CAS CS 132

Gaussian Elimination (+ Numerics) Geometric Algorithms Lecture 4

Practice Problem

Write down the general forms solution of the above linear system.

$x + z = 1$ *x* + *y* + 3*z* = 3 *x* − *y* − *z* = − 1

Solution

 $x + z = 1$ $x + y + 3z = 3$ *x* − *y* − *z* = − 1

Objectives

- 1. (Finally) discuss Gaussian elimination
- 2. Think more carefully about number representations
- 3. Look at the consequences of floating point representations
- 4. Introduce NumPy and talk about best best practices

Keywords

forward elimination back substitution floating point numbers $IEEE-754$ relative error numpy.isclose ill-conditioned problems

Defining the Gaussian Elimination (GE) Algorithm

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eliminations + back-substitution

eliminations + back-substitution *we've already done this*

eliminations + back-substitution *we've already done this* the algorithm as pseudocode

but we'll take one step further and write down

- eliminations + back-substitution *we've already done this*
- but we'll take one step further and write down the algorithm as pseudocode
- **Keep in mind.** How do we turn our intuitions into a formal procedure?

The details of Gaussian elimination are tricky.

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The goal is not to understand it entirely, but

-
- to get enough intuition to emulate it.
- **You should roughly use Gaussian Elimination when solving a system by hand.**

demo (step-throughs)

The Algorithm

Gaussian Elimination (Specification)

FUNCTION GE(A): # **INPUT:** m × n matrix A # **OUTPUT:** equivalent m × n RREF matrix

...

Gaussian Elimination (High Level)

 FUNCTION fwd_elim(A): # **INPUT:** m × n matrix A # **OUTPUT:** equivalent m × n echelon form matrix ...

 FUNCTION back_sub(A): # **INPUT:** m × n echelon form matrix A # **OUTPUT:** equivalent m × n RREF matrix ...

 FUNCTION GE(A): **RETURN** back_sub(fwd_elim(A))

Elimination Stage

Elimination Stage (High Level)

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Input: matrix A of size $m \times n$ **Output:** echelon form of *A*

Elimination Stage (High Level)

Input: matrix *A* of size *m* × *n*

Output: echelon form of *A*

starting at the top left and move down, find a

leading entry and eliminate it from latter equations

-
-

What if the first equation doesn't have the variable *x*1?

What if the first equation doesn't have the variable x_1 ?

Swap rows with an equation that does.

What if the first equation doesn't have the variable *x*1?

What if *none* of the equations have the variable x_1 ?

Swap rows with an equation that does.

- What if the first equation doesn't have the variable x_1 ?
- **Swap rows with an equation that does.**
- x_1 ?
- **of the remaining equations.**

What if *none* of the equations have the variable

Find the *leftmost* **variable which appears in** *any*

FUNCTION fwd_elim(A):

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FOR [i from 1 to m]: # for each row from top to bottom

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-
- **IF** [rows i...m are all-zeros]: # if remaining rows are zero

Elimination Stage (Pseudocode) FUNCTION fwd_elim(A): **FOR** [i from 1 to m]: # for each row from top to bottom **IF** [rows i...m are all-zeros]: # if remaining rows are zero **RETURN** A

-
-

FUNCTION fwd_elim(A):

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Elimination Stage (Pseudocode) FUNCTION fwd_elim(A): **FOR** [i from 1 to m]: # for each row from top to bottom **IF** [rows i...m are all-zeros]: # if remaining rows are zero **RETURN** A **ELSE**:

 (j, k) ← [position of leftmost entry in the rows i...m]

-
-

FUNCTION fwd_elim(A):

- **FOR** [i from 1 to m]: # for each row from top to bottom
	- **IF** [rows i...m are all-zeros]: # if remaining rows are zero

 (j, k) ← [position of leftmost entry in the rows i...m] [swap row i and row j]

RETURN A

ELSE:
Elimination Stage (Pseudocode)

FUNCTION fwd_elim(A):

- **FOR** [i from 1 to m]: # for each row from top to bottom
	- **IF** [rows i...m are all-zeros]: # if remaining rows are zero **RETURN** A

(j, k) \leftarrow [position of leftmost entry in the rows i...m] [swap row i and row j] **FOR** [l from i + 1 to m]: # for all remaining rows

ELSE:

Elimination Stage (Pseudocode)

FUNCTION fwd_elim(A):

 FOR [i from 1 to m]: # for each row from top to bottom **IF** [rows i...m are all-zeros]: # if remaining rows are zero

