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General linear non autonomous systems.

$\dot{x} = A(t)x$, $x(t_0) = x_0$, $A \in C(I, \mathbb{R}^{n \times n})$,
- a matrix - continuous in an interval I . $A(t)x$ is always lipschitz

Theorem 3.9. The system has a unique solution satisfying initial condition $x(t_0) = x_0$, on the whole interval $t \in I$, $t_0 \in I$.

It follows directly from Th. 2.17 on equations $\dot{x} = \delta(x, t)$ with $\delta(t, x)$ rising not faster than linearly in x .

An independent proof. Let $C = \max_{t \in [t_0, T]} \|A(t)\|$

Let $\varphi(t)$ = solution and $\|\varphi(t)\| = r(t)$

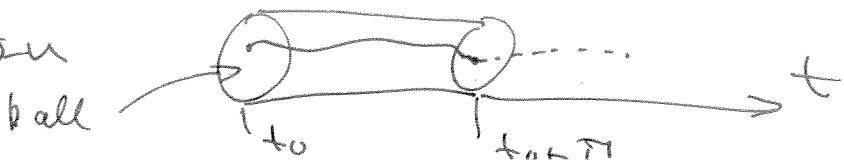
$$\text{If } b(r) = \ln(r^2) \quad t_0 < t < T$$

$$L \leq \frac{2r\dot{r}}{r^2} \leq 2C$$

$$L(t) \leq L(t_0) + 2C(t-t_0)$$

$$\Rightarrow \|\varphi(t)\| \leq e^{C(t-t_0)} \|\varphi(t_0)\|$$

It means that solution $\varphi(t)$ does not leave a ball with radius $\|e^{C(T-t_0)}\| \|\varphi(t_0)\|$ and must have an extension



Superposition principle.

A sum of solutions to a linear system
 $\dot{x} = A(t)x$ is also a solution.

Solutions to this equation form a vector space.

Theorem 3.10 The solutions to $\dot{x} = A(t)x$ form an n -dimensional vector space.

There is a matrix-valued solution $\Pi(t, t_0)$ such that the solution satisfying initial data $x(t_0) = x_0$ is given by

$$x(t) = \Pi(t, t_0)x_0 ; \quad \Pi(t, t_0) = A(t)\Pi(t, t_0)$$

$\Pi(t, t_0)$ is called principal matrix solution. $\Pi(t_0, t_0) = I$ - unit matrix.

Proof. Suppose there are n solutions $x_1(t), x_2(t), \dots, x_n(t)$ that are linearly independent at $t_0 = t$ and are linearly dependent at some $t = t^*$.

Then $\{c_i\}_{i=1}^n$ such that $c_i \neq 0$ all together,

$$\sum_{i=1}^n c_i x_i(t_0) \neq 0$$

and $\sum_{i=1}^n c_i x_i(t^*) = 0$ (linear dependent)

We observe that $X(t) - \sum_{i=1}^n c_i x_i(t)$ is the solution to

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$\dot{x} = A(t)x$, because of the superposition principle and

$$X(t_0) \neq 0; \quad X(t_*) = 0.$$

The last is impossible, therefore $X_1(t), \dots, X_n(t)$ must be linearly independent for all $t \in I$ if they are linearly independent at time t_0 . It proves that the space of solutions is n -dimensional, because the space of initial data $x(t_0) = x_0$ is n -dimensional.

The second clause in the theorem follows by the explicit construction and the superposition principle.

We solve the equation $\dot{x} = A(t)x$ with initial data δ_j ; $j=1, \dots, n$, where δ_j - standard basis vectors in \mathbb{R}^n $(1, \dots, 0)$, $(0, 1, \dots, 0)$, $(0, 0, 1, \dots, 0)$ etc. that are columns in the unit matrix, I .

Corresponding solutions $\phi(t, t_0, \delta_j)$ put in this order in a matrix

$$\Pi(t, t_0) = [\phi(t, t_0, \delta_1), \dots, \phi(t, t_0, \delta_n)]$$

give the principal matrix solution

Any vector $x_0 = I x_0$ and

$\phi(t, t_0, x_0) = [\Pi(t, t_0) x_0]$ for an arbitrary initial data at time t_0 ,

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$\Pi(t, t_0)$ is an exponent $\exp(At)$ in the case when $A(t)$ is independent of time (constant matrix.)

In the general case there is no explicit expression for the principal matrix solution $\Pi(t, t_0)$.

The matrix $\Pi(t, t_0)$ is non-degenerate because it's non-degenerate for $t=t_0$ (unit matrix $\Pi(t_0, t_0) = \mathbb{I}$) and its column solutions are linearly independent for all other times because of the proof above.

