## Appendix A DiVe Learning Equations

In this section we will derive the equations that will compose the learning of DiVe. The steps will only be from DiVe Single Point, this steps for DiVe Dual Point will follow the same principles.

## A.1 Activation Sigmoid

We will minimize our  $NLL_{NS}(\mathcal{D})$  function using the descending gradient technique, so for each variable you have to find the  $\Delta$  gradient, thus considering  $NLL_{NS}(\mathcal{D})$  as:

$$NLL_{NS}(\mathcal{D}) = -\sum_{s_n \in D} \sum_{i=j}^{k_{s_n}} \left[ \log \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) - \sum_{v \in \mathcal{V}'} \log \sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(1)

Reminder that:

$$\sigma(x) = \frac{1}{1 + e^{-x}} \tag{2}$$

$$f(\mathcal{X}(w_i), \mathcal{X}(c_i)) = -\frac{1}{2}||\mathcal{X}(w_i) - \mathcal{X}(c_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(w_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(c_i)||_2^2$$
(3)

Thus, the gradient for the  $\mathcal{X}(w_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(w_i)} \text{NLL}(\mathcal{D}) = \left[ -\log \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \sum_{v \in \mathcal{V}'} \log \sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(4)

$$= \left[ -\Delta_{\mathcal{X}(w_i)} \log \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \Delta_{\mathcal{X}(w_i)} \sum_{v \in \mathcal{V}'} \log \sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(5)

$$= \left[ -\frac{\Delta_{\mathcal{X}(w_i)} \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))}{\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} + 0 \right]$$
(6)

$$= \left[ -\frac{(1 - \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))))\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))\Delta_{\mathcal{X}(w_i)}f(\mathcal{X}(w_i), \mathcal{X}(c_i))}{\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} \right]$$
(7)

$$= -(1 - \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))))[-\mathcal{X}(w_i) + \mathcal{X}(c_i) + \alpha \mathcal{X}(w_i)]$$
(8)

Thus, the gradient for the  $\mathcal{X}(c_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(c_i)} \text{NLL}(\mathcal{D}) = \left[ -\log \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \sum_{v \in \mathcal{V}'} \log \sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(9)

$$= \left[ -\Delta_{\mathcal{X}(c_i)} \log \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \Delta_{\mathcal{X}(c_i)} \sum_{v \in \mathcal{V}'} \log \sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(10)

$$= \left[ -\frac{\Delta_{\mathcal{X}(c_i)}\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))}{\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} + \sum_{v \in \mathcal{V}'} \frac{\Delta_{\mathcal{X}(c_i)}\sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}{\sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))} \right]$$
(11)

$$= -\frac{(1 - \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))))\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))\Delta_{\mathcal{X}(c_i)}f(\mathcal{X}(w_i), \mathcal{X}(c_i))}{\sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} + \sum_{v \in \mathcal{Y}} \frac{-\sigma(f(\mathcal{X}(w_v), \mathcal{X}(c_i)))\sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) - \Delta_{\mathcal{X}(c_i)}f(\mathcal{X}(w_v), \mathcal{X}(c_i))}{\sigma(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}$$

$$(12)$$

$$= -(1 - \sigma(f(\mathcal{X}(w_i), \mathcal{X}(c_i))))[\mathcal{X}(w_i) - \mathcal{X}(c_i) + \alpha \mathcal{X}(c_i)]$$

$$+ \sum_{v \in \mathcal{V}'} -\sigma(f(\mathcal{X}(w_v), \mathcal{X}(c_i)))[-(\mathcal{X}(w_v) - \mathcal{X}(c_i) + \alpha \mathcal{X}(c_i))]$$
(13)

## A.2 Activation Tanh

$$NLL_{NS}(\mathcal{D}) = -\sum_{s_n \in D} \sum_{i=j}^{k_{s_n}} \left[ \log \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) - \sum_{v \in \mathcal{V}'} \log \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(14)

$$f(\mathcal{X}(w_i), \mathcal{X}(c_i)) = -\frac{1}{2}||\mathcal{X}(w_i) - \mathcal{X}(c_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(w_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(c_i)||_2^2$$
(15)

Thus, the gradient for the  $\mathcal{X}(w_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(w_i)} \text{NLL}(\mathcal{D}) = \left[ -\log \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \sum_{v \in \mathcal{V}'} \log \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(16)

$$= \left[ -\Delta_{\mathcal{X}(w_i)} \log \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \Delta_{\mathcal{X}(w_i)} \sum_{v \in \mathcal{V}'} \log \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(17)

$$= \left[ -\frac{\Delta_{\mathcal{X}(w_i)} \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} + 0 \right]$$
(18)

$$= \left[ -\frac{(1 - \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))^2) \Delta_{\mathcal{X}(w_i)} f(\mathcal{X}(w_i), \mathcal{X}(c_i))}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} \right]$$
(19)

$$= -\frac{(1 - \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))^2)}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} [-\mathcal{X}(w_i) + \mathcal{X}(c_i) + \alpha \mathcal{X}(w_i)]$$
(20)

