## Задание №18

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

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

## Задание №19

$$g_k(s_1, s_2, ..., s_K) = \frac{e^{s_k}}{\sum_{l=1}^K e^{s_l}}$$

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

$$\frac{\partial g_k}{\partial s_l} \stackrel{k \neq l}{=} \frac{\partial}{\partial s_l} \left( \frac{e^{s_k}}{\sum_{i=1}^K e^{s_i}} \right) = \frac{-e^{s_k} e^{s_l}}{(\sum_{i=1}^K e^{s_i})^2} = \frac{e^{s_k}}{\sum_{i=1}^K e^{s_i}} \left( -\frac{e^{s_l}}{\sum_{i=1}^K e^{s_i}} \right) = g_k(0 - g_l)$$

$$\frac{\partial g_k}{\partial s_l} \stackrel{k=l}{=} \frac{\partial g_l}{\partial s_l} = \frac{\partial}{\partial s_l} \left( \frac{e^{s_l}}{\sum_{i=1}^K e^{s_i}} \right) = e^{s_l} \left( \sum_{i=1}^K e^{s_i} \right)^{-1} + e^{s_l} \left( -\frac{e^{s_l}}{(\sum_{i=1}^K e^{s_i})^2} \right) =$$

$$= \frac{e^{s_l}}{\sum_{i=1}^K e^{s_i}} \left( 1 - \frac{e^{s_l}}{\sum_{i=1}^K e^{s_i}} \right) = g_l(1 - g_l) \stackrel{l=k}{=} g_k(1 - g_l)$$

$$\frac{\partial g_k}{\partial s_l} = g_k(I(k = l) - g_l)$$

$$\begin{split} \frac{\partial R^{(i)}}{\partial g_k} &= \frac{\partial}{\partial g_k} \left( -\sum_{k=1}^K I(y^{(i)} = k) \cdot \ln g_k \right) = \frac{\partial}{\partial g_k} (-I(y^{(i)}) \ln g_k) = \\ &= -I(y^{(i)}) \frac{\partial}{\partial g_k} \sum_{k=1}^K \ln g_k = -I(y^{(i)}) \cdot \frac{1}{g_k} \\ \frac{\partial R^{(i)}}{\partial g_k} &= -\frac{I(y^{(i)})}{g_k} \end{split}$$

3)

$$R^{(i)} = -\sum_{k=1}^{K} I(y^{(i)} = k) \ln g_k = -\sum_{k=1}^{K} I(y^{(i)} = k) \ln \left(\frac{e^{s_k}}{(\sum_{i=1}^{K} e^{s_i})^2}\right) =$$

$$= -\sum_{k=1}^{K} I(y^{(i)} = k) (\ln e^{s_k} - \ln(\sum_{i=1}^{K} e^{s_i})) = -\sum_{k=1}^{K} I(y^{(i)} = k) (s_k - \ln(\sum_{i=1}^{K} e^{s_i}))$$

$$\frac{\partial R^{(i)}}{\partial s_l} = \frac{\partial}{\partial s_l} \left(-\sum_{k=1}^{K} I(y^{(i)} = k) (s_k - \ln(\sum_{i=1}^{K} e^{s_i})\right)$$

$$\frac{\partial R^{(i)}}{\partial s_l} \stackrel{k=l}{=} \frac{\partial}{\partial s_l} (-I(y^{(i)} = l) (s_l - \ln(\sum_{i=1}^{K} e^{s_i})) = -I(y^{(i)} = l) \left(1 - \frac{e^{s_l}}{\sum_{i=1}^{K} e^{s_i}}\right) =$$

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

$$\frac{\partial R^{(i)}}{\partial s_l} \stackrel{k\neq l}{=} \frac{\partial}{\partial s_l} (-I(y^{(i)} = k) (s_k - \ln(\sum_{i=1}^{K} e^{s_i})) = -I(y^{(i)} = k) \left(-\frac{e^{s_l}}{\sum_{i=1}^{K} e^{s_i}}\right) =$$

$$= I(y^{(i)} = l) \cdot g_l$$

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

## Задание №21

$$R^{(i)} = \operatorname{logloss}(\operatorname{softmax}(B(\sigma(Ax))))$$
Обозначим функцию logloss как  $L$ , функцию softmax как  $g$ . Тогда: 
$$R^{(i)} = L(g(B \cdot \sigma(Ax)))$$

$$v(x) = B \cdot \sigma(Ax)$$

$$w(x) = Ax$$

$$R^{(i)} = L(g(v(x)))$$

$$\frac{\partial R^{(i)}}{\partial x} = \frac{\partial L}{\partial g} \frac{\partial g}{\partial x} = \frac{\partial L}{\partial g} \frac{\partial g}{\partial v} \frac{\partial v}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial v}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial \sigma} \frac{\partial \sigma}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial \sigma} \frac{\partial \sigma}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial \sigma} \frac{\partial \sigma}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial \sigma} \frac{\partial \sigma}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial (B \cdot \sigma(w))}{\partial \sigma} \frac{\partial \sigma}{\partial x} = \frac{\partial L}{\partial v} \frac{\partial \sigma(Ax)}{\partial x} = \frac{\partial \sigma(Ax)}{\partial x} =$$