

딥러닝 활용

자동차게임 자율주행하기





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딥러닝 창업동아리(D-Lab) 운영중

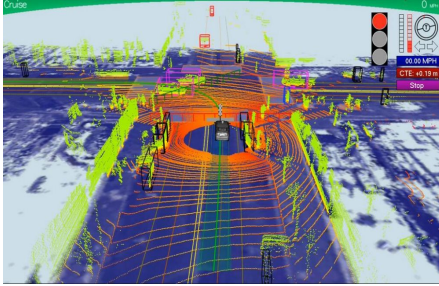
BDA[BigData Analysis] Lab

대학원 진학 준비중

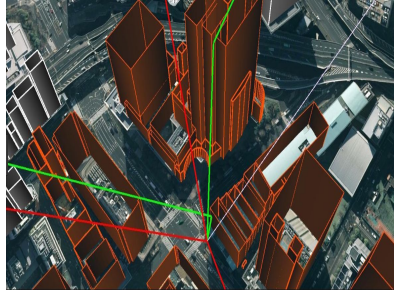
자율주행을 하려면?



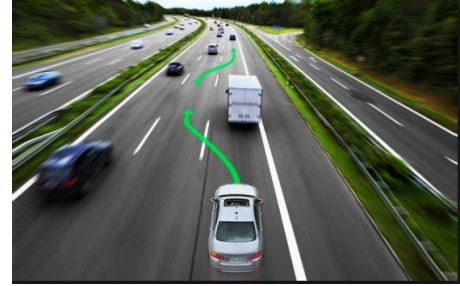
Self-driving technology



mapping

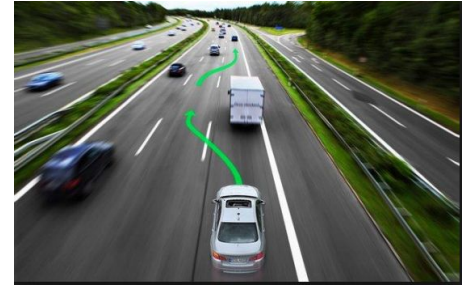
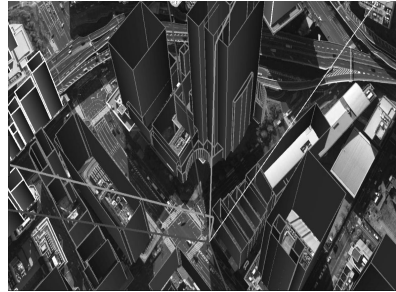
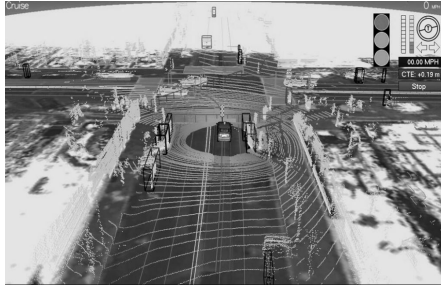


localization



path planning

Self-driving technology



path planning



Why? Deep Learning?



<https://www.youtube.com/watch?v=zxxfvP8-lrU>



End to End Learning for Self-Driving Cars

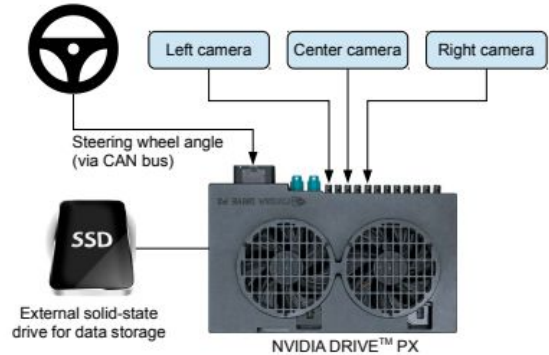


Figure 1: High-level view of the data collection system.

