

Recap: • The EIGENVALUES of an $n \times n$ matrix A are solutions λ of $\det(A - \lambda I) = 0$. ← degree n polynomial

• For each eigenvalue λ of A , the collection of EIGENVECTORS is given by the nonzero vectors in $N(A - \lambda I)$.

• If the eigenvalues $\lambda_1, \dots, \lambda_n$ (possibly repeated!) of A have n independent eigenvectors $\vec{v}_1, \dots, \vec{v}_n$, then A is DIAGONALIZABLE.

• That is to say,

$$A = SDS^{-1}$$

where $D = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$ is diagonal,

and $S = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix}$ is invertible.

• Upshot: ability to compute $A^k = SD^k S^{-1}$.

Notes: If A is diagonalizable, then

$$\begin{aligned} A) \quad \det(A) &= \det(SDS^{-1}) = \det(S) \det(D) \frac{1}{\det(S)} \\ &= \det(D) = \underbrace{\lambda_1 \cdots \lambda_n}_{\text{product of eigenvalues!}} \end{aligned}$$

B) But be careful!

(i) Row operations do NOT preserve eigenvalues.

$$\begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix}$$

$$\lambda_1 = 0$$

$$\lambda_2 = 5$$

$$\lambda_1 = 0$$

$$\lambda_2 = 1$$

(ii) So, $\left(\begin{array}{c} \text{product} \\ \text{of} \\ \text{pivots} \end{array} \right) = \det(A) = \left(\begin{array}{c} \text{product} \\ \text{of} \\ \text{eigenvalues} \end{array} \right)$

Even though there may be no direct relation between pivots & eigenvalues!

This is Important

DIFFERENTIAL EQUATIONS (Linear)

Maybe the easiest differential equations in 1-d take on the form $\frac{dx}{dt} = \alpha x$ along with an initial condition,

$$x(0) = x_0.$$

The solution only requires integration:

$$\int_0^t \frac{dx}{x} = \int_0^t \alpha dt,$$

$$\text{so } \ln x(t) - \ln x(0) = \alpha t$$

$$\text{so } \ln x(t) = \alpha t + \ln x_0.$$

ONE DIMENSIONAL

$$\text{So } x(t) = e^{\alpha t + \ln x_0}$$

$$= e^{\alpha t} e^{\ln x_0}$$

$$\Rightarrow x(t) = e^{\alpha t} x_0$$

This solves the diff'l equation

$$\frac{dx}{dt} = \alpha x$$

$$x(0) = x_0$$

COMPLETELY!

(Where $e^y = 1 + y + \frac{y^2}{2!} + \dots = \sum_{j=0}^{\infty} \frac{y^j}{j!}$)

A standard (but silly) model for population growth is given by such an equation:

$$p' = \alpha p$$

rate of change of population at time t

fraction of reproducing "adults"
 $0 \leq \alpha \leq 1$

current population at time t

"The more people there are, the more babies they will make"

If the population at $t=0$ is $p(0)$, then the population at time t is given by

$$p(t) = e^{\alpha t} p(0) \leftarrow \text{BAM! exponential growth.}$$

A slightly more reasonable model of growth takes into account the natural resources that are available to the population at any given time.

Letting $r(t)$ be the resources at time t , we make the following assumptions:

$p'(t)$ is proportional to $p(t)$ AND $r(t)$ (+vely)
 $r'(t)$ is proportional to $p(t)$ (-vely)

The first assumption says we need people AND resources to make more people. The second says that the more people we've got, the faster the resources will deplete. So:

Two linear differential equations "coupled" together.

$$\begin{cases} p'(t) = \alpha p(t) + \beta r(t) \\ r'(t) = -\gamma p(t) \end{cases} \quad \text{for } \alpha, \beta, \gamma \text{ positive constants.}$$

In matrix form: $\frac{d}{dt} \begin{bmatrix} p(t) \\ r(t) \end{bmatrix} = \begin{bmatrix} \alpha & \beta \\ -\gamma & 0 \end{bmatrix} \begin{bmatrix} p(t) \\ r(t) \end{bmatrix}$

Def A system $\vec{x}' = A\vec{x}$ where A is $n \times n$ is called a LINEAR DIFFERENTIAL SYSTEM (of dimension n).

When $n=1$, we are back to $x' = \alpha x$ and know the solution given any initial $x(0) = x_0$. So:

Q How to solve the n -dim'l system given an initial vector $\vec{x}(0)$?

