# Forward Propagation 例人 MNIST Classification Model

# 수업 목표

# 이번 수업의 핵심:

- MNIST Dataset 구성
- Neural Network 기반 MNIST Classification 모델
- Forward Propagation 코드 작성
- Error/Loss의 개념 확인

# 핵심 개념

- MNIST Dataset
- Classification Model
- Squared Error / Loss

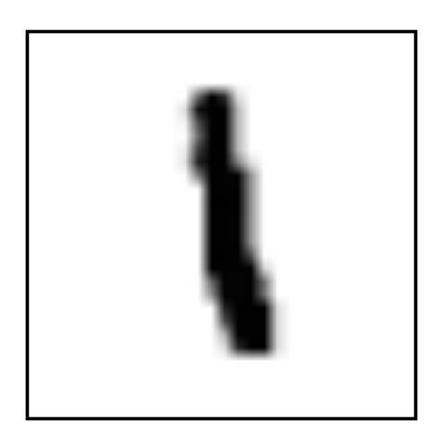
### **MNIST Dataset**

MNIST (Modified National Institute of Standards and Technology)

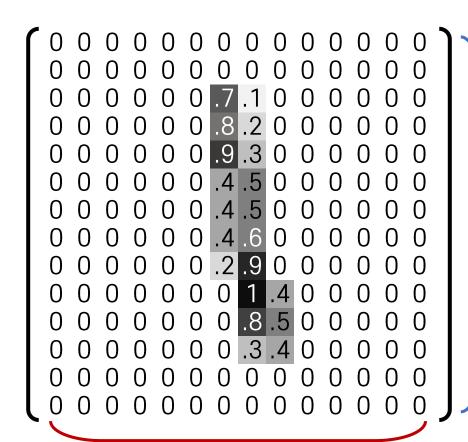
```
000000000000000
/11/11/11/11/11/11
2222222222222
5555555555555555
6666666666666
ファチ17ククフフフフフフ)ク
888888888888888888
99999999999
```

- 이에서 9까지 손글씨 숫자 사진
  - 55,000개의 Training Examples
  - 10,000개의 Test Examples
- 각 숫자 사진들은 전처리됨
  - 숫자가 사진의 중앙에 정렬
  - 숫자가 비슷한 크기로 조절
  - 각 사진이 28 × 28 픽셀 크기로 고정
  - → 0부터 1사이의 실수 행렬

