

Set 20: Systems of First Order Linear ODEs
Part 1

Kyle A. Gallivan
Department of Mathematics
Florida State University
Ordinary Differential Equations
Fall 2009

Systems of Linear ODEs

- $x_i : \mathbb{R} \rightarrow \mathbb{R}, i = 1, 2, 3$
- Suppose x'_i depends on $x_i, i = 1, 2, 3$ linearly

$$x'_i(t) = p_{i1}(t)x_1(t) + p_{i2}(t)x_2(t) + p_{i3}(t)x_3(t)$$

- System of 3 linear ODES defining 3 functions

$$x'_1(t) = p_{11}(t)x_1(t) + p_{12}(t)x_2(t) + p_{13}(t)x_3(t) + g_1(t)$$

$$x'_2(t) = p_{21}(t)x_1(t) + p_{22}(t)x_2(t) + p_{23}(t)x_3(t) + g_2(t)$$

$$x'_3(t) = p_{31}(t)x_1(t) + p_{32}(t)x_2(t) + p_{33}(t)x_3(t) + g_3(t)$$

Matrix Form

$$x'(t) = P(t)x(t) + g(t)$$

$$P(t) = \begin{pmatrix} p_{11}(t) & p_{12}(t) & p_{13}(t) \\ p_{21}(t) & p_{22}(t) & p_{23}(t) \\ p_{31}(t) & p_{32}(t) & p_{33}(t) \end{pmatrix}$$

$$x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} \quad x'(t) = \begin{pmatrix} x'_1(t) \\ x'_2(t) \\ x'_3(t) \end{pmatrix} \quad g(t) = \begin{pmatrix} g_1(t) \\ g_2(t) \\ g_3(t) \end{pmatrix}$$

Matrix Vector Multiplication

$$x'(t) = P(t)x(t) + g(t)$$

$$P(t)x(t) = \begin{pmatrix} p_{11}(t) \\ p_{21}(t) \\ p_{31}(t) \end{pmatrix} x_1(t) + \begin{pmatrix} p_{12}(t) \\ p_{22}(t) \\ p_{32}(t) \end{pmatrix} x_2(t) + \begin{pmatrix} p_{13}(t) \\ p_{23}(t) \\ p_{33}(t) \end{pmatrix} x_3(t)$$

Linear combination of columns of the matrix with coefficients given by the corresponding elements of the vector.

Existence and Uniqueness

Theorem 20.1 (Textbook page 359). *If $p_{ij}(t)$, $1 \leq i, j \leq n$, and $g_i(t)$, $1 \leq i \leq n$, are continuous on an open interval $\alpha < t < \beta$ then there exists a unique solution $x_1(t) = \phi_1(t), \dots, x_n(t) = \phi_n(t)$ to the initial value problem*

$$x'(t) = P(t)x(t) + g(t), \quad x(t_0) = x^{(0)}$$

for any $\alpha < t_0 < \beta$ and $x^{(0)} \in \mathbb{R}^n$, that exists on the entire open interval.

Higher Order Equations

- A single higher order linear equation can be transformed into a system of first order equations.
- $u''' + p(t)u'' + q(t)u' + r(t)u = f(t)$
- $x_1 = u, x_2 = x_1' = u', x_3 = x_2' = u''$
- System of dimension 3

$$x_1'(t) = x_2(t)$$

$$x_2'(t) = x_3(t)$$

$$x_3'(t) = -r(t)x_1(t) - q(t)x_2(t) - p(t)x_3(t) + f(t)$$

Higher Order Equations

$$x_1'(t) = x_2(t)$$

$$x_2'(t) = x_3(t)$$

$$x_3'(t) = -r(t)x_1(t) - q(t)x_2(t) - p(t)x_3(t) + f(t)$$

$$\begin{pmatrix} x_1'(t) \\ x_2'(t) \\ x_3'(t) \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -r(t) & -q(t) & -p(t) \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ f(t) \end{pmatrix}$$

