

Math 2270-1
Tuesday 11/29

4.7.6 Stability

Example $\vec{x}(t+1) = A \vec{x}(t)$

if $A_{n \times n}$ diagonalizable (at least over \mathbb{C})

then $B = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ is a complex eigenbasis, evals $\lambda_1, \lambda_2, \dots, \lambda_n$

$$\vec{x}_0 = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n$$

$$A^t \vec{x}_0 = c_1 \lambda_1^t \vec{v}_1 + c_2 \lambda_2^t \vec{v}_2 + \dots + c_n \lambda_n^t \vec{v}_n$$

$$\lambda_k = r_k e^{i\theta_k} \quad r_k = |\lambda_k|, \quad \theta_k = \text{polar angle}$$

$$\lambda_k^t = r_k^t e^{it\theta_k}$$

Deduce: if each $|\lambda_k| < 1$ then $A^t \vec{x}_0 \rightarrow \vec{0}$ "geometrically" as $t \rightarrow \infty$.

if at least one $|\lambda_k| > 1$ then $\exists \vec{x}_0$ with $\|\vec{x}_0\|$ arbitrarily small so that $\|A^t \vec{x}_0\| \rightarrow \infty$ as $t \rightarrow \infty$.

Def $\vec{0}$ is an asymptotically stable equilibrium of a discrete dynamical system

$$\vec{x}(t+1) = A \vec{x}(t)$$

iff $\lim_{t \rightarrow \infty} \vec{x}(t) = \vec{0}$ for all \vec{x}_0 .

[think of the Glucose-Insulin model, in which $\vec{x} = \begin{bmatrix} G(t) \\ H(t) \end{bmatrix}$ measured deviations from equilibrium].

if this is
all you need to
know, the test is easy

Theorem: For the discrete dynamical system

$$\vec{x}(t+1) = A \vec{x}(t)$$

$\vec{0}$ is an asymptotically stable equilibrium iff all evals of A satisfy $|z| < 1$.

Pf: Well, we just did the proof in case of diagonalizable A .

The general proof relies on the Jordan canonical form of A ,

in case A is not diagonalizable. (We will talk about Jordan form in 2280)

Examples :

① Since the only $A_{2 \times 2}$ for which we don't know the stability theorem is the case of equal eigenvalues, let's focus there:

$$\begin{aligned} |A - \lambda I| &= \lambda^2 - \text{trace}(A)\lambda + |A| \\ &= (\lambda - \frac{1}{2} \text{trace}(A))^2 \quad (\text{discrim}=0) \\ &= (\lambda - \lambda_1)^2 \end{aligned}$$

Case 1 $\dim E_{\lambda_1} = 2$: this is the diagonalizable case A similar to $\lambda_1 I$ ($\Rightarrow A = \lambda_1 I$)

Case 2 $\dim E_{\lambda_1} = 1$

Let $E_{\lambda_1} = \text{span}\{\vec{u}\}$.

Let $B = \{\vec{u}, \vec{v}\}$ be a basis for \mathbb{R}^2

$$L(\vec{x}) = A\vec{x}$$

$$\begin{aligned} B = [L]_B &= \begin{bmatrix} \lambda_1 & b_1 \\ 0 & b_2 \end{bmatrix} \quad \text{but } |B - \lambda I| \text{ also } = (\lambda - \lambda_1)^2 \\ &\Rightarrow b_2 = \lambda_1 \\ &= \begin{bmatrix} \lambda_1 & b_1 \\ 0 & \lambda_1 \end{bmatrix}, b_1 \neq 0 \text{ since } \dim(E_{\lambda_1}) = 1 \end{aligned}$$

$$\Rightarrow L(\vec{v}) = b_1 \vec{u} + \lambda_1 \vec{v}$$

$$\Rightarrow L(\frac{1}{b_1} \vec{v}) = \vec{u} + \lambda_1 \left(\frac{\vec{v}}{b_1}\right)$$

Let $C = \{\vec{u}, \frac{1}{b_1} \vec{v}\}$

$$C = [L]_C = \begin{bmatrix} \lambda_1 & 1 \\ 0 & \lambda_1 \end{bmatrix} \quad \text{this is known as a}$$

2×2 Jordan block.

