

Nonlinear Systems

(Chap. 9)

Systems of the form

$$\begin{cases} \frac{dx}{dt} = f(x, y) \\ \frac{dy}{dt} = g(x, y) \end{cases} \quad (2\text{-d system})$$

Where f & g contain terms like x^2 , xy , y^3 , etc. (nonlinear terms).

CAN'T write as matrix.

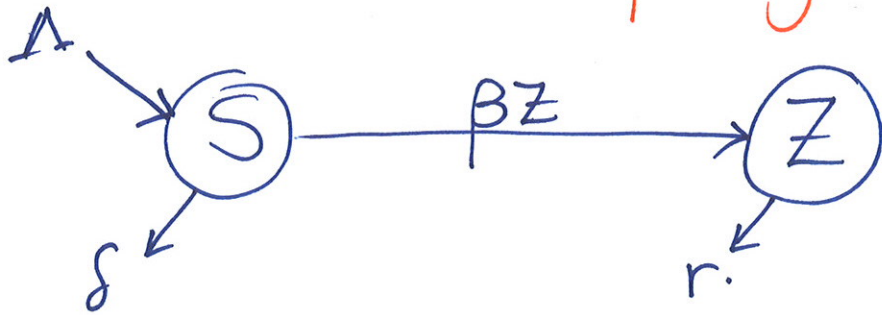
Sometimes write as

$$\vec{x}' = f(\vec{x})$$

Note: This is an AUTONOMOUS system $\Rightarrow f, g$ don't depend on t .

Motivating Example

→ Zombie plague.



S = susceptibles
Z = zombies.

(Based on epidemic models
→ Kermack & McKendrick, 1920s)

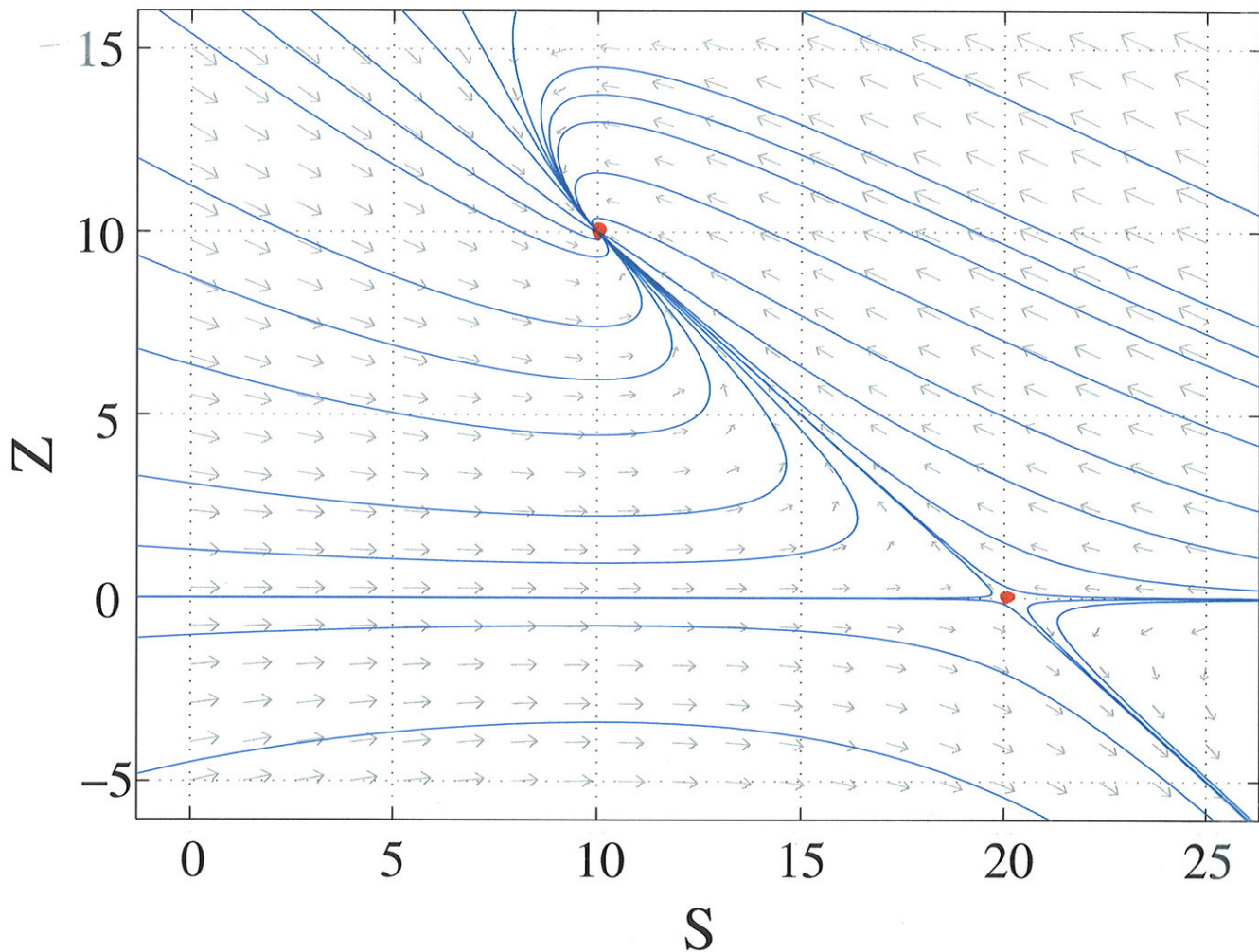
$$\begin{cases} \frac{dS}{dt} = \Lambda - \beta SZ - \delta S \\ \frac{dZ}{dt} = \beta SZ - rZ \end{cases}$$

