



Fitting the model

Cost function and log-loss



Logistic Regression Model

$$P(y=1 | X) = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_n x_n)}}$$

- X are the independent variables.
- Y is the target, which is class target.
- β are the parameters of the model.

Probabilities

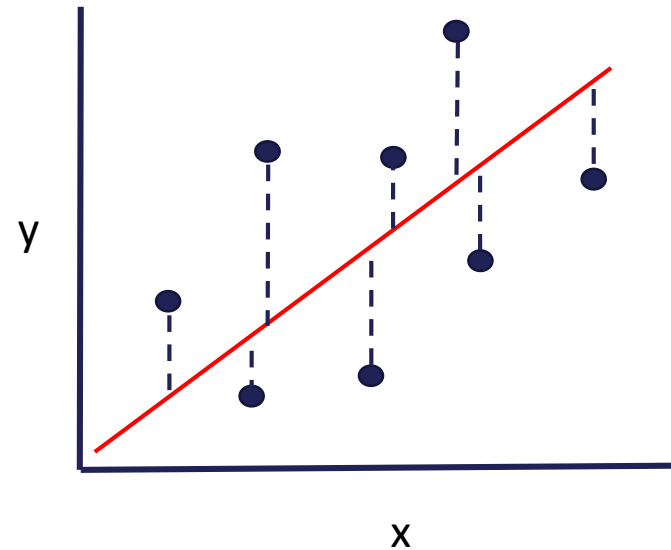
$$P(y=1 | X) = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_n x_n)}}$$

- The sum of probabilities must equal 1.
- If we know $P(y=1) \rightarrow$ we know $P(y=0) = 1 - P(y=1)$

Linear regression: residuals

$$\beta = \min \sum (y - \text{predictions})^2$$

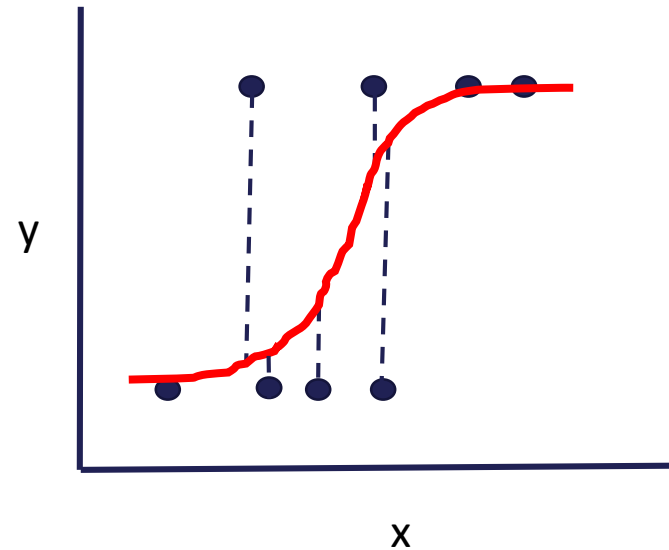
OLS → Find the coefficients that minimize the squared difference of the target and the predictions.



Residuals – not useful

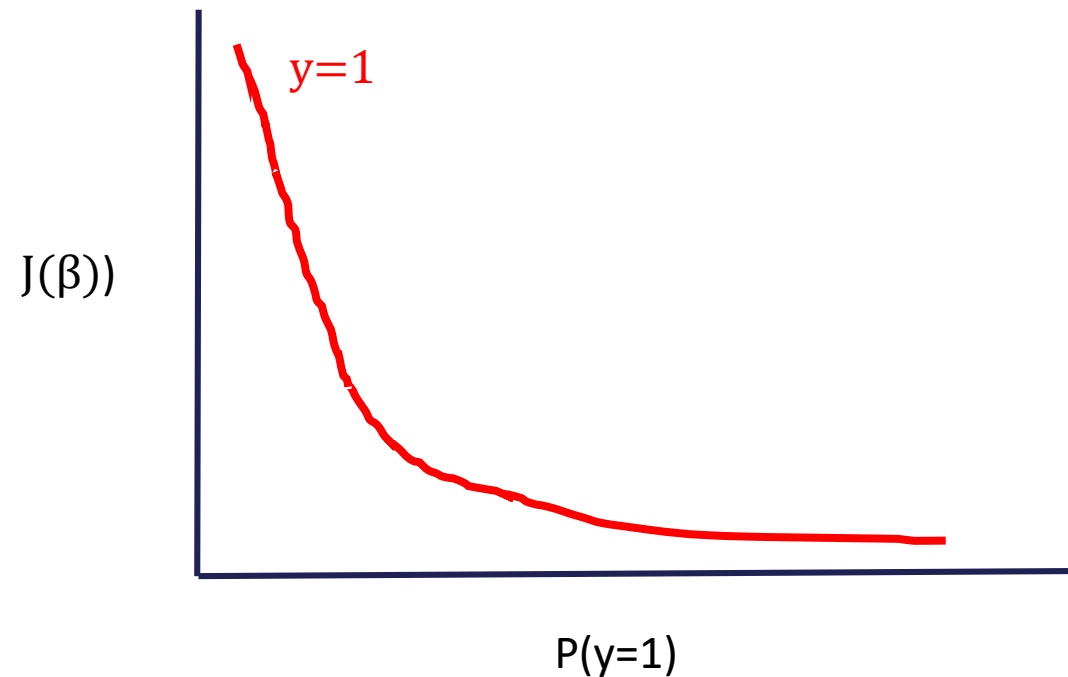
$$\beta = \min \sum (y - \text{predictions})^2$$

