#### Introduction to Neural Networks 2

CS 486/686

University of Waterloo

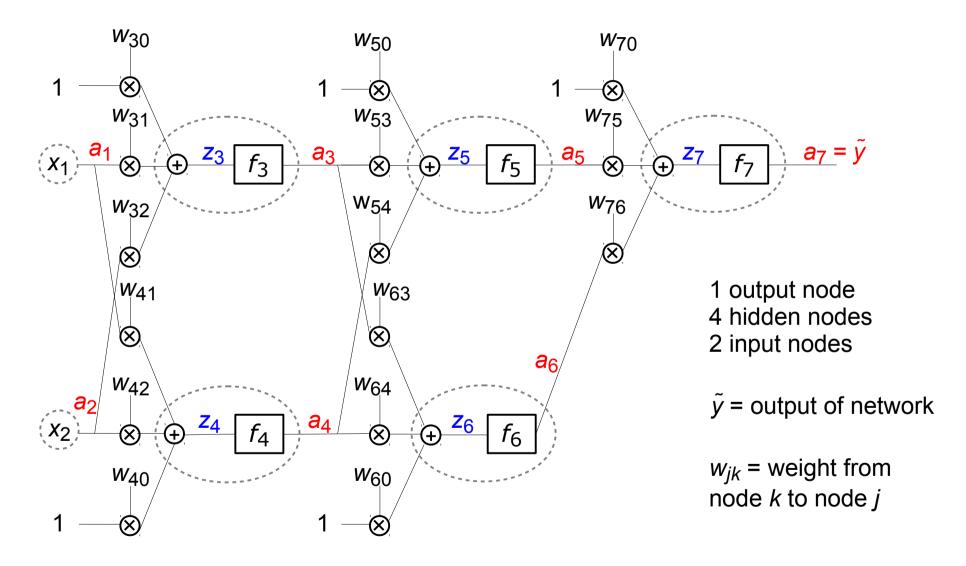
Lecture 20: July 9, 2015

#### Multi-Layer Neural Networks

Perceptron can only represent linear separators

 Need multiple layers to represent more complicated separators

### Example: Two Hidden Layers



## Learning Multi-Layer Network

Minimize squared error:

$$E(w, x, y) = \frac{1}{2}(y - \tilde{y})^{2}$$
$$\hat{w} = \operatorname*{arg\,min}_{w} E(w, x, y)$$

- Solution: gradient descent
- Just like what we did with sigmoid perceptron.

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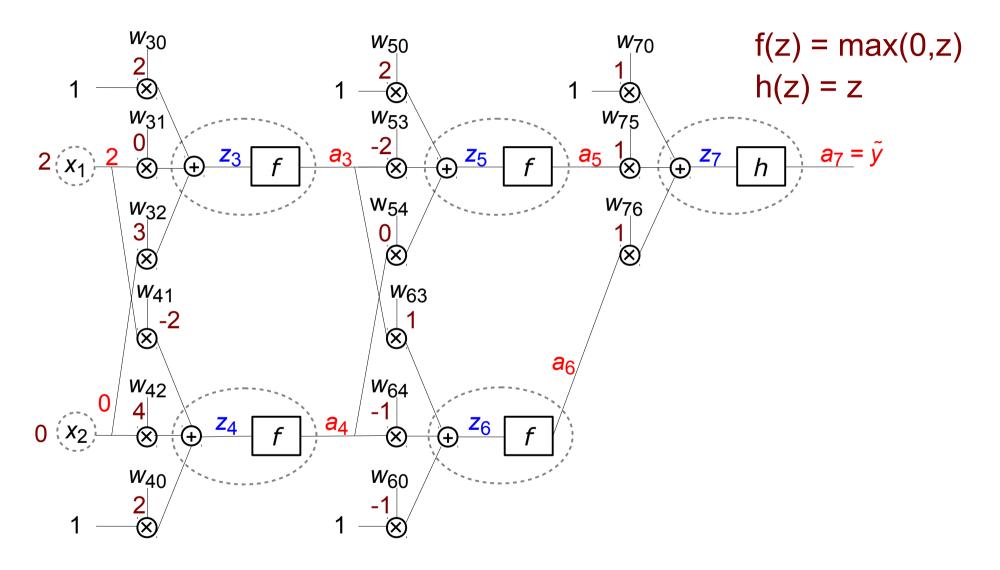
- Solution: gradient descent
- Just like what we did with sigmoid perceptron.
- Problem: gradient much harder to compute
  - Solution: compute gradient with backpropagation

