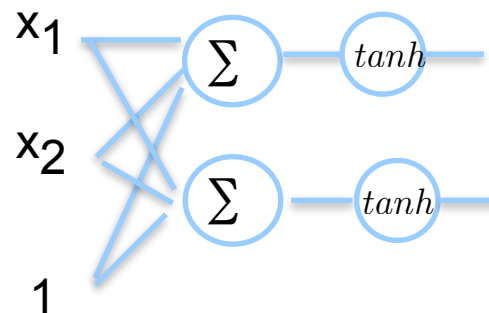


# Least Square Method

## Nonlinear Transformation Reconstruction

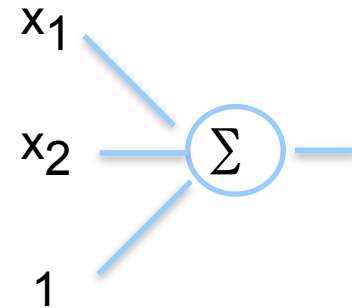
## Learning nonlinear transformation



```
x=rand(400,2);  
z(:,1) = 2*x(:,1)+x(:,2)-1;  
z(:,2)=x(:,1)-x(:,2)+1;  
a(1,:)=gradient_descent(x,z(:,1 ))  
a(2,:)=gradient_descent(x,z(:,2 ))
```

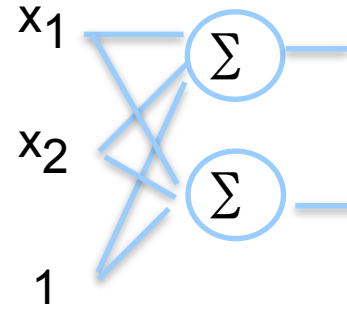
# Learning a linear relation

```
function a=gradient_descent(x,y)
max_loop=2000;
[N,d]=size(x);
X=[x ones(N,1)];
a=rand(1,d+1); hc=0; c=0.1;e=y-X*a';
E=100; loop=1;
while ~hc
    G=mean(X.*(e*ones(1,d+1))));
    a_new=a-c*G;
    y_hat=X*a_new';
    e_new=y_hat-y;
    E_new=mean(e_new.^2);
    if mod(loop,100)==0
        fprintf('loop %d mse %f\n',loop,E_new);
    end
    if E_new < E & loop < max_loop
        a=a_new;e=e_new;
        E=E_new;
    else
        hc=1;
    end
    loop=loop+1;
end
```



# Two linear relations

- Two-input-two-output
- Revise the flow chart



```
z(:,1) = 2*x(:,1)+x(:,2)-1;  
z(:,2)=x(:,1)-x(:,2)+1;  
a=gradient_descent(x,z );
```

# Vector Form of gradients

*Minimize*

$$E(\mathbf{a}) = \frac{1}{2n} \sum_{i=1}^n \sum_{j=1}^2 (\mathbf{x}_i^T \mathbf{a}_j - b_{ji})^2$$

$$\nabla(\mathbf{a}) = \frac{dE(\mathbf{a})}{d\mathbf{a}}$$

$$= \left[ \frac{dE(\mathbf{a})}{da_1} \quad \frac{dE(\mathbf{a})}{da_2} \right]$$

$$= \left[ \frac{1}{n} \sum_{i=1}^n (\mathbf{x}_i^T \mathbf{a}_1 - b_{1i}) \mathbf{x}_i \quad \frac{1}{n} \sum_{i=1}^n (\mathbf{x}_i^T \mathbf{a}_2 - b_{2i}) \mathbf{x}_i \right]$$

$$= \frac{1}{n} \left[ \sum_{i=1}^n e_{1i} \mathbf{x}_i \quad \sum_{i=1}^n e_{2i} \mathbf{x}_i \right]$$