$$x'(t) = P(t)x(t) + g(t)$$

Systems of ODEs

Given the ODE $u^{(n)} = F(t, u, u', \dots, u^{(n-1)})$, the variable definitions $x_1 = u, x_2 = x'_1 = u', x_3 = x'_2 = u'', \dots, x_n = x'_{n-1} = u^{(n-1)}$ convert the single nonlinear ODE of order n into a system of n nonlinear first order ODEs

$$x'_1 = x_2$$

$$x'_2 = x_3$$

$$\vdots$$

$$x'_{n-1} = x_n$$

$$x'_n = F(t, x_1, x_2, \dots, x_{n-1})$$

Systems of ODEs

The general form of a system of first order ODEs is

$$x'_1 = F_1(t, x_1, x_2, \dots, x_{n-1}, x_n)$$

$$x'_2 = F_2(t, x_1, x_2, \dots, x_{n-1}, x_n)$$

⋮

$$x'_{n-1} = F_{n-1}(t, x_1, x_2, \dots, x_{n-1}, x_n)$$

$$x'_n = F_n(t, x_1, x_2, \dots, x_{n-1}, x_n)$$

Linear Constant Coefficient First Order System

Suppose $P(t)$ is a constant matrix A and for simplicity take $n = 2$.

$$x'(t) = P(t)x(t) + g(t) = Ax(t) + g(t)$$

$$A = \begin{pmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{pmatrix}$$

$$x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} \quad x'(t) = \begin{pmatrix} x'_1(t) \\ x'_2(t) \end{pmatrix} \quad g(t) = \begin{pmatrix} g_1(t) \\ g_2(t) \end{pmatrix}$$

Homogeneous Problem

If $g(t) = 0 \in \mathbb{R}^k$ the problem is an homogeneous linear constant coefficient first order system.

$$x'(t) = Ax(t)$$

$$A = \begin{pmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{pmatrix}$$

$$x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} \quad x'(t) = \begin{pmatrix} x'_1(t) \\ x'_2(t) \end{pmatrix}$$

A Solution

Suppose we have a scalar r and a (constant) vector $v \in \mathbb{R}^2$ such that

$$Av = vr$$

i.e., the action of A on the vector v is equivalent to scaling by r .

r is an eigenvalue and v is an associated eigenvector of A .

A Solution

Consider:

$$x(t) = ve^{rt}$$

$$x(t) = \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = \begin{pmatrix} v_1e^{rt} \\ v_2e^{rt} \end{pmatrix}$$

$$x'(t) = \begin{pmatrix} x'_1(t) \\ x'_2(t) \end{pmatrix} = \begin{pmatrix} v_1re^{rt} \\ v_2re^{rt} \end{pmatrix} = x(t)r$$

$$x' = x(t)r = vre^{rt} = Ave^{rt} = Ax$$

So an eigenvalue, eigenvector pair solves the homogeneous problem.

General Solution

- one solution

$$x(t) = ve^{rt}$$

- general solution to the homogeneous problem has a fundamental set of solutions so that all solutions to the homogeneous system can be written, for $n = 2$,

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

- What is the fundamental set of solutions for

$$x' = Ax$$

General Solution

- Want the general solution to the homogeneous system

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

- Need eigenvectors and eigenvalues
- Need linear independence
- We proceed as with scalar homogeneous problems but use vector-valued functions as solutions.

General Solution

- Suppose $x(t)$ is a solution to $x' = Ax$ on the open interval $\alpha < t < \beta$
- For any $\alpha < t_0 < \beta$ $x(t)$ is the unique solution with the value $x(t_0)$.
- Find unique constants so that $x(t) = x^{(1)}(t)c_1 + x^{(2)}(t)c_2$ for two solutions $x^{(1)}(t)$ and $x^{(2)}(t)$

General Solution

First note that given two solutions $x^{(1)}(t)$ and $x^{(2)}(t)$ and two constants c_1 and c_2 we have

$$\begin{aligned}x(t) &= x^{(1)}(t)c_1 + x^{(2)}(t)c_2 \\x'(t) &= [x^{(1)}(t)]'c_1 + [x^{(2)}(t)]'c_2 \\&= Ax^{(1)}(t)c_1 + Ax^{(2)}(t)c_2 = A[x^{(1)}(t)c_1 + x^{(2)}(t)c_2] \\&= Ax(t)\end{aligned}$$

So $x(t) = x^{(1)}(t)c_1 + x^{(2)}(t)c_2$ is also a solution to the homogeneous system.