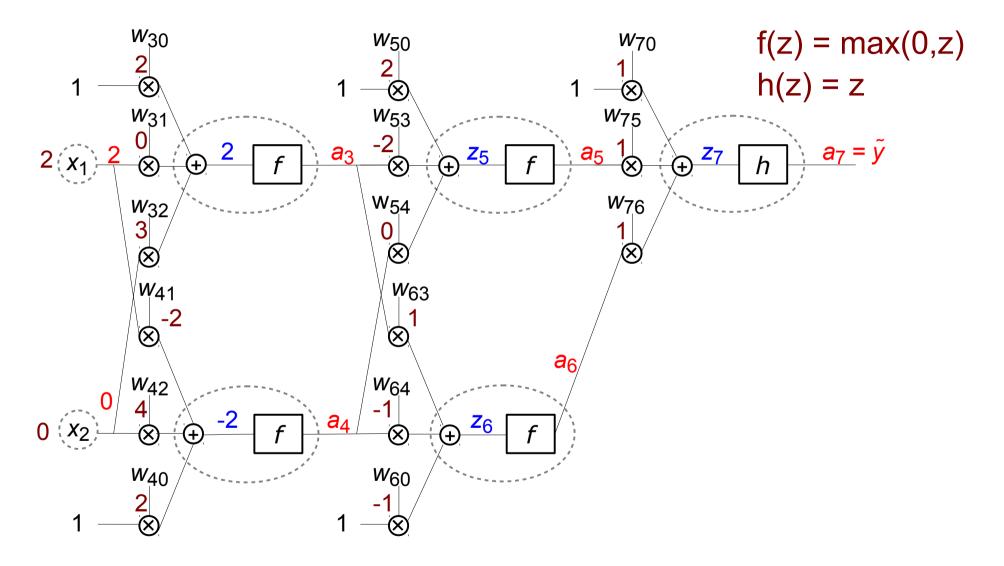
# Learning Multi-Layer Network by Gradient Descent

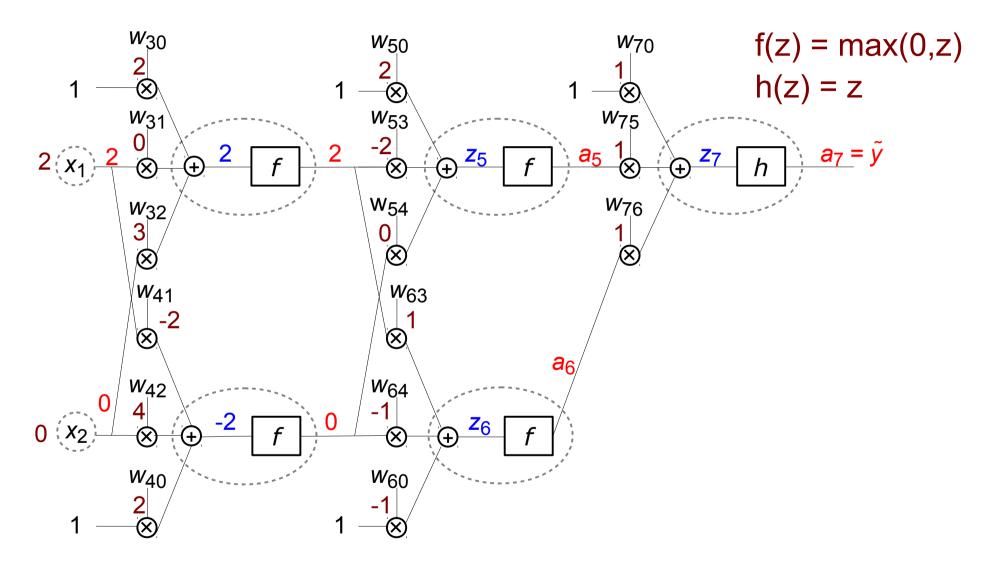
- Initialize weights w
- For each training example (x,y) do
  - Compute gradient  $\nabla E(w)$  by backpropagation
  - Update weights  $w \leftarrow w \alpha \nabla E(w)$
- Repeat until stopping criteria satisfied

