Differential Equations

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Solution Techniques

Higher Order ODEs & Systems of ODEs: Summary

Higher-Order ODEs

$$y^{(m)}(t) = f(t, y, y', ..., y^{(m-1)}), \quad a \le t \le b$$

 $y(a) = \alpha_1, y'(a) = \alpha_2, ..., y^{(m-1)}(a) = \alpha_m$
System of ODEs
 $\frac{du_1}{dt} = f_1(t, u_1, u_2, ..., u_m),$
 $\frac{du_2}{dt} = f_2(t, u_1, u_2, ..., u_m),$
 \vdots
 $\frac{du_m}{dt} = f_m(t, u_1, u_2, ..., u_m),$
 $u_1(a) = \alpha_1, u_2(a) = \alpha_2, ..., u_m(a) = \alpha_m$
System for Higher-Order ODEs
 $\frac{du_1}{dt} = \frac{dy}{dt} = u_2,$
 $\frac{du_2}{dt} = \frac{dy'}{dt} = u_3,$
 \vdots
 $\frac{du_{m-1}}{dt} = \frac{dy^{(m-2)}}{dt} = u_m$
 $\frac{du_m}{dt} = \frac{dy^{(m-1)}}{dt} = y^{(m)} = f(t, y, y', ..., y^{(m-1)})$
 $= f(t, u_1, u_2, ..., u_m)$
 $u_1(a) = y(a) = \alpha_1, u_2(a) = y'(a) = \alpha_2,$
 $\cdots u_m(a) = y^{(m-1)}(a) = \alpha_m$

Systems of First Order Equations Linear Algebra Review

Linear System of Equations

System with n equations and n unknowns

```
a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1

a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2

\vdots

a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n
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Solve for x in Ax = b.

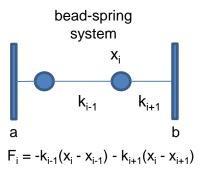
Need theory to determine when this is possible (existence / uniqueness).

Methods to solve

- algebriac approaches (gaussian elimination, factorizations)
- · computational methods (direct methods, iterative methods)
- issues: tractability, robustness to small errors.

Example

For bead-spring system, find locations X_i of beads that balances the forces.





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Eigenvalues & Eigenvectors

Def: A vector
$$v$$
 is called an eigenvector of a matrix A
if there exists a scalar $\lambda \in \mathbb{R}$ so that
 $Av = \lambda v$.

Def! The scalar XEIR is called an <u>eigenvalue</u> of the matrix A.

Def: determinent 2x2 system det [ab] = ad-bc.



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Operations that preserve solution x:
(i) Row E; can be multiplied by any
non-zero constant Y, E;
$$\leq$$
 YE;.
(ii) Row E; can be multiplied by any
non-zero constant & and added to now E;
E; \leq E; $+$ YE;
(iii) Rows E; and E; can always be
transpositioned (exchanged) in order
E; \leq E; \leq E; \leq E;

How can one use these operations to obtain a linear system that is easier to solve?

$$A \underline{X} = b$$

$$a_{11} \underline{X}_{1} + a_{12} \underline{X}_{2} = b_{1}$$

$$a_{21} \underline{X}_{1} + a_{22} \underline{X}_{2} = b_{2}$$

$$A = \begin{bmatrix} E_{1} \\ -E_{2} \end{bmatrix}, b = \begin{bmatrix} b_{1} \\ b_{2} \end{bmatrix}$$



We can try to pat the matrix in an upper triangular form. Gaussian Elimination Method.





