DEEP LEARNING MODEL FOR SELF DRIVING CARS

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PROBLEM DEFINITION

▶ Take the video of the front scenario using the front camera on the car and process it through a system to determine the output. In the scope of this project, the output is the steering angle. The video is received as a sequence of images and the output is a sequence of steering angle corresponding to the images. The model is to be made in such a way so as to reduce the problem of overfitting and thus making a more generic model.

OBJECTIVES

To apply CNN along with 'dropout' to make a more efficient end-to-end learning model to predict the steering angle for a self driving car and overcoming the problem of overfitting.

INTRODUCTION

Self-driving cars, also called 'driverless' cars, are cars that perform the various functions of driving a car by itself. This includes steering, accelerating, braking, amongst others. The input is the images and distances retrieved using the different sensors on the car.

Convolution Neural Networks (CNN) became popular in 2012 and have ever since completely transformed the pattern recognition.

Before CNNs, for image processing, feature extraction and classification was needed to be hand-crafted. However, CNN learns features on its own from the training samples.

NVidia first came up with an algorithm to steer an automated vehicle, after years of research, in 2016.

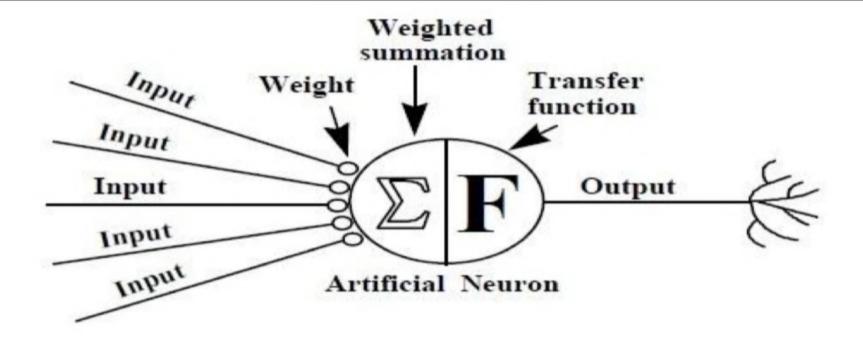
INTRODUCTION

CNN is a Deep Learning algorithm. Especially in Deep Learning, there is a huge tendency to 'overfit'. Overfitting is when the model learns too much to adjust to just the training data and thus is not generic enough. There are different techniques to overcome this problem, one being 'dropout'.

WHAT IS ARTIFICIAL NEURAL NETWORK (ANN)?

ANN is like copying the brain's network on the computers. It consists of neurons that have input and output, and are connected by weights. These weights control the importance the connection has. The network is made of 2 or more layers: first being the input layer and the last one being the output layer. All the layers in between the input and the output layers are called as the 'hidden' layers. Every layer has an 'activation function' which is responsible to generate the output from the neurons of that layer. Some of them are step function, sigmoid function, tan hyperbolic function, ReLU, etc.

A TYPICAL ARTIFICIAL NEURON:

