

3. ELEMENTARY MATRIX OPERATIONS AND SYSTEMS OF LINEAR EQUATIONS

goal: computing rank of matrix
solving systems of linear equations

3.1. Elementary matrix operations and elementary matrices.

goal: transform a matrix by elementary operations
into a simpler one of same rank

Definition 3.1. An elementary row-column operation is

- (1) interchanging two rows/columns
- (2) multiplying a row /column by a non-zero scalar
- (3) adding a scalar multiple of a row /column to another row /column

$$\begin{array}{lll}
 \text{Ex } A = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 1 & -1 & 3 \\ 4 & 0 & 1 & 2 \end{pmatrix} & \begin{array}{l} \text{interchange} \\ \text{type 1 : rows 1 and 2} \end{array} & \begin{pmatrix} 2 & 1 & -1 & 3 \\ 1 & 2 & 3 & 4 \\ 4 & 0 & 1 & 2 \end{pmatrix} \stackrel{\text{need later}}{\downarrow} = A_1 \\
 \left(\begin{array}{l} \text{Rem If } P \rightarrow Q \text{ by elementary} \\ \text{operation, then} \\ Q \rightarrow P \text{ by el. oper.} \end{array} \right) & \begin{array}{l} \text{multiply} \\ \text{type 2 : second column by 3} \end{array} & \begin{pmatrix} 1 & 6 & 3 & 4 \\ 2 & 3 & -1 & 3 \\ 4 & 0 & 1 & 2 \end{pmatrix} = A_2 \\
 & \begin{array}{l} \text{add } 4 \times \\ \text{type 3 : third row to first row} \end{array} & \begin{pmatrix} 17 & 2 & 7 & 12 \\ 2 & 1 & -1 & 3 \\ 4 & 0 & 1 & 2 \end{pmatrix} = A_3
 \end{array}$$

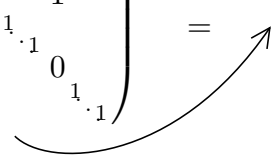
Definition 3.2. An $n \times n$ elementary matrix is one obtained from I_n by performing an elementary operation.

$$\begin{array}{ll}
 \text{Ex. } \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \nearrow \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \text{exch. row 1 and 2} \\
 & \searrow \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} & \text{add } -2 \cdot \text{row 3 to row 1}
 \end{array}$$

Theorem 3.3. Let $A \in M_{m \times n}(\mathbf{F})$ $A \xrightarrow{[col.]} B$ by elementary row operation.
 $\implies \exists$ elementary matrix E with $B = EA$ [$B = AE$].
 with $Id \rightarrow E$ by the same elementary operation.

$$A_1 = \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} A \quad A_2 = A \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad A_3 = \begin{pmatrix} 1 & 0 & 4 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} A$$

Theorem 3.4. Elementary matrices are invertible, and their inverses are of the same type.

Proof. $\begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & \lambda & \\ & & & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 1/\lambda & \\ & & & 1 \end{pmatrix}$ $\begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 0 & 1 \\ & & 1 & 0 \end{pmatrix}^{-1} =$ 

$$\begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & \lambda & \\ & & & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & -\lambda & \\ & & & 1 \end{pmatrix}.$$
 □

3.2. Rank of matrix and matrix inverses.

Definition 3.5. $A \in M_{m \times n}(\mathbf{F})$ $\text{rk}(A) := \text{rk}(L_A : \mathbf{F}^n \rightarrow \mathbf{F}^m).$

Theorem 3.6. Let V, W be VS / \mathbf{F} $T : V \rightarrow W$ linear transformation.

$\begin{matrix} \cup & \cup \\ \beta & \gamma \\ OB \end{matrix}$

Then $\text{rk}(T) = \text{rk}([T]_{\beta}^{\gamma}).$

Proof. The following diagram commutes.

$$\begin{array}{ccc} V & \xrightarrow{T} & W \\ \phi_{\beta} \downarrow & & \downarrow \phi_{\gamma} \\ \mathbf{F}^n & \xrightarrow{L_{[T]_{\beta}^{\gamma}}} & \mathbf{F}^m \end{array} \quad \begin{array}{l} \phi_{\beta}(\mathbf{x}) = [\mathbf{x}]_{\beta} \\ \phi_{\beta}, \phi_{\gamma} \text{ isomorphism} \end{array}$$

$$\text{rk}(T) \underset{\substack{\uparrow \\ \text{def of rank}}}{=} \dim(\text{Im}T) \underset{\substack{\uparrow \\ \phi_{\gamma} \text{ iso}}}{=} \dim(\phi_{\gamma}(\text{Im}T)) \underset{\substack{\uparrow \\ (1)}}{=} \dim(\text{Im}(L_{[T]_{\beta}^{\gamma}})) \overset{\text{def of rank}}{\downarrow} = \text{rk}(L_{[T]_{\beta}^{\gamma}}) \overset{\text{previous def}}{\downarrow} =: \text{rk}([T]_{\beta}^{\gamma})$$

using

$$\phi_{\gamma}(\text{Im}T) = \text{Im}(\phi_{\gamma} \circ T) \underset{\substack{\uparrow \\ \text{diagram commutes}}}{=} \text{Im}(L_{[T]_{\beta}^{\gamma}} \circ \phi_{\beta}) = L_{[T]_{\beta}^{\gamma}}(\text{Im}\phi_{\beta}) \underset{\substack{\uparrow \\ \phi_{\beta} \text{ iso}}}{=} L_{[T]_{\beta}^{\gamma}}(\mathbf{F}^n) = \text{Im}(L_{[T]_{\beta}^{\gamma}}) \quad (1)$$

Theorem 3.7. Let $A \in M_{m \times n}(\mathbf{F})$ $\begin{matrix} P, & Q \\ \cap & \cap \\ M_{m \times m} & M_{n \times n} \end{matrix}$ invertible.

