

Probabilistic discriminative models:

$$P(c_i | \underline{x}) = \sigma(\underline{\omega}^T \underline{x})$$

$$\underline{\omega} \in \mathbb{R}^D$$

Computes $\underline{\omega}$ using ML approach for
D parameters is called "Logistic regression"

Goal: Given $\{\underline{x}_i, y_i\}_{i=1}^N$ with $y_i \in \{0, 1\}$

$$P(y | \underline{x}; \underline{\omega}) = \text{Ber}(y | \sigma(\underline{\omega}^T \underline{x}))$$

$$P(\underline{y} | \underline{x}; \underline{\omega}) = \prod_{i=1}^N P(y_i | \underline{x}_i; \underline{\omega})$$

$$= \prod_{n=1}^N [\sigma(\underline{\omega}^T \underline{x})]^{y_n} [1 - \sigma(\underline{\omega}^T \underline{x})]^{1-y_n}$$

Negative likelihood function:

$$\begin{aligned} \ell(\underline{\omega}) := & - \sum_{n=1}^N \left[y_n \log \sigma(\underline{\omega}^T \underline{x}) \right. \\ & \left. + (1 - y_n) \log (1 - \sigma(\underline{\omega}^T \underline{x})) \right] \end{aligned}$$

Cross-entropy function

$$\sigma(-t) = 1 - \sigma(t)$$

Alternative view:

$$y_i \in \{0, 1\}$$

$$\tilde{y}_i = 2y_i - 1$$

$$\tilde{y}_i \in \{-1, +1\}$$

$$J(\underline{w}) = \sum_{i=1}^N \log (1 + \exp (-\tilde{y}_i \underline{w}^T \underline{x}_i))$$

When

$$y_i = 1$$

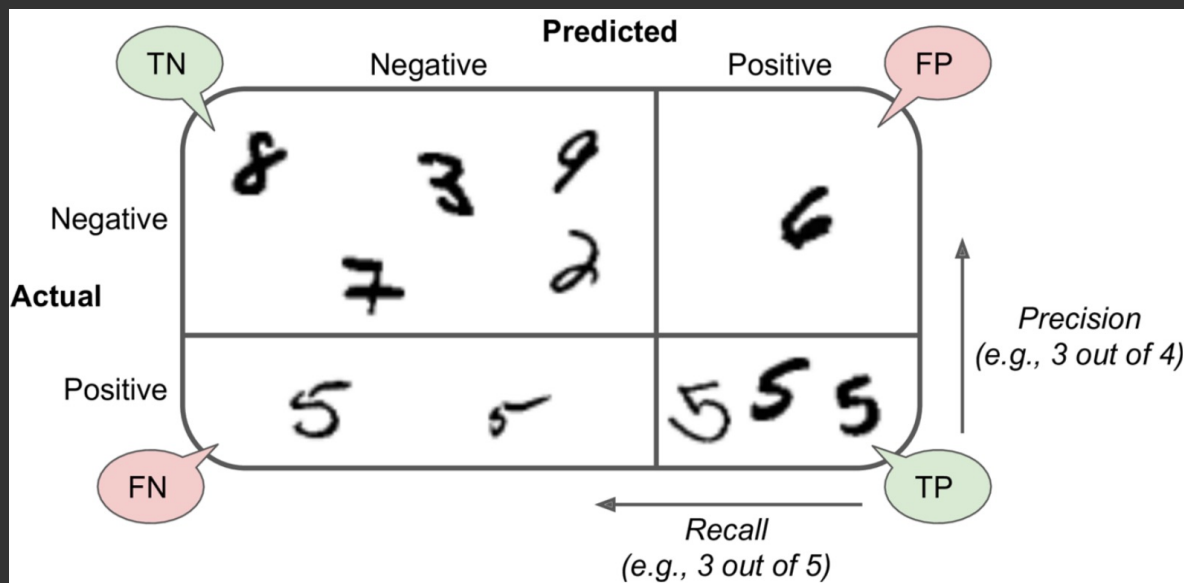
$$\begin{aligned} -\log \sigma(\underline{w}^T \underline{x}_i) &= \log (1 + \exp (-\underline{w}^T \underline{x}_i)) \\ &= \log (1 + \exp (-\tilde{y}_i \underline{w}^T \underline{x}_i)) \end{aligned}$$

$$y_i = 0$$

$$\begin{aligned} -\log (1 - \sigma(\underline{w}^T \underline{x}_i)) &= \log (1 + \exp (\underline{w}^T \underline{x}_i)) \\ &= \log (1 + \exp (-\tilde{y}_i \underline{w}^T \underline{x}_i)) \end{aligned}$$

Multiclass logistic regression:

$$f(\mathbf{W}) = - \sum_{n=1}^N \sum_{k=1}^C y_{nk} \log \frac{\exp(\omega_k^T x_n)}{\sum_j \exp(\omega_j^T x_n)}$$



Accuracy:

$$\frac{TP + TN}{TP + TN + FP + FN}$$

Not very useful with
class imbalance $\begin{bmatrix} 990 +ve \\ 10 -ve \end{bmatrix}$
99%

$$\text{Precision} = \frac{TP}{TP + FP}$$

$$\text{Recall} = \frac{TP}{TP + FN}$$

$$\text{F1 Score} = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}} \quad (\text{harmonic mean})$$

Trade off:

