

Q2) Sigmoid function  $\sigma(a) = \frac{1}{1 + \exp(-a)} = \frac{1}{1 + e^{-a}}$

Probit function  $\Phi(a) = \int_{-\infty}^a N(\theta | 0, 1) d\theta$

$N(\theta | 0, 1) \rightarrow$  standard normal distribution ( $\mu$ )  
mean = 0  
variance = 1

Standard distribution function

$$f(\theta | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(\theta - \mu)^2}{2\sigma^2}} = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(\theta - 0)^2}{2(1)^2}} = \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}}$$

Using Probit function instead of sigmoid function  
given training data  $\{(x_i, y_i) : 1 \leq i \leq n\}$  from  
distribution  $\mathcal{D}$

so, for Probit regression negative conditional likelihood loss  
is:

$$\begin{aligned} \therefore \hat{L}(\omega) &= -\frac{1}{n} \sum_{y_i=1} \log \phi(\omega^T x_i) - \frac{1}{n} \sum_{y_i=0} \log [1 - \Phi(\omega^T x_i)] \\ &= -\frac{1}{n} \sum_{y_i=1} \log \int_{-\infty}^{\omega^T x_i} N(\theta | 0, 1) d\theta - \frac{1}{n} \sum_{y_i=0} \log \left[ 1 - \int_{-\infty}^{\omega^T x_i} N(\theta | 0, 1) d\theta \right] \end{aligned}$$

$$\begin{aligned} \text{where } N(\theta | 0, 1) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} \\ &= -\frac{1}{n} \sum_{y_i=1} \log \int_{-\infty}^{\omega^T x_i} \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} d\theta - \frac{1}{n} \sum_{y_i=0} \log \left[ 1 - \int_{-\infty}^{\omega^T x_i} \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} d\theta \right] \end{aligned}$$

So, our final equation for negative conditional  
likelihood loss using probit function:

$$\boxed{\hat{L}(\omega) = -\frac{1}{n} \sum_{y_i=1} \log \int_{-\infty}^{\omega^T x_i} \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} d\theta - \frac{1}{n} \sum_{y_i=0} \log \left[ 1 - \int_{-\infty}^{\omega^T x_i} \frac{1}{\sqrt{2\pi}} e^{-\frac{\theta^2}{2}} d\theta \right]}$$