```
Listing of original (and were noted modified from other source) code used in this project
# 1. augmentation.py
# Note: code modified from original in
# https://github.com/naokishibuya/car-behavioral-cloning/blob/master/utils.py
# Available for audit in audit files/naoki from sharepoint link
import cv2, os
import numpy as np
import matplotlib.image as mpimg
import conf
IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = conf.image height, conf.image width, conf.image
INPUT SHAPE = (IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS)
IMAGE HEIGHT NET, IMAGE WIDTH NET = conf.image height net, conf.image width net
def load image(image path):
  Load RGB images from a file
  return mpimg.imread(image path)
def crop(image):
  Crop the image (removing the sky at the top and the car front at the bottom)
  # this breaks nvidia baseline
  return image[60:-25, :, :] # remove the sky and the car front
def resize(image):
  Resize the image to the input shape used by the network model
  return cv2.resize(image, (IMAGE WIDTH NET, IMAGE HEIGHT NET), cv2.INTER AREA)
def rgb2yuv(image):
  Convert the image from RGB to YUV (This is what the NVIDIA model does)
  return cv2.cvtColor(image, cv2.COLOR RGB2YUV)
```

def preprocess(image):

image = crop(image)

Combine all preprocess functions into one

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image = resize(image)
  image = rgb2yuv(image)
  return image
"
# only one camera in our case
def choose image(data dir, center, left, right, steering angle):
  Randomly choose an image from the center, left or right, and adjust
  the steering angle.
  choice = np.random.choice(3)
  if choice == 0:
     return load image(data dir, left), steering angle + 0.2
  elif choice == 1:
     return load image(data dir, right), steering angle - 0.2
  return load image(data dir, center), steering angle
def random flip(image, steering angle):
  Randomly flipt the image left <-> right, and adjust the steering angle.
  if np.random.rand() < 0.5:
     image = cv2.flip(image, 1)
     steering angle = -steering angle
  return image, steering angle
def random translate(image, steering angle, range x, range y):
  Randomly shift the image vertically and horizontally (translation).
  trans x = range x * (np.random.rand() - 0.5)
  trans y = range y * (np.random.rand() - 0.5)
  steering angle += trans x * 0.002
  trans m = np.float32([[1, 0, trans x], [0, 1, trans y]])
  height, width = image.shape[:2]
  image = cv2.warpAffine(image, trans m, (width, height))
  return image, steering angle
def random shadow(image):
  Generates and adds random shadow
  \# (x1, y1) and (x2, y2) forms a line
  # xm, ym gives all the locations of the image
  x1, y1 = IMAGE WIDTH * np.random.rand(), 0
  x2, y2 = IMAGE WIDTH * np.random.rand(), IMAGE HEIGHT
  xm, ym = np.mgrid[0:IMAGE HEIGHT, 0:IMAGE WIDTH]
  # mathematically speaking, we want to set 1 below the line and zero otherwise
  # Our coordinate is up side down. So, the above the line:
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\# (ym-y1)/(xm-x1) \ge (y2-y1)/(x2-x1)
  # as x^2 == x^1 causes zero-division problem, we'll write it in the below form:
  \# (ym-y1)*(x2-x1) - (y2-y1)*(xm-x1) > 0
  mask = np.zeros like(image[:, :, 1])
  mask[(ym - y1) * (x2 - x1) - (y2 - y1) * (xm - x1) > 0] = 1
  # choose which side should have shadow and adjust saturation
  cond = mask == np.random.randint(2)
  s ratio = np.random.uniform(low=0.2, high=0.5)
  # adjust Saturation in HLS(Hue, Light, Saturation)
  hls = cv2.cvtColor(image, cv2.COLOR RGB2HLS)
  hls[:, :, 1][cond] = hls[:, :, 1][cond] * s ratio
  return cv2.cvtColor(hls, cv2.COLOR HLS2RGB)
def random brightness(image):
  Randomly adjust brightness of the image.
  # HSV (Hue, Saturation, Value) is also called HSB ('B' for Brightness).
  hsv = cv2.cvtColor(image, cv2.COLOR_RGB2HSV)
  ratio = 1.0 + 0.4 * (np.random.rand() - 0.5)
  hsv[:,:,2] = hsv[:,:,2] * ratio
  return cv2.cvtColor(hsv, cv2.COLOR HSV2RGB)
def augment(image, steering angle, range x=100, range y=10):
  Generate an augumented image and adjust steering angle.
  (The steering angle is associated with the center image)
  # resize according to expected input shape e.g. AlexNet 224x224, Udacity 320x160, Unity 160x120, etc
  # set in conf.py
  image = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
  #image, steering angle = choose image(data dir, center, left, right, steering angle)
  image, steering angle = random flip(image, steering angle)
  image, steering angle = random translate(image, steering angle, range x, range y)
  image = random shadow(image)
  image = random brightness(image)
  return image, steering angle
# 2. Augmentation.py
# Note: code modified from original in
# https://github.com/naokishibuya/car-behavioral-cloning/blob/master/utils.py
# Available for audit in audit files/naoki from sharepoint link
```

import cv2, os import numpy as np import matplotlib.image as mpimg import conf

```
class Augmentation():
  Augmentation methods
  img dims = []
  def __init__(self, model):
    Set image dimensions for model
    Inputs
       model: string, model name
    self.img dims = self.get image dimensions(model)
  def get image dimensions(self, model):
    Get the required dimensions for image model, used for resizing and cropping images
       model: string, name of network model
    Output
       int: IMAGE WIDTH, IMAGE HEIGHT, IMAGE WIDTH NET, IMAGE HEIGHT NET, TOP CROP,
BOTTOM CROP
    if (model == conf.ALEXNET):
       return conf.alexnet img dims
    elif (model == conf.NVIDIA1):
       return conf.nvidia1 img dims
    elif (model == conf.NVIDIA2):
       return conf.nvidia2 img dims
    elif (model == conf.NVIDIA BASELINE):
       return conf.nvidia baseline img dims
    else:
       # default to nvidia1
       return conf.nvidia1 img dims
  def load image(self, image path):
    Load RGB images from a file
    return mpimg.imread(image path)
  def crop(self, image):
    Crop the image (removing the sky at the top and the car front at the bottom)
    # this breaks nvidia baseline
    # return image[60:-25, :, :] # remove the sky and the car front
    return image[self.img_dims[conf.IMG_TOP_CROP_IDX]:self.img_dims[conf.IMG_BOTTOM_CROP_IDX],
:,
        :] # remove the sky and the car front
```

def resize(self, image):

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Resize the image to the input shape used by the network model
    return cv2.resize(image, (self.img_dims[conf.IMG_WIDTH_NET_IDX], self.img_dims[conf.IMG_HEIGHT_
NET IDX]),
               cv2.INTER AREA)
  def resize expected(self, image):
    Resize the image to the expected original shape
    return cv2.resize(image, (self.img_dims[conf.IMG_WIDTH_IDX], self.img_dims[conf.IMG_HEIGHT_IDX]),
cv2.INTER AREA)
  def rgb2yuv(self, image):
    Convert the image from RGB to YUV (This is what the NVIDIA model does)
    return cv2.cvtColor(image, cv2.COLOR RGB2YUV)
  def preprocess(self, image):
    Combine all preprocess functions into one
    image = self.crop(image)
    image = self.resize(image)
    image = self.rgb2yuv(image)
    return image
  def random flip(self, image, steering angle):
    Randomly flipt the image left <-> right, and adjust the steering angle.
    if np.random.rand() < 0.5:
       image = cv2.flip(image, 1)
       steering angle = -steering angle
    return image, steering angle
  def random translate(self, image, steering angle, range x, range y):
    Randomly shift the image vertically and horizontally (translation).
    trans x = range x * (np.random.rand() - 0.5)
    trans y = range y * (np.random.rand() - 0.5)
    steering angle += trans x * 0.002
    trans m = np.float32([[1, 0, trans x], [0, 1, trans y]])
    height, width = image.shape[:2]
    image = cv2.warpAffine(image, trans_m, (width, height))
    return image, steering angle
  def random shadow(self, image):
    Generates and adds random shadow
```

```
\# (x1, y1) and (x2, y2) forms a line
    # xm, ym gives all the locations of the image
    x1, y1 = self.img dims[conf.IMG WIDTH IDX] * np.random.rand(), 0
    x2, y2 = self.img dims[conf.IMG WIDTH IDX] * np.random.rand(), self.img dims[conf.IMG HEIGHT ID
X
    xm, ym = np.mgrid[0:self.img dims[conf.IMG HEIGHT IDX], 0:self.img dims[conf.IMG WIDTH IDX]]
    # mathematically speaking, we want to set 1 below the line and zero otherwise
    # Our coordinate is up side down. So, the above the line:
    \# (ym-y1)/(xm-x1) > (y2-y1)/(x2-x1)
    # as x^2 == x^1 causes zero-division problem, we'll write it in the below form:
    \# (ym-y1)*(x2-x1) - (y2-y1)*(xm-x1) > 0
    mask = np.zeros like(image[:, :, 1])
    mask[(ym - y1) * (x2 - x1) - (y2 - y1) * (xm - x1) > 0] = 1
    # choose which side should have shadow and adjust saturation
    cond = mask == np.random.randint(2)
    s ratio = np.random.uniform(low=0.2, high=0.5)
    # adjust Saturation in HLS(Hue, Light, Saturation)
    hls = cv2.cvtColor(image, cv2.COLOR RGB2HLS)
    hls[:, :, 1][cond] = hls[:, :, 1][cond] * s ratio
    return cv2.cvtColor(hls, cv2.COLOR HLS2RGB)
  def random brightness(self, image):
    Randomly adjust brightness of the image.
    # HSV (Hue, Saturation, Value) is also called HSB ('B' for Brightness).
    hsv = cv2.cvtColor(image, cv2.COLOR_RGB2HSV)
    ratio = 1.0 + 0.4 * (np.random.rand() - 0.5)
    hsv[:, :, 2] = hsv[:, :, 2] * ratio
    return cv2.cvtColor(hsv, cv2.COLOR HSV2RGB)
  def augment(self, image, steering angle, range x=100, range y=10):
    Generate an augumented image and adjust steering angle.
    (The steering angle is associated with the center image)
    # resize according to expected input shape e.g. AlexNet 224x224, Udacity 320x160, Unity 160x120, etc
    # set in conf.pv
    #image = cv2.resize(image, (self.img_dims[conf.IMG_WIDTH_IDX], self.img_dims[conf.IMG_HEIGHT_ID
X]),
                 cv2.INTER AREA)
    image = self.resize expected(image)
    # image, steering angle = choose image(data dir, center, left, right, steering angle)
    image, steering angle = self.random flip(image, steering angle)
    image, steering angle = self.random translate(image, steering angle, range x, range y)
    image = self.random shadow(image)
    image = self.random brightness(image)
    return image, steering angle
```

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# 3. augment.ipynb.py
#!/usr/bin/env python
# coding: utf-8
# In[1]:
# start jupyter notebooks from prompt to load all required libraries
# jupyter notebook
import cv2
# In[3]:
import matplotlib.image as mpimg
fp = '/home/simbox/git/msc-data/unity/log2/logs_Fri_Jul_10_09_16_18_2020/10000_cam-image_array_.jpg'
# fp = '/home/simbox/Downloads/IMG/center 2020 11 10 22 02 48 622.jpg'
img = mpimg.imread(fp)
# adapt to naoki net, we have 160w x 120h, first scale to 200
img = cv2.resize(img, (320,160), cv2.INTER AREA)
import matplotlib pyplot as plt
plt.imshow(img)
#plt.imshow(img)\n",
# plt.imshow(img[61:-25, :, :]) # image is 120h160w3d: 60:-25 ~ start at h pixel index 60, end at index (120) - 25
                 # equivalent to img[60:95, :, :]
                 # plain english: remove sky and car shadow
# print(img[59:-25, :, :].shape)
  # img[50:-25, :, :].shape"
# In[8]:
# first resize
image = load image(fp)
image = cv2.resize(image, (320, 160), cv2.INTER AREA)
image = crop(image)
plt.imshow(image)
# In[33]:
image = load image(fp)
image = cv2.resize(image, (320, 160), cv2.INTER AREA)
steering angle = 0.07
# image, steering angle = augment(image, steering angle)
# image = crop(image)
# image = crop(image)
# image = resize(image)
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# image = rgb2yuv(image)
# print(steering angle)
plt.imshow(image)
# In[2]:
import cv2, os
import numpy as np
import matplotlib.image as mpimg
# Udacity
IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 160, 320, 3
# Alexnet
#IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 224, 224, 3
# IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 66, 200, 3
# IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 120, 160, 3
INPUT SHAPE = (IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS)
# Dimensions expected by network
# Udacity
IMAGE WIDTH NET, IMAGE HEIGHT NET = 200, 66
# Alexnet
# IMAGE WIDTH NET, IMAGE HEIGHT NET = 224, 224
def load image(image path):
  Load RGB images from a file
  return mpimg.imread(image path)
def crop(image):
  Crop the image (removing the sky at the top and the car front at the bottom)
  ** ** **
  # unity
  # return image[60:-25, :, :] # remove the sky and the car front
  # alexnet
  return image[109:-40, :, :] # remove the sky and the car front
def resize(image):
  Resize the image to the input shape used by the network model
  return cv2.resize(image, (IMAGE_WIDTH_NET, IMAGE_HEIGHT_NET), cv2.INTER_AREA)
def rgb2yuv(image):
```

```
Convert the image from RGB to YUV (This is what the NVIDIA model does)
  return cv2.cvtColor(image, cv2.COLOR RGB2YUV)
def preprocess(image):
  Combine all preprocess functions into one
  image = crop(image)
  image = resize(image)
  image = rgb2yuv(image)
  return image
def choose image(data dir, center, left, right, steering angle):
  Randomly choose an image from the center, left or right, and adjust
  the steering angle.
  choice = np.random.choice(3)
  if choice == 0:
     return load image(data dir, left), steering angle + 0.2
  elif choice == 1:
     return load image(data dir, right), steering angle - 0.2
  return load image(data dir, center), steering angle
def random flip(image, steering angle):
  Randomly flipt the image left <-> right, and adjust the steering angle.
  if np.random.rand() < 0.5:
     image = cv2.flip(image, 1)
     steering angle = -steering angle
  return image, steering angle
def random translate(image, steering angle, range x, range y):
  Randomly shift the image virtially and horizontally (translation).
  trans x = range x * (np.random.rand() - 0.5)
  trans y = range y * (np.random.rand() - 0.5)
  steering angle += trans x * 0.002
  trans m = np.float32([[1, 0, trans x], [0, 1, trans y]])
  height, width = image.shape[:2]
  image = cv2.warpAffine(image, trans m, (width, height))
  return image, steering angle
def random shadow(image):
  Generates and adds random shadow
```

```
*****
  \# (x1, y1) and (x2, y2) forms a line
  # xm, ym gives all the locations of the image
  x1, y1 = IMAGE WIDTH * np.random.rand(), 0
  x2, y2 = IMAGE WIDTH * np.random.rand(), IMAGE HEIGHT
  # could this be a bug?
  xm, ym = np.mgrid[0:IMAGE HEIGHT, 0:IMAGE WIDTH]
  # xm, ym = np.mgrid[0:IMAGE WIDTH, 0:IMAGE HEIGHT]
  # mathematically speaking, we want to set 1 below the line and zero otherwise
  # Our coordinate is up side down. So, the above the line:
  \# (ym-y1)/(xm-x1) > (y2-y1)/(x2-x1)
  # as x^2 == x^1 causes zero-division problem, we'll write it in the below form:
  \# (ym-y1)*(x2-x1) - (y2-y1)*(xm-x1) > 0
  mask = np.zeros like(image[:, :, 1])
  mask[(ym - y1) * (x2 - x1) - (y2 - y1) * (xm - x1) > 0] = 1
  # choose which side should have shadow and adjust saturation
  cond = mask == np.random.randint(2)
  s ratio = np.random.uniform(low=0.2, high=0.5)
  # adjust Saturation in HLS(Hue, Light, Saturation)
  hls = cv2.cvtColor(image, cv2.COLOR RGB2HLS)
  hls[:, :, 1][cond] = hls[:, :, 1][cond] * s ratio
  return cv2.cvtColor(hls, cv2.COLOR HLS2RGB)
def random brightness(image):
  Randomly adjust brightness of the image.
  # HSV (Hue, Saturation, Value) is also called HSB ('B' for Brightness).
  hsv = cv2.cvtColor(image, cv2.COLOR RGB2HSV)
  ratio = 1.0 + 0.4 * (np.random.rand() - 0.5)
  hsv[:,:,2] = hsv[:,:,2] * ratio
  return cv2.cvtColor(hsv, cv2.COLOR HSV2RGB)
def augment(image, steering angle, range x=100, range y=10):
  Generate an augumented image and adjust steering angle.
  (The steering angle is associated with the center image)
  # resize - we start with assumed image capture size
  image = cv2.resize(image, (320,160), cv2.INTER AREA)
  # image = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
  image, steering angle = random flip(image, steering angle)
  image, steering angle = random translate(image, steering angle, range x, range y)
  image = random shadow(image)
  image = random brightness(image)
  return image, steering angle
```

def batch_generator(data_dir, image_paths, steering_angles, batch_size, is_training):

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```

```
Generate training image give image paths and associated steering angles
  images = np.empty([batch_size, IMAGE_HEIGHT, IMAGE_WIDTH, IMAGE_CHANNELS])
  steers = np.empty(batch_size)
  while True:
    i = 0
    for index in np.random.permutation(image paths.shape[0]):
       center, left, right = image paths[index]
       steering angle = steering angles[index]
       # argumentation
       if is training and np.random.rand() < 0.6:
         image, steering angle = augument(data dir, center, left, right, steering angle)
         image = load image(data dir, center)
       # add the image and steering angle to the batch
       images[i] = preprocess(image)
       steers[i] = steering angle
       i += 1
       if i == batch_size:
         break
    yield images, steers
# In[11]:
# cropping test nvidia1
def crop(image, top crop, bot crop):
  Crop the image (removing the sky at the top and the car front at the bottom)
  # unity
  return image[top crop:bot crop, :, :] # remove the sky and the car front
  # alexnet
  # return image[100:-50, :, :] # remove the sky and the car front
import matplotlib.pvplot as plt
IMAGE WIDTH, IMAGE HEIGHT = 160, 120
fp = '/home/simbox/git/msc-data/unity/log2/logs Fri Jul 10 09 16 18 2020/10000 cam-image array .jpg'
image = load image(fp)
image resized = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (120, 160, 3) nvidia1")
# In[12]:
image nvidial crop = crop(image resized, 60, -25)
# cropping test
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```
print("nvidia1 crop (35, 160, 3)")
print(image nvidial crop.shape)
plt.imshow(image nvidia1 crop)
# In[14]:
image nvidia crop resized = cv2.resize(image nvidia1 crop, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER
AREA)
print(image nvidia crop resized.shape)
print("nvidia1 resized")
plt.imshow(image nvidia crop resized)
# In[25]:
# cropping test nvidia2
import matplotlib.pyplot as plt
IMAGE WIDTH, IMAGE HEIGHT = 320, 160
image resized = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (160, 320, 3) nvidia1")
# In[26]:
image nvidia2 crop 70 35 = \text{crop}(\text{image resized}, 70, -35)
# cropping test
print("nvidia2 70 -35 crop (55, 320, 3)")
print(image nvidia2 crop 70 35.shape)
plt.imshow(image nvidia2 crop 70 35)
# In[27]:
image nvidia crop 70 35 resized = cv2.resize(image nvidia2 crop 70 35, (IMAGE WIDTH, IMAGE HEIGHT
), cv2.INTER AREA)
print(image nvidia crop 70 35 resized.shape)
print("nvidia1 crop 70 -35 resized")
plt.imshow(image nvidia crop 70 35 resized)
# In[20]:
image nvidia2 crop 91 35 = \text{crop}(\text{image resized}, 91, -35)
# cropping test
```

