

from the German adjective *eigen* ‘own.’



# Eigenvector and Eigenvalue Eigen-decomposition with change of basis

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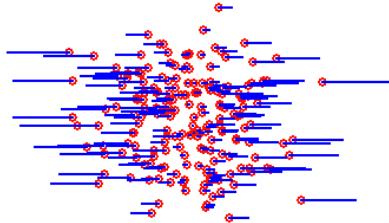
Some slides modified from Prof. Michale Fee's slides for  
9.40 - Intro to Neural Computation, MIT

# Preview

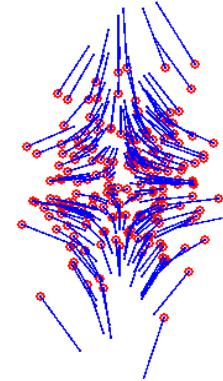
- To facilitate this section, I asked students to watch two youtube videos with animations:
- **Change of Basis:**  
[https://www.youtube.com/watch?v=P2LTAUO1TdA&t=0s&index=13&list=PLZHQBObOWTQDPD3MizzM2xVFitgF8hE\\_ab](https://www.youtube.com/watch?v=P2LTAUO1TdA&t=0s&index=13&list=PLZHQBObOWTQDPD3MizzM2xVFitgF8hE_ab)
- **Eigenvectors and Eigenvalues:**  
[https://www.youtube.com/watch?v=PFDu9oVAE-g&list=PLZHQBObOWTQDPD3MizzM2xVFitgF8hE\\_ab&index=13](https://www.youtube.com/watch?v=PFDu9oVAE-g&list=PLZHQBObOWTQDPD3MizzM2xVFitgF8hE_ab&index=13)

# Intro to Eigenvectors and eigenvalues

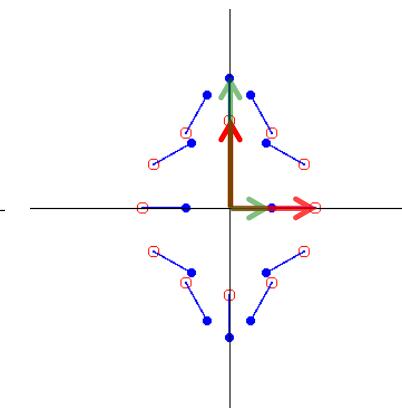
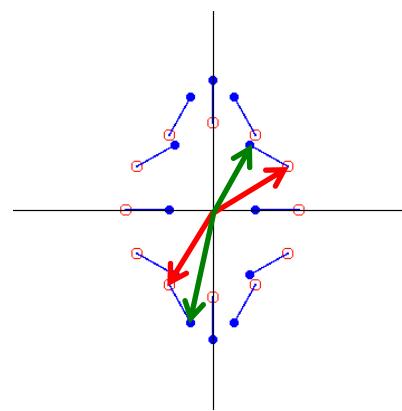
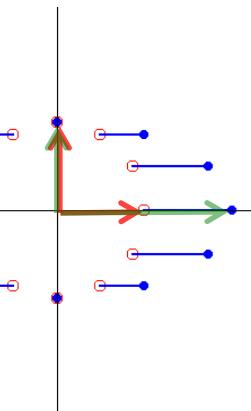
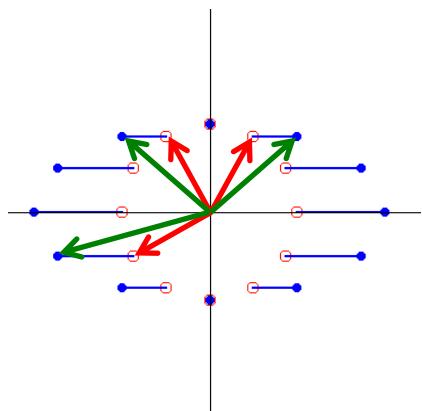
$$A = \begin{pmatrix} 1+\delta & 0 \\ 0 & 1 \end{pmatrix}$$



