아티스트를 위한 머신러닝 & 딥러닝

텐서플로를 활용한 딥러닝 #1

서울대학교 & V.DO / 김대식

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딥러닝
Blah blah 이전에..

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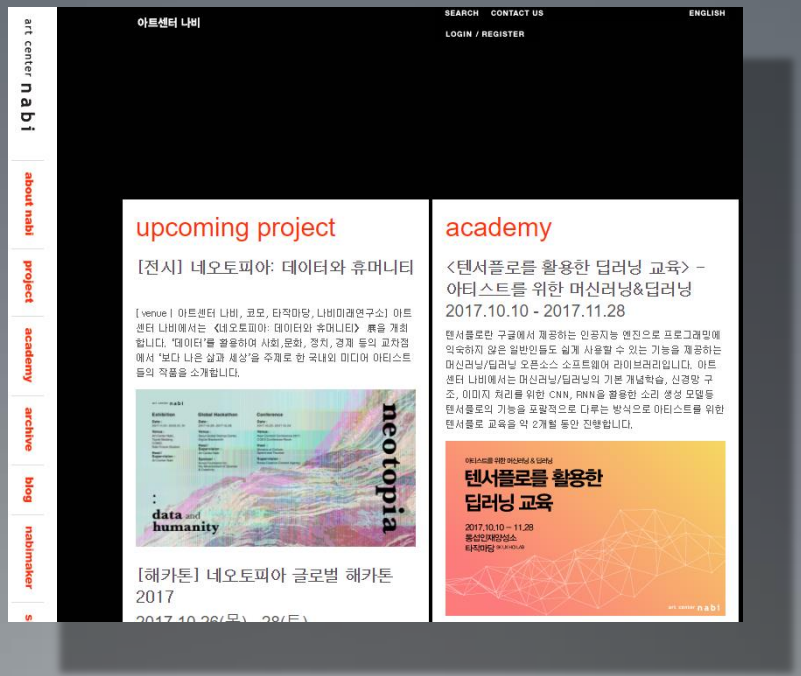
누가 이 수업을 들을까?

과연 들으려는 사람이 있을까?



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Why Here?

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저의 소개

PhD candidate in SNU
CTO in V.DO

Research Interests

- Machine Learning, Multi-modal Deep Learning, Knowledge based Deep learning

Education

- B.Sc. Industrial Engineering and Management, Pohang University of Science and Technology, 2007

Employment

- System developer, Programmer(Powerbuilder, Oracle), Mega IT. 2002-2005
- Assistant Manager ,Tresury department, Kookmin bank, 2007-2014
- Founder & CTO in V.DO, 2017~

Publication

- Seo, J., Kim, D., Suh, B., & Lee, J. (2015, April). Design of a Smart TV Logging System Using Beacons and Smartphones. In Proceedings of the 33rd Annual ACM Conference Extended Abstracts on Human Factors in Computing Systems (pp. 2157-2162). ACM.
- 멀티모달 딥러닝을 이용한 음악 장르 분류. HCI KOREA 2016
- Daesik Kim, Myunggi Lee and Nojun Kwak, "Matching Video Net: Memory-based embedding for video action recognition", International Joint Conference on Neural Networks 2017 (IJCNN2017), Anchorage, AK, May 2017.
- Sangkuk Lee, Daesik Kim, Myunggi Lee, Jihye Hwang and Nojun Kwak, "Where to Play: Retrieval of video segments using natural-language queries", arXiv, July 2017

Competition

- 4th place of "Textbook Question Answering Challenge", CVPR 2017 Workshop on Visual Understanding Across Modalities

Teaching

- Tensorflow tutorial at OSIA (Aug. 2016)
- Tensorflow tutorial at CONTECH (Apr. 2017)

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인공지능
딥러닝

현재 가장 핫한 키워드

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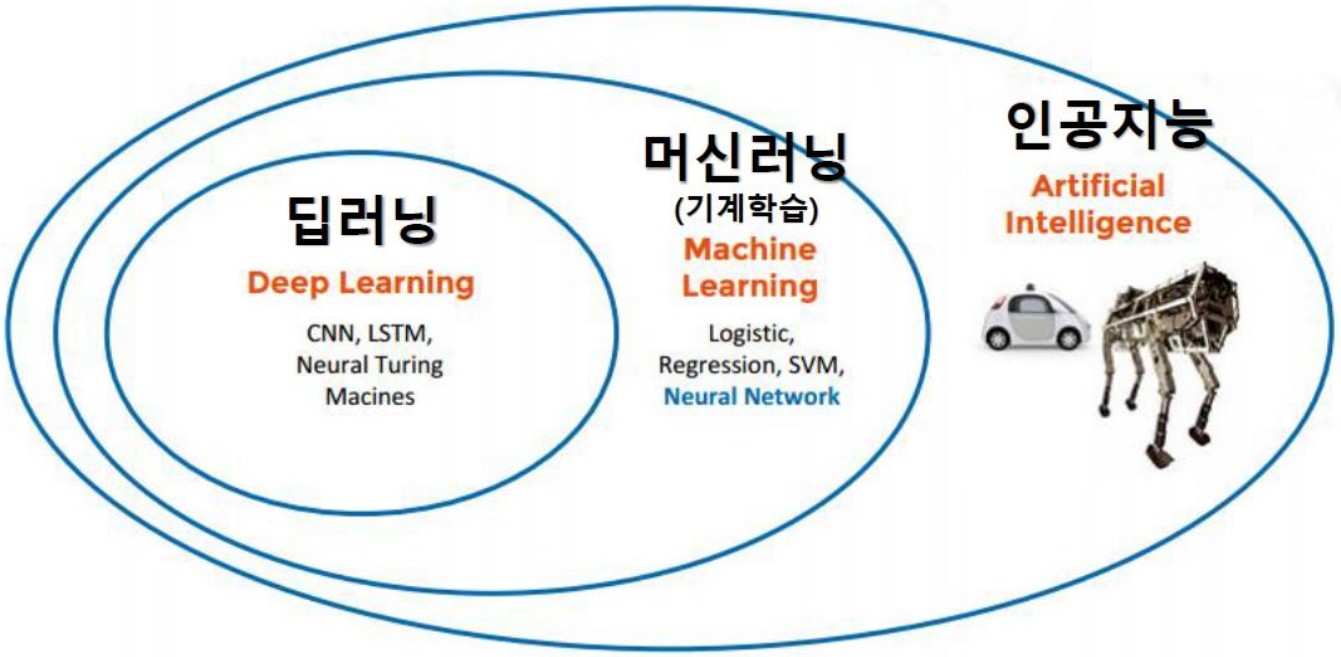
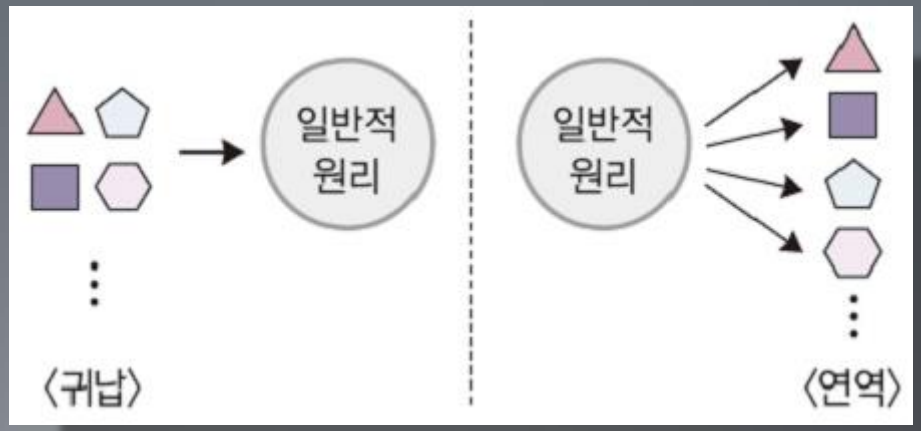
인공지능 ≠ 딥러닝

