

LOW-DEGREE TESTING (version 1)

Input Table of some function $f: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$
 (or $f: \{0,1\}^n \rightarrow \{0,1\}$ - identify $\mathbb{F}_2 = \{0,1\}$)

Problem Decide whether f is a polynomial
 in $\mathbb{F}_2[x_1, \dots, x_n]$ of total degree $\leq r$

Constraint Run very fast
 Ideally in time $O(1)$ in terms of n
 (Running time might depend on r)

Two observations

(1) Every $f: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ is a polynomial,
 so the only question is what the degree is

Define for $a \in \mathbb{F}_2$

$$\chi_a(x) = \begin{cases} 1-x & \text{if } a=0 \\ x & \text{if } a=1 \end{cases}$$

For $\vec{a} \in \mathbb{F}_2^n$

$$\chi_{\vec{a}}(\vec{x}) = \prod_{i=1}^n \chi_{a_i}(x_i)$$

Clearly $\chi_{\vec{a}}$ degree- n polynomial such that

$$\chi_{\vec{a}}(\vec{x}) = \begin{cases} 1 & \text{if } \vec{x} = \vec{a} \\ 0 & \text{otherwise} \end{cases}$$

Then

$$f(\vec{x}) = \sum_{\vec{a} \in \mathbb{F}_2^n} f(\vec{a}) \cdot \chi_{\vec{a}}(\vec{x})$$

(2) Problem as posed above impossible! II

Suppose we have low-degree testing algorithm A
Check what positions $U \subseteq \mathbb{F}_2^n$ are read by A
(will say A QUERIES these positions)

$|U| \ll 2^n$, so algorithm can't detect if
we corrupt positions outside of U

Have to allow RANDOMNESS

Still doesn't help. Consider function table
of $\chi_{\vec{a}}(\vec{x})$ for some fixed $\vec{a} \in \mathbb{F}_2^n$.

Even randomized A will see $\vec{a}$ with probability
 2^{-n} . Apart from $\vec{a}$, function looks like
zero polynomial $\Rightarrow$ low degree
But has degree n !

But $\chi_{\vec{a}}(\vec{x})$ is zero polynomial "except
for noise" in point $\vec{a}$.

Low-degree testing algorithm:

- Should be allowed to use randomness
- Should determine whether f is a low-degree polynomial except possibly for noise.

The DISTANCE between f and g is

$$\delta(f, g) = \Pr_{\vec{x} \in \mathbb{F}_2^n} [f(\vec{x}) \neq g(\vec{x})] = \frac{|\{\vec{x} \in \mathbb{F}_2^n : f(\vec{x}) \neq g(\vec{x})\}|}{|\mathbb{F}_2^n|}$$

Distance from class of functions $\mathcal{G}$

$$\delta(f, \mathcal{G}) = \min_{g \in \mathcal{G}} \{\delta(f, g)\}$$

Let $P_r = \{g \in \mathbb{F}_2[x_1, \dots, x_n] \mid \deg(g) \leq r\}$ III

LOW-DEGREE TESTING (version 2)

Input Table of $f: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$, parameters r, ϵ

Constraint: Query f in constant number of positions (i.e., independent of n , but can depend on r and ϵ)

Task: (a) If $\deg(f) \leq r$, ^{i.e., $f \in P_r$} output YES
(always)
(b) If $\delta(f, P_r) \geq \epsilon$, output NO
with probability at least $2/3$
(probability over internal randomness,
not over input)

