

Behavioral_Cloning_Report

Overview

This project is about behavioral cloning. We trained neural network to drive a car (simulator) around the track on its own.

The design will require building a neural network model, training it using data collected from human driving behavior. Then use the trained neural network model to drive the car.

The goals / steps of this project are the following:

- Use the simulator to collect data of good driving behavior
- Build, a convolution neural network in Keras that predicts steering angles from images
- Train and validate the model with a training and validation set
- Test that the model successfully drives around track one without leaving the road

Collecting data from good driving car behavior

Using provided simulator to drive around the tract to collect data to data directory (data), we drive:

1. 2 laps on center lane
2. 1 lap focusing on smooth drive around the curve
3. 1 lap to recover driving from side to side around the track.

Create convolution Neural Network Models from Keras

We experienced with 4 neural network models :

1. nvidia,
2. lenet1,
3. lenet2,
4. lenet3.

With many iteration the model from Nvidia performed the best on the test drive through simulator.

```
In [ ]: Nvidia model
```

Layer (type)	Output Shape	Param #	Connected to
01. lambda_5 (Lambda)	(None, 160, 320, 3)	0	lambda_input_5[0][0]
02. cropping2d_4 (Cropping2D)	(None, 65, 320, 3)	0	lambda_5[0][0]
03. convolution2d_15 (Convolution2D)	(None, 31, 158, 24)	1824	cropping2d_4[0][0]
04. convolution2d_16 (Convolution2D)	(None, 14, 77, 36)	21636	convolution2d_15[0][0]
05. convolution2d_17 (Convolution2D)	(None, 5, 37, 48)	43248	convolution2d_16[0][0]
06. convolution2d_18 (Convolution2D)	(None, 3, 35, 64)	27712	convolution2d_17[0][0]
07. convolution2d_19 (Convolution2D)	(None, 1, 33, 64)	36928	convolution2d_18[0][0]
08. flatten_5 (Flatten)	(None, 2112)	0	convolution2d_19[0][0]
09. dense_13 (Dense)	(None, 100)	211300	flatten_5[0][0]
10. dense_14 (Dense)	(None, 10)	1010	dense_13[0][0]
11. dense_15 (Dense)	(None, 1)	11	dense_14[0][0]

Reading driving data:

After collecting all driving data, we will have to design away to read the data, to make sure we can handle large amount of data, we use generator to take in one batch of data at a time. Model.py implemented both ways to get data to analyze:

1. create_train_model (data_dir, lambda: nvidia_model(), modelfile, epoch)
2. create_train_model_generator (data_dir, lambda: nvidia_model(), modelfile, epoch) The first function read everything into memory, the second one read 32 data set at a time. Therefore, in order to train the neural network model: Python model.py -m -i training_data_dir/ -o modelsaved.h5 -e epoch

Compiling neural network model:

When it comes to compiling our neural network model, we are using optimizer with parameters set :

1. loss='mean_squared_error'
2. lr=0.001,
3. beta_1=0.9,
4. beta_2=0.999,
5. epsilon=1e-08,
6. decay=0.0.

Preprocessing Data stream:

1. At this time data stream is augmented using augmentation_data()
2. We crop out images to reduce confusion to model at the first stages of processing
3. Test data is 20% of the entire data set.

Example of Augmented Images

```
In [7]: from myLib.BehavioralCloning_ImportFile import *  
data_dir = './data1_4/'
```

Using TensorFlow backend.

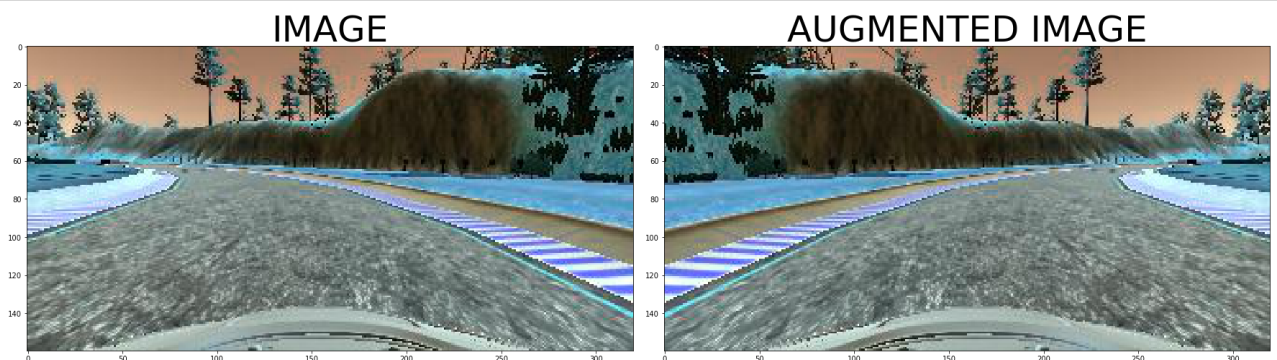
```
In [8]: images, measurements = load_data (data_dir)
```

Data Set Length is : 1341 4023 4023

```
In [11]: augmented_images, augmented_measurements = [], []  
augmented_images, augmented_measurements = augmentation_data(images, measurements)
```

