

Motivation: The pendulum

Pendulum equation:

$$(1) \quad \ddot{x} + \omega^2 \sin x = 0.$$

No solution in terms of elementary functions.

Phase plane analysis:

① **Equivalent system** ($y = \dot{x}$):

$$(2) \quad \begin{cases} \dot{x} = y, \\ \dot{y} = -\omega^2 \sin x. \end{cases}$$

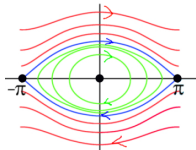
② Implicit solution = **phase trajectory** (here C is **energy**):

$$\frac{dy}{dx} = \frac{\dot{y}}{\dot{x}} = -\frac{\omega^2 \sin x}{y} \implies \frac{1}{2}y^2 - \omega^2 \cos x = C = \text{const.} \quad \text{i.e. a curve for each } C.$$

③ **Equilibrium points** $(x_e, y_e) = \text{constant solutions of (2)}$:

$$\Leftrightarrow (y_e, -\omega^2 \cos x_e) = (0, 0) \Leftrightarrow (x_e, y_e) = (n\pi, 0), \quad n \in \mathbb{Z}.$$

④ **Phase diagram:** All phase trajectories (periodic in x , symmetric about $y = 0$)



⑤ Interpretation: **Red** = rotations, **green** = swinging back and forth.

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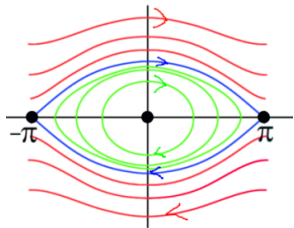
Stability:

A solution is **stable** if all solutions starting “near” remain “near” the solution for all times.

Stability for the pendulum:

Since solutions follow phase trajectories in the direction of the arrows, we see from the phase portrait that:

- 1 The **equilibrium points** $(2m\pi, 0)$ are **stable**.
- 2 The **equilibrium points** $((2m + 1)\pi, 0)$ are **unstable**.
- 3 The **separatrices** (blue curve) are **unstable**.
- 4 For **all other solutions** more powerful methods are needed!



Due to small disturbances, physical systems tend over time to be (near) their stable equilibrium solutions (not necessarily equilibrium points).

Linear 2×2 systems of ODEs

Initial value problem:

$$(1) \quad \begin{cases} \frac{dx_1}{dt} = a_{11}x_1 + a_{12}x_2 \\ \frac{dx_2}{dt} = a_{21}x_1 + a_{22}x_2, \end{cases} \quad \text{or} \quad \frac{d\vec{x}}{dt} = A\vec{x} \quad \text{or} \quad \frac{dx}{dt} = Ax,$$

$$(2) \quad \vec{x}(0) = \vec{x}_0.$$

where $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \text{constant}$ and $\vec{x} = x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$.

Autonomous equation (does not depend on t) in **normal form** (like $\dot{x} = f(x, t)$).

Results:

- Theorem 1:** There is only one solution of (1) and (2) for all $t \in \mathbb{R}$.
- Theorem 2:** x_1 and x_2 solve (1) $\Rightarrow c_1\vec{x}_1 + c_2\vec{x}_2$ solve (1) for all $c_1, c_2 \in \mathbb{R}$.
- Theorem 3:** x_1 and x_2 solve (1) and are linear independent
 \Rightarrow any solution of (1) can be written as $c_1\vec{x}_1 + c_2\vec{x}_2$ for some $c_1, c_2 \in \mathbb{R}$.

In Tm. 3 x_1, x_2 is a **basis**, and $\vec{x} = c_1\vec{x}_1 + c_2\vec{x}_2$ a **general solution** of (1).

Solving equation (1)

$$(1) \quad \frac{d\vec{x}}{dt} = A\vec{x} \quad \text{where} \quad A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \text{constant}.$$

Idea: Find a basis of two independent solutions.

Test solution: $\vec{x} = \vec{v}e^{\lambda t}$, $\vec{v} \neq \vec{0}$.

- 1 Solves (2) iff $(A - \lambda I)\vec{v} = \vec{0}$... Eigenvalue problem.
- 2 Non-zero solutions \vec{v} iff characteristic equation holds

$$\det(A - \lambda I) = \lambda^2 - p\lambda + q = 0, \quad \text{where} \quad \begin{cases} p = \text{tr } A = a_{11} + a_{22}, \\ q = \det A = a_{11}a_{22} - a_{12}a_{21}. \end{cases}$$

- 3 Hence always solutions (λ_1, \vec{v}_1) , (λ_2, \vec{v}_2) , but complex solutions possible.

Solving linear 2x2 systems

$$(1) \quad \dot{x} = Ax \quad \text{or} \quad \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Test solution: $x = ve^{\lambda t}$, $v \neq 0$.

- ① Solves (1) iff $(A - \lambda I)v = 0 \dots$ **Eigenvalue problem** \dots solutions iff:

$$\det(A - \lambda I) = \lambda^2 - p\lambda + q = 0, \quad q = a_{11} + a_{22}, \quad q = a_{11}a_{22} - a_{12}a_{21}.$$

- ② Hence always solutions (λ_1, v_1) , (λ_2, v_2) . All possible cases:

- ① $\lambda_1 \neq \lambda_2$ **real**: $v_1, v_2 \in \mathbb{R}^2$ are linearly independent.
- ② $\lambda_1 = \lambda_2$ **real**: $v_1, v_2 \in \mathbb{R}^2$ may or may not be lin. independent.
- ③ $\lambda_1 = \bar{\lambda}_2$ **complex**: $v_1 = \bar{v}_2 \in \mathbb{C}^2$ are linearly independent.

