

**Linear mapping**

**Neural mapping**

**Numerical and Symbolic approaches**

# array

```
A = sym('a',[1 3])
```

```
A =
```

```
[ a1, a2, a3]
```

```
>> A = sym('a',[2 3])
```

```
A =
```

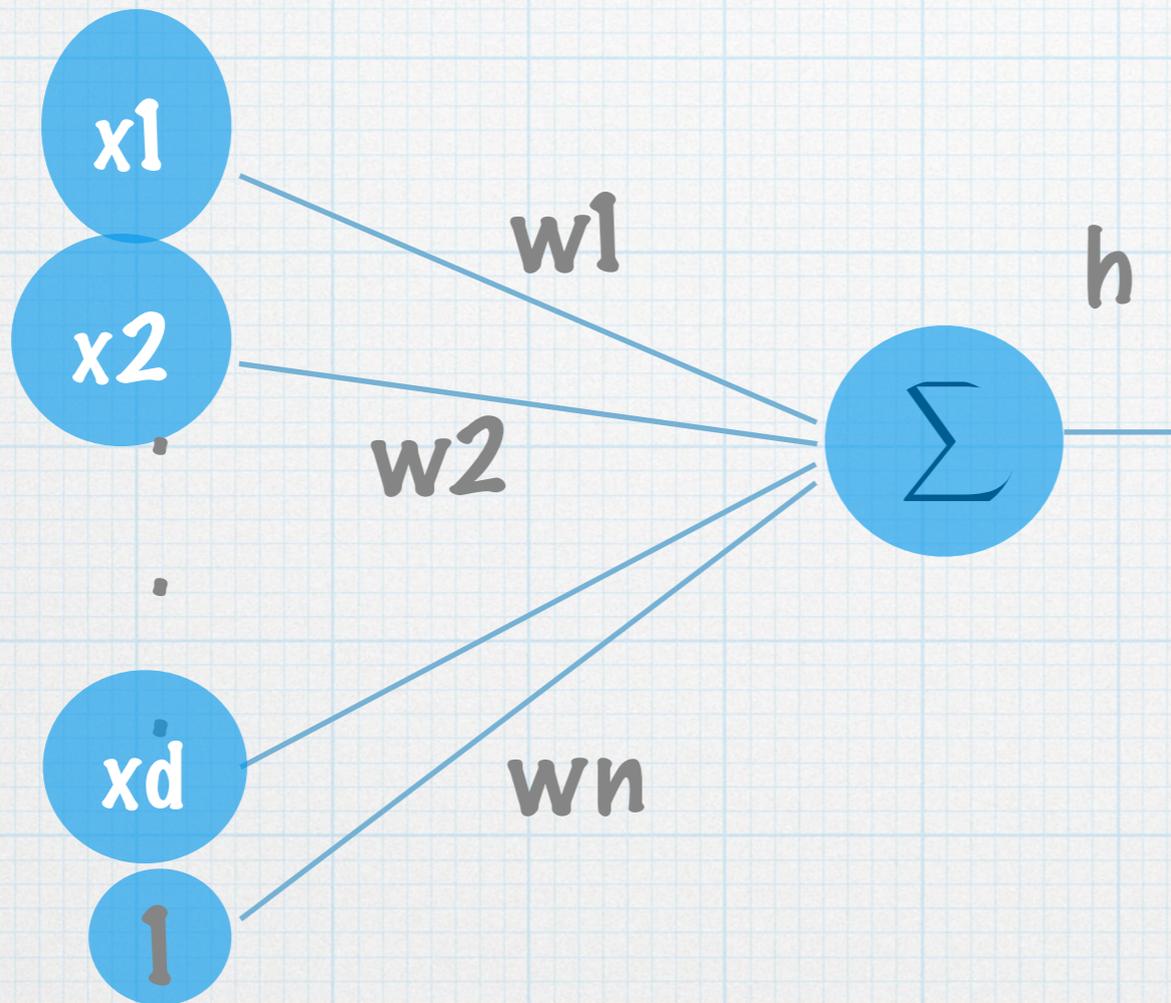
```
[ a1_1, a1_2, a1_3]  
[ a2_1, a2_2, a2_3]
```

```
>> A(1,:)
```

```
ans =
```

```
[ a1_1, a1_2, a1_3]
```

