Classification (Short Version)

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Classification

- To learn more ……

https://youtu.be/fZAZUYEeIMg

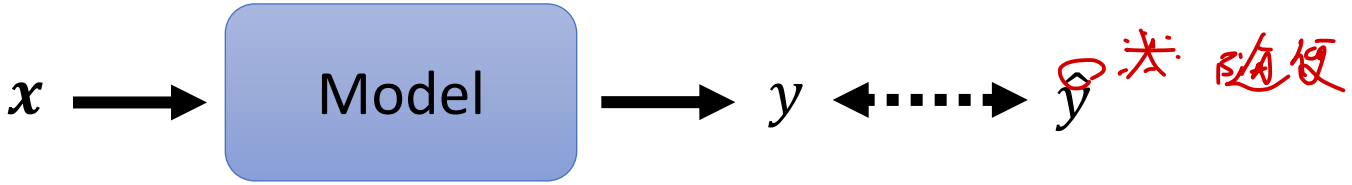
(in Mandarin)

https://youtu.be/hSXFuypLukA

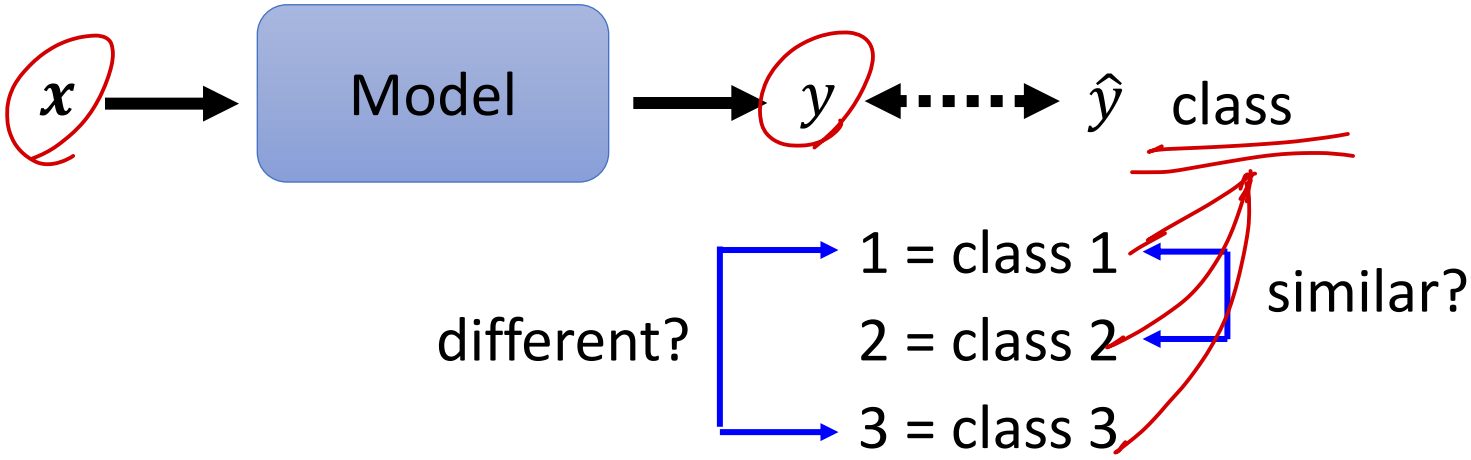
(in Mandarin)

Classification as Regression?