$\begin{array}{c} \underbrace{\mathsf{Exi}}_{i} A = \begin{bmatrix} -1 \\ 1 \\ -1 \end{bmatrix}, \ b = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \\ \underbrace{\mathsf{E}_{i}}_{i} = \underbrace{\mathsf{E}_{i}}_{i} + 1 \\ \underbrace{\mathsf{Exi}}_{i} = \underbrace{\mathsf{Exi}$

Systems of First Order Equations Theory

Systems of First Order Equations

First Order Linear Systems of Differential Equations

$$x'_{1} = p_{11}(t)x_{1} + \dots + p_{1n}(t)x_{n} + g_{1}(t),$$

$$\vdots$$

$$x'_{n} = p_{n1}(t)x_{1} + \dots + p_{nn}(t)x_{n} + g_{n}(t)$$

vector notation
$$\mathbf{x}' = \mathbf{P}(t)\mathbf{x} + \mathbf{g}(t)$$

Homogeneous Case: Initial Value Problem

 $\mathbf{x}' = \mathbf{P}(t)\mathbf{x}$ (*) $\mathbf{x}(t_0) = \boldsymbol{\xi}$

Theorem

If the vector functions $\mathbf{x}^{(1)}$ and $\mathbf{x}^{(2)}$ are solutions of the system (\mathbf{x}), then the linear combination $c_1\mathbf{x}^{(1)} + c_2\mathbf{x}^{(2)}$ is also a solution for any constants c_1 and c_2 .

Theorem

If the vector functions $\mathbf{x}^{(1)}, \ldots, \mathbf{x}^{(n)}$ are linearly independent solutions of the system (*) for each point in the interval $\alpha < t < \beta$, then each solution $\mathbf{x} = \boldsymbol{\phi}(t)$ of the system (*) can be expressed as a linear combination of $\mathbf{x}^{(1)}, \ldots, \mathbf{x}^{(n)}$

$$\boldsymbol{\phi}(t) = c_1 \mathbf{x}^{(1)}(t) + \dots + c_n \mathbf{x}^{(n)}(t)$$

in exactly one way.

solutions $\mathbf{x}^{(1)}(t) = \begin{pmatrix} x_{11}(t) \\ x_{21}(t) \\ \vdots \\ x_{n1}(t) \end{pmatrix}, \quad \dots, \quad \mathbf{x}^{(k)}(t) = \begin{pmatrix} x_{1k}(t) \\ x_{2k}(t) \\ \vdots \\ x_{nk}(t) \end{pmatrix}$

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Solution Techniques (Constant Coefficients)

First Order Systems: Constant Coefficients

First Order Linear Systems with Constant Coefficients

$$\begin{pmatrix} x'_1 = p_{11} \cdot x_1 + \dots + p_{1n} \cdot x_n \\ \vdots \\ x'_n = p_{n1} \cdot x_1 + \dots + p_{nn} \cdot x_n \end{pmatrix} \longrightarrow \qquad \text{vector notation}$$

$$\mathbf{x}' = \mathbf{A}\mathbf{x}$$

Solution Candidates

$$\mathbf{x} = \boldsymbol{\xi} e^{rt} \longrightarrow r\boldsymbol{\xi} e^{rt} = \mathbf{A}\boldsymbol{\xi} e^{rt} \longrightarrow (\mathbf{A} - r\mathbf{I})\boldsymbol{\xi} = \mathbf{0}$$

 $\det(\mathbf{A} - r\mathbf{I}) = 0.$

Roots of the characteristic polynomial

$$\rho(\mathbf{r}) = \det(\mathbf{A} - r\mathbf{I})$$

Cases:

- real and distinct eigenvalues
- complex-valued eigenvalues
- **repeated** eigenvalues
- mixture of the cases above.

First Order Systems with Constant Coefficients (Real and Distinct Eigenvalues)

Homogeneous First Order System

 $\mathbf{x}' = \mathbf{A}\mathbf{x},$ $\mathbf{x}(t_0) = \mathbf{\xi}$

Summary of Solution Technique:

Find roots of the characteristic polynomial:

 $\rho(r) = \det(\mathbf{A} - r\mathbf{I})$

Are the roots distinct real?

Yes, then fundamental solutions are

$$\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)} e^{r_1 t}, \quad \dots, \quad \mathbf{x}^{(n)}(t) = \boldsymbol{\xi}^{(n)} e^{r_n t}$$

Use solution of the form

 $\mathbf{x} = c_1 \boldsymbol{\xi}^{(1)} e^{r_1 t} + \dots + c_n \boldsymbol{\xi}^{(n)} e^{r_n t}$

Find coefficients using initial conditions by solving $c_1 \mathbf{x}^{(1)}(t_0) + c_2 \mathbf{x}^{(2)}(t_0) + \cdots + c_n \mathbf{x}^{(n)}(t_0) = \mathbf{x}(t_0)$

No, then need to use another method.