```
print("nvidia2 91 -35 crop (35, 160, 3)")
print(image nvidia2 crop_91_35.shape)
plt.imshow(image nvidia2 crop 91 35)
# In[35]:
IMAGE WIDTH, IMAGE HEIGHT = 320, 160
image resized = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (160, 320, 3) nvidia2")
# In[36]:
image nvidia2 crop 77 35 = \text{crop}(\text{image resized}, 77, -35)
# cropping test
print("nvidia2 77 -35 crop (55, 320, 3)")
print(image nvidia2 crop 77 35.shape)
plt.imshow(image nvidia2 crop 77 35)
# In[37]:
image resized = cv2.resize(image nvidia2 crop 77 35, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER ARE
A)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (160, 320, 3) nvidia2")
# In[21]:
# cropping test nvidia2 - higher crop to include all road markings
import matplotlib.pyplot as plt
IMAGE WIDTH, IMAGE HEIGHT = 320, 160
image resized = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (160, 320, 3) nvidia1")
# In[32]:
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```
image nvidia2 crop 81 35 = \text{crop}(\text{image resized}, 81, -35)
# cropping test
print("nvidia2 81 -35 crop (35, 160, 3)")
print(image nvidia2_crop_81_35.shape)
plt.imshow(image nvidia2 crop 81 35)
# In[33]:
IMAGE WIDTH, IMAGE HEIGHT = 320, 160
image resized = cv2.resize(image nvidia2 crop 81 35, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER ARE
A)
# plt.imshow(image)
plt.imshow(image resized)
print(image resized.shape)
print("Expected (160, 320, 3) nvidia2")
# In[34]:
IMAGE WIDTH, IMAGE HEIGHT = 320, 160
image resized = cv2.resize(image nvidia2 crop 81 35, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER ARE
A)
image nvidia2 crop 77 35 = \text{crop}(\text{image resized}, 77, -35)
# cropping test
print("nvidia2 77 -35 crop (35, 160, 3)")
print(image nvidia2 crop 77 35.shape)
plt.imshow(image nvidia2 crop 77 35)
# In[64]:
# images for dissertation - Augmentation section
# Finally move from RGB to YUV colour space
image = load image(fp)
image = cv2.resize(image, (320, 160), cv2.INTER AREA)
steering angle = 0.07
image, steering angle = augment(image, steering angle)
image = crop(image)
image = resize(image)
image = rgb2yuv(image)
# print(steering angle)
plt.imshow(image)
print(image.shape) # original size of this image: 120 x 160
# In[103]:
```

images for dissertation - Augmentation section

```
image = load image(fp)
image1 = cv2.resize(image, (320, 160), cv2.INTER AREA) # original NVIDIA capture size
steering angle = 0.07
image2, steering angle = augment(image1, steering angle) # augmented
image3 = crop(image2) # cropped
image4 = resize(image3) # resized to network design
image5 = rgb2yuv(image4) # RGB to YUV transform
# print(steering angle)
#plt.imshow(image)
#print(image.shape) # original size of this image: 120 x 160
# plot for dissertation - cannot adjust padding between rows so further image processing required
import matplotlib.pyplot as plt
import numpy as np
import matplotlib
matplotlib.rcParams['font.size'] = 12.0
fig, axs = plt.subplots(2, 3)
fig.set figheight(15)
fig.set figwidth(15)
# Raw
axs[0, 0].title.set text('Raw' + str(image1.shape))
axs[0, 0].imshow(image1)
# Augmented
axs[0, 1].title.set text('Augmented' + str(image2.shape))
axs[0, 1].imshow(image2)
# cropped
axs[0, 2].title.set text('Cropped' + str(image3.shape))
axs[0, 2].imshow(image3)
# resized
axs[1, 0].title.set text('Resized' + str(image4.shape))
axs[1, 0].imshow(image4)
# RGBtoYUV
axs[1, 1].title.set_text('RGB to YUV' + str(image5.shape))
axs[1, 1].imshow(image5)
# Dummy to better format
axs[1, 2].title.set text('RGB to YUV ' + str(image5.shape))
axs[1, 2].imshow(image5)
plt.show()
# In[91]:
```

import numpy as np

```
import matplotlib.pyplot as plt
from matplotlib import gridspec
nrow = 2
ncol = 3
fig = plt.figure(figsize=(10, 10))
gs = gridspec.GridSpec(nrow, ncol, width ratios=[1, 1, 1],
     wspace=0.0, hspace=0.0, top=0.95, bottom=0.05, left=0.17, right=0.845)
for i in range(2):
  for j in range(3):
    im = np.random.rand(28,28)
    ax = plt.subplot(gs[i,j])
    ax.imshow(image)
    ax.set xticklabels([])
    ax.set yticklabels([])
#plt.tight layout() # do not use this!!
plt.show()
#4. conf.py
# Note: code based on
# https://github.com/tawnkramer/sdsandbox/blob/master/src/conf.py
# Available for audit in audit files/tawn from sharepoint link
import math
training patience = 6
training default epochs = 100
training default aug mult = 1
training default aug percent = 0.0
learning rate = 0.00001 \# 0.00001
# default model name
# model name = 'nvidia1'
#nvidia 1 - use this size for both
#nvidia 1 and 2, just change nvidia size before
# presenting to neural network
#image width = 160
#image height = 120
#nvidia 2
# size augmentation process is expecting, i.e. what came from camera
# AlexNet 224x224, Udacity 320x160, Unity 160x120, etc
# IMAGE DIMS INDEXES
# expected original image size
```

```
IMG WIDTH IDX = 0
IMG HEIGHT IDX = 1
IMG DEPTH IDX = 2
# size to be presented to network
IMG WIDTH NET IDX = 3
IMG HEIGHT NET IDX = 4
# to crop road from image
IMG TOP CROP IDX = 5
IMG BOTTOM CROP IDX = 6
# What these lists mean:
# Expected width, height and depth of acquired image,
# width and height of image expected by network
# top crop and bottom crop to remove car and sky from image
# ALEXNET
ALEXNET = 'alexnet'
alexnet img dims = [224,224,3,224,224,60,-25]
# NVIDIA1
NVIDIA1 = 'nvidia1' # a.k.a. TawnNe,
nvidia1 img dims = [160,120,3,160,120,60,-25]
# NVIDIA2
NVIDIA2 = 'nvidia2' # a.k.a. NaokiNet
nvidia2 img dims = [320,160,3,200,66,81,-35]
#NVIDIA BASELINE
NVIDIA BASELINE = 'nvidia baseline' # a.k.a. NaokiNet
nvidia baseline img dims = [160,120,3,200,66,60,-25]
# Alexnet
image width alexnet = 224
image height alexnet = 224
# nvidia
image width = 160
image_height = 120
#nvidia2 (Udacity NaokiNet)
#image width = 160
#image height = 120
# size network is expecting
image width net = 160
image height net = 120
# same for all
image depth = 3
row = image height net
col = image width net
ch = image depth
# training for steering and throttle:
num outputs = 2
# steering alone:
\# num outputs = 1
throttle out scale = 1.0
# alexnet
```

```
batch size = 64
# Using class members to avoid passing same parameters through various functions
# The original NVIDIA paper mentions augmentation but no cropping i.e. road only
# augmentation
aug = False
# pre-process image: crop, resize and rgb2yuv
preproc = False
# image normalization constant, Unity model maximum steer
norm const = 25
# rain type and slant
rt = "
st = 0
# video recording
VIDEO WIDTH, VIDEO_HEIGHT = 800, 600
IMAGE STILL WIDTH, IMAGE STILL HEIGHT = 800, 600
record = False
def setdims(modelname):
  Set image dimensions for training and predicting
  Inputs
    modelname: string, network name
  Outputs
    none
  Example
  setdims('alexnet') # set width and height to 224
  if(modelname=='alexnet'):
    conf.row = image width alexnet
    self.col = image height alexnet
# 5. data predict.py
# Predict steering angles for a dataset for which a ground truth exists
# dsikar@gmail.com
from __future__ import print_function
import argparse
import fnmatch
import ison
import os
import pickle
import random
from datetime import datetime
from time import strftime
import numpy as np
```

```
from PIL import Image
from tensorflow import keras
import tensorflow as tf
import conf
import models
from helper functions import hf mkdir
from augmentation import augment, preprocess
import cv2
from train import load ison, get files
from augmentation import preprocess
from utils.steerlib import gos, plotSteeringAngles
from pathlib import Path
from tensorflow.python.keras.models import load model
# 1. get a list of files to predict (sequential)
# 2. read the steering angle (ison file)
#3. generate a prediction
#4. Store results in list
# 5. Generate a "goodness of steer" value (average steering error)
#6. Generate graph
def predict drive(datapath, modelpath, nc):
  Generate predictions from a model for a dataset
  Inputs
     datapath: string, path to data
     modelpath: string, path to trained model
     nc: steering angle normalization constant
  print("loading model", modelpath)
  model = load model(modelpath)
  # In this mode, looks like we have to compile it
  # NB this is a bit tricky, do need to use optimizer and loss function used to train model?
  model.compile("sgd", "mse")
  files = get files(datapath, True)
  outputs = \prod
  for fullpath in files:
    frame number = os.path.basename(fullpath).split(" ")[0]
    json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
     data = load json(json filename)
     # ground truth
     steering = float(data["user/angle"]) # normalized - divided by nc by simulator
     # prediction
     image = cv2.imread(fullpath)
     # The image will be 1. resized to expected pre-processing size and 2.resized to expected
     # size to be presented to network. This is network architecture and dataset dependant and
     # currently managed in conf.py
     image = preprocess(image)
     image = image.reshape((1,) + image.shape)
```

```
mod pred = model.predict(image)
     # append prediction and ground truth to list
     outputs.append([mod_pred[0][1], steering])
  # get goodness of steer
  sarr = np.asarray(outputs)
  p = sarr[:, 0]
  g = sarr[:, 1]
  gs = gos(p,g,nc)
  print(gs)
  # def plotSteeringAngles(p, g=None, n=1, save=False, track= "Track Name", mname="model name", title='title'):
  gss = "\{:.2f\}".format(gs)
  modelpath = modelpath.split('/')
  datapath = datapath.split('/')
  plotSteeringAngles(p, g, nc, True, datapath[-2], modelpath[-1], 'Gs' + gss)
# dataset ../dataset/unity/jungle1/
# model ../trained models/nvidia2/20201124032017 nvidia2.h5
  # calculate gos (average steering error)
  # plot graph unormalized angles. + average steering error
  # save graph.
  # done
if __name__ == "__main__":
  parser = argparse.ArgumentParser(description='prediction server')
  parser.add argument('--datapath', type=str, default='/home/simbox/git/msc-data/unity/genTrackOneLap 3/*.jpg',
help='model filename')
  parser.add argument('--modelpath', type=str, default='/home/simbox/git/sdsandbox/trained models/sanity/202011
20171015 sanity.h5', help='Model')
  parser.add argument('--nc', type=int, default=1, help='Steering Angle Normalization Constant')
  # time allowing, set image sizes based on model name. For now, these have to be managed in conf.py
  #parser.add argument('--model', type=str, default='nvidia1', help='model name')
  # set dimensions
  args = parser.parse args()
  predict drive(args.datapath, args.modelpath, args.nc)
  # max value for slant is 20
  # Example
  # python3 predict client.py --model=../trained models/sanity/20201120171015 sanity.h5 --rain=light --slant=0
```

```
# coding: utf-8
# In[22]:
import json
import numpy as np
import matplotlib.pyplot as plt
import statistics
import seaborn as sns
import os
import fnmatch
def GetSteeringFromtcpflow(filename):
  Get a tcpflow log and extract steering values obtained from network communication between.
  Note, we only plot the predicted steering angle isondict['steering']
  and the value of jsondict['steering angle'] is ignored. Assumed to be the steering angle
  calculated by PID given the current course.
  sim and prediction engine (predict client.py)
  Inputs
     filename: string, name of tcpflow log
  Returns
     sa: list of arrays, steering angle predicton and actual value tuple.
  Example
  *****
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
     \#readline = f.read()
     lines = file.splitlines()
     for line in lines:
       # print(line)
       start = line.find('{')
       if(start == -1):
          continue
       jsonstr = line[start:]
       # print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in jsondict:
          # predicted
          pred = jsondict['steering']
          # jsondict['steering angle']
          # sa.append([float(pred), act])
          sa.append([float(pred), float(pred)]) # append twice to keep code from breaking
       #if "steering angle" in jsondict:
          # actual
        # act = jsondict['steering angle']
```

```
# save pair, only keep last pred in case two were send as it does happen i.e.:
         # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "t
hrottle": "0.08249988406896591", "brake": "0.0"}
          # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "t
hrottle": "0.08631626516580582", "brake": "0.0"}
         # 127.000.000.001.09091-127.000.000.001.59460: {"msg_type":"telemetry","steering_angle":-0.0719603
7,(...)
        # if(len(pred) > 0):
             sa.append([float(pred), act])
             pred = " # need to save this image
          # deal with image later, sort out plot first
          #imgString = jsondict["image"]
          #image = Image.open(BytesIO(base64.b64decode(imgString)))
          #img arr = np.asarray(image, dtype=np.float32)
  except Exception as e:
     print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
  return sa
def plotBinsFromArray(svals, nc=25, pname=None, logname = "tcpflow log name"):
  svalscp = [element * nc for element in svals]
  values = len(svalscp)
  mymean = ("%.2f" % statistics.mean(svalscp))
  mystd = ("%.2f" % statistics.stdev(svalscp))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalscp, bins=nc*2, kde=False, norm hist=True,
  axlabel= "tcpflow: " + logname + ", model: " + pname + ' steer. degs. norm. hist. ' + str(values) + ' values, mean =
' + mymean + ' std = ' + mystd)
  #if(save):
  # sns.save("output.png")
  plt.savefig(pname + '.png')
# Steering angle predictions by model 20201107210627 nvidia1.h5
def plotSteeringAngles(p, g, n, save=False, track= "Track Name", mname="model.h5"):
  Plot predicted steering angles
  plt.rcParams["figure.figsize"] = (18,3)
  plt.plot(p*25)
  # plt.plot(sarr[:,1]*25, label="simulator")
  plt.ylabel('Steering angle')
  plt.xlabel('Frame number')
  # Set a title of the current axes.
  mytitle = 'tcpflow log predicted steering angles: track ' + str(track) + ' model ' + str(mname)
  plt.title(mytitle)
  # show a legend on the plot
  #plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
```

```
# set limit
  plt.xlim([-5,len(p)+5])
  plt.gca().invert yaxis()
  plt.show()
def plotMultipleSteeringAngles(p, n, save=False, track= "Track Name", mname="model.h5", w=18, h=3):
  Plot multiple predicted steering angles
  Inputs
     p: list of tuples, steering angles and labels
     n: integer,
  plt.rcParams["figure.figsize"] = (18,3)
  plt.plot(p*25)
  # plt.plot(sarr[:,1]*25, label="simulator")
  plt.ylabel('Steering angle')
  plt.xlabel('Frame number')
  # Set a title of the current axes.
  mytitle = 'tcpflow log predicted steering angles: track ' + str(track) + ' model ' + str(mname)
  plt.title(mytitle)
  # show a legend on the plot
  #plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
  # set limit
  plt.xlim([-5,len(p)+5])
  plt.gca().invert yaxis()
  plt.show()
def GetJSONSteeringAngles(filemask):
  Get steering angles stored as 'user/angle' attributes in .json files
  Inputs:
     filemask: string, path and mask
  Outputs
     svals: list, steering values
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # sort by create date
  matches = sorted(matches, key=os.path.getmtime)
  # steering values
  svals = []
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split(" ")[0]
```

```
json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
       jobj = load json(json filename)
       svals.append(jobj['user/angle'])
  return svals
def load json(filepath):
  Load a json file
  Inputs
     filepath: string, path to file
  Outputs
     data: dictionary, json key, value pairs
  Example
  path = "~/git/msc-data/unity/roboRacingLeague/log/logs_Sat_Nov_14_12_36_16_2020/record_11640.json"
  js = load json(path)
  with open(filepath, "rt") as fp:
     data = ison.load(fp)
  return data
# In[2]:
sa = GetSteeringFromtcpflow('../trained models/nvidia2/tcpflow/20201207124146 nvidia2 tcpflow.log')
sarr = np.asarray(sa)
p = sarr[:,0]
g = sarr[:,0]
plotSteeringAngles(p, g, 25, False, "Generated Track", "20201207124146 nvidia2.h5")
plotBinsFromArray(p, 25, "20201207124146 nvidia2.h5", "20201207124146 nvidia2 tcpflow.log")
# In[28]:
# plot ground truth steering angles for
filemask = '../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020/*.jpg'
g = GetJSONSteeringAngles(filemask)
# print(type(g)) # list
g = np.asarray(g)
# print(type(g)) # <class 'numpy.ndarray'>
plt.rcParams["figure.figsize"] = (18,3)
nc = 25 # norm. constant, maximum steering angle
plt.plot(g*nc)
# plt.plot(sarr[:,1]*25, label="simulator")
plt.ylabel('Steering angle')
plt.xlabel('Frame number')
# Set a title of the current axes.
mytitle = 'Ground truth steering for logs Wed Nov 25 23 39 22 2020 - One lap recorded from Generated Track i
n "Auto Drive w Rec" mode'
plt.title(mytitle)
plt.grid(axis='y')
```

```
# set limit
plt.xlim([-5,len(g)+5])
plt.gca().invert yaxis()
plt.show()
# In[2]:
# Steering angle predictions by model 20201107210627 nvidia1.h5
def plotSteeringAngles(p, g, n, save=False, track= "Track Name", mname="model.h5"):
  Plot predicted steering angles
  import matplotlib.pyplot as plt
  plt.rcParams["figure.figsize"] = (18,3)
  plt.plot(p*25)
  # plt.plot(sarr[:,1]*25, label="simulator")
  plt.ylabel('Steering angle')
  plt.ylabel('Frame number')
  # Set a title of the current axes.
  plt.title('tcpflow log predicted steering angles: track ' + track + ' model ' + mname)
  # show a legend on the plot
  #plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
  # set limit
  plt.xlim([-5,len(p)+5])
  plt.gca().invert yaxis()
  plt.show()
sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201120184912 sanity.log')
sarr = np.asarray(sa)
p = sarr[:,0]
g = sarr[:,1]
plotSteeringAngles(p, g, 25, False, "Generated Road, ", "20201120184912 sanity.h5")
#sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201120184912 sanity.log')
#plotBinsFromArray(p, 25, "20201120184912 sanity.h5", "tcpflow/20201120184912 sanity.log")
#type(g)
# TypeError: 'str' object is not callable - cleared by restarting kernel (clearing variables?)
# In[3]:
def gos(p, g, n):
  Calculate the goodness-of-steer between a prediction and a ground truth array.
```