$$A = \begin{pmatrix} 1-\delta & 0 \\ 0 & 1+\delta \end{pmatrix}$$



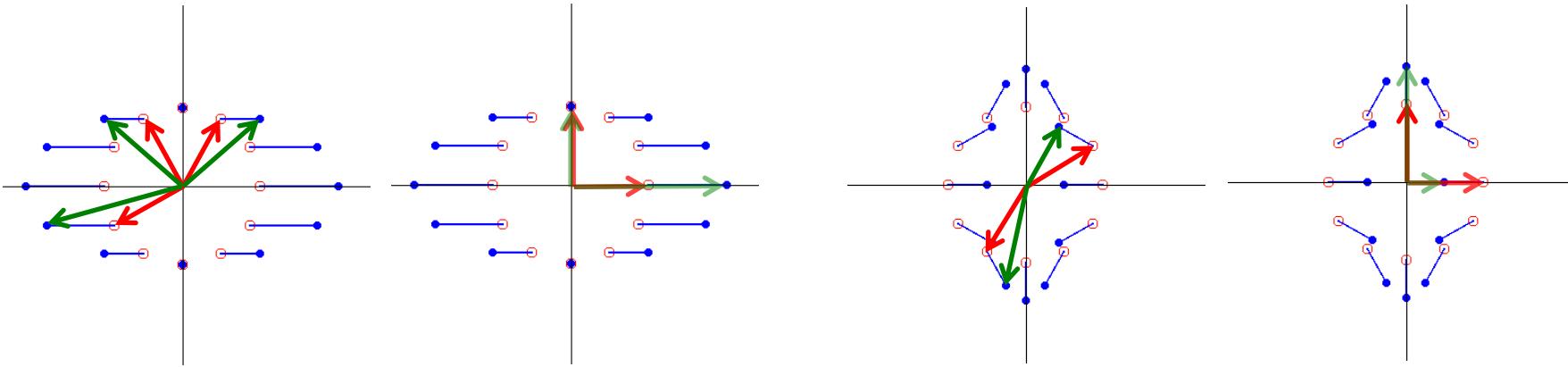
- Some vectors are rotated, some are not.
- Matrix transformations have special directions



These are all diagonal matrices

$$\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

- Some vectors are rotated, some are not.



- For a diagonal matrix, vectors along the axes are scaled, but not rotated.

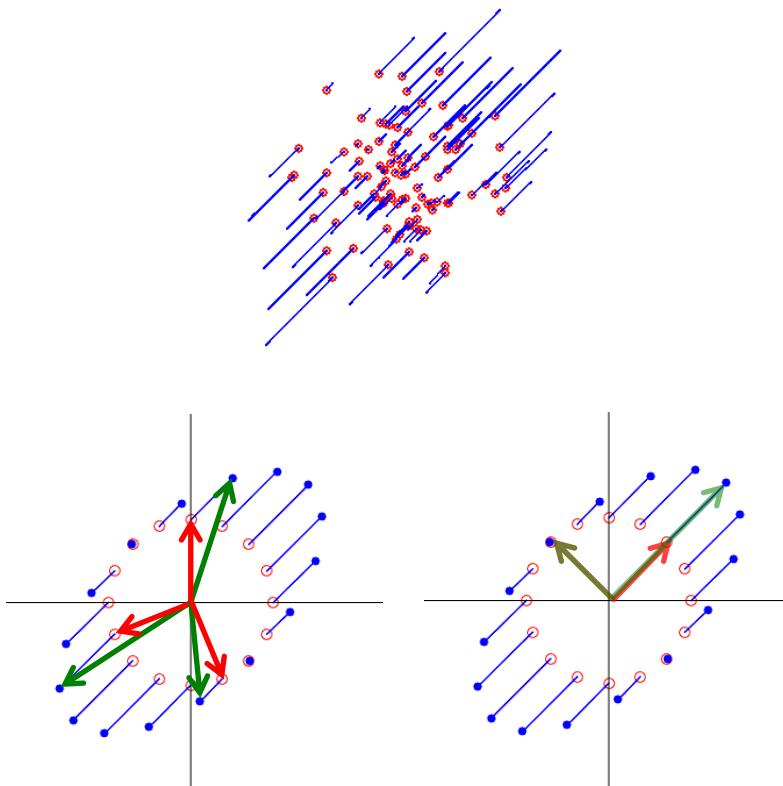
$$\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \lambda_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \Lambda \hat{e}_1 = \lambda_1 \hat{e}_1$$

$$\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \lambda_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad \Lambda \hat{e}_2 = \lambda_2 \hat{e}_2$$

