

# Machine Learning

## Multivariate Chain Rule

**Mostafa S. Ibrahim**

*Teaching, Training and Coaching for more than a decade!*

*Artificial Intelligence & Computer Vision Researcher*

*PhD from Simon Fraser University - Canada*

*Bachelor / MSc from Cairo University - Egypt*

*Ex-(Software Engineer / ICPC World Finalist)*



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# Function Composition

- An operation where two functions say  $f$  and  $g$  generate a new function say  $h$  in such a way that  $h(x) = g(f(x))$ .
- It means here function  $g$  is applied to **the output of** function of  $x$
- Example:  $y = \sin(\text{sigmoid}(\text{sqrt}(x)))$ 
  - Given  $x$
  - Compute  $s = \text{sqrt}(x)$
  - Then Compute  $t = \text{sigmoid}(s)$
  - Then Compute  $y = \sin(t)$

# Chain Rule

- A rule that makes our life easy when we compute the derivative of a composition of functions
- Example:
  - Let  $y = \sin(\text{sigmoid}(\text{sqrt}(x)))$
  - Compute  $\partial y / \partial x$
- Rule

- $$\frac{d}{dx}[f(g(x))] = f'(g(x))g'(x)$$

$$\frac{d}{dx}[f(g(h(x)))] = f'(g(h(x)))g'(h(x))h'(x)$$

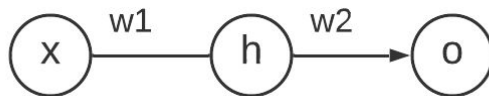
# Example 1

- Compute  $\partial y / \partial x$  where  $y = (3x+5)^4$
- Use series of **symbols** and compute partial derivatives relative to them then multiply their results
  - $y = a^4$
  - $a = 3x+5$
- Compute  $\partial y / \partial x = \partial y / \partial a * \partial a / \partial x$
- $\partial y / \partial a = 4a^3$
- $\partial a / \partial x = 3$
- $\partial y / \partial x = 4a^3 * 3 = 4(3x+5)^3 * 3 = 12 (3x+5)^3$

## Example 2

- Compute  $\partial y / \partial x$  where  $y = 2x^3 + (3x+5)^4$
- The rule here just **add** the parts together
- $\partial / \partial x 2x^3 + \partial / \partial x (3x+5)^4$
- $6x^2 + 12 (3x+5)^3$

# Example 3



Assume h and o are followed by activation  
 $f(a) = a^3$

$$E = (o - t)^2$$

Compute  $\partial E / \partial w_1$

- Let's express E ( a simple NN) fully mathematically

- $E = (f(f(x * w_1) * w_2) - t)^2$

- Express as series of symbols

[tip start from the inner  $x * w_1 = a$ ]