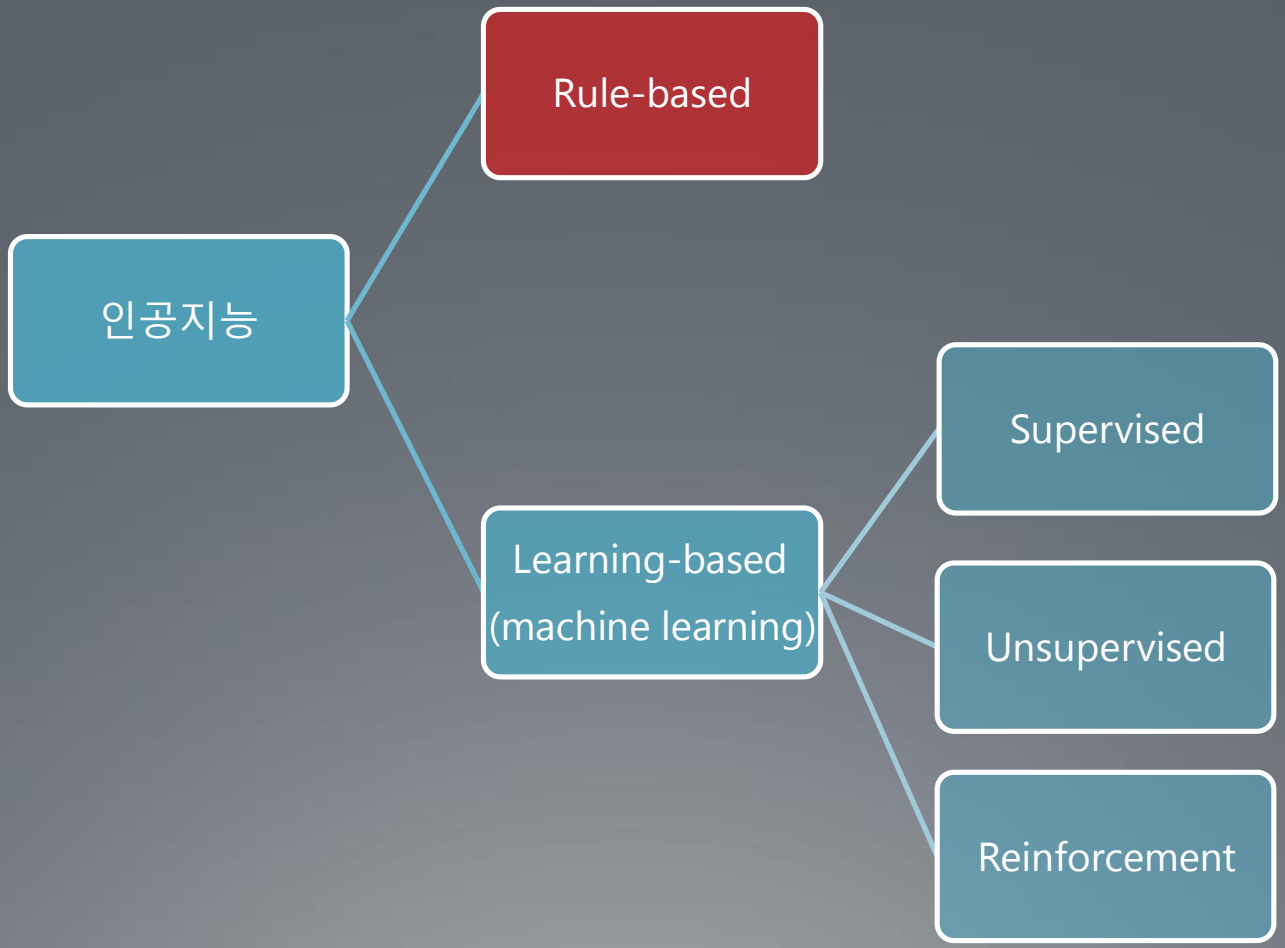


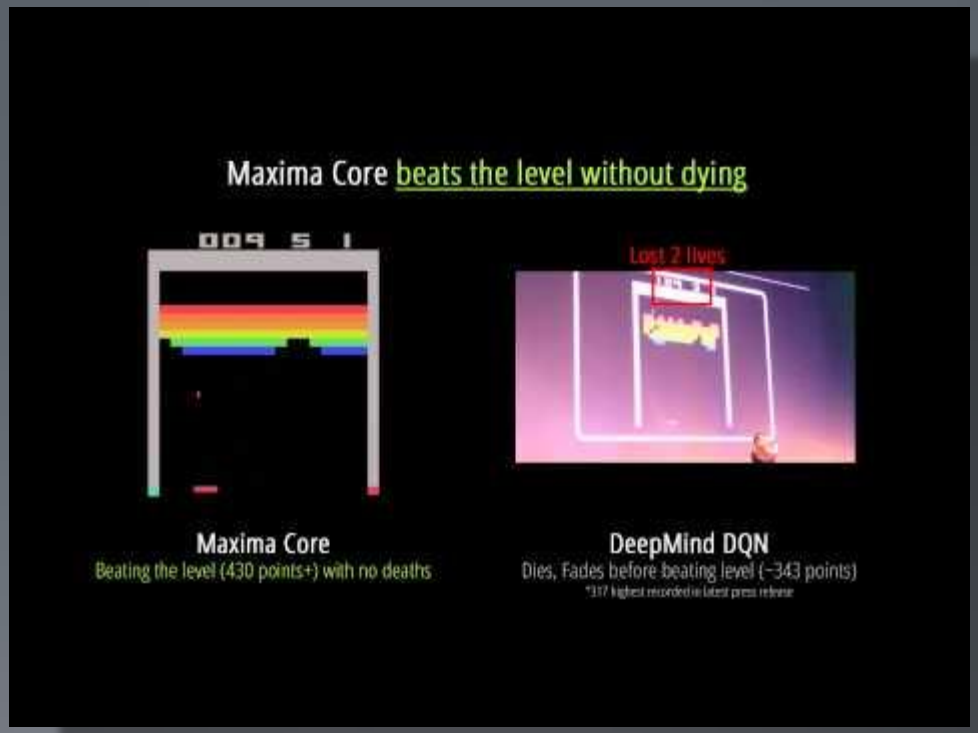
Slide by Jiqiong Qiu at DevFest 2016
http://www.slideshare.net/SfeirGroup/first-step-deep-learning-by-jiqiong-qiu-devfest-2016