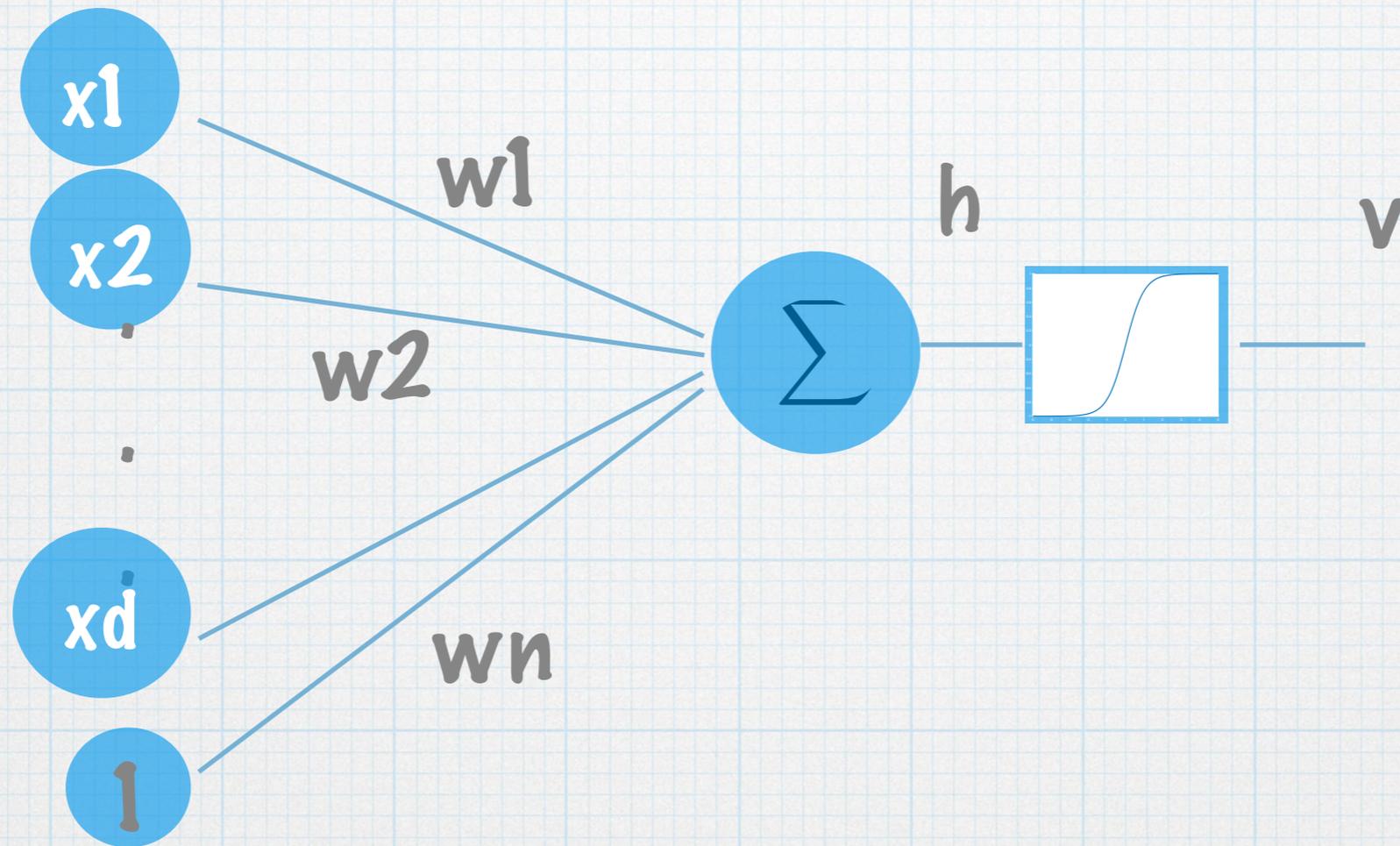
# Linear projection



```
x=sym('x',[3 1]);  
w=sym('w',[3 1]);  
s=char(sum(w.*x));  
ss=vector_input(s,'w','x');  
f=inline(ss,'x','w')  
f[1 1 1],[2 2 2])
```

```
function ss=vector_input(s,c1,c2)
    L=length(s);i=1;ss=[];
    while i <= L-1
        if ((s(i)==c1 | s(i)==c2) & (s(i+1)<='9' & s(i+1) >= '0'))
            start_i=i;
            while i<=L-1 & s(i+1) <= '9' & s(i+1) >= '0'
                i=i+1;
            end
            end_i=i;
            ss=[ss s(start_i) '(' s(start_i+1:end_i) ')'];
            i=i+1;
        else
            ss=[ss s(i)];
            i=i+1;
        end
    end
end
```

# A perceptron



```
>> x=sym('x',[3 1])
s=char(sum(w.*x))
ss=vector_input(s,'w','x')

f=inline(ss,'x','w')
```

```
x =
```

```
 x1
 x2
 x3
```

```
s =
```

```
 'w1*x1 + w2*x2 + w3*x3'
```

```
ss =
```

```
 'w(1)*x(1) + w(2)*x(2) + w(3)*x(3)'
```

```
f =
```

```
 Inline function:
```

```
 f(x,w) = w(1)*x(1) + w(2)*x(2) + w(3)*x(3)
```

```
>> f([1 1 1],[2 2 2])
```

