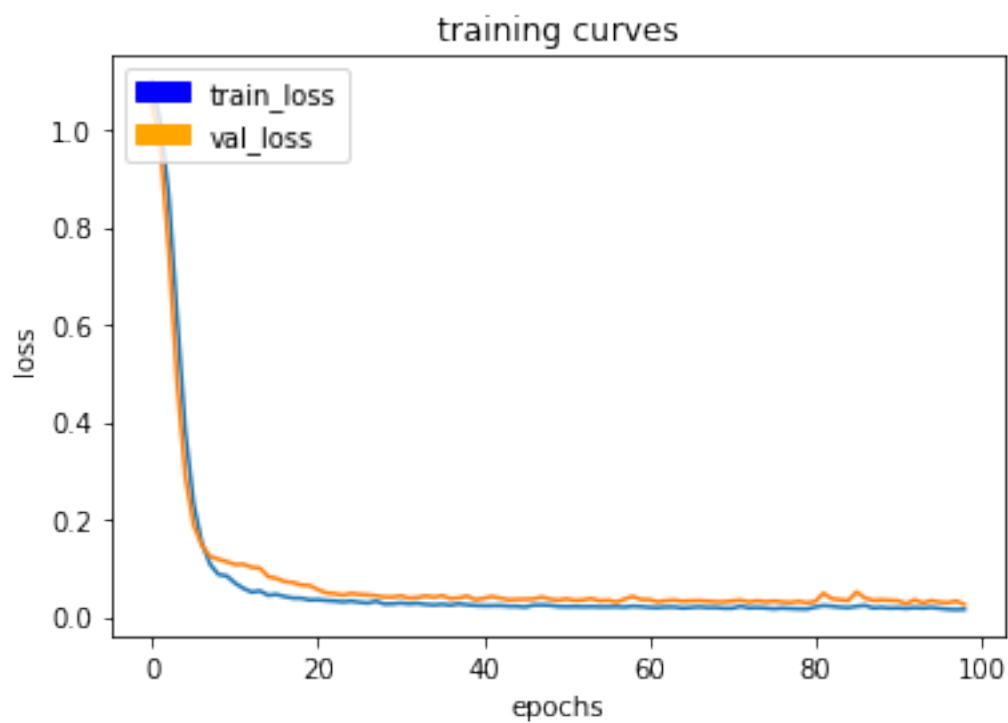


30/30 [=====] - 23s - loss: 0.0154 - val\_loss: 0.0287  
Epoch 98/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0146

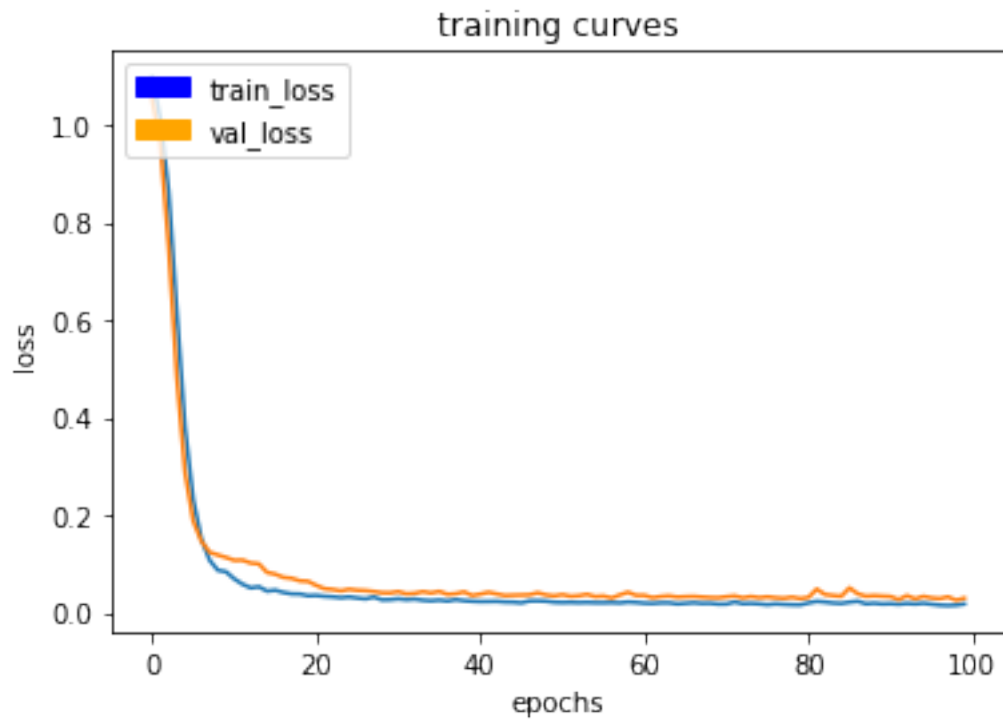


```

30/30 [=====] - 23s - loss: 0.0147 - val_loss: 0.0324
Epoch 99/200
29/30 [=====>.] - ETA: 0s - loss: 0.0154
  
```

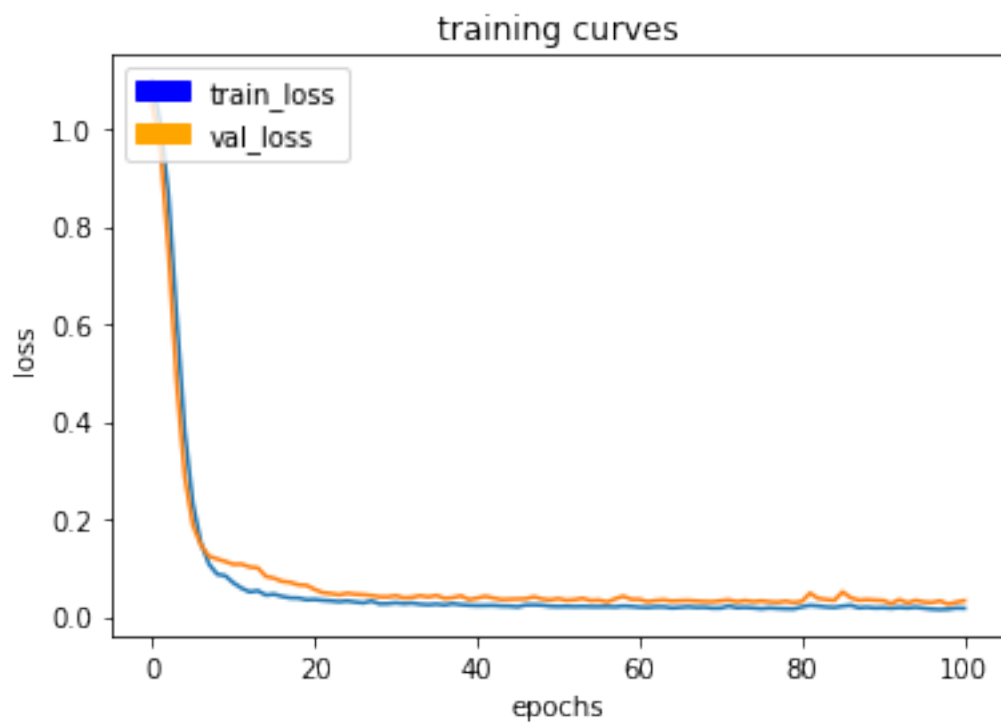


30/30 [=====] - 23s - loss: 0.0154 - val\_loss: 0.0254  
Epoch 100/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0178



30/30 [=====] - 23s - loss: 0.0178 - val\_loss: 0.0296  
Epoch 101/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0172

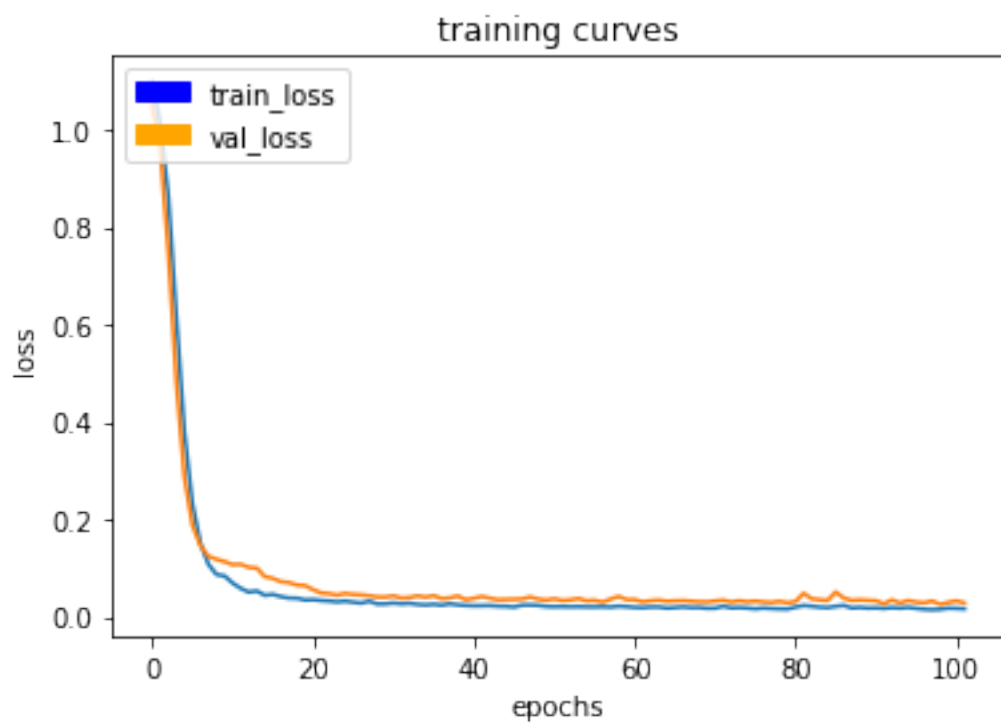




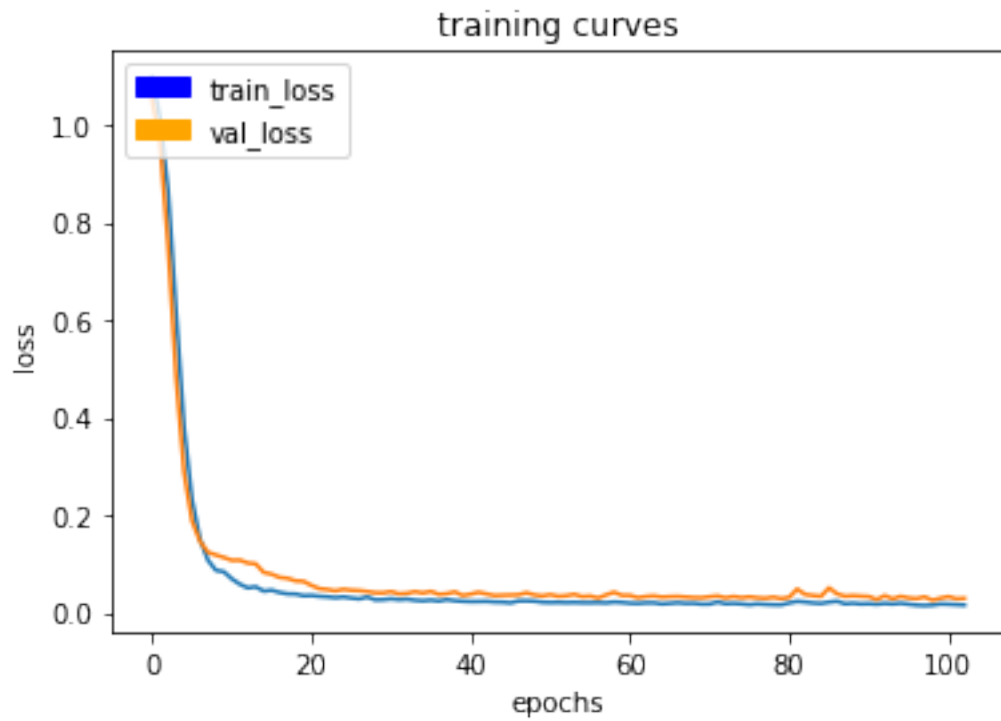
30/30 [=====] - 23s - loss: 0.0172 - val\_loss: 0.0328

Epoch 102/200

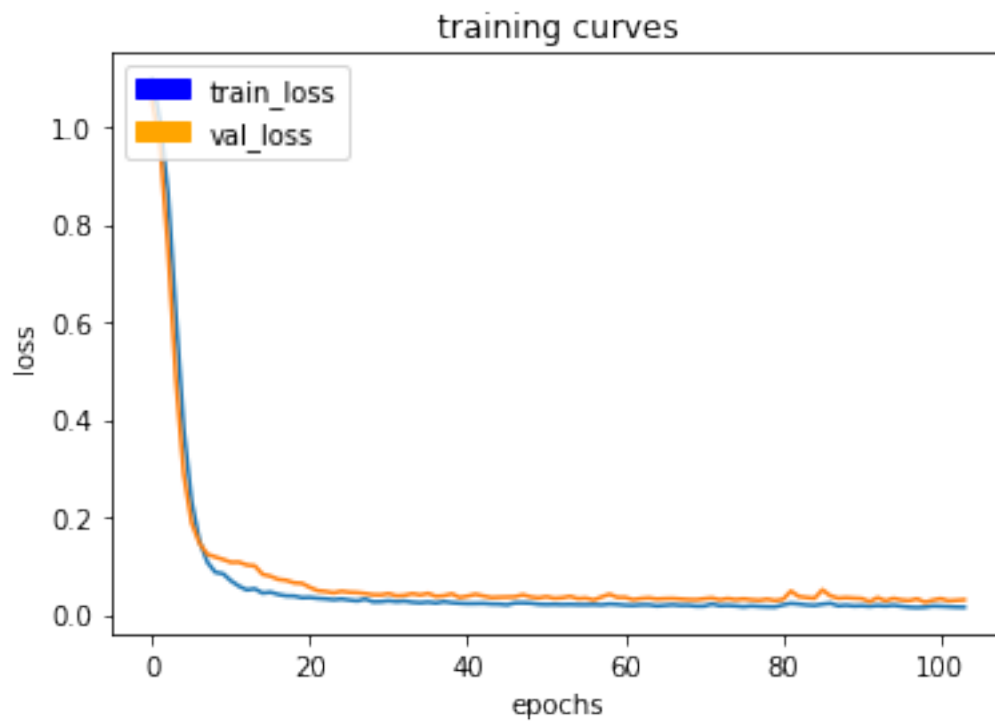
29/30 [=====>.] - ETA: 0s - loss: 0.0163



30/30 [=====] - 23s - loss: 0.0163 - val\_loss: 0.0284  
Epoch 103/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0155



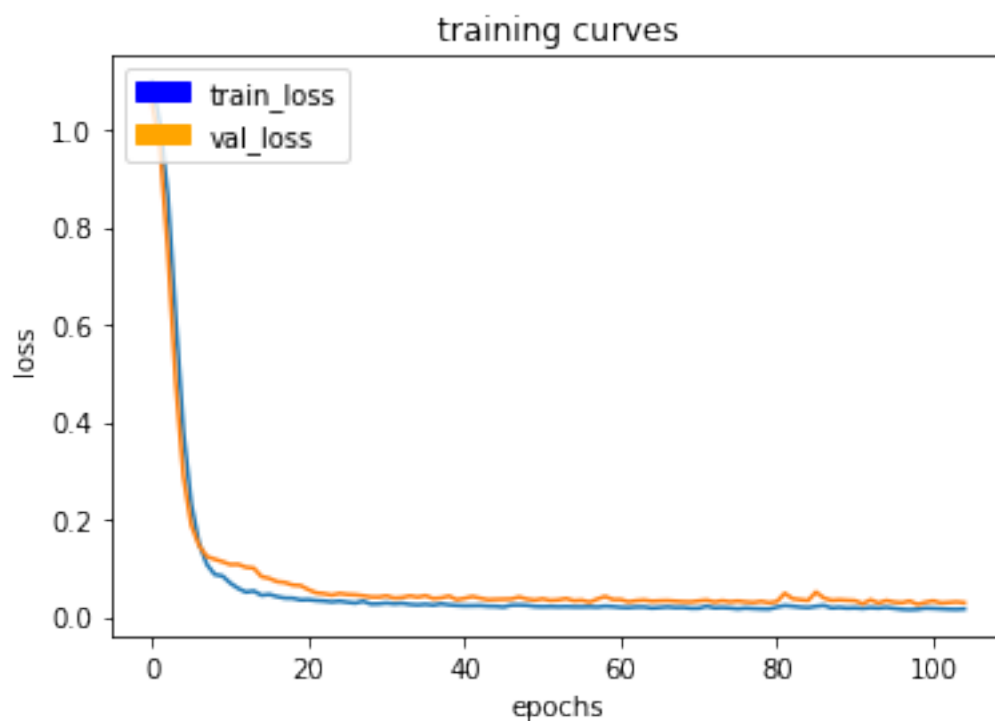
30/30 [=====] - 23s - loss: 0.0155 - val\_loss: 0.0293  
Epoch 104/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0151



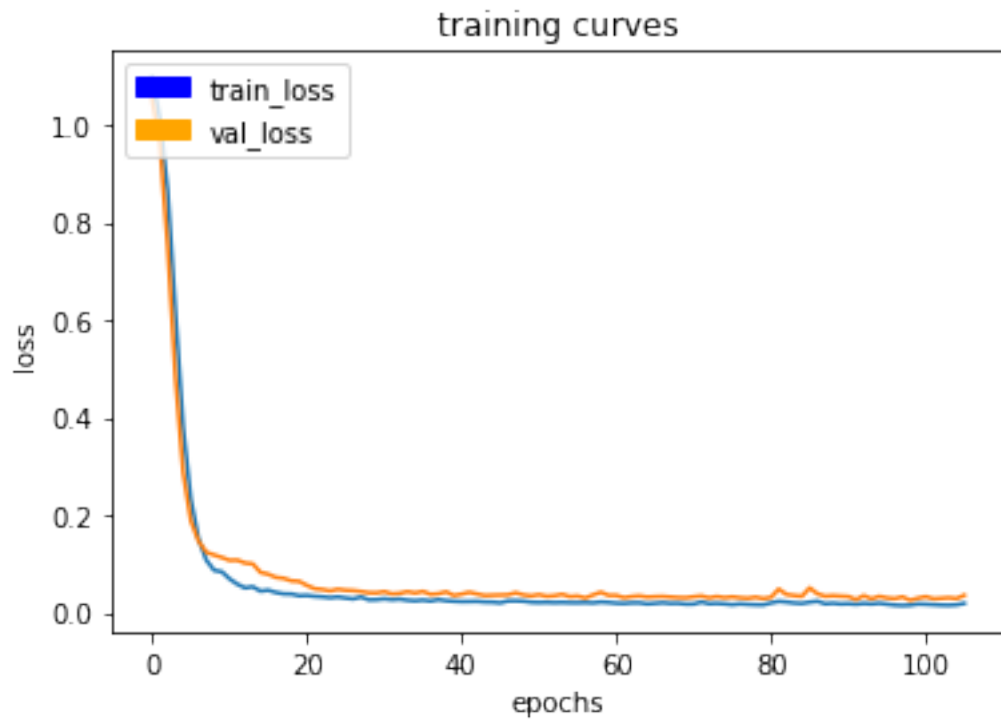
```

30/30 [=====] - 23s - loss: 0.0152 - val_loss: 0.0305
Epoch 105/200
29/30 [=====>.] - ETA: 0s - loss: 0.0161

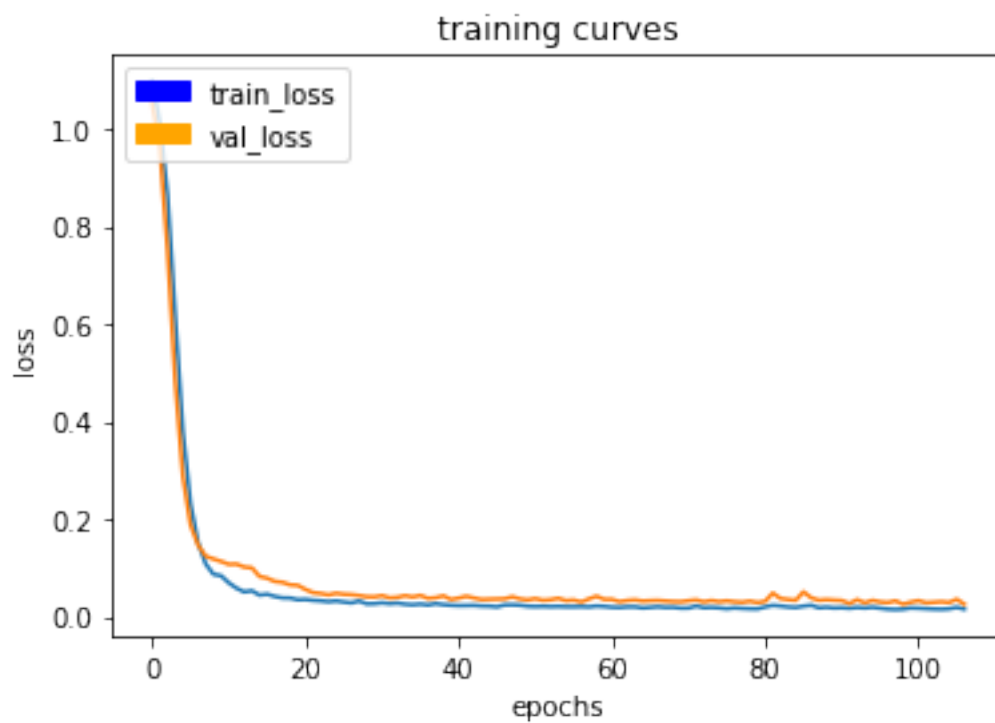
```



30/30 [=====] - 24s - loss: 0.0160 - val\_loss: 0.0289  
Epoch 106/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0191



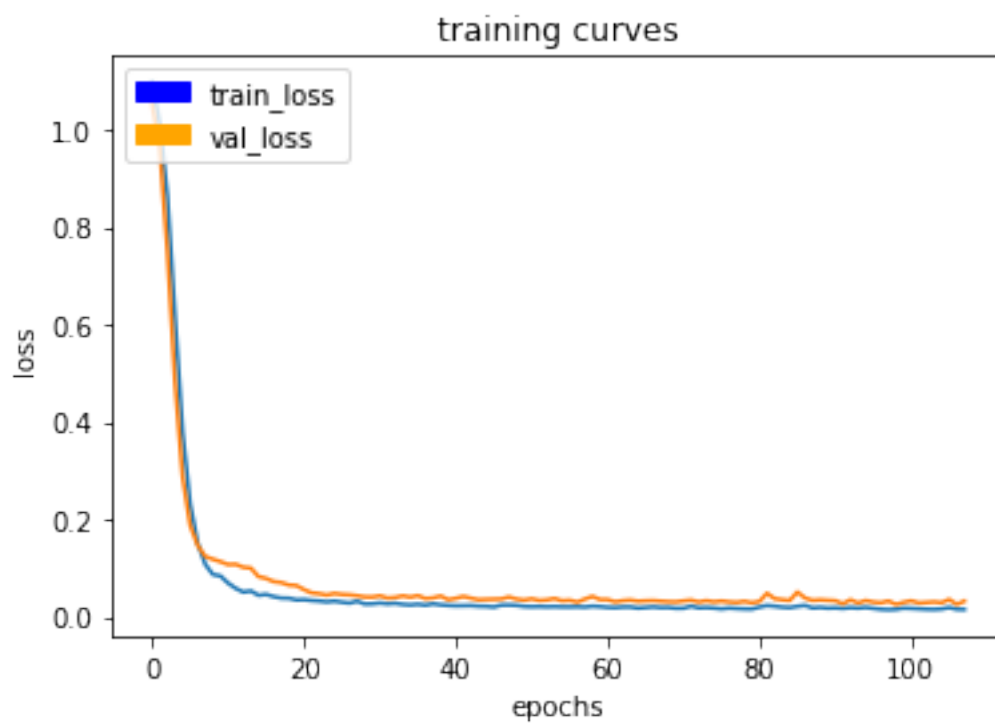
30/30 [=====] - 23s - loss: 0.0191 - val\_loss: 0.0357  
Epoch 107/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0155



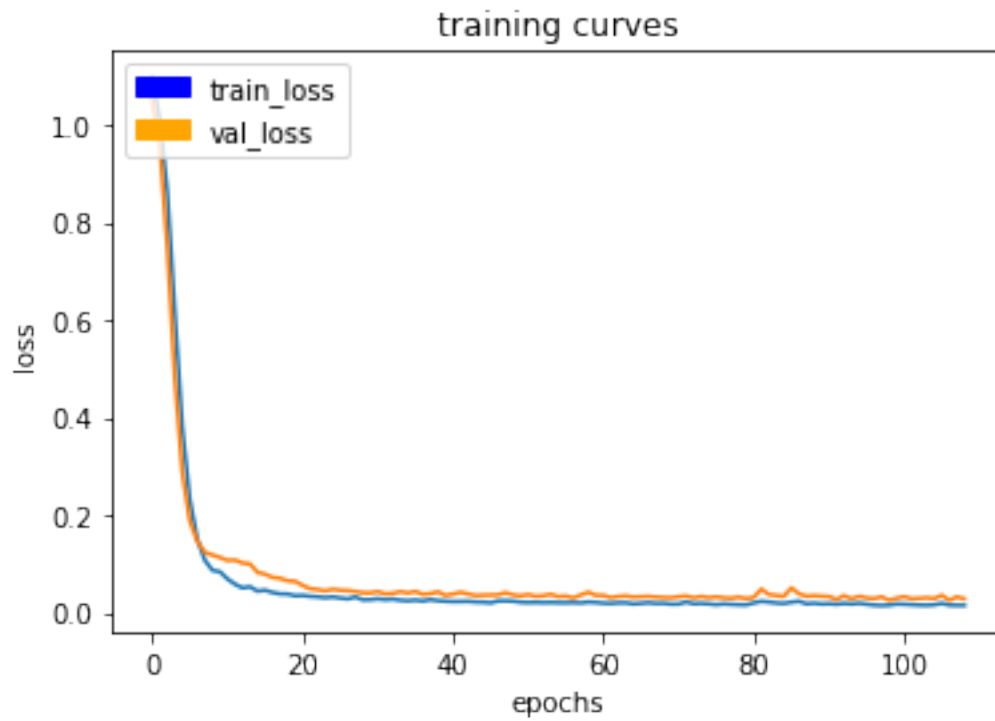
```

30/30 [=====] - 23s - loss: 0.0154 - val_loss: 0.0254
Epoch 108/200
29/30 [=====>.] - ETA: 0s - loss: 0.0151

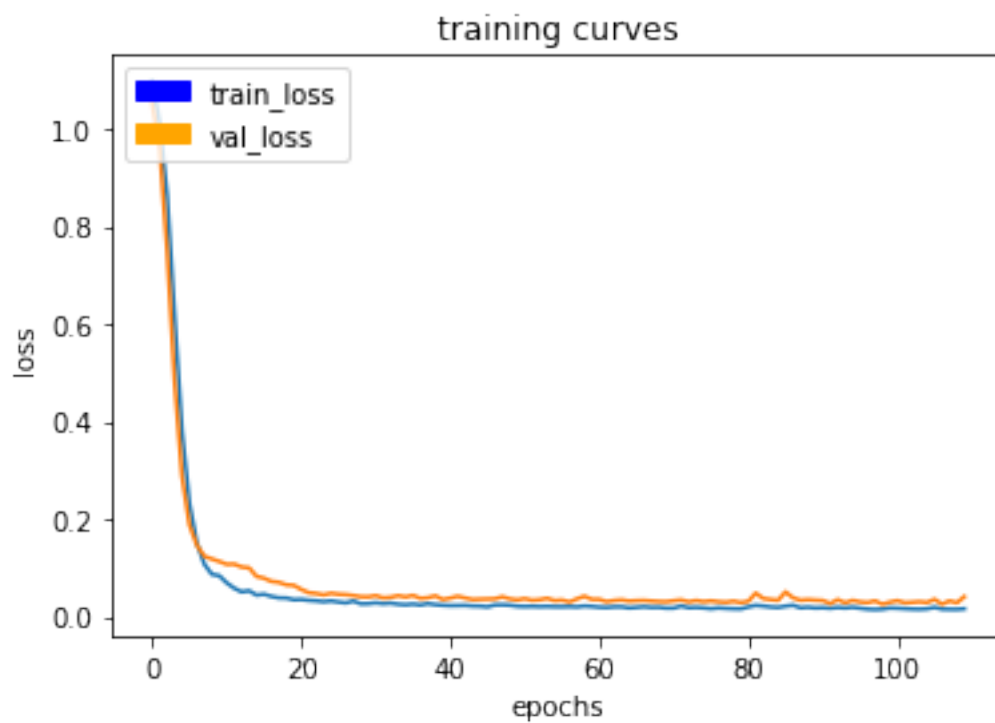
```



30/30 [=====] - 23s - loss: 0.0151 - val\_loss: 0.0322  
Epoch 109/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0149



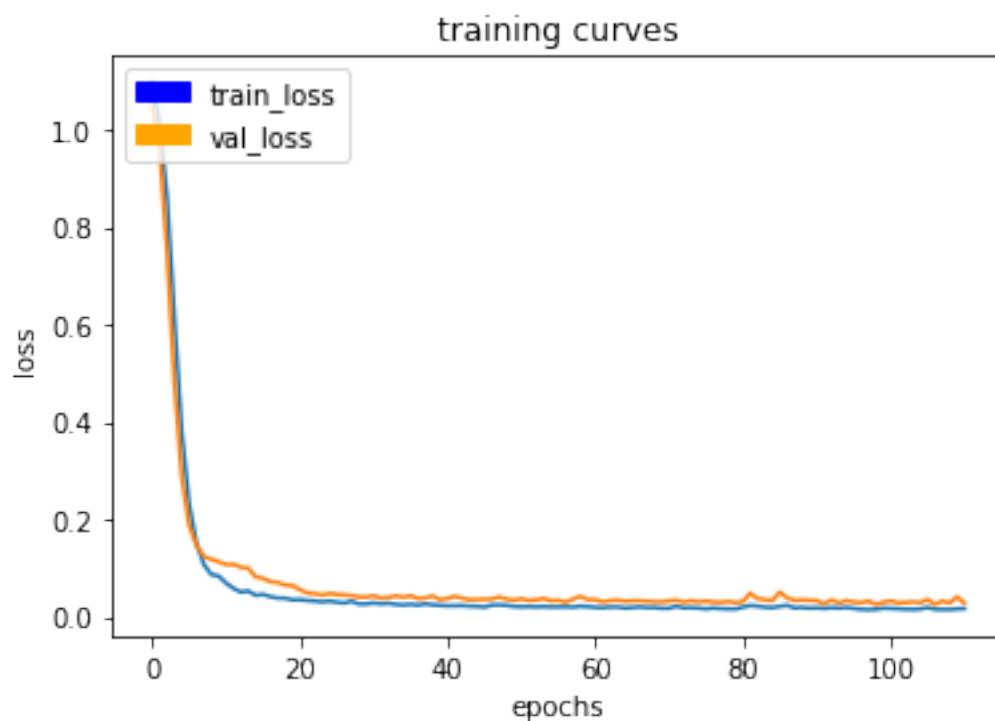
30/30 [=====] - 23s - loss: 0.0149 - val\_loss: 0.0284  
Epoch 110/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0160



