Randomness in Computing

CS 537

LECTURE 1

Randomness in Computing

- Course information
- Verifying polynomial identities
- Probability Amplification
  Probability Review

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Course information

1. Course staff
2. Course website(s)
3. Piazza bonus
4. Prerequisites
5. Textbook(s)
6. Syllabus
7. Homework logistics
8. Collaboration policy
9. Exams and grading

Tips for the course

- Concepts in this course take some time to sink in: be careful not to fall behind.
- Do the assigned reading on each topic before the corresponding lecture.
- Attend the lectures: most of the material will be presented on the blackboard (and some of it is not in the book).
- Attend the discussions: practice problem solving.
- Take advantage of office hours.
- Be active in lectures and on piazza.
- Allocate lots of time for the course: comparable to a project course, but spread more evenly.

Tips for the course: HW

- Start working on HW early.
- Spread your HW time over multiple days.
- You can work in groups (up to 4 people), but spend 1-2 hours thinking about it on your own before your group meeting.

Tips: learning problem solving

To learn problem solving, you have to do it:

- Try to think how you would solve any presented problem before you read/hear the answer.
- Do exercises in addition to HW.

Tips: how to read a math text

- Not like reading a mystery novel.
- The goal is not to get the answers, but to learn the techniques.
- Always try to foresee what is coming next.
- Always think how you would approach a problem before reading the solution.
- This applies to things that are not explicitly labeled as problems.

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Skills we will work on

- Mathematical reasoning
- Expressing your ideas
  - abstractly (suppress inessential details)
  - precisely (rigorously)
- Probabilistic thinking
- Algorithmic thinking
- Problem solving
- Having **FUN** with all of the above!!!

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Uses of Randomness in Computing

- To speed up algorithms.
- To enable new applications:
  - Symmetry breaking in distributed algorithms, cryptography, privacy, online games and gambling.
- To simulate real world events in physical systems: model them as happening randomly.
- To analyze algorithms when data is generated from some distribution:
  - learning theory, data compression.
- To analyze algorithms when errors happen randomly
  - error-correcting codes.
- Analyzing statistics from sampling.

§1.1 (MU) Verifying Polynomial Identities

- $(x+1)(x-2)(x+3)(x-4)(x+5)(x-6) \equiv^{?} x^6 - 7x + 37$

Task: Given two polynomials $F(x)$ and $G(x)$, verify if $F(x) \equiv G(x)$.

Idea 1 (deterministic): Convert both polynomials to canonical form $\sum_{i=0}^{d} c_i x^i$.

Faster with Fourier Transform

- If $F(x)$ is given as $\Pi_{i=1}^{d}(x - a_i)$, conversion by consecutively multiplying monomials requires $\Theta(d^2)$ multiplications of coefficients.

§1.1 (MU) Verifying Polynomial Identities

Task: Given two polynomials $F(x)$ and $G(x)$, verify if $F(x) \equiv G(x)$.

Idea 2 (randomized): Evaluate the polynomials on random integers.

Let d = max degree of $F(x)$ and $G(x)$

1. Pick r uniformly from $\{1, \dots, 100d\}$.
2. Compute $F(r)$ and $G(r)$.
3. **reject** if $F(r) \neq G(r)$; o. w. **accept**.

O(d) ops for product form

Error Analysis: Probability of accepting incorrectly

1-sided error

Fundamental Theorem of Algebra

A polynomial of degree d has at most d roots.

$\Pr[\text{error}] =$

Review: Axioms of Probability

Probability space has three components

- Sample space Ω
- Family of allowable events $E \subseteq \Omega$
- A probability function Pr that maps events E to $\mathbb{R}$ such that
  - $\Pr(E) \in [0,1]$ for any event E;
  - $\Pr(\Omega) = 1$;
  - For any finite or countable sequence of pairwise disjoint events $E_1, E_2, \ldots,$

$$\Pr\left(\bigcup_{i \geq 1} E_i\right) = \sum_{i \geq 1} \Pr(E_i)\,.$$

Review: Inclusion-Exclusion Principle

For any two events E_1 and E_2,

$$\Pr(E_1 \cup E_2) = \Pr(E_1) + \Pr(E_2) - \Pr(E_1 \cap E_2).$$

For any three events E_1, E_2 and E_3,

$$\Pr(E_1 \cup E_2 \cup E_3) = \Pr(E_1) + \Pr(E_2) + \Pr(E_3) - \Pr(E_1 \cap E_2) - \Pr(E_1 \cap E_3) - \Pr(E_2 \cap E_3) + \Pr(E_1 \cap E_2 \cap E_3)$$