- Regression



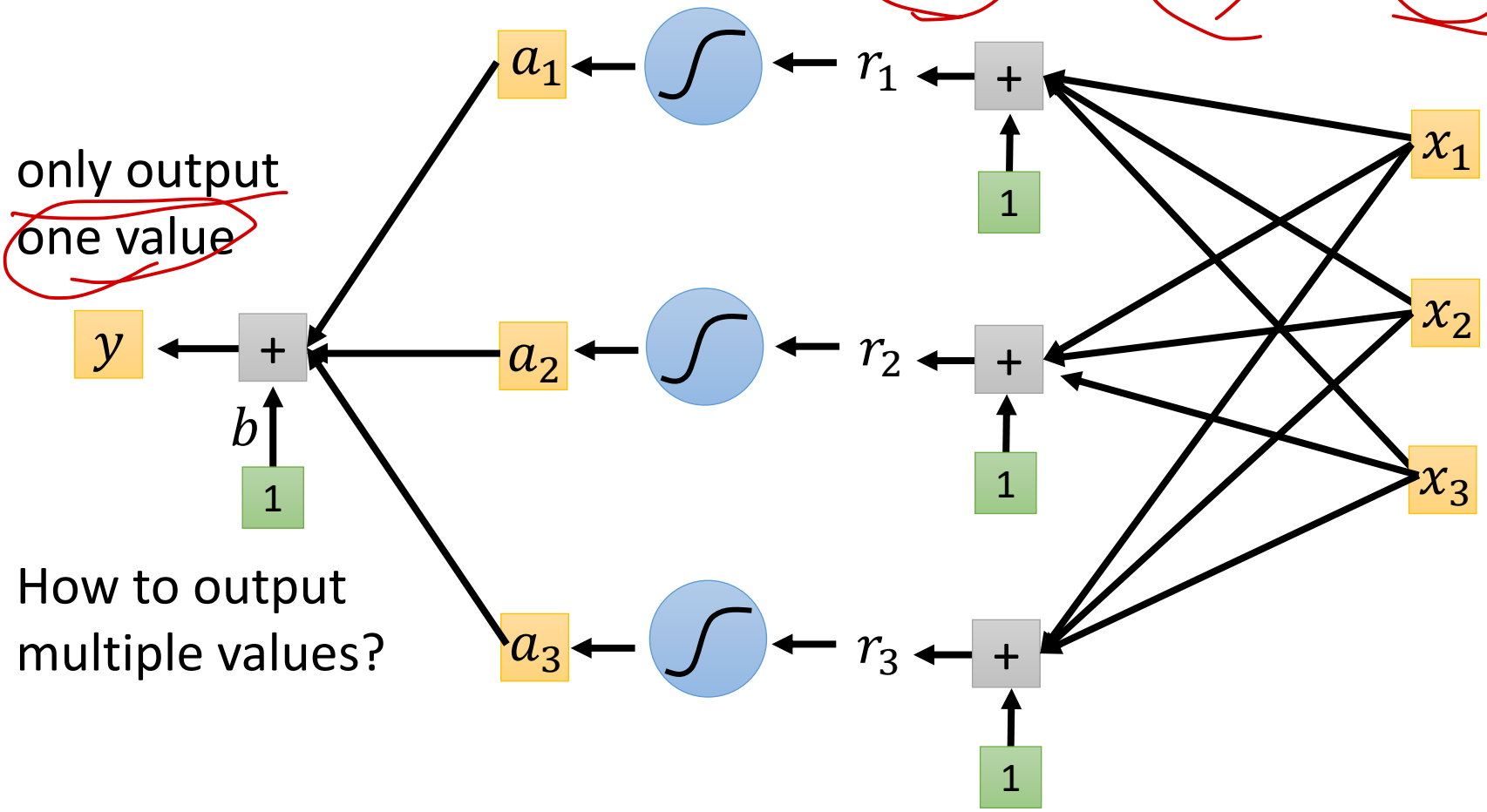
- Classification as regression?



Class as one-hot vector

Class 1 Class 2 Class 3

$\hat{y} = \begin{bmatrix}1\\0\\0\end{bmatrix}$ or $\begin{bmatrix}0\\1\\0\end{bmatrix}$ or $\begin{bmatrix}0\\0\\1\end{bmatrix}$