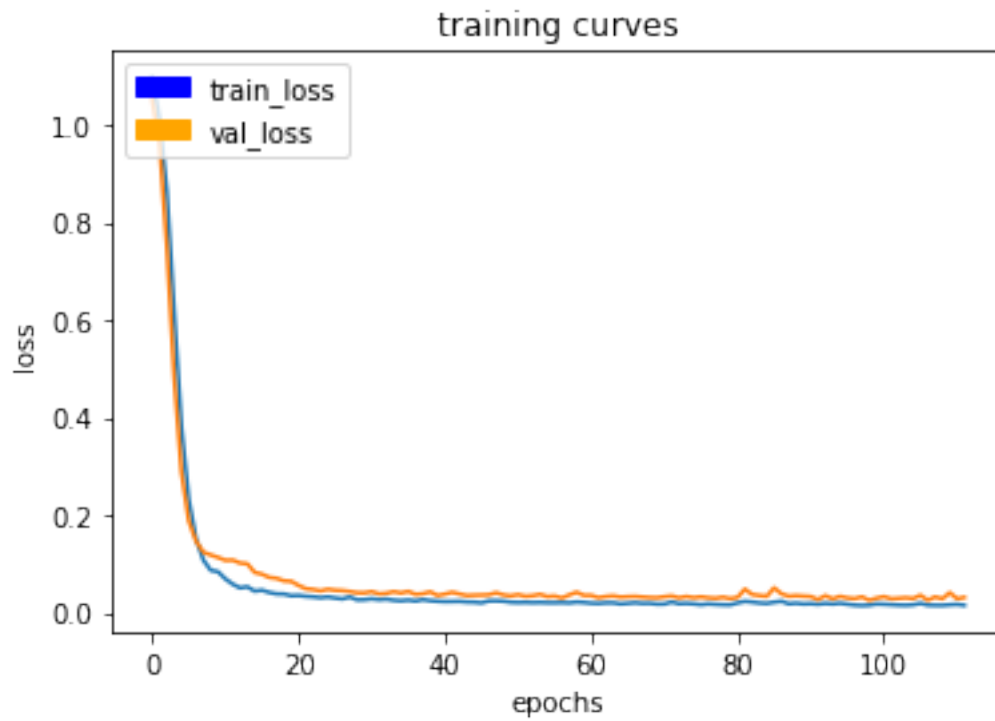
```

30/30 [=====] - 23s - loss: 0.0161 - val_loss: 0.0410
Epoch 111/200
29/30 [=====>.] - ETA: 0s - loss: 0.0168

```

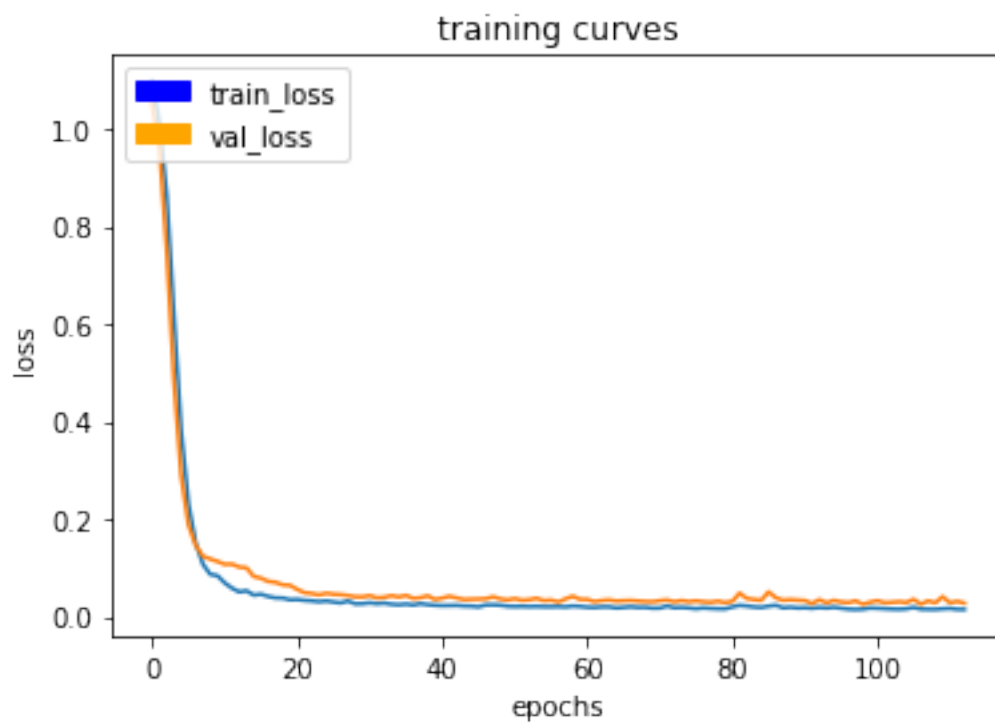


30/30 [=====] - 23s - loss: 0.0169 - val\_loss: 0.0285  
Epoch 112/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0150



30/30 [=====] - 23s - loss: 0.0150 - val\_loss: 0.0313  
Epoch 113/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0148

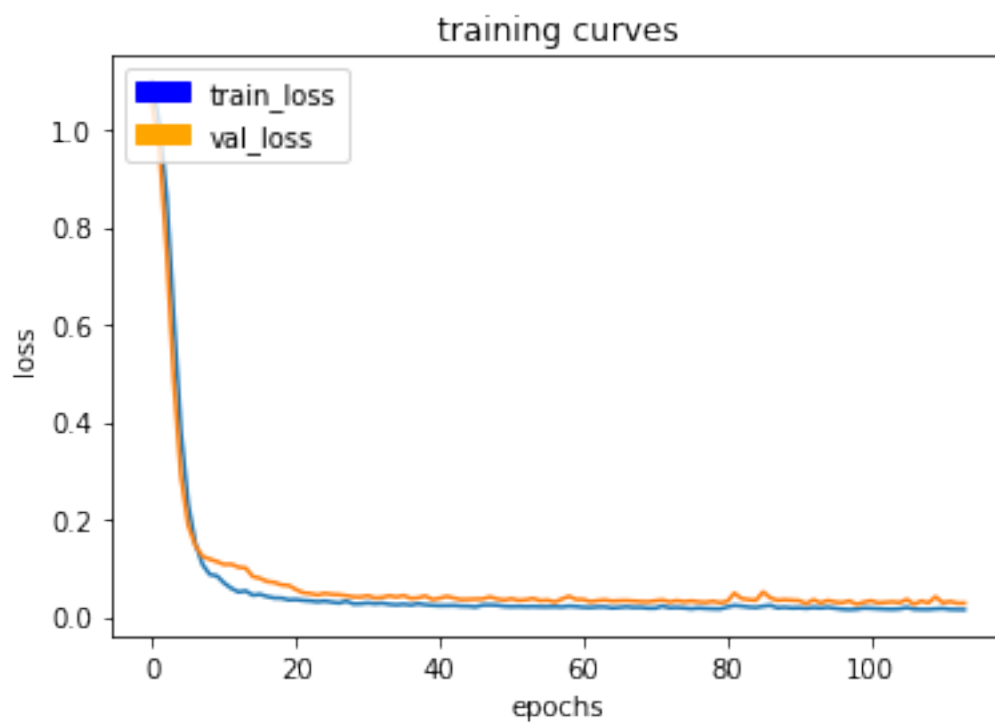




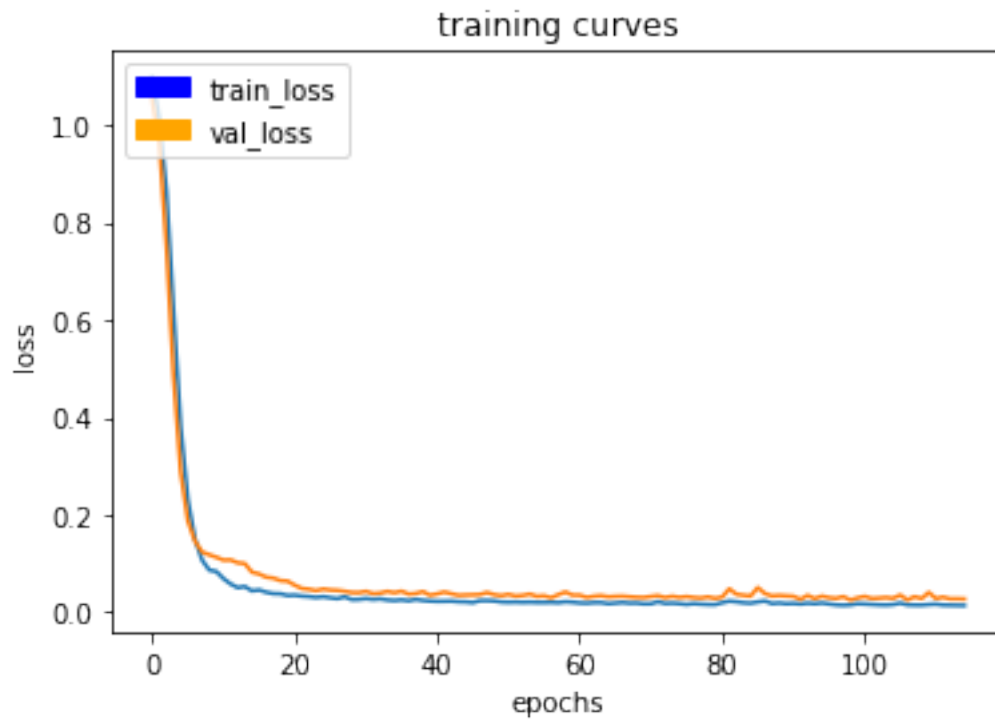
30/30 [=====] - 23s - loss: 0.0147 - val\_loss: 0.0278

Epoch 114/200

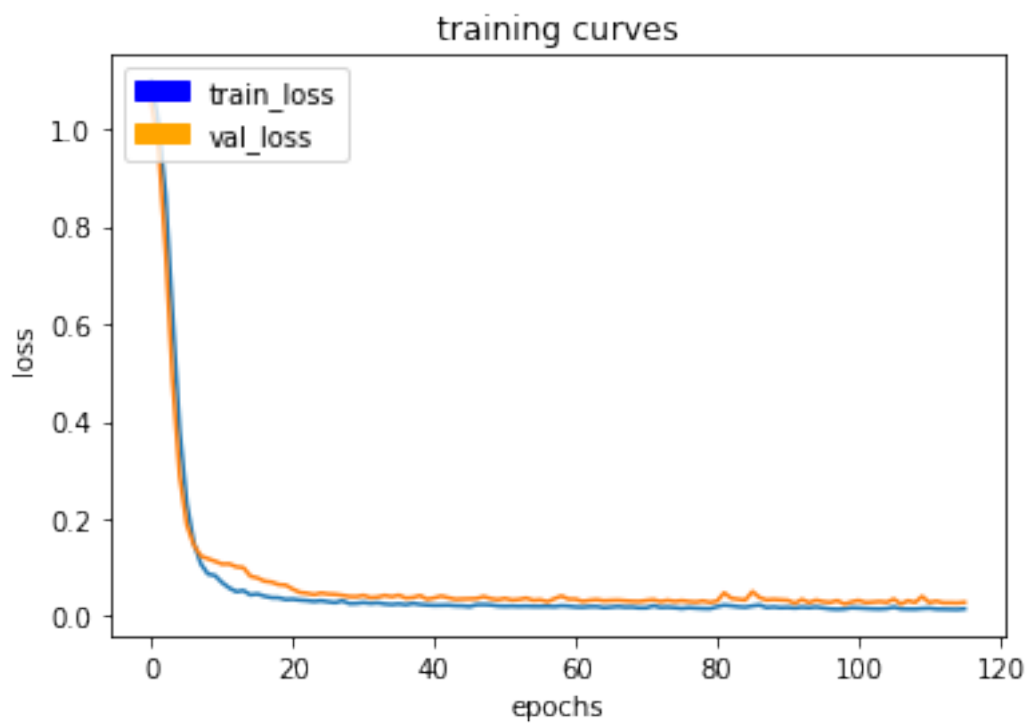
29/30 [=====>.] - ETA: 0s - loss: 0.0146



30/30 [=====] - 23s - loss: 0.0145 - val\_loss: 0.0279  
Epoch 115/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0141



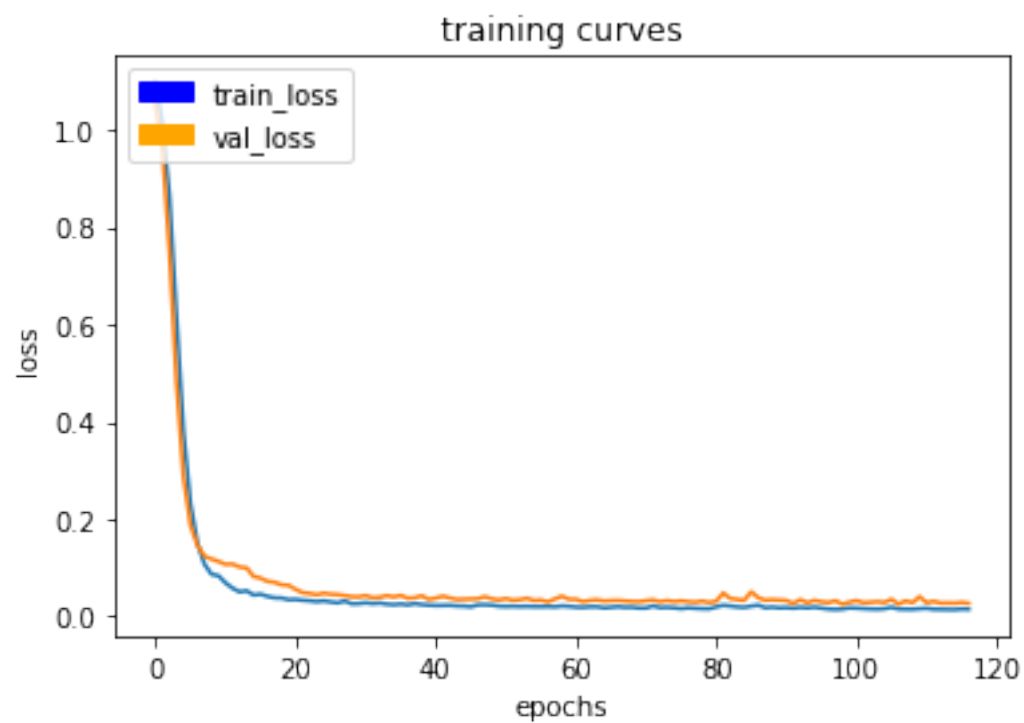
30/30 [=====] - 23s - loss: 0.0142 - val\_loss: 0.0275  
Epoch 116/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0152



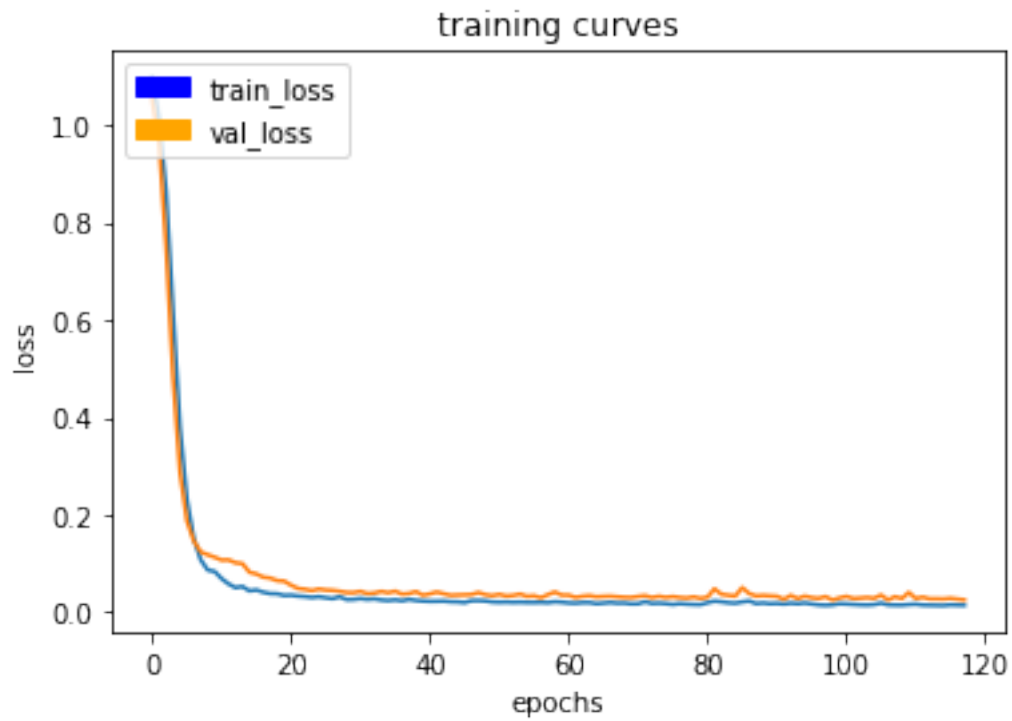
30/30 [=====] - 23s - loss: 0.0155 - val\_loss: 0.0289

Epoch 117/200

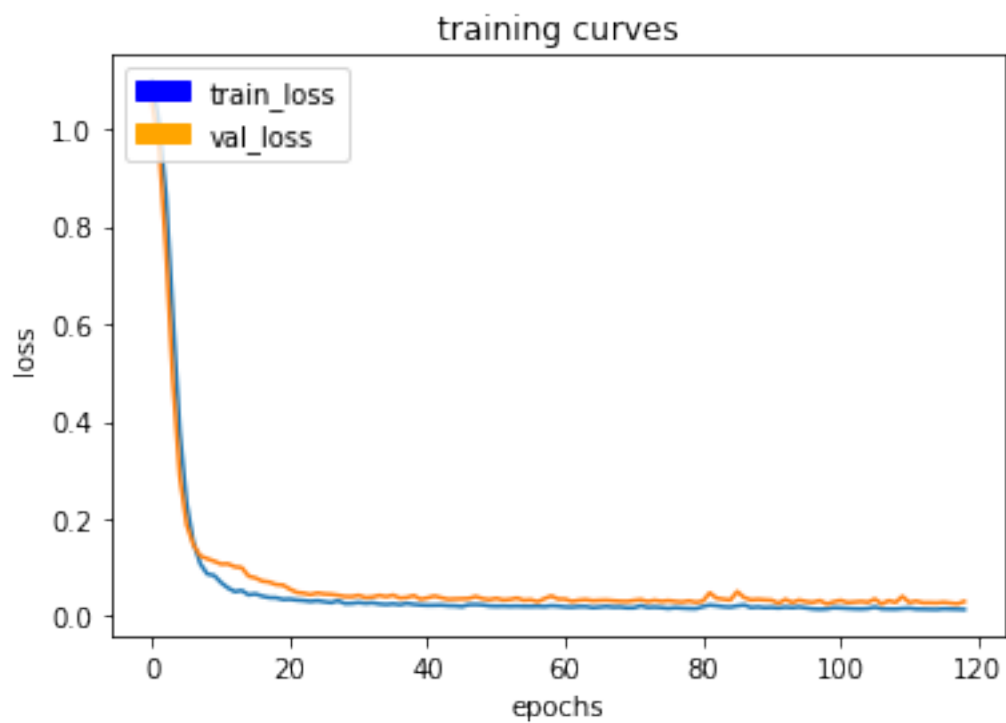
29/30 [=====>.] - ETA: 0s - loss: 0.0149



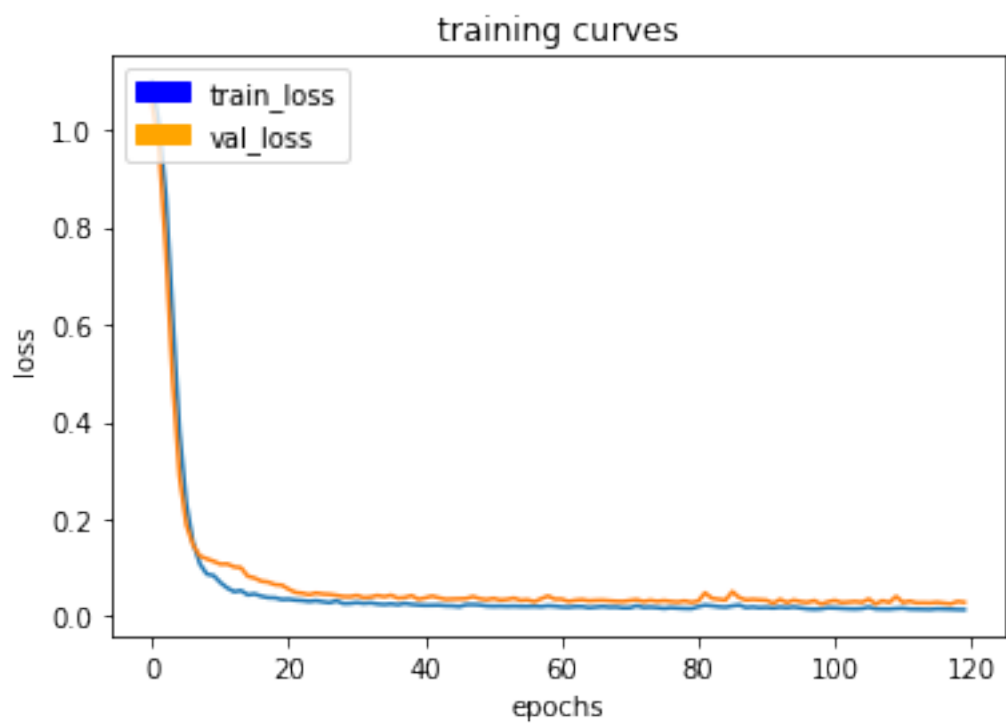
30/30 [=====] - 23s - loss: 0.0150 - val\_loss: 0.0272  
Epoch 118/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0147



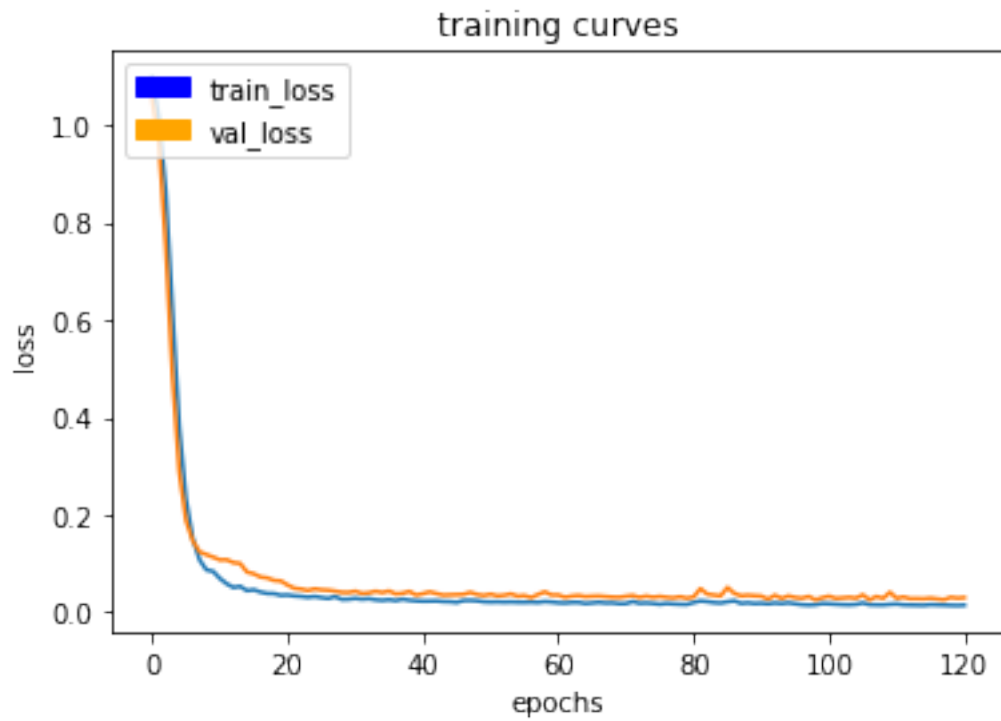
30/30 [=====] - 23s - loss: 0.0147 - val\_loss: 0.0257  
Epoch 119/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0142



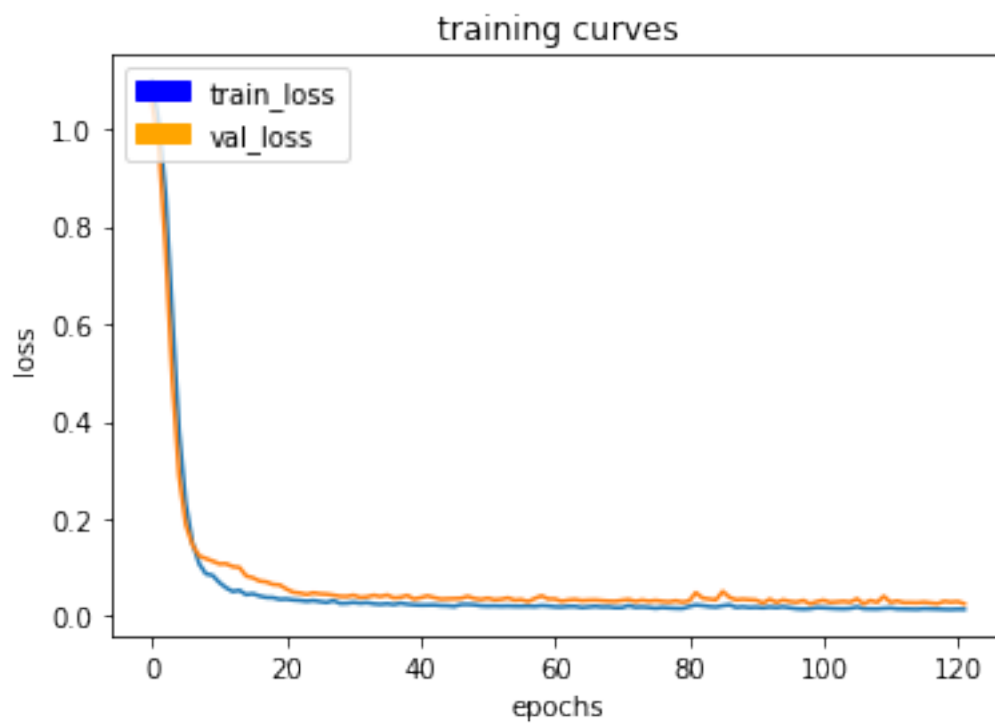
```
30/30 [=====] - 23s - loss: 0.0141 - val_loss: 0.0306
Epoch 120/200
29/30 [=====>.] - ETA: 0s - loss: 0.0138
```



30/30 [=====] - 23s - loss: 0.0138 - val\_loss: 0.0289  
Epoch 121/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0144



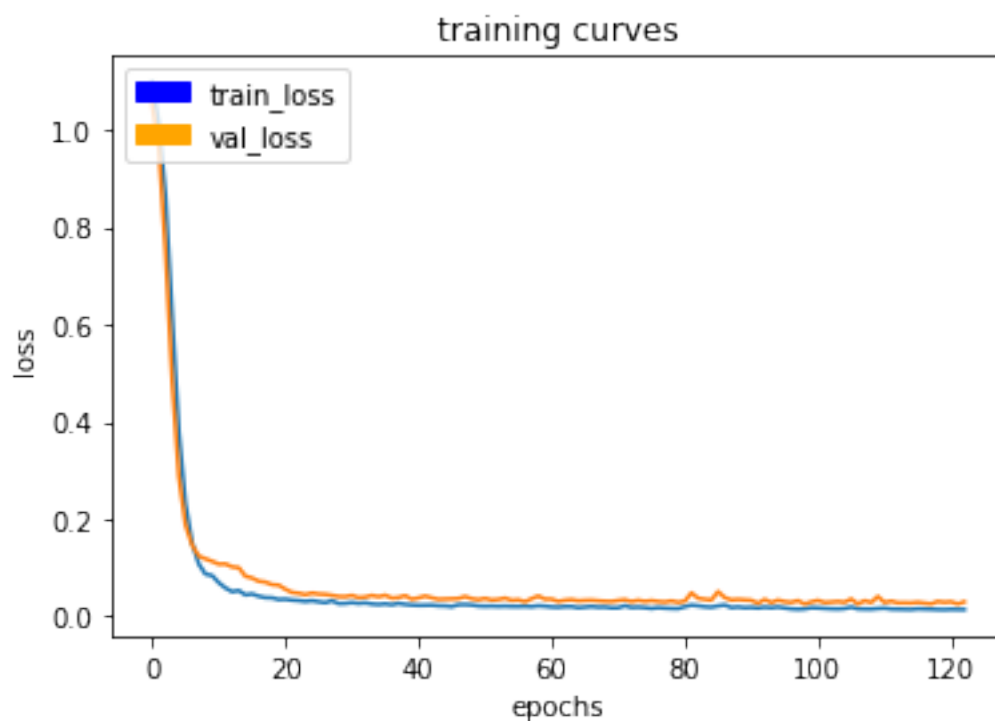
30/30 [=====] - 23s - loss: 0.0144 - val\_loss: 0.0300  
Epoch 122/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0142



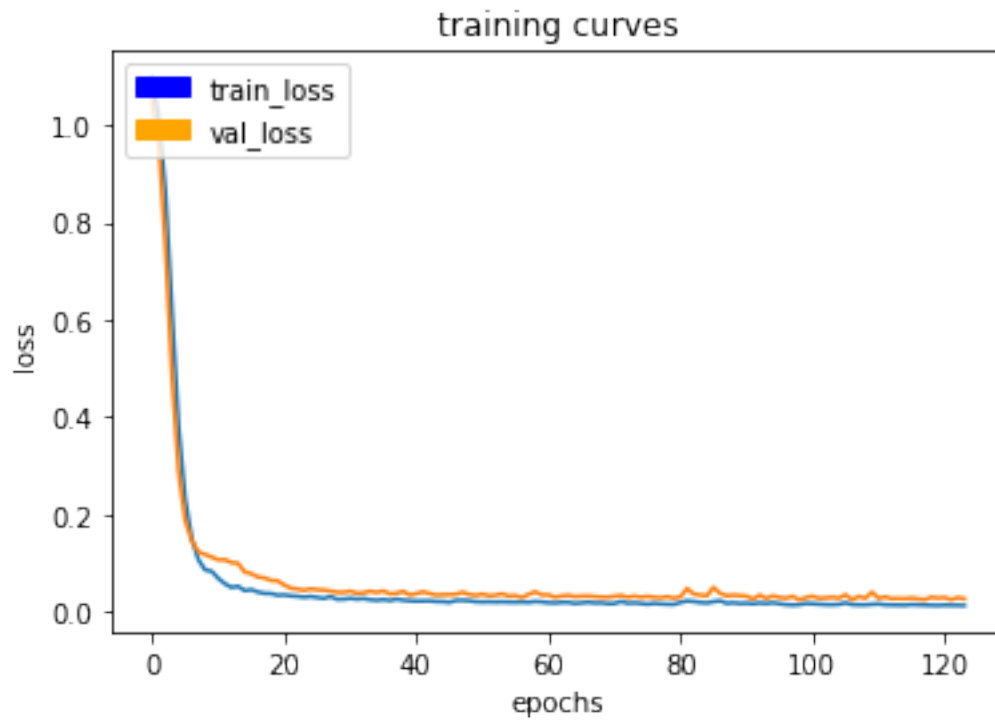
```

30/30 [=====] - 23s - loss: 0.0141 - val_loss: 0.0255
Epoch 123/200
29/30 [=====>.] - ETA: 0s - loss: 0.0139

```

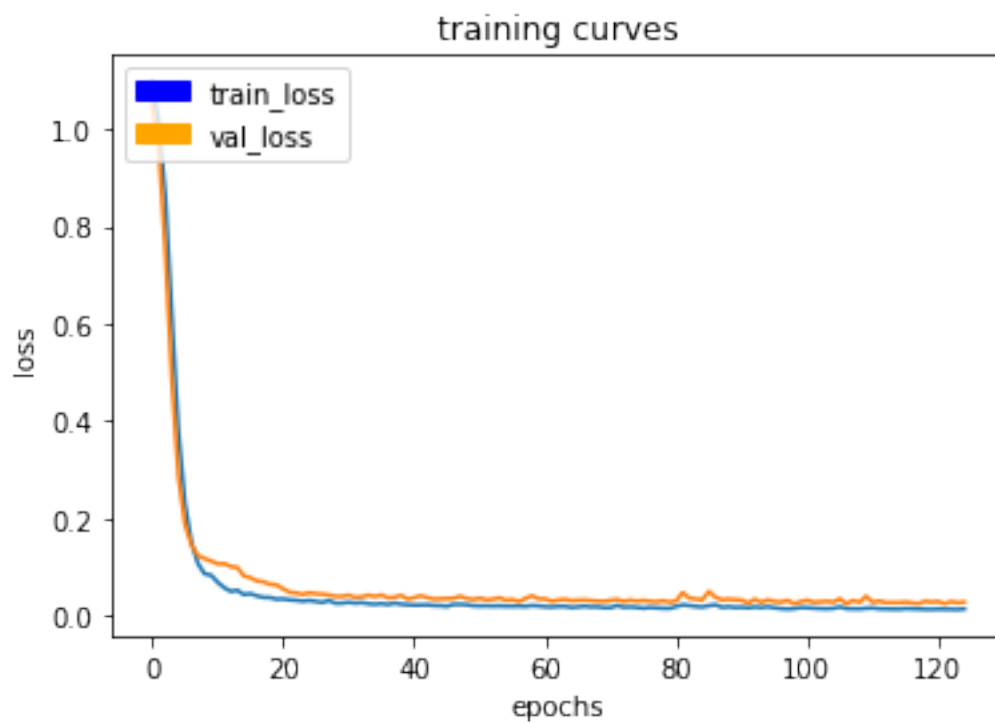


30/30 [=====] - 23s - loss: 0.0140 - val\_loss: 0.0299  
Epoch 124/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0136