- Time-stamped video from the cameras is captured simultaneously with the steering angle applied by the human driver.
- Training data contains single images sampled from the video, paired with the corresponding steering command (1/r)

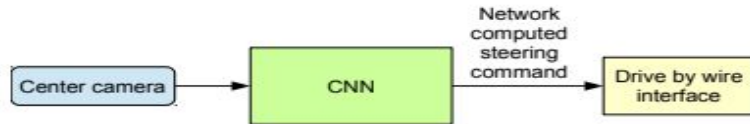
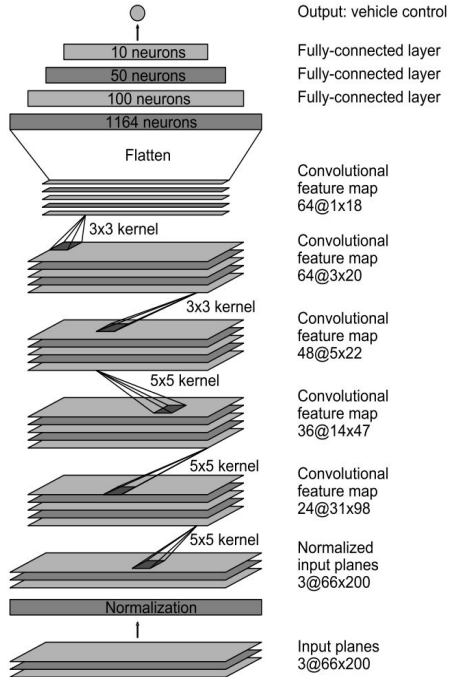


Figure 3: The trained network is used to generate steering commands from a single front-facing center camera.