```
Inputs
     p: array of floats, steering angle prediction
     g: array of floats, steering angle ground truth.
     n: float, normalization constant
  Output
  gos: float, average of absolute difference between ground truth and prediction arrays
  # todo add type assertion
  assert len(p) == len(g), "Arrays must be of equal length"
  return sum(abs(p - g)) / len(p) * n
\#p = sarr[:,0]
\#g = sarr[:,1]
\#sterr = gos(p,g, 25)
#print("Goodness of steer: {:.2f}".format(sterr))
# In[4]:
import statistics
import matplotlib.pyplot as plt
import seaborn as sns
def plotBinsFromArray(svals, nc=25, pname=None, logname = "tcpflow log name"):
  svalscp = [element * nc for element in svals]
  values = len(svalscp)
  mean = ("%.2f" % statistics.mean(svalscp))
  std = ("%.2f" % statistics.stdev(svalscp))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalscp, bins=nc*2, kde=False, norm hist=True,
  axlabel= "tcpflow: " + logname + ", model: " + pname + ' steer. degs. norm. hist. ' + str(values) + ' values, mean =
' + mean + ' std = ' + std)
  #if(save):
  # sns.save("output.png")
  plt.savefig(pname + '.png')
#sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201120184912 sanity.log')
\#sarr = np.asarray(sa)
\#p = sarr[:,0]
\#p = sarr[:,0]
#plotBinsFromArray(p, 25, "20201120184912_sanity.h5", "tcpflow/20201120184912_sanity.log")
# In[5]:
#### ../dataset/unity/genRoad/tcpflow/20201123162643 sanity.log
sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201120184912 sanity.log')
sarr = np.asarray(sa)
p = sarr[:,0]
p = sarr[:,0]
```

```
plotSteeringAngles(p, g, 25, False, "Generated Road", "20201120184912 sanity.h5")
plotBinsFromArray(p, 25, "20201120184912 sanity.h5", "tcpflow/20201123162643 sanity.log")
# In[6]:
#### ../dataset/unity/genRoad/tcpflow/20201123162643 sanity.log
sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201123162643 sanity.log')
sarr = np.asarray(sa)
p = sarr[:,0]
p = sarr[:,0]
plotSteeringAngles(p, g, 25, False, "Generated Road", "20201123162643 sanity.h5")
plotBinsFromArray(p, 25, "20201123162643 sanity.h5", "tcpflow/20201123162643 sanity.log")
# In[]:
#### ../dataset/unity/genRoad/tcpflow/20201123162643 sanity pp.log
sa = GetSteeringFromtcpflow('../dataset/unity/genRoad/tcpflow/20201123162643 sanity pp.log')
sarr = np.asarray(sa)
p = sarr[:,0]
p = sarr[:,0]
plotSteeringAngles(p, g, 25, False, "Generated Road", "20201123162643 sanity pp.h5")
plotBinsFromArray(p, 25, "20201123162643 sanity.h5", "tcpflow/20201123162643 sanity pp.log")
# In[7]:
sa = GetSteeringFromtcpflow('../trained models/nvidia2/tcpflow/20201207111940 nvidia2 tcpflow.log')
sarr = np.asarray(sa)
p = sarr[:,0]
p = sarr[:,0]
plotSteeringAngles(p, g, 25, False, "Generated Track", "20201207111940 nvidia2.h5")
plotBinsFromArray(p, 25, "20201207111940 nvidia2.h5", "20201207111940 nvidia2 tcpflow.log")
# In[]:
#7. GetSteering.py
import argparse
import fnmatch
import json
```

```
import os
from io import BytesIO
from PIL import Image
import base64
import numpy as np
import matplotlib.pyplot as plt
from augmentation import preprocess
def GetSteering(filename):
  Get a topflow log and extract steering values received from and sent to sim
  Inputs
     filename: string, name of tcpflow log
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
     \#readline = f.read()
     lines = file.splitlines()
     for line in lines:
       #print(line)
       start = line.find('{')
       if(start == -1):
          continue
       jsonstr = line[start:]
       #print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in jsondict:
          # predicted
          pred = jsondict['steering']
       if "steering angle" in jsondict:
          # actual
          act = jsondict['steering angle']
          # save pair, only keep last pred in case two were send as it does happen i.e.:
          # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "t
hrottle": "0.08249988406896591", "brake": "0.0"}
          # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "t
hrottle": "0.08631626516580582", "brake": "0.0"}
          # 127.000.000.001.09091-127.000.000.001.59460: {"msg_type":"telemetry", "steering_angle":-0.0719603
7,(...)
          if(len(pred) > 0):
            sa.append([float(pred), act])
            pred = " # need to save this image
          # deal with image later, sort out plot first
            imgString = jsondict["image"]
            image = Image.open(BytesIO(base64.b64decode(imgString)))
            img arr = np.asarray(image, dtype=np.float32)
            img arr proc = preprocess(img arr)
            stitch = stitchImages(img arr, img arr proc, 160, 120)
            plt.imshow(stitch)
```

```
except Exception as e:
    print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
def stitchImages(a, b, w, h):
  Stitch two images together side by side
  Inputs
    a, b: floating point image arrays
    w, h: integer width and height dimensions
  Output
    c: floating point stitched image array
  # https://stackoverflow.com/questions/30227466/combine-several-images-horizontally-with-python
  total width = w * 2
  max height = h
  a = Image.fromarray(a.astype('uint8'), 'RGB')
  b = Image.fromarray(b.astype('uint8'), 'RGB')
  new im = Image.new('RGB', (total width, max height))
  new im.paste(a, (0,0))
  new im.paste(b, (w,0))
  return new im # new im
if name == " main ":
  parser = argparse.ArgumentParser(description='train script')
  parser.add_argument('--filename', type=str, help='tcpflow log')
  args = parser.parse args()
  GetSteering(args.filename)
#8. isonclean.py
Go though Unity3D generated files, delete .jpg if a corresponding .json file does not exist
$ python3 jsonclean.py --inputs=../../dataset/unity/log/*.jpg --delete=[False|True]
import os
import argparse
import fnmatch
import json
def load json(filename):
  with open(filename, "rt") as fp:
```

```
data = json.load(fp)
  return data
def cleanjson(filemask, delete):
  # filemask = '~/git/sdsandbox/dataset/unity/log/*.jpg'
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
    for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # deleted file count
  dc = 0
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split(" ")[0]
      json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
      try:
         load json(json filename)
       except:
         print('No matching .json file for: ', fullpath)
         # No matching .json file for: ../../dataset/unity/log/logs Mon Jul 13 09 03 21 2020/35095 cam-image
_array_.jpg
         if(delete):
           print("File deleted.")
           os.remove(fullpath)
           dc += 1
         continue
  print("Files deleted:", dc)
def parse bool(b):
  return b == "True"
if name == " main ":
  parser = argparse.ArgumentParser(description='JSON missing file handler/cleaner')
  parser.add argument('--inputs', default='../dataset/unity/jungle1/log/*.jpg', help='input mask to gather images')
  parser.add argument('--delete', type=parse bool, default=False, help='image deletion flag')
  args = parser.parse args()
  cleanjson(args.inputs, args.delete)
# 9. makebins.pv
Go though Unity3D generated files, get all steering angles and plot a histogram
$ python3 makebins.py --inputs=../../dataset/unity/log/*.jpg
import os
import argparse
```

```
import fnmatch
import json
import seaborn as sns
import os
def load json(filename):
  with open(filename, "rt") as fp:
    data = ison.load(fp)
  return data
def cleanjson(filemask):
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
    for filename in fnmatch.filter(filenames, mask):
      matches.append(os.path.join(root, filename))
  # steering values
  svals = []
  for fullpath in matches:
      frame number = os.path.basename(fullpath).split("_")[0]
      json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
      jobj = load json(json filename)
      svals.append(jobj['user/angle'])
  print("Files deleted:", dc)
def parse bool(b):
  return \overline{b} == "True"
if name == " main ":
  parser = argparse.ArgumentParser(description='JSON missing file handler/cleaner')
  parser.add argument('--inputs', default='../dataset/unity/jungle1/log/*.jpg', help='input mask to gather images')
  args = parser.parse args()
  cleanison(args.inputs)
# 10. MakeVideoFromtcpflow.ipynb.py
#!/usr/bin/env python
# coding: utf-8
# In[123]:
#import argparse
#import fnmatch
import json
```

```
#import os
import base64
import datetime
from io import BytesIO
from PIL import Image
import sys
import numpy as np
import matplotlib.pyplot as plt
# convert image array to PIL image, to see if we can concatenate them
# from PIL import Image
from matplotlib import cm
def GetSteeringFromtcpflow(filename):
  Get a tcpflow log and extract steering values obtained from network communication between
  sim and prediction engine (predict client.py)
  Inputs
     filename: string, name of tcpflow log
  Returns
     sa: list of arrays, steering angle predicton and actual value tuple.
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
     \#readline = f.read()
     lines = file.splitlines()
     for line in lines:
       # print(line)
       start = line.find('{')
       if(start == -1):
          continue
       jsonstr = line[start:]
       # print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in isondict:
          # predicted
          pred = jsondict['steering']
       if "steering angle" in jsondict:
          # actual
          act = jsondict['steering angle']
          # save pair, only keep last pred in case two were send as it does happen i.e.:
          # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "t
hrottle": "0.08249988406896591", "brake": "0.0"}
          # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "t
hrottle": "0.08631626516580582", "brake": "0.0"}
          # 127.000.000.001.09091-127.000.000.001.59460: {"msg_type":"telemetry", "steering_angle":-0.0719603
7,(...)
          if(len(pred) > 0):
```

```
sa.append([float(pred), act])
            pred = " # need to save this image
            imgString = jsondict["image"]
            image = Image.open(BytesIO(base64.b64decode(imgString)))
            img arr = np.asarray(image, dtype=np.float32)
            img arr proc = preprocess(img arr)
            img arr = crop(img arr)
            img arr = resize(img arr)
            # something happens here which "fixes" the image, i.e.
            img arr = rgb2yuv(img arr)
            #im = Image.fromarray(np.uint8(cm.gist_earth(img_arr)))
            #stitch = stitchImages(img arr, img arr proc, 160, 120)
            #print(img arr proc.shape)
            #print(type(image))
            # plt.imshow(img arr proc)
            print(type(img arr))
            myconc = np.concatenate((img_arr_proc, img_arr), axis = 1)
            # plt.imshow(img arr)
            plt.imshow(myconc)
            # print(image.size[0])
            sys.exit(0)
  except Exception as e:
     print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
  return sa
sa = GetSteeringFromtcpflow('/tmp/tcpflow.log')
# In[52]:
def stitchImages(a, b, w, h):
  Stitch two images together side by side
     a, b: floating point image arrays
     w, h: integer width and height dimensions
  Output
     c: floating point stitched image array
  # https://stackoverflow.com/questions/30227466/combine-several-images-horizontally-with-python
  total width = w * 2
  max height = h
  # convert to PIL image to paste
  a = Image.fromarray(a.astype('uint8'), 'RGB')
  b = Image.fromarray(b.astype('uint8'), 'YCbCr')
  new im = Image.new('RGB', (total width, max height))
  new im.paste(a, (0,0))
```

```
# convert back to float array
  return new im # np.asarray(new im, dtype=np.float32) # new im
# In[124]:
myimg = np.asarray([[[1,1,1], [1,1,1], [1,1,1]],
            [[1,1,1],[1,1,1],[1,1,1]],
            [[1,1,1],[1,1,1],[1,1,1]])
myimg2 = np.asarray([[[2,2,2], [2,2,2], [2,2,2]],
            [[2,2,2], [2,2,2], [2,2,2]],
            [[2,2,2], [2,2,2], [2,2,2]]]
myimg3 = np.asarray([[[1,1,1], [1,1,1], [1,1,1]],
            [[1,1,1],[1,1,1],[1,1,1]],
            [[1,1,1],[1,1,1],[1,1,1]],
            [[2,2,2], [2,2,2], [2,2,2]],
            [[2,2,2], [2,2,2], [2,2,2]],
            [[2,2,2], [2,2,2], [2,2,2]])
myimg4 = np.asarray([[[1,1,1], [1,1,1], [1,1,1], [2,2,2], [2,2,2], [2,2,2]],
            [[1,1,1],[1,1,1],[1,1,1],[2,2,2],[2,2,2],[2,2,2]],
            [[1,1,1],[1,1,1],[1,1,1],[2,2,2],[2,2,2],[2,2,2]]]
print(myimg.shape)
print(myimg2.shape)
print(myimg3.shape)
print(myimg4.shape)
# np.append(myimg, myimg2, axis=0)
import numpy as np
myconc = np.concatenate((myimg, myimg2), axis = 1)
myconc.shape
# In[127]:
# quick test, can we concatenate if we read from disk?
myimg = load image('../dataset/unity/jungle1/log/logs_Sat_Nov__7_11_11_06_2020/10000_cam-image_array_.jpg'
)
myprocimg = preprocess(myimg)
myconc = np.concatenate((myimg, myprocimg), axis = 1)
plt.imshow(myconc)
# In[4]:
```

new im.paste(b, (w,0))

```
import cv2, os
import numpy as np
import matplotlib.image as mpimg
IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 160, 120, 3
# IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 66, 200, 3
# IMAGE HEIGHT, IMAGE WIDTH, IMAGE CHANNELS = 120, 160, 3
INPUT_SHAPE = (IMAGE_HEIGHT, IMAGE_WIDTH, IMAGE_CHANNELS)
# Dimensions expected by network
IMAGE WIDTH NET, IMAGE HEIGHT NET = 160, 120
def load image(image path):
  Load RGB images from a file
  return mpimg.imread(image path)
def crop(image):
  Crop the image (removing the sky at the top and the car front at the bottom)
  return image[60:-25, :, :] # remove the sky and the car front
def resize(image):
  Resize the image to the input shape used by the network model
  return cv2.resize(image, (IMAGE WIDTH NET, IMAGE HEIGHT NET), cv2.INTER AREA)
def rgb2yuv(image):
  Convert the image from RGB to YUV (This is what the NVIDIA model does)
  return cv2.cvtColor(image, cv2.COLOR RGB2YUV)
def preprocess(image):
  Combine all preprocess functions into one
  image = crop(image)
  image = resize(image)
  image = rgb2yuv(image)
  return image
def choose image(data dir, center, left, right, steering angle):
  Randomly choose an image from the center, left or right, and adjust
```

```
the steering angle.
  choice = np.random.choice(3)
  if choice == 0:
     return load image(data dir, left), steering angle + 0.2
  elif choice == 1:
     return load image(data dir, right), steering angle - 0.2
  return load image(data dir, center), steering angle
def random flip(image, steering angle):
  Randomly flipt the image left <-> right, and adjust the steering angle.
  if np.random.rand() < 0.5:
     image = cv2.flip(image, 1)
     steering angle = -steering angle
  return image, steering angle
def random translate(image, steering_angle, range_x, range_y):
  Randomly shift the image virtually and horizontally (translation).
  trans x = range x * (np.random.rand() - 0.5)
  trans y = range y * (np.random.rand() - 0.5)
  steering angle += trans x * 0.002
  trans m = np.float32([[1, 0, trans x], [0, 1, trans y]])
  height, width = image.shape[:2]
  image = cv2.warpAffine(image, trans m, (width, height))
  return image, steering angle
def random shadow(image):
  Generates and adds random shadow
  \# (x1, y1) and (x2, y2) forms a line
  # xm, ym gives all the locations of the image
  x1, y1 = IMAGE WIDTH * np.random.rand(), 0
  x2, y2 = IMAGE WIDTH * np.random.rand(), IMAGE HEIGHT
  # could this be a bug?
  xm, ym = np.mgrid[0:IMAGE HEIGHT, 0:IMAGE WIDTH]
  # xm, ym = np.mgrid[0:IMAGE WIDTH, 0:IMAGE HEIGHT]
  # mathematically speaking, we want to set 1 below the line and zero otherwise
  # Our coordinate is up side down. So, the above the line:
  \# (ym-y1)/(xm-x1) \ge (y2-y1)/(x2-x1)
  # as x^2 == x^1 causes zero-division problem, we'll write it in the below form:
  \# (ym-y1)*(x2-x1) - (y2-y1)*(xm-x1) > 0
  mask = np.zeros like(image[:, :, 1])
  mask[(ym - y1) * (x2 - x1) - (y2 - y1) * (xm - x1) > 0] = 1
  # choose which side should have shadow and adjust saturation
```

```
cond = mask == np.random.randint(2)
  s ratio = np.random.uniform(low=0.2, high=0.5)
  # adjust Saturation in HLS(Hue, Light, Saturation)
  hls = cv2.cvtColor(image, cv2.COLOR_RGB2HLS)
  hls[:, :, 1][cond] = hls[:, :, 1][cond] * s ratio
  return cv2.cvtColor(hls, cv2.COLOR HLS2RGB)
def random brightness(image):
  Randomly adjust brightness of the image.
  # HSV (Hue, Saturation, Value) is also called HSB ('B' for Brightness).
  hsv = cv2.cvtColor(image, cv2.COLOR RGB2HSV)
  ratio = 1.0 + 0.4 * (np.random.rand() - 0.5)
  hsv[:,:,2] = hsv[:,:,2] * ratio
  return cv2.cvtColor(hsv, cv2.COLOR HSV2RGB)
def augment(image, steering angle, range x=100, range y=10):
  Generate an augumented image and adjust steering angle.
  (The steering angle is associated with the center image)
  # resize - we start with assumed image capture size
  image = cv2.resize(image, (320,160), cv2.INTER AREA)
  # image = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
  image, steering angle = random flip(image, steering angle)
  image, steering angle = random translate(image, steering angle, range x, range y)
  image = random shadow(image)
  image = random brightness(image)
  return image, steering angle
def batch generator(data dir, image paths, steering angles, batch size, is training):
  Generate training image give image paths and associated steering angles
  images = np.empty([batch_size, IMAGE_HEIGHT, IMAGE_WIDTH, IMAGE_CHANNELS])
  steers = np.empty(batch_size)
  while True:
    i = 0
     for index in np.random.permutation(image_paths.shape[0]):
       center, left, right = image paths[index]
       steering angle = steering angles[index]
       # argumentation
       if is training and np.random.rand() < 0.6:
         image, steering angle = augument(data dir, center, left, right, steering angle)
       else:
         image = load image(data dir, center)
       # add the image and steering angle to the batch
       images[i] = preprocess(image)
       steers[i] = steering angle
```

```
i += 1
       if i == batch size:
          break
     yield images, steers
# In[5]:
# Steering angle predictions by model 20201107210627 nvidia1.h5
def plotSteering(p,g,n):
  *****
  Plot
  Inputs
     p: array of floats, steering angle prediction
     g: array of floats, steering angle ground truth.
     n: float, normalization constant
  Output
  import matplotlib.pyplot as plt
  plt.rcParams["figure.figsize"] = (18,3)
  plt.plot(p*n, label="model")
  plt.plot(g*n, label="simulator")
  plt.ylabel('Steering angle')
  # Set a title of the current axes.
  plt.title('Predicted and actual steering angles: SDSandbox simulator and model 20201107210627 nvidia1.h5')
  # show a legend on the plot
  plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
  # set limit
  plt.xlim([-5,len(sarr)+5])
  # invert axis so if seen sideways plot corresponds to direction of steering wheel
  plt.gca().invert yaxis()
  plt.show()
p = sarr[:,0]
g = sarr[:,1]
n = 25 # 25 frames per second
plotSteering(p,g,n)
\# \ g \ s(p,g) = \frac{n^n \ln p(i)-g(i) \cdot p(i)}{N} \times n \ c \ 
# In[2]:
def gos(p, g, n):
  Calculate the goodness-of-steer between a prediction and a ground truth array.
```

