



Overview

- Orthogonal projectors
- Gaussian elimination
- Row echelon reduction
- Matrix rank from row echelon reduction
- LU -factorization

- Consider the linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ with $\mathbf{A} \in \mathbb{R}^{m \times n}$, $\mathbf{x} \in \mathbb{R}^n$, $\mathbf{b} \in \mathbb{R}^m$. Orthogonal projectors and knowledge of the four fundamental matrix subspaces allows us to succinctly express whether there exist no solutions, a single solution or an infinite number of solutions:
 - Consider the factorization $\mathbf{Q}\mathbf{R} = \mathbf{A}$, the orthogonal projector $\mathbf{P} = \mathbf{Q}\mathbf{Q}^T$, and the complementary orthogonal projector $\mathbf{I} - \mathbf{P}$
 - If $\|(\mathbf{I} - \mathbf{P})\mathbf{b}\| \neq 0$, then \mathbf{b} has a component outside the column space of \mathbf{A} , and $\mathbf{A}\mathbf{x} = \mathbf{b}$ has no solution
 - If $\|(\mathbf{I} - \mathbf{P})\mathbf{b}\| = 0$, then $\mathbf{b} \in C(\mathbf{Q}) = C(\mathbf{A})$ and the system has at least one solution
 - If $N(\mathbf{A}) = \{\mathbf{0}\}$ (null space only contains the zero vector, i.e., null space of dimension 0) the system has a unique solution
 - If $\dim N(\mathbf{A}) = n - r > 0$, then a vector $\mathbf{y} \in N(\mathbf{A})$ in the null space is written as

$$\mathbf{y} = c_1\mathbf{z}_1 + \dots + c_{n-r}\mathbf{z}_{n-r}$$

and if \mathbf{x} is a solution of $\mathbf{A}\mathbf{x} = \mathbf{b}$, so is $\mathbf{x} + \mathbf{y}$, since

$$\mathbf{A}(\mathbf{x} + \mathbf{y}) = \mathbf{A}\mathbf{x} + c_1\mathbf{A}\mathbf{z}_1 + \dots + c_{n-r}\mathbf{A}\mathbf{z}_{n-r} = \mathbf{b} + \mathbf{0} + \dots + \mathbf{0} = \mathbf{b}$$

The linear system has an $(n - r)$ -parameter family of solutions



- Idea: make one fewer unknown appear in each equation. Use first equation to eliminate x_1 in equations 2,3

$$\begin{cases} x_1 + 2x_2 - x_3 = 2 \\ 2x_1 - x_2 + x_3 = 2 \\ 3x_1 - x_2 - x_3 = 1 \end{cases} \Rightarrow \begin{cases} x_1 + 2x_2 - x_3 = 2 \\ -5x_2 + 3x_3 = -2 \\ -7x_2 + 2x_3 = -5 \end{cases}$$

- Use second equation to eliminate x_2 in equation 3

$$\begin{cases} x_1 + 2x_2 - x_3 = 2 \\ -5x_2 + 3x_3 = -2 \\ -7x_2 + 2x_3 = -5 \end{cases} \Rightarrow \begin{cases} x_1 + 2x_2 - x_3 = 2 \\ -5x_2 + 3x_3 = -2 \\ -\frac{11}{5}x_3 = -\frac{11}{5} \end{cases}$$

- Start finding components from last to first to obtain $x_3 = 1$, $x_2 = 1$, $x_1 = 1$

- Explicitly writing the unknowns x_1, x_2, x_3 is not necessary. Introduce the “bordered” matrix

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & -1 & 1 & 2 \\ 3 & -1 & -1 & 1 \end{bmatrix}$$

- Define allowed operations:
 - multiply a row by a non-zero scalar
 - add a row to another
- Bordered matrices obtained by the allowed operations are said to be *similar*, in that the solution of the linear system stays the same

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & -1 & 1 & 2 \\ 3 & -1 & -1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & -7 & 2 & -5 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & -\frac{11}{5} & -\frac{11}{5} \end{bmatrix}$$



- To find solution, use allowed operations to make an identity matrix appear

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & -\frac{11}{5} & -\frac{11}{5} \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & 1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

- The above constitute “Gaussian elimination”

