Intro

Machine Learning

Microsoft Student Partner

백진헌

Overview

- 1. Concept of Machine Learning
- 2. Supervised Learning
- 3. Unsupervised Learning

✓ Machine Learning?

"Filed of study that gives computers the ability to learn without being explicitly programmed.", Arthur Samuel (1959)



✓ Machine Learning?

"A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E.", Tom Mitchell (1997)



✓ Machine Learning?

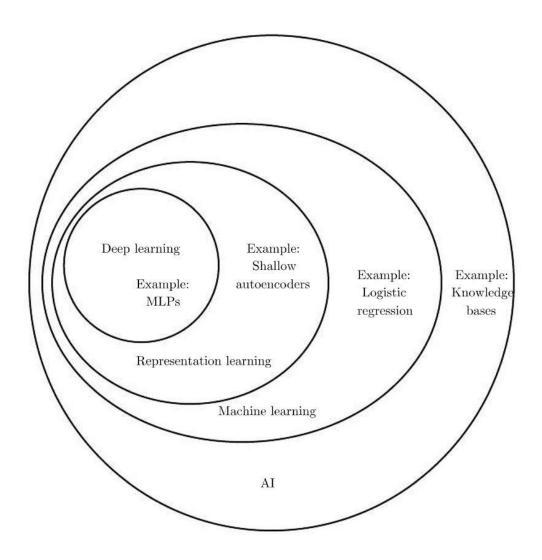
Ex) Spam mail filter,

E(Experience): Watching you label emails as spam or not spam.

T(Task): Classifying emails as spam or not spam.

P(Performance): The number (or fraction) of emails correctly classified as spam / not spam.





√Al

"How to get this informal knowledge into a computer!"



√Al

"How to get this informal knowledge into a computer!"

Knowledge base approach





None of these projects has led to a major success.



✓ Machine Learning

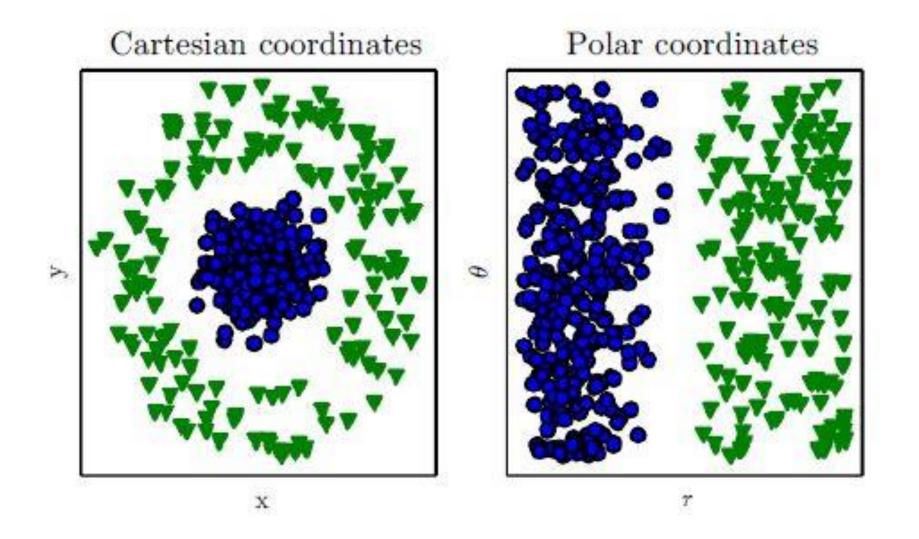
"Extract patterns from raw data."



✓ Machine Learning

"Difficult to know what **features** should be extracted."





✓ Representation Learning

"Discover not only the mapping from representation to output but also the representation itself."



✓ Representation Learning - Autoencoder

Autoencoder: encoder + decoder

Encoder: input data into a different representation

Decode: new representation back into the original format



✓ Representation Learning

"What if difficult to obtain a representation"



✓ Deep Learning

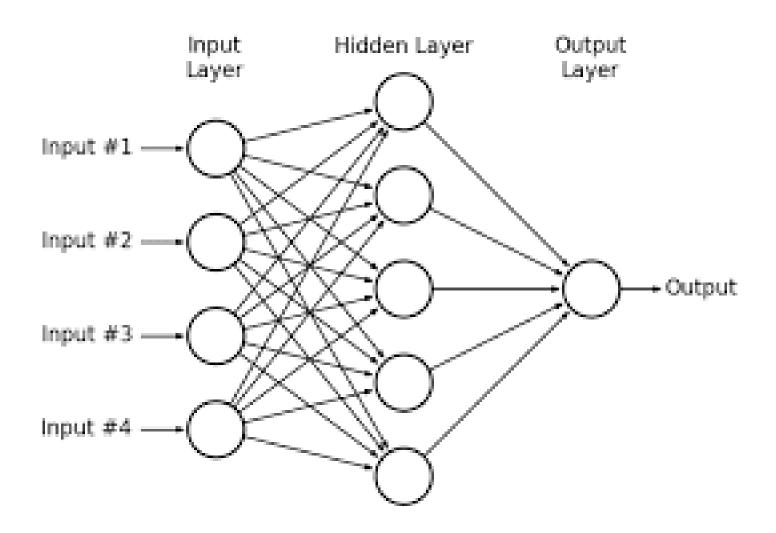
"Build complex concepts out of simpler concepts(representation)"

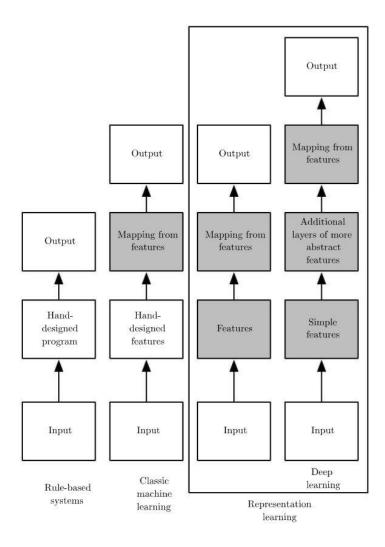


✓ Deep Learning - MLP (multilayer perceptron)

"Each function provides a new representation"







Learning

- ✓ Supervised Learning
- : Experience a dataset containing features, but each example is also associated with a label or target.
- ✓ Unsupervised Learning
- : Experience a dataset containing many features, then learn useful properties of the structure of this dataset.