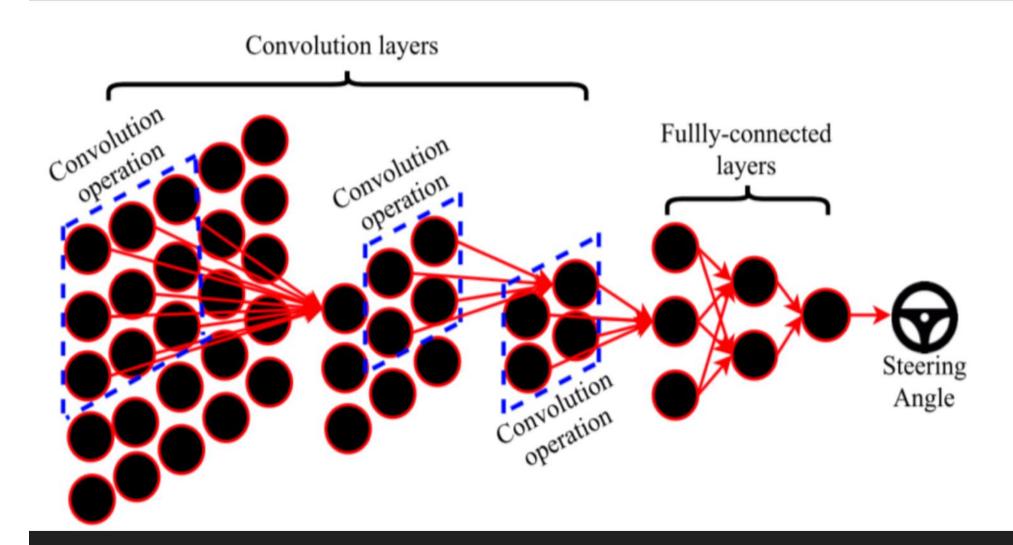


DEEP NEURAL NETWORKS (DNN)

Networks that have multiple hidden layers (generally 2 or more) are called Deep Neural Networks. A hidden layer is any layer in a neural network between the input and the output layer.

WHAT IS CNN?

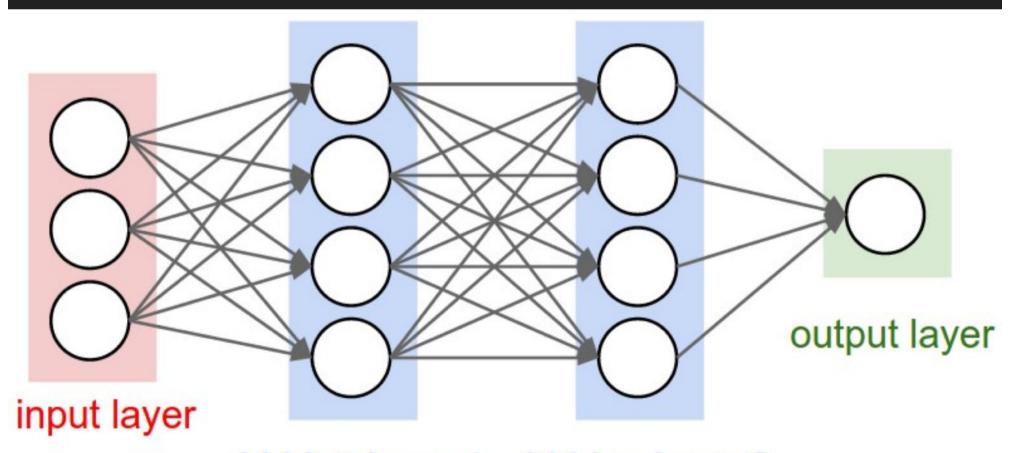
CNNs are fundamentally different from DNN (deep NN) in that they contain one or more convolution layers. In it, each input image will be passed through a series of convolution layers with filters or Kernels, generally followed by a pooling layer, then by fully connected layers (FC) and a loss layer. Convolution is the first layer to extract features from an input image. Convolution preserves the relationship between pixels by learning image features using small squares of input data. It is a mathematical operation that takes two inputs such as image matrix and a filter or kernel/ Feature Map. CNN has revolutionised image processing and pattern recognition.



FULLY CONNECTED LAYERS

Fully connected layers are layers in which each of the neurons of one layer are connected to all the neurons of the next layer. These are generally present after the convolution layers in CNN.