General solution: $x = C_1x_1 + C_2x_2$, $C_1, C_2 \in \mathbb{R}$, where

- ① λ_1, λ_2 **real**, v_1, v_2 lin. independent: $x_1(t) = v_1 e^{\lambda_1 t}$ | $x_2(t) = v_2 e^{\lambda_2 t}$

- ② $\lambda_1 = \lambda_2$ **real**, only one lin. indep. v : $x_1(t) = v e^{\lambda_1 t}$ and

$$x_2 = (vt + u)e^{\lambda_1 t} \quad \text{where} \quad (A - \lambda_1 I)u = v.$$

- ③ $\lambda_1 = \bar{\lambda}_2 = \alpha - i\beta$ **complex**: $x_1(t) = \text{Re}(v_1 e^{\lambda_1 t})$ | $x_2(t) = \text{Im}(v_1 e^{\lambda_1 t})$

Linear 2x2 systems

$$(1) \quad \dot{x} = Ax \quad \text{or} \quad \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

- ① **Thm:** All solutions of (1) can be written in the form

$$x(t) = C_1 x_1(t) + C_2 x_2(t),$$

where x_1 and x_2 are defined as above in each case.

- ② **Thm:** All solutions of (1) are linear combinations of products of trigonometric, exponential, and polynomial functions.

- ③ **Thm:** The initial value problem (1) and

$$(2) \quad x(0) = x_0,$$

has a solution for all $x_0 \in \mathbb{R}^2$ and all $t \in \mathbb{R}$.

Fundamental matrix and Stability

$$(1) \quad \dot{x}(t) = Ax(t),$$

$$(2) \quad x(0) = x_0.$$

where $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ is constant and $x \in \mathbb{R}^2$.

Concepts:

- ① **Fundamental matrix:** $\Phi(t) = [x_1(t), x_2(t)]$ where x_1, x_2 basis for (1)
- ② **Flow ϕ :** $x(t) = \phi(t; x_0)$ solve (1) and (2).
- ③ **Lemma:** $\phi(t; x_0) = \Phi(t)\Phi(0)^{-1}x_0$.

Stability: Let $x(t) = \phi(t; x_0)$.

- ① $x(t), t > 0$, **stable** if for all $\varepsilon > 0$ there is $\delta > 0$ such that
$$|x_1 - x_0| < \delta \quad \Rightarrow \quad |\phi(t; x_1, t_0) - \phi(t; x_0, t_0)| < \varepsilon \quad \text{for all } t > 0.$$
- ② $x(t), t > 0$, **asymptotic stable** if stable and there is $\eta > 0$ s.t.
$$|x_1 - x_0| < \eta \quad \Rightarrow \quad |\phi(t; x_1, t_0) - \phi(t; x_0, t_0)| \rightarrow 0 \quad \text{as } t \rightarrow \infty.$$

Obs: The definitions hold for general non-linear $n \times n$ systems!

Stability of solutions of linear systems

$$(1) \quad \dot{x}(t) = Ax(t),$$

$$(2) \quad x(0) = x_0.$$

where $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ is constant and $x \in \mathbb{R}^2$.

Theorem 1: Let Φ be any fundamental matrix of (1).

- 1 $\|\Phi(t)\| \leq M < \infty$ for all $t \geq 0 \Rightarrow$ all solutions of (1) are **stable**.
- 2 $\lim_{t \rightarrow 0} \|\Phi(t)\| = 0 \Rightarrow$ all solutions of (1) are **asymptot. stable**.
- 3 $\lim_{t \rightarrow 0} \|\Phi(t)\| = \infty \Rightarrow$ all solutions of (1) are **unstable**.

Theorem 2: Let λ_1, λ_2 be the eigenvalues of A .

- 1 all solutions of (1) are **stable** $\Rightarrow \max_i \operatorname{Re}(\lambda_i) \leq 0$.
- 2 $\max_i \operatorname{Re}(\lambda_i) \leq 0$ and $\lambda_1 \neq \lambda_2 \Rightarrow$ all solutions of (1) are **stable**.
- 3 $\max_i \operatorname{Re}(\lambda_i) < 0 \Rightarrow$ all solutions of (1) are **asymptot. stable**.
- 4 $\max_i \operatorname{Re}(\lambda_i) \geq 0 \Rightarrow$ all solutions of (1) are **unstable**.

Autonomous linear 2x2 systems

$$(1) \quad \dot{x} = Ax; \quad A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$

Equilibrium points:

- Constant solutions x_e of (1), i.e. solutions of $Ax_e = 0$.
- $\lambda_1, \lambda_2 \in \mathbb{R}$: **Node** if $\lambda_1 \lambda_2 > 0$, and **saddle** if $\lambda_1 \lambda_2 < 0$.
- $\lambda_1 = \bar{\lambda}_2 \in \mathbb{C}$: **Spiral** if $\operatorname{Re} \lambda_1 \neq 0$, and **center** if $\operatorname{Re} \lambda_1 = 0$.
- $\lambda_1 = \lambda_2$ or either $\lambda_1 = 0$ or $\lambda_2 = 0$: **Degenerate cases**.

Phase diagrams/portraits:

- ① **Phase trajectory** through x_0 : $\{x(t) : t \in \mathbb{R}, x \text{ solves (1), } x(0) = x_0\}$
Tangent at x : $Ax [= \dot{x}]$.

Direction: Direction of x as time increases = tangent direction

- Direction in one pt. \rightarrow all directions by **continuity of directions**

Equation: $\frac{dx_2}{dx_1} = \frac{\dot{x}_2}{\dot{x}_1} = \frac{a_{21}x_1 + a_{22}x_2}{a_{11}x_1 + a_{12}x_2}$

Equilibrium point: Trajectory = one point

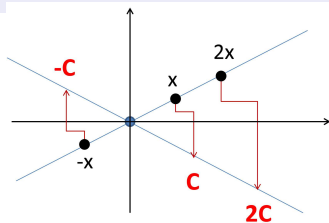
- ② **Phase plane/diagram:** x_1x_2 -plane/sketch of “all” phase trajectories.

Remark: One trajectory through every $x_0 \in \mathbb{R}^2$; trajectories do not cross!

