

Deep Learning Assignment 1

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Abstract. This is the report for deep learning assignment 1

1. Warmup. Write $\frac{\partial E}{\partial X_{in}}$ in terms of $\frac{\partial E}{\partial X_{out}}$

$$\frac{\partial E}{\partial X_{in}} = \frac{\partial E}{\partial X_{out}} \frac{\partial F(X_{in}, W_i)}{\partial X_{in}} = \frac{\partial E}{\partial X_{out}} \frac{e^{X_{in}}}{(1 + e^{X_{in}})^2} = \frac{\partial E}{\partial X_{out}} X_{out}(1 - X_{out}) \quad (1.1)$$

2. Multinomial logistic regression. Write the expression of $\frac{\partial(X_{out})_i}{\partial(X_{in})_j}$ if $i = j$, , and let $C = \sum_k e^{(X_I)_k} - e^{(X_I)_i}$

$$(X_o)_i = \frac{e^{(X_I)_i}}{\sum_k e^{(X_I)_k}} = \frac{e^{(X_I)_i}}{e^{(X_I)_0} + e^{(X_I)_i} + \dots e^{(X_I)_i} + \dots + e^{(X_I)_k}} = \frac{e^{(X_I)_i}}{C + e^{(X_I)_i}} \quad (2.1)$$

$$\frac{\partial(X_o)_i}{\partial(X_I)_i} = \frac{\partial}{\partial(X_I)_i} \left(\frac{e^{(X_I)_i}}{C + e^{(X_I)_i}} \right) = \frac{-\beta e^{-\beta(X_I)_i}}{C + e^{(X_I)_i}} + \frac{\beta e^{-2\beta(X_I)_i}}{(C + e^{(X_I)_i})^2} = \beta X_o(-1 + X_o) \quad (2.2)$$

if $i \neq j$, and let $K = \sum_k e^{(X_I)_k} - e^{(X_I)_j}$

$$\frac{\partial(X_o)_i}{\partial(X_I)_j} = \frac{\partial}{\partial(X_I)_j} \left(\frac{e^{(X_I)_i}}{K + e^{(X_I)_j}} \right) = \frac{\beta e^{-\beta(X_I)_i} e^{-\beta(X_I)_j}}{(K + e^{(X_I)_j})^2} = \beta(X_o)_i(X_o)_j \quad (2.3)$$

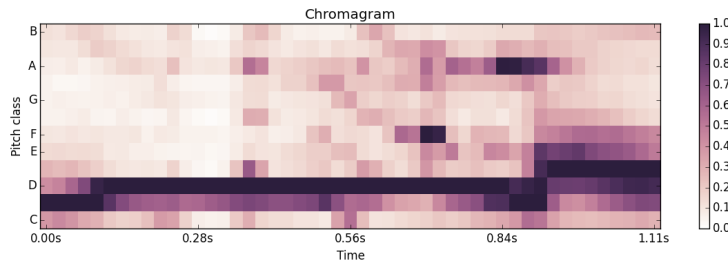


Figure 2.1. Histogram of accuracy result from search space. Top row is baseline method 5, and the remaining rows are new feature extraction method 1, 2, and 3 respectively

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