VV286 RC2

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Slope parametrization

Given y'' exists and $y'' \neq 0$, y' is monotonic function of x. We can use slope to parametrize the solution curve.

$$p = y'(x) = y'(x(p))$$

$$\frac{dy(p)}{dp} = \frac{d}{dp}y(x(p)) = \frac{dy}{dx}\Big|_{x=x(p)} \cdot \frac{dx(p)}{dp} = p \cdot \frac{dx(p)}{dp}$$

$$F(y,y';x)=0$$

1. Try to use slope parametrization. Solve

$$F(y(p), p; x(p)) = 0, y'(p) = px'(p)$$

2. Straight line solution.



General Implicit Differential Equation

Use slope parametrization,

$$F(y, y'; x) = 0$$

$$\Rightarrow F(y(p), p; x(p)) = 0$$

$$\Rightarrow F_x \dot{x} + F_y \dot{y} + F_p = 0$$

$$\xrightarrow{\underline{y'(p) = px'(p)}} \dot{x} = -\frac{F_p}{F_x + pF_y}, \quad \dot{y} = -\frac{pF_p}{F_x + pF_y}$$

y = xy' + g(y') (Clairaut's equation)

Assume $g \in C^1(I)$ for some interval I.

1. Use slope parametrization, $y(p) = x(p) \cdot p + g(p)$, then

$$y'(p) = px'(p) + x(p) + g'(p)$$

Since y'(p) = px'(p),

$$x(p) = -g'(p), \quad y(p) = -pg'(p) + g(p)$$

2. Straight line solution: $y = cx + g(c), c \in I$

The solution of Clairaut's equation obtained from a slope parametrization of the integral curve is always the envelope of the straight-line solutions.

Given a one-parameter family of smooth curves in \mathbb{R}^2

$$\textit{F} = \{\mathcal{C}_{\textit{s}}, \textit{s} \in \textit{I} \subset \mathbb{R}\}$$

with each curve C_p parameterized by a function

$$\gamma(s,\cdot): J \to \mathcal{C}_p, \quad t \mapsto \gamma(s,t)$$

Then the envelope \mathcal{E} of F which is parametrized by $\gamma(s, \psi(s))$ can be found through

$$\frac{\partial \gamma_1}{\partial s} \frac{\partial \gamma_2}{\partial t} = \frac{\partial \gamma_1}{\partial t} \frac{\partial \gamma_2}{\partial s}, \quad t = \psi(s)$$

Another way to solve Clairaut's equation

1. Straight line solution: $y = cx + g(c), c \in I$

2.
$$\gamma(c,x) = \begin{pmatrix} x \\ cx + g(c) \end{pmatrix}$$

$$\frac{\partial \gamma_1}{\partial c} \frac{\partial \gamma_2}{\partial x} = \frac{\partial \gamma_1}{\partial x} \frac{\partial \gamma_2}{\partial c} \Rightarrow 0 = x + g'(c)$$

We obtain the parametrization of ${\mathcal E}$ as $\gamma({\boldsymbol c}, -{\boldsymbol g}'({\boldsymbol c}))$. So

$$y(c) = -cg'(c) + g(c)$$

$$y = xf(y') + g(y')$$
 (d'Alembert's equation)

Assume $f, g \in C^1(I)$ for some interval I.

1. Use slope parametrization, $y(p) = x(p) \cdot f(p) + g(p)$, then

$$y'(p) = f(p)x'(p) + f'(p)x(p) + g'(p)$$

Since
$$y'(p) = px'(p)$$
, $x'(p) = \frac{f'(p)x(p) + g'(p)}{p - f(p)}$

2. Straight line y = cx + d is solution if and only if c = f(c), d = g(c)

Example

$$y = (y \cdot y' + 2x) \cdot y'$$

Use slope parametrization,

$$y(p) = p^{2}y(p) + 2x(p)p, \quad y'(p) = px'(p)$$

$$\Rightarrow y(p) = \frac{2px(p)}{1 - p^{2}}$$

$$\Rightarrow px'(p) = y'(p) = \frac{(2x(p) + 2px'(p))(1 - p^{2}) - 2px(p)(-2p)}{(1 - p^{2})^{2}}$$

$$\Rightarrow ((1 - p^{2})p - 2p)x'(p) = \frac{2(1 + p^{2})}{1 - p^{2}}x(p)$$

$$\Rightarrow x'(p) = -\frac{2}{p(1 - p^{2})}x(p) = -\frac{1}{p}\left(\frac{1}{1 - p} + \frac{1}{1 + p}\right)x(p)$$

$$\Rightarrow x'(p) = -\left(\frac{1}{p} + \frac{1}{1 - p} + \frac{1}{p} - \frac{1}{1 + p}\right)x(p)$$