# Gradient descent method

- Updating

$$\nabla(\mathbf{a}) = \frac{dE(\mathbf{a})}{da}$$

$$\mathbf{a} \leftarrow \mathbf{a} - \lambda \nabla(\mathbf{a})$$

# Learning many linear relations

error, e, is  
a column  
vector

```
...
a=rand(1,d+1); hc=0; c=0.1; e=y-X*a';
E=100; loop=1;
while ~hc
    G=mean(X.*(e*ones(1,d+1)));
    a_new=a-c*G;
    y_hat=X*a_new';
    e_new=y_hat-y;
    E_new=mean(e_new.^2);
```

error matrix, e,  
contains two  
column vectors

1. One linear relation
2. Multiple input single output

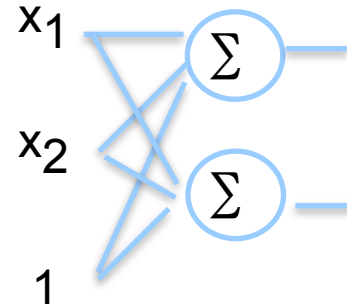
1. Many linear relations
2. Multiple input multiple output

```
a=rand(2,d+1);
hc=0; c=0.1; e=y-X*a';
E=100; loop=1;
while ~hc
    G=[mean(X.*(e(:,1)*ones(1,d+1))); mean(X.*(e(:,2)*ones(1,d+1)))]';
    a_new=a-c*G;
    y_hat=X*a_new';
    e_new=y_hat-y;
    E_new=sum(mean(e_new.^2));
```

$$\frac{1}{n} \left[ \sum_{i=1}^n e_{1i} \mathbf{x}_i \quad \sum_{i=1}^n e_{2i} \mathbf{x}_i \right]$$

# MIMO

- Two-input-two-output
- Revise the flow chart



```
z(:,1) = 2*x(:,1)+x(:,2)-1;  
z(:,2)=x(:,1)-x(:,2)+1;  
a=gradient_descent(x,z );
```



# Linear MIMO transformation

function a=gradient\_descent\_mimo(x,y)

start

```
[N,d]=size(x);  
X=[x ones(N,1)];  
a=rand(2,d+1); hc=0; c=0.01; e=y-X*a'  
E=sum(mean(e.^2))
```

1. Many linear relations
2. Multiple input multiple output

~hc

exit

```
if E < E_new  
    a=a_new; e=e_new  
    E=E_new  
else  
    hc=1;  
end
```

```
G=∇(a)  
a_new=a-c*G  
y_hat=X*a_new'  
e_new=y_hat-y  
E_new=mean(e_new.^2)
```

```
G=mean(X.*(e(:,1))*ones(1,d+1)));  
G=[G; mean(X.*(e(:,2))*ones(1,d+1))];
```

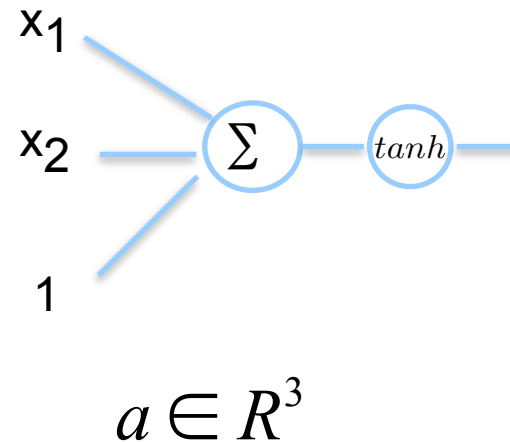
# Nonlinear Transformation

- tanh based nonlinear relation

$$y = \tanh(\mathbf{x}^T \mathbf{a})$$

$$\frac{dy}{d\mathbf{a}} = \frac{d \tanh(\mathbf{x}^T \mathbf{a})}{d\mathbf{a}}$$

