

I. Polynomial Identity Testing

A. Review

1. Problem statement

- a) Q: Given a polynomial $h \in F[x_1, \dots, x_n]$, is $h=0$?

2. Clarifying the problem.

- a) Def: A polynomial is a finite linear combination of monomials with coefficients in F .

- b) A polynomial is the zero polynomial if all of its coefficients are equal to zero.

(i) Remark: This is the algebraic definition of the zero polynomial.

(ii) The computer-scientific definition would be if it is identically zero as a function: for all choice of input variables, the polynomial evaluates to zero.

(iii) These are equivalent over fields of infinite characteristic, but not for finite fields, e.g.

$$f = x_1^2 + x_1 \text{ over } F_2.$$

(iv) Lead to different problems: we focus on the algebraic definition.

B. Oracle-access model.

1. How is our polynomial presented to us?

- a) If as in the definition — a linear combination of monomials — this problem is trivial

- (i) Simply check each coefficient to see if any of them are non-zero.
- b) (Un)fortunately, there are many ways to skin a cat, and ~~there~~ almost as many ways to be given a polynomial.

~~A polynomial could be given in factored form~~

- (i) e.g. polynomial could be a linear combination of products of factors; e.g.

$$f(x_1, x_2) = (x_1 + x_2)^2 + (x_1 + 2x_2)(2x_1 + x_2) \\ = 0 \text{ over } \mathbb{F}_3.$$

- (ii) multiplying out can be exponentially large \rightarrow not efficient.

2. Oracle access

- a) We shall assume that we do not have access to the polynomial itself, but rather just the polynomial function
- b) i.e. we can ask what value the polynomial takes at given points

C. Zero sets of polynomials

1. Algorithmic strategy.

- a) Essentially the only thing we can do is test the polynomial at a number of different points, and see if we get any non-zero values.
- b) How many zeros do we need to see before we can be satisfied that the polynomial is indeed the zero polynomial?

2. One-dimensional setting

a) Fact: If $f \in \mathbb{F}[x]$ is a non-zero polynomial of degree at most d , then f can have at most d zeros.

b) Pf sketch:

(i) For every zero, have a linear factor of f .

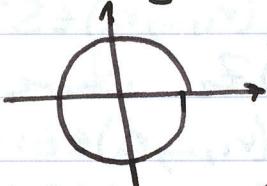
(ii) Induct on quotient (degree)

(iii) Remark: holds for polynomials over any integral domain: don't need a field.

3. Multi-dimensional setting

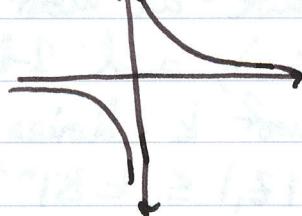
a) zero sets of multivariate polynomials can be much larger

(i) e.g. $f(x,y) = x^2 + y^2 - 1 \in \mathbb{R}[x,y]$



infinitely many zeroes!

(ii) $f(x,y) = xy - 1 \in \mathbb{R}[x,y]$



4. Schwartz-Zippel lemma

a) Idea: zero sets are structured: cannot intersect a cube too often.

b) Theorem (The Schwartz-Zippel lemma)

Let $f \in \mathbb{F}[x_1, \dots, x_n]$ be a non-zero polynomial of degree $d \geq 0$, and $S \subseteq \mathbb{F}^n$ a finite set. Then $| \{(r_1, \dots, r_n) \in S : f(r_1, \dots, r_n) = 0\} | \leq d |S|^{n-1}$.

c) historical remarks

- (i) Discovered for the purposes of polynomial identity testing
- (ii) Similar results obtained independently by DeMillo-Lipton (1978), Zippel (1979), Schwartz (1980).

d) Proof of Schwartz-Zippel

- (i) Induction on n .
- (ii) Base case: $n=1 \equiv$ our earlier fact ✓
- (iii) Induction step. Let x_n be a variable, and let $k \leq d$ be the highest power of x_n appearing in f .
- (iv) $\Rightarrow f(x_1, \dots, x_n) = \sum_{i=0}^k f_i(x_1, \dots, x_{n-1}) x_n^i$, where each $f_i \in \mathbb{F}[x_1, \dots, x_{n-1}]$.