So $\ln |x(p)| = -2 \ln |p| + \ln |p-1| + \ln |p+1| + C$, i.e.

$$x(p) = C \cdot \left(1 - \frac{1}{p^2}\right)$$

and therefore $y(p) = \frac{2px(p)}{1 - p^2} = -\frac{2C}{p}$ So

$$x = C \cdot \left(1 - \left(\frac{y}{2C}\right)^2\right) = C - \frac{y^2}{4C}$$

Straight line solution is given by y = 0

System of Equations

Higher Order Equations

Given an explicit ODE of order n,

$$x^{(n)}(t) = f(x, x', x'', \cdots, x^{(n-1)}, t)$$

set

$$x_1(t) = x(t), x_2(t) = x'(t), x_3(t) = x''(t), \dots, x_n(t) = x^{(n-1)}(t)$$

and we obtain

$$\begin{pmatrix} x'_{1}(t) \\ x'_{2}(t) \\ x'_{3}(t) \\ \vdots \\ x'_{n}(t) \end{pmatrix} = \begin{pmatrix} x_{2}(t) \\ x_{3}(t) \\ x_{4}(t) \\ \vdots \\ f(x_{1}, x_{2}, \cdots, x_{n}, t) \end{pmatrix}$$





Initial Value Problem

$$\frac{dx}{dt} = F(x,t), \quad x(t_0) = x_0 \in \mathbb{R}^n$$

Picard Iteration

- 1. Guess a function $x^{(0)}(t) = x_0$ (constant)
- 2. Set $x^{(k+1)}(t) := x_0 + \int_{t_0}^t F(x^{(k)}(s), s) ds, k \in \mathbb{N}$

Under suitable conditions on the function F, the sequence of functions $(x^{(k)})$ will converge to a unique function x(t) which satisfies

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

Theorem of Picard-Lindelöf (Existence and Uniqueness)

Let $x_0 \in \Omega$, where $\Omega \subset \mathbb{R}^n$ is open and let $t_0 \in I$, where $I \subset \mathbb{R}$ is an interval. Suppose $F: \Omega \times I \to \mathbb{R}^n$ is a continuous function satisfying a Lipschitz estimate in x: there exists an L > 0 such that for all $x, y \in \Omega$ and all $t \in I$

$$||F(x,t)-F(y,t)|| \leqslant L||x-y||$$

then the initial value problem has a unique solution in some t-interval containing t_0 .

Gronwall's Inequality (Stability)

Under same conditions, x, y satisfies

$$x'(t) = F(x, t), x(t_0) = x_0, y'(t) = F(y, t), y(t_0) = y_0$$

Then

$$||x(t) - y(t)|| \le e^{L \cdot |t - t_0|} ||x_0 - y_0||$$



Linear Systems of Equations (Explict Linear Higher Order ODE)

$$\frac{dx}{dt} = A(t)x + b(t), \quad t \in I \subset \mathbb{R}$$

where $A: I \rightarrow Mat(n \times n, \mathbb{R})$ is a matrix-valued function of t and $b: I \rightarrow \mathbb{R}^n$

Construction of Solutions

$$x(t) = x_{\mathsf{hom}}(t) + x_{\mathsf{part}}(t)$$

$$x_{\text{hom}}(t) = \sum_{k=1}^{n} \lambda_k x^{(k)}(t)$$

where $x^{(k)}(t)$, $k = 1, \dots, n$ satisfy

$$\frac{dx^{(k)}}{dt} = A(t)x^{(k)}, \left(x^{(k)}(t_0) = b_k, x_0 = \sum_{k=1}^n \lambda_k b_k\right)$$

for some numbers $\lambda_1, \cdots, \lambda_n \in \mathbb{R}$ and

$$\left(\underset{t \in I}{\forall} \sum_{k=1}^{n} \lambda'_{k} x^{(k)}(t) = 0 \right) \Rightarrow \lambda'_{1} = \dots = \lambda'_{n} = 0$$



Solution Space

A vector space contains all solution.