Then $\text{rk}(PA) = \text{rk}(A) = \text{rk}(AQ) = \text{rk}(PAQ).$

Proof. $R(L_{AQ}) = R(L_A \circ L_Q) = L_A L_Q(\mathbf{F}^n) = L_A(L_Q(\mathbf{F}^n)) \overset{L_Q \text{ onto}}{\downarrow} = L_A(\mathbf{F}^n) = R(L_A)$

$$\implies \text{rk}(AQ) = \dim(R(L_{AQ})) = \dim(R(L_A)) = \text{rk}(A).$$

$$\begin{array}{ccc} \dim R(L_{PA}) & = \dim L_P(L_A(\mathbf{F}^n)) & \underset{\substack{\uparrow \\ L_P \text{ iso}}}{=} \dim(L_A(\mathbf{F}^n)) = \dim R(L_A) = \text{rk}(A). \\ \parallel & & \\ \text{rk}(PA) & & \end{array}$$

Corollary 3.8. *Elementary row and column operations on a matrix are rank-preserving.*

Theorem 3.9. *The rank of a matrix is the maximal number of linearly independent columns = $\dim(\text{subspace generated by columns})$.*

Proof. Let $A \in M_{m \times n}(\mathbf{F})$. Then

$$L_A : \mathbf{F}^n \rightarrow \mathbf{F}^m \quad L_A(\mathbf{x}) = A \cdot \mathbf{x}$$

$$L_A(\mathbf{e}_i) = A \cdot \mathbf{e}_i \quad i\text{-th column of } A.$$

L_A linear

$$\text{So } \text{Im}(L_A) = L_A(\text{span}(\{\mathbf{e}_i\})) \xrightarrow{L_A \text{ linear}} \text{span}(\{L_A(\mathbf{e}_i)\}) = \underline{\text{span}(\text{columns of } A)}. \quad \square$$

Example 3.10.

$$\text{rk} \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix} = 2 \quad \begin{matrix} \text{first two cols are linearly independent,} \\ \begin{pmatrix} 7 \\ 8 \\ 9 \end{pmatrix} = 2 \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} - \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \end{matrix}$$

Corollary 3.11. $A \in M_{m \times n}(\mathbf{F})$ is invertible $\iff \text{rk}(A) = m = n$.

Theorem 3.12. (important!)

$$A \in M_{m \times n}(\mathbf{F}). \text{ Then } A \xrightarrow[\text{el. op.}]{m-r} \left\{ \left(\begin{array}{c|c} Id_r & \overbrace{0}^{n-r} \\ \hline 0 & 0 \end{array} \right) \right\} \Rightarrow \underline{r = \text{rk}(A)}.$$

\Downarrow
 $I_{r,m,n}$

Proof. By example.

$$\begin{pmatrix} \boxed{0} & 2 & 4 & 2 & 2 \\ 4 & 4 & 4 & 8 & 0 \\ 8 & 2 & 0 & 10 & 2 \\ 6 & 3 & 2 & 9 & 1 \end{pmatrix}$$

① $p := \text{pivot element row} := \text{pivot element col} := 1$

if all elements in pivot rectangle are 0 \Rightarrow end
 $\left. \begin{matrix} \text{row} \\ \text{col} \end{matrix} \right\} \geq p$

$$\begin{pmatrix} \boxed{4} & 4 & 4 & 8 & 0 \\ 0 & 2 & 4 & 2 & 2 \\ 8 & 2 & 0 & 10 & 2 \\ 6 & 3 & 2 & 9 & 1 \end{pmatrix}$$

① exchange rows and columns so that pivot $\neq 0$
(type 1)

$$\cdot - 6 \quad \left(\begin{array}{c} \boxed{1} \\ \cdot - 8 \end{array} \right) \begin{pmatrix} 1 & 1 & 2 & 0 \\ 0 & 2 & 4 & 2 \\ 8 & 2 & 0 & 10 \\ 6 & 3 & 2 & 9 \end{pmatrix}$$

② divide row by pivot element so it gets = 1
(type 2)

4

$$\begin{pmatrix} 1 & 1 & 1 & 2 & 0 \\ 0 & 2 & 4 & 2 & 2 \\ 0 & -6 & -8 & -6 & 2 \\ 0 & -3 & -4 & -3 & 1 \end{pmatrix} \quad \textcircled{3} \text{ zero out below pivot by type 3 row operations}$$

Diagram showing row operations: Row 2 - Row 1, Row 3 - (-6)Row 1, Row 4 - (-3)Row 1.

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 4 & 2 & 2 \\ 0 & -6 & -8 & -6 & 2 \\ 0 & -3 & -4 & -3 & 1 \end{pmatrix} \quad \textcircled{4} \text{ zero out right of pivot by type 3 col operations (make } \boxed{} \text{ entries to 0)}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & 1 \\ 0 & -6 & -8 & -6 & 2 \\ 0 & -3 & -4 & -3 & 1 \end{pmatrix} \quad \textcircled{5} \text{ move pivot one row down \& one col right} \rightarrow \textcircled{1}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & 1 \\ 0 & 0 & 4 & 0 & 8 \\ 0 & 0 & 2 & 0 & 4 \end{pmatrix} \xrightarrow{\textcircled{3-5}} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 4 & 0 & 8 \\ 0 & 0 & 2 & 0 & 4 \end{pmatrix} \xrightarrow{\textcircled{2}} \begin{pmatrix} 1 & & & & \\ & 1 & & & \\ & & 4 & 0 & 8 \\ & & 2 & 0 & 4 \end{pmatrix} \xrightarrow{\textcircled{3-5}} \begin{pmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & 0 & 0 \\ & & 0 & \boxed{0} & 0 \end{pmatrix}$$

$\xrightarrow{\textcircled{0}} I_{3,4,5} \text{ done} \quad \square$

Corollary 3.13. $A \in M_{m \times n}(\mathbf{F}) \quad \text{rk}(A) = r$ then

$$\begin{matrix} \exists & P, & Q & \text{products of elementary matrices:} \\ \cap & & \cap & \\ M_{m \times m} & & M_{n \times n} & \end{matrix} \quad A = P I_{r,m,n} Q.$$

Proof. $P^{-1} = E_1 \cdots E_k$ product of elementary matrices E_i for row operations in Q^{-1} col

$\textcircled{1} - \textcircled{5}$. Then $P = E_k^{-1} \cdots E_1^{-1}$, E_i^{-1} are elementary. \square

Corollary 3.14. Let $A \in M_{m \times n}(\mathbf{F})$.

- (a) $\text{rk}(A^T) = \text{rk}(A)$
- (b) $\text{rk}(A^T) = \dim \text{span}(\text{columns of } A) = \dim \text{span}(\text{rows of } A)$

Corollary 3.15. $A \in M_{n \times n}(\mathbf{F})$ is invertible \iff A is a product of elementary matrices

$$\begin{matrix} I_{n,n,n} \\ \parallel \\ I_n \end{matrix} \quad \text{Proof. "}\implies\text{" } A \text{ invertible} \implies \text{rk}(A) = n \xrightarrow{\text{cor. 3.13}} A = P \begin{matrix} I_{n,n,n} \\ \parallel \\ I_n \end{matrix} Q = PQ,$$

P, Q products of elementary matrices

" \impliedby " elementary matrices are invertible \Rightarrow their products are invertible \square

Corollary 3.16. Let $A \in M_{m \times n}(\mathbf{F})$ $\begin{matrix} M_{m \times m} & M_{n \times n} \\ \cup & \cup \end{matrix}$
 $\text{rk}(A) = r \iff \exists \begin{matrix} P & Q \end{matrix}$, $\begin{matrix} P & Q \end{matrix}$ invertible with $A = PI_{r,m,n}Q$.