where  $d = f(f(x * w_1) * w_2)$

- $E = (d - t)^2$

- $d = f(c)$

- $c = b * w_2$

- $b = f(a)$

- $a = x * w_1$

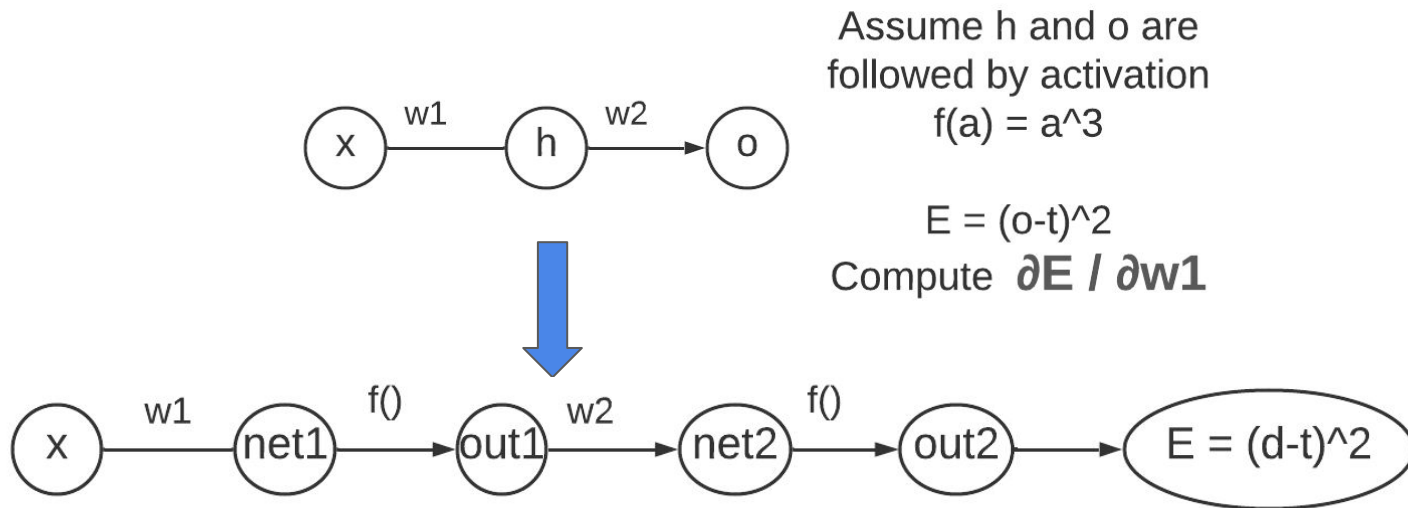
where  $c = f(x * w_1) * w_2$

node h represents **2 operations**: a and b

- $\partial E / \partial w_1 = \partial E / \partial d * \partial d / \partial c * \partial c / \partial b * \partial b / \partial a * \partial a / \partial w_1$

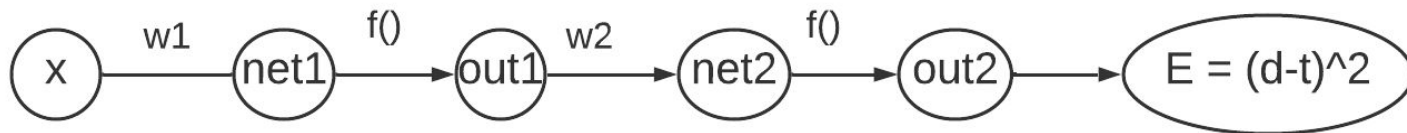
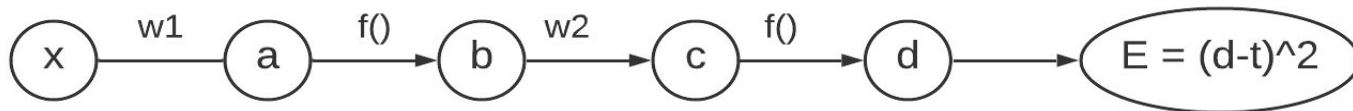
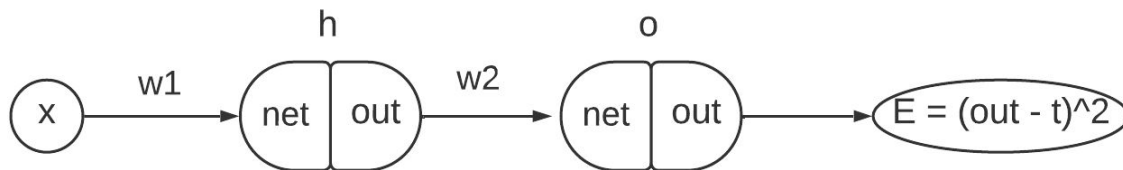
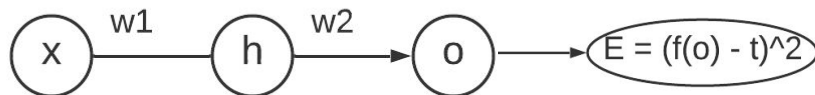
- $\partial E / \partial w_1 = 2d * 3c^2 * w_2 * 3a^2 * w_1$

