

Behavioral Cloning for Udacity Drive Simulator

This Project was prepared for Udacity Self Driving Car ND - Term one. 12 Feb 2019

Notes:

Project Uses Nvidia model published in NVIDIA End To End Driving Paper - Thanks and Acknowledgements

There are two environments. One for Drive simulator Model building Environment (this file) and second environment is for simulation interface bridge for realtime simulation (drive.py)

I have used two different workstations, a mac and a tensorflow gpu on windows 10 pro Xeon W workstation on GPU 1080, unfortunately this environment broke down with pip issues when installing pip (12.0 - released Feb 2019)along with eventlet and socket io which we use for drive.py. I switched between mac dual core mojave tensorflow cpu (conda based). I finished off in mac. Final movie is edited in apple imovie. Music track is american native pipe.

There are two ipynb jupyter notebooks. One shows all the experiments from the start and this one is the second notebook where i finalised a model for final testing and submission

Date:12 Feb 2019

Change request for release : 1

Release no:

Prepared for : Udacity ND Term 1 BC

Prepared by : ARV

Design doc ref:

Release doc ref:

Tech doc ref:

Testing doc ref:

User Read me ref :

Admin and Dev read me ref:

In [45]:

```
!conda list
```

```
# packages in environment at /miniconda3/envs/udrivesimul:
#
# Name          Version   Build Channel
absl-py        0.7.0     py36_1000  conda-f
orge
appnope        0.1.0     py36hf537a9a_0
asn1crypto     0.24.0    py36_1003  conda-f
orge
astor         0.7.1     py_0      conda-f
orge
backcall       0.1.0     py36_0
blas           1.0       mkl      anacond
a
bleach         3.1.0     pypi_0   pypi
bzip2          1.0.6     h1de35cc_1002
orge
c-ares         1.15.0    h1de35cc_1001
orge
ca-certificates 2019.1.23 0
cairo          1.14.12   h9d4d9ac_1005
orge
certifi        2018.11.29 py36_0
cffi           1.11.5    py36h342befbf_1001
orge
-forge
click          7.0       py_0      conda-f
orge
cloudpickle    0.7.0     pypi_0   pypi
cryptography   2.5       py36ha12b0ac_0
cycler         0.10.0    py_1      conda-f
orge
dask            1.1.1     pypi_0   pypi
dbus            1.13.6    h90a0687_0
decorator       4.3.2     py36_0
entrypoints     0.3       py36_0
eventlet        0.23.0    py36_1000
orge
expat           2.2.6     h0a44026_0
ffmpeg          4.1       heb45b42_1000
orge
flask           1.0.2     py_2      conda-f
orge
fontconfig      2.13.1    h1e4e890_1000
orge
freetype         2.9.1     hb4e5f40_0
anacond
a
gast             0.2.1.post0 py_0      conda-f
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gettext         0.19.8.1   hcca000d_1001
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giflib          5.1.4     h1de35cc_1001
orge
glib            2.56.2    h67dad55_1001
orge
gmp             6.1.2     h0a44026_1000
orge
gnutls          3.6.5     h53004b3_1001
orge
graphite2       1.3.13   h2098e52_1000
orge
greenlet        0.4.13   py36_0   conda-f
orge
```

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h5py	2.8.0	py36h878fce3_3	anacond
a			
harfbuzz	1.9.0	h9889186_1001	conda-f
orge			
hdf5	1.10.2	hfale0ec_1	anacond
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icu	58.2	h0a44026_1000	conda-f
orge			
idna	2.8	py36_1000	conda-f
orge			
imageio	2.5.0	pypi_0	pypi
imgaug	0.2.8	pypi_0	pypi
intel-openmp	2019.1	144	
ipykernel	5.1.0	py36h39e3cac_0	
ipython	7.2.0	py36h39e3cac_0	
ipython_genutils	0.2.0	py36h241746c_0	
ipywidgets	7.4.2	py36_0	
itsdangerous	1.1.0	py_0	conda-f
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jasper	1.900.1	h636a363_1006	conda-f
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jedi	0.13.2	py36_0	
jinja2	2.10	py_1	conda-f
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jpeg	9c	h1de35cc_1001	conda-f
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jsonschema	2.6.0	py36hb385e00_0	
jupyter	1.0.0	py36_7	
jupyter_client	5.2.4	py36_0	
jupyter_console	6.0.0	py36_0	
jupyter_core	4.4.0	py36_0	
keras	2.2.4	py36_0	conda-f
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keras-applications	1.0.4	py_1	conda-f
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keras-preprocessing	1.0.2	py_1	conda-f
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kiwisolver	1.0.1	py36h04f5b5a_1002	conda
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libcxx	7.0.0	h2d50403_2	conda-f
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libffi	3.2.1	h0a44026_1005	conda-f
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libgfortran	3.0.1	h93005f0_2	
libgpuarray	0.7.6	h1de35cc_1003	conda-f
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libiconv	1.15	h1de35cc_1004	conda-f
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libopenblas	0.3.3	hdc02c5d_3	
libpng	1.6.36	ha441bb4_0	anacond
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libprotobuf	3.6.1	hd9629dc_1000	conda-f
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libsodium	1.0.16	h3efe00b_0	
libtiff	4.0.10	hcb84e12_2	anacond
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libwebp	1.0.2	h801f6e5_1	conda-f
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libxml2	2.9.8	hf14e9c8_1005	conda-f
orge			

llvm-meta	7.0.0	0	conda-f
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mako	1.0.7	py_1	conda-f
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markdown	2.6.11	py_0	conda-f
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markupsafe	1.1.0	py36h1de35cc_1000	conda
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matplotlib	3.0.2	py36_1002	conda-f
orge			
matplotlib-base	3.0.2	py36hf043ca5_1002	conda
-forge			
mistune	0.8.4	py36h1de35cc_0	
mkl	2019.1	144	
mkl_fft	1.0.10	py36_0	conda-f
orge			
mkl_random	1.0.2	py36_0	conda-f
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nbconvert	5.3.1	py36_0	
nbformat	4.4.0	py36h827af21_0	
ncurses	6.1	h0a44026_1002	conda-f
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nettle	3.4.1	h1de35cc_1002	conda-f
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networkx	2.2	pypi_0	pypi
notebook	5.7.4	py36_0	
numpy	1.15.4	py36hac dab7b_0	anacond
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numpy-base	1.15.4	py36h6575580_0	anacond
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olefile	0.46	py36_0	anacond
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openblas	0.3.3	hdc02c5d_1001	conda-f
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opencv	3.4.1	py36h6fd60c2_1	anacond
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opencv-python	4.0.0.21	pypi_0	pypi
openh264	1.8.0	hd9629dc_1000	conda-f
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openssl	1.1.1a	h1de35cc_0	
pandas	0.24.1	py36h0a44026_0	anacond
a			
pandoc	2.2.3.2	0	
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parso	0.3.2	py36_0	
pcre	8.41	h0a44026_1003	conda-f
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pexpect	4.6.0	py36_0	
pickleshare	0.7.5	py36_0	
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prompt_toolkit	2.0.8	py_0	
protobuf	3.6.1	py36h0a44026_1001	conda
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ptyprocess	0.6.0	py36_0	
pyparser	2.19	py_0	conda-f
orge			

