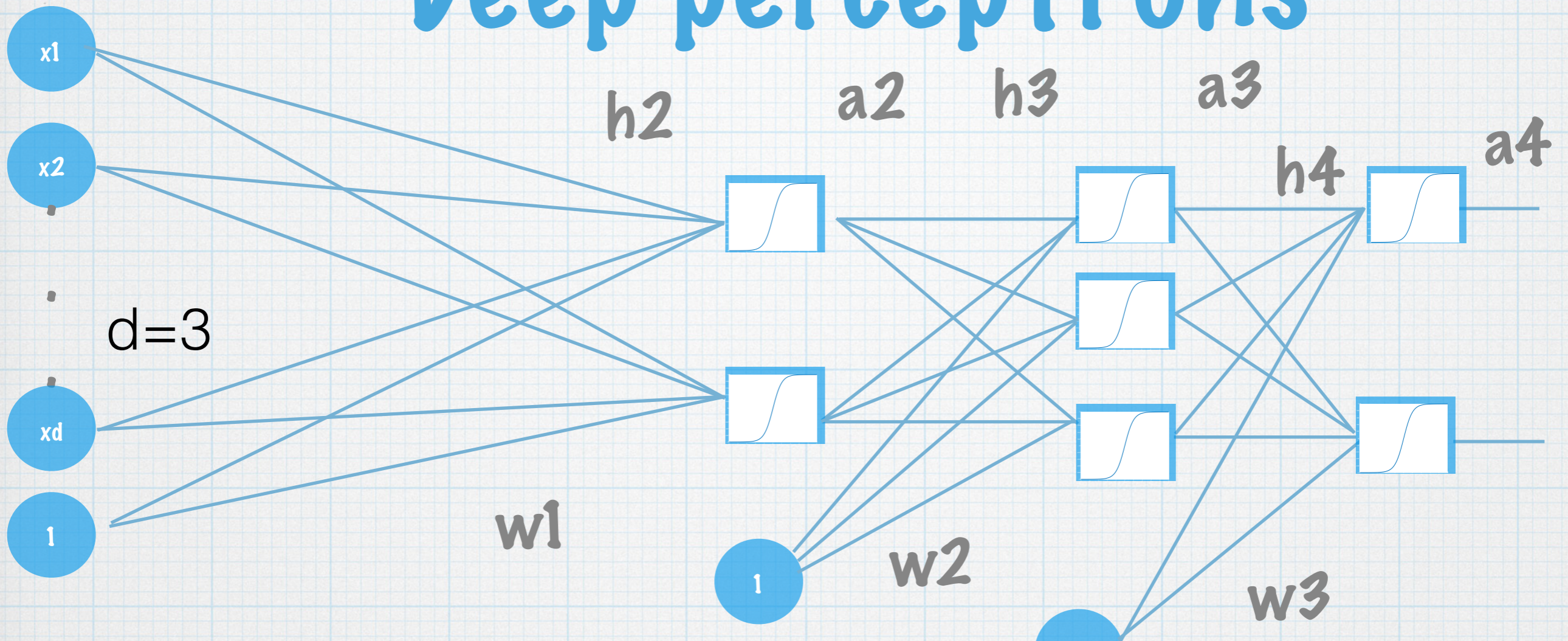


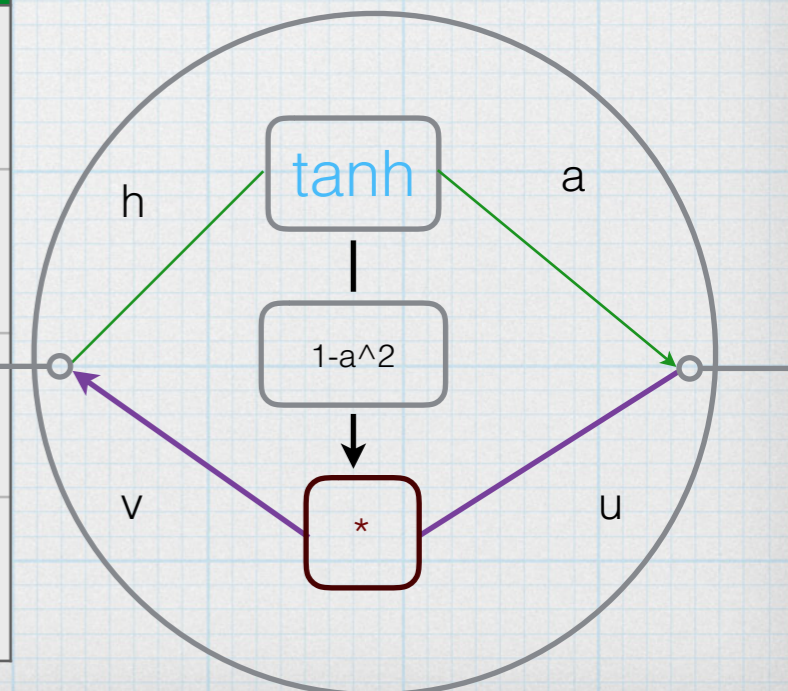
# Gradients in Neural mapping

Numerical and Symbolic approaches

# Deep perceptrons



Size	u	v	backward. $w'$	h	W	a	
4	$2 \times 2 \rightarrow 2 \times 2$	$2 \times 2$		$2 \times 1$	$2 \times 1$	$2 \times 1$	Output
3	$3 \times 2 \rightarrow 3 \times 2$	$3 \times 2$	$3 \times 2$	$3 \times 1$	$2 \times 4$	$3 \times 1$	
2	$2 \times 2 \rightarrow 2 \times 2$	$2 \times 2$	$2 \times 3$	$2 \times 1$	$3 \times 3$	$2 \times 1$	
1			$3 \times 2$		$2 \times 4$	$3 \times 1$	



