

29 January - MATH 345

January 29, 2015 1:55 PM

- HW 2 due February 5 (next Thursday)

- 2nd-order equations (even non-linear) can be expressed as systems of 1st-order equations

Example 2.1

mass-spring system with viscous damping



$$\underbrace{m \frac{d^2x}{dt^2}}_{\text{inertia term}} = -\underbrace{kx}_{\text{spring restoring force}} - \underbrace{b \frac{dx}{dt}}_{\text{viscous damping}}$$

Hooke's "law"

define $y = \frac{dx}{dt}$

$$\frac{dx}{dt} = y \quad (\text{by definition!})$$

$$\frac{dy}{dt} = \frac{d^2x}{dt^2} = -\frac{k}{m}x - \frac{b}{m} \left(\frac{dx}{dt} \right)$$

get a system:

$$\begin{cases} \frac{dx}{dt} = y \\ \frac{dy}{dt} = -\frac{k}{m}x - \frac{b}{m}y \end{cases}$$

OR

$$\frac{d\vec{x}}{dt} = A\vec{x}, \quad \vec{x} = \begin{pmatrix} x \\ y \end{pmatrix}, \quad A = \begin{pmatrix} 0 & 1 \\ -\frac{k}{m} & -\frac{b}{m} \end{pmatrix}$$

- in general, consider

$$\dot{\vec{x}} = A\vec{x}, \quad \vec{x} \in \mathbb{R}^2$$

- fixed points:

$$\vec{0} = A\vec{x}$$

- $\vec{x}^* = \vec{0}$ is a fixed point (there could be others)

Classification of 2-dim linear systems

- are there solutions of the form $\vec{x}(t) = e^{\lambda t} \vec{v}$?

$$\begin{aligned}\dot{\vec{x}}(t) &= A\vec{x}(t) \\ \lambda e^{\lambda t} \vec{v} &= A e^{\lambda t} \vec{v}\end{aligned}\quad \begin{matrix} \text{plug in } \vec{x} = e^{\lambda t} \vec{v} \\ \text{& take derivative}\end{matrix}$$

$$\lambda \vec{v} = A \vec{v}$$

- if λ is an eigenvalue and \vec{v} is an eigenvector, then

$$\vec{x}(t) = e^{\lambda t} \vec{v} \text{ is a solution}$$

- if $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$, then the characteristic equation
(to find eigenvalues) is

$$(A - \lambda I)\vec{v} = \vec{0}, \vec{v} \neq 0$$

identity matrix

matrix must be singular (i.e. determinant is zero)

$$0 = \det(A - \lambda I) = \begin{vmatrix} a-\lambda & b \\ c & d-\lambda \end{vmatrix} = \lambda^2 - (\text{tr} A)\lambda + (\det A)$$

$$\text{tr} A = \text{trace}(A) = T$$

$$\det A = \det(A) = \Delta$$

$$\lambda^2 - T\lambda + \Delta = 0$$

- roots of this are eigenvalues

$$\lambda_{1,2} = \frac{T \pm \sqrt{T^2 - 4\Delta}}{2}$$

Exercise show that $\lambda_1 + \lambda_2 = T$
 $\lambda_1 \lambda_2 = \Delta$

- diagonalised matrix is $\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$
- trace & determinant are invariant
- eigenvalues determine the stability
- trace & determinant of A determine stability

recall: if eigenvalues are real and distinct, then
the general solution is

$$\vec{x}(t) = c_1 e^{\lambda_1 t} \vec{v}_1 + c_2 e^{\lambda_2 t} \vec{v}_2$$

Exercise show that if $\Delta < 0$ then both eigenvalues
are real and have opposite signs