pygments	2.3.1	py36_0	
pygpu	0.7.6	py36h917ab60_1000	conda
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pyopenssl	19.0.0	py36_0	conda-f
orge			
pyparsing	2.3.1	py_0	conda-f
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python	3.6.8	haf84260_0	
python-dateutil	2.7.5	py36_0	
python-engineio	3.0.0	py_0	conda-f
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python-socketio	3.1.2	py_0	conda-f
orge			
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pywavelets	1.0.1	pypi_0	pypi
pyyaml	3.13	py36h1de35cc_1001	conda
-forge			
pyzmq	17.1.2	py36h0a44026_2	
qt	5.9.7	h468cd18_1	
qtconsole	4.4.3	py36_0	
readline	7.0	hcfe32e1_1001	conda-f
orge			
scikit-image	0.14.2	pypi_0	pypi
scikit-learn	0.20.2	py36h27c97d8_0	anacond
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scipy	1.2.0	py36h1410ff5_0	anacond
a			
send2trash	1.5.0	py36_0	
setuptools	40.8.0	py36_0	conda-f
orge			
shapely	1.6.4.post2	pypi_0	pypi
sip	4.19.8	py36h0a44026_0	
six	1.12.0	py36_1000	conda-f
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sqlite	3.26.0	h1765d9f_1000	conda-f
orge			
tensorboard	1.10.0	py36_0	conda-f
orge			
tensorflow	1.10.0	py36_0	conda-f
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termcolor	1.1.0	py_2	conda-f
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terminado	0.8.1	py36_1	
testpath	0.4.2	py36_0	
theano	1.0.3	py36_0	conda-f
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tk	8.6.9	ha441bb4_1000	conda-f
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werkzeug	0.14.1	py_0	conda-f
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wheel	0.32.3	py36_0	conda-f
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yaml	0.1.7	h1de35cc_1001	conda-f
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zeromq	4.3.1	h0a44026_3	
zlib	1.2.11	h1de35cc_1004	conda-f
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zstd	1.3.7	h5bba6e5_0	anacond
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Let us do the imports. All the imports done were collected and collated here

In [17]:

```
import os
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import keras
from keras.models import Sequential
from keras.optimizers import Adam
from keras.layers import Convolution2D, MaxPooling2D, Dropout, Flatten, Dense
from sklearn.utils import shuffle
from sklearn.model_selection import train_test_split
from imgaug import augmenters as iaa
import cv2
import pandas as pd
import ntpath
import random
```

Handle the CSV file containing ref to snapshot of manual driving files simulating real cameras on a car

In [18]:

```
datadir = 'new_track'  
columns =['center','left','right','steering','throttle','reverse','speed']  
data=pd.read_csv('driving_log.csv',names=columns)  
data.head
```

Out[18]:

```
<bound method NDFrame.head of
    center \
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13     C:\Users\Vijy\Desktop\new_track\IMG\right_2018... -0.250000
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1.000000
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24     C:\Users\Vijy\Desktop\new_track\IMG\right_2018... -0.417654
1.000000
25     C:\Users\Vijy\Desktop\new_track\IMG\right_2018... -0.148133
1.000000

```

26 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
1.000000
27 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
1.000000
28 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... -0.150000
1.000000
29 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... -0.203962
1.000000
... ...
...
4023 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.150000
1.000000
4024 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.300000
1.000000
4025 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.097232
0.811835
4026 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.566915
4027 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.380225
4028 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.132538
4029 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4030 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4031 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4032 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4033 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4034 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4035 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4036 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4037 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4038 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4039 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4040 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4041 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4042 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4043 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4044 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4045 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4046 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4047 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4048 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000

0.000000
4049 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4050 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4051 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000
4052 C:\Users\Vijy\Desktop\new_track\IMG\right_2018... 0.000000
0.000000

	reverse	speed
0	0.0	0.649786
1	0.0	0.627942
2	0.0	0.622910
3	0.0	0.619162
4	0.0	0.615438
5	0.0	0.610506
6	0.0	0.606834
7	0.0	0.601971
8	0.0	0.598350
9	0.0	0.620654
10	0.0	0.707000
11	0.0	0.946799
12	0.0	1.434013
13	0.0	2.173052
14	0.0	2.864847
15	0.0	3.791584
16	0.0	4.489107
17	0.0	5.422441
18	0.0	6.111838
19	0.0	7.026899
20	0.0	7.709986
21	0.0	8.614678
22	0.0	9.288085
23	0.0	10.162890
24	0.0	10.792000
25	0.0	11.502830
26	0.0	12.418390
27	0.0	13.250700
28	0.0	14.023000
29	0.0	14.644770
...
4023	0.0	30.184000
4024	0.0	30.169180
4025	0.0	30.153680
4026	0.0	30.089560
4027	0.0	30.040620
4028	0.0	29.987340
4029	0.0	29.788930
4030	0.0	29.608500
4031	0.0	29.369460
4032	0.0	29.132300
4033	0.0	28.955670
4034	0.0	28.721800
4035	0.0	28.490560
4036	0.0	28.318310
4037	0.0	28.090210
4038	0.0	27.920720
4039	0.0	27.752260
4040	0.0	27.529250
4041	0.0	27.363170

```

4042      0.0  27.143300
4043      0.0  26.979560
4044      0.0  26.762780
4045      0.0  26.601340
4046      0.0  26.387600
4047      0.0  26.228420
4048      0.0  26.017680
4049      0.0  25.808640
4050      0.0  25.652960
4051      0.0  25.446850
4052      0.0  17.007810

```

[4053 rows x 7 columns]>

Handle path data and verifying first few lines with 'head'

In [19]:

```

def path_leaf(path):
    head,tail = ntpath.split(path)
    return tail
data['center'] = data['center'].apply(path_leaf)
data['left'] = data['left'].apply(path_leaf)
data['right'] = data['right'].apply(path_leaf)
data.head()

```

Out[19]:

	center	left	
0	center_2018_07_16_17_11_43_382.jpg	left_2018_07_16_17_11_43_382.jpg	right_2018
1	center_2018_07_16_17_11_43_670.jpg	left_2018_07_16_17_11_43_670.jpg	right_2018
2	center_2018_07_16_17_11_43_724.jpg	left_2018_07_16_17_11_43_724.jpg	right_2018
3	center_2018_07_16_17_11_43_792.jpg	left_2018_07_16_17_11_43_792.jpg	right_2018
4	center_2018_07_16_17_11_43_860.jpg	left_2018_07_16_17_11_43_860.jpg	right_2018

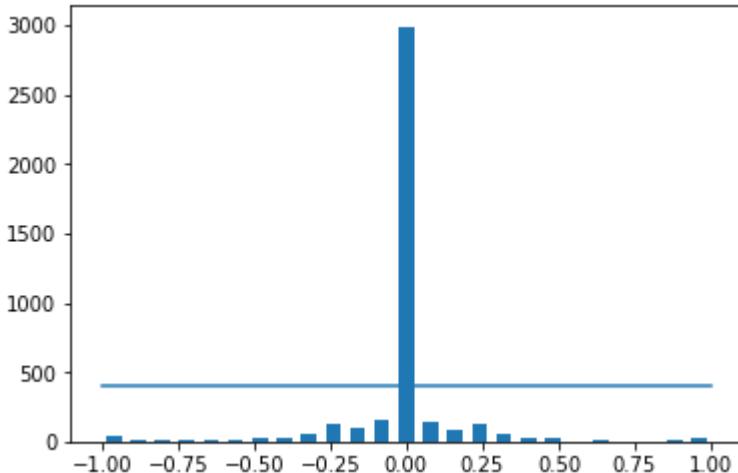
Plotting steering angle as bins between values of -1 and +1, against number of occurrences in the training

In [20]:

```
num_bins = 25
samples_per_bin = 400
hist, bins = np.histogram(data['steering'], num_bins)
center = (bins[:-1] + bins[1:]) * 0.5
plt.bar(center, hist, width=0.05)
plt.plot((np.min(data['steering']), np.max(data['steering'])), (samples_per_bin,
 samples_per_bin))
```