$$= (1 - \tanh^2(\mathbf{x}^T \mathbf{a})) \mathbf{x}$$

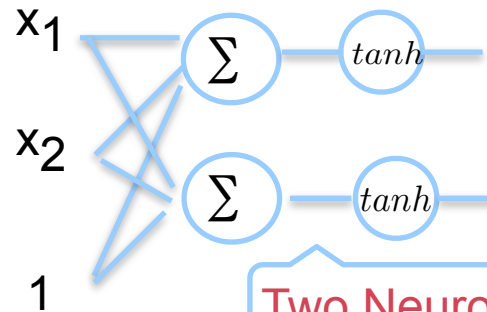


# Two tanh based nonlinear relations

$$\begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} \tanh(x^T a_1) \\ \tanh(x^T a_2) \end{bmatrix}$$

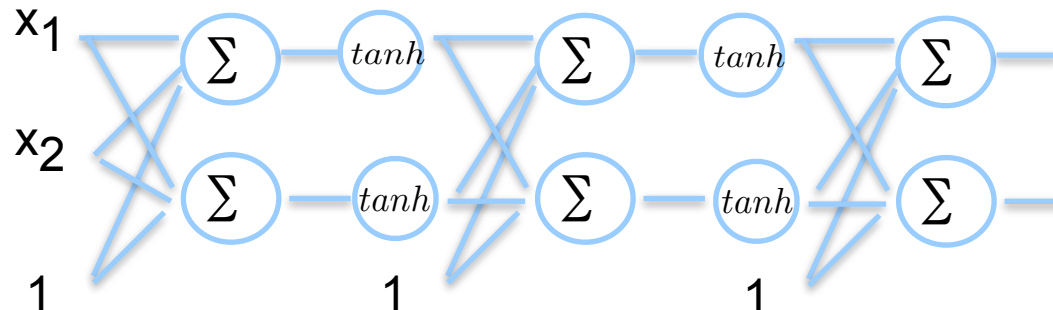
$$z(:,1) = \tanh(2x(:,1) + x(:,2) - 1);$$

$$z(:,2) = \tanh(x(:,1) - x(:,2) + 1);$$



Two Neurons  
Two neural elements

Deep Neural Networks



# Learning two nonlinear relations

*Minimize*

$$E(\mathbf{a}) = \frac{1}{2n} \sum_{i=1}^n \sum_{j=1}^2 (\tanh(\mathbf{x}_i^T \mathbf{a}_j) - b_{ji})^2$$

a 3x2  
matrix

$$\nabla(\mathbf{a}) = \frac{dE(\mathbf{a})}{d\mathbf{a}}$$

$$= \left[ \frac{dE(\mathbf{a})}{da_1} \quad \frac{dE(\mathbf{a})}{da_2} \right]$$

$$= \left[ \frac{1}{n} \sum_{i=1}^n \frac{d}{da_1} (\tanh(\mathbf{x}_i^T \mathbf{a}_1) - b_{1i})^2 \quad \frac{1}{n} \sum_{i=1}^n \frac{d}{da_2} (\tanh(\mathbf{x}_i^T \mathbf{a}_2) - b_{2i})^2 \right]$$

$$= \frac{1}{n} \left[ \sum_{i=1}^n e_{1i} \mathbf{x}_i \quad \sum_{i=1}^n e_{2i} \mathbf{x}_i \right]$$

# Learning two nonlinear relations

$$\begin{aligned}\nabla(\mathbf{a}) &= \frac{dE(\mathbf{a})}{d\mathbf{a}} \\ &= \left[ \frac{dE(\mathbf{a})}{da_1} \quad \frac{dE(\mathbf{a})}{da_2} \right] \\ &= \left[ \frac{1}{n} \sum_{i=1}^n \frac{d}{da_1} (\tanh(\mathbf{x}_i^T \mathbf{a}_1) - b_{1i})^2 \quad \frac{1}{n} \sum_{i=1}^n \frac{d}{da_2} (\tanh(\mathbf{x}_i^T \mathbf{a}_2) - b_{2i})^2 \right] \\ &= \frac{1}{n} \left[ \sum_{i=1}^n e_{1i} \mathbf{x}_i \quad \sum_{i=1}^n e_{2i} \mathbf{x}_i \right]\end{aligned}$$