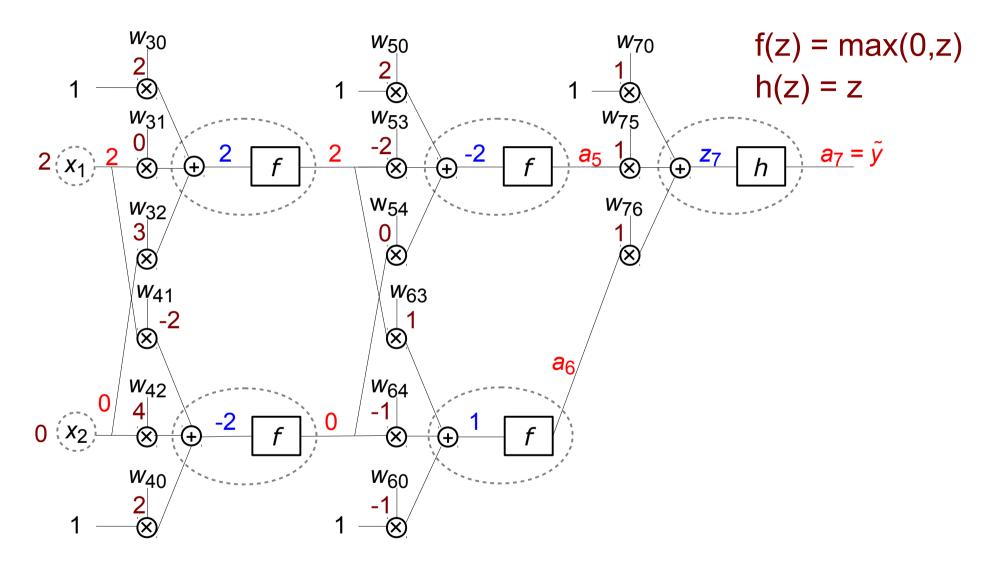
## Backpropagation

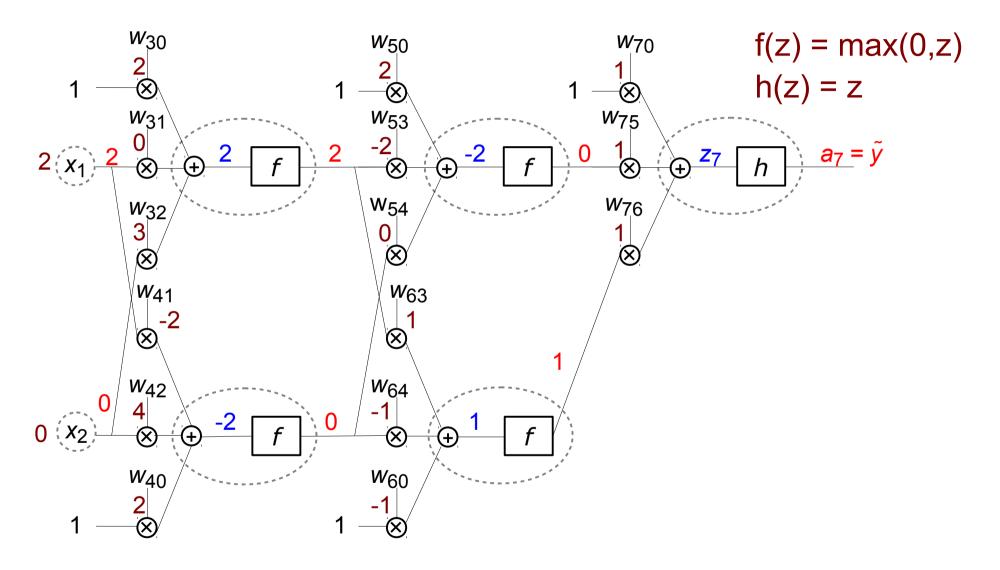
- Given training example (x, y)
- Forward phase
  - Starting from input nodes, compute all  $z_k$ 's and  $a_k$ 's by forward propagation
- Backward phase
  - Starting from output nodes, compute  $D_k = -\frac{\partial E}{\partial z_k}$ :
    - $D_k = f'(z_k)(a_k y_k)$  if k is an output node
    - $D_k = f'(z_k) \sum_j w_{jk} D_j$  if k is a hidden node
  - Compute  $\frac{\partial E}{\partial w_{ik}} = -D_j a_k$  for all  $w_{jk}$
- Return  $\nabla E(w)$

