

$$E = \sum_{i=1}^N \exp[-y_i(f_{m-1}(x_i) + \beta G(x_i))]$$

$$E = \sum_{i=1}^N w_i^{(m)} \exp[-\beta y_i G(x_i)]$$

$$E = e^{-\beta} \cdot \sum_{y_i = G(x_i)} w_i^{(m)} + e^{\beta} \cdot \sum_{y_i \neq G(x_i)} w_i^{(m)}$$

$$E = (e^{\beta} - e^{-\beta}) \cdot \sum_{i=1}^N w_i^{(m)} I(y_i \neq G(x_i)) + e^{-\beta} \cdot \sum_{i=1}^N w_i^{(m)}$$

$$\frac{\partial E}{\partial \beta} = (e^{\beta} + e^{-\beta}) \cdot \sum_{i=1}^N w_i^{(m)} I(y_i \neq G(x_i)) - e^{-\beta} \cdot \sum_{i=1}^N w_i^{(m)} = 0$$

$$(e^{2\beta} + 1) \cdot \sum_{i=1}^N w_i^{(m)} I(y_i \neq G(x_i)) = \sum_{i=1}^N w_i^{(m)}$$

$$e^{2\beta} = \frac{\sum_{i=1}^N w_i^{(m)}}{\sum_{i=1}^N w_i^{(m)} I(y_i \neq G(x_i))} - 1$$

$$\beta = \frac{1}{2} \log \left[ \frac{\sum_{i=1}^N w_i^{(m)}}{\sum_{i=1}^N w_i^{(m)} I(y_i \neq G(x_i))} - 1 \right]$$

$$\beta = \frac{1}{2} \log \left[ \frac{1 - \text{err}_m}{\text{err}_m} \right]$$

$$\text{where } \text{err}_m = \frac{\sum_{i=1}^N w_i^{(m)} I(y_i \neq G_m(x_i))}{\sum_{i=1}^N w_i^{(m)}}$$