# Lecture 3 (Chapter 2)

Linear Algebra, Course 124A, Fall, 2009

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Ch. 2: Lec. 2

Solving  $A\vec{x} = \vec{k}$ 

# Solving $A\vec{x} = \vec{b}$ :

- ▶ We (people + computers) solve systems of linear equations by a systematic method of Elimination followed by Back substitution
- ▶ Due to our man Gauss, hence Gaussian elimination.
- ▶ Our first example:

$$\begin{array}{rcl}
-x_1 & + & 3x_2 & = & 1 \\
2x_1 & + & x_2 & = & 5
\end{array}$$

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

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#### Gaussian elimination:

## Basic elimination rules (roughly):

- 1. Strategically, mechanically remove unwanted entries by subtracting a multiple of a row from another.
- 2. Swap rows if needed to create an 'upper triangular form'

e.g.

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

# Gaussian elimination:

Solve:

$$2x_1-3x_2=3$$

$$4x_1 - 5x_2 + x_3 = 7$$

$$2x_1 - x_2 - 3x_3 = 5$$

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Frame 4/8



### Gaussian elimination:

## Summary:

Using row operations, we turned this problem:

$$A\vec{x} = \vec{b} : \begin{bmatrix} 2 & -3 & 0 \\ 4 & -5 & 1 \\ 2 & -1 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 7 \\ 5 \end{bmatrix}$$

into this problem:

$$U\vec{x} = \vec{d} : \begin{bmatrix} 2 & -3 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & -5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ 0 \end{bmatrix}$$

and the latter is easy to solve using back substitution.

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#### Gaussian elimination:

#### Defn:

The entries along *U*'s main diagonal are the pivots of *A*. (The pivots are hidden—elimination finds them.)

#### Defn:

A matrix with only zeros below the main diagonal is called upper triangular. A matrix with only zeros above the main diagonal is called lower triangular. We get from A to U and the latter is always upper triangular.

#### Defn:

Singular means a system has no unique solution.

- It may have no solutions or infinitely many solutions.
- Singular = archaic way of saying 'messed up.'

#### Truth:

If at least one pivot is zero, the matrix will be singular. (but the reverse is not necessarily true).

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### Gaussian elimination:

#### The one true method:

- We simplify A using elimination in the same way every time.
- ► Eliminate entries one column at a time, moving left to right, and down each column.

$$X + X + X + X = X$$
  
 $1 \downarrow + X + X + X = X$   
 $2 \downarrow + 4 \downarrow + X + X = X$   
 $3 \nearrow + 5 \rightarrow + 6 + X = X$ 

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

Frame 7/8

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# Gaussian elimination:

- ▶ To eliminate entry in row i of jth column, subtract a multiple  $\ell_{ij}$  of the jth row from i.
- For example:

$$2x_1 + 3x_2 + -2x_3 + x_4 = 1$$

$$x_1 - 7x_2 + 3x_3 + x_4 = 1$$

$$-x_1 - 3x_2 - x_3 + 5x_4 = -2$$

$$2x_1 + x_2 - 2x_3 + 2x_4 = 0$$

$$\ell_{21} = 1/2, \, \ell_{31} = -1/2, \, \ell_{41} = ?.$$

- Note: we cannot find  $\ell_{32}$  etc., until we are finished with row 1. Pivots are hidden!
- Note: the denominator of each  $\ell_{ij}$  multiplier is the pivot in the *j*th column.

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

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