

14. Conditioning of a problem and stability of an algorithm

Summary of the previous class

- The Singular-Value Decomposition (SVD)
 - $A = U\Sigma V^T$
- Calculation via eigen-decomposition of symmetric matrix
 - AA^T
- Application to understanding and compressing data
- PCA and best-fitting k -dimensional subspace

Goals for today's class

- **Conditioning**: How sensitive is a **problem**
- Catastrophic cancellation
- **Stability** of an **algorithm**

Conditioning of a problem

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- Example: Intersection of two thick lines at an angle

Collaboration I

Perturbing input and output

- 1 What is the relationship between Δx and Δy ?
- 2 What does the form of the resulting equation suggest?
- 3 What might be a better way of measuring the perturbation in x ?

Absolute errors

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- Define the error $\Delta x := \hat{x} - x$

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- Δx and Δy are **absolute errors**

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- So $\hat{x} = x + \Delta x = x(1 + \delta x)$
- And $\hat{y} = y + \Delta y = y(1 + \delta y)$

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- How **accurate** is it?
- The relative error is $|\delta x| = \left| \frac{\hat{x} - x}{x} \right|$
- We can express this in terms of the number of **significant** or **accurate digits**:

$$d = -\log_{10} |\delta x|$$

- This gives the number of digits after which \hat{x} and x differ
- Independently of the absolute size of x

Collaboration II

- 1 Calculate the number of accurate digits for the approximation $22/7$ of π
- 2 Does that make sense?
- 3 How could you get better approximations?

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- For others, perturbing the input can drastically change the output: they are **ill-conditioned**
- How can we **measure** the conditioning of a *problem*?

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- In face we need $\hat{\kappa}_{\phi}(x, \epsilon)$, taking the maximum over inputs with $\|\Delta x\| < \epsilon$

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- Provided ϕ is differentiable

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Collaboration III

Condition number of addition

Addition of two numbers is a very basic operation in numerical computing. We hope that it is well-conditioned. Is it?

- 1 Calculate the absolute condition number for the problem $\phi(x) = x + a$ with a fixed a .
- 2 Calculate the relative condition number for addition.
- 3 Can the relative condition number be large?

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- But we really care about the *relative* condition number

Condition number of addition II

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- The relative condition number is

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- Note that $\phi(x)$ appears in the *denominator*

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- Note that $\phi(x)$ appears in the *denominator*
- So κ_{ϕ} is *large* when $x \simeq -a$
- **Catastrophic cancellation:** Loss of accuracy when *subtracting* two numbers that are close
- Common task: **identify + eliminate** catastrophic cancellation

Collaboration IV

Quadratic equations

- 1 Formulate “solving a quadratic equation” as a problem. What are the inputs and outputs?
- 2 What can go wrong?

Quadratic equations

- Let's look at solving a quadratic equation
$$ax^2 + bx + c = 0$$
- What is the problem ϕ ?
- The input \mathcal{X} is (a, b, c)
- The output \mathcal{Y} is the set of two roots x_{\pm} given by

$$x_{\pm} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

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- What can go wrong?
- If $b > 0$ and ac is small, then x_{+} has catastrophic cancellation

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- Take the input $\mathcal{X} = a$ and output $\mathcal{Y} = x_+$
- Problem: $\Phi(a) = x_+(a)$ (with b and c fixed)
- Suppose we move a to $a + \Delta a$
- Then the root x_+ moves to $x_+ + \Delta x_+$

Condition number of quadratic equations II

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- So the condition number is *large* when the roots are *close*
- We can visualize this by perturbing the quadratic function

