How to Simulate a Self-Driving Car

Inspired by YouTuber Siraj Raval.

Initialize

- This workshop is for aspiring and active developers
- Please let's help each other out
- Ask who is a Python beginner/comfortable/pro? (show hands)
- Ask minority to disperse
- Feel free to ask questions at any time.
- Including, perhaps even especially, among yourselves.

Start with a demo



Three step process

- 1) Generate Data
- 2) Train model
- 3) Test

1) Data Generation

NVIDIA trained an autonomous car in real life.

Udacity built a self-driving car simulator to perform the same task.

1) Data Generation

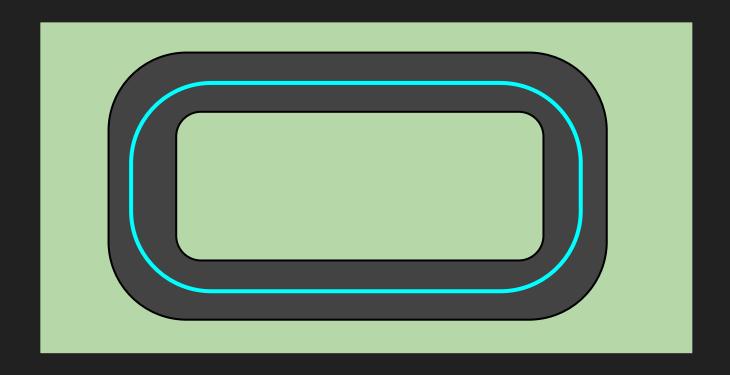
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Udacity built a self-driving car simulator to perform the same task.

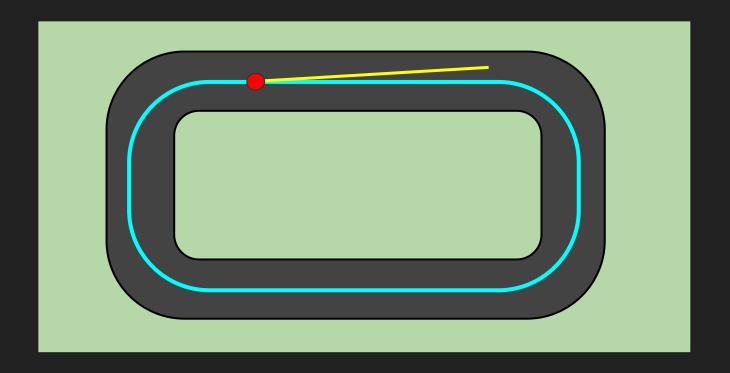
THEY DID THE EXACT SAME THING WE ARE GOING TO DO.

- Three cameras to the head of the car
- Human driver
- Collect data (i.e. steering angle, speed, throttle, brake and images)
- Train a neural net
- Drive car using a single camera

The three cameras



The three cameras



The three cameras



left center right

...steers you back on course after minor model errors

machine learning approach = behavioural cloning.

- NVIDIA = 72 hours all kinds of conditions
- We will do a few laps on a single circuit.

- steering angle
- throttle
- brake
- speed

'how to drive the car'

- steering angle
- throttle
- brake
- speed

'how to drive the car'

This requires a lot of training data so we are going to simplify...

.. and just train the steering angle.

label images kinda like this..



