

Brief Announcement: Erasure-Resilience Versus Tolerance to Errors

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Abstract

We describe work in progress on providing a separation between erasure-resilient and tolerant property testing. Specifically, we are able to exhibit a property which is testable (with the number of queries independent of the length of the input) in the presence of erasures, but is not testable tolerantly.

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1 Description

In this brief announcement, we describe our recent investigation of the effects of adversarial corruption to inputs on the complexity of sublinear-time algorithms. Input corruption occurs either in the form of errors (when some values get changed) or in the form of erasures (when some values go missing). Understanding the relative difficulty of designing algorithms that work in the presence of different forms of corruption is a problem of fundamental importance. It is with this motivation in mind that property testing [5, 7], one of the most widely studied models of sublinear-time algorithms, was generalized to erasure-resilient testing [3] and (error) tolerant testing [6].

Erasure-resilient property testing falls between (standard) property testing and tolerant testing. Specifically, an erasure-resilient tester for a property, in the special case when no erasures occur, is a standard tester for this property. Also, a tolerant tester for a property implies the existence of an erasure-resilient tester with comparable parameters for the same property. Dixit, Raskhodnikova, Thakurta and Varma [3] separate standard and erasure-resilient testing by describing a property that is *easy* to test in the standard model and *hard* to test in the erasure-resilient model. Their separation is based on an earlier result by Fischer and Fortnow [4] that separates standard property testing from tolerant property testing in the same sense. Their main tool is PCPs of proximity (also known as assignment testers) defined by Ben-Sasson, Goldreich, Harsha, Sudan and Vadhan [1] and by Dinur and Reingold [2]. Dixit et al. [3] asked whether it is possible to obtain a separation between



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41 erasure-resilient and tolerant testing. Here, we announce such a separation. Specifically,
 42 we are able to describe a property testable in the erasure-resilient model with the query
 43 complexity independent of the input size, but for which the query complexity of tolerant
 44 testing grows with the input size.

45 1.1 Erasure-Resilient and Tolerant Testing: Definitions

46 We now describe the erasure-resilient and tolerant models of testing. A *property* \mathcal{P} is a set of
 47 strings. A string is α -erased for $\alpha \in [0, 1)$ if at most an α fraction of its values are erasures
 48 (denoted by \perp). A *completion* of an α -erased string $x \in \{0, 1, \perp\}^n$ is a string $y \in \{0, 1\}^n$ that
 49 agrees with x on all the positions where x is nonerased. An α -erasure-resilient ε -tester [3]
 50 for a property \mathcal{P} is a randomized algorithm that, given parameters $\alpha \in [0, 1), \varepsilon \in (0, 1)$
 51 and oracle access to an α -erased string x , accepts with probability at least $2/3$ if x has a
 52 completion in \mathcal{P} and rejects with probability at least $2/3$ if, in every completion of x , at
 53 least an ε fraction of the **nonerased** positions has to be changed to get a string in \mathcal{P} . The
 54 property \mathcal{P} is α -erasure-resiliently ε -testable if there exists an α -erasure-resilient ε -tester
 55 for \mathcal{P} with query complexity that depends only on the parameters α and ε (but not on the
 56 length of the input string).

57 A string $x \in \{0, 1\}^n$ is ε' -far (α -close) from (to, respectively) a property \mathcal{P} , if the
 58 normalized Hamming distance of x from \mathcal{P} is at least ε' (at most α , respectively). An (α, ε') -
 59 tolerant tester [6] for \mathcal{P} is a randomized algorithm that, given parameters $\alpha \in (0, 1), \varepsilon' \in (\alpha, 1)$
 60 and oracle access to a string x , accepts with probability at least $\frac{2}{3}$, if x is α -close to \mathcal{P} and
 61 rejects with probability at least $\frac{2}{3}$, if x is ε' -far from \mathcal{P} . The property \mathcal{P} is (α, ε') -tolerantly
 62 testable if there exists an (α, ε') -tolerant tester for \mathcal{P} with query complexity that depends
 63 only on the parameters α and ε' (but not on the length of the input string).

64 1.2 Comparison of parameters

65 We remark that, while comparing the above two models, it is appropriate to compare
 66 $(\alpha, \alpha + \varepsilon(1 - \alpha))$ -tolerant testing of a property \mathcal{P} with α -erasure-resilient ε -testing of \mathcal{P}
 67 for the same values of α and ε . The parameter α in both the models is an upper bound on
 68 the fraction of corruptions (erasures, or errors) that an adversary can make to an input.
 69 An α -erasure-resilient ε -tester rejects with probability at least $\frac{2}{3}$ if, for every way of
 70 completing an input string, one needs to change at least an ε fraction of the remaining part
 71 of the input to make it satisfy \mathcal{P} . Similarly, an $(\alpha, \alpha + \varepsilon(1 - \alpha))$ -tolerant tester rejects with
 72 probability at least $\frac{2}{3}$ if, for every way of *correcting* α fraction of the input values, one needs
 73 to change at least an ε fraction of the remaining $(1 - \alpha)$ fraction of the input to make it
 74 satisfy \mathcal{P} .

75 1.3 Our Results

76 Our main contribution is the following theorem which states that there exists a property
 77 that is erasure-resiliently testable and is not tolerantly testable. This proves that tolerant
 78 testing is, in general, a harder problem than erasure-resilient testing.

79 ► **Theorem 1** (Main Theorem). *There exists a property \mathcal{P} and constants $\