연역법 vs 귀납법
Deduction vs Induction

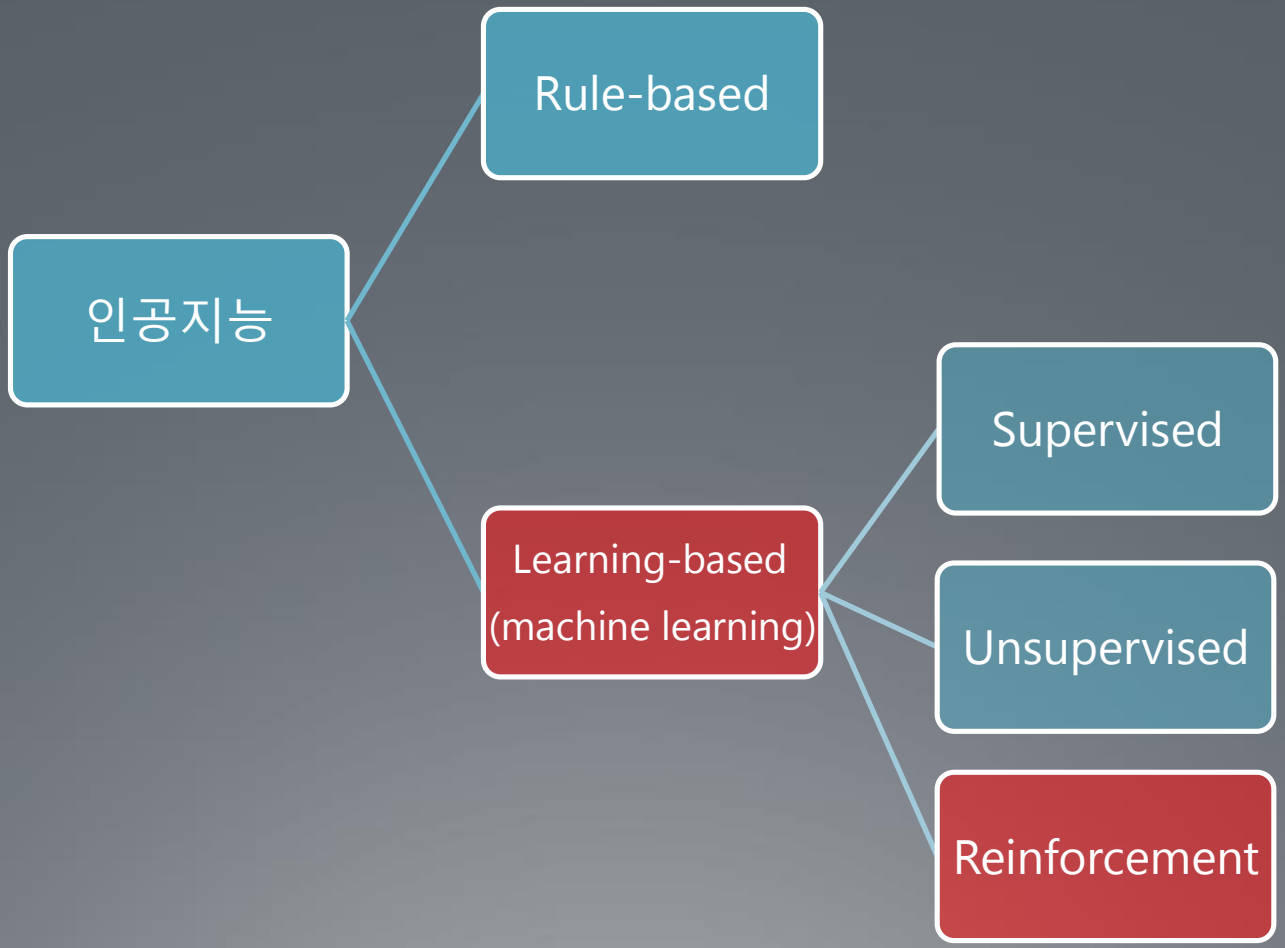


인공지능?


Rule-based
인공지능
Learning-based
(machine learning)
Supervised
Unsupervised
Reinforcement


Maxima Core beats the level without dying
009 5 1
Lost 2 lives
Maxima Core
Beating the level (430 points+) with no deaths
DeepMind DQN
Dies, Fades before beating level (~343 points)

인공지능?


Rule-based
인공지능
Learning-based
(machine learning)
Supervised
Unsupervised
Reinforcement

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AlphaGo

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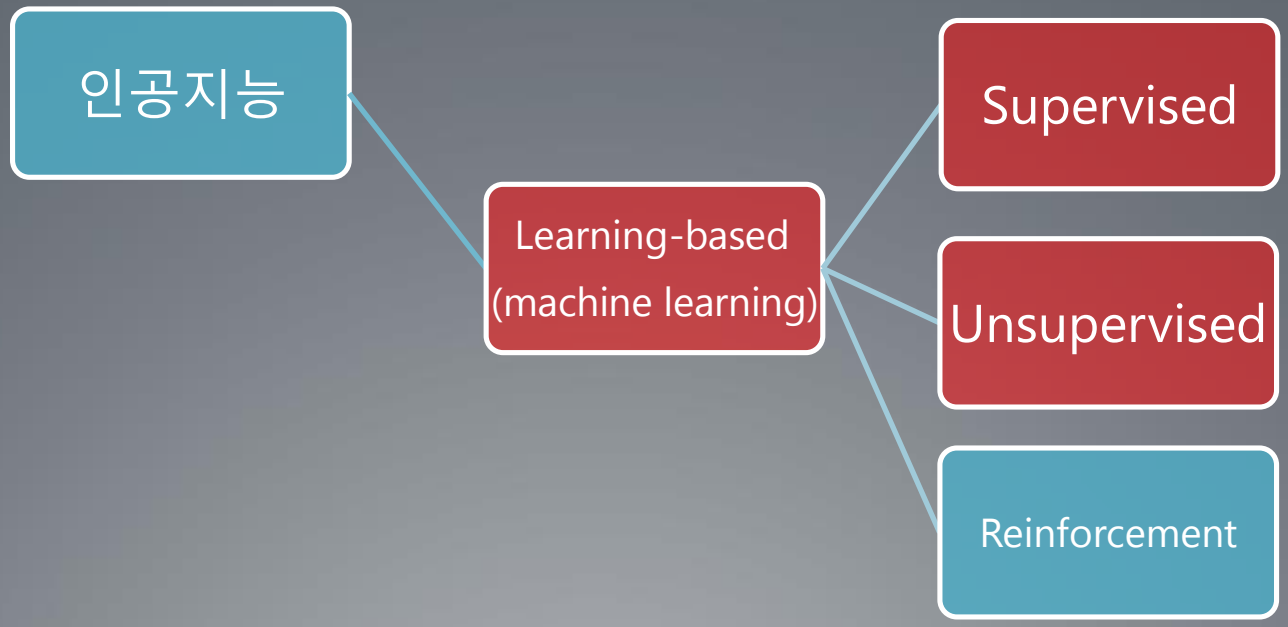
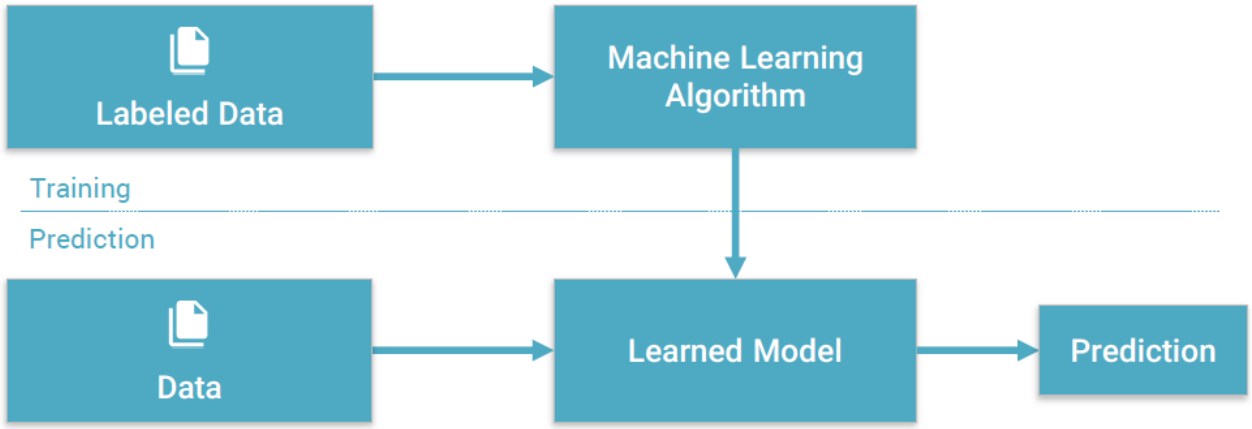
Rule-based + Reinforcement learning

