

# Stochastic gradient descent for logistic regression smoothing

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## 1 The log-logistic regression model

We consider the four parameter log-logistic dose-response model of the real-valued random variable,  $Y$  given  $x$ . The model is defined as

$$f(x|\alpha, \beta, \gamma, \rho) = \gamma + \frac{\rho - \gamma}{1 + e^{\beta \log(x) - \alpha}}$$

With  $\alpha, \beta, \gamma, \rho \in \mathbb{R}$ . We observe the noised variables

$$Y_i = f(x_i|\alpha, \beta, \gamma, \rho) + \varepsilon_i$$

for  $i = 1, \dots, N$  and  $\varepsilon_i \sim \mathcal{N}(0, \sigma^2)$  independent. We estimate  $(\alpha, \beta, \gamma, \rho) \in \mathbb{R}^4$  by minimizing the non-linear least squares

$$H_N = \frac{1}{N} \sum_{i=1}^N (y_i - f(x_i|\alpha, \beta, \gamma, \rho))^2 = \frac{1}{N} \sum_{i=1}^N \left( y_i - \gamma + \frac{\gamma - \rho}{1 + e^{\beta \log(x_i) - \alpha}} \right)^2$$

We calculate the gradient of  $H_N$  by computing the partial derivatives w.r.t the all the parameters.

$$\begin{aligned} \frac{\partial H_N}{\partial \alpha} &= \frac{2}{N} \sum_{i=1}^N \left( y_i - \gamma + \frac{\gamma - \rho}{1 + e^{\beta \log(x_i) - \alpha}} \right) \frac{(\gamma - \rho) e^{\beta \log(x_i) - \alpha}}{(e^{\beta \log(x_i) - \alpha} + 1)^2} \\ \frac{\partial H_N}{\partial \beta} &= -\frac{2}{N} \sum_{i=1}^N \left( y_i - \gamma + \frac{\gamma - \rho}{1 + e^{\beta \log(x_i) - \alpha}} \right) \frac{(\gamma - \rho) \log(x_i) e^{\beta \log(x_i) - \alpha}}{(e^{\beta \log(x_i) - \alpha} + 1)^2} \\ \frac{\partial H_N}{\partial \gamma} &= \frac{2}{N} \sum_{i=1}^N \left( y_i - \gamma + \frac{\gamma - \rho}{1 + e^{\beta \log(x_i) - \alpha}} \right) \left( \frac{1}{1 + e^{\beta \log(x_i) - \alpha}} - 1 \right) \\ \frac{\partial H_N}{\partial \rho} &= -\frac{2}{N} \sum_{i=1}^N \left( y_i - \gamma + \frac{\gamma - \rho}{1 + e^{\beta \log(x_i) - \alpha}} \right) \left( \frac{1}{1 + e^{\beta \log(x_i) - \alpha}} \right) \end{aligned}$$