

from the German adjective *eigen* ‘own.’



Eigenvector and Eigenvalue

Eigen-decomposition

with change of basis

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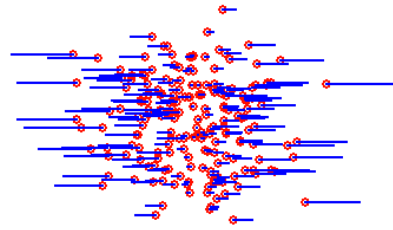
Some slides modified from Prof. Michale Fee's slides for
9.40 - Intro to Neural Computation, MIT

Preview

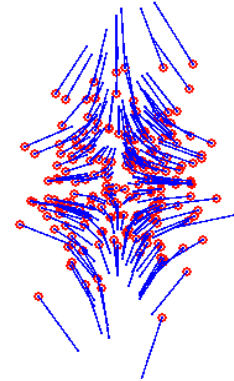
- To facilitate this section, I asked students to watch two youtube videos with animations:
- **Change of Basis:**
https://www.youtube.com/watch?v=P2LTAUO1TdA&t=0s&index=13&list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE_ab
- **Eigenvectors and Eigenvalues:**
https://www.youtube.com/watch?v=PFDu9oVAE-g&list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE_ab&index=13

Intro to Eigenvectors and eigenvalues

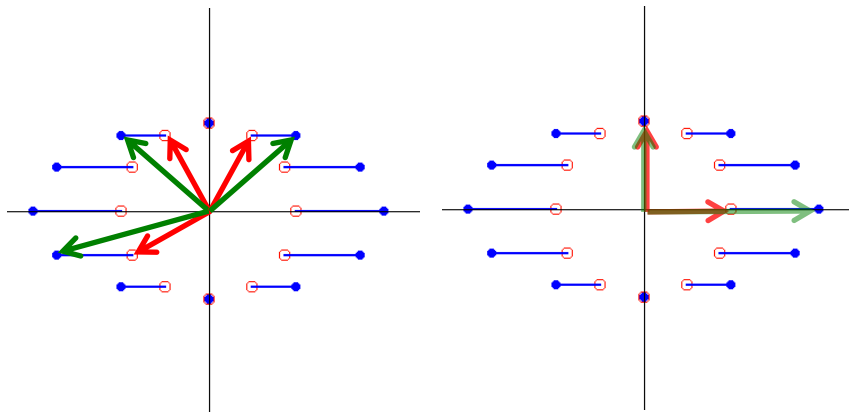
$$A = \begin{pmatrix} 1+\delta & 0 \\ 0 & 1 \end{pmatrix}$$



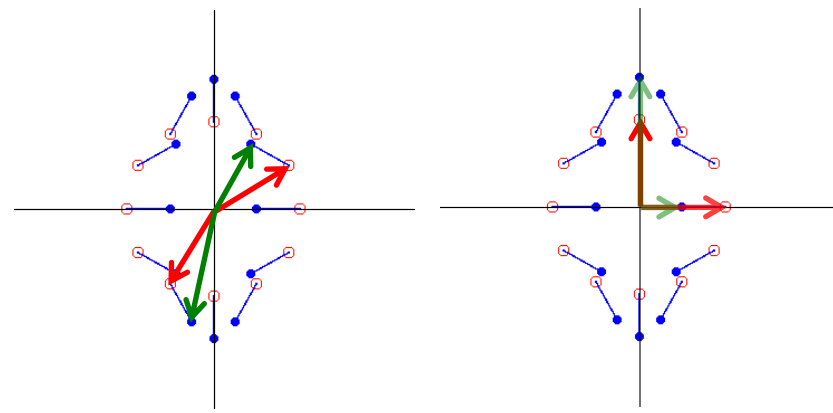
$$A = \begin{pmatrix} 1-\delta & 0 \\ 0 & 1+\delta \end{pmatrix}$$