Proof. “ \implies ” By Corollary 3.13, $\exists P, Q$ products of elementary matrices.

elementary matrices are invertible $\Rightarrow P, Q$ invertible

“ \impliedby ” assume $A = PI_{r,m,n}Q$, P, Q invertible

by Corollary 3.15, P, Q are products of elementary matrices

and $\text{rk}(EA) = \text{rk}(A) = \text{rk}(AE')$ when E, E' elementary

so $\text{rk}(A) = \text{rk}(PI_{r,m,n}Q) = \text{rk}(I_{r,m,n}) = r$. □

Proof of Cor 3.14. (a) When P is invertible, $(P^T)^{-1} = (P^{-1})^T$,

$$\text{so } A = PI_{r,m,n}Q \iff \text{rk}(A) = r$$

\Downarrow

$$\begin{array}{ccc} A^T = Q^T I_{r,n,m} P^T & \implies & \text{rk}(A^T) = r. \\ \uparrow & & \uparrow \\ \text{invertible} & & \text{invertible} \end{array}$$

(b) follows from (a). □

Theorem 3.17. Let $T : V \rightarrow W$, $U : W \rightarrow Z$ linear transformations,
 A, B matrices such that AB defined.

$$\underline{\text{Then}} \quad \begin{array}{ll} \text{(a) } \text{rk}(UT) \leq \text{rk}(U) & \text{(b) } \text{rk}(AB) \leq \text{rk}(A) \\ & \text{rk}(T) \qquad \qquad \text{rk}(B) \end{array}$$

Proof. (a) $\text{rk}(UT) = \dim(U(T(V))) \leq \dim(U(W)) = \dim \text{Im}(U) = \text{rk}(U)$.

$$U : W \rightarrow Z. \text{ Let } \tilde{U} = U|_{T(V)} : T(V) \rightarrow Z$$

$$\dim \text{Im}(\tilde{U}) + \underbrace{\dim(\ker(\tilde{U}))}_{\geq 0} = \dim(T(V))$$

$$\dim \text{Im}(\tilde{U}) \leq \dim(T(V))$$

$$(*) = \dim(\tilde{U}(T(V))) \leq \dim(T(V)) = \dim \text{Im}(T) = \text{rk}(T)$$

(b) for matrices apply to L_A, L_B . □

The inverse of a matrix

Definition 3.18. $A \in M_{m \times n}$, $B \in M_{m \times p}$, $\begin{matrix} (A|B) \\ \parallel \\ (AB) \in M_{m \times (n+p)} \end{matrix}$ $(A|B)$
 augmented matrix

Let $A \in M_{n \times n}(\mathbf{F})$ consider $(A|I_n)$.

Assume A is invertible. Then A^{-1} is product of elementary matrices
 (Cor 3.15 above)

$$A^{-1} = E_1 \cdot \dots \cdot E_g.$$

$$(A|B) \longrightarrow E(A|B) = (EA|EB) \quad \begin{array}{l} \text{for an elementary matrix} \\ E \text{ is an elementary row operation} \end{array}.$$

Now do this for E_g, E_{g-1}, \dots, E_1 on $(A|Id_n)$.

$$(A|Id_n) \longrightarrow A^{-1}(A|Id_n) = (A^{-1}A|A^{-1}Id_n) = (Id|A^{-1}).$$

Corollary 3.19. A is invertible \iff there exists a sequence of row operations that turn $(A|Id_n)$ into $(Id_n|B)$. In this case $B = A^{-1}$.

Proof. “ \implies ” we proved that $(A|Id_n) \rightarrow (Id_n|A^{-1})$ by row operations.

Now assume it is possible.

$$(A|Id_n) \rightarrow (Id_n|X) \quad \text{by row operations} \quad (2)$$

row operations on $(A|X)$ preserve $X^{-1}A$: “invariant”

$$\begin{array}{ccc} (EA|EX) & & (A|X) \\ \downarrow & & \downarrow \\ (X^{-1}E^{-1})(EA) & = & X^{-1}A. \end{array}$$

$$\begin{array}{lcl} \text{Thus if (2), then } X^{-1}Id_n = Id_n^{-1}A & \Rightarrow & A = X^{-1} \\ & \Rightarrow & X = A^{-1} \end{array}$$

“ \Leftarrow ” If $(A|Id_n) \rightarrow (Id_n|B)$ by row operations, then
forget right block: $A \rightarrow Id_n$ by row operations,
so $\text{rk}(A) = \text{rk}(Id_n) = n \Rightarrow A$ is invertible. □

$$\left[\begin{array}{l} \text{this means: if } A \text{ is non-invertible, any attempt to} \\ \text{turn } (A|Id_n) \xrightarrow{\text{row}} (Id_n|B) \text{ will produce a left} \\ \text{block with a zero row/column.} \end{array} \right]$$

How to find row operations for $(A|Id_n) \rightarrow (Id_n|B)$?

(this is the first method to compute $B = A^{-1}$; we
may see another)

$$\underline{\text{Ex}} \quad \begin{pmatrix} 0 & 2 & 4 \\ 2 & 4 & 2 \\ 3 & 3 & 1 \end{pmatrix}^{-1} = ? \quad (\text{We start as in the algorithm to determine rank.})$$

$$(A|I) = \left(\begin{array}{ccc|ccc} \boxed{0} & 2 & 4 & 1 & 0 & 0 \\ 2 & 4 & 2 & 0 & 1 & 0 \\ 3 & 3 & 1 & 0 & 0 & 1 \end{array} \right)$$

$$\left(\begin{array}{ccc|ccc} \boxed{2} & 4 & 2 & 0 & 1 & 0 \\ 0 & 2 & 4 & 1 & 0 & 0 \\ 3 & 3 & 1 & 0 & 0 & 1 \end{array} \right)$$

- ① exch row 1 and 2 \Rightarrow pivot $\neq 0$
if pivot + all below pivot = 0
then stop. column := 0 \Rightarrow

matrix not invertible

$$\cdot -3 \left\{ \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \left(\begin{array}{ccc|ccc} 1 & 2 & 1 & 0 & 1/2 & 0 \\ 0 & 2 & 4 & 1 & 0 & 0 \\ 3 & 3 & 1 & 0 & 0 & 1 \end{array} \right) \right.$$

② multiply first row by $1/2 \Rightarrow \text{pivot}=1$

$$\left(\begin{array}{ccc|ccc} 1 & 2 & 1 & 0 & 1/2 & 0 \\ 0 & 2 & 4 & 1 & 0 & 0 \\ 0 & -3 & -2 & 0 & -3/2 & 1 \end{array} \right)$$

③ zero out below pivot & above pivot (if pivot is in column > 1)

comes first important difference to
row-column reduction:

cannot zero out right of pivot
by column operation!