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & -1 & 1 & 2 \\ 3 & -1 & -1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & -7 & 2 & -5 \end{bmatrix}$$

```
∴ A=[1. 2 -1 2; 2 -1 1 2; 3 -1 -1 1]; A[2,:]=A[2,:]-2*A[1,:]; A[3,:]=A[3,:]-3*A[1,:];
```

```
∴ A
```

```
∴ A[3,:]=A[3,:]- (7/5)*A[2,:]; A
```

- To find solution, use allowed operations to make an identity matrix appear

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & -\frac{11}{5} & -\frac{11}{5} \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & 1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

- The above constitute “Gaussian elimination”

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & -1 & 1 & 2 \\ 3 & -1 & -1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & -7 & 2 & -5 \end{bmatrix}$$

```
∴ A=[1. 2 -1 2; 2 -1 1 2; 3 -1 -1 1]; A[2,:]=A[2,:]-2*A[1,:]; A[3,:]=A[3,:]-3*A[1,:];
```

```
∴ A
```

$$\begin{bmatrix} 1.0 & 2.0 & -1.0 & 2.0 \\ 0.0 & -5.0 & 3.0 & -2.0 \\ 0.0 & -7.0 & 2.0 & -5.0 \end{bmatrix} \quad (1)$$

```
∴ A[3,:]=A[3,:]-(-7/5)*A[2,:]; A
```

- To find solution, use allowed operations to make an identity matrix appear

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & -\frac{11}{5} & -\frac{11}{5} \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & 0 & 1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

- The above constitute “Gaussian elimination”

$$\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & -1 & 1 & 2 \\ 3 & -1 & -1 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 2 \\ 0 & -5 & 3 & -2 \\ 0 & -7 & 2 & -5 \end{bmatrix}$$

```
∴ A=[1. 2 -1 2; 2 -1 1 2; 3 -1 -1 1]; A[2,:]=A[2,:]-2*A[1,:]; A[3,:]=A[3,:]-3*A[1,:];
```

```
∴ A
```

$$\begin{bmatrix} 1.0 & 2.0 & -1.0 & 2.0 \\ 0.0 & -5.0 & 3.0 & -2.0 \\ 0.0 & -7.0 & 2.0 & -5.0 \end{bmatrix} \quad (2)$$

```
∴ A[3,:]=A[3,:]- (7/5)*A[2,:]; A
```

$$\begin{bmatrix} 1.0 & 2.0 & -1.0 & 2.0 \\ 0.0 & -5.0 & 3.0 & -2.0 \\ 0.0 & 0.0 & -2.1999999999999993 & -2.2 \end{bmatrix} \quad (3)$$

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix} \in \mathbb{R}^{3 \times 3}, \mathbf{b} = \begin{bmatrix} 3 \\ 1 \\ 3 \end{bmatrix} \in \mathbb{R}^3, \mathbf{c} = \begin{bmatrix} 3 \\ 1 \\ 4 \end{bmatrix} \in \mathbb{R}^3.$$

$$[\mathbf{A} \ \mathbf{b}] = \left[\begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 1 & 2 & 3 & 3 \end{array} \right] \sim [\mathbf{A}_1 \ \mathbf{b}_1] = \left[\begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right] \Leftrightarrow \begin{cases} x_1 + 2x_2 + 3x_3 = 3 \\ x_2 + x_3 = 1 \\ 0 = 0 \end{cases},$$

$$[\mathbf{A} \ \mathbf{c}] = \left[\begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \end{array} \right] \sim [\mathbf{A}_1 \ \mathbf{c}_1] = \left[\begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{array} \right] \Leftrightarrow \begin{cases} x_1 + 2x_2 + 3x_3 = 3 \\ x_2 + x_3 = 1 \\ 0 = 1 \end{cases}.$$



- Use similarity transformations to *reduced row echelon form*:
 - All zero rows are below non-zero rows
 - First non-zero entry on a row is called the *leading entry*
 - In each non-zero row, the leading entry is to the left of lower leading entries
 - Each leading entry equals 1 and is the only non-zero entry in its column
- *Row echelon form*:
 - Allow additional non-zero elements in a column, above the leading entry
- After carrying out rref on bordered matrix $[\mathbf{A} \mid \mathbf{b}]$, if:
 - there is a row with $[0 \ 0 \ \dots \ 0 \mid 1] \Rightarrow$ No solutions
 - the result is of form $[\mathbf{I} \mid \mathbf{c}] \Rightarrow$ Unique solution
 - there is no row of form $[0 \ 0 \ \dots \ 0 \mid 1]$, and there is a row of all zeros $[0 \ 0 \ \dots \ 0 \mid 0] \Rightarrow$ Infinitely many solutions