Scaling and symmetry for linear systems

Scaling: $s \in \mathbb{R}, \neq 0$

- 1 Point x : sx
- 2 Curve C : $sC = \{sx : x \in C\}$



Observe:

- 1 $\dot{x} = Ax \iff \dot{y} = Ay$ (scale invariant)
 $y=sx$
- 2 C part of trajectory through $x_0 \Rightarrow sC$ part of trajectory through sx_0

Consequences:

- 1 In any sector from 0 :
 - 1 One trajectory determines the whole diagram by scaling.
 - 2 Equal transit time for all trajectory-segments in sector (check it!)
- 2 Phase diagram symmetric about 0 .
- 3 No isolated closed curves.

Read for yourself Jordan-Smith chp. 2.6

Linear non-autonomous nxn systems

$$(1) \begin{cases} \dot{x} = A(t)x + b(t), \\ x(t_0) = x_0, \end{cases} \quad A = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix}, \quad b = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

We always assume: A, b continuous for all $t \in \mathbb{R}$.

Results and definitions:

- 1 Non-homogeneous, non-autonomous equation in normal form.
- 2 There exists a unique solution of (1) for all $x_0 \in \mathbb{R}^n$ and $t \in \mathbb{R}$.
- 3 Flow $\phi(t; x_0, t_0)$: The solution of eq'n (1)₁ and initial data (1)₂.

Homogeneous equation $b \equiv 0$:

- 1 **Basis:** n linearly independent solutions x_1, \dots, x_n of $\dot{x} = Ax$
- 2 **Fundamental matrix** $\Phi \in \mathbb{R}^{n \times n}$, a for all $t \in \mathbb{R}$ invertible solution of $\dot{\Phi} = A\Phi$.
- 3 **OBS:** $\Phi = [x_1, \dots, x_n]$ and $b \equiv 0 \Rightarrow \phi(t; x_0, t_0) = \Phi(t)\Phi(t_0)^{-1}x_0$
- 4 There always exists a fundamental matrix (and a basis) for (1).

Linear $n \times n$ systems

Non-autonomous equation $\dot{x} = A(t)x + b(t)$; $A \in \mathbb{R}^{n \times n}$.

$x(t) = \Phi(t)\Phi^{-1}(t_0)x(t_0) + \Phi(t) \int_{t_0}^t \Phi^{-1}(s)b(s)ds$ if Φ fund. matrix.

Autonomous equation $\dot{x} = Ax$; A constant, eigenval's/-vec's λ_i/r_i .

① $x = re^{\lambda t}$ solve (1) iff $(A - \lambda I)r = 0$.

Basis when n lin. independent r_i : $x_1 = r_1 e^{\lambda_1 t}, \dots, x_n = r_n e^{\lambda_n t}$.

② Basis in general: There are polynomial p_i (order $\leq n$) such that

$$x_1(t) = p_1(t)e^{\lambda_1 t}, \dots, x_n(t) = p_n(t)e^{\lambda_n t}.$$

③ Real basis: $\operatorname{Re} x_i, \operatorname{Im} x_i, i = 1, \dots, n$ (give n lin. indep. solutions).

Jordan form and exponential

Jordan form: $A \in \mathbb{C}^{n \times n}$

- 1 Eigenvalues/-vectors λ_i/r_i .
- 2 **Complex diagonalization:** When n lin. independent r_i 's,

$$A = P\Lambda P^{-1} \quad \text{where} \quad \Lambda = \text{diag}(\lambda_i) \in \mathbb{C}^{n \times n}, \quad P = [r_1 \dots r_n].$$

- 3 **Jordan form:** For any **real** matrix A ! There is a P s.t.

$$A = P^{-1}JP \quad \text{where} \quad J = \text{diag}(B_1, \dots, B_m) \in \mathbb{R}^{n \times n},$$

$$B_i = \lambda_i, \quad \begin{pmatrix} \text{Re } \lambda_i & \text{Im } \lambda_i \\ -\text{Im } \lambda_i & \text{Re } \lambda_i \end{pmatrix} =: D_i, \quad \begin{bmatrix} \lambda_i & 1 & & 0 \\ & \ddots & \ddots & \\ & & \lambda_i & 1 \\ 0 & & & \lambda_i \end{bmatrix}, \quad \text{or} \quad \begin{bmatrix} D_i & I & & 0 \\ & \ddots & \ddots & \\ & & D_i & I \\ 0 & & & D_i \end{bmatrix}.$$

Matrix exponential: $A \in \mathbb{C}^{n \times n}$

- 1 $e^A := \lim_{n \rightarrow \infty} \left(I + A + \frac{1}{2}A^2 + \dots + \frac{1}{n!}A^n \right)$.
- 2 $\|e^A\| \leq e^{\|A\|}$ and the e^A is well-defined.
- 3 $\Phi(t) = e^{tA}$ solve the following initial value problem

$$\dot{\Phi} = A\Phi \quad \text{and} \quad \Phi(0) = I.$$

Matrix exponential and stability

$$(1) \quad \dot{x} = A(t)x + b(t); \quad A \in \mathbb{R}^{n \times n}; \quad b, x \in \mathbb{R}^n.$$

Matrix exponential:

① $A = PJP^{-1} \Rightarrow e^A = Pe^J P^{-1}$

② e^{tA} fundamental matrix; the solution of (1) when $A = \text{konst}$ is

$$x(t) = e^{(t-t_0)A}x(t_0) + \int_{t_0}^t e^{(t-s)A}b(s)ds.$$

Stability:

① **Theorem:** Φ fundamental matrix, $\dot{\Phi} = A(t)\Phi$.

(a) Φ bounded, $t \geq t_0 \Rightarrow$ all solutions of (1) **stable**

(b) $\Phi \rightarrow 0$ as $t \rightarrow \infty \Rightarrow$ all solutions of (1) **asympt. stable**

(c) Φ unbounded, $t \geq t_0 \Rightarrow$ all solutions of (1) **unstable**

② **Theorem:** $A = \text{konst}$, eigenvalues λ_j .