$$\left(\begin{array}{ccc|ccc} 1 & \boxed{2} & 1 & 0 & 1/2 & 0 \\ 0 & \boxed{2} & 4 & 1 & 0 & 0 \\ 0 & -3 & -2 & 0 & -3/2 & 1 \end{array} \right)$$

④ change shape of pivot rectangle

(now what is above pivot is also
in rectangle)

① unnecessary

② divide by 2 $\Rightarrow \text{pivot}=1$

$$\cdot -2 \left\{ \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \left(\begin{array}{ccc|ccc} 1 & \boxed{2} & 1 & 0 & 1/2 & 0 \\ 0 & \boxed{1} & 2 & 1/2 & 0 & 0 \\ 0 & -3 & -2 & 0 & -3/2 & 1 \end{array} \right) \right.$$

$$\cdot -3 \left\{ \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \left(\begin{array}{ccc|ccc} 1 & \boxed{2} & 1 & 0 & 1/2 & 0 \\ 0 & \boxed{1} & 2 & 1/2 & 0 & 0 \\ 0 & -3 & -2 & 0 & -3/2 & 1 \end{array} \right) \right.$$

$$\left(\begin{array}{ccc|ccc} 1 & 0 & \boxed{-3} & -1 & 1/2 & 0 \\ 0 & 1 & \boxed{2} & 1/2 & 0 & 0 \\ 0 & 0 & \boxed{4} & 3/2 & -3/2 & 1 \end{array} \right)$$

③ zero above and below pivot

④ change pivot

$$\cdot -2 \left\{ \begin{array}{l} \rightarrow \\ \rightarrow \end{array} \left(\begin{array}{ccc|ccc} 1 & 0 & -3 & -1 & 1/2 & 0 \\ 0 & 1 & 2 & 1/2 & 0 & 0 \\ 0 & 0 & 1 & 3/8 & -3/8 & 1/4 \end{array} \right) \right.$$

① skip

②

$$\left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 1/8 & -5/8 & 3/4 \\ 0 & 1 & 0 & -1/4 & 3/4 & -1/2 \\ 0 & 0 & 1 & 3/8 & -3/8 & 1/4 \end{array} \right)$$

③

$$\Rightarrow \text{done} \quad \left(\begin{array}{ccc} 0 & 2 & 4 \\ 2 & 4 & 2 \\ 3 & 3 & 1 \end{array} \right)^{-1} = \left(\begin{array}{ccc} \frac{1}{8} & -\frac{5}{8} & \frac{3}{4} \\ -\frac{1}{4} & \frac{3}{4} & -\frac{1}{2} \\ \frac{3}{8} & -\frac{3}{8} & \frac{1}{4} \end{array} \right).$$

$$\begin{array}{rcl}
a_{11}x_1 + \dots + a_{1n}x_n & = & b_1 \\
\vdots & & \vdots \\
a_{m1}x_1 + \dots + a_{mn}x_n & = & b_m
\end{array} \quad (3) \quad \begin{array}{l} a_{ij}, b_i \in \mathbf{F} \text{ field const} \\ x_j \in \mathbf{F} \text{ unknowns} \end{array}$$

system of m linear equations
with n unknowns over \mathbf{F}

\uparrow

we had this before!
given $\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{x}$,
find λ_i with $\sum \lambda_i \mathbf{v}_i = \mathbf{x}$.
Now we do this more systematically

$A = (a_{ij})_{i=1, j=1}^m, n$ coefficient matrix.

$$\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \quad \mathbf{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$

(3) $\iff A\mathbf{x} = \mathbf{b}$. linear equation system (LES)

$\mathbf{s} \in \mathbf{F}^n$ solution $\iff A\mathbf{s} = \mathbf{b}$ $\{\text{solutions}\} =:$ solution set

$$(3) \begin{cases} \text{consistent} : \iff \text{solution set} \neq \emptyset \\ \text{inconsistent} : \iff \text{solution set} = \emptyset \end{cases}$$

$$\begin{array}{lll}
x_1 + x_2 = 3 & 2x_1 + 3x_2 + x_3 = 1 & x_1 + x_2 = 0 \\
x_1 - x_2 = -1 & x_1 - x_2 + 2x_3 = 6 & x_1 + x_2 = 1 \\
x_1 = 1 \quad x_2 = 2 & (x_1, x_2, x_3) = \begin{cases} (-6, 2, 7) \\ (8, -4, -3) \end{cases} \text{ sol.} & \text{no sol. } \underline{\text{inconsistent}} \\
\text{unique sol} & &
\end{array}$$

How do we find solution?

Definition 3.20. A system $A\mathbf{x} = \mathbf{b}$ is $\begin{cases} \text{homogeneous} & \text{if } \mathbf{b} = \mathbf{0} \\ \text{nonhomogeneous} & \text{if } \mathbf{b} \neq \mathbf{0} \end{cases}$

Theorem 3.21. K sol. set of $A\mathbf{x} = \mathbf{0}$. $K = \ker(L_A) =: \ker A$.
 $\dim K = n - \text{rk}(L_A)$

$$L_A : \mathbf{F}^n \rightarrow \mathbf{F}^m$$

Proof. $\dim \ker(L_A) = \dim \mathbf{F}^n - \dim \text{Im}(L_A)$. □

fewer eqn than variables

\downarrow

Corollary 3.22. If $m < n$, then $K \neq \{\mathbf{0}\}$.

Theorem 3.23. Let $A\mathbf{x} = \mathbf{b}$ (4) be a (consistent) LES and $A\mathbf{x} = \mathbf{0}$ (5) be the homogeneous system corresponding to (4). Then

$$K_{(4)} = \mathbf{s} + K_{(5)} = \{ \mathbf{s} + \mathbf{k} : \mathbf{k} \in K_{(5)} \},$$

where \mathbf{s} is a concrete solution of (4).

Proof. Let $\mathbf{s} \in K_{(4)}$ $A\mathbf{s} = \mathbf{b}$
 $\mathbf{s}' \in K_{(4)}$ $A\mathbf{s}' = \mathbf{b}$ $A(\mathbf{s} - \mathbf{s}') = \mathbf{0} \Rightarrow \mathbf{s}' - \mathbf{s} \in K_{(5)}$
 $\Rightarrow \mathbf{s}' = \underbrace{(-\mathbf{s} + \mathbf{s}')}_{\in K_{(5)}} + \mathbf{s} \in \mathbf{s} + K_{(5)}.$

$$\Rightarrow K_{(4)} \subseteq \mathbf{s} + K_{(5)}.$$

Now let $\mathbf{s}' \in \mathbf{s} + K_{(5)}$ $\mathbf{s}' = \mathbf{s} + \mathbf{k}$ $A\mathbf{k} = \mathbf{0}$

$$A\mathbf{s}' = A\mathbf{s} + A\mathbf{k} = A\mathbf{s} = \mathbf{b} \Rightarrow \mathbf{s}' \in K_{(4)} \Rightarrow \mathbf{s} + K_{(5)} \subseteq K_{(4)}. \quad \square$$

