

# FIFTY SHADES OF ADAPTIVITY (IN PROPERTY TESTING)

An Adaptivity Hierarchy Theorem for Property Testing

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Joint work with **Tom Gur** (Weizmann Institute)

“PROPERTY TESTING?”

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# WHY?

Sublinear,

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Sublinear, approximate,

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Sublinear, approximate, randomized

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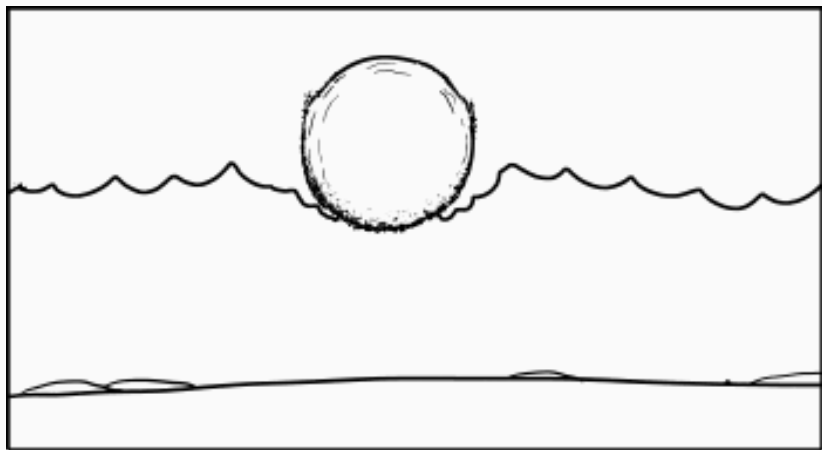
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- Good Enough: **a priori** knowledge

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Sublinear, approximate, randomized decision algorithms that make queries

- Big object: **too** big
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- Good Enough: **a priori** knowledge

Need to infer information – **one bit** – from the data: **quickly**, or with **very few lookups**.



# HOW?

Known space (say,  $\{0, 1\}^N$ )

Property  $\mathcal{P} \subseteq \{0, 1\}^N$

Query (oracle) access to **unknown**  $x \in \{0, 1\}^N$

Proximity parameter  $\varepsilon \in (0, 1]$

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(and be correct on any  $x$  with probability at least  $2/3$ )

# HOW?

Property Testing:



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Property Testing:

in an (egg)shell.

**Many flavors...**

... one-sided vs. two-sided,

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... one-sided vs. two-sided, query-based vs. sample-based,

## Many flavors...

... one-sided vs. two-sided, query-based vs. sample-based, uniform vs. distribution-free,



# ADAPTIVITY

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# OUR FOCUS: ADAPTIVITY

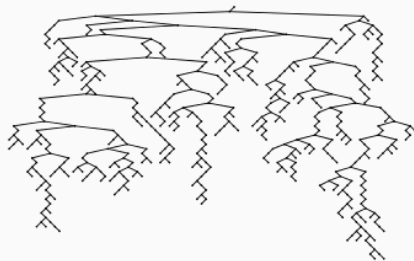
## Non-adaptive algorithm

Makes all its queries **upfront**:

$$Q \subseteq [N] = Q(\epsilon, r) = \{i_1, \dots, i_q\}$$

## Adaptive algorithm

Each query can **depend arbitrarily** on the previous answers:





### Dense graph model

At most a quadratic gap between adaptive and non-adaptive algorithms:  $q$  vs.  $2q^2$  [AFKS00, GT03],[GR11]

### Boolean functions

At most an exponential gap between adaptive and non-adaptive algorithms:  $q$  vs.  $2^q$

### Bounded-degree graph model

Everything is possible:  $O(1)$  vs.  $\Omega(\sqrt{n})$ . [RS06]

## WHY SHOULD WE CARE?

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Fewer queries is **always** better.

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Understanding the **benefits and tradeoffs** of adaptivity is crucial.

## A closer look

Does the power of testing algorithms smoothly grow with the “amount of adaptivity?”

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Does the power of testing algorithms smoothly grow with the “amount of adaptivity?”

(and what does “amount of adaptivity” even mean?)

### Definition (Round-Adaptive Testing Algorithms)

Let  $\Omega$  be a domain of size  $n$ , and  $k, q \leq n$ . A randomized algorithm is said to be a **(k, q)-round-adaptive** tester for  $\mathcal{P} \subseteq 2^\Omega$ , if, on input  $\varepsilon \in (0, 1]$  and granted query access to  $f: \Omega \rightarrow \{0, 1\}$ :

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- (i) Query Generation: The algorithm proceeds in  $k + 1$  rounds, such that at round  $\ell \geq 0$ , it produces a set of queries  $Q_\ell := \{x^{(\ell),1}, \dots, x^{(\ell),|Q_\ell|}\} \subseteq \Omega$ , based on its own internal randomness and the answers to the previous sets of queries  $Q_0, \dots, Q_{\ell-1}$ , and receives  $f(Q_\ell) = \{f(x^{(\ell),1}), \dots, f(x^{(\ell),|Q_\ell|})\}$ ;



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- (ii) Completeness: If  $f \in \mathcal{P}$ , then it outputs **accept** with probability  $2/3$ ;
- (iii) Soundness: If  $\text{dist}(f, \mathcal{P}) > \varepsilon$ , then it outputs **reject** with probability  $2/3$ .

