

it layer (input)

ith layer (Hidden) Lth layer (suppt)

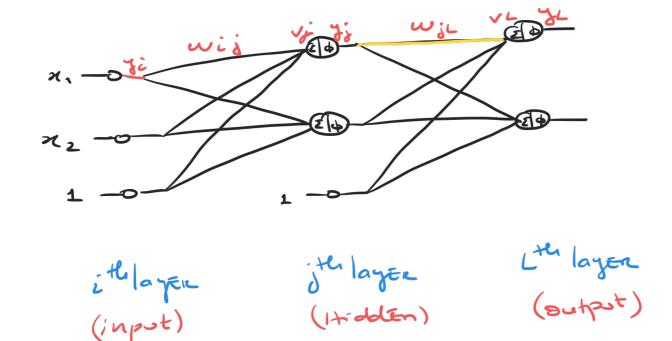
OBJECTIVE fet:
$$J(\omega) = \frac{1}{2} \sum_{i=1}^{N} \varepsilon_{i}^{2}$$

output layer). Ei = ti - yi Eznor (ct

$$Ei = ti - gi$$

output
$$J_L = \Phi(\sum_i w_{iL}, J_i) = \Phi(v_L)$$

NO 2

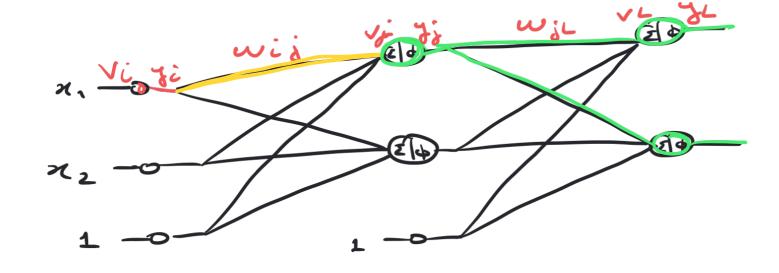


$$\frac{\partial J}{\partial w_{jl}} = \frac{\partial J}{\partial E_{L}} \cdot \frac{\partial E_{L}}{\partial y_{L}} \cdot \frac{\partial y_{L}}{\partial v_{L}} \cdot \frac{\partial v_{L}}{\partial w_{jl}}$$

$$= E_{L} \cdot (-1) \cdot \varphi^{2}(v_{L}) \cdot y_{j}$$

online tearning
it uses a single sample to make constections

Gradient Descent: (t) (t)



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accumulate

Error stern

 $S_i = -\frac{2J}{2v_i}$

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in the output legton

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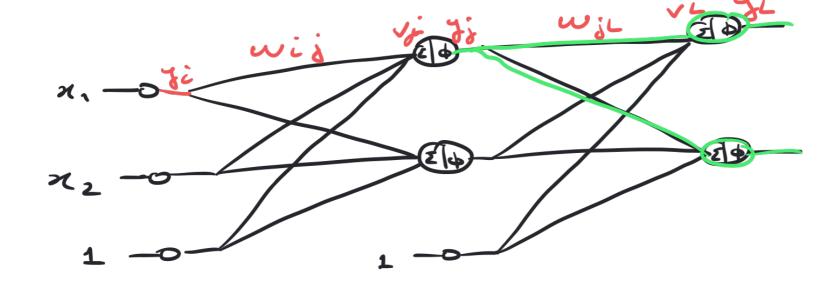
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- SL

 $\frac{\partial T}{\partial w_{jl}} = -S_{l}. \frac{\partial v_{l}}{\partial w_{jl}} = -S_{l}. \gamma_{i}$

in the Now, for the weights hidden lagen, wij.

 $\frac{\partial J}{\partial w_{ij}} = -J_{ij} \cdot J_{ij} = \phi(v_{ij})$ $\frac{\partial J}{\partial w_{ij}} = -J_{ij} \cdot J_{ij} = \phi(v_{ij})$



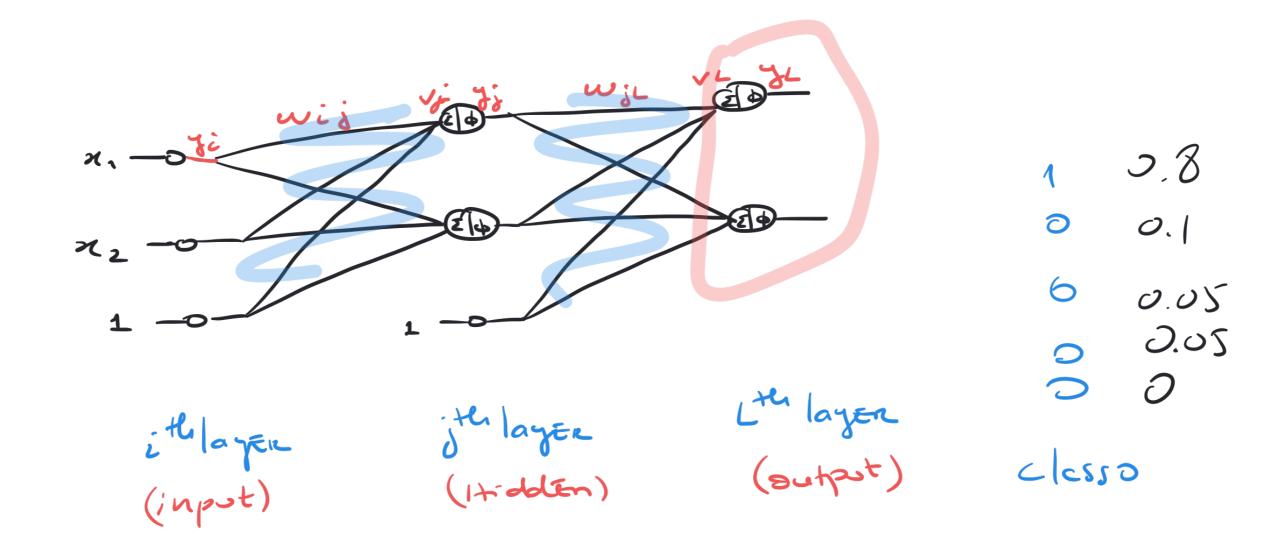
$$\frac{\partial J}{\partial j_{0}} = \sum_{L}^{J} \frac{\partial J}{\partial E_{L}} \cdot \frac{\partial E_{L}}{\partial \gamma_{L}} \cdot \frac{\partial \gamma_{L}}{\partial \gamma_{L}} \cdot \frac{\partial \gamma_{L}}{\partial \gamma_{0}} \cdot \frac{\partial \gamma_{L}}{\partial \gamma_{0}$$

Putting it all together

$$\delta_{\delta} = \left(\sum_{L} \in_{L} , \varphi'(v_{L}), \omega_{\delta L} \right), \varphi'(v_{\delta})$$

DJ - Sj. Zi Dwij

$$=-(2=(4)(v_{i}).\psi_{i}).\psi_{i}).y_{i}$$



$$2 - z - z$$