✓ Supervised Learning

: Regression (Continuous Value)

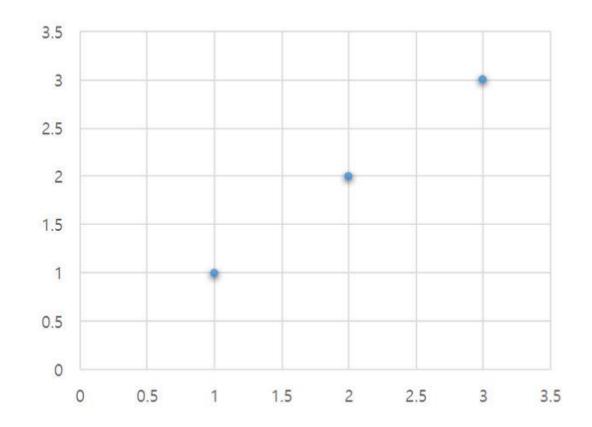
: Classification (Discrete Value)

In Deep Learning? CNN / RNN

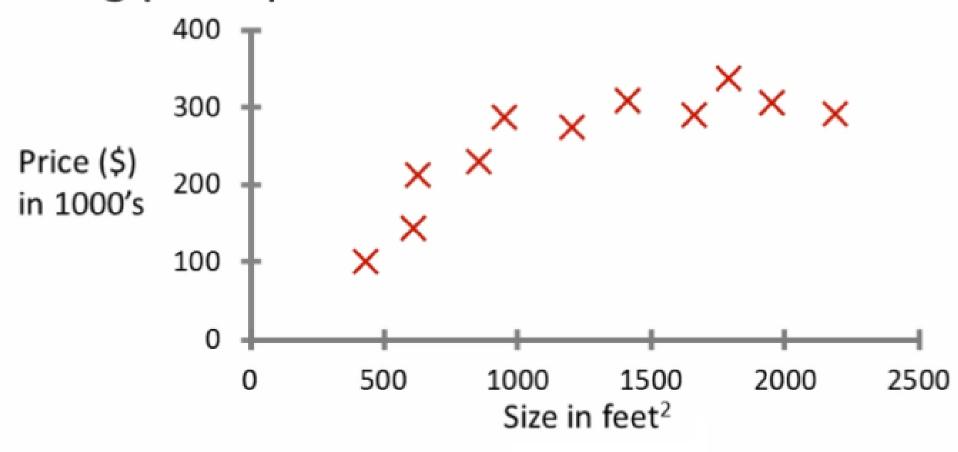


✓ Supervised Learning - Regression

x	У
1	1
2	2
3	3

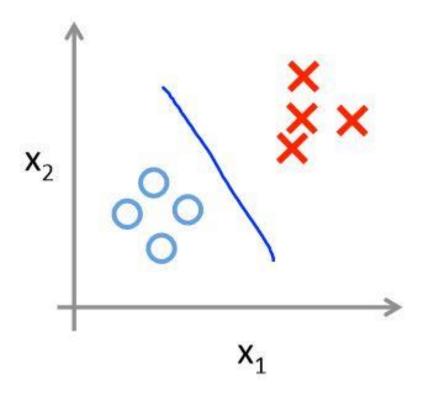


Housing price prediction.



✓ Supervised Learning - Classification

Binary classification:



Unsupervised Learning

✓ Unsupervised Learning

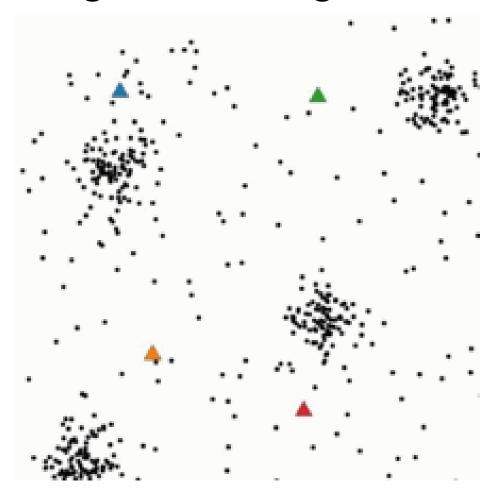
: Clustering

In Deep Learning? AutoEncoders

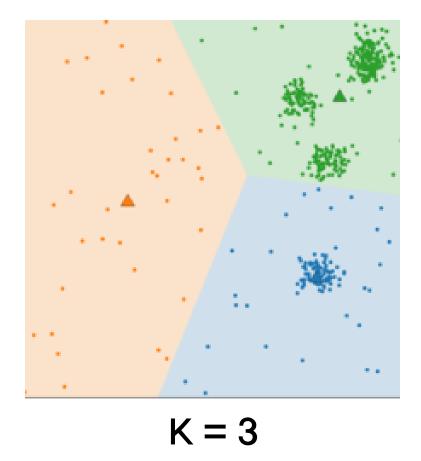


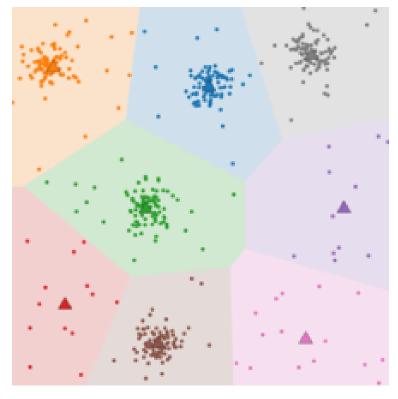
Learning

✓ Unsupervised Learning - Clustering



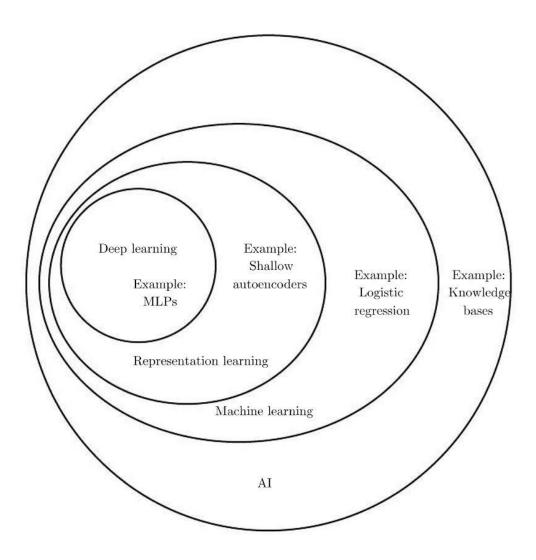
Learning





$$K = 8$$

Warm Up



Warm Up

✓ Supervised Learning

✓ Unsupervised Learning

Supervised Learning & HOL

Machine Learning

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Overview

1. Linear Regression

(HOL with Azure Machine Learning)

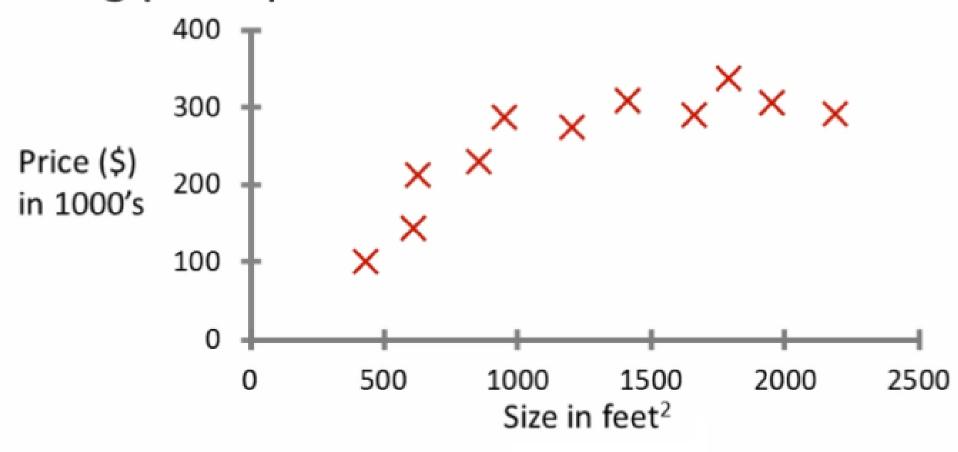
2. Neural Network

(HOL with Azure Notebook)