- (v) let $Z = \{(r_1, \dots, r_n) \in S^n : f(r_1, \dots, r_n) = 0\}$.
- (vi) $Z = Z_1 \cup Z_2$, where

$$Z_1 = \{(r_1, \dots, r_n) \in Z : f_k(r_1, \dots, r_{n-1}) \neq 0\}$$

$$Z_2 = \{(r_1, \dots, r_n) \in Z : f_k(r_1, \dots, r_{n-1}) = 0\}$$

- (vii) In Z_1 , have at most $|S|^{n-1}$ choices for (r_1, \dots, r_{n-1}) , and 1-D case \Rightarrow at most k choices for r_n
 $\Rightarrow |Z_1| \leq k |S|^{n-1}$.

- (viii) f_k is a polynomial of degree $\leq d-k$
I.H. $\rightarrow \leq (d-k) |S|^{n-2}$ choices for (r_1, \dots, r_{n-1})
 $\leq |S|$ choices for r_n
 $\Rightarrow |Z_2| \leq (d-k) |S|^{n-1}$
- (ix) $\Rightarrow |Z| \leq d |S|^{n-1}$ □.

D. Randomised algorithm

1. Testing random points

- a) Schwartz-Zippel \rightarrow suffices to test a bounded (but large) number of points.

b) Using randomness gives huge efficiency boost.

2. Algorithm

- a) Fix \mathbb{F} input: polynomial $h \in \mathbb{F}[x_1, \dots, x_n]$ of degree $\leq d$. Q: is $h = 0$?
- b) Fix $S \subseteq \mathbb{F}$, $|S| = 2d$, arbitrarily.
- c) Choose $r_1, \dots, r_n \in S^n$ uniformly at random.
- d) If $h(r_1, \dots, r_n) = 0$, return $h = 0$. Otherwise, return $h \neq 0$.

3. Analysis

- a) No false negatives, only false positives.
- b) Schwartz-Zippel
$$\Rightarrow P(\text{false positive}) \leq \frac{d|S|^n}{|S|^n} = \frac{1}{2}.$$
- c) Testing at k independent points in S^n
$$\Rightarrow P(\text{false positive}) \leq \frac{1}{2^k}.$$

4. Small fields

- a) What if $|\mathbb{F}| < 2d$, so we cannot choose S ?
- b) If $d \geq |\mathbb{F}|$, can make a non-zero polynomial that vanishes everywhere
 - (i) \Rightarrow impossible to distinguish over \mathbb{F}
 - (ii) either restrict $d \leq |\mathbb{F}|$, or work over a field extension instead.

II. Bipartite Matchings

A. Motivation

1. Application of polynomial identity testing: determining if a bipartite graph has a perfect matching.
2. Already have the Augmenting Path Algorithm,

but that is a bit long and unwieldy.

3. This algorithm is simple, and more easily generalised to the non-bipartite setting.

B. Framework

- Given bipartite graph $G = (U \sqcup V, E)$,

$$U = \{u_1, \dots, u_n\}, V = \{v_1, \dots, v_n\}.$$

- Can define a (bipartite) adjacency matrix

$$A = (a_{ij})_{1 \leq i, j \leq n},$$

$$a_{ij} = \begin{cases} 1 & \text{if } \{u_i, v_j\} \in E \\ 0 & \text{otherwise.} \end{cases}$$

C. Permanents, Determinants and Matchings

- Perfect matching in G

Permutation $\sigma \in S_n$ s.t.

$$\{\{u_1, v_{\sigma(1)}\}, \{u_2, v_{\sigma(2)}\}, \dots, \{u_n, v_{\sigma(n)}\}\} \subseteq E(G)$$

Permutation $\sigma \in S_n$ s.t. $\prod_{i=1}^n a_{i\sigma(i)} = 1$

- $\Rightarrow \text{per}(A) := \sum_{\sigma \in S_n} \prod_{i=1}^n a_{i\sigma(i)} = \# \text{ of perfect matchings in } G.$

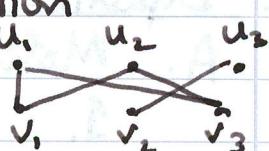
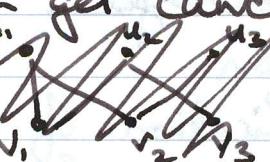
a) Problem: permanent is ~~NP-hard~~
NP-hard to compute!