Example 3.24.

$$\boxed{1} \quad \begin{array}{rcl} -x_1 + 2x_2 - x_3 & = & 0 \\ x_1 + x_2 - x_3 & = & 0 \end{array} \quad A = \begin{pmatrix} -1 & 2 & -1 \\ 1 & 1 & -1 \end{pmatrix} \quad \text{rk}(A) = 2$$

$$: \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

$$\dim(\ker A) = \dim(\mathbb{R}^3) - \text{rk}(A) = 1.$$

$$\text{check that } \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \in \ker A \quad \ker A = \left\{ t \cdot \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} : t \in \mathbb{R} \right\}$$

$$\uparrow$$

$$(\text{how to find - later})$$

$$\downarrow$$

$$\boxed{2} \quad \begin{array}{rcl} -x_1 + 2x_2 - x_3 & = & -3 \\ x_1 + x_2 - x_3 & = & -2 \end{array} \quad \begin{pmatrix} 1 \\ 1 \\ 4 \end{pmatrix} \quad \text{special solution}$$

$$\Rightarrow K = \left\{ \begin{pmatrix} 1 \\ 1 \\ 4 \end{pmatrix} + t \cdot \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} : t \in \mathbb{R} \right\} \quad \begin{array}{l} \text{solution} \\ \text{set} \end{array}$$

One special case

Theorem 3.25. Let $A\mathbf{x} = \mathbf{b}$ be a LES with $A \in M_{n \times n}$.

Then $A\mathbf{x} = \mathbf{b}$ has exactly one solution $\iff A$ is invertible

In that case the solution is $\mathbf{x} = A^{-1}\mathbf{b}$.

Proof. “ \Leftarrow ” Let A be invertible $\mathbf{x} = A^{-1}\mathbf{b}$ is a solution.

$$\{\text{sol. of } A\mathbf{x} = \mathbf{b}\} = A^{-1}\mathbf{b} + \underbrace{\{\text{sol of } A\mathbf{x} = \mathbf{0}\}}_{\ker A = \{\mathbf{0}\}} = \{A^{-1}\mathbf{b}\}$$

$$\uparrow$$

$$A \text{ invertible} \Rightarrow \text{injective}$$

“ \Rightarrow ” Let \mathbf{s} be the unique solution of $A\mathbf{x} = \mathbf{b}$.

$$\{\text{solutions to } A\mathbf{x} = \mathbf{b}\} = \mathbf{s} + \underbrace{\{\text{sol of } A\mathbf{x} = \mathbf{0}\}}_{\ker A} \stackrel{\substack{\text{s unique} \\ \downarrow}}{=} \{\mathbf{s}\}.$$

$$\Rightarrow \ker(A) = \{\mathbf{0}\}$$

$$\Rightarrow A \text{ injective} \Rightarrow A : \mathbf{F}^n \rightarrow \mathbf{F}^n \text{ bijective} \Rightarrow A \text{ invertible} \quad \square$$

$$\underline{\text{Ex.}} \quad \begin{array}{rcl} 2x_2 + 4x_3 & = & 3 \\ 2x_1 + 4x_2 + 2x_3 & = & 3 \\ 3x_1 + 3x_2 + x_3 & = & 1 \end{array} \quad \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = A^{-1}\mathbf{b} = \begin{pmatrix} 1/8 & -5/8 & 3/4 \\ -1/4 & 3/4 & -1/2 \\ 3/8 & -3/8 & 1/4 \end{pmatrix} \begin{pmatrix} 3 \\ 3 \\ 1 \end{pmatrix} = \begin{pmatrix} -3/4 \\ 1 \\ 1/4 \end{pmatrix}.$$

Definition 3.26. $(A|\mathbf{b})$ the augmented matrix of LES $A\mathbf{x} = \mathbf{b}$.

Theorem 3.27. $A\mathbf{x} = \mathbf{b}$ is consistent LES $\iff \text{rk}(A|\mathbf{b}) = \text{rk}(A)$.

$$\begin{aligned}
 \exists \mathbf{x} : A\mathbf{x} = \mathbf{b} &\iff \mathbf{b} \in \text{Im}(A) = \text{span}(\text{columns of } A) \\
 &\quad \quad \quad \begin{array}{c} \parallel \\ (\text{Im } L_A) \end{array} \\
 &\iff \overbrace{\text{span}(\text{columns of } A \cup \{\mathbf{b}\})}^V = \overbrace{\text{span}(\text{columns of } A)}^W \\
 \left[\begin{array}{l} \text{if } V \supseteq W, \text{ then} \\ V = W \iff \dim V = \dim W \end{array} \right] \longrightarrow &\iff \dim(\text{span}(\text{col of } A)) = \dim(\text{span}(\text{col of } A \cup \{\mathbf{b}\})) \\
 &\quad \quad \quad \parallel \qquad \qquad \qquad \parallel \\
 &\iff \text{rk}(A) = \text{rk}(A|\mathbf{b}) \quad \square
 \end{aligned}$$

Example 3.28.
$$\begin{array}{rcl}
 x_1 + x_2 & = & 1 \\
 x_1 + x_2 & = & 0
 \end{array}
 \quad A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \quad \mathbf{b} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad (A|\mathbf{b}) = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix}.$$

$$\text{rk}(A) = 1 \qquad \text{rk}(A|\mathbf{b}) = 2$$

\Rightarrow inconsistent

Example 3.29. Is $\begin{pmatrix} 3 \\ 3 \\ 2 \end{pmatrix}$ a linear combination of $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$?

$$\begin{pmatrix} 3 \\ 3 \\ 2 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + x_2 \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \quad \begin{array}{rcl} x_1 + x_2 & = & 3 \\ x_1 - x_2 & = & 3 \\ x_1 & = & 2 \end{array} \quad \text{no solution} \Rightarrow \underline{\text{no}}$$

closed economic model (Application)

producer : Farmer (food), Tailor (cloth), carpenter (house)
& consumer

no capital enters and exits (closed model)

	Food	Clothing	Housing	
Farmer	0.4	0.2	0.2	} consumption
Tailor	0.1	0.7	0.2	
Carpenter	0.5	0.1	0.6	
	<hr/>	<hr/>	<hr/>	
(1.0	1.0	1.0)

e.g. Farmer consumes 40% of food and 20% of cloth.