# Example 3: Observe



- Observe, every original node actually consists of **2 operations** (due to activation), then it is actually **2 nodes** (NN notation is to merge)
  - Step 1) Compute net:  $\sum w_i * in_i$       step2:  $activation(net)$

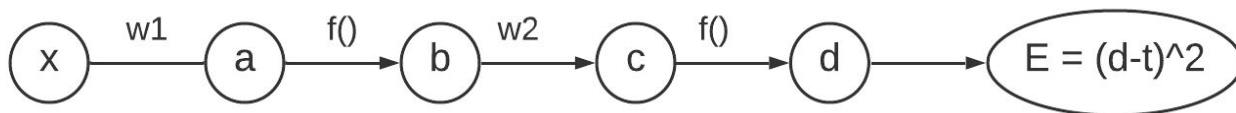
## Example 3: Correspondance





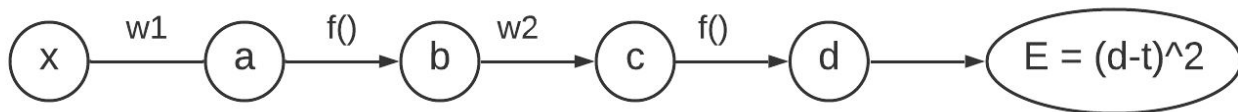
# Multivariate Chain Rule

- The previous example is actually about multivariables ( $w_1$  and  $w_2$ )
- It highlights this connection between complex functions and DAG
- We see that From  $E$  to  $W_1$  we need to pass with many steps (**nodes**)



- We found that its chain rule is the multiplication of all these  $\partial/\partial$ 
  - $\partial E / \partial w_1 = \partial E / \partial d * \partial d / \partial c * \partial c / \partial b * \partial b / \partial a * \partial a / \partial w_1$
- In fact, we can generalize that to a **tree diagram / computational graph**

# Chain Components



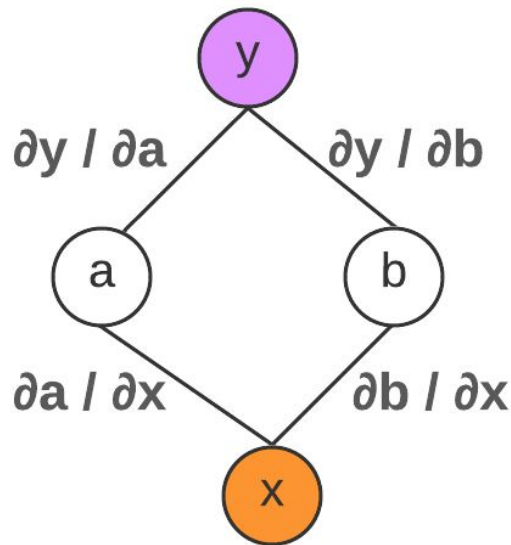
- $\partial E / \partial w1 = \partial E / \underline{\partial d} * \underline{\partial d} / \partial c * \partial c / \partial b * \partial b / \partial a * \partial a / \partial w1$
- $\partial E / \partial w1 = \partial E / \underline{\partial c} * \underline{\partial c} / \partial b * \partial b / \partial a * \partial a / \partial w1$
- $\partial E / \partial w1 = \partial E / \underline{\partial b} * \underline{\partial b} / \partial a * \partial a / \partial w1$
- $\partial E / \partial w1 = \partial E / \underline{\partial a} * \underline{\partial a} / \partial w1$
- $\partial E / \partial w1 = \partial E / \partial d * \partial d / \partial c * \partial c / \partial b * \partial b / \partial w1$  [canceled  $\partial b / \underline{\partial a} * \underline{\partial a} / \partial w1$ ]
- $\partial E / \partial w1 = \partial E / \partial d * \partial d / \partial a * \partial a / \partial w1$
- Keep this observation in mind: we can create several sub-path of derivatives from a single chain