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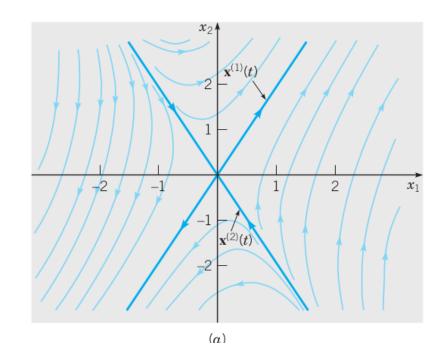
First Order Systems with Constant Coefficients (Real and Distinct Eigenvalues)

Ex: $\underline{x}' = \begin{bmatrix} 1 \\ 4 \end{bmatrix} \underbrace{x}$ Find the general solution. $p(r) = det(A - rI), A = \begin{bmatrix} 1 \\ 4 \end{bmatrix}$

$$\begin{split} \rho(r) &= \begin{vmatrix} 1 - r & 1 \\ 4 & | - r \end{vmatrix} = (1 - r)^{2} - 4 = r^{2} - 2r - 3 = 0 \\ r_{1} &= 3, r_{2} = -1, \quad X^{(1)} &= \int (1 - r)^{2} e^{3t}, \quad X^{(2)} &= \int (1 - r)^{2} e^{3t} e^{3t} \\ (A - r I) &= 0, \quad \begin{bmatrix} 1 - r & 1 \\ 4 & | - r \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 5 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ r_{1} = 3 : \begin{bmatrix} -3 & 1 \\ 4 & -2 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 5 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \in \int [1 - r]^{2} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \in \int [1 - r]^{2} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ = \begin{bmatrix} 1 - 2 \\ 5 \end{bmatrix}$$

 $Y_{j} = -1 : \begin{bmatrix} y \\ y \\ z \end{bmatrix} = \begin{bmatrix} y \\ z \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \notin \begin{bmatrix} y \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$

$$X^{(1)} = \begin{bmatrix} i \\ b \end{bmatrix} e^{3t}, X^{(1)} = \begin{bmatrix} i \\ -b \end{bmatrix} e^{-t}$$
$$X(t) = c_1 \begin{bmatrix} i \\ b \end{bmatrix} e^{5t} + c_2 \begin{bmatrix} i \\ -b \end{bmatrix} e^{-t}$$



First Order Systems with Constant Coefficients (Complex-valued Eigenvalues)

Homogeneous First Order System

 $\mathbf{x}' = \mathbf{A}\mathbf{x},$ $\mathbf{x}(t_0) = \mathbf{\xi}$

Summary of Solution Technique:

Find roots of the characteristic polynomial:

 $\rho(\mathbf{r}) = \det(\mathbf{A} - r\mathbf{I})$

Are the roots complex-valued?