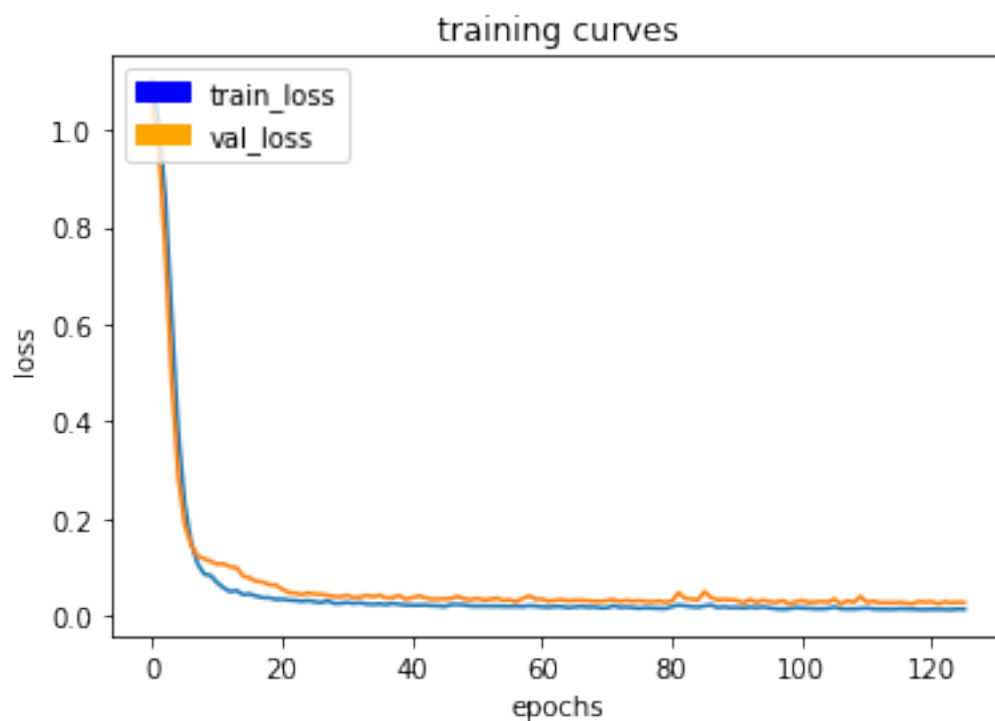
30/30 [=====] - 23s - loss: 0.0136 - val\_loss: 0.0275  
Epoch 125/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0150



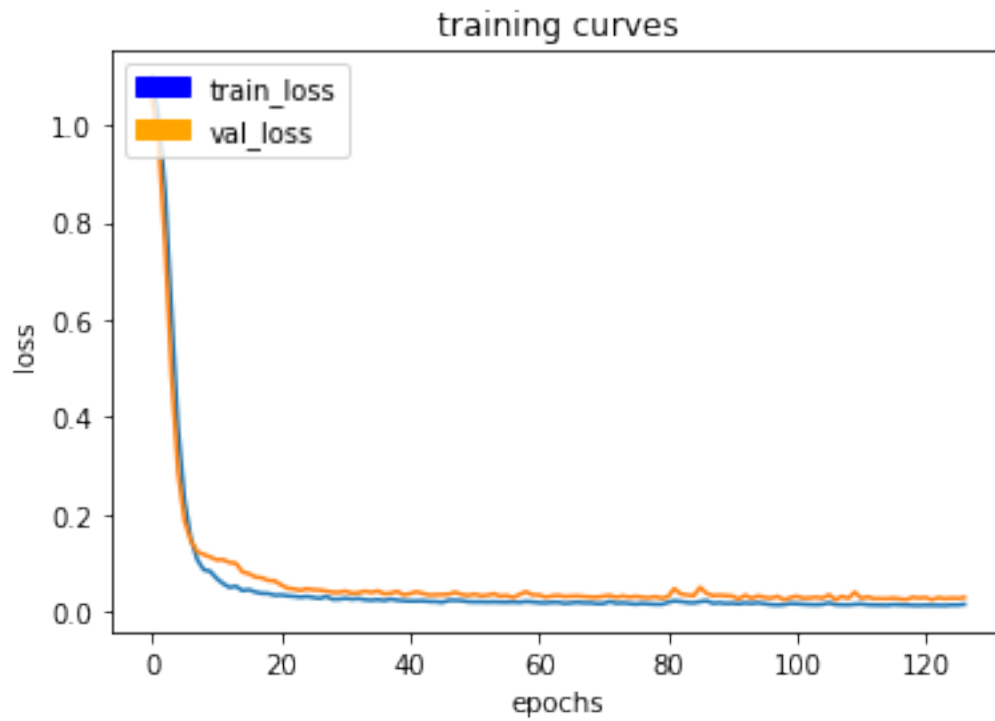


```

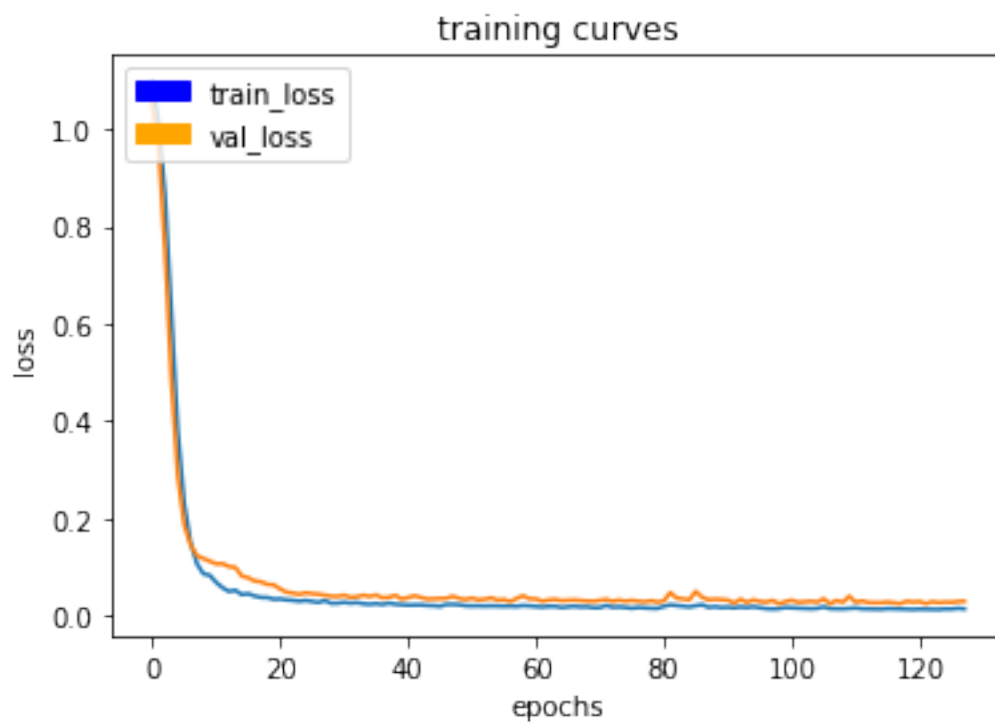
30/30 [=====] - 23s - loss: 0.0148 - val_loss: 0.0290
Epoch 126/200
29/30 [=====>.] - ETA: 0s - loss: 0.0145
  
```



30/30 [=====] - 23s - loss: 0.0145 - val\_loss: 0.0282  
Epoch 127/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0161



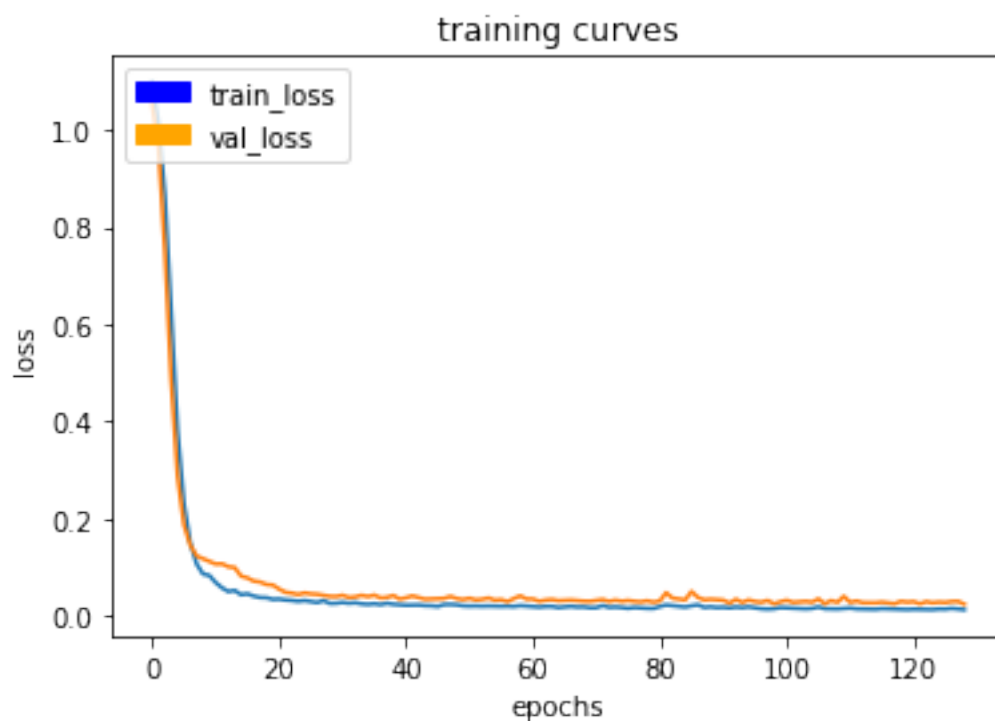
30/30 [=====] - 23s - loss: 0.0161 - val\_loss: 0.0301  
Epoch 128/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0148



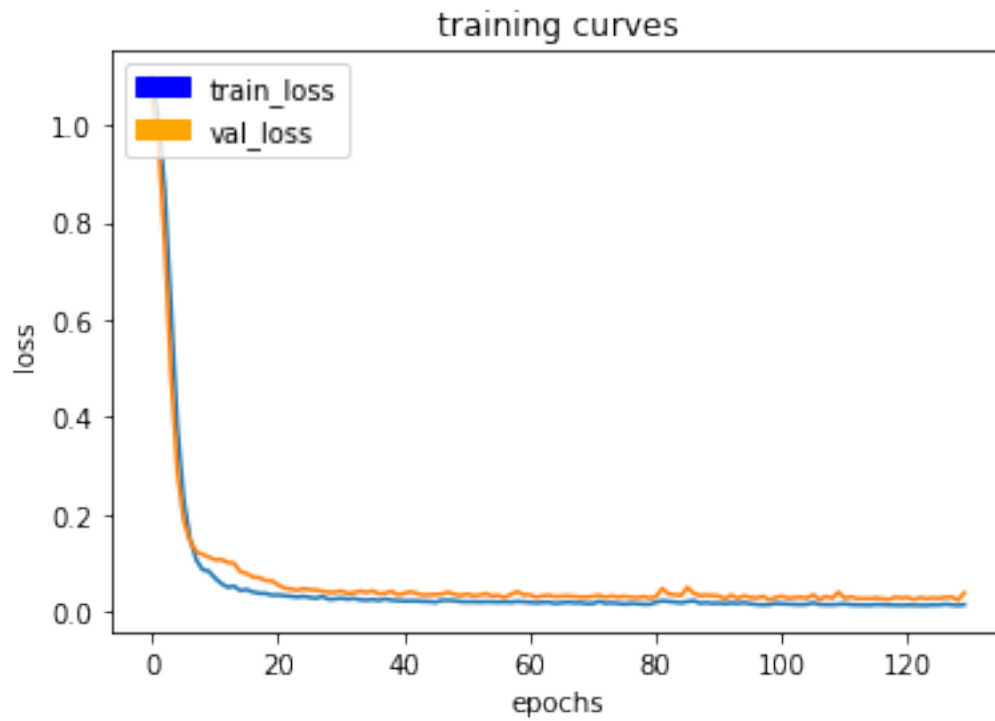
```

30/30 [=====] - 23s - loss: 0.0148 - val_loss: 0.0305
Epoch 129/200
29/30 [=====>.] - ETA: 0s - loss: 0.0134

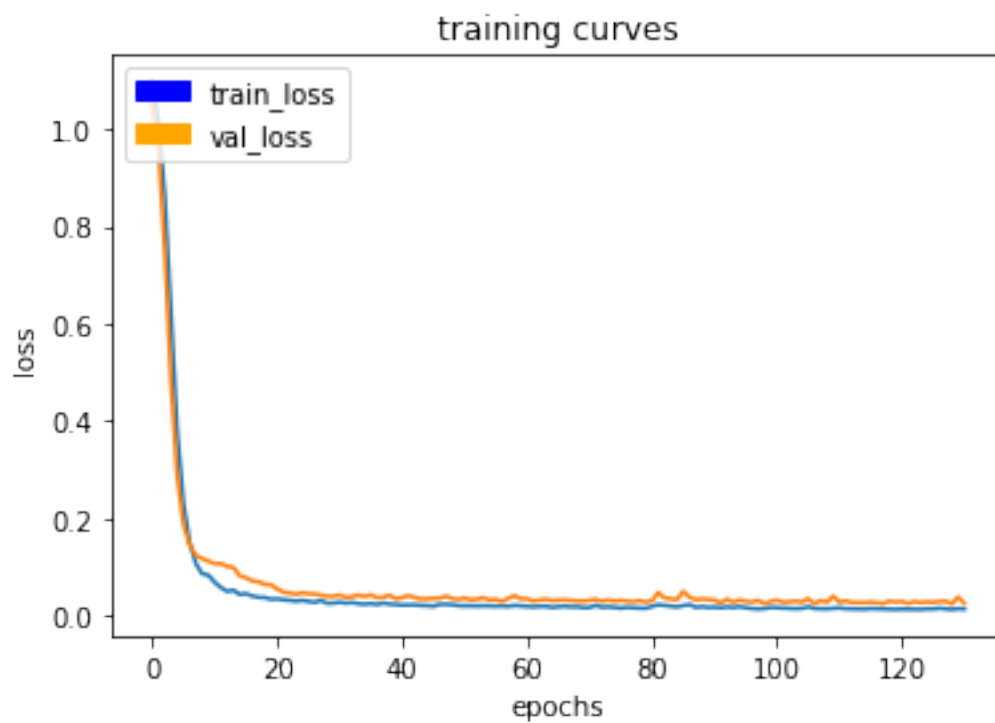
```



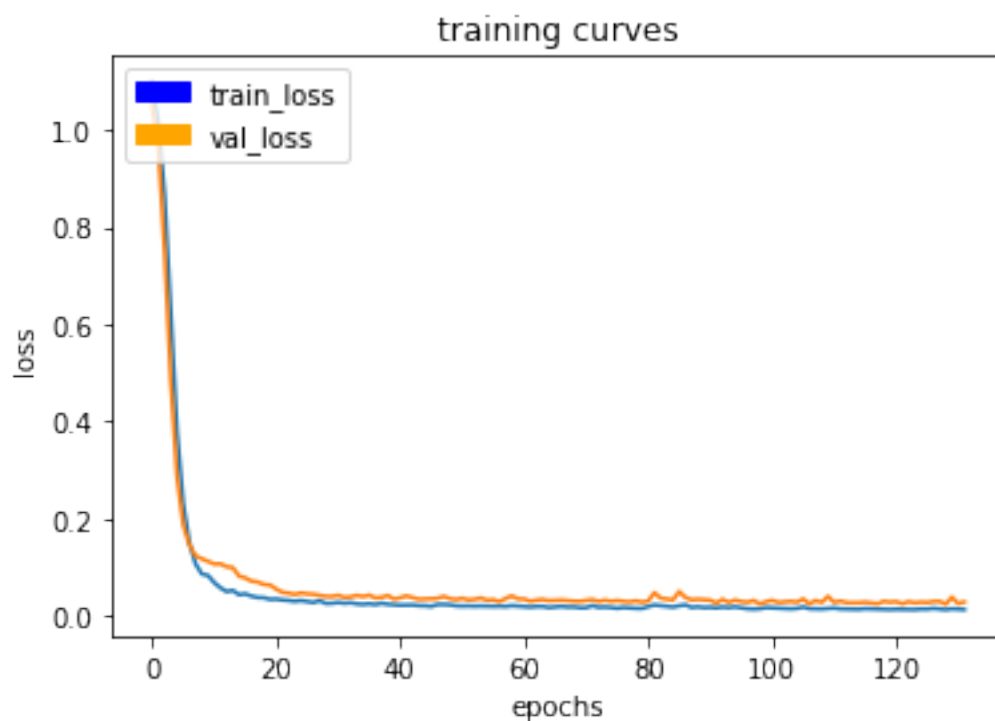
30/30 [=====] - 23s - loss: 0.0135 - val\_loss: 0.0247  
Epoch 130/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0152



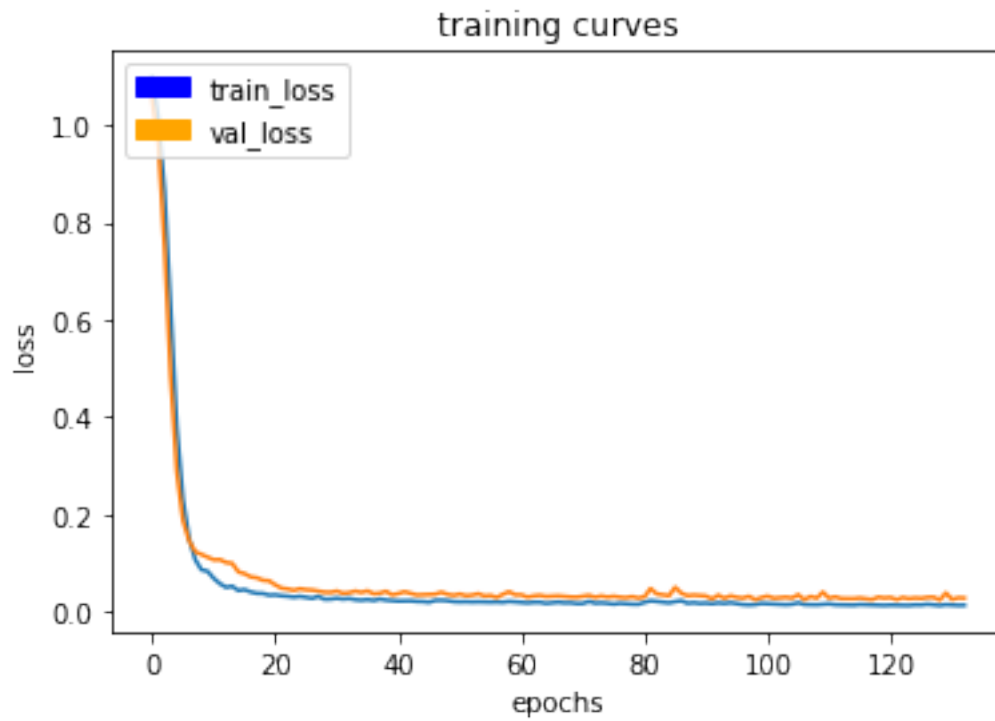
30/30 [=====] - 23s - loss: 0.0152 - val\_loss: 0.0393  
Epoch 131/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0145



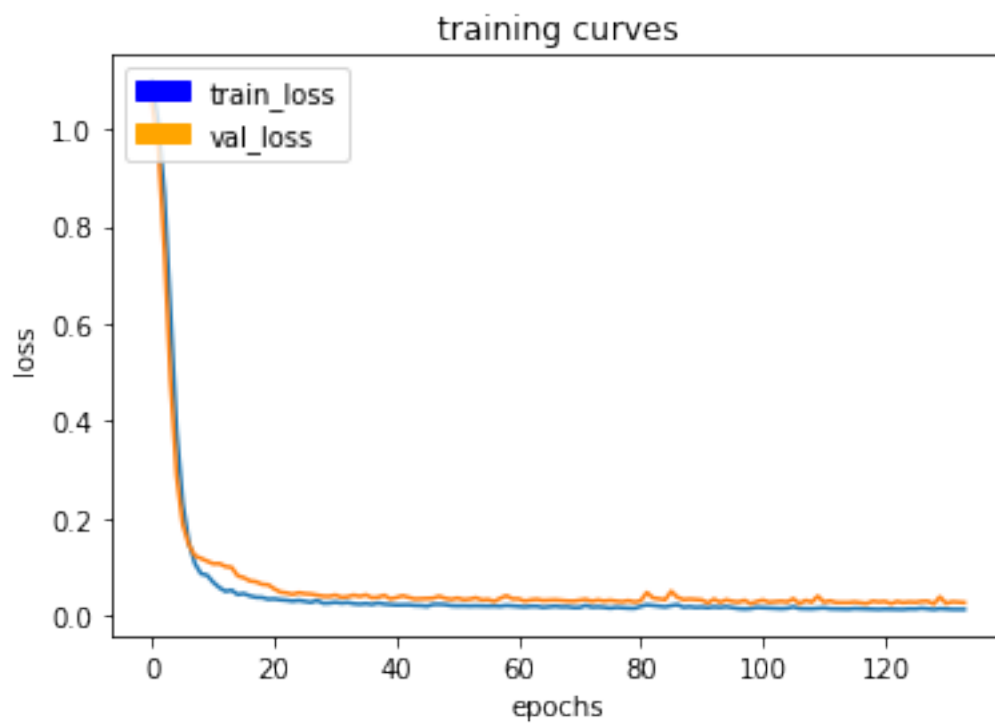
30/30 [=====] - 23s - loss: 0.0145 - val\_loss: 0.0259  
Epoch 132/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0136



30/30 [=====] - 23s - loss: 0.0138 - val\_loss: 0.0293  
Epoch 133/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0141



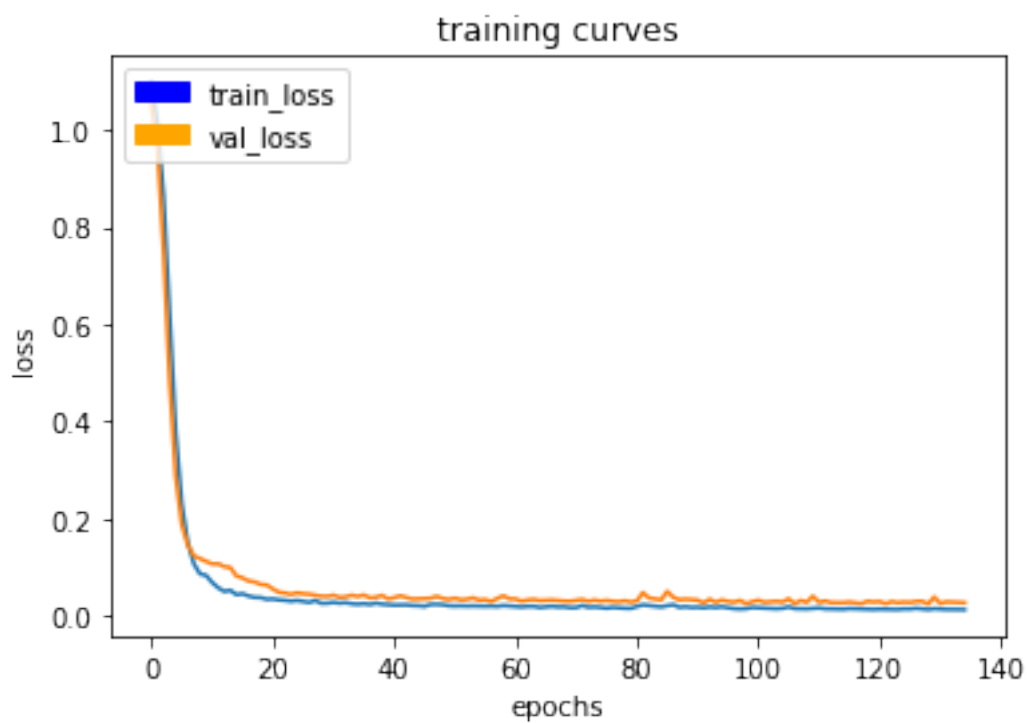
30/30 [=====] - 23s - loss: 0.0140 - val\_loss: 0.0288  
Epoch 134/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0136



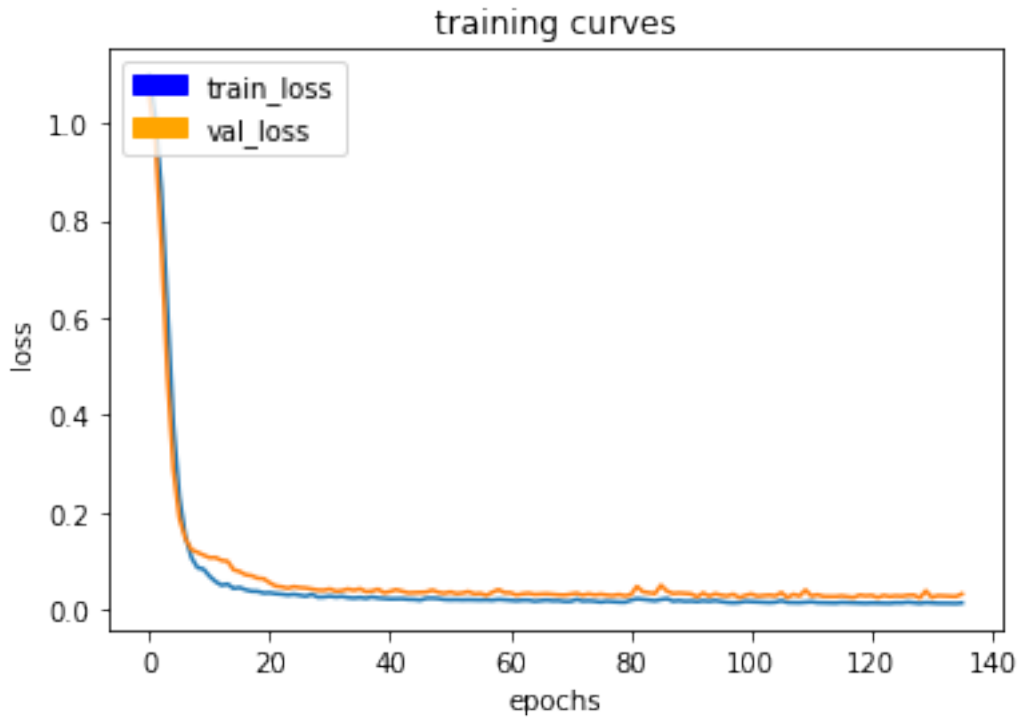
30/30 [=====] - 23s - loss: 0.0136 - val\_loss: 0.0276

Epoch 135/200

29/30 [=====>.] - ETA: 0s - loss: 0.0133

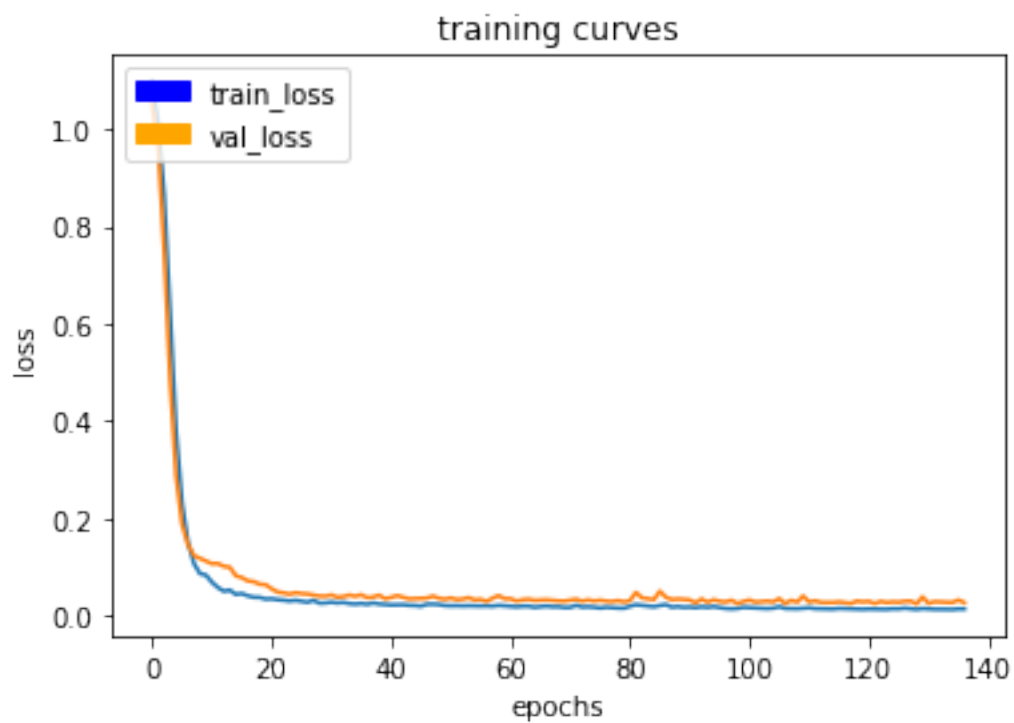


30/30 [=====] - 23s - loss: 0.0133 - val\_loss: 0.0275  
Epoch 136/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0143



30/30 [=====] - 23s - loss: 0.0145 - val\_loss: 0.0329  
Epoch 137/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0143