[Machine Learning]

- **✓** Regression
- : Linear Regression

Housing price prediction.



$$Y = W*X + b$$

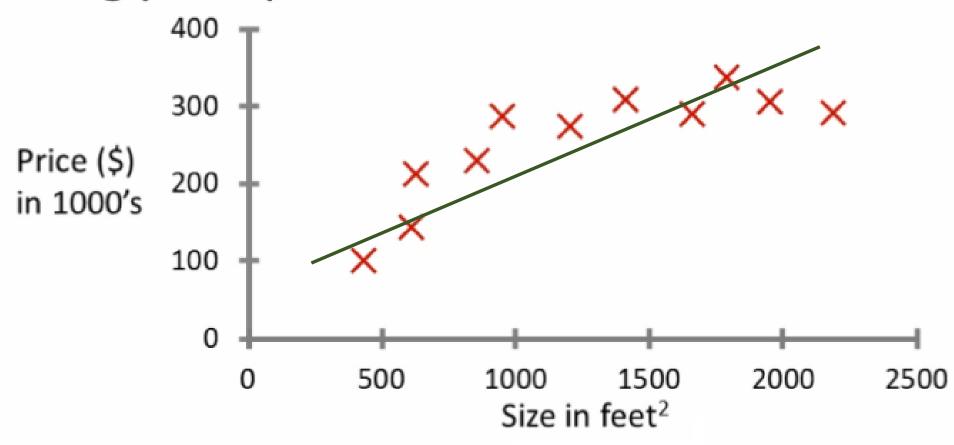
X: Y를 예측할 때 필요한 유의미한 피처 (집의 크기)

W, b: Y를 예측할 때 사용되는 각 피처에 대한 무게 값

Y: 학습된 모델을 통해 예측한 값



Housing price prediction.



$$Y = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$$

X: Y를 예측할 때 필요한 유의미한 피처 (집의 크기, 집의 층 수 등)

W, b: Y를 예측할 때 사용되는 각 피처에 대한 무게 값

Y: 학습된 모델을 통해 예측한 값



$$Y = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$$

그럼 우리가 학습을 통해 계산해야 하는 것은?



$$Y = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$$

계산해야 하는 W(weight) 는 어떻게 계산할까 ?



$$Y_{(predict)} = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$$

Y_predict: 학습된 모델을 통해 예측한 값

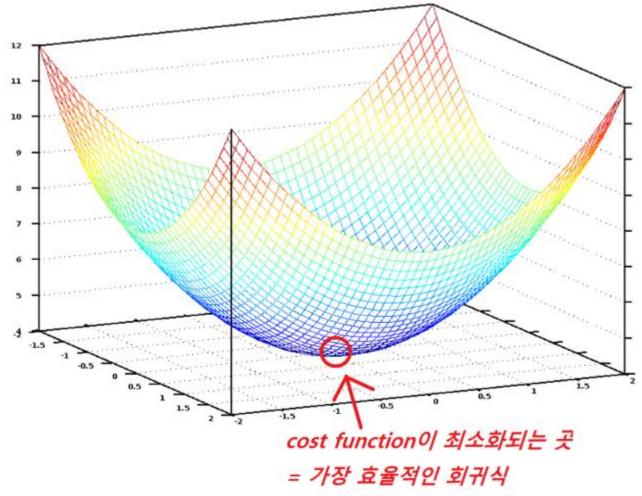
Y: 실제 값 (Label)



$$Y_{(predict)} = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$$



✓ Graph of Cost Function



✓ Gradient, Gradient Descent

✓ Derivative

: positive direction -> increase

: negative direction -> increase

✓ Partial Derivative

Suppose $f: \mathbb{R}^n \to \mathbb{R}$

The *partial* derivative of $f(x_1, ..., x_n)$ at x_i

$$\frac{\partial}{\partial x_i} f(x_1, \dots, x_n) = \lim_{h \to 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}$$

✓ Partial Derivative

Suppose $f: \mathbb{R}^n \to \mathbb{R}$

The *partial* derivative of $f(x_1, ..., x_n)$ at x_i

$$\frac{\partial}{\partial x_i} f(x_1, \dots, x_n) = \lim_{h \to 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}$$

 $\frac{\partial f}{\partial x_i}$ 방향으로 가면, x_i 축에서 증가하는 방향으로 향한다.



✓ Gradient

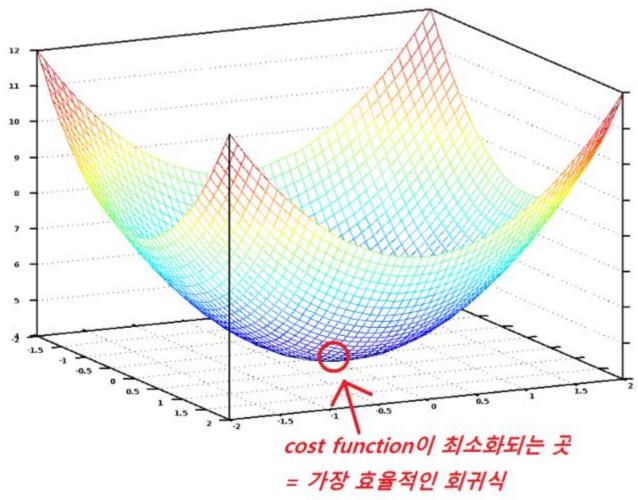
Suppose $f: \mathbb{R}^n \to \mathbb{R}$ Gradient $\nabla f(x)$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix}$$

- A vector containing the partial derivatives of the function f at certain point x
- Direction of steepest ascent (i.e., maximum increase of the function f) from x
 - Thus, $-\nabla f$ is the direction of steepest descent (maximum decrease)

$Y_{(predict)} = (W1*X1) + (W2*X2) + \cdots + (Wn*Xn) + b$ Cost Function : $(Y - Y_{(predict)}) \wedge 2$

✓ Gradient Descent

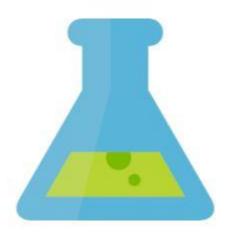


✓ Stochastic Gradient Descent

"Gradient Descent is computationally expensive"



√HOL



Automobile Price Data



√HOL

"Azure Machine Learning Studio"