# Intro to Eigenvectors and eigenvalues

- Another example

$$\begin{pmatrix} 1.5 & 0.5 \\ 0.5 & 1.5 \end{pmatrix}$$



# Eigenvectors and eigenvalues

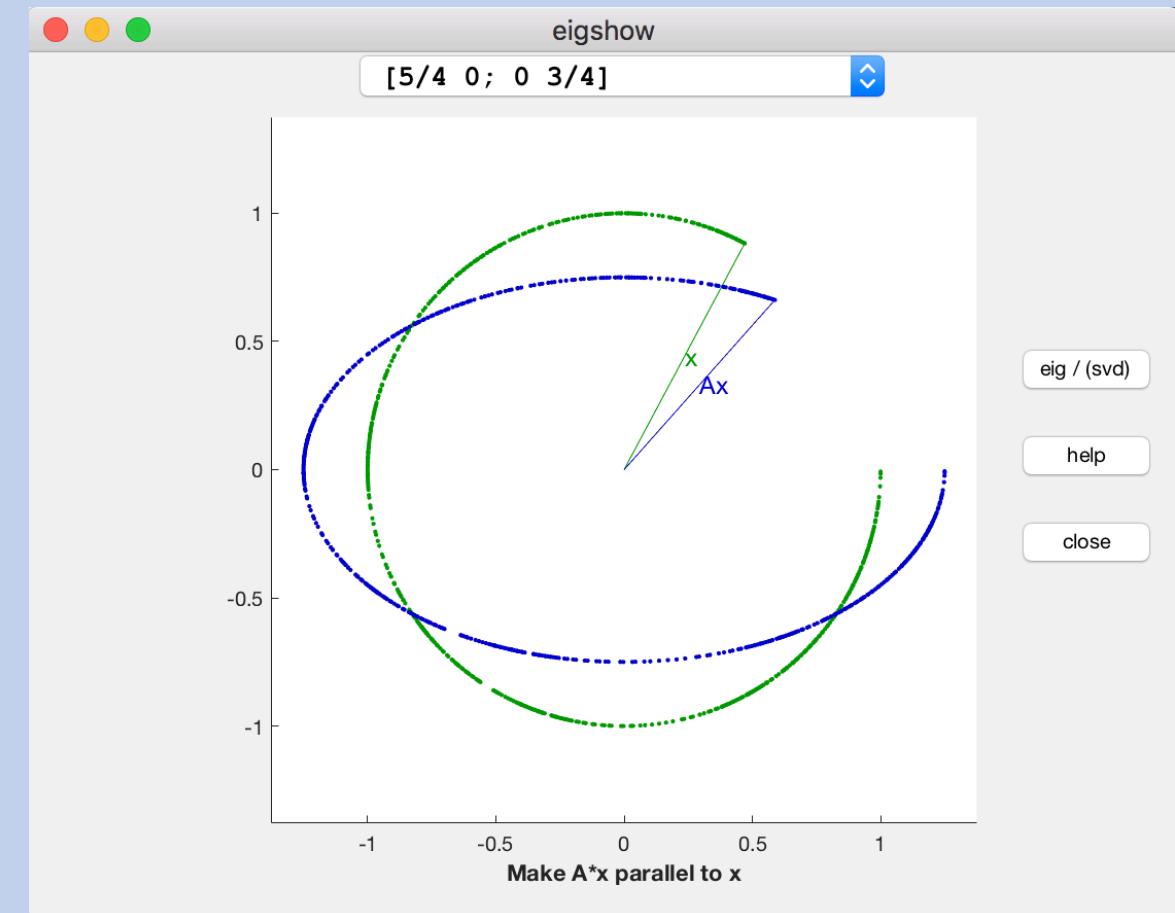
- Any vector  $\vec{x}$  that is mapped by matrix  $A$  onto a parallel vector  $\lambda\vec{x}$  is called an eigenvector of  $A$ . The scale factor  $\lambda$  is called the eigenvalue of vector  $\vec{x}$ .
- $A\vec{x} = \lambda\vec{x}$

Why do we need to find out eigenvectors and eigenvalues?

- To simplify calculation:
- $A\vec{x} = \lambda\vec{x}$  : what is  $A^{100}\vec{x}$  ?

# Optional: Some visualization (no need to dig into it)

- You can take a look at  
<https://blogs.mathworks.com/cleve/2013/07/08/eigshow-week-1/>
- Code for eigshow() is emailed to you.



