## 1 9/5 Matrices, vectors, and their applications

Algebra: study of objects and operations on them.

Linear algebra: object: matrices and vectors. operations: addition, multiplication etc.

Algorithms/Geometric intuition/sets and maps

 $m \times n$  matrix: numbers forming a rectangular grid, m rows and n columns. Motivation: coefficients of a system of linear equations. Data tables in statistics.

(i, j)-th entry of a matrix.

Vectors: matrices with one row/column. Motivation: coordinates in plane and space.

Operations: (1) Addition. (2) Scalar multiplication. (3) Matrix-vector multiplication. (4) Transpose.

Example: 
$$A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$$
,  $x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .  $Ax$ ,  $A(Ax)$ .

Example: Averaging over columns. Covariance? Other statistical concepts?

Laws: The usual laws one may expect. e.g. A(x+y) = Ax + Ay,  $(A+B)^T = A^T + B^T$ ,  $(A^T)^T = A$ . Note:  $A(Bx) \neq B(Ax)$ !

Zero and one matrix. Standard vectors.

Example: Rotation by 60 degrees (or  $\pi/3$ ).

Consequence: Matrix is completely determined by its action on the standard vectors! Matrix-matrix multiplication.

Example:  $2 \times 2$  case.

The concept of linear combination. Relationship with matrix-vector multiplication.

Example: Rotation and Translation.

Example: Random walk on graphs.

# 2 9/8 Linear equations

Review:

- Matrix multiplications
- Transposes
- Standard vectors
- Identity Matrix
- Rotation matrix

• Stochastic matrix

\*\*\*\*

Linear systems as matrix equations. Coefficient matrix and augmented matrices

Elementary row operations: swap, multiply, add. Property: reversible, and preserves solution set.

Row echelon form: The first non-zero entry (called pivot) of each row is to the right of the previous. Reduced row echelon form: The first non-zero entry is 1 and is the only non-zero entry in that column. Uniqueness under row operations.

Algorithm (Gaussian elimination):

- Write augmented matrix.
- Use row operations, turn it into reduced echelon form.
- General solution from RREF (Example:  $x_1 + 2x_2 + x_3 + x_4 = 3$ ,  $x_1 + 3x_3 x_4 = 8$ ).

		Pivot at last col.	No pivot at last col.
Structure of solutions:	All coefficient col. have pivot	None	Inf
	Some coeff. col. have no pivot	None	One
Examples of the 4 cases.			

True or false:

- A system of 3 linear equations with 6 variables can not have just one solution.
- A system of 3 linear equations with 6 variables must have infinitely many solutions.

\*\*\*\*

Counting: number of arbitrary constants and the number of pivots. Rank and dimension.

Explicit algorithm from RREF to general solutions.

# 3 9/12 Linear equations cont.

### 3.1 Review

- Augmented matrix, row operations.
- RREF.
- Condition for no/one/infinitely many solutions.
- General solution: write basic variables in terms of free variables, or the vector form.

### 3.2 Gaussian elimination

Augment matrices to REF or RREF through finitely many elementary row operations.

For r=1, 2, ... n:

Find the left-most non-zero entry among the  $r, r+1, \ldots n$  rows. If there aren't any, terminate. Exchange rows to move this entry to the r-th row.

Multiply the r-th row and add it to the  $r+1, \ldots$  rows to eliminate all entries on the left-most non-zero column.

To Further turn it into a RREF (backward pass):

Multiply to each non-zero row to make the first entry 1.

For each non-zero row, multiply and add it to each of the rows above it to turn the entries on pivot columns 0.

Reason for distinguish forward/backward passes: forward pass is a permutation matrix with a lower triangular matrix with 1 on the diagonals, backward pass is a upper triangular matrix. Row pivoting.

Example: 
$$\begin{pmatrix} 0 & 1 & 2 & 3 \\ 2 & 2 & 4 & 7 \\ 2 & 0 & 1 & 0 \end{pmatrix}$$
. RREF? General solution?

## 3.3 Uniqueness of RREF

Key idea: read the RREF from matrix using linear combinations of rows or columns!

Appendix E uses columns. One can also use rows as follows: Let R be the space of linear combination (span) of the row vectors. The last non-zero row in RREF is the one in R with the most number of 0 entries on the left and the first non-zero entry 1. Let the index of the first non-zero entry be  $c_1$ . The preceding row in RREF is the one in R with  $c_1$ -th entry 0, first non-zero entry 1, and the most possible number of 0 on the left, etc.

### 3.4 Rank and nullity

Rank of A: num. of pivots in A=num. of non-zero rows in REF of A=num of basic variables in Ax = bNullity of A: num of non-pivot columns in A=num. of columns of A-rank of A=num of free variables in Ax = b

True or false:

- The rank of  $[A \ B]$  must be no smaller than the sum of the ranks of A and B.
- The nullity of  $[A \ B]$  must be no smaller than the sum of the ranks of A and B.
- The RREF of a square matrix of no nullity must be the identity matrix
- The nullity of A is non-zero iff some row of A is a linear combination of the others.
- $[A \ B]$  has the same rank as B iff the columns of A are linear combinations of the columns of B.