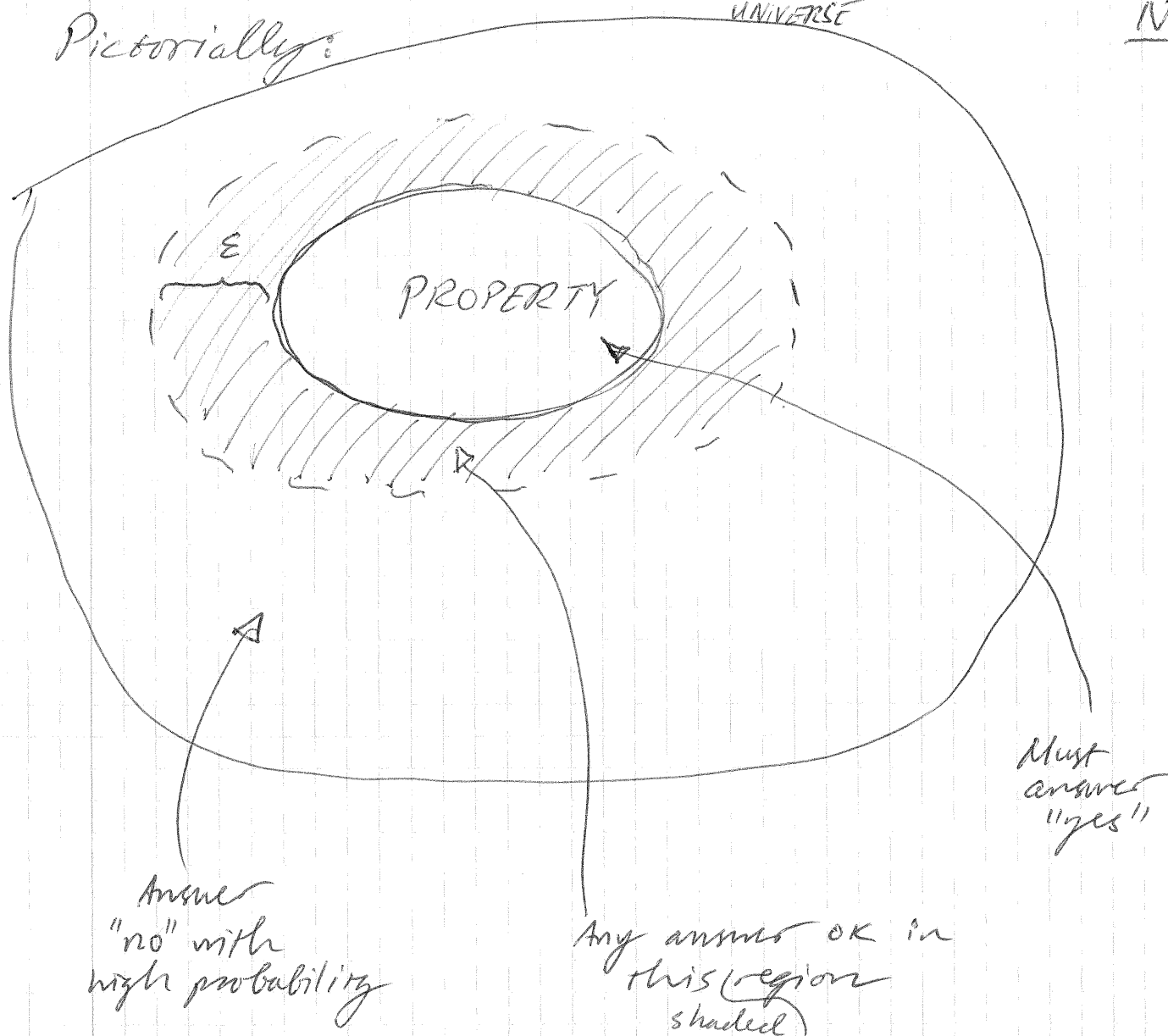
Comments:

- An algorithm A solving this problem is a PROPERTY TESTER with PERFECT COMPLETENESS (no error if $f \in P_r$) and SOUNDNESS ERROR $1/3$, also known as a ONE-SIDED (PROPERTY) TESTER.
- The exact constant $2/3$ ($1/3$) is arbitrary — can go from any constant to any other (smaller) constant by repeating test.
- This is a PROMISE PROBLEM. If neither (a) or (b) applies, then algorithm can give any answer (so we are "promised" that either (a) or (b) applies).

Pictorially:

UNIVERSE

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- There is an implicit parametrization in terms of n , which is usually ignored in the notation. That is, we have

$$\mathcal{P}_r^n = \{g \in \mathbb{F}_2[x_1, \dots, x_n] \mid g \text{ } n\text{-variate, } \deg(g) \leq r\}$$

and

$$\mathcal{P}_r = \bigcup_{n=L}^{\infty} \mathcal{P}_r^n$$

All asymptotic statements are w.r.t $n \rightarrow \infty$
 From now on, we will follow the custom to suppress n in the notation

WHY CARE?