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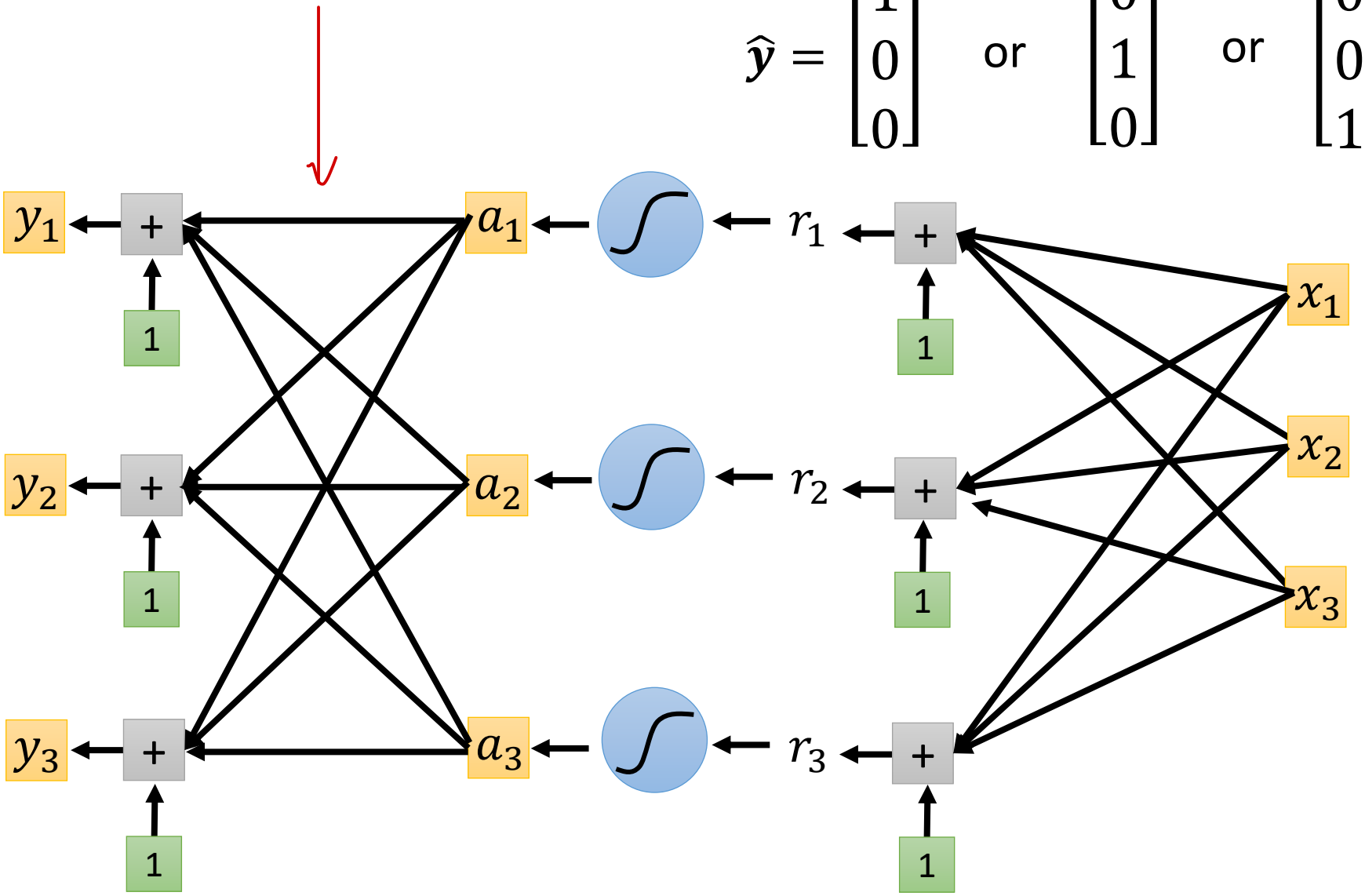