(a) $\max_i \operatorname{Re} \lambda_i \leq 0$ and $\lambda_i \neq \lambda_j \Rightarrow$ all solutions of (1) **stable**

(b) $\max_i \operatorname{Re} \lambda_i < 0 \Rightarrow$ all solutions of (1) **asympt. stable**

(c) $\max_i \operatorname{Re} \lambda_i > 0 \Rightarrow$ all solutions of (1) **unstable**

Stability when A is non-constant

$$(1) \quad \dot{x} = A(t)x + b(t); \quad A \in \mathbb{R}^{n \times n}; \quad b, x \in \mathbb{R}^n.$$

① **Assume:** $A(t) = B + C(t)$ where B constant, $\int_{t_0}^{\infty} \|C(s)\| ds < \infty$.

② **Theorem:** Φ fundamental matrix, $\dot{\Phi} = B\Phi$.

(a) Φ bounded, $t \geq t_0 \Rightarrow$ all solutions of (1) **stable**

(b) $\Phi \rightarrow 0$ as $t \rightarrow \infty \Rightarrow$ all solutions of (1) **asympt. stable**

③ **Theorem:** B has eigenvalues λ_j .

(a) $\max_j \operatorname{Re} \lambda_j \leq 0$ and $\lambda_i \neq \lambda_j \Rightarrow$ all solutions of (1) **stable**

(b) $\max_j \operatorname{Re} \lambda_j < 0 \Rightarrow$ all solutions of (1) **asympt. stable**

Lipschitz condition

Definitions: Let $f = (f_1, \dots, f_n) \in \mathbb{R}^n$.

- ① $f(x)$ is Lipschitz in a set $\Omega \subset \mathbb{R}^n$ if there is $L > 0$ such that

$$|f(x) - f(y)| \leq L|x - y| \quad \text{for all } x, y \in \Omega.$$

- ② $f(x, t)$ is x -Lipschitz in a set $R \subset \mathbb{R}^{n+1}$ if there is $L > 0$ such that

$$|f(x, t) - f(y, t)| \leq L|x - y| \quad \text{for all } (x, t), (y, t) \in R.$$

- ③ $f(x, t)$ is locally Lipschitz on Ω if they are Lipschitz on each closed bounded subset $K \subset \Omega$ (or – on each closed bounded ball $\bar{B} \subset \Omega$).

- ④ Locally x -Lipschitz defined similarly.

- ⑤ Lipschitz constant: The smallest L in the above inequalities.

- ⑥ $[\sqrt{x_1^2 + x_3^2}, 5, \frac{x_1^2}{1+x_1^2+x_3^2}]$ Lip. in \mathbb{R}^3 ; x^2 only loc. Lip./ \sqrt{x} not in \mathbb{R}

Lipschitz condition

- ① $f(x, t)$ is x -Lipschitz in a set $R \subset \mathbb{R}^{n+1}$ if there is $L > 0$ such that

$$|f(x, t) - f(y, t)| \leq L|x - y| \quad \text{for all } (x, t), (y, t) \in R.$$

- ② $f(x)$ is locally x -Lipschitz in R if it is x -Lipschitz on each closed bounded subset $K \subset \Omega$ (or – on each closed ball $\bar{B} \subset \Omega$).

③ Jacobi matrix: $Df(x, t) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \cdots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}.$

- ④ f is C^1 in bounded, convex $R \Rightarrow f$ is x -Lip. in R and $L \leq \max_R \|Df\|.$

Uniqueness and continuous dependence

$$(1) \quad \dot{x}(t) = f(x(t), t),$$

$$(2) \quad x(t_0) = x_0.$$

Uniqueness:

If f is continuous and **locally x -Lipschitz** in a domain R , then there are not more than one solution of (1) and (2) in R .

Continuous dependence on initial data:

Assume

- 1 f is continuous and x -Lipschitz in a domain R ,
- 2 x, y solutions of (1) for $t \in (t_0, b)$,
- 3 $(x(t), t), (y(t), t) \in R$ for $t \in [t_0, b)$,
- 4 $\lim_{t \rightarrow t_0} x(t) = x_0$ and $\lim_{t \rightarrow t_0} y(t) = y_0$.

Then

$$|x(t) - y(t)| \leq e^{L(t-t_0)} |x_0 - y_0| \quad \text{for } t \in [t_0, b).$$

Remark: $x(t) = \phi(t; x_0)$ is continuous in x_0 .

Continuous dependence on data

Theorem:

Assume

- 1 $\dot{x} = f(x, t)$ and $\dot{y} = g(y, t)$ for $t \in (t_0, b)$,
- 2 $(x(t), t), (y(t), t) \subset R$ for $t \in [t_0, b)$ (R domain),
- 3 f, g is continuous and f x -Lipschitz in R ,
- 4 $|f(x, t) - g(x, t)| \leq \varepsilon$ ($\varepsilon > 0$) for all $(x, t) \in R$,
- 5 $\lim_{t \rightarrow t_0} x(t) = x_0$ and $\lim_{t \rightarrow t_0} y(t) = y_0$.

Then

$$|x(t) - y(t)| \leq e^{2L(t-t_0)} |x_0 - y_0| + \frac{\varepsilon}{L} \sqrt{e^{2L(t-t_0)} - 1}, \quad t \in [t_0, b),$$

where L is the Lipschitz constant of f , i.e. $L \leq \max_R \|Df(x, t)\|$.

Remark: The solution $x(t) = \phi(t; x_0, t_0, f)$ of

$$\dot{x} = f(x, t) \quad \text{and} \quad x(t_0) = x_0$$

is continuous in x_0 and f (and $t_0!$).

Existence of solutions:

$$(1) \quad \dot{x}(t) = f(x(t), t),$$

$$(2) \quad x(t_0) = x_0.$$

Theorem 1: (“Global” existence)

If f is **continuous and x -Lipschitz** for $x \in \mathbb{R}^n$ and $|t - t_0| \leq T$, then there **exists** a solution of (1) and (2) for $|t - t_0| \leq T$.