Out[20]:

[<matplotlib.lines.Line2D at 0x1c2f1f7390>]



Data study for feature engineering and data wrangling. Removing too much bias for 0 deg straight angle and attempting at generalising beyond this predominantly straight track road

In [21]:

```
print('total data:', len(data))
remove_list = []
for j in range(num_bins):
    list_ = []
    for i in range(len(data['steering'])):
        if data['steering'][i] >= bins[j] and data['steering'][i] <= bins[j+1]:
            list_.append(i)
    list_ = shuffle(list_)
    list_ = list_[samples_per_bin:]
    remove_list.extend(list_)

print('removed:', len(remove_list))
data.drop(data.index[remove_list], inplace=True)
print('remaining:', len(data))
```

```
total data: 4053
removed: 2590
remaining: 1463
```

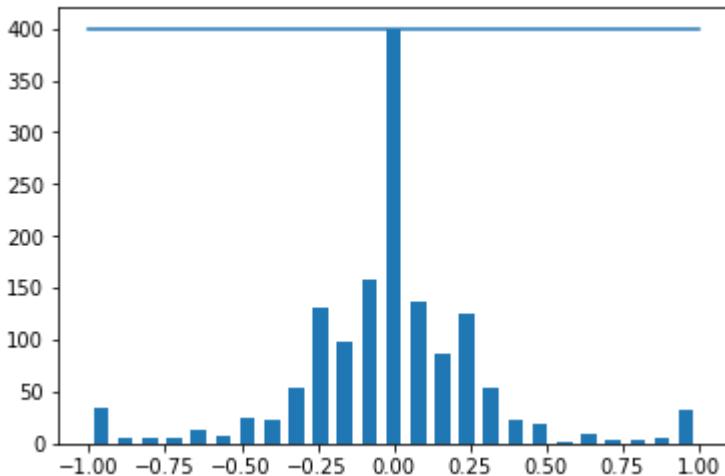
Histogramming steering after removal

In [22]:

```
hist, _ = np.histogram(data['steering'], (num_bins))
plt.bar(center, hist, width=0.05)
plt.plot((np.min(data['steering']), np.max(data['steering'])), (samples_per_bin,
 samples_per_bin))
```

Out[22]:

```
[<matplotlib.lines.Line2D at 0x1c2e5140b8>]
```



Printing a sample and verifying data pattern

In [23]:

```
print(data.iloc[1])
```

center	center_2018_07_16_17_11_44_485.jpg
left	left_2018_07_16_17_11_44_485.jpg
right	right_2018_07_16_17_11_44_485.jpg
steering	-0.25
throttle	0.863326
reverse	0
speed	2.17305

Name: 13, dtype: object

Function for picking up images and returning image paths and steering angle value from csv file

In [24]:

```
def load_img_steering(datadir, df):
    image_path = []
    steering = []
    for i in range(len(data)):
        indexed_data = data.iloc[i]
        center, left, right = indexed_data[0], indexed_data[1], indexed_data[2]
        image_path.append(os.path.join(datadir, center.strip()))
        steering.append(float(indexed_data[3]))
    # left image append
    image_path.append(os.path.join(datadir, left.strip()))
    steering.append(float(indexed_data[3])+0.15)
    # right image append
    image_path.append(os.path.join(datadir, right.strip()))
    steering.append(float(indexed_data[3])-0.15)
    image_paths = np.asarray(image_path)
    steerings = np.asarray(steering)
    return image_paths, steerings
```

Adding real file system point for dataset. Splitting training and validation set from collected data

In [25]:

```
image_paths, steerings = load_img_steering(datadir + '/IMG', data)
X_train, X_valid, y_train, y_valid = train_test_split(image_paths, steerings, test_size=0.2, random_state=6)
```

Printing out our training and validation split set, plotting histogram to visualise steering angles

In [26]:

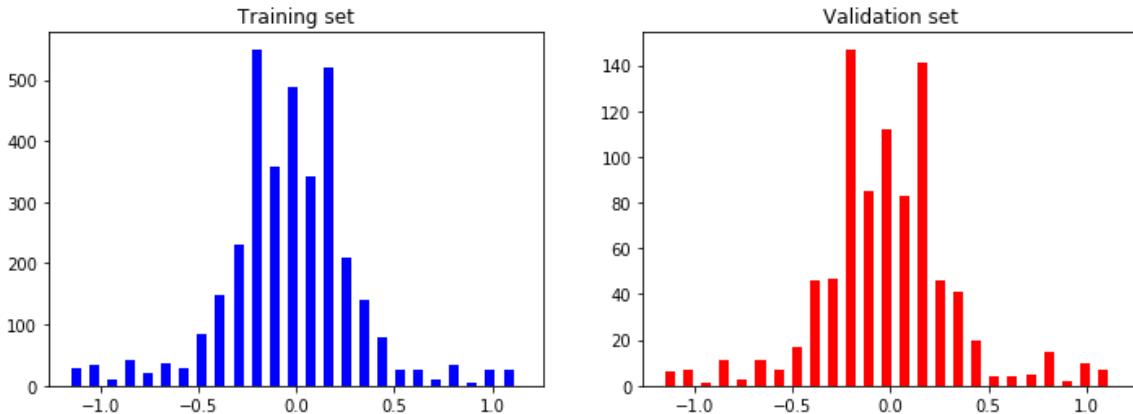
```
print('Training Samples: {} \nValid Samples: {}'.format(len(X_train), len(X_valid)))
fig, axes = plt.subplots(1, 2, figsize=(12, 4))
axes[0].hist(y_train, bins=num_bins, width=0.05, color='blue')
axes[0].set_title('Training set')
axes[1].hist(y_valid, bins=num_bins, width=0.05, color='red')
axes[1].set_title('Validation set')
```

Training Samples: 3511

Valid Samples: 878

Out[26]:

Text(0.5, 1.0, 'Validation set')



Augmentation Techniques

We attempt various augmentation methods to try and generalise trained model from the collected training data.

Zoom affine augmentation

In [27]:

```
def zoom(image):
    zoom = iaa.Affine(scale=(1, 1.3))
    image = zoom.augment_image(image)
    return image
```

In [28]:

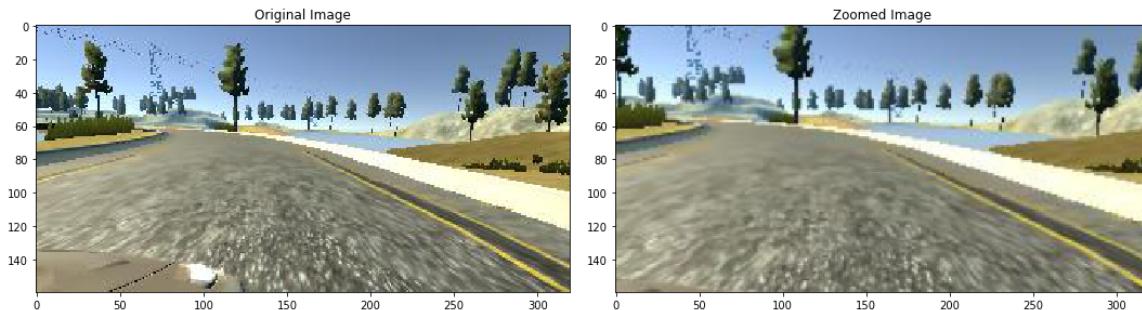
```
print(image)  
IMG/IMG/right_2018_07_16_17_11_47_988.jpg
```