FULLY CONNECTED LAYERS



hidden layer 1 hidden layer 2

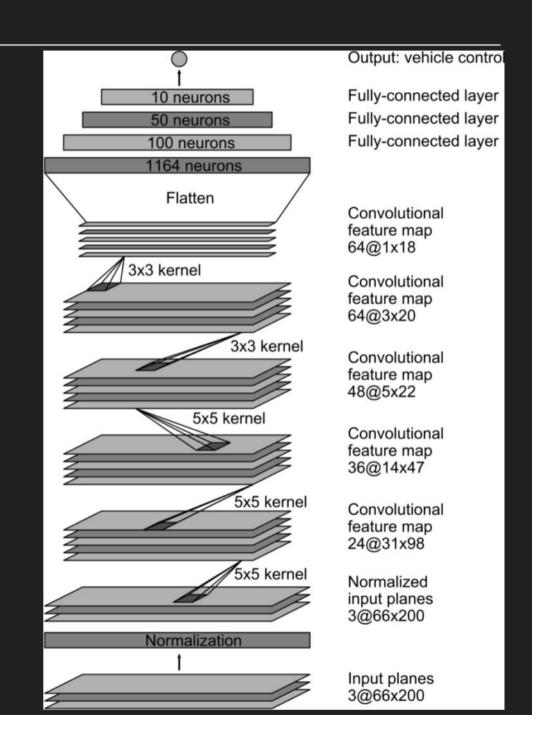
METHODOLOGY

- Select CNN architecture and training process parameters.
- Collect data to train and test the CNN (video from front camera and the corresponding steering angles at that time).
- Make the model and train and test it.
- Find area of improvement and make possible corrections.

NVIDIA ARCHITECTURE:

CNN architecture. It consists of a normalisation layer, 5 convolution

layers, 3 fully connected layers

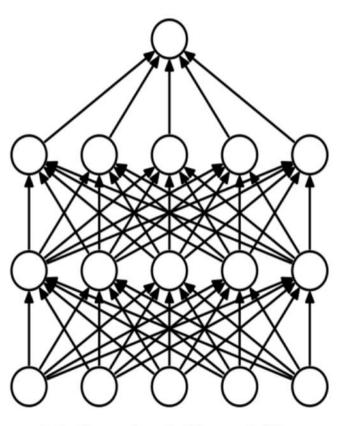


DROPOUT:

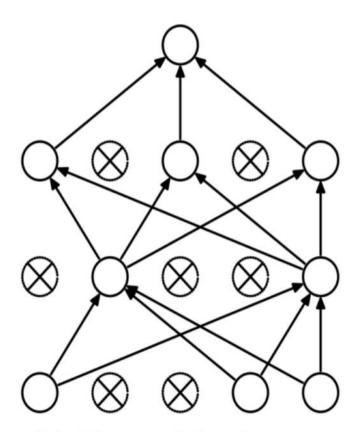
Dropout is a technique used to overcome the problem of overfitting, which is a very common issue, specially in deep neural networks. Dropout means to literally temporarily drop out units from the NN. When we drop a unit, we temporarily remove all its connections, both to and fro from the neuron. The selection of the unit is made at random, thus not partial to any unit.

We decide a 'keep probability' which can be thought of as the percentage of neurons to be kept in the network at any time. This probability is generally closer to 1 and it chosen to be anywhere from 0.5 to 1. So if the keep probability is 0.8, 80% of the neurons are kept in the network while training, or 20% are dropped out. Every time the NN is run, these neurons are chosen at random, thus every neuron has 'p' (keep probability) probability of staying in the network. This can be thought of as making a 'thinner' network for every run. Dropout is not applied while testing.

NN BEFORE AND AFTER DROPOUT



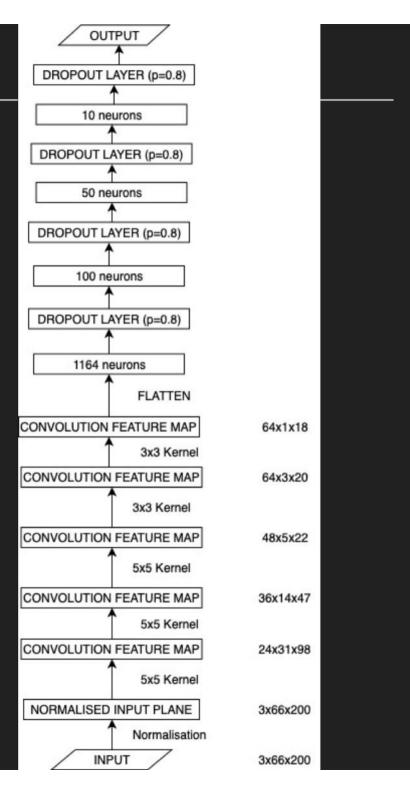
(a) Standard Neural Net



(b) After applying dropout.