Regression

feature

label

$$\hat{y} \longleftrightarrow y = b + c^T \sigma(b + W x)$$

Classification

feature

向量

$$y = b' + W' \sigma(b + W x)$$

label

$$\hat{y} \longleftrightarrow y' = softmax(y)$$

0 or 1

Make all values between 0 and 1

通过

Can have any value

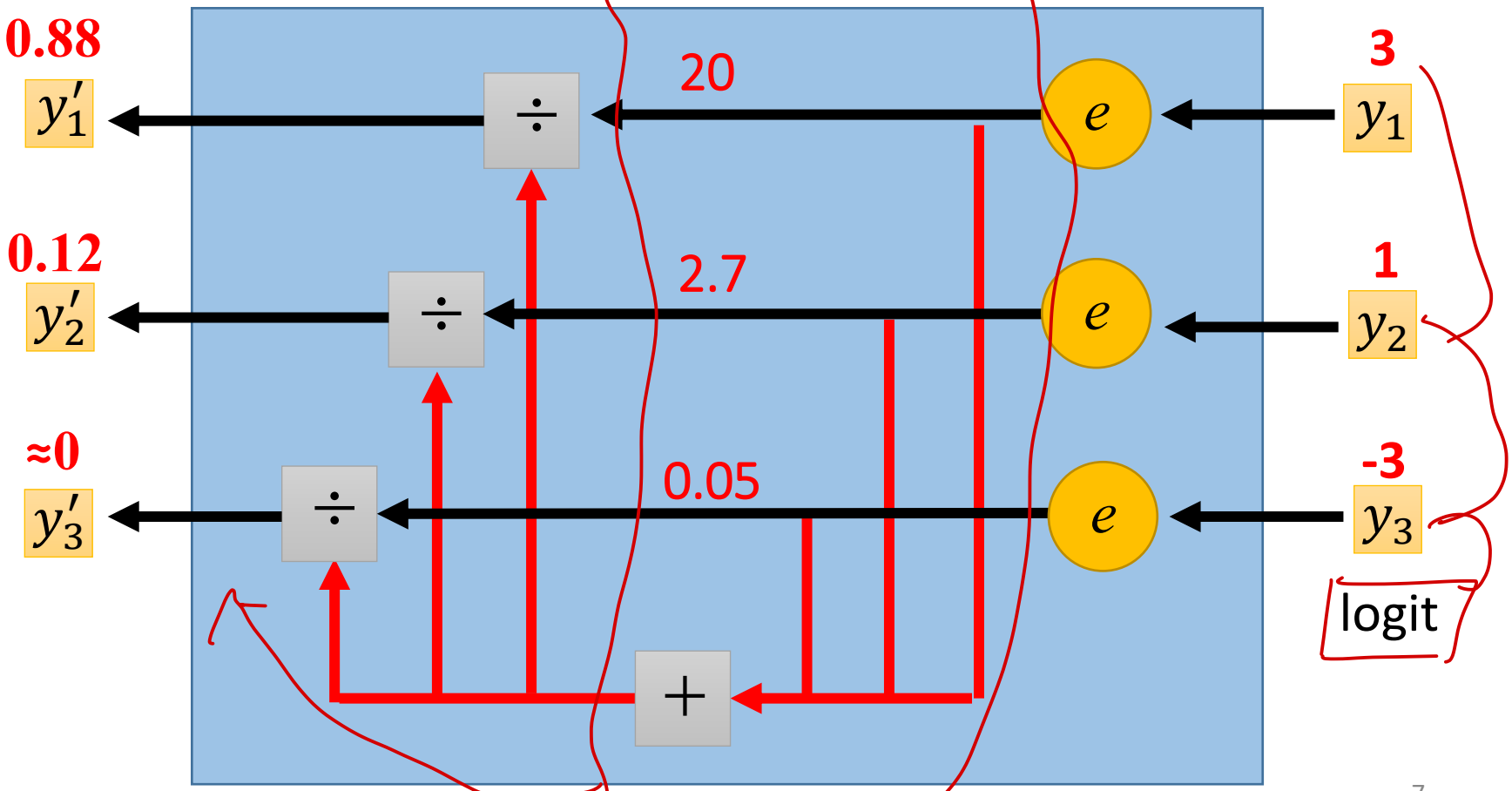
normalize → 归一化

Soft-max

$$y_i' = \frac{exp(y_i)}{\sum_j exp(y_i)}$$