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

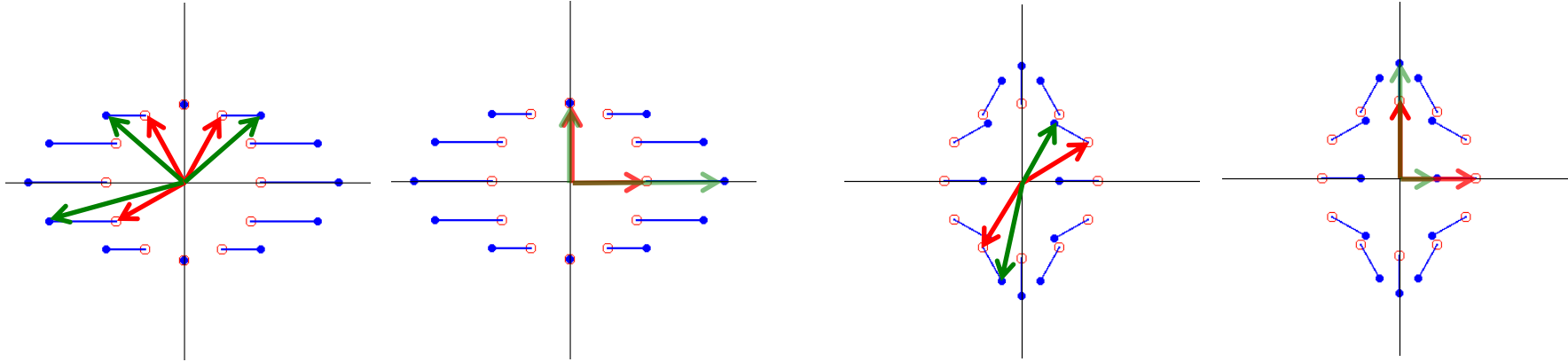


These are all diagonal matrices



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

- Some vectors are rotated, some are not.



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

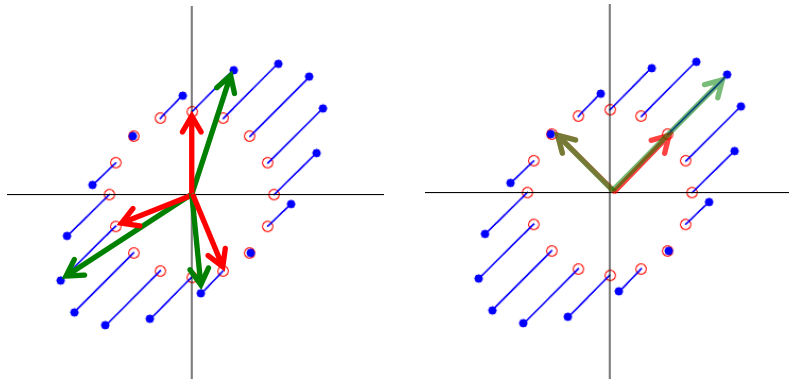
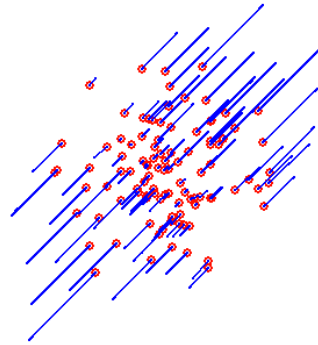
$$\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \lambda_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \Lambda \hat{e}_1 = \lambda_1 \hat{e}_1$$

