## Exercise 3a) Why must more than three vectors in $\mathbb{R}^3$ be linearly dependent? consider C, Y, + G V, + C, V, + C, V, + ... + C, V, = 0 n34 vectors augmented matrix 3 { \[ \vec{v}\_1 \vec{v}\_2 \\ \vec{v}\_n \] \[ \frac{0}{0} \\ \rightarrow \] at most 3 pivots at least n-3 free variables <u>3b</u>) How about more than *m* vectors in $\mathbb{R}^m$ ? $m \left\{ \begin{bmatrix} \vec{\lambda}_1 & \vec{\lambda}_2 & \cdots & \vec{\lambda}_n \end{bmatrix} \right\}$ so, lots of dependencies. at most in pivots at clast nom free variables n>m So, lots of dependencies <u>3c</u>) If you are given a set of exactly n vectors in $\mathbb{R}^n$ how can you check whether or not they are linearly independent? n } | \vec{v}\_1 | \vec{v}\_2 | \dots \vec{v}\_n \ref | \oldsymbol{0} \cdots \oldsymbol{0} \cdots \oldsymbol{0} \dots \oldsymbol{ $\begin{array}{c} c_1 \vec{v}_1 + c_2 \vec{v}_2 + \ldots + c_n \vec{v}_n = \vec{0} \\ 3d) \text{ If you have a set of fewer than } n \text{ vectors in } \mathbb{R}^n \text{ can they span } \mathbb{R}^n? \\ \text{Friday Warm-ap} \\ \text{P<n:} \qquad x_1 \vec{v}_1 + x_2 \vec{v}_2 + \ldots + x_p \vec{v}_p = \vec{b} \in \mathbb{R}^n \\ \text{NO} \\ \text{NO} \end{array}$ n { [v, vz. vp | b] at most p [000-01ch] inconsistent unless ch=0. $\frac{3e}{n}$ If $\{\underline{v}_1, \underline{v}_2, \dots \underline{v}_n\}$ is a set of exactly n vectors in $\mathbb{R}^n$ what condition on the reduced row echelon form of $A = [v_1, v_2, \dots v_n]$ of $A = [v_1, \dots v_n]$ of A =of $A = [\underline{v}_1, \underline{v}_2, \dots \underline{v}_n]$ guarantees and is required so that the vectors span $\mathbb{R}^n$ ? Compare with $\underline{3c}$ . if exactly n vectors in R" need reef A = I in order for the vectors to span IR" (else there's a bottom Friday warm-up rows of O's in ref (A)

## Fri Sept 7

• 1.7 reduced row echelon form as encoding linear independence/dependence of matrix columns; introduction to linear transformations, section 1.8.

Announcements: returned Hw1 & Quiz 3

To some of the problems are graded

- see rubric on public the page

Solutions to non-text problems are on CANVAS.

'til 12:57 Warm-up Exercise: 3d, 3e; the last two exercises in yesterday's moles Exercise 1) Consider the homogeneous matrix equation A = 0, with the matrix A (and its reduced row echelon form) shown below:

$$A := \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 2 & 4 & 1 & 4 & 1 \\ -2 & -4 & 0 & -2 & -2 \end{bmatrix} \stackrel{\bigcirc}{O} \rightarrow \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} - \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{O} = \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \stackrel{\bigcirc}{$$

Find and express the solutions to this system in linear combination form. Note that you are finding all of the columns, the dependencies for the collection of vectors that are the columns of A, namely the set

for the collection of vectors that are the columns of 
$$A$$
, namely the set
$$\begin{bmatrix}
\mathbf{q}_1 \\
\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \\ -2 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} 2 \\ 4 \\ -4 \end{bmatrix}, \mathbf{v}_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \mathbf{v}_4 = \begin{bmatrix} 1 \\ 4 \\ -2 \end{bmatrix}, \mathbf{v}_5 = \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix}.$$
Let  $\mathbf{q}_5$  which disapped with how  $\mathbf{q}_5$  is a small probability  $\mathbf{q}_5$  and  $\mathbf{q}_5$  is a small probability  $\mathbf{q}_5$  and  $\mathbf{q}_5$  is a small probability  $\mathbf{q}_5$ .