→ nonlinear term.

Interested in solution trajectories → phase plane.

$$\left(\begin{array}{l} \Lambda = 20 \text{ ppl/day, } r = 1 \text{ per day,} \\ \delta = 1 \text{ per day, } \beta = 0.1 \text{ per contact per day} \end{array} \right)$$

Phase portrait:



Each trajectory (blue line) is a solution.

Notice:

- multiple critical points $\vec{x}_0 \neq 0$ s.t.

$$f(\vec{x}_0) = 0 \quad (\text{in red})$$

- $S = 20, Z = 0$ \Leftarrow uninfected eq.
- $S = 10, Z = 10$ \Leftarrow endemic eq.

More careful definition of stability.

Given system $\vec{x}' = f(\vec{x})$ with solution $\vec{x}(t) = \vec{\phi}(t)$.

- A critical point \vec{x}_0 s.t. $f(\vec{x}_0) = 0$ is **STABLE** if, given $\varepsilon > 0$, there is a $\delta > 0$ s.t. every solution $\vec{\phi}(t)$ ^{that} satisfies at $t=0$

$$\|\vec{\phi}(0) - \vec{x}_0\| < \delta \quad \leftarrow \text{starts "close enough"}$$

exists + satisfies for all $t \geq 0$

$$\|\vec{\phi}(t) - \vec{x}_0\| < \varepsilon.$$

↪ solution bounded, stays around \vec{x}_0 .

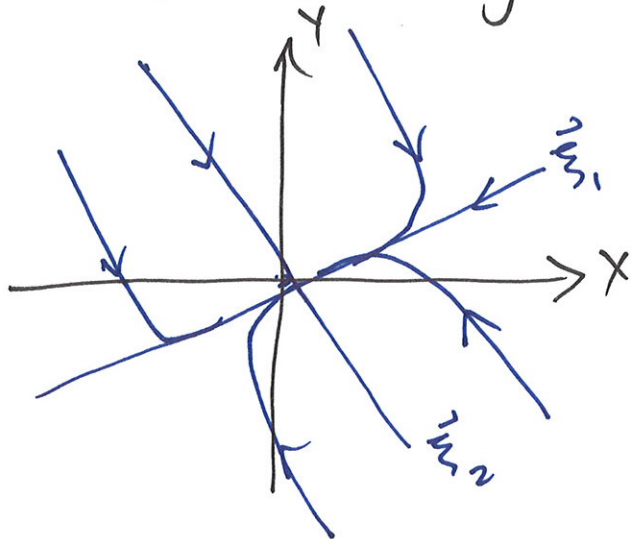
A critical point \vec{x}_0 is **asymptotically stable** if it is stable AND there ~~is~~ exists some $\delta_0 > 0$ s.t. $\vec{x}(t) = \vec{\phi}(t)$ satisfies $\|\vec{\phi}(0) - \vec{x}_0\| < \delta_0$