# MNIST 예시

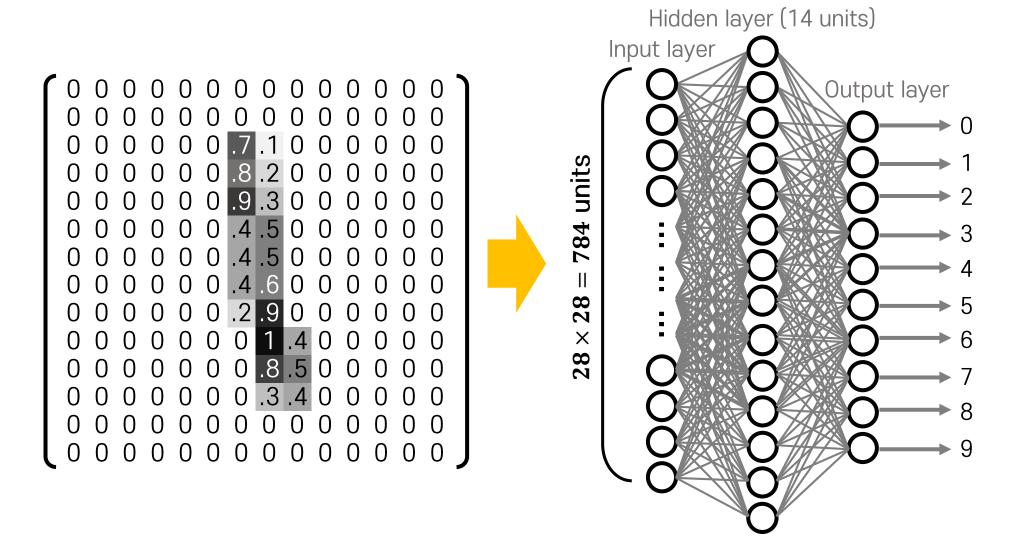


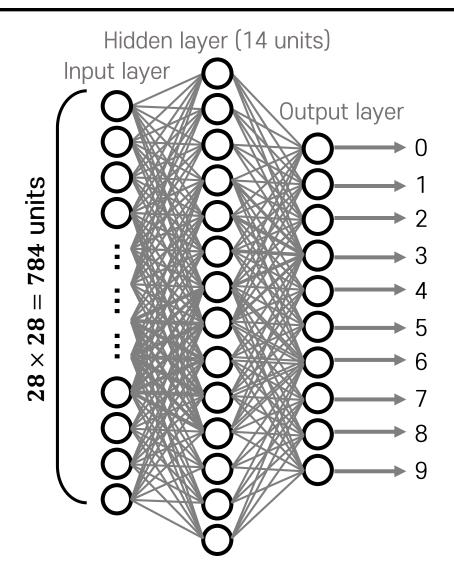




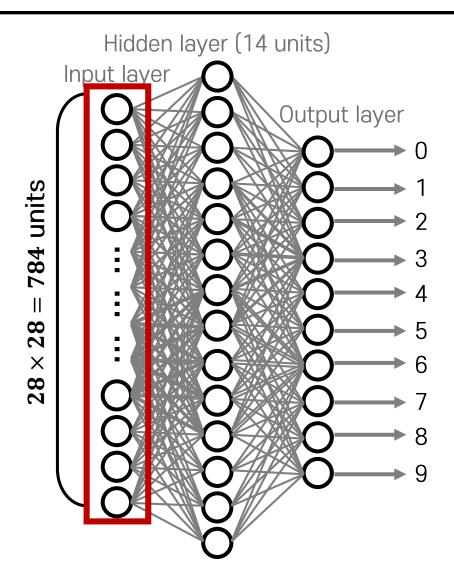
28

28

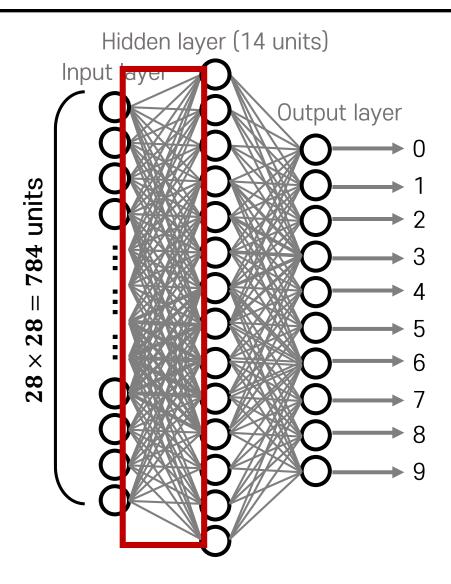




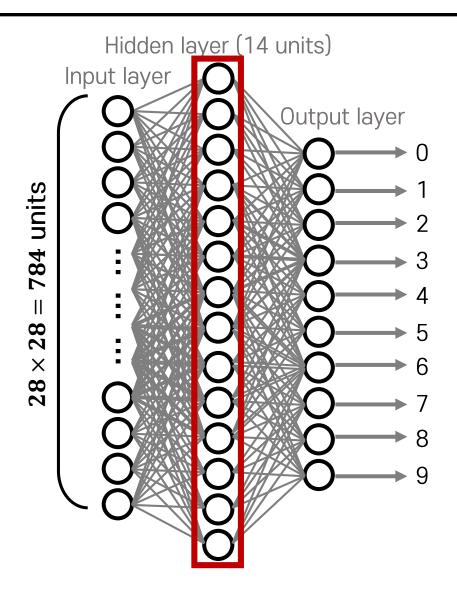
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import numpy as np
def sigmoid(x):
    return 1 / (1 + np.exp(-x))
def neural_network(x, w1, w2):
    # x: 784 x 1
    # w1: 14 x (784 + 1)
    # w2: 10 \times (14 + 1)
    a1 = np.concatenate([x, [[1]]], 0)
