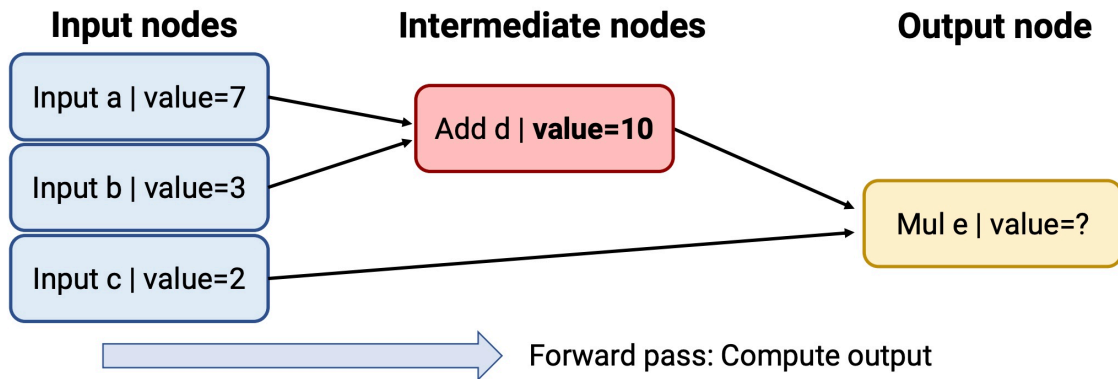


Computation Graph

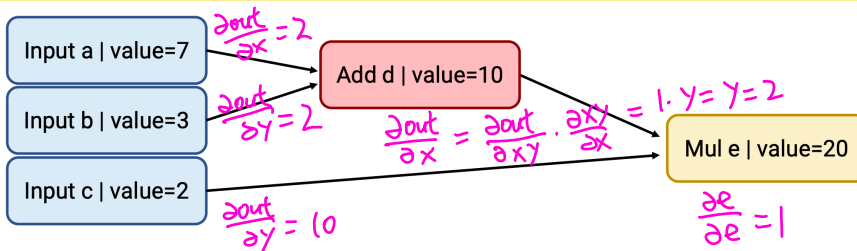
Computation graph for $(a + b) * c$ when $a=7, b=3, c=2$



Gradient checking

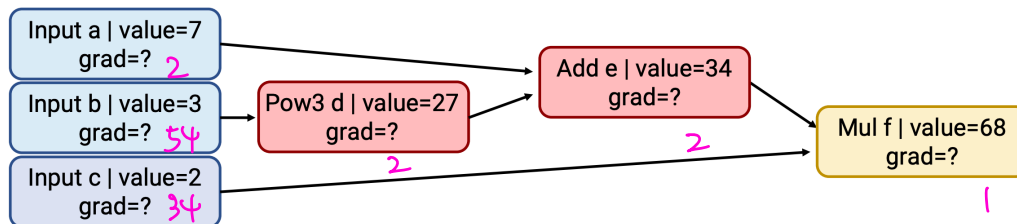
- numerical gradients
- Easy to implement
- Slow - $O(\#inputs)$

Backpropagation on trees



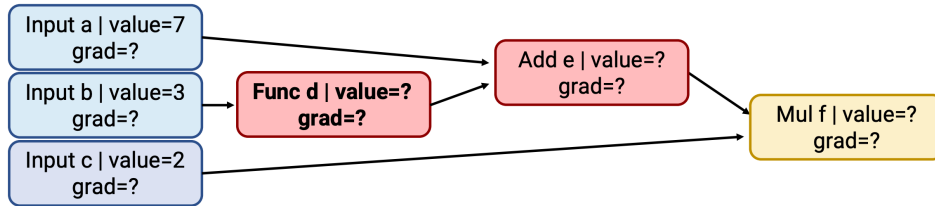
- For now: assume that the computation graph is a tree
 - Each node is only used in a single computation
 - Root of tree is output
 - Leaves of tree are inputs
- Idea: Recursively compute $\partial(output)/\partial(node)$ for each node, starting at output

PowerNode



New function: $(a + b^3) * c$ where $a=7, b=3, c=2$

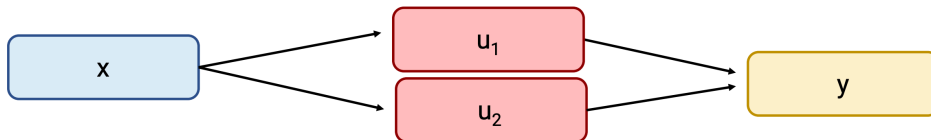
Generic Unary Function



New function: $(a + \text{Func}(b)) * c$ where $a=7, b=3, c=2$

- Steps 1-3 are the same
- Step 4: Func (generic function)
 - $\text{child.grad} = \text{parent.grad} * \partial(\text{Func}(\text{child}))/\partial\text{child}$

Multivariate chain rule

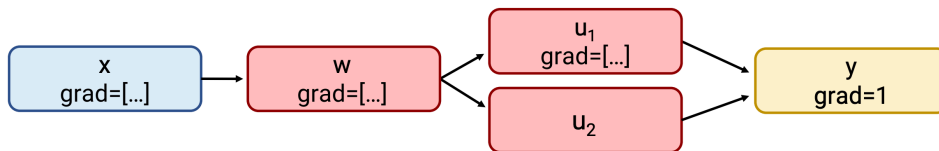


- Minimal example
 - u_1 and u_2 depend on variable x (i.e., x has two parents)
 - y depends on both u_1 and u_2
- By Multivariate Chain Rule: $\partial y / \partial x = \partial y / \partial u_1 * \partial u_1 / \partial x + \partial y / \partial u_2 * \partial u_2 / \partial x$

Dot product of $[\partial y / \partial u_1, \partial y / \partial u_2]$ & $[\partial u_1 / \partial x, \partial u_2 / \partial x]$

 - Changing x by epsilon changes each parent a bit; Effects add for small epsilon
 - Generalization of multiplying derivatives is matrix multiplication of Jacobians

What order of traversal?



Better code:

- `topo_order = [x, w, u1, u2, y]`
- Iterate in reverse order:
 - `y.backward()`
 - `u2.backward()`
 - `u1.backward()`
 - `w.backward()`
 - `x.backward()`
- Going recursively double-counts
 - First call to `w.backward()` makes final `x.grad` too large
- Solution: **Topological sort the nodes**
 - Iterate in reverse order, starting from output
 - Ensures that we process each node after all of its parents