$$\begin{aligned}y &= \tanh(\mathbf{x}^T \mathbf{a}) \\ \frac{dy}{d\mathbf{a}} &= \frac{d \tanh(\mathbf{x}^T \mathbf{a})}{d\mathbf{a}} \\ &= (1 - \tanh^2(\mathbf{x}^T \mathbf{a})) \mathbf{x}\end{aligned}$$

# Gradient descent method

- Updating

$$\nabla(\mathbf{a}) = \frac{dE(\mathbf{a})}{da}$$

$$\mathbf{a} \leftarrow \mathbf{a} - \lambda \nabla(\mathbf{a})$$

# MIMO

- Two-input-two-output
- Revise the flow chart
- Execute

```
z(:,1) = tanh( 2*x(:,1)+x(:,2)-1);  
z(:,2) = tanh(x(:,1)-x(:,2)+1);  
a=gradient_descent_mimo(x,z);
```

```
a=rand(2,d+1);
```

```
hc=0; c=0.1; e=y-X*a';
```

error  
calculation

```
E=100; loop=1;
```

```
while ~hc
```

```
G=[mean(X.*(e(:,1)*ones(1,d+1))); mean(X.*(e(:,2)*ones(1,d+1)))];
```

```
a_new=a-c*G;
```

```
y_hat=X*a_new';
```

y\_hat  
calculation

```
e_new=y_hat-y;
```

gardent  
calculation

Learning  
two linear  
relations

Learning two  
non-linear  
relations

```
a=rand(2,d+1);
```

```
hc=0; c=0.2; e=y-tanh(X*a');
```

```
E=100; loop=1;
```

```
while ~hc
```

```
dy=1-tanh(X*a').^2;
```

```
G=[mean(X.*(dy(:,1)*ones(1,d+1)).*(e(:,1)*ones(1,d+1)))];
```

```
G=[G; mean(X.*(dy(:,2)*ones(1,d+1)).*(e(:,2)*ones(1,d+1)))];
```

```
a_new=a-c*G;
```

```
y_hat=tanh(X*a_new');
```