Then $\lim_{t \rightarrow \infty} \vec{\phi}(t) = \vec{x}_0$.

"starts close enough"

(~~our~~ def'n we had been using).

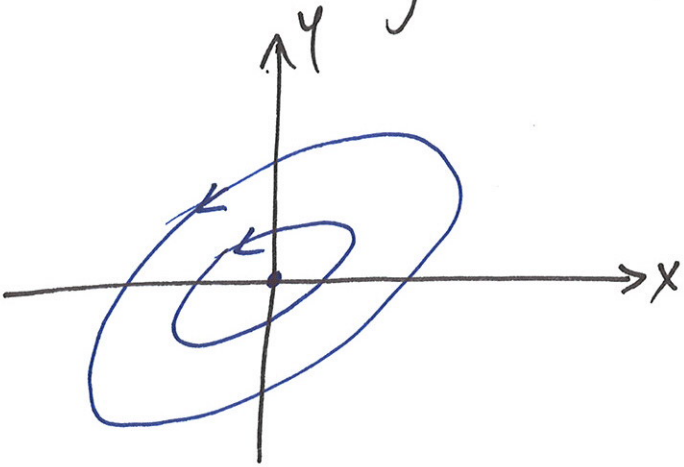
Eg] 2 negative, real eigenvalues
 (2×2 system) ($r_1 \neq r_2$)



$$\vec{x} = C_1 \vec{z}_1 e^{r_1 t} + C_2 \vec{z}_2 e^{r_2 t}$$

here the origin is **asymptotically stable**.

Eg 2 purely imaginary e-val.
(2×2 system)



The origin is
stable
but not
asymptotically
stable.

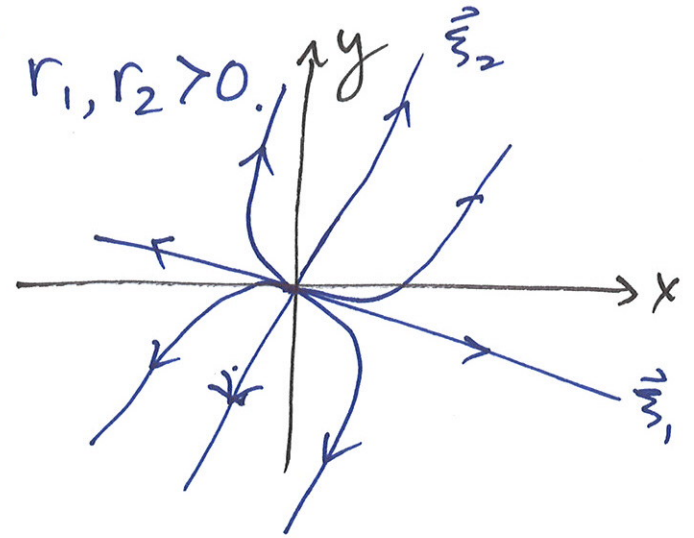
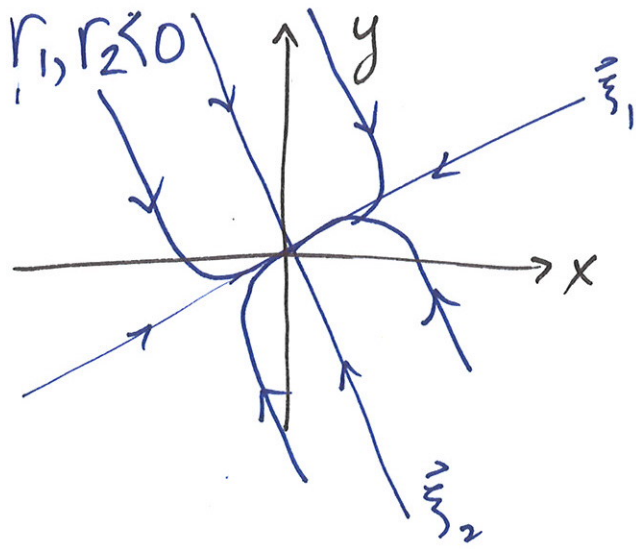
Phase portraits of linear systems \rightarrow review