Span
$$\{x^{(1)}, \cdots, x^{(n)}\}$$

Fundamental System of Solutions

A set of functions giving a bassis of the solution space.

$$\{x^{(1)},\cdots,x^{(n)}\}$$

Fundamental Matrix

$$X(t) = (x^{(1)}(t), \cdots, x^{(n)}(t))$$

Systems of Linear ODEs with Constant Coefficients

$$\frac{dx}{dt} = Ax, \quad x(0) = x_0$$

Try

$$x(t) \stackrel{?}{=} e^{At} x_0 \stackrel{\Delta}{=} x_0 \left(\mathbb{1} + \sum_{k=1}^{\infty} \frac{A^k t^k}{k!} \right)$$





Well-defined

Use operator norm with euclidean norm in \mathbb{R}^n

$$||A|| = \sup_{x \in \mathbb{R}^n \setminus \{0\}} \frac{|Ax|}{|x|}$$

Then

$$\left| \sum_{k=1}^{\infty} \left| \left| \frac{A^k t^k}{k!} \right| \right| = \sum_{k=1}^{\infty} \frac{||A^k|| \cdot |t^k|}{k!} \le \sum_{k=1}^{\infty} \frac{||A||^k \cdot |t|^k}{k!} = e^{|t| \cdot ||A||} - 1 < \infty$$

The series is absolutely convergent and therefore it's convergent to some matrix in $Mat(n \times n, \mathbb{R})$

Satisfy the Differential Equation

$$\frac{d}{dt}e^{At} = \frac{d}{dt}\sum_{k=0}^{\infty} \frac{A^k t^k}{k!} = \sum_{k=0}^{\infty} \frac{d}{dt} \frac{A^k t^k}{k!}$$
$$= \sum_{k=1}^{\infty} \frac{A^k t^{k-1}}{(k-1)!} = A \sum_{k=0}^{\infty} \frac{A^k t^k}{k!}$$
$$= Ae^{At}$$

And for
$$t = 0$$

$$e^{At}=1\Rightarrow x(0)=x_0$$

Eigenvalue Problem

Eigenvalue

Let V be a real or complex vector space and $L \in \mathcal{L}(V, V)$. Then a number $\lambda \in \mathbb{F}$ such that for some $x \neq 0$

$$Lx = \lambda x$$

Eigenvector

Any x such that $Lx = \lambda x$ holds is called an eigenvector for the eigenvalue λ .

Eigenspace

The subspace

$$V_{\lambda} = \{ x \in V : Lx = \lambda x \}$$





The Eigenvalue Problem for Matrices

For a matrix $A \in Mat(n \times n, \mathbb{R})$

$$Ax = \lambda x \Leftrightarrow (A - \lambda 1)x = 0 \Rightarrow det(A - \lambda 1) = 0$$

 $p(\lambda) := det(A - \lambda 1)$ is a polynomial of degree n. It has at most n distinct roots.

If A has n distinct eigenvalues $\lambda_1, \dots, \lambda_n$, then it has precisely n independent eigenvectors v_1, \dots, v_n and $\mathbb{R}^n = \bigoplus_{j=1}^n V_{\lambda_j}$.

Set
$$U = (v_1, \dots, v_n), D = U^{-1}AU$$

$$De_k = U^{-1}AUe_k = U^{-1}Av_k = U^{-1}(\lambda_k v_k) = \lambda_k e_k$$

So

$$D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & \lambda_n \end{pmatrix}$$