Examples

$$\left[\begin{array}{cccc|c} 1 & -1 & 0 & 1 & 2 \\ 0 & 0 & 1 & -1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow \text{Infinitely many solutions}$$

$$\left[\begin{array}{ccc|c} 1 & 2 & -4 & -4 \\ 0 & 3 & -1 & 2 \\ 0 & 0 & 8 & 8 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 0 & 0 & -2 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{array} \right] \Rightarrow \text{Unique solution}$$

- Recall the basic operation in row echelon reduction: constructing a linear combination of rows to form zeros beneath the main diagonal, e.g.

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ a_{31} & a_{32} & \dots & a_{3m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mm} \end{pmatrix} \sim \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ 0 & a_{22} - \frac{a_{21}}{a_{11}}a_{12} & \dots & a_{2m} - \frac{a_{21}}{a_{11}}a_{1m} \\ 0 & a_{32} - \frac{a_{31}}{a_{11}}a_{12} & \dots & a_{3m} - \frac{a_{31}}{a_{11}}a_{1m} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & a_{m2} - \frac{a_{m1}}{a_{11}}a_{12} & \dots & a_{mm} - \frac{a_{m1}}{a_{11}}a_{1m} \end{pmatrix}$$

- This can be stated as a matrix multiplication operation, with $l_{i1} = a_{i1}/a_{11}$

$$\begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ -l_{21} & 1 & 0 & \dots & 0 \\ -l_{31} & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -l_{m1} & 0 & 0 & \dots & 1 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ a_{31} & a_{32} & \dots & a_{3m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mm} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ 0 & a_{22} - l_{21}a_{12} & \dots & a_{2m} - l_{21}a_{1m} \\ 0 & a_{32} - l_{31}a_{12} & \dots & a_{3m} - l_{31}a_{1m} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & a_{m2} - l_{m1}a_{12} & \dots & a_{mm} - l_{m1}a_{1m} \end{pmatrix}$$



- Denote a permutation by

$$\sigma = \begin{pmatrix} 1 & 2 & \dots & m \\ i_1 & i_2 & \dots & i_m \end{pmatrix}$$

with $i_1, \dots, i_m \in \{1, \dots, m\}$, $i_j \neq i_k$ for $j \neq k$

- The sign of a permutation, $\nu(\sigma)$ is the number of pair swaps needed to obtain the permutation starting from the identity permutation

$$\begin{pmatrix} 1 & 2 & \dots & m \\ 1 & 2 & \dots & m \end{pmatrix}$$

- A permutation can be specified by a permutation matrix \mathbf{P} obtained from \mathbf{I} by swapping rows and columns $k \leftrightarrow i_k$

Definition. *The matrix*

$$\mathbf{L}_k = \begin{pmatrix} 1 & \dots & 0 & \dots & 1 \\ 0 & \ddots & 0 & \dots & 0 \\ 0 & \dots & 1 & \dots & 0 \\ 0 & \dots & -l_{k+1,k} & \dots & 0 \\ 0 & \dots & -l_{k+2,k} & \dots & 0 \\ \vdots & \dots & \vdots & \ddots & \vdots \\ 0 & \dots & -l_{m,k} & \dots & 1 \end{pmatrix}$$

with $l_{i,k} = a_{i,k}^{(k)} / a_{k,k}^{(k)}$, and $\mathbf{A}^{(k)} = (a_{i,j}^{(k)})$ the matrix obtained after step k of row echelon reduction (or, equivalently, Gaussian elimination) is called a Gaussian **multiplier matrix**.

Permutation and Gaussian multiplier matrices are **elementary matrices**.