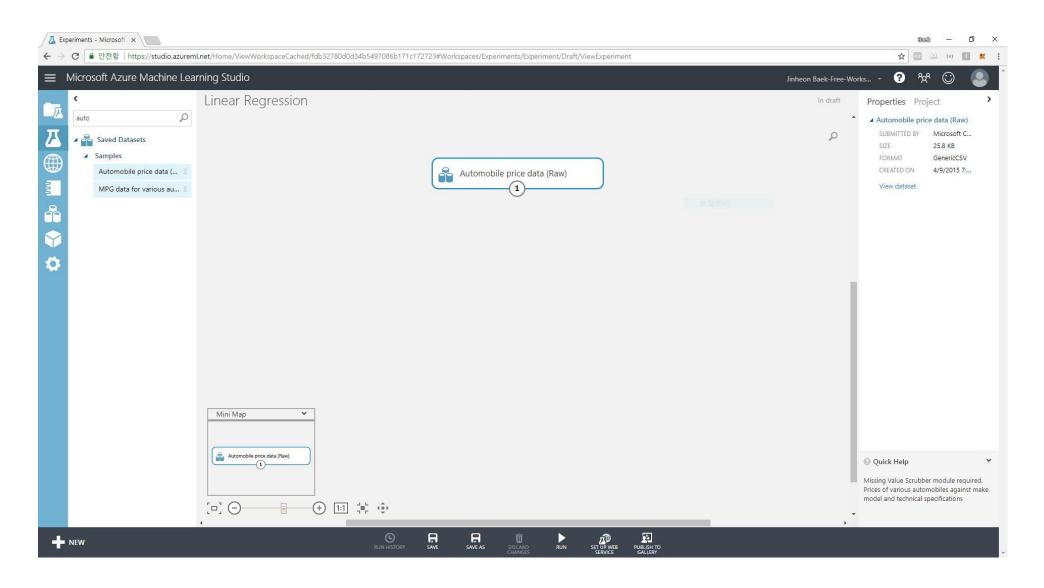
(https://studio.azureml.net/)



√HOL

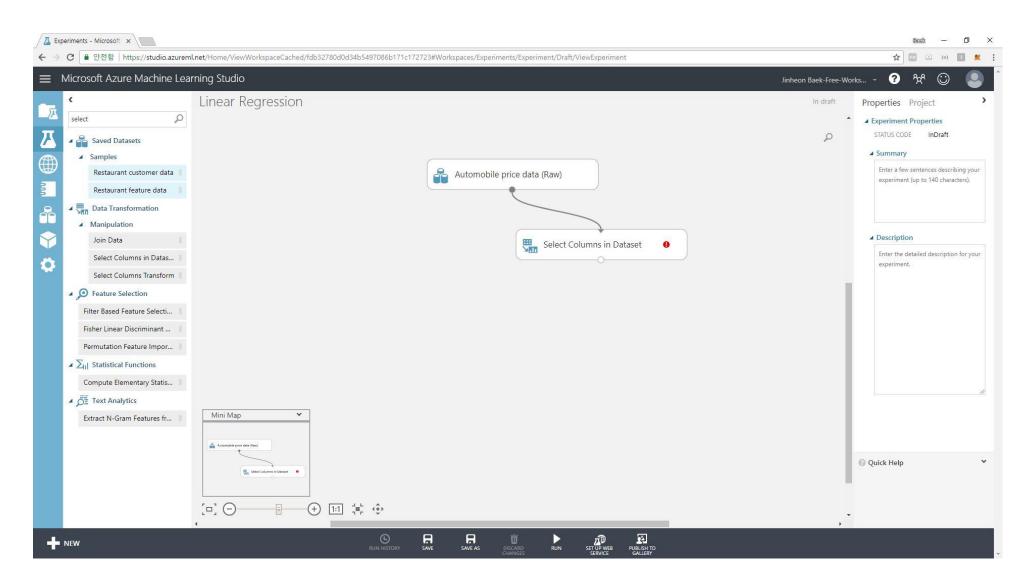
- 1) New
- 2) EXPERIMENT
- 3) Blank Experiment

검색창: Automobile price data 선택





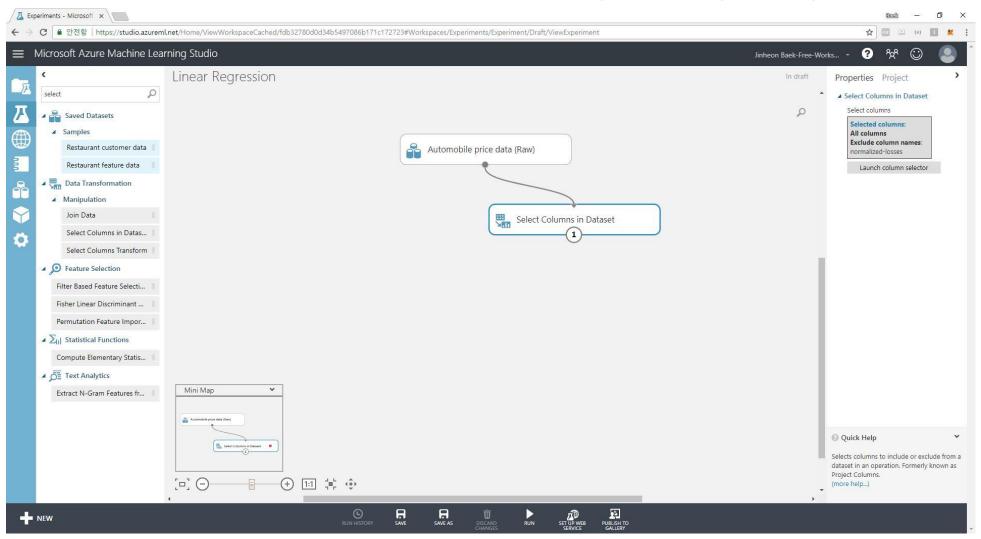
검색창: Select Columns in Dataset 선택 Automobile price data -> Select Columns ··· 이어주기





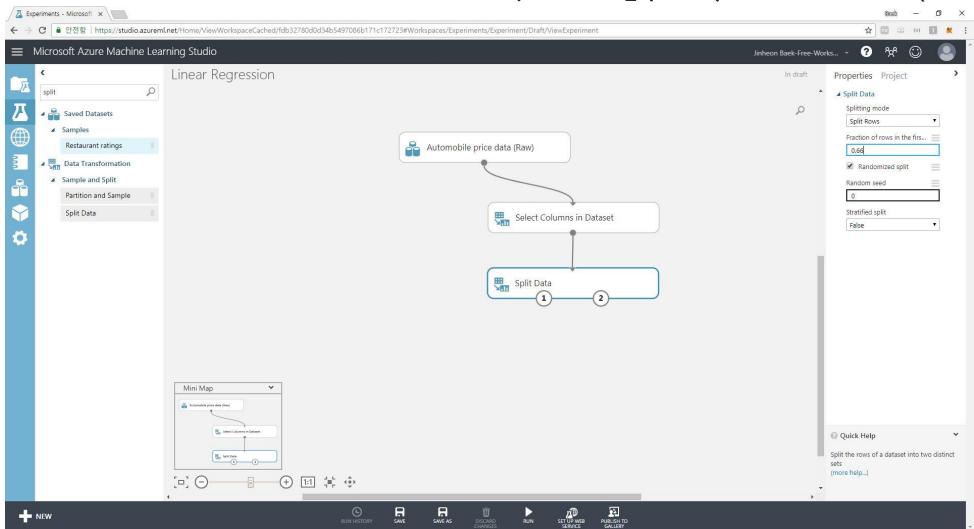
Select Columns in Dataset 클릭 (오른쪽 메뉴 확인) Launch column selector 클릭

With rules −⟩ All columns −⟩ Exclude −⟩ normalized-losses (column names)





