Gaussian Elimination + Numerics

Geometric Algorithms
Lecture 3

Outline

- >> Finish our discussion of Gaussian Elimination
- Think more carefully about number representations, and look at the consequences of floating point representations
- » If there's time: Analyze the running time of
 Gaussian Elimination

Keywords

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

Practice Problem

lem
$$k = \frac{1}{2}$$

$$k = 4$$

$$x + hy = 3$$

$$2x - 5y = k$$

For what values of h and k is the above system inconsistent?

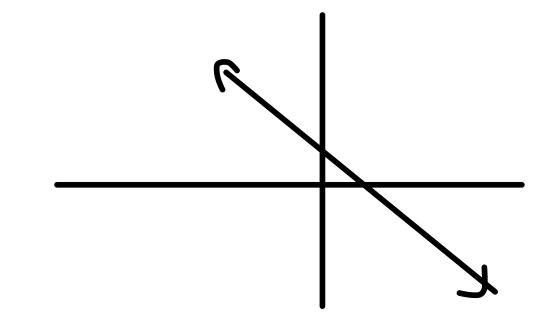
$$\begin{array}{ccc}
R_{1} & -2R_{1} & h = -2.5 \\
X & + hy & = 3
\end{array}$$

$$2x - 5y = k$$

$$2x+2y=2$$

 $x+4y=1$

$$h = -2.5$$
 $K \neq 6$



Recap

Recap: Echelon Form

```
\begin{bmatrix} 0 & \blacksquare & * & * & * & * & * & * & * & * \\ 0 & 0 & 0 & \blacksquare & * & * & * & * & * \\ 0 & 0 & 0 & 0 & \blacksquare & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & \blacksquare & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \blacksquare & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}
```

= nonzero, * = anything

Recap: Echelon Form

```
next leading entry
   to the right
                        all-zero rows at
                           the bottom
```

= nonzero, * = anything

Recap: Reduced Echelon Form

```
\begin{bmatrix} 0 & 1 & * & 0 & 0 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 1 & 0 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 0 & 1 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 0 & 1 & * & * & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}
```

Recap: Reduced Echelon Form

other column entries are 0

leading entries are 1

Recap: The Fundamental Points

Recap: The Fundamental Points

Point 1. we can "read off" the solutions of a system of linear equations from its RREF

Recap: The Fundamental Points

Point 1. we can "read off" the solutions of a system of linear equations from its RREF

Point 2. every matrix is row equivalent to a unique matrix in reduced echelon form

Recap: General Form Solution

$$\begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 0 & 1 \end{bmatrix}$$

$$x_1 = 2 - x_3$$

$$x_2 = 1$$

$$x_3 \text{ is free}$$

$$x_4 = 2 - 2 = 0$$

$$x_3 = 7 - 7 = -5$$

$$x_4 = 1$$

$$x_3 = 2$$

$$x_4 = 1$$

$$x_4 = 1$$

$$x_3 = 2$$

Recap: General Form Solution

$$\begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 0 & 1 \end{bmatrix}$$

$$x_1 = 2 - x_3$$

$$x_2 = 1$$

$$x_3 \text{ is free}$$

1. For each pivot position (i,j), isolate x_i in the equation in row i

Recap: General Form Solution

$$\begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 0 & 1 \end{bmatrix}$$

$$x_1 = 2 - x_3$$

$$x_2 = 1$$

$$x_3 \text{ is free}$$

- 1. For each pivot position (i,j), isolate x_i in the equation in row i
- 2. If x_i is not in a pivot column then write

 x_i is free

1. Write your system as an augmented matrix

1. Write your system as an augmented matrix

2. Find the RREF of that matrix

1. Write your system as an augmented matrix

2. Find the RREF of that matrix

3. Read off the solution from the RREF

the goal of <u>back-substitution</u> is to reduce an echelon form matrix to a **reduced** echelon form

the goal of <u>back-substitution</u> is to reduce an echelon form matrix to a **reduced** echelon form

the goal of <u>Gaussian elimination</u> is to reduce an **augmented** matrix to a **reduced** echelon form

the goal of <u>back-substitution</u> is to reduce an echelon form matrix to a **reduced** echelon form

the goal of <u>Gaussian elimination</u> is to reduce an **augmented** matrix to a **reduced** echelon form

reduced echelon forms describe solutions to linear equations

Gaussian Elimination

eliminations + back-substitution

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

eliminations + back-substitution

we've already done this

but we'll take one step further and write down the algorithm as <u>pseudocode</u>

eliminations + back-substitution

we've already done this

but we'll take one step further and write down the algorithm as <u>pseudocode</u>

Keep in mind. How do we turn our intuitions into a formal procedure?