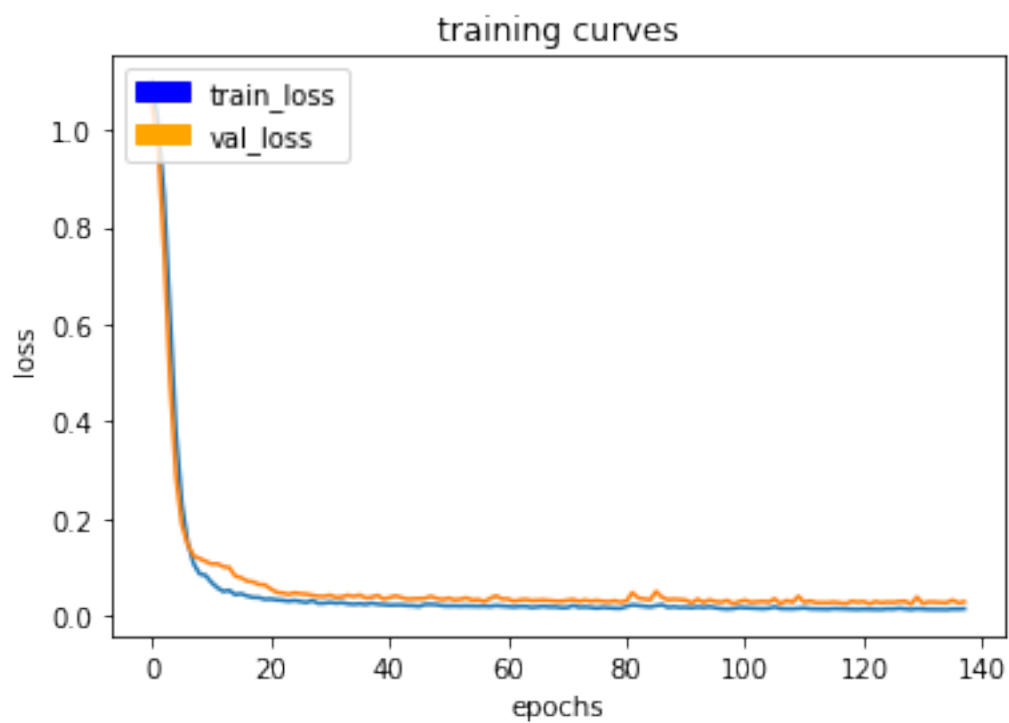




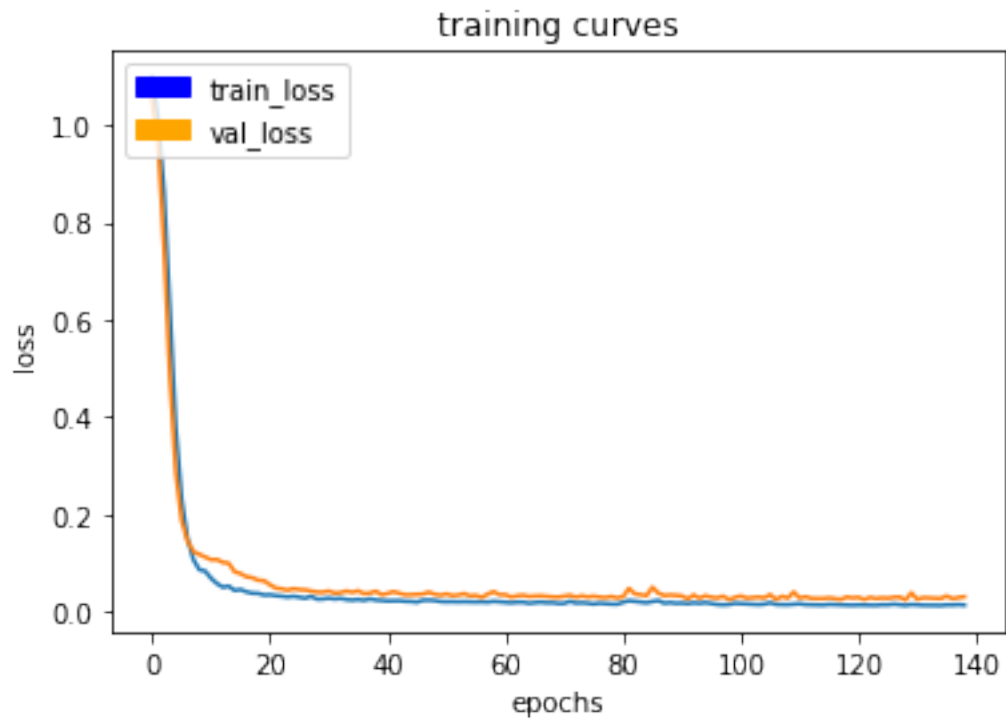
```

30/30 [=====] - 23s - loss: 0.0143 - val_loss: 0.0271
Epoch 138/200
29/30 [=====>.] - ETA: 0s - loss: 0.0152

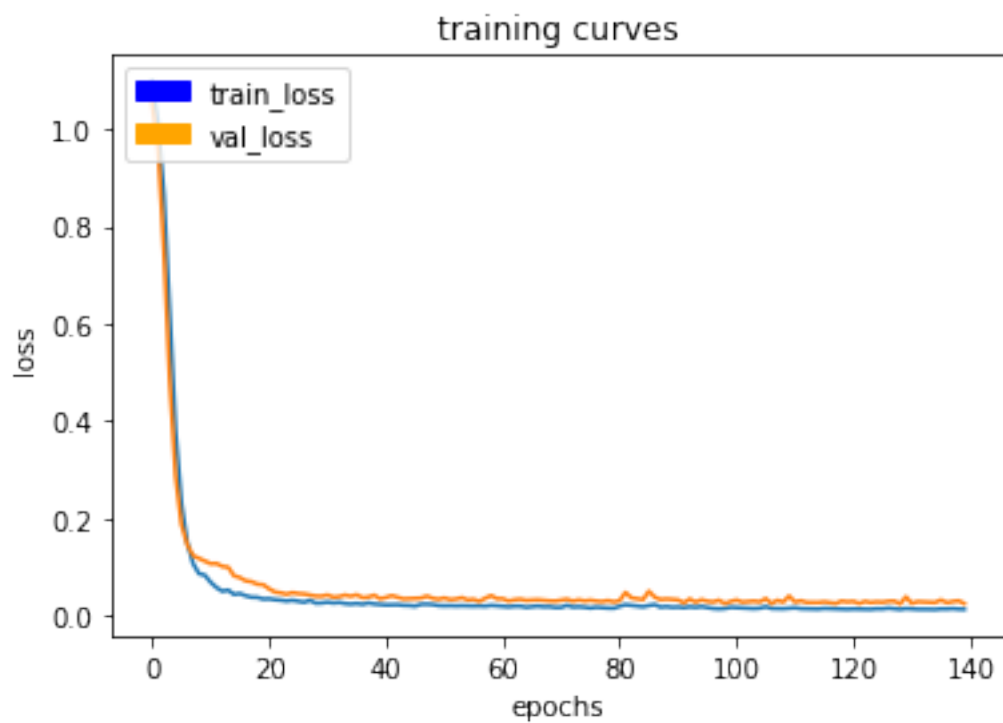
```



30/30 [=====] - 23s - loss: 0.0151 - val\_loss: 0.0296  
Epoch 139/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0141



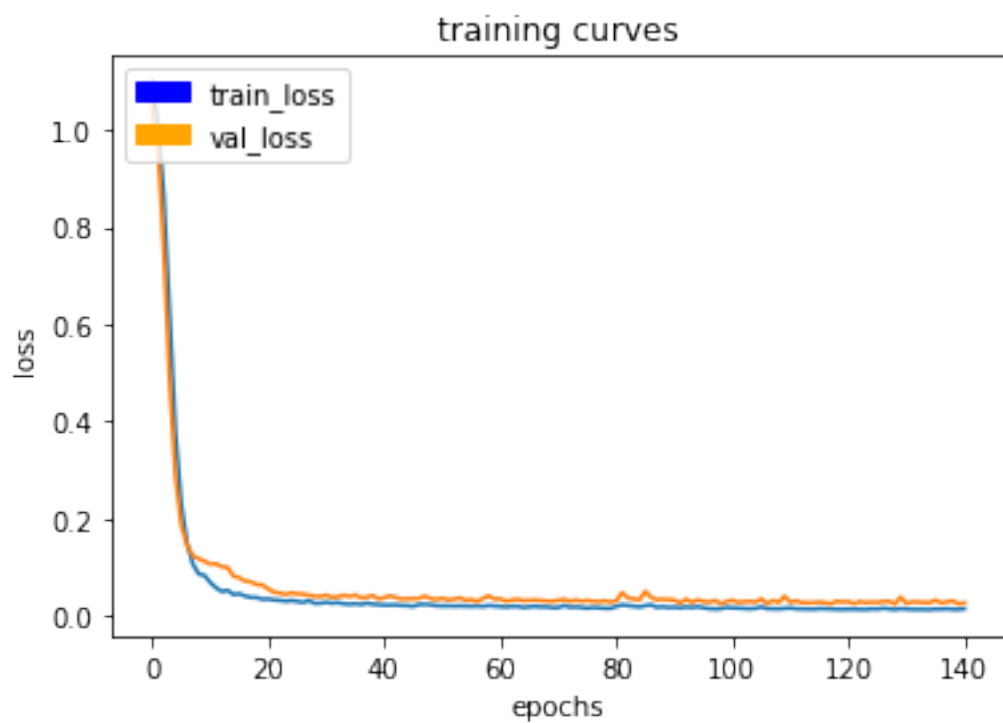
30/30 [=====] - 23s - loss: 0.0141 - val\_loss: 0.0312  
Epoch 140/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0136



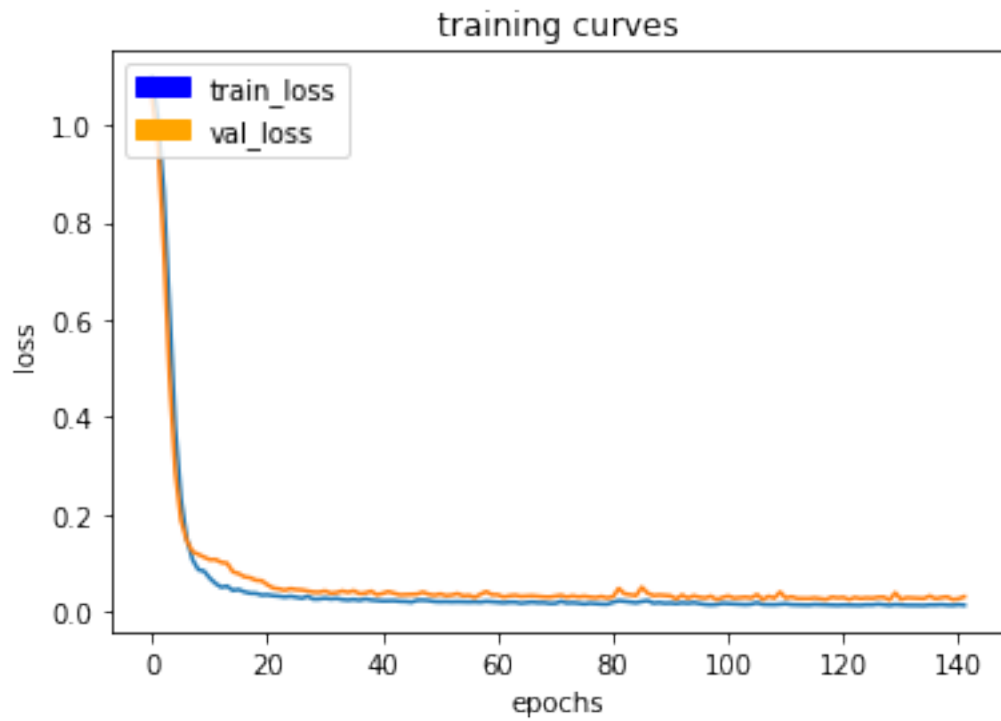
```

30/30 [=====] - 23s - loss: 0.0137 - val_loss: 0.0254
Epoch 141/200
29/30 [=====>.] - ETA: 0s - loss: 0.0147

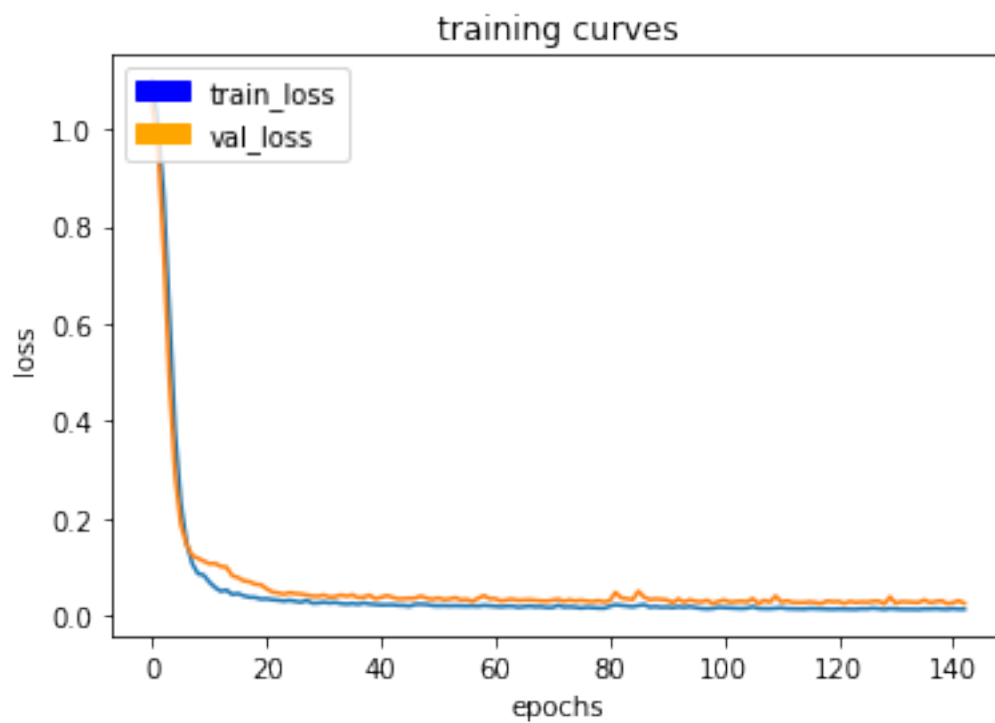
```



30/30 [=====] - 23s - loss: 0.0150 - val\_loss: 0.0272  
Epoch 142/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0139



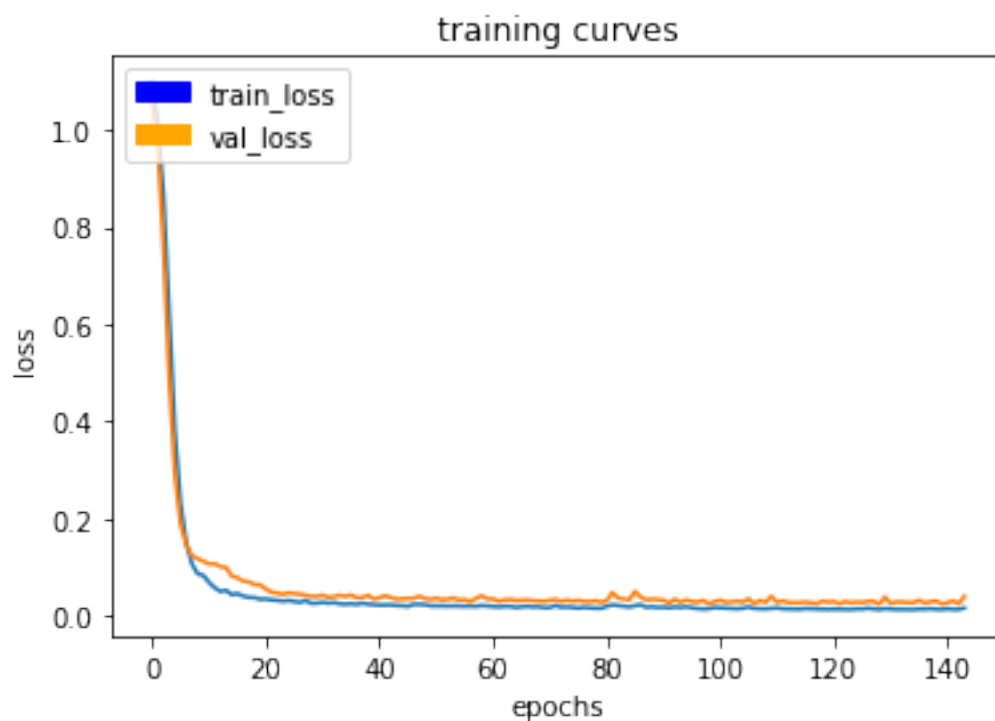
30/30 [=====] - 23s - loss: 0.0139 - val\_loss: 0.0313  
Epoch 143/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0139



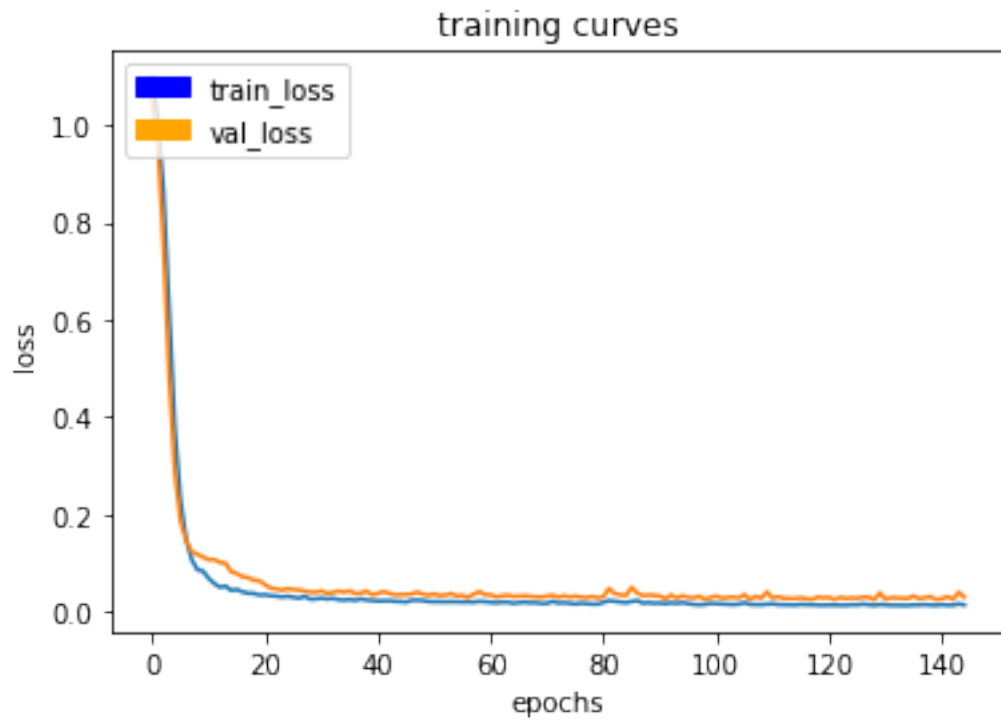
```

30/30 [=====] - 23s - loss: 0.0139 - val_loss: 0.0259
Epoch 144/200
29/30 [=====>.] - ETA: 0s - loss: 0.0166

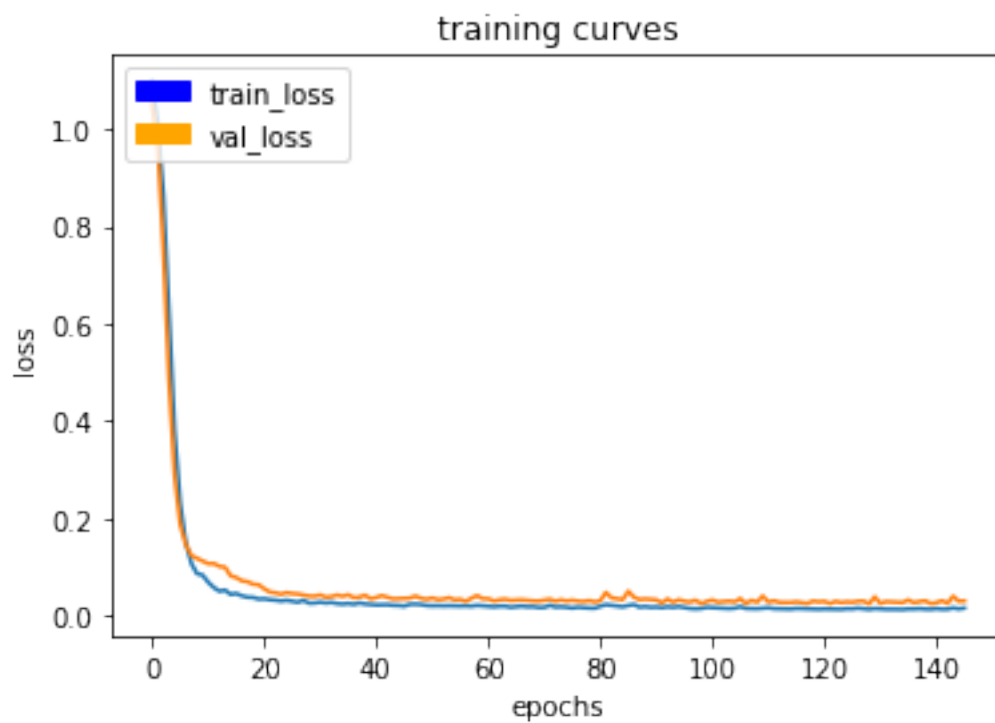
```



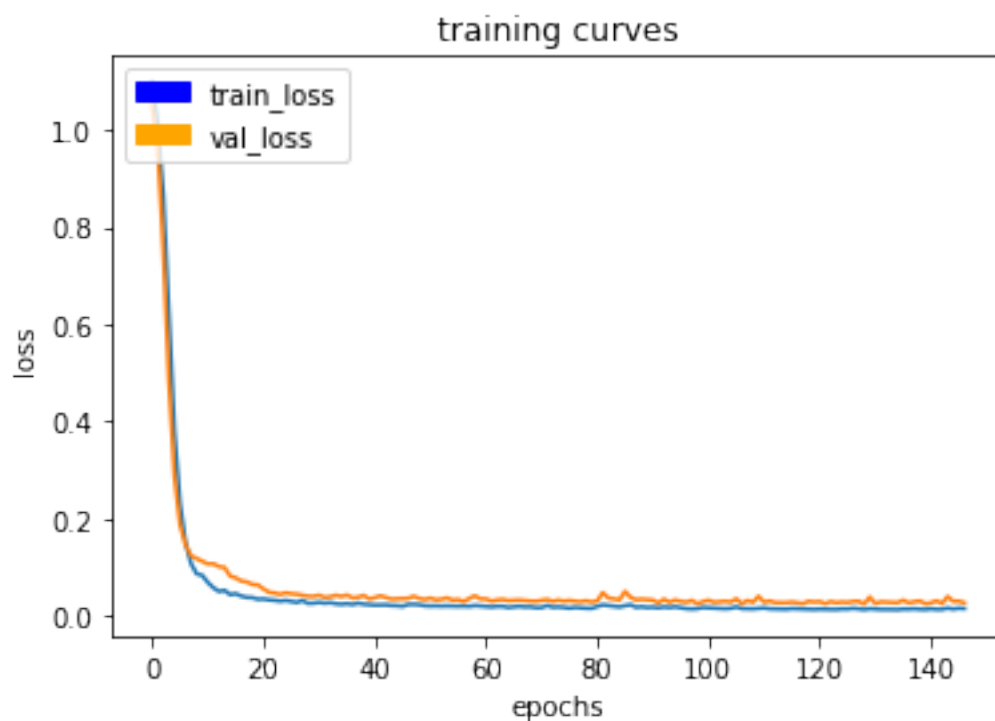
30/30 [=====] - 23s - loss: 0.0166 - val\_loss: 0.0404  
Epoch 145/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0144



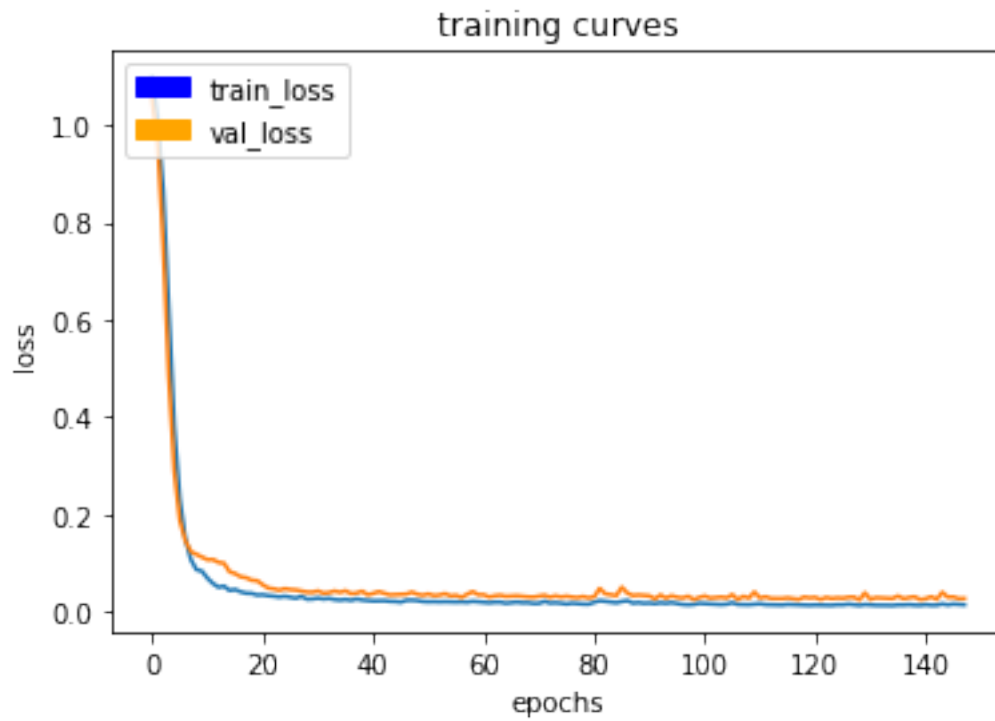
30/30 [=====] - 23s - loss: 0.0144 - val\_loss: 0.0303  
Epoch 146/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0162



30/30 [=====] - 23s - loss: 0.0162 - val\_loss: 0.0308  
 Epoch 147/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0155

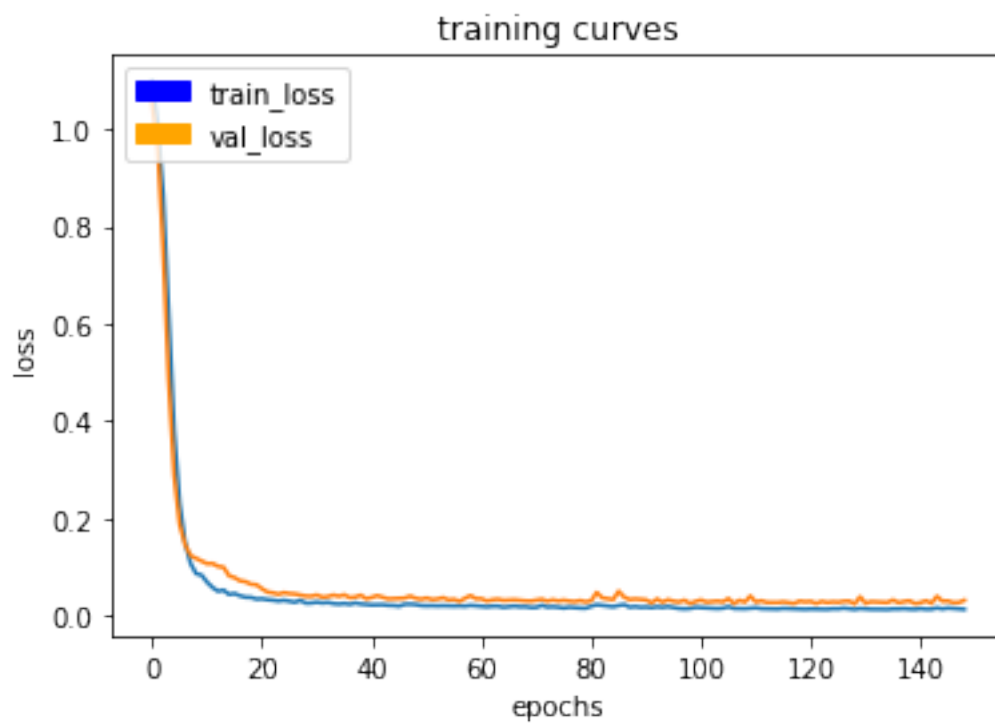


30/30 [=====] - 23s - loss: 0.0154 - val\_loss: 0.0273  
Epoch 148/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0148

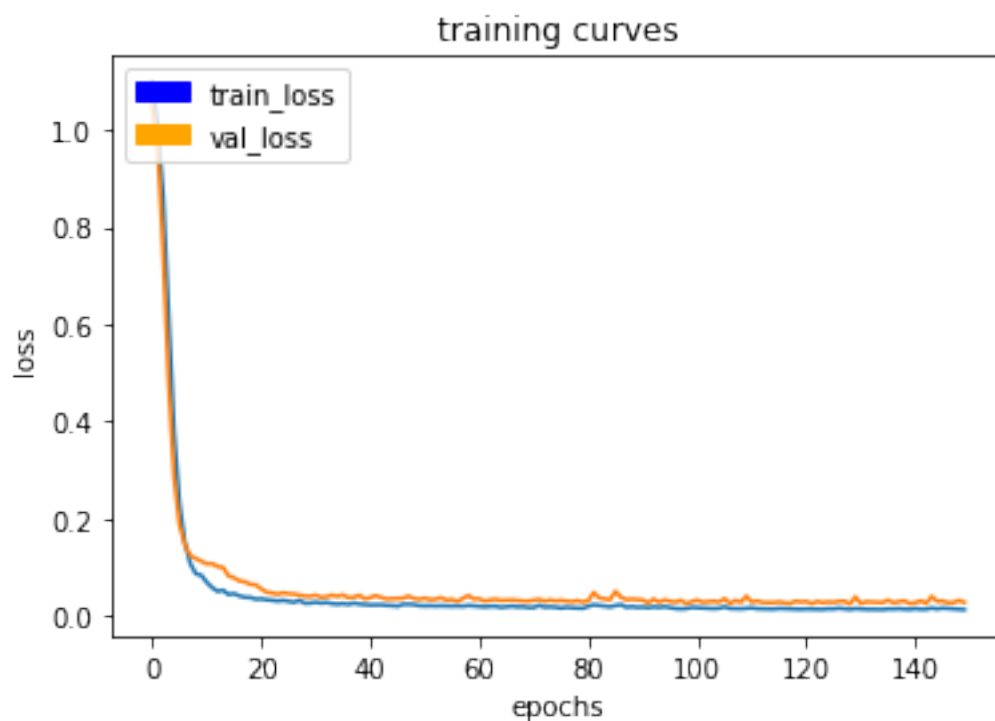


30/30 [=====] - 23s - loss: 0.0147 - val\_loss: 0.0276  
Epoch 149/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0136

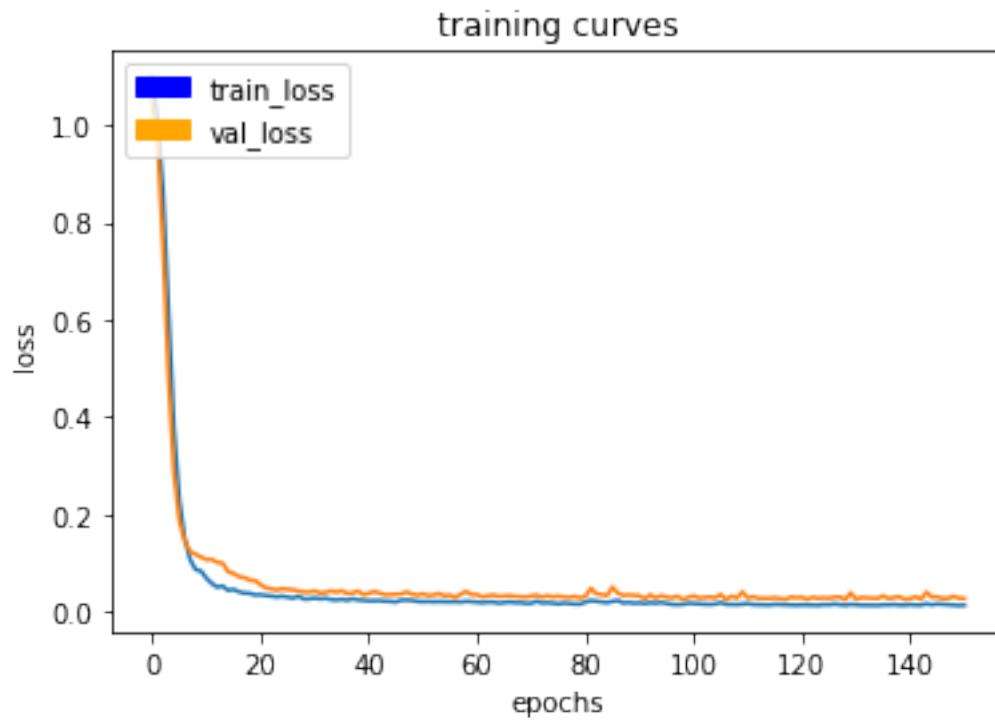




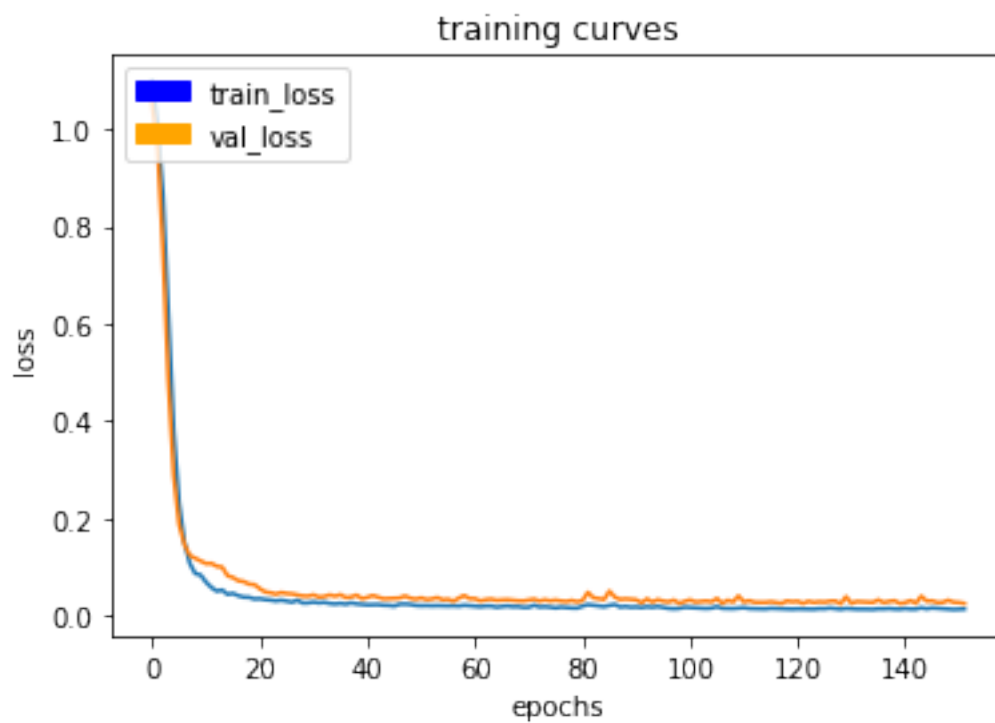
```
30/30 [=====] - 23s - loss: 0.0136 - val_loss: 0.0321
Epoch 150/200
29/30 [=====>.] - ETA: 0s - loss: 0.0131
```