RETURN A

ELSE:

 (j, k) ← [position of leftmost entry in the rows i...m] [swap row i and row j] **FOR** [l from i + 1 to m]: # for all remaining rows [zero out A[l, k] using a replacement operation]

Elimination Stage (Pseudocode)

FUNCTION fwd_elim(A):

 FOR [i from 1 to m]: # for each row from top to bottom **IF** [rows i...m are all-zeros]: # if remaining rows are zero

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ELSE:

- (j, k) ← [position of leftmost entry in the rows i...m] [swap row i and row j]
- **FOR** [l from i + 1 to m]: # for all remaining rows [zero out A[l, k] using a replacement operation]
	-

RETURN A

leftmost nonzero entry

Swap R_1 and R_3

leftmost nonzero entry

−9 12 −9 6 15 −7 8 −5 8 9 3 −6 6 4 −5

3 −9 12 −9 6 15 3 −7 8 −5 8 9 0 3 −6 6 4 −5

next entry to zero

3 −9 12 −9 6 15 3 −7 8 −5 8 9 0 3 −6 6 4 −5 next entry to zero

 $R_3 \leftarrow R_3 - R_1$

−9 12 −9 6 15 2 −4 4 2 −6 3 −6 6 4 −5

3 −9 12 −9 6 15 0 2 −4 4 2 −6 0 3 −6 6 4 −5

leftmost nonzero entry

3 −9 12 −9 6 15 0 2 −4 4 2 −6 0 3 −6 6 4 −5 leftmost nonzero entry

swap *R*2 with *R*2

−9 12 −9 6 15 2 −4 4 2 −6 3 −6 6 4 −5

next entry $\begin{bmatrix} 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$

3 −9 12 −9 6 15 0 2 −4 4 2 −6

next entry $\begin{bmatrix} 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$

3 −9 12 −9 6 15 0 2 −4 4 2 −6

 $R_3 \leftarrow R_3 - \frac{3R_2}{2}$ 2

−9 12 −9 6 15 2 −4 4 2 −6

0 0 0 1 4

0 0 0 0 1 4 leftmost nonzero entry

3 −9 12 −9 6 15 0 2 −4 4 2 −6

3 −9 12 −9 6 15 0 2 −4 4 2 −6

0 0 0 0 1 4 leftmost nonzero entry

swap *R*3 with *R*3

−9 12 −9 6 15 2 −4 4 2 −6

0 0 0 1 4

3 −9 12 −9 6 15 0 2 −4 4 2 −6

0 0 0 0 1 4

done with elimination stage going to back substitution stage Back Substitution Stage

Back Substitution Stage (High Level)

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Input: matrix *A* of size *m* × *n* in echelon form **Output:** reduced echelon form of *A*

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Back Substitution Stage (High Level)

Input: matrix *A* of size *m* × *n* in echelon form **Output:** reduced echelon form of *A* scale pivot positions and eliminate the variables for that column from the other equations

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FUNCTION back_sub(A):

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-
-

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-
-
-

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- FOR [k from 1 to i 1]: # for the rows above the current one

 FUNCTION back_sub(A): **FOR** [i from 1 to m]: # for each row from top to bottom **IF** [row i has a leading entry]: j ← index of leading entry of row i $R_i(\textsf{A}) \leftarrow R_i(\textsf{A})$ / $\textsf{A}[\texttt{i}, \texttt{j}]$ # divide by leading entry $R_k(A) \leftarrow R_k(A) - R[k, j] \times R_i(A)$

-
-
-
- FOR [k from 1 to i 1]: # for the rows above the current one
	-
	- # zero out R[k, j] above the leading entry

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RETURN A

-
-
-
- FOR [k from 1 to i 1]: # for the rows above the current one
	-
	-

You will have to implement this part in HW2...