```
Inputs
    p: array of floats, steering angle prediction
    g: array of floats, steering angle ground truth.
    n: float, normalization constant
  Output
  gos: float, average of absolute difference between ground truth and prediction arrays
  # todo add type assertion
  assert len(p) == len(g), "Arrays must be of equal length"
  return sum(abs(p - g)) / len(p) * n
p = sarr[:,0]
g = sarr[:,1]
sterr = gos(p,g, 25)
print("Goodness of steer: {:.2f}".format(sterr))
# In[2]:
import os
path = '~/git/sdsandbox/dataset/ford/2017-08-04-V2-Log1-Center'
for root, dirnames, filenames in os.walk(path):
  print(os.path.join(root, filename))
# In[]:
#11. MakeVideo.py
import argparse
import fnmatch
import json
import os
from io import BytesIO
from PIL import Image
import base64
import numpy as np
import matplotlib.pyplot as plt
from augmentation import preprocess
```

import cv2 import conf

import debug.RecordVideo as RecordVideo

```
def MakeVideo(filename, model, preproc=False):
  Make video from tcpflow logged images.
  video.avi is written to disk
  Inputs
    filename: string, name of tcpflow log
    model: name of model to stamp onto video
    preproc: boolean, show preprocessed image next to original
  Output
    none
  # video name
  video name = model + '.avi'
  VIDEO WIDTH, VIDEO HEIGHT = 800, 600
  IMAGE WIDTH, IMAGE HEIGHT = 800, 600
  if(preproc == True): # wide angle
     VIDEO WIDTH = IMAGE WIDTH*2
  video = cv2.VideoWriter(video name, 0, 11, (VIDEO WIDTH, VIDEO HEIGHT)) # assumed 11fps
  # font
  font = cv2.FONT HERSHEY SIMPLEX
  # normalization constant
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
    \#readline = f.read()
    lines = file.splitlines()
    for line in lines:
       #print(line)
       start = line.find('{')
       if(start == -1):
         continue
       jsonstr = line[start:]
       #print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in jsondict:
         # predicted
         pred = jsondict['steering']
       if "steering angle" in jsondict:
         # actual
         act = jsondict['steering angle']
         # save pair, only keep last pred in case two were send as it does happen i.e.:
         # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "t
hrottle": "0.08249988406896591", "brake": "0.0"}
         # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "t
hrottle": "0.08631626516580582", "brake": "0.0"}
         # 127.000.000.001.09091-127.000.000.001.59460: {"msg_type":"telemetry", "steering_angle":-0.0719603
7,(...)
         if(len(pred) > 0):
            # save steering angles
```

```
pred = " # need to save this image
            # process image
            imgString = jsondict["image"]
            # decode string
            image = Image.open(BytesIO(base64.b64decode(imgString)))
            # try to convert to jpg
            #image = np.array(image) # sky colour turns orange (TODO investigate)
            # save
            image.save('frame.jpg')
            # reopen with user-friendlier cv2
            image = cv2.imread('frame.jpg') # 120x160x3
            image copy = image
            # resize so we can write some info onto image
            image = cv2.resize(image, (IMAGE WIDTH, IMAGE HEIGHT), cv2.INTER AREA)
            # add Info to frame
            cv2.putText(image, model, (50, 50), font, 1, (255, 255, 255), 2, cv2.LINE AA)
            # Predicted steering angle
            pst = sa[len(sa)-1][0]
            pst *= conf.norm const
            simst = "Predicted steering angle: {:.2f}".format(pst)
            cv2.putText(image, simst, (50, 115), font, 1, (255, 255, 255), 2, cv2.LINE AA)
            # create a preprocessed copy to compare what simulator generates to what network "sees"
            if (preproc == True): # wide angle
              image2 = preprocess(image copy)
              image2 = cv2.resize(image2, (IMAGE_WIDTH, IMAGE_HEIGHT), cv2.INTER_AREA)
              cv2.putText(image2, 'Network Image', (50, 50), font, 1, (255, 255, 255), 2, cv2.LINE AA)
            # concatenate
            if (preproc == True): # wide angle
              cimgs = np.concatenate((image, image2), axis=1)
              image = cimgs
            # model name
            # model
            video.write(image);
            pred = "
  except Exception as e:
    print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
  cv2.destroyAllWindows()
  video.release()
  return "DummyName.mp4"
if name == " main ":
  parser = argparse.ArgumentParser(description='Make Video script')
  parser.add argument('--filename', type=str, help='tcpflow log')
  parser.add argument('--model', type=str, help='model name for video label')
  args = parser.parse args()
  MakeVideo(args.filename, args.model, True)
  # example
  # python MakeVideo.py --filename=/tmp/tcpflow.log --model=20201120184912 sanity.h5
```

sa.append([float(pred), act])

```
# 12. mech turk.py
import os
import fnmatch
from shutil import copyfile
def MechTurkBatch(filemask, op):
       files = 4445 # we want 50 uniformly distributed files
       mod = 88 \# 445 / 50
       count = 1
       filemask = os.path.expanduser(filemask)
       path, mask = os.path.split(filemask)
       #make path
       os.makedirs(path + op, exist_ok=True)
       for root, dirnames, filenames in os.walk(path):
              for filename in fnmatch.filter(filenames, mask):
                     # matches.append(os.path.join(root, filename))
                     if(count \% mod == 0):
                             src = os.path.join(root, filename)
                            dst = root + '/out1/' + filename
                            #print("cp " + source + "./mech turk-v2-1")
                            copyfile(src, dst)
                     count += 1
if __name__ == "__main__":
       \overline{path} = \frac{1}{100} \frac{1}
       op = 'out1'
       MechTurkBatch(path, op)
#13. models.py
# Note: code based on
# https://github.com/tawnkramer/sdsandbox/blob/master/src/models.py
# Available for audit in audit files/tawn from sharepoint link
# and
# https://github.com/naokishibuya/car-behavioral-cloning/blob/master/model.py
# Available for audit in audit files/naoki from sharepoint link
111
Models
Define the different NN models we will use
```

Author: Tawn Kramer

```
from future import print function
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Input
from tensorflow.keras.layers import Dense, Lambda, ELU
from tensorflow.keras.layers import Activation, Dropout, Flatten, Dense, BatchNormalization
from tensorflow.keras.layers import Cropping2D
from tensorflow.keras.optimizers import Adadelta, Adam, SGD
from tensorflow.keras import initializers, regularizers
# Alexnet
import conf
def show model summary(model):
  model.summary()
  for layer in model.layers:
     print(layer.output shape)
def nvidia baseline(num outputs):
  this model is approximately equal to:
  https://images.nvidia.com/content/tegra/automotive/images/2016/solutions/pdf/end-to-end-dl-using-px.pdf
  Although nothing is said about dropout or activation, which is assumed to be RELU
  Hi Daniel,
  We used the following settings (we haven't documented them in any publication):
  loss function: MSE
  optimizer: adadelta
  learning rate: 1e-4 (but not really used in adadelta)
  dropout: 0.25
  Best regards,
  Urs
  # note row and col values are now from NET
  # Adjust sizes accordingly in conf.py
  row, col, ch = conf.row, conf.col, conf.ch
  drop = 0.1 # spreading droupout
  # batch init = initializers.glorot uniform # Original AlexNet initializers.RandomNormal(mean=0., stddev=0.01);
  img in = Input(shape=(row, col, ch), name='img in')
  x = img in
  # RGB values assumed to be normalized and not centered i.e. x/127.5 - 1.
  x = Lambda(lambda x: x / 255.0)(x)
  x = \text{Conv2D}(24, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d} 1\text{''})(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(36, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d} 2\text{''})(x) \#2\text{nd}
  x = Dropout(drop)(x)
  x = \text{Conv2D}(48, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d } 3\text{''})(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(64, (3, 3), \text{ strides}=(1, 1), \text{ activation}='\text{relu'}, \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1) \# 4\text{th}
  x = Dropout(drop)(x)
```

```
x = \text{Conv2D}(64, (3, 3), \text{strides}=(1, 1), \text{activation}=\text{'relu'}, \text{name}=\text{''conv2d} 5'')(x) #5th
  x = Dropout(drop)(x)
  x = Flatten(name='flattened')(x)
  \# x = Dense(1164, activation='relu', name="dense 1", kernel initializer=batch init, bias initializer='ones')(x)
  \#x = Dropout(drop)(x)
  x = Dense(100, activation='relu', name="dense 2")(x)
  \#x = Dropout(drop)(x)
  x = Dense(50, activation='relu', name="dense 3")(x) # Added in Naoki's model
  \#x = Dropout(drop)(x)
  \# x = Dense(10, activation='relu', name="dense 4", kernel initializer=batch init, bias initializer='zeros')(x)
  \#x = Dropout(drop)(x)
  outputs = []
  # outputs.append(Dense(num outputs, activation='linear', name='steering throttle')(x))
  outputs.append(Dense(num outputs, activation='linear', name='steering')(x))
  model = Model(inputs=[img in], outputs=outputs)
  # opt = Adadelta(learning rate=0.001, rho=0.95, epsilon=1e-07, name="Adadelta")
  opt = Adam(lr=conf.learning rate)
  model.compile(optimizer=opt, loss="mse", metrics=['acc'])
  # add weight decay
  # https://stackoverflow.com/questions/41260042/global-weight-decay-in-keras
  #alpha = 0.0005 # weight decay coefficient
  #for layer in model.layers:
      if isinstance(layer, Conv2D) or isinstance(layer, Dense):
         layer.add loss(lambda: regularizers.12(alpha)(layer.kernel))
      if hasattr(layer, 'bias regularizer') and layer.use bias:
         layer.add loss(lambda: regularizers.12(alpha)(layer.bias))
  return model
def nvidia model1(num outputs):
  This model expects image input size 160hx120w
  this model is inspired by the NVIDIA paper
  https://images.nvidia.com/content/tegra/automotive/images/2016/solutions/pdf/end-to-end-dl-using-px.pdf
  Activation is RELU
  # row, col, ch = conf.row, conf.col, conf.ch
  # better albeit less readable
  row, col, ch = conf.nvidia1 img dims[conf.IMG HEIGHT NET IDX], conf.nvidia1_img_dims[conf.IMG_WID
TH_NET_IDX], conf.nvidia1 img dims[conf.IMG DEPTH IDX]
  drop = 0.1
  img in = Input(shape=(row, col, ch), name='img in')
  x = img in
  \# x = Cropping2D(cropping=((10,0), (0,0)))(x) \#trim 10 pixels off top
  \# x = Lambda(lambda x: x/127.5 - 1.)(x) \# normalize and re-center
  x = Lambda(lambda x: x/255.0)(x)
  x = \text{Conv2D}(24, (5,5), \text{ strides}=(2,2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d} 1\text{''})(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(32, (5,5), \text{ strides}=(2,2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d} 2\text{''})(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(64, (5,5), \text{ strides}=(2,2), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d } 3\text{''})(x)
  x = Dropout(drop)(x)
```

```
x = Conv2D(64, (3,3), strides=(1,1), activation='relu', name="conv2d 4")(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(64, (3,3), \text{ strides}=(1,1), \text{ activation}=\text{'relu'}, \text{ name}=\text{''conv2d} 5\text{''})(x)
  x = Dropout(drop)(x)
  x = Flatten(name='flattened')(x)
  x = Dense(1064, activation='relu')(x)
  x = Dense(100, activation='relu')(x)
  \#x = Dropout(drop)(x)
  x = Dense(50, activation='relu')(x)
  \#x = Dropout(drop)(x)
  x = Dense(10, activation = 'relu')(x)
  outputs = []
  outputs.append(Dense(num outputs, activation='linear', name='steering throttle')(x))
  model = Model(inputs=[img in], outputs=outputs)
  opt = Adam(lr=0.0001)
  model.compile(optimizer=opt, loss="mse", metrics=['acc'])
  # might want to try metrics=['acc', 'loss'] https://stackoverflow.com/questions/51047676/how-to-get-accuracy-of-
model-using-keras
  return model
def nvidia model2(num outputs):
  A.K.A. NaokiNet - https://github.com/naokishibuya/car-behavioral-cloning/blob/master/model.py
  This model expects images of size 66,200,3
  # row, col, ch = conf.row, conf.col, conf.ch
  row, col, ch = conf.nvidia2 img dims[conf.IMG HEIGHT NET IDX], conf.nvidia2 img dims[conf.IMG WID
TH NET IDX], \
            conf.nvidia2 img dims[conf.IMG DEPTH IDX]
  drop = 0.1
  img in = Input(shape=(row, col, ch), name='img in')
  x = img in
  \# x = Cropping2D(cropping=((10,0), (0,0)))(x) \#trim 10 pixels off top
  \# x = Lambda(lambda x: x/127.5 - 1.0)(x) \# normalize and re-center
  x = Lambda(lambda x: x / 255.0)(x)
  x = Conv2D(24, (5, 5), strides=(2, 2), activation='elu', name="conv2d 1")(x)
  x = Dropout(drop)(x)
  x = Conv2D(32, (5, 5), strides=(2, 2), activation='elu', name="conv2d 2")(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(48, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'elu'}, \text{ name}=\text{''conv2d } 3\text{''})(x)
  x = Dropout(drop)(x)
  x = \text{Conv2D}(64, (3, 3), \text{ activation='elu'}, \text{ name="conv2d 4"})(x) \# \text{ default strides=}(1,1)
  x = Dropout(drop)(x)
  x = Conv2D(64, (3, 3), activation='elu', name="conv2d 5")(x)
  x = Dropout(drop)(x)
  x = Flatten(name='flattened')(x)
```

```
x = Dense(100, activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(50, activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(10, activation='elu')(x) # Added in Naoki's model
  outputs = []
  # outputs.append(Dense(num outputs, activation='linear', name='steering throttle')(x))
  outputs.append(Dense(num outputs, name='steering throttle')(x))
  model = Model(inputs=[img in], outputs=outputs)
  opt = Adam(lr=0.0001)
  model.compile(optimizer=opt, loss="mse", metrics=['acc'])
  return model
def nvidia model3(num outputs):
  This model expects images of size 66,200,3
  row, col, ch = conf.row, conf.col, conf.ch
  drop = 0.1
  img in = Input(shape=(row, col, ch), name='img in')
  x = img in
  \# x = Cropping2D(cropping=((10,0), (0,0)))(x) \#trim 10 pixels off top
  x = Lambda(lambda x: x/127.5 - 1.0)(x) # normalize and re-center
  \# x = Lambda(lambda x: x / 255.0)(x)
  x = \text{Conv2D}(24, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'elu'}, \text{ name}=\text{''conv2d} 1\text{''})(x)
  \# x = Dropout(drop)(x)
  x = Conv2D(36, (5, 5), strides=(2, 2), activation='elu', name="conv2d 2")(x)
  \#x = Dropout(drop)(x)
  x = Conv2D(48, (5, 5), strides=(2, 2), activation='elu', name="conv2d 3")(x)
  \#x = Dropout(drop)(x)
  x = \text{Conv2D}(64, (3, 3), \text{activation='elu'}, \text{name=''conv2d 4''})(x) \# \text{ default strides=}(1,1)
  \#x = Dropout(drop)(x)
  x = Conv2D(64, (3, 3), activation='elu', name="conv2d 5")(x)
  x = Dropout(drop)(x)
  x = Flatten(name='flattened')(x)
  x = Dense(100, activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(50, activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(10, activation='elu')(x) # Added in Naoki's model
  outputs = []
  # outputs.append(Dense(num outputs, activation='linear', name='steering throttle')(x))
  outputs.append(Dense(num outputs, name='steering throttle')(x))
  model = Model(inputs=[img in], outputs=outputs)
  opt = Adam(lr=0.0001)
```

```
model.compile(optimizer=opt, loss="mse", metrics=['acc'])
     return model
def get_alexnet(num_outputs):
     this model is also inspired by the NVIDIA paper
     https://images.nvidia.com/content/tegra/automotive/images/2016/solutions/pdf/end-to-end-dl-using-px.pdf
     but taken from
     https://github.com/naokishibuya/car-behavioral-cloning/blob/master/model.py
     model = Sequential()
     model.add(Lambda(lambda x: x/127.5-1.0, input shape=INPUT SHAPE))
     model.add(Conv2D(24, 5, 5, activation='elu', strides=2))
     model.add(Conv2D(36, 5, 5, activation='elu', strides=2))
     model.add(Conv2D(48, 5, 5, activation='elu', strides=2))
     model.add(Conv2D(64, 3, 3, activation='elu'))
     model.add(Conv2D(64, 3, 3, activation='elu'))
     model.add(Dropout(args.keep prob))
     model.add(Flatten())
     model.add(Dense(100, activation='elu'))
     model.add(Dense(50, activation='elu'))
     model.add(Dense(10, activation='elu'))
     model.add(Dense(1))
     model.summary()
     NB Tawn Kramer's model uses dropout = 0.1 on five layers, Naoki uses
     0.5 on a single layer
     #row, col, ch = conf.image width alexnet, conf.image height alexnet, conf.ch
     row, col, ch = conf.alexnet img dims[conf.IMG HEIGHT NET IDX], conf.nvidia2 img dims[conf.IMG WID
TH_NET IDX], \
     conf.nvidia2 img dims[conf.IMG DEPTH IDX]
     drop = 0.5
     # read https://stackoverflow.com/questions/58636087/tensorflow-valueerror-failed-to-convert-a-numpy-array-to-a
-tensor-unsupporte
     # to work out shapes
     img in = Input(shape=(row, col, ch), name='img in')
     x = img in
     \# x = Cropping2D(cropping=((10,0), (0,0)))(x) \#trim 10 pixels off top
     x = Lambda(lambda x: x/127.5 - 1.0)(x) # normalize and re-center
     \# x = Lambda(lambda x: x / 255.0)(x)
     x = \text{Conv2D}(48, (8, 8), \text{ strides}=(4, 4), \text{ padding}='valid', activation='relu', name="conv2d 1")}(x)
     \# x = Dropout(drop)(x)
     x = MaxPooling2D(48, (1, 1), padding="same", name="maxpool2d 1")(x)
     x = \text{Conv2D}(128, (3, 3), \text{ strides}=(2, 2), \text{ padding}='valid', \text{ activation}='relu', \text{ name}=''conv2d 2'')(x)
     \#x = Dropout(drop)(x)
     x = MaxPooling2D(128, (1, 1), padding="same", name="maxpool2d 2")(x)
     x = Conv2D(192, (3, 3), strides=(2, 2), padding='valid', activation='relu', name="conv2d 3")(x)
     \#x = Dropout(drop)(x)
     x = \text{Conv2D}(192, (3, 3), \text{ strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 4'')(x) \# \text{ default strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{same}', \text{ activation}
(1,1)
     \#x = Dropout(drop)(x)
     x = \text{Conv2D}(128, (3, 3), \text{ strides}=(1, 1), \text{ padding}='\text{same}', \text{ activation}='\text{relu}', \text{ name}=''\text{conv2d} 5'')(x)
```