- This kind of property testing is an entire area of research in TCS with many fascinating results.
- Can be viewed as extreme form of decoding-like question for error-correcting codes.

Reed-Muller codewords are evaluations of low-degree polynomials. We are asking whether RM codes are LOCALLY TESTABLE (check in sublinear time whether received word is a codeword or far from ^{all} codewords).

Other flavours of this problem:

- o LOCALLY DECODABLE CODES - decode individual positions in message in sublinear time
- o LOCALLY CORRECTABLE CODES - correct individual positions in codeword correctly (even if this particular position was corrupted in transit)

- Very important for hardness of approximation
Low-degree testing at the heart of the PCP Theorem

([Arora, Lund, Motwani, Sudan, & Szegedy '98] and other papers)

saying, roughly, that any NP-witness (for satisfiability of a CNF formula, say) can be encoded in such a way that it can be checked by randomly querying a constant number of positions.

- Finally, seems like a fairly natural and interesting question about polynomials. An in this course, we care about polynomials, so we care, in particular, about low-degree testing. :-)

TODAY AND NEXT LECTURE:

Discuss result that provides almost optimal tester for low-degree polynomials

Alon, Kaufman, Krivelevich, Litman, & Ron
"Testing Reed-Muller Codes"

IEEE Transactions on Information Theory 2005
(Conference version in APPROX-RANDOM 2003)

Two ways to view testing algorithm

① Code theoretic view

We want to test Reed-Muller code C

Look at dual code $C^\perp = \{ \vec{x} \in \mathbb{F}_2^n \mid \langle \vec{x}, \vec{c} \rangle = 0 \forall \vec{c} \in C \}$

Given received word y

Pick random codewords $\vec{x}_i$ of minimum (constant) also Reed-Muller code

Hamming weight for $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$, $k = O_n(1)$.

Accept y if $\forall \vec{x}_i: \langle \vec{x}_i, y \rangle = 0$

Works if $\vec{y} \in \mathbb{F}_2^n$ if s.t. $\Delta(\vec{y}, C) \geq \epsilon n$ has

high probability of having $\langle \vec{x}_i, \vec{y} \rangle \neq 0$ for random min-weight $\vec{x}_i \in C^\perp$

(Will not use this view from now on.)

② Polynomial view

Pick random subspace S of dimension $r+1$ in $\mathbb{F}_2^n$
 Check if f restricted to S is degree- r polynomial — accept if yes, reject otherwise

For historic reasons, [AKKLRO5] focuses on degree- r polynomials with constant term 0, i.e., $f(\vec{0}) = 0$

So we will do so as well. Let

$$\mathcal{P}_r^* = \{ g \in \mathbb{F}_2[x_1, \dots, x_n] \mid \deg(g) \leq r, g(\vec{0}) = 0 \}$$

We will solve low-degree testing problem for $\mathcal{P}_r^*$ instead. Difference is not very important.

A BRIEF HISTORY

Blum - Luby - Rubinfeld '93

Testing of Hadamard codes

= linear functions over $\mathbb{F}_2^n$ = degree-1 polynomials

Test:

Repeat a suitable number of times:

Pick $\vec{y}_1, \vec{y}_2 \in \mathbb{F}_2^n$ uniformly and independently at random

Reject if $f(\vec{y}_1) + f(\vec{y}_2) \neq f(\vec{y}_1 + \vec{y}_2)$

Accept if all tests passed.

This test clearly accepts all linear functions.

Less obviously, test is good at detecting functions that are far from being linear, as shown in [BLR93]

(Improved analysis in [Bellare, Cooper-Smith, Hastad, Kim, & Sudan '96] using discrete Fourier analysis.)

Also well-studied for fields $\mathbb{F}_q$ of characteristic $q > \text{degree bound}$

General idea:

- Pick random line L
- Check if $f|_L$ (f restricted to L) is univariate degree- r poly
- Checking $r+1$ points on L tells us what $p(t) = f|_L$ should be; now evaluate at some other point to check consistency

This tests illustrate why we should expect dependence on r in # queries - f evaluated on too few points always looks consistent with some polynomial of degree $\leq r$.