Non-convex function → not guaranteed to find the minimum.



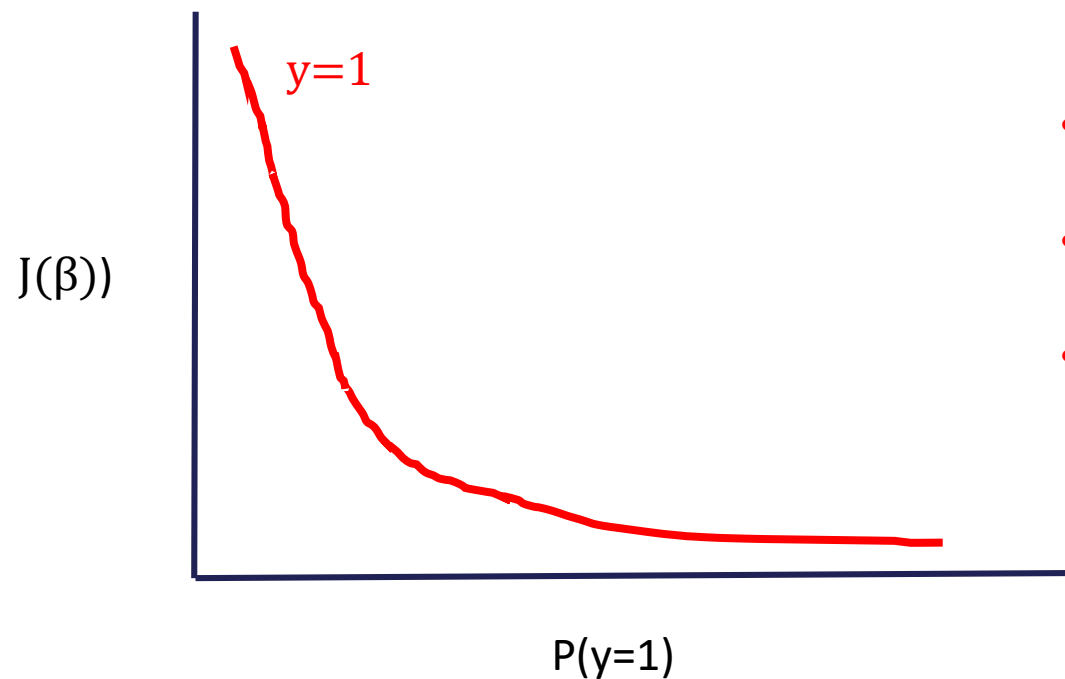
Cost

$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \end{cases}$$



Cost

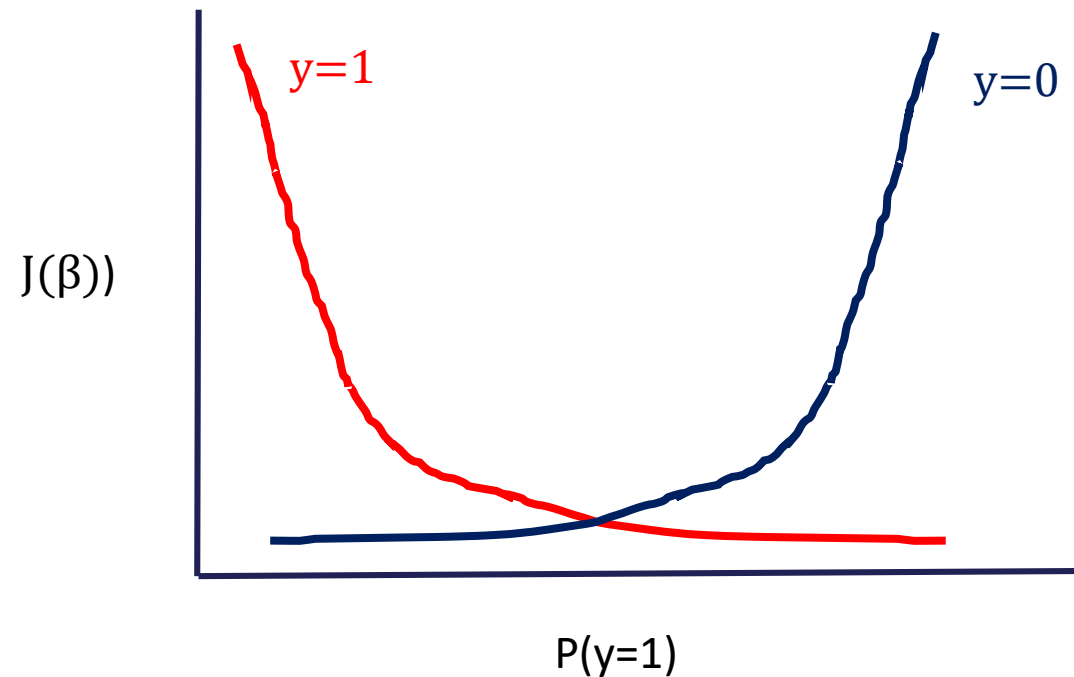
$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \end{cases}$$