Example 2.2

$$\dot{\vec{x}} = \begin{pmatrix} 1 & 1 \\ 4 & -2 \end{pmatrix} \vec{x}$$

$$\Delta = (1)(-2) - (1)(4) = -6 < 0$$

so λ_1, λ_2 will be real and have opposite signs

$$\tau = 1 - 2 = -1$$

$$\lambda^2 - (-1)\lambda - 6 = 0$$

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

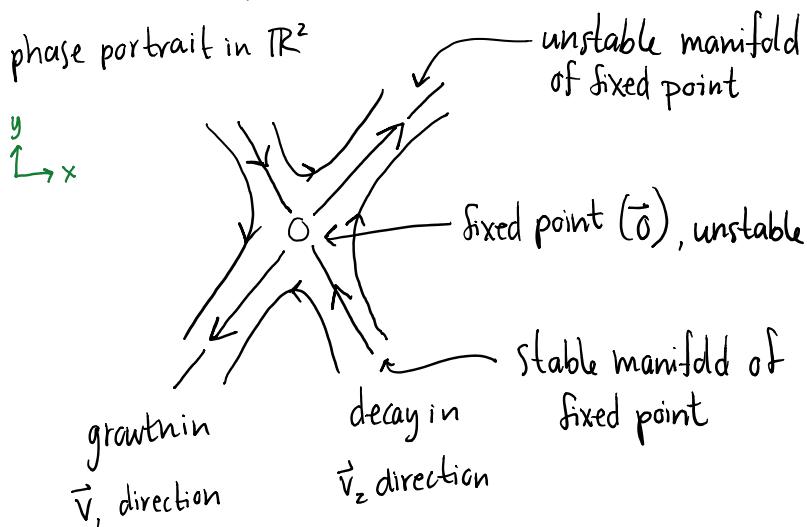
$$(\lambda - 2)(\lambda + 3) = 0$$

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

Exercise find eigenvectors \vec{v}_1, \vec{v}_2 for λ_1, λ_2

general solution:

$$\vec{x}(t) = \underbrace{c_1 e^{2t} \vec{v}_1}_{\text{growth}} + \underbrace{c_2 e^{-3t} \vec{v}_2}_{\text{decay}}$$

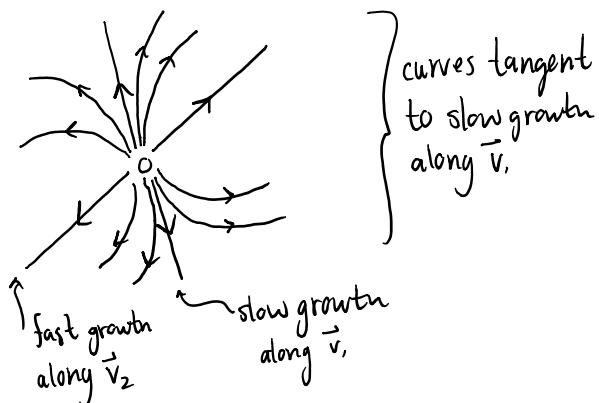


. this type of fixed point is called a saddle point and it is unstable

- . If $\tau^2 - 4\Delta > 0$ and $\Delta > 0$ then λ_1, λ_2 are real, distinct and both have the same sign as τ . In this case, the fixed point \vec{o} is called a node

Exercise Show i) $\tau > 0$ is stable
ii) $\tau < 0$ is unstable

e.g. $0 < \lambda_1 < \lambda_2$



Exercise Carefully plot phase portrait of

$$\vec{x} = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \vec{x}$$

. if $\tau^2 - 4\Delta < 0$, then the eigenvalues are complex

$$\lambda_1, \lambda_2 = \frac{\tau \pm i\sqrt{4\Delta - \tau^2}}{2} = \alpha + i\omega$$

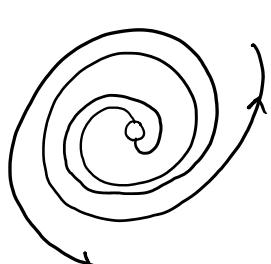
$$\alpha = \operatorname{Re}(\lambda_1) = \tau/2$$

$$\omega = \operatorname{Im}(\lambda_1) = \frac{\sqrt{4\Delta - \tau^2}}{2}$$

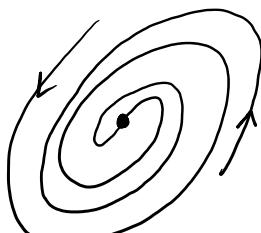
$$e^{\lambda_1 t} = e^{(\alpha + i\omega)t} = e^{\alpha t} e^{i\omega t} = e^{\alpha t} (\cos \omega t + i \sin \omega t)$$

- components of (real) general solution $\vec{x}(t)$ are linear combinations of $e^{\alpha t} \cos \omega t$ and $e^{\alpha t} \sin \omega t$
- α has the same sign as τ
- if $\tau \neq 0$ then the fixed point $\vec{0}$ in this case is called a spiral
 - stable if $\tau < 0$
 - unstable if $\tau > 0$

e.g.