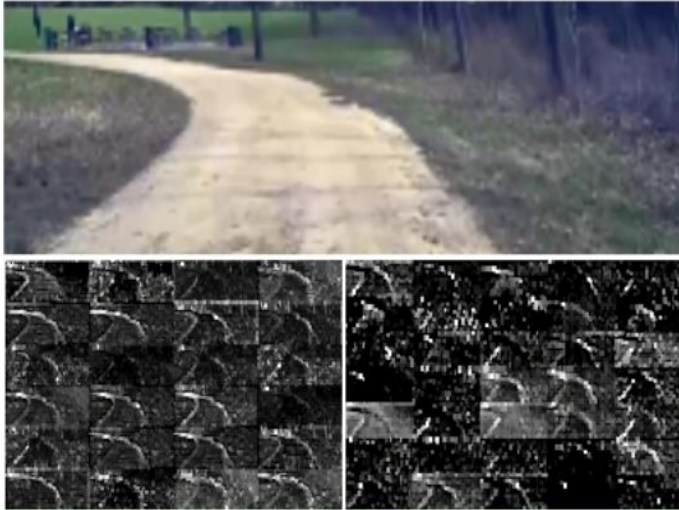
Nvidia model



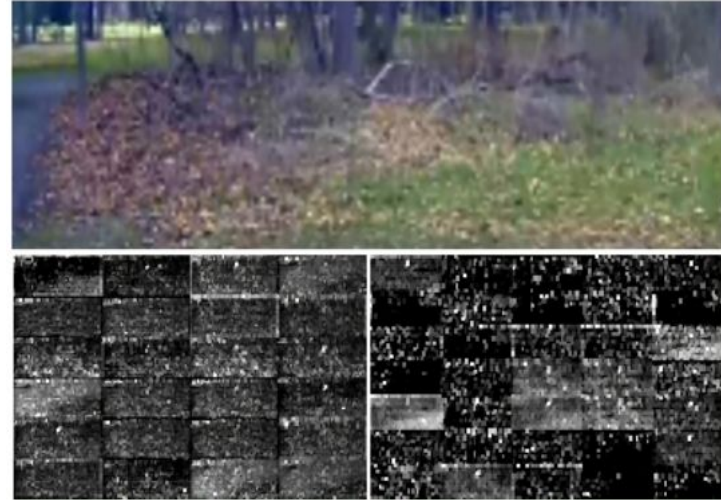
```
def get_model():
    pool_size = (2, 2)
    stride = (2, 2)
    dropout = 0.2
    model = Sequential()
    model.add(Lambda(lambda x: x/127.5 - 1.,
                      input_shape=(66, 200, 3),
                      output_shape=(66, 200, 3)))
    model.add(Convolution2D(5, 5, 24, subsample=stride, border_mode="same"))
    model.add(MaxPooling2D(pool_size = pool_size))
    model.add(ELU())
    model.add(Convolution2D(5, 5, 36, subsample=stride, border_mode="same"))
    model.add(ELU())
    model.add(Convolution2D(5, 5, 48, subsample=stride, border_mode="same"))
    model.add(ELU())
    model.add(Convolution2D(3, 3, 64, subsample=stride, border_mode="same"))
    model.add(ELU())
    model.add(Convolution2D(3, 3, 64, subsample=stride, border_mode="same"))
    model.add(ELU())
    model.add(Flatten())
    model.add(Dropout(.2))
    model.add(ELU())
    model.add(Dense(1164))
    model.add(Dropout(dropout))
    model.add(ELU())
    model.add(Dense(100))
    model.add(Dropout(dropout))
    model.add(ELU())
    model.add(Dense(50))
    model.add(Dropout(dropout))
    model.add(ELU())
    model.add(Dense(10))
    model.add(Dropout(dropout))
    model.add(ELU())
    model.add(Dense(1))
    adam = Adam(lr = 0.00001)
    model.compile(optimizer=adam, loss="mse", metrics = ['accuracy'])
    return model
```

Figure 4: CNN architecture. The network has about 27 million connections and 250 thousand parameters.





Bottom left: Activation of the first layer feature maps.
Bottom right: Activation of the second layer feature maps



The activations of the first two feature maps appear to contain mostly noise

Video Test



First

Let's get it using the game



GTA5



Data Collection

Image



Label

A, W, S, D

방향키



Data Collection

[0]	[1]
	<p data-bbox="1097 452 1619 484">[0, 1, 0, 0, 0, 0, 0, 0, 0]</p>

[0, 0, 0, 0, 0, 0, 0, 0, 0]

[A, W, D, S, AW, WD, AS, SD, 0]



Make the custom data

Window API사용
화면 캡처

Window API사용
키보드 방향키 입력



Numpy형식으로 저장

이미지 데이터	라벨
[200, 200, 200, 199, 199, 199, 199, 198, 198,...]	[0, 1, 0]
[198, 200, 195, 199, 199, 199, 199, 198, 198,...]	[0, 0, 1]
[192, 191, 191, 191, 191, 191, 191, 191, 190,...]	[0, 0, 1]

```

def grab_screen(region=None, title=None):
    hwin = win32gui.GetDesktopWindow()
    if region:
        left, top, x2, y2 = region
        width = x2 - left + 1
        height = y2 - top + 1
    elif title:
        gtawin = win32gui.FindWindow(None, title)
        if not gtawin:
            raise Exception('window title not found')
        #get the bounding box of the window
        left, top, x2, y2 = win32gui.GetWindowRect(gtawin)
        width = x2 - left + 1
        height = y2 - top + 1
    else:
        width = win32api.GetSystemMetrics(win32con.SM_CXVIRTUALSCREEN)
        height = win32api.GetSystemMetrics(win32con.SM_CYVIRTUALSCREEN)
        left = win32api.GetSystemMetrics(win32con.SM_XVIRTUALSCREEN)
        top = win32api.GetSystemMetrics(win32con.SM_YVIRTUALSCREEN)

    hwindc = win32gui.GetWindowDC(hwin)
    srcdc = win32ui.CreateDCFromHandle(hwindc)
    memdc = srcdc.CreateCompatibleDC()
    bmp = win32ui.CreateBitmap()
    bmp.CreateCompatibleBitmap(srcdc, width, height)
    memdc.SelectObject(bmp)
    memdc.BitBlt((0, 0), (width, height), srcdc, (left, top), win32con.SRCCOPY)

    signedIntsArray = bmp.GetBitmapBits(True)
    img = np.fromstring(signedIntsArray, dtype='uint8')
    img.shape = (height, width, 4)

    srcdc.DeleteDC()
    memdc.DeleteDC()
    win32gui.ReleaseDC(hwin, hwindc)
    win32gui.DeleteObject(bmp.GetHandle())

    return cv2.cvtColor(img, cv2.COLOR_BGRA2RGB)