PROPOSED ARCHITECTURE:

 Dropout layers with keep probability of 0.8 after every fully connected layer to reduce overfitting



TRAINING PARAMETERS

- Optimiser used Adam
- Activation units ReLU
- Loss calculated as Mean Square Error with L2 Norm Regularisation.
- Training performed using Mini Batch SGD

COLLECTING THE DATA

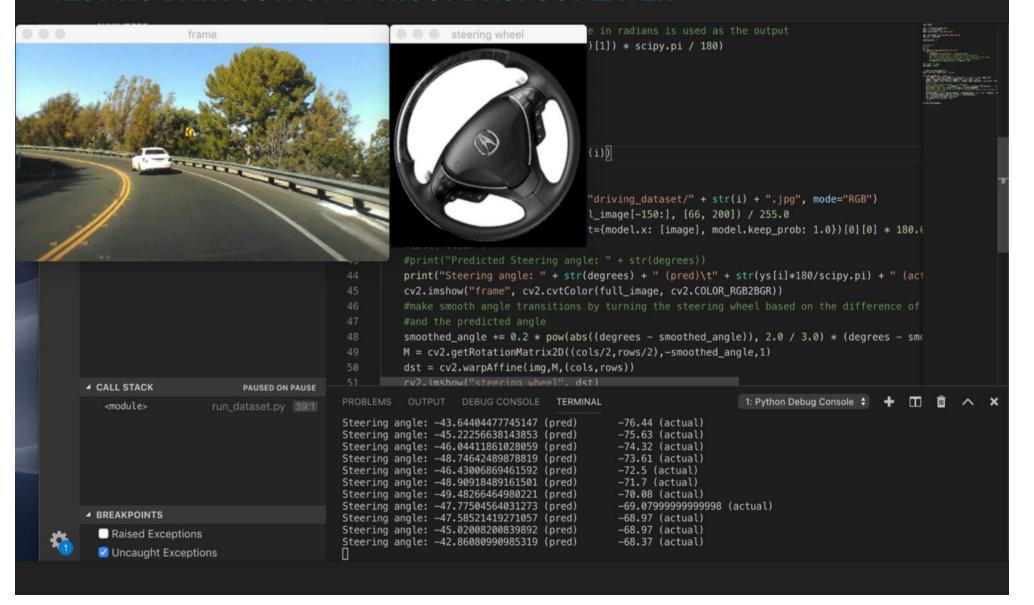
▶ The data collected is from SullyChen's GitHub and consists of 25min long drive across his city. There are a total of 45406 images in the dataset (at the rate of 30 frames per second). The data is divided into training and testing data in 80:20 ratio. So the first 80% of the data (20 mins) is used for training and validation while the last 20% (5 mins) data is used for testing the model.