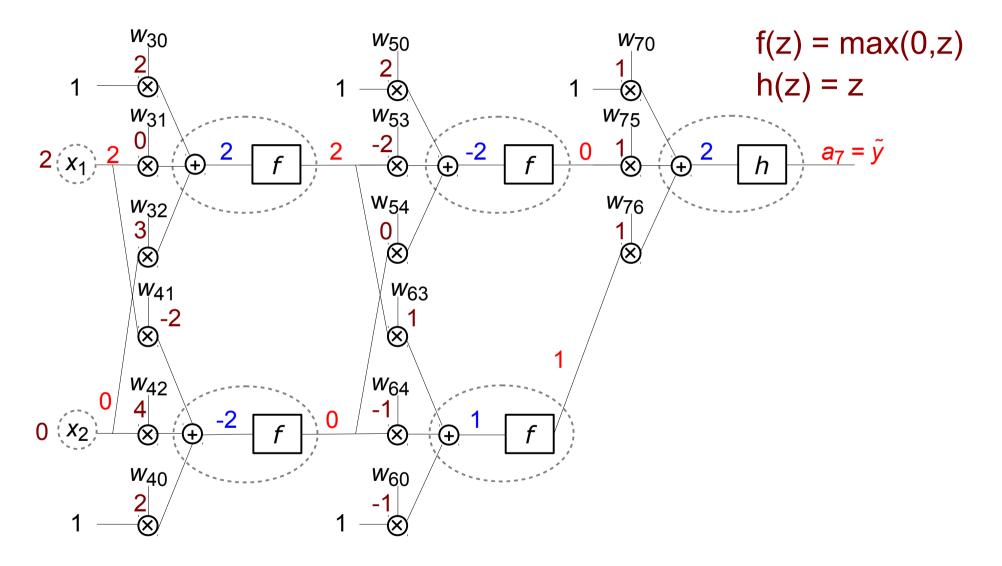


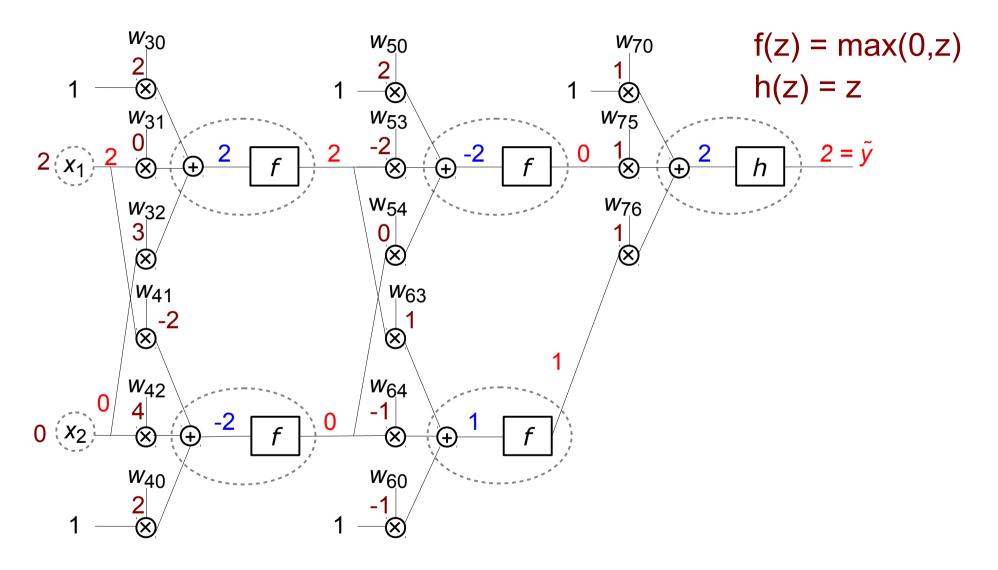




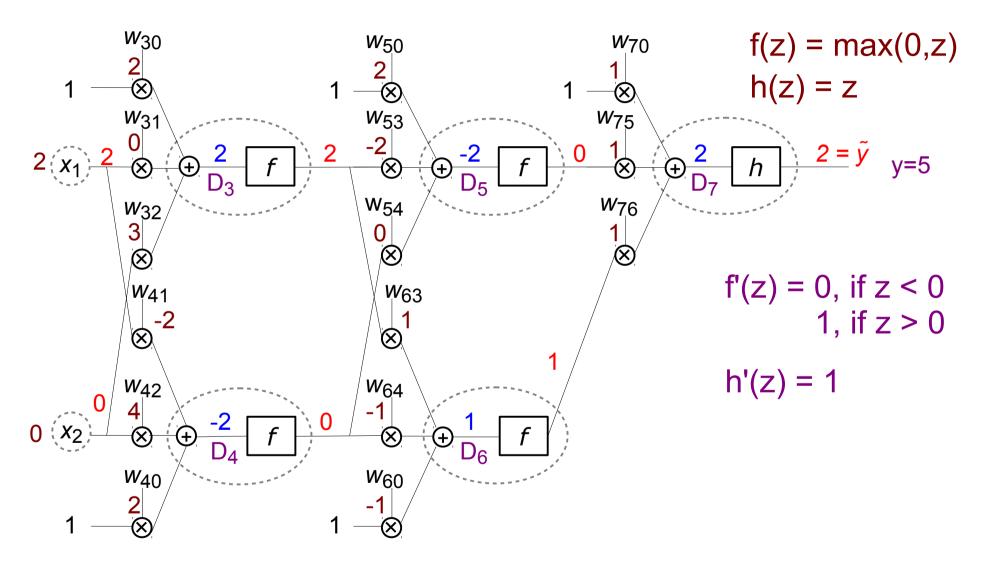