검색창: Split Data
Select Columns in Dataset -> Split Data (연결)
Split Data 오른쪽 메뉴 -> Fraction of rows ... (0.66 설정)

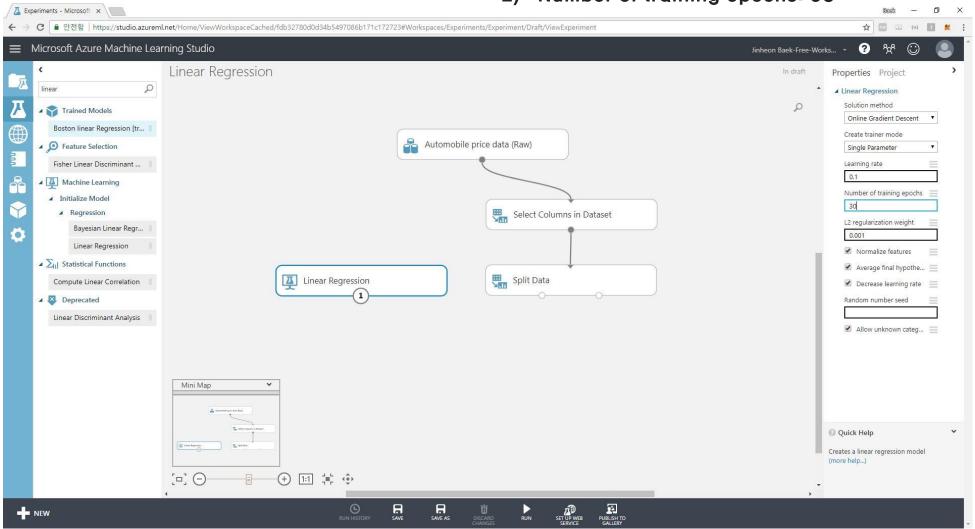




검색창: Linear Regression

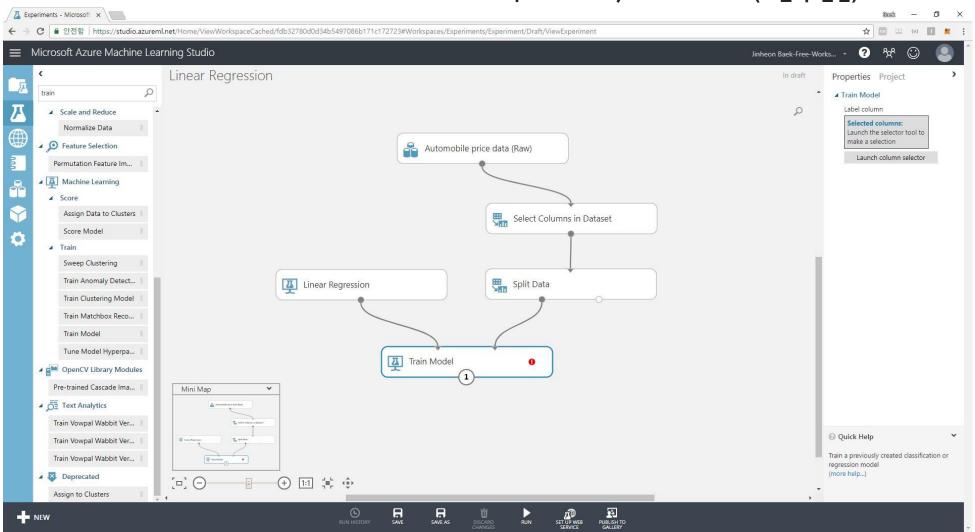
Linear Regression 오른쪽 메뉴 ->

- 1) Solution Method: Online Gradient Descent
- 2) Number of training epochs: 30



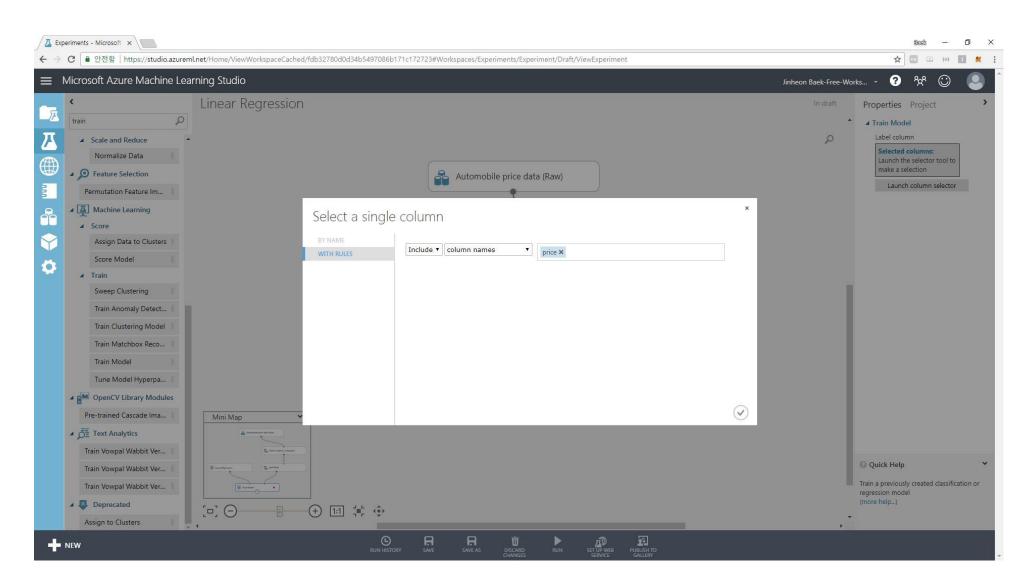


검색창: Train Model Linear Regression -> Train Model (왼쪽 연결) Split Data -> Train Model (오른쪽 연결)



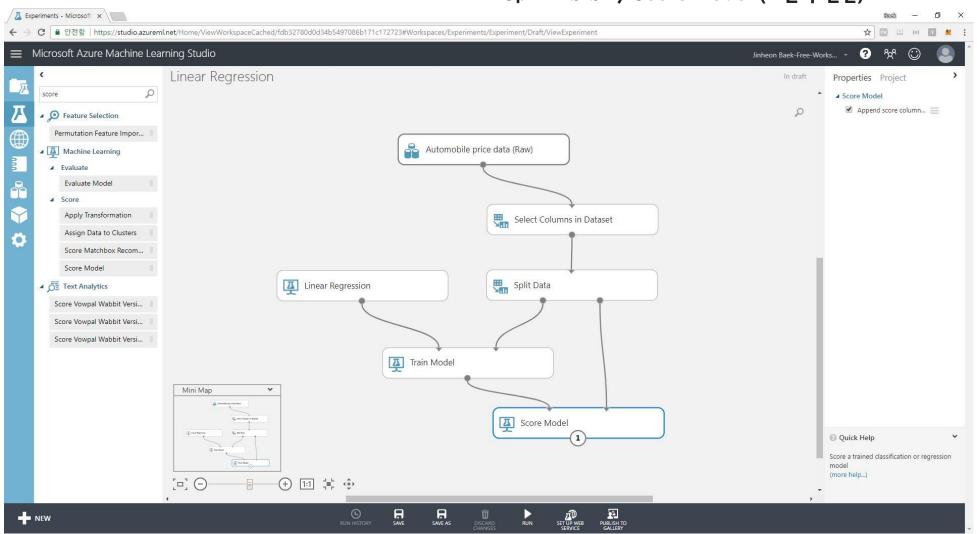


Train Model 오른쪽 메뉴 -> Launch column selector With Rules -> Include -> column names -> price