Qu How much must F,T,C produce to attain an equilibrium (income = spending \implies society survives)?

food cloth house
Let p_1 , p_2 , p_3 be incomes of farmer/tailor/carpenter
when farmer must buy 20% of housing,
he spends $0.2p_3$

$$\begin{array}{rcl}
 \underbrace{\hspace{2cm}}_{\text{spending}} & & \underbrace{\hspace{1cm}}_{\text{income}} \\
 0.4p_1 + 0.2p_2 + 0.2p_3 & = & p_1 \\
 0.1p_1 + 0.7p_2 + 0.2p_3 & = & p_2 \\
 0.5p_1 + 0.1p_2 + 0.6p_3 & = & p_3
 \end{array}
 \quad
 \begin{array}{c}
 \text{consumption matrix} \\
 \downarrow \\
 A = \begin{pmatrix} 0.4 & 0.2 & 0.2 \\ 0.1 & 0.7 & 0.2 \\ 0.5 & 0.1 & 0.6 \end{pmatrix} \\
 \mathbf{p} = \begin{pmatrix} p_1 \\ p_2 \\ p_3 \end{pmatrix} \quad A\mathbf{p} = \mathbf{p}.
 \end{array}$$

In this example $\mathbf{p} = \begin{pmatrix} 0.25 \\ 0.35 \\ 0.4 \end{pmatrix}$ $\mathbf{p} \in \ker(A - Id) \iff$ equilibrium condition
 so income of farmer : taylor : carpenter = 5 : 7 : 8

Rem. what we want is $\mathbf{p} \geq \mathbf{0}$ (positive; all entries $p_i \geq 0$)

Thm. (B.Noble 1971) If $A = \left(\begin{array}{c|c} \overbrace{B}^{n-1} & \overbrace{C}^1 \\ \hline D & E \end{array} \right) \begin{matrix} \} n-1 \\ 1 \end{matrix}$ with $C, D > \mathbf{0}$, then

$(I - A)\mathbf{x} = \mathbf{0}$ has a 1-dimensional solution set generated by a non-negative vector.

open model outside demand d_1 food d_2 cloth d_3 house $\mathbf{d} = \begin{pmatrix} d_1 \\ d_2 \\ d_3 \end{pmatrix}$

income(carpenter) = spending(carpenter) + $\overbrace{\text{outside demand (house)}}^{\text{win(carpenter)}}$
 farmer farmer food
 taylor taylor cloth
 \Downarrow

$$\mathbf{d} + A\mathbf{p} = \mathbf{p} \implies \mathbf{p} = (I - A)^{-1}\mathbf{d} \quad (\text{and again } \mathbf{p} \text{ must be non-negative})$$

3.4. LES - computational aspects.

we know: to solve $A\mathbf{x} = \mathbf{b}$ we need / 1. one sol
 \ 2. basis of sol. set of $A\mathbf{x} = \mathbf{0}$

we use row operations to accomplish 1 & 2 by transforming $A\mathbf{x} = \mathbf{b}$ into
 a simple system where we see solutions

Definition 3.30. $(A|\mathbf{b}) \iff (A'|\mathbf{b}')$ equivalent if \mathbf{x} sol. of $A\mathbf{x} = \mathbf{b} \iff \mathbf{x}$ sol. of $A'\mathbf{x} = \mathbf{b}'$.

Theorem 3.31. $C \in M_{m \times m}$ invertible, $A \in M_{m \times n}$, $b \in \mathbf{F}^m$

$$(CA|C\mathbf{b}) \iff (A|\mathbf{b}) \quad (\text{same solutions})$$

Corollary 3.32. Row operations on $(A|\mathbf{b})$ give equivalent system.

Qu: How to find proper row operations?

Gaussian elimination

$$\text{by example} \quad \left. \begin{array}{rrcr} 3x_1 + 2x_2 + 3x_3 - 2x_4 & = & 1 \\ x_1 + x_2 + x_3 & = & 3 \\ x_1 + 2x_2 + x_3 - x_4 & = & 2 \end{array} \right\} \quad (6)$$

I part

Forward pass

$$\left(\begin{array}{cccc|c} \boxed{3} & 2 & 3 & -2 & 1 \\ 1 & 1 & 1 & 0 & 3 \\ 1 & 2 & 1 & -1 & 2 \end{array} \right)$$

$$\cdot -3 \quad \left(\begin{array}{cccc|c} \boxed{1} & 2 & 1 & -1 & 2 \\ 1 & 1 & 1 & 0 & 3 \\ 3 & 2 & 3 & -2 & 1 \end{array} \right)$$

$$\cdot -1 \quad \left(\begin{array}{cccc|c} 1 & 2 & 1 & -1 & 2 \\ 0 & \boxed{-1} & 0 & 1 & 1 \\ 0 & -4 & 0 & 1 & -5 \end{array} \right)$$

$$\cdot 4 \quad \left(\begin{array}{cccc|c} 1 & 2 & 1 & -1 & 2 \\ 0 & 1 & 0 & -1 & -1 \\ 0 & -4 & 0 & 1 & -5 \end{array} \right)$$

$$\left(\begin{array}{cccc|c} 1 & 2 & 1 & -1 & 2 \\ 0 & 1 & 0 & -1 & -1 \\ 0 & 0 & 0 & \boxed{-3} & -9 \end{array} \right)$$

$$\left(\begin{array}{cccc|c} \textcircled{1} & 2 & 1 & -1 & 2 \\ 0 & \textcircled{1} & 0 & -1 & -1 \\ 0 & 0 & 0 & \textcircled{1} & 3 \end{array} \right)$$

(but not down) (but not above)

① move pivot right as long as pivot + below = 0

① make pivot to 1

can divide by 3 but better exch row 1 & 3

② zero out below pivot

③ move pivot 1 right & 1 down

① skip

① make pivot = 1 ($\cdot -1$ second row)

① pivot + below is zero \rightarrow move right

①

we achieved now that the first non-zero entry in each row = 1 & always right of first entry in the previous row

II part

Backward pass

zero out above ① using row operations of type 3 from right to left

(so that in columns with ① this is the only non-zero entry)

$$\begin{pmatrix} 1 & 2 & 1 & -1 & | & 2 \\ 0 & 1 & 0 & -1 & | & -1 \\ 0 & 0 & 0 & \textcircled{1} & | & 3 \end{pmatrix} \begin{array}{l} \leftarrow \\ \leftarrow \\ \leftarrow \end{array} \begin{array}{l} \\ \\ + \end{array} + \\
\begin{pmatrix} 1 & 2 & 1 & 0 & | & 5 \\ 0 & \textcircled{1} & 0 & 0 & | & 2 \\ 0 & 0 & 0 & 1 & | & 3 \end{pmatrix} \begin{array}{l} \leftarrow \\ \leftarrow \\ \leftarrow \end{array} \begin{array}{l} \\ \\ -2 \end{array} \\
\begin{pmatrix} 1 & 0 & 1 & 0 & | & 1 \\ 0 & 1 & 0 & 0 & | & 2 \\ 0 & 0 & 0 & 1 & | & 3 \end{pmatrix} \iff (7) \quad \left| \begin{array}{rrc} x_1 & +x_3 & = 1 \\ & x_2 & = 2 \\ & & x_4 = 3 \end{array} \right.$$

Definition 3.33. A matrix is in reduced row echelon form (r.r.e. form, RREF) if the following conditions are satisfied:

- (a) any non-zero row lies above any zero row
- (b) first (=leftmost) non-zero entry in a row is the only non-zero entry in its column
- (c) first (=leftmost) non-zero entry in a row is = 1 and occurs in a column (above) right to leftmost entry in previous row

What is the solution of a LES in r.r.e. form?