Yes, then fundamental solutions are

$$\mathbf{u}(t) = e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t)$$
$$\mathbf{v}(t) = e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$$

Use solution of the form

 $\mathbf{x}(t) = c_1 \mathbf{u}(t) + c_2 \mathbf{v}(t)$

Find coefficients using initial conditions by solving $c_1 \mathbf{u}(t_0) + c_2 \mathbf{v}(t_0) = \mathbf{x}(t_0)$

Complex Eigenvalues $r = \lambda + i\mu$ $(\mathbf{A} - r_1 \mathbf{I})\boldsymbol{\xi}^{(1)} = \mathbf{0}$ $(\mathbf{A} - \overline{r}_1 \mathbf{I})\overline{\boldsymbol{\xi}}^{(1)} = \mathbf{0}$ $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)}e^{r_1 t}, \quad \mathbf{x}^{(2)}(t) = \overline{\boldsymbol{\xi}}^{(1)}e^{\overline{r}_1 t}$

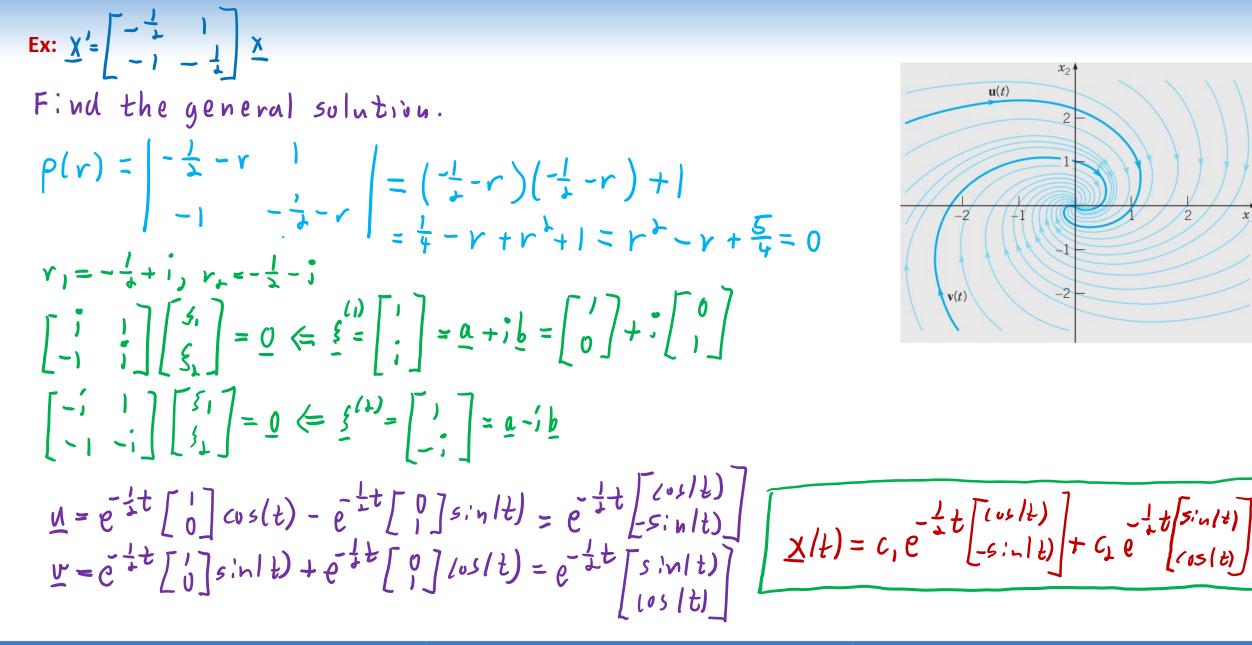
Euler Formula $e^{i\theta} = \cos(\theta) + i\sin(\theta)$

Formulation using Euler Formula
$$\boldsymbol{\xi}^{(1)} = \mathbf{a} + i\mathbf{b}$$

 $\mathbf{x}^{(1)}(t) = e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t) + ie^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$
 $\mathbf{x}^{(1)}(t) = \mathbf{u}(t) + i\mathbf{v}(t), \quad \mathbf{x}^{(2)}(t) = \mathbf{u}(t) - i\mathbf{v}(t)$
 $\mathbf{u}(t) = e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t)$
 $\mathbf{v}(t) = e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$
 $\checkmark \qquad \tilde{c}_1 \mathbf{x}^{(1)} + \tilde{c}_2 \mathbf{x}^{(2)} = c_1 \mathbf{u}(t) + c_2 \mathbf{v}(t)$

No, then need to use another method.

First Order Systems: Constant Coefficients (Complex-Valued Eigenvalues)



First Order Systems with Constant Coefficients (Repeated Eigenvalues, s = 2, k=1)

Homogeneous First Order System

 $\mathbf{x}' = \mathbf{A}\mathbf{x},$ $\mathbf{x}(t_0) = \mathbf{\xi}$

Summary of Solution Technique:

Find roots of the characteristic polynomial:

 $\rho(\mathbf{r}) = \det(\mathbf{A} - r\mathbf{I})$

Are the roots repeated?