General Solution

For any solution $x(t)$ to have this form we must have for any $\alpha < t_0 < \beta$ $x(t)$ a unique pair of constants c_1 and c_2 :

$$x(t_0) = x^{(1)}(t_0)c_1 + x^{(2)}(t_0)c_2$$

$$\begin{pmatrix} x_1(t_0) \\ x_2(t_0) \end{pmatrix} = \begin{pmatrix} x_1^{(1)}(t_0) & x_1^{(2)}(t_0) \\ x_2^{(1)}(t_0) & x_2^{(2)}(t_0) \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$

$$X(t_0)c = x(t_0)$$

Linear Algebra

Lemma 20.2. *An $n \times n$ linear system of equations $Av = b$ has a unique solution $v = A^{-1}b$ for any vector $b \in \mathbb{R}^n$ if any of the following equivalent conditions are true:*

- *All of the eigenvalues of A are nonzero.*
- $\det(A) \neq 0$
- *The linear system $Av = 0$ has a unique solution $v = 0$, i.e., the zero vector in $b \in \mathbb{R}^n$*
- *The columns of A are n linearly independent vectors.*

Note that if $Av = 0$ has a solution $v \neq 0$ then the columns of A are called linearly dependent.

General Solution

Definition 20.1. $x^{(1)}(t)$ and $x^{(2)}(t)$ are a fundamental set of solutions to the homogeneous system of ODEs $x' = Ax$ with $n = 2$ and $x(t) = x^{(1)}(t)c_1 + x^{(2)}(t)c_2$ is its general solution if

- At any point, t_0 , in $\alpha < t < \beta$ the determinant $\det(X(t_0)) \neq 0$ where

$$X(t_0) = \begin{bmatrix} x^{(1)}(t_0) & x^{(2)}(t_0) \end{bmatrix} \in \mathbb{R}^{2 \times 2}$$

- or equivalently the vectors $x^{(1)}(t_0)$ and $x^{(2)}(t_0)$ are linearly independent.

Note $\det(X(t)) = W[x^{(1)}(t), x^{(2)}(t)]$ is called the Wronskian of the set of vectors.

Fundamental Theorem

Theorem 20.3 (Textbook p. 387). *If the vector functions $x^{(1)}(t), \dots, x^{(n)}(t)$ are linearly independent vectors at each point in an open interval $\alpha < t < \beta$ and each solve*

$$x' = P(t)x$$

then any solution $x(t)$ can be expressed as a unique linear combination of the $x^{(i)}(t)$. The vector functions $x^{(1)}(t), \dots, x^{(n)}(t)$ are a fundamental set of solutions for the homogeneous ODE.

Fundamental Theorem

Theorem 20.4 (Textbook p. 387). *If the vector functions $x^{(1)}(t), \dots, x^{(n)}(t)$ solve in an open interval $\alpha < t < \beta$ the ODE*

$$x' = P(t)x$$

then the Wronskian $W[x^{(1)}(t), \dots, x^{(n)}(t)]$ either is identically 0 or always nonzero on the entire interval.

In other words, the solutions are linearly independent vectors at all points in the interval if they are linearly independent vectors at any single point.

Determinant Test $n = 2$

Definition 20.2. The determinant of the 2×2 matrix A is given by

$$\det(A) = \begin{vmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{vmatrix} = \alpha_{11}\alpha_{22} - \alpha_{21}\alpha_{12}$$

$$\det \begin{pmatrix} 1 & 1 \\ 2 & -3 \end{pmatrix} = \begin{vmatrix} 1 & 1 \\ 2 & -3 \end{vmatrix} = -3 - 2 = -5$$

$$\det \begin{pmatrix} 1 & -2 \\ 2 & -4 \end{pmatrix} = \begin{vmatrix} 1 & -2 \\ 2 & -4 \end{vmatrix} = -4 + 4 = 0$$

Special Case for Two Vectors

- Note that two vectors, z_1 and z_2 are linearly dependent if and only if one is a scalar multiple of the other, i.e., $z_1 = \alpha z_2$.
- For $n = 2$ this means the matrix with z_1 and z_2 as its columns has a determinant 0.