30/30 [=====] - 23s - loss: 0.0131 - val\_loss: 0.0279  
Epoch 151/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0135

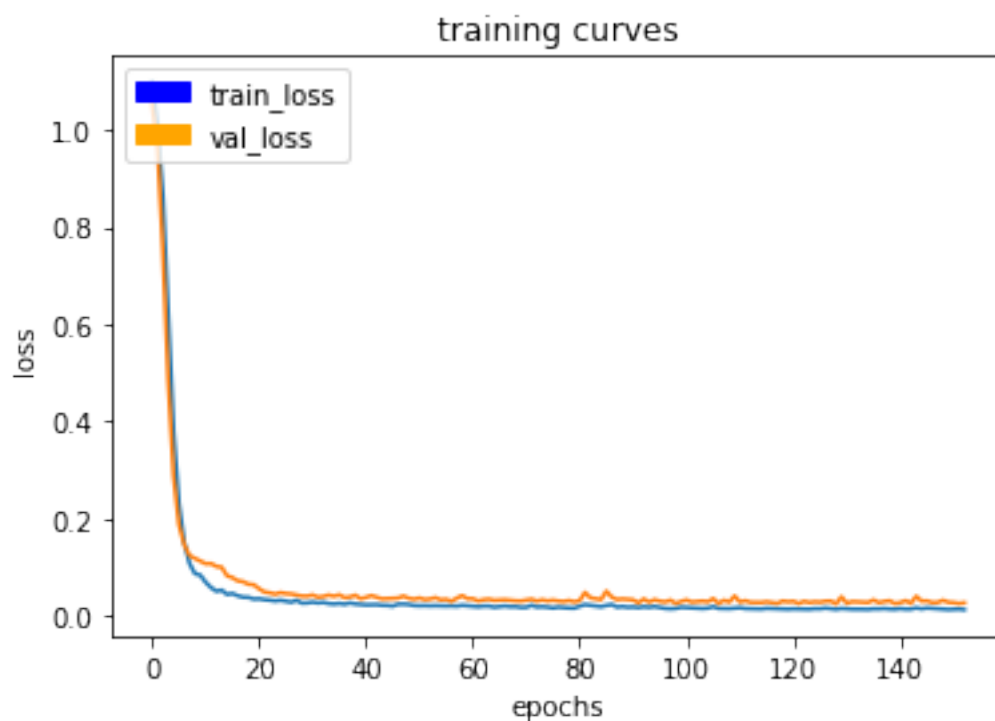


30/30 [=====] - 23s - loss: 0.0136 - val\_loss: 0.0273  
Epoch 152/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0141

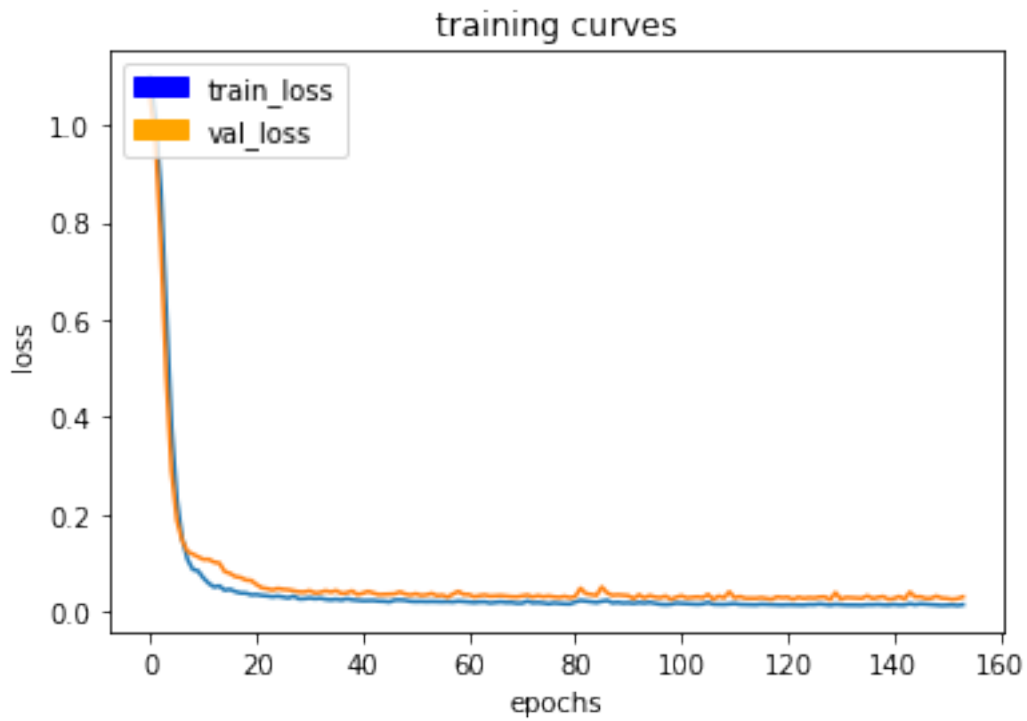


```

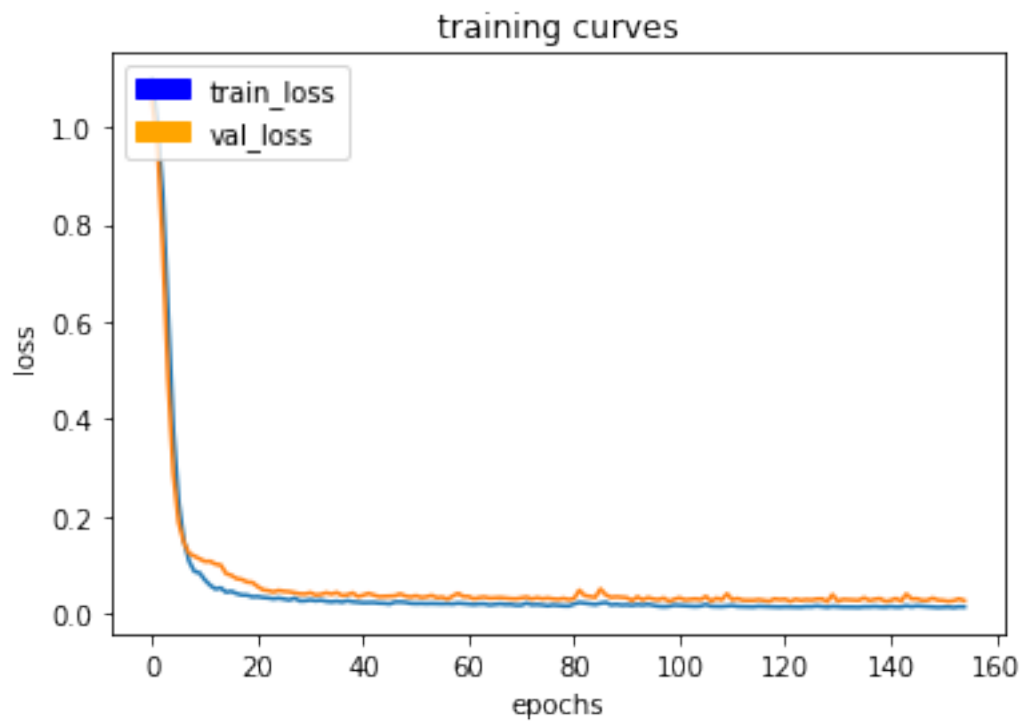
30/30 [=====] - 23s - loss: 0.0140 - val_loss: 0.0254
Epoch 153/200
29/30 [=====>.] - ETA: 0s - loss: 0.0130
  
```



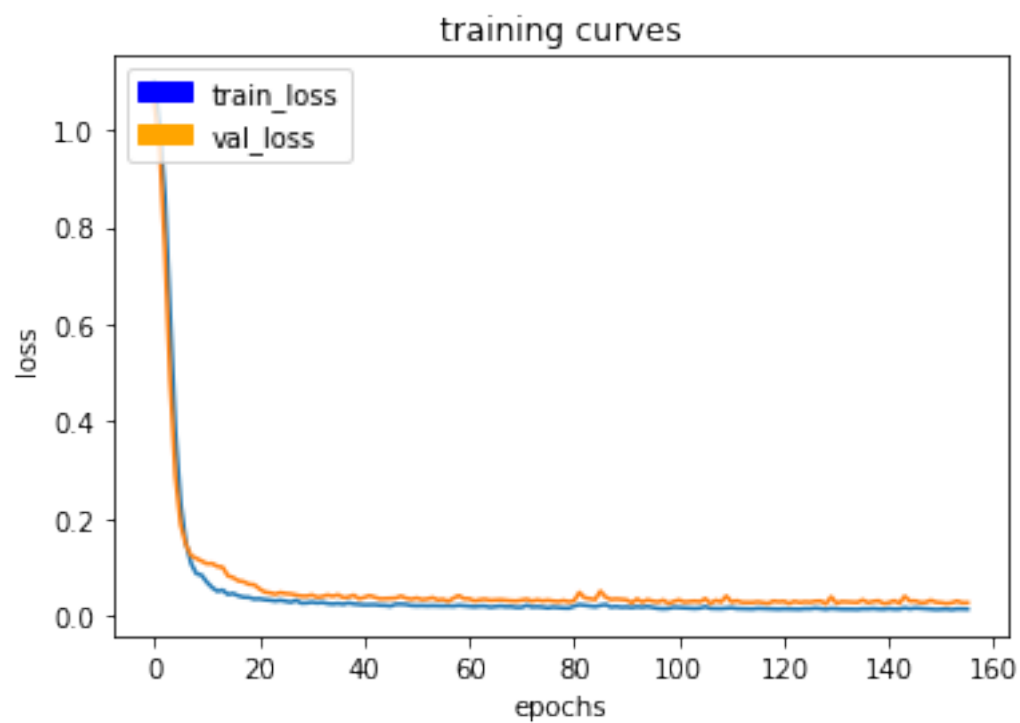
30/30 [=====] - 23s - loss: 0.0129 - val\_loss: 0.0269  
Epoch 154/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0145



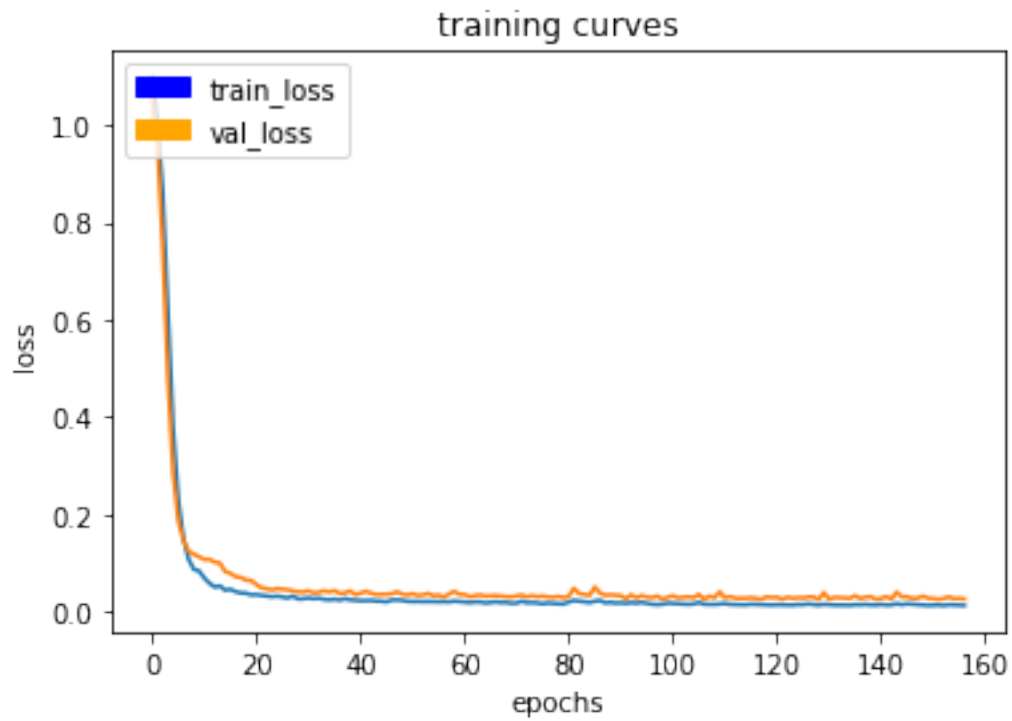
30/30 [=====] - 23s - loss: 0.0144 - val\_loss: 0.0305  
Epoch 155/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0140



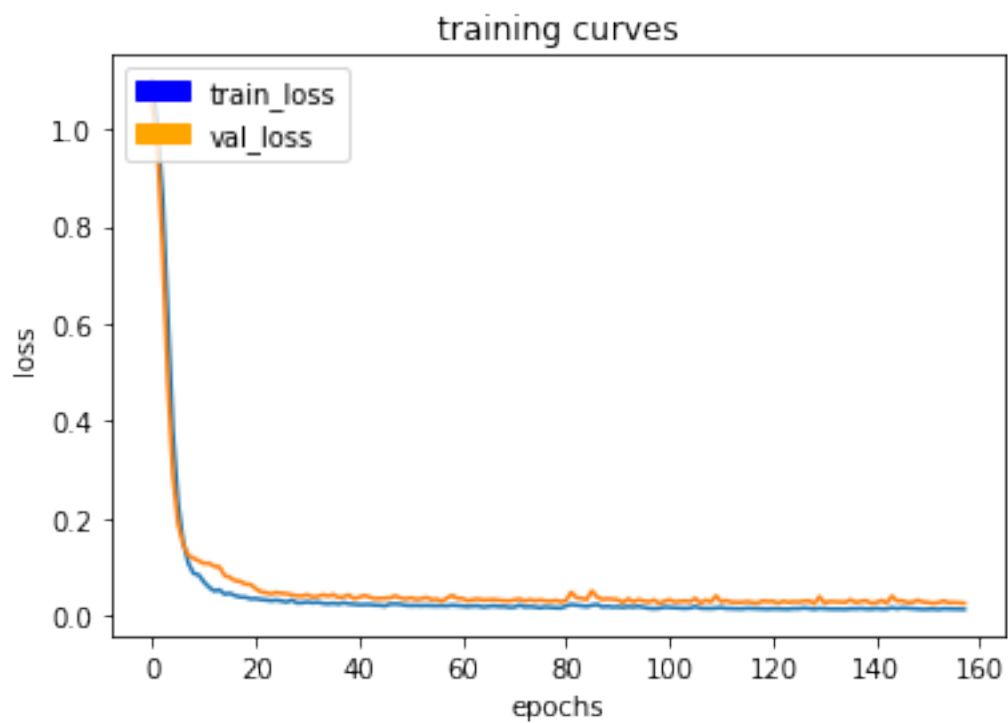
30/30 [=====] - 23s - loss: 0.0139 - val\_loss: 0.0269  
 Epoch 156/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0139



30/30 [=====] - 23s - loss: 0.0138 - val\_loss: 0.0271  
Epoch 157/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0129



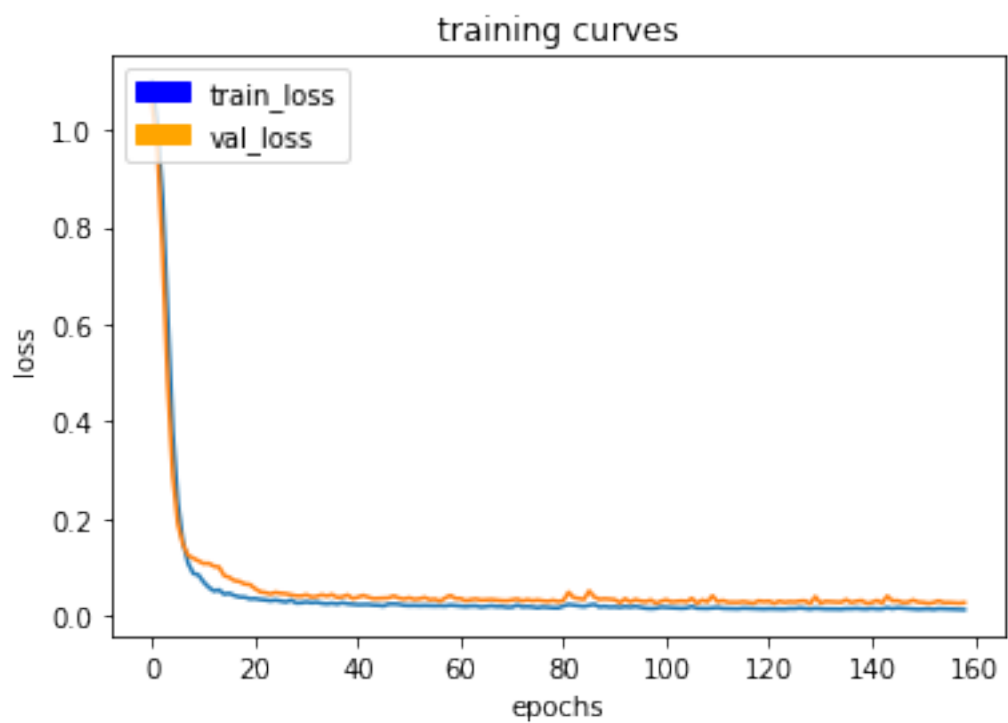
30/30 [=====] - 23s - loss: 0.0129 - val\_loss: 0.0262  
Epoch 158/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0131



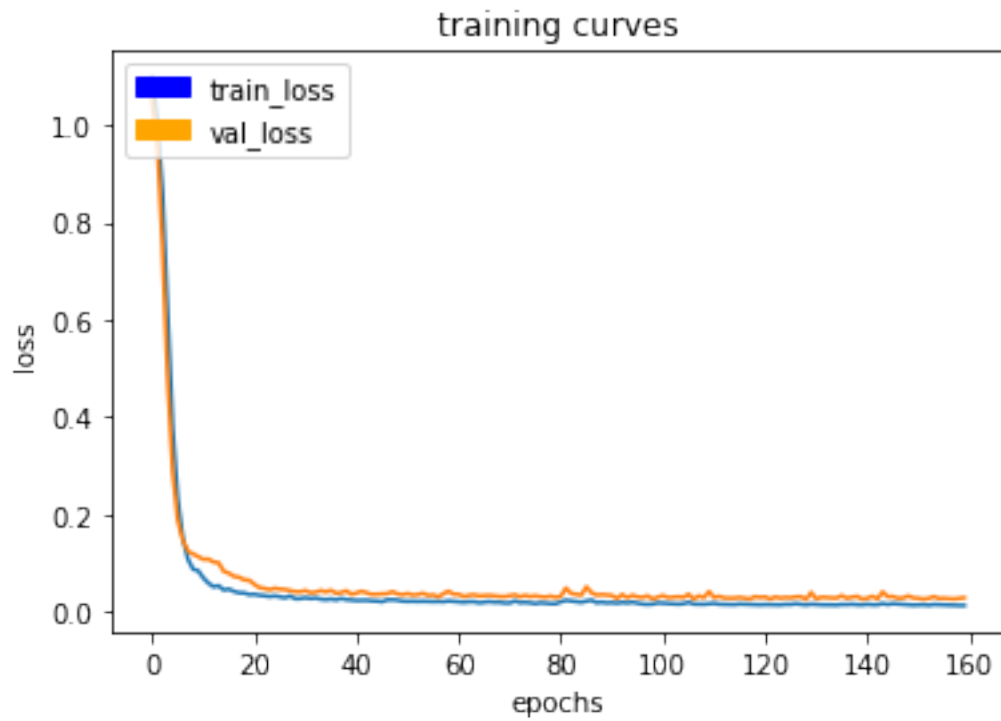
30/30 [=====] - 23s - loss: 0.0131 - val\_loss: 0.0258

Epoch 159/200

29/30 [=====>.] - ETA: 0s - loss: 0.0125

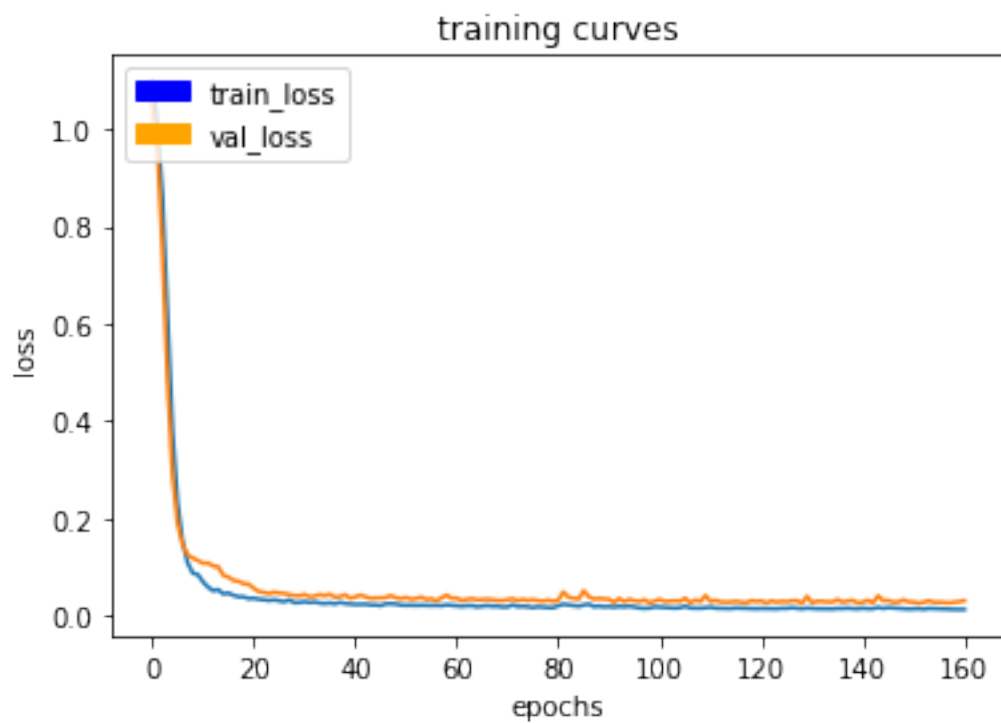


30/30 [=====] - 23s - loss: 0.0124 - val\_loss: 0.0269  
Epoch 160/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0124

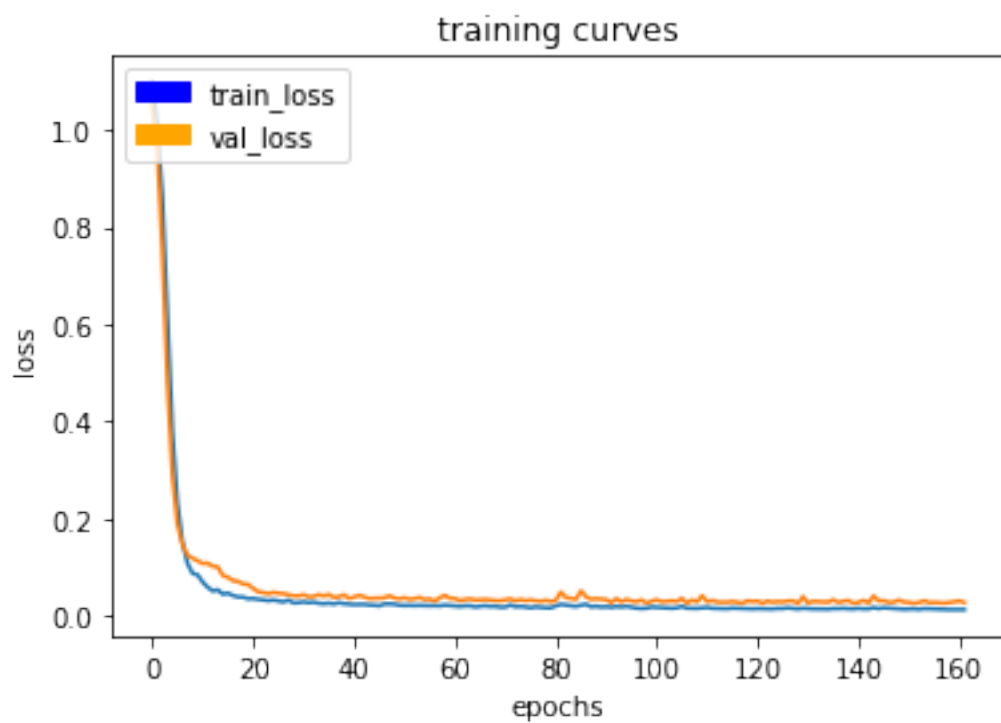


30/30 [=====] - 23s - loss: 0.0124 - val\_loss: 0.0283  
Epoch 161/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0122

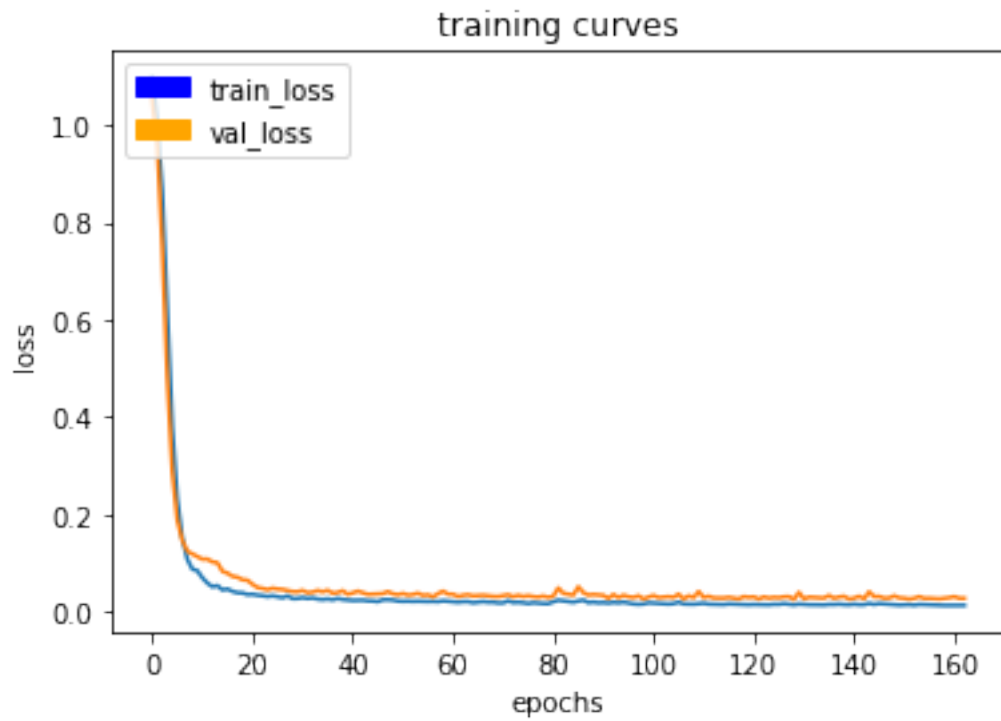




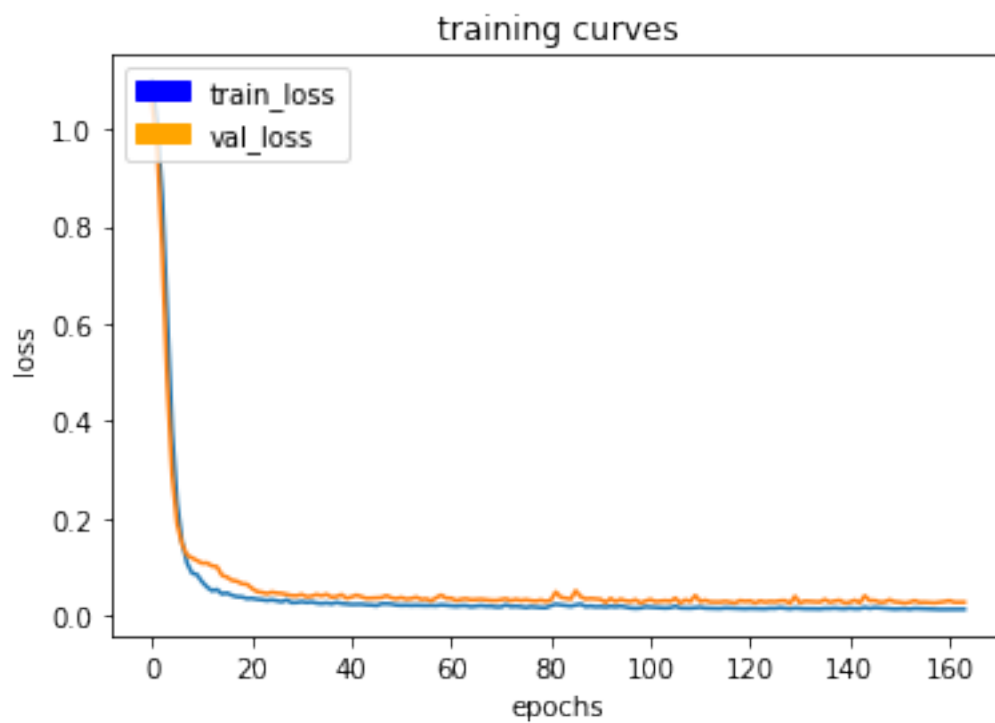
30/30 [=====] - 23s - loss: 0.0122 - val\_loss: 0.0304  
 Epoch 162/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0125



30/30 [=====] - 23s - loss: 0.0124 - val\_loss: 0.0270  
Epoch 163/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0125



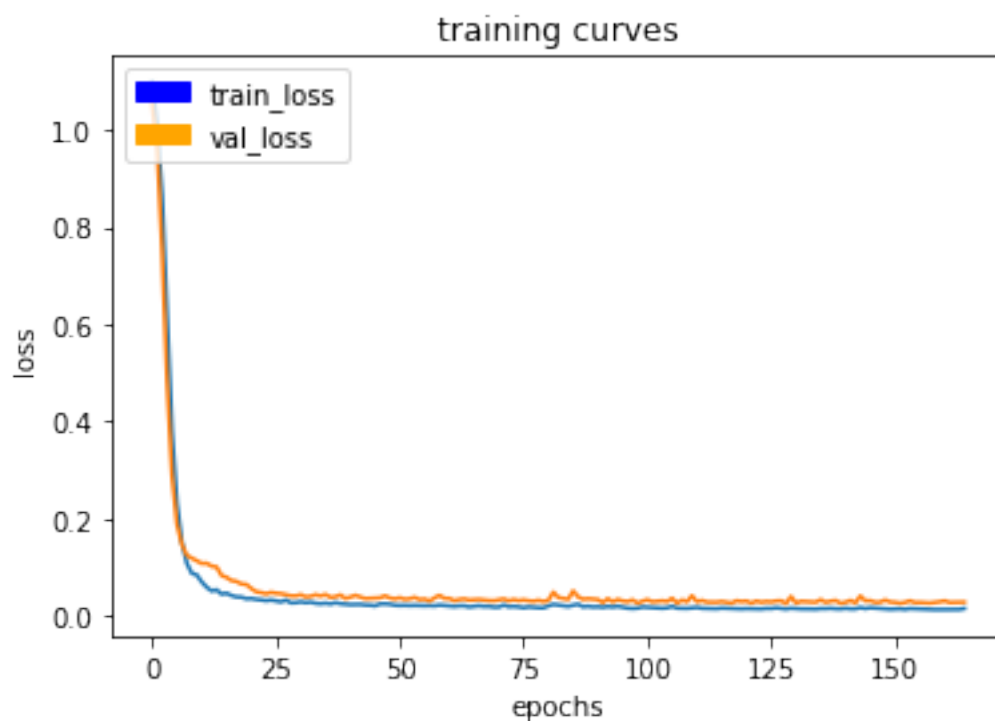
30/30 [=====] - 23s - loss: 0.0125 - val\_loss: 0.0270  
Epoch 164/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0126