It makes that $\Pi(t, t_0)$ is a one to one mapping from \mathbb{R}^n to \mathbb{R}^n and again proves that the set of all solutions is an n -dimensional vector space.

$$\text{Example. } \begin{cases} x_1' = x_1 + t x_2 \\ x_2' = 2x_2 \end{cases}$$

Find $\Pi(t, t_0)$. $\Pi(t, t_0)$ has the first column-solution with $x_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ and the second column - with $x_0 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

$x_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ gives $x_2 = 0; x_1 = e^{t-t_0}$

$x_0 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ gives $x_2 = e^{2(t-t_0)}; x_1 = e^{(t-t_0)} - e^{(t-t_0)(t_0-1)}$

$$\text{Therefore } \square(t_1, t_0) = \begin{pmatrix} e^{t-t_0} & e^{2(t-t_0)} \\ 0 & e^{2(t-t_0)} \end{pmatrix}^{t-t_0}$$

We observe that $t_0 \leq t_1 \leq t$

$$\square(t_1, t_0) \square(t_1, t_0) = \square(t_1, t_0)$$

because both expressions solve the equation $\dot{\square}(t, t_0) = A(t) \square(t)$ and they coincide at the point $t = t_0$, therefore must be equal everywhere by uniqueness of solutions. The meaning of the equation is simple: the system moves from time t to t_1 and then from t_1 to t and it is the same as to move all the way from t_0 to t , because of the uniqueness.

Definition

Fundamental matrix solution is an arbitrary matrix solution such that $\det(\bar{V}(t)) \neq 0$; $\dot{V} = A(t)V$

It is easy to observe that for arbitrary fundamental matrix solution $V(t)$

$$V(t)(V(t_0))^{-1} = \square(t_1, t_0), \text{ because}$$

$$V(t_0)(V(t_0))^{-1} = I.$$

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It's easy to observe that

$$(\Sigma(t_1, t_0))^{-1} = \Sigma(t_0, t)$$

Lemma 3.11 For any fundamental matrix solution

$$\det(V(t)) = \det(V(t_0)) \exp \left\{ \int_{t_0}^t \text{tr}(A(s)) ds \right\}$$

$$\text{tr } A = \sum_{i=1}^n A_{ii}.$$

Direct proof. Let $A = a_{jk}(t)$; $V(t) = y_k^j(t)$

$$\Delta(t) = \det V(t)$$

$$\Delta(t) = \sum_{j=1}^n \det(V_j(t)) \quad \text{where } V_j(t)$$

is a matrix taken from $V(t)$ by exchanging its j -th row (y_1^j, \dots, y_n^j) by derivatives $(y_1^j)', \dots, (y_n^j)'$.

V is a solution to $\dot{V} = A(t)V$, therefore $(y_k^j(t))' = \sum_{i=1}^n a_{ji} y_k^i$ — j -th row in V_j consists of j -th row in $V(t)$ times a_{jj} and a linear combination of other rows in V_j \Rightarrow

$$\det V_j = a_{jj} \det(V); \Rightarrow \Delta'(t) = (\text{tr } A) \cdot \Delta(t) \Rightarrow \Delta(t) = \Delta(t_0) e^{\int_{t_0}^t \text{tr } A ds}$$

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Inhomogeneous linear systems.

Theorem 3.12 Solution to inhomogeneous system with initial condition

$$\dot{x} = A(t) + g(t), \quad x(t_0) = x_0$$

$$\text{is } x(t) = \Gamma(t, t_0)x_0 + \int_{t_0}^t \Gamma(t, s)g(s)ds.$$

(Variant of Duhamel's formula)

Proof. Check by direct calculation and substituting to the equation.

One can guess it by variation of parameter. Let

$$x(t) = \Gamma(t, t_0)c(t); \quad c(t_0) = x_0.$$

$$\dot{x} = A(t)x(t) + \Gamma(t, t_0)\dot{c}(t)$$

$$\Rightarrow \dot{c}(t) = \Gamma(t_0, t)g(t)$$

$$c(t) = x_0 + \int_{t_0}^t \Gamma(t_0, s)g(s)ds$$

$$\Rightarrow x(t) = \Gamma(t, t_0)x_0 + \int_{t_0}^t \Gamma(t, s)g(s)ds$$

$$\underbrace{\Gamma(t, t_0)}_{\Gamma(t_0, s)} \underbrace{\Gamma(t_0, s)}_{\Gamma(t, s)} = \underbrace{\Gamma(t, s)}_{!}$$