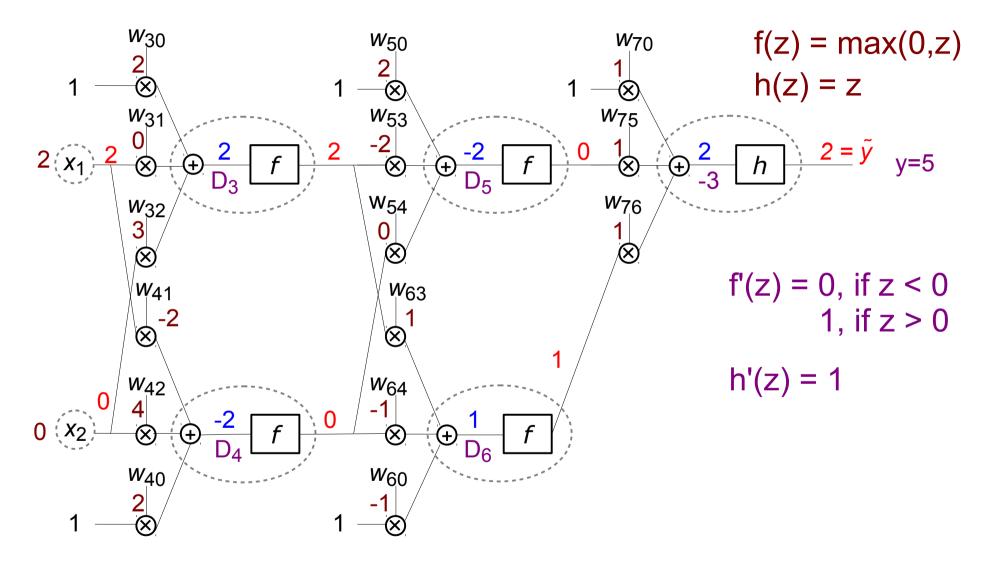


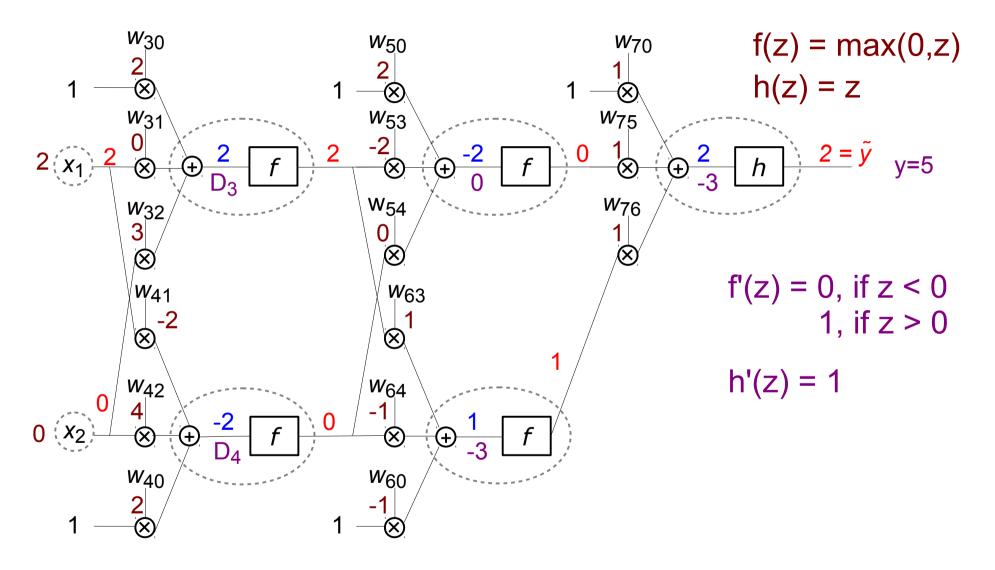


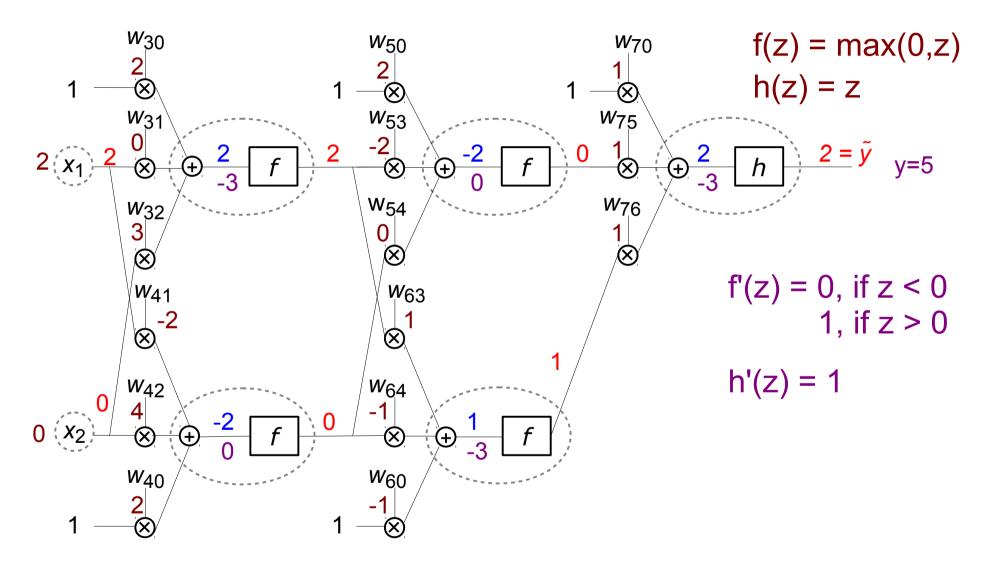