 $\begin{array}{lll} x_{1} = -2 \times_{2} - x_{4} - x_{5} & & & \\ x_{2} = x_{2} & \text{free} & & & \\ x_{3} = -2 \times_{4} + x_{5} & & & \\ x_{4} = x_{4} & \text{free} & & \\ x_{5} = x_{5} & \text{free} & & \\ \end{array}$ 

these homog. sol'this also encode column dependencies

e.g. 
$$x_z=1$$
,  $x_y=x_s=0$ 
 $\begin{array}{c} -2\vec{a}_1+1\vec{a}_2=\vec{0} \\ -2\vec{a}_1+1\vec{a}_2=\vec{0} \end{array}$ 
 $\begin{array}{c} v_z=2\vec{v}_1 \\ v_z=2\vec{v}_1 \\ v_z=2\vec{v}_1 \end{array}$ 

e.g.  $x_z=x_s=0$ ,  $x_y=1$ 
 $\begin{array}{c} -2\vec{a}_1+2\vec{a}_2=\vec{0} \\ -2\vec{a}_1+2\vec{v}_2=\vec{0} \end{array}$ 
 $\begin{array}{c} -2\vec{a}_1+2\vec{v}_2=\vec{0} \\ -2\vec{v}_1+2\vec{v}_2 \end{array}$ 
 $\begin{array}{c} v_z=2\vec{v}_1 \\ v_y=\vec{v}_1+2\vec{v}_2 \end{array}$ 
 $\begin{array}{c} v_z=2\vec{v}_1 \\ v_y=2\vec{v}_1+2\vec{v}_2 \end{array}$ 

Exercise 2) Focus on the idea that solutions to homogeneous matrix equations correspond exactly to linear dependencies between the columns of the matrix. Now, since the solution set to a homogeneous linear system does not change as you do elementary row operations to the augmented matrix, column dependencies also do not change. Therefore the vectors that span the space of homogeneous solutions in Exercise 1 are encoding the key column dependencies in  $\mathbb{R}^3$ , for both the original matrix, and for the reduced row echelon form.

Check this, by reading off "easy" column dependencies in the reduced matrix; seeing that they are also dependencies in the original matrix; and that they correspond to the basis of the homogeneous solution space. Magic! We will use this magic in important interesting ways, later in the course.

$$A := \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 2 & 4 & 1 & 4 & 1 \\ -2 & -4 & 0 & -2 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\vec{\nabla}_{1} \vec{\nabla}_{2} \vec{\nabla}_{3} \vec{\nabla}_{4} \vec{\nabla}_{5} \qquad \vec{\nabla}_{2} = 2 \vec{\nabla}_{1} \qquad \begin{bmatrix} 2 & 0 & 1 & 1 \\ 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\vec{\nabla}_{1}, \vec{\nabla}_{2} \vec{\nabla}_{3} \vec{\nabla}_{4} \vec{\nabla}_{5} \qquad \vec{\nabla}_{2} = 2 \vec{\nabla}_{1} \qquad \begin{bmatrix} 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\vec{\nabla}_{1}, \vec{\nabla}_{3} \text{ are ind.} \qquad \vec{\nabla}_{2} = 2 \vec{\nabla}_{1} \qquad \begin{bmatrix} 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\vec{\nabla}_{1}, \vec{\nabla}_{3} \text{ are ind.} \qquad \vec{\nabla}_{2} = 2 \vec{\nabla}_{1} \qquad \begin{bmatrix} 2 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\vec{\nabla}_{1}, \vec{\nabla}_{3} \text{ are ind.} \qquad \vec{\nabla}_{4} = \vec{\nabla}_{1} + 2 \vec{\nabla}_{2} \vec{\nabla}_{3} \qquad \vec{\nabla}_{4} = \vec{\nabla}_{1} + 2 \vec{\nabla}_{3} \vec{\nabla}_{4} \qquad \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{3} \qquad \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{2} \qquad \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{2} = \vec{\nabla}_{1} - \vec{\nabla}_{3} \qquad \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{5} = \vec{\nabla}_{1} - \vec{\nabla}_{2} = \vec{\nabla}_{1} - \vec{\nabla}_{3} = \vec{\nabla}_{1} - \vec{\nabla}_{2} = \vec{\nabla}_{1} - \vec{\nabla}_{3} = \vec{\nabla}_{1} - \vec{$$