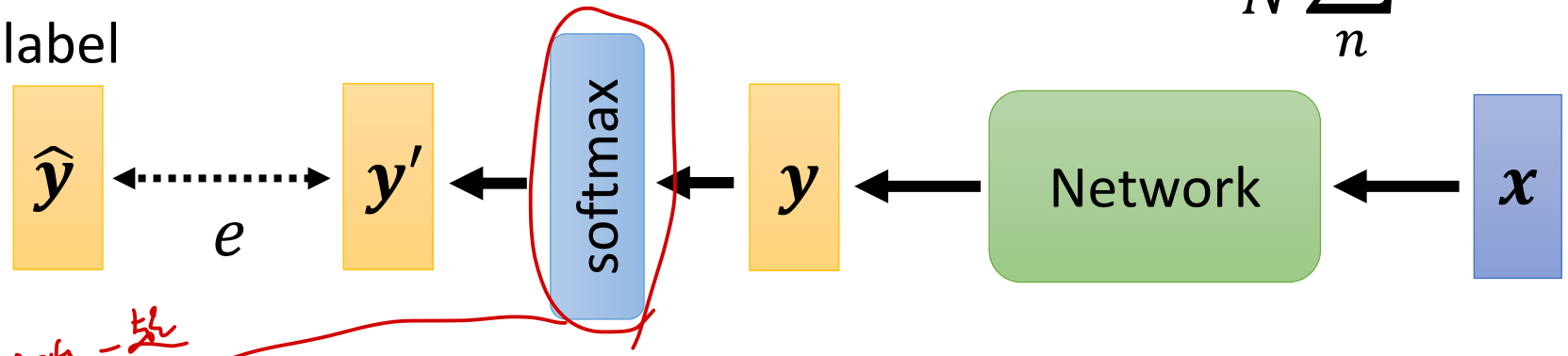
- $1 > y_i' > 0$
- $\sum_i y_i' = 1$

How about **binary classification**? ☺



Loss of Classification

$$L = \frac{1}{N}\sum_{n} e_n$$

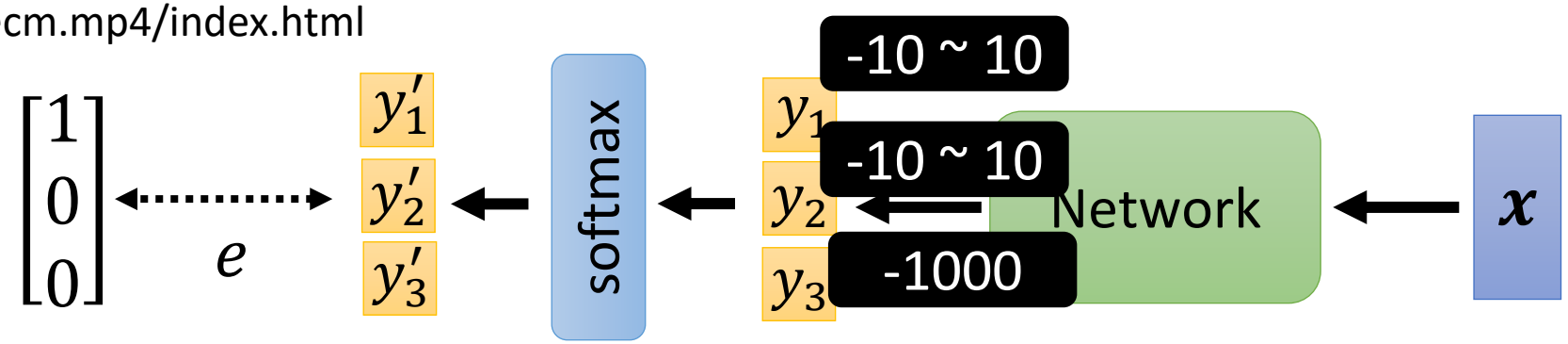


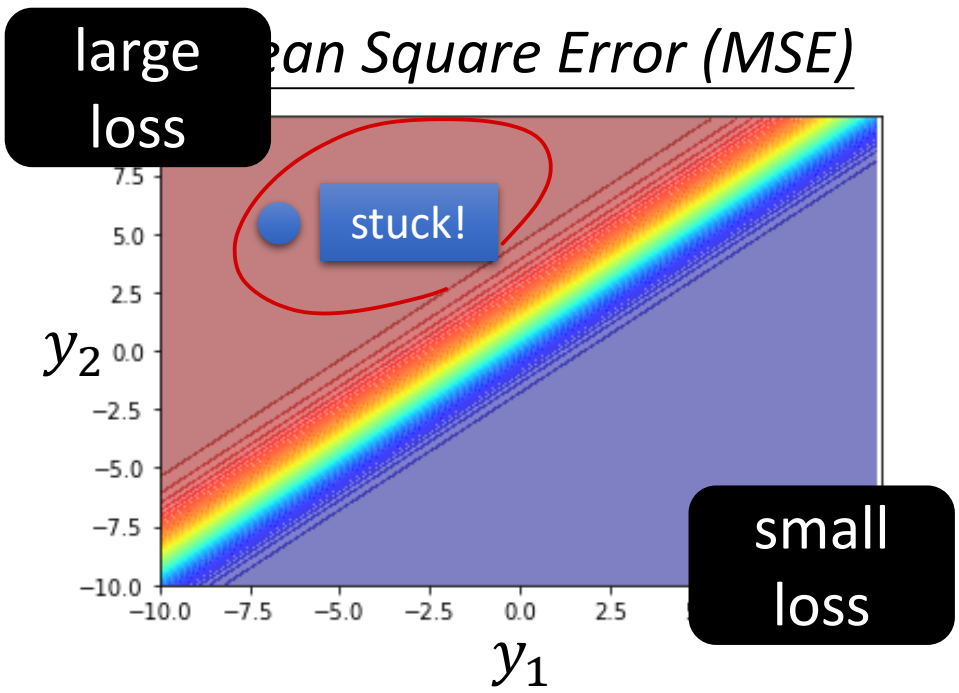
Mean Square Error (MSE) $e = \sum_i (\hat{y}_i - y'_i)^2$

Cross-entropy WIN $e = -\sum_i \hat{y}_i ln y'_i$

