#### CS 357: Numerical Methods

#### Lecture 4: Matrix Norms

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## Vector Norms...

Vector norms are functions that map a vector to a real number

You can think of it as measuring the magnitude of the vector The norm you know is the 2-norm:  $\|v\|_2 = \sqrt{\mathring{a}v_i^2}$ 

$$\boldsymbol{v} = \left\langle v_0, v_1, \dots, v_{d-1} \right\rangle$$

You can use it to measure the distance between two points

Compute a vector v = p2-p1 and take the norm of v

## Requirements for a Norms...

All of the following must be true for some function to be a norm

$$||x|| \ge 0$$
  
$$||x + y|| \le ||x|| + ||y||$$
  
$$||\partial x|| = |\partial|||x||$$
  
$$||x|| = 0 \Leftrightarrow x = 0$$

#### Vector Norms...

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Vectors:

$$\|x\|_{p} = (|x_{1}|^{p} + |x_{2}|^{p} + \dots + |x_{n}|^{p})^{1/p}$$
$$\|x\|_{1} = |x_{1}| + |x_{2}| + \dots + |x_{n}| = \sum_{i=1}^{n} |x_{i}|$$
$$\|x\|_{\infty} = \max(|x_{1}|, |x_{2}|, \dots, |x_{n}|) = \max_{i}(|x_{i}|)$$

## Visualizing 1-ball of Norms

Where does the norm equal 1 in the 2D Euclidean plane



# Measuring Error

- Norms measure distance between vectors
  - $\square$  The distance between v1 and v2 is  $\|v1 v2\|$
- □ Imagine we have an approximation  $x_0$  to some true value x So...x=  $x_0 + \Delta x$
- $\square$  The absolute error is  $\Delta x$
- □ ...for vectors we could use  $||\Delta x||$
- Can this tell us if an answer is good?

# Absolute Error - Example

$$|X - X_{0}| = 0.1$$
  
if  $x = 10000000$   
if  $x = 0.1$  Abad

## Relative Error

Relative error is defined as the absolute error divide by the true value

$$error_{rel} = \frac{\|\Delta x\|}{\|x\|}$$

This allows us some way to assess the magnitude or error

Any obvious issues with computing error?

## Relative Error

- We usually don't know the true value x
- □ So....
  - If we know an error range, compute an upper bound
  - Take the largest possible error value use that

### Precision and Accuracy

- **Precision:** number of digits with which a number is expressed
- Accuracy:

number of correct significant digits in approximating a value

- □ What about saying  $\pi = 3.1425289898?$  4 = 6, 601 ish
- Rule of thumb, if a number has a relative error of about 10<sup>-p</sup> then it has about p correct significant digits in its decimal representation

## Relative Error - Example

$$1/6g = 0.125 £ .12$$
  
 $A \times = 0.005$   
rel =  $\frac{0.005}{0.125} = 0.04$ 

y's error

# **Condition Number**

- We'll be interested in numerical methods (algorithms)
- How much does a method amplify the error in an input
  - Every input has some error
  - Every output has some error...

$$C = Ma \gamma$$
  
 $\chi_{in}$ 

ILAXIN 1

## Matrices Transformations (3D Graphics version)

#### General

$$\begin{bmatrix} d & e & f & a \\ g & h & i & b \\ j & k & l & c \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} dx + ey + fz + a \\ gx + hy + iz + b \\ jx + ky + lz + c \\ 1 \end{bmatrix}$$

Translation

$$\begin{array}{cccc} 1 & & & & a \\ & 1 & & & b \\ & & 1 & c \\ & & & & 1 \end{array} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x + a \\ y + b \\ z + c \\ 1 \end{bmatrix}$$

# 3-D Coordinates

- Points represented by 4-vectors
- Need to decide orientation of coordinate axes





## Scale



# 3-D Rotations



Rotations do not commute!

## Norms of Linear Functions

What is a linear function?

 $f(x_1,\ldots,x_k)=b+a_1x_1+\ldots+a_kx_k$ 

...a hyperplane in dimension k

■ We can define the norm of a linear function as  $\|f\| = \max \frac{\|f(x)\|}{\|x\|}$ 

With x ≠0 (Why?)

This is like a maximum amplification factor

## Matrix Norm

We'll define a matrix norm to be the max possible stretching a matrix can perform on a vector....

 $||A|| = \max ||Ax||$  for all x with ||x|| = 1

## Matrix Norms

$$\|A\|_{p} = \max_{x \neq 0} \frac{\|Ax\|_{p}}{\|x\|_{p}}$$
$$\|A\|_{1} = \max_{1 \leq j \leq n} \sum_{i=1}^{m} |a_{ij}|$$
$$\|A\|_{\infty} = \max_{1 \leq i \leq m} \sum_{j=1}^{n} |a_{ij}|$$

Matrix norms are defined in terms of an underlying vector norm

In general matrix p-norms are complicated to compute but a couple of them are easy...

What is the 1-norm computing?

What is the inf-norm computing?

### Matrix Norms - Example



## Solving Linear Systems and Error