Gaussian Elimination (Example)

−9 12 −9 6 15 2 −4 4 2 −6

0 0 0 1 4

Gaussian Elimination (Example)

3 −9 12 −9 6 15 0 2 −4 4 2 −6

0 0 0 0 1 4

pivot position
3 −9 12 −9 6 15 0 2 −4 4 2 −6

 $R_1 \leftarrow R_1 / 3$

0 0 0 0 1 4

pivot position

pivot position

1 −3 4 −3 2 5 0 2 −4 4 2 −6

 $R_2 \leftarrow R_2 / 2$

0 0 0 0 1 4 pivot position

next entry to zero

1 −3 4 −3 2 5 0 1 −2 2 1 −3 0 0 0 0 1 4 next entry to zero

 $R_1 \leftarrow R_1 + 3R_2$

pivot
psition $\begin{bmatrix} 0 & 0 & 0 & 1 & 4 \end{bmatrix}$ position

 $R_3 \leftarrow R_3 / 1$

position

next entry to zero

 $R_2 \leftarrow R_2 - R_1$

next entry to zero

next entry to zero

 $R_1 \leftarrow R_1 - 5R_3$

next entry to zero

$\overline{}$

1 0 −2 3 0 −24 0 1 −2 2 0 −7 $\begin{bmatrix} 0 & 1 & 2 & 2 & 0 \\ 0 & 0 & 0 & 1 & 4 \end{bmatrix}$

$\overline{}$

1 0 −2 3 0 −24 0 1 −2 2 0 −7 $\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$ 4

done with back substitution phase

Question

$\overline{}$

1 0 −2 3 0 −24 0 1 −2 2 0 −7 $\begin{bmatrix} 0 & 1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 1 & 4 \end{bmatrix}$

Write down the general form solution from the given RREF.

Solution

\mathbf{I} $1 \quad 0 \quad -2 \quad 3 \quad 0 \quad -24$ 0 1 -2 2 0 -7 0 0 0 0 1 4

Solution

 $x₃$ is free *x*⁴ is free $x_{5} = 4$

$x_1 = (-24) + 2x_3 - 3x_4$ $x_2 = (-7) + 2x_3 - 2x_4$

-
-

1. Write your system as an augmented matrix

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2. Find the RREF of that matrix

-
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2. Find the RREF of that matrix

3. Read off the solution from the RREF

-
-
-

1. Write your system as an augmented matrix

2. Find the RREF of that matrix **Gaussian elimination**

3. Read off the solution from the RREF

Numerics

demo (mini-GE)

Have you ever been docked points in a science

class for having incorrect sig figs?

Have you ever been docked points in a science class for having incorrect sig figs?

when you use a ruler, you can't do better than ± 1 mm, so we can't say anything about nanometer differences

- *Have you ever been docked points in a science class for having incorrect sig figs?*
- when you use a ruler, you can't do better than ± 1 mm, so we can't say anything about nanometer differences
- we run into a similar problem with decimal numbers in programs

your computer is a collection of fixed size registers

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each register holds a sequence of bits

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each register holds a sequence of bits

The Goal. represent numbers so they fit in those registers
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each register holds a sequence of bits

The Goal. represent numbers so they fit in those registers

this is, of course, a lie an abstraction

Question. How do we slice up our fixed sequence to represent numbers?

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things to consider:

- simple idea (easy to understand)
- maximize coverage (not too redundant)
- simple numeric operations (easy to use)

Unsigned Integers

binary value (we should know this by now) e.g. 10001010 represents $1(2^7) + 0(2^6) + 0(2^5) + 0(2^4) + 0(2^3) + 1(2^2) + 0(2^1) + 1(2^0)$

value

Signed Integers

sign bit + binary value e.g. 10001010 represents $-1 \times (0(2^6) + 0(2^5) + 0(2^4) + 0(2^3) + 1(2^2) + 0(2^1) + 1(2^0))$

sign value

-
-
- -
-
-

floats in python use 64 bits

floats in python use 64 bits That's 1.8×10^{19} possible values

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We can't represent everything. We'll have to choose and then round

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We can't represent everything. We'll have to choose and then round

Question. Which ones should we represent?

Integers work because they are **discrete and evenly spaced**

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What if we evenly discretize a range of values?

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What if we evenly discretize a range of values?

i.e., represent

..., -0.001, 0, 0.0001, 0.002, 0.003, 0.004,...

Question

Discuss the advantages and disadvantages of this approach

like scientific notation, but binary

like scientific notation, but binary the equation:

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fra}}{2}\right)
$$

like scientific notation, but binary the equation:

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fra}}{2}\right)
$$

it's an accepted standard, not perfect, but it works well

Question

Any ideas why this is better/worse? And why not have a sign bit for the exponent?

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2)}
$$

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2^{10}-1)}
$$

Definition. step size is the space between two floating-point representations

 $(-1)^{\text{sign}} \times (1 +$ fraction $\frac{252}{252}$ \times 2^{exponent-(2¹⁰-1)}

step size increases with magnitude \longrightarrow

for fixed exponent *n* two numbers are at least $0.00...001 \times 2^n = 2^{-52} \times 2^n$

Definition. step size is the space between two floating-point representations

away (why?)