In [29]:

```
image = image_paths[random.randint(0, 1000)]  
original_image = mpimg.imread(image)  
zoomed_image = zoom(original_image)  
  
fig, axs = plt.subplots(1, 2, figsize=(15, 10))  
fig.tight_layout()  
  
axs[0].imshow(original_image)  
axs[0].set_title('Original Image')  
  
axs[1].imshow(zoomed_image)  
axs[1].set_title('Zoomed Image')
```

Out[29]:

```
Text(0.5, 1.0, 'Zoomed Image')
```



Pan augmentation

In [30]:

```
def pan(image):
    pan = iaa.Affine(translate_percent= {"x" : (-0.1, 0.1), "y": (-0.1, 0.1)})
    image = pan.augment_image(image)
    return image
image = image_paths[random.randint(0, 1000)]
original_image = mpimg.imread(image)
panned_image = pan(original_image)

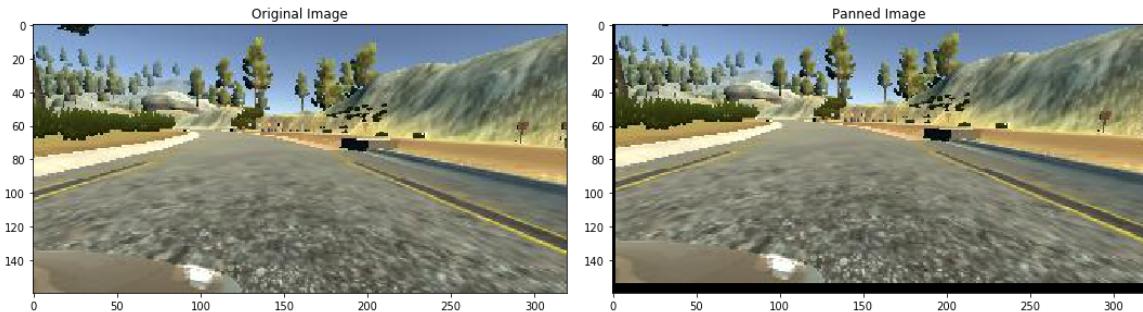
fig, axs = plt.subplots(1, 2, figsize=(15, 10))
fig.tight_layout()

axs[0].imshow(original_image)
axs[0].set_title('Original Image')

axs[1].imshow(panned_image)
axs[1].set_title('Panned Image')
```

Out[30]:

Text(0.5, 1.0, 'Panned Image')



Altered Brightness augmentation

In [31]:

```
def img_random_brightness(image):
    brightness = iaa.Multiply((0.2, 1.2))
    image = brightness.augment_image(image)
    return image
image = image_paths[random.randint(0, 1000)]
original_image = mpimg.imread(image)
brightness_altered_image = img_random_brightness(original_image)

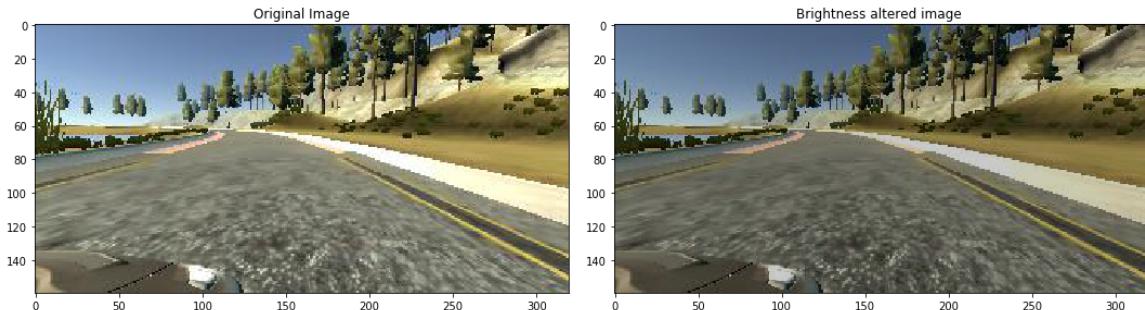
fig, axs = plt.subplots(1, 2, figsize=(15, 10))
fig.tight_layout()

axs[0].imshow(original_image)
axs[0].set_title('Original Image')

axs[1].imshow(brightness_altered_image)
axs[1].set_title('Brightness altered image ')
```

Out[31]:

Text(0.5, 1.0, 'Brightness altered image ')



Random Flip Augmentation

In [32]:

```

def img_random_flip(image, steering_angle):
    image = cv2.flip(image,1)
    steering_angle = -steering_angle
    return image, steering_angle
random_index = random.randint(0, 1000)
image = image_paths[random_index]
steering_angle = steerings[random_index]

original_image = mpimg.imread(image)
flipped_image, flipped_steering_angle = img_random_flip(original_image, steering_angle)

fig, axs = plt.subplots(1, 2, figsize=(15, 10))
fig.tight_layout()

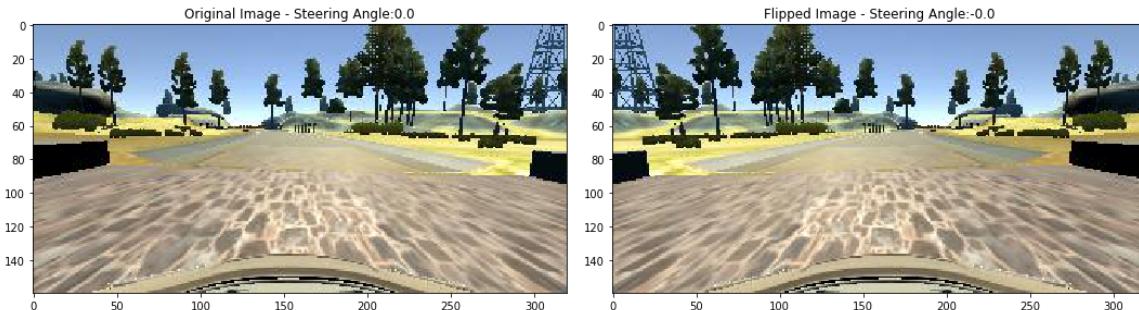
axs[0].imshow(original_image)
axs[0].set_title('Original Image - ' + 'Steering Angle:' + str(steering_angle))

axs[1].imshow(flipped_image)
axs[1].set_title('Flipped Image - ' + 'Steering Angle:' + str(flipped_steering_angle))

```

Out[32]:

Text(0.5, 1.0, 'Flipped Image - Steering Angle:-0.0')