Theorem 2: (Local existence 1 – Piccard-Lindelöf)

If f is **continuous + x -Lipschitz** for $|x - x_0| \leq K$, $|t - t_0| \leq T$, then there **exists** a sol'n of (1) + (2) for $|t - t_0| \leq \min(T, \frac{K}{M})$, $M = \max_{\substack{|t - t_0| \leq T \\ |x - x_0| \leq K}} |f(x, t)|$

Theorem 3: (Local existence 2 – Peano)

If f is **continuous** for $|x - x_0| \leq K$ and $|t - t_0| \leq T$, then there **exists** a solution of (1) and (2) for $|t - t_0| \leq \min(T, \frac{K}{M})$ where M as in Thm. 2.

Remark: In Thm's 1 and 2, the solution is unique – but not in Thm 3!

Existence interval and C^1 dependence on x_0 .

- (1) $\dot{x}(t) = f(x(t), t),$
(2) $x(t_0) = x_0.$

Theorem 1: f continuous and x -Lipschitz on domain $R \ni (x_0, t_0)$.

(a) There exists a unique solution of (1) and (2) on a maximal existence interval $(t_0 - a, t_0 + b)$ for some $a, b > 0$.

(b) If $b < \infty$, then either $(x(t), t) \rightarrow \partial R$ or $|x(t)| \rightarrow \infty$ as $t \rightarrow b^-$.

Remark:

- 1 3 possibilities: $b = \infty$, or x blows up in finite time, or $(x(t), t)$ leaves in finite time the region R of well-posedness.
- 2 $b =$ existence/life time of the solution of (1) and (2).

Theorem 2: $f \in C^1$ and $\phi(t; x_0)$ (unique) solution of (1) and (2).

Then ϕ is C^1 in x_0 (and t), and $w = \frac{\partial \phi}{\partial x_j}$ is the unique solution of

$$w_t = Df(\phi(t; x_0), t)w, \quad w(0) = e_j.$$

Phasediagram for autonomous systems

- (1) $\dot{x} = f(x), \quad f \in \mathbb{R}^n$ independent of t
(2) $x(t_0) = x_0.$

- ① **Phase trajectory** through x_0 : $\{x(t) : t \in \mathbb{R}, x \text{ solves (1), } x(0) = x_0\}$
Tangent (direction) at x : $f(x) [= \dot{x}]$.

- f continuous \Rightarrow **continuity of directions**

$$\text{Equation: } \frac{dx_2}{dx_1} = \frac{\dot{x}_2}{\dot{x}_1} = \frac{f_2(x)}{f_1(x)}, \dots, \frac{dx_n}{dx_1} = \frac{\dot{x}_n}{\dot{x}_1} = \frac{f_n(x)}{f_1(x)}$$

Equilibrium point x_e : $f(x_e) = 0$, and trajectory = one point

- ② **Phase diagram**: Sketch of “all” phase trajectories.
③ **Well-posedness** for (1) and (2) for all x_0, t
 \Rightarrow trajectories exist + do not cross + pass through every $x_0 \in \mathbb{R}^n$.
④ **Separatrix**: Trajectory separating regions w. different sol'n behaviour
⑤ **Only in non-linear systems**: Many isolated equilibrium points, limit cycles, separatrix cycles, and chaos (in $\mathbb{R}^n, n \geq 3$).

Linearization

$$(1) \quad \dot{x} = f(x).$$

Linearized about equi. pt. x_0 : $f(x) = f(x_0) + Df(x_0)(x - x_0) + \dots$

$$(2) \quad \dot{y} = Df(x_0)y \quad (x - x_0 = y).$$

- ① Near x_0 , $f(x_0) \neq 0$: Phase diagram \approx straight lines (if $f \in C^1$).
- ② Near x_0 , $f(x_0) = 0$ and hyperbolic: $\operatorname{Re} \lambda_i \neq 0$ for eig.val's of $Df(x_0)$
 - ① Phasediagram (1) looks like phasediagram (2) near $y = 0$
 - ② (1) and (2) has same types of equilibrium points in x_0 and $y = 0$ (including asymptotic lines etc.)
 - ③ Justification: Hartman-Grobman + Hartman thm's (need $f \in C^2$!).
- ③ Near x_0 , $f(x_0) = 0$ and non-hyperbolic:
No information from linearization – need higher order theory.

Linearization

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No information from linearization – need higher order theory.

Reminder: Stability

$$(1) \quad \dot{x}(t) = f(x(t), t),$$

$$(2) \quad x(t_0) = x_0.$$

Flow ϕ : $x(t) = \phi(t; x_0, t_0)$ solution of (1) and (2).

Stability: Let $x(t) = \phi(t; x_0, t_0)$.

- ① $x(t)$ **stable** for $t \geq t_0$ if for all $\varepsilon > 0$ there is $\delta > 0$ such that

$$|x_1 - x_0| < \delta \quad \Rightarrow \quad |\phi(t; x_1, t_0) - \phi(t; x_0, t_0)| < \varepsilon \quad \text{for all } t > t_0.$$

- ② $x(t)$ **asymptotic stable** for $t \geq t_0$ if stable and there is $\eta > 0$ s.t.

$$|x_1 - x_0| < \eta \quad \Rightarrow \quad |\phi(t; x_1, t_0) - \phi(t; x_0, t_0)| \rightarrow 0 \quad \text{as } t \rightarrow \infty.$$

Stability of equilibrium points

Nearly linear systems:

$$(1) \quad \dot{x} = Ax + h(x, t); \quad A \in \mathbb{R}^{n \times n}, \quad x, h \in \mathbb{R}^n.$$

Theorem 1:

Assume λ_i eigenvalues of A , $h \in C^1$, and

$$|h(x, t)| = o(|x|) \quad \text{as } x \rightarrow 0 \quad \text{uniformly in } t.$$

$\max_i \operatorname{Re} \lambda_i < 0 \Rightarrow 0$ is an **asympt. stable** equilibrium pt. of (1).

