

# LDA AS LINEAR CLASSIFIER



$$\begin{aligned} & \pi_k \cdot p(\mathbf{x}|y = k) \\ \propto & \pi_k \exp\left(-\frac{1}{2}\mathbf{x}^T \Sigma^{-1} \mathbf{x} - \frac{1}{2}\boldsymbol{\mu}_k^T \Sigma^{-1} \boldsymbol{\mu}_k + \mathbf{x}^T \Sigma^{-1} \boldsymbol{\mu}_k\right) \\ = & \exp\left(\log \pi_k - \frac{1}{2}\boldsymbol{\mu}_k^T \Sigma^{-1} \boldsymbol{\mu}_k + \mathbf{x}^T \Sigma^{-1} \boldsymbol{\mu}_k\right) \exp\left(-\frac{1}{2}\mathbf{x}^T \Sigma^{-1} \mathbf{x}\right) \\ = & \exp\left(\mathbf{w}_{0k} + \mathbf{x}^T \mathbf{w}_k\right) \exp\left(-\frac{1}{2}\mathbf{x}^T \Sigma^{-1} \mathbf{x}\right) \\ \propto & \exp\left(\mathbf{w}_{0k} + \mathbf{x}^T \mathbf{w}_k\right) \end{aligned}$$

by defining  $\mathbf{w}_{0k} := \log \pi_k - \frac{1}{2}\boldsymbol{\mu}_k^T \Sigma^{-1} \boldsymbol{\mu}_k$  and  $\mathbf{w}_k := \Sigma^{-1} \boldsymbol{\mu}_k$ .

By finally taking the log, we can write our transformed scores as linear:

$$f_k(\mathbf{x}) = \mathbf{w}_{0k} + \mathbf{x}^T \mathbf{w}_k$$

- The above is a little bit "tax" so lets carefully check
- We left out several (pos) multiplicative constants
- $\exp\left(-\frac{1}{2}\mathbf{x}^T \Sigma^{-1} \mathbf{x}\right)$  contains  $\mathbf{x}$  but is the same for all classes
- $\log(at + b)$  is still isotonic for  $a > 0$