(Classify critical points)

CASE I: eigenvalues real, same sign.

Sol. $\vec{x} = C_1 \vec{\xi}_1 e^{r_1 t} + C_2 \vec{\xi}_2 e^{r_2 t}$

$\left. \begin{array}{l} \uparrow \\ \leftarrow \end{array} \right\} \begin{array}{l} \text{e-val} \\ \text{e-vect.} \end{array}$

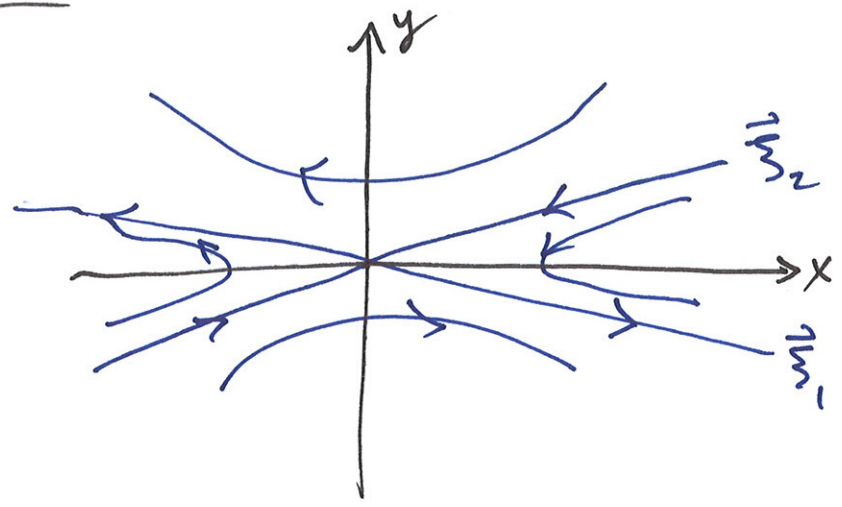


- stable, improper node.
- book \rightarrow "nodal sink"
- asymptotically stable.

- unstable, improper node
- book \rightarrow "nodal source".
- unstable.

Case II: eigenvalues real + opposite sign ($r_1 > 0, r_2 < 0$).

Sol: $\vec{X} = C_1 \vec{v}_1 e^{r_1 t} + C_2 \vec{v}_2 e^{r_2 t}$.



- saddle point
- unstable.

- NEAR critical points, trajectories "look like" linear systems.

trajectories.

$S = 20, Z = 0 \rightarrow$ saddle point
(1 +ve, 1 -ve).

$S = 10, Z = 10 \rightarrow$ repeated e-val, < 0 ,
with 1 linearly indep.
e-vect.