We do a 50% probability for selective augmentation on various images

In [33]:

```

def random_augment(image, steering_angle):
    image = mpimg.imread(image)
    if np.random.rand() < 0.5:
        image = pan(image)
    if np.random.rand() < 0.5:
        image = zoom(image)
    if np.random.rand() < 0.5:
        image = img_random_brightness(image)
    if np.random.rand() < 0.5:
        image, steering_angle = img_random_flip(image, steering_angle)

    return image, steering_angle

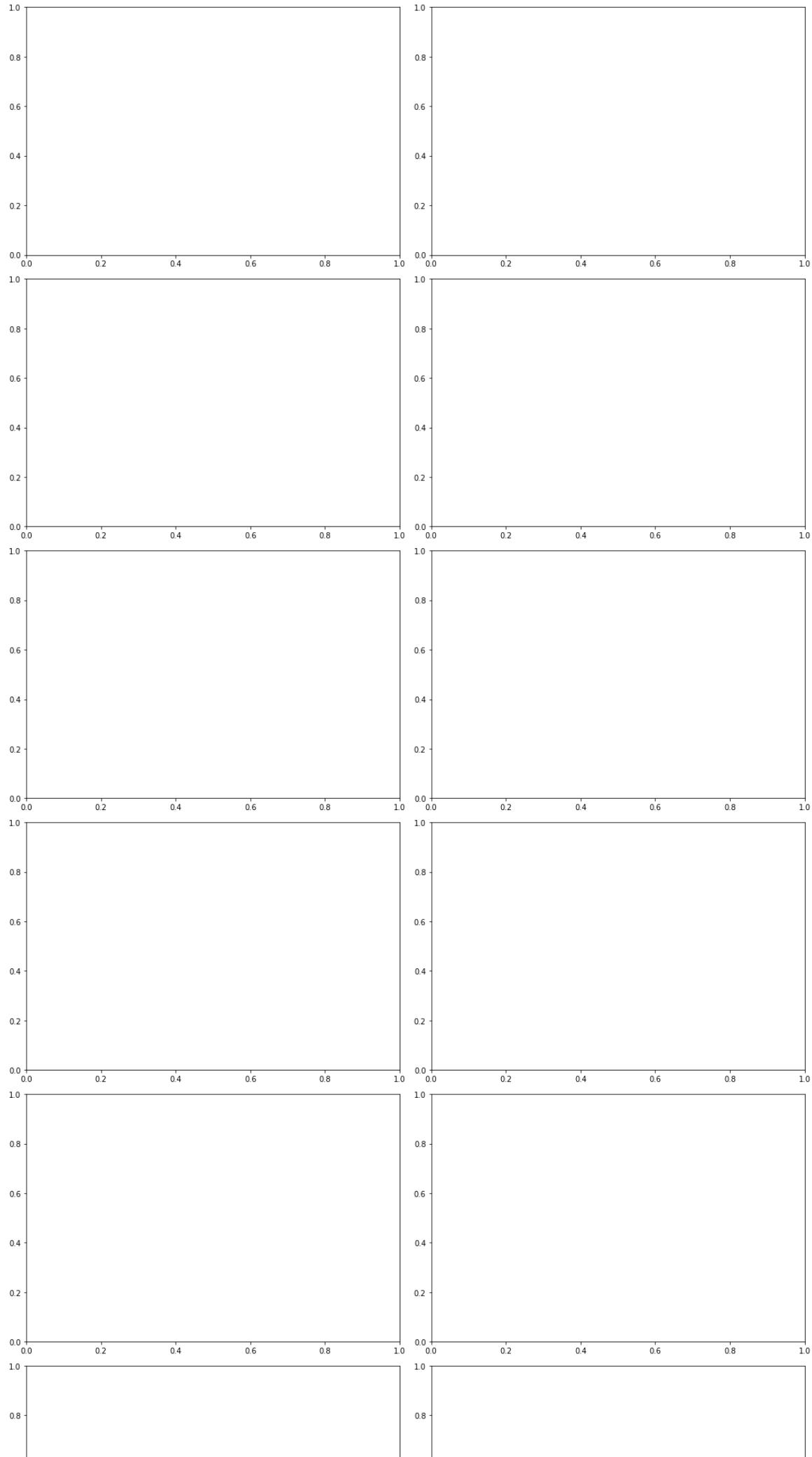
```

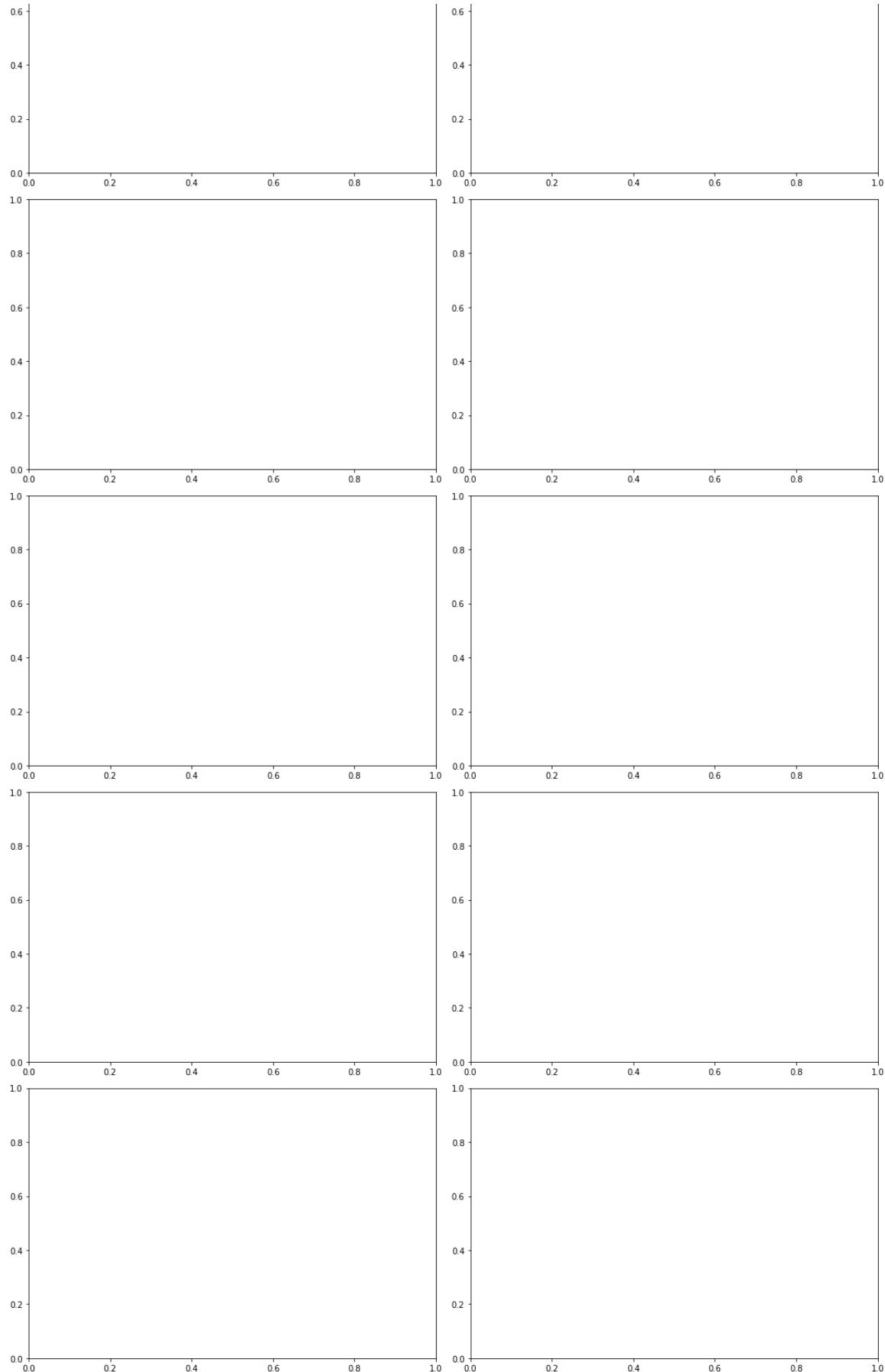
Preparing for plot areas to visualise augmentations

In [34]:

```
ncol = 2
nrow = 10

fig, axs = plt.subplots(nrow, ncol, figsize=(15, 50))
fig.tight_layout()
```





