

Градиент на функцията softmax

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Функцията softmax : $\mathbb{R}^n \rightarrow \mathbb{R}^n$ се дефинира като:

$$\text{softmax}(\mathbf{z})_i = \frac{e^{z_i}}{\sum_j e^{z_j}}$$

или векторно:

$$\text{softmax}(\mathbf{z}) = \frac{e^{\mathbf{z}}}{\sum_j e^{z_j}}$$

Градиентът ѝ е матрица $\in \mathbb{R}^{n \times n}$, чиито елементи са:

$$\begin{aligned} \left(\frac{\partial \text{softmax}(\mathbf{z})}{\partial \mathbf{z}} \right)_{ik} &= \frac{\partial \text{softmax}(\mathbf{z})_i}{\partial z_k} \\ &= \frac{\left(\frac{\partial e^{z_i}}{\partial z_k} \right) \left(\sum_j e^{z_j} \right) - e^{z_i} \left(\frac{\partial \sum_j e^{z_j}}{\partial z_k} \right)}{\left(\sum_j e^{z_j} \right)^2} \\ &= \frac{e^{z_i} \delta_{ik} \left(\sum_j e^{z_j} \right) - e^{z_i} e^{z_k}}{\left(\sum_j e^{z_j} \right)^2} \\ &= \frac{e^{z_i}}{\sum_j e^{z_j}} \left(\delta_{ik} - \frac{e^{z_k}}{\sum_j e^{z_j}} \right) \\ &= \text{softmax}(\mathbf{z})_i (\delta_{ik} - \text{softmax}(\mathbf{z})_k) \end{aligned}$$

Същата сметка може да се напише за целия ред на градиента:

$$\begin{aligned} \frac{\partial \text{softmax}(\mathbf{z})_i}{\partial \mathbf{z}} &= \frac{\left(\frac{\partial e^{z_i}}{\partial \mathbf{z}} \right) \left(\sum_j e^{z_j} \right) - e^{z_i} \left(\frac{\partial \sum_j e^{z_j}}{\partial \mathbf{z}} \right)}{\left(\sum_j e^{z_j} \right)^2} \\ &= \frac{e^{z_i} \text{eye}(i) \left(\sum_j e^{z_j} \right) - e^{z_i} e^{\mathbf{z}}}{\left(\sum_j e^{z_j} \right)^2} \\ &= \frac{e^{z_i}}{\sum_j e^{z_j}} \left(\text{eye}(i) - \frac{e^{\mathbf{z}}}{\sum_j e^{z_j}} \right) \\ &= \text{softmax}(\mathbf{z})_i (\text{eye}(i) - \text{softmax}(\mathbf{z})) \end{aligned}$$

, където $\text{eye}(i)$ е вектор с 1 на позиция i и 0 навсякъде другаде (`numpy.eye(n, k=i)`). Така може да напишем и матричния израз за градиента на softmax:

$$\frac{\partial \text{softmax}(\mathbf{z})}{\partial \mathbf{z}} = \text{softmax}(\mathbf{z})^T \cdot (\text{eye}(n) - \text{softmax}(\mathbf{z}))$$