우린 Learning 배웁니다



Machine Learning



Provides various techniques that can learn from and make predictions on data

Machine Learning

Supervised Learning: Learning with a labeled training set
Example: email spam detector with training set of already labeled emails

Unsupervised Learning: Discovering patterns in unlabeled data
Example: cluster similar documents based on the text content

Reinforcement Learning: learning based on feedback or reward
Example: learn to play chess by winning or losing

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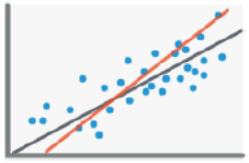
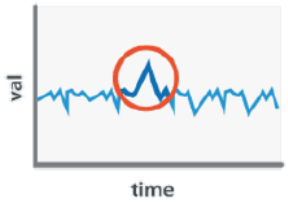
Types of Machine Learning

Classification
(supervised – predictive)



Regression
(supervised – predictive)

Clustering
(unsupervised – descriptive)



Anomaly Detection
(unsupervised– descriptive)

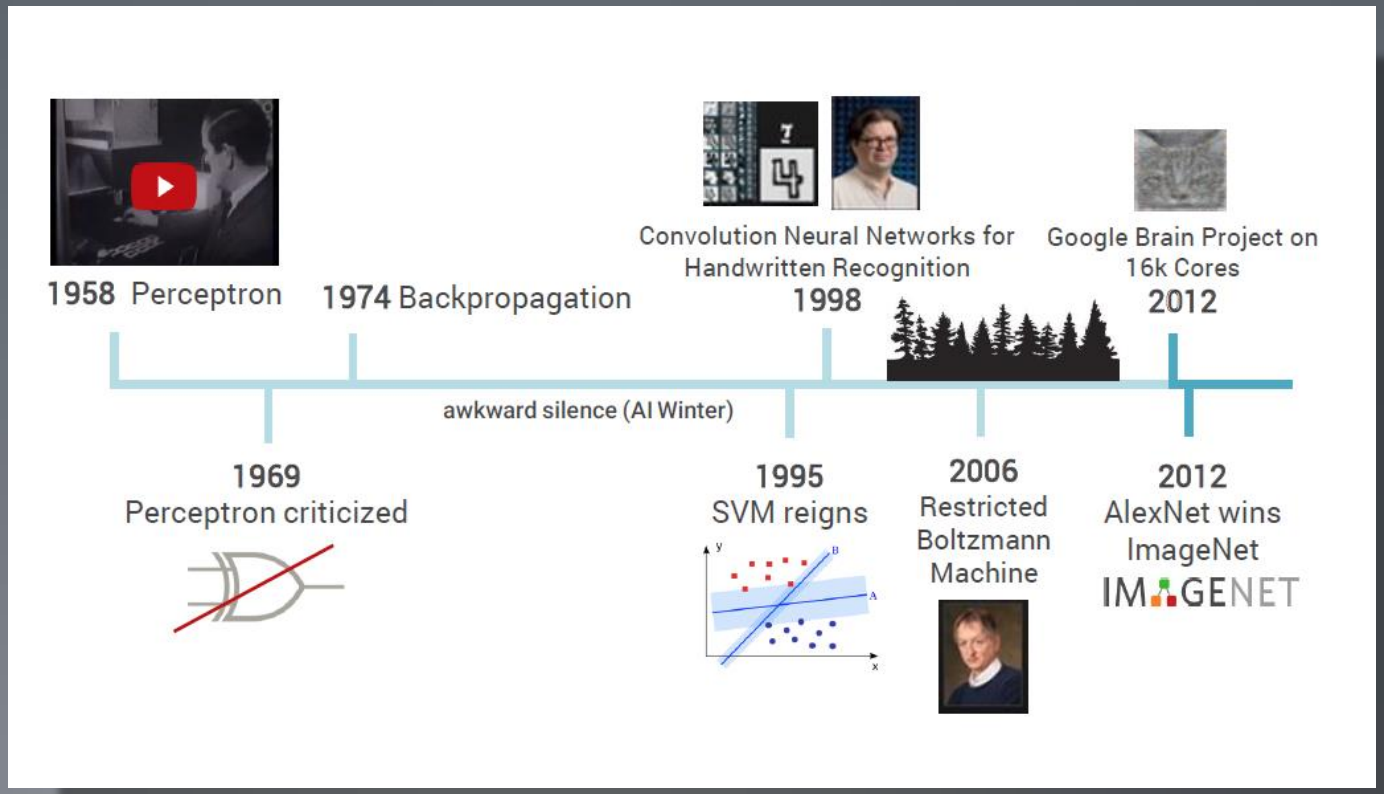
딥러닝의 위엄

https://www.youtube.com/watch?v=pW6nZXeWlGM&feature=youtu.be

https://www.youtube.com/watch?v=gjGOcI17bzg

https://www.youtube.com/watch?v=YRm_kqClsFY

왜 기존에 안되었던 것들이 되기 시작 할까?



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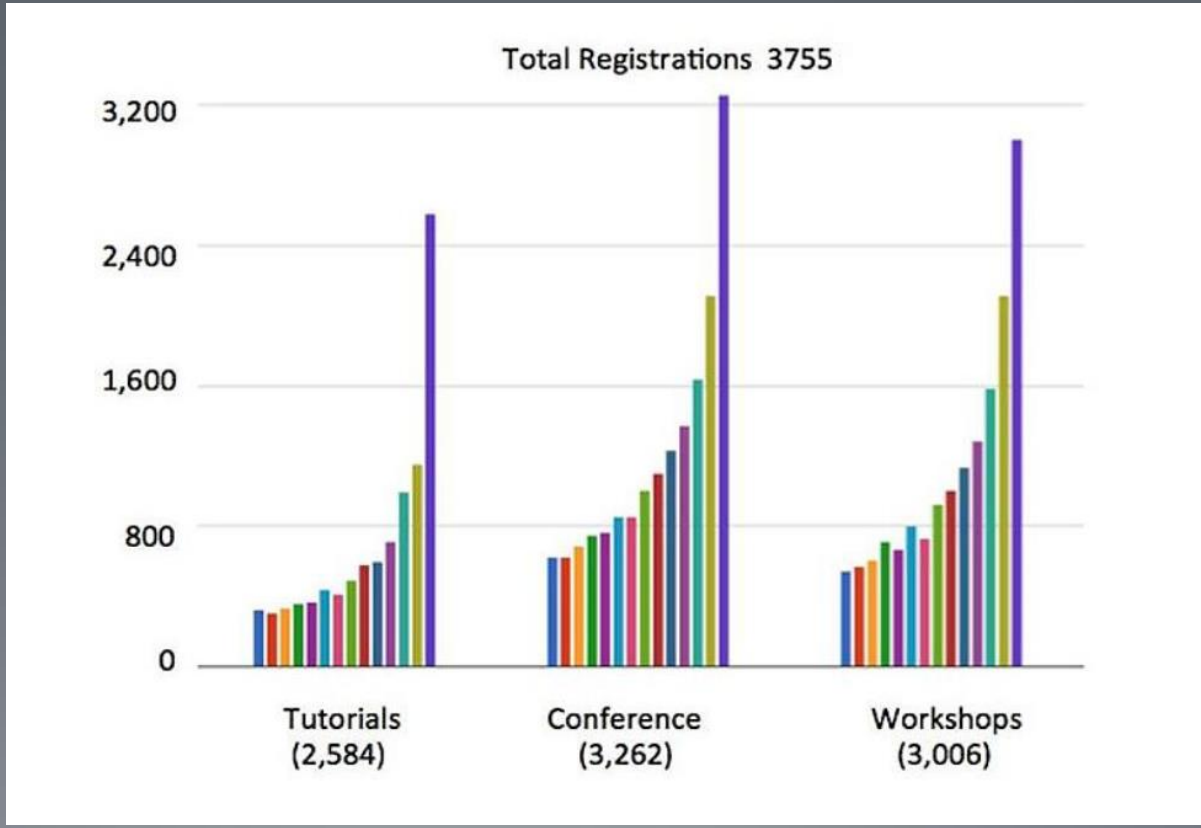
NIPS: 딥러닝 최고 학회