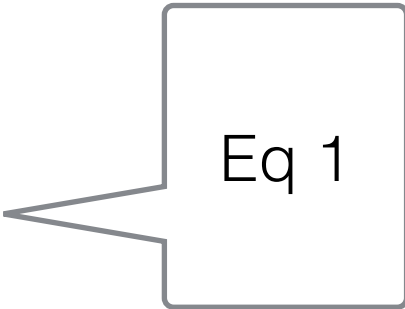
$$W\{n\} = \begin{bmatrix} w\{n\} & b\{n\} \end{bmatrix}$$

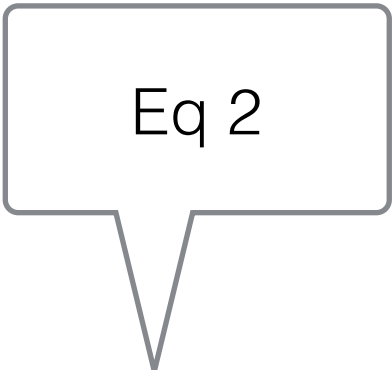
$$h\{n+1\} = W\{n\} \begin{bmatrix} a\{n\} \\ 1 \end{bmatrix}$$

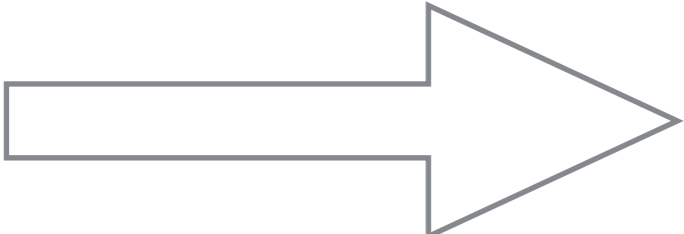
$$a\{n\} = \tanh(h\{n\}), \text{ for } n > 1$$

# Backward Linear transformation

- For  $n=L-1:1$ , where  $L$  denotes the number of layers

$$\frac{dh\{n+1\}}{da\{n\}} = w\{n\}$$


$$\frac{dy}{da\{n\}} = \frac{dy}{dh\{n+1\}} \frac{dh\{n+1\}}{da\{n\}}$$



$$u\{n\} = w^T\{n\}v\{n+1\}$$

# Backward Nonlinear transformation

$$a\{n\} = \tanh(h\{n\})$$

Eq 3

$$\frac{dy}{dh\{n\}} = \frac{dy}{da\{n\}} \frac{da\{n\}}{dh\{n\}}$$

$$v\{n\} = u\{n\} \frac{da\{n\}}{dh\{n\}}$$

# Sizes of matrices

$$h\{n+1\} = w\{n\} * a\{n\}$$

$$a\{n\} = \tanh(h\{n\}), \text{ for } n > 1$$

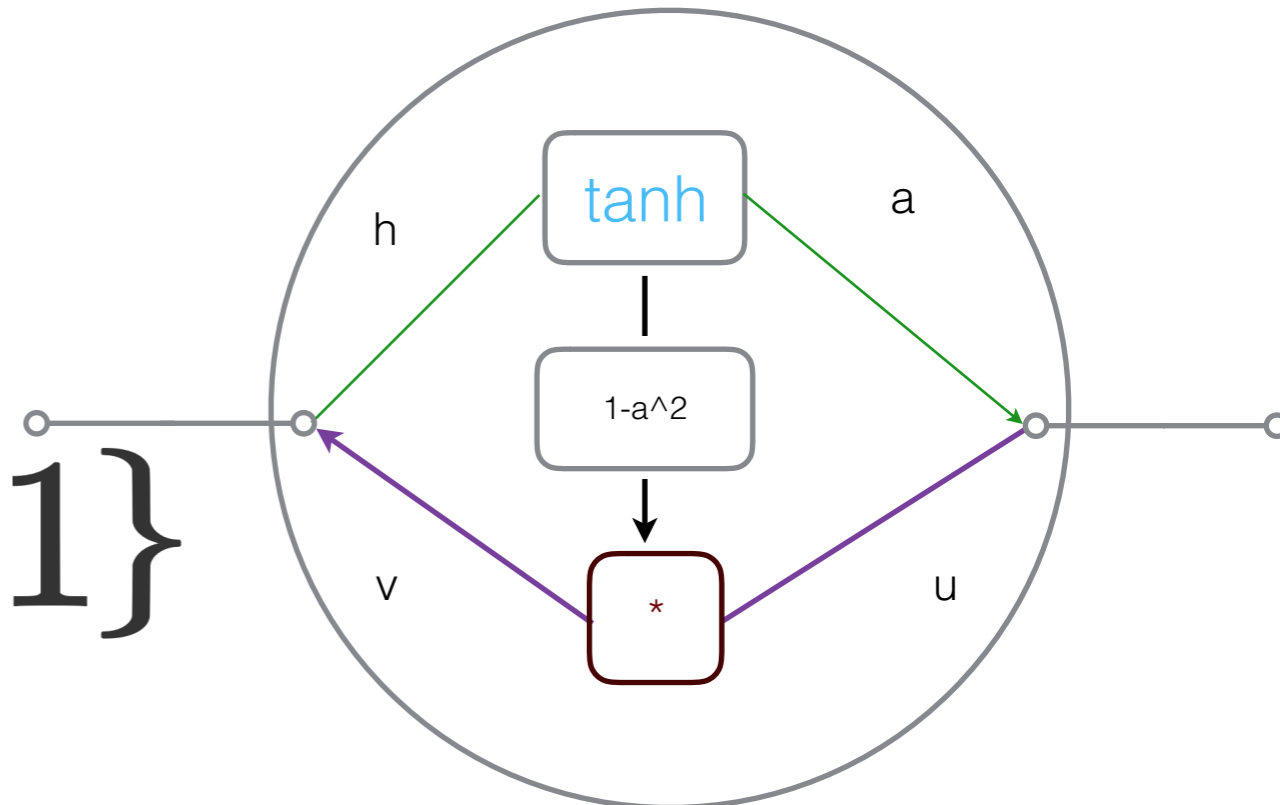
$$u\{n\} \equiv \frac{dy}{da\{n\}}$$

$$v\{n\} \equiv \frac{dy}{dh\{n\}}$$

Size	u	v	backward. w'	h	W	a	
4	2x2	2x2		2x1		2x1	Output
3	3x2	3x2		3x1	2x4	3x1	
2	2x2	2x2		2x1	3x3	2x1	
1			3x2		2x4	3x1	