# 求解 Eigenvectors and eigenvalues

- What are the eigenvalues of a general 2-dim matrix A ?

$$A\vec{x} = \lambda \vec{x}$$

we only want solutions where

$$A\vec{x} = \lambda I \vec{x}$$

$$\vec{x} \neq 0$$

$$(A - \lambda I)\vec{x} = 0$$

Which means  $A - \lambda I$  has a non-empty nullspace, thus it is not invertible:  $\det(A - \lambda I) = 0$

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$A - \lambda I = \begin{pmatrix} a - \lambda & b \\ c & d - \lambda \end{pmatrix}$$

$$\det(A - \lambda I) = (a - \lambda)(d - \lambda) - bc$$

$$\det(A - \lambda I) = (a - \lambda)(d - \lambda) - bc = 0$$

$$ad - \lambda(a + d) + \lambda^2 - bc = 0$$

Characteristic equation of matrix A

$$\lambda^2 - (a + d)\lambda + (ad - bc) = 0$$

$$\lambda_1 \lambda_2 = \det(A) ; \lambda_1 + \lambda_2 = \text{trace}$$

# 求解 Eigenvalues

- What are the eigenvalues of a general 2-dim matrix?

Characteristic equation of matrix A

$$\lambda^2 - (a+d)\lambda + (ad - bc) = 0$$

$$\lambda_1 \lambda_2 = \det A; \lambda_1 + \lambda_2 = \text{trace}$$

- Solutions are given by the quadratic formula

$$\lambda_{\pm} = \frac{1}{2}(a+d) \pm \frac{1}{2}\sqrt{(a-d)^2 + 4bc}$$



eigenvalues can be real, complex, imaginary

(one way to understand them will be explored in Chapter 6.3)

- For matrix not invertible: there must be one eigenvalue equal to 0.  
(zero is an eigenvalue iff A is not invertible)

# 求解 Eigenvalues

$$\lambda_{\pm} = \frac{1}{2}(a+d) \pm \frac{1}{2}\sqrt{(a-d)^2 + 4bc}$$

- For a symmetric matrix

$$A = \begin{pmatrix} a & b \\ b & d \end{pmatrix}$$

$$A = \begin{pmatrix} 3/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

$$\lambda_{\pm} = \frac{1}{2}(a+d) \pm \frac{1}{2}\sqrt{(a-d)^2 + 4b^2}$$

$\underbrace{\hspace{10em}}$   $\geq 0$

$$\lambda_{\pm} = \frac{1}{2}\left(\frac{3}{2} + \frac{1}{2}\right) \pm \frac{1}{2}\sqrt{\left(\frac{3}{2} - \frac{1}{2}\right)^2 + 4\left(\frac{1}{2}\right)^2}$$

The eigenvalues of a symmetric matrix are always real.

$$\lambda_{\pm} = 1 \pm \frac{\sqrt{2}}{2}$$

# 求解 Eigenvectors

- 算出eigenvalue后，分别带入原式求出对应的eigenvector

$$(A - \lambda I) \vec{x} = 0$$

- Some tricks on problem solving:
  - When dealing with 2 by 2 matrices: both rows of  $(A - \lambda I)$  are multiples of a vector  $(a,b)$ . The eigenvector is any multiple of  $(b,a)$ .
  - Use the product of eigenvalues = det; the sum of eigenvalues = trace to check your answers.

# Eigenvectors

- Note that eigenvectors are not unique...

If  $\vec{x}_i$  is an eigenvector of A      then so is  $a\vec{x}_i$

$$A\vec{x}_i = \lambda_i \vec{x}_i \quad A(a\vec{x}_i) = \lambda_i(a\vec{x}_i)$$

... so we usually write eigenvectors as unit vectors

- Chapter 6.4: For a symmetric matrix, the eigenvectors of A are orthogonal (and we write them as unit vectors)...

... the eigenvectors of A form a complete orthonormal basis set.

- For eigenvectors of  $A$ :  $A\vec{x} = \lambda \vec{x}$ , then nicely we have  $A^n\vec{x} = \lambda^n \vec{x}$
- What about other arbitrary vectors? Can we simplify?