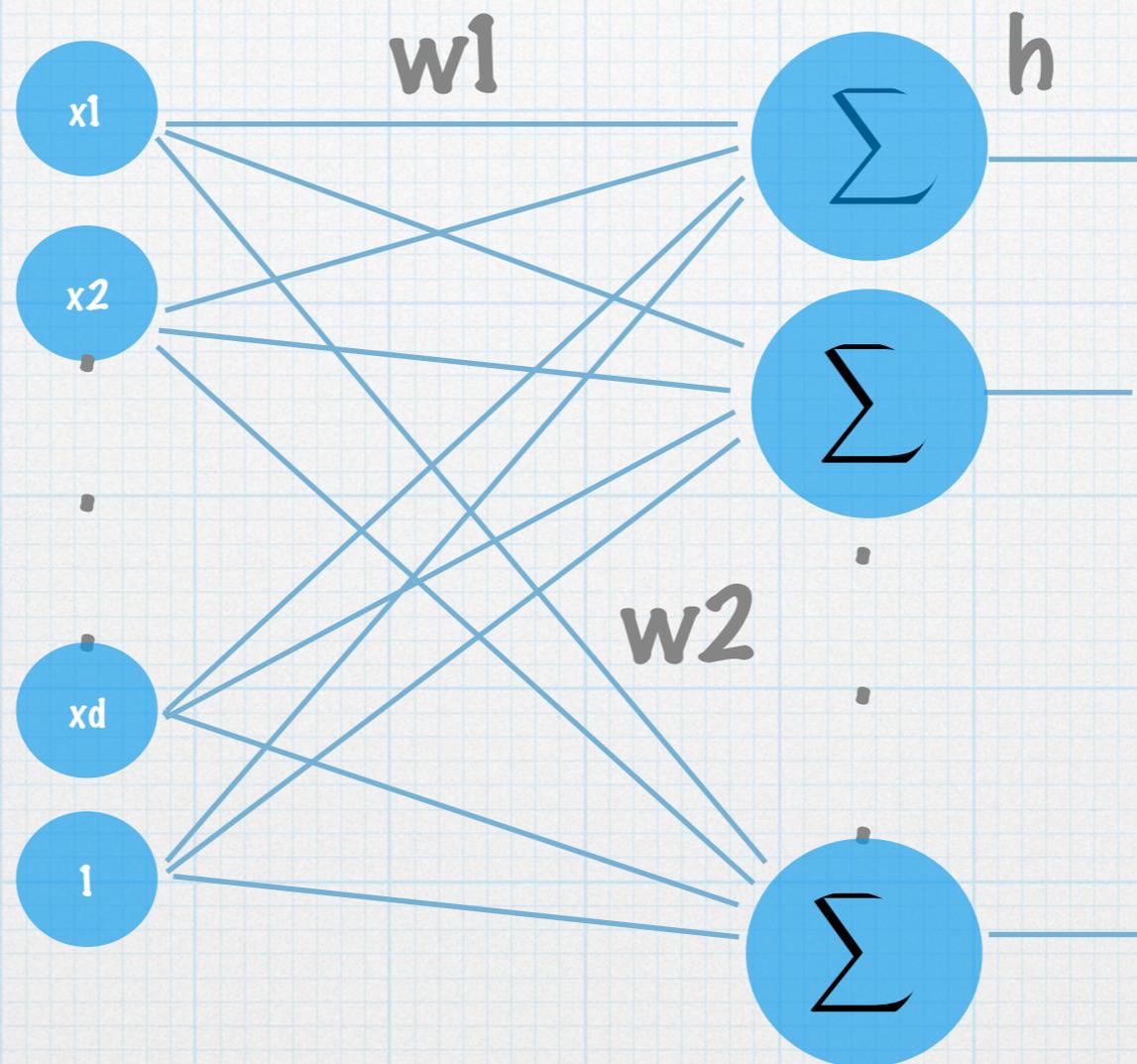
```
ans =
```

```
 6
```

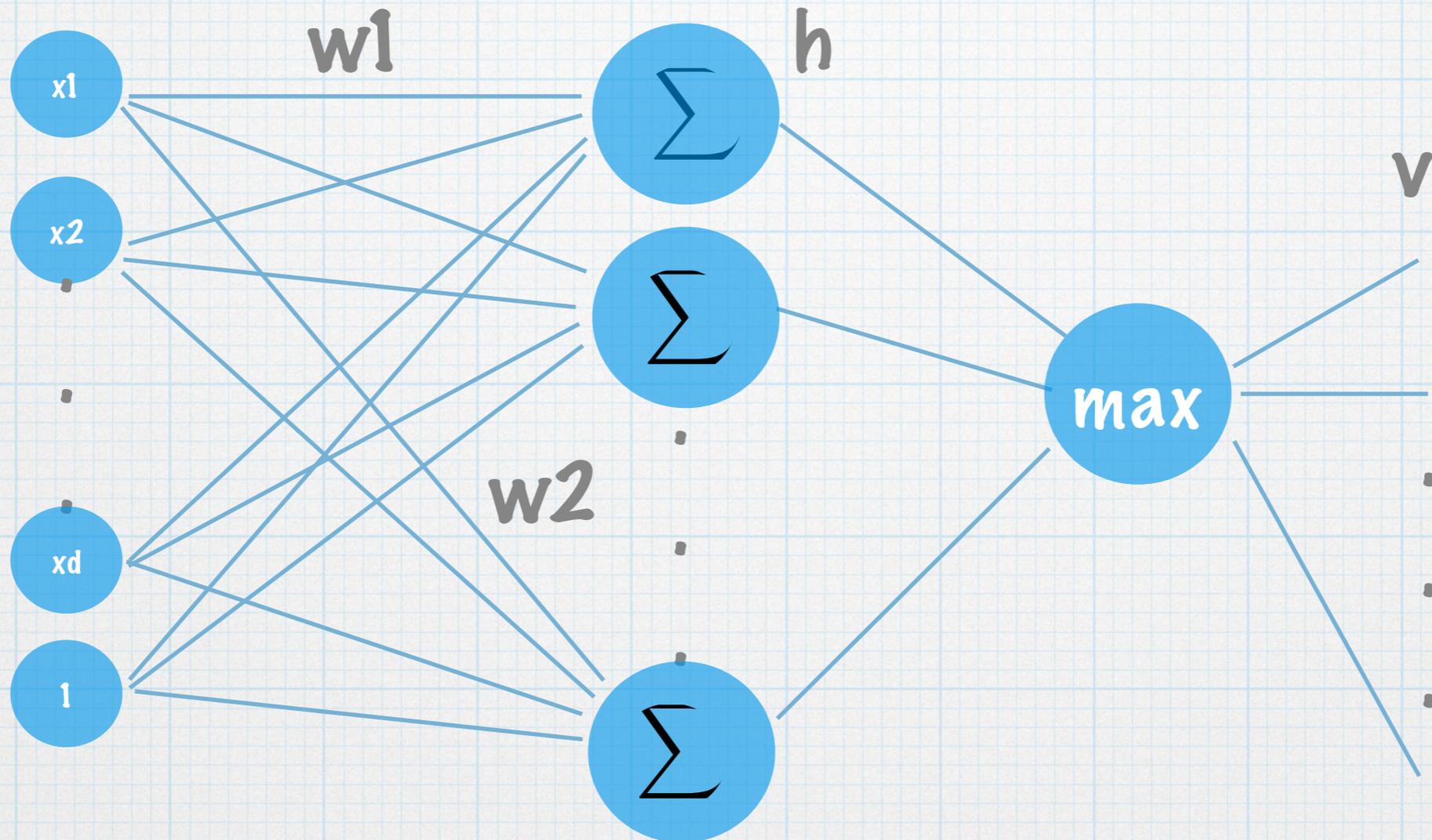
```
>>
```

```
x=sym('x',[3 1]);  
w=sym('w',[3 1]);  
s=char(tanh(sum(w.*x)) );  
ss=vector_input(s,'w','x');  
f=inline(ss,'x','w')  
f([1 1 1],[2 2 2])
```