```

Win32.GetAsyncKeyState()사용

Image

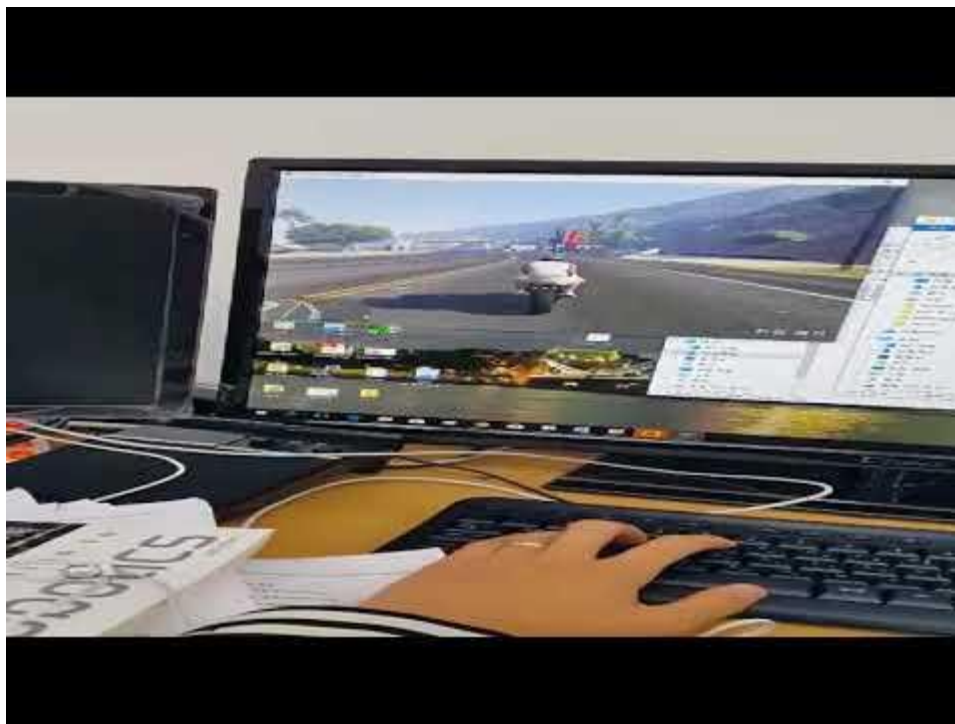
key value



Now you have to drive hard.

아주 많이....





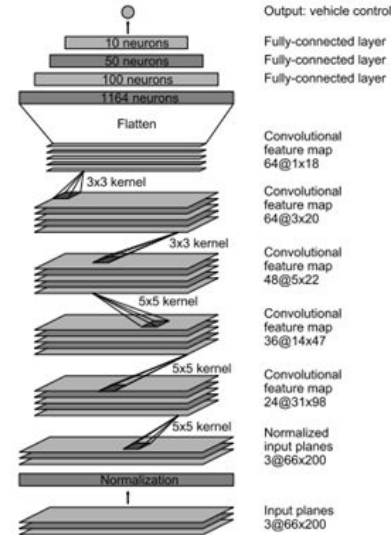
<https://www.youtube.com/watch?v=xJ7PWTNdcnk>



Train

AlexNet, Inception, VGG, Resnet 등등 → from keras.applications.'모델명' import '모델명'

