

ML/DL for Everyone Season2

with  TensorFlow

04 - Multi-variable linear regression

다변수 선형 회기

Code: <https://github.com/deeplearningzerotoall/TensorFlow>

Slides: <http://bit.ly/2LQMKvk>

Lecturer: 이승준 plusjune@financedata.kr



Recap

- Hypothesis
- Cost function
- Gradient descent

Recap

- Hypothesis
예측

$$H(x) = Wx + b$$

- Cost function
예측과 실제의 차이

cost가 되는 W 를 찾아가는 과정

$$cost(W) = \frac{1}{m} \sum_{i=1}^m (Wx_i - y_i)^2$$

-값이 나오지 않도록 제공

- Gradient descent

$$W := W - \alpha \frac{1}{m} \sum_{i=1}^m (W(x_i) - y_i)x_i$$

Predicting exam score: regression using one input (x)

One-variable
(One-feature)

x (hours)	y (score)
10	90
9	80
3	50
2	60
11	40

Predicting exam score:

regression using three inputs (x_1 , x_2 , x_3)

Multi-variable (Multi-feature)

x_1 (quiz 1)	x_2 (quiz 2)	x_3 (midterm 1)	y (final)
73	80	75	152
93	88	93	185
89	91	90	180
96	98	100	196
73	66	70	142

Hypothesis

$$H(x) = Wx + b$$

Hypothesis

$$H(x) = Wx + b$$

$$H(x_1, x_2, x_3) = w_1 x_1 + w_2 x_2 + w_3 x_3 + b$$

변수가 늘어난 개수만큼 가중치(x)의 개수도 늘어나

Cost function

$$H(x_1, x_2, x_3) = w_1 x_1 + w_2 x_2 + w_3 x_3 + b$$

$$\text{cost}(W, b) = \frac{1}{m} \sum_{i=1}^m (H(x_1, x_2, x_3) - y_i)^2$$

Multi-variable

$$H(x_1, x_2, x_3) = w_1 x_1 + w_2 x_2 + w_3 x_3 + b$$

$$H(x_1, x_2, x_3, \dots, x_n) = w_1 x_1 + w_2 x_2 + w_3 x_3 + \dots + w_n x_n + b$$

Matrix

$$w_1 x_1 + w_2 x_2 + w_3 x_3 + \dots + w_n x_n$$

Matrix multiplication

메트릭스의 곱
"Dot Product"

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \times \begin{bmatrix} 7 & 8 \\ 9 & 10 \\ 11 & 12 \end{bmatrix} = \begin{bmatrix} 58 \\ \end{bmatrix}$$

Hypothesis using matrix

$$w_1 x_1 + w_2 x_2 + w_3 x_3 + \dots + w_n x_n$$

$$(x_1 \quad x_2 \quad x_3) \cdot \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = (x_1 w_1 + x_2 w_2 + x_3 w_3)$$

$$H(X) = XW$$

일반적으로 메트릭스는 대문자

행과 열을 연산하기 때문에 X가 앞에 온다

Hypothesis using matrix

x_1	x_2	x_3	y
73	80	75	152
93	88	93	185
89	91	90	180
96	98	100	196
73	66	70	142

Test Scores for General Psychology

$$H(x_1, x_2, x_3) = w_1 x_1 + w_2 x_2 + w_3 x_3$$

Hypothesis using matrix

x_1	x_2	x_3	y
73	80	75	152
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Many x instances

x_1	x_2	x_3	y
73	80	75	152
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Test Scores for General Psychology

$$(x_1 \quad x_2 \quad x_3) \cdot \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = (x_1 w_1 + x_2 w_2 + x_3 w_3)$$

x_1	x_2	x_3	y
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Hypothesis using matrix

매트릭스를 사용하면 데이터의 개수가 많은 적든 상관 없다

$$w_1 x_1 + w_2 x_2 + w_3 x_3 + \dots + w_n x_n$$

$$\begin{matrix} \downarrow \end{matrix} \begin{pmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \\ x_{41} & x_{42} & x_{43} \\ x_{51} & x_{52} & x_{53} \end{pmatrix} \cdot \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = \begin{pmatrix} x_{11}w_1 + x_{12}w_2 + x_{13}w_3 \\ x_{21}w_1 + x_{22}w_2 + x_{23}w_3 \\ x_{31}w_1 + x_{32}w_2 + x_{33}w_3 \\ x_{41}w_1 + x_{42}w_2 + x_{43}w_3 \\ x_{51}w_1 + x_{52}w_2 + x_{53}w_3 \end{pmatrix}$$

$$H(X) = XW$$

Hypothesis using matrix

$$\begin{pmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \\ x_{41} & x_{42} & x_{43} \\ x_{51} & x_{52} & x_{53} \end{pmatrix} \cdot \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = \begin{pmatrix} x_{11}w_1 + x_{12}w_2 + x_{13}w_3 \\ x_{21}w_1 + x_{22}w_2 + x_{23}w_3 \\ x_{31}w_1 + x_{32}w_2 + x_{33}w_3 \\ x_{41}w_1 + x_{42}w_2 + x_{43}w_3 \\ x_{51}w_1 + x_{52}w_2 + x_{53}w_3 \end{pmatrix}$$