Thus, the gradient for the  $\mathcal{X}(c_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(c_i)} \text{NLL}(\mathcal{D}) = \left[ -\log \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \sum_{v \in \mathcal{V}'} \log \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(21)

$$= \left[ -\mathbf{\Delta}_{\mathcal{X}(c_i)} \log \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \mathbf{\Delta}_{\mathcal{X}(c_i)} \sum_{v \in \mathcal{V}'} \log \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(22)

$$= \left[ -\frac{\Delta_{\mathcal{X}(c_i)} \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} + \sum_{v \in \mathcal{V}'} \frac{\Delta_{\mathcal{X}(c_i)} \tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}{\tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))} \right]$$
(23)

$$= -\frac{(1 - \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))^2) \Delta_{\mathcal{X}(c_i)} f(\mathcal{X}(w_i), \mathcal{X}(c_i))}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i))} + \sum_{v \in \mathcal{V}'} \frac{-\frac{4 \exp(2 * f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}{(\exp(2 * f(\mathcal{X}(w_v), \mathcal{X}(c_i))) + 1)^2} - \Delta_{\mathcal{X}(c_i)} f(\mathcal{X}(w_v), \mathcal{X}(c_i))}{\tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}$$

$$(24)$$

$$= -\frac{(1 - \tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))^2)}{\tanh(f(\mathcal{X}(w_i), \mathcal{X}(c_i)))} [\mathcal{X}(w_i) - \mathcal{X}(c_i) + \alpha \mathcal{X}(c_i)] + \sum_{v \in \mathcal{V}'} \frac{-\frac{4 \exp(2 * f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}{(\exp(2 * f(\mathcal{X}(w_v), \mathcal{X}(c_i))) + 1)^2} [-(\mathcal{X}(w_v) - \mathcal{X}(c_i) + \alpha \mathcal{X}(c_i))]}{\tanh(-f(\mathcal{X}(w_v), \mathcal{X}(c_i)))}$$

$$(25)$$

## A.3 Activation Exp

$$NLL_{NS}(\mathcal{D}) = -\sum_{s_n \in D} \sum_{i=j}^{k_{s_n}} \left[ \log \exp(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) - \sum_{v \in \mathcal{V}'} \log \exp(-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(26)

Reminder that:

$$f(\mathcal{X}(w_i), \mathcal{X}(c_i)) = -\frac{1}{2}||\mathcal{X}(w_i) - \mathcal{X}(c_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(w_i)||_2^2 + \frac{\alpha}{2}||\mathcal{X}(c_i)||_2^2$$
(27)

Thus, the gradient for the  $\mathcal{X}(w_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(w_i)} \text{NLL}(\mathcal{D}) = \left[ -f(\mathcal{X}(w_i), \mathcal{X}(c_i)) + \sum_{v \in \mathcal{V}'} -f(\mathcal{X}(w_v), \mathcal{X}(c_i)) \right]$$
(28)

$$= \left[ -\Delta_{\mathcal{X}(w_i)}(f(\mathcal{X}(w_i), \mathcal{X}(c_i))) + \Delta_{\mathcal{X}(w_i)} \sum_{v \in \mathcal{V}'} (-f(\mathcal{X}(w_v), \mathcal{X}(c_i))) \right]$$
(29)

$$= -[-\mathcal{X}(w_i) + \mathcal{X}(c_i) + \alpha \mathcal{X}(w_i)] \tag{30}$$

Thus, the gradient for the  $\mathcal{X}(c_i)$  vector considering a batch size of 1:

$$\Delta_{\mathcal{X}(c_{i})} \text{NLL}(\mathcal{D}) = -(f(\mathcal{X}(w_{i}), \mathcal{X}(c_{i}))) + \sum_{v \in \mathcal{V}'} (-f(\mathcal{X}(w_{v}), \mathcal{X}(c_{i})))$$

$$= [-\Delta_{\mathcal{X}(c_{i})} (f(\mathcal{X}(w_{i}), \mathcal{X}(c_{i}))) + \Delta_{\mathcal{X}(c_{i})} \sum_{v \in \mathcal{V}'} (-f(\mathcal{X}(w_{v}), \mathcal{X}(c_{i})))]$$

$$= -[\mathcal{X}(w_{i}) - \mathcal{X}(c_{i}) + \alpha \mathcal{X}(c_{i})] + \sum_{v \in \mathcal{V}'} -[\mathcal{X}(w_{i}) - \mathcal{X}(c_{i}) + \alpha \mathcal{X}(c_{i})]$$
(31)