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

$$x + hy = 3$$
$$2x - 5y = k$$

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

Solution

$$x + hy = 3$$
$$2x - 5y = k$$

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

Recap: Reduced Echelon Form

leading entries are 1

Recap: The Fundamental Points

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

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

Recap: General Form Solution

$$x_1 = 2 - x_3$$

$$x_2 = 1$$

$$x_3 ext{ is free}$$

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

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

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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

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we've already done this

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

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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

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

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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)

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

Output: echelon form of A

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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 ?

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Swap rows with an equation that does.

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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]
```

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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
```

```
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]
```

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

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

```
\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

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

 $R_3 \leftarrow R_3 - R_1$

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

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}
```

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

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

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

```
\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
```

```
\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)

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

Elimination Stage (Example)

done with elimination stage going to back substitution stage

Back Substitution Stage

Back Substitution Stage (High Level)

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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

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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}
```

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

 $R_1 \leftarrow R_1 / 3$

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

$$R_2 \leftarrow R_2 / 2$$

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

```
\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

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3. Read off the solution from the RREF

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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?

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

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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

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Each register holds a sequence of bits

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Each register holds a sequence of bits

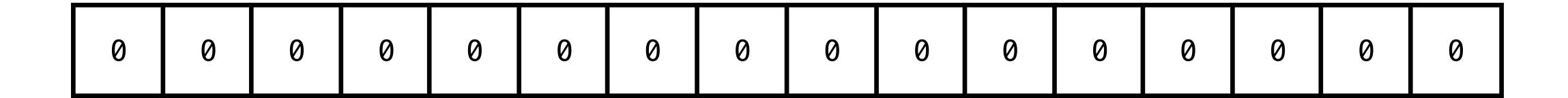
The Goal. represent numbers so they fit in those registers

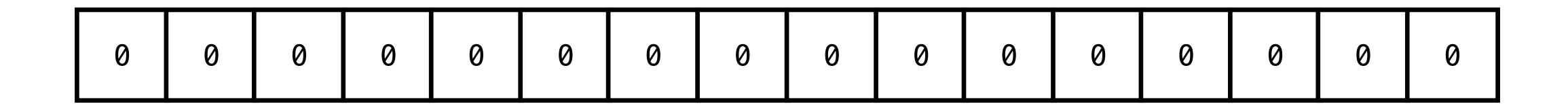
Your computer is a collection of fixed size registers

Each register holds a sequence of bits

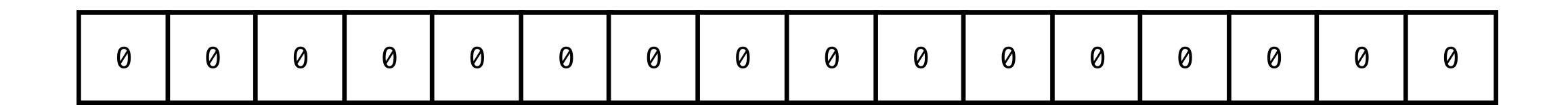
<u>The Goal.</u> represent numbers so they fit in those registers

this is, of course, a lie an abstraction





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

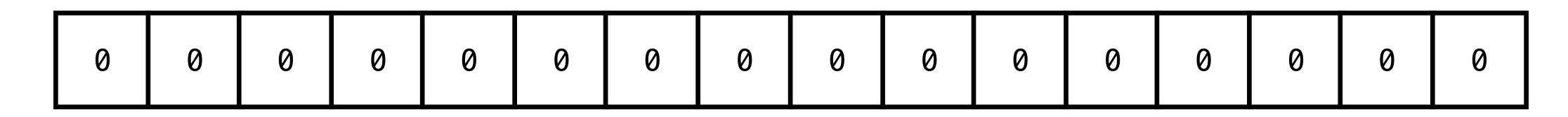


Question. How do we slice up our fixed sequence to represent numbers?

things to consider:

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

Unsigned Integers



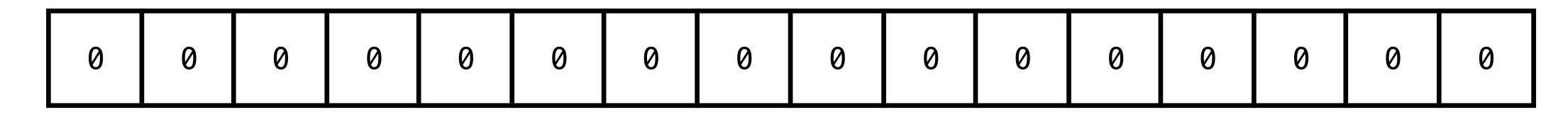
value

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

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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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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?

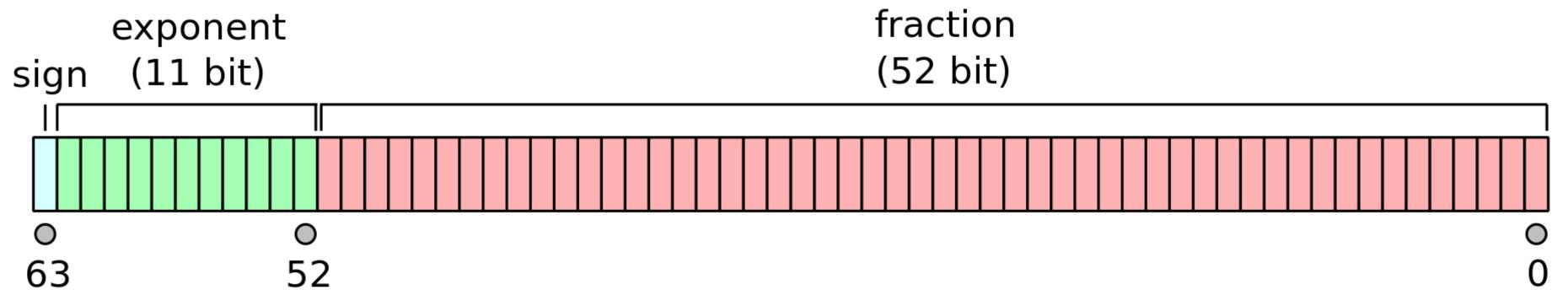
i.e., represent

 $-0.001, 0, 0.0001, 0.002, 0.003, 0.004, \dots$

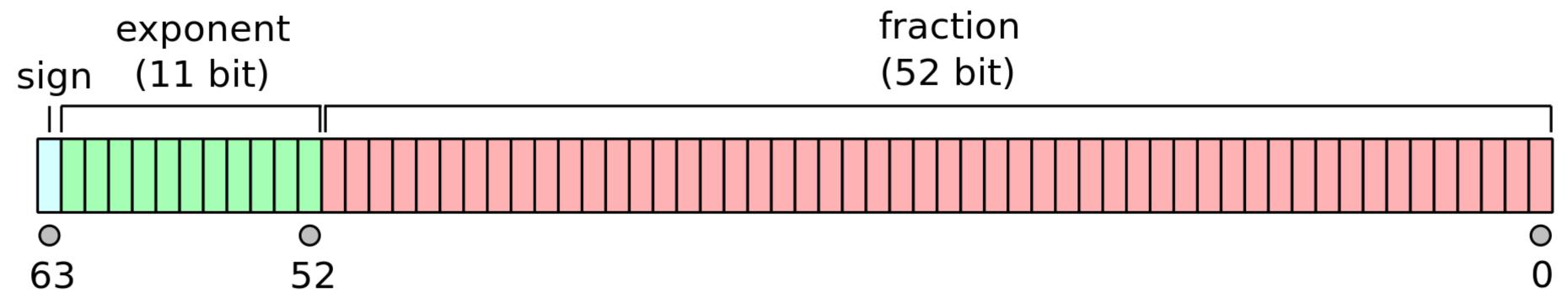
Question

Discuss the advantages and disadvantages of this approach

Floating-Point Numbers (IEEE-754)

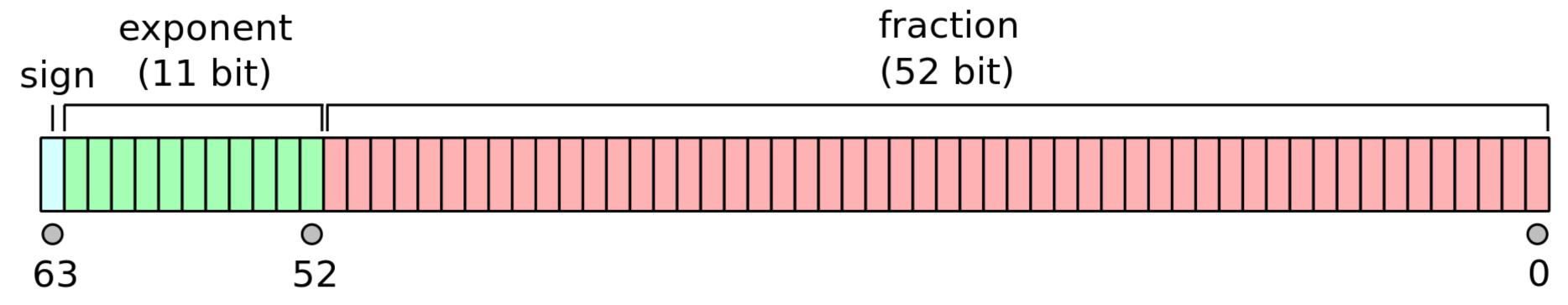


Floating-Point Numbers (IEEE-754)



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

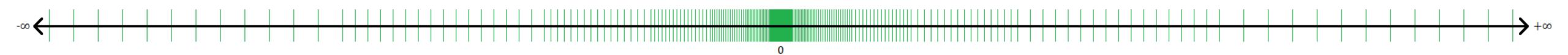
$$(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2^{10}-1)}$$

Any ideas why this is better/worse?

And why not have a sign bit for the exponent?

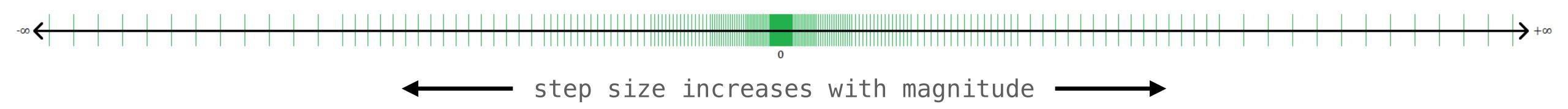
Step Size

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



Step Size

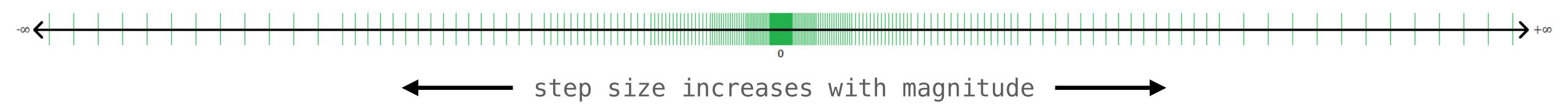
$$(-1)^{\text{sign}} \times \left(1 + \frac{\text{fraction}}{2^{52}}\right) \times 2^{\text{exponent}-(2^{10}-1)}$$



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

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

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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}

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Relative Error.

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

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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) (1+ fraction 252) × 2 exponent-(210-1)

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

(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) (1 + fraction 252) × 2 exponent-(210-1)

(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

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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

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

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In Practice.

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

demo

Principle 2: Ill-Conditioned Problems

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Make sure your problem is not sensitive to small errors.

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In Practice. for example, don't divide by numbers much smaller than your error tolerance

demo

Principle 3: Small Differences

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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!

