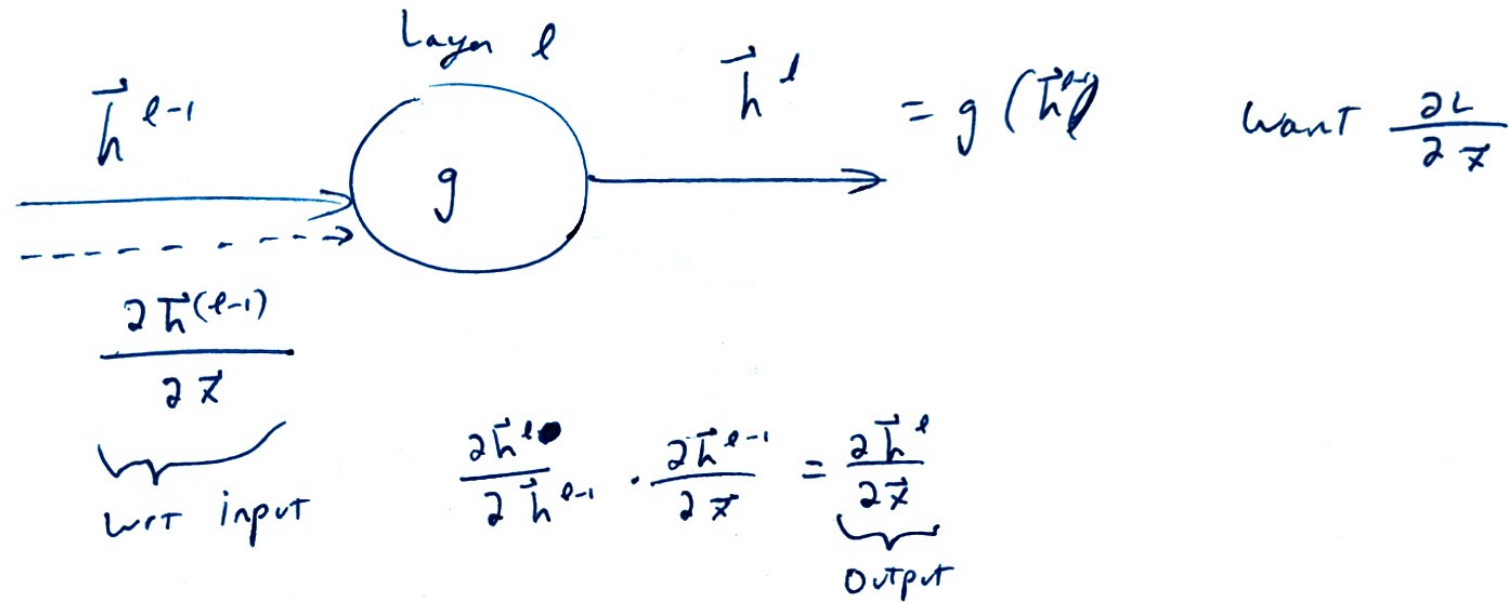
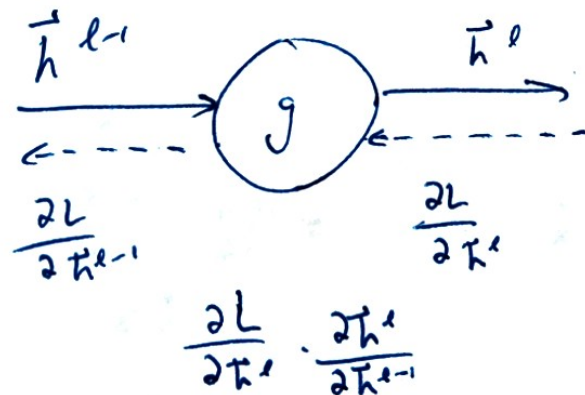


Forward Mode



Reverse Mode



Key Difference :

- Forward propagate derivative output wrt input
 (At end output is loss)
 • No need for 2 passes, but
- Backwards propagate input ^{# passes} derivative of loss wrt. input
 • Increased storage in proportion to # ops
 Backward mode preferred for $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ $n \gg m$