Left a bit



Straight on



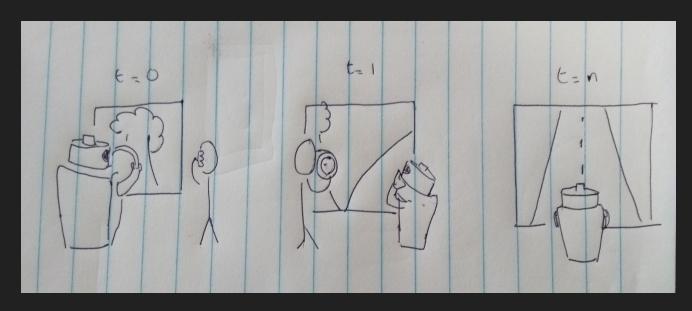
Right

This is an example of a supervised learning algorithm

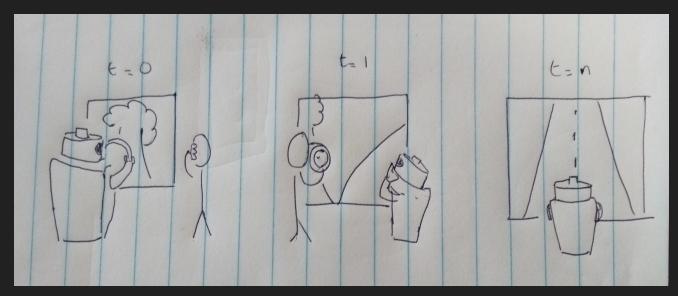
Reduce the difference between:

- what human did
- 2) what the computer thinks it should do

This is an example of a supervised learning algorithm

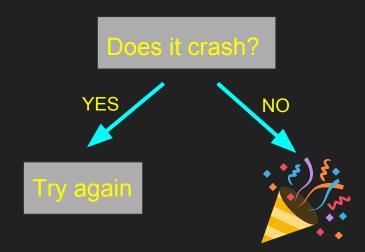


This is an example of a supervised learning algorithm



GIVEN A NEW IMAGE I HAVEN'T SEEN YET, HOW SHOULD I DRIVE?

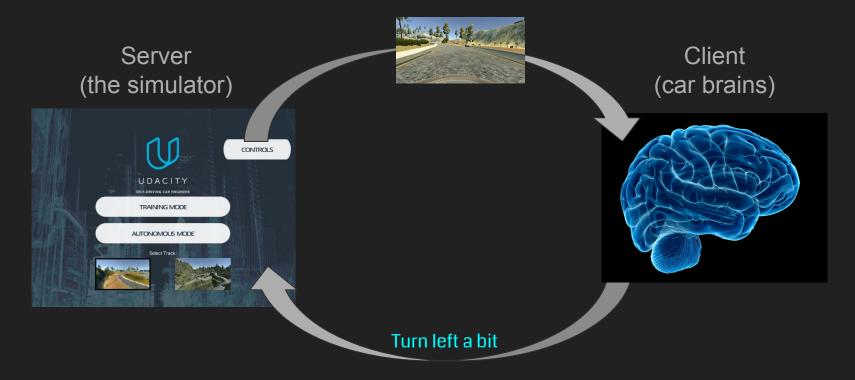
Drive the car using the neural net.



We will communicate using a server-client architecture



server-client architecture

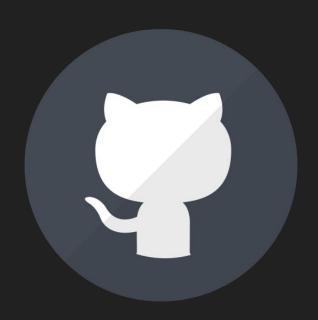


server-client architecture





Fork, clone and cd into source code repo



Download Self-Driving Car Simulator

Welcome to Udacity's Self-Driving Car Simulator

This simulator was built for Udacity's Self-Driving Car Nanodegree, to teach students how courses using deep learning. See more project details here.

All the assets in this repository require Unity. Please follow the instructions below for the

Avaliable Game Builds (Precompiled builds of the simulator)

Instructions: Download the zip file, extract it and run the executable file.

Version 2, 2/07/17

Linux Mac Windows

Version 1, 12/09/10

Linux Mac Windows 32 Windows 64

Install miniconda



Install dependencies

conda env create -f environments.yml
source activate car-behavioral-cloning

Install TensorFlow

conda install tensorflow

ALVIN (1992)



(Courtesy of Deat Formethes)

NVIDIA (2016)



Generate Data

- Run car simulator in TRAINING MODE
- Press 'r' key to select <u>fixed</u> output directory
- Press 'r' again to record driving skillz
- Complete between 3 to 5 laps.

