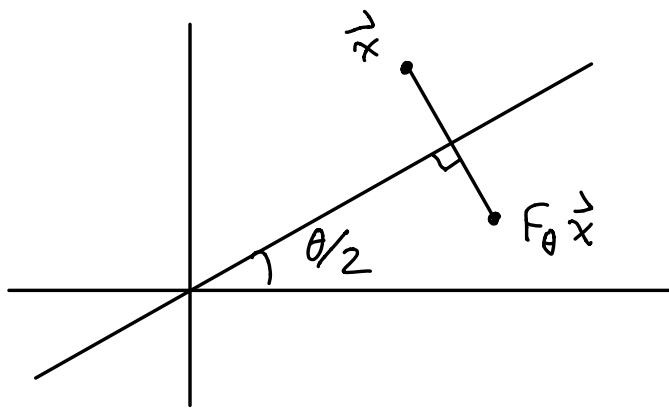


HW5 due Monday
Quiz5 on Tuesday.

Current Topic: Diagonalization.
(eigenvalues & eigenvectors).

Conceptual Example:

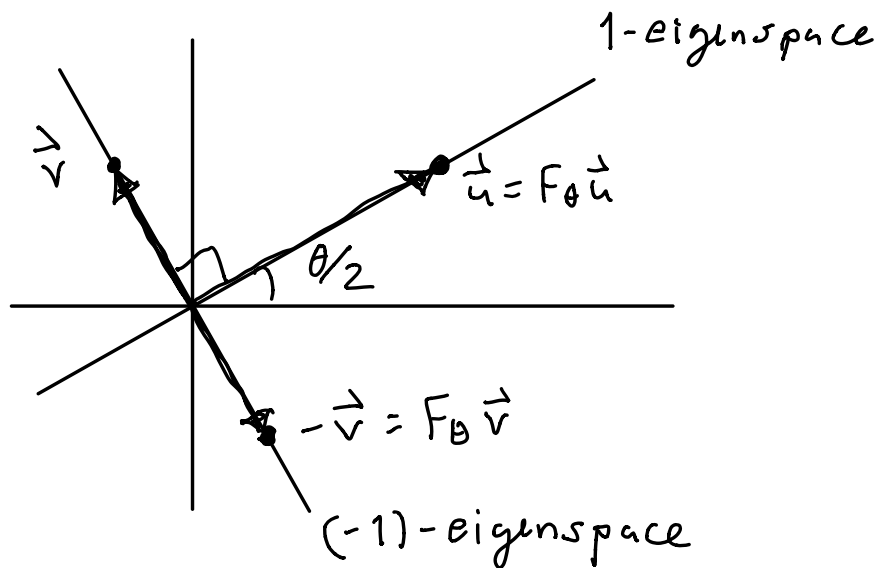
$$F_{\theta} = \begin{pmatrix} \cos \theta & \sin \theta \\ \sin \theta & -\cos \theta \end{pmatrix}.$$



What are the eigenvalues & eigenvectors?

$$F_{\theta} \vec{u} = \lambda \vec{u} \quad \text{and} \quad \vec{u} \neq \vec{0}$$

Example: If \vec{u} is on the line
then $F_{\theta} \vec{u} = \vec{u} = 1\vec{u}$:



Also, for any vector $\vec{v} \perp$ to the line we have $F_\theta \vec{v} = -\vec{v} = (-1)\vec{v}$.
 And this matrix has NO other eigenvectors.

Conclusion:

F_θ has eigenvalues 1 & -1.

1-eigenspace = line $t(\cos \frac{\theta}{2}, \sin \frac{\theta}{2})$

(-1)-eigenspace = line $t(\sin \frac{\theta}{2}, -\cos \frac{\theta}{2})$.

Remark: Sometimes we can learn everything about eigenvalues & eigenvectors

by thinking about the geometry.

But sometimes we need to do calculations.

Example: Find all e. values & vectors of

$$A = \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}.$$

[Recall: λ is an eigenvalue of A
if and only if $\det(A - \lambda I) = 0$.]

$$\det(A - \lambda I) = 0$$

$$\det \left[\begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right] = 0$$

$$\det \begin{pmatrix} -\lambda & 3 \\ 2 & 5 - \lambda \end{pmatrix} = 0$$

$$(-\lambda)(5 - \lambda) - 2 \cdot 3 = 0$$

$$\lambda^2 - 5\lambda - 6 = 0$$

$$(\lambda - 6)(\lambda + 1) = 0 \quad \text{inspection}$$

$$\text{or } \lambda = \frac{5 \pm \sqrt{25 + 24}}{2} = \frac{5 \pm 7}{2} = 6 \text{ or } -1$$