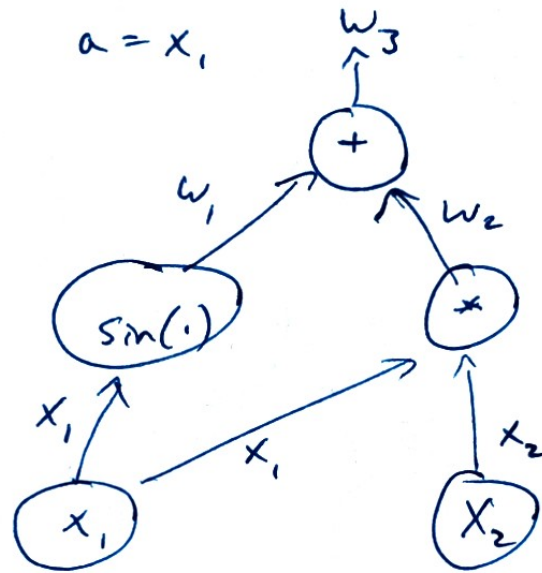
20 Example Forward Mode AD

$$F(x_1, x_2) = x_1 x_2 + \sin(x_1)$$

$$\frac{\partial F}{\partial x} = \left[\frac{\partial F}{\partial x_1} \quad \frac{\partial F}{\partial x_2} \right]$$

Pick one input: $a = x_1$

Label intermediate



Add-distributor
 Max-router
 Mul-switcher

Calculate derivative of each variable w.r.t $x_1 (=a)$

$$\left(= \frac{\partial x_i}{\partial a} \right)$$

$$\dot{x}_1 = \frac{\partial x_1}{\partial a} = 1$$

$$\dot{x}_2 = \frac{\partial x_2}{\partial a} = 0$$

$$\dot{w}_1 = \frac{\partial w_1}{\partial a} = \frac{\partial w_1}{\partial x_1} \cdot \frac{\partial x_1}{\partial a} = \cos(x_1) \dot{x}_1$$

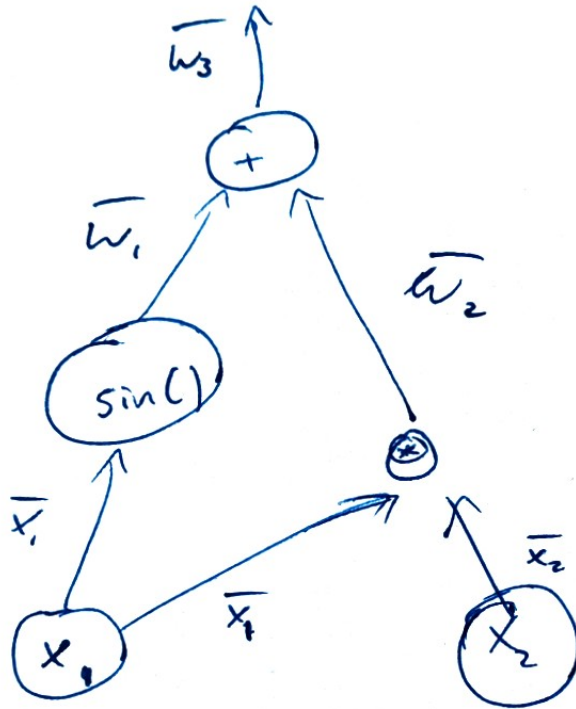
$$\dot{w}_2 = \dot{x}_1 x_2 + x_1 \dot{x}_2$$

$$\dot{w}_3 = \dot{w}_1 + \dot{w}_2$$

$$\left[\frac{\partial w_2}{\partial a} = \frac{\partial w_2}{\partial x_1} \cdot \frac{\partial x_1}{\partial a} + \frac{\partial w_2}{\partial x_2} \cdot \frac{\partial x_2}{\partial a} \right]$$

$$\left[\frac{\partial w_3}{\partial w_1} \cdot \frac{\partial w_1}{\partial x_1} + \frac{\partial w_3}{\partial w_2} \cdot \frac{\partial w_2}{\partial x_1} \right]$$

(21) Reverse Mode AD



$$\bar{w}_3 = \frac{\partial F}{\partial w_3} = 1$$

Notice: Can choose order of evaluation!

$$\bar{w}_1 = \bar{w}_3 \quad \frac{\partial F}{\partial w_1} = \frac{\partial F}{\partial w_3} \cdot \frac{\partial w_3}{\partial w_1} = 1$$

$$\bar{w}_2 = \bar{w}_3$$

$$\bar{x}_1 = \bar{w}_1 \cdot \cos(x_1) \quad \frac{\partial F}{\partial w_1} = \frac{\partial F}{\partial w_1} \cdot \frac{\partial w_1}{\partial x_1} = \bar{w}_1 \cdot \cos(x_1)$$

$$\bar{x}_1 = \bar{w}_2 \cdot x_2$$

$$\bar{x}_2 = \bar{w}_2 \cdot x_1 \quad \frac{\partial F}{\partial x_2} = \frac{\partial F}{\partial w_2} \cdot \frac{\partial w_2}{\partial x_2} = \bar{w}_2 \cdot x_1$$

Duplicate x_1 .

Computes gradients for all inputs!

Add - distributor

Max - router

Multiply - switches