Unfortunately, this idea does not work for $\mathbb{F}_2$

Instead, generalize BLR test:

AKKLR-TESTER(f, r, ϵ)

Repeat $\Theta\left(\frac{1}{\epsilon \cdot 2^r} + r \cdot 2^r\right)$ times:

- o uniformly and independently at random choose $\vec{y}_1, \dots, \vec{y}_{r+1} \in \mathbb{F}_2^n$

- o Evaluate f on all nontrivial linear combinations of the $\vec{y}_j$ and reject if

$$\sum_{\emptyset \neq I \subseteq [r+1]} f\left(\sum_{i \in I} \vec{y}_i\right) \neq 0 \quad (1)$$

Accept if all tests passed

THEOREM 1 [AKKLR05]

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The AKKLR - tester:

- (a) has query complexity $\Theta\left(\frac{1}{\epsilon} + r \cdot 2^{2r}\right)$
- (b) always accepts $f \in \mathcal{P}_r^*$
- (c) rejects f that are ϵ -far from $\mathcal{P}_r^*$ with probability $\geq 2/3$ (over internal randomness of algorithm).

We have to prove

- completeness (b) - usually not too hard
- soundness (c) - usually trickier part

Claim (a) about # evaluations of f will follow from closer description

Since we are going to write the basic AKKLR-test in (1) many times, we want some notation

For $f: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ and $\vec{y}_1, \dots, \vec{y}_\ell \in \mathbb{F}_2^n$, let $T_f(\vec{y}_1, \dots, \vec{y}_\ell)$ denote sum of f evaluated at all points in subspace $\text{span}(\vec{y}_1, \dots, \vec{y}_\ell)$ except $\vec{0}$, i.e.,

$$T_f(\vec{y}_1, \dots, \vec{y}_\ell) = \sum_{\emptyset \neq I \subseteq [\ell]} f\left(\sum_{i \in I} \vec{y}_i\right) \quad (2)$$

LEMMA 2 (Completeness)

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$f \in \mathcal{P}_r^*$ if and only if for all $\vec{y}_1, \dots, \vec{y}_r \in \mathbb{F}_2^n$ it holds that $T_f(\vec{y}_1, \dots, \vec{y}_{r+1}) = 0$.

Proof: Need to show

$$(\Rightarrow) f \in \mathcal{P}_r^* \Rightarrow T_f(\vec{y}_1, \dots, \vec{y}_{r+1}) = 0$$

sufficient to prove for monomial m of degree $\leq r$

$$(\Leftarrow) \text{ Write } f(\vec{x}) = \sum_{I \subseteq [n]} a_I \prod_{i \in I} x_i$$

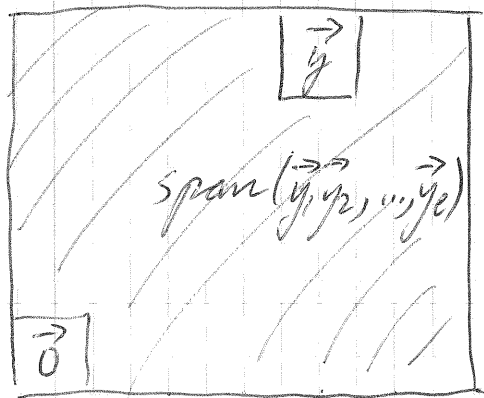
Show that $a_\emptyset = 0$ and $a_I = 0$ for $|I| > r$ if f satisfies $T_f(\vec{y}_1, \dots, \vec{y}_{r+1}) = 0 \forall \vec{y}_1, \dots, \vec{y}_{r+1}$.

Detailed proof left as an exercise. 

Let us also define

$$T_f^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_r) = T_f(\vec{y}, \vec{y}_2, \dots, \vec{y}_r) + f(\vec{y}) \quad (3)$$

$T_f^{\vec{y}}$ evaluates f everywhere on subspace $\text{span}(\vec{y}, \vec{y}_2, \dots, \vec{y}_r)$ except $\vec{0}$ and $\vec{y}$ (since $f(\vec{y})$ -evaluations cancel in $\mathbb{F}_2$).



Or in pictures...