If we can write  $\vec{u}_0$  as:  $\vec{u}_0 = c_1\vec{x}_1 + c_2\vec{x}_2$ ,

$$\begin{aligned}\text{then } A^n\vec{u}_0 &= A^n(c_1\vec{x}_1 + c_2\vec{x}_2) \\ &= c_1A^n\vec{x}_1 + c_2A^n\vec{x}_2 \\ &= c_1\lambda_1^n\vec{x}_1 + c_2\lambda_2^n\vec{x}_2\end{aligned}$$

- Thus, as long as we find  $c_1$  and  $c_2$ , we can simplify as:

$$A^n\vec{u}_0 = c_1\lambda_1^n\vec{x}_1 + c_2\lambda_2^n\vec{x}_2$$

- Prerequisite: there are enough independent eigenvectors.
  - Independent  $\vec{x}$ 's from different  $\lambda$ 's (proved in your textbook, read it by yourself)
  - (digression) Question: (True or False) different  $\lambda$ 's from Independent  $\vec{x}$ 's ?
- Finding  $c_1$  and  $c_2$  is essentially a process called 'change of basis'

# Change of Basis

Decompose other non-eigenvectors into combinations of eigenvectors  
(Finding  $c_1$  and  $c_2$ )

- Let's look at an example (6.2 Example 3):

$$\begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix} \text{ has } \lambda_1 = 2 \text{ and } \vec{x}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}; \lambda_2 = -1 \text{ and } \vec{x}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

- Express  $\vec{u}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  by  $c_1 \vec{x}_1 + c_2 \vec{x}_2$  (find  $c_1$  and  $c_2$ )

You can do it easily by expressing the above equation as:

$$\vec{u}_0 = [\vec{x}_1 \vec{x}_2] \mathbf{c}$$

$$\text{Then } \mathbf{c} = [\vec{x}_1 \vec{x}_2]^{-1} \vec{u}_0$$

$$\mathbf{c} = S^{-1} \vec{u}_0, S = [\vec{x}_1 \vec{x}_2]$$

# Eigen-decomposition

- Now, find  $A^n \vec{u}_0$

Substitute it into the equation we derived previously:

$$A^n \vec{u}_0 = c_1 \lambda_1^n \vec{x}_1 + c_2 \lambda_2^n \vec{x}_2$$

$$A^n \vec{u}_0 = [\lambda_1^n \vec{x}_1, \lambda_2^n \vec{x}_2] c$$

$$A^n \vec{u}_0 = [\lambda_1^n \vec{x}_1, \lambda_2^n \vec{x}_2] S^{-1} \vec{u}_0 , S = [\vec{x}_1 \vec{x}_2]$$

$$A^n \vec{u}_0 = [\vec{x}_1, \vec{x}_2] \begin{bmatrix} \lambda_1^n & \\ & \lambda_2^n \end{bmatrix} S^{-1} \vec{u}_0$$

$$A^n \vec{u}_0 = S \begin{bmatrix} \lambda_1^n & \\ & \lambda_2^n \end{bmatrix} S^{-1} \vec{u}_0$$

Explain intuitively what it means.

Eigen-decomposition

$$A^n \vec{u}_0 = S \begin{bmatrix} \lambda_1^n & \\ & \lambda_2^n \end{bmatrix} S^{-1} \vec{u}_0$$

# Homework assistant: MATLAB

- Matlab has a function 'eig' to calculate eigenvectors and eigenvalues

```
>> A=[1.5 0.5;0.5 1.5]
A =
    1.5000    0.5000
    0.5000    1.5000
>> [F,V]=eig(A)
F =
    -0.7071    0.7071
    0.7071    0.7071
V =
    1         0
    0         2
>> F*V*F'
ans =
    1.5000    0.5000
    0.5000    1.5000
```

$$A = F V F^T$$

The Transpose here is because A is symmetric. For now, understand it as  $F^{-1}$

# Diagonalizing

Eigen-decomposition

$$A^n = S \begin{bmatrix} \lambda_1^n & \\ & \lambda_2^n \end{bmatrix} S^{-1}$$

Diagonalizing

$$S^{-1} A^n S = \begin{bmatrix} \lambda_1^n & \\ & \lambda_2^n \end{bmatrix}$$

# Application for diagonalization

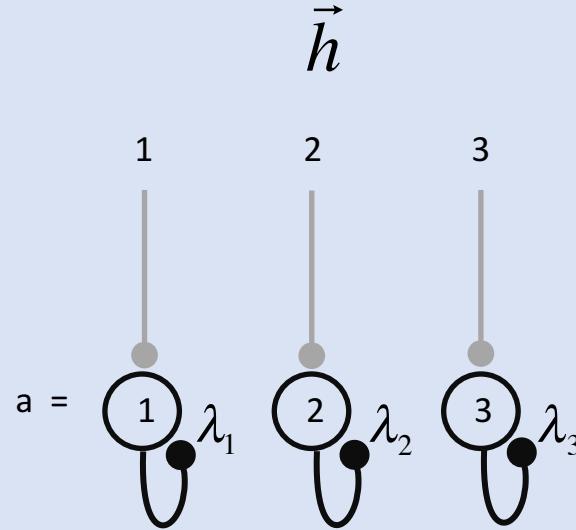
## Recurrent networks

- $M$  is a diagonal matrix

$$M = \Lambda = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \lambda_3 & \ddots \end{pmatrix}$$

