

Quantum Property Testing

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Property Testing

In classical Algorithms the typical decision problems is:

For a fixed property \mathcal{P} and a given input x , decide whether x belongs to \mathcal{P} or not.

Sometimes we don't care about an exact answer as there is a 'gray area', or, sometimes, there is not enough time for exact decision. Then the following 'approximation' may be used:

decide whether x has \mathcal{P} or it is very 'far' from having \mathcal{P} .

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- Working with huge data, e.g. genome data, WWW.:
 - difficult to store.
 - cannot look at entire input.
- 'Decide if the election was won by A (versus B)'.

Property Testing - definitions

We encode inputs as strings; $x \in \{0, 1\}^n$ and a property is just a collection of inputs (these that have the property). Namely, $\mathcal{P} \subseteq \{0, 1\}^n$.

Being far: is measured by hamming distance, namely $dist(x, y) = |\{i \mid x_i \neq y_i\}|$ and $dist(x, \mathcal{P}) = \min_{y \in \mathcal{P}} dist(x, y)$.

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For a $\epsilon < 1$ we say that x is ϵ -far from \mathcal{P} if $dist(x, \mathcal{P}) \geq \epsilon n$.

ϵ -Tests

An (ϵ, q) -test for a fixed property \mathcal{P} is a **randomized algorithm** that for unknown input x queries at most q bits of x and:

- If $x \in \mathcal{P}$ then the algorithm accepts it with probability $\geq 2/3$.
- If x is ϵ -far from \mathcal{P} then it gets rejected with probability $\geq 2/3$.

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Interested: $q = o(n)$, better yet $q = \text{poly}(\log n)$ or even better $q = O(1)$.

A concrete Example

Given: a list x_1, x_2, \dots, x_n of Integers.

Property: The list is sorted, $x_1 \leq x_2 \leq \dots \leq x_n$.

Require $\Omega(n)$ time (= queries) for probabilistic algorithms.

Can be done by $\Theta(\sqrt{n})$ quantum algorithm.

Approximation

Given: a list x_1, x_2, \dots, x_n of Integers.

Question: Is the list (almost) sorted, i.e, can change at most ϵ fraction of the numbers to make it sorted.

Can test in $O(1/\epsilon \cdot \log n)$ queries, [Ergun, Kannan, Kumar, Rubinfeld, Viswanathan 2000, Fischer 2001].

Background

Property testing was first defined by Rubinfeld and Sudan [96] who were mainly motivated by the connection to program checking.

The study of this object for combinatorial objects (mainly for graph properties) was introduced by Goldreich, Goldwasser and Ron [96], pointing the connection to approximation algorithms, PAC learning, PCP, etc.

Goldreich et al. showed that the graph property of **being bipartite** is testable in $O(1)$ queries.

Since then property testing became a very active area with many interesting results.

Classically Testable Properties

- **Linearity test** ($\forall x, y, f(x) + f(y) = f(x + y)$) [Blum, Luby and Rubinfeld 93, Bellare, Coppersmith, Hastad, Kiwi and Sudan 95].
- **Graph Properties**—colorability, not containing a forbidden subgraph, connectivity, acyclicity, rapidly mixing, max cut, ... [Goldreich, Goldwasser and Ron 87, Alon, Fischer, Krivelevich and Szegedy 99, Parnas and Ron 99, Bender and Ron 2000, Fischer 2001, Alon 2001]....
- **Monotonicity** [Goldreich, Goldwasser, Lehman, Ron, Dodis, Raskhodnikova and Samorodnitsky 99, Lehman, Fischer, Newman, Rubinfeld, Raskhodnikova and Samorodnitsky 2002 ..].
- **Set properties**—equality, distinctness, ... [Ergun, Kannan, Kumar, Rubinfeld and Viswanathan 98..].
- **Geometric properties**—metrics, clustering, convex hulls,... [Parnas and Ron 99, Alon, Dar, Parnas and Ron 2000, Czumaj and Sohler 2002...].

- Membership in low-complexity languages—regular languages, constant-width branching programs, context-free languages [Alon, Krivelevich, Newman, and Szegedy 99, ...].

Quantum Circuits / Algorithms

We think of an algorithm as a **state transformer**.

Classical algorithm:

- A state is a **bit-vector**; values of all variables / intermediate gates.
- k variables - states are vectors in F_2^k .

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Randomized Algorithm:

- A state is a convex combination of basic-states.
- k variabls - states are vectors in R^{2^k} ,

$$\psi = \sum_{j \in \{0,1\}^k} \alpha_j v_j$$

with $\alpha_j \in \mathbb{R}$ and $\sum_{j \in \{0,1\}^k} \alpha_j = 1$.

Quantum Circuits / Algorithms

Quantum: k -qbits - 2^k basic states, v_j , $j \in \{0, 1\}^k$.

- **States:** are 2^k dimensional vectors, that are linear combinations of basic states. $|\psi\rangle = \sum_{j \in \{0,1\}^k} \alpha_j |j\rangle$ with $\alpha_j \in \mathbb{C}$ and $\sum_{j \in \{0,1\}^k} |\alpha_j|^2 = 1$.