# From equation to a tree/dag

- Assume we are given a multivariate equation
  - E.g.  $z = f(x,y)$ , where  $x$  and  $y$  themselves depend on more variable
- We will draw a tree where its lowest leaves represents our given variables
- We will keep grouping basic operations and create new variables
- We will keep doing the same until reaching a single variable
- This is the root of the tree represents our expression

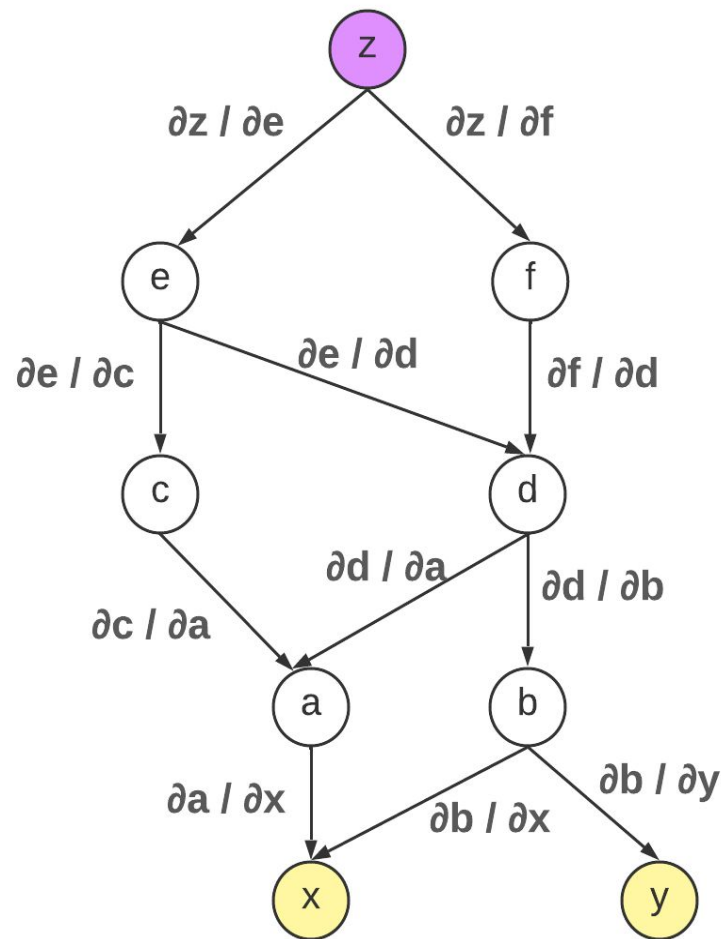
# Example 4

- Let  $y = 2x^5 - 4x^2$ 
  - This is actually a **univariate** variable ( $x$ )
  - So the bottom leave is just  $x$
  - We can create 2 new variables (nodes)
  - $a = 2x^5$  and  $b = 4x^2$
  - Finally we create a higher level  $y$
- Observe every edge is a derivative
  - Edge ( $g \Rightarrow h$ ) represents  $\partial g / \partial h$
  - There are 2 paths
  - $y \Rightarrow a \Rightarrow x$ : a **chain rule** with value  $\partial y / \partial x$
  - $y \Rightarrow b \Rightarrow x$ : a **chain rule** with value  $\partial y / \partial x$
  - Then to compute  $\partial y / \partial x$ : **sum** the results from the 2 paths