In view of Lem 2, if f is a degree- r poly then $T_f^{\vec{y}}$ should be a way to evaluate $f(\vec{y})$ without looking at $\vec{y}$.

OBSERVATION 3

If $f \in \mathcal{P}_r^*$, then for all $\vec{y}_2, \dots, \vec{y}_{r+1} \in \mathbb{F}_2^n$ it holds that $T_f(\vec{y}, \vec{y}_2, \dots, \vec{y}_{r+1}) = f(\vec{y})$

Proof Adding $f(\vec{y})$ to both sides yields $T_f(\vec{y}, \vec{y}_2, \dots, \vec{y}_{r+1}) = 0$, which holds if f is a degree- r poly with constant term 0 by Lem 2 $\square$

It remains to prove soundness, i.e. that if f is far from being a degree- r poly, then AKLR-tester rejects with (sufficiently) high probability.

Usually proven in contrapositive direction: Suppose that tester for property $\mathcal{P}$ accepts f with "too high" probability.

Then f is inside the "shaded region" in our figure, i.e., there is a $g \in \mathcal{P}$ such that the distance $\delta(f, g)$ is small.

This is how the AKLR proof goes as well

Let us suppose that f is indeed close to some polynomial g

Which polynomial should this be?

How can we recover it?

Idea: Take "majority vote"

Let $g: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ be defined by

$$g(\vec{y}) = \begin{cases} 1 & \text{if } \Pr_{\vec{y}_2, \dots, \vec{y}_{r+1} \sim \mathbb{F}_2^n} [T_{\vec{y}}^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_{r+1}) = 1] \geq 1/2 \\ 0 & \text{otherwise} \end{cases} \quad (4)$$

where $\vec{y}_i \sim \mathbb{F}_2^n$ denotes uniformly random and independent samples.

That is, if we sample a ^{random} subspace containing $\vec{y}$ and evaluate everywhere except $\vec{0}$ and $\vec{y}$, then $g(\vec{y})$ is majority vote on what value at $\vec{y}$ should be to satisfy local constraint that ^{we} have a degree- r poly.

Define also

$$\eta = \Pr_{\vec{y}_1, \dots, \vec{y}_{r+1} \sim \mathbb{F}_2^n} [T_f(\vec{y}_1, \dots, \vec{y}_{r+1}) \neq 0] \quad (5)$$

to be probability that single iteration of AKKLK-tester detects that f is not a degree- r polynomial

OBSERVATION 4

$$\eta = \Pr_{\vec{y}, \vec{y}_2, \dots, \vec{y}_{r+1} \sim \mathbb{F}_2^n} [T_{\vec{y}}^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_{r+1}) \neq f(\vec{y})]$$

Proof Just substitute (3) into (5).

Now the rest of the proof goes roughly as follows:

Suppose that ^{one iteration of} AKKLR-test rejects f with probability $\eta \geq \varepsilon$.

If so, test is doing the right thing — $\eta > 0$ only if f is not degree- r poly.

Furthermore, repeating test $\sim 1/\varepsilon$ times will boost reject probability to constant (which can then be further boosted to $2/3$).

Suppose, however, that $\eta < \varepsilon$. Then we want to find "excuse" for AKKLR test to be reluctant to reject, i.e., show that f is close to some degree- r poly.

This part divides into two claims

- ① f is close to g as defined in (4)
- ② If η is small enough, then g ~~is~~ is a degree- r polynomial.

claim ① is relatively easy.

claim ② is substantially more work (will be the focus of next lecture).

LEMMA 5 (f and g are close)

Let $f: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ be any function, and let $g: \mathbb{F}_2^n \rightarrow \mathbb{F}_2$ be defined by (4) and γ by (5). Then f and g are 2γ -close, i.e., $\delta(f, g) \leq 2\gamma$.

Proof Consider subset $\mathcal{U} \subseteq \mathbb{F}_2^n$ such that for all $\vec{y} \in \mathcal{U}$ it holds that

$$\Pr_{\vec{y}_2, \dots, \vec{y}_{r+1}} [f(\vec{y}) \neq T_f^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_{r+1})] > \frac{1}{2}.$$

Details are left as an exercise. $\square$

The roadmap for the rest of the proof is now as follows.