# Learning two nonlinear relations

$$\begin{aligned}
 \nabla(\mathbf{a}) &= \frac{dE(\mathbf{a})}{d\mathbf{a}} \\
 &= \left[ \frac{dE(\mathbf{a})}{da_1} \quad \frac{dE(\mathbf{a})}{da_2} \right] \\
 &= \left[ \frac{1}{n} \sum_{i=1}^n \frac{d}{da_1} (\tanh(\mathbf{x}_i^T \mathbf{a}_1) - b_{1i})^2 \quad \frac{1}{n} \sum_{i=1}^n \frac{d}{da_2} (\tanh(\mathbf{x}_i^T \mathbf{a}_2) - b_{2i})^2 \right] \\
 &= \frac{1}{n} \left[ \sum_{i=1}^n e_{1i} \mathbf{x}_i \quad \sum_{i=1}^n e_{2i} \mathbf{x}_i \right]
 \end{aligned}$$

```

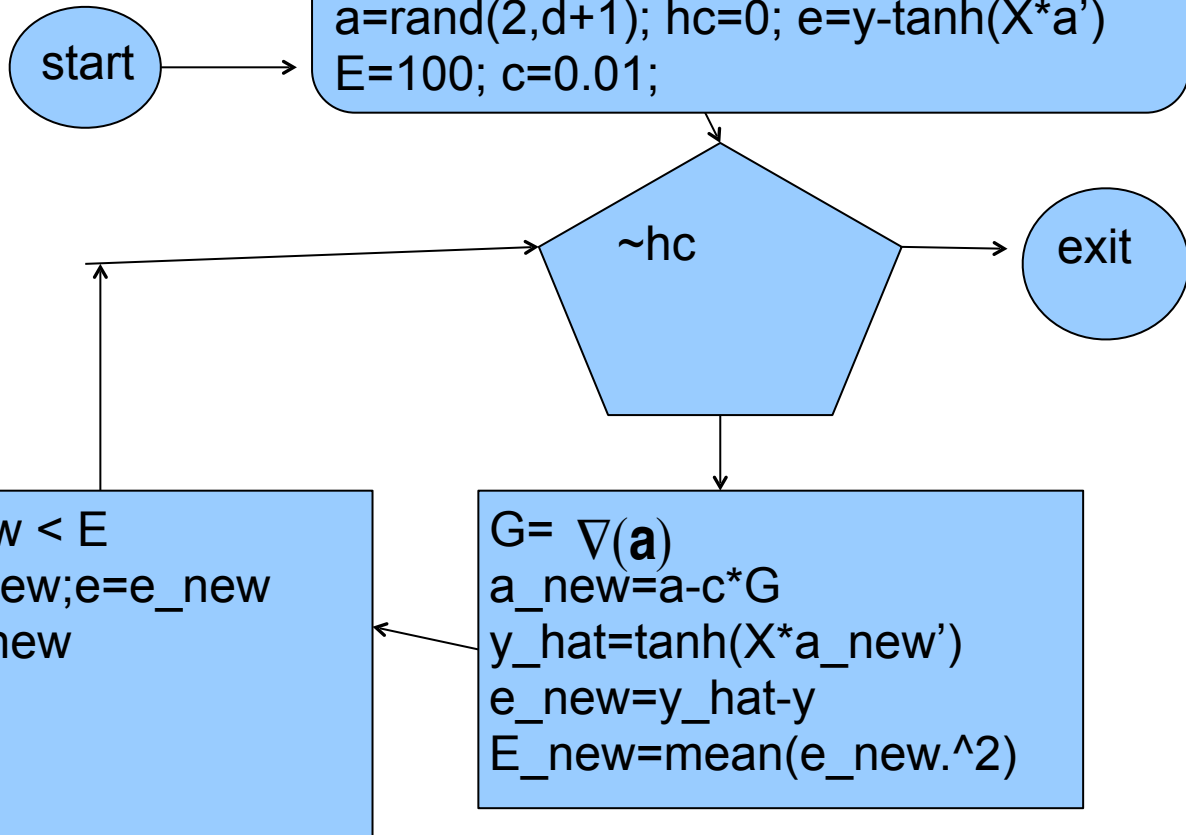
dy=1-tanh(X*a') .^2;
G=[mean(X.*(dy(:,1)*ones(1,d+1)).*(e(:,1)*ones(1,d+1)))]';
G=[G; mean(X.*(dy(:,2)*ones(1,d+1)).*(e(:,2)*ones(1,d+1)))]';
    
```

$$\begin{aligned}
 y &= \tanh(\mathbf{x}^T \mathbf{a}) \\
 \frac{dy}{d\mathbf{a}} &= \frac{d \tanh(\mathbf{x}^T \mathbf{a})}{d\mathbf{a}} \\
 &= (1 - \tanh^2(\mathbf{x}^T \mathbf{a})) \mathbf{x}
 \end{aligned}$$

# Learning multiple nonlinear relations

function a=gradient\_descent\_mimo(x,y)

```
[N,d]=size(x);  
X=[x ones(N,1)];  
a=rand(2,d+1); hc=0; e=y-tanh(X*a')  
E=100; c=0.01;
```



```
if E_new < E  
  a=a_new;e=e_new  
  E=E_new  
else  
  hc=1;  
end
```

```
G= ∇(a)  
a_new=a-c*G  
y_hat=tanh(X*a_new')  
e_new=y_hat-y  
E_new=mean(e_new.^2)
```

# Deep learning

How to learn a deep neural network ?