Note

If $\vec{x}' = D\vec{x}$ where D is DIAGONAL,
say $D = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$ then there is
no trouble: the n -dimensional system just
decomposes into n one-dimensional linear
differential equations:

$$\left\{ \begin{array}{l} x_1'(t) = \lambda_1 x_1(t) \\ x_2'(t) = \lambda_2 x_2(t) \\ \vdots \\ x_n'(t) = \lambda_n x_n(t) \end{array} \right.$$

which are solved
by using

$$x_j(t) = e^{\lambda_j t} x_j(0)$$

But wait:
this means

$$\vec{x}(t) = \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} \vec{x}(0)$$

(This looks a lot like $x(t) = e^{xt} x(0)$, the 1D solution)

Def

The EXPONENTIAL e^A of an $n \times n$ matrix A is
defined by the series

$$e^A = \text{Id} + A + \frac{A^2}{2!} + \dots$$
$$= \sum_{j=0}^{\infty} \frac{A^j}{j!}$$

When

$A = D = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$, then we just have

$$e^D = \begin{bmatrix} e^{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n} \end{bmatrix}$$

If A is NOT diagonal, then e^A looks like a MESS to compute. BUT this is what eigenvalues and eigenvectors are good for! IF A is diagonalizable, then $A = SDS^{-1}$, and

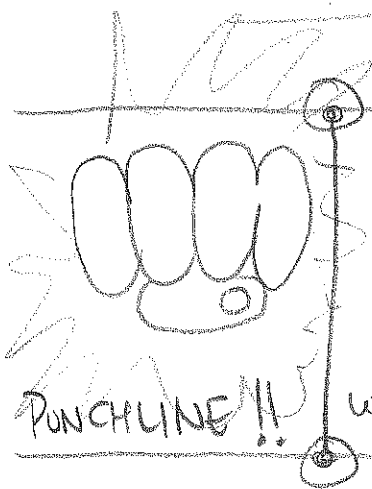
$$e^A = \sum_{j=0}^{\infty} \frac{A^j}{j!} = \sum_{j=0}^{\infty} \frac{(SDS^{-1})^j}{j!}$$

$$= \sum_{j=0}^{\infty} SD^j S^{-1} \frac{1}{j!} = S \left(\sum_{j=0}^{\infty} \frac{D^j}{j!} \right) S^{-1}$$

$$= \frac{S e^D S^{-1}}$$

So, $e^A = S \begin{bmatrix} e^{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n} \end{bmatrix} S^{-1}$

where λ 's are eigenvalues of A , and S is the corresponding eigenvector matrix.



The n -dimensional system $\vec{x}' = A\vec{x}$ with initial conditions $\vec{x}(0)$ is solved by

$$\vec{x}(t) = S \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} S^{-1} \vec{x}(0) \quad \left(\begin{smallmatrix} \text{eigen} \\ \text{vals} \end{smallmatrix} \right)$$

PUNCHLINE!! whenever $A = SDS^{-1}$ is diagonalizable; $D = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$

Okay, enough theory: Example Time!

Eg Solve the 2D linear differential system

$$\vec{x}' = \begin{bmatrix} 7 & 5 \\ -10 & -8 \end{bmatrix} \vec{x},$$

with initial conditions $\vec{x}(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$.

STEP 1 Compute eigenvalues of $A = \begin{bmatrix} 7 & 5 \\ -10 & -8 \end{bmatrix}$

$$A - \lambda I = \begin{bmatrix} (7-\lambda) & 5 \\ -10 & (-8-\lambda) \end{bmatrix}$$

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

$$(7-\lambda)(-8-\lambda) + 50 = 0$$

$$\Rightarrow \lambda^2 + \lambda - 6 = 0 = (\lambda - 2)(\lambda + 3) = 0$$

So, $\lambda_1 = 2$ and $\lambda_2 = -3$

STEP 2 Compute eigenvectors for each λ .

$$\lambda_1 = 2, \quad A - 2I = \begin{bmatrix} 5 & 5 \\ -10 & -10 \end{bmatrix}, \quad \vec{v}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_2 = -3, \quad A + 3I = \begin{bmatrix} 10 & 5 \\ -10 & -5 \end{bmatrix}, \quad \vec{v}_2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

STEP 3 Write $D = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$, $S = [\vec{v}_1 \ \vec{v}_2]$ and compute S^{-1} .

$$D = \begin{bmatrix} 2 & 0 \\ 0 & -3 \end{bmatrix}, \quad S = \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}, \quad S^{-1} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

STEP 4

$$\text{Ans} = Se^{Dt}S^{-1}\vec{x}(0)$$
$$= \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{-3t} \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \dots$$