Eq 2

$$u\{n\} = w^T\{n\}v\{n+1\}$$



set  $u\{L\}$  to an identity matrix

for  $n=L-1:-1:2$

apply eq(3) to calculate  $v\{n+1\}$

apply eq(2) to calculate  $u\{n\}$

$$h\{n + 1\} = W\{n\}$$

$$\begin{bmatrix} a\{n\} \\ 1 \end{bmatrix}$$

$y\{m\}$  denotes an output component

$$\frac{dy\{m\}}{dW\{n\}} = \frac{dy\{m\}}{dh\{n+1\}} \frac{dh\{n+1\}}{dW\{n\}}$$

$$= v_m\{n + 1\} \frac{dh\{n+1\}}{dW\{n\}}$$

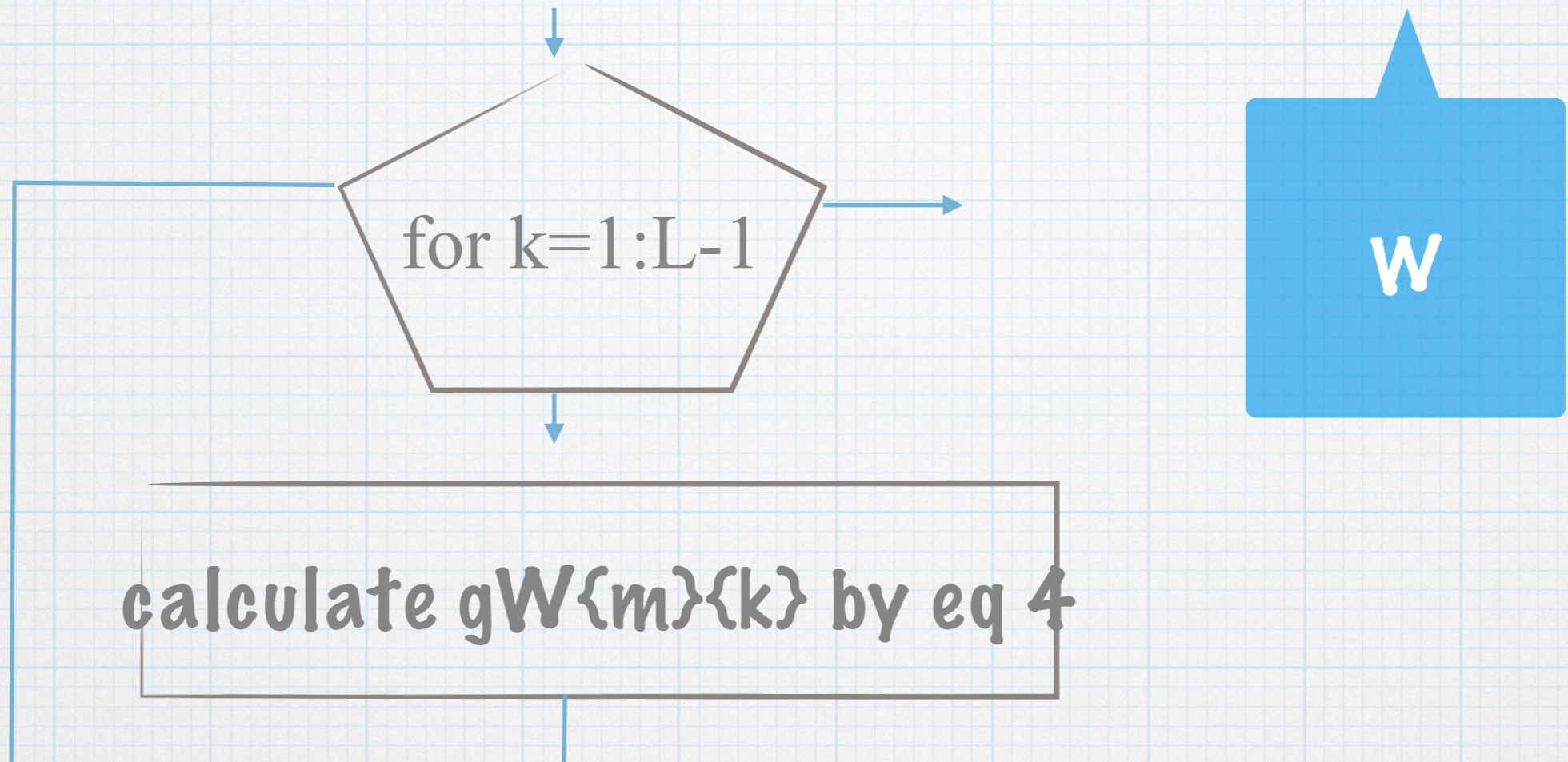
denoted by  $v_m\{n + 1\}$

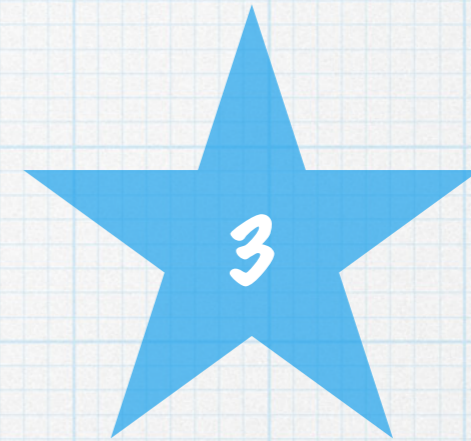
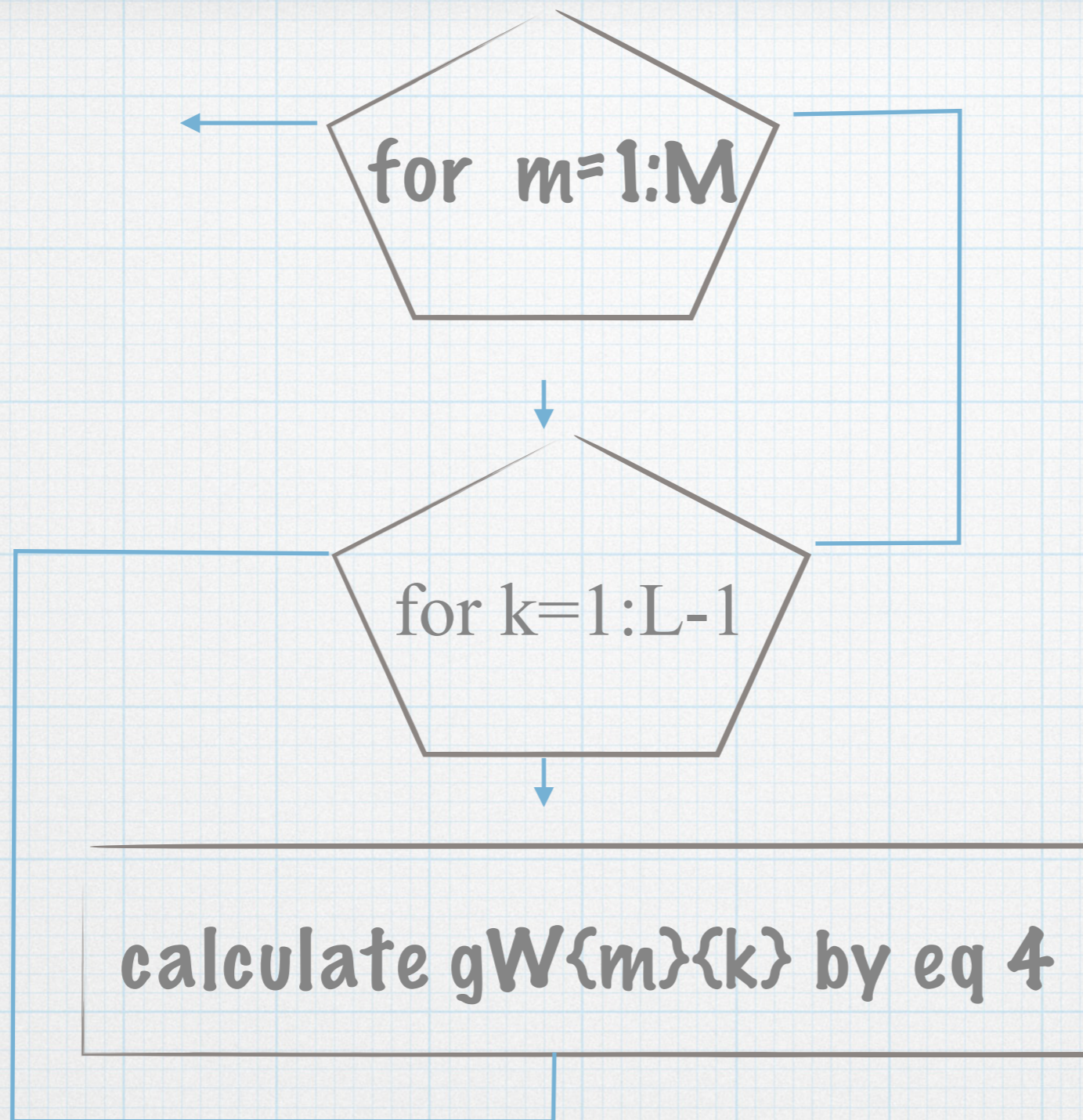
$$= v_m\{n + 1\} \begin{bmatrix} a\{n\} \\ 1 \end{bmatrix}^T$$

