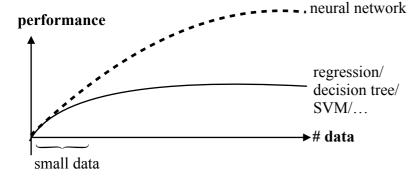
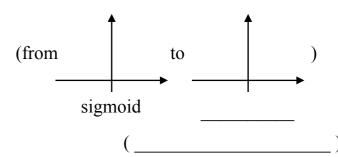
# SC201 Lecture 9

## **Deep Learning**

Why is it taking off?

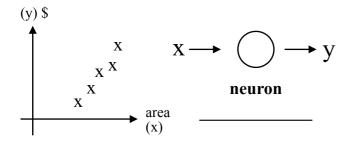


- Digitalization era makes data booming
- \_\_\_\_\_ enables complex algorithm
- Better math function

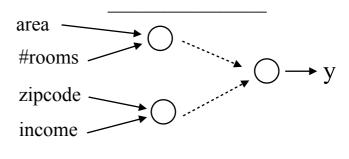


replace for loops

### Neuron



## Neural Network



- (1) feature extractor
- ② 2低維非線性組合出超高維model



The \_\_\_\_\_\_ it goes, the more \_\_\_\_\_\_ it can extract.



## **Matrix for Machine Learning** from Python dict → \_\_\_\_\_

(1) <

for epoch in range(num epochs):

for x, y in trainigExamples:

for i in range(num features):

$$W = \begin{vmatrix} W_1 \\ W_2 \\ \vdots \\ W_{nf} \end{vmatrix}$$

$$Y = [y_1, y_2 ... y_m] = [0 1 ... 1]$$

(especially \_\_\_\_\_) is good at matrix operations!

### **Basic Matrix Operation in Python**

 $W = [0.1, 0.2, 0.3, 0.4] # 1D ____/$ 

$$X_2 = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0.1 & 0.2 & 0.3 & 0.4 \\ 0.01 & 0.02 & 0.03 & 0.04 \end{bmatrix} 3x4$$

#### < dot Product >

$$X_{1} \bullet X_{2} \rightarrow \underbrace{ \begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \end{bmatrix} }_{\text{CO1 0.02 0.03 0.04}}$$

$$\begin{bmatrix}
2 & 3 & 4 & 5 \\
2 & 3 & 4 & 5 \\
2 & 3 & 4 & 5 \\
2 & 3 & 4 & 5
\end{bmatrix}$$

$$X_{1} = \begin{bmatrix} 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 \end{bmatrix} \quad X_{2} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0.1 & 0.2 & 0.3 & 0.4 \\ 0.01 & 0.02 & 0.03 & 0.04 \end{bmatrix} 3x4$$

< Transpose>

$$X_2.shape == (3, 4)$$
  
 $X_2._...shape == (__, __)$ 

< rank 1 array>

- 1D vector \_\_\_\_\_ transpose operation
- Use \_\_\_\_\_ to change it to 2D array \\_\_\_

np.random.rand(d1, d2)
reshape(\_\_\_\_\_)
np.zeros(\_\_\_\_\_)

< Element-wise> (\*,+,-,/,//)

$$X_{3} = \begin{bmatrix} 0.1 & 0.1 & 0.1 & 0.1 \\ 0.1 & 0.1 & 0.1 & 0.1 \\ 0.1 & 0.1 & 0.1 & 0.1 \end{bmatrix} \qquad X_{1} * X_{3} \rightarrow \underbrace{\qquad \qquad }_{((3x4) \text{ x } (3x4))}$$

$$X_{2} * X_{3} \rightarrow \underbrace{\qquad \qquad }_{((3x4) \text{ x } (3x4))}$$

$$\begin{bmatrix} 0.1 & 0.1 & 0.1 & 0.1 \\ 0.1 & 0.1 & 0.1 & 0.1 \\ 0.1 & 0.1 & 0.1 & 0.1 \end{bmatrix} * \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0.1 & 0.2 & 0.3 & 0.4 \\ 0.01 & 0.02 & 0.03 & 0.04 \end{bmatrix} =$$

< Vectorization>

Take advantage of \_\_\_\_\_ on CPU/GPU, we can speed up Python code by matrix operations.

$$Z = \begin{bmatrix} Z_0 \\ Z_1 \\ \vdots \\ \vdots \\ Z_{nf} \end{bmatrix}$$
  $= \begin{bmatrix} e^{Z_0} \\ e^{Z_1} \\ \vdots \\ e^{Z_{nf}} \end{bmatrix} = = \begin{bmatrix} = \\ = \\ = \\ = \end{bmatrix}$ 

< broadcasting> for \_\_\_\_\_ array (or \_\_\_\_\_)

(not

# Vecterizing Logistic Regression

• Fowardprop (\_\_\_\_\_

$$X = \begin{bmatrix} \vdots & \vdots & & \vdots \\ x_1 & x_2 & \cdots & x_m \\ \vdots & \vdots & & \vdots \end{bmatrix} \quad W = \begin{bmatrix} W_1 \\ W_2 \\ \vdots \\ W_{nf} \end{bmatrix} \quad Y = \begin{bmatrix} 0 & 1 & \dots & 1 \end{bmatrix}$$

$$H = 1/(1+np.exp(-k))$$

$$L = \underline{\qquad}$$

$$J = \underline{\qquad}$$

$$W = W - \alpha \frac{dJ}{dW}$$

$$W = W - \alpha \frac{dJ}{dW}$$

$$\frac{dJ}{dW} =$$

$$\frac{dJ}{dH} =$$

$$\frac{dH}{dK} =$$

$$\frac{dK}{dW} =$$