Yes, then fundamental solutions are

$$\mathbf{x}^{(1)} = \mathbf{\xi}^{(1)} e^{rt}$$

 $\mathbf{x}^{(2)} = \eta^{(1)} e^{rt} + \mathbf{\xi}^{(1)} t e^{rt}$

Use solution of the form

 ${f x}(t)=c_1{f x}^{(1)}(t)+c_2{f x}^{(2)}(t)$

Find coefficients using initial conditions by solving $c_1 \mathbf{x}^{(1)}(t_0) + c_2 \mathbf{x}^{(2)}(t_0) = \mathbf{x}(t_0).$

Repeated Eigenvalues (s=2, k=1)

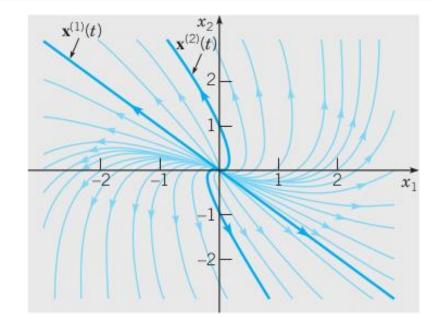
(i) find eigenvector for r, $\boldsymbol{\xi}^{(1)} \leftarrow (A - r\mathbf{I}) \boldsymbol{\xi} = 0$ (ii) construct the generalized eigenvector $\boldsymbol{\eta}^{(1)} \leftarrow (A - r\mathbf{I})^2 \boldsymbol{\eta} = 0$ (iii) this gives fundamental solutions $\mathbf{x}^{(1)} = \boldsymbol{\xi}^{(1)} e^{rt}$ $\mathbf{x}^{(2)} = \boldsymbol{\eta}^{(1)} e^{rt} + \boldsymbol{\xi}^{(1)} t e^{rt}$ $(A - r\mathbf{I}) \boldsymbol{\eta} = \boldsymbol{\xi}$ $(A - r\mathbf{I}) \boldsymbol{\xi} = 0$

s: algebraic multiplicity k: ge

k: geometric multiplicity

First Order Systems: Constant Coefficients (Repeated Eigenvalues)

Ex: $\chi' = \begin{bmatrix} 1 & -1 \\ 1 & 3 \end{bmatrix} \chi'$ Find the general solution. $p(r) = \begin{vmatrix} 1 - r & -1 \\ 1 & 3 - r \end{vmatrix} = (1 - r)(3 - r) + 1 = r^{2} - 4r + 4 = 0$ r=2, r=2, algebraic multiplicity s=2 $\begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 5 \\ 5 \end{bmatrix} = 0 = \underbrace{5} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad (A - rI) = 0$ $(A-rI)^{*}\eta = 0 \in (A-rI)n = \frac{3}{2},$ $\begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} N_1 \\ N_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{pmatrix} = \\ N = \end{bmatrix} \begin{bmatrix} -1 \\ -1 \end{bmatrix}$ $\underline{X}^{(1)} = \begin{bmatrix} -1 \\ -1 \end{bmatrix} e^{\lambda t}, \quad \underline{X}^{(\lambda)} = \underline{n} e^{\lambda t} + \underline{s} t e^{\lambda t} = \begin{bmatrix} -\frac{1}{2} + t \\ -\frac{1}{2} + t \end{bmatrix} e^{\lambda t}$





First Order Systems with Constant Coefficients (Summary)

Homogeneous First Order System

 $\mathbf{x}' = \mathbf{A}\mathbf{x},$ $\mathbf{x}(t_0) = \boldsymbol{\xi}$

Summary of Solution Technique:

Are the coefficients constant?

Yes, then find roots of the characteristic polynomial:

 $\rho(\mathbf{r}) = \det(\mathbf{A} - r\mathbf{I})$

Construct the **fundamental solution set** using the cases

 $\{\mathbf{x}^{(1)}(t), \mathbf{x}^{(2)}(t), \ldots, \mathbf{x}^{(n)}(t)\}$

Use solution of the form

 $\mathbf{x}(t) = c_1 \mathbf{x}^{(1)}(t) + c_2 \mathbf{x}^{(2)}(t) + \cdots + c_n \mathbf{x}^{(n)}(t).$

Find coefficients using initial conditions by solving $c_1 \mathbf{x}^{(1)}(t_0) + c_2 \mathbf{x}^{(2)}(t_0) + \cdots + c_n \mathbf{x}^{(n)}(t_0) = \mathbf{x}(t_0)$

No, then need to use another method.

