

Задача 18. Пусть  $\sigma(z) = \frac{1}{1+e^{-z}}$  - сигмоида

Проверить:  $\sigma' = \sigma(1-\sigma)$

$$\sigma'(z) = \frac{-(-e^{-z})}{(1+e^{-z})^2} = \frac{e^{-z}}{(1+e^{-z})^2}$$

$$\begin{aligned}\sigma(z) \cdot (1-\sigma(z)) &= \frac{1}{1+e^{-z}} \cdot \left(1 - \frac{1}{1+e^{-z}}\right) = \frac{1}{1+e^{-z}} \cdot \left(\frac{1+e^{-z}-1}{1+e^{-z}}\right) = \\ &= \frac{1}{1+e^{-z}} \cdot \frac{e^{-z}}{1+e^{-z}} = \frac{e^{-z}}{(1+e^{-z})^2}\end{aligned}$$

Отсюда  $\sigma'(z) = \sigma(z) \cdot (1-\sigma(z))$

Задача 19 Задача классификации  $K$  классов  $\{1, 2, \dots, K\}$   
Последний слой нейросети вычисляет softmax функцию.

$$g_k(s_1, s_2, \dots, s_k) = \frac{e^{s_k}}{\sum_{l=1}^K e^{s_l}} - \text{softmax-функция}$$

В качестве потерь logloss-функция:

$$R^{(i)} = - \sum_{k=1}^K I(y^{(i)} = k) \cdot \ln g_k(s_1, s_2, \dots, s_k)$$

① Док-ть:  $\frac{\partial g_k}{\partial s_l} = g_k \cdot (I(k=l) - g_l)$

$$\frac{\partial g_k}{\partial s_l} = \frac{\partial}{\partial s_l} \left( \frac{e^{s_k}}{\sum_{j=1}^K e^{s_j}} \right) = e^{s_k} \cdot \frac{-e^{s_l}}{\left(\sum_{j=1}^K e^{s_j}\right)^2} = \frac{e^{s_k}}{\sum_{j=1}^K e^{s_j}} \cdot \frac{-e^{s_l}}{\sum_{j=1}^K e^{s_j}} \quad \text{②}$$



$$\textcircled{=} -g_k \cdot g_l = g_k (I(k=l) - g_l) \leftarrow \text{индикатор} = 0 \text{ (} k \neq l \text{)}$$

$$k=l: \frac{\partial g_k}{\partial s_k} = \frac{\partial}{\partial s_k} \left( \frac{l^{s_k}}{\sum_{j=1}^K l^{s_j}} \right) = \frac{l^{s_k}}{\sum_{j=1}^K l^{s_j}} + \frac{l^{s_k} \cdot (-l^{s_k})}{\left( \sum_{j=1}^K l^{s_j} \right)^2} =$$

$$= g_k - g_k^2 = g_k (1 - g_k) = g_k (I(k=l) - g_l) \leftarrow \text{индикатор} = 1$$

$$\textcircled{2} \text{ Док-мб: } \frac{\partial R^{(i)}}{\partial g_k} = - \frac{I(y^{(i)}=k)}{g_k}$$

$$\frac{\partial R^{(i)}}{\partial g_k} = \frac{\partial}{\partial g_k} \left( - \sum_{k=1}^K I(y^{(i)}=k) \cdot \ln g_k(s_1, s_2, \dots, s_K) \right) =$$

$$= - \sum_{j=1}^K I(y^{(i)}=j) \cdot \frac{\partial}{\partial g_k} (\ln g_j(s_1, s_2, \dots, s_K)) =$$

$$= I(y^{(i)}=1) \cdot \frac{\partial}{\partial g_k} \ln g_1 - I(y^{(i)}=2) \cdot \frac{\partial}{\partial g_k} \ln g_2 - \dots - I(y^{(i)}=k) \cdot \frac{\partial}{\partial g_k} \ln g_k - \dots$$

$$- I(y^{(i)}=k) \cdot \frac{\partial}{\partial g_k} \ln g_k = - I(y^{(i)}=k) \cdot \frac{\partial}{\partial g_k} \ln g_k =$$

$$= - I(y^{(i)}=k) \cdot \frac{1}{g_k}$$

$$\textcircled{3} \text{ Док-мб: } \frac{\partial R^{(i)}}{\partial s_l} = g_l - I(l=y^{(i)})$$

$$\frac{\partial R^{(i)}}{\partial s_l} = \sum_{j=1}^K \frac{\partial R^{(i)}}{\partial g_j} \cdot \frac{\partial g_j}{\partial s_l} = \sum_{j=1}^K \left( - \frac{I(y^{(i)}=j)}{g_j} \right) \cdot g_j (I(j=l) - g_l) =$$



$$= \sum_{j=1}^K I(y^{(i)}=j) \cdot (g_l - I(j=l)) = g_l \sum_{j=1}^K I(y^{(i)}=j) -$$

$$- \sum_{j=1}^K I(y^{(i)}=j) \cdot I(j=l) = g_l - I(y^{(i)}=l)$$