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

Intro to Eigenvectors and eigenvalues

- Another example

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



Eigenvectors and eigenvalues

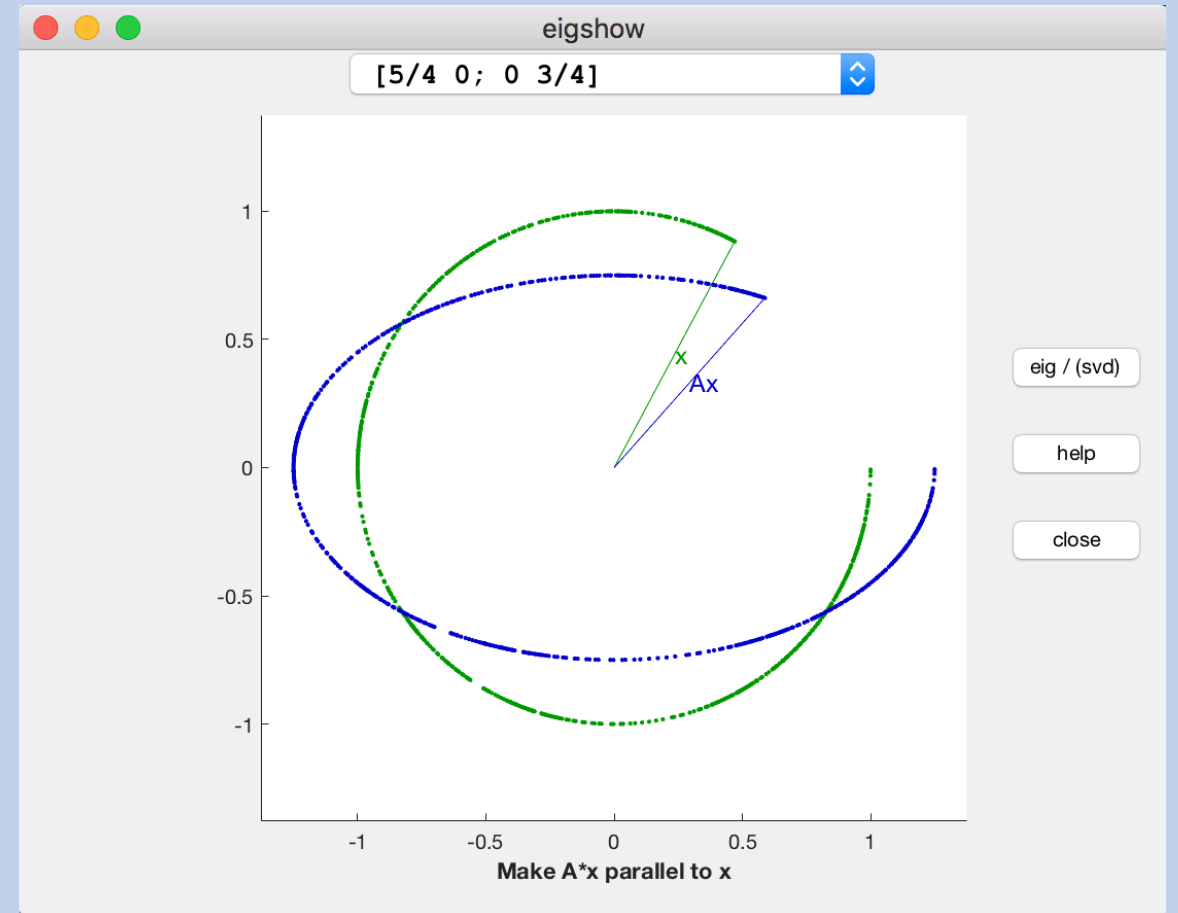
- Any vector \vec{x} that is mapped by matrix A onto a parallel vector $\lambda\vec{x}$ is called an eigenvector of A . The scale factor λ is called the eigenvalue of vector \vec{x} .
- $A\vec{x} = \lambda \vec{x}$

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

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

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

Which means $A - \lambda I$ has a non-empty nullspace, thus it is

$$(A - \lambda I)\vec{x} = 0 \quad \text{not invertible: } \det(A - \lambda I) = 0$$

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \quad A - \lambda I = \begin{pmatrix} a - \lambda & b \\ c & d - \lambda \end{pmatrix} \quad \det(A - \lambda I) = (a - \lambda)(d - \lambda) - bc$$

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

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

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

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

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

$$\lambda_{\pm} = \frac{1}{2}(a+d) \pm \frac{1}{2}\sqrt{(a-d)^2 + 4b^2} \geq 0$$

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

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 $a\vec{x}_i$

$$A\vec{x}_i = \lambda_i \vec{x}_i$$

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

... 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: $\vec{u}_0 = c_1 \vec{x}_1 + c_2 \vec{x}_2$,

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

- 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 λ 's (proved in your textbook, read it by yourself)
 - (digression) Question: (True or False) different λ '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: $A^n \vec{u}_0 = c_1 \lambda_1^n \vec{x}_1 + c_2 \lambda_2^n \vec{x}_2$