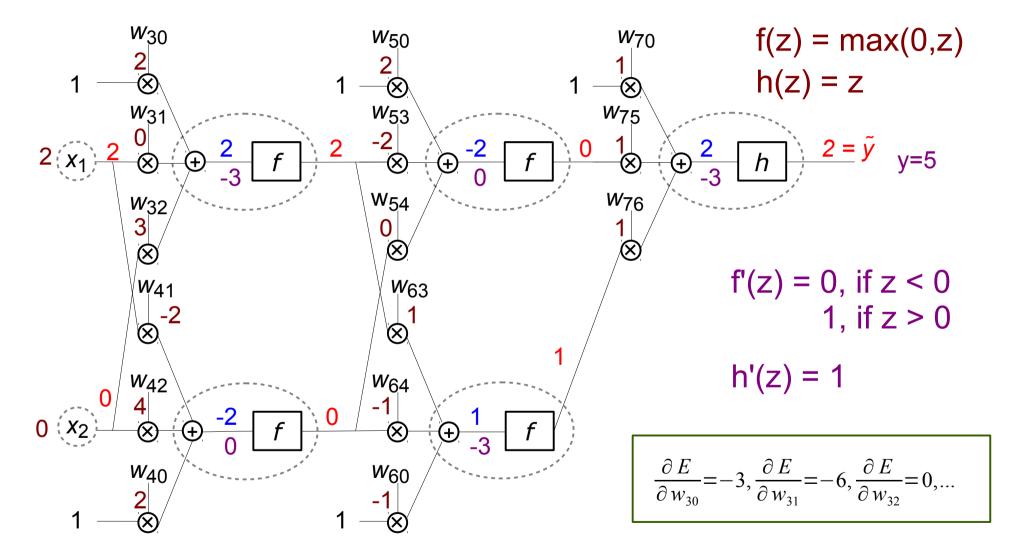
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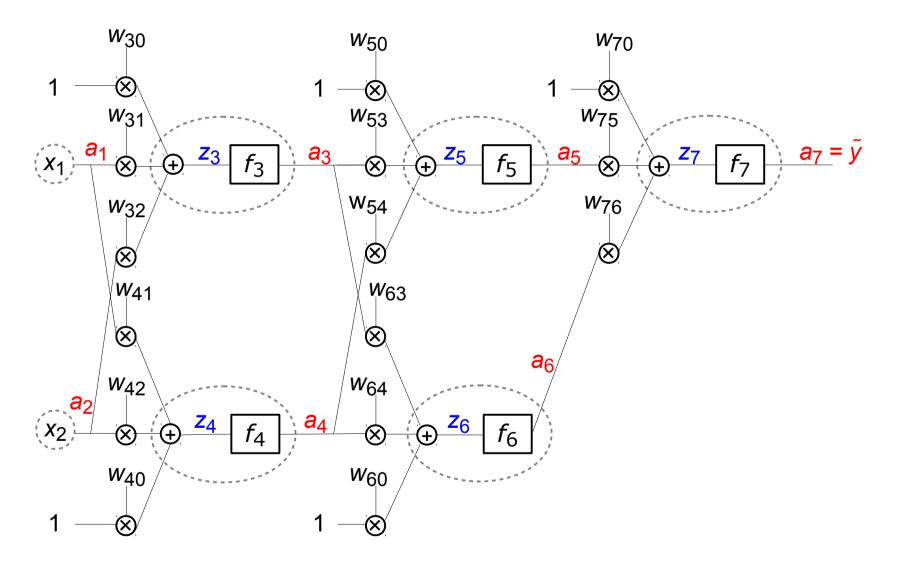