The details of Gaussian elimination are tricky

The details of Gaussian elimination are tricky

The goal is not to understand it entirely, but to get enough intuition to emulate it

The details of Gaussian elimination are tricky

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

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)

Elimination Stage (High Level)

Input: matrix A of size $m \times n$

Output: echelon form of A

Elimination Stage (High Level)

Input: matrix A of size $m \times n$

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 ?

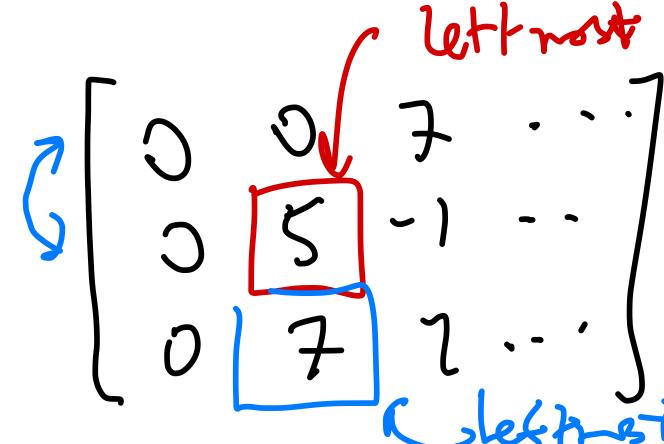
What if the first equation doesn't have the variable x_1 ?

Swap rows with an equation that does.

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

Swap rows with an equation that does.

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

Find the *leftmost* variable which appears in *any* of the remaining equations.

FUNCTION fwd_elim(A):

```
FUNCTION fwd_elim(A):
   FOR [i from 1 to m]: # for each row from top to bottom
```

```
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
```

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

```
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) \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
        [zero out A[l, k] using a replacement operation]
```

```
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) \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
        [zero out A[l, k] using a replacement operation]
 RETURN A
```

```
\begin{bmatrix} 0 & 3 & -6 & 6 & 4 & -5 \\ 3 & -7 & 8 & -5 & 8 & 9 \\ \hline 3 & -9 & 12 & -9 & 6 & 15 \end{bmatrix}
entry
```

```
\begin{bmatrix} 0 & 3 & -6 & 6 & 4 & -5 \\ 3 & -7 & 8 & -5 & 8 & 9 \\ \hline 3 & -9 & 12 & -9 & 6 & 15 \end{bmatrix}
entry
```

Swap R_1 and R_3

$$\begin{bmatrix}
3 & -9 & 12 & -9 & 6 & 15 \\
3 & -7 & 8 & -5 & 8 & 9 \\
0 & 3 & -6 & 6 & 4 & -5
\end{bmatrix}$$