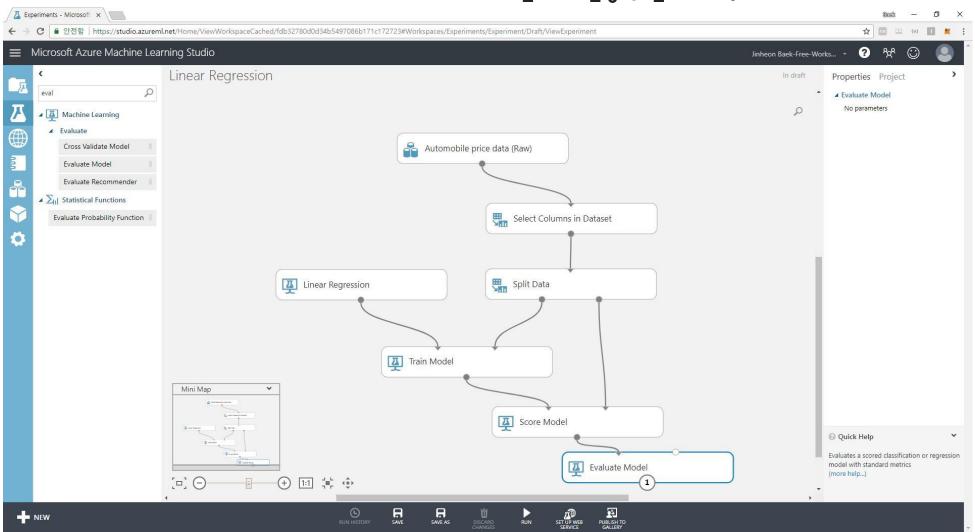


검색창: Score Model
Train Model -> Score Model (왼쪽 연결)
Split Data -> Score Model (오른쪽 연결)





검색창: Evaluate Model Score Model -> Evaluate Model (왼쪽 연결) 하단 Run 실행 후 결과 보기!



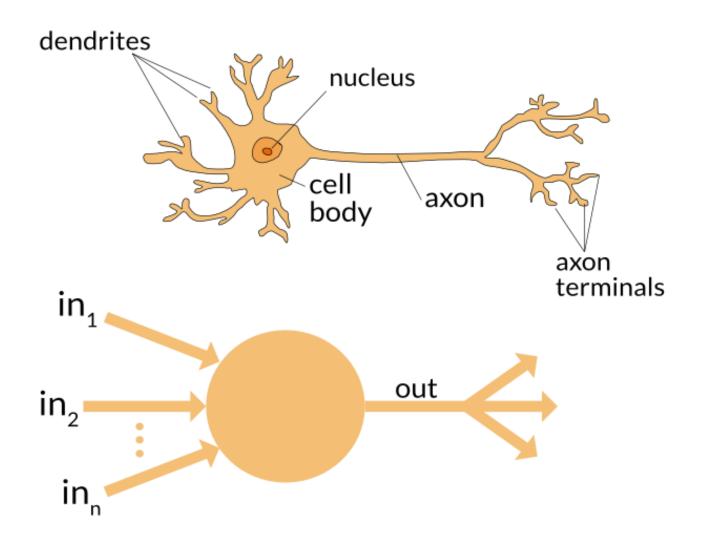


[Deep Learning]

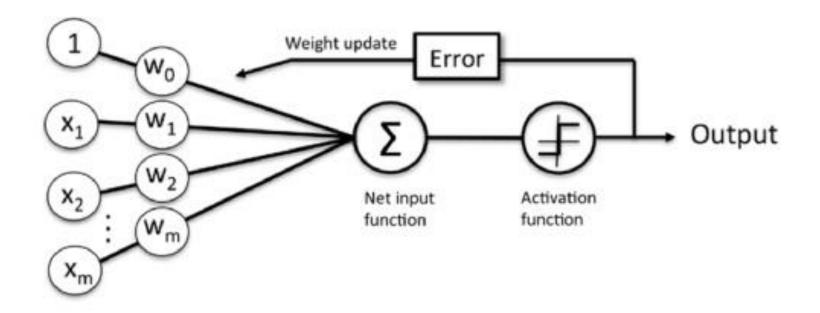
- ✓ Neural Network
- : Simple MLP (Multi-Layer Perceptron)

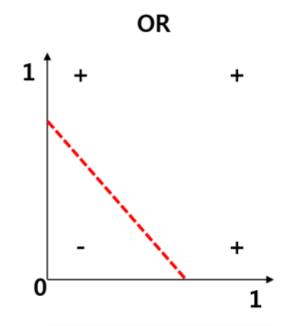
Note) 이건 mnist 로 하자!!!! 젠장, Neural Network 설명도 해야해…