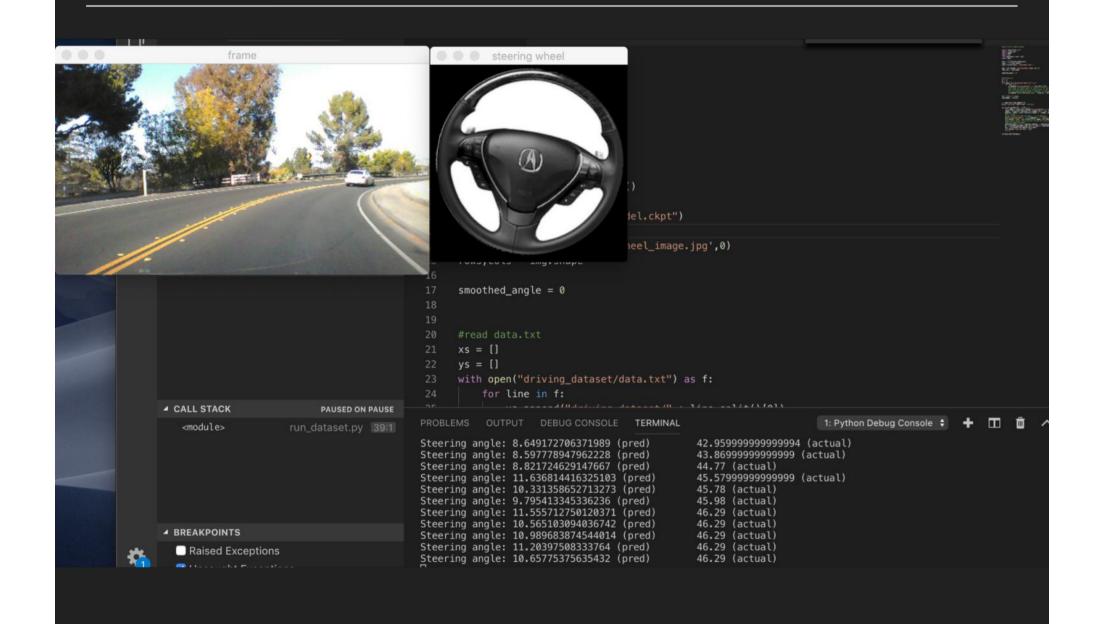
TRAINING ERRORS FOR MODEL WITHOUT DROPOUT

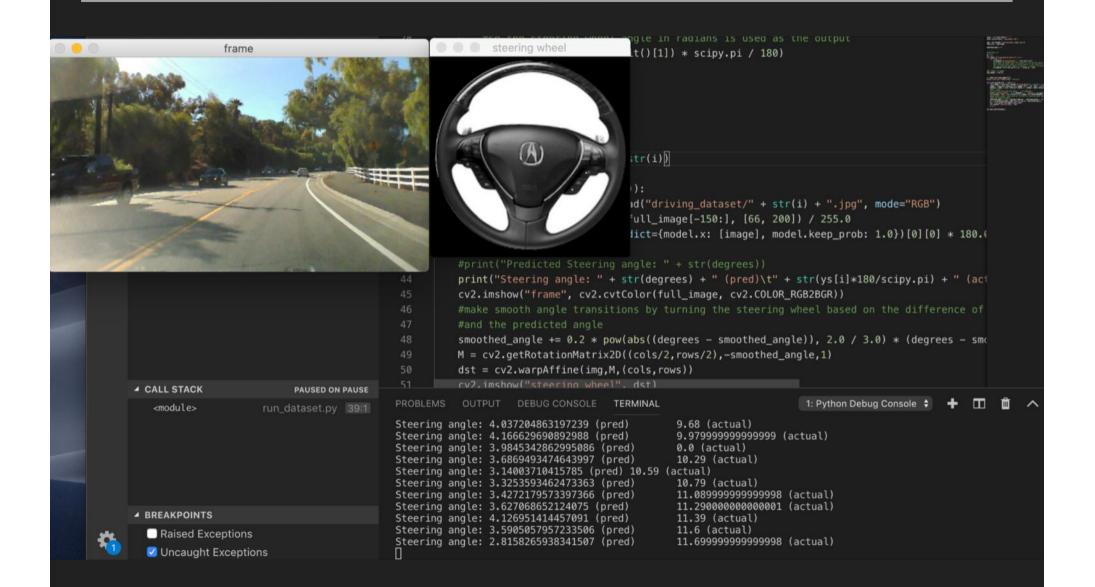
```
Epoch: 0, Step: 0, Loss: 6.46984
WARNING:tensorflow:TensorFlow's V1 checkpoint format has been deprecated.
WARNING: tensorflow: Consider switching to the more efficient V2 format:
WARNING:tensorflow: `tf.train.Saver(write_version=tf.train.SaverDef.V2)`
WARNING:tensorflow:now on by default.
WARNING: tensorflow: ************************
Epoch: 0, Step: 10, Loss: 6.43449
Epoch: 0, Step: 20, Loss: 6.24551
Epoch: 0, Step: 30, Loss: 6.25654
Epoch: 0, Step: 40, Loss: 6.23714
Epoch: 0, Step: 50, Loss: 6.23179
Epoch: 0, Step: 60, Loss: 6.10271
Epoch: 0, Step: 70, Loss: 5.99958
Epoch: 0, Step: 80, Loss: 6.05258
Epoch: 0, Step: 90, Loss: 6.02651
Epoch: 0, Step: 100, Loss: 5.83032
```