First, by definition of g we know that for at least half of the r -tuples $\vec{y}_2, \dots, \vec{y}_{r+1} \in \mathbb{F}_2^n$ it holds that

$$g(\vec{y}) = T_f^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_{r+1})$$

But if γ is small, then this ^{in fact} holds for a vast majority of the r -tuples, not just half of them.

LEMMA 6

For every $\vec{y} \in \mathbb{F}_2^n$ it holds that

$$\Pr_{\vec{y}_2, \dots, \vec{y}_{r+1}} [g(\vec{y}) = T_f^{\vec{y}}(\vec{y}_2, \dots, \vec{y}_{r+1})] \geq 1 - 2r\gamma$$

then we will prove that if detection probability γ of one iteration of AKKL-tern is small enough, then g is a low-degree poly

LEMMA 7

If $\gamma < \frac{1}{(4r+2)2^r}$, then $g \in \mathcal{P}_r^*$.

This would already give us a better if we are willing to do $\Omega(1/\epsilon)$ iterations. But we are shooting for only $\Omega(1/(2^r \epsilon))$ iterations.

For this we need:

LEMMA 8

Suppose that $0 < \gamma < \frac{1}{(4r+2)2^r}$. Then for

$\vec{y}_1, \dots, \vec{y}_{r+1}$ chosen uniformly and independently at random, the probability that exactly one point $\vec{v} \in \left\{ \sum_{i \in I} \vec{y}_i \mid \emptyset \neq I \subseteq [r+1] \right\}$ satisfies $f(\vec{v}) \neq g(\vec{v})$ is at least

$$\frac{1}{3} (2^{r+1} - 1) \delta(f, g)$$

COROLLARY 9

If $\gamma < \frac{1}{(4r+2)2^r}$, then $\gamma \geq \frac{1}{3} (2^{r+1} - 1) \delta(f, g)$

Proof g is a polynomial, so $T_f^g(\vec{y}_1, \vec{y}_2, \dots, \vec{y}_{r+1}) = 0$ always. If $f(\vec{v}) \neq g(\vec{v})$ in exactly one point, $T_f^g(\vec{y}_1, \vec{y}_2, \dots, \vec{y}_{r+1}) \neq 0$ and the AKKL-tern rejects f . □

We can now prove the main theorem,

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THEOREM 1 [AKKLROS]

The AKKLK-tester

- (a) has query complexity $\Theta\left(\frac{1}{\epsilon} \cdot r \cdot 2^{2r}\right)$
- (b) always accepts $f \in \mathcal{P}_r^*$
- (c) rejects f s.t. $\delta(f, \mathcal{P}_r^*) \geq \epsilon$ with probability $\geq 2/3$

Proof: If $f \in \mathcal{P}_r^*$, then AKKLK-tester accepts by Lemma 2.

Suppose that $\delta(f, \mathcal{P}_r^*) \geq \epsilon$.

If we run AKKLK-tester for $\Theta(1/\eta)$ iterations, then tester will reject with any constant probability we choose. (Expected #iterations needed = $1/\eta$), and if we do $C \cdot 1/\eta$ iterations then probability of detection failure tails off ^{slowly} like $\exp(-C)$.

So it is sufficient to show that

$$\eta = \Omega\left(\min\left(2^r \cdot \epsilon, \frac{1}{r \cdot 2^r}\right)\right).$$

If $\eta \geq \frac{1}{(4r+2)2^r}$ we're done, so suppose

$$\eta < \frac{1}{(4r+2)2^r}$$

By (Lem 8 and) Cor 9, we have

$$\eta \geq \frac{1}{3} (2^{r+1} - 1) \delta(f, g)$$

By Lemma 5 we have $g \in P_r^*$. Since $\delta(f, P_r^*) \geq \varepsilon$, it follows that

$$\delta(f, g) \geq \varepsilon$$

hence

$$\eta \geq \frac{1}{3} (2^{r+1} - 1) \cdot \varepsilon = \Omega(2^r \cdot \varepsilon)$$

and the theorem follows $\square$

Oh, and # iterations

$$O\left(\frac{1}{2^r \cdot \varepsilon} + r \cdot 2^r\right)$$

plus $2^{r+1} - 1$ evaluations per iteration
yields bound on query complexity