[5, 3]

[3, 1]

[5, 1]

동일해야 함

$$H(X) = XW$$

Hypothesis using matrix

$$\begin{pmatrix} \mathbf{X} \end{pmatrix} \times \begin{pmatrix} \mathbf{W} \end{pmatrix} = \begin{pmatrix} \mathbf{H}(\mathbf{X}) \end{pmatrix}$$

$[5, 3] \quad \quad [?, ?] \quad \quad [5, 1]$

$$H(X) = XW$$

Hypothesis using matrix

$$\begin{pmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \\ x_{41} & x_{42} & x_{43} \\ x_{51} & x_{52} & x_{53} \end{pmatrix} \cdot \begin{pmatrix} w_1 \\ w_2 \\ w_3 \end{pmatrix} = \begin{pmatrix} x_{11}w_1 + x_{12}w_2 + x_{13}w_3 \\ x_{21}w_1 + x_{22}w_2 + x_{23}w_3 \\ x_{31}w_1 + x_{32}w_2 + x_{33}w_3 \\ x_{41}w_1 + x_{42}w_2 + x_{43}w_3 \\ x_{51}w_1 + x_{52}w_2 + x_{53}w_3 \end{pmatrix}$$

$[n, 3]$

동일

$[3, 1]$

$[n, 1]$

$$H(X) = XW$$

Hypothesis using matrix (n output)

$$\begin{pmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \\ x_{41} & x_{42} & x_{43} \\ x_{51} & x_{52} & x_{53} \end{pmatrix} \cdot \text{?} = \begin{pmatrix} x_{11}w_{11} + x_{12}w_{21} + x_{13}w_{31} & x_{11}w_{12} + x_{12}w_{22} + x_{13}w_{32} \\ x_{21}w_{11} + x_{22}w_{21} + x_{23}w_{31} & x_{21}w_{12} + x_{22}w_{22} + x_{23}w_{32} \\ x_{31}w_{11} + x_{32}w_{21} + x_{33}w_{31} & x_{31}w_{12} + x_{32}w_{22} + x_{33}w_{32} \\ x_{41}w_{11} + x_{42}w_{21} + x_{43}w_{31} & x_{41}w_{12} + x_{42}w_{22} + x_{43}w_{32} \\ x_{51}w_{11} + x_{52}w_{21} + x_{53}w_{31} & x_{51}w_{12} + x_{52}w_{22} + x_{53}w_{32} \end{pmatrix}$$

Weight의 크기는 입력 데이터의 column 개수와 출력 데이터와 column 개수에 따라 결정

[n, 3]

[?, ?]

[n, 2]

변수(feature)의 개수, ^{3, 2}출력의 개수

$$H(X) = XW$$

Hypothesis using matrix (n output)

$$\begin{pmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \\ x_{41} & x_{42} & x_{43} \\ x_{51} & x_{52} & x_{53} \end{pmatrix} \cdot \begin{pmatrix} w_{11} & w_{12} \\ w_{21} & w_{22} \\ w_{31} & w_{32} \end{pmatrix} = \begin{pmatrix} x_{11}w_{11} + x_{12}w_{21} + x_{13}w_{31} & x_{11}w_{12} + x_{12}w_{22} + x_{13}w_{32} \\ x_{21}w_{11} + x_{22}w_{21} + x_{23}w_{31} & x_{21}w_{12} + x_{22}w_{22} + x_{23}w_{32} \\ x_{31}w_{11} + x_{32}w_{21} + x_{33}w_{31} & x_{31}w_{12} + x_{32}w_{22} + x_{33}w_{32} \\ x_{41}w_{11} + x_{42}w_{21} + x_{43}w_{31} & x_{41}w_{12} + x_{42}w_{22} + x_{43}w_{32} \\ x_{51}w_{11} + x_{52}w_{21} + x_{53}w_{31} & x_{51}w_{12} + x_{52}w_{22} + x_{53}w_{32} \end{pmatrix}$$

[n, 3]

[3, 2]

[n, 2]

$$H(X) = XW$$

입력 개수와 출력 개수에 상관없이 우리의 가설(H(x))은 항상 XW로 변함이 없다

WX vs XW

- Lecture (theory)

$$H(x) = Wx + b$$

왼쪽의 $Wx + b$ 와 동일한 의미

$$h_{\theta}(x) = \theta_1 x + \theta_0$$

$$f(x) = ax + b$$

- Implementation (TensorFlow)

$$H(X) = XW$$

텐서플로우 코드에서는 matrix의 곱으로 XW 로 씀

What's Next?

- Logistic (Regression) Classification