

머신러닝 스터디 3rd week

보조 자료

20180119 김성현

summary

- ch.5 Logistic (regression) classifier

- logistic hypothesis

$$H(X) = \frac{1}{1 + e^{-(W^T X)}}$$

- cost

$$\text{cost}(W) = -\frac{1}{m} \sum y \log(H(x)) + (1 - y) \log(1 - H(x))$$

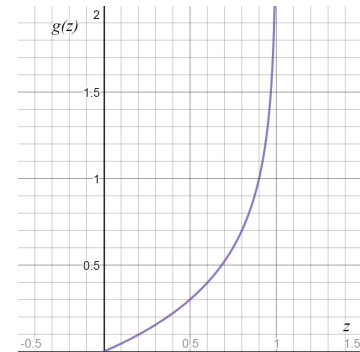
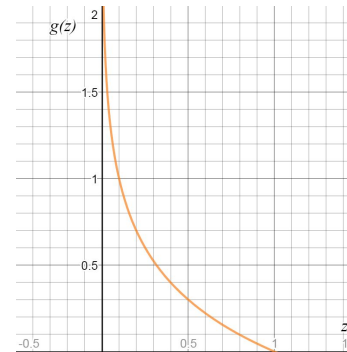
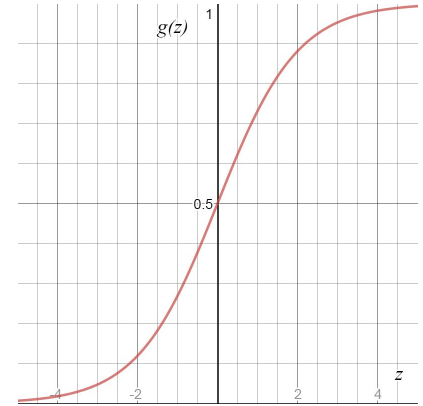
- Gradient descent algorithm

$$W := W - \alpha \frac{\partial}{\partial W} \text{cost}(W)$$

```
# cost function
cost = tf.reduce_mean(-tf.reduce_sum(Y*tf.log(hypothesis) + (1-Y)*tf.log(1-hypothesis)))

# Minimize
a = tf.Variable(0.1) # Learning rate, alpha
optimizer = tf.train.GradientDescentOptimizer(a)
train = optimizer.minimize(cost)
```

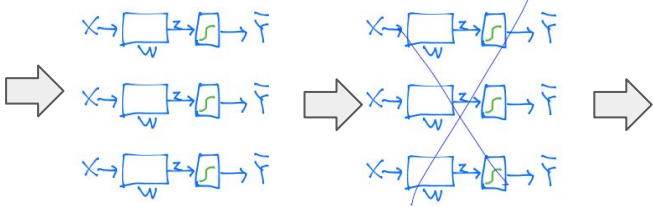
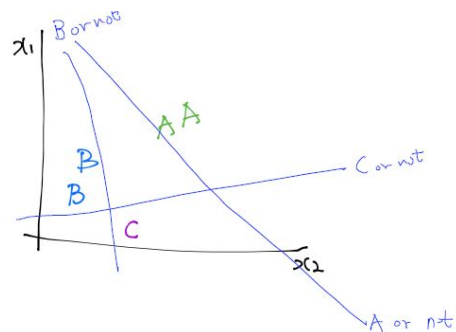
$$g(z) = \frac{1}{(1 + e^{-z})}$$



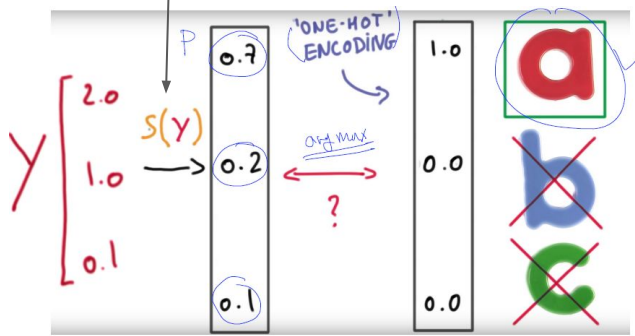
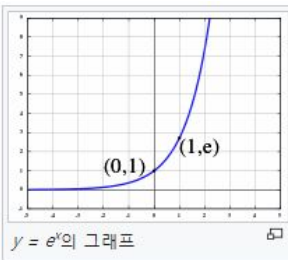
$$\text{cost}(H(x), y) = \begin{cases} -\log(H(x)) & : y = 1 \\ -\log(1 - H(x)) & : y = 0 \end{cases}$$

summary

- ch.6 Softmax Classifier



$$S(y_i) = \frac{e^{y_i}}{\sum_j e^{y_j}}$$



Cost function

Hypothesis

$$s(w \cdot x_i + b)$$

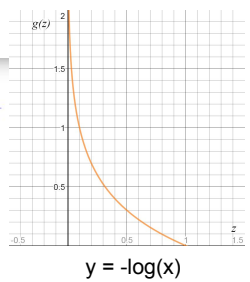
LOSS

$$\mathcal{L} = \frac{1}{N} \sum_i \mathcal{D}(s(w \cdot x_i + b), L_i)$$

TRAINING SET

CROSS-ENTROPY

$$\mathcal{D}(S, L) = - \sum_i L_i \log(S_i)$$



Gradient descent

$$-\alpha \Delta \mathcal{L}(w_1, w_2)$$

DERIVATIVE

decision boundary

- tensorflow 사용한 decision boundary
 - <https://mubaris.com/2017/10/21/tensorflow-101/>
- Neural Network decision boundary
 - <http://www.wildml.com/2015/09/implementing-a-neural-network-from-scratch/>
- Some Deep Learning with Python, TensorFlow and Keras
 - <https://sandipanweb.wordpress.com/2017/11/25/some-deep-learning-with-python-tensorflow-and-keras/>