eq 4



computing the gradient of  $y\{m\}$  with respect to  $w\{k\}$  for all  $k$



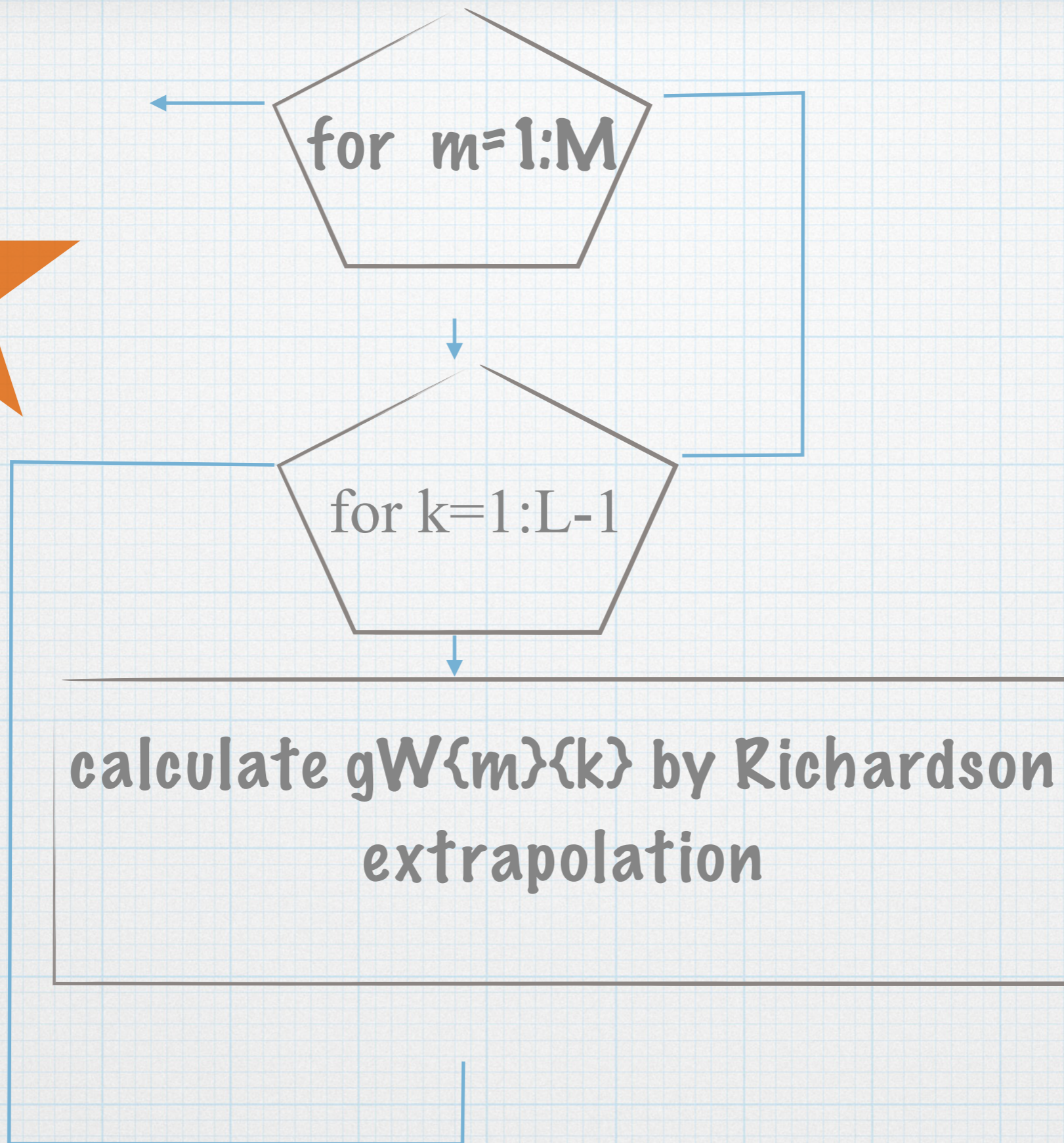


# Gradient check

## Richardson extrapolation

$$f'(x) \approx \varphi\left(\frac{h}{2}\right) + \frac{1}{3} \left[ \varphi\left(\frac{h}{2}\right) - \varphi(h) \right]$$

$$\varphi(h) = \frac{f(x+h) - f(x-h)}{2h}$$



## Richardson extrapolation

# Richardson Extrapolation

$$f'(x) \approx \varphi\left(\frac{h}{2}\right) + \frac{1}{3} \left[ \varphi\left(\frac{h}{2}\right) - \varphi(h) \right]$$

$$\varphi(h) = \frac{f(x+h) - f(x-h)}{2h}$$

set small  $z$

4b

for  
each element in  
 $W\{m\}\{k\}$

$$g1 = (f1 - f2) / (2z)$$

$$g2 = (f3 - f4) / z$$

Perturb it by adding  $z, -z, z/2, -z/2$  respectively  
and determine four corresponding output  $y\{m\}$ ,  
denoted by  $f1, f2, f3, f4$

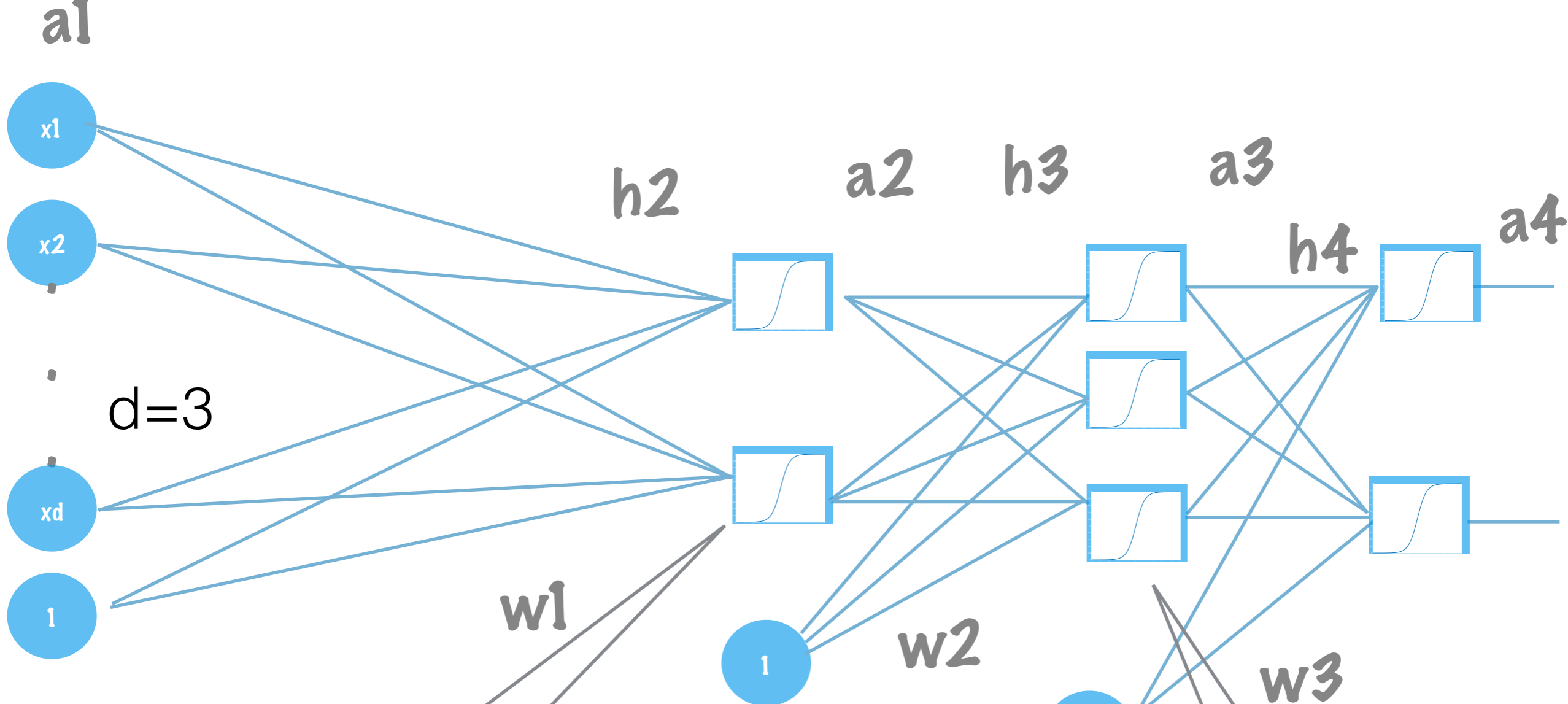
set corresponding element in  $gW\{m\}\{k\}$  to  
extrapolation determined by  $f1-f4$

