

# Backpropagation za logističku regresiju

May 2, 2023

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Metoda optimizacije je stohastički gradijentni spust:

$$w'_i = w_i - \eta \frac{\partial \mathbb{E}}{\partial w_i}$$

# Osnovna pravila deriviranja

1 LD:

$$[f(x)a + g(x)b]' = f(x)'a + g(x)'b$$

2 REC:

$$\left[\frac{1}{f(x)}\right]' = -\frac{f'(x)}{f(x)^2}$$

3 CONST:

$$c' = 0$$

4 VAR:

$$\frac{dy}{dz}z = 1$$

5 EXP:

$$[f(x)^n]' = n \cdot f(x)^{n-1} \cdot f'(x)$$

6 CEXP:

$$[e^{f(x)}]' = e^{f(x)} \cdot f'(x)$$

7 CHAIN:

$$\frac{dx}{dz} = \frac{dx}{dy} \frac{dy}{dz}$$

# Weight update za logističku regresiju

Za puni GD:

$$w'_i = w_i - \eta \frac{\partial \mathbb{E}}{\partial w_i} = w_i + \sum_n \eta x_i^{(n)} (t^{(n)} - \hat{y}^{(n)})$$

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No zašto  $\frac{\partial \mathbb{E}}{\partial w_i} = -x_i(t - \hat{y})$ ?

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Prvo s CHAIN raskomadamo:

$$\frac{\partial \mathbb{E}}{\partial w_i} = \frac{\partial \hat{y}}{\partial w_i} \frac{d\mathbb{E}}{d\hat{y}}$$

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Na desni dio ćemo se još vratiti, za sad gledamo  $\frac{\partial \hat{y}}{\partial w_i}$ . Ponovno primjenimo CHAIN i dobijemo

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Sve što još trebamo pokazati je  $\frac{d\hat{y}}{dz} = \hat{y}(1 - \hat{y})$

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$$\frac{d\hat{y}}{dz} \frac{1}{1 + e^{-z}} = - \frac{\frac{d\hat{y}}{dz} (1 + e^{-z})}{(1 + e^{-z})^2} \quad (REC)$$

$$= - \frac{\frac{d\hat{y}}{dz} 1 + \frac{d\hat{y}}{dz} e^{-z}}{(1 + e^{-z})^2} \quad (LD)$$

$$= - \frac{0 + \frac{d\hat{y}}{dz} (e^{-z})}{(1 + e^{-z})^2} \quad (CONST)$$

$$= - \frac{e^{-z} \cdot \frac{d\hat{y}}{dz} (-z)}{(1 + e^{-z})^2} \quad (CEXP)$$

$$= - \frac{e^{-z} \cdot \frac{d\hat{y}}{dz} (z) \cdot (-1)}{(1 + e^{-z})^2} \quad (LD)$$

$$= - \frac{e^{-z} \cdot (-1)}{(1 + e^{-z})^2} = \frac{e^{-z}}{(1 + e^{-z})^2} \quad (VAR)$$

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Ajmo ovo faktorizirati na sljedeći način:

$$\frac{d\hat{y}}{dz} = \frac{e^{-z}}{(1 + e^{-z})^2} = \frac{1}{1 + e^{-z}} \cdot \frac{e^{-z}}{1 + e^{-z}}$$

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Dakle,  $\frac{d\hat{y}}{dz} = \hat{y}(1 - \hat{y})$

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Prisjetimo se da koristimo SSE, i to u varijanti za stohastički gradijentni spust,  $\mathbb{E} = \frac{1}{2}(t - y)^2$  (za običan gradijentni spust ide još suma po svim predikcijama na datasetu).

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$$\begin{aligned}\frac{d\mathbb{E}}{d\hat{y}}\left[\frac{1}{2}(t - \hat{y})^2\right] &=_{LD} \frac{1}{2} \frac{d\mathbb{E}}{d\hat{y}}(t - \hat{y})^2 =_{EXP} \frac{1}{2} \cdot 2 \cdot (t - \hat{y}) \cdot \frac{d\mathbb{E}}{d\hat{y}}(t - \hat{y}) \\ &=_{LD} (t - \hat{y})\left[\frac{d\mathbb{E}}{d\hat{y}}t - \frac{d\mathbb{E}}{d\hat{y}}y\right] =_{CONST,VAR} (t - \hat{y}) \cdot (0 - 1) = -1 \cdot (t - \hat{y})\end{aligned}$$

# Konačni weight update za logističku regresiju prema specifikacijama (SSE, Logistička, SGD)

Prisjetimo se:

$$w'_i = w_i - \eta \frac{\partial \mathbb{E}}{\partial w_i} \quad \frac{\partial \mathbb{E}}{\partial w_i} = \frac{\partial z}{\partial w_i} \frac{d\hat{y}}{dz} \frac{d\mathbb{E}}{d\hat{y}}$$

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Sve sto nam preostaje jest sastaviti weight update formulu:

$$\begin{aligned} w'_i &= w_i - \eta \frac{\partial \mathbb{E}}{\partial w_i} = w_i - \eta \frac{\partial z}{\partial w_i} \frac{d\hat{y}}{dz} \frac{d\mathbb{E}}{d\hat{y}} = w_i - \eta (x_i \cdot \hat{y}(1 - \hat{y}) \cdot -(t - \hat{y})) = \\ &= w_i + \eta x_i \hat{y}(1 - \hat{y})(t - \hat{y}) \end{aligned}$$