- for every variable that does not occur leftmost in an equation take arbitrary value;

in our example $x_3 = t$

- then solve in terms of t

$$x_1 = 1 - x_3 = 1 - t \quad x_3 = t$$

$$x_2 = 2 \quad x_4 = 3$$

$$\begin{aligned} \Rightarrow \quad & \text{solution set of (7)} \\ & \parallel \\ & \text{solution set of (6)} \end{aligned} = \left\{ \begin{pmatrix} 1-t \\ 2 \\ t \\ 3 \end{pmatrix} \right\} = \left\{ \begin{pmatrix} 1 \\ 2 \\ 0 \\ 3 \end{pmatrix} + t \cdot \begin{pmatrix} -1 \\ 0 \\ 1 \\ 0 \end{pmatrix} : t \in \mathbb{R} \right\}$$

Example 3.34.

$$\begin{pmatrix} \textcircled{1} & 0 & 2 & 0 & -2 & | & 3 \\ 0 & \textcircled{1} & -1 & 0 & 1 & | & 1 \\ 0 & 0 & 0 & \textcircled{1} & -2 & | & 2 \\ 0 & 0 & 0 & 0 & 0 & | & 0 \end{pmatrix} \quad \begin{array}{rrc} x_1 & +2x_3 & -2x_5 = 3 \\ x_2 & -x_3 & +x_5 = 1 \\ & & x_4 - 2x_5 = 2 \end{array}$$

entries that do not occur first in a row: x_3 and x_5

$$x_3 = s, \quad x_5 = t$$

and solve for x_1, x_2, x_4

$$\begin{aligned} x_4 &= 2 + 2x_5 = 2 + 2t \\ x_2 &= 1 - x_5 + x_3 = 1 - t + s \\ x_1 &= 3 + 2x_5 - 2x_3 = 3 + 2t - 2s \end{aligned} \quad \Rightarrow \quad \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \begin{pmatrix} 3 + 2t - 2s \\ 1 - t + s \\ s \\ 2 + 2t \\ t \end{pmatrix}$$

$$= \begin{pmatrix} 3 \\ 1 \\ 0 \\ 2 \\ 0 \end{pmatrix} + s \cdot \begin{pmatrix} -2 \\ 1 \\ 1 \\ 0 \\ 0 \end{pmatrix} + t \cdot \begin{pmatrix} 2 \\ -1 \\ 0 \\ 2 \\ 1 \end{pmatrix}$$

general
solution
($s, t \in \mathbb{R}$ arbitrary)

Theorem 3.35. *(A|b) LES in r.r.e. form with r non-zero equations.*

- (a) $\text{rk}(A) = r$
- (b) *the above procedure gives the general solution*

$$x = x_0 + sx_1 + tx_2 + ux_3 + \cdots \quad s, t, u, \cdots \in \mathbb{R}$$

where x_0 is one solution of non-homogeneous system \mathcal{E}

$sx_1 + tx_2 + \cdots$ is the general solution of corresponding homogeneous system

if $(A|b)$ *consistent*

- (c) $(A|b)$ is inconsistent $\iff \exists$ zero row in A with non-zero element in \mathbf{b} ;

$$\left(\begin{array}{cc|c} \dots & \dots & \dots \\ 0 & 0 \dots 0 & \neq 0 \\ \dots & \dots & \dots \end{array} \right)$$

Rem r.r.e. form is unique. (Why?)

An interpretation of r.r.e.f.

Let A be a matrix and B be the r.r.e.f. of A

Let $\textcircled{1}$ be the leftmost non-zero entries in each row of B .

$$\begin{aligned} \#(1) &= \text{rk } B = \# \text{ non-zero rows in } B \\ \text{r.r.e.f.} \implies \underset{\substack{!! \\ j_k}}{\text{column of }} \#k(1) \text{ are } \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} \leftarrow k\text{-th row} &= \mathbf{e}_k . \end{aligned}$$

Claim 1: j_1, \dots, j_r -columns of A are linearly independent

because

$$c_1 a_{j_1} + \cdots + c_r a_{j_r} = 0$$

$$B = MA$$

invertible mtx
 \downarrow

because $A \rightarrow B$ by row

$$c_1 Ma_{j_1} + \cdots + c_r Ma_{j_r} = 0 \implies c_i = 0$$

operations

$$\parallel$$

$$\mathbf{e}_1$$

$$\parallel$$

$$\mathbf{e}_r$$

Let A be invertible $\mathbf{x} = A^{-1}\mathbf{b}$ is a solution.

$$\{\text{sol. of } A\mathbf{x} = \mathbf{b}\} = A^{-1}\mathbf{b} + \underbrace{\{\text{sol of } A\mathbf{x} = \mathbf{0}\}}_{\substack{\ker A = \{\mathbf{0}\} \\ \uparrow \\ A \text{ invertible} \Rightarrow \text{injective}}} = \{A^{-1}\mathbf{b}\}$$

Claim 2: when $\mathbf{b}_k = \begin{pmatrix} d_1 \\ \vdots \\ d_r \\ 0 \\ \vdots \\ 0 \end{pmatrix}$ (8) is the k -th column of B then

$\mathbf{a}_k = \sum_{i=1}^r d_i \mathbf{a}_{j_i}$ is the k -th column of A
i.e., all other $\neq j_i$ -th columns
of A are linear combinations of \mathbf{a}_{j_i}
because $\mathbf{b}_k = \sum_{i=1}^r d_i \mathbf{e}_i$, so

$$\mathbf{a}_k = M^{-1}\mathbf{b}_k \stackrel{(8)}{=} M^{-1} \left(\sum_{i=1}^r d_i \mathbf{e}_i \right) \stackrel{\mathbf{b}_{j_i} = \mathbf{e}_i}{=} \sum_{i=1}^r d_i M^{-1} \mathbf{b}_{j_i} = \sum_{i=1}^r d_i \mathbf{a}_{j_i}$$

\Rightarrow all columns of A are linear combinations of \mathbf{a}_{j_i}

$\Rightarrow \{\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_r}\}$ are a basis of the image of A

Theorem 3.36. Let $A \in M_{m \times n}$ $\text{rk}(A) = r$ ($r > 0$) and let. Then
 B be r.r.e.f. of A

- (a) The number of non-zero rows in B is r .
- (b) $\forall i = 1, 2, \dots, r \exists$ a column \mathbf{b}_{j_i} of B s.t. $\mathbf{b}_{j_i} = \mathbf{e}_i$
- (c) $\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_r}$ (columns of A numbered j_1, \dots, j_r) are linearly independent
- (d) For each $k = 1, 2, \dots, n$: column k of B is

$$\mathbf{b}_k = d_1 \mathbf{e}_1 + \dots + d_r \mathbf{e}_r \iff \mathbf{a}_k = d_1 \mathbf{a}_{j_1} + \dots + d_r \mathbf{a}_{j_r}$$

Corollary 3.37. r.r.e.f of A is unique.