```
x = MaxPooling2D(128, (1, 1), padding="same", name="maxpool2d 3")(x)
  \#x = \text{Conv2D}(64, (3, 3), \text{ activation='relu', name="conv2d 6"})(x)
  # error Negative dimension size caused by subtracting 128 from 10 for '{{node max pooling2d/MaxPool}} = Ma
xPool[T=DT FLOAT,
  # data format="NHWC", ksize=[1, 128, 128, 1], padding="VALID", strides=[1, 3, 3, 1]](conv2d 4/Identity)'
  # with input shapes: [?,10,10,192].
  \# x = MaxPooling2D(128, (3, 3), padding="SAME")(x)
  # By commenting out line above, error is:
  # Input to reshape is a tensor with 1843200 values, but the requested shape requires a multiple of 101568
  # [[node model/flattened/Reshape (defined at /git/sdsandbox/src/train.py:272) ]] [Op: inference train function
1081]
  x = Dropout(drop)(x)
  *****
     x = \text{Conv2D}(24, (5, 5), \text{ strides}=(2, 2), \text{ activation}=\text{'elu'}, \text{ name}=\text{''conv2d} 1\text{''})(x)
  \# x = Dropout(drop)(x)
  x = Conv2D(36, (5, 5), strides=(2, 2), activation='elu', name="conv2d 2")(x)
  \#x = Dropout(drop)(x)
  x = Conv2D(64, (5, 5), strides=(2, 2), activation='elu', name="conv2d 3")(x)
  \#x = Dropout(drop)(x)
  x = Conv2D(64, (3, 3), activation='elu', name="conv2d_4")(x) # default strides=(1,1)
  \#x = Dropout(drop)(x)
  x = MaxPooling2D(64, (3, 3), name="maxpool2d_5")(x)
  x = Dropout(drop)(x)
  x = Flatten(name='flattened')(x)
  x = Dense(2048, activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(2048, activation='elu')(x)
  \# x = Dropout(drop)(x)
  \# x = Dense(10, activation='elu')(x) \# Added in Naoki's model
  x = Flatten(name='flattened')(x) \# error when followed by
  x = Dense(2048, name='Dense 1', activation='relu')(x) # 2048, 2048 \sim Input to reshape is a tensor with 442368 va
lues, but the requested shape requires a multiple of 21632
 # [[node model/flattened/Reshape (defined at /git/sdsandbox/src/train.py:272) ]] [Op: inference train function 1
192]
  \# x = Dropout(drop)(x)
  x = Dense(50, name='Dense 2', activation='elu')(x)
  \# x = Dropout(drop)(x)
  x = Dense(10, activation='elu')(x) # Added in Naoki's model
  outputs = []
  # outputs.append(Dense(num outputs, activation='linear', name='steering throttle')(x))
  outputs.append(Dense(num outputs, name='steering throttle')(x))
  model = Model(inputs=[img in], outputs=outputs)
```

```
opt = Adam(lr=0.0001)
  model.compile(optimizer=opt, loss="mse", metrics=['acc'])
  return model
def alexnet model(img shape=(224, 224, 3), n classes=10, 12 reg=0.,
weights=None):
# Initialize model
alexnet = Sequential()
# Layer 1
alexnet.add(Conv2D(96, (11, 11), input shape=img shape,
 padding='same', kernel regularizer=12(12 reg)))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
alexnet.add(MaxPooling2D(pool size=(2, 2)))
# Layer 2
alexnet.add(Conv2D(256, (5, 5), padding='same'))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
alexnet.add(MaxPooling2D(pool size=(2, 2)))
# Layer 3
alexnet.add(ZeroPadding2D((1, 1)))
alexnet.add(Conv2D(512, (3, 3), padding='same'))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
alexnet.add(MaxPooling2D(pool size=(2, 2)))
# Layer 4
alexnet.add(ZeroPadding2D((1, 1)))
alexnet.add(Conv2D(1024, (3, 3), padding='same'))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
# Layer 5
alexnet.add(ZeroPadding2D((1, 1)))
alexnet.add(Conv2D(1024, (3, 3), padding='same'))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
alexnet.add(MaxPooling2D(pool size=(2, 2)))
# Layer 6
alexnet.add(Flatten())
alexnet.add(Dense(3072))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
alexnet.add(Dropout(0.5))
# Layer 7
alexnet.add(Dense(4096))
alexnet.add(BatchNormalization())
alexnet.add(Activation('relu'))
```

```
alexnet.add(Dropout(0.5))
# Layer 8
alexnet.add(Dense(n classes))
alexnet.add(BatchNormalization())
alexnet.add(Activation('softmax'))
if weights is not None:
 alexnet.load weights(weights)
return alexnet
** ** **
# 14. predict client.py
# Note: code based on
# https://github.com/tawnkramer/sdsandbox/blob/master/src/predict_client.py
# Available for audit in audit files/tawn from sharepoint link
Predict Server
Create a server to accept image inputs and run them against a trained neural network.
This then sends the steering output back to the client.
Author: Tawn Kramer
from future__import print_function
import os
import sys
import argparse
import time
import ison
import base64
import datetime
from io import BytesIO
import signal
import tensorflow as tf
from tensorflow.python import keras
from tensorflow.python.keras.models import load model
from PIL import Image
import numpy as np
from gym donkeycar.core.fps import FPSTimer
from gym donkeycar.core.message import IMesgHandler
from gym donkeycar.core.sim client import SimClient
# same preprocess as for training
from augmentation import augment, preprocess
import conf
from helper functions import parse bool
import utils.RecordVideo as RecordVideo
import Augmentation
if tf. version == '1.13.1':
```

```
from tensorflow import ConfigProto, Session
```

```
# Override keras session to work around a bug in TF 1.13.1
  # Remove after we upgrade to TF 1.14 / TF 2.x.
  config = ConfigProto()
  config.gpu options.allow growth = True
  session = Session(config=config)
  keras.backend.set session(session)
# need to import file TODO
import Automold as am
import Helpers as hp
import numpy as np
# helper function for prediction
def add rain(image arr, rt=None, st=0):
  Add rain to image
  Inputs:
     image arr: numpy array containing image
     rt: string, rain type "heavy" or "torrential"
     st: range to draw a random slant from
  Output
     image arr: numpy array containing image with rain
  # print("Adding rain...")
  if(st != 0):
    # draw a random number for slant
     st = np.random.randint(-1 * st, st)
  if(rt!='light'): # heavy or torrential
     image arr = am.add rain single(image arr, rain type=rt, slant=st)
  else:
     # no slant
     image arr = am.add rain single(image arr)
  return image arr
class DonkeySimMsgHandler(IMesgHandler):
  STEERING = 0
  THROTTLE = 1
  def init (self, model, constant throttle, image cb=None, rand seed=0):
     self.model = model
     self.constant throttle = constant throttle
     self.client = None
     self.timer = FPSTimer()
     self.img arr = None
     self.image cb = image cb
     self.steering angle = 0.
     self.throttle = 0.
     self.rand seed = rand seed
    self.fns = {'telemetry' : self.on_telemetry,\
```

```
'car loaded' : self.on car created,\
         'on_disconnect' : self.on_disconnect,
         'aborted' : self.on aborted}
  # images to record
  self.img orig = None
  self.img add rain = None
  self.img processed = None
  self.frame count = 0
  # model name
def on connect(self, client):
  self.client = client
  self.timer.reset()
def on aborted(self, msg):
  self.stop()
def on disconnect(self):
  pass
def on recv message(self, message):
  self.timer.on frame()
  if not 'msg type' in message:
    print('expected msg type field')
    print("message:", message)
    return
  msg type = message['msg type']
  if msg type in self.fns:
    self.fns[msg type](message)
  else:
    print('unknown message type', msg_type)
def on car created(self, data):
  if self.rand seed != 0:
    self.send regen road(0, self.rand seed, 1.0)
def on telemetry(self, data):
  imgString = data["image"]
  image = Image.open(BytesIO(base64.b64decode(imgString)))
  img arr = np.asarray(image, dtype=np.float32)
  self.frame count += 1
  self.img orig = img arr
  # set to same image size expected from acquisition process
  img arr = ag.resize expected(img arr)
  # check for rain
  if(conf.rt != "):
    img arr = add rain(img arr, conf.rt, conf.st)
    self.img add rain = img arr
```

```
# same preprocessing as for training
  img arr = ag.preprocess(img arr)
  self.img processed = img arr
  #if(conf.record == True):
  # text = (['Network Image', 'No Rain'])
  # rv.add image(img arr, text)
  # if we are testing the network with rain
  self.img arr = img arr.reshape(1,) + img arr.shape)
  if self.image cb is not None:
     self.image cb(img arr, self.steering angle)
def update(self):
  if self.img arr is not None:
     self.predict(self.img arr)
     self.img arr = None
def predict(self, image array):
  outputs = self.model.predict(image array)
  # check if we are recording
  if (conf.record == True):
     # Add first image, with name of network and frame number
     # TODO, get network name from argument
     text = ([rv.modelname, 'Intensity Multiplie: 1', 'Acquired image', 'Frame: ' + str(self.frame count)])
     rv.add image(self.img orig, text)
     # Add second image, preprocessed with rain or without
     # text = (['Network image', 'No rain'])
     # rv.add image(self.img processed, text)
     # if rain added
     # check for rain
     if (conf.rt != "):
       rtype = 'Type: ' + conf.rt
       s = 'Slant: -+' + str(conf.st)
       text = (['Added rain', rtype, s])
       rv.add image(self.img add rain, text)
       # add third image with prediction
     steering = outputs[0][0]
     steering *= conf.norm const
     st_str = "{:.2f}".format(steering)
     st str = "Predicted steering angle: " + st str
     # st str = "Predicted steering angle: 20"
     #rtype = 'Type: ' + conf.rt
     \#s = 'Slant: -+' + str(conf.st)
     text = (["Network image", st str])
     rv.add image(image array[0], text)
     rv.add frame()
  self.parse outputs(outputs)
```

```
def parse outputs(self, outputs):
  res = []
  # Expects the model with final Dense(2) with steering and throttle
  for i in range(outputs.shape[1]):
     res.append(outputs[0][i])
  self.on parsed outputs(res)
def on parsed outputs(self, outputs):
  self.outputs = outputs
  self.steering angle = 0.0
  self.throttle = 0.2
  if len(outputs) > 0:
     self.steering angle = outputs[self.STEERING]
  if self.constant throttle != 0.0:
     self.throttle = self.constant throttle
  elif len(outputs) > 1:
     self.throttle = outputs[self.THROTTLE] * conf.throttle out scale
  self.send control(self.steering angle, self.throttle)
def send control(self, steer, throttle):
  # print("send st:", steer, "th:", throttle)
  msg = { 'msg type' : 'control', 'steering': steer. str (), 'throttle':throttle. str (), 'brake': '0.0' }
  self.client.queue message(msg)
def send regen road(self, road style=0, rand seed=0, turn increment=0.0):
  Regenerate the road, where available. For now only in level 0.
  In level 0 there are currently 5 road styles. This changes the texture on the road
  and also the road width.
  The rand seed can be used to get some determinism in road generation.
  The turn increment defaults to 1.0 internally. Provide a non zero positive float
  to affect the curviness of the road. Smaller numbers will provide more shallow curves.
  msg = { 'msg type' : 'regen road',
     'road style': road style.__str__(),
     'rand seed': rand seed. str (),
     'turn increment': turn increment. str () }
  self.client.queue message(msg)
def stop(self):
  self.client.stop()
def del (self):
  self.stop()
```

```
for client in arr:
     if not client.is connected():
       return False
  return True
def go(filename, address, constant throttle=0, num cars=1, image cb=None, rand seed=None):
  print("loading model", filename)
  model = load model(filename)
  # In this mode, looks like we have to compile it
  model.compile("sgd", "mse")
  clients = []
  for in range(0, num cars):
     # setup the clients
     handler = DonkeySimMsgHandler(model, constant throttle, image cb=image cb, rand seed=rand seed)
     client = SimClient(address, handler)
     clients.append(client)
  while clients connected(clients):
     try:
       time.sleep(0.02)
       for client in clients:
          client.msg handler.update()
     except KeyboardInterrupt:
       # unless some hits Ctrl+C and then we get this interrupt
       print('stopping')
       break
def stop exec(signum, frame):
  # restore the original signal handler as otherwise evil things will happen
  # in raw input when CTRL+C is pressed, and our signal handler is not re-entrant
  signal.signal(signal.SIGINT, original sigint)
  try:
     # changed raw input to input
     if input("\nFinish recording video? (y/n)>").lower().startswith('y'):
       print("*** CTRL+C to stop ***")
       rv.save video()
       sys.exit(1)
  except KeyboardInterrupt:
     print("Ok ok, quitting")
     sys.exit(1)
  # restore the exit gracefully handler here
  signal.signal(signal.SIGINT, stop exec)
if name == " main ":
  parser = argparse.ArgumentParser(description='prediction server')
  parser.add argument('--model', type=str, help='model filename')
```

```
parser.add argument('--modelname', type=str, default='nvidia1', help='model filename')
  parser.add argument('--host', type=str, default='127.0.0.1', help='server sim host')
  parser.add argument('--port', type=int, default=9091, help='bind to port')
  parser.add argument('--num cars', type=int, default=1, help='how many cars to spawn')
  parser.add argument('--constant throttle', type=float, default=0.0, help='apply constant throttle')
  parser.add argument('--rand seed', type=int, default=0, help='set road generation random seed')
  parser.add argument('--rain', type=str, default=", help='type of rain [light|heavy|torrential')
  parser.add argument('--slant', type=int, default=0, help='Rain slant deviation')
  parser.add argument('--record', type=parse bool, default="False", help='Record video of raw and processed imag
es')
  # parser.add argument('--img cnt', type=int, default=3, help='Number of side by side images to record')
  args = parser.parse args()
  address = (args.host, args.port)
  conf.rt = args.rain
  conf.st = args.slant
  conf.record = args.record
  ag = Augmentation.Augmentation(args.modelname)
  if conf.record == True:
    print("*** When finished, press CTRL+C and y to finish recording, then CTRL+C to quit ***")
    original sigint = signal.getsignal(signal.SIGINT)
    signal.signal(signal.SIGINT, stop exec)
    rv = RecordVideo.RecordVideo(args.model, conf.rt)
  go(args.model, address, args.constant throttle, num cars=args.num cars, rand seed=args.rand seed)
  # max value for slant is 20
  # Example
  # python3 predict client.py --model=../trained models/sanity/20201120171015 sanity.h5 --rain=light --slant=0
#15. RecordVideo.py
# https://docs.python.org/3/tutorial/classes.html
import cv2
import conf
import numpy as np
class RecordVideo():
  Record video class, used to record videos from tepflow log or still images
  def __init__(self, model, rt):
    Record video while running predictions
```

```
Inputs
       model: string, model name
       rt: type of rain
    model = model.split('/')
    self.modelname = model[-1]
    videoname = self.modelname + '.avi'
    # 3 images side by side in the rain
    img cnt = 3
    if(rt == "): # 2 images in the dry
       img cnt = 2
    self.img cnt = img cnt
    self.VIDEO WIDTH, self.VIDEO HEIGHT = conf.VIDEO WIDTH, conf.VIDEO HEIGHT # 800, 600
    self.IMAGE WIDTH, self.IMAGE HEIGHT = conf.IMAGE STILL WIDTH, conf.IMAGE STILL HEIGH
T
    self.VIDEO WIDTH = self.IMAGE WIDTH * img cnt
    self.video = cv2.VideoWriter(videoname, 0, 11,
                      (self.VIDEO WIDTH, self.VIDEO HEIGHT)) # assumed 11fps approximately
    self.font = cv2.FONT HERSHEY SIMPLEX
    # video line spacing
    self.images = []
  def add image(self, image, text):
    \# self.img arr 1 =
    # cv2.putText(image, model, (50, 50), font, 1, (255, 255, 255), 2, cv2.LINE_AA)
    # Predicted steering angle
    \# pst = sa[len(sa) - 1][0]
    # pst *= conf.norm const
    # simst = "Predicted steering angle: {:.2f}".format(pst)
    # cv2.putText(image, simst, (50, 115), font, 1, (255, 255, 255), 2, cv2.LINE AA)
    self.images.append([image, text])
  def add frame(self):
    y offset = 50;
    x 	ext{ offset} = 50
    step = 65;
    img cnt = len(self.images)
    # prepare images
    for i in range(0, img cnt):
       # resize image
       image = self.images[i][0]
       image = cv2.resize(image, (self.IMAGE WIDTH, self.IMAGE HEIGHT), cv2.INTER AREA)
       # add text
       lines = self.images[i][1]
       for line in lines:
         cv2.putText(image, line, (x offset, y offset), self.font, 1, (255, 255, 255), 2, cv2.LINE AA)
         v offset += step
         # print(line)
       # set start to top for printing lines in next frame
       y offset = x offset
       # store processed image
       self.images[i][0] = image
    # concatenate
    output image = self.images[0][0]
```

```
for i in range(1, img cnt):
      output image = np.concatenate((output image, self.images[i][0]), axis=1)
    # append
    try:
      self.video.write(np.uint8(output image)) # catch error Assertion failed) image.depth() == CV 8U
    except Exception as e:
      print("Exception raise: " + str(e))
    # blank images
    self.images = []
  def save video(self):
    # save video as videoname
    cv2.destroyAllWindows()
    self.video.release()
# 16. result plots.py
import numpy as np
from utils.steerlib import GetSteeringFromtcpflow, plotMultipleSteeringAngles
def plot1():
  Intensity multiplier 1, 20201207091932 nvidia1.h5
  # intensity multiplier = 1
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 no rain tcpflow.lo
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'no rain'])
  # light mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 light rain mult 1
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 heavy 10 mult 1
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 torrential 20 mult
1 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
```

```
p.append([pa, 'torrential rain slant +-20'])
  plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 1", "20201207091932 nvidia1.h5"
, 'tcpflow log predicted')
def plot2():
  Intensity multiplier 4, 20201207091932 nvidia1.h5
  # intensity multiplier = 4
  p = []
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('.../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 no rain tcpflow.lo
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'no rain'])
  # light mult 1
  sa = GetSteeringFromtcpflow('.../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 light rain mult 4
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 heavy 10 mult 4
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 torrential 20 mult
 4 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'torrential rain slant +-20'])
  plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 4", "20201207091932 nvidia1.h5"
                   'tcpflow log predicted')
def plot3():
  Intensity multiplier 8, 20201207091932 nvidia1.h5
  # intensity multiplier = 8
  p = \lceil \rceil
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 no rain tcpflow.lo
g')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'no rain'])
  # light mult 1
```

```
sa = GetSteeringFromtcpflow(
    '../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 light rain mult 8 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow(
    '../../trained models/nvidia1/tcpflow/20201207091932_nvidia1_heavy_10_mult_8_tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow(
    '../../trained models/nvidia1/tcpflow/20201207091932 nvidia1 torrential 20 mult 8 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'torrential rain slant +-20'])
  plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 8", "20201207091932 nvidia1.h5"
                   'tcpflow log predicted')
def plot4():
  Intensity multiplier 1, 20201207192948 nvidia2.h5
  # intensity multiplier = 1
  p = []
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 no rain tcpflow.lo
g')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'no rain'])
  # light mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 light rain mult 1
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow('../../trained_models/nvidia2/tcpflow/20201207192948_nvidia2_heavy_10_mult_1_
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 torrential 20 mult
1 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'torrential rain slant +-20'])
```

```
plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 1", "20201207192948 nvidia2.h5"
, 'tcpflow log predicted')
def plot5():
  Intensity multiplier 4, 20201207192948 nvidia2.h5
  # intensity multiplier = 4
  p = []
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 no rain tcpflow.lo
g')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'no rain'])
  # light mult 1
  sa = GetSteeringFromtcpflow('..././trained models/nvidia2/tcpflow/20201207192948 nvidia2 light rain mult 4
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow('.../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 heavy 10 mult 4
tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 torrential 20 mult
4 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'torrential rain slant +-20'])
  plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 4", "20201207192948 nvidia2.h5"
                   'tcpflow log predicted')
def plot6():
  Intensity multiplier 8, 20201207192948 nvidia2.h5
  # intensity multiplier = 8
  p = []
  # No rain mult 1 (shaw we have a "no rain" for every multiplier ?)
  sa = GetSteeringFromtcpflow('../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 no rain tcpflow.lo
g')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'no rain'])
  # light mult 1
  sa = GetSteeringFromtcpflow(
     '../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 light rain mult 8 tcpflow.log')
```

