

Exam 1 Friday in Ungar 411.
Office Hours Today & Tomorrow 3-4.



Today: Finish Proof of The
Fundamental Theorem:

$$\dim R(A) = \dim C(A)$$

$$\begin{array}{ccc} \text{row rank} & \text{row rank} = \text{column rank} & \text{column rank} \\ \nearrow & \text{row rank} & \nwarrow \\ \text{row rank} & = & \text{column rank} \end{array}$$

Strategy:

We showed for invertible E, F :

$$R(EA) = R(A)$$

$$R(AF) \cong R(A).$$

$$\begin{aligned} \text{So } \dim R(EA) &= \dim(AF) \\ &= \dim(EAF) \\ &= \dim R(A) = r_R \end{aligned}$$

Similarly,

$$C(AF) = C(A)$$

$$C(EA) \cong C(A)$$

$$\text{So } \dim C(AF) = \dots = r_c.$$

To complete the proof we will find inv. matrices E & F such that

$$EAF = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$$

The matrix is partitioned into four quadrants: top-left is I_r , top-right is 0 , bottom-left is 0 , and bottom-right is 0 . Brackets indicate dimensions: a top bracket over the entire matrix is labeled n ; a left bracket over the entire matrix is labeled m ; a bottom bracket under the first r columns is labeled r ; a bottom bracket under the remaining $n-r$ columns is labeled $n-r$; a right curly bracket over the top r rows is labeled r ; and a right curly bracket over the bottom $m-r$ rows is labeled $m-r$.

Then it follows that

$$\begin{aligned} r_R &= \dim R(EAF) \\ &= r \\ &= \dim C(EAF) \\ &= r_c. \end{aligned}$$

invertible:

$$D_i(\lambda)^{-1} = D_i(1/\lambda) \quad \lambda \neq 0$$

$$L_{ij}(\lambda)^{-1} = L_{ij}(-\lambda)$$

$$T_{ij}^{-1} = T_{ij}$$

Can use elem. mats. to perform
"row elimination" & "col elimination".

Recall:

(ith row EA) is a linear combo.
of rows of A . E exact:

$$(\text{ith row } E) = (e_i, e_{i2} \dots e_{in})$$

$$(\text{ith row } A) = \vec{a}_i^T$$

$$(\text{ith row } EA) = (\text{ith row } E) A$$

$$= (e_{i1} \dots e_{in}) \begin{pmatrix} -\vec{a}_1^T - \\ \vdots \\ -\vec{a}_n^T - \end{pmatrix}$$

$$= e_{i1} \vec{a}_1^T + e_{i2} \vec{a}_2^T + \dots + e_{in} \vec{a}_n^T \quad \checkmark$$

Now let $E =$ elementary.

$$\text{ith row } D_j(\lambda) A = \begin{cases} \vec{a}_i^T & j \neq i, \\ \lambda \vec{a}_j^T & j = i. \end{cases}$$

$$\text{ith row } L_{jk}(\lambda) A = \begin{cases} \vec{a}_i^T & j \neq i, \\ \vec{a}_j^T + \lambda \vec{a}_k & j = i. \end{cases}$$

$$\text{ith row } T_{jk} A = \begin{cases} \vec{a}_i & i \neq j, k, \\ \vec{a}_j & i = k, \\ \vec{a}_k & i = j. \end{cases}$$

Similar formulas for

$A E$, E elementary.

performs a column operation on A .

Finally, the reduction algorithm.

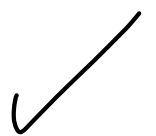
- IF $A = 0$ stop.
- Swap rows/cols to get nonzero entry in top left.
- Scale 1st row (or col) to make top left entry = 1:

$$\left(\begin{array}{c|cccc} 1 & * & - & * & * \\ \hline * & & & & \\ \vdots & & & & \\ * & & & & \end{array} \right)$$

- Apply L_{ij} matrices to left & right to eliminate entries in 1st row/col:

$$\left(\begin{array}{c|cccc} 1 & 0 & - & - & 0 & 0 \\ \hline 0 & & & & & \\ \vdots & & & & & \\ 0 & & & & & \end{array} \right) \begin{array}{c} \\ \\ A' \\ \end{array}$$

- Repeat on smaller matrix A' .