Matrix Powers

$$D^{k} = (U^{-1}AU)^{k} = \underbrace{(U^{-1}AU)(U^{-1}AU)\cdots(U^{-1}AU)}_{k \text{ times}} = U^{-1}A^{k}U$$

$$A^k = UD^kU^{-1}$$

Functional Calculus

For a power series has infinite redius of convergence

$$f(x) = \sum_{j=0}^{\infty} c_j x^j$$

$$f(A) = \sum_{j=0}^{\infty} c_j A^j = \sum_{j=0}^{\infty} c_j (UD^j U^{-1}) = U \left(\lim_{N \to \infty} \sum_{j=0}^{N} c_j D^j \right) U^{-1}$$

$$= U \begin{pmatrix} \sum_{j=0}^{\infty} c_j \lambda_1^j & 0 \\ & \ddots & \\ 0 & & \sum_{j=0}^{\infty} c_j \lambda_n^j \end{pmatrix} U^{-1}$$

$$= U \begin{pmatrix} f(\lambda_1) & 0 \\ & \ddots & \\ 0 & & f(\lambda_n) \end{pmatrix} U^{-1}$$



Matrix Exponential

For a diagonlizable matrix A,

$$e^{At}=Uegin{pmatrix} e^{\lambda_1 t} & 0 \ & \ddots & \ 0 & e^{\lambda_n t} \end{pmatrix} U^{-1}$$





The Spectral Theorem

Every self-adjoint matrix A is diagonalizable.

Self-adjoint

For a inner product space $(V, \langle x, Ay \rangle)$ and $A : V \to V$, the adjoint of A is defined through the relation

$$\langle x, Ay \rangle = \langle A^*x, y \rangle$$
 for all $x, y \in V$

If $A = A^*$, the map A is said to be self-adjoint. In the case of matrices, if $A \in \text{Mat}(n \times n, \mathbb{C})$

$$A = A^* = \overline{A}^T$$





Non-Diagonalizable Matrices

"Bottom-up" Method

For some eigenvalue λ , dim $V_{\lambda} < a_{\lambda}$,

$$E_1 = V_{\lambda} = \ker(A - \lambda \mathbb{1})$$

$$E_k = ker(A - \lambda \mathbb{1})^k = \{ v \in V : (A - \lambda \mathbb{1})^k v = 0 \}$$

Choose (dim $V_{\lambda}-1$) different eigenvector for the eigenvalue λ . Then start from k=1. Choose a suitable $v^{(k)} \in E_k$ and solve $(A-\lambda \mathbb{1})v^{(k+1)}=v^{(k)}$ to find one $v^{(k+1)} \in E_{k+1} \setminus E_k$. k=k+1. Repeat until dim $V_{\lambda}+k=a_{\lambda}$.

U is formed with all independent (generalized) eigenvectors. For some eigenvalue λ , dim $V_{\lambda} < a_{\lambda}$, assume its generalized eigenvectors are at $i_{\lambda}th$ to $(i_{\lambda}+a_{\lambda}-1)th$ column in U

For
$$i_{\lambda} \leqslant j \leqslant i_{\lambda} + \dim V_{\lambda} - 1$$
, $Ue_{j} \in V_{\lambda}$, so $U^{-1}AUe_{j} = \lambda e_{j}$.

For
$$i_{\lambda}+\dim V_{\lambda}\leqslant j\leqslant i_{\lambda}+a_{\lambda}-1$$
, set $k=j-i_{\lambda}-\dim V_{\lambda}+2$
$$U^{-1}AUe_{j}=U^{-1}Av^{(k)}=U^{-1}(\lambda\mathbb{1}v^{(k)}+v^{(k-1)})$$
$$=\lambda e_{j}+e_{j-1}$$

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"Top-down" Method

For some eigenvalue λ , dim $V_{\lambda} < a_{\lambda}$, set $m = a_{\lambda} - \dim V_{\lambda} + 1$, then solve

$$(A - \lambda 1)^m v = 0, (A - \lambda 1)^{m-1} v \neq 0$$
 as $v^{(m)}$

For $2 \le k \le m-1$, set $v^{(k)} = (A - \lambda \mathbb{1})v^{(k+1)}$ which (naturally) satisfies

$$(A - \lambda 1)^k v^{(k)} = 0, (A - \lambda 1)^{k-1} v^{(k)} \neq 0$$

then
$$v^{(1)} = (A - \lambda \mathbb{1})v^{(2)} \in V_{\lambda}$$
.

Choose $(\dim V_{\lambda} - 1)$ independent eigenvector for the eigenvalue λ .