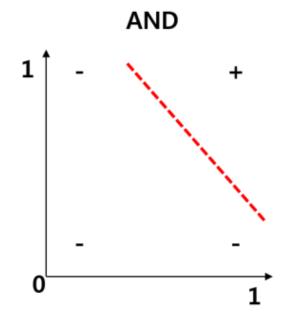


Single Layer Neural Network

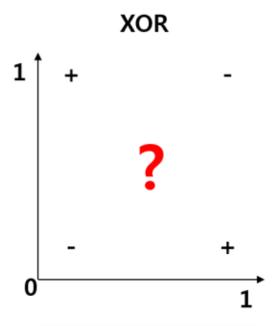




x_1	x_2	у
0	0	0
0	1	1
1	0	1
1	1	1

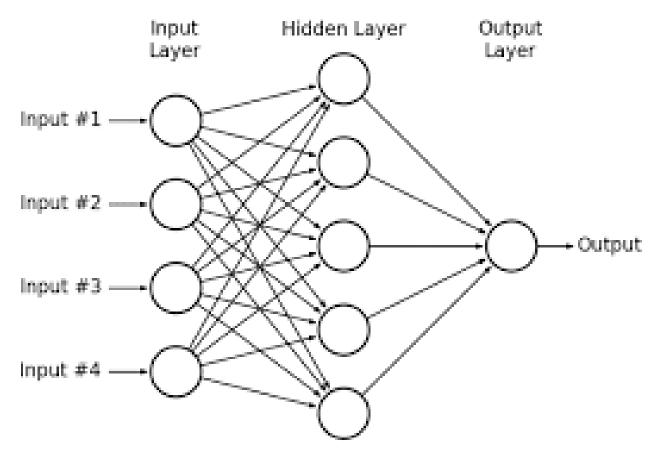


x_1	x_2	y
0	0	0
0	1	0
1	0	0
1	1	1

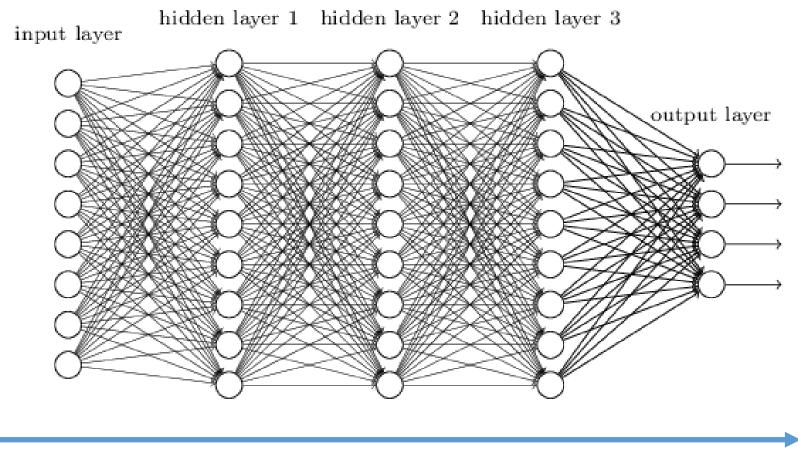


x_1	x_2	y
0	0	0
0	1	1
1	0	1
1	1	0

Multi-Layer Neural Network



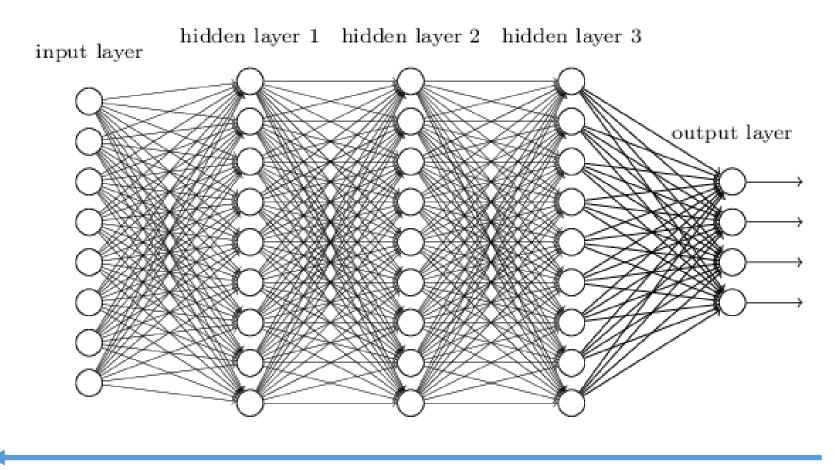
Neural-network



결과 예측



Neural-network



Layer 값 조정



✓HOL

"MNIST Data Set?"

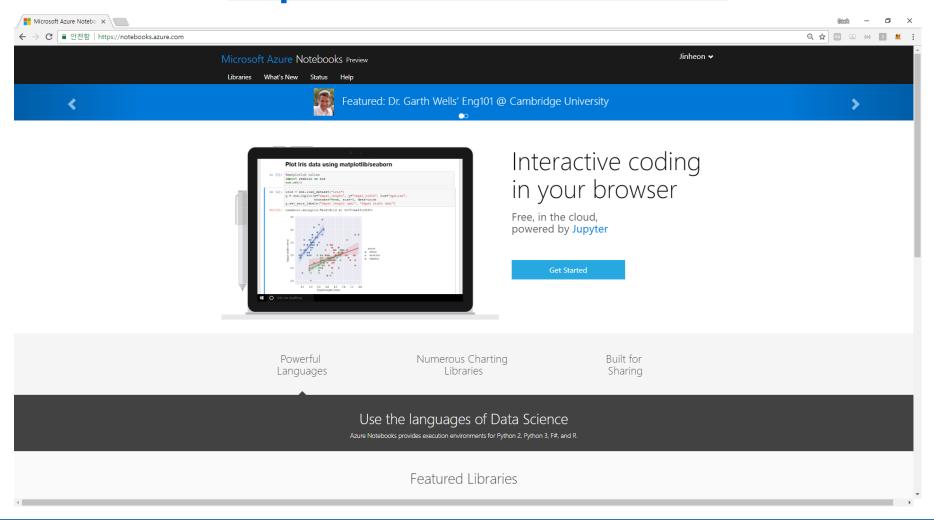


0000000000000000 / 1 | | / 1 | / 7 1 | / / / / | 222222222222 555555555555555 6666666666666 ファチ17ァファファファファ 8888888888888888 9999999999999

28 x 28, 필기체 숫자 (O ~ 9)



✓ Azure Notebook (https://notebooks.azure.com)

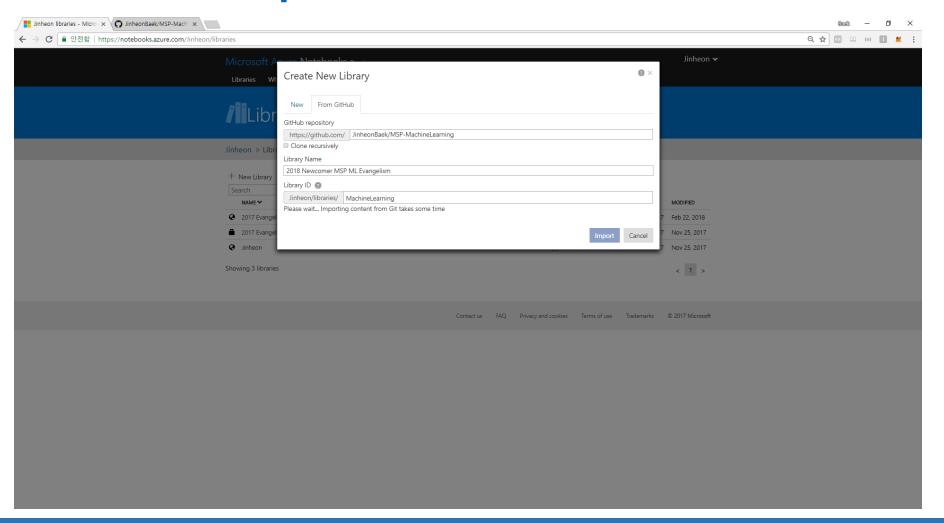




- ✓ Azure Notebook (https://notebooks.azure.com)
- 1. Logic
- 2. Libraries 메뉴
- 3. New Library

GitHub repository: JinheonBaek/MSP-MachineLearning Library Name (예시): 2018 Newcomer MSP ML Evangelism Library ID (예시): MachineLearning

✓Azure Notebook (https://notebooks.azure.com)





Neural Network

✓ HOL - Challenge

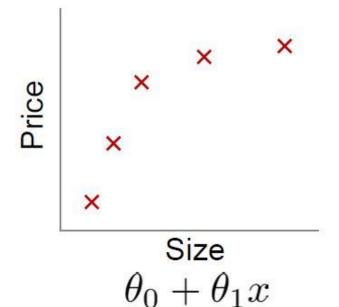
"Testing score is above 0.98"