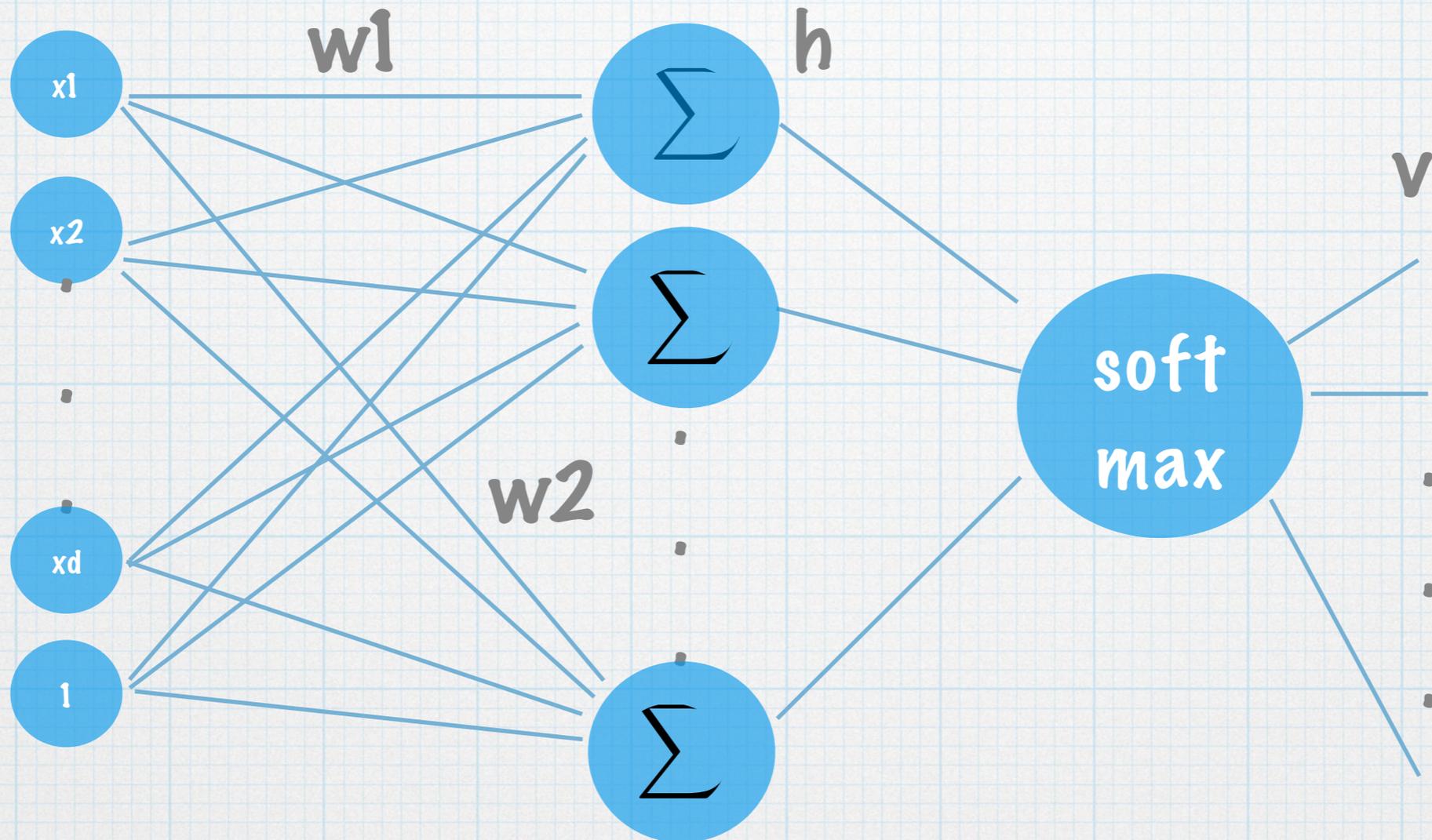
# Linear mapping



# Winner-take-all



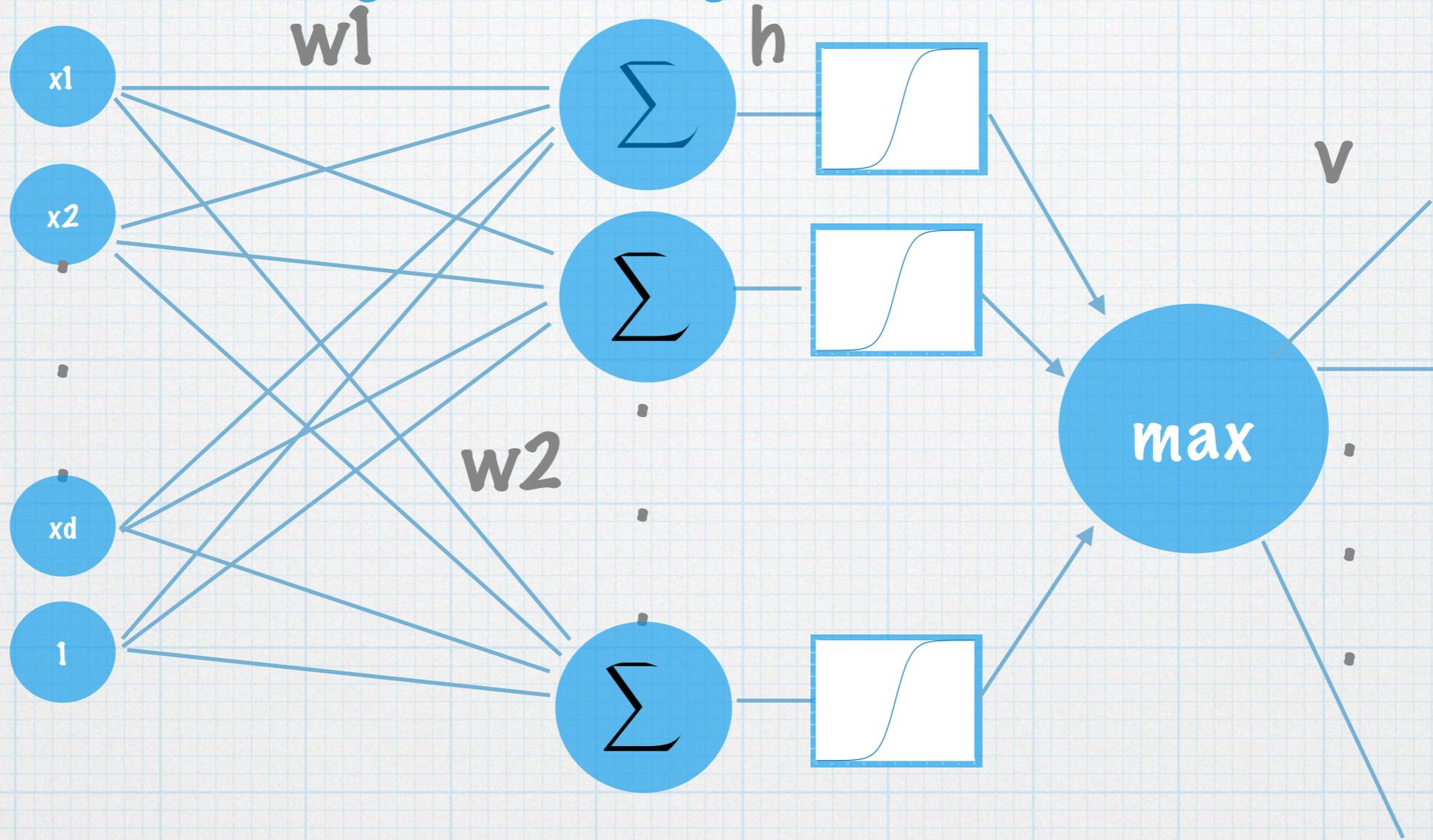
# Softmax



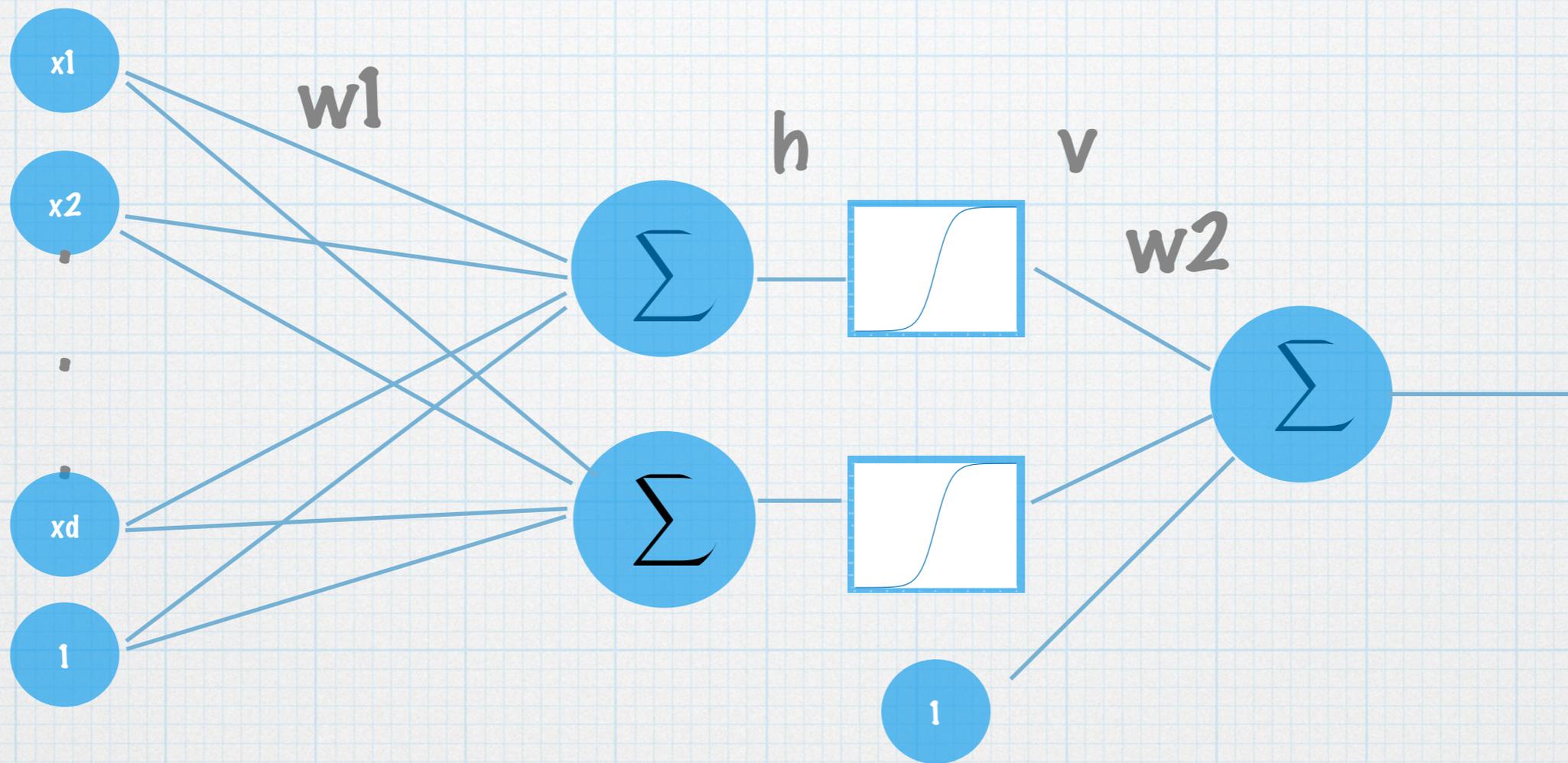


$$v_i = \frac{\exp(h_i)}{\sum_j \exp(h_j)}$$

# perceptrons



# perceptrons



$$y = f(x) = w_2 \tanh(w_1 x)$$

# object oriented programming

```
classdef perceptrons
    %UNTITLED12 Summary of this class goes here
    % Detailed explanation goes here

    properties
    end

    methods
    end

end
```

```
classdef perceptrons
```

```
    %UNTITLED Summary of this class goes here
```

```
    % Detailed explanation goes here
```

```
    properties
```

```
        layers;
```

```
        w;
```

```
        nL;
```