Autonomous systems: Linearization

$$(2) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^n.$$

Theorem 2:

Assume $f \in C^1$, $f(x_0) = 0$, and λ_i eigenvalues of $Df(x_0)$.

(a) $\max_i \operatorname{Re} \lambda_i < 0 \Rightarrow x_0$ is an **asympt. stable** equilibrium pt. of (2)

(b) $\max_i \operatorname{Re} \lambda_i > 0 \Rightarrow x_0$ is an **unstable** equilibrium pt. of (2)

(c) $\max_i \operatorname{Re} \lambda_i = 0 \Rightarrow$ NO CONCLUSION!!

OBS: Stability based on Df = stability by linearization $\dot{y} = Df(x_0)y$

Liapunov's direct method

- More general method than linearization.
- Need to construct (find) a Liapunov function.

$$(2) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^n.$$

$V(x)$ is a **strong Liapunov function** for (2) in domain $B \ni 0$ if

- 1 $V \in C^1(B)$,
- 2 $V(0) = 0$ and $V(x) > 0$ for $x \in B \setminus \{0\}$,
- 3 $\dot{V}(0) = 0$ and $\dot{V}(x) < 0$ for $x \in B \setminus \{0\}$, where

$$\dot{V}(x) = \frac{d}{dt} V(x(t)) = \nabla V(x) \cdot \dot{x} = \nabla V(x) \cdot f(x).$$

Theorem 2: Assume $f(0) = 0$ and f Lipschitz in some domain $B \ni 0$.

If $V(x)$ is a **strong Liapunov function** for (2) in B ,
then $x = 0$ is an **asymptotically stable** equilibrium point for (2).

Liapunov's direct method

$$(1) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^n.$$

Weak (strong) Liapunov function $V(x)$ for (1) in domain $B \ni 0$:

- 1 $V \in C^1(B)$,
- 2 $V(0) = 0$ and $V(x) > 0$ for $x \in B \setminus \{0\}$,
- 3 $\dot{V}(0) = 0$ and $\dot{V}(x) \leq 0$ ($V(x) < 0$) for $x \in B \setminus \{0\}$, where
- 4 $\dot{V}(x) = \frac{d}{dt} V(x(t)) = \nabla V(x) \cdot \dot{x} = \nabla V(x) \cdot f(x)$.

Candidates:

$V(x) = \sum_i c_i x_i^2$, or $V(x) = x^T A x$ for $A = A^T$ positive definite, or $V = \text{energy/Hamiltonian}$ etc.

Theorem: Assume $f(0) = 0$ and f Lipschitz in some domain $B \ni 0$.

- (a) If $V(x)$ is a **weak Liapunov function** for (1) in B , then $x = 0$ is an **stable** equilibrium point for (1).
- (b) If $V(x)$ is a **strong Liapunov function** for (1) in B , then $x = 0$ is an **asymptotically stable** equilibrium point for (1).

Liapunov methods for autonomous systems

$$(1) \quad \dot{x} = f(x); \quad f, x \in \mathbb{R}^n.$$

Definitions:

- 1 $\dot{V}(x) = \frac{d}{dt} V(x(t)) = \nabla V(x) \cdot f(x)$
- 2 $V(x)$ weak (strong) **Liapunov function** of (1) in domain $B \ni 0$ if
 - 1 $V \in C^1(B)$,
 - 2 $V(0) = 0$ and $V(x) > 0$ for $x \neq 0$
 - 3 $\dot{V}(0) = 0$ and $\dot{V}(x) \leq 0$ ($\dot{V}(x) < 0$) for $x \neq 0$

Theorem: Assume $f(0) = 0$ and f Lipschitz in a domain $B \ni 0$.

- 1 $V(x)$ weak **Liapunov function** for (1) $\Rightarrow x(t) \equiv 0$ is **stable**
- 2 $V(x)$ strong **Liapunov function** for (1) $\Rightarrow x(t) \equiv 0$ is **asympt. stable**
- 3 $U(x)$ satisfy 1–3 below $\Rightarrow x(t) \equiv 0$ is **unstable**
 - 1 $U \in C^1(B)$,
 - 2 $U(0) = 0$; for any $\delta > 0$, there is $x \in B$ s.t. $|x| < \delta$ and $U(x) > 0$,
 - 3 there is $\eta > 0$ such that $\dot{U}(x) > 0$ for all x s.t. $|x| < \eta$ and $U(x) > 0$

Examples: $V = \frac{1}{2}(x_1^2 + x_2^2)$, = energy/Hamiltonian; $U = x_1x_2$, = $x_1^2 - x_2^2$

Invariant domains and domains of attraction

$$(1) \quad \dot{x} = f(x); \quad f, x \in \mathbb{R}^n.$$

① Flow: $\phi(t; x_0)$ solution of (1) and $x(0) = x_0$.

② $\Omega \subset \mathbb{R}^n$ invariant (positive invariant) under (1) if $\phi(t; x_0) \in \Omega$ for all $t \in \mathbb{R}$ ($t \geq 0$) and $x_0 \in \Omega$.

$\Omega_a = \{x \in \mathbb{R}^n : \lim_{t \rightarrow \infty} \phi(t; x) = x_0\}$ domain of attraction of x_0 .

③ Lemma: $\Omega = \{x : V(x) \leq c\}$ is positive invariant if $f, V \in C^1$ and $\nabla V(x) \neq 0$ and $\dot{V}(x) = \nabla V(x) \cdot f(x) \leq 0$ on $V(x) = c$.

④ Lasalle's invariance principle. Assume:

① $f(0) = 0$ and f Lipschitz in $\Omega \ni 0$,

② Ω positive invariant, closed and bounded,

③ $V(x)$ is a weak Liapunov function on Ω ,

④ there is no global in time solution of (1) such that

$$x(t) \in \Omega \setminus \{0\} \quad \text{and} \quad V(x(t)) = \text{constant} \quad \text{for all} \quad t \in \mathbb{R}.$$

Then $x = 0$ is asymptotically stable and $\Omega \subset \Omega_a$ (attraction).