Exercise 3) (This exercise explains why each matrix has only one reduced row echelon form, no matter what sequence of elementary row operations one uses to find it. We didn't have the tools to explain why this fact was true earlier in the chapter) Let  $B_{4 \times 5}$  be a matrix whose columns satisfy the following dependencies:

$$col_1(B) \neq \underline{\mathbf{0}} \quad \text{(i.e. is independent)}$$
 
$$col_2(B) = 3 \ col_1(B)$$
 
$$col_3(B) \text{ is independent of column 1}$$
 
$$col_4(B) \text{ is independent of columns 1,3.}$$
 
$$col_5(B) = -3 \ col_1(B) + 2 \ col_3(B) - col_4(B).$$
 What is the reduced row echelon form of  $B$ ?

$$B = \begin{bmatrix} \vec{b}_1 & \vec{b}_2 & \vec{b}_3 & \vec{b}_4 & \vec{b}_5 \end{bmatrix} \implies \text{ref}(B) = \begin{bmatrix} 1 & 3 & 0 & 0 & -3 \\ 0 & 0 & 1 & 0 & 2 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\xrightarrow{-3} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + 2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

What if: 1300
0010 would violate
0000 ref conditions!

depending on Sth
column would
either violate

\* zero rows at bottom
or \* pivots move to
right as you go
down the rows

1.8 Introduction to linear transformations.

<u>Definition</u>: A function T which has domain equal to  $\mathbb{R}^n$  and whose range lies in  $\mathbb{R}^m$  is called a *linear* transformation if it transforms sums to sums, and scalar multiples to scalar multiples. Precisely,  $T: \mathbb{R}^n \to \mathbb{R}^m$  is *linear* if and only if

$$T(\underline{\boldsymbol{u}} + \underline{\boldsymbol{v}}) = T(\underline{\boldsymbol{u}}) + T(\underline{\boldsymbol{v}}) \qquad \forall \underline{\boldsymbol{u}}, \underline{\boldsymbol{v}} \in \mathbb{R}^n T(c\underline{\boldsymbol{u}}) = c T(\underline{\boldsymbol{u}}) \qquad \forall c \in \mathbb{R}, \underline{\boldsymbol{u}} \in \mathbb{R}^n .$$

Notation In this case we call  $\mathbb{R}^m$  the *codomain*. We call  $T(\underline{\boldsymbol{u}})$  the *image of*  $\underline{\boldsymbol{u}}$ . The *range* of T is the collection of all images  $T(\underline{u})$ , for  $\underline{u} \in \mathbb{R}^n$ .

<u>Important connection to matrices:</u> Each matrix  $A_{m \times n}$  gives rise to a linear transformation  $T: \mathbb{R}^n \to \mathbb{R}^m$ , namely  $T(\underline{x}) := A \underline{x} \qquad \forall \ \underline{x} \in \mathbb{R}^n.$ 

This is because

Theorem 5 (p. 39 Matrix multiplication is linear) If A is an  $m \times n$  matrix,  $\underline{\boldsymbol{u}}, \underline{\boldsymbol{v}} \in \mathbb{R}^n$ , c a scalar, then

a) 
$$A(\underline{u} + \underline{v}) = A \underline{u} + A \underline{v}$$

b) 
$$A(c \mathbf{u}) = c A \mathbf{u}$$

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b)  $A(c \underline{u}) = c A \underline{u}$ 

b)  $A(c \underline{u}) = c A \underline{u}$ 

$$A(\underline{v} + \underline{v}) = a \underbrace{u}_{n-1} + a \underbrace{u}_{n-$$

Remark: One reason that the word "linear" is appropriate for these sorts of functions is that linear transformations transform lines to lines (or points); and families of parallel lines are transformed into families of parallel lines (or points). to be untinued