Generate Data

Inspect

- Images (IMG/)
- driving log.csv

Train the model

python model.py --help

y = brains(X)

```
y = brains(X)

ALL MACHINE LEARNING MODELS EVER)

LABEL INPUT

HOW TO IMAGES (OUR MODEL)

DRIVE
```

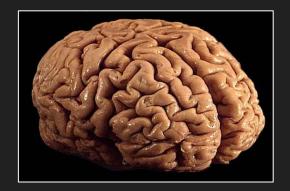
```
y = brains(X)

ALL MACHINE LEARNING MODELS EVER)

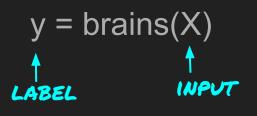
LABEL INPUT

HOW TO IMAGES (OUR MODEL)

DRIVE
```



BUT WHAT ARE BRAINS?

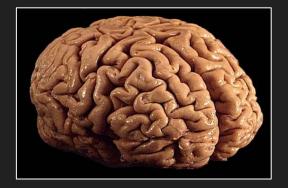


(ALL MACHINE LEARNING MODELS EVER)

HOW TO DRIVE

IMAGE5

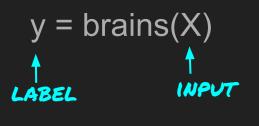
(OUR MODEL)



BUT WHAT ARE BRAINS?

a way of converting some input into something else that has meaning for humans

= cat



(ALL MACHINE LEARNING MODELS EVER)

HOW TO DRIVE

IMAGE5

(OUR MODEL)



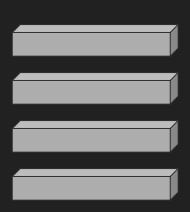
BUT WHAT ARE BRAINS?

a way of converting some input into something else that has meaning for humans

also = arrangement of mathematical functions

Convolutional Neural Network (CNN)

- Keras
 - o python machine learning library
- everything arranged in layers
 - o input is first layer
 - output is last layer
- Model specifics
 - 2D Convolutional layers
 - Dropout layer
 - Fully connected layers



- image input
- filter (a.k.a. kernel)
- dot product
- move along
- repeat
- outputs a new layer

1	<u>O</u>	2	1	2	<u>0</u>
<u>0</u>	2	1	2	2	1
1	1	<u>0</u>	2	<u>3</u>	4
4	<u>5</u>	2	1	2	2
1	<u>0</u>	<u>0</u>	<u>0</u>	1	2
<u>3</u>	2	1	<u>0</u>	<u>4</u>	<u>2</u>

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l	1	<u>0</u>	<u>2</u>	<u>1</u>	2	<u>0</u>
l	<u>0</u>	2	<u>1</u>	<u>2</u>	2	1
l	1	1	<u>0</u>	<u>2</u>	<u>3</u>	<u>4</u>
	<u>4</u>	<u>5</u>	<u>2</u>	1	2	2
	1	<u>0</u>	<u>0</u>	<u>0</u>	1	2
	<u>3</u>	2	1	<u>0</u>	<u>4</u>	2

0	1	0
1	1	1
0	1	0

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<u>0</u>	2	<u>1</u>	<u>2</u>	2	1
1	1	<u>0</u>	<u>2</u>	<u>3</u>	4
<u>4</u>	<u>5</u>	<u>2</u>	1	2	<u>2</u>
1	<u>0</u>	<u>0</u>	<u>0</u>	1	2
<u>3</u>	2	1	<u>0</u>	<u>4</u>	2

0	1	0	
1	1	1	= 4
0	1	0	

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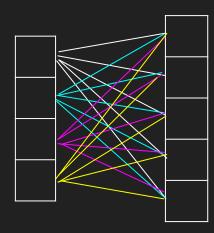
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1	<u>0</u>	<u>0</u>	<u>0</u>	1	2
<u>3</u>	2	1	<u>0</u>	<u>4</u>	2

4	7	8	10
9	6	8	13
12	8	7	9
8	3	2	9

- Neurons are only locally connected (i.e. in width and height)
- Less computationally expensive (images too large for full connections)
- Can find spatial patterns
- Much better at generalized learning (fully connected overfit)

What is a fully connected layer?