[is composition of scaling & shear].

$$C = S^{-1}AS$$

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

$$C = \lambda_1 I + N$$

$$N = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$A^t = SC^{t^{-1}}S^{-1}$$

$$C^n = \lambda_1^n I +$$

$$N^2 = 0$$

$$\begin{aligned} n\lambda_1^{n-1}N \\ + 0 + 0 + 0 \end{aligned}$$

(binomial expansion)

[matrices for which a power of the matrix is zero are called nilpotent].

$$A^t = S \left[\lambda_1^t I + n\lambda_1^{t-1}N \right] S^{-1}$$

$$\rightarrow 0 \text{ if } |\lambda_1| < 1$$



② The general case of complex roots for real $A \in \mathbb{R}^{2 \times 2}$ [with glucose-insulin was typical]

$$A\vec{w} = \lambda \vec{w} \quad \lambda = p+qi \quad (\text{assume } b \neq 0)$$

$$A\vec{w} = \bar{\lambda} \vec{w} \quad \bar{\lambda} = p-qi$$

$$\vec{w} = \vec{u} + i\vec{v}$$

$$A\vec{w} = (p+qi)(\vec{u} + i\vec{v})$$

$$A\vec{u} + iA\vec{v} \quad (p\vec{u} - q\vec{v}) + i(p\vec{v} + q\vec{u}) ; \quad \text{so } A\vec{u} = p\vec{u} - q\vec{v}$$

$$A\vec{v} = p\vec{v} + q\vec{u}$$

for $B = \{v, u\}$

$$L(x) = Ax$$

$$B = [L]_B = \begin{bmatrix} p & -q \\ q & p \end{bmatrix} \quad \text{rotation-dilatation}$$

$$= r \begin{bmatrix} \cos\phi & -\sin\phi \\ \sin\phi & \cos\phi \end{bmatrix} ; \quad r = \sqrt{p^2 + q^2}$$

$$B = S^{-1}AS$$

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

$$A^t = S B^t S^{-1}$$

$$= r^t S \begin{bmatrix} \cos(t\phi) & -\sin(t\phi) \\ \sin(t\phi) & \cos(t\phi) \end{bmatrix} S^{-1} \quad t \in \mathbb{Z}$$

$$A^t \vec{x}_0 = r^t S \begin{bmatrix} \cos(t\phi) & -\sin(t\phi) \\ \sin(t\phi) & \cos(t\phi) \end{bmatrix} S^{-1} \vec{x}_0$$

$\underbrace{\hspace{10em}}_{\text{points on a circle of radius } \|S^{-1}\vec{x}_0\|}$
 $\underbrace{\hspace{10em}}_{\text{converts circle to an ellipse}}$

scales points to spiral

outward if $r > 1$

inward if $r < 1$

stays ellipse if $r = 1$

$$r = \sqrt{p^2 + q^2} = 1.21.$$

Let's revisit glucose-insulin from this point of view:

$$A = \begin{bmatrix} .9 & -.4 \\ -.1 & .9 \end{bmatrix}$$

$$\lambda = p + iq$$

$$\vec{w} = \begin{bmatrix} -2 \\ i \end{bmatrix} = \begin{bmatrix} -2 \\ 0 \end{bmatrix} + i \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$B = \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \end{bmatrix} \right\}$$

$$B = [L]_B = \begin{bmatrix} .9 & -.2 \\ .2 & .9 \end{bmatrix} = \sqrt{.85} \begin{bmatrix} \cos\phi & -\sin\phi \\ \sin\phi & \cos\phi \end{bmatrix} \quad \phi = \arctan(\frac{2}{9})$$