- The Gaussian multiplier matrix ...

$$\mathbf{L}_k = \begin{pmatrix} 1 & \dots & 0 & \dots & 0 \\ 0 & \ddots & 0 & \dots & 0 \\ 0 & \dots & 1 & \dots & 0 \\ 0 & \dots & -l_{k+1,k} & \dots & 0 \\ \vdots & \dots & \vdots & \ddots & \vdots \\ 0 & \dots & -l_{m,k} & \dots & 1 \end{pmatrix}$$

- ... has inverse (matrix that “undoes” the linear transformation)

$$\mathbf{L}_k^{-1} = \begin{pmatrix} 1 & \dots & 0 & \dots & 0 \\ 0 & \ddots & 0 & \dots & 0 \\ 0 & \dots & 1 & \dots & 0 \\ 0 & \dots & l_{k+1,k} & \dots & 0 \\ \vdots & \dots & \vdots & \ddots & \vdots \\ 0 & \dots & l_{m,k} & \dots & 1 \end{pmatrix}$$

- Consider elementary matrices

$$\mathbf{E}_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix}, \mathbf{E}_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix}, \mathbf{E}_1 \mathbf{E}_2 = \mathbf{E}_2 \mathbf{E}_1 = \mathbf{I},$$

stating that \mathbf{E}_1 undoes the effect of \mathbf{E}_2 .

- $\mathbf{A} \in \mathbb{R}^{m \times m}$ is invertible if there exists $\mathbf{X} \in \mathbb{R}^{m \times m}$ such that

$$\mathbf{A} \mathbf{X} = \mathbf{X} \mathbf{A} = \mathbf{I}$$

- Notation $\mathbf{X} = \mathbf{A}^{-1}$, is the *inverse* of \mathbf{A} .



- What about general square matrices $A \in \mathbb{R}^{m \times m}$? How to find inverse
- X is inverse if $AX = I$ or

$$A[\mathbf{x}_1 \ \mathbf{x}_2 \ \dots \ \mathbf{x}_m] = [A\mathbf{x}_1 \ A\mathbf{x}_2 \ \dots \ A\mathbf{x}_m] = [\mathbf{e}_1 \ \mathbf{e}_2 \ \dots \ \mathbf{e}_m]$$

- Find the inverse is equivalent to solving systems $A\mathbf{x}_1 = \mathbf{e}_1, \dots, A\mathbf{x}_m = \mathbf{e}_m$
- Gauss Jordan algorithm generalizes Gaussian elimination that solves a single linear system to solving m systems simultaneously by forming the bordered matrix $[A \mid I]$

$$[A \mid I] \sim [I \mid X]$$

- When does a matrix inverse exist? $A \in \mathbb{R}^{m \times m}$
 - a A invertible
 - b $Ax = b$ has a unique solution for all $b \in \mathbb{R}^m$
 - c $Ax = 0$ has a unique solution
 - d The reduced row echelon form of A is I
 - e A can be written as product of elementary matrices

$$a \Rightarrow b \Rightarrow c \Rightarrow d \Rightarrow e \Rightarrow a$$

$a \Rightarrow b$ A invertible $\Rightarrow A^{-1}$ exists, and $x = A^{-1}b$ is a solution $A(A^{-1}b) = (AA^{-1})b = b$. If there were two solutions x, y , then

$$x - y = (A^{-1}A)(x - y) = A^{-1}(Ax - Ay) = A^{-1}(b - b) = A^{-1}0 = 0.$$

$b \Rightarrow c$ Choose $b = 0$

$c \Rightarrow d$ $[A \mid 0] \sim [U \mid 0]$. If $U \neq I$ there is a row of zeros, and solution is not unique. If solution is unique then $U = I$

$d \Rightarrow e$ $[A \mid 0] \sim [I \mid 0]$ implies $E_k \dots E_1 A = I \Rightarrow A = E_1^{-1} \dots E_k^{-1}$

$e \Rightarrow a$ $A = E_1^{-1} \dots E_k^{-1} \Rightarrow A^{-1} = E_k \dots E_1$.



- The inverse of a product $(AB)^{-1} = B^{-1}A^{-1}$

$$(AB)B^{-1}A^{-1} = A(BB^{-1})A^{-1} = AIA^{-1} = AA^{-1} = I$$

$$B^{-1}A^{-1}(AB) = B^{-1}(A^{-1}A)B = B^{-1}IB = B^{-1}B = I$$

- If $A \in \mathbb{R}^{m \times m}$ invertible so are: cA , A^T , A^k

$$(A^T)^{-1} = (A^{-1})^T$$

Verify

$$A^T (A^{-1})^T = (A^{-1}A)^T = I$$

$$(A^{-1})^T A^T = (AA^{-1})^T = I$$