```
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```

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

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

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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

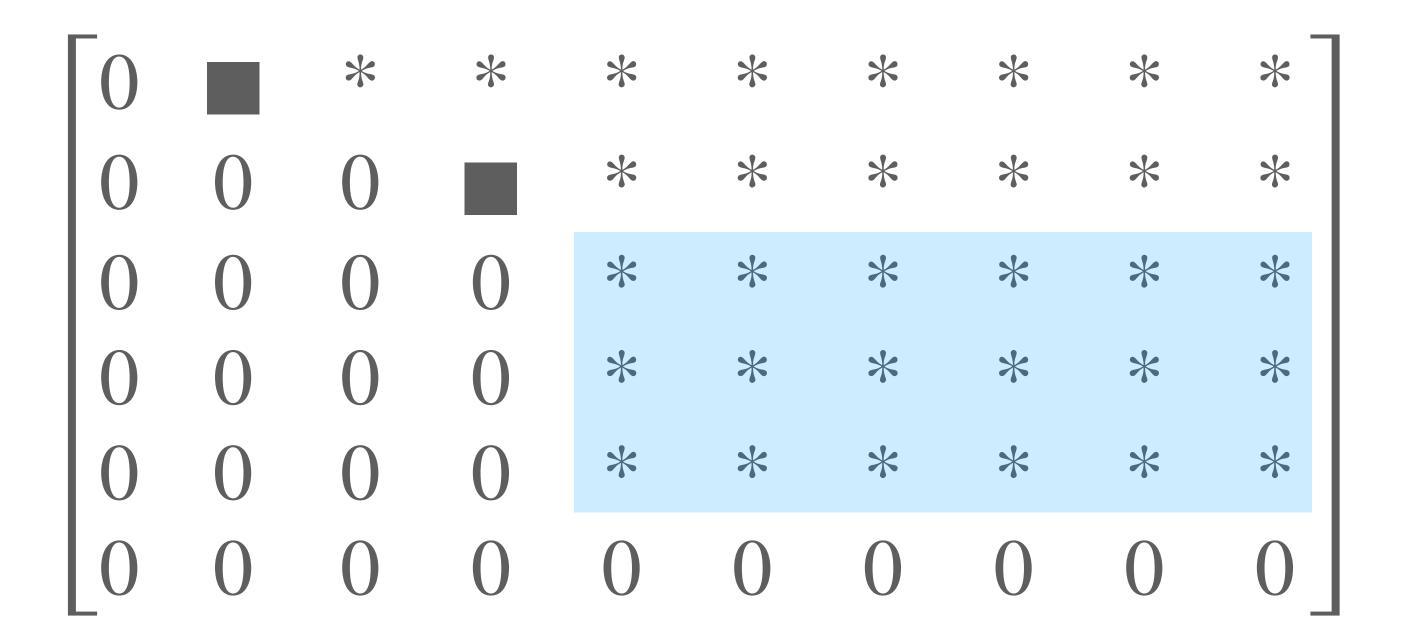
Rough Cost of Elimination

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