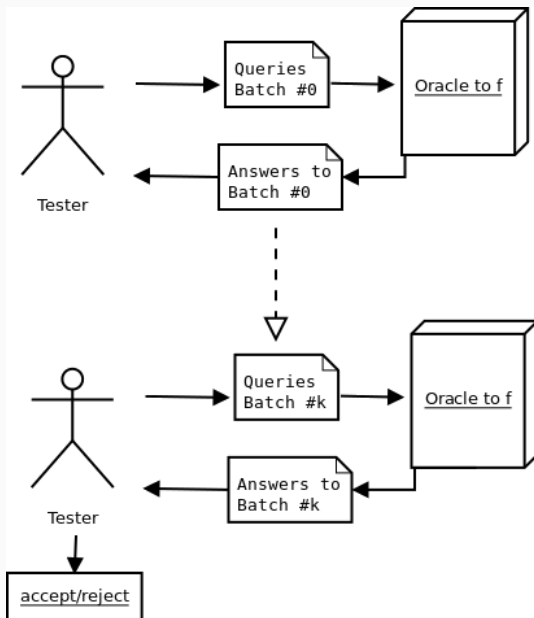
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The **query complexity**  $q$  of the tester is the total number of queries made to  $f$ , i.e.,  $q = \sum_{\ell=0}^k |Q_\ell|$ .

# THAT WAS A MOUTHFUL, BUT... (I CAN'T DRAW)



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- Similar in spirit to...

... now, what do we do with it?

Does the power of testing algorithms smoothly grow with the  
“~~amount of adaptivity~~” **number of rounds** of adaptivity?

## OUR RESULTS

---



... and we have an answer.

Yes, the power of testing algorithms smoothly grows with the number of rounds of adaptivity.

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## Theorem (Hierarchy Theorem I)

For every  $n \in \mathbb{N}$  and  $0 \leq k \leq n^{0.33}$  there is a property  $\mathcal{P}_{n,k}$  of strings over  $\mathbb{F}_n$  such that:

- (i) there exists a  $k$ -round-adaptive tester for  $\mathcal{P}_{n,k}$  with query complexity  $\tilde{O}(k)$ , yet
- (ii) any  $(k - 1)$ -round-adaptive tester for  $\mathcal{P}_{n,k}$  must make  $\tilde{\Omega}(n/k^2)$  queries.

## CAN WE HAVE SOMETHING A BIT LESS CONTRIVED?

It's only natural.

Yes, that also happens for actual things people care about.

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## Theorem (Hierarchy Theorem II)

Let  $k \in \mathbb{N}$  be a constant. Then,

- (i) there exists a  $k$ -round-adaptive tester with query complexity  $O(1/\epsilon)$  for  $(2k + 1)$ -cycle freeness in the bounded-degree graph model; yet
- (ii) any  $(k - 1)$ -round-adaptive tester for  $(2k + 1)$ -cycle freeness in the bounded-degree graph model must make  $\Omega(\sqrt{n})$  queries, where  $n$  is the number of vertices in the graph.

# OUTLINE OF THE PROOF

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## Main Idea

Getting a hierarchy theorem directly for property testing seems hard; but we know how to get one easily in the **decision tree complexity** model. **Can we lift it to property testing?**

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Function  $f$  hard to compute in  $k$  rounds (but easy in  $k + 1$ )



Property  $C_f$  hard to test in  $k$  rounds (but easy in  $k + 1$ )

## OUTLINE OF THE PROOF, CT'D

Fix any  $\alpha > 0$ . Let  $C: \mathbb{F}_n^n \rightarrow \mathbb{F}_n^m$  be a code with constant relative distance  $\delta(C) > 0$ , with

- **linearity**:  $\forall i \in [m]$ , there is  $a^{(i)} \in \mathbb{F}_n^n$  s.t.  $C(x)_i = \langle a^{(i)}, x \rangle$  for all  $x$ ;
- **rate**:  $m \leq n^{1+\alpha}$ ;
- **testability**:  $C$  is a one-sided LTC\* with **non-adaptive** tester;
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**Theorem ([GGK15])**

These things exist.\*

YOU HAVE AWAKENED ME  
FROM THE LAMP. YOU MAY  
HAVE THREE WISHES. WHAT  
DOES YOUR HEART DESIRE?



SWEET!



For any  $f: \mathbb{F}_n^n \rightarrow \{0, 1\}$ , consider the subset of codewords

$$\mathcal{C}_f := C(f^{-1}(1)) = \{ C(x) : x \in \mathbb{F}_n^n, f(x) = 1 \} \subseteq \mathcal{C}$$

**Lemma.** (LDT  $\rightsquigarrow$  PT)

$k$ -round-adaptive tester for  $\mathcal{C}_f$  with query complexity  $q$  implies  $k$ -round-adaptive LDT\* algorithm for  $f$  with query complexity  $q$ .

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$k$ -round-adaptive DT algorithm for  $f$  with query complexity  $q$  implies  $k$ -round-adaptive tester for  $\mathcal{C}_f$  with query complexity  $\tilde{O}(q)$ .

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Transference lemmas

## Putting it together

Apply the above for  $f$  being the **k-iterated address** function

$$f_k: \mathbb{F}_n^n \rightarrow \{0, 1\}.$$

## Lemma

For every  $0 \leq k \leq \tilde{O}(n^{1/3})$ , no  $k$ -round-adaptive LDT algorithm can compute  $f_{k+1}$  with  $o(n/(k^2 \log n))$  queries..

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## Proof.

Reduction to communication complexity,\* lower bound of [NW93] on the “pointer-following” problem. □

## OTHER RESULTS

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THE END IS NIGH

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- Can we **swap the quantifiers** in the theorems? ( $\forall k \exists \mathcal{P}_k \rightsquigarrow \exists \mathcal{P} \forall k$ )

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- Other applications of the **transference lemmas**?

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- Also holds for some **natural** properties
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- **Codes are great!**

THANK YOU



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