Proof. go from $k = 1, k = 2, \dots, k = n$

Let l be the number of linearly independent columns $1, \dots, k-1$.

k -th column \mathbf{a}_k of A is not lin. combination of previous columns

$\iff k$ -th column of B is not linear combination of previous columns

$$\stackrel{B \text{ r.r.e.f.}}{\iff} \mathbf{b}_k = \mathbf{e}_{l+1}$$

if so, $1 + l$ is uniquely determined: it is the number in order of such k

if not, $\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_l}$ is linearly independent

so \exists uniquely determined d_i with $\sum_{m=1}^l d_m \mathbf{a}_{j_m} = \mathbf{a}_k$

$$\text{then } \sum_{m=1}^l d_m \underset{\substack{\parallel \\ \mathbf{e}_m}}{\mathbf{b}_{j_m}} = \mathbf{b}_k = \begin{pmatrix} d_1 \\ d_2 \\ \vdots \\ d_l \\ 0 \end{pmatrix}$$

Applications: (1) choose a basis from a generating set
(2) complete a linearly independent set to a basis

Example 3.38.

$$A = \begin{pmatrix} 2 & 4 & 6 & 2 & 4 \\ 1 & 2 & 3 & 1 & 1 \\ 2 & 4 & 8 & 0 & 0 \\ 3 & 6 & 7 & 5 & 9 \end{pmatrix} \quad B = \begin{pmatrix} \textcircled{1} & 2 & 0 & 4 & 0 \\ 0 & 0 & \textcircled{1} & -1 & 0 \\ 0 & 0 & 0 & 0 & \textcircled{1} \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$\mathbf{b}_1, \mathbf{b}_3, \mathbf{b}_5$ linearly independent, $\mathbf{b}_2 = 2\mathbf{b}_1$, $\mathbf{b}_4 = 4\mathbf{b}_1 - \mathbf{b}_3$

$\Rightarrow \mathbf{a}_1, \mathbf{a}_3, \mathbf{a}_5$ linearly independent, $\mathbf{a}_2 = 2\mathbf{a}_1$, $\mathbf{a}_4 = 4\mathbf{a}_1 - \mathbf{a}_3$

$\{\mathbf{a}_1, \mathbf{a}_3, \mathbf{a}_5\}$ basis of $\text{span}\{\mathbf{a}_1, \dots, \mathbf{a}_5\}$

$$\textcircled{1} \text{ Question: find a basis of span } \overbrace{\left\{ \begin{pmatrix} 2 \\ 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 4 \\ 2 \\ 4 \\ 6 \end{pmatrix}, \begin{pmatrix} 6 \\ 3 \\ 8 \\ 7 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \\ 0 \\ 5 \end{pmatrix}, \begin{pmatrix} 4 \\ 1 \\ 0 \\ 9 \end{pmatrix} \right\}}^S$$

which is a subset of S : answer $\{*1, *2, *3\}$.

Solution: – write the vectors in S as columns of a matrix A

– find reduced r.e.f. B of A

– choose the columns of A which correspond to columns of B with $\textcircled{1}$

$\textcircled{2}$ Question: Complete a linearly independent set of vectors $S \subseteq V$ to a basis of V .

– write S as columns of a matrix A

– append to A on the right as columns some generating set H of A , $A' = (A|H)$

– find reduced r.e.f. B of A'

– choose the columns of A' which correspond to columns with $\textcircled{1}$ of B :

$S \cup \{\dots\}$ is a basis

Example 3.39. $V = \{ (x_1, \dots, x_5) \in \mathbb{R}^5 : x_1 + 7x_2 + 5x_3 - 4x_4 + 2x_5 = 0 \}$
 $S = \{(-2, 0, 0, -1, -1), (1, 1, -2, -1, -1), (-5, 1, 0, 1, 1)\}$
 can check using ① that S is linearly independent
 i.e. if we write vectors in S into a matrix A
 then every column in r.r.e.f. B of A has a ①:

$$\begin{array}{ccc} \begin{pmatrix} -2 & 1 & -5 \\ 0 & 1 & 1 \\ 0 & -2 & 0 \\ -1 & -1 & 1 \\ -1 & -1 & 1 \end{pmatrix} & \longrightarrow & \begin{pmatrix} \textcircled{1} & 0 & 0 \\ 0 & \textcircled{1} & 0 \\ 0 & 0 & \textcircled{1} \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \\ A & & B \end{array}$$

find a generating set H of $V = \{x_1 = -7x_2 - 5x_3 + 4x_4 - 2x_5\}$

$$\begin{aligned} (x_1, x_2, x_3, x_4, x_5) &= (-7t_1 - 5t_2 + 4t_3 - 2t_4, t_1, t_2, t_3, t_4) = \\ &= t_1 \underbrace{\begin{pmatrix} -7 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}}_{*1} + t_2 \underbrace{\begin{pmatrix} -5 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}}_{*2} + t_3 \underbrace{\begin{pmatrix} 4 \\ 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}}_{*3} + t_4 \underbrace{\begin{pmatrix} -2 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}}_{*4} \quad H = \{*1, *2, *3, *4\} \end{aligned}$$

$$\begin{array}{ccc} \underbrace{\begin{pmatrix} -2 & 1 & -5 \\ 0 & 1 & 1 \\ 0 & -2 & 0 \\ -1 & -1 & 1 \\ -1 & -1 & 1 \end{pmatrix}}_S & \underbrace{\begin{pmatrix} -7 & -5 & 4 & -2 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}}_H & = A' \\ & & \downarrow \\ & & B' \text{ r.r.e.f.} \end{array}$$

$$\begin{pmatrix} \textcircled{1} & 0 & 0 & 1 & 1 & 0 & -1 \\ 0 & \textcircled{1} & 0 & 0 & -1/2 & 0 & 0 \\ 0 & 0 & \textcircled{1} & 1 & 1/2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \textcircled{1} & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

answer: basis of $V \supseteq S$ is

$$S \cup \{(4, 0, 0, 1, 0)\}$$

summary: we can determine rank
 determine inverse
 solve a LES
 complete a linearly independent set to a basis
 choose a basis from a generating set