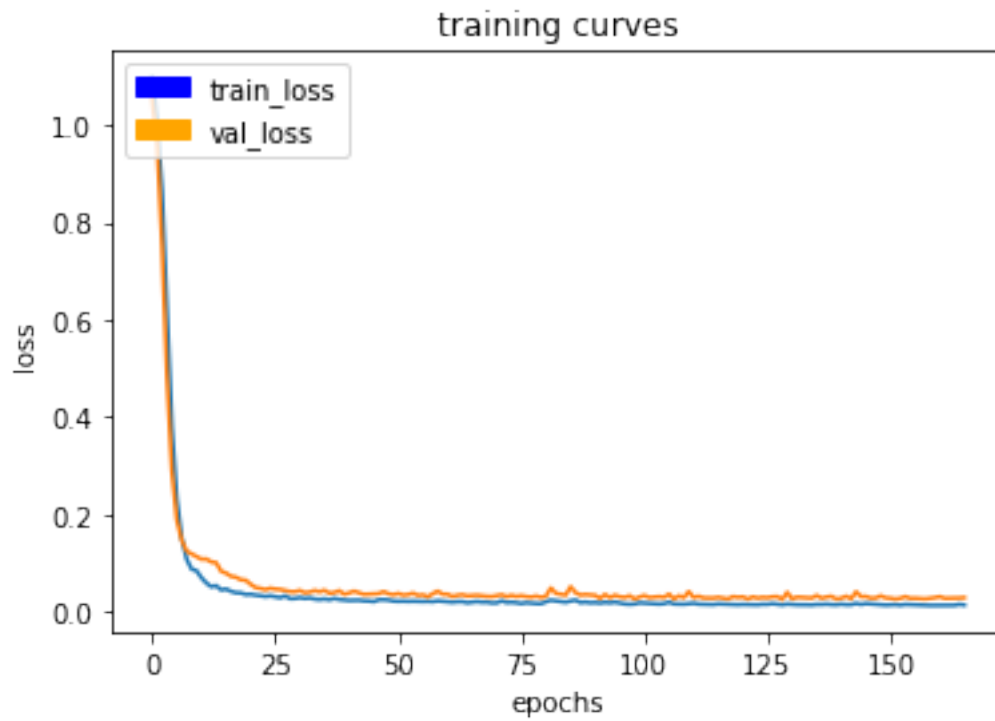
```

30/30 [=====] - 23s - loss: 0.0126 - val_loss: 0.0272
Epoch 165/200
29/30 [=====>.] - ETA: 0s - loss: 0.0147

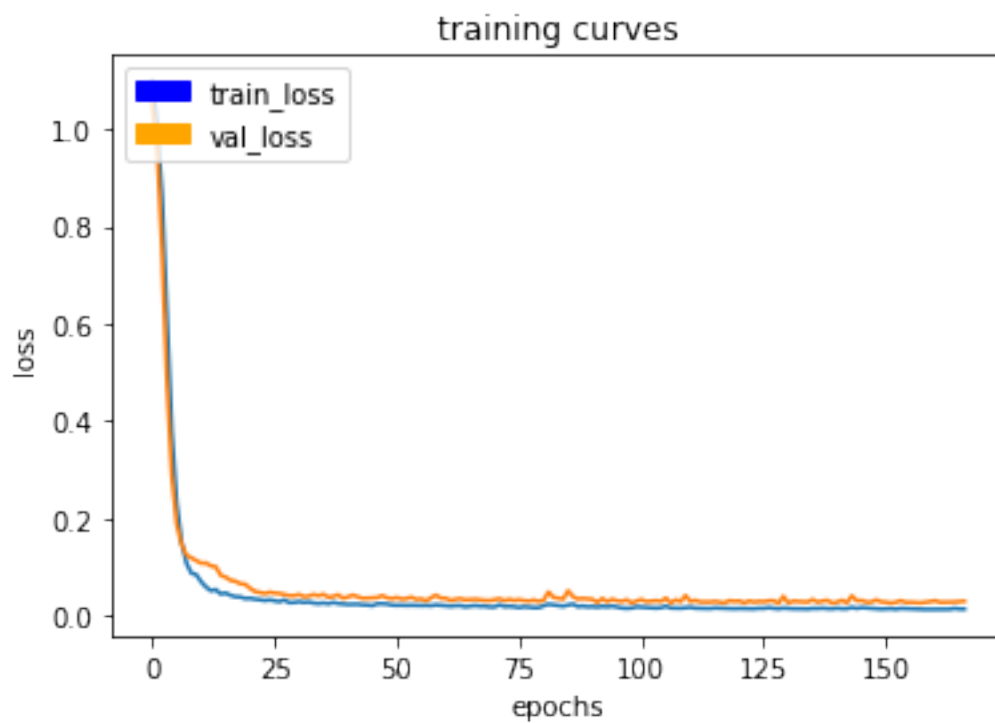
```



30/30 [=====] - 23s - loss: 0.0146 - val\_loss: 0.0276  
Epoch 166/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0131



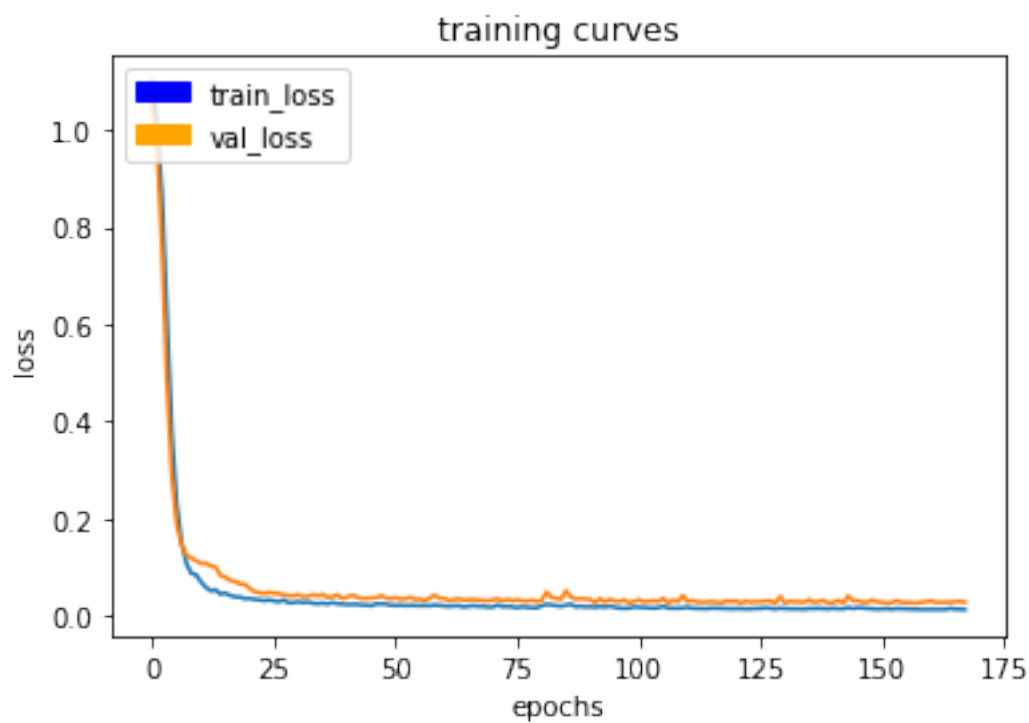
30/30 [=====] - 23s - loss: 0.0131 - val\_loss: 0.0284  
Epoch 167/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0131



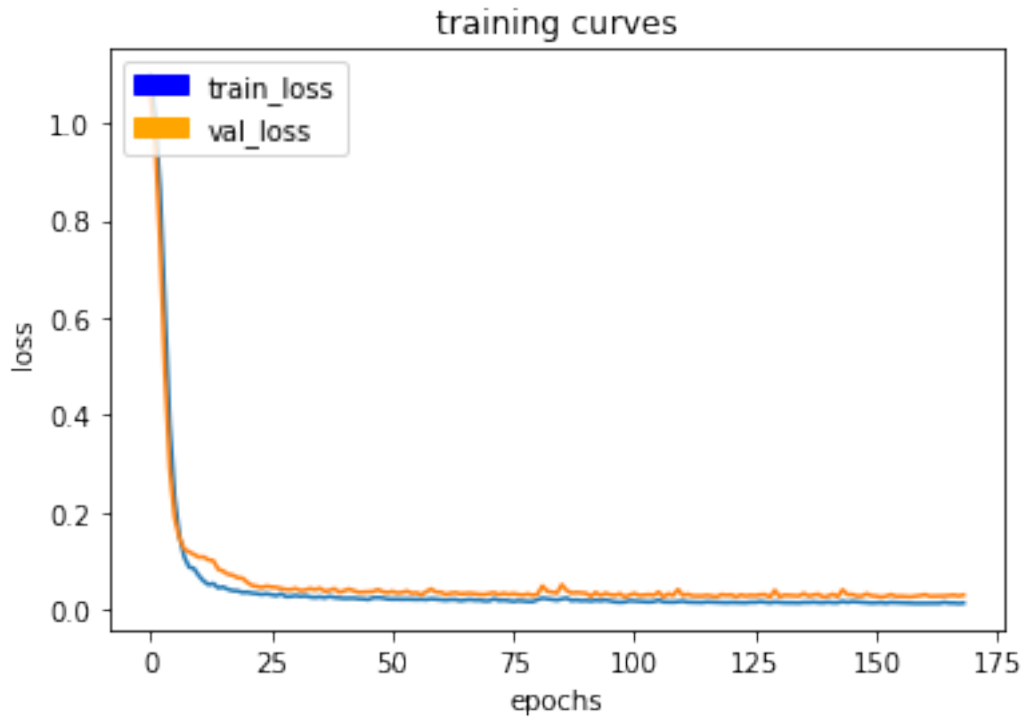
30/30 [=====] - 23s - loss: 0.0131 - val\_loss: 0.0293

Epoch 168/200

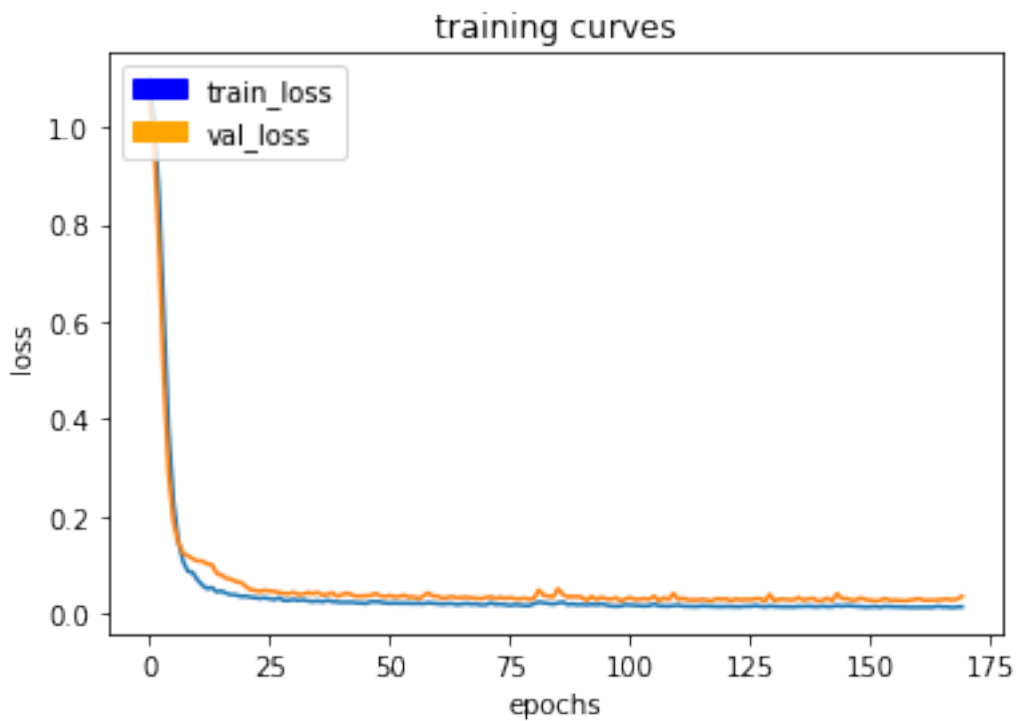
29/30 [=====>.] - ETA: 0s - loss: 0.0121



30/30 [=====] - 23s - loss: 0.0121 - val\_loss: 0.0274  
Epoch 169/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0130



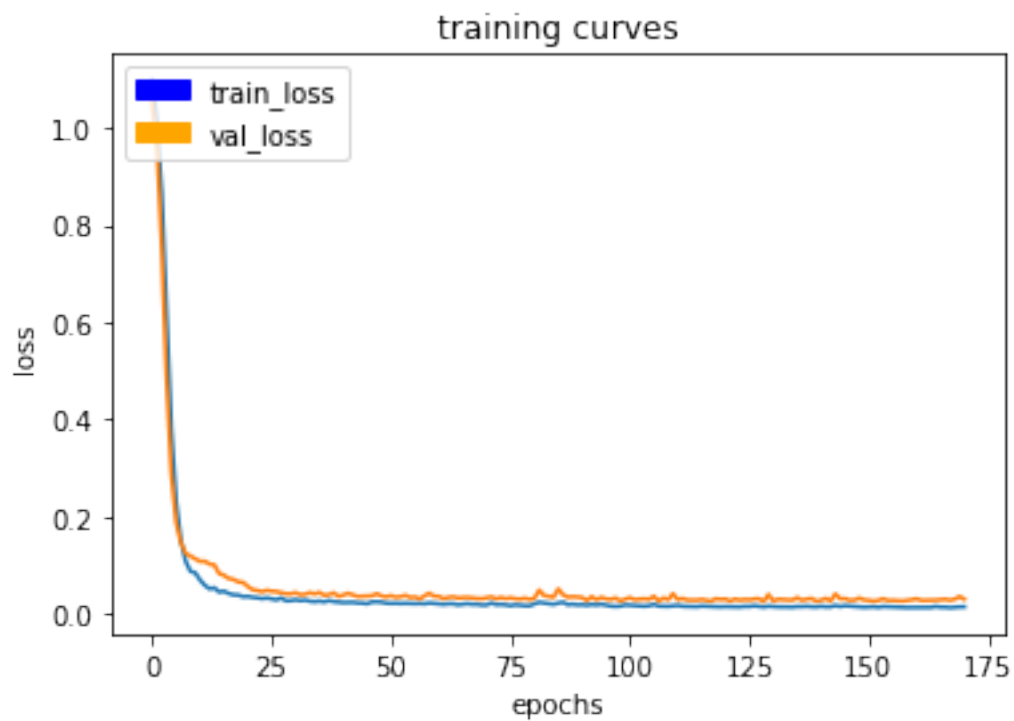
30/30 [=====] - 23s - loss: 0.0130 - val\_loss: 0.0295  
Epoch 170/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0135



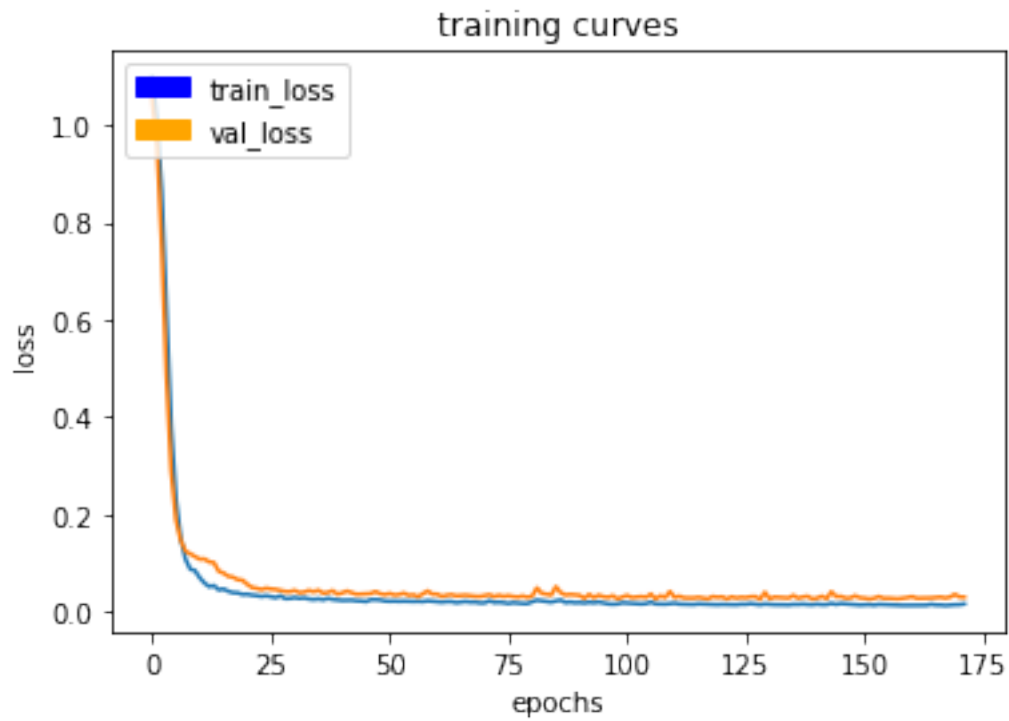
30/30 [=====] - 23s - loss: 0.0136 - val\_loss: 0.0351

Epoch 171/200

29/30 [=====>.] - ETA: 0s - loss: 0.0136

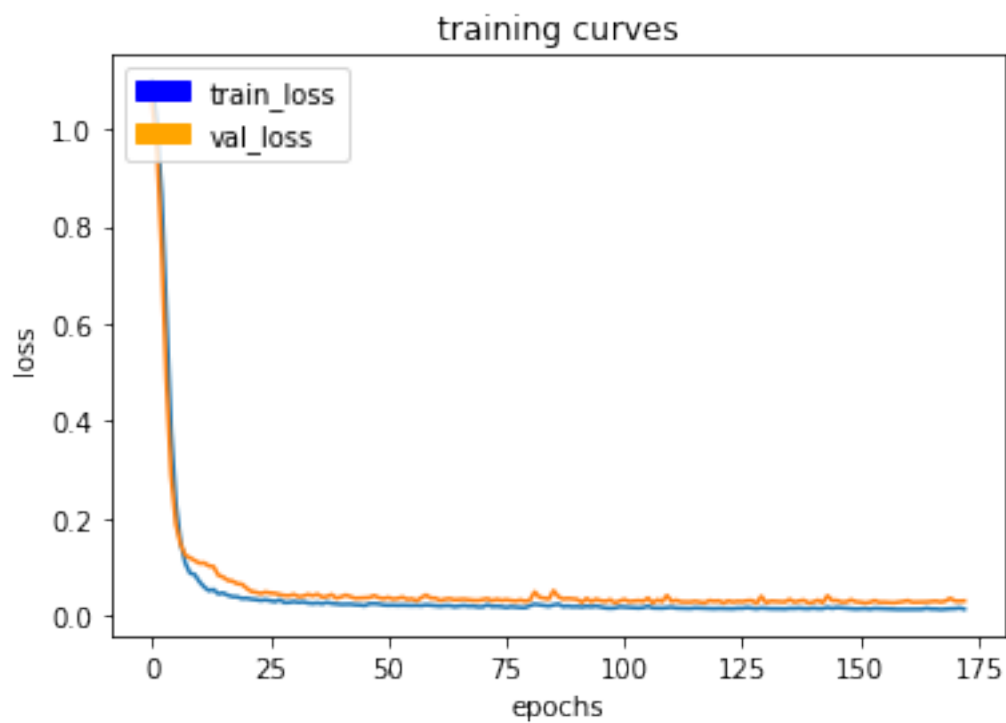


30/30 [=====] - 23s - loss: 0.0138 - val\_loss: 0.0295  
Epoch 172/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0156



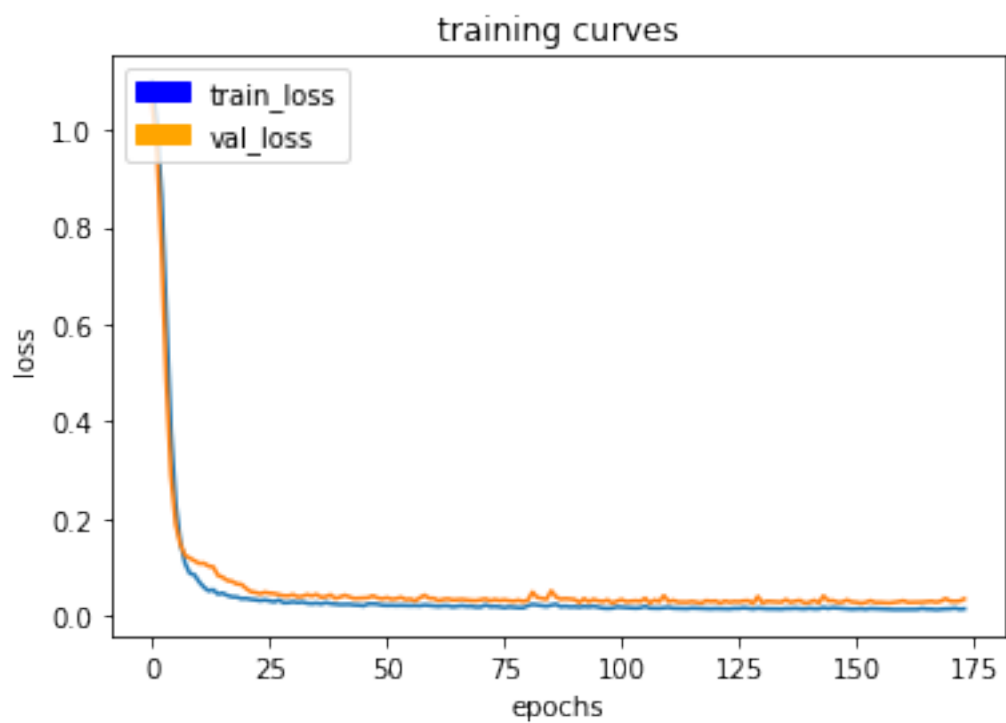
30/30 [=====] - 23s - loss: 0.0155 - val\_loss: 0.0292  
Epoch 173/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0128



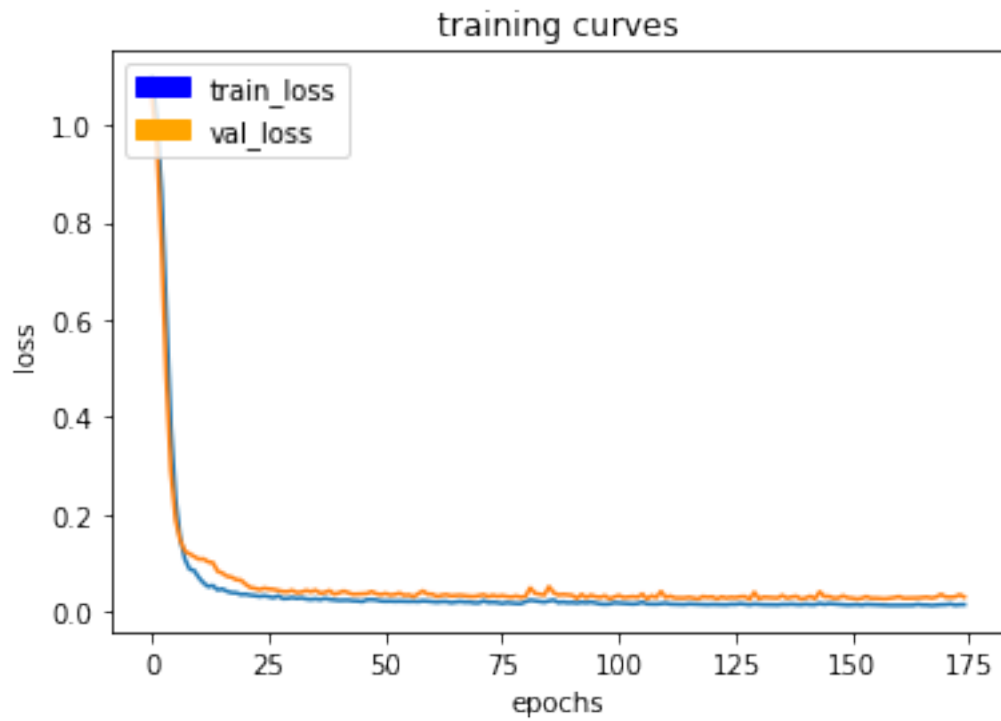


```

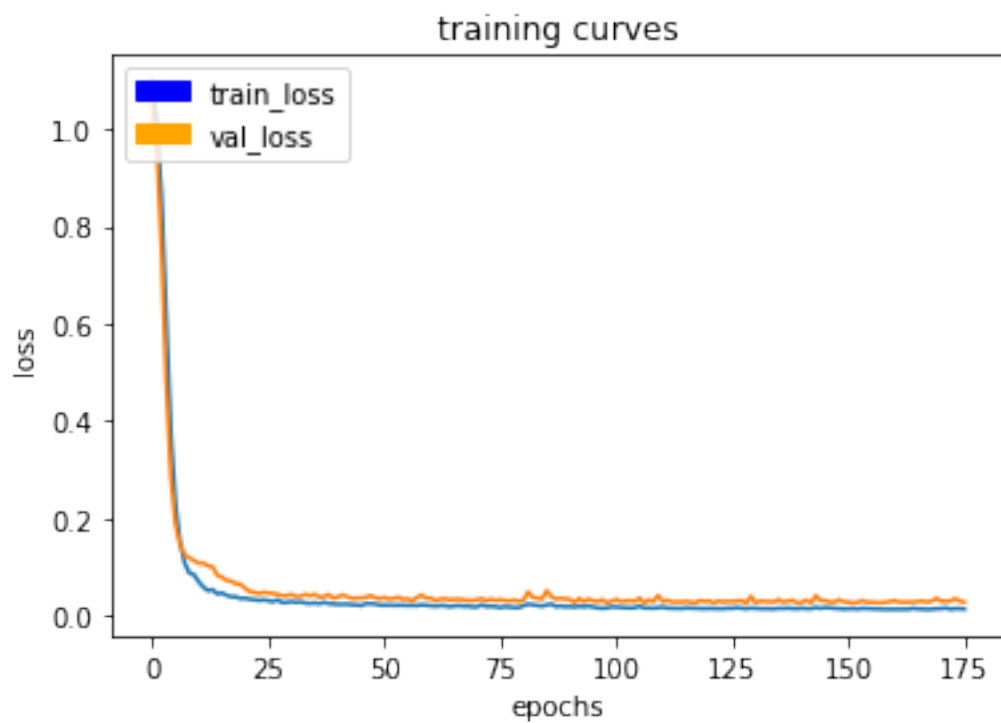
30/30 [=====] - 23s - loss: 0.0127 - val_loss: 0.0298
Epoch 174/200
29/30 [=====>.] - ETA: 0s - loss: 0.0138
  
```



30/30 [=====] - 23s - loss: 0.0138 - val\_loss: 0.0343  
Epoch 175/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0140



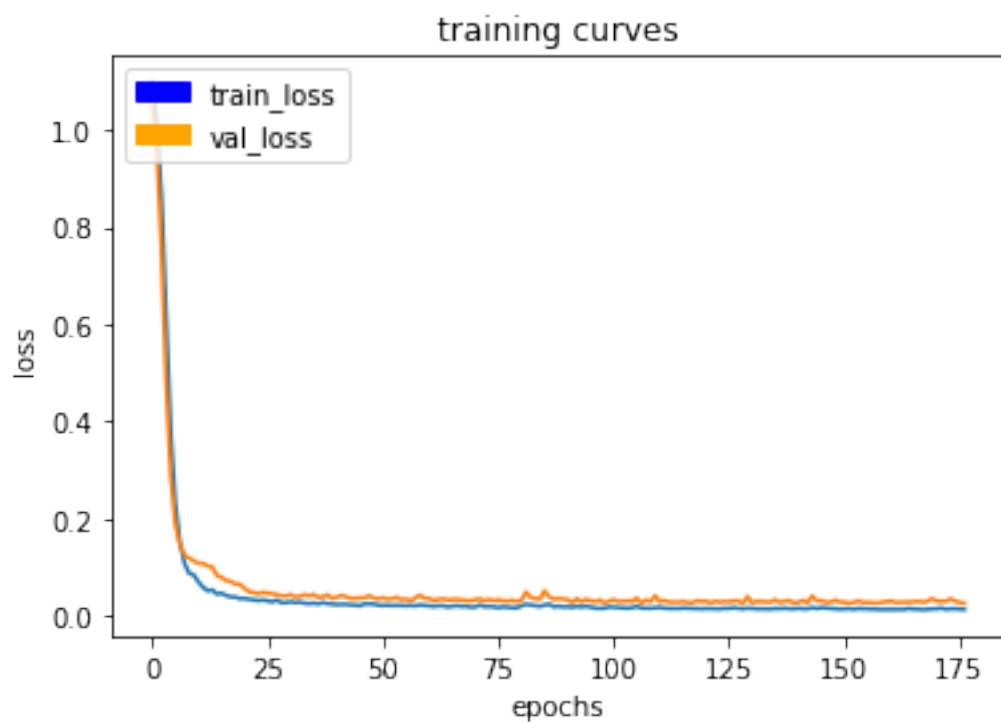
30/30 [=====] - 23s - loss: 0.0139 - val\_loss: 0.0295  
Epoch 176/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0130



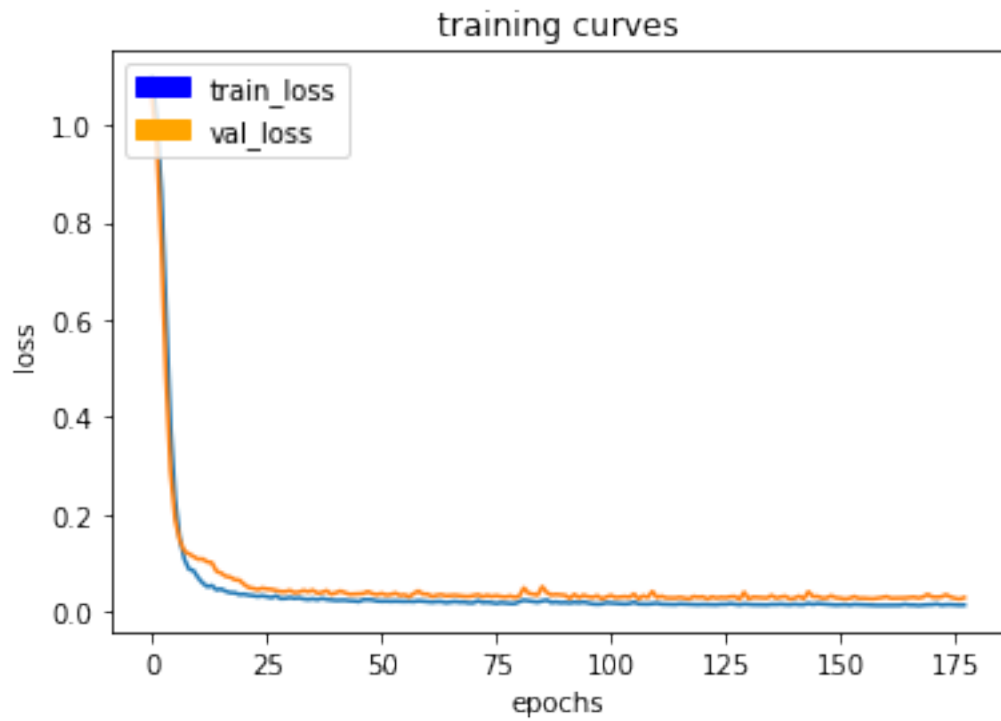
30/30 [=====] - 23s - loss: 0.0130 - val\_loss: 0.0269