$$\frac{dy}{dW_3} = ?$$

$$\frac{dy}{dW_2} = ?$$

$$\frac{dy}{dW_1} = ?$$

$dw_{\{n\}} = ?$  for all  $n$



$$\frac{dy}{dW_{m_2}[i,j]} = ?$$

# Batch

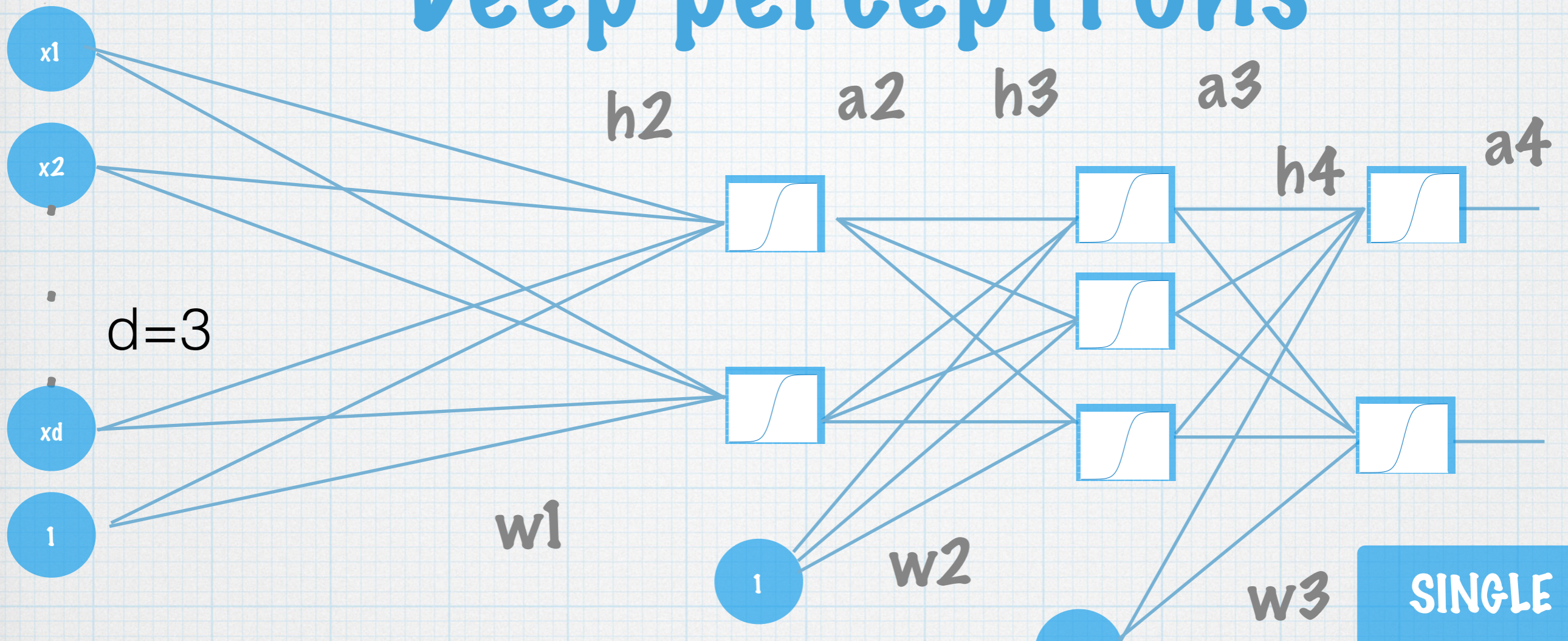
A batch contains a set of inputs.

Let  $u\{m\}$  and  $v\{m\}$  denote gradients corresponding the  $m$ th output component.

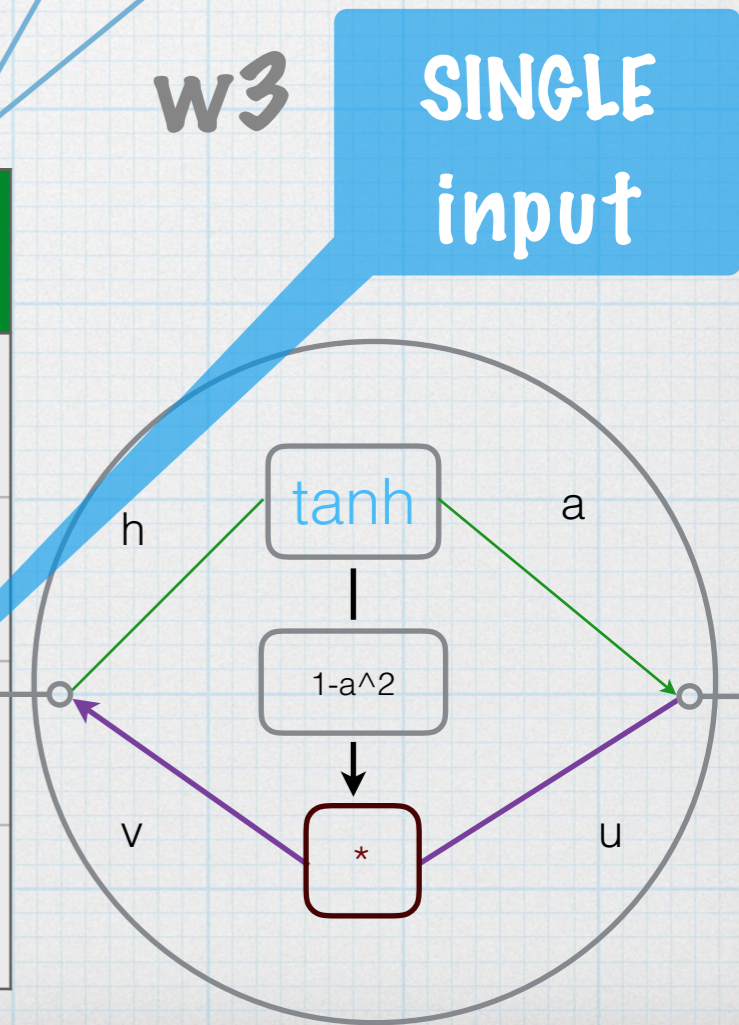
Extend representations for a set of inputs