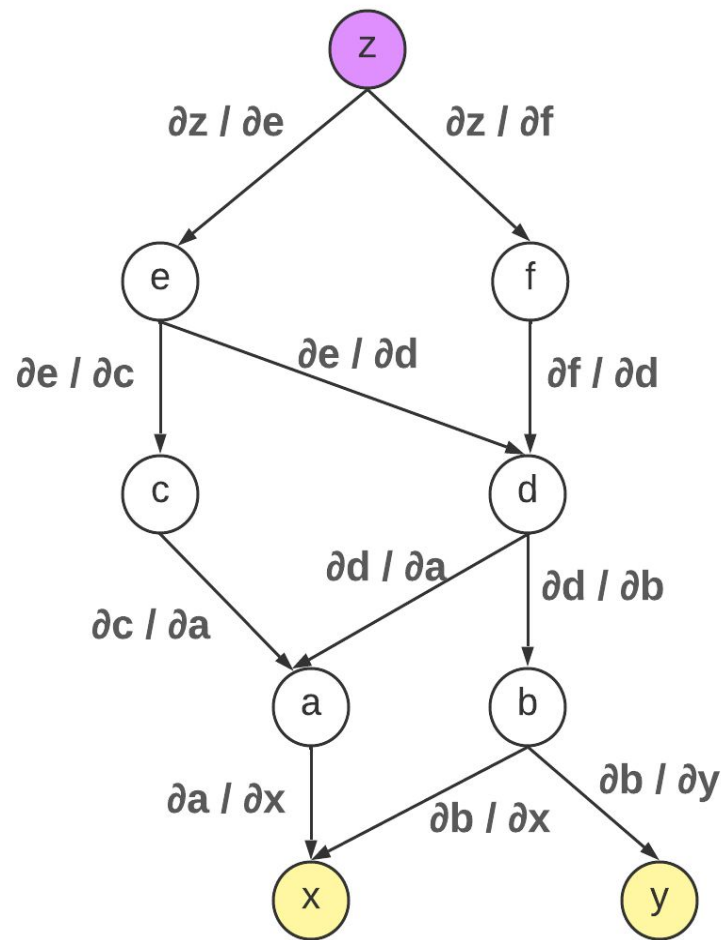
# Example 5

- Assume we have  $z = f(x, y)$ 
  - Put  $x$  and  $y$  in the leaves
  - Build the tree up to  $z$
- To compute any **partial derivative** from node( $m$ ) to node( $n$ )
  - Find all paths from  $m$  to  $n$ 
    - Each path is a simple chain
    - Multiply path value  $\Rightarrow$  chain rule value
  - Sum all the paths



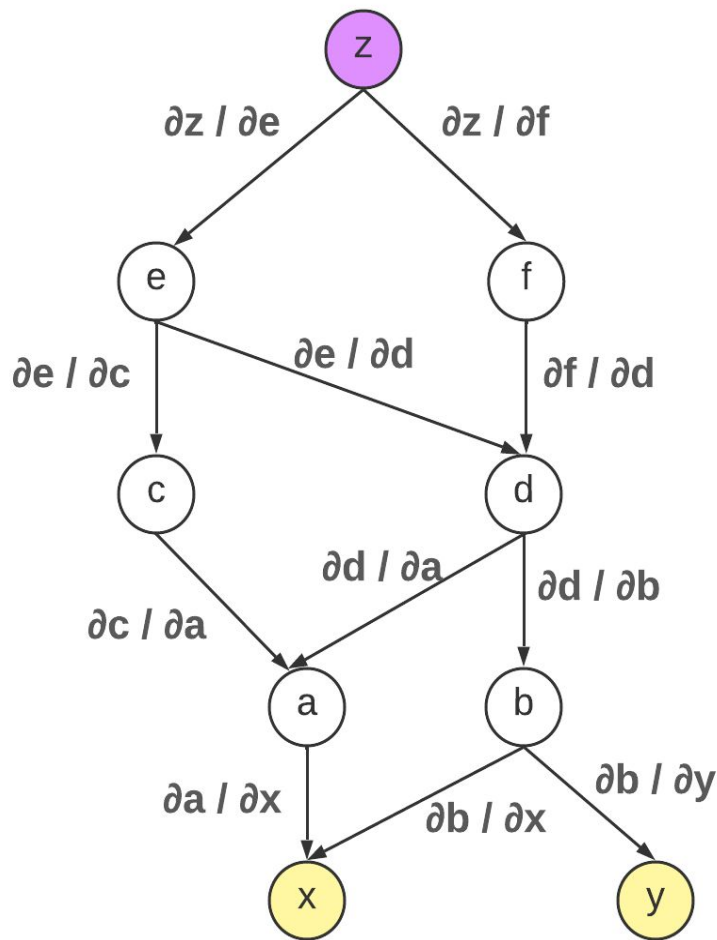
# Example 5A

- Compute  $\partial d / \partial x$
- We have 2 paths
- $d \Rightarrow a \Rightarrow x$ 
  - Represents:  $\partial d / \partial a * \partial a / \partial x$
- $d \Rightarrow b \Rightarrow x$ 
  - Represents:  $\partial d / \partial b * \partial b / \partial x$
- Let's pretend that our calculations are
- $\partial d / \partial x = 4$