Data Set Length is : 8046 8046

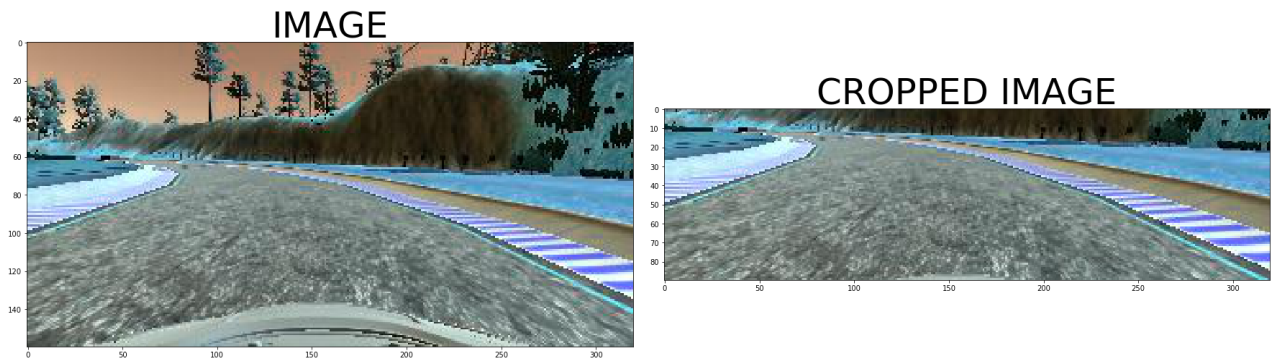
```
In [12]: %matplotlib inline  
plot_2_images (images[0], augmented_images[1], "IMAGE", "AUGMENTED IMAGE")
```



Example of Cropping Images

```
In [13]: crop_img = images[0]
crop_img1 = crop_img[50:140, 0:320] # Crop from x, y, w, h -> 100, 200, 300, 400
```

```
In [14]: plot_2_images (images[0], crop_img1, "IMAGE", "CROPPED IMAGE")
```



Discussion

As a old farmer come to the city, it was amazing to see just by

1. training the simulator,
2. collecting the training data,
3. creating a convolutional network model,
4. train the model with collected data,
5. the car now can be controlled by software !

In this project, changing different models their hyper parameters but using same training data we could not see performing change. However, a set of correct training data is the key toward right solution.

Improvement on the operation was significant when we crop the image, augmented images and use generator to create model. Would like to thank my awesome reviewer on

1. fixing drive.py on BGR issues
2. cropping the images earlier in process
3. implement check point to stop the training process earlier

Saved me hours of work.

References

In []: LeNet3 model

Layer (type)	Output Shape	Param #	Connected to
lambda_1 (Lambda)	(None, 160, 320, 3)	0	lambda_input_1[0][0]
cropping2d_1 (Cropping2D)	(None, 65, 320, 3)	0	lambda_1[0][0]
convolution2d_1 (Convolution2D)	(None, 61, 316, 6)	456	cropping2d_1[0][0]
maxpooling2d_1 (MaxPooling2D)	(None, 30, 158, 6)	0	convolution2d_1[0][0]
convolution2d_2 (Convolution2D)	(None, 26, 154, 6)	906	maxpooling2d_1[0][0]
maxpooling2d_2 (MaxPooling2D)	(None, 13, 77, 6)	0	convolution2d_2[0][0]
flatten_1 (Flatten)	(None, 6006)	0	maxpooling2d_2[0][0]
dense_1 (Dense)	(None, 120)	720840	flatten_1[0][0]
dense_2 (Dense)	(None, 84)	10164	dense_1[0][0]
dense_3 (Dense)	(None, 1)	85	dense_2[0][0]

In []: LetNet2 model

Layer (type)	Output Shape	Param #	Connected to
lambda_8 (Lambda)	(None, 160, 320, 3)	0	lambda_input_8[0][0]
cropping2d_7 (Cropping2D)	(None, 65, 320, 3)	0	lambda_8[0][0]
convolution2d_27 (Convolution2D)	(None, 61, 316, 6)	456	cropping2d_7[0][0]
maxpooling2d_7 (MaxPooling2D)	(None, 30, 158, 6)	0	convolution2d_27[0][0]
convolution2d_28 (Convolution2D)	(None, 26, 154, 16)	2416	maxpooling2d_7[0][0]
maxpooling2d_8 (MaxPooling2D)	(None, 13, 77, 16)	0	convolution2d_28[0][0]
flatten_8 (Flatten)	(None, 16016)	0	maxpooling2d_8[0][0]
dense_20 (Dense)	(None, 128)	2050176	flatten_8[0][0]
dropout_1 (Dropout)	(None, 128)	0	dense_20[0][0]
dense_21 (Dense)	(None, 84)	10836	dropout_1[0][0]
dropout_2 (Dropout)	(None, 84)	0	dense_21[0][0]
dense_22 (Dense)	(None, 1)	85	dropout_2[0][0]

In []: LeNet1 model

Layer (type)	Output Shape	Param #	Connected to
lambda_2 (Lambda)	(None, 160, 320, 3)	0	lambda_input_2[0][0]
convolution2d_3 (Convolution2D)	(None, 156, 316, 6)	456	lambda_2[0][0]
maxpooling2d_3 (MaxPooling2D)	(None, 78, 158, 6)	0	convolution2d_3[0][0]
convolution2d_4 (Convolution2D)	(None, 74, 154, 6)	906	maxpooling2d_3[0][0]
maxpooling2d_4 (MaxPooling2D)	(None, 37, 77, 6)	0	convolution2d_4[0][0]
flatten_2 (Flatten)	(None, 17094)	0	maxpooling2d_4[0][0]
dense_4 (Dense)	(None, 120)	2051400	flatten_2[0][0]
dense_5 (Dense)	(None, 84)	10164	dense_4[0][0]
dense_6 (Dense)	(None, 1)	85	dense_5[0][0]