

# Pushforward / Jvp rule for the $L_2$ -loss

↳ squared  $L_2$  norm

$$\ell(\underline{y}, \underline{y}^r) = \frac{1}{2} \|\underline{y} - \underline{y}^r\|_2^2 =: \mathcal{L}$$

$$\underline{y} \in \mathbb{R}^N$$

$$\underline{y}^r \in \mathbb{R}^N$$

$$\mathcal{L} \in \mathbb{R}$$

$$\ell: \mathbb{R}^N \times \mathbb{R}^N \rightarrow \mathbb{R}$$

↳ typical in regression

task: forward propagate tangents on the input

$\underline{\dot{y}} \in \mathbb{R}^N$  &  $\underline{\dot{y}}^r \in \mathbb{R}^N$  to the output  $\dot{\mathcal{L}} \in \mathbb{R}$

in general

$$\dot{\mathcal{L}} = \left( \frac{\partial \ell}{\partial \underline{y}} \right)^T \underline{\dot{y}} + \frac{\partial \ell}{\partial \underline{y}^r} \underline{\dot{y}}^r$$

$$\frac{\partial \ell}{\partial \underline{y}} = \frac{\partial}{\partial \underline{y}} \left( \frac{1}{2} \|\underline{y} - \underline{y}^r\|_2^2 \right) = \frac{\partial}{\partial \underline{y}} \left( \frac{1}{2} (\underline{y} - \underline{y}^r)^T (\underline{y} - \underline{y}^r) \right)$$

index notation

$$\frac{\partial \ell}{\partial y_i} = \frac{\partial}{\partial y_i} \left( \frac{1}{2} \sum_{j=1}^N (y_j - y_j^r)^2 \right)$$

$$= \frac{1}{2} \sum_{j=1}^N \frac{\partial}{\partial y_i} (y_j - y_j^r)^2$$

$$= \frac{1}{2} \sum_{j=1}^N 2 (y_j - y_j^r) \left( \frac{\partial y_j}{\partial y_i} \right)$$

$= \delta_{ji}$

$$\left( \frac{\partial -y_j^r}{\partial y_i^r} \right)$$

$$= \frac{1}{2} \sum_{j=1}^N 2 (y_j - y_j^r) \delta_{ji}$$

$$= \frac{1}{2} 2 (y_i - y_i^r)$$

$$= y_i - y_i^r$$

back to symbolic notation

$$\frac{\partial \ell}{\partial \underline{y}} = (\underline{y} - \underline{y}^r)^T$$

$$\frac{\partial \ell}{\partial \underline{y}^r} = (\underline{y}^r - \underline{y})^T = -(\underline{y} - \underline{y}^r)^T$$

$$\dot{\mathcal{L}} = (\underline{y} - \underline{y}^r)^T \underline{\dot{y}} + (\underline{y}^r - \underline{y})^T \underline{\dot{y}}^r$$

Full pushforward rule

$$\mathcal{J}(\ell, (\underline{y}, \underline{y}^r), (\underline{\dot{y}}, \underline{\dot{y}}^r)) = \left( \underbrace{\left( \frac{1}{2} \|\underline{y} - \underline{y}^r\|_2^2 \right)}_{\mathcal{L}}, \underbrace{\left( (\underline{y} - \underline{y}^r)^T \underline{\dot{y}} + (\underline{y}^r - \underline{y})^T \underline{\dot{y}}^r \right)}_{\dot{\mathcal{L}}} \right)$$

often this  $\underline{\dot{y}}^r$  is 0

$$\Delta \underline{y} = \underline{y} - \underline{y}^r$$