6 - Eigenspace :

$$(A - 6I) \vec{u} = \vec{0}$$

$$\begin{pmatrix} -6 & 3 \\ 2 & -1 \end{pmatrix} \vec{u} = \vec{0}$$

$$\left(\begin{array}{cc|c} -6 & 3 & 0 \\ 2 & -1 & 0 \end{array} \right)$$

$$\left(\begin{array}{cc|c} -6 & 3 & 0 \\ 0 & 0 & 0 \end{array} \right)$$

$$\left(\begin{array}{cc|c} 1 & -\frac{1}{2} & 0 \\ 0 & 0 & 0 \end{array} \right) \quad x - \frac{1}{2}y = 0$$

$$\vec{u} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \frac{1}{2}y \\ y \end{pmatrix} = y \begin{pmatrix} \frac{1}{2} \\ 1 \end{pmatrix}$$

(-1) - Eigenspace :

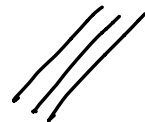
$$(A + 1I) \vec{v} = \vec{0}$$

$$\left(\begin{array}{cc|c} 1 & 3 & 0 \\ 2 & 6 & 0 \end{array} \right)$$

(

$$\begin{pmatrix} 1 & 3 & | & 0 \\ 0 & 0 & | & 0 \end{pmatrix} \quad x + 3y = 0$$

$$\vec{v} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -3y \\ y \end{pmatrix} = y \begin{pmatrix} -3 \\ 1 \end{pmatrix}$$



[Remark : Computations with 3×3 are worse, and in general we need a computer to find e. values & vectors of large square matrices !!]



What can we do with this ?

Consider a linear recurrence equation :

$$\vec{x}_0 = \begin{pmatrix} 4 \\ 1 \end{pmatrix} \quad \& \quad \vec{x}_{n+1} = \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix} \vec{x}_n .$$

Find an explicit formula for the vector \vec{x}_n .

Step 1: Find e. values & e. vectors.

Pick your favorites:

$$\begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 6 \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix} \begin{pmatrix} 3 \\ -1 \end{pmatrix} = -1 \begin{pmatrix} 3 \\ -1 \end{pmatrix} \quad \checkmark$$

Step 2: Express the initial condition as a linear combination of eigenvectors:

$$\begin{pmatrix} 4 \\ 1 \end{pmatrix} = a \begin{pmatrix} 1 \\ 2 \end{pmatrix} + b \begin{pmatrix} 3 \\ -1 \end{pmatrix}$$

By inspection, $a=1$ & $b=1$.

$$\begin{pmatrix} 4 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 3 \\ -1 \end{pmatrix}.$$

Step 3: Solve. IF $\vec{x}_{n+1} = A \vec{x}_n$

$$\begin{aligned} \text{then } \vec{x}_n &= A \vec{x}_{n-1} \\ &= AA \vec{x}_{n-2} \\ &\vdots \end{aligned}$$

$$\begin{aligned} &= AAA \dots A \vec{x}_0 \\ &= A^n \vec{x}_0. \end{aligned}$$

In our case,

$$\begin{aligned} \vec{x}_n &= \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}^n \begin{pmatrix} 4 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}^n \left[\begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 3 \\ -1 \end{pmatrix} \right] \\ &= \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}^n \begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}^n \begin{pmatrix} 3 \\ -1 \end{pmatrix} \\ &= 6^n \begin{pmatrix} 1 \\ 2 \end{pmatrix} + (-1)^n \begin{pmatrix} 3 \\ -1 \end{pmatrix}. \end{aligned}$$

Done



Why did it work?

For any $A\vec{u} = \lambda\vec{u}$ we have

$$\begin{aligned} A^n \vec{u} &= A^{n-1} A \vec{u} \\ &= A^{n-1} (\lambda \vec{u}) \end{aligned}$$

|

$$\begin{aligned}
& \downarrow \\
& = \lambda A^{n-1} \vec{u} \\
& = \lambda \lambda A^{n-2} \vec{u} \\
& = \lambda \lambda \dots \lambda \vec{u} \\
& = \lambda^n \vec{u}. \quad \checkmark
\end{aligned}$$

What about

$$\begin{aligned}
& (A^3 - 2A^2 + A - 5I) \vec{u} \quad ? \\
& = A^3 \vec{u} - 2A^2 \vec{u} + A \vec{u} - 5 \vec{u} \\
& = \lambda^3 \vec{u} - 2\lambda^2 \vec{u} + \lambda \vec{u} - 5 \vec{u} \\
& = (\lambda^3 - 2\lambda + \lambda - 5) \vec{u}.
\end{aligned}$$