```
sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'light rain'])
  # heavy mult 1
  sa = GetSteeringFromtcpflow(
    '../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 heavy 10 mult 8 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'heavy rain slant +-10'])
  # torrential mult 1
  sa = GetSteeringFromtcpflow(
    '../../trained models/nvidia2/tcpflow/20201207192948 nvidia2 torrential 20 mult 8 tcpflow.log')
  sarr = np.asarray(sa)
  pa = sarr[:, 0]
  p.append([pa, 'torrential rain slant +-20'])
  plotMultipleSteeringAngles(p, 25, True, "Generated Track intensity multiplier 8", "20201207192948 nvidia2.h5"
                 'tcpflow log predicted')
          == " main ":
if name
  # 20201207091932 nvidia1.h5
  # mult 1
  # plot1()
  # mult 4
  # plot2()
  # mult 8
  # plot3()
  # 20201207192948_nvidia2.h5
  # mult 1
  # plot4()
  # mult 4
  # plot5()
  # mult 8
  # plot6()
# 17. SteeringAngleBins.ipynb.py
#!/usr/bin/env python
# coding: utf-8
# In[1]:
import fnmatch
import json
import seaborn as sns
import os
import numpy as np
import matplotlib.pyplot as plt
import statistics
```

```
def load json(filepath):
  Load a json file
  Inputs
     filepath: string, path to file
  Outputs:
     data: dictionary, json key, value pairs
  with open(filepath, "rt") as fp:
     data = json.load(fp)
  return data
def GetJSONSteeringAngles(filemask):
  Get steering angles stored as 'user/angle' attributes in .json files
  Inputs:
     filemask: string, path and mask
  Outputs
     svals: list, steering values
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # matches = sorted(matches, key=os.path.getmtime)
  # steering values
  svals = []
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split(" ")[0]
       json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
       jobj = load json(json filename)
       svals.append(jobj['user/angle'])
  return svals
# In[5]:
#als = GetJSONSteeringAngles('~/git/msc-data/unity/genRoad/*.jpg')
\#es = len(svals)
# Unity steering angle
#nityMaxSteering = 25#
#valscp = [element * UnityMaxSteering for element in svals]
# NB Plotted as normalized histogram
#ns.distplot(svalscp, bins=50, kde=False, norm hist=True,
#
         axlabel='Steering Angles (degrees) norm. hist. ' + str(values) + " data points \n \
        mean = 0.00 \text{ std} = 0.00"
#
# sns plot.savefig("output.png")
```

```
def jsonSteeringBins(filemask, pname="output", save=True, nc=25):
  Plot a steering values' histogram
  Inputs
     filemask: string, where to search for images, and corresponding ison files
     pname: string, output plot name
     save: boolean, save plot to disk
     nc: int, normalization constant, used in the simulator to put angles in range
     -1, 1. Default is 25.
  Outputs
     svals: list containing non-normalized steering angles
  svals = GetJSONSteeringAngles(filemask)
  svalscp = [element * nc for element in svals]
  values = len(svalscp)
  mean = ("%.2f" % statistics.mean(svals))
  std = ("%.2f" % statistics.stdev(svals))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalscp, bins=nc*2, kde=False, norm hist=True,
  axlabel= pname + 'steer. degs. norm. hist. ' + str(values) + 'values, mean = ' + mean + 'std = ' + std)
  #if(save):
  # sns.save("output.png")
  plt.savefig(pname + '.png')
  # return for downstream processing if required
  return svals
def listSteeringBins(svals, pname="output", save=True, nc=25):
  Plot a steering values' histogram
  Inputs
     svals: list, array of normalized steering values
     pname: string, output plot name
     save: boolean, save plot to disk
     nc: int, normalization constant, used in the simulator to put angles in range
     -1, 1. Default is 25.
  Outputs
     none
  svalscp = [element * nc for element in svals]
  values = len(svalscp)
  mean = ("%.2f" % statistics.mean(svals))
  std = ("%.2f" % statistics.stdev(svals))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalscp, bins=nc*2, kde=False, norm hist=True,
  axlabel= pname + 'steer. degs. norm. hist. ' + str(values) + 'values, mean = ' + mean + 'std = ' + std)
  #if(save):
  # sns.save("output.png")
  plt.savefig(pname + '.png')
  # return for downstream processing if required
  return svals
```

```
svals = jsonSteeringBins('~/git/msc-data/unity/genRoad/*.jpg', 'genRoad')
# In[]:
# In[52]:
def printTimeMS(nf, fr):
  Print time in minutes and seconds.
    nf: integer, number of frames
     fr: float, frame rate
  Outputs
    None
  ,,,,,,,
  total secs = nf / fr
  minutes = int(nf / fr / 60)
  seconds = round(nf / fr / 60 % minutes * 60)
  print("{}m{}s".format(minutes, seconds))
printTimeMS(45410, 24)
printTimeMS(280727, 24)
# In[54]:
# In[87]:
# simpler
import datetime
str(datetime.timedelta(seconds=280727/24))
# In[4]:
svals = GetJSONSteeringAngles('~/git/msc-data/unity/genRoad/*.jpg')
values = len(svals)
# Unity steering angle
UnityMaxSteering = 25
svalscp = [element * UnityMaxSteering for element in svals]
```

```
# NB Plotted as normalized histogram
sns.distplot(svalscp, bins=50, kde=False, norm hist=True,
        axlabel='Steering Angles (degrees) norm. hist. ' + str(values) + " data points")
# In[99]:
def printJSONSteeringAngles(filemask):
  Get steering angles stored as 'user/angle' attributes in .json files
  Inputs:
     filemask: string, path and mask
  Outputs
     svals: list, steering values
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # steering values
  svals = []
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split("_")[0]
       json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
       jobj = load json(json filename)
       svals.append(jobj['user/angle'])
  return svals
  # return svals
svals = printJSONSteeringAngles('~/git/sdsandbox/sdsim/log/*.jpg')
cnt = 0
sec = 0
for vals in svals:
  cnt += 1
  if(cnt \% 24 == 0):
     sec += 1
     print("Frame {} ========".format(sec))
  print(vals)
# In[116]:
UnityMaxSteering = 25
svalscp = [element * UnityMaxSteering for element in svals]
my iterator = filter(lambda svalscp: svalscp <= 20 and svalscp >= -20, svalscp)
svals2020 = list(my iterator)
rm = len(svals) - len(svals2020)
```

```
pctrm = rm * 100 / len(svals)
print("Removed entries outside -20 + 20 range: ", len(svals) - len(svals2020))
print("%.2f" % pctrm)
# In[121]:
# mean and std
import statistics
print("%.2f" % statistics.mean(svals2020))
print("%.2f" % statistics.stdev(svals2020))
# In[6]:
# SkewClenaup.png
# genRoad outlier cleanup
# NB Name is last one plotted
# plot everything
# used in SkewCleanup.png
import os
# plot all - nice effect but breaks kernel
# Now let's do one plot per folder to find out where these outliers are
path = '~/git/msc-data/unity/genRoad' # both work
path = '../../dataset/unity/genRoad/'
dirs = os.walk(path)
for mydir in dirs:
  if (len(mydir[1]) == 0):
    fn = mydir[0].split('/')
    path = mydir[0] + \frac{1}{*}.jpg'
    svals = jsonSteeringBins(path, fn[-1])
# deleted empties logs_Thu_Jul__9_16_12_28_2020
# both ~/git ... and ../.. work ok
#\.../../dataset/unity/genRoad/logs Thu Jul 9 12 25 38 2020/*.jpg', 'test')
# In[20]:
import os
# plot all - nice effect but breaks kernel
# Now let's do one plot per folder to find out where these outliers are
path = '~/git/msc-data/unity/genRoad' # both work
path = '../../dataset/unity/genRoad/'
dirs = os.walk(path)
for mydir in dirs:
```

```
if (\operatorname{len}(\operatorname{mydir}[1]) == 0):
    fn = mydir[0].split('/')
    path = mydir[0] + '/*.jpg'
    print("svals = jsonSteeringBins('{}', '{}')".format(path, fn[-1]))
    \# svals = jsonSteeringBins(path, fn[-1])
# In[34]:
# find skewer
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Fri Jul 10 09 29 13 2020/*.jpg', 'logs Fri Jul 10 09
29 13 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 15 47 22 2020/*.jpg', 'logs Thu Jul 9 1
5 47 22 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 16 06 19 2020/*.jpg', 'logs Thu Jul 9 1
6 06 19 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 15 53 22 2020/*.jpg', 'logs Thu Jul 9 1
5 53 22 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Fri Jul 10 09 32 12 2020/*.jpg', 'logs Fri Jul 10 09
32 12 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 12 57 15 2020/*.jpg', 'logs Thu Jul 9 1
2 57 15 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 13 03 17 2020/*.jpg', 'logs Thu Jul 9 1
3 03 17 2020')
# svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 12 33 37 2020/*.jpg', 'logs Thu Jul 9
12 33 37 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 14 59 57 2020/*.jpg', 'logs Thu Jul 9 1
4 59 57 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Fri Jul 10 09 22 57 2020/*.jpg', 'logs Fri Jul 10 09
22 57 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 16 08 45 2020/*.jpg', 'logs Thu Jul 9 1
6 08 45 2020')
svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 16 00 15 2020/*.jpg', 'logs Thu Jul 9 16
00 15 2020')
#svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020/*.jpg', 'logs Fri Jul 10 09
16 18 2020')
# svals = jsonSteeringBins('../../dataset/unity/genRoad/logs Thu Jul 9 12 25 38 2020/*.jpg', 'logs Thu Jul 9
12 25 38 2020')
# In[35]:
# moved skewing logs
# mv genRoad/logs Thu Jul 9 16 00 15 2020 quarantine/
svals = jsonSteeringBins('.../../dataset/unity/genRoad/*.jpg', 'logs Fri Jul 10 09 29 13 2020')
# In[]:
```

```
# 18. steerlib.ipynb.py
#!/usr/bin/env python
# coding: utf-8
# In[10]:
# steerlib - Helper library to create videos and plots
# modules
import fnmatch
import json
import seaborn as sns
import os
import numpy as np
import argparse
import matplotlib.pyplot as plt
import statistics
import seaborn as sns
import pickle
# In[15]:
def load json(filepath):
  Load a json file
  Inputs
    filepath: string, path to file
  Outputs
    data: dictionary, json key, value pairs
  Example
  path = "~/git/msc-data/unity/roboRacingLeague/log/logs_Sat_Nov_14_12_36_16_2020/record_11640.json"
  js = load json(path)
  with open(filepath, "rt") as fp:
    data = json.load(fp)
  return data
# In[16]:
def GetJSONSteeringAngles(filemask):
  Get steering angles stored as 'user/angle' attributes in .json files
```

Inputs:

```
filemask: string, path and mask
  Outputs
     svals: list, steering values
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # steering values
  svals = []
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split(" ")[0]
       json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
       jobj = load json(json filename)
       svals.append(jobj['user/angle'])
  return svals
# In[64]:
def jsonSteeringBins(filemask, pname="output", save=True, nc=25, rmout=0):
  Plot a steering values' histogram
  Inputs
     filemask: string, where to search for images, and corresponding ison files
     pname: string, output plot name
     save: boolean, save plot to disk
     nc: int, normalization constant, used in the simulator to put angles in range
     -1, 1. Default is 25.
     rmout: integer, outlier range to remove
  Outputs
     svals: list containing non-normalized steering angles
  Example:
  # svals = jsonSteeringBins('~/git/msc-data/unity/genRoad/*.jpg', 'genRoad', save=True, nc=25, rmout=20)
  svals = GetJSONSteeringAngles(filemask)
  values = len(svals)
  svalscp = [element * nc for element in svals]
  if(rmout>0):
     my iterator = filter(lambda svalscp: svalscp <= rmout and svalscp >= (-1 * rmout), svalscp)
     svalsrmout = list(my iterator)
     svalscp = svalsrmout
     values = len(svalsrmout)
     print("Removed {} records".format(len(svals) - len(svalsrmout)))
     svals = svalsrmout
  mean = ("%.2f" % statistics.mean(svalscp))
  std = ("%.2f" % statistics.stdev(svalscp))
  plt.title=(pname)
  # NB Plotted as normalized histogram
```

```
sns.distplot(svalsep, bins=nc*2, kde=False, norm hist=True,
  axlabel= pname + 'st. degs. norm. hist. '+ str(values) + 'values, mean = '+ mean + 'std = '+ std)
  #if(save):
  # sns.save("output.png")
  if(save):
     plt.savefig(pname + '.png')
  plt.show()
  # return for downstream processing if required
  return svals
# In[63]:
def removeOutliers(svals, rmout, nc):
  Remove outliers from a list
  Inputs
     svals: double, steering values
     rmout: integer, -+ range tp remove
     nc: steering normalization constant - same as used in simulator (max steering)
  Output
     svals: list, list with values excluded
  svalscp = [element * nc for element in svals]
  my iterator = filter(lambda svalscp: svalscp <= rmout and svalscp >= (-1 * rmout), svalscp)
  svalsrmout = list(my iterator)
  svalscp = svalsrmout
  values = len(svalsrmout)
  print("Removed {} records".format(len(svalscp) - len(svalsrmout)))
  svals = svalsrmout
  return svals
# In[65]:
def listSteeringBins(svals, pname="output", save=True, nc=25, rmout=0):
  Plot a steering values' histogram
  Inputs
     svals: list, array of normalized steering values
     pname: string, output plot name
     save: boolean, save plot to disk
     nc: int, normalization constant, used in the simulator to put angles in range
     -1, 1. Default is 25.
     rmout: integer, outlier range to remove
  Outputs
     none
  svalscp = [element * nc for element in svals]
  values = len(svals)
  # remove outliers
```

```
if(rmout>0):
    #my iterator = filter(lambda svalscp: svalscp <= rmout and svalscp >= (-1 * rmout), svalscp)
    #svalsrmout = list(my iterator)
    #svalscp = svalsrmout
    #values = len(svalsrmout)
    #print("Removed {} records".format(len(svals) - len(svalsrmout)))
    #svals = svalsrmout
    svals = removeOutliers(svalscp, rmout, nc)
    values = len(svals)
  mean = ("%.2f" % statistics.mean(svalscp))
  std = ("%.2f" % statistics.stdev(svalscp))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalscp, bins=nc*2, kde=False, norm hist=True,
  axlabel= pname + 'steer. degs. norm. hist. '+ str(values) + 'values, mean = '+ mean + 'std = '+ std)
  #if(save):
  # sns.save("output.png")
  if(save):
    plt.savefig(pname + '.png')
  plt.show()
# In[66]:
filemask = '~/git/msc-data/unity/genRoad/*.jpg'
svals = GetJSONSteeringAngles(filemask)
listSteeringBins(svals, pname="outputExOut20", save=True, nc=25, rmout=20)
# In[58]:
filemask = '~/git/msc-data/unity/genRoad/*.jpg'
svals = GetJSONSteeringAngles(filemask)
listSteeringBins(svals, pname="output", save=True, nc=25, rmout=0)
# In[68]:
# generate training dataset latex tables for report
for folder in ['genRoad', 'log sample', 'roboRacingLeague', 'smallLoop', 'smallLoopingCourse', 'warehouse']:
  print(folder)
SDSandbox unity/smallLoopingCourse/log/* 34443 from small\ looping\ course
SDSandbox unity/warehouse/* 41126 From Warehouse course
SDSandbox unity/smallLoop/* 45422 From small\ looping\ course
SDSandbox unity/roboRacingLeague/* 12778 From "Robot Racing League" course
SDSandbox unity/log\ sample 25791 From small\ looping\ course
SDSandbox unity/genRoad 280727 From "Generated Road" course
*****
```

```
# In[]:
\begin{table}[]
\begin{center}
\begin{tabular} {||1||1||1|}
\hline
\multicolumn{4}{|c|}{Deliverables - Datasets} \\ \hline
get_ipython().run_line_magic('ID', 'Task Deliverable Description')
get ipython().run line magic('1', 'Download D1 Udacity real world dataset')
get ipython().run line magic('2', 'Generate D2 Unity3D simulator data')
get ipython().run line magic('3', 'Combine D3 Udacity real and simulator data')
get ipython().run line magic('4', 'Mechanical Turk dry/rainy Ford dataset')
ID & Task & Deliverable & Description \\ \hline\hline
1 & Download & D1 & Udacity real world dataset \\\hline
2 & Generate & D2 & Udacity simulator data \\\hline
3 & Combine & D3 & Udacity real and simulator data \\\hline
4 & Gather & D4 & Mechanical Turk dry/rainy Ford dataset \\ \hline
\end{tabular}
\end{center}
\caption{Datasets used to train models}
\label{Deliverables-Datasets}
\end{table}
# 19. steerlib.py
# Helper library to create videos and plots
import fnmatch
import ison
import os
import numpy as np
import matplotlib.pyplot as plt
import statistics
import seaborn as sns
import pickle
from PIL import Image
# prediction
import tensorflow as tf
from tensorflow.python import keras
from tensorflow.python.keras.models import load model
# rain
from predict client import add rain
# Augmentation library
```

import Augmentation

```
def load json(filepath):
  Load a json file
  Inputs
     filepath: string, path to file
  Outputs
     data: dictionary, json key, value pairs
  Example
  path = "~/git/msc-data/unity/roboRacingLeague/log/logs Sat Nov 14 12 36 16 2020/record 11640.json"
  js = load json(path)
  with open(filepath, "rt") as fp:
     data = ison.load(fp)
  return data
def GetSteeringFromtcpflow(filename):
  Get a tcpflow log and extract steering values obtained from network communication between.
  Note, we only plot the predicted steering angle jsondict['steering']
  and the value of jsondict['steering angle'] is ignored. Assumed to be the steering angle
  calculated by PID given the current course.
  sim and prediction engine (predict client.py)
  Inputs
     filename: string, name of tcpflow log
  Returns
     sa: list of arrays, steering angle predicton and actual value tuple.
  Example
  ** ** **
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
     \# readline = f.read()
     lines = file.splitlines()
     for line in lines:
       # print(line)
       start = line.find('{')
       if (start == -1):
          continue
       jsonstr = line[start:]
       # print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in jsondict:
          # predicted
          pred = jsondict['steering']
          # jsondict['steering angle']
          # sa.append([float(pred), act])
          sa.append([float(pred), float(pred)]) # append twice to keep code from breaking
```

```
# if "steering angle" in jsondict:
       # actual
       # act = jsondict['steering angle']
       # save pair, only keep last pred in case two were send as it does happen i.e.:
       # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "thr
ottle": "0.08249988406896591", "brake": "0.0"}
       # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "thr
ottle": "0.08631626516580582", "brake": "0.0"}
       # 127.000.000.001.09091-127.000.000.001.59460: {"msg_type":"telemetry", "steering_angle":-0.07196037,(...
..)
       # if(len(pred) > 0):
            sa.append([float(pred), act])
            pred = " # need to save this image
       # deal with image later, sort out plot first
       # imgString = jsondict["image"]
       # image = Image.open(BytesIO(base64.b64decode(imgString)))
       # img arr = np.asarray(image, dtype=np.float32)
  except Exception as e:
     print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
  return sa
def GetJSONSteeringAngles(filemask):
  Get steering angles stored as 'user/angle' attributes in .json files
  Inputs:
     filemask: string, path and mask
  Outputs
     svals: list, steering values
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # sort by create date
  matches = sorted(matches, key=os.path.getmtime)
  # steering values
  svals = []
  for fullpath in matches:
       frame number = os.path.basename(fullpath).split(" ")[0]
       json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
       jobj = load json(json filename)
       svals.append(jobj['user/angle'])
  return svals
def jsonSteeringBins(filemask, pname="output", save=True, nc=25):
  Plot a steering values' histogram
  Inputs
     filemask: string, where to search for images, and corresponding .json files
```