 $(-1)^{\text{sign}} \times (1 +$ fraction $\frac{252}{252}$ \times 2^{exponent-(2¹⁰-1)}

step size increases with magnitude \longrightarrow

for fixed exponent *n* two numbers are at least $0.00...001 \times 2^n = 2^{-52} \times 2^n$

Definition. step size is the space between two floating-point representations

away (why?)

 $(-1)^{\text{sign}} \times (1 +$ fraction $\frac{252}{252}$ \times 2^{exponent-(2¹⁰-1)}

step size increases with magnitude \longrightarrow

Step size doubles for each exponent **image source**

IEEE-754 defines a subset of decimal numbers

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operations on floating point numbers attempt to give you the closest to the actual value, though there will be errors.

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operations on floating point numbers attempt to give you the closest to the actual value, though there will be errors.

we can assume when we write down a number like '0.3' we get the closest IEEE-754 value

Relative Error

massive for 10−²⁰

Observation. ± 0.001 is $tiny$ error for 10^{20} but

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Relative Error.

 $err_{rel} =$ err val

Observation. ± 0.001 is $tiny$ error for 10^{20} but

Relative Error

massive for 10−²⁰

Relative Error.

 $err_{rel} =$

IEEE-754 keeps relative error small

Observation. ± 0.001 is $tiny$ error for 10^{20} but

err val

Relative Error (Calculation)

(fix an exponent *n*)

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$

error is determined by step-size

Relative Error (Calculation)

(fix an exponent *n*)

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}} \right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$

$err \leq 2^{-52} \times 2^n$

Relative Error (Calcu

(fix an exponent *n*)

 1.0×2^n

the smallest number we can represent at least

 $val \geq 1.0 \times 2^n$

(why do we care about a lower bound on val?)

$$
\textbf{1} \text{ation} \tag{1 + \frac{\text{fraction}}{2^{52}} \times 2^{\text{exponent} - (2^{10} - 1)}
$$

Relative Error (Calculation)

(fix an exponent *n*)

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$
Relative Error (Calculation)

the relative error is *small* (fix an exponent *n*)

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$

 $val \geq 1.0 \times 2^n$ $err \leq 2^{-52} \times 2^n$

Relative Error (Calculation)

2−⁵² × 2*ⁿ* 1.0 × 2*ⁿ* $= 2^{-52} \approx 10^{-16}$

the relative error is *small* (fix an exponent *n*)

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$

$val \geq 1.0 \times 2^n$ $err < 2^{-52} \times 2^n$

Relative Error (Calculation)

$val \geq 1.0 \times 2^n$ 𝖾𝗋𝗋 ≤ 2−⁵² × 2*ⁿ*

the relative error is *small* (fix an exponent *n*)

$$
\frac{2^{-52} \times 2^n}{1.0 \times 2^n} = 2^{-52} \approx 10^{-16}
$$

$$
(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent} - (2^{10} - 1)}
$$

16 digits of accuracy Not bad, but also not great

≈

demo (example from the notes)

operations on floating-point numbers are not exact

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not hold

$\mathsf{property}$ like $(ab)c = a(bc)$ (associativity) may

operations on floating-point numbers are not exact

 $\mathsf{property}$ like $(ab)c = a(bc)$ (associativity) may not hold

it's a trade-off for large range and low relative error

operations on floating-point numbers are not exact

not hold

it's a trade-off for large range and low relative error

$\mathsf{property}$ like $(ab)c = a(bc)$ (associativity) may

What do we do about it?

Best Practices

1. don't compare floating points for equality

-
- 2. be aware of ill-conditioned problems
- 3. be aware of small differences

Principle 1: Closeness

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When doing floating-point calculations in a program, define an error margin and use that for equality checking

Principle 1: Closeness

Replace x == y with numpy.isclose(x, y)

When doing floating-point calculations in a program, define an error margin and use that for equality checking

In Practice.

demo

Principle 2: Ill-Conditioned Problems

Principle 2: Ill-Conditioned Problems

Make sure your problem is not sensitive to small errors.

Principle 2: Ill-Conditioned Problems

Make sure your problem is not sensitive to small errors.

In Practice. for example, don't divide by

numbers much smaller than your error tolerance

demo

Principle 3: Small Differences

Principle 3: Small Differences

Make sure you understand your error tolerance

when looking that the small differences of large numbers.