Conditioning for linear systems

Norms

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Norms

- How sensitive is the solution of $\mathbf{Ax} = \mathbf{b}$ to perturbations of the problem?
- We need to be able to talk about the **norm** of vectors and of matrices
- For vectors $\mathbf{v} \in \mathbb{R}^m$ we will use the 2-norm:

$$\|\mathbf{v}\|_2 := \sqrt{\sum_{i=1}^m v_i^2}$$

Induced matrix norm

- There are various ways of measuring the size (norm) of a matrix that are useful for different purposes
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- We will use the matrix norm **induced** from a given vector norm
- Suppose A is a square $(m \times m)$ matrix
- For each \mathbf{x} we can measure $\|\mathbf{x}\|$ and $\|A\mathbf{x}\|$. . .
- We define the induced matrix norm as the smallest C such that

$$\|A\| \leq C \|\mathbf{x}\| \quad \forall \mathbf{x} \in \mathbb{R}^m$$

Induced matrix norm II

■ Hence

$$\|A\| = \sup_{\|x\|=1} \|Ax\|$$

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- Note that $\|A\mathbf{x}\| \leq \|A\|\|\mathbf{x}\|$

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- I.e. perturbations in A or \mathbf{b} ?
- Simplest case: Perturb the input \mathbf{b} to $\mathbf{b} + \Delta\mathbf{b}$
- How much is the output \mathbf{x} perturbed?

Condition number of linear systems II

- We have

$$A(\mathbf{x} + \Delta\mathbf{x}) = \mathbf{b} + \Delta\mathbf{b}$$

- Since $A\mathbf{x} = \mathbf{b}$, we find

$$A(\Delta\mathbf{x}) = \Delta\mathbf{b}$$

- Hence

$$\Delta\mathbf{x} = A^{-1}(\Delta\mathbf{b})$$

Condition number III

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- Using $\mathbf{b} = A\mathbf{x}$ we get

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- Define the **condition number of matrix A**:

$$\kappa(A) := \|A\| \|A^{-1}\|$$

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Residual

- Suppose we are solving $A\mathbf{x} = \mathbf{b}$
- Suppose have approximate solution $\hat{\mathbf{x}}$
- As usual, do not know how far \mathbf{x} is from true solution
- i.e. **forward error** $\hat{\mathbf{x}} - \mathbf{x}$
- Instead, only know the **residual**

$$\mathbf{r} := A\hat{\mathbf{x}} - \mathbf{b}$$

Backward error

- We see that $\hat{\mathbf{x}}$ **exactly** solves a **perturbed problem**:

$$A\hat{\mathbf{x}} = \mathbf{b} - \mathbf{r}$$

- Relative error is

$$\frac{\|\Delta \mathbf{x}\|}{\|\mathbf{x}\|} = \frac{\|\hat{\mathbf{x}} - \mathbf{x}\|}{\|\mathbf{x}\|} \leq \kappa(A) \frac{\|\mathbf{r}\|}{\|\mathbf{b}\|}$$

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- So forward error can be large even when backward error is small, if $\kappa(A)$ is large

Stability of algorithms

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- The notion of “condition number” refers to a **problem**
- Solving a problem on the computer requires an **algorithm**
 - a sequence of steps = mini-problems
- Conditioning of a problem is *independent* of any algorithm for solving that problem
- An algorithm replaces ϕ with an alternative problem $\hat{\phi}$
- The result cannot be better than given by the condition number
- If an algorithm is much **worse** than expected from conditioning, the algorithm is **unstable**; otherwise it's

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- x_- is not affected by catastrophic cancellation (sum of two quantities of a similar size)
- Can we calculate x_+ from x_- ?
- Factor the quadratic as $f(x) = a(x - x_-)(x - x_+)$
- Then $a x_- x_+ = c$, so $x_+ = \frac{c}{a x_-}$

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- Suppose an algorithm $\hat{\phi}$ approximates a problem ϕ
- Then $\hat{y} := \hat{\phi}(x)$ approximates the exact result $y := \phi(x)$
- So far we have studied the **forward error** $\Delta y := \hat{y} - y$
- Instead, we can take an alternative viewpoint:
 - “calculated result = exact solution for nearby input?”
- i.e. want **backward error** Δx such that

$$\hat{y} = f(x + \Delta x)$$

Summary

- Condition number for a problem: Sensitivity of output to input
- Catastrophic cancellation in subtraction and its avoidance
- Condition number for summation, quadratics and linear solve
- Stability
- Backward error