Epoch 177/200

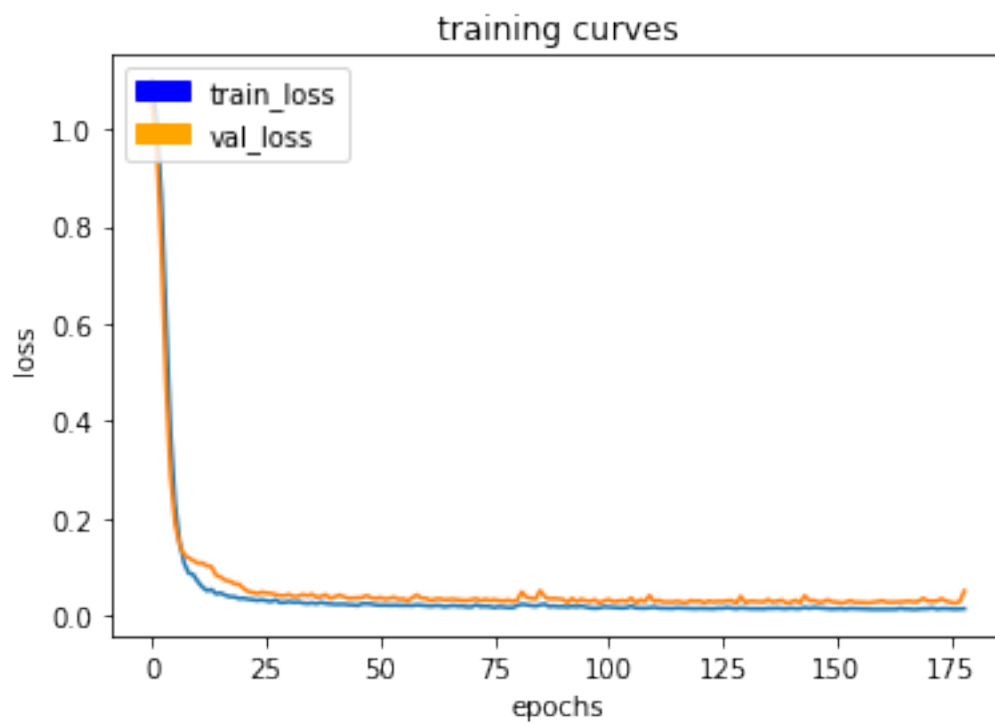
29/30 [=====>.] - ETA: 0s - loss: 0.0126



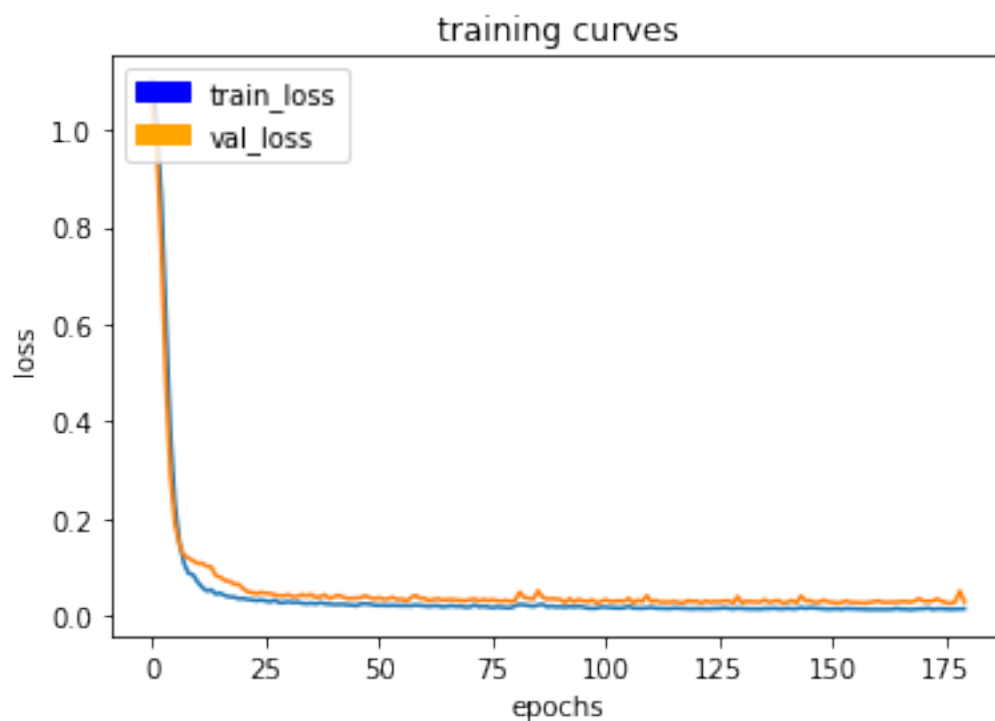
30/30 [=====] - 23s - loss: 0.0126 - val\_loss: 0.0252  
Epoch 178/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0131



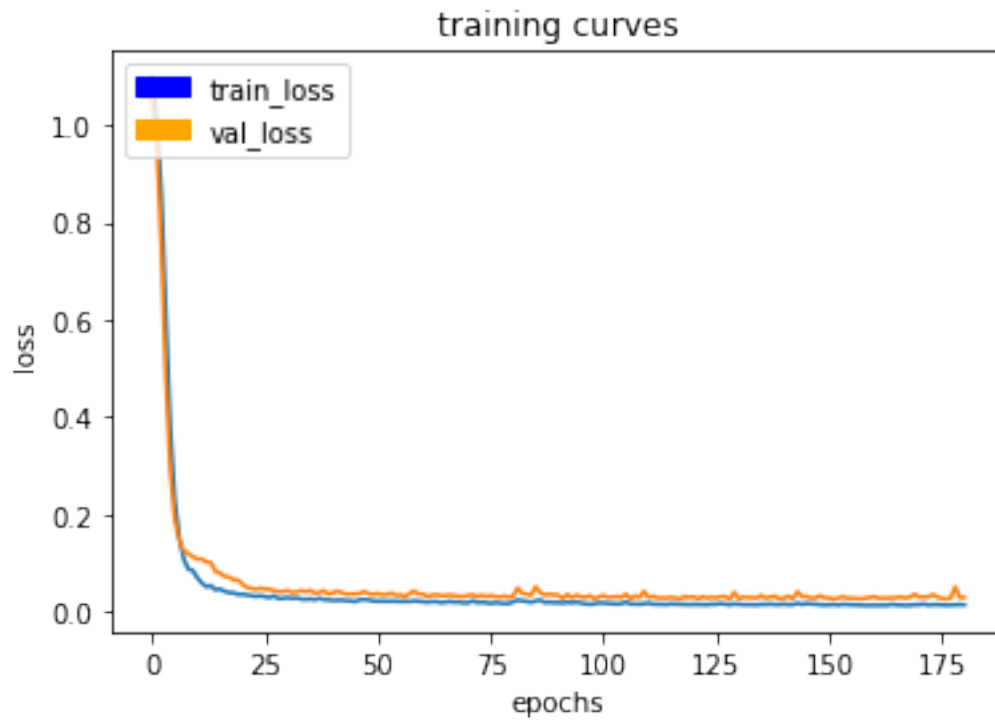
30/30 [=====] - 23s - loss: 0.0131 - val\_loss: 0.0284  
Epoch 179/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0137



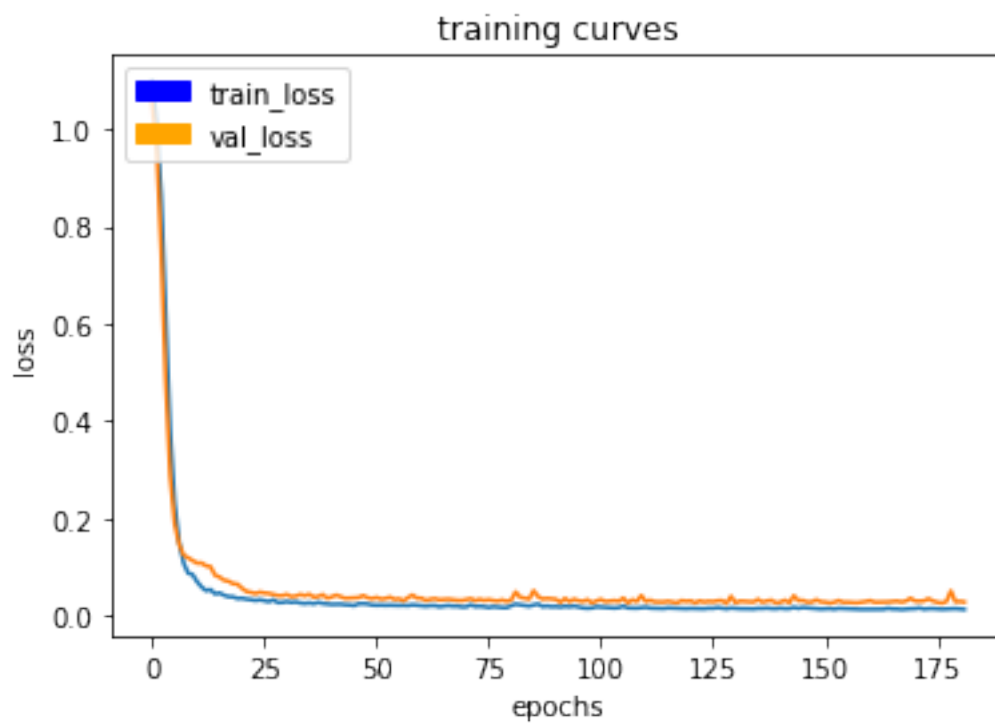
30/30 [=====] - 23s - loss: 0.0137 - val\_loss: 0.0503  
 Epoch 180/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0139



30/30 [=====] - 23s - loss: 0.0139 - val\_loss: 0.0280  
Epoch 181/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0133



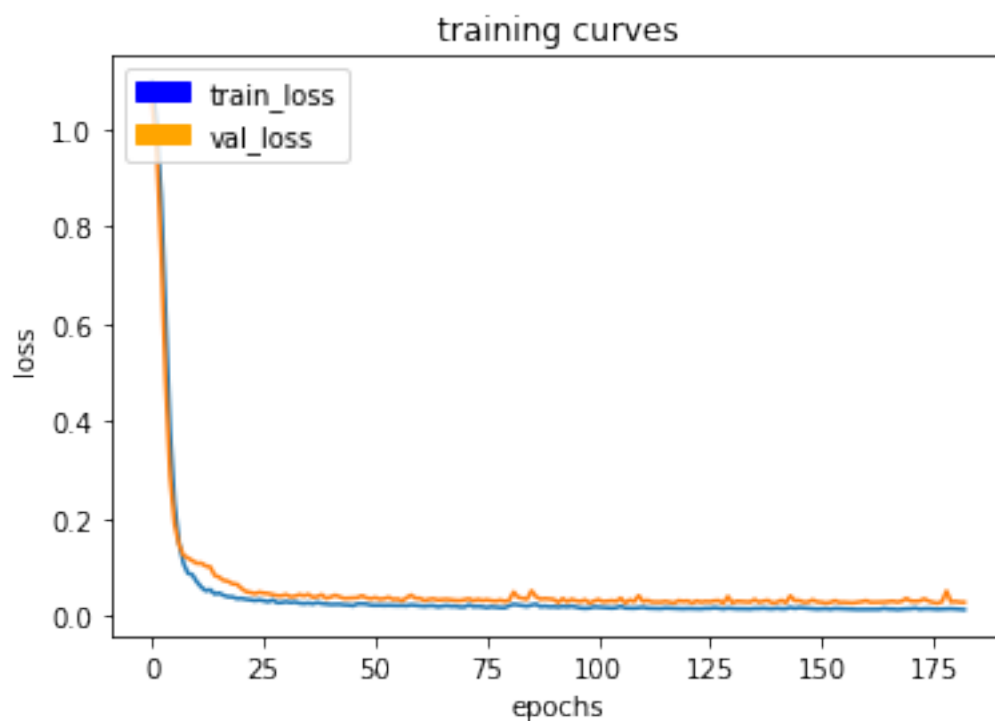
30/30 [=====] - 23s - loss: 0.0133 - val\_loss: 0.0287  
Epoch 182/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0123



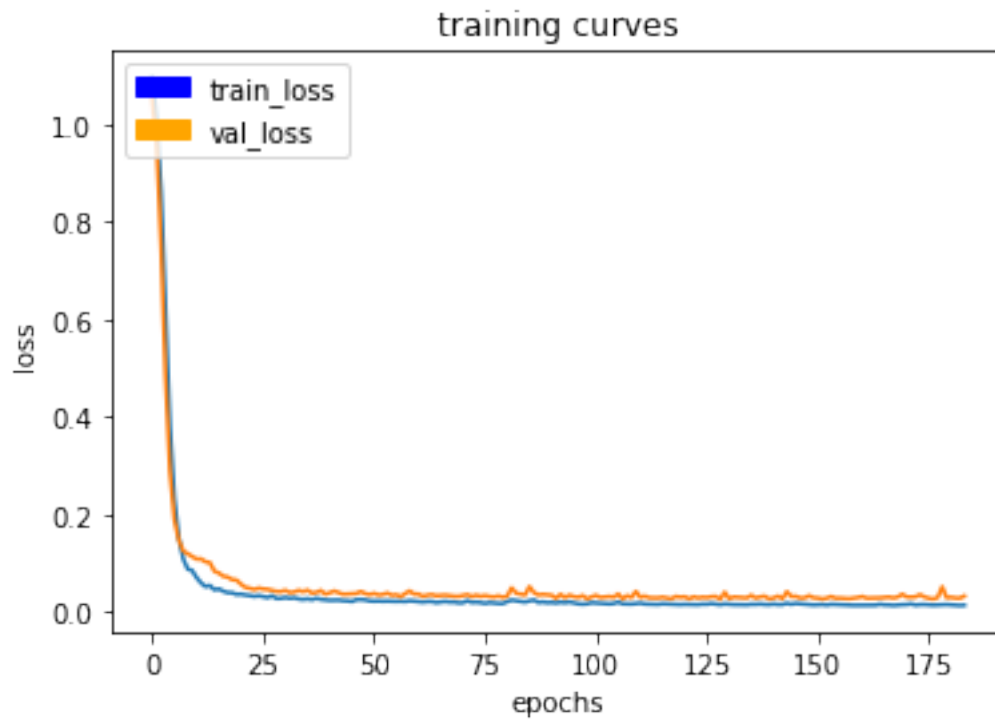
```

30/30 [=====] - 23s - loss: 0.0123 - val_loss: 0.0273
Epoch 183/200
29/30 [=====>.] - ETA: 0s - loss: 0.0120

```

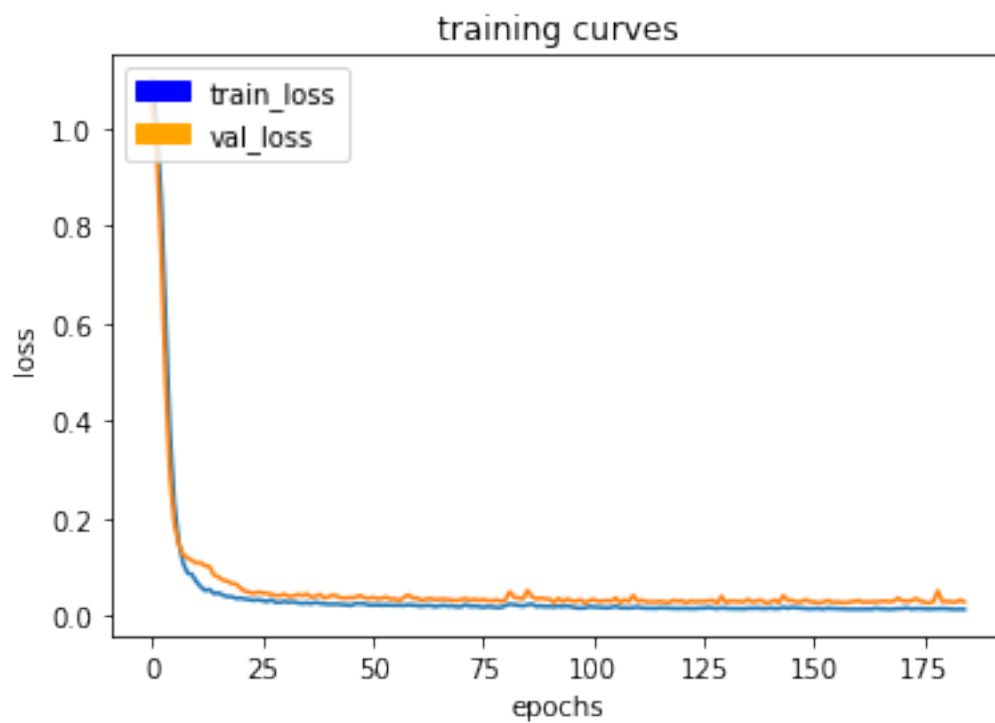


30/30 [=====] - 23s - loss: 0.0120 - val\_loss: 0.0271  
Epoch 184/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0124



30/30 [=====] - 23s - loss: 0.0125 - val\_loss: 0.0313  
Epoch 185/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0119

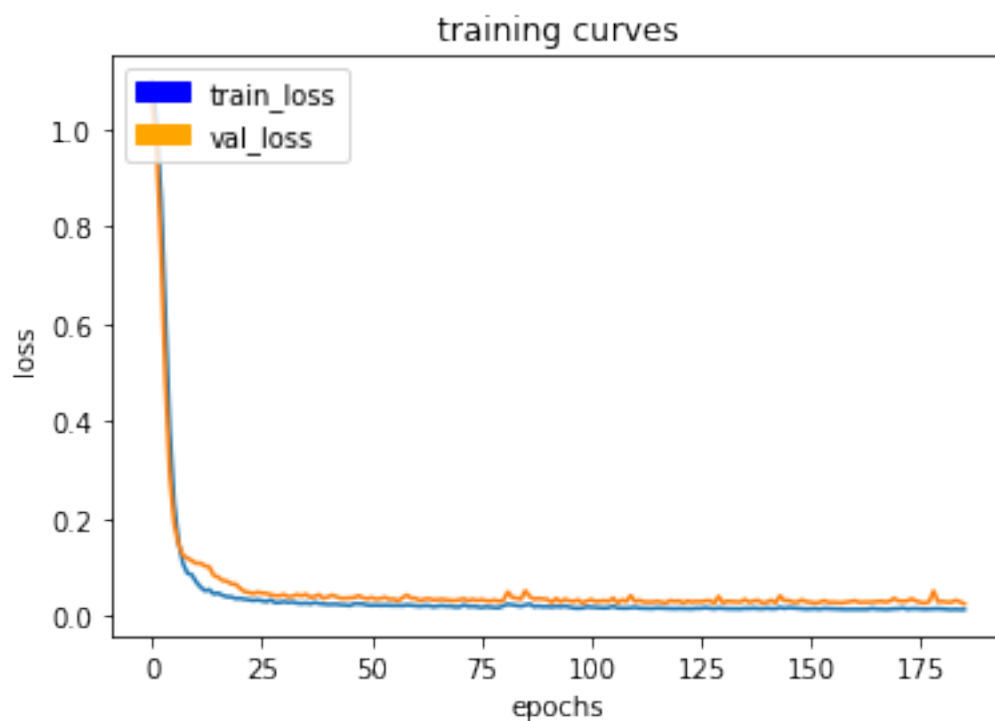




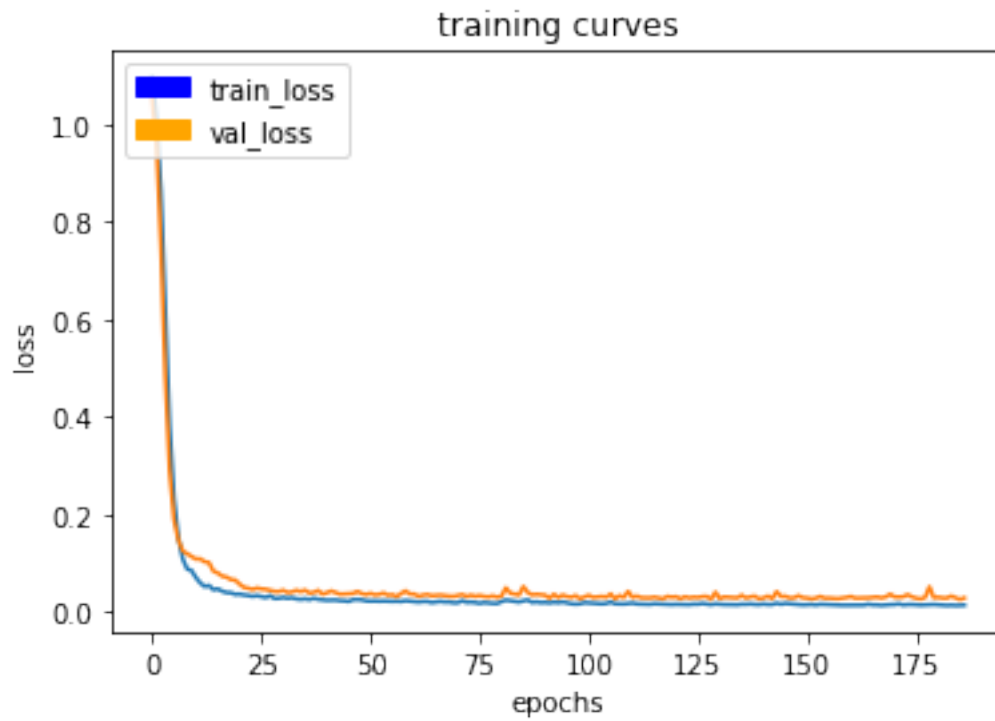
```

30/30 [=====] - 23s - loss: 0.0119 - val_loss: 0.0275
Epoch 186/200
29/30 [=====>.] - ETA: 0s - loss: 0.0122

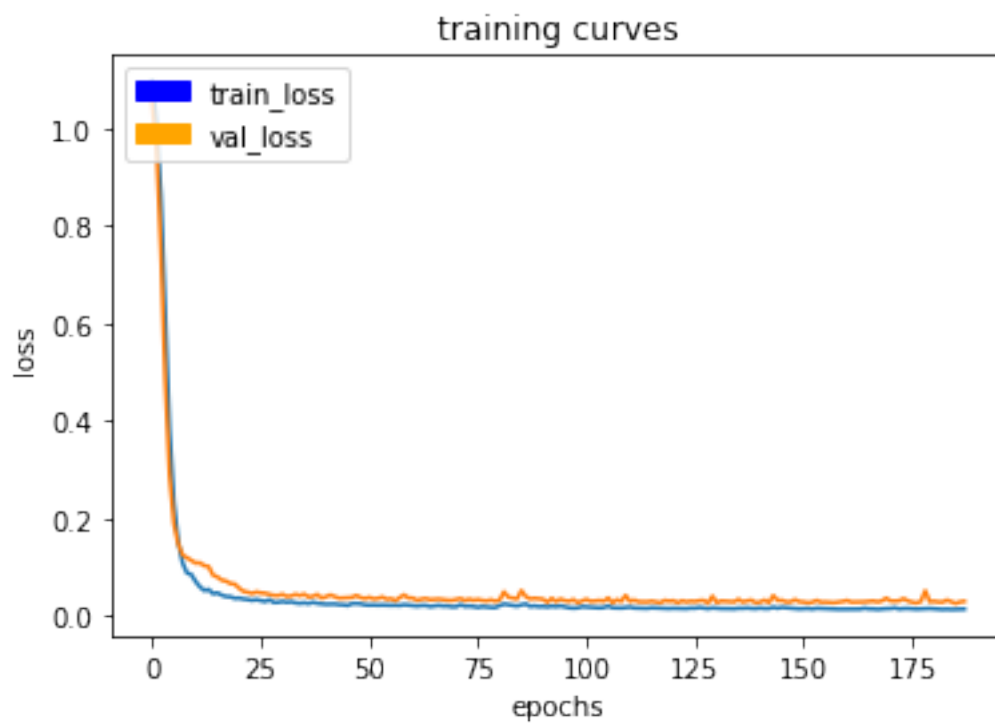
```



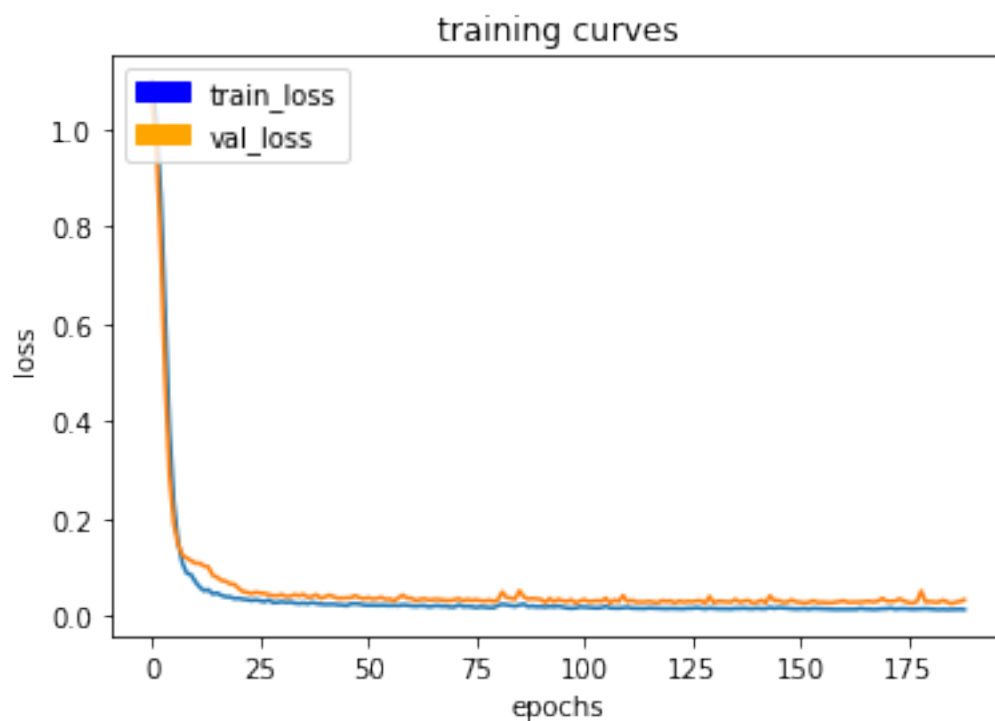
30/30 [=====] - 23s - loss: 0.0122 - val\_loss: 0.0242  
Epoch 187/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0126



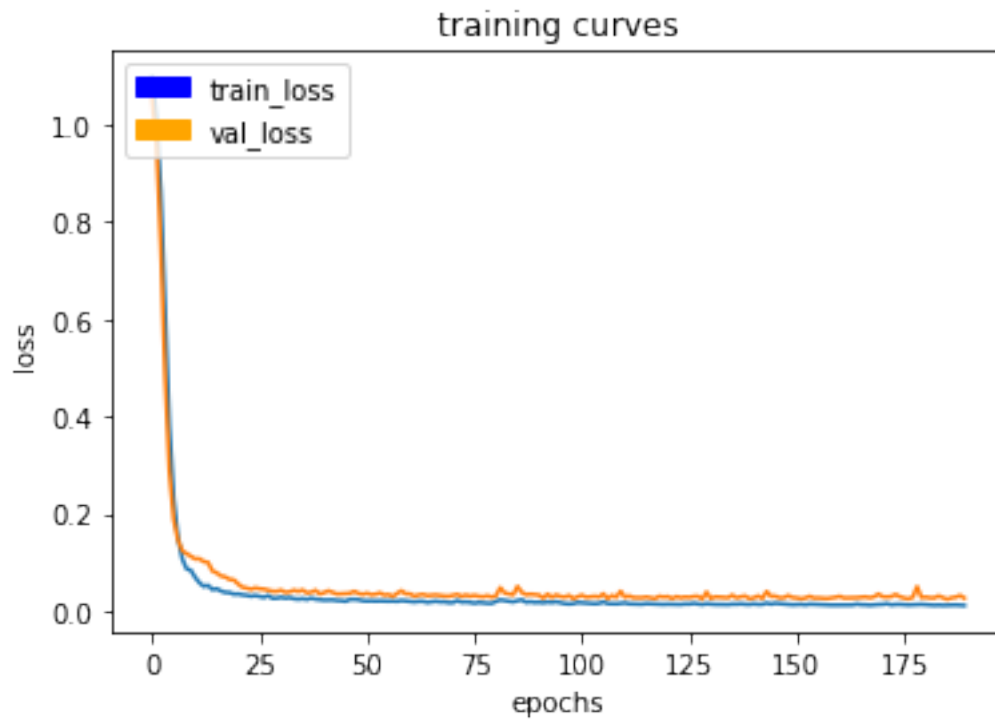
30/30 [=====] - 23s - loss: 0.0125 - val\_loss: 0.0274  
Epoch 188/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0127



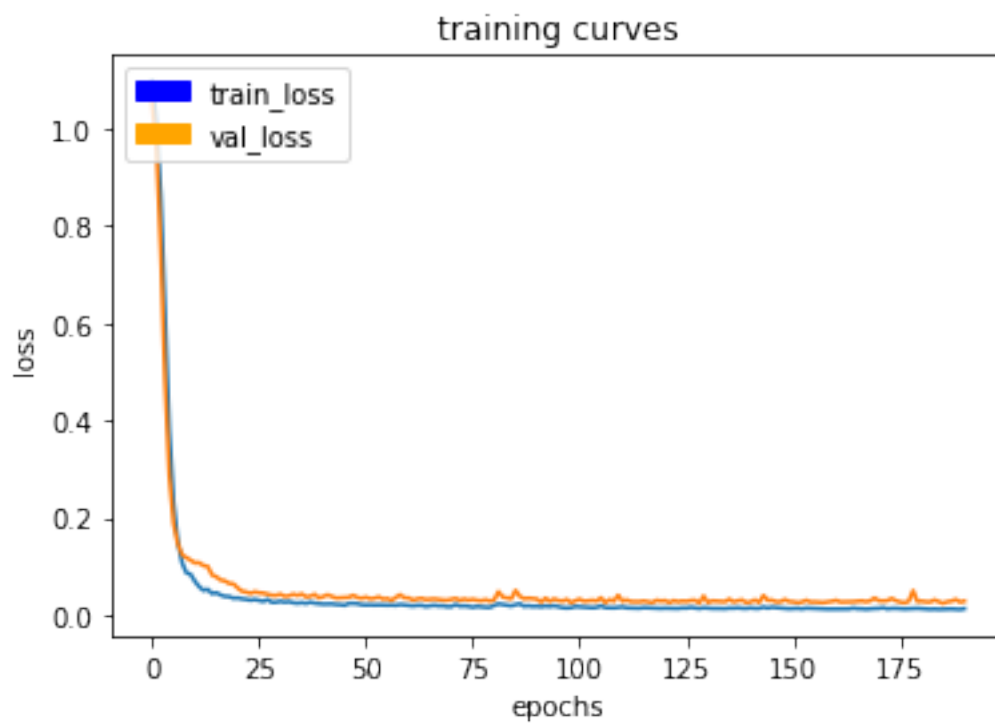
30/30 [=====] - 24s - loss: 0.0127 - val\_loss: 0.0287  
 Epoch 189/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0120



30/30 [=====] - 23s - loss: 0.0121 - val\_loss: 0.0322  
Epoch 190/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0117