 $R_{1} \leftarrow R_{1} - R_{1}$

leftmost nonzero entry $\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$

swap R_2 with R_2

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

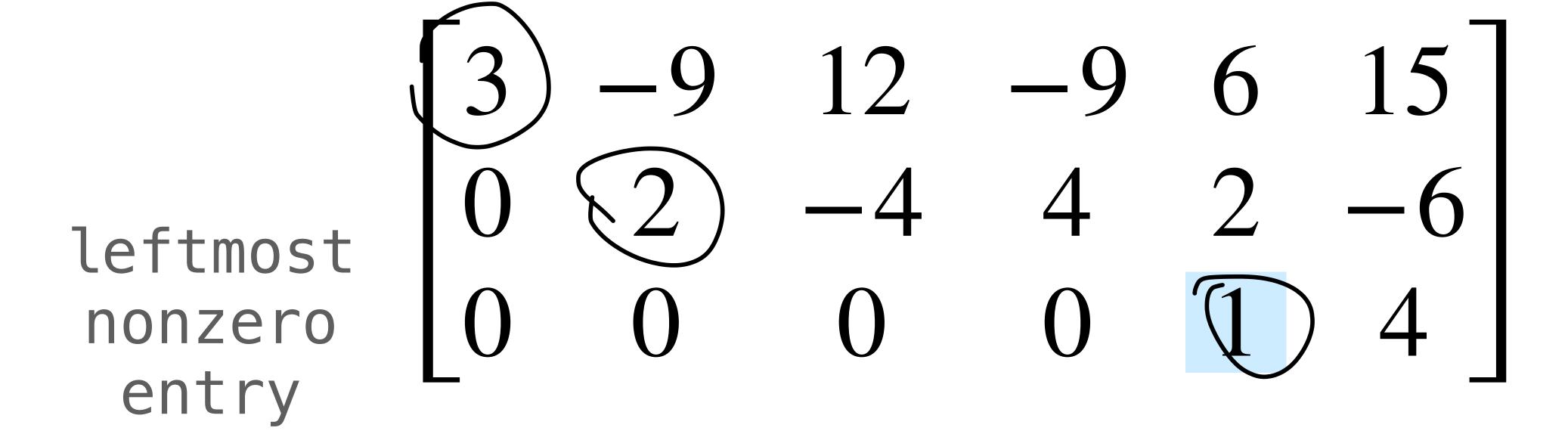
$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

$$R_{3} \leftarrow R_{3} - R_{2}$$

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

$$R_{3} \leftarrow R_{3} - \frac{3R_{2}}{2}$$

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$



$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$
leftmost nonzero entry

swap R_3 with R_3

Elimination Stage (Example)

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

Elimination Stage (Example)

done with elimination stage going to back substitution stage

Back Substitution Stage

Back Substitution Stage (High Level)

Back Substitution Stage (High Level)

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

Output: reduced echelon form of A

Back Substitution Stage (High Level)

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

Output: reduced echelon form of A

scale pivot positions and eliminate the variables for that column from the other equations

FUNCTION back_sub(A):

```
FUNCTION back_sub(A):
   FOR [i from 1 to m]: # for each row from top to bottom
```

```
FUNCTION back_sub(A):
    FOR [i from 1 to m]: # for each row from top to bottom
        IF [row i has a leading entry]:
```

```
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
```

```
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 \leftarrow index \ of \ leading \ entry \ of \ row \ i

R_i(A) \leftarrow R_i(A) \ / \ A[i, j] \ # \ divide \ by \ leading \ entry
```

```
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<sub>i</sub>(A) ← R<sub>i</sub>(A) / A[i, j] # divide by leading entry

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(A) \leftarrow R_i(A) / A[i, j] \# divide by leading entry
      FOR [k from 1 to i - 1]: # for the rows above the current one
        R_k(A) \leftarrow R_k(A) - R[k, j] * R_i(A)
        # zero out R[k, j] above the leading entry
```

```
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(A) \leftarrow R_i(A) / A[i, j] \# divide by leading entry
      FOR [k from 1 to i - 1]: # for the rows above the current one
        R_k(A) \leftarrow R_k(A) - R[k, j] * R_i(A)
        # zero out R[k, j] above the leading entry
  RETURN A
```

```
\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}
```

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

 $R_1 \leftarrow R_1 / 3$

$$\Rightarrow \begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 2 & -4 & 4 & 2 & -6 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

 $R_2 \leftarrow R_2 / 2$

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

```
\begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}
```

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

 $R_1 \leftarrow R_1 + 3R_2$

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

```
\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}
```

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

 $R_3 \leftarrow R_3 / 1$

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

$$R_2 \leftarrow R_2 - R_1$$

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

```
\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}
```

```
\begin{bmatrix} 1 & 0 & -2 & 3 & 5 & -4 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}
```

$$R_1 \leftarrow R_1 - 5R_3$$

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

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

done with back substitution phase

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

 $x_2 = (-7) + 2x_3 - 2x_4$
 x_3 is free
 x_4 is free
 $x_5 = 4$

1. Write your system as an augmented matrix

1. Write your system as an augmented matrix

2. Find the RREF of that matrix

1. Write your system as an augmented matrix

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

Do you remember sig figs from science class?

