1. Обратное распространение ошибки из оригинальной статьи про Batch normalisation (https://arxiv.org/pdf/1502.03167.pdf):

$$\frac{\partial l}{\partial \hat{x}_{i}} = \frac{\partial l}{\partial y_{i}} \cdot \gamma$$

$$\frac{\partial l}{\partial \sigma_{B}^{2}} = \sum_{i=1}^{m} \frac{\partial l}{\partial \hat{x}_{i}} \cdot (x_{i} - \mu_{B}) \cdot \left(-\frac{1}{2}\right) \left(\sigma_{B}^{2} + \epsilon\right)^{-3/2}$$

$$\frac{\partial l}{\partial \mu_{B}} = \left(\sum_{i=1}^{m} \frac{\partial l}{\partial \hat{x}_{i}} \cdot \left(-\frac{1}{\sqrt{\sigma_{B}^{2} + \epsilon}} \right) \right) + \frac{\partial l}{\partial \sigma_{B}^{2}} \cdot \frac{\sum_{i=1}^{m} -2 (x_{i} - \mu_{B})}{m}$$

$$\frac{\partial l}{\partial x_{i}} = \frac{\partial l}{\partial \hat{x}_{i}} \cdot \frac{1}{\sqrt{\sigma_{B}^{2} + \varepsilon}} + \frac{\partial l}{\partial \sigma_{B}^{2}} \cdot \frac{2 (x_{i} - \mu_{B})}{m} + \frac{\partial l}{\partial \mu_{B}^{2}} \cdot \frac{1}{m}$$

$$\frac{\partial l}{\partial \gamma} = \sum_{i=1}^{m} \frac{\partial l}{\partial y_i} \cdot \hat{x}_i$$

$$\frac{\partial l}{\partial \beta} = \sum_{i=1}^{m} \frac{\partial l}{\partial y_i}$$

2. Соберем наиболее важные обозначения из оригинальной статьи:

$$\hat{X}_i = \frac{X_i - \mu_B}{\sigma_B + \epsilon} - z - \text{score ДЛЯ } X_i$$

$$y_i = \gamma x_i + \beta$$

$$\frac{\partial l}{\partial \hat{x}_{i}} = \frac{\partial l}{\partial y_{i}} \cdot \gamma$$

3. Разберем $\frac{\partial l}{\partial x_i}$ для упрощения вычслений обратного распространения ошибки :

$$\frac{\partial l}{\partial x_i} = \frac{\partial l}{\partial \hat{x}_i} \cdot \frac{1}{\sqrt{\sigma_B^2 + \epsilon}} + \frac{\partial l}{\partial \sigma_B^2} \cdot \frac{2 (x_i - \mu_B)}{m} + \frac{\partial l}{\partial \mu_B^2} \cdot \frac{1}{m} \left(\text{Развернём } \frac{\partial l}{\partial \sigma_B^2} \right)$$

$$\frac{\partial l}{\partial x_i} = \frac{\partial l}{\partial \hat{x}_i} \cdot \frac{1}{\sqrt{\sigma_B^2 + \epsilon}} + \Big[\sum_{i=1}^m \frac{\partial l}{\partial \hat{x}_i} \cdot (x_i - \mu_B) \cdot \left(-\frac{1}{2} \right) \left(\sigma_B^2 + \epsilon \right)^{-3/2} \Big] \cdot \frac{2 (x_i - \mu_B)}{m} + \frac{\partial l}{\partial \mu_B^2} \cdot \frac{1}{m}$$

$$\frac{\partial l}{\partial x_i} = \frac{\partial l}{\partial \hat{x}_i} \cdot \frac{1}{\sqrt{\sigma_B^2 + \varepsilon}} - \Big[\sum_{j=1}^m \frac{\partial l}{\partial \hat{x}_j} \cdot (x_j - \mu_B) \cdot \frac{1}{\left(\sqrt{\sigma_B^2 + \varepsilon}\right)^3} \Big] \cdot \frac{(x_i - \mu_B)}{m} + \frac{\partial l}{\partial \mu_B^2} \cdot \frac{1}{m}$$

$$\frac{\partial \, l}{\partial \, x_{\, i}} \, = \, \frac{\partial \, l}{\partial \, \hat{x}_{\, i}} \, \cdot \, \frac{1}{\sqrt{\sigma_B^2 + \varepsilon}} \, - \, \Big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \cdot \, \frac{(x_j - \mu_B)}{\sqrt{\sigma_B^2 + \varepsilon}} \, \cdot \, \frac{1}{\left(\sqrt{\sigma_B^2 + \varepsilon}\,\right)^2} \Big] \, \cdot \, \frac{(x_i - \mu_B)}{m} \, + \, \frac{\partial \, l}{\partial \, \mu_B^2} \, \cdot \, \frac{1}{m}$$

$$\frac{\partial}{\partial x_{i}} = \frac{\partial}{\partial \hat{x}_{i}} \cdot \frac{1}{\sqrt{\sigma_{0}^{2} + \epsilon}} - \left[\sum_{j=1}^{N} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \cdot \frac{1}{\left(\sqrt{\sigma_{0}^{2} + \epsilon} \right)^{2}} \right] \cdot \frac{(x_{i} - \mu_{0})}{m} + \frac{\partial}{\partial \mu_{0}^{2}} \cdot \frac{1}{m}$$