- More traditional neural network
- Every single neuron in a layer is connected to every other neuron







Sequential is a linear stack of layers from keras.models import Sequential

this creates a model instance
model = Sequential()

For an 8-bit image (i.e. from 0 to 255) this will normalize values between -1 to 1. model.add(Lambda(lambda x: x/127.5-1.0, input_shape=INPUT_SHAPE))



```
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```

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1

must define input shape in the first layer



model.add(Conv2D(24, 5, 5, activation='elu', subsample=(2, 2)))



Translation to English..

A 2D convolutional layer with 24 sets of random weights and a 5 by 5 pixel kernel. Neurons will fire based on an exponential linear unit activation function. The kernel will skip 2 pixels vertically and horizontally during the convolution.



```
model.add(Conv2D(24, 5, 5, activation='elu', subsample=(2, 2)))
model.add(Conv2D(36, 5, 5, activation='elu', subsample=(2, 2)))
model.add(Conv2D(48, 5, 5, activation='elu', subsample=(2, 2)))
model.add(Conv2D(64, 3, 3, activation='elu'))
model.add(Conv2D(64, 3, 3, activation='elu'))
```

A bunch of convolutional layers

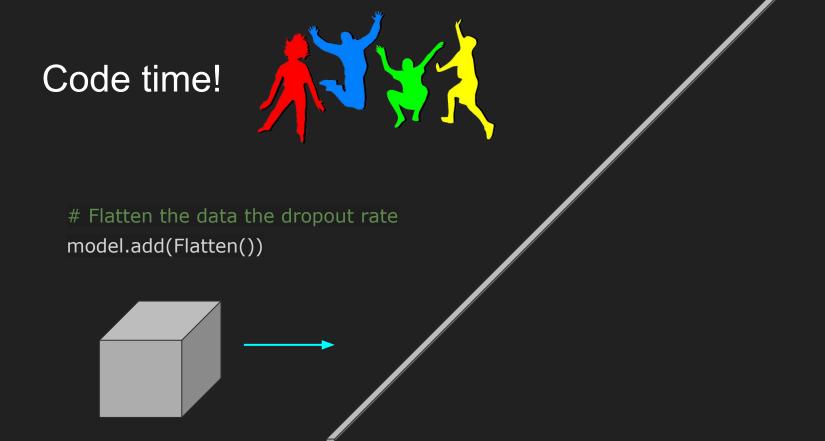


model.add(Dropout(args.keep_prob))

The dropout rate (e.g. 50%) will define how many nodes are randomly ignored

- Helps to prevent over-fitting
- Speeds up training

Essentially it forces the network to not really too heavily on a single set of neurons.



..because we are going to feed in a series of fully connected layers which expect a 1D vector



model.add(Dense(100, activation='elu'))
model.add(Dense(50, activation='elu'))
model.add(Dense(10, activation='elu'))
model.add(Dense(1))



Dense = fully connected layer

Series of fully connected layers condensing the output into a single value, i.e. the steering angle.

Training the model

- Minimize root mean squared error (RMSE)
 - o Between model output and human output

- Adam optimizer (gradient descent)
 - o Adaptive moment estimation.

Self Driving



Self Driving

```
# Start the 3D simulation
(double click using mouse)
# Ready for instructions
```

(run in autonomous mode)

Self Driving

```
# Move into source code repository

cd {path_to_your_local_repo}

# Start the car brain

python drive.py {model name.h5}
```

Well Done!



NASCAR Bonus Challenge

Who can drive the fastest?

max speed on line 41 of drive.py