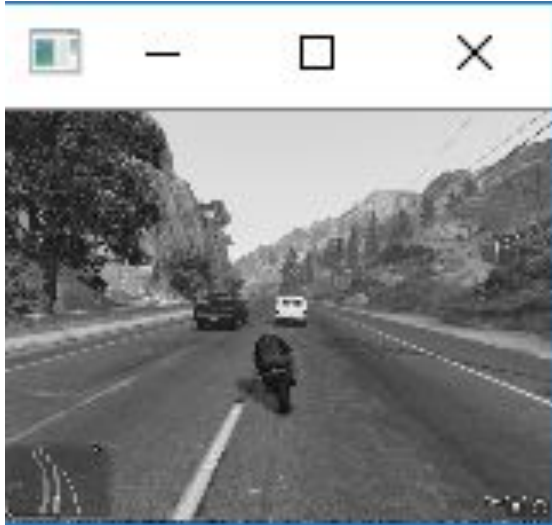
Model	Size	Top-1 Accuracy	Top-5 Accuracy	Parameters	Depth
Xception	88 MB	0.790	0.945	22,910,480	126
VGG16	528 MB	0.715	0.901	138,357,544	23
VGG19	549 MB	0.727	0.910	143,667,240	26
ResNet50	99 MB	0.759	0.929	25,636,712	168
InceptionV3	92 MB	0.788	0.944	23,851,784	159
InceptionResNetV2	215 MB	0.804	0.953	55,873,736	572
MobileNet	17 MB	0.665	0.871	4,253,864	88
DenseNet121	33 MB	0.745	0.918	8,062,504	121
DenseNet169	57 MB	0.759	0.928	14,307,880	169
DenseNet201	80 MB	0.770	0.933	20,242,984	201



Karol Zieba. End to End Learning for Self-Driving Cars 25 Apr 2016



Train Data



gray image

Size: 120x80

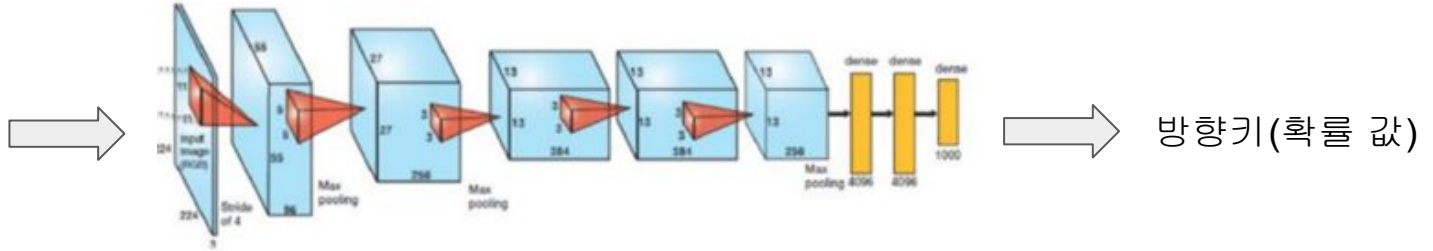
300K

Train time : 60H
(CPU : i5
GPU: X)