    z2 = np.dot(w1, a1)
    a2 = sigmoid(z2)
    a2 = np.concatenate([a2, [[1]]], 0)
    z3 = np.dot(w2, a2)
    a3 = sigmoid(z3)
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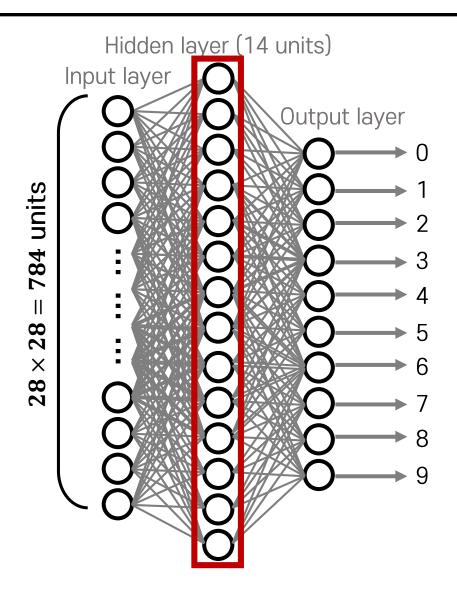
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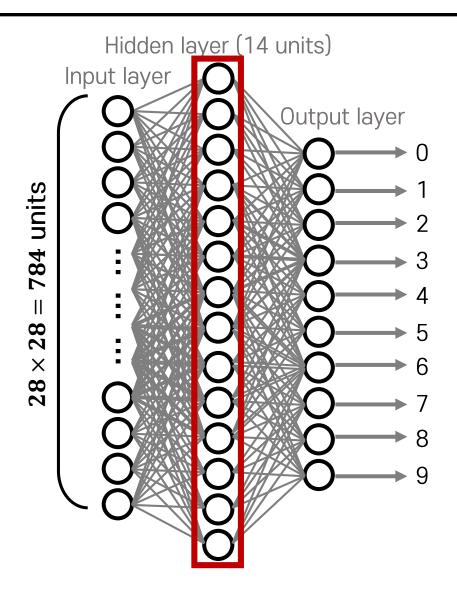
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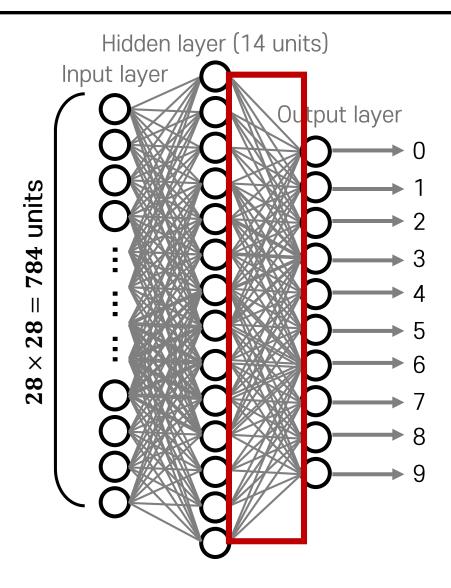
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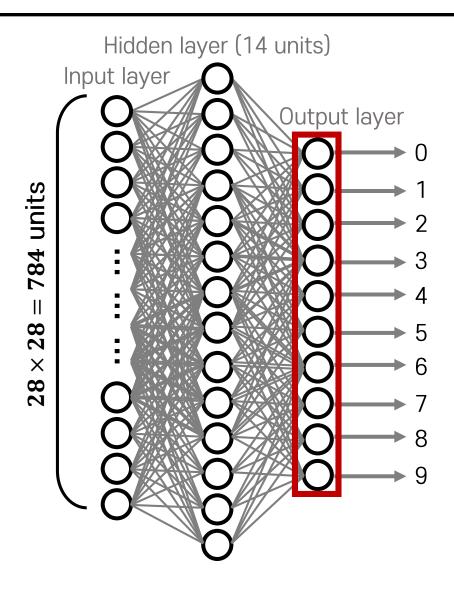