Superstars

Geoffrey Hinton: University of Toronto & Google

Yann LeCun: New York University & Facebook

Andrew Ng: Stanford & Baidu

Yoshua Bengio: University of Montreal

Jürgen Schmidhuber: Swiss AI Lab & NNAISENSE

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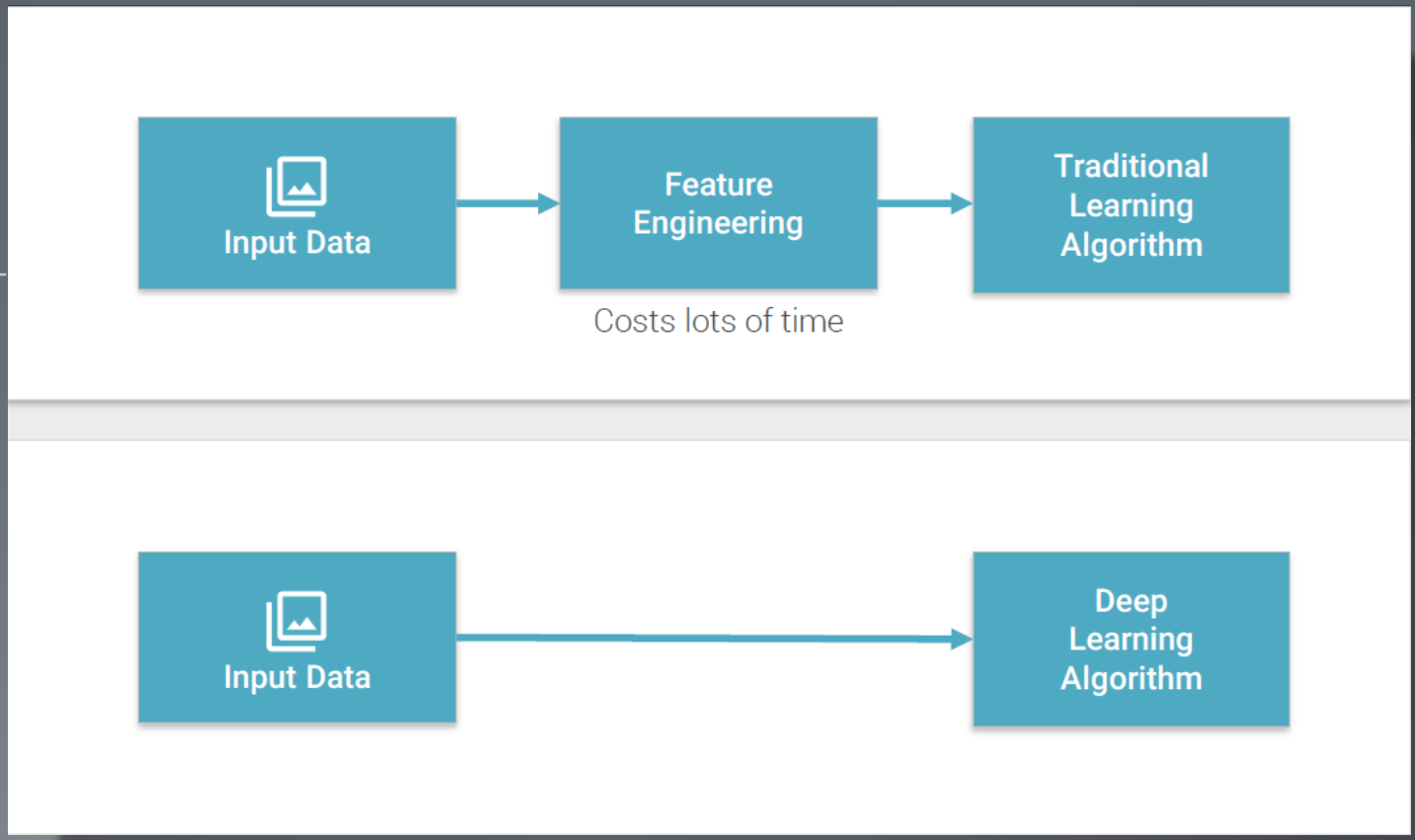
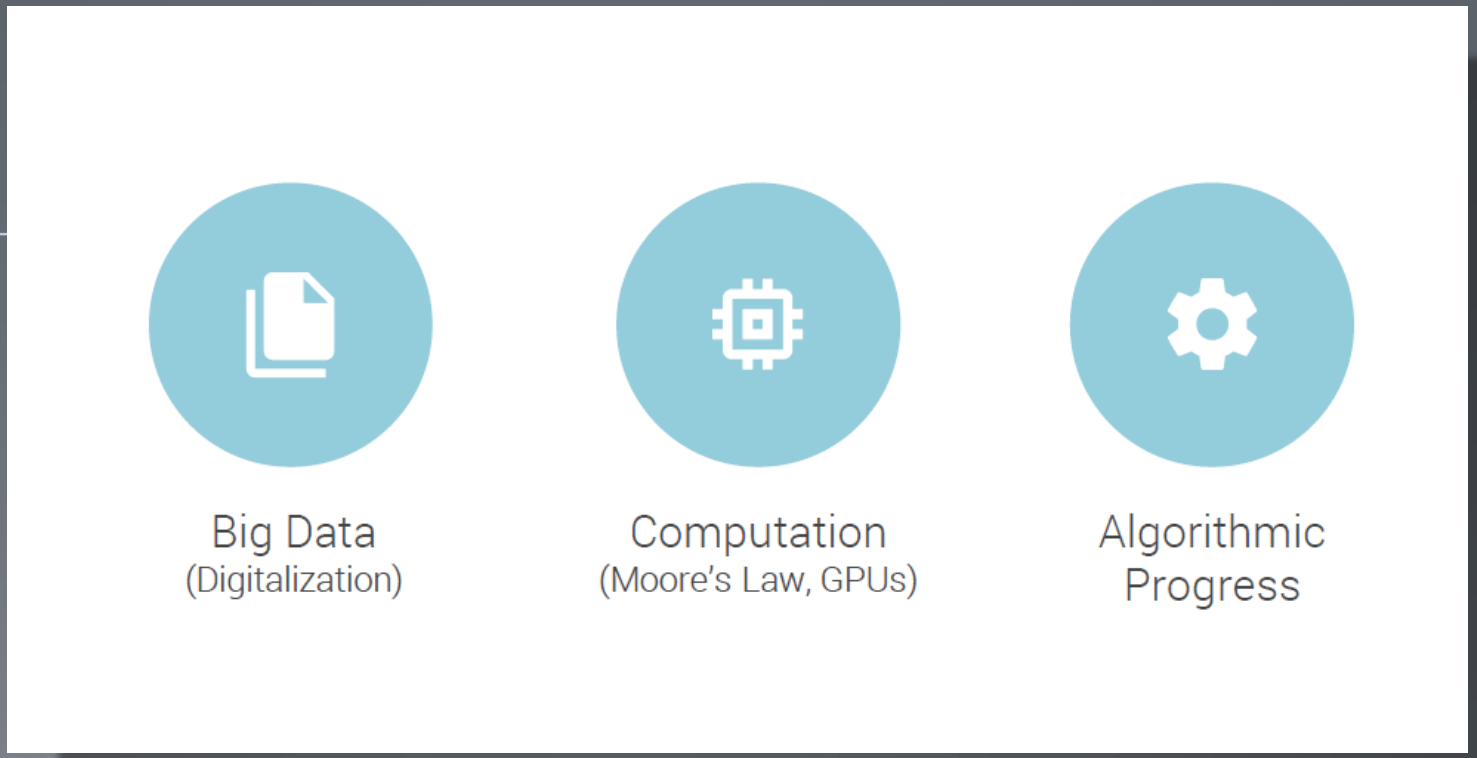
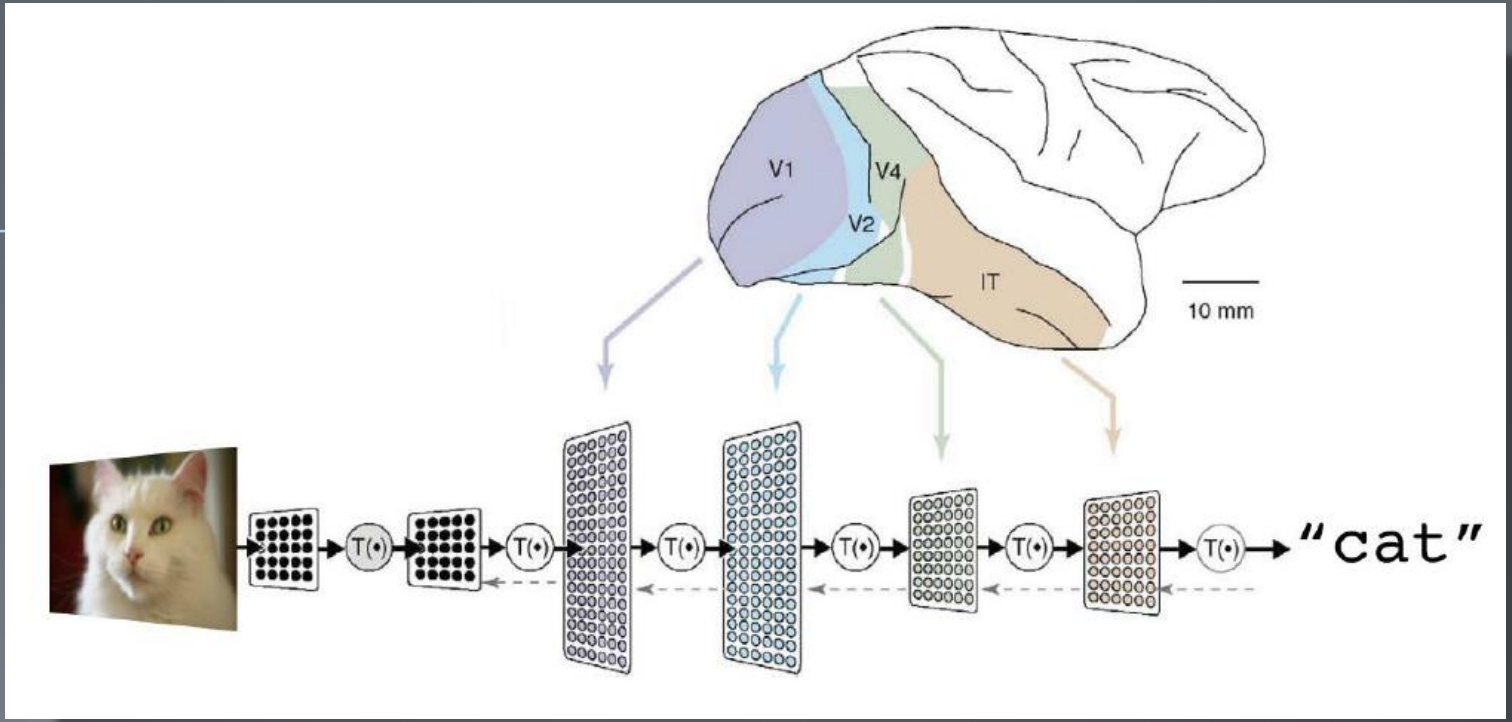
Paradigm의 변화