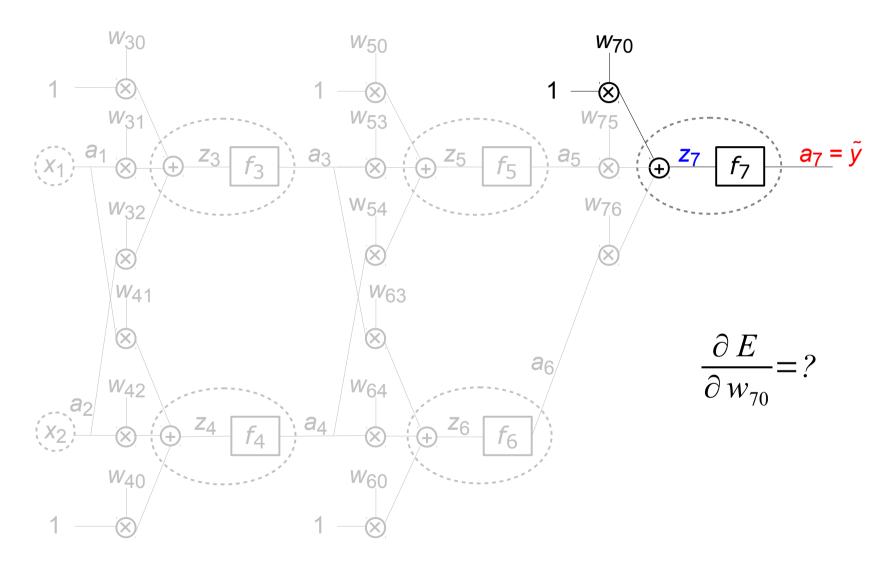


July 9, 2015 Wilson Hsu 20

## **Backpropagation Derivation**



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