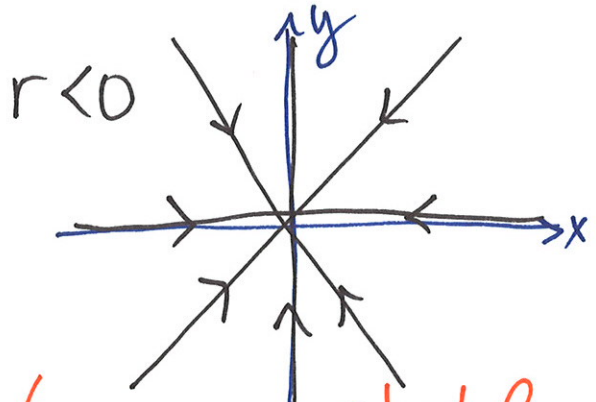
- ~~the~~ In understanding systems, stability is key.

Therefore, our work in linear systems is a tool for understanding solution trajectories, critical point stability in nonlinear systems.

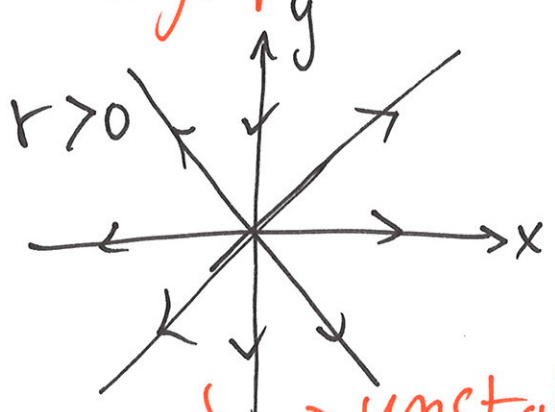
Case III: repeated real e-val. $r_1 = r_2 = r$.

(i) ^{lin.} 2 indep e-veets

$$\vec{x} = C_1 \vec{\xi}_1 e^{rt} + C_2 \vec{\xi}_2 e^{rt}$$



↳ asymp. stable



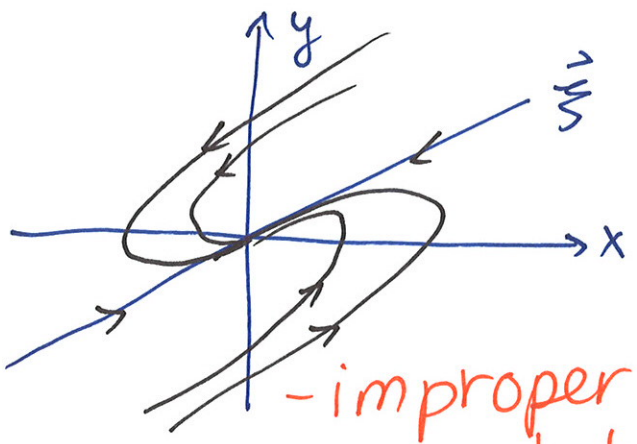
↳ unstable

"star"
- proper node.

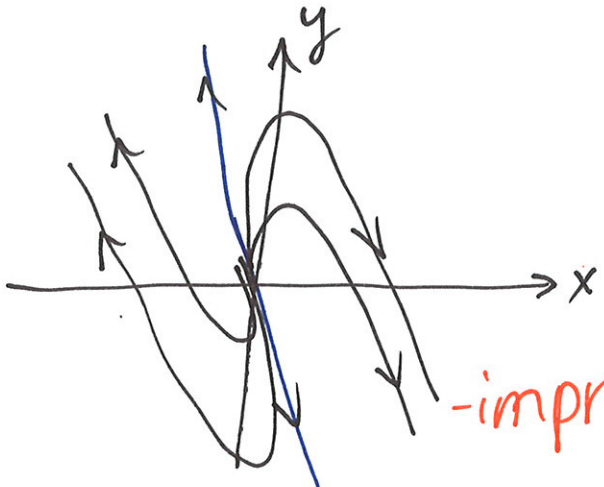
-20%

(ii) only 1 lin. indep. e-veet.

$$\vec{x} = C_1 \vec{\xi} e^{rt} + C_2 (\vec{\xi}t + \vec{\eta}) e^{rt}$$



- improper node
- asymptotically stable.



- improper node
- unstable.