Hamiltonian 2×2 systems

$$(1) \quad \dot{x} = f(x); \quad x = (x_1, x_2), \quad f = (f_1, f_2).$$

Definition: (1) **Hamiltonian** if there is a **Hamilton function** $H(x) \in C^2$ s.t.

$$f_1 = \frac{\partial H}{\partial x_2} \quad \text{and} \quad f_2 = -\frac{\partial H}{\partial x_1}.$$

- 1 (1) **Hamiltonian** $\Leftrightarrow \nabla \cdot f = 0$ (f **divergence free**)
- 2 H **Hamilton function**, $x(t)$ **solution of (1)** $\Rightarrow H(x(t)) = \text{constant}$
- 3 **Equilibrium points of (1)** = **critical points of H** ($\nabla H = 0$)
- 4 **Classification of equilibrium point x_0 via 2nd derivative test for H :**
 $q(x_0) > 0$ **center**, $q(x_0) < 0$ **saddle**, $q(x_0) = 0$ **no conclusion**,
where $q = \det D^2 H = H_{x_1 x_1} H_{x_2 x_2} - H_{x_1 x_2} H_{x_2 x_1} = \lambda_1(D^2 H)\lambda_2(D^2 H)$.

OBS: Similar results hold for $2n \times 2n$ systems: $f_1 = \nabla_{x_2} H$, $f_2 = -\nabla_{x_1} H$

Index theory for 2×2 systems

$$(1) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^2.$$

① Polar angle of f : $\phi = \arctan \frac{f_2}{f_1}$

② Curve Γ : Simple, closed, p.w. C^1 , oriented counter cl.wise, $f|_{\Gamma} \neq 0$

③ Index of Γ : $I_{\Gamma} = \frac{1}{2\pi} \oint_{\Gamma} d\phi$

④ $\Gamma = \{y(s) : s_0 \leq s \leq s_T\}$:

$$\bullet I_{\Gamma} = \frac{1}{2\pi} \int_{s_0}^{s_T} \frac{f_1 \frac{d}{ds} f_2 - f_2 \frac{d}{ds} f_1}{f_1^2 + f_2^2} ds; \quad f_i(s) = f_i(y(s)),$$

$$\bullet I_{\Gamma} = \frac{1}{2\pi} \int_{y(s_0)}^{y(s_T)} \nabla \phi \cdot dy = \frac{\phi(y(s_T)) - \phi(y(s_0))}{2\pi} \in \mathbb{Z} \text{ since } f(s_0) = f(s_T)$$

⑤ Winding number of $\Gamma_f = \{f(x) : x \in \Gamma\}$: $\frac{1}{2\pi} \int_{\Gamma_f} \frac{X_1 dX_2 - X_2 dX_1}{X_1^2 + X_2^2} = I_{\Gamma}$.

$$I_{\Gamma} = \frac{1}{2}(p + q) \text{ where}$$

p = number of times $f(x)$ is parallel with any given axis and crosses it counter clockwise as x traverses Γ counter clockwise

q = number of times f cross the same axis clockwise on Γ .

Index theory for 2×2 systems

$$(1) \quad \dot{x} = f(x); \quad x = (x_1, x_2), \quad f = (f_1, f_2).$$

Definition:

- 1 Index of curve Γ : $I_\Gamma = \frac{1}{2\pi} \oint_\Gamma d\phi$
- 2 Index of equilibrium point x_e :
 $I_{x_e} = I_\Gamma$ for any Γ encircling x_e but no other equilibrium point of (1).

Remarks:

- 1 $\phi = \arctan \frac{f_2}{f_1}$ polar angle of f .
- 2 Γ : Simple, closed, p.w. C^1 curve oriented counter cl.wise, $f|_\Gamma \neq 0$.
- 3 Saddle: $I = -1$; node, spiral, center, closed phase trajectory: $I = 1$
- 4 There are equilibrium points with $I = n$ for any $n \in \mathbb{Z}$.

Theorem: Assume f is C^2 and $f = 0$ in Ω only at x_1, \dots, x_n . Then

$$I_\Gamma = I_{x_1} + I_{x_2} + \dots + I_{x_n}$$

for any curve Γ in Ω encircling x_1, \dots, x_n .

Closed trajectories and Poincare sequences

$$(1) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^2.$$

Closed trajectories

- 1 Closed phase trajectories \Leftrightarrow periodic solutions (autonomous syst.)
- 2 Limit cycles: Isolated closed phase trajectories.
- 3 Index test: Closed curve Γ surround equilibrium points x_i , index I_i :

$$\sum I_i \neq 1 \Rightarrow \Gamma \text{ is not a phase trajectory.}$$

- 4 Dulac's test: Ω open, simply connected; $\rho, f = (f_1, f_2) \in C^1(\Omega)$:
 $\nabla \cdot (\rho f) < 0$ in Ω (or > 0) \Rightarrow no closed phase trajectory in Ω .
- 5 $\rho \equiv 1 \rightarrow$ Bendixon's negative criterion

Poincare sequences:

- 1 Poincare cross section Σ : curve transversal (=non-parallel) to f
- 2 Poincare map P_Σ of $x_0 \in \Sigma$: Point of first return of flow $\phi(t; x_0)$ to Σ
- 3 Poincare sequence: $x_0, P_\Sigma(x_0), \dots, P_\Sigma^n(x_0), \dots$ ($P^n = P \circ P \circ \dots \circ P$)

Poincare Bendixon's theorem

$$(1) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^2.$$

- ① $\Gamma_{x_0}^{\pm} = \{\phi(t, x_0) : t \in [0, \pm\infty)\}$ ($\Gamma_{x_0} = \Gamma_{x_0}^+ \cup \Gamma_{x_0}^-$ trajectory; ϕ flow)
- ② $\Omega \subset \mathbb{R}^2$ positive invariant under (1): $\Gamma_{x_0}^+ \subset \Omega$ for all $x_0 \in \Omega$.
- ③ ω -limit set of Γ_{x_0} : All z s.t. $x(t_n) = \phi(t_n; x_0) \xrightarrow{t_n \rightarrow \infty} z$ for some $\{t_n\}_n$.
- ④ Cycle: A periodic trajectory but no equi. point (limit or center c.).