Stop when you hit a matrix
of zeros:

$$E_n \cdots E_1 A F_1 \cdots F_k$$

$$= \left(\begin{array}{c|c} 1 & 0 \\ \vdots & \\ \dots & \\ 1 & 0 \\ \hline 0 & 0 \end{array} \right)$$

QED.

Remark: IF $A \in \mathbb{Z}^{m \times n}$ then
a similar algorithm over \mathbb{Z} produces
(much harder)

$$EAF = \left(\begin{array}{c|c} d_1 & 0 \\ d_2 & \\ \vdots & \\ d_r & 0 \\ \hline 0 & 0 \end{array} \right)$$

where $d_1, \dots, d_r \in \mathbb{Z}$ satisfy

$$d_1 \mid d_2 \mid \dots \mid d_r.$$

Called the "Smith Normal Form".



Recall: Solving a linear system

$$A \vec{x} = \vec{b}$$

$$A \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$

is related to finding a left inverse
for A : $CA = I$.

$$A \vec{x} = \vec{b}$$

$$CA \vec{x} = C \vec{b}$$

$$\vec{x} = C \vec{b}.$$

In fact we can take C as
a product of elementary row ops:

$$E_1 (A | \vec{b})$$

$$(E_1 A | E_1 \vec{b})$$

$$\leadsto E_2 (E_1 A | E_1 \vec{b})$$

maybe don't
look at this
right now.
We'll discuss
later.

$$(E_2 E_1 A \mid E_2 E_1 \vec{b})$$

$\rightsquigarrow \dots$

$$(I \mid \underbrace{E_n \dots E_2 E_1}_{\text{solution}} \vec{b}),$$

\rightsquigarrow

Apply Fund. Thm. to matrix inversion. Note: Linear system

$$A \vec{x} = \vec{b} \text{ has a solution}$$

$$\iff \vec{b} \in C(A).$$

Indeed,

$$A \vec{x} = \begin{pmatrix} \vec{a}_1 & \dots & \vec{a}_n \\ | & & | \\ 1 & & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$= x_1 \vec{a}_1 + \dots + x_n \vec{a}_n$$

linear combo
of cols of A

More generally, matrix eqn

$$\begin{matrix} A & X & = & I_m \\ m \times n & n \times m & & m \times m \end{matrix}$$

$$A \begin{pmatrix} \begin{matrix} | \\ \vec{x}_1 \\ | \end{matrix} \cdots \begin{matrix} | \\ \vec{x}_m \\ | \end{matrix} \end{pmatrix} = \begin{pmatrix} \begin{matrix} | \\ \vec{e}_1 \\ | \end{matrix} \cdots \begin{matrix} | \\ \vec{e}_m \\ | \end{matrix} \end{pmatrix}$$

$$\left(\begin{matrix} | \\ A \vec{x}_1 \\ | \end{matrix} \cdots \begin{matrix} | \\ A \vec{x}_m \\ | \end{matrix} \right) = \left(\begin{matrix} | \\ \vec{e}_1 \\ | \end{matrix} \cdots \begin{matrix} | \\ \vec{e}_m \\ | \end{matrix} \right)$$

has a solution

$$\Leftrightarrow \vec{e}_i \in C(A) \text{ for all } i.$$

$$\Leftrightarrow C(A) = \mathbb{R}^m$$

$$\Leftrightarrow r_c = m$$

(full column-rank)

Similarly, matrix A has a left inverse

$$XA = I_n \Leftrightarrow R(A) = \mathbb{R}^n$$

$$\Leftrightarrow r_R = n$$

(full row-rank)

By the Fund. Thm. (A $m \times n$)

A has right inverse

$$\Leftrightarrow r_c = m$$

$$\Leftrightarrow r = m.$$

A has left inverse

$$\Leftrightarrow r_r = n$$

$$\Leftrightarrow r = n.$$

A has both left & right inverse

$$\Leftrightarrow m = r = n.$$

(A is a square matrix
of full rank)