

1 Natural gradient - alternate derivation

$$\begin{aligned}
 de &= \sum_{i,j} \frac{\partial e}{\partial w_{ij}} dw_{ij} \\
 &= \sum_{i,j} \varphi_i x_j dw_{ij} + \sum_{i,j} (\underline{\mathbf{w}}^{-1})_{ji} dw_{ij} \\
 &= \underline{\varphi}^T d\underline{\mathbf{w}}\underline{\mathbf{x}} + \text{Tr}(d\underline{\mathbf{w}} \cdot \underline{\mathbf{w}}^{-1}) \\
 &= \underline{\varphi}^T \underbrace{(d\underline{\mathbf{w}} \cdot \underline{\mathbf{w}}^{-1}) \overbrace{\widehat{\underline{\mathbf{s}}}^{\widehat{\underline{\mathbf{s}}=\underline{\mathbf{w}}\underline{\mathbf{x}}}}} + \text{Tr}(d\underline{\mathbf{w}} \cdot \underline{\mathbf{w}}^{-1})}_{d\underline{\mathbf{Z}}}
 \end{aligned} \tag{1}$$

with

$$d\underline{\mathbf{Z}} = d\underline{\mathbf{w}} \cdot \underline{\mathbf{w}}^{-1} \tag{2}$$

we obtain

$$de = \underline{\varphi}^T \widehat{\underline{\mathbf{s}}} d\underline{\mathbf{Z}} + \text{Tr}(d\underline{\mathbf{Z}}) \tag{3}$$

gradient ascent learning

$$\begin{aligned}
 \Delta Z_{ij} &= \eta \frac{\partial e}{\partial z_{ij}} \\
 &= \eta (\varphi_i \widehat{s}_j + \delta_{ij}) \\
 &= \eta \left(\varphi_i \sum_k w_{jk} x_k + \delta_{ij} \right) \\
 &= \sum_k \Delta w_{jk} (\underline{\mathbf{w}}^{-1})_{kj}
 \end{aligned} \tag{4}$$

$$\Delta w_{il} = \eta \left(w_{il} + \varphi_i \sum_{k,j} w_{jl} w_{jk} x_k \right) \tag{5}$$