TRAINING ERRORS FOR MODEL WITH DROPOUT

TESTING DATA OUTPUT WITHOUT DROPOUT LAYER





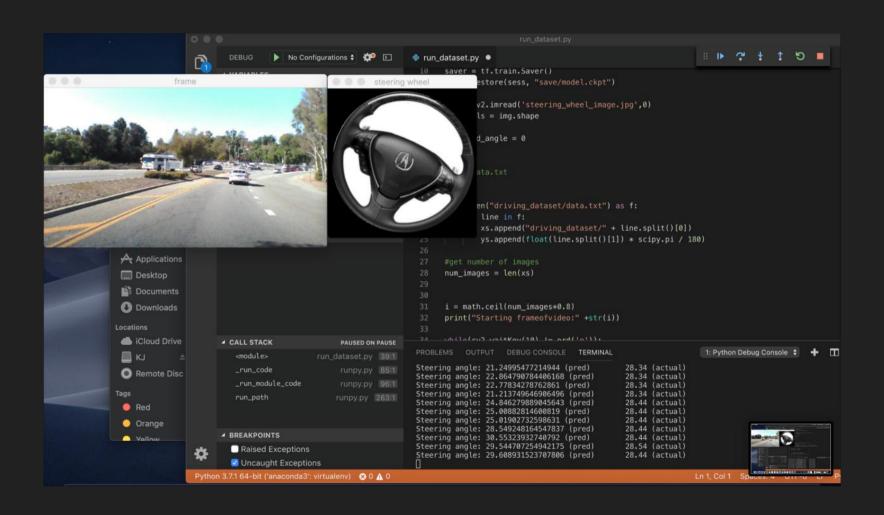


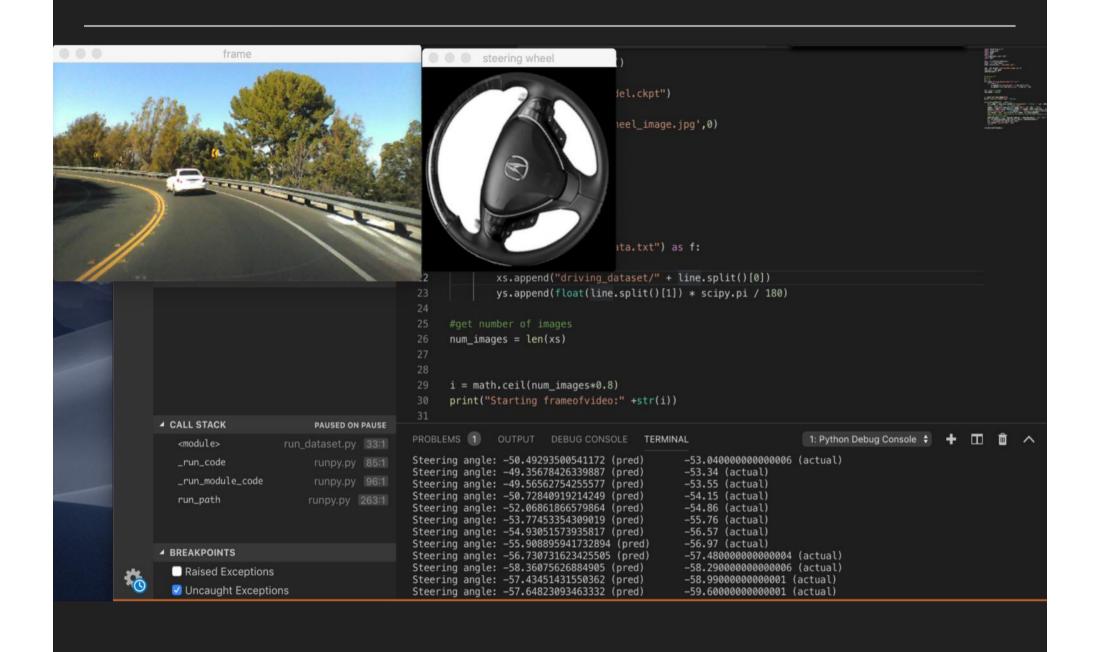
AVERAGE TEST LOSS WITHOUT DROPOUT LAYERS

The average test loss expressed as average deviation in degrees for this model is 13.812

```
2.22 (actual)
Steering angle: 18.62599416851647 (pred)
Steering angle: 18.62599416851647
                                  (pred)
                                                 2.22 (actual)
Steering angle: 18.62599416851647 (pred)
                                                 2.22 (actual)
Steering angle: 18.62599416851647 (pred)
                                                 2.22 (actual)
Steering angle: 18.62599416851647
                                  (pred)
                                                 2.22 (actual)
Steering angle: 18.62599416851647 (pred)
                                                2.22 (actual)
Steering angle: 18.62599416851647 (pred)
                                                2.22 (actual)
Steering angle: 18.62599416851647 (pred)
                                                 0.0 (actual)
Average loss = 13.811814953226108
Terminated: 15
(bace) Khyatic MacRook Pro:Solf Driving Care khyatiquntat
```

TESTING DATA OUTPUT WITH THE DROPOUT LAYERS





AVERAGE TEST LOSS WITH DROPOUT

The average test loss expressed as deviation in degrees for the model with dropout layers is 3.327, 10.5 degrees lesser than that without dropout model.

```
occurring ungeer orgotite potents, (preu)
Steering angle: 0.5372797555861281 (pred)
                                                2.22 (actual)