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- **Gate:** unitary operator U (length-preserving matrix).
- **Output:** measurement \mathcal{M} ; for final state $\sum_{j \in \{0,1\}^k} \beta_j |j\rangle$

$$\Pr[\text{output } 1] = \sum_{j \in 1\{0,1\}^{k-1}} |\beta_j|^2$$

Quantum Black-Box Algorithms

Gates:

- **computational gates**, e.g.,

$$\text{NOT} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \text{CNOT} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}, H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

- queries to **oracle** [Beals, Buhrman, Cleve, Mosca and de Wolf 98] for $x \in \{0, 1\}^n$

$$O_x : |j, b\rangle \mapsto |j, b \oplus x_j\rangle \quad (j \in \{0, 1\}^{\log n}, b \in \{0, 1\})$$

Quantum Property Tester

Given a fixed property $\mathcal{P} \subseteq \{0, 1\}^n$.

- Input: n bits/values $x = x_1x_2 \dots x_n$.
- **Quantum tester circuit**, that starts with the state $|00\dots 0\rangle$. Uses O_x oracle gates.
- If $x \in \mathcal{P}$, tester **accepts**.
- If x is ϵ -far from \mathcal{P} , tester **rejects** w.h.p.
- Complexity: number of **# oracle query gates** O_x

Motivation

Show gaps between quantum algorithms and Classical Ones.

Results

- Complexity separations: give properties s.t.

quantum	classical	
$O(1)$	$\Omega(\log n)$	(random Hadamard codewords)
$O(\log n)$	$n^{\Omega(1)}$	(Simon)
$n^{\Omega(1)}$		(pseudo-random numbers)

n = number of values in input

First attempt for a 1 – vs. $\log n$ gap

Inner Product Over F_2 : For $x, y \in \{0, 1\}^k$:

$$\langle x, y \rangle := \sum_{\ell=1}^k y_{\ell} j_{\ell} (\text{mod } 2).$$

Hadamard code of $y \in \{0, 1\}^{\log n}$:

$$h(y) := x_0 \dots x_{n-1} \text{ with } x_j = \langle y, j \rangle$$

Candidate for a ‘classically hard’ property: Being a Hadamard codeword, namely $\mathcal{P} = \{h(y) \mid y \in \{0, 1\}^{\log n}\}$.

\exists quantum black-box algorithm to find y with **one** application of $O_{h(y)}$ [Bernstein Vazirani].

Classically: need $\log n$ queries in order to find y from $h(y)$ (information theory).

Testing Hadamard Codewords

Catch: Classical Tester (does not need to know y).

- for $O(1/\epsilon)$ many pairs j, j' : query x_j , $x_{j'}$, and $x_{j \oplus j'}$.
- **reject** if $x_j \oplus x_{j'} \neq x_{j \oplus j'}$ for any of the pairs. **otherwise:**
accept.

A better candidate

For a subset $A \subseteq \{0, 1\}^{\log n}$, let $P_A = \{h(y) \mid y \in A\}$. Namely, P_A contains the Hadamard codewords of vectors in a predefined subset A .

The subset of Choice - random.

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Quantum Test

- In one query find y such that $h(y) = x$.
- Check that $y \in A$
- Test for random $i \leq n$ that $x_i = \langle y, i \rangle$.

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A $\Omega(\log n)$ classical lower bound can be proven.

Exponential Separation

A property \mathcal{P} such that

\mathcal{P} is quantum-testable with $O(\log n)$ queries.

Any classical tester need $n^{\Omega(1)}$ queries.

Exponential Separation

Simon's Promise problem:

Input: $f : \{0, 1\}^n \longrightarrow \{0, 1\}^n$, such that.

- f is 2 to 1.
- There is an $s \in \{0, 1\}^n - \{(0, \dots, 0)\}$, such that for every x , $f(x) = f(x \oplus s)$.

Goal: Find $s \neq (0, \dots, 0)$.

Quantum - in $O(n)$ queries [Simon 97, Brassard Høyer 97] .

Classical - $\Omega(2^{n/2})$ (birthday paradox).

Brassard Høyer Algorithm

- There are $n - 1$ rounds.
- The i th round produces a vector $z_i \in \{0, 1\}^n$ for which,
 - (a) $\langle z, s \rangle = 0$
 - (b) z_i is linearly independent of $\{z_1, \dots, z_{i-1}\}$.

After $n - 1$ times can find s

Exact !

Our Property

$\mathcal{P} = \{f : \{0, 1\}^n \longrightarrow \{0, 1\} \mid \text{such that } \exists s \neq (0, \dots, 0), \forall x, f(x) = f(x \oplus s)\}.$

Quantum - in $O(n \log n)$ queries.

Classical - $\Omega(2^{n/2})$.

Quantum Lower bounds for Property Testing

- Most of the properties \mathcal{P} of n bit strings, of size $2^{n/20}$ require $\Omega(n)$ quantum queries.
- The range of d -wise independent n -bit generator requires $d/2$ random queries.

we use the Polynomial method.

Polynomial method

Let $f : \{0, 1\}^n \longrightarrow \{0, 1\}$.

[Beals, Buhrman, Cleave, Mosca, d' Wolf 98] If f has a q -query quantum algorithm, then there is a multilinear polynomial p that approximates f .

For all $x \in \{0, 1\}^n$,

$$|p(x) - f(x)| \leq 1/3$$

This is true for promise problems too - in particular for property testing

Proving Q-lower bounds for property testing

To prove lower bounds for a property \mathcal{P} , need to show that every real, multilinear polynomial p for which,

- For all $x \in \mathcal{P}$, $p(x) \geq 2/3$.
- For all x , $\text{dist}(x, \mathcal{P}) \geq \epsilon n$, $p(x) \leq 1/3$.
- For all $x \in \{0, 1\}^n$, $p(x) \in [0, 1]$

Has high degree.

Open Problems

- Gaps of 1 vs. $\Omega(n)$?
- Natural properties.
- Characterization of efficient Q-testers in terms of polynomials ?