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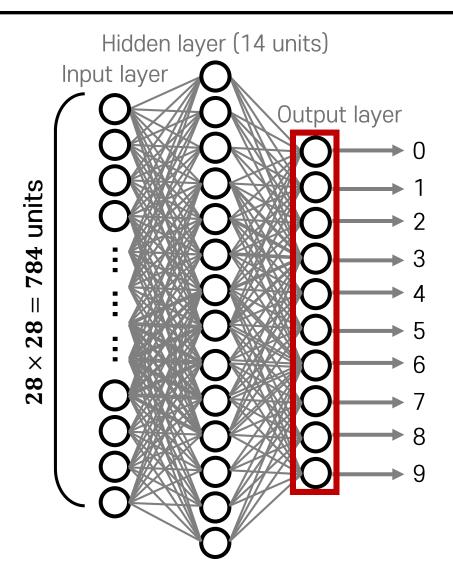
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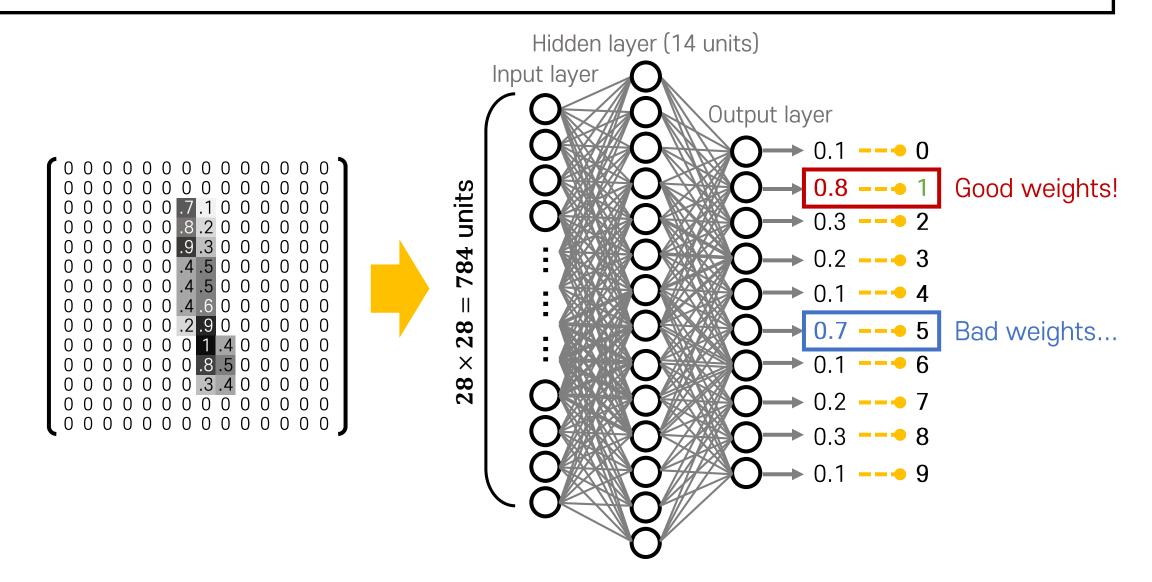
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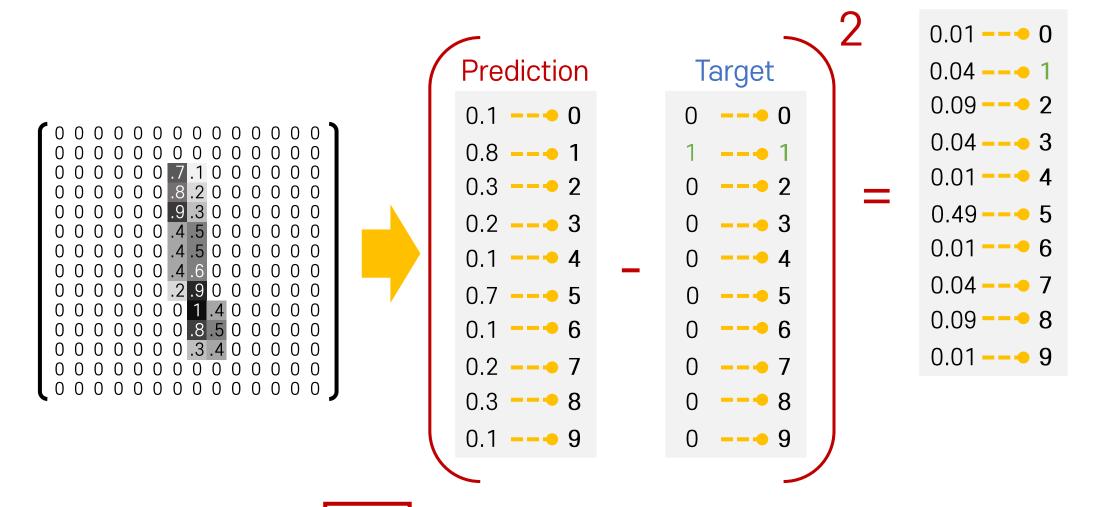


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Squared Error (Loss): 0.83 Error/Loss를 최소화하는 Weight를 찾자!

# 요약

- MNIST Dataset Classification 모델 제작
- Neural Network를 Forward Propagation하는 예시 코드 확인
- Error / Loss를 최소화하는 목표 제시