Minimizing cross-entropy is equivalent to **maximizing likelihood**.

http://speech.ee.ntu.edu.tw/~tlkagk/courses/MLDS_2015_2/Lecture/Deep%20More%20(v2).ecm.mp4/index.html





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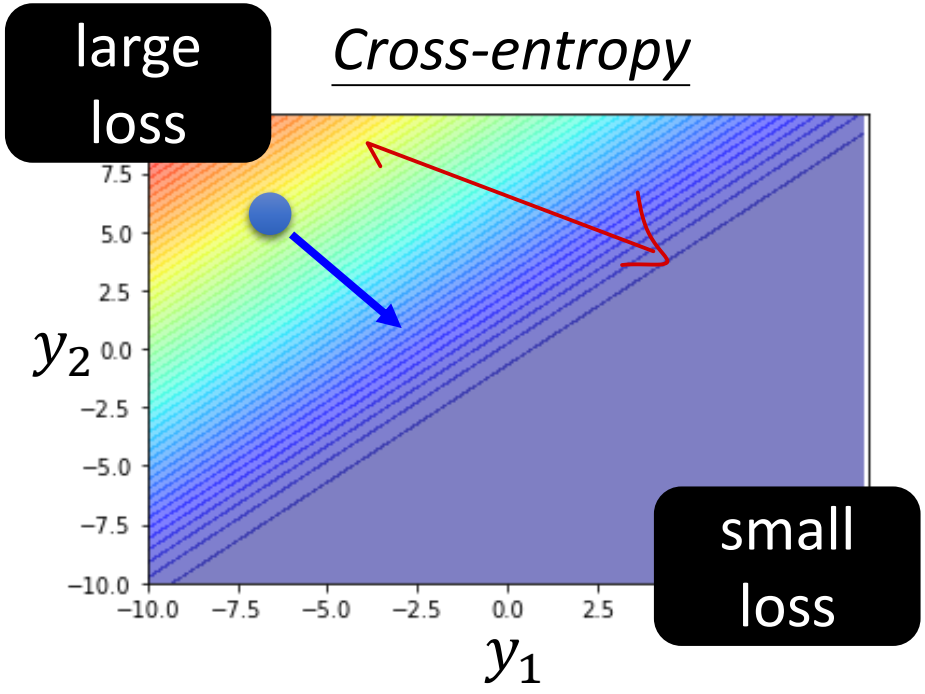

Cross-entropy

Changing the loss function can change the difficulty of optimization.