Jordan Matrices

For $\lambda \in \mathbb{C}$ we define the Jordan block of size $k \in \mathbb{N} \setminus \{0\}$

$$J_k(\lambda) = egin{pmatrix} \lambda & 1 & & 0 \ & \ddots & \ddots & \ & & \ddots & 1 \ 0 & & & \lambda \end{pmatrix} \in \mathsf{Mat}(k imes k, \mathbb{C})$$

A block matrix of the form

$$J = egin{pmatrix} J_{k_1}(\lambda_1) & 0 & 0 \ & \ddots & \ 0 & J_{k_m}(\lambda_m) \end{pmatrix}$$

with not necessarily distinct $\lambda_1, \dots, \lambda_m \in \mathbb{C}$ and $k_1, \dots, k_m \in \mathbb{N}$ is called Jordan matrix.

$$J = D + N$$

There exists some $k \in \mathbb{N}$ such that $N^k = 0$. (Nipotent Matrices) Assume $D, N \in \mathsf{Mat}(n \times n, \mathbb{C})$.

For
$$1 \leqslant i \leqslant j \leqslant n$$

$$(DN)_{ji} = D_{j.} \cdot N_{.i} = D_{jj} N_{ji} + D_{j,i-1} N_{i-1,i} = 0$$

$$(ND)_{ji} = N_{j.} \cdot D_{.i} = N_{j,j+1} D_{j+1,i} + N_{ji} D_{ii} = 0$$
 For $1 \leqslant j \leqslant n-2, j+2 \leqslant i \leqslant n$
$$(DN)_{ji} = D_{j.} \cdot N_{.i} = D_{jj} N_{ji} + D_{j,i-1} N_{i-1,i} = 0$$

$$(ND)_{ij} = N_{i.} \cdot D_{.i} = N_{i,j+1} D_{i+1,i} + N_{ji} D_{ji} = 0$$

For
$$1 \le i \le n - 1, j = i + 1$$

$$(DN)_{i,i+1} = D_{i\cdot} \cdot N_{\cdot,i+1} = D_{ii}N_{i,i+1}$$

$$(ND)_{i,i+1} = N_{i\cdot} \cdot D_{\cdot,i+1} = N_{i,i+1}D_{i+1,i+1}$$

If $D_{ii} \neq D_{i+1,i+1}$, $N_{i,i+1} = 0$ and therefore $(DN)_{i,i+1} = (ND)_{i,i+1}$

So $\forall i, j \in [1, n] \cap \mathbb{N}$, $(DN)_{ij} = (ND)_{ij}$. So DN = ND and we can prove that

$$e^J = e^{D+N} = e^D \cdot e^N$$

For
$$e^N = \sum_{i=0}^{\infty} \frac{1}{i!} N^i = \sum_{i=0}^{k-1} \frac{1}{i!} N^i$$
,

$$e^{At} = Ue^{Jt}U^{-1} = U(e^{Dt} \cdot \sum_{i=0}^{k-1} \frac{1}{i!} (Nt)^i)U^{-1}$$





Homogeneous Solution

For any basis $\mathcal{B} = \{v_1, \cdots, v_n\}$ of \mathbb{R}^n , the systems of functions

$$\mathcal{F} = \{e^{At}v_1, \cdots, e^{At}v_n\}$$

is a fundamental system for $\frac{dx}{dt} = Ax$.