Complex Eigenvalues $r = \lambda + i\mu$ $(\mathbf{A} - r_1 \mathbf{I})\boldsymbol{\xi}^{(1)} = \mathbf{0}$ $(\mathbf{A} - \overline{r}_1 \mathbf{I})\overline{\boldsymbol{\xi}}^{(1)} = \mathbf{0}$ $\mathbf{x}^{(1)}(t) = \boldsymbol{\xi}^{(1)}e^{r_1 t}, \qquad \mathbf{x}^{(2)}(t) = \overline{\boldsymbol{\xi}}^{(1)}e^{\overline{r}_1 t}$

Formulation using Euler Formula $\xi^{(1)} = \mathbf{a} + i\mathbf{b}$ $\mathbf{x}^{(1)}(t) = e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t) + ie^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$ $\mathbf{x}^{(1)}(t) = \mathbf{u}(t) + i\mathbf{v}(t), \quad \mathbf{x}^{(2)}(t) = \mathbf{u}(t) - i\mathbf{v}(t)$ $\mathbf{u}(t) = e^{\lambda t} (\mathbf{a} \cos \mu t - \mathbf{b} \sin \mu t)$ $\mathbf{v}(t) = e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$ $\mathbf{v}(t) = e^{\lambda t} (\mathbf{a} \sin \mu t + \mathbf{b} \cos \mu t)$

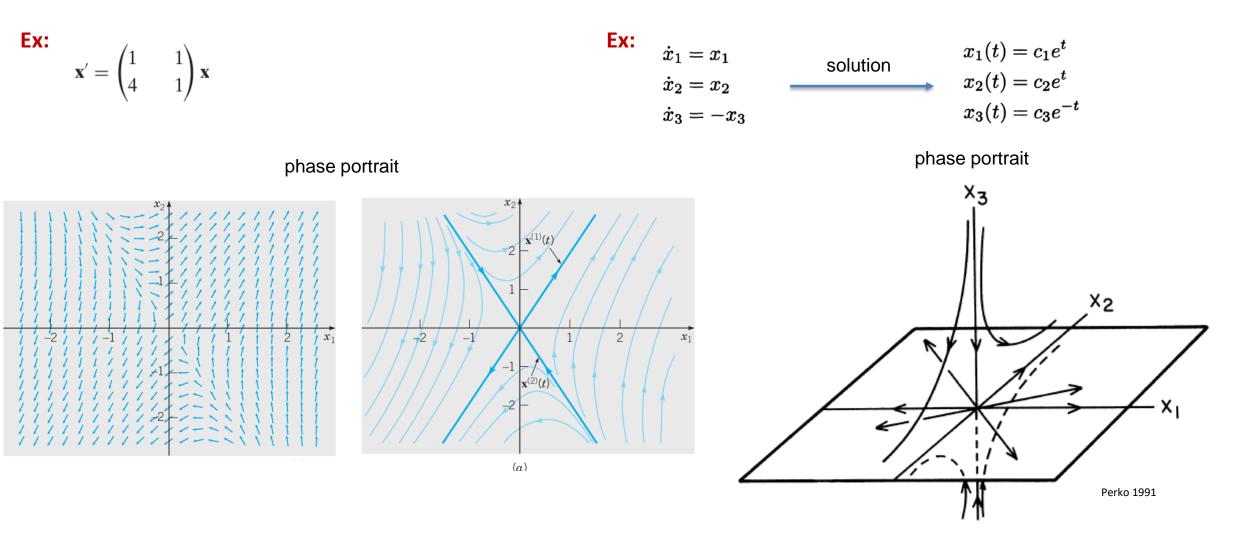
Repeated Eigenvalues (s=2,k=1)

(i) find eigenvector for r, $\xi^{(1)} \leftarrow (A - r\mathbf{I}) \xi = 0$ (ii) construct the generalized eigenvector $\eta^{(1)} \leftarrow (A - r\mathbf{I})^2 \eta = 0$ (iii) this gives fundamental solutions $\mathbf{x}^{(1)} = \xi^{(1)} e^{rt}$ $\mathbf{x}^{(2)} = \eta^{(1)} e^{rt} + \xi^{(1)} t e^{rt}$ s: algebraic multiplicity k: geometric multiplicity

Systems of First Order Equations Phase Portraits Dynamical Behaviors

Phase Portraits

A **phase portrait** shows representative trajectories and vector field for the evolution of the state $\mathbf{x}(t)$ of the differential equation.