Theorem (Poincare-Bendixon):

$\Gamma^+ \subset$ closed bounded K ; no equilibrium points in $\omega_{\Gamma} \Rightarrow \omega_{\Gamma}$ is a cycle.

Lemma 1: $\Gamma^+ \subset K$, K closed, bounded (=compact)

$\Rightarrow \omega_{\Gamma} \subset K$, $\neq \emptyset$, closed, bounded, connected, invariant under (1).

Lemma 2: Assume L transversal line segment ($L \nparallel f$)

- ① Γ crosses L in two different points $\Leftrightarrow \Gamma$ is not closed.
- ② Crossing points of Γ and L are ordered the same way along L as along Γ .

Corollary: $x_0 \in \omega_{\Gamma} \cap \Gamma \Leftrightarrow \Gamma$ closed trajectory.

Application of Poincare Bendixon's theorem

$$(1) \quad \dot{x} = f(x); \quad x, f \in \mathbb{R}^2.$$

Theorem (Poincare-Bendixon):

$\Gamma^+ \subset$ closed bounded K ; no equilibrium points in $\omega_\Gamma \Rightarrow \omega_\Gamma$ is a cycle.

Cycle: A periodic trajectory but not equilibrium point (limit or center c.)
Corresponds to a periodic solution.

Corollary 1: $K \subset \mathbb{R}^2$ closed, bounded, pos. invariant, with no equi. pt's
 \Rightarrow at least one cycle in K .

Lemma: $\Omega = \{x : V(x) \leq c\}$ is positive invariant if $f, V \in C^1$ and
 $\nabla V(x) \neq 0$ and $\dot{V}(x) = \nabla V(x) \cdot f(x) \leq 0$ on $V(x) = c$.

Corollary 2: There is at least one cycle in $K = \{x : c_1 \leq V(x) \leq c_2\}$ if

- 1 K is bounded and contains no equilibrium points,
- 2 $\nabla V \neq 0$ and $\dot{V} = \nabla V \cdot f \geq 0$ on $V(x) = c_1$,
- 3 $\nabla V \neq 0$ and $\dot{V} = \nabla V \cdot f \leq 0$ on $V(x) = c_2$.

The Lienard equation

$$(2) \quad \ddot{x} + f(x, \dot{x})\dot{x} + g(x) = 0 \quad \Leftrightarrow \quad \begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -f(x_1, x_2)x_2 - g(x_1) \end{cases}$$

Obs:

- 1 **Energy:** $E(x_1, x_2) = \frac{1}{2}x_2^2 + G(x_1)$, where $G(x) = \int_0^x g(s)ds$.
- 2 $\frac{d}{dt}E(x(t)) = \nabla E(x) \cdot \dot{x} = -f(x_1, x_2)x_2^2 \leq 0$ (≥ 0) if $f \geq 0$ ($f \leq 0$)

Theorem (cycles): Equation (2) has at least one cycle if

- (a) $f(x_1, x_2)$ continuous, $f < 0$ for $|x| < r$, $f > 0$ for $|x| > R$,
- (b) $g(x_1)$ continuous, $g < 0$ for $x_1 < 0$, $g > 0$ for $x_1 > 0$,
- (c) $G(x_1) = \int_0^{x_1} g(s)ds \rightarrow \infty$ as $|x_1| \rightarrow \infty$.

Idea: $x = 0$ only equilibrium point, $E(x) = c_1$ and $E(x) = c_2$ bounds bounded invariant region for c_1 small and c_2 big, use [Poincare-Bendixon](#).

The Lienard equation

$$(1) \quad \ddot{x} + f(x, \dot{x})\dot{x} + g(x) = 0 \quad \Leftrightarrow \quad \begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -f(x_1, x_2)x_2 - g(x_1) \end{cases}$$

Energy: $E = \frac{1}{2}x_2^2 + \int_0^x g(s)ds$; $\dot{E} = -f(x_1, x_2)x_2^2$.

Theorem (cycle): Equation (1) has at least one cycle if

- (a) $f(x_1, x_2)$ continuous, $f < 0$ for $|x| < r$, $f > 0$ for $|x| > R$,
- (b) $g(x_1)$ continuous, $g < 0$ for $x_1 < 0$, $g > 0$ for $x_1 > 0$,
- (c) $G(x_1) = \int_0^{x_1} g(s)ds \rightarrow \infty$ as $|x_1| \rightarrow \infty$.

Theorem (centre): Equation (1) has a centre at $x = 0$ if for $|x| < R$:

- (a) $f = f(x_1)$ continuous, odd, one sign for $x_1 < 0$,
- (b) $g(x_1)$ continuous, odd, $g > 0$ for $x_1 > 0$,
- (c) $g(x_1) > \alpha f(x_1) \int_0^{x_1} f(s)ds$ for $\alpha > 1$.

Obs: Orbits loose energy in $\{x_1 \geq 0\}$ but gain same amount in $\{x_1 < 0\}$.

Theorem (limit cycle): Equation (1) has one and only one cycle if

- (a) $F(x) = \int_0^x f(s)ds$, g locally Lipschitz,
- (b) g odd, $g > 0$ for $x > 0$,
- (c) F odd, $= 0$ only at $x = 0, \pm a$, $F \rightarrow \infty$ monotonically for $x > a$.

Obs: The proof uses the Lienard plane: $\dot{x}_1 = x_2 - F(x_1)$, $\dot{x}_2 = -g(x_1)$