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

$$A t \vec{x}_0 = r^t \begin{bmatrix} 0 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \cos(t\phi) & -\sin(t\phi) \\ \sin(t\phi) & \cos(t\phi) \end{bmatrix} \underbrace{\begin{bmatrix} 0 & 1 \\ -\frac{1}{2} & 0 \end{bmatrix}}_{\begin{bmatrix} 0 \\ -\sin\phi \end{bmatrix}} \begin{bmatrix} 100 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 50 \sin(t\phi) \\ -50 \cos(t\phi) \end{bmatrix}$$

on circle $x^2 + y^2 = 50^2$

$$\begin{bmatrix} 100 \cos(t\phi) \\ 50 \sin(t\phi) \end{bmatrix}$$

on ellipse $\frac{x^2}{100^2} + \frac{y^2}{30^2} = 1$.

$$(.85)^{t_2} \begin{bmatrix} 100 \cos(t\phi) \\ 50 \sin(t\phi) \end{bmatrix}$$

Spiral

Math 2270-1
November 29, 2005

Phase portraits, eigenvectors and eigenvalues, stability

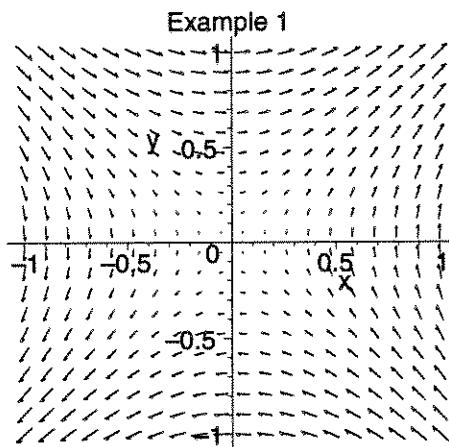
```

> with(plots):with(linalg):
> A:=matrix(2,2,[1,2,2,1]):
eigenvects(A);

$$A := \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$$

[3, 1, {[1, 1]}], [-1, 1, {[ -1, 1]}]
> pict1:=fieldplot([0*x+2*y,2*x+0*y],x=-1..1,y=-1..1):
display({pict1},title='Example 1');

```



```

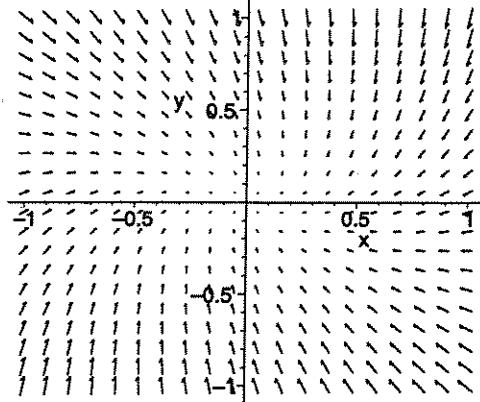
> B:=matrix(2,2,[-1,1,-1,-3]):
eigenvects(B);

$$B := \begin{bmatrix} -1 & 1 \\ -1 & -3 \end{bmatrix}$$

[-2, 2, {[1, -1]}]
> pict2:=fieldplot([-2*x+y,-1*x-4*y],x=-1..1,y=-1..1):
display({pict2},title='Example 2');

```

Example 2

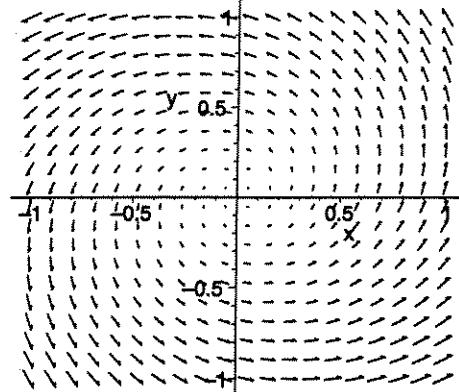


```
> C:=matrix(2,2,[2,-3,3,2]);
  eigenvecs(C);

$$C := \begin{bmatrix} 2 & -3 \\ 3 & 2 \end{bmatrix}$$

[2+3*I, 1, {[1, -I]}], [2-3*I, 1, {[1, I]}]
> pict3:=fieldplot([x-3*y,3*x+y],x=-1..1,y=-1..1):
  display({pict3},title='Example 3');
```

Example 3



>