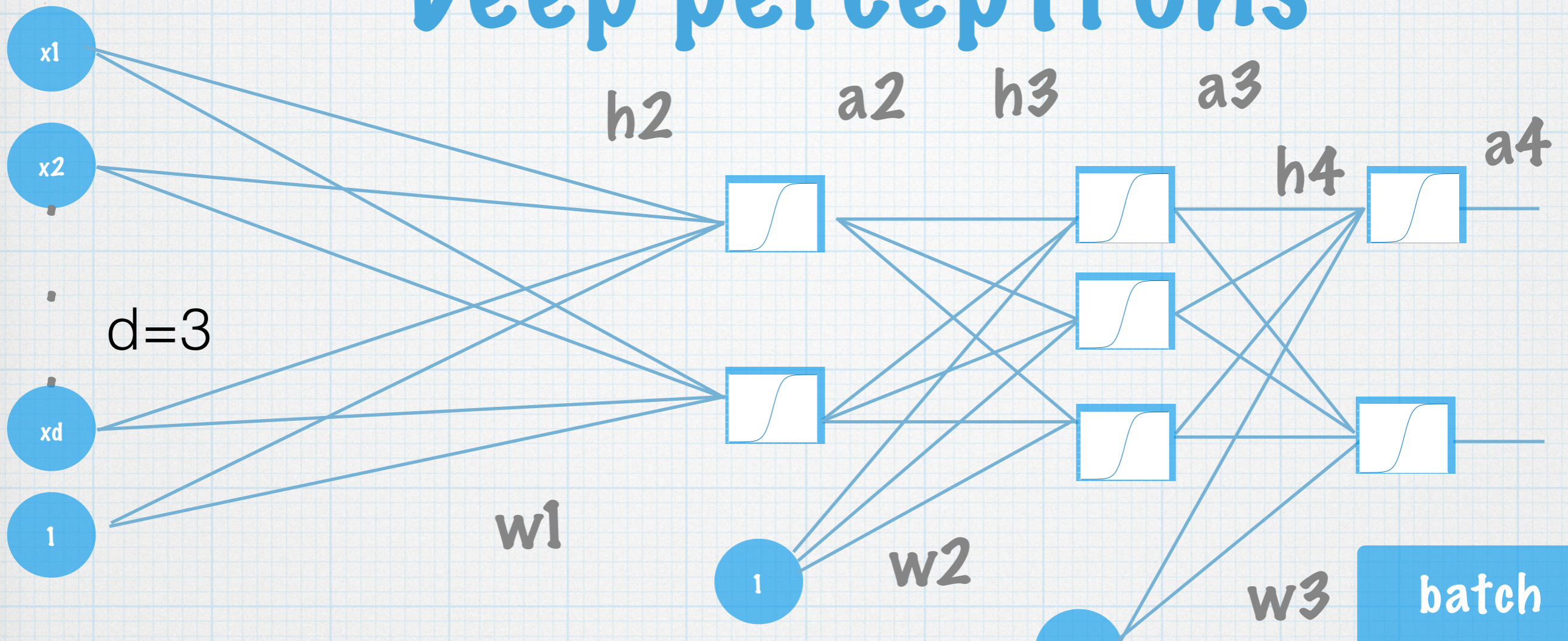
# Deep perceptrons



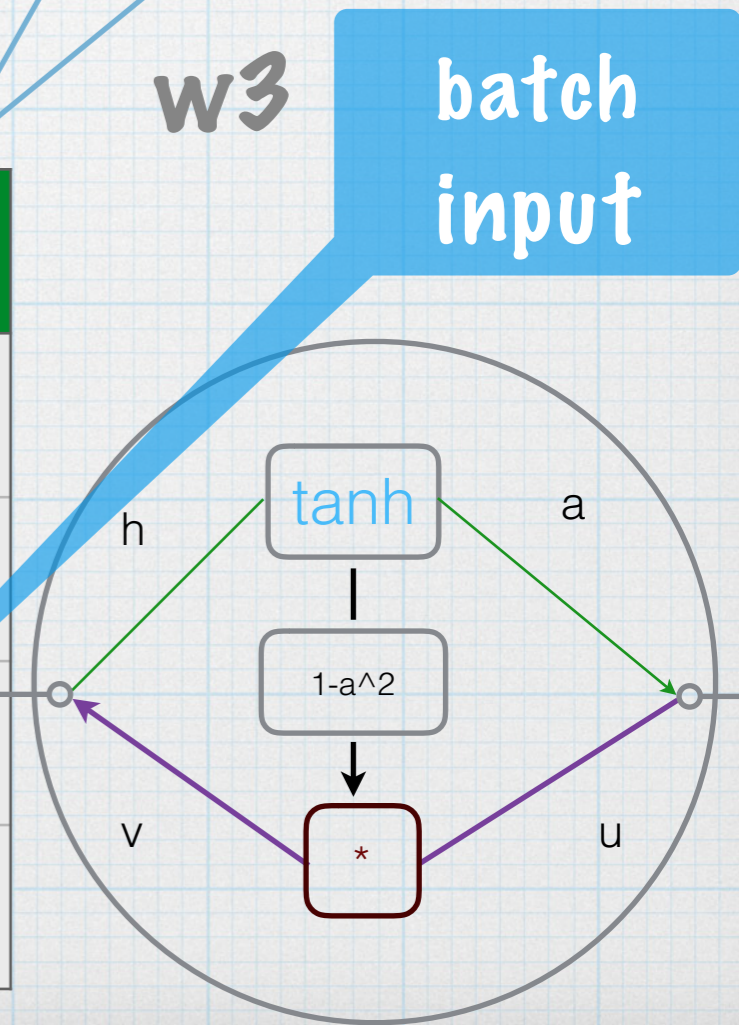
Size	$u\{m\}$	$v\{m\}$	backward. $w'$	$h$	$W$	$a$	
4	$2 \times 1 \rightarrow 2 \times 1$	$2 \times 1$		$2 \times 1$	$2 \times 1$	$2 \times 1$	Output
3	$3 \times 1 \rightarrow 3 \times 1$	$3 \times 1$	$3 \times 2$	$3 \times 1$	$2 \times 4$	$3 \times 1$	
2	$2 \times 1 \rightarrow 2 \times 1$	$2 \times 1$	$2 \times 3$	$2 \times 1$	$3 \times 3$	$2 \times 1$	
1			$3 \times 2$		$2 \times 4$	$3 \times 1$	