Structure of the general solution in terms of rank or nullity:

```
If rank(A) < rank([A, b]):
```

No solution.

Else:

\*\*\*\*\*

If nullity(A) = 0:

One solution.

Else:

Infinitely many solutions.

Example: 
$$\begin{pmatrix} a & b & c \\ e & f & g \end{pmatrix}$$
.

## 4 9/15 Span

#### Review:

- Augmented matrix and row operations
- REF, RREF, pivot
- free and basic variables
- Rank and Nullity

\*\*\*\*\*\*

Linear combination: S is a set of matrices of the same size, v is called a linear combination of S iff there exist finitely many matrices  $A_1 
ldots A_n$  in S, and scalars  $a_1, 
ldots a_n$ , so that  $v = \sum_k a_k A_k$ .

Span: The span of a set is the set of all linear combinations of that set. S is called a generating set of the set Span(S).

Example: span of the standard vectors.

Span closed under addition and scalar multiplication.

Transitivity.

b is in the span of columns of A iff Ax = b has a solution.  $\mathbb{R}^n$ : all vectors of n entries. Span is  $\mathbb{R}^n$  iff matrix is full rank iff ...

Example: use linear equation to detect spans.

Implication on the rank of the matrices while adding columns.

\*\*\*\*

Algorithm for minimal generating set. Example.

True or false:

Row operation changes the span of the column vectors.

A matrix is in REF, then the span of the columns are the span of some standard vectors.

# 5 9/18 Linear dependency

### 5.1 Review

Notation: when A and B has the same number of rows, by  $[A \ B]$  we mean a larger matrix formed by stacking them together horizontally.

relationship between matrices, system of equations Ax = b, and the column vectors:

The followings are equivalent:

- Ax = b has a solution (is **consistent**).
- b lies in the **span** of the columns vectors of A.

- The span of the columns of A is the same as the span of the columns of A and b.
- $rank([A \ b]) = rank(A)$ .
- Nullity( $[A \ b]$ ) = Nullity(A) + 1.
- In the **RREF** of  $[A \ b]$ , the last column does not contain a **pivot**.

Examples.

The followings are equivalent:

- Ax = b has a solution (is **consistent**) for all b.
- The span of the columns vectors of A is  $\mathbb{R}^m$ .
- $\operatorname{rank}(A) = m$ .
- Nullity(A) = n m.
- In the **RREF** of A, every row contain a **pivot**.
- The **RREF** of A does not contain zero rows.

Examples.

## 5.2 Linear dependence/independence

A set S is called **linearly independent**, if for any sequence of distinct elements  $x_1, \ldots x_k \in S$ ,  $c_1x_1 + \ldots c_kx_k = 0$  implies that  $c_1 = c_2 = \cdots = 0$ . If a set is not linearly independent it is linearly dependent.

 $a_1 
dots a_n$  are linearly dependent if and only if  $[a_1 
dots a_n]x = 0$  (the **homogeneous eq.**) has one (hence infinitely many) non-zero solutions. (hence  $[a_1 
dots a_n]x = b$  has infinitely many solutions for some b, hence has free variables, hence the nullity of A is non-zero).

Example: 1 or 2 vectors.

Linear dependency in standard vectors.

Linear dependency in RREF.

Linear dependency in vector form of the general solution.

### 5.3 Number of rows and columns

m > n: column vectors may or may not be linearly dependent, but can never span  $\mathbb{R}^m$ .

m < n: column vectors may or may not span  $\mathbb{R}^m$ , but can never be linearly independent.

m=n: column vectors span  $\mathbb{R}^m$  iff they are linearly independent.

## 5.4 Adding and removing vectors

If S is linearly independent, any subset of S is linearly independent and has a smaller span,  $S \cap \{v\}$  is linearly independent iff v is in the span of S.

If S is linearly dependent, so is any set larger than S.

Examples.

\*\*\*\*\*\*Optional\*\*\*\*\*

Row vectors under row operation.

Rank=num. of linearly independent column vectors.

Vertical stacks of matrices.

Relationship between homogeneous and non-homogeneous equations.

## 6 9/22 Review of Chapter 1, more on matrix multiplication

Important concepts to remember:

- Matrix
- Identity matrix
- Zero matrix
- Vector
- Standard vectors
- $\bullet$   $\mathbb{R}^n$
- Scalar multiplication
- Addition
- Multiplication
- Transpose
- Symmetric matrix
- System of linear equation
- Homogeneous equation
- Augmented matrix
- Coefficient matrix
- Row operation
- REF

- RREF
- Pivot
- Rank
- Nullity
- Free variable
- Basic variable
- General solution
- General solution in vector form
- Linear combination
- Span
- ullet Linear independence

True or false:

A set of 3 vectors in  $\mathbb{R}^2$  can not be linearly independent. A set of 3 vectors in  $\mathbb{R}^3$  is either linearly dependent or spans  $\mathbb{R}^3$ .

The idenitiy matrix is in REEF.

3 linear equations with 4 variables can not have finitely many solutions.

If the nullity of A is greater than 0, then Ax = 0 has infinitely many solutions.