

$$p = \frac{1}{1 + e^{-\theta^T x}}$$

$$\Leftrightarrow \log \frac{p}{1-p} = \theta^T x$$

$$\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$\psi(z) = \frac{1}{1 + e^{-z}}$$

$$\psi'(z) = -\frac{1}{(1+e^{-z})^2} (-e^{-z}) = \frac{e^{-z}}{(1+e^{-z})^2} = e^{-z} \psi(z)^2$$

$$= \psi(z) (1 - \psi(z))$$

$$p_i = \psi(\theta^T x_i)$$

$$\frac{\partial p_i}{\partial \theta_j} = \psi'(\theta^T x_i) (x_i)_j = \psi(\theta^T x_i) (1 - \psi(\theta^T x_i)) (x_i)_j$$

$$= p_i (1 - p_i) (x_i)_j$$

$$\psi(\theta^T x_i) = \underbrace{\psi(\theta^T x_i)}_{p_i} \underbrace{(1 - \psi(\theta^T x_i))}_{1-p_i} = p_i (1 - p_i)$$

$p_i$  depends on  $\theta$

$$\log L(\theta) = \log \prod_{i=1}^n p_i^{y_i} (1-p_i)^{1-y_i}$$

$$= \sum_{i=1}^n (\log p_i^{y_i} (1-p_i)^{1-y_i}) = \sum y_i \log p_i + \sum (1-y_i) \log (1-p_i)$$

$$\frac{\partial \log L(\theta)}{\partial \theta_j} = \sum_i \frac{y_i}{p_i} \frac{\partial p_i}{\partial \theta_j} + \sum_i \frac{1-y_i}{1-p_i} \frac{\partial (1-p_i)}{\partial \theta_j}$$

$$= \sum_i \left( \frac{y_i}{p_i} - \frac{1-y_i}{1-p_i} \right) \frac{\partial p_i}{\partial \theta}$$

$$= \sum_i \left( \frac{y_i}{p_i} - \frac{1-y_i}{1-p_i} \right) p_i (1-p_i) (x_i)$$

$$= \sum_i (y_i (1-p_i) - (1-y_i) p_i) (x_i)$$

$$= \sum_i (y_i - p_i) (x_i) = \sum_i (y_i - p_i) x_{i,j}$$

$$= X^T (y - p)$$

$$\frac{\partial}{\partial \theta_k} \frac{\partial \log L(\theta)}{\partial \theta_j} = \frac{\partial}{\partial \theta_k} \sum_i (y_i - p_i) x_{i,j}$$

$$= - \sum_i \frac{\partial p_i}{\partial \theta_k} x_{i,j} = - \sum_i p_i (1 - p_i) x_{i,j} x_{i,k}$$

$$= - X^T \Lambda X$$

$$\Lambda = \begin{pmatrix} p_1(1-p_1) & & \\ & \ddots & \\ & & p_n(1-p_n) \end{pmatrix}$$

Trick - Calculate posterior density without  
constants

- Recognize certain distribution
- Fill in constants from known  
distribution