General Solution $n = 2$

Suppose we have two eigenvalue/eigenvector pairs, $(v^{(1)}, r_1)$ and $(v^{(2)}, r_2)$ we therefore have

$$x^{(1)}(t) = v^{(1)} e^{r_1 t} = \begin{pmatrix} v_1^{(1)} e^{r_1 t} \\ v_2^{(1)} e^{r_1 t} \end{pmatrix} \quad \text{and} \quad x^{(2)}(t) = v^{(2)} e^{r_2 t} = \begin{pmatrix} v_1^{(2)} e^{r_2 t} \\ v_2^{(2)} e^{r_2 t} \end{pmatrix}$$

$$\begin{aligned} \det(X(t)) &= \begin{vmatrix} v_1^{(1)} e^{r_1 t} & v_1^{(2)} e^{r_2 t} \\ v_2^{(1)} e^{r_1 t} & v_2^{(2)} e^{r_2 t} \end{vmatrix} = v_1^{(1)} v_2^{(2)} e^{(r_1+r_2)t} - v_1^{(2)} v_2^{(1)} e^{(r_1+r_2)t} \\ &= [v_1^{(1)} v_2^{(2)} - v_1^{(2)} v_2^{(1)}] e^{(r_1+r_2)t} = \det(X(0)) e^{(r_1+r_2)t} \end{aligned}$$

So $\det(X(t)) \neq 0 \leftrightarrow \det(X(0)) \neq 0$ or equivalently, $v^{(1)}$ and $v^{(2)}$ are linearly independent.

Independence

Theorem 20.5. *If $\lambda_1 \neq \lambda_2 \neq \dots \neq \lambda_n$ are eigenvalues of $A \in \mathbb{R}^{n \times n}$ with eigenvectors $v^{(1)}, v^{(2)}, \dots, v^{(n)}$, respectively, then*

- *the eigenvectors are linearly independent;*
- *the matrix $V \in \mathbb{R}^{n \times n}$ whose i -th column is $v^{(i)}$ is nonsingular, i.e., V^{-1} exists uniquely;*
- *The system of linear equations $Vc = b$ has a unique solution $c = V^{-1}b$ for any $b \in \mathbb{R}^{n \times n}$.*

Fundamental Theorem

Theorem 20.6. *The homogeneous linear constant coefficient system of first order ODEs*

$$x' = Ax$$

has a fundamental set of solutions

$$x^{(i)}(t) = v^{(i)} e^{r_i t}, \quad 1 \leq i \leq n$$

and general solution

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

if $(v^{(1)}, r_1), \dots, (v^{(n)}, r_n)$ are eigenvector/eigenvalue pairs with $r_1 \neq r_2 \neq \cdots \neq r_n$.

Independent or Dependent

Tests for linear independence or dependence of a set of vectors $x^{(i)} \in \mathbb{R}^n$ for $i = 1, \dots, n$

- Solve $Xc = 0$ for c .
 - If there is a nonzero solution then the vectors are dependent. In this case there will be more than one $c \neq 0$.
 - If $c = 0$ is the only solution then the vectors are independent.
- Compute the determinant of X , $\det(X)$.
 - If $\det(X) = 0$ then the vectors are dependent.
 - If $\det(X) \neq 0$ then the vectors are independent.
- Check the eigenvalues of X .
 - If at least one eigenvalue is 0 then the vectors are dependent.
 - If all eigenvalues are nonzero then the vectors are independent.

Determinant Test

Definition 20.3. The determinant of the 2×2 matrix A is given by

$$\det(A) = \begin{vmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{vmatrix} = \alpha_{11}\alpha_{22} - \alpha_{21}\alpha_{12}$$

Example

$$\det \begin{pmatrix} 1 & 1 \\ 2 & -3 \end{pmatrix} = \begin{vmatrix} 1 & 1 \\ 2 & -3 \end{vmatrix} = -3 - 2 = -5$$

$$\det \begin{pmatrix} 1 & -2 \\ 2 & -4 \end{pmatrix} = \begin{vmatrix} 1 & -2 \\ 2 & -4 \end{vmatrix} = -4 + 4 = 0$$

Note it is also easy to see that one column is or is not a scalar multiple of the other.