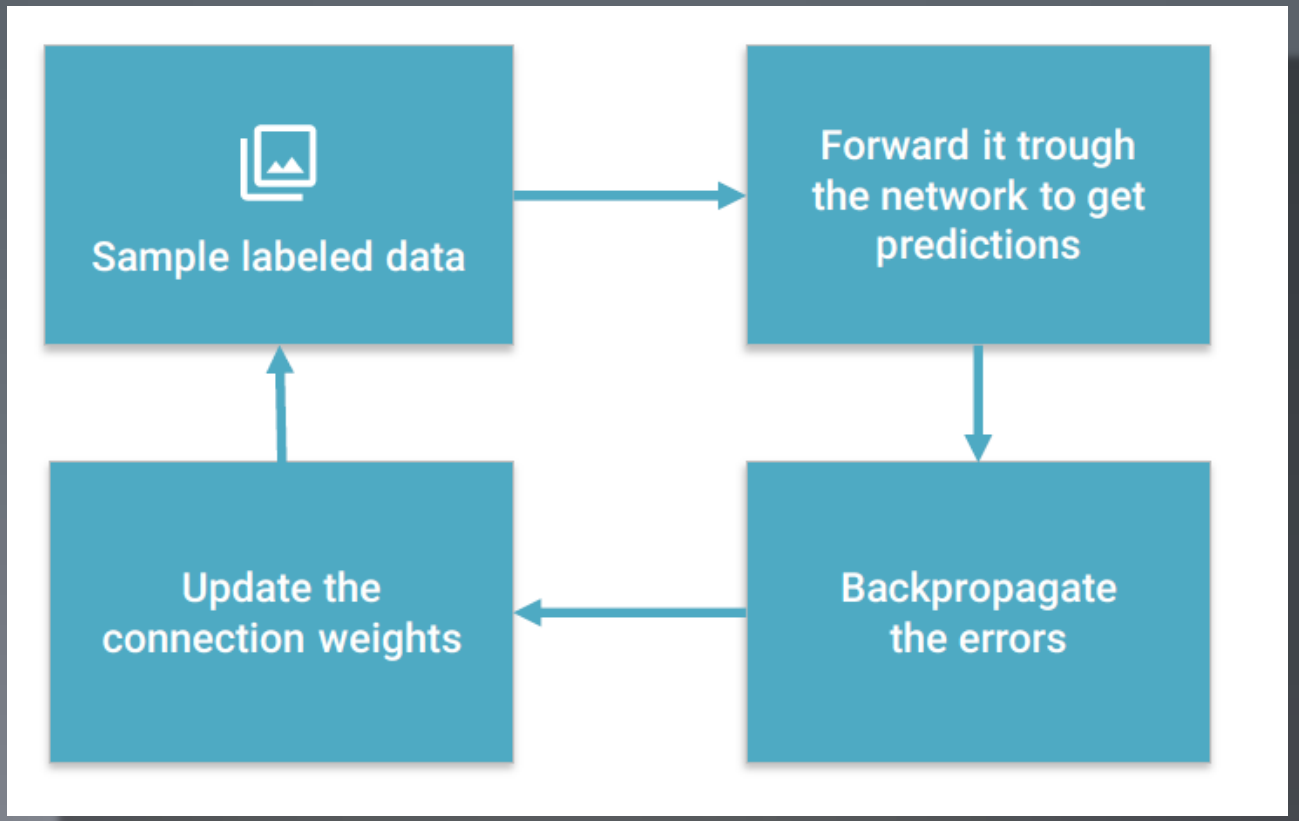
Paradigm의 변화



아키텍처



딥러닝 학습 프로세스


Sample labeled data
Forward it trough the network to get predictions
Backpropagate the errors
Update the connection weights