Augmentations visualised

In [35]:

```
ncol = 2
nrow = 10

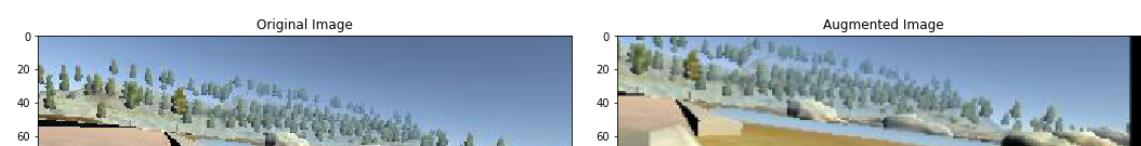
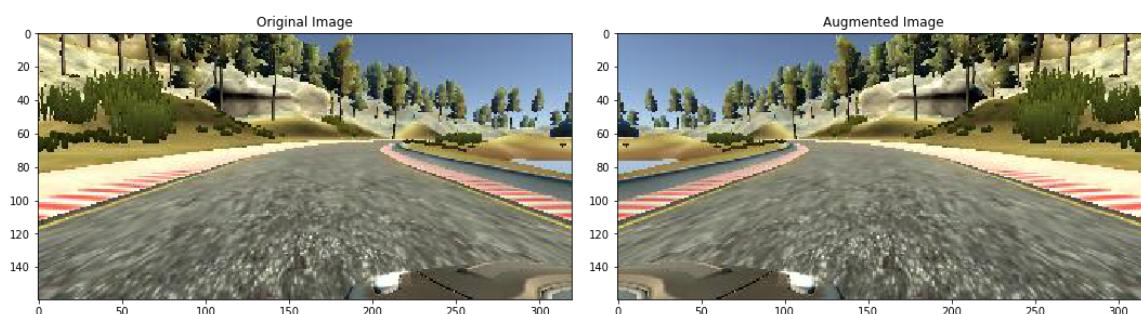
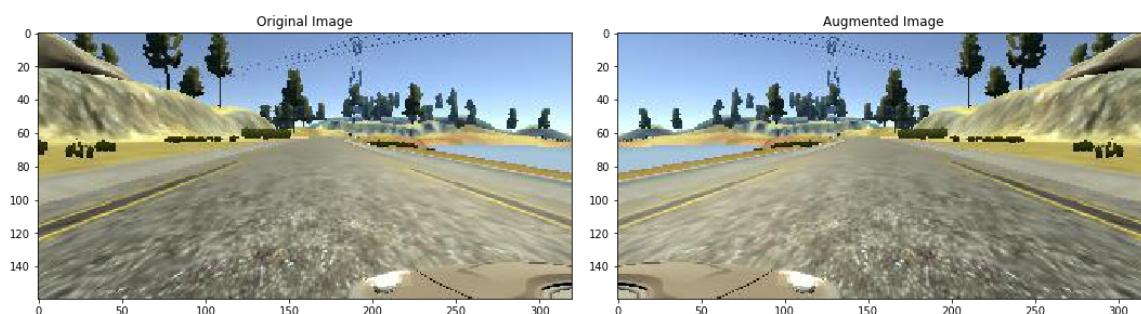
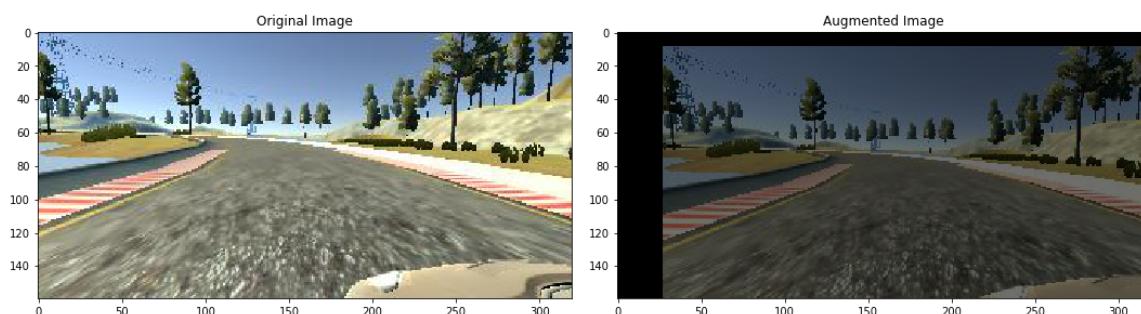
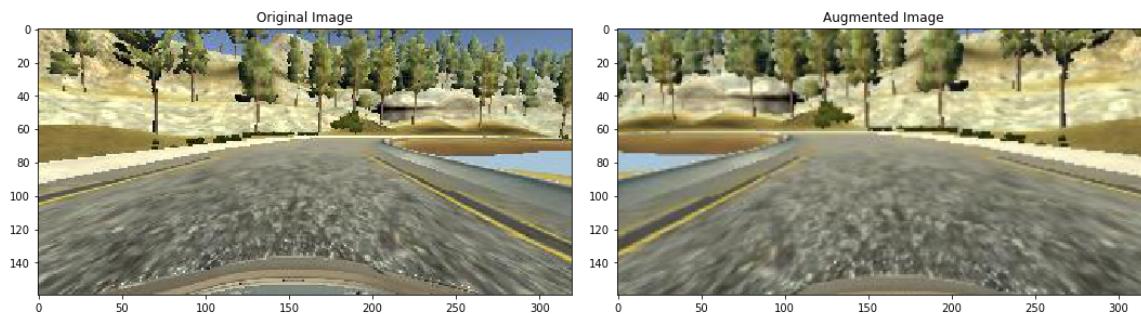
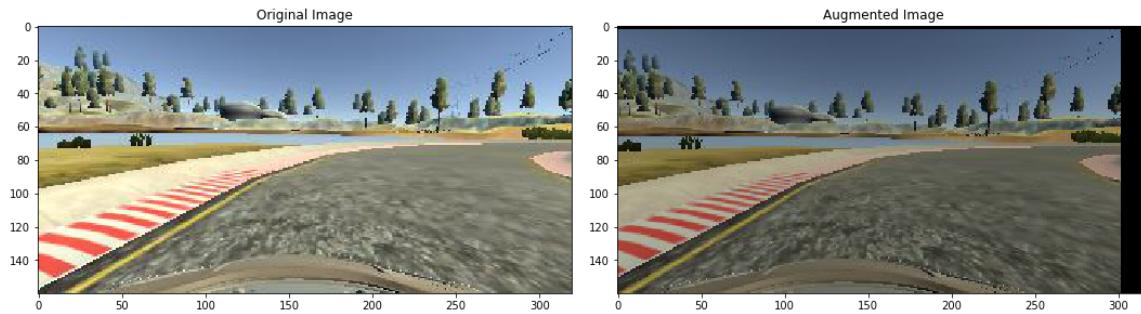
fig, axs = plt.subplots(nrow, ncol, figsize=(15, 50))
fig.tight_layout()

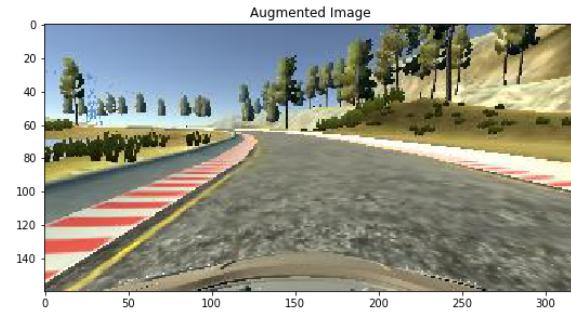
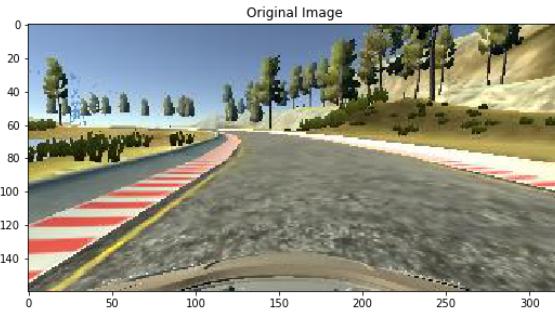
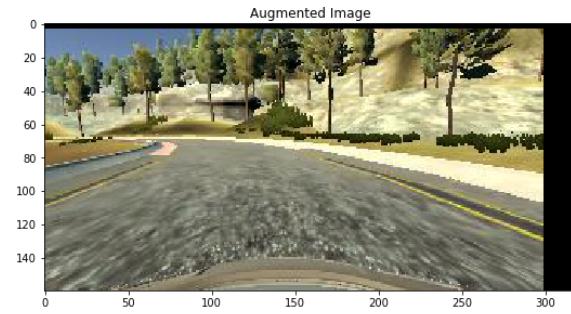
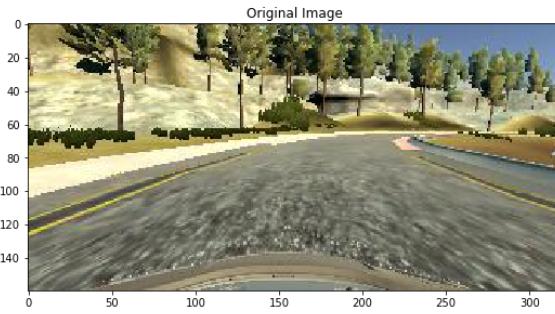
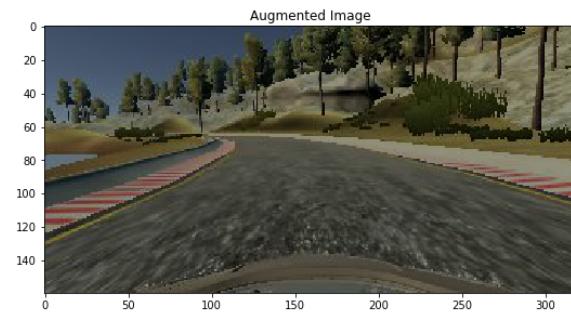
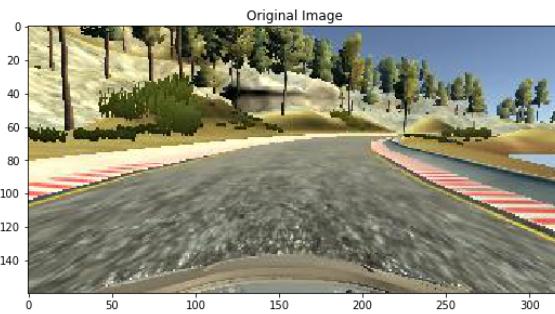
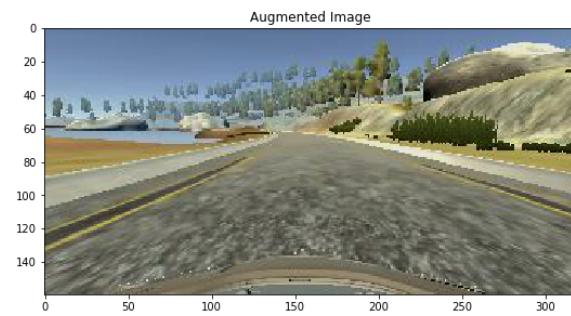
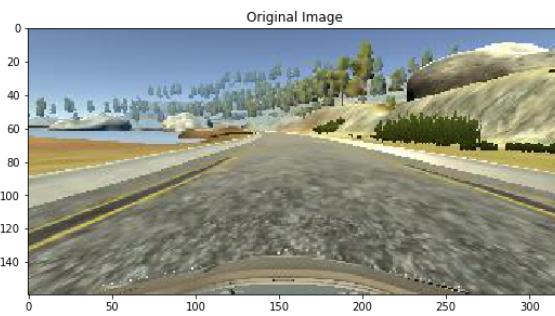
for i in range(10):
    randnum = random.randint(0, len(image_paths) - 1)
    random_image = image_paths[randnum]
    random_steering = steerings[randnum]

    original_image = mpimg.imread(random_image)
    augmented_image, steering = random_augment(random_image, random_steering)

    axs[i][0].imshow(original_image)
    axs[i][0].set_title("Original Image")

    axs[i][1].imshow(augmented_image)
    axs[i][1].set_title("Augmented Image")
```