For any n events $E_1, E_2, \ldots, E_n$,

$$\Pr\left(\bigcup_{i\in[n]} E_i\right) = \sum_{i\in[n]} \Pr(E_i) - \sum_{i<j} \Pr(E_i \cap E_j) + \sum_{i<j<k} \Pr(E_i \cap E_j \cap E_k) - \cdots + (-1)^{n+1} \Pr\left(\bigcap_{i\in[n]} E_i\right).$$

Review: Union Bound

Union Bound

For finite or countable sequence of events $E_1, E_2, \ldots,$

$$\Pr\left(\bigcup_{i\geq 1} E_i\right) \leq \sum_{i\geq 1} \Pr(E_i)\,.$$

Probability Amplification

- Our algorithm for verifying polynomial identities accepts incorrectly with probability $\leq \frac{1}{100}$

Idea: Repeat the algorithm and accept if all iterations accept.

$$\Pr[\text{error in all } k \text{ iterations}] \leq \left(\frac{1}{100}\right)^k$$

Review: Independence

Independent events

- Two events E_1 and E_2 are independent if
$$\Pr(E_1 \cap E_2) = \Pr(E_1) \cdot \Pr(E_2)\,.$$
- Events $E_1, \ldots, E_n$ are mutually independent if,
for every subset $I \in [n]$,
$$\Pr\left(\bigcap_{i \in I} E_i\right) = \prod_{i \in I} \Pr(E_i)\,.$$
- If all pairs of events among $E_1, \ldots, E_n$ are independent then $E_1, \ldots, E_n$ are pairwise independent.

$[n]$ ***denotes*** $\{1, 2, \ldots, n\}$

- Pairwise independence does not necessarily imply mutual independence!

Independence: Example

- We toss a fair coin twice.
- Let A be the event that the 1st flip is HEADS.
- Let B be the event that the 2nd flip is HEADS.
- Let C be the event that both flips are the same.

Are events A,B,C pairwise independent?

Are they mutually independent?

Review: Conditional Probability

Conditional Probability

The conditional probability of event E given event F is

$$\Pr(E \mid F) = \frac{\Pr(E \cap F)}{\Pr(F)}.$$

Well defined only if $\Pr(F) \neq 0$

- When E and F are independent,

$$\Pr(E \mid F) = \frac{\Pr(E \cap F)}{\Pr(F)} = \frac{\Pr(E) \cdot \Pr(F)}{\Pr(F)} = \Pr(E).$$

Review question

- Event E: the numbers on two dice sum to 8.
- Event F: the numbers on two dice are both even.
- What is the probability of E given that F occurred, $\Pr(E|F)$?

A. Less than 1/3

B. 1/3

C. Greater than 1/3, but less than 2/3

D. 2/3

E. Greater than 2/3

Review question

Card dealing

We deal two cards. What is the probability that the second card is an ace, given that the first is an ace?

A. 3/52

B. 3/51

C. 4/52

D. 5/52

E. None of the answers above are correct.

Product rule

For any two events E_1 and E_2,

$$\Pr(E_1 \cap E_2) = \Pr(E_1) \cdot \Pr(E_2|E_1)\,.$$

For all events $E_1, \ldots, E_n$,

$$\Pr(\cap_{i=1}^{n} E_i) = \Pr(E_1) \cdot \Pr(E_2|E_1) \cdot \ldots \cdot (E_n| \cap_{i=1}^{n-1} E_i)$$

Sampling without replacement

- Let E_i be the event that we choose a root in iteration i

$$\Pr[\text{error in all } k \text{ iterations}]$$
$$= \Pr[E_1 \cap \cdots \cap E_k]$$
$$= \Pr[E_1] \cdot \Pr[E_2|E_1] \cdot \ldots \cdot \Pr[E_k|E_1 \cap \cdots \cap E_{k-1}]$$

- It is 0 if $k > d$.
- If $k \leq d$, then $\Pr[E_j|E_1 \cap \cdots \cap E_{j-1}] = \dfrac{d-(j-1)}{100d-(j-1)}$

$$\Pr[\text{error in all } k \text{ iterations}] \leq \left(\frac{1}{100}\right)^k$$