Do you remember sig figs from science class?

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

Do you remember sig figs from science class?

When you use a ruler, you can't do better than ±1mm, 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

Your computer is a collection of fixed size registers

Each register holds a sequence of bits

Your computer is a collection of fixed size registers

Each register holds a sequence of bits

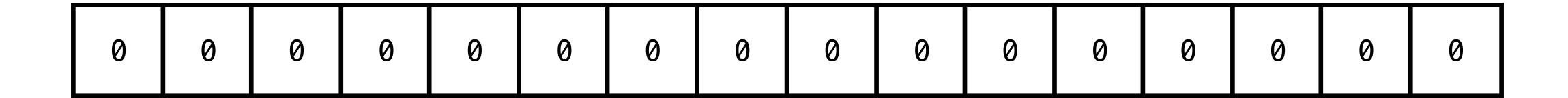
<u>The Goal.</u> represent numbers so they fit in those registers

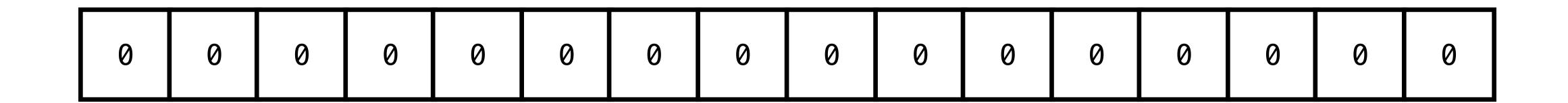
Your computer is a collection of fixed size registers

Each register holds a sequence of bits

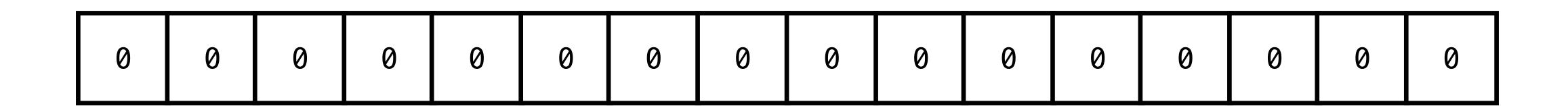
<u>The Goal.</u> represent numbers so they fit in those registers

this is, of course, a lie an abstraction





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

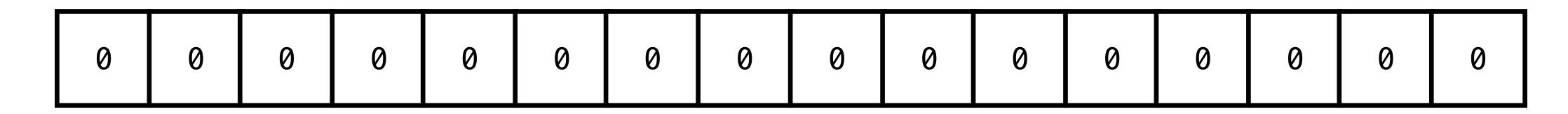


Question. How do we slice up our fixed sequence to represent numbers?

things to consider:

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

Unsigned Integers



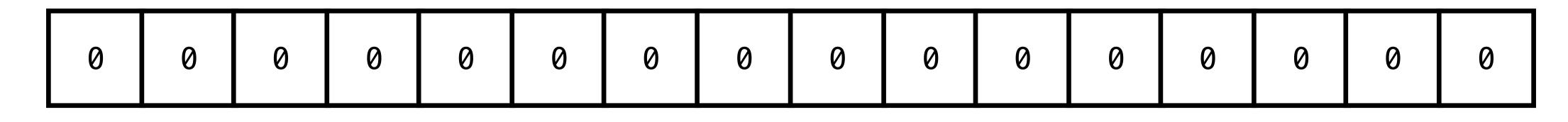
value

binary value (we should know this by now)

e.g. 10001010 represents

$$1(2^7) + 0(2^6) + 0(2^5) + 0(2^4) + 0(2^3) + 1(2^2) + 0(2^1) + 1(2^0)$$

Signed Integers



sign value

sign bit + binary value

e.g. 10001010 represents

$$-1 \times (0(2^6) + 0(2^5) + 0(2^4) + 0(2^3) + 1(2^2) + 0(2^1) + 1(2^0)$$

floats in python use 64 bits

floats in python use 64 bits

That's 1.8×10^{19} possible values

floats in python use 64 bits

That's 1.8×10^{19} possible values

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

floats in python use 64 bits

That's 1.8×10^{19} possible values

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

Integers work because they are discrete and evenly spaced

What if we evenly discretize a range of values?