Eigenvalues and Eigenvector

Definition 20.4. If A is an $n \times n$ matrix then the nonzero vector x and scalar λ are an eigenvector/eigenvalue pair if

$$Ax = x\lambda$$

We have

$$Ax = x\lambda \rightarrow (A - \lambda I)x = 0$$

$$\therefore \det(A - \lambda I) = 0$$

Eigenvalues and Eigenvector

- $\det(A - \lambda I)$ is a polynomial of degree n in the variable λ
- The eigenvalues are the roots of the polynomial.
- Any nonzero vector that solves, the system of equations

$$(A - \lambda I)x = 0$$

is an eigenvector associated with λ .

- If x is an eigenvector associated with eigenvalue λ then so is αx for any scalar α .
- If x_1 and $x_2 \neq \alpha x_1$ are eigenvectors associated with eigenvalue λ then so is $\alpha_1 x_1 + \alpha_2 x_2$.

Example

Example 4 Textbook page 379

$$A = \begin{pmatrix} 3 & -1 \\ 4 & -2 \end{pmatrix}$$

$$\det \begin{pmatrix} 3 - \lambda & -1 \\ 4 & -2 - \lambda \end{pmatrix} = -(3 - \lambda)(2 + \lambda) + 4$$

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

$$\therefore \lambda_1 = 2 \quad \text{and} \quad \lambda_2 = -1$$

Example

Example 4 Textbook page 379

$$A - 2I = \begin{pmatrix} 1 & -1 \\ 4 & -4 \end{pmatrix}$$

add -4 times row 1 to row 2

$$\left(\begin{array}{cc|c} 1 & -1 & 0 \\ 4 & -4 & 0 \end{array} \right) \rightarrow \left(\begin{array}{cc|c} 1 & -1 & 0 \\ 0 & 0 & 0 \end{array} \right) \rightarrow x_1 = x_2$$

So any vector with $x_1 = x_2$ with x_2 an arbitrary value is an eigenvector for $\lambda_1 = 2$.

Example

Example 4 Textbook page 379

$$A + I = \begin{pmatrix} 4 & -1 \\ 4 & -1 \end{pmatrix}$$

add row 1 to row 2

$$\left(\begin{array}{cc|c} 4 & -1 & 0 \\ 4 & -1 & 0 \end{array} \right) \rightarrow \left(\begin{array}{cc|c} 4 & -1 & 0 \\ 0 & 0 & 0 \end{array} \right) \rightarrow 4x_1 = x_2$$

So any vector with $x_1 = 0.25x_2$ with x_2 an arbitrary value is an eigenvector for $\lambda_1 = -1$.

Independence

Note that

$$\lambda_1 = 2, \quad x^{(1)} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \text{and} \quad \lambda_2 = -1, \quad x^{(2)} = \begin{pmatrix} 1 \\ 4 \end{pmatrix}$$

$x^{(1)}$ and $x^{(2)}$ are linearly independent

This is consistent with Theorem 20.5.

Initial Value Problem for the Example

Suppose we impose the initial conditions $x_1(0) = 1$ and $x_2(0) = 1$

$$x(t) = x^{(1)}(t)c_1 + x^{(2)}(t)c_2 = \begin{pmatrix} e^{2t} & e^{-t} \\ e^{2t} & 4e^{-t} \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$

$$\begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 4 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}$$

$$c_1 = 1, \quad c_2 = 0$$

$$x(t) = x^{(1)}(t) = \begin{pmatrix} e^{2t} \\ e^{2t} \end{pmatrix} \text{ solves the IVP}$$

Problems with $n \geq 3$

- For this class exams will only cover ODE systems with $n = 2$.
- The textbook has discussions and examples of determining eigenvalues/eigenvectors and solving linear systems with $n = 3$ and $n = 4$. Read Sections 7.2 and 7.3.
- The rest of this set gives some examples.
- The homework will include other problems with $n = 3$.