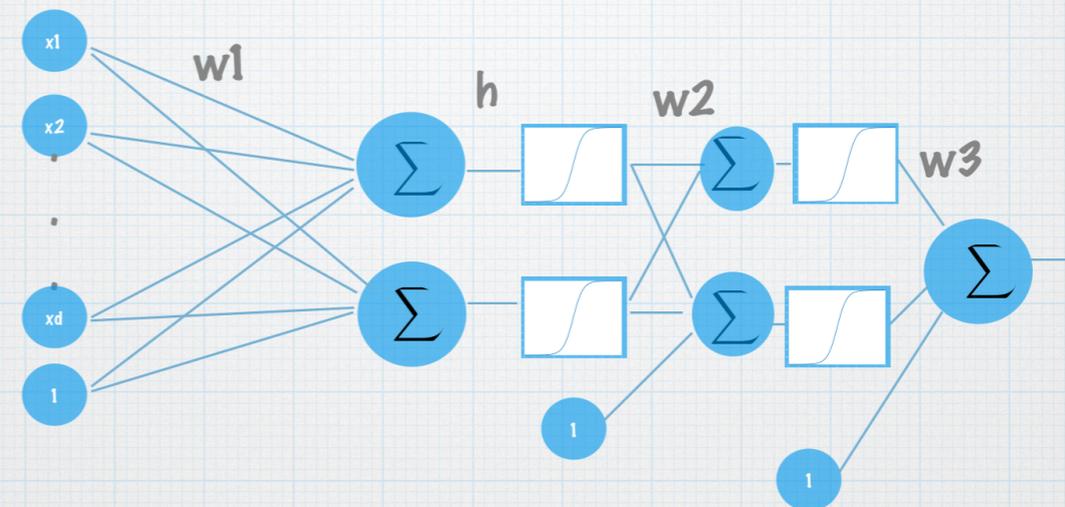
```
        a;
```

```
    end
```

```
    methods
```

```
    end
```

## Deep perceptrons



$$y = f(x) = w_3 \tanh(w_2 \tanh(w_1 x))$$

```
methods
```

```
function obj=perceptrons(layers,w,nL)
```

```
    % layers include input, hidden and output layers
```

```
    obj.w=w;
```

```
    obj.nL=nL;
```

```
    obj.layers=layers;
```

```
end
```

```
function obj=ff(obj,x)
```

```
    % exercise today
```

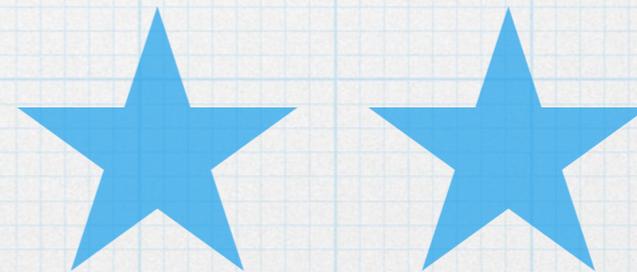
```
    % .
```

```
    % .
```

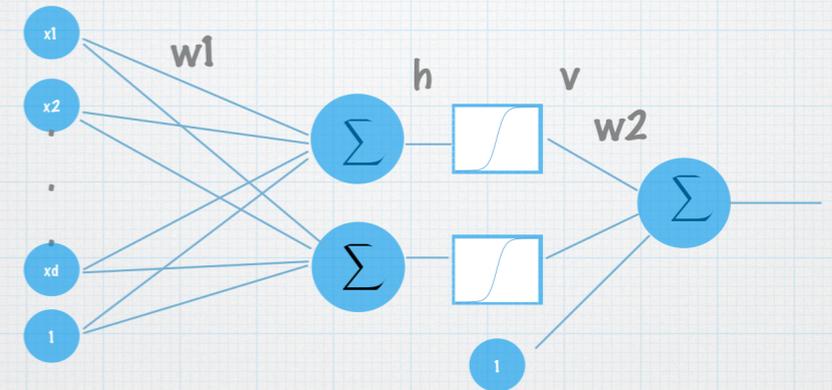
```
    % .
```

```
end
```

```
end
```