딥러닝 + ART

Machine Learning for Creativity and Design

NIPS 2017 Workshop, Long Beach, California, USA

Friday December 8th 8:00 — 18:30

https://nips2017creativity.github.io/

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딥러닝 + ART

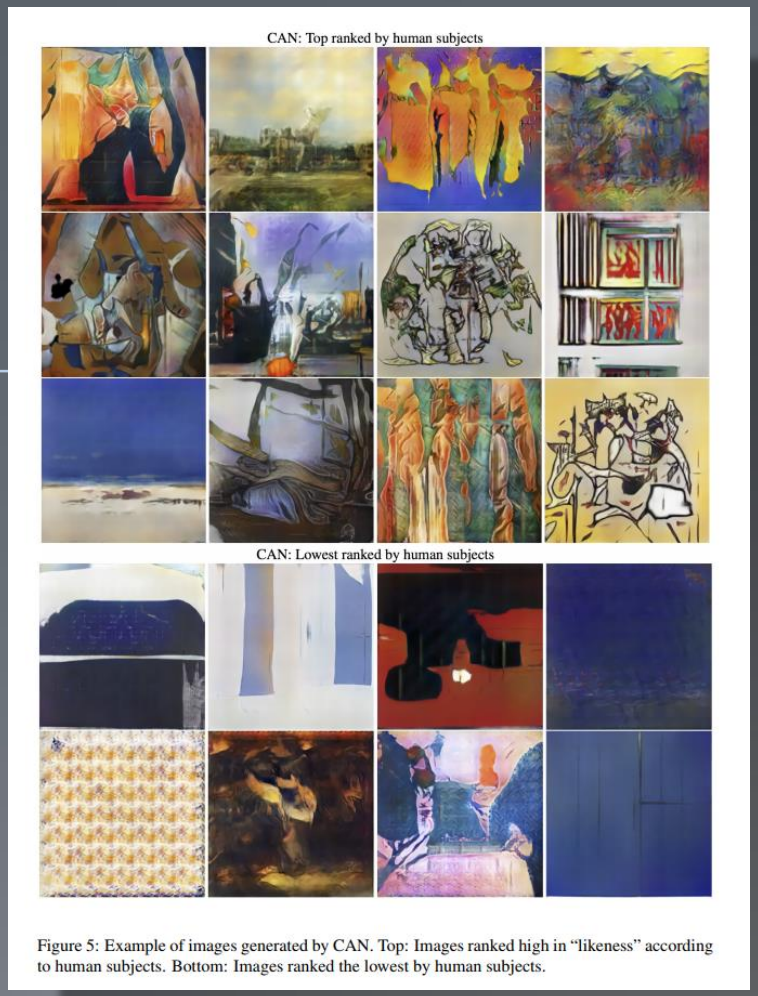

Figure 5: Example of images generated by CAN. Top: Images ranked high in "likeness" according to human subjects. Bottom: Images ranked the lowest by human subjects.

Table 1: Artistic Styles Used in Training

Style name	Image number	Style name	Image number
Abstract-Expressionism	2782	Mannerism-Late-Renaissance	1279
Action-Painting	98	Minimalism	1337
Analytical-Cubism	110	Naive Art-Primitivism	2405
Art-Nouveau-Modern	4334	New-Realism	314
Baroque	4241	Northern-Renaissance	2552
Color-Field-Painting	1615	Pointillism	513
Contemporary-Realism	481	Pop-Art	1483
Cubism	2236	Post-Impressionism	6452
Early-Renaissance	1391	Realism	10733
Expressionism	6736	Rococo	2089
Fauvism	934	Romanticism	7019
High-Renaissance	1343	Synthetic-Cubism	216
Impressionism	13060	Total	75753

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CAN : Creative Adversarial Network
https://arxiv.org/pdf/1706.07068.pdf

딥러닝님께서
모든 문제를 해결?
새로운 것을 창조?

딥러닝

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매우 높은 차원의 함수

+

매우 많은 양의 데이터

창조

모방

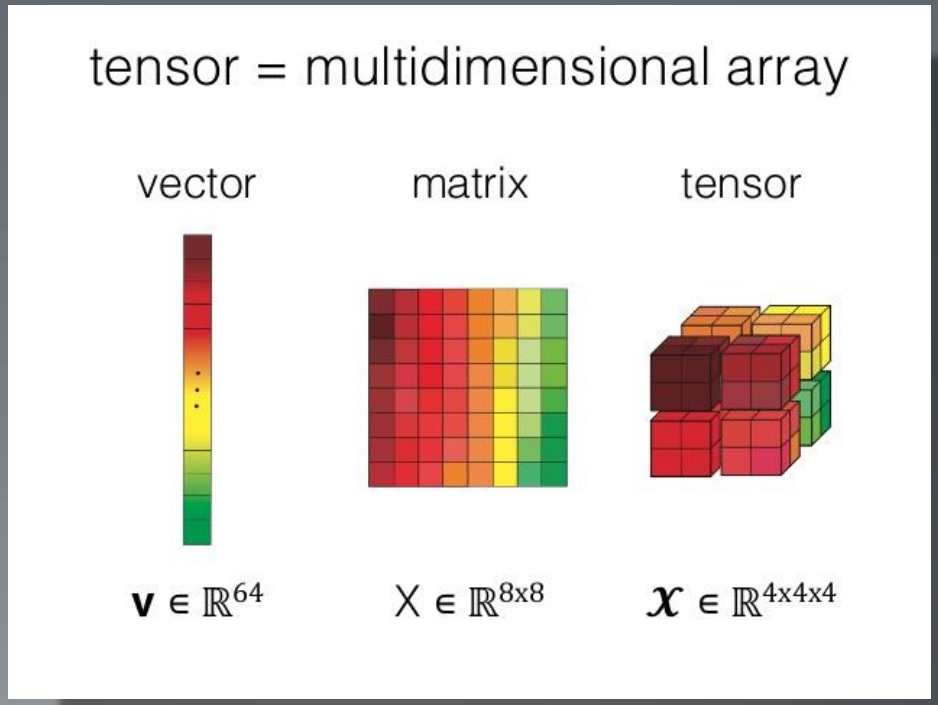
텐서플로(Tensorflow)