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

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

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

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

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
```

$$A = F V F^T$$

```
>> F*V*F'
```

```
ans =
```

```
1.5000    0.5000
0.5000    1.5000
```

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

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

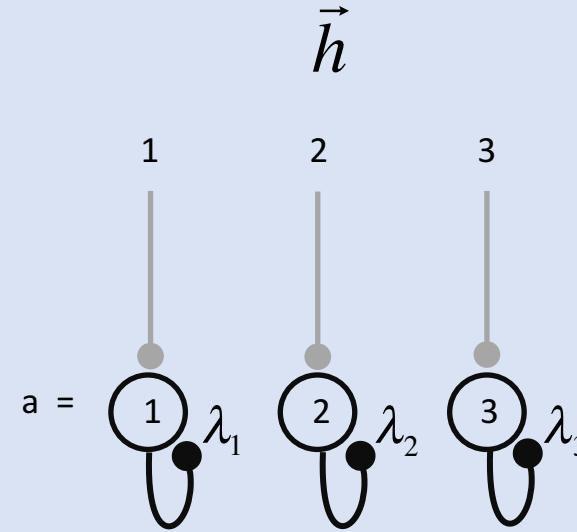
Application for diagonalization

Recurrent networks

- M is a diagonal matrix

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

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



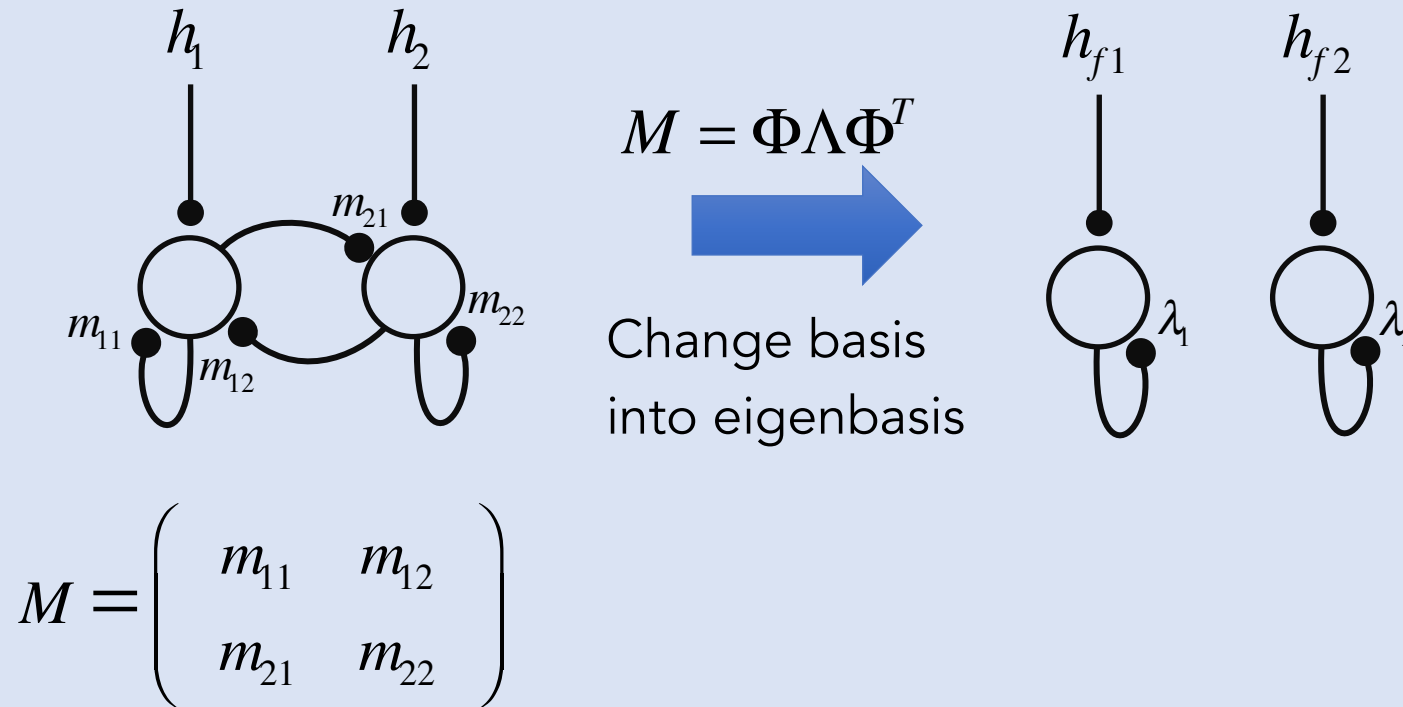
$$\tau_n \frac{dv_a}{dt} = -v_a + \lambda_a v_a + h_a$$

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!



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



$$ABx = \beta \lambda x$$

Similarly, $BAx = \beta \lambda x$

Therefore, $AB = BA$



If $AB = BA$

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

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

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 is the eigenvalues and eigenvectors of matrix A^{-1} ? (Strang 6.1-A)

- What are the eigenvalues of $\begin{bmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{bmatrix}; \begin{bmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{bmatrix}$ (Strang 6.1-29)

- 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