More generally, for any polynomial expression

$$f(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_d x^d$$

we can define the matrix

$$f(A) = a_0 I + a_1 A + a_2 A^2 + \dots + a_d A^d.$$

Then for any $A\vec{u} = \lambda\vec{u}$ we find

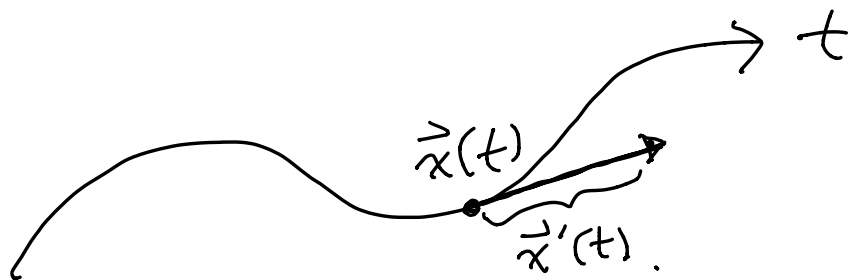
$$\underbrace{f(A)}_{\text{matrix}} \vec{u} = \underbrace{f(\lambda)}_{\text{scalar}} \vec{u} .$$

I will show you an amazing application of this.



Vector Fields.

Let $\vec{x}(t)$ be a parametrized path in \mathbb{R}^n :



The derivative is the instantaneous velocity at time t .

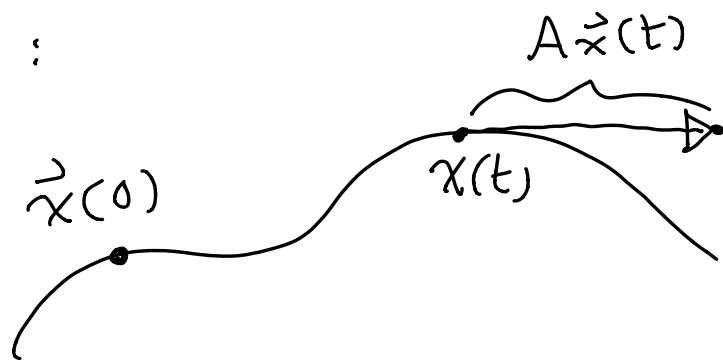
$$\vec{x}(t) = (x_1(t), x_2(t), \dots, x_n(t))$$

$$\vec{x}'(t) = (x_1'(t), x_2'(t), \dots, x_n'(t)).$$

A vector field tells us a vector at every point. A "linear vector field" has the form:

$$\vec{x}'(t) = A \vec{x}(t).$$

Idea:



Our problem: Given a starting point $\vec{x}(0)$, find the unique path $\vec{x}(t)$ such that

$$\vec{x}'(t) = A \vec{x}(t)$$

Such a path is called a flow line of the vector field. The solution to this problem is surprisingly nice.

Theorem: For any square matrix A ,
the "linear system of ordinary
differential equations with constant
coefficients"

$$\vec{x}'(t) = A \vec{x}(t)$$

has the unique solution

$$\vec{x}(t) = e^{At} \vec{x}(0)$$

where e^{At} is the square matrix
defined as follows:

$$\begin{aligned} e^{At} &= I + At + A^2 \frac{t^2}{2} + A^3 \frac{t^3}{6} + \dots \\ &= \sum_{k=0}^{\infty} A^k \frac{t^k}{k!} \end{aligned}$$

It turns out that this series
always converges!



Before we try to apply this, it's easy to see why it works:

$$\begin{aligned}\frac{d}{dt}(e^{At}) &= \frac{d}{dt}\left(\mathbf{I} + At + A^2 \frac{t^2}{2} + A^3 \frac{t^3}{6} + \dots\right) \\ &= 0 + A + A^2 \frac{2t}{2} + A^3 \frac{3t^2}{6} + \dots \\ &= A + A^2 t + A^3 \frac{t^2}{2} + A^4 \frac{t^3}{6} + \dots \\ &= A \left(\mathbf{I} + At + A^2 \frac{t^2}{2} + A^3 \frac{t^3}{6} + \dots\right) \\ &= \underbrace{A}_{\text{matrix}} \underbrace{e^{At}}_{\text{matrix}}.\end{aligned}$$

So we observe that $\vec{x}(t) = e^{At} \vec{x}(0)$ satisfies:

$$\vec{x}(0) = e^0 \vec{x}(0) = \mathbf{I} \vec{x}(0) \quad \checkmark$$

$$\vec{x}'(t) = \frac{d}{dt}(e^{At} \vec{x}(0))$$

$$= A e^{At} \vec{x}(0) = A \vec{x}(t) \quad \checkmark$$

You might worry that the matrix

$$e^{At} = I + At + A^2 \frac{t^2}{2} + \dots$$

is hard to compute. The good news is we don't need to compute it. We just need to know the eigenvalues & eigenvectors, and the following theorem makes this easy.