https://www.tensorflow.org/

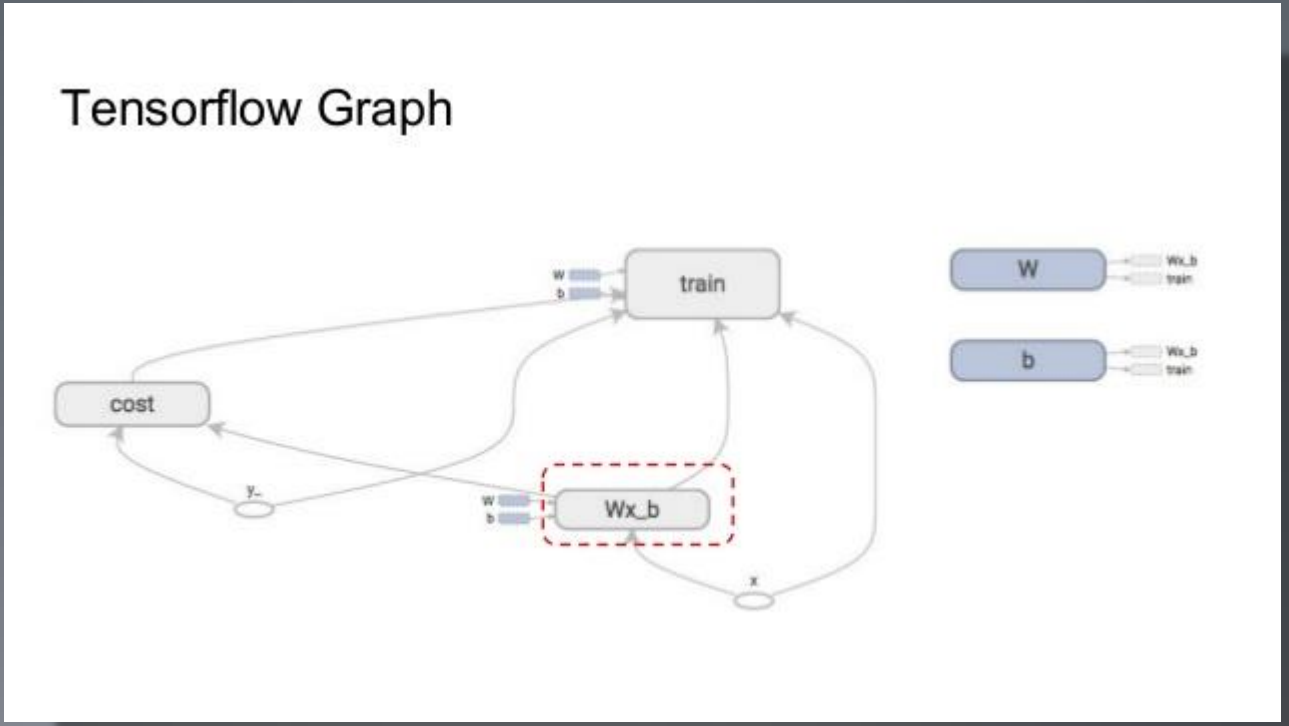
Tensorflow
=Tensor + Flow

Tensor :

3차원 이상의 matrix



Flow into Graph



Tensor : 3차원 이상의 matrix
모양의 DATA
Flow : Graph형태의 Algorithm

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텐서플로를 한다는 것:

주어진 DATA 를
구현한 Algorithm 에
흘려 학습하는 과정

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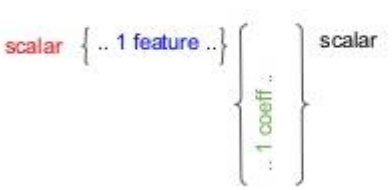
Tensorflow Code

```
# Model linear regression y = Wx + b
x = tf.placeholder(tf.float32, [None, 1])
W = tf.Variable(tf.zeros([1,1]))
b = tf.Variable(tf.zeros([1]))
product = tf.matmul(x,W)
y = product + b
y_ = tf.placeholder(tf.float32, [None, 1])

# Cost function 1/n * sum((y_-y)**2)
cost = tf.reduce_mean(tf.square(y_-y))

# Training using Gradient Descent to minimize cost
train_step = tf.train.GradientDescentOptimizer(0.0000001).minimize(cost)
```

Tensorflow Graph: Basic, Multi-feature, Multi-class

1 Feature

```
y = tf.matmul(x, W) + b

W = tf.Variable(tf.zeros[1,1])
b = tf.Variable(tf.zeros[1])

x = tf.placeholder(tf.float, [None, 1])
y_ = tf.placeholder(tf.float, [None, 1])
```

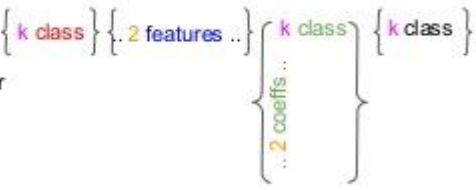


2 Features

```
y = matmul(x, W) + b

W = tf.Variable(tf.zeros[2,1])
b = tf.Variable(tf.zeros[1])

x = tf.placeholder(tf.float, [None, 2])
y_ = tf.placeholder(tf.float, [None, 1])
```

2 Features, 10 Classes

```
y = matmul(x, W) + b

W = tf.Variable(tf.zeros[2,10])
b = tf.Variable(tf.zeros[10])

x = tf.placeholder(tf.float, [None, 2])
y_ = tf.placeholder(tf.float, [None, 10])
```



```python
import tensorflow as tf
import numpy as np

x_data = np.random.rand(100).astype(np.float32)
y_data = x_data * 0.1 + 0.3

W = tf.Variable(tf.random_uniform([1], -1.0, 1.0))
b = tf.Variable(tf.zeros([1]))
y = W * x_data + b

loss = tf.reduce_mean(tf.square(y-y_data))
optimizer = tf.train.GradientDescentOptimizer(0.5)
train = optimizer.minimize(loss)

init = tf.initialize_all_variables()

sess = tf.Session()
sess.run(init)

for step in xrange(201):
    sess.run(train)
    if step % 20 == 0:
        print(step, sess.run(W), sess.run(b))
```

Tensor

Flow

```
(0, array([-0.02093266], dtype=float32), array([ 0.47095078], dtype=float32))
(20, array([ 0.05205543], dtype=float32), array([ 0.32374111], dtype=float32))
(40, array([ 0.08607709], dtype=float32), array([ 0.30689433], dtype=float32))
(60, array([ 0.09595685], dtype=float32), array([ 0.3020021], dtype=float32))
(80, array([ 0.09882588], dtype=float32), array([ 0.3005814], dtype=float32))
(100, array([ 0.09965905], dtype=float32), array([ 0.30016884], dtype=float32))
(120, array([ 0.09990099], dtype=float32), array([ 0.30004904], dtype=float32))
(140, array([ 0.09997127], dtype=float32), array([ 0.30001423], dtype=float32))
(160, array([ 0.09999166], dtype=float32), array([ 0.30000415], dtype=float32))
(180, array([ 0.09999759], dtype=float32), array([ 0.3000012], dtype=float32))
(200, array([ 0.09999931], dtype=float32), array([ 0.30000037], dtype=float32))
```

Tensorflow 설치

https://www.tensorflow.org/install/

Window 10

http://naver.me/59PQvkxa

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Week	Date	Title	Detail
1	10/10	- 딥러닝 시대의 의미 - 텐서플로 기본	머신러닝 분야에서 딥러닝의 의미 텐서플로 설치 및 기본 개념 학습
2	10/17	- 선형회귀분석	기본적인 머신러닝 개념 학습
3	10/24	- 단일/다중 계층신경망	신경망 구조 설명 및 응용
4	10/31	- CNN	이미지 처리를 위한 Convolution neural network 소개
5	11/7	- generative model	다양한 형태의 이미지 생성모델 소개
6	11/14	- style transfer, colorization	이미지 관련 응용 예시 소개
7	11/21	- RNN	Recurrent neural network 소개
8	11/28	- audio application	RNN의 활용한 소리 생성 모델 등 소개

감사합니다