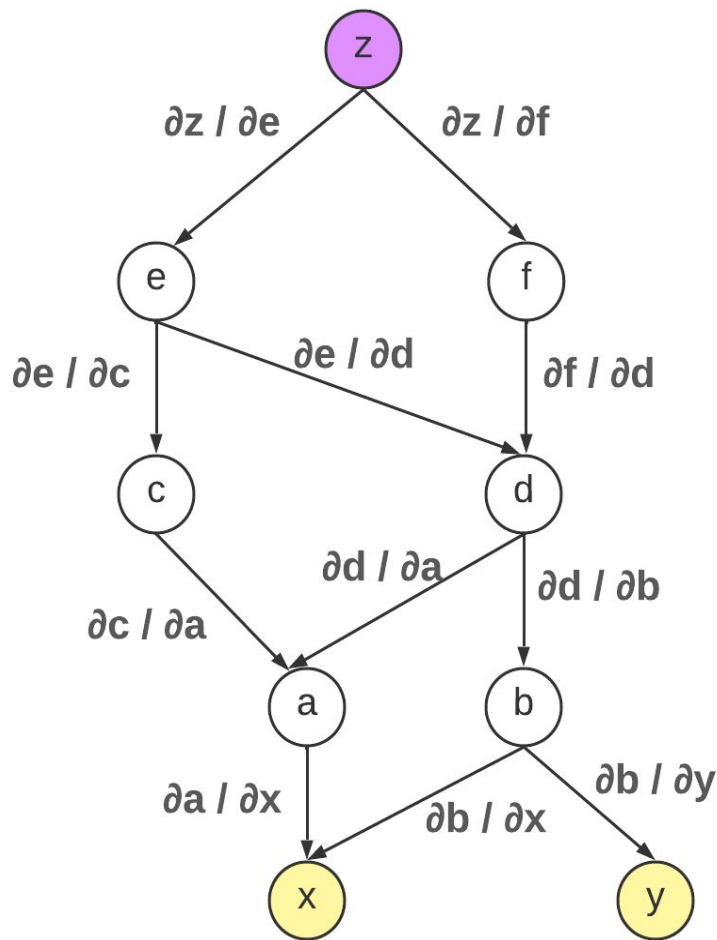
# Example 5B

- Compute  $\partial f / \partial x$ 
  - Assume  $\partial f / \partial d = 3$
- We have 2 paths
- $f \Rightarrow d \Rightarrow a \Rightarrow x$
- $f \Rightarrow d \Rightarrow b \Rightarrow x$
- Multiply and sum
- But this is waste of time!
- Can you find it **faster**!



# Example 5C

- Compute  $\partial f / \partial x$ 
  - Assume  $\partial f / \partial d = 3$
- $\partial f / \partial x = \partial f / \partial d * \partial d / \partial x$ 
  - We already computed  $\partial d / \partial x = 4$
  - Then  $\partial f / \partial x = 3 * 4 = 12$
  - This caching trick is the core of backpropagation algorithm
  - It is simply based on bottom-up processing starting from **x and y up to z**





# Relevant Materials

- Try to solve some of the examples in this [page](#)
  - Solve #: 3, 11, 12, 13
  - This [online calculator](#) might be helpful
- [Link](#)
- [Link](#)
- [Link](#)

*“Acquire knowledge and impart it to the people.”*

*“Seek knowledge from the Cradle to the Grave.”*