Integers work because they are discrete and evenly spaced

What if we evenly discretize a range of values?

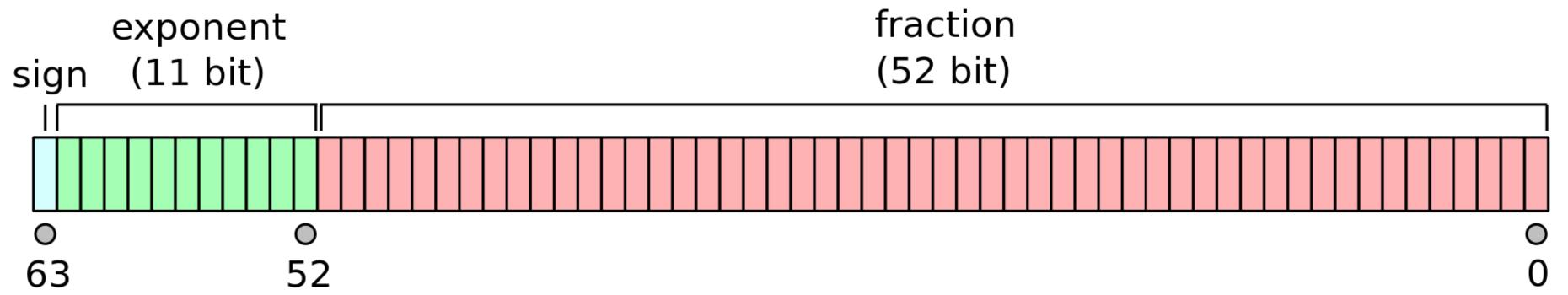
i.e., represent

-0.001, 0, 0.00, 0.002, 0.003, 0.004, ...

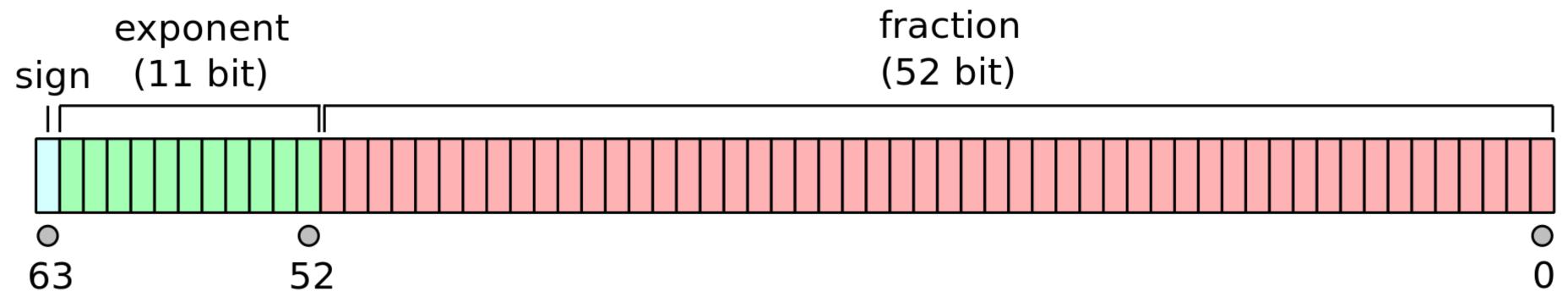
Question

Discuss the advantages and disadvantages of this approach

Floating-Point Numbers (IEEE-754)

