

# Derivative

## 1 Last Time

### 1.1 Polynomial Function

$$\dagger \quad y = x^n \rightarrow \frac{dy}{dx} = nx^{n-1}$$

$$\dagger \quad y = x^2 \rightarrow \frac{dy}{dx} = 2x$$

### 1.2 Chain Rule

$$\dagger \quad y = (3x + 4)^2 \rightarrow \frac{dy}{dx} = 2(3x + 4) \cdot \frac{d(3x+4)}{dx} = 2(3x + 4) \cdot 3$$

### 1.3 Partial Derivative

$$\dagger \quad y = (3x + 4w)^2 \rightarrow \frac{\partial y}{\partial x} = 2(3x + 4w) \cdot \frac{\partial(3x+4w)}{\partial x} = 2(3x + 4) \cdot 3$$

$$\dagger \quad y = (3x + 4w)^2 \rightarrow \frac{\partial y}{\partial w} = 2(3x + 4w) \cdot \frac{\partial(3x+4w)}{\partial w} = 2(3x + 4) \cdot 4$$

### 1.4 Squared Error Function

$$\dagger \quad E(w, b) = \frac{1}{2} \sum_{i=1}^n (w \cdot x_i + b - y_i)^2 \rightarrow$$
$$\frac{\partial E(w, b)}{\partial w} = \frac{1}{2} \sum_{i=1}^n 2(w \cdot x_i + b - y_i) \cdot \frac{\partial(w \cdot x_i + b - y_i)}{\partial w} = \sum_{i=1}^n (w \cdot x_i + b - y_i) \cdot x_i$$
$$\frac{\partial E(w, b)}{\partial b} = \frac{1}{2} \sum_{i=1}^n 2(w \cdot x_i + b - y_i) \cdot \frac{\partial(w \cdot x_i + b - y_i)}{\partial b} = \sum_{i=1}^n (w \cdot x_i + b - y_i) \cdot 1$$

## 2 New Function

### 2.1 Exponential Function

$$\dagger \quad y = e^x \rightarrow \frac{dy}{dx} = e^x$$

### 2.2 Chain Rule

$$\dagger \quad y = e^{wx} \rightarrow \frac{dy}{dx} = e^{wx} \cdot \frac{d(wx)}{dx} = e^{wx} \cdot w$$

### 2.3 Partial Derivative

$$\dagger \quad y = e^{wx} \rightarrow \frac{\partial y}{\partial w} = e^{wx} \cdot \frac{\partial(wx)}{\partial w} = e^{wx} \cdot x$$

### 3 Logistic Function

#### 3.1 Basic

$$\begin{aligned}\dagger \quad y = \sigma(x) &= \frac{1}{1+e^{-x}} = (1 + e^{-x})^{-1} \rightarrow \\ \frac{dy}{dx} &= (-1) \cdot (1+e^{-x})^{-2} \cdot \frac{d(1+e^{-x})}{dx} = (-1) \cdot (1+e^{-x})^{-2} \cdot e^{-x} \cdot \frac{d(-x)}{dx} = (-1) \cdot (1+e^{-x})^{-2} \cdot e^{-x} \cdot (-1) \\ &= (1 + e^{-x})^{-2} \cdot e^{-x} = \frac{1}{1 + e^{-x}} \cdot \frac{e^{-x}}{1 + e^{-x}} = \frac{1}{1 + e^{-x}} \cdot \left(1 - \frac{1}{1 + e^{-x}}\right) = \sigma(x)(1 - \sigma(x))\end{aligned}$$

#### 3.2 Partial Derivative

$$\begin{aligned}\dagger \quad y = \sigma(w) &= \frac{1}{1+e^{-wx}} = (1 + e^{-wx})^{-1} \rightarrow \\ \frac{\partial y}{\partial w} &= (-1) \cdot (1+e^{-wx})^{-2} \cdot \frac{\partial(1+e^{-wx})}{\partial w} = (-1) \cdot (1+e^{-wx})^{-2} \cdot e^{-wx} \cdot \frac{\partial(-wx)}{\partial w} = (-1) \cdot (1+e^{-wx})^{-2} \cdot e^{-wx} \cdot (-x) \\ &= (1+e^{-wx})^{-2} \cdot e^{-wx} \cdot x = \frac{1}{1 + e^{-wx}} \cdot \frac{e^{-wx}}{1 + e^{-wx}} \cdot x = \frac{1}{1 + e^{-wx}} \cdot \left(1 - \frac{1}{1 + e^{-wx}}\right) \cdot x = \sigma(w)(1 - \sigma(w)) \cdot x\end{aligned}$$

#### 3.3 Error Function

$$\begin{aligned}\dagger \quad E(w) &= \frac{1}{2} \sum_{i=1}^n \left( \frac{1}{1+e^{-wx_i}} - y_i \right)^2 \rightarrow \\ \frac{\partial E(w)}{\partial w} &= \sum_{i=1}^n \left( \frac{1}{1+e^{-wx_i}} - y_i \right) \cdot \left( \frac{1}{1+e^{-wx_i}} \right) \cdot \left( 1 - \frac{1}{1+e^{-wx_i}} \right) \cdot x_i\end{aligned}$$