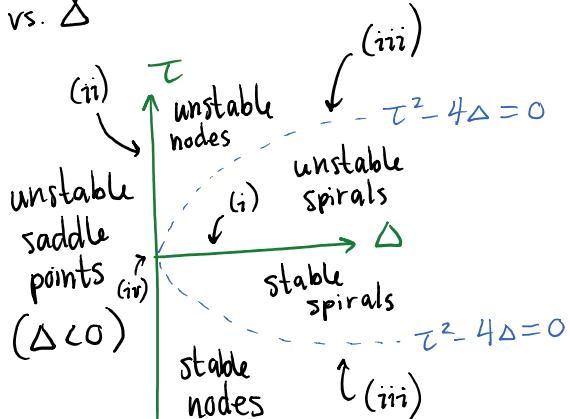
$\tau > 0$



$\tau < 0$

- classification:

τ vs. Δ

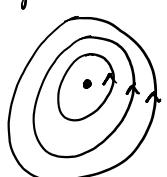


- "borderline" cases

(i) $\tau = 0, \Delta > 0$

purely imaginary eigenvalues $\pm i\omega$

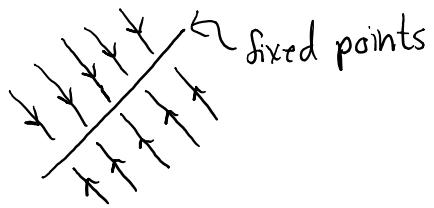
"linear centre"



(ii) $\Delta = 0, \tau \neq 0$

simple zero eigenvalue

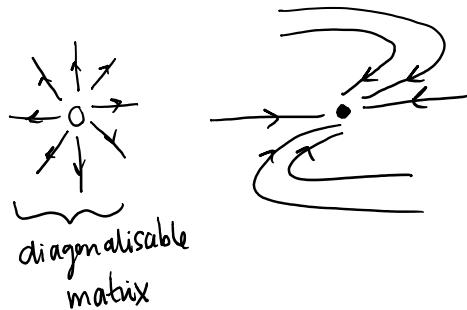
"non-isolated fixed points"



$$(iii) \tau^2 - 4\Delta = 0, \tau \neq 0$$

double non-zero eigenvalue $\lambda_{1,2} = \tau/2$

"stars, degenerate nodes"



$$(iv) \tau=0, \Delta=0$$

double zero eigenvalue

Stability

$$\dot{\vec{x}} = \vec{f}(\vec{x}), \vec{x} \in \mathbb{R}^2, \text{ i.e. } \begin{cases} \dot{x}_1 = f_1(x_1, x_2) \\ \dot{x}_2 = f_2(x_1, x_2) \end{cases} \text{ OR } \begin{cases} \dot{x} = f(x, y) \\ \dot{y} = g(x, y) \end{cases}$$

• fixed point \vec{x}^* is a solution of $\vec{0} = \vec{f}(\vec{x})$

• Lyapunov stable if for all $\vec{x}_0 = \vec{x}(0)$ sufficiently close to \vec{x}^* ,

$\vec{x}(t)$ remains arbitrarily close to \vec{x}^* for all $t \geq 0$

• Unstable if not Lyapunov stable (equivalent to definition from Jan. 8 for 1D)

• Attracting if for all \vec{x}_0 sufficiently close to \vec{x}^* , $\vec{x}(t) \rightarrow \vec{x}^*$ as $t \rightarrow \infty$

• Stable (or asymptotically stable) if Lyapunov stable and attracting

• Neutrally stable if Lyapunov stable but not attracting