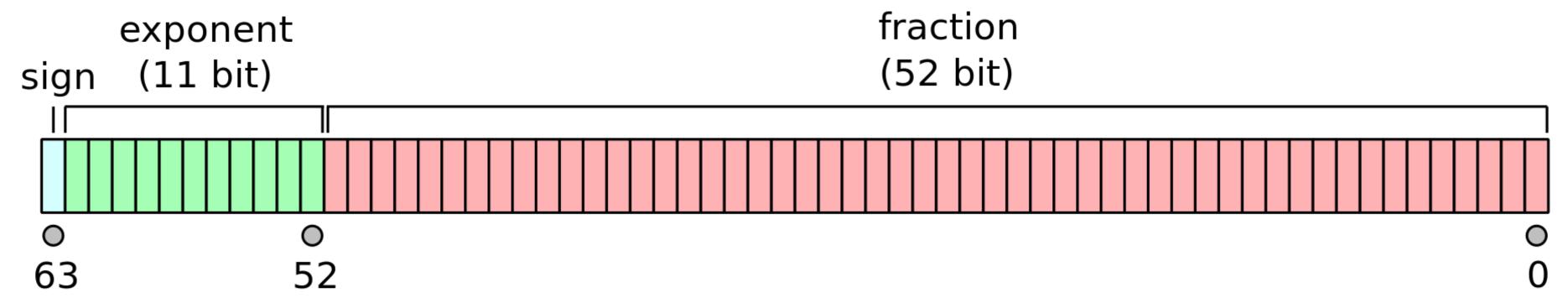


Floating-Point Numbers (IEEE-754)



This is like scientific notation, but binary:

Floating-Point Numbers (IEEE-754)



This is like scientific notation, but binary:

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

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

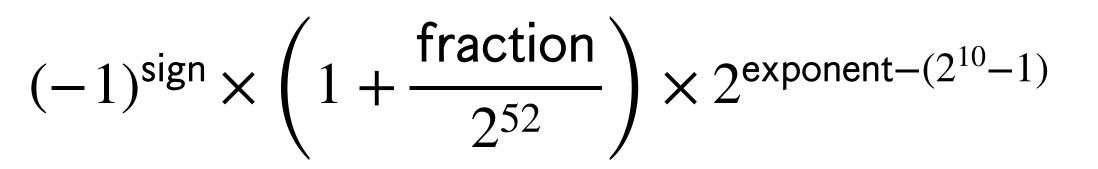
Question

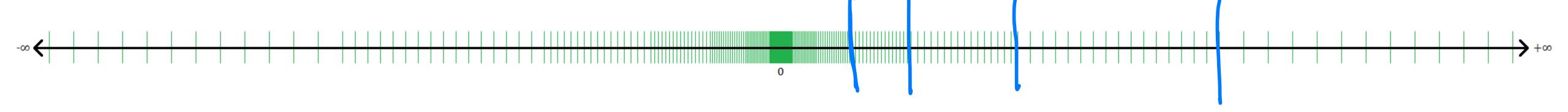
$$(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2^{10}-1)}$$

Any ideas why this is better/worse?

And why not have a sign bit for the exponent?

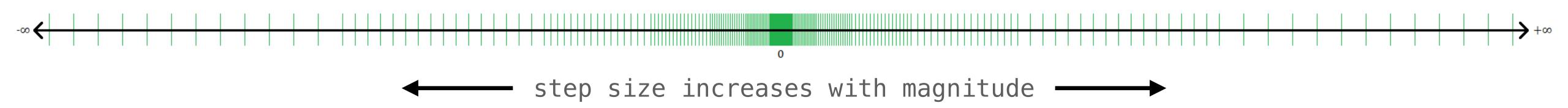
Step Size





Step Size

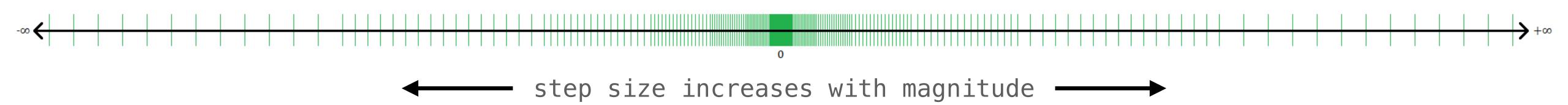
$$(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2^{10}-1)}$$



Definition. <u>step size</u> is the space between two floating-point representations

Step Size

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



Definition. <u>step size</u> is the space between two floating-point representations

for fixed exponent n two numbers are at least

$$0.00...001 \times 2^n = 2^{-52} \times 2^n$$

away (why?)