$$\frac{\partial}{\partial x_{i}} = \frac{\partial}{\partial \hat{x}_{i}} \cdot \frac{1}{\sqrt{\sigma_{0}^{2} + \epsilon}} - \frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] \cdot \hat{x}_{i} + \frac{\partial}{\partial \mu_{0}^{2}} \cdot \frac{1}{m} \quad \left[\operatorname{Passephēm} \frac{\partial}{\partial \mu_{0}^{2}} \right]$$

$$\frac{\partial}{\partial x_{i}} = \frac{\partial}{\partial \hat{x}_{i}} \cdot \frac{1}{\sqrt{\sigma_{0}^{2} + \epsilon}} - \frac{\hat{x}_{i}}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \cdot \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] +$$

$$\frac{1}{m} \cdot \left[\left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \left(-\frac{1}{\sqrt{\sigma_{0}^{2} + \epsilon}} \right) \right] + \frac{\partial}{\partial \sigma_{0}^{2}} \cdot \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] +$$

$$\frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \left(-\frac{1}{\sqrt{\sigma_{0}^{2} + \epsilon}} \right) \right] + \frac{\partial}{\partial \sigma_{0}^{2}} \cdot \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] -$$

$$\frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \left(-\frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \right) \right] + \frac{\partial}{\partial \sigma_{0}^{2}} \cdot \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] -$$

$$\frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \left(-\frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \right) \right] + \frac{\partial}{\partial \sigma_{0}^{2}} \cdot \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] -$$

$$\frac{\partial}{\partial x_{i}} = \frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] - \frac{1}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \right] \right] + \frac{(-2)}{m^{2}} \left[\frac{\partial}{\partial \sigma_{0}^{2}} \cdot \sum_{j=1}^{m} (x_{j} - \mu_{0}) \right]$$

$$\frac{\partial}{\partial x_{i}} = \frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\frac{\partial}{\partial \hat{x}_{i}} \cdot \frac{\hat{x}_{i}}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] - \frac{1}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \right] \right) + \frac{(-2)}{m^{2}} \left[\frac{\partial}{\partial \sigma_{0}^{2}} \cdot \left(\sum_{j=1}^{m} (x_{j} - \mu_{0}) \right]$$

$$\frac{\partial}{\partial x_{i}} = \frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\frac{\partial}{\partial \hat{x}_{i}} \cdot \frac{\hat{x}_{i}}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right] - \frac{1}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \right] \right] + \frac{(-2)}{m} \left[\frac{\partial}{\partial \sigma_{0}^{2}} \left(\sum_{j=1}^{m} (x_{j} - \mu_{0}) \right) \right]$$

$$\frac{\partial}{\partial x_{i}} = \frac{1}{m \sqrt{\sigma_{0}^{2} + \epsilon}} \left[\frac{\partial}{\partial x_{i}} \cdot \frac{\hat{x}_{i}}{m} \left[\sum_{j=1}^{m} \frac{\partial}{\partial \hat{x}_{j}} \cdot \hat{x}_{j} \right]$$

$$\begin{split} &\frac{\partial \, l}{\partial \, x_i} \, = \, \frac{1}{\sqrt{\sigma_B^2 + \varepsilon}} \, \left(\frac{\partial \, l}{\partial \, \hat{x}_i} - \frac{\hat{x}_i}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \hat{x}_j \, \big] - \frac{1}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \big] \right) + 0 \\ &\frac{\partial \, l}{\partial \, x_i} \, = \, \frac{1}{\sqrt{\sigma_B^2 + \varepsilon}} \, \left(\frac{\partial \, l}{\partial \, \hat{x}_i} - \frac{\hat{x}_i}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \hat{x}_j \, \big] - \frac{1}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \big] \right) \\ &\frac{\partial \, l}{\partial \, x_i} \, = \, \frac{1}{\sqrt{\sigma_B^2 + \varepsilon}} \, \left(\frac{m}{m} \, \frac{\partial \, l}{\partial \, \hat{x}_i} - \frac{\hat{x}_i}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \hat{x}_j \, \big] - \frac{1}{m} \, \big[\sum_{j=1}^m \frac{\partial \, l}{\partial \, \hat{x}_j} \, \big] \right) \end{split}$$

$$\frac{\partial \textbf{l}}{\partial \textbf{x}_{i}} = \frac{\textbf{1}}{\textbf{m} \sqrt{\sigma_{B}^{2} + \varepsilon}} \left(\textbf{m} \; \frac{\partial \textbf{l}}{\partial \hat{\textbf{x}}_{i}} - \hat{\textbf{x}}_{i} \left[\sum_{j=1}^{m} \frac{\partial \textbf{l}}{\partial \hat{\textbf{x}}_{j}} \; \hat{\textbf{x}}_{j} \right] - \left[\sum_{j=1}^{m} \frac{\partial \textbf{l}}{\partial \hat{\textbf{x}}_{j}} \right] \right)$$

Последнее выражение является упрощенной версией вычисления $\frac{\partial l}{\partial x}$.

Полезные материалы:



https://kevinzakka.github.io/2016/09/14/batch_normalization/http://cthorey.github.io./backpropagation/