$$\tau_n \frac{d\vec{v}}{dt} = -\vec{v} + \Lambda \vec{v} + \vec{h}$$

$$\tau_n \frac{dv_a}{dt} = -v_a + \lambda_a v_a + h_a$$

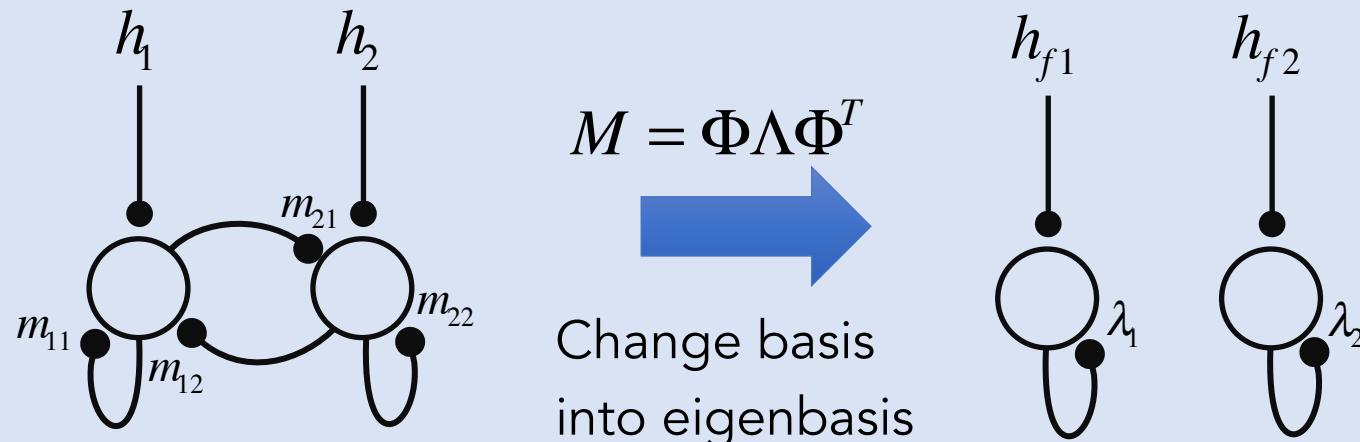


We have  $n$  independent equations –  
each neuron acts independently of all  
the others

- A simple system to solve

# Recurrent networks

- What if the system is not described by a diagonal matrix?
- We can let's make  $M$  diagonal!



$$M = \begin{pmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{pmatrix}$$

# Eigenvectors of AB

If  $x$  is an eigenvector of  $B$

$$ABx = A\beta x$$

$$ABx = \beta Ax \quad (1)$$

If  $x$  is also an eigenvector of  $A$



If  $AB = BA$

Similarly,

$$ABx = \beta\lambda x$$

Therefore,

$$BAx = \beta\lambda x$$

$$AB = BA$$

$$ABx = BAx \quad (2)$$

$$\beta Ax = BAx \text{ by (1) and (2)}$$

All  $Ax$ 's are also eigenvectors of  $B$ ,  
So eigenvectors of  $A$ 's are eigenvectors  
of  $B$

This side of reasoning is  
flawed. To prove the  
inverse, some advanced  
linear algebra is needed, we  
will skip them.

Suppose both  $A$  and  $B$  can be diagonalized, they share  
the same eigenvector matrix  $S$  if and only if  $AB=BA$

## More exercises

- If  $A$  has eigenvalues 1 and 3, and eigenvectors  $x_1$  and  $x_2$ , what are the eigenvalues and eigenvectors of matrix  $A-1$ ? (Strang 6.1-A)

$$\begin{matrix} 1 & 2 & 3 & 2 & 2 & 2 \\ 0 & 4 & 5 & 2 & 2 & 2 \\ 0 & 0 & 6 & 2 & 2 & 2 \end{matrix}$$

- Try it yourself (when you review for quiz):
  - Strang 6.1-21: **The eigenvalues of  $A$  equal the eigenvalues of  $A^T$**
  - Strang 6.2-4, 31, 34