- Magically, introducing a sign factor gives the determinant, which is easy to compute!

$$\det(A) = \sum_{\sigma \in S_n} \text{sgn}(\sigma) \prod_{i=1}^n a_{i\sigma(i)}$$

a) Problem: can get cancellation

- Example: $G =$



$$A = \begin{pmatrix} \textcircled{I} & 0 & \textcircled{I} \\ \textcircled{I} & 0 & \textcircled{I} \\ 0 & \textcircled{I} & 0 \end{pmatrix}$$

$$\text{per}(A) = 2$$

$$\det(A) = 1 - 1 = 0.$$

5. $\therefore \det(A) \neq 0 \Rightarrow \exists$ perfect matching,
 but $\det(A) = 0 \not\Rightarrow \nexists$ perfect matching.

D. Introducing variables

1. The fix

a) To prevent this harmful cancellation, we replace the entries of A with variables.

b) Let $\tilde{A} = (\tilde{a}_{ij})_{1 \leq i, j \leq n}$, where
 $\tilde{a}_{ij} = \begin{cases} x_{ij} & \text{if } \{u_i, v_j\} \in E \\ 0 & \text{otherwise.} \end{cases}$

c) Now $\det(\tilde{A})$ is a polynomial in $\mathbb{F}[x_{11}, x_{12}, \dots, x_{nn}]$ of degree at most n .

3. Characterisation

a) Lemma: $\det(\tilde{A}) \neq 0 \Leftrightarrow G$ has a perfect matching.

b) Proof

(i) \Rightarrow : If $\det(\tilde{A}) \neq 0$, there is some monomial with a non-zero coefficient.

monomial: $\prod_{i=1}^n x_{i\sigma(i)}$ for some $\sigma \in S_n$ defining a matching ✓

(ii) \Leftarrow : If we have a matching corresponding to $\sigma \in S_n$, let

$$x_{ij} = \begin{cases} 1 & \text{if } j = \sigma(i) \\ 0 & \text{otherwise} \end{cases}$$

$$\Rightarrow \det(\tilde{A})(x) = \pm 1 \neq 0 \quad \square \quad \blacksquare.$$

2. Example: $G = \begin{array}{c} \text{graph} \\ \text{with edges} \end{array}$

$$\tilde{A} = \begin{pmatrix} x_{11} & 0 & x_{13} \\ x_{21} & 0 & x_{23} \\ 0 & x_{32} & 0 \end{pmatrix}, \det(\tilde{A}) = x_{21}x_{43}x_{32} - x_{11}x_{23}x_{32}.$$

E. The algorithm

1. Fix some set $S \subseteq F$, $|S|=2n$.
2. Construct \tilde{A} , substituting a uniformly random value from S for each edge variable x_{ij} .
3. Compute $\det(\tilde{A})$ with these values.
4. If $\det(\tilde{A}) \neq 0$, then G has a perfect matching.
5. If $\det(\tilde{A}) = 0$, return: G has no perfect matching.

F. Analysis and closing remarks

1. Since $\det(\tilde{A})$ is a polynomial of degree $\leq n$, Schwartz-Zippel \Rightarrow prob. of a false negative is at most $\frac{1}{2}$.
2. Which field?
 - a) If $F = \mathbb{R}$ and we take $S = [2n]$, we could deal with numbers as large as $(2n)^n \rightarrow$ not good for computers
 - b) Simpler to work over a finite field, eg. \mathbb{F}_p for some prime $2n \leq p \leq 4n$
 \hookrightarrow bounded calculations.
3. Non-bipartite setting
 - a) In the general case, an n -vertex graph has an $n \times n$ adjacency matrix, but only $\frac{n^2}{2}$ edges in a matching
 \rightarrow no one-to-one correspondence b/w determinant monomials and matchings.
 - b) Need to be a bit cleverer - see HW.
4. Any questions?