- If $P(y=1) = 0.3 \rightarrow -\log(0.3) = 1.2$
- If $P(y=1) = 0.5 \rightarrow -\log(0.5) = 0.69$
- If $P(y=1) = 0.8 \rightarrow -\log(0.8) = 0.22$

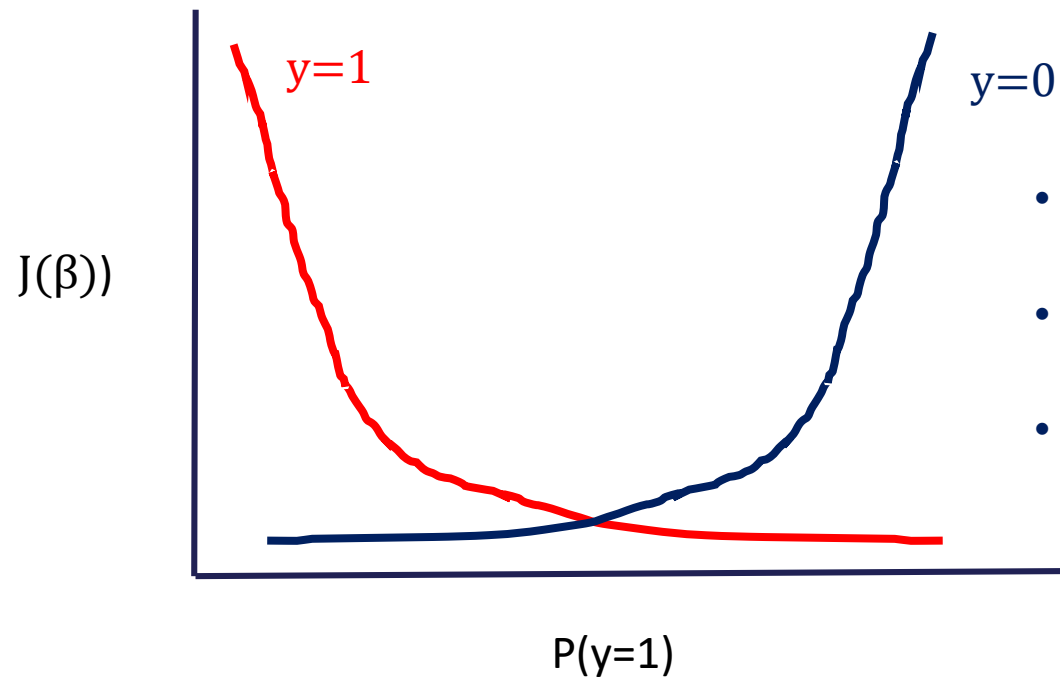
Cost

$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \\ -\log(1-P(y=1)) & \text{if } y=0 \end{cases}$$



Cost

$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \\ -\log(1-P(y=1)) & \text{if } y=0 \end{cases}$$



- If $P(y=1) = 0.7 \rightarrow -\log(1-0.7) = 1.2$
- If $P(y=1) = 0.4 \rightarrow -\log(1-0.4) = 0.51$
- If $P(y=1) = 0.2 \rightarrow -\log(1-0.2) = 0.22$

Cost

$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \\ -\log(1-P(y=1)) & \text{if } y=0 \end{cases}$$

$$\text{Cost}(\text{model}, y) = \sum -y \log(P(y=1)) - (1-y) \log(1-P(y=1))$$

Log-loss

$$\text{Cost}(\text{model}, y) = \begin{cases} -\log(P(y=1)) & \text{if } y=1 \\ -\log(1-P(y=1)) & \text{if } y=0 \end{cases}$$

$$\text{Cost}(\text{model}, y) = \sum -y \log(P(y=1)) - (1-y) \log(1-P(y=1))$$

$$J(\beta) = -\frac{1}{m} \sum y \log(P(y=1)) + (1-y) \log(1-P(y=1))$$

THANK YOU

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