

# Text skew correction

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### Introduction

It happens a lot in OCR projects that the image isn't captured correctly or it is rotated manually and we need to skew it correctly and rotate so that the lines become horizontal. In this task, we are asked to correct the image containing textual data and rotate it correctly.

### Challenges

- Provide a dataset for the model to train.
- Design proper model.

### Dataset

For providing training and test Dataset we use two methods.

First, we collect some pictures from the internet and also I myself took some pictures.

Then we resize the image to 512\*512 format and also we use augmentation methods to increase the size of the dataset we rotate each picture with 32 different angles. The data set is provided in the following link [data\\_set](#)

Code for the process of augmentation and saving is in [data\\_generation.py](#)

### [Model](#)

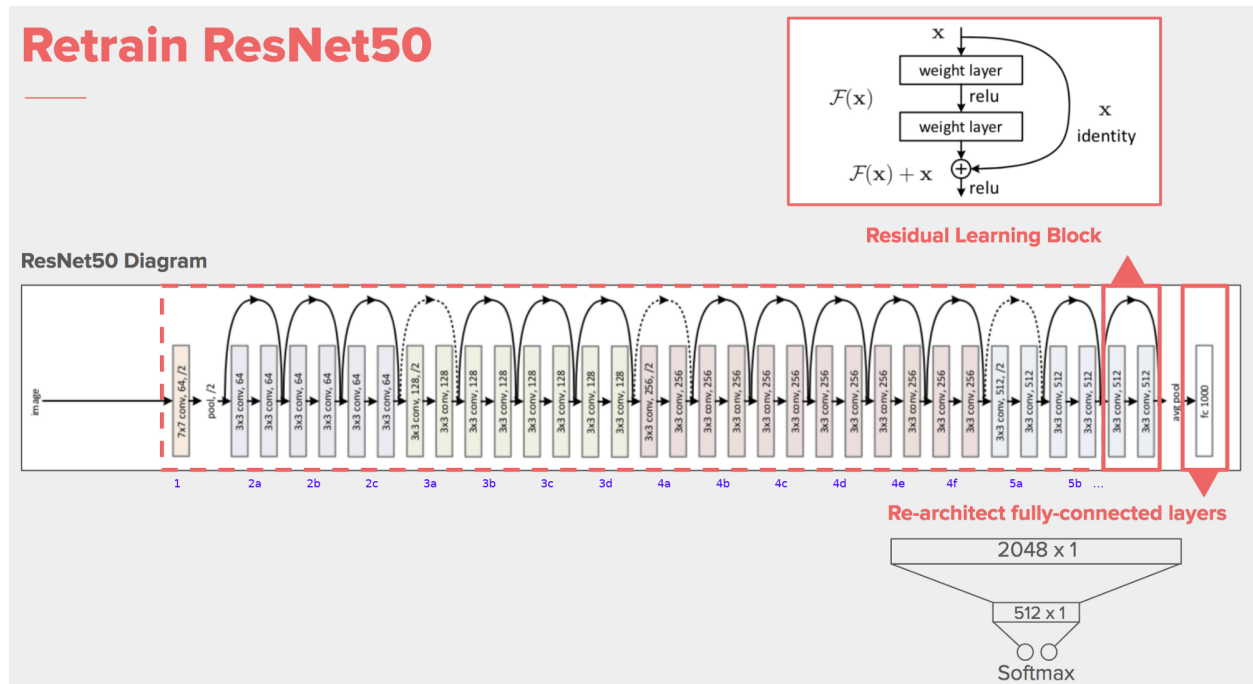
We design a model that recognizes the angle that image's been rotated or captured badly.

So we can rotate the image with the opposite angle and have the corrected image.

With this definition of the problem our problem becomes the **regression** type problem.

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The idea is Transfer Learning. Using ResNet50 and removing the last layer of FC we replace the last layer with FC with only one output\_features with bounded Relu as activation (bounded between 0 and 359)



Instead of Softmax and 1000-FC output, we have FC with only one output, and instead of softmax we use bounded Relu activation.

We chose MSE loss for training. I tried logarithmic loss but I didn't get a good result, i think it's because logarithmic loss doesn't create a good gradient for training but MSE create a greater gradient for training.

[Mode](#) summary is like below

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```
Model: "sequential_4"
```

Layer (type)	Output Shape	Param #
=====		
resnet50 (Functional)	(None, 16, 16, 2048)	23587712
flatten_2 (Flatten)	(None, 524288)	0
dense_2 (Dense)	(None, 1)	524289
activation_2 (Activation)	(None, 1)	0
=====		
Total params: 24,112,001		
Trainable params: 524,289		
Non-trainable params: 23,587,712		

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## Train model

We only train the last FC and freeze the other layer in pre-trained ResNet with an ADAM optimizer.

With 20 epoch and a validation set of 10 percent.

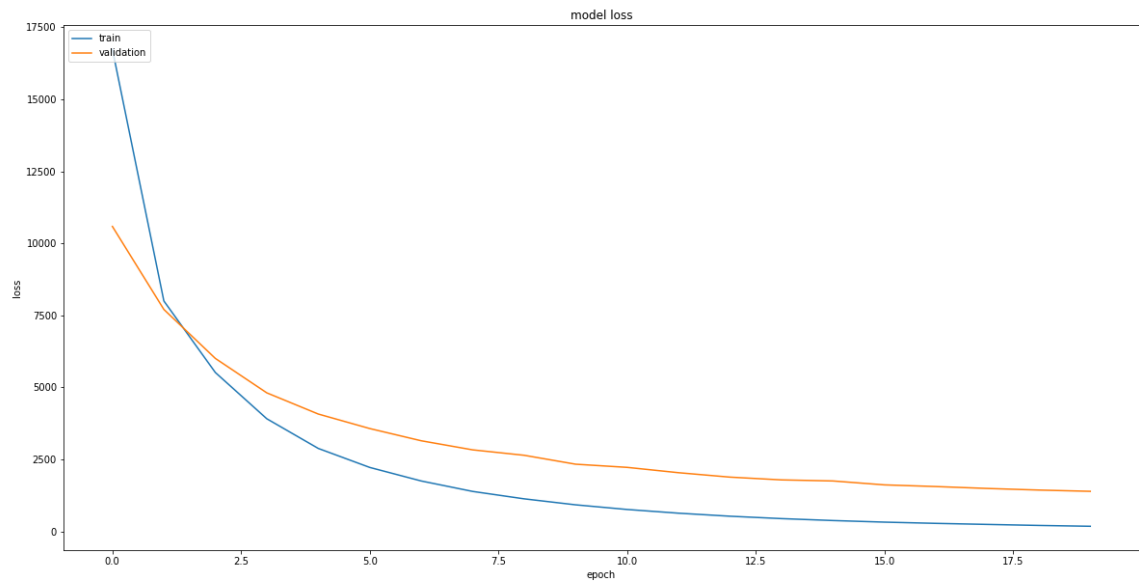
[Python Code](#)

[Colab](#)

## Results

the model is trained with an ADAM optimizer with a learning rate of 10e-4.

The results plot for loss is below



As you see the learning rate is chosen properly. So we have a smooth decrease in loss.

The other thing worth mentioning is that with more 20 epochs the model would be overfitted with train data.