Theorem: If $A\vec{u} = \lambda\vec{u}$ for some square matrix A , then

$$\underbrace{e^{At}}_{\text{matrix}} \underbrace{\vec{u}}_{\text{vector}} = \underbrace{e^{\lambda t}}_{\text{scalar}} \underbrace{\vec{u}}_{\text{vector}}.$$

That is, the matrix e^{At} has the same eigenvectors as A . Only the eigenvalues change:

$$\lambda \text{ is an e.value of } A \iff e^{\lambda t} \text{ is an e.value of } e^{At}.$$

Let's apply this.
Suppose that functions $x(t)$, $y(t)$
satisfy the following initial conditions
and differential equations:

$$\begin{cases} x(0) = 4 \\ y(0) = 1 \end{cases} \quad \begin{cases} x'(t) = 3y(t) \\ y'(t) = 2x(t) + 5y(t). \end{cases}$$

Find explicit formulas for $x(t)$ & $y(t)$.
I claim that we have already done
all the work!

Define $\vec{x}(t) = (x(t), y(t))$.

so that $\vec{x}'(t) = A \vec{x}(t)$

has the solution

$$\vec{x}(t) = e^{At} \vec{x}(0).$$

In our case:

$$A = \begin{pmatrix} 0 & 3 \\ 2 & 5 \end{pmatrix}$$

has e.values & e.vectors

$$A\vec{u} = 6\vec{u} \quad \& \quad A\vec{v} = -1\vec{v}$$

$$A\begin{pmatrix} 1 \\ 2 \end{pmatrix} = 6\begin{pmatrix} 1 \\ 2 \end{pmatrix} \quad \& \quad A\begin{pmatrix} 3 \\ -1 \end{pmatrix} = -\begin{pmatrix} 3 \\ -1 \end{pmatrix}$$

Express the initial condition in terms of these:

$$\vec{x}(0) = \begin{pmatrix} 4 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 3 \\ -1 \end{pmatrix} = \vec{u} + \vec{v}.$$

Therefore, the solution is:

$$\begin{aligned} \vec{x}(t) &= e^{At} \vec{x}(0) \\ &= e^{At} (\vec{u} + \vec{v}) \\ &= e^{At} \vec{u} + e^{At} \vec{v} \\ &= e^{6t} \vec{u} + e^{-t} \vec{v} \\ &= e^{6t} \begin{pmatrix} 1 \\ 2 \end{pmatrix} + e^{-t} \begin{pmatrix} 3 \\ -1 \end{pmatrix} \end{aligned}$$

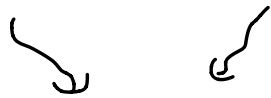
Conclusion:

$$\begin{aligned} x(t) &= e^{6t} + 3e^{-t} \\ y(t) &= 2e^{6t} - e^{-t}. \end{aligned}$$

See how easy that was ?

Well, at least it's no harder than solving linear recurrence equations.

$$\vec{x}_{n+1} = A \vec{x}_n \quad \vec{x}'(t) = A \vec{x}(t)$$



Same method of solution.

Namely,

Step 1: Find all e. values & e. vectors of A:

$$A \vec{u}_1 = \lambda_1 \vec{u}_1, \quad A \vec{u}_2 = \lambda_2 \vec{u}_2, \quad \dots, \quad A \vec{u}_d = \lambda_d \vec{u}_d.$$

Step 2: Express initial conditions in terms of eigenvectors:

$$\vec{x}_0 = a_1 \vec{u}_1 + a_2 \vec{u}_2 + \dots + a_d \vec{u}_d$$

$$\vec{x}(0) = b_1 \vec{u}_1 + b_2 \vec{u}_2 + \dots + b_d \vec{u}_d.$$

Step 3: Solve,

$$\vec{x}_n = A^n \vec{x}_0$$

{

$$\begin{aligned}
&\downarrow \\
&= A^n (a_1 \vec{u}_1 + \dots + a_d \vec{u}_d) \\
&= a_1 A^n \vec{u}_1 + \dots + a_d A^n \vec{u}_d \\
&= a_1 \lambda_1^n \vec{u}_1 + \dots + a_d \lambda_d^n \vec{u}_d
\end{aligned}$$

Done.

$$\begin{aligned}
\vec{x}(t) &= e^{At} \vec{x}(0) \\
&= e^{At} (b_1 \vec{u}_1 + \dots + b_d \vec{u}_d) \\
&= b_1 e^{At} \vec{u}_1 + \dots + b_d e^{At} \vec{u}_d \\
&= b_1 e^{\lambda_1 t} \vec{u}_1 + \dots + b_d e^{\lambda_d t} \vec{u}_d.
\end{aligned}$$

Done.

This is the power of diagonalization. See HWS. 5.