Usual regular Preprocessing of RGB to YUV, Gauss Blurr, and resizing

In [36]:

```
def img_preprocess(img):
    img = img[60:135,:,:]
    img = cv2.cvtColor(img, cv2.COLOR_RGB2YUV)
    img = cv2.GaussianBlur(img, (3, 3), 0)
    img = cv2.resize(img, (200, 66))
    img = img/255
    return img
```

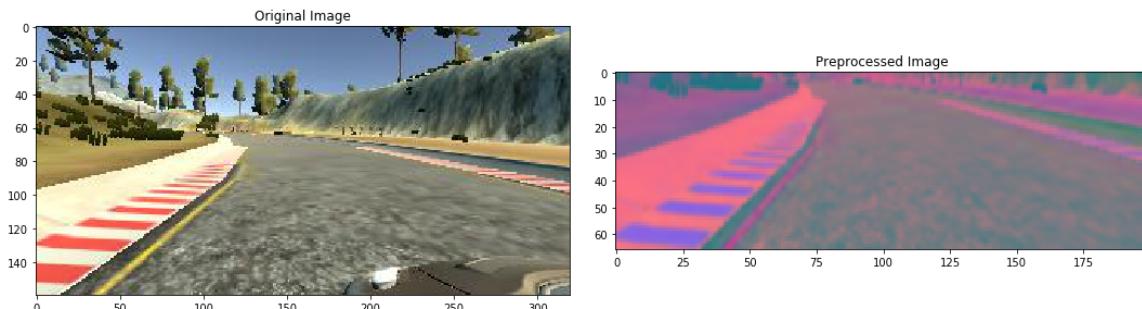
In [37]:

```
image = image_paths[100]
original_image = mpimg.imread(image)
preprocessed_image = img_preprocess(original_image)

fig, axs = plt.subplots(1, 2, figsize=(15, 10))
fig.tight_layout()
axs[0].imshow(original_image)
axs[0].set_title('Original Image')
axs[1].imshow(preprocessed_image)
axs[1].set_title('Preprocessed Image')
```

Out[37]:

Text(0.5, 1.0, 'Preprocessed Image')



A Realtime batch processor with 'Yield' as return to handle realtime simulation

In [38]:

```
def batch_generator(image_paths, steering_ang, batch_size, istraining):

    while True:
        batch_img = []
        batch_steering = []

        for i in range(batch_size):
            random_index = random.randint(0, len(image_paths) - 1)

            if istraining:
                im, steering = random_augment(image_paths[random_index], steering_ang[random_index])

            else:
                im = mpimg.imread(image_paths[random_index])
                steering = steering_ang[random_index]

            im = img_preprocess(im)
            batch_img.append(im)
            batch_steering.append(steering)
        yield (np.asarray(batch_img), np.asarray(batch_steering))
```

Training in batches using batch generator

In [39]:

```
x_train_gen, y_train_gen = next(batch_generator(X_train, y_train, 1, 1))
x_valid_gen, y_valid_gen = next(batch_generator(X_valid, y_valid, 1, 0))
```

Attempting visual verification of training and validation image from batch generator output

In [40]:

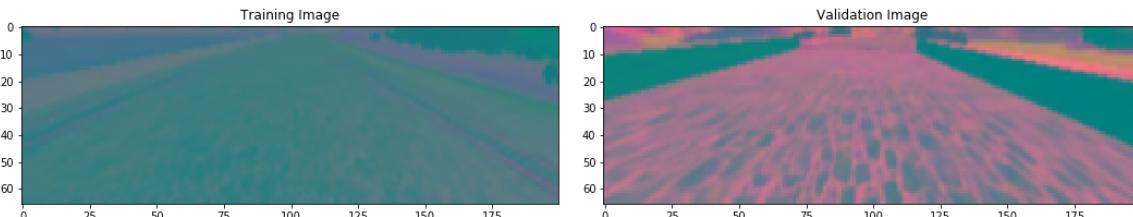
```
fig, axs = plt.subplots(1, 2, figsize=(15, 10))
fig.tight_layout()

axs[0].imshow(x_train_gen[0])
axs[0].set_title('Training Image')

axs[1].imshow(x_valid_gen[0])
axs[1].set_title('Validation Image')
```

Out[40]:

Text(0.5, 1.0, 'Validation Image')



Building NVIDIA Model

Thanks to NVIDIA for paper,figs & model are from two nvidia papers and references mentioned. Fully acknowledged, appreciated and thanked for sharing with public.