1. Take $v_i = e_i$, the fundamental matrix is given by

$$X(t) = e^{At}$$

2. Take $v_i = u_i$, the fundamental matrix is given by

$$X(t) = Ue^{Jt}$$

2.1 If *A* is diagonalizable, $X(t) = (e^{\lambda_1 t} u_1, \dots, e^{\lambda_n t} u_n)$



Particular Solution

$$\frac{dx}{dt} = Ax + b(t)$$

$$\Rightarrow e^{-At} \frac{dx}{dt} = Ae^{-At} + e^{-At}b(t)$$

$$\Rightarrow \frac{d}{dt}(e^{-At}x) = e^{-At}b(t)$$

$$\Rightarrow x_{part} = e^{At} \int e^{-As}b(s)ds$$





To solve linear explicit higher order ODE with constant coefficient,

- 1. Transform into systems of linear ODEs with constant coefficients $\frac{dx}{dt} = Ax + b(t)$
- 2. Find all eigenvalues of A, calculate their eigenvectors
 - 2.1 For those $\dim V_{\lambda} < a_{\lambda}$, find $a_{\lambda} \dim V_{\lambda}$ more generalized eigenvectors $v_{\lambda}^{(2)}, \cdots, v_{\lambda}^{(a_{\lambda} \dim V_{\lambda} + 1)}$
- 3. Form U with all eigenvectors in order, write down $J(=U^{-1}AU)$ directly. Obtain the fundamental matrix X(t).
- 4. Find particular solution,

4.1
$$x_{part} = e^{At} \int e^{-As} b(s) ds (X(t) = e^{At})$$

4.2 Use Cramer's rule to solve X(t)c'(t) = b(t), then

$$x_{\text{part}}(t) = \sum_{k=1}^{n} c_k(t) x^{(k)}(t)$$



Linear Systems with Variable Coefficients

$$\frac{dx}{dt} = A(t)x + b(t)$$

No general method to solve variable-coefficient homogeneous, linear systems in terms of elementary functions.

Given a fundamental systems of solutions to an associated homogeneous equation, a solution to an inhomogeneous equation can be found.

Variation of Parameters for Linear Systems

$$\frac{dx}{dt} = A(t)x + b(t), A : \mathbb{R} \to Mat(n \times n, \mathbb{R}), b : \mathbb{R} \to \mathbb{R}^n$$

Given a fundamental system $x^{(1)}, x^{(2)}, \dots, x^{(n)}$, then

$$X_{\text{hom}}(t) = c_1 X^{(1)}(t) + \cdots + c_n X^{(n)}(t), \quad c_1, \cdots, c_n \in \mathbb{R}$$

Set

$$x_{\text{part}}(t) = c_1(t)x^{(1)}(t) + \cdots + c_n(t)x^{(n)}(t)$$

$$\frac{dx_{\text{part}}}{dt} = \sum_{k=1}^{n} (c'_{k}(t)x^{(k)}(t) + c_{k}(x^{(k)})'(t)) = A(t)x_{\text{part}} + b(t)$$

$$\Rightarrow \sum_{k=1}^{n} c'_{k}(t)x^{(k)}(t) = b(t)$$

$$x^{(k)} = \begin{pmatrix} x_1^{(k)} \\ \vdots \\ x_n^{(k)} \end{pmatrix}, \quad c(t) = \begin{pmatrix} c_1(t) \\ \vdots \\ c_n(t) \end{pmatrix}$$

$$\sum_{k=1}^{n} (c'_{k}(t)x^{(k)}(t)) = \begin{pmatrix} c'_{1}(t)x_{1}^{(1)}(t) + \dots + c'_{n}(t)x_{1}^{(n)}(t) \\ \vdots \\ c'_{1}(t)x_{n}^{(1)}(t) + \dots + c'_{n}(t)x_{n}^{(n)}(t) \end{pmatrix}$$

$$= \begin{pmatrix} x_{1}^{(1)}(t) & x_{2}^{(2)}(t) & \dots & x_{1}^{(n)}(t) \\ \vdots & & \vdots \\ x_{n}^{(1)}(t) & x_{n}^{(2)}(t) & \dots & x_{n}^{(n)}(t) \end{pmatrix} \begin{pmatrix} c'_{1}(t) \\ \vdots \\ c'_{n}(t) \end{pmatrix}$$

$$= X(t)c'(t)$$

Use Grainer's rule to solve
$$\lambda(t)C(t) = D(t)$$
,

$$c_k'(t) = \frac{\det X^{(k)}(t)}{\det X(t)}$$

where $X^{(k)}$ is the fundamental matrix with the kth column replaced with b

The Wronskian of *n* Solutions of a System

$$W(t) = \det(x^{(1)}(t), \cdots, x^{(n)}(t)) = \det X(t)$$
 $\frac{dW}{dt} = \operatorname{tr} A(t) \cdot W, \quad W(t) = W(t_0) e^{\int_{t_0}^t \operatorname{tr} A(s) ds}$ $W(t) = 0$ for all t or $W(t) \neq 0$ for all t