Step Size

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



Definition. <u>step size</u> is the space between two floating-point representations

for fixed exponent n two numbers are at least

$$0.00...001 \times 2^n = 2^{-52} \times 2^n$$

away (why?)

Step size <u>doubles</u> for each exponent

IEEE-754 defines a <u>subset</u> of decimal numbers

IEEE-754 defines a <u>subset</u> of decimal numbers

operations on floating point numbers attempt to give you the <u>closest</u> to the actual value, though there will be errors

IEEE-754 defines a <u>subset</u> of decimal numbers

operations on floating point numbers attempt to give you the <u>closest</u> 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

Observation. ± 0.001 is *tiny* error for 10^{20} but *massive* for 10^{-20}

Relative Error

Observation. ± 0.001 is *tiny* error for 10^{20} but *massive* for 10^{-20}

Relative Error.

$$err_{rel} = \frac{err}{val}$$

Relative Error

Observation. ± 0.001 is *tiny* error for 10^{20} but *massive* for 10^{-20}

Relative Error.

$$err_{rel} = \frac{err}{val}$$

IEEE-754 keeps relative error <u>small</u>

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

(fix an exponent n)

Relative Error (Calculation) (1 + fraction 252) × 2 exponent-(210-1)

(fix an exponent n)

error is determined by step-size

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

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

(fix an exponent n)

the smallest number we can represent at least 1.0×2^n

$$val \geq 1.0 \times 2^n$$

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

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

(fix an exponent n)

Relative Error (Calculation) (1+ fraction 252) × 2 exponent-(210-1)

(fix an exponent n)

the relative error is small

$$val \ge 1.0 \times 2^n$$

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

Relative Error (Calculation) (1 + fraction) × 2 exponent-(2 10-1)

(fix an exponent n)

the relative error is *small*

$$val \ge 1.0 \times 2^n$$

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

$$err_{rel} = \frac{err}{val} \le \frac{2^{-52} \times 2^n}{1.0 \times 2^n} = 2^{-52} \approx 10^{-16}$$

Relative Error (Calculation) (1+ fraction) × 2^{exponent-(2¹⁰-1)}

(fix an exponent n)

the relative error is small

$$val \ge 1.0 \times 2^n$$

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

$$\operatorname{err}_{\text{rel}} = \frac{\operatorname{err}}{\operatorname{val}} \le \frac{2^{-52} \times 2^n}{1.0 \times 2^n} = 2^{-52} \approx 10^{-16}$$

≈16 digits of accuracy

Not bad, but also not great

demo

(example from the notes)

operations on floating-point numbers are not exact

operations on floating-point numbers are not exact

properties like (ab)c = a(bc) (associativity) may not hold

operations on floating-point numbers are not exact

properties 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

properties like (ab)c = a(bc) (associativity) may not hold

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

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

Principle 1: Closeness

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

Principle 1: Closeness

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

In Practice.

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

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

```
(we can't already represent 0?)
```

nan stands for not a number, .e.g, sqrt(-2)

inf symbolic infinity, behaves as expected

Extra Topic: Analyzing the Algorithm

We will not use $O(\cdot)$ notation!

```
We will not use O(\cdot) notation!
```

For numerics, we care about number of **FL**oating-oint **OP**erations (FLOPs):

- >> addition
- >> subtraction
- >> multiplication
- >> division
- >> square root

```
We will not use O(\cdot) notation!
```

For numerics, we care about number of **FL**oating-oint **OP**erations (FLOPs):

- >> addition
- >> subtraction
- >> multiplication
- >> division
- >> square root

```
2n vs. n is very different when n \sim 10^{20}
```

that said, we don't care about exact bounds

that said, we don't care about exact bounds

A function f(n) is asymptotically equivalent to g(n) if

$$\lim_{i \to \infty} \frac{f(i)}{g(i)} = 1$$

that said, we don't care about exact bounds

A function f(n) is asymptotically equivalent to g(n) if

$$\lim_{i \to \infty} \frac{f(i)}{g(i)} = 1$$

for polynomials, they are equivalent to their dominant term

The **dominant term** of a polynomial is the monomial with the highest degree

$$\lim_{i \to \infty} \frac{3x^3 + 100000x^2}{3x^3} = 1$$

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

Parameters

n: number of variables

m : number of equations (we will assume m=n)