Steering angle: 0.4920055798684341 (pred)
                                                2.22 (actual)
Steering angle: 0.5671486530742698 (pred)
                                                2.22 (actual)
Steering angle: 0.739416962916834 (pred)
                                                2.22 (actual)
Steering angle: 0.5327562493648755 (pred)
                                                2.22 (actual)
Steering angle: 0.483554341348454 (pred)
                                                2.22 (actual)
Steering angle: 0.43468975396358117 (pred)
                                                2.22 (actual)
Steering angle: 0.6795118283058851 (pred)
                                                0.0 (actual)
Average loss = 3.32730183185685
Terminated: 15
(base) Khyatis-MacBook-Pro:Self Driving Cars khyatigupta$
```

THANK YOU!

REFERENCES

- 1.Manish Mishra, Monika Srivastava, "A View of Artificial Neural Network", IEEE International Conference on Advances in Engineering & Technology Research (ICAETR 2014), 2014.
- 2.Daniel Dworakowski, Mathew Monfort, Urs Muller, Jiakai Zhang, "End to End Learning for Self-Driving Cars", arXiv:1604.07316v1 [cs.CV], CoRR, vol. abs/1604.07316, April 2016.
- 3.Neena Aloysius and Geetha M, "A Review on Deep Convolutional Neural Networks", International Conference on Communication and Signal Processing, April 2017.
- 4.Nitish Srivastava, Geoffrey Hinton, Alex Krizhevsky, Ilya Sutskever, Ruslan Salakhutdinov, "Dropout: A Simple Way to Prevent Neural Networks from Overfitting", Journal of Machine Learning Research 15, 2014.
- 5.https://github.com/SullyChen/Autopilot-TensorFlow