Principle 3: Small Differences

Make sure you understand your error tolerance

when looking that the small differences of large numbers.

In Practice. Don't expect $a - b$ to be small when a and b are "close" but very large.

demo

One Last Note: Special Numbers

0 (we can't already represent 0?) nan stands for not a number, $.e.g.,$ sqrt (-2) inf symbolic infinity, behaves as expected

NumPy is a library for doing linear algebra in Python.

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Its **fast** and very widely used.

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Python.

Its **fast** and very widely used.

We will primarily be using numpy (and scipy) instead of sympy in this course.

NumPy vs. Sympy

NumPy is **fast** NumPy is **approximate** NumPy is **widely used in applications**

Sympy is **slow** Sympy is **exact** Sympy is a **teaching tool** (and useful in symbolic computation research)

NumPy vs. Sympy

- numpy.array(...)
- a[i] #row access
- a[:,j] #col access
- a.shape[0]
- a.shape[1]

Matrix(...) a[i,:] #row access a[:,j] #col access a.rows a.cols

demo

Extra Topic: Analyzing Gaussian Elimination

Analyzing the Algorithm

Analyzing the Algorithm

We will not use $O($ ·) notation!

Analyzing the Algorithm

- We will not use $O(\cdot)$ notation! For numerics, we care about number of **FL**oatingoint **OP**erations (FLOPs):
	- >> addition
	- >> subtraction
	- >> multiplication
	- >> division
	- >> square root
Analyzing the Algorithm

- We will not use $O(\cdot)$ notation! For numerics, we care about number of **FL**oatingoint **OP**erations (FLOPs):
	- >> addition
	- >> subtraction
	- >> multiplication
	- >> division
	- >> square root

2*n* vs. *n* is very different when *n* ∼ 1020

-
-
-
-
-
- -

that said, we don't care about *exact* bounds

that said, we don't care about *exact* bounds $g(n)$ if

> *f*(*i*) *g*(*i*) $= 1$

lim *i*→∞

A function $f(n)$ is asymptotically equivalent to

that said, we don't care about *exact* bounds $g(n)$ if

> *f*(*i*) *g*(*i*) $= 1$

lim *i*→∞

for polynomials, they are equivalent to their dominant term

A function $f(n)$ is asymptotically equivalent to

the dominant term of a polynomial is the monomial with the

$3x^3 + 100000x^2$ 3*x*³ $= 1$

 $3x^3$ dominates the function even though the coefficient for x^2 is so large

highest degree

lim *i*→∞

Parameters

- : number of variables *n*
-
-

m : number of equations (we will assume $m = n$) $n+1$: number of rows in the augmented matrix

The Cost of a Row Operation

$n+1$ multiplications for the scaling $n+1$ additions for the row additions

Tally: 2(*n* + 1) FLOPS

$R_i \leftarrow R_i + aR_j$

Cost of First Iteration of Elimination

- $R_n \leftarrow R_n + a_n R_1$
- repeated row operations for each row except the

first

Tally: $\approx 2n(n+1)$ FLOPS

 $R_2 \leftarrow R_2 + a_2 R_1$ $R_3 \leftarrow R_3 + a_3 R_1$

Rough Cost of Elimination

repeating this last process at most n times gives us a dominant term 2*n*³

we can give a better estimation...

 $Tally: \approx 2n^2(n+1)$ FLOPS

-
-

Cost of Elimination

At iteration *i*, we're only interested in rows after *i* And to the right of column *i*

Cost of Elimination

Iteration 1: $2n(n+1)$ Iteraiton Iteration

 $\ddot{\bullet}$

$$
\frac{2n(n+1)}{\tan 2: 2(n-1)n}
$$
\n
$$
\frac{2n(n+1)(2n-1)}{2n(n+1)(2n+1)} \leftarrow (2/3)n^3
$$

Tally: ~ (2/3)n³ FLOPS

Cost of Back Substitution

- (Let's assume no free variables)
- for each pivot, we only need to:
	- >> zero out a position in 1 row (0 FLOPS) >> add a value to the last row (1 FLOP)
		- **at most 1 FLOP per row per pivot** ∼ *n*²

Tally: ∼ (2/3)n³ FLOPS

Cost of Gaussian Elimination

Tally:∼(2/3)n³ FLOPS

(dominated by elimination)

Summary

floating point numbers are represented in your computer

floating point operations are <u>not</u> exact

this can have unintended consequences

we get 16 digits of accuracy