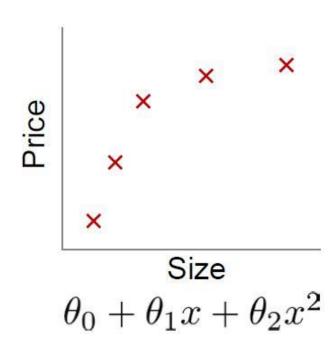
FittingMachine Learning

Microsoft Student Partner 백진현

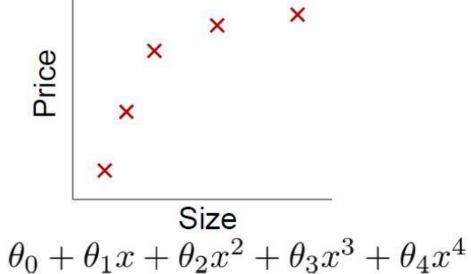
√Fitting



High bias (underfit)



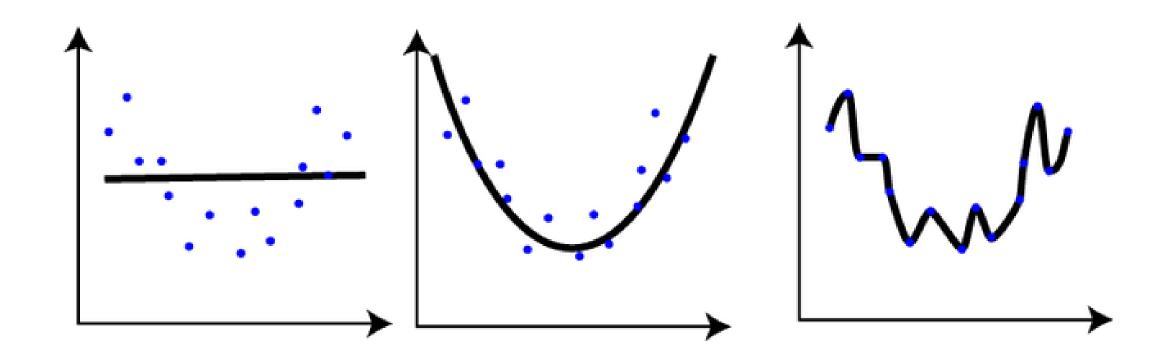
"Just right"



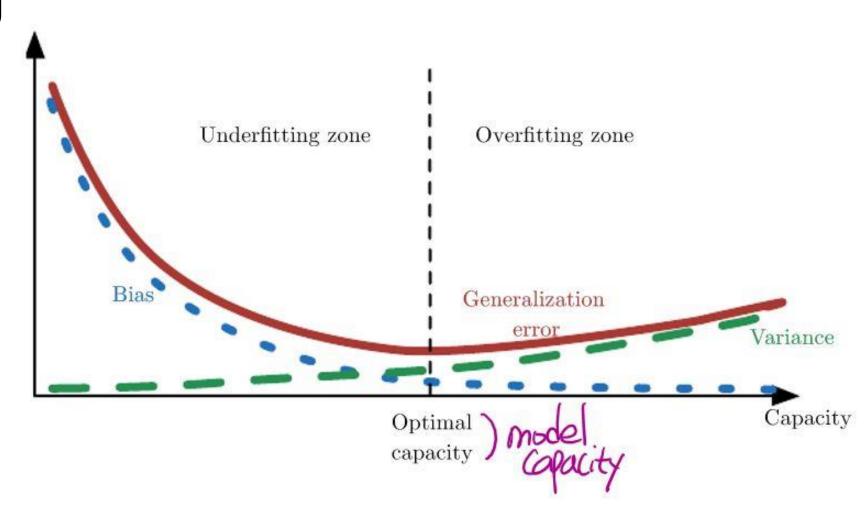
High variance (overfit)



√Fitting



√ Fitting





✓ Fitting HOL

"Connect Four"



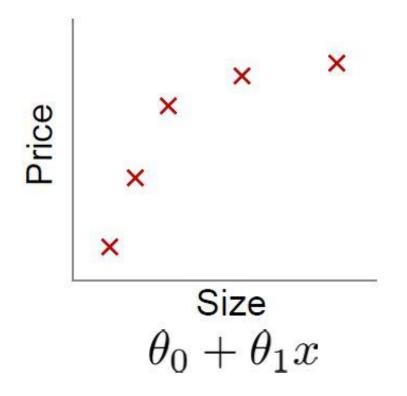
Machine Learning

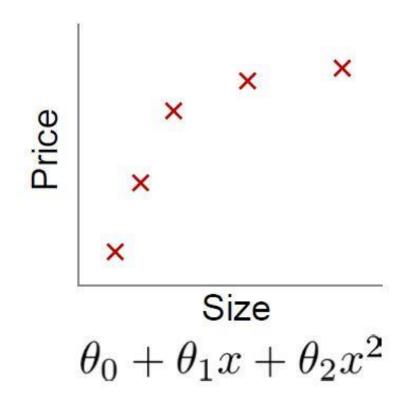
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백진헌

- ✓ Linear vs Nonlinear
- **✓** Black box vs Descriptive
- √ First-principle vs Data-driven

✓ Linear vs Nonlinear





✓ Blackbox vs Descriptive Model

Linear Regression

: Models are descriptive, because one can see which variables are weighed heaviest.

Neural Network

: Models are generally opaque.



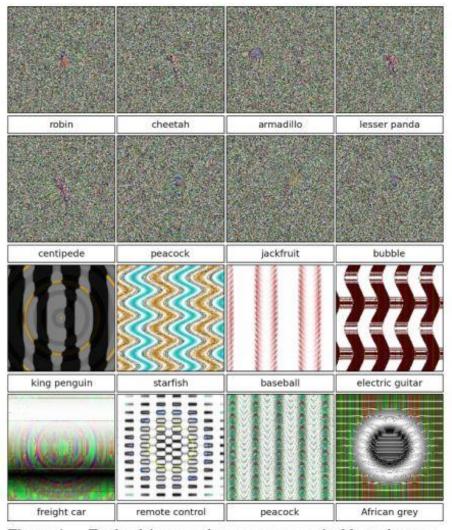
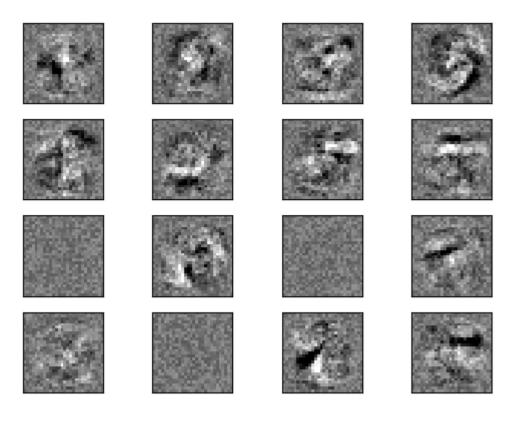


Figure 1. Evolved images that are unrecognizable to humans, but that state-of-the-art DNNs trained on ImageNet believe with $\geq 99.6\%$ certainty to be a familiar object. This result highlights differences between how DNNs and humans recognize objects. Images are either directly (top) or indirectly (bottom) encoded.





√ First-principle vs Data-driven

First-principle

: Models based on a theoretical explanation.

Data-driven

: Models based on observed data correlations.



References

- ✓ Deep Learning Book MIT
- ✓ Machine Learning Coursera
- √ Korea Univ. Cose432 (Data Science)



Thanks