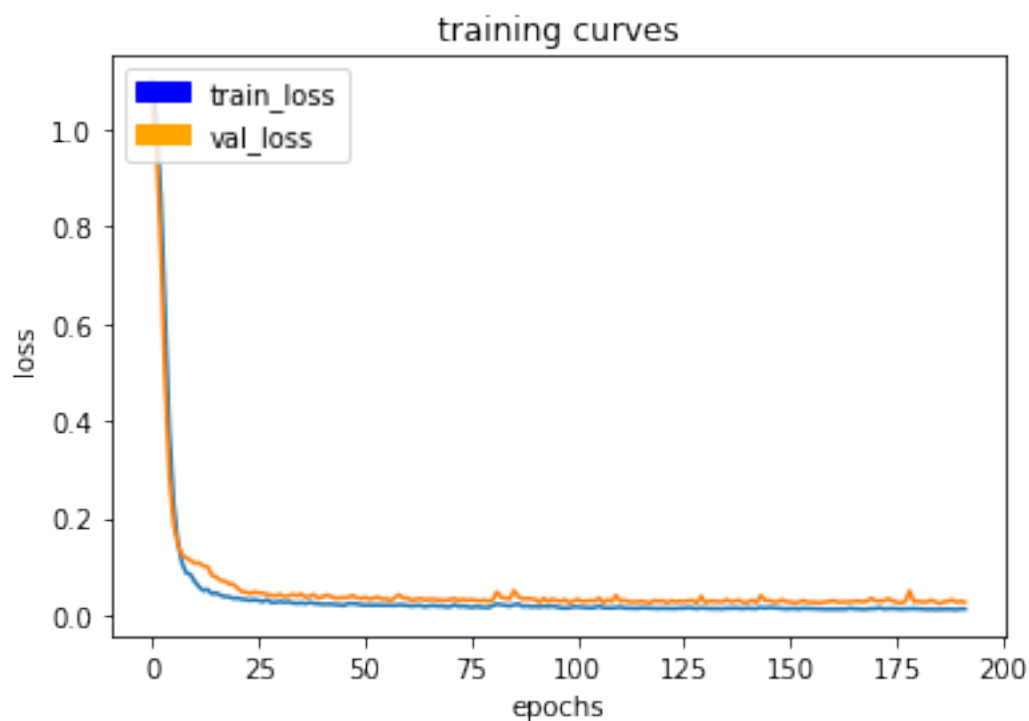
30/30 [=====] - 23s - loss: 0.0116 - val\_loss: 0.0267  
Epoch 191/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0133



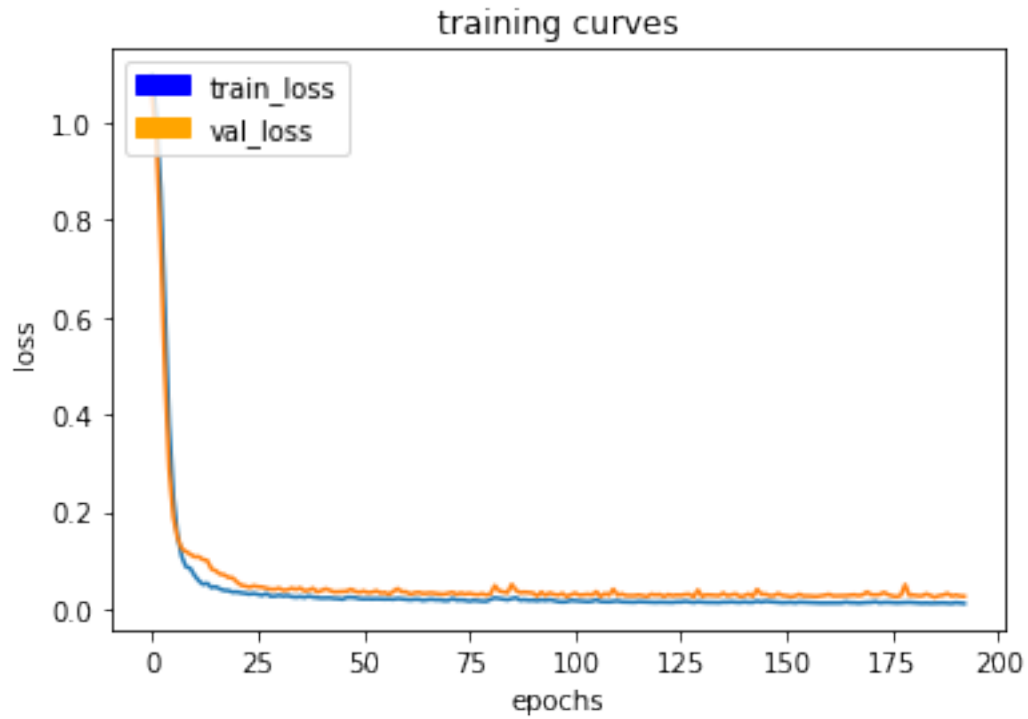
```

30/30 [=====] - 25s - loss: 0.0132 - val_loss: 0.0288
Epoch 192/200
29/30 [=====>.] - ETA: 0s - loss: 0.0127

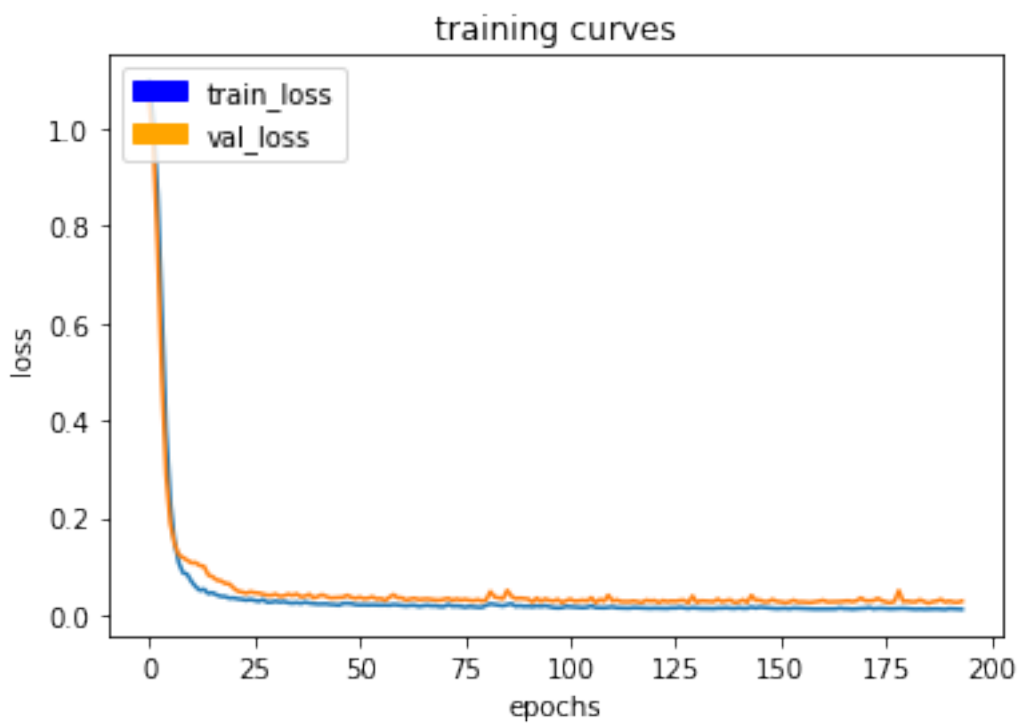
```



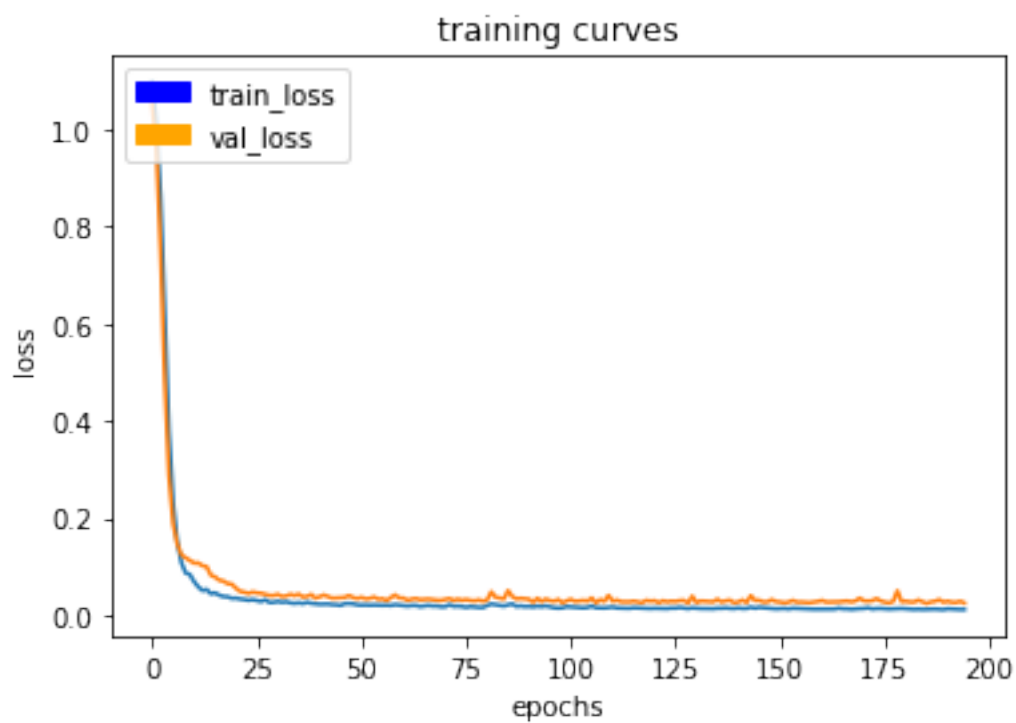
30/30 [=====] - 23s - loss: 0.0126 - val\_loss: 0.0267  
Epoch 193/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0114



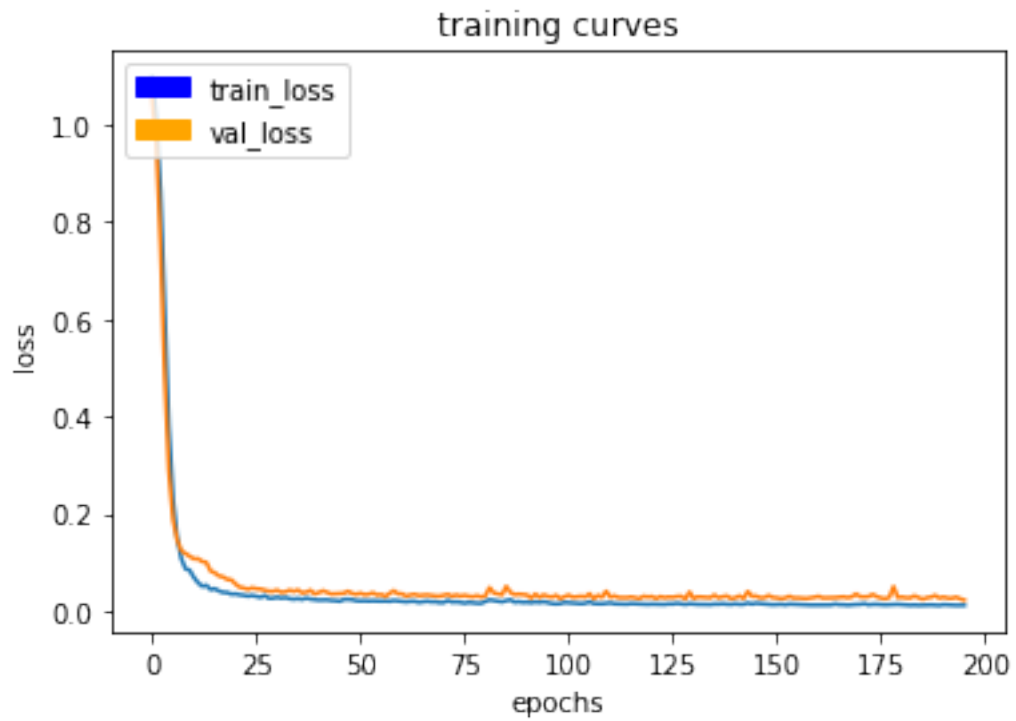
30/30 [=====] - 23s - loss: 0.0115 - val\_loss: 0.0262  
Epoch 194/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0119



30/30 [=====] - 23s - loss: 0.0118 - val\_loss: 0.0286  
 Epoch 195/200  
 29/30 [=====>.] - ETA: 0s - loss: 0.0119

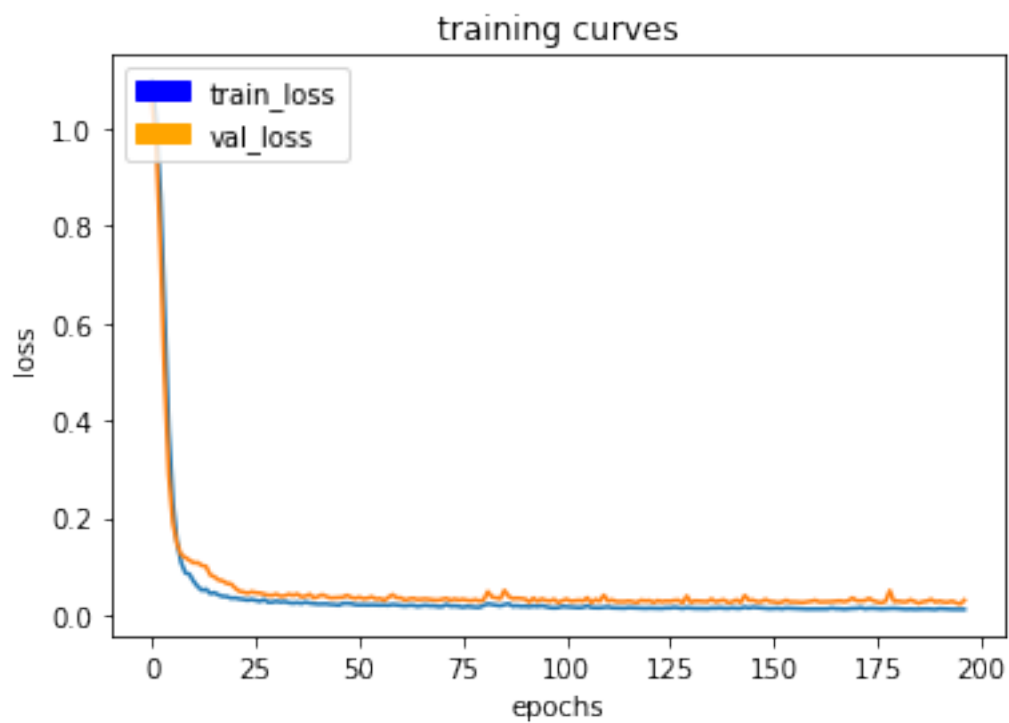


30/30 [=====] - 23s - loss: 0.0119 - val\_loss: 0.0252  
Epoch 196/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0120



30/30 [=====] - 23s - loss: 0.0120 - val\_loss: 0.0225  
Epoch 197/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0116

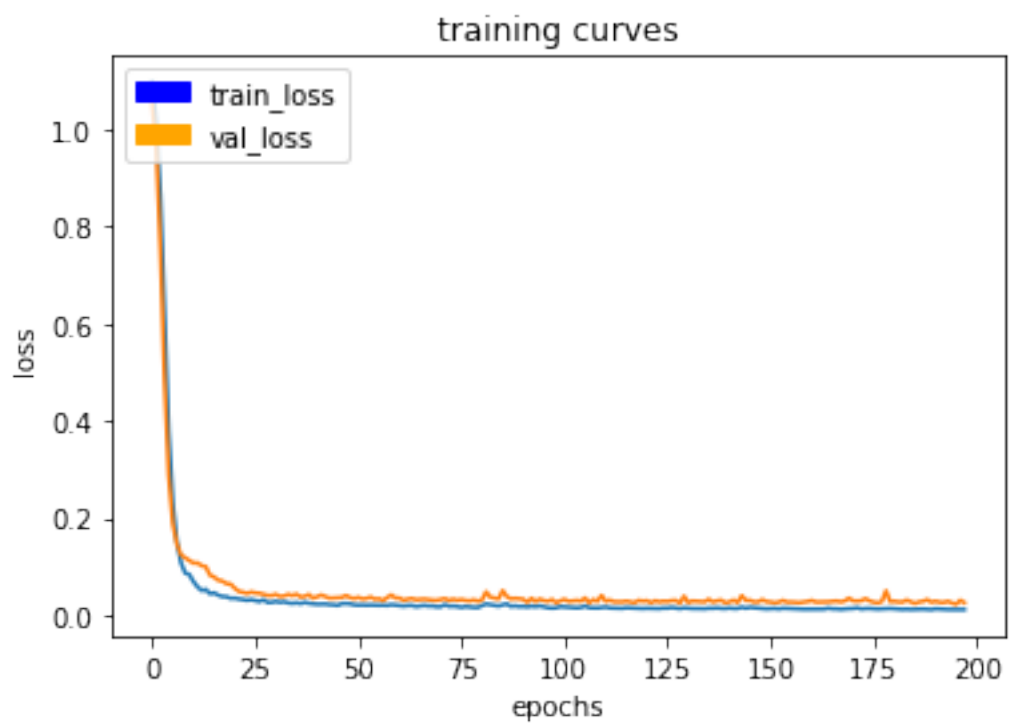




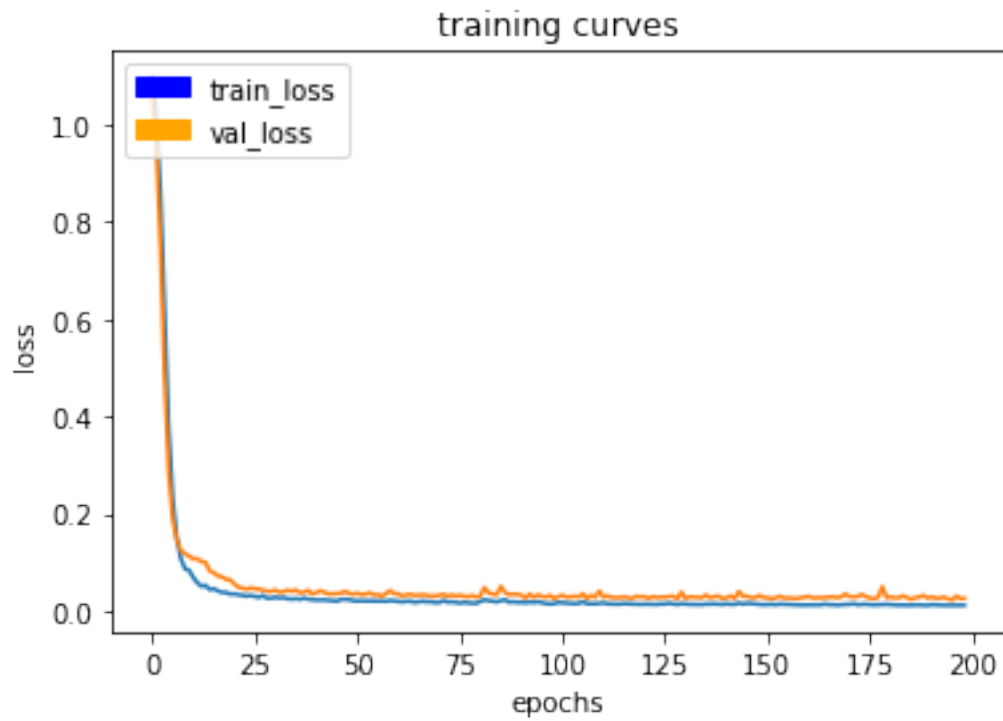
```

30/30 [=====] - 23s - loss: 0.0116 - val_loss: 0.0308
Epoch 198/200
29/30 [=====>.] - ETA: 0s - loss: 0.0116

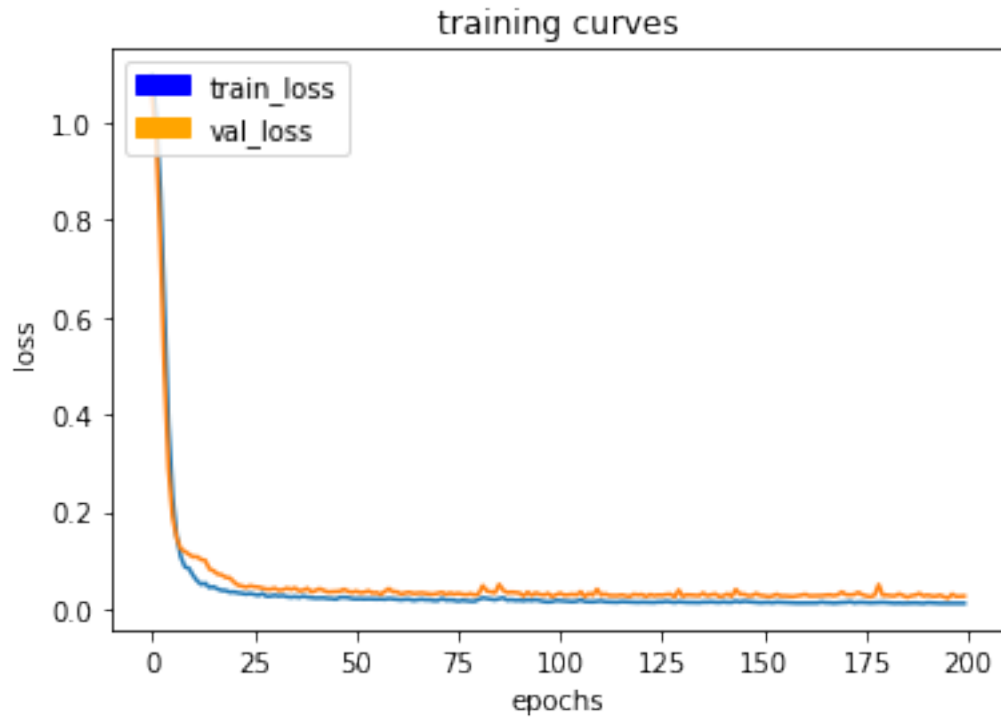
```



30/30 [=====] - 23s - loss: 0.0117 - val\_loss: 0.0258  
Epoch 199/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0119



30/30 [=====] - 23s - loss: 0.0119 - val\_loss: 0.0260  
Epoch 200/200  
29/30 [=====>.] - ETA: 0s - loss: 0.0118



30/30 [=====] - 23s - loss: 0.0118 - val\_loss: 0.0267

Out[118]: <tensorflow.contrib.keras.python.keras.callbacks.History at 0x7f48525cf160>

```
In [119]: # Save your trained model weights
weight_file_name = 'model_weights'
model_tools.save_network(model, weight_file_name)
```

## 1.5 Prediction

Now that you have your model trained and saved, you can make predictions on your validation dataset. These predictions can be compared to the mask images, which are the ground truth labels, to evaluate how well your model is doing under different conditions.

There are three different predictions available from the helper code provided: - **patrol\_with\_targ**: Test how well the network can detect the hero from a distance. - **patrol\_non\_targ**: Test how often the network makes a mistake and identifies the wrong person as the target. - **following\_images**: Test how well the network can identify the target while following them.

In [120]: *# If you need to load a model which you previously trained you can uncomment the code*

```
weight_file_name = 'model_weights'
restored_model = model_tools.load_network(weight_file_name)
```

The following cell will write predictions to files and return paths to the appropriate directories. The `run_num` parameter is used to define or group all the data for a particular model run. You can change it for different runs. For example, 'run\_1', 'run\_2' etc.

```
In [121]: run_num = 'run_1'
```

```
val_with_targ, pred_with_targ = model_tools.write_predictions_grade_set(model,
                                                                    run_num, 'patrol_with_targ', 'sample_evaluation_data')

val_no_targ, pred_no_targ = model_tools.write_predictions_grade_set(model,
                                                                    run_num, 'patrol_non_targ', 'sample_evaluation_data')

val_following, pred_following = model_tools.write_predictions_grade_set(model,
                                                                    run_num, 'following_images', 'sample_evaluation_data')
```

```
-----

TypeError                                Traceback (most recent call last)
```

```
<ipython-input-121-dafef4af7b78> in <module>()
    2
    3 val_with_targ, pred_with_targ = model_tools.write_predictions_grade_set(model,
----> 4                                     run_num, 'patrol_with_targ', 'sample_evaluation_data')
    5
    6 val_no_targ, pred_no_targ = model_tools.write_predictions_grade_set(model,
```

```
TypeError: write_predictions_grade_set() takes 3 positional arguments but 4 were given
```

Now lets look at your predictions, and compare them to the ground truth labels and original images. Run each of the following cells to visualize some sample images from the predictions in the validation set.

```
In [ ]: # images while following the target
im_files = plotting_tools.get_im_file_sample('sample_evaluation_data', 'following_images')
for i in range(3):
    im_tuple = plotting_tools.load_images(im_files[i])
    plotting_tools.show_images(im_tuple)
```

```
In [ ]: # images while at patrol without target
im_files = plotting_tools.get_im_file_sample('sample_evaluation_data', 'patrol_non_targ',)
for i in range(3):
    im_tuple = plotting_tools.load_images(im_files[i])
    plotting_tools.show_images(im_tuple)
```

```
In [ ]:
    # images while at patrol with target
    im_files = plotting_tools.get_im_file_sample('sample_evaluation_data','patrol_with_targ')
    for i in range(3):
        im_tuple = plotting_tools.load_images(im_files[i])
        plotting_tools.show_images(im_tuple)
```

## 1.6 Evaluation

Evaluate your model! The following cells include several different scores to help you evaluate your model under the different conditions discussed during the Prediction step.

```
In [ ]: # Scores for while the quad is following behind the target.
    true_pos1, false_pos1, false_neg1, iou1 = scoring_utils.score_run_iou(val_following, pred_
```

```
In [ ]: # Scores for images while the quad is on patrol and the target is not visible
    true_pos2, false_pos2, false_neg2, iou2 = scoring_utils.score_run_iou(val_no_targ, pred_
```

```
In [ ]: # This score measures how well the neural network can detect the target from far away
    true_pos3, false_pos3, false_neg3, iou3 = scoring_utils.score_run_iou(val_with_targ, pre
```

```
In [ ]: # Sum all the true positives, etc from the three datasets to get a weight for the score
    true_pos = true_pos1 + true_pos2 + true_pos3
    false_pos = false_pos1 + false_pos2 + false_pos3
    false_neg = false_neg1 + false_neg2 + false_neg3

    weight = true_pos/(true_pos+false_neg+false_pos)
    print(weight)
```

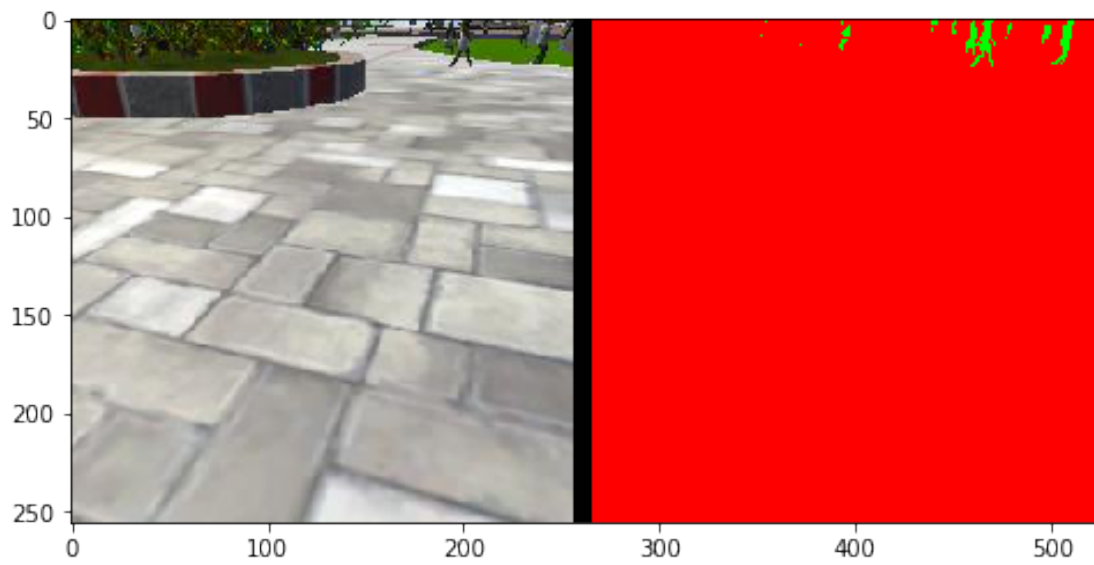
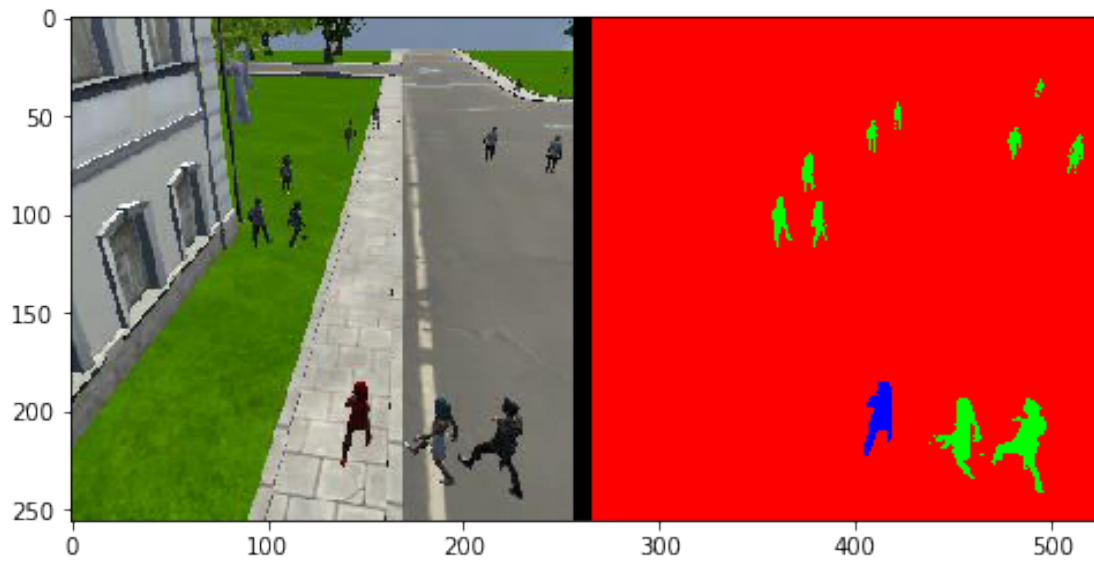
```
In [ ]: # The IoU for the dataset that never includes the hero is excluded from grading
    final_IoU = (iou1 + iou3)/2
    print(final_IoU)
```

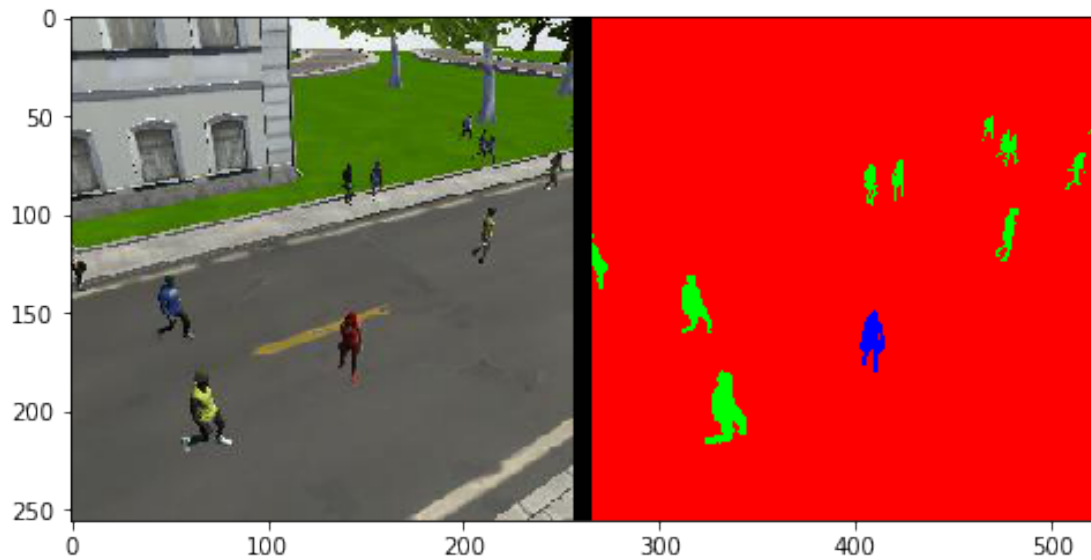
```
In [ ]: # And the final grade score is
    final_score = final_IoU * weight
    print(final_score)
```

```
In [122]: #Test from segmentation lab

    run_number = 'run1'
    validation_path, output_path = model_tools.write_predictions_grade_set(model,run_numbe
```

```
In [123]: # take a look at predictions
    # validation_path = 'validation'
    im_files = plotting_tools.get_im_file_sample(run_number,validation_path)
    for i in range(3):
        im_tuple = plotting_tools.load_images(im_files[i])
        plotting_tools.show_images(im_tuple)
```





```
In [124]: scoring_utils.score_run(validation_path, output_path)
```

```
number of validation samples intersection over the union evaulated on 1184
average intersection over union for background is 0.9931917692038189
average intersection over union for other people is 0.3952889774577268
average intersection over union for hero is 0.16170155977126485
global average intersection over union is 0.5167274354776036
```