Determinant Test

Definition 20.5. The determinant, $\det(A)$, of a 3×3 matrix A is given by:

$$\det(A) = \begin{vmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} \\ \alpha_{21} & \alpha_{22} & \alpha_{23} \\ \alpha_{31} & \alpha_{32} & \alpha_{33} \end{vmatrix}$$

$$\begin{aligned} &= (-1)^{i+1} \det(A_{i1})\alpha_{i1} + (-1)^{i+2} \det(A_{i2})\alpha_{i2} + (-1)^{i+3} \det(A_{i3})\alpha_{i3} \\ &= (-1)^{1+j} \det(A_{1j})\alpha_{1j} + (-1)^{2+j} \det(A_{2j})\alpha_{2j} + (-1)^{3+j} \det(A_{3j})\alpha_{3j} \end{aligned}$$

where A_{ij} is the 2×2 matrix resulting from removing row i and column j from A .

Example

Use row 1:

$$\begin{aligned}\det(A) &= \begin{vmatrix} 1 & -2 & 3 \\ -1 & 2 & -2 \\ 2 & -1 & -1 \end{vmatrix} \\ &= (-1)^{1+1}(1) \begin{vmatrix} 2 & -2 \\ -1 & -1 \end{vmatrix} + (-1)^{1+2}(-2) \begin{vmatrix} -1 & -2 \\ 2 & -1 \end{vmatrix} + (-1)^{1+3}(3) \begin{vmatrix} -1 & 2 \\ 2 & -1 \end{vmatrix} \\ &= 0 + 10 - 9 = 1 \neq 0\end{aligned}$$

Example

Use column 1:

$$\begin{aligned}\det(A) &= \begin{vmatrix} 1 & -2 & 3 \\ -1 & 2 & -2 \\ 2 & -1 & -1 \end{vmatrix} \\ &= (-1)^{1+1}(1) \begin{vmatrix} 2 & -2 \\ -1 & -1 \end{vmatrix} + (-1)^{2+1}(-1) \begin{vmatrix} -2 & 3 \\ -1 & -1 \end{vmatrix} + (-1)^{3+1}(2) \begin{vmatrix} -2 & 3 \\ 2 & -2 \end{vmatrix} \\ &= 0 + 5 - 4 = 1 \neq 0\end{aligned}$$

Example

Use row 1:

$$\begin{aligned}\det(A) &= \begin{vmatrix} 1 & -2 & 3 \\ -1 & 1 & -2 \\ 2 & -1 & 3 \end{vmatrix} \\ &= (-1)^{1+1}(1) \begin{vmatrix} -2 & 3 \\ -1 & -2 \end{vmatrix} + (-1)^{1+2}(-2) \begin{vmatrix} -1 & -2 \\ 2 & 3 \end{vmatrix} + (-1)^{1+3}(3) \begin{vmatrix} -1 & 1 \\ 2 & -1 \end{vmatrix} \\ &= (3 - 2) + 2(-3 + 4) + 3(1 - 2) \\ &= 1 + 2 - 3 = 0\end{aligned}$$

Example

Find the eigenvalues.

$$A = \begin{pmatrix} 3 & 2 & 2 \\ 1 & 4 & 1 \\ -2 & -4 & -1 \end{pmatrix} \quad \det \begin{pmatrix} 3 - \lambda & 2 & 2 \\ 1 & 4 - \lambda & 1 \\ -2 & -4 & -1 - \lambda \end{pmatrix}$$

Example

$$\det(A - \lambda I) = (-1)^{1+1}(3 - \lambda) \begin{vmatrix} 4 - \lambda & 1 \\ -4 & -1 - \lambda \end{vmatrix} \\ + (-1)^{1+2}(2) \begin{vmatrix} 1 & 1 \\ -2 & -1 - \lambda \end{vmatrix} + (-1)^{1+3}(2) \begin{vmatrix} 1 & 4 - \lambda \\ -2 & -4 \end{vmatrix}$$

$$= (3 - \lambda)(\lambda^2 - 3\lambda) + 2(\lambda - 1) + 2(-2\lambda + 4)$$

$$= -\lambda^3 + 6\lambda^2 - 11\lambda + 6$$

$$\lambda_1 = 1, \quad \lambda_2 = 2, \quad \lambda_3 = 3$$

Example

Let $n = 3$ and consider

$$x^{(1)} = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}, \quad x^{(2)} = \begin{pmatrix} -2 \\ 1 \\ -1 \end{pmatrix}, \quad x^{(3)} = \begin{pmatrix} 3 \\ -2 \\ -1 \end{pmatrix}$$