Test



Game screen



Test Video



https://www.youtube.com/watch?v=32ywp_y1QEc



Future Work



Driving Data



480 x 270 RGB Image

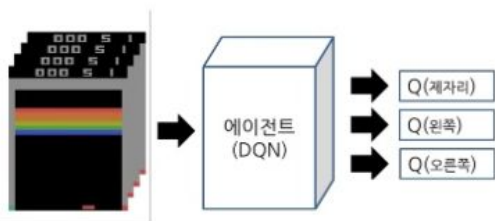
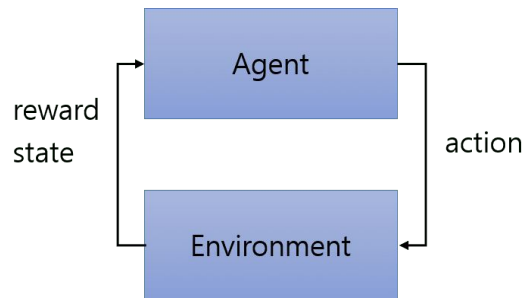
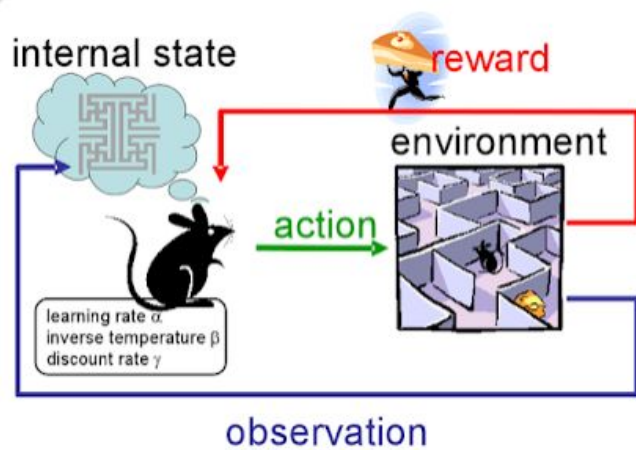
600,000 Image

Data collection after navigation setup

Object Detection

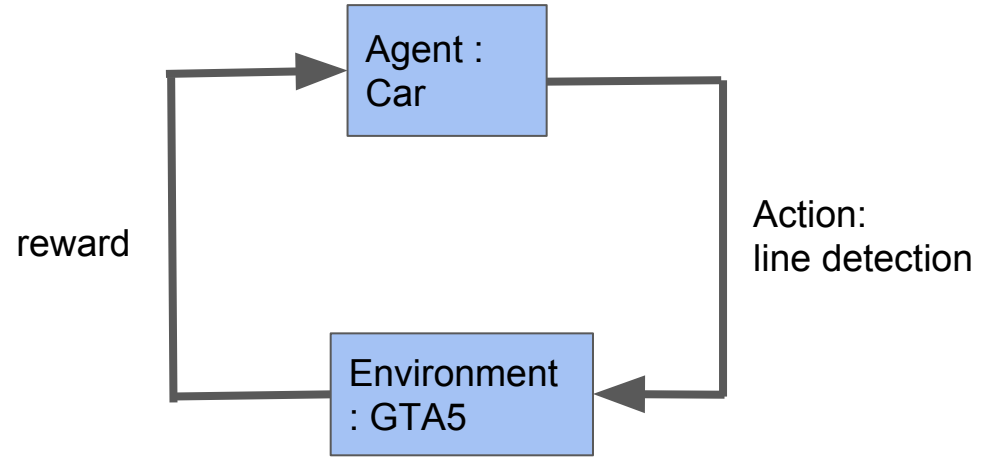
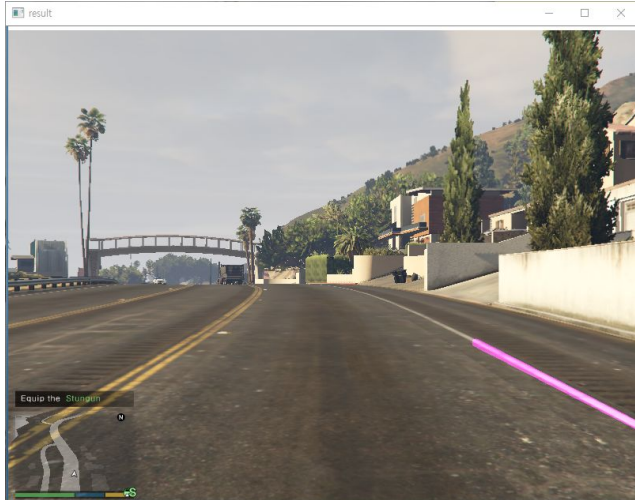


Reinforcement Learning



https://www.slideshare.net/WoongwonLee/ss-78783597?from_m_app=android

Reinforcement Learning



Q & A



감사합니다.