## perceptrons

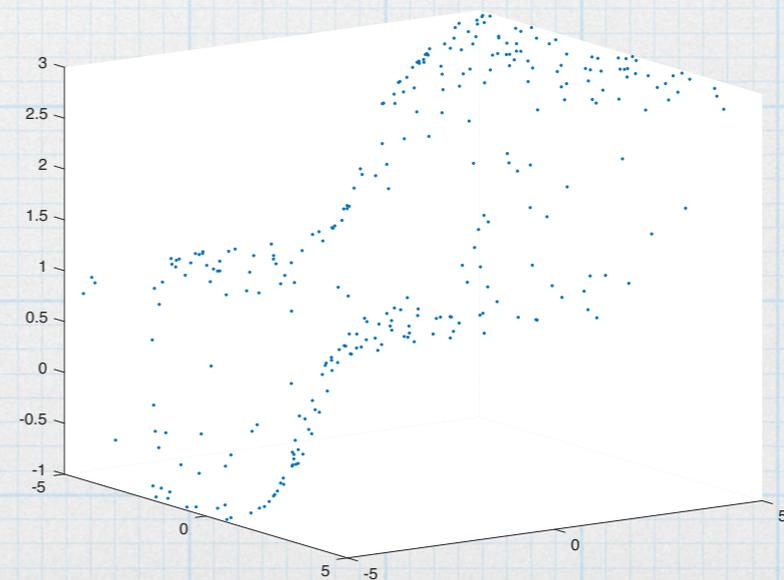


$$y = f(x) = w_2 \tanh(w_1 x)$$

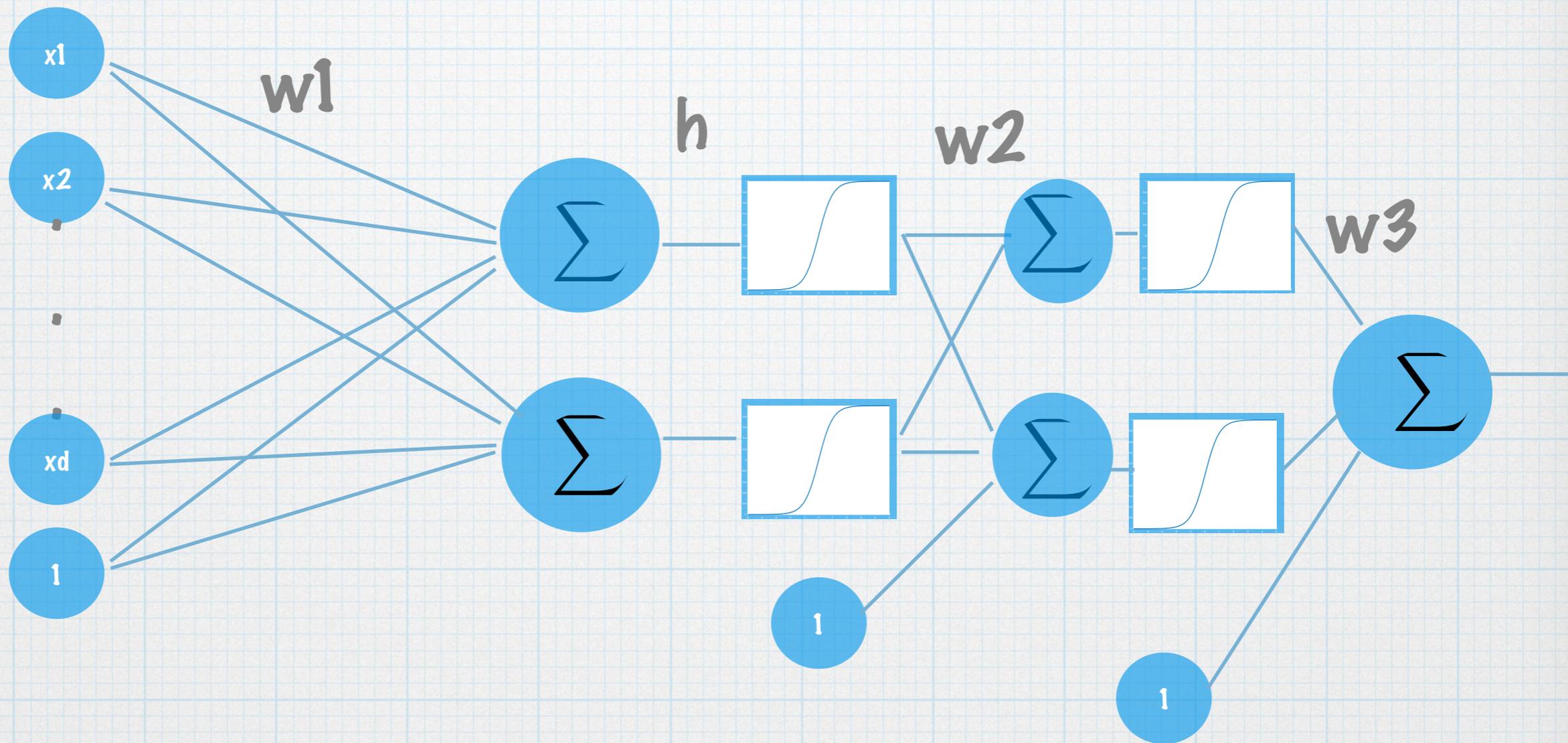
```
w{1}=[1 1 2;1 -1 -1]';nL{1}='tanh';  
w{2}=[1 -1 1]';nL{2}='linear';  
layers=3;
```

```
net=perceptrons(layers,w,nL);
```

```
x=rand(300,2)*10-5;  
net=net.ff(x);  
y=net.a{layers};  
plot3(x(:,1),x(:,2),y,'.')
```



# Deep perceptrons



$$y = f(x) = w_3 \tanh(w_2 \tanh(w_1 x))$$

```

w{1}=[1 1 2;1 -1 -1]';nL{1}='tanh';
w{2}=[1 -1 1;-1 -1 -1]';nL{2}='tanh';
w{3}=[1 1 1/2]';nL{3}='linear';
layers=4;

```

```

net=perceptrons(layers,w,nL);
x=rand(300,2)*10-5;
net=net.ff(x);
y=net.a{layers};
plot3(x(:,1),x(:,2),y,'.')

```