n+1: number of rows in the augmented matrix

The Cost of a Row Operation

$$R_i \leftarrow R_i + aR_j$$

n+1 multiplications for the scaling

n+1 additions for the row additions

Tally: 2(n+1) FLOPS

Cost of First Iteration of Elimination

$$R_2 \leftarrow R_2 + a_2 R_1$$

$$R_3 \leftarrow R_3 + a_3 R_1$$

$$\vdots$$

$$R_n \leftarrow R_n + a_n R_1$$

Repeated row operations for each row except the first

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

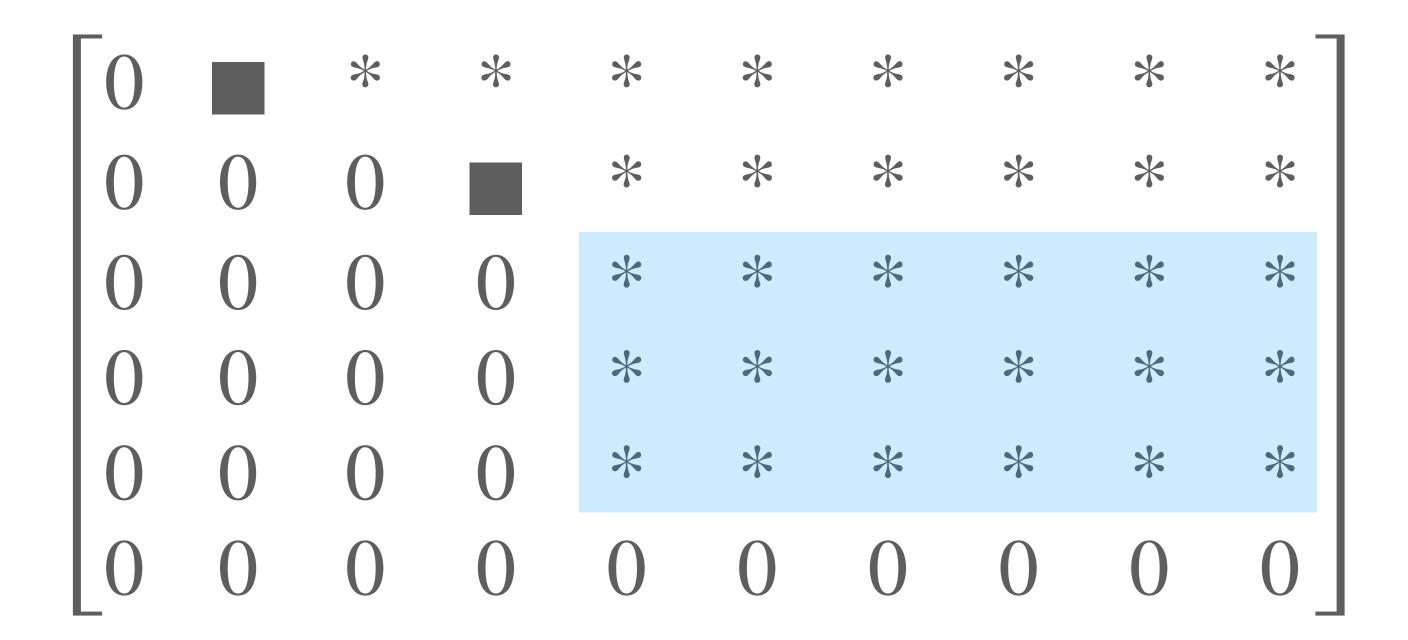
Rough Cost of Elimination

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

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)
Iteration 2: 2(n-1)n
Iteration 3: 2(n-2)(n-1)
\vdots
```

$$\sum_{k=1}^{n} 2k(k+1) \approx \frac{2n(n+1)(2n+1)}{6} \sim (2/3)n^3$$

Tally: $\sim (2/3)n^3$ 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 ~ n²
```

Tally: $\sim (2/3)n^3$ FLOPS

Cost of Gaussian Elimination

Tally:
$$\sim (2/3)n^3$$
 FLOPS

(dominated by elimination)

Summary

floating point numbers are **represented** in your computer

Floating point operations are *not* exact, and this can have unintended consequences

we get 16 digits of accuracy