```
pname: string, output plot name
     save: boolean, save plot to disk
     nc: int, normalization constant, used in the simulator to put angles in range
     -1, 1. Default is 25.
  Outputs
     svals: list containing non-normalized steering angles
  Example:
  # svals = jsonSteeringBins('~/git/msc-data/unity/genRoad/*.jpg', 'genRoad')
  svals = GetJSONSteeringAngles(filemask)
  values = len(svals)
  svalscp = [element * nc for element in svals]
  mean = ("%.2f" % statistics.mean(svals))
  std = ("%.2f" % statistics.stdev(svals))
  plt.title=(pname)
  # NB Plotted as normalized histogram
  sns.distplot(svalsep, bins=nc*2, kde=False, norm hist=True,
  axlabel= pname + 'steer. degs. norm. hist. ' + str(values) + 'values, mean = ' + mean + 'std = ' + std)
  #if(save):
  # sns.save("output.png")
  plt.savefig(pname + '.png')
  # return for downstream processing if required
  return svals
def plot hist(path):
  Create loss/accuracy plot for training and save to disk
  Inputs
     path: file to history pickle file
  Outputs
     none
  Example:
  path = "/home/simbox/git/sdsandbox/trained models/nvidia1/20201107144927 nvidia1.history"
  plot hist(path)
  or
  $ python steerlib.py
  $ python
  $>>> import steerlib
  $>>> path = "/home/simbox/git/sdsandbox/trained models/nvidia1/20201107144927_nvidia1.history"
  $>>> plot hist(path)
  history = pickle.load(open(path, "rb"))
  # type(history)
  # <class 'dict'>
  # history.keys()
  # dict keys(['loss', 'acc', 'val loss', 'val acc'])
  do plot = True
  try:
     if do plot:
       fig = plt.figure() # when loading dictionary keys, we omit .history (see train.py)
       plot name = path.split('/')[-1]
       sp = plot name \setminus
          + ' - ' + '(1,v1,a,va)' + '\{0:.3f\}'.format(history['loss'][-1]) \setminus
```

```
+','+'\{0:.3f\}'.format(history['val loss'][-1])\
          +','+'{0:.3f}'.format(history['acc'][-1]) \
          +','+'\{0:.3f\}'.format(history['val acc'][-1])
       fig.suptitle(sp, fontsize=9)
       ax = fig.add subplot(111)
       ax.plot(history['loss'], 'r-', label='Training Loss', )
       ax.plot(history['val loss'], 'm-', label='Validation Loss')
       ax2 = ax.twinx()
       ax2.plot(history['acc'], '-', label='Training Accuracy')
       ax2.plot(history['val acc'], '-', label='Validation Accuracy')
       ax.legend(loc=2) # https://matplotlib.org/3.1.1/api/ as gen/matplotlib.pyplot.legend.html
       ax.grid()
       ax.set xlabel("Epoch")
       ax.set ylabel(r"Loss")
       ax2.set ylabel(r"Accurary")
       ax.set ylim(0, 0.2)
       ax2.set vlim(0.5, 1)
       ax2.legend(loc=1)
       aimg = plot name.split('.')[0]
       aimg = str(aimg) + '_accuracy.png'
       plt.savefig(aimg)
  except Exception as e:
     print("Failed to save accuracy/loss graph: " + str(e))
def gos(p, g, n):
  Calculate the goodness-of-steer between a prediction and a ground truth array.
     p: array of floats, steering angle prediction
     g: array of floats, steering angle ground truth.
     n: float, normalization constant
  Output
     gos: float, average of absolute difference between ground truth and prediction arrays
  # todo add type assertion
  assert len(p) == len(g), "Arrays must be of equal length"
  return sum(abs(p - g)) / len(p) * n
  # print("Goodness of steer: {:.2f}".format(steer))
def plotSteeringAngles(p, g=None, n=1, save=False, track= "Track Name", mname="model name", title='title'):
  Plot predicted and (TODO) optionally ground truth steering angles
  Inputs
     p, g: prediction and ground truth float arrays
     n: float, steering normalization constant
     save: boolean, save plot flag
     track, mname, title: string, track (data), trained model and title strings for plot
  Outputs
     plt: pyplot, plot
  Example
  # set argument variables (see data predict.py)
  plotSteeringAngles(p, g, nc, True, datapath[-2], modelpath[-1], 'Gs' + gss)
```

```
plt.rcParams["figure.figsize"] = (18,3)
  plt.plot(p*n, label="predicted")
  try:
     if (g is not None):
       plt.plot(g*n, label="ground truth")
  except Exception as e:
     print("problems plotting: " + str(e))
  plt.ylabel('Steering angle')
  plt.xlabel('Frame number')
  # Set a title of the current axes.
  # plt.title('tcpflow log predicted steering angles: track ' + track + ' model ' + mname)
  plt.title(title + ' Steering angles: track ' + track + ', model ' + mname)
  # show a legend on the plot
  plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
  # set limit
  plt.xlim([-5,len(p)+5])
  plt.gca().invert yaxis()
  # plt.show()
  if(save==True):
     plt.savefig('sa '+ track + ' '+ mname + '.png')
  # if need be
  return plt
def plotMultipleSteeringAngles(p, n=25, save=False, track="Track Name", mname="model name", title='title', w=1
8, h=3):
  ,,,,,,,
  Plot multiple predicted and (TODO) optionally ground truth steering angles
  Inputs
     p: list of tuples, prediction and labels
     n: float, steering normalization constant
     save: boolean, save plot flag
     track, mname, title: string, track (data), trained model and title strings for plot
     w: integer, plot width
     h: integer, plot height
  Outputs
     plt: pyplot, plot
  Example
  # get some steering angles
  sa = GetSteeringFromtcpflow('../trained models/nvidia1/tcpflow/20201207091932 nvidia1 no rain tcpflow.log'
  sarr = np.asarray(sa)
  pa = sarr[:,0]
  p.append([pa, 'no rain'])
  plotSteeringAngles(p, g, nc, True, datapath[-2], modelpath[-1], 'Gs' + gss)
  import matplotlib.pyplot as plt # local copy
  plt.rcParams["figure.figsize"] = (w,h)
```

```
for i in range (0, len(p)):
     plt.plot(p[i][0]*n, label=p[i][1])
  #try:
    # if (g is not None):
         plt.plot(g*n, label="ground truth")
  #except Exception as e:
      print("problems plotting: " + str(e))
  plt.ylabel('Steering angle')
  plt.xlabel('Frame number')
  # Set a title of the current axes.
  # plt.title('tcpflow log predicted steering angles: track ' + track + ' model ' + mname)
  plt.title(title + ' Steering angles: track ' + track + ', model ' + mname)
  # show a legend on the plot
  plt.legend()
  # Display a figure.
  # horizontal grid only
  plt.grid(axis='y')
  # set limit
  plt.xlim([-5,len(p[0][0])+5])
  plt.gca().invert yaxis()
  # plt.show()
  if(save==True):
     plt.savefig('sa_' + track + '_' + mname + '.png')
  # if need be
  return plt
def getSteeringFromtcpflow(filename):
  Get a tcpflow log and extract steering values obtained from network communication between sim and predict clie
nt.py.
  Note, we only plot the predicted steering angle jsondict['steering']
  and the value of jsondict['steering angle'] is ignored. Assumed to be the steering angle
  calculated by PID given the current course.
  sim and prediction engine (predict client.pv)
  Inputs
     filename: string, name of tcpflow log
  Returns
     sa: list of arrays, steering angle predicton and actual value tuple.
  Example
  ** ** **
  # open file
  sa = []
  # initialize prediction
  pred = "
  f = open(filename, "r")
  file = f.read()
  try:
     \# readline = f.read()
     lines = file.splitlines()
     for line in lines:
        # print(line)
```

```
start = line.find('{')
       if (start == -1):
          continue
       jsonstr = line[start:]
       # print(jsonstr)
       jsondict = json.loads(jsonstr)
       if "steering" in jsondict:
          # predicted
          pred = jsondict['steering']
          # jsondict['steering angle']
          # sa.append([float(pred), act])
          sa.append([float(pred), float(pred)]) # append twice to keep code from breaking
       # if "steering angle" in jsondict:
       # actual
       # act = jsondict['steering angle']
       # save pair, only keep last pred in case two were send as it does happen i.e.:
       # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.071960375", "thr
ottle": "0.08249988406896591", "brake": "0.0"}
       # 127.000.000.001.59460-127.000.000.001.09091: {"msg_type": "control", "steering": "-0.079734944", "thr
ottle": "0.08631626516580582", "brake": "0.0"}
       # 127.000.000.001.09091-127.000.000.001.59460: \{\text{"msg type":"telemetry",\text{"steering angle":-0.07196037,\(\).
..)
       \# if(len(pred) > 0):
            sa.append([float(pred), act])
            pred = " # need to save this image
       # deal with image later, sort out plot first
       # imgString = jsondict["image"]
       # image = Image.open(BytesIO(base64.b64decode(imgString)))
       # img arr = np.asarray(image, dtype=np.float32)
  except Exception as e:
     print("Exception raise: " + str(e))
  # file should be automatically closed but will close for good measure
  f.close()
  return sa
# get image
# get steering
# add / don't add rain
outputs = self.model.predict(image array)
Append groundtruth, predicted to list.
def PrintLatexRowModelGOS(filemask, modelpath, modelname, rt=", st=0):
  Generate a "goodness of fit value" for a model on a given track
  Inputs
     filemask: string, path and mask
     modelpath: string, path to keras model
     modelname: string, canonical model name e.g. nvidia1, nvidia2, nvidia baseline
```

```
rt: string, rain type e.g. drizzle/light, heavy torrential
  st: integer, -+20 degree rain slant
Outputs
  svals: list, ground truth steering values and predictions
# load augmentation library for correct model geometry
ag = Augmentation.Augmentation(modelname)
# load model
print("loading model", modelpath)
model = load model(modelpath)
# In this mode, looks like we have to compile it
model.compile("sgd", "mse")
filemask = os.path.expanduser(filemask)
path, mask = os.path.split(filemask)
matches = []
for root, dirnames, filenames in os.walk(path):
  for filename in fnmatch.filter(filenames, mask):
     matches.append(os.path.join(root, filename))
# steering values
svals = []
for fullpath in matches:
  frame number = os.path.basename(fullpath).split("_")[0]
  json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
  jobj = load json(json filename)
  # steering ground truth
  steer gt = jobj['user/angle']
  # open the image
  img arr = Image.open(fullpath)
  # Convert PIL Image to numpy array
  img arr = np.array(img arr, dtype=np.float32)
  # add rain if need be
  if rt != ":
     img arr = add rain(img arr, rt, st)
  # apply same preprocessing
  # same preprocessing as for training
  img arr = ag.preprocess(img arr)
  # put in correct format
  img arr = img arr.reshape((1,) + img arr.shape)
  # generate prediction
  outputs = model.predict(img arr)
  # store predictions
  steer pred = outputs[0][0]
  # store ground truth and prediction
  svals.append([steer pred, steer gt])
# get goodness of fit
sarr = np.asarray(svals)
p = sarr[:, 0]
g = sarr[:, 1]
nc = 25 # unity maximum steering angle / normalization constant - should hold in conf.py and managed with Aug
```

```
mentation
  mygos = gos(p, g, nc)
  # format to human readable/friendlier 2 decimal places
  gos str = "{:.2f}".format(round(mygos, 2))
  # strip path from modelpath
  modelfile = modelpath.split('/')
  modelfile = modelfile[-1]
  # print latex formated data
  # header
  hd str = 'Filename & Model & Rain Type & Slant & gos \\\\\ \hline'
  # log file
  print('Log: ', path, '\\\\ \hline')
  print(hd str)
  # results
  res str = f'\{modelfile\} \& \{modelname\} \& \{rt\} \& \{gos str\} \ \\\\\ \hline'
  print(res str)
def GetPredictedSteeringAngles(filemask, model, modelname, rt=", st=0):
  Generate a "goodness of fit value" for a model on a given track
  Inputs
     filemask: string, path and mask
     modelpath: string, path to keras model
     modelname: string, canonical model name e.g. nvidia1, nvidia2, nvidia baseline
     rt: string, rain type e.g. drizzle/light, heavy torrential
     st: integer, -+20 degree rain slant
  Outputs
     svals: list, ground truth steering values and predictions
  # load augmentation library for correct model geometry
  ag = Augmentation.Augmentation(modelname)
  # load model
  # print("loading model", modelpath)
  # assume model is loaded and compiled
  # model = load model(modelpath)
  # In this mode, looks like we have to compile it
  # model.compile("sgd", "mse")
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  # sort by create date
  matches = sorted(matches, key=os.path.getmtime)
  # steering values
  svals = []
  for fullpath in matches:
     frame number = os.path.basename(fullpath).split(" ")[0]
     json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
```

```
jobj = load json(json filename)
    # steering ground truth
    steer gt = jobj['user/angle']
    # open the image
    img arr = Image.open(fullpath)
    # Convert PIL Image to numpy array
    img arr = np.array(img arr, dtype=np.float32)
    # add rain if need be
    if rt != ":
      img arr = add rain(img arr, rt, st)
    # apply same preprocessing
    # same preprocessing as for training
    img arr = ag.preprocess(img arr)
    # put in correct format
    img arr = img arr.reshape((1,) + img arr.shape)
    # generate prediction
    outputs = model.predict(img arr)
    # store predictions
    steer pred = outputs[0][0]
    # store ground truth and prediction
    svals.append([steer pred, steer gt])
  return svals
def printGOSRows():
  Print GOS rows for results report.
  We are comparing mainly the two best models for nvidia1 and nvidia2, would have been nice to also test
  the driveable nvidia baseline, and sanity models
  # models
  # nvidia2 - ../../trained models/nvidia2/20201207192948 nvidia2.h5
  # nvidia1 - ../../trained models/nvidia1/20201207091932 nvidia1.h5
  # 20201120124421\ nvidia\ baseline.h5
  # log logs Wed Nov 25 23 39 22 2020
  #1. nvidia2 Generated track
  # log = '.././dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020/*.jpg'
  #modelpath = '../../trained models/nvidia2/20201207192948 nvidia2.h5'
  #modelname = 'nvidia2'
  #rt = 'torrential'
  \#st = 20
  #PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
  # Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & & 0 & 1.68 \\ \hline
  #Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & light & 0 & 2.12 \\ \hline
  # Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & heavy & 10 & 2.17 \\ \hline
  # Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
```

```
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207192948 nvidia2.h5 & nvidia2 & torrential & 20 & 2.30 \\ \hline
# 2. nvidia1 Generated track
# modelpath = '../../trained models/nvidia1/20201207091932 nvidia1.h5'
#modelname = 'nvidia1'
#rt = 'torrential'
\#st = 20
#PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207091932 nvidia1.h5 & nvidia1 & & 0 & 1.82 \\ \hline
#Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207091932 nvidia1.h5 & nvidia1 & light & 0 & 2.11 \\ \hline
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207091932 nvidia1.h5 & nvidia1 & heavy & 10 & 2.13 \\ \hline
#Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207091932 nvidia1.h5 & nvidia1 & torrential & 20 & 2.28 \\ \hline
# 3. nvidia baseline Generated track
#modelpath = '../../trained models/nvidia baseline/20201207201157 nvidia baseline.h5'
#modelname = 'nvidia2 baseline'
#rt = 'torrential'
\#st = 20
#PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207201157 nvidia baseline.h5 & nvidia2 baseline & & 0 & 2.32 \\ \hline
#Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207201157 nvidia baseline.h5 & nvidia2 baseline & light & 0 & 3.12 \\ \hline
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207201157 nvidia baseline.h5 & nvidia2 baseline & heavy & 10 & 3.17 \\ \hline
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed_Nov_25_23_39_22_2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201207201157 nvidia baseline.h5 & nvidia2 baseline & torrential & 20 & 3.39 \\ \hline
# 4. sanity Generated track 20201120171015\_ sanity.h5
#modelpath = '../../trained models/sanity/20201120171015 sanity.h5'
#modelname = 'nvidia1'
#rt = 'torrential'
\#st = 20
#PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
# Filename & Model & Rain Type & Slant & gos \\ \hline
# 20201120171015 sanity.h5 & nvidia1 & & 0 & 5.03 \\ \hline
# Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
```

```
# Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & light & 0 & 3.11 \\ \hline
  # Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & heavy & 10 & 3.07 \\ \hline
  #Log: ../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & torrential & 20 & 3.00 \\ \hline
  #5. nvidia2 Generated Road
  log = '../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020/*.jpg'
  #modelpath = '../../trained models/nvidia2/20201207192948 nvidia2.h5'
  #modelname = 'nvidia2'
  #rt = 'torrential'
  \#st = 20
  #PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & & 0 & 2.99 \\ \hline # drove 16 minutes on this road https://youtu.be
/z9nILq9dQfI
  #Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & light & 0 & 3.20 \\ \hline
  #Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & heavy & 10 & 3.22 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207192948 nvidia2.h5 & nvidia2 & torrential & 20 & 3.27 \\ \hline
  # 6. nvidia1 Generated Road
  #modelpath = '../../trained models/nvidia1/20201207091932 nvidia1.h5'
  #modelname = 'nvidia1'
  # rt = 'torrential'
  \# st = 20
  # PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207091932 nvidia1.h5 & nvidia1 & & 0 & 3.87 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207091932 nvidia1.h5 & nvidia1 & light & 0 & 3.75 \\ \hline
  #Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207091932 nvidia1.h5 & nvidia1 & heavy & 10 & 3.70 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207091932 nvidia1.h5 & nvidia1 & torrential & 20 & 3.57 \\ \hline
```

```
#modelpath = '../../trained models/nvidia baseline/20201207201157 nvidia baseline.h5'
  #modelname = 'nvidia2 baseline'
  #rt = 'torrential'
  \#st = 20
  #PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207201157 nvidia baseline.h5 & nvidia2 baseline & & 0 & 5.51 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207201157 nvidia baseline.h5 & nvidia2 baseline & light & 0 & 4.97 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207201157 nvidia baseline.h5 & nvidia2 baseline & heavy & 10 & 4.98 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201207201157 nvidia baseline.h5 & nvidia2 baseline & torrential & 20 & 5.05 \\ \hline
  # 8. sanity Generated track 20201120171015\ sanity.h5
  modelpath = '../../trained models/sanity/20201120171015 sanity.h5'
  modelname = 'nvidia1'
  rt = 'torrential'
  st = 20
  PrintLatexRowModelGOS(log, modelpath, modelname, rt, st)
  #Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & & 0 & 3.85 \\ \hline
  #Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & light & 0 & 3.06 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & heavy & 10 & 3.05 \\ \hline
  # Log: ../../dataset/unity/genRoad/logs Fri Jul 10 09 16 18 2020 \\ \hline
  # Filename & Model & Rain Type & Slant & gos \\ \hline
  # 20201120171015 sanity.h5 & nvidia1 & torrential & 20 & 3.02 \\ \hline
def printMultiPlots(model1 nvidia2, model2 nvidia1, model3 nvidia baseline, model4 sanity):
  Print multiple plots to reference first 16 results in Goodness of steer tables
  model1 nvidia2, model2 nvidia1, model3 nvidia baseline, model4 sanity: the required keras models
  # init plot list
  plot list = []
  # define log
  log = '../../dataset/unity/genTrack/genTrackOneLap/logs Wed Nov 25 23 39 22 2020/*.jpg'
  # Get ground truth values
  gt = GetJSONSteeringAngles(log)
  gt = np.asarray(gt)
  # get predictions
```