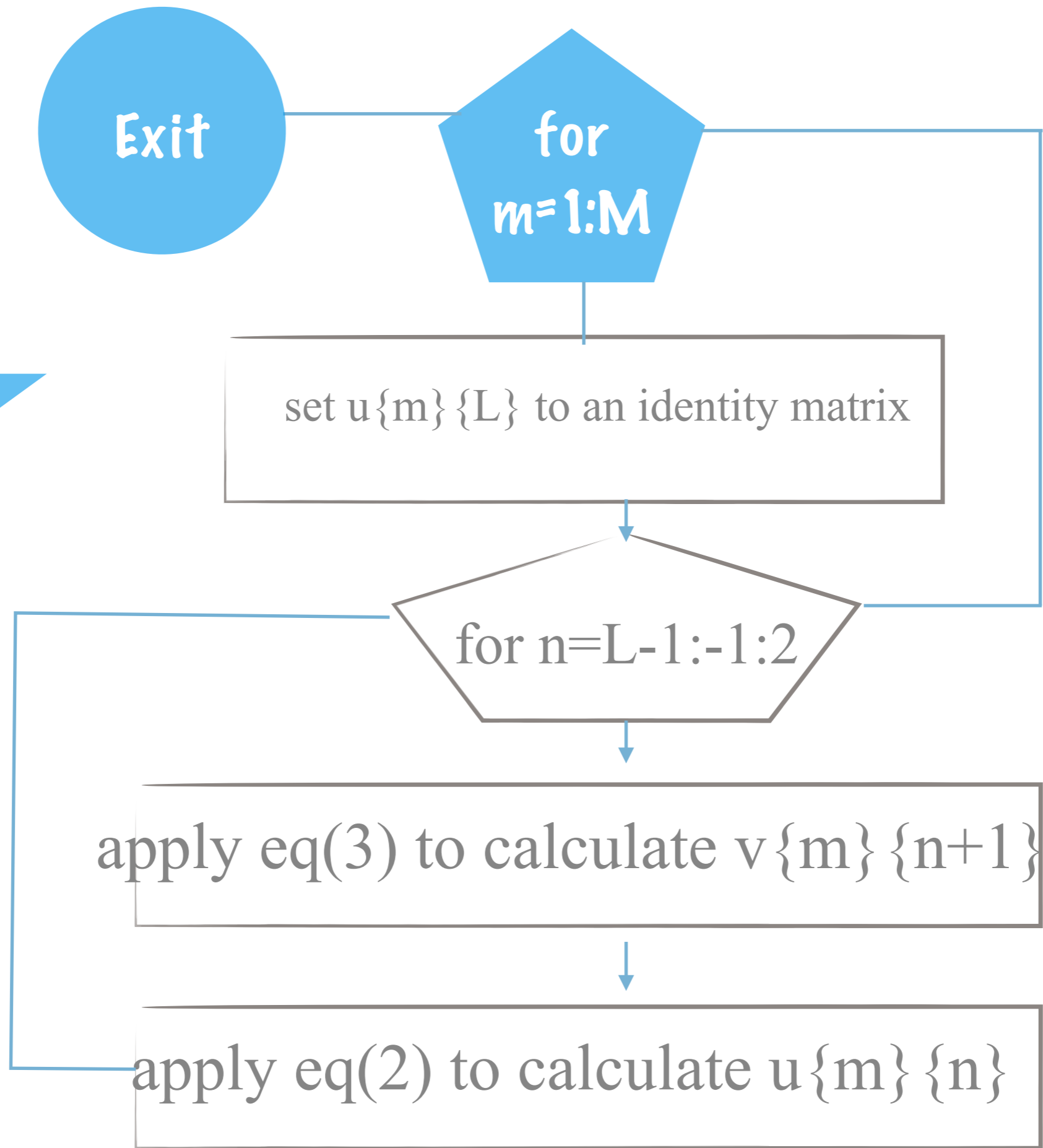
# Deep perceptrons



Size	$u\{m\}$	$v\{m\}$	backward. $w'$	$h$	$W$	$a$	
4	$2 \times n$	$2 \times n$		$2 \times n$		$2 \times n$	Output
3	$3 \times n$	$3 \times n$	$3 \times 2$	$3 \times n$	$2 \times 4$	$3 \times n$	
2	$2 \times n$	$2 \times n$	$2 \times 3$	$2 \times n$	$3 \times 3$	$2 \times n$	
1			$3 \times 2$		$2 \times 4$	$3 \times n$	



2



$$U \{n\} = W^T * v \{n+1\} \quad (2)$$

$$v \{n\} = U \{n\} * (1 - a \{n\}^2) \quad (3)$$

$$gW_m \{n\} = v_m \{n+1\} \begin{matrix} \text{expand} \\ \left[ \begin{array}{c} a \{n\} \\ 1 \end{array} \right]^T \end{matrix} \quad (\text{eq 4})$$

# Square error

$$E(W) = \sum_t \sum_m e_m^2[t]$$

$$e_m[t] = y[t] - \hat{y}[t]$$

## project

Submit your codes to [ett2012@gmail.com](mailto:ett2012@gmail.com) pm9 Dec. 1

Accomplish two methods of classdef of perceptrons

1. Calculate gradients of outputs with respect to weight matrices
2. Compare calculated gradients with numerical gradients for realization of Gradient-check