Solve the system

$$\begin{pmatrix} 1 & -2 & 3 \\ -1 & 1 & -2 \\ 2 & -1 & -1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Example

Use row combinations to transform system of equations (this also works for nonzero righthand side vectors):

add row 1 to row 2; add -2 times row 1 to row 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ -1 & 1 & -2 & | & 0 \\ 2 & -1 & -1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & -1 & 1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix}$$

multiply row 2 by -1

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & -1 & 1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix}$$

Example

add 3 times row 2 to row 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 0 & -4 & | & 0 \end{pmatrix}$$

divide row 3 by -4

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 0 & -4 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 0 & 1 & | & 0 \end{pmatrix}$$

Example

$$\begin{pmatrix} 1 & -2 & 3 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$c_1 - 2c_2 + 3c_3 = 0$$

$$c_2 - c_3 = 0$$

$$c_3 = 0$$

$c_1 = c_2 = c_3 = 0$ is unique solution so the vectors are independent.

Example

Let $n = 3$ and consider

$$x^{(1)} = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}, \quad x^{(2)} = \begin{pmatrix} -2 \\ 1 \\ -1 \end{pmatrix}, \quad x^{(3)} = \begin{pmatrix} 3 \\ -2 \\ 3 \end{pmatrix}$$

Solve the system

$$\begin{pmatrix} 1 & -2 & 3 \\ -1 & 1 & -2 \\ 2 & -1 & 3 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Example

due to the position modified in the matrix the same row combinations transform system of equations:

add row 1 to row 2; add -2 times row 1 to row 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ -1 & 1 & -2 & | & 0 \\ 2 & -1 & 3 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & -1 & 1 & | & 0 \\ 0 & 3 & -3 & | & 0 \end{pmatrix}$$

multiply row 2 by -1

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & -1 & 1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 3 & -3 & | & 0 \end{pmatrix}$$

Example

add 3 times row 2 to row 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 3 & -3 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{pmatrix}$$

Example

$$\begin{pmatrix} 1 & -2 & 3 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$c_1 - 2c_2 + 3c_3 = 0$$

$$c_2 - c_3 = 0$$

$$0 = 0$$

c_3 is arbitrary and then c_1 and c_2 follow.

So $c_1 = -1$, $c_2 = 1$, and $c_3 = 1$ is a nonzero solution. The vectors are dependent.

Example

$$x^{(1)} = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}, \quad x^{(2)} = \begin{pmatrix} -2 \\ 2 \\ -1 \end{pmatrix}, \quad x^{(3)} = \begin{pmatrix} 3 \\ -2 \\ -1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -2 & 3 \\ -1 & 2 & -2 \\ 2 & -1 & -1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Example

add row 1 to row 2; add -2 times row 1 to row 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ -1 & 2 & -2 & | & 0 \\ 2 & -1 & -1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 0 & 1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix}$$

swap rows 2 and 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 0 & 1 & | & 0 \\ 0 & 3 & -7 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 3 & -7 & | & 0 \\ 0 & 0 & 1 & | & 0 \end{pmatrix}$$

Example

Divide row 2 by 3

$$\begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 3 & -7 & | & 0 \\ 0 & 0 & 1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -2 & 3 & | & 0 \\ 0 & 1 & -7/3 & | & 0 \\ 0 & 0 & 1 & | & 0 \end{pmatrix}$$

$$c_1 - 2c_2 + 3c_3$$

$$c_2 - \frac{7}{3}c_3 = 0$$

$$c_3 = 0$$

So $c_1 = c_2 = c_3 = 0$ is the unique solution.