```
# 1. nvidia2 Generated track
  # 1.1 dry
  print('Predicting nvidia2 dry...')
  sa = GetPredictedSteeringAngles(log, model1 nvidia2, 'nvidia2', rt=", st=0)
  sarr = np.asarray(sa)
  p = sarr[:,0]
  g = sarr[:, 1]
  plot list.append([g, 'ground truth'])
  plot list.append([p, 'prediction - no rain'])
  # 1.2 light rain
  print('Predicting nvidia2 light rain...')
  sa = GetPredictedSteeringAngles(log, model1 nvidia2, 'nvidia2', rt='light', st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - light rain'])
  # 1.3 heavy rain, slant = +-10
  print('Predicting nvidia2 heavy rain...')
  sa = GetPredictedSteeringAngles(log, model1 nvidia2, 'nvidia2', rt='heavy', st=10)
  p = sarr[:,0]
  plot list.append([p, 'prediction - heavy rain +-10'])
  # 1.4 torrential rain, slant = +-20
  print('Predicting nvidia2 torrential rain...')
  sa = GetPredictedSteeringAngles(log, model1 nvidia2, 'nvidia2', rt='torrential', st=20)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - torrential rain +-20'])
  print('Plotting...')
  plotMultipleSteeringAngles(plot list, 25, True, "Generated Track", "20201207192948 nvidia2.h5", title='SDSan
dbox log genTrackOneLap/logs Wed Nov 25 23 39 22 2020', w=18, h=4)
  # 2. nvidia1 Generated track
  # 2.1 dry
  print('Predicting nvidia1 dry...')
  sa = GetPredictedSteeringAngles(log, model2 nvidia1, 'nvidia1', rt=", st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  g = sarr[:, 1]
  plot list.append([g, 'ground truth'])
  plot list.append([p, 'prediction - no rain'])
  #2.2 light rain
  print('Predicting nvidia1 light rain...')
  sa = GetPredictedSteeringAngles(log, model2 nvidia1, 'nvidia1', rt='light', st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - light rain'])
  # 2.3 heavy rain, slant = +-10
  print('Predicting nvidia1 heavy rain...')
  sa = GetPredictedSteeringAngles(log, model2 nvidia1, 'nvidia1', rt='heavy', st=10)
  sarr = np.asarray(sa);
  p = sarr[:,0]
```

```
plot list.append([p, 'prediction - heavy rain +-10'])
  # 2.4 torrential rain, slant = +-20
  print('Predicting nvidia1 torrential rain...')
  sa = GetPredictedSteeringAngles(log, model2 nvidia1, 'nvidia1', rt='torrential', st=20)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - torrential rain +-20'])
  print('Plotting...')
  plotMultipleSteeringAngles(plot list, 25, True, "Generated Track", "20201207091932 nvidia1.h5", title='SDSan
dbox log genTrackOneLap/logs Wed Nov 25 23 39 22 2020', w=18, h=4)
  # 3. nvidia baseline Generated track
  # 3.1 dry
  print('Predicting model3 nvidia baseline dry...')
  sa = GetPredictedSteeringAngles(log, model3 nvidia baseline, 'nvidia1', rt=", st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  g = sarr[:, 1]
  plot list.append([g, 'ground truth'])
  plot list.append([p, 'prediction - no rain'])
  #3.2 light rain
  print('Predicting model3 nvidia baseline light rain...')
  sa = GetPredictedSteeringAngles(log, model3 nvidia baseline, 'nvidia1', rt='light', st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - light rain'])
  # 3.3 heavy rain, slant = +-10
  print('Predicting model3 nvidia baseline heavy rain...')
  sa = GetPredictedSteeringAngles(log, model3 nvidia baseline, 'nvidia1', rt='heavy', st=10)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - heavy rain +-10'])
  # 3.4 torrential rain, slant = +-20
  print('Predicting model3 nvidia baseline torrential rain...')
  sa = GetPredictedSteeringAngles(log, model3 nvidia baseline, 'nvidia1', rt='torrential', st=20)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - torrential rain +-20'])
  print('Plotting...')
  plotMultipleSteeringAngles(plot list, 25, True, "Generated Track", "20201207201157 nvidia baseline.h5", title
='SDSandbox log genTrackOneLap/logs Wed Nov 25 23 39 22 2020', w=18, h=3)
  *****
  # 4. sanity Generated track
  # 4.1 dry
  print('Predicting model4 sanity dry...')
  sa = GetPredictedSteeringAngles(log, model4 sanity, 'nvidia1', rt=", st=0)
  sarr = np.asarray(sa);
```

```
p = sarr[:,0]
  g = sarr[:, 1]
  plot list.append([g, 'ground truth'])
  plot list.append([p, 'prediction - no rain'])
  # 4.2 light rain
  print('Predicting model4 sanity light rain...')
  sa = GetPredictedSteeringAngles(log, model4 sanity, 'nvidia1', rt='light', st=0)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - light rain'])
  # 4.3 heavy rain, slant = +-10
  print('Predicting model4 sanity heavy rain...')
  sa = GetPredictedSteeringAngles(log, model4 sanity, 'nvidia1', rt='heavy', st=10)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - heavy rain +-10'])
  # 4.4 torrential rain, slant = +-20
  print('Predicting model4 sanity torrential rain...')
  sa = GetPredictedSteeringAngles(log, model4 sanity, 'nvidia1', rt='torrentia1', st=20)
  sarr = np.asarray(sa);
  p = sarr[:,0]
  plot list.append([p, 'prediction - torrential rain +-20'])
  print('Plotting...')
  plotMultipleSteeringAngles(plot list, 25, True, "Generated Track", "20201120171015 sanity.h5", title='SDSand
box log genTrackOneLap/logs Wed Nov 25 23 39 22 2020', w=18, h=3)
if name == " main ":
  # plot hist("/home/simbox/git/sdsandbox/trained models/nvidia1/20201107144927 nvidia1.history")
\#if name == " main ":
  parser = argparse.ArgumentParser(description='Plot Steering Utils')
   parser.add argument('--inputs', type=str, help='file path')
  # args = parser.parse args()
  #svals = jsonSteeringBins('~/git/msc-data/unity/genRoad/*.jpg', 'genRoad')
  # PrintLatexRowModelGOS('../../dataset/unity/genTrack/genTrackOneLap/logs_Wed_Nov_25_23_39_22_2020/
*.jpg', '../../trained models/nvidia2/20201207192948 nvidia2.h5', 'nvidia2')
  # printGOSRows()
  # load models
  model1 nvidia2 = "
  model2 nvidia1 = "
  model3 nvidia baseline = "
  model4 sanity = "
  # modelpath = '../../trained models/sanity/20201120171015 sanity.h5'
  # modelpath = '../../trained models/nvidia baseline/20201207201157 nvidia baseline.h5'
  #modelpath = '../../trained models/nvidia1/20201207091932 nvidia1.h5'
  # modelpath = '../../trained models/nvidia2/20201207192948 nvidia2.h5'
  # assume model is loaded and compiled
  # nvidia2
  modelpath = '../../trained models/nvidia2/20201207192948 nvidia2.h5'
  model1 nvidia2 = load model(modelpath)
  model1 nvidia2.compile("sgd", "mse")
  # nvidia1
  modelpath = '../../trained models/nvidia1/20201207091932 nvidia1.h5'
```

```
model2 nvidia1 = load model(modelpath)
  model2 nvidia1.compile("sgd", "mse")
  # nvidia baseline
  modelpath = '../../trained models/nvidia baseline/20201207201157_nvidia_baseline.h5'
  model3 nvidia baseline = load model(modelpath)
  model3 nvidia baseline.compile("sgd", "mse")
  # sanity
  modelpath = '../../trained models/sanity/20201120171015 sanity.h5'
  model4 sanity = load model(modelpath)
  model4 sanity.compile("sgd", "mse")
  printMultiPlots(model1 nvidia2, model2 nvidia1, model3 nvidia baseline, model4 sanity)
  \#path = 'record 11640.json'
  \#is = load ison(path)
  #print(is)
  # plotSteeringAngles(p, None, 25, True, "Generated Track", "20201120171015" sanity.h5", 'tcpflow log predicted'
  # plots1()
# 20. train.py
# Note: code based on
# https://github.com/tawnkramer/sdsandbox/blob/master/src/train.py
# Available for audit in audit files/tawn from sharepoint link
***
Train
Train your nerual network
Author: Tawn Kramer
To fix missing .jpeg files, see utils/
from future import print function
import argparse
import fnmatch
import json
import os
import pickle
import random
from datetime import datetime
from time import strftime
import numpy as np
from PIL import Image
from tensorflow import keras
# import tensorflow as tf
import conf
import models
from helper functions import hf mkdir, parse bool
from augmentation import augment, preprocess
# import cv2
```

```
import Augmentation
"
matplotlib can be a pain to setup. So handle the case where it is absent. When present,
use it to generate a plot of training results.
try:
  import matplotlib
  # Force matplotlib to not use any Xwindows backend.
  matplotlib.use('Agg')
  import matplotlib.pyplot as plt
  do plot = True
except:
  do plot = False
def shuffle(samples):
  randomly mix a list and return a new list
  ret arr = []
  len samples = len(samples)
  while len samples > 0:
     iSample = random.randrange(0, len_samples)
     ret arr.append(samples[iSample])
     del samples[iSample]
     len samples -= 1
  return ret arr
def load json(filename):
  with open(filename, "rt") as fp:
     data = ison.load(fp)
  return data
def generator(samples, is training, batch size=64):
  Rather than keep all data in memory, we will make a function that keeps
  it's state and returns just the latest batch required via the yield command.
  As we load images, we can optionally augment them in some manner that doesn't
  change their underlying meaning or features. This is a combination of
  brightness, contrast, sharpness, and color PIL image filters applied with random
  settings. Optionally a shadow image may be overlayed with some random rotation and
  opacity.
  We flip each image horizontally and supply it as a another sample with the steering
  negated.
  num samples = len(samples)
  while 1: # Loop forever so the generator never terminates
     samples = shuffle(samples)
     #divide batch size in half, because we double each output by flipping image.
     for offset in range(0, num samples, batch size):
```

```
batch samples = samples[offset:offset+batch size]
images = []
controls = []
for fullpath in batch samples: # not sure this is doing anything, as images are not being flipped
  try:
    frame number = os.path.basename(fullpath).split(" ")[0]
    json filename = os.path.join(os.path.dirname(fullpath), "record " + frame number + ".json")
    data = load ison(ison filename)
    steering = float(data["user/angle"])
    throttle = float(data["user/throttle"])
    try:
       image = Image.open(fullpath)
    except:
       print('failed to open', fullpath)
       continue
    #PIL Image as a numpy array
    image = np.array(image, dtype=np.float32)
    \# image cp = image
    # resize for nvidia
    # nvidia 2
    # image = cv2.resize(image, (200, 66), cv2.INTER AREA)
    # augmentation only for training
    if(conf.aug):
       if is training and np.random.rand() < 0.6:
         image, steering = ag.augment(image, steering)
    # This provides this actual size network is expecting, so must run
    if (conf.preproc):
       image = ag.preprocess(image) # preprocess(image)
    # assert (preprocess(image)==ag.preprocess(image))
    # for nvidia2 model
    # 224 224 Alexnet
    # image = cv2.resize(image, (224, 224), cv2.INTER AREA)
    # for NVIDIA should be 200x66
    images.append(image)
    if conf.num outputs == 2:
       controls.append([steering, throttle])
    elif conf.num outputs == 1:
       controls.append([steering])
    else:
       print("expected 1 or 2 outputs")
  except Exception as e:
    print(e)
    print("we threw an exception on:", fullpath)
    yield [], []
```

```
X train = np.array(images)
       y_train = np.array(controls)
       yield X train, y train
def get files(filemask, s=False):
  Use a filemask and search a path recursively for matches
  Inputs
     filemask: string passed as command line option, must not be enclosed in quotes
     s: boolean, sort by create date flag
  filemask = os.path.expanduser(filemask)
  path, mask = os.path.split(filemask)
  matches = []
  for root, dirnames, filenames in os.walk(path):
     for filename in fnmatch.filter(filenames, mask):
       matches.append(os.path.join(root, filename))
  if(s == True):
     matches = sorted(matches, key=os.path.getmtime)
  return matches
def train test split(lines, test perc):
  split a list into two parts, percentage of test used to seperate
  train = []
  test = []
  for line in lines:
     if random.uniform(0.0, 1.0) < \text{test perc}:
       test.append(line)
     else:
       train.append(line)
  return train, test
def make generators(inputs, limit=None, batch size=conf.batch size):
  load the job spec from the csv and create some generator for training
  #get the image/steering pairs from the csv files
  lines = get files(inputs)
  print("found %d files" % len(lines))
  if limit is not None:
     lines = lines[:limit]
     print("limiting to %d files" % len(lines))
  train_samples, validation_samples = train_test_split(lines, test_perc=0.2)
```

```
print("num train/val", len(train samples), len(validation samples))
  # compile and train the model using the generator function
  train generator = generator(train samples, True, batch size=batch size)
  validation generator = generator(validation samples, False, batch size=batch size)
  n train = len(train samples)
  n val = len(validation samples)
  return train generator, validation generator, n train, n val
def go(model_name, outdir, epochs=50, inputs='./log/*.jpg', limit=None):
  print('working on model', model name)
  hf mkdir(outdir)
  outdir += '/' + model name
  hf mkdir(outdir)
  # https://docs.python.org/3/library/datetime.html#strftime-strptime-behavior
  dt = strftime("\%Y\%m\%d\%H\%M\%S")
  fp = outdir + '/' + dt + ' ' + model name;
  model name = fp + '.h5'
  modify config.json to select the model to train.
  # model = models.get nvidia model naoki(conf.num outputs)
  # interpreter seems to be playing up, dummy assignment to appease
  if(conf.model name=='nvidia1'):
     model = models.nvidia model1(conf.num outputs)
  elif(conf.model name=='nvidia2'):
     model = models.nvidia model2(conf.num outputs)
  elif(conf.model name == 'nvidia baseline'):
     model = models.nvidia baseline(conf.num outputs)
  elif(conf.model name == 'alexnet'):
     try:
       model = models.get alexnet(conf.num_outputs)
     except Exception as e:
       print("Failed to save accuracy/loss graph: " + str(e))
     # adjust image size
     # conf.row = conf.image width net = conf.image width alexnet
     # conf.col = conf.image height net = conf.image height alexnet
  else:
     try:
       raise ValueError
     except ValueError:
       print('No valid model name given. Please check command line arguments and model.py')
  callbacks = [
     # running with naoki's model
```

```
keras.callbacks.EarlyStopping(monitor='val loss', patience=conf.training patience, verbose=0),
     keras.callbacks.ModelCheckpoint(model name, monitor='val loss', save best only=True, verbose=0),
     # keras.callbacks.ModelCheckpoint(('model-{epoch:03d}' +' ' + model name), monitor='val loss', save best o
nly=True, verbose=0),
  batch size = conf.batch size
  #Train on session images
  train generator, validation generator, n train, n val = make generators(inputs, limit=limit, batch size=batch siz
e)
  if n train == 0:
     print('no training data found')
     return
  steps per epoch = n train // batch size
  validation steps = n \text{ val } / / \text{ batch size}
  print("steps per epoch", steps per epoch, "validation steps", validation steps)
  s1 = strftime("\%Y\%m\%d\%H\%M\%S")
  #history = model.fit generator(train generator,
      steps per epoch = steps per epoch,
      validation data = validation generator,
     validation steps = validation steps,
     epochs=epochs,
     verbose=1,
  # callbacks=callbacks)
  history = []
  try:
     history = model.fit(train generator,
       steps per epoch = steps per epoch,
       validation data = validation generator,
       validation steps = validation steps,
       epochs=epochs,
       verbose=1,
       callbacks=callbacks)
  except Exception as e:
     print("Failed to run model: " + str(e))
  # e = "Input to reshape is a tensor with 147456 values, but the requested shape requires a multiple of 27456". errp
r rao with jungle1 dataset
  # [[node model/flattened/Reshape (defined at /git/sdsandbox/src/train.py:250)]] [Op: inference train function
_2398]
  #
  # Function call stack:
  # train function
  s2 = strftime("%Y\%m\%d\%H\%M\%S")
  FMT = "\%Y\%m\%d\%H\%M\%S"
  tdelta = datetime.strptime(s2, FMT) - datetime.strptime(s1, FMT)
  tdelta = "Total training time: " + str(tdelta)
```

```
# save info
  \log = \text{fp} + '.\log'
  logfile = open(log, 'w')
  logfile.write("Model name: " + model_name + "\r\n")
  logfile.write(tdelta)
  logfile.write('\r\n');
  logfile.write("Training loss: " + '{0:.3f}'.format(history.history['loss'][-1]) + "\r\n")
  logfile.write("Validation loss: " + '{0:.3f}'.format(history.history['val loss'][-1]) + "\r\n")
  logfile.write("Training accuracy: " + '{0:.3f}'.format(history.history['acc'][-1]) + "\r\n")
  logfile.write("Validation accuracy: " + '{0:.3f}'.format(history.history['val_acc'][-1]) + "\r\n")
  logfile.close()
  # save history
  histfile = fp + '.history'
  with open(histfile, 'wb') as file pi:
     pickle.dump(history.history, file pi)
  try:
     if do plot:
        fig = plt.figure()
        sp = (1,v1,a,va)' + (0..3f)'.format(history.history['loss'][-1]) \setminus
           +','+'{0:.3f}'.format(history.history['val loss'][-1]) \
           +','+'\{0:.3f\}'.format(history.history['acc'][-1]) \
           +'', +'' (0:.3f)'.format(history.history['val acc'][-1]) \
           +'-'+ model name.split('/')[-1]
        fig.suptitle(sp, fontsize=9)
        ax = fig.add subplot(111)
        #ax.plot(time, Swdown, '-', label='Swdown')
        ax.plot(history.history['loss'], 'r-', label='Training Loss', )
        ax.plot(history.history['val loss'], 'c-', label='Validation Loss')
        ax2 = ax.twinx()
        ax2.plot(history.history['acc'], 'm-', label='Training Accuracy')
        ax2.plot(history.history['val acc'], 'y-', label='Validation Accuracy')
        ax.legend(loc=2) # https://matplotlib.org/3.1.1/api/ as gen/matplotlib.pyplot.legend.html
        ax.grid()
        ax.set xlabel("Epoch")
        ax.set ylabel(r"Loss")
        ax2.set ylabel(r"Accurary")
        ax.set ylim(0, 0.2)
        ax2.set ylim(0.5, 1)
        ax2.legend(loc=1)
        aimg = fp + ' accuracy.png'
        plt.savefig(aimg)
  except Exception as e:
     print("Failed to save accuracy/loss graph: " + str(e))
# moved to helper functions
#def parse bool(b):
# return b == "True"
if name == " main ":
  parser = argparse.ArgumentParser(description='train script')
  parser.add argument('--model', type=str, help='model name')
```

```
parser.add_argument('--outdir', type=str, help='output directory')
parser.add_argument('--epochs', type=int, default=conf.training_default_epochs, help='number of epochs')
parser.add_argument('--inputs', default='../dataset/unity/genRoad/*.jpg', help='input mask to gather images')
parser.add_argument('--limit', type=int, default=None, help='max number of images to train with')
parser.add_argument('--aug', type=parse_bool, default=False, help='image augmentation flag')
parser.add_argument('--preproc', type=parse_bool, default=True, help='image preprocessing flag')

args = parser.parse_args()

conf.aug = args.aug
conf.preproc = args.preproc
conf.model_name = args.model
#print(tf.__version___) 2.2.0
ag = Augmentation.Augmentation(args.model)
go(args.model, args.outdir, epochs=args.epochs, limit=args.limit, inputs=args.inputs)

#python train.py ..\outputs\mymodel aug 90 x4 e200 --epochs=200
```