

# Classification

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1 Logistic Regression

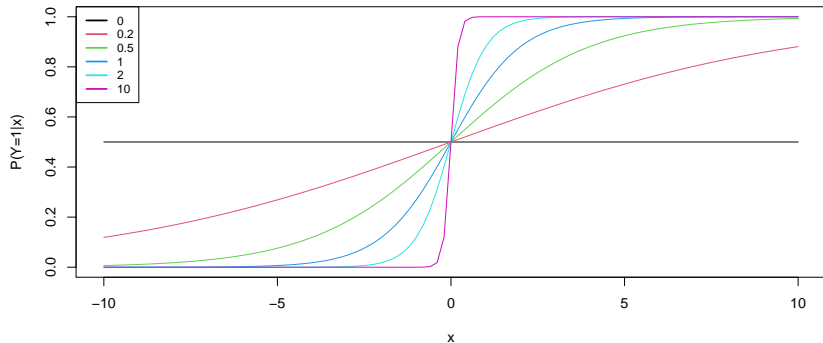
2 Newton Raphson Method

3 Linear and Quadratic Discrimination

- The probabilities of  $y = 1$  and  $y = -1$  are expressed by  $\frac{e^{\beta_0 + x\beta}}{1 + e^{\beta_0 + x\beta}}$  and  $\frac{1}{1 + e^{\beta_0 + x\beta}}$

$$\frac{1}{1 + e^{-y(\beta_0 + x\beta)}}$$

Logistic Curve



$$f'(x) = \beta \frac{e^{-(\beta_0 + x\beta)}}{(1 + e^{-(\beta_0 + x\beta)})^2} \geq 0$$
$$f''(x) = -\beta^2 \frac{e^{-(\beta_0 + x\beta)}[1 - e^{-(\beta_0 + x\beta)}]}{(1 + e^{-(\beta_0 + x\beta)})^3}$$

- We see that  $f(x)$  is increasing monotonically and is convex and concave when  $x < -\beta_0/\beta$  and  $x > -\beta_0/\beta$ , they change at  $x = 0$ , when  $\beta = 0$

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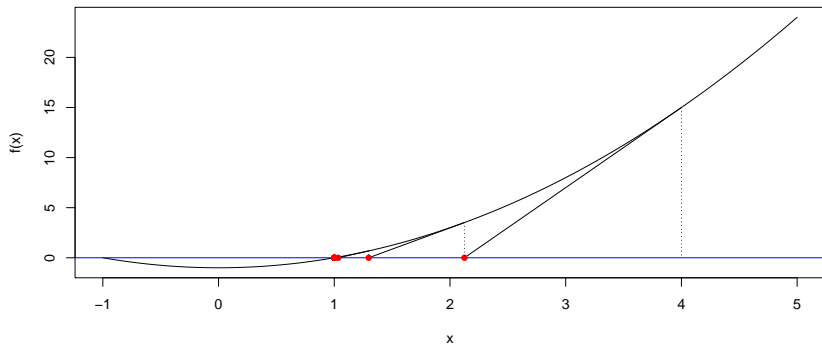
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# Newton Raphson

- Tangent line is  $y - f(x_i) = f'(x_i)(x - x_i)$  the intersection with  $y = 0$

$$x_{i+1} \triangleq x_i - \frac{f(x_i)}{f'(x_i)}$$

- Example  $f(x) = x^2 - 1$  and  $x_0 = 4$



# Newton Raphson Method for two variables

- We can even be applied to two variables and two equations

$$\begin{cases} f(x, y) = 0, \\ g(x, y) = 0 \end{cases}$$

- We can see Newton Raphson is extended to

$$\begin{bmatrix} x \\ y \end{bmatrix} \leftarrow \begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} \frac{\partial f(x, y)}{\partial x} & \frac{\partial f(x, y)}{\partial y} \\ \frac{\partial g(x, y)}{\partial x} & \frac{\partial g(x, y)}{\partial y} \end{bmatrix}^{-1} \begin{bmatrix} f(x, y) \\ g(x, y) \end{bmatrix}$$

- We apply the same method to the problem of finding  $\beta_0 \in \mathbb{R}$  and  $\beta \in \mathbb{R}^p$  such that  $\nabla l(\beta_0, \beta) = 0$

$$(\beta_0, \beta) \leftarrow (\beta_0, \beta) - \{\nabla^2 l(\beta_0, \beta)\}^{-1} \nabla l(\beta_0, \beta)$$

- If we differentiate the negative log-likelihood  $l(\beta_0, \beta) \in \mathbb{R} \times \mathbb{R}^{p+1}$  and we let  $v_i = e^{-y_i(\beta_0 + x_i \beta)}$ ,  $i = 1, \dots, N$
- The vector  $\nabla(\beta_0, \beta) \in \mathbb{R}^{p+1}$  such that the  $j$ th element is  $\frac{\partial l(\beta_0, \beta)}{\partial \beta_j}$  can be expressed by  $\nabla l(\beta_0, \beta) = -X^T u$  with

$$u = \begin{bmatrix} \frac{y_1 v_1}{1+v_1} \\ \vdots \\ \frac{y_N v_N}{1+v_N} \end{bmatrix}$$

- Note  $y_i = \pm 1$ , i.e.,  $y^2 = 1$ , the matrix  $\nabla^2 l(\beta_0, \beta)$  such that the  $(i, k)$ th element is  $\frac{\partial^2 l(\beta_0, \beta)}{\partial \beta_j \partial \beta_k}$ ,  $j, k = 0, 1, \dots, p$  can be expressed by  $\nabla^2 l(\beta_0, \beta) = X^T W X$

$$W = \begin{bmatrix} \frac{v_1}{(1+v_1)^2} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \frac{v_N}{(1+v_N)^2} \end{bmatrix}$$



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