Title :End to End Learning for Self-Driving Cars

REF :<https://arxiv.org/pdf/1604.07316v1.pdf> (<https://arxiv.org/pdf/1604.07316v1.pdf>)

I also found another interesting but advanced paper from NVIDIA

Title : Explaining How a Deep Neural Network Trained with End-to-End Learning Steers a Car

<https://arxiv.org/pdf/1704.07911.pdf> (<https://arxiv.org/pdf/1704.07911.pdf>)

In [41]:

```
def nvidia_model():
    model = Sequential()
    model.add(Convolution2D(24, 5, 5, subsample=(2, 2), input_shape=(66, 200, 3),
activation='elu'))
    model.add(Convolution2D(36, 5, 5, subsample=(2, 2), activation='elu'))
    model.add(Convolution2D(48, 5, 5, subsample=(2, 2), activation='elu'))
    model.add(Convolution2D(64, 3, 3, activation='elu'))

    model.add(Convolution2D(64, 3, 3, activation='elu'))
#    model.add(Dropout(0.5))

    model.add(Flatten())

    model.add(Dense(100, activation = 'elu'))
#    model.add(Dropout(0.5))

    model.add(Dense(50, activation = 'elu'))
#    model.add(Dropout(0.5))

    model.add(Dense(10, activation = 'elu'))
#    model.add(Dropout(0.5))

    model.add(Dense(1))

optimizer = Adam(lr=1e-3)
model.compile(loss='mse', optimizer=optimizer)
return model
model = nvidia_model()
print(model.summary())
```

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 31, 98, 24)	1824
conv2d_2 (Conv2D)	(None, 14, 47, 36)	21636
conv2d_3 (Conv2D)	(None, 5, 22, 48)	43248
conv2d_4 (Conv2D)	(None, 3, 20, 64)	27712
conv2d_5 (Conv2D)	(None, 1, 18, 64)	36928
flatten_1 (Flatten)	(None, 1152)	0
dense_1 (Dense)	(None, 100)	115300
dense_2 (Dense)	(None, 50)	5050
dense_3 (Dense)	(None, 10)	510
dense_4 (Dense)	(None, 1)	11

Total params: 252,219

Trainable params: 252,219

Non-trainable params: 0

None

```
/miniconda3/envs/udrivesimul/lib/python3.6/site-packages/ipykernel_launcher.py:3: UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(24, (5, 5), input_shape=(66, 200, ..., activation="elu", strides=(2, 2))`
```

This is separate from the ipykernel package so we can avoid doing imports until

```
/miniconda3/envs/udrivesimul/lib/python3.6/site-packages/ipykernel_launcher.py:4: UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(36, (5, 5), activation="elu", strides=(2, 2))`
```

after removing the cwd from sys.path.

```
/miniconda3/envs/udrivesimul/lib/python3.6/site-packages/ipykernel_launcher.py:5: UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(48, (5, 5), activation="elu", strides=(2, 2))`
```

```
"""
```

```
/miniconda3/envs/udrivesimul/lib/python3.6/site-packages/ipykernel_launcher.py:6: UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(64, (3, 3), activation="elu")`
```

```
/miniconda3/envs/udrivesimul/lib/python3.6/site-packages/ipykernel_launcher.py:8: UserWarning: Update your `Conv2D` call to the Keras 2 API: `Conv2D(64, (3, 3), activation="elu")`
```

Training Time

In [42]:

```
history = model.fit_generator(batch_generator(X_train, y_train, 100, 1),
                               steps_per_epoch=300,
                               epochs=10,
                               validation_data=batch_generator(X_valid, y_val
id, 100, 0),
                               validation_steps=200,
                               verbose=1,
                               shuffle = 1)
```

```
Epoch 1/10
300/300 [=====] - 343s 1s/step - loss: 0.23
25 - val_loss: 0.1145
Epoch 2/10
300/300 [=====] - 334s 1s/step - loss: 0.10
38 - val_loss: 0.0730
Epoch 3/10
300/300 [=====] - 329s 1s/step - loss: 0.07
58 - val_loss: 0.0586
Epoch 4/10
300/300 [=====] - 330s 1s/step - loss: 0.08
37 - val_loss: 0.0595
Epoch 5/10
300/300 [=====] - 322s 1s/step - loss: 0.06
75 - val_loss: 0.0577
Epoch 6/10
300/300 [=====] - 330s 1s/step - loss: 0.05
86 - val_loss: 0.0420
Epoch 7/10
300/300 [=====] - 325s 1s/step - loss: 0.05
31 - val_loss: 0.0393
Epoch 8/10
300/300 [=====] - 333s 1s/step - loss: 0.05
01 - val_loss: 0.0357
Epoch 9/10
300/300 [=====] - 323s 1s/step - loss: 0.04
79 - val_loss: 0.0356
Epoch 10/10
300/300 [=====] - 332s 1s/step - loss: 0.04
69 - val_loss: 0.0435
```

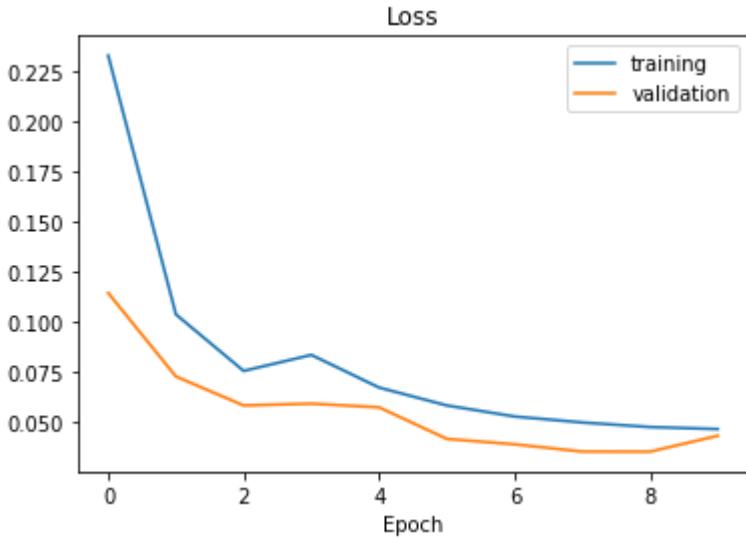
Plotting Model loss

In [43]:

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.legend(['training', 'validation'])
plt.title('Loss')
plt.xlabel('Epoch')
```

Out[43]:

```
Text(0.5, 0, 'Epoch')
```



Saving Model to File system.

In [44]:

```
model.save('selected_model_BC1.h5')
```

End