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Differential Equations

2D Linear Systems: Phase Portraits

Linear Dynamical System

 $\dot{\mathbf{x}} = A\mathbf{x}$

Consider the following standard forms

$$\dot{\mathbf{x}} = B\mathbf{x}$$

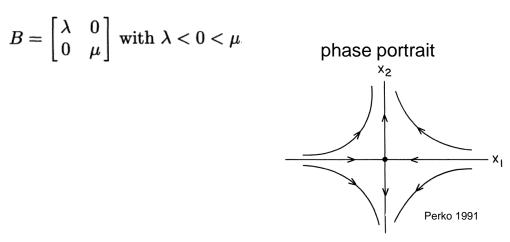
with

$$B = egin{bmatrix} \lambda & 0 \ 0 & \mu \end{bmatrix}, \quad B = egin{bmatrix} \lambda & 1 \ 0 & \lambda \end{bmatrix}, \quad ext{or} \quad B = egin{bmatrix} a & -b \ b & a \end{bmatrix}$$

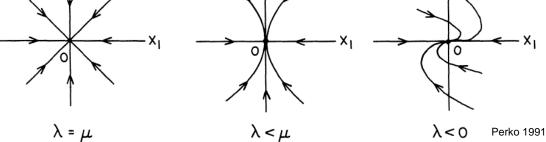
This has the solutions

$$\mathbf{x}(t) = \begin{bmatrix} e^{\lambda t} & 0\\ 0 & e^{\mu t} \end{bmatrix} \mathbf{x}_0, \ \mathbf{x}(t) = e^{\lambda t} \begin{bmatrix} 1 & t\\ 0 & 1 \end{bmatrix} \mathbf{x}_0, \ \mathbf{x}(t) = e^{at} \begin{bmatrix} \cos bt & -\sin bt\\ \sin bt & \cos bt \end{bmatrix} \mathbf{x}_0$$

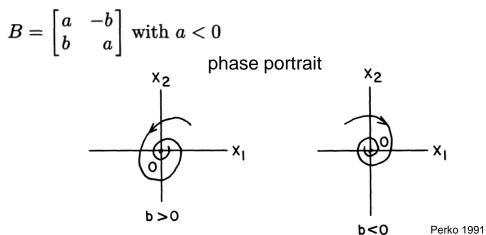
Case I:



Case II: $B = \begin{bmatrix} \lambda & 0 \\ 0 & \mu \end{bmatrix} \text{ with } \lambda \le \mu < 0 \text{ or } B = \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} \text{ with } \lambda < 0$ phase portrait x_{2} x_{2} x_{2} x_{2} x_{2} x_{3} x_{4} x_{5} x_{4} x_{5} x_{4} x_{5} x_{4} x_{5} x



Case III:



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2D Linear Systems: Phase Portraits

Linear Dynamical System

$$\dot{x} = Ax.$$
 $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

Linear Dynamic System

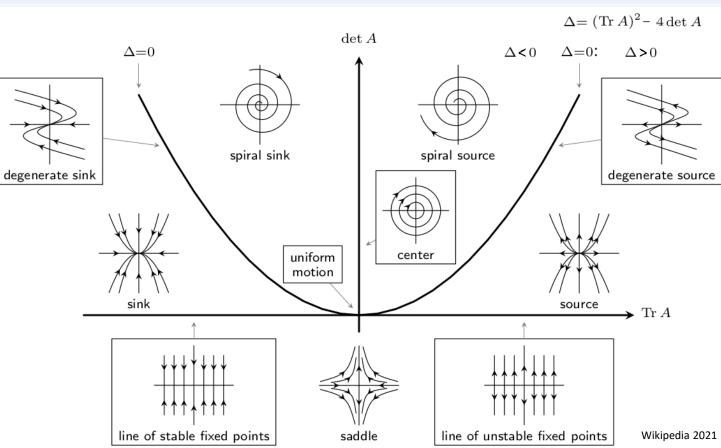
 $\begin{array}{rcl} \dot{x}_1 &=& ax_1 + bx_2 \\ \dot{x}_2 &=& cx_1 + dx_2 \end{array}$

Characterization

 $\det(A) = ad - bc$ $\operatorname{Tr}(A) = a + d$

$$p(r) = r^{2} - Tr(A)r + det(A).$$

$$r = \frac{Tr(A) \pm \sqrt{(T - TA)^{2} - 4 det(A)}}{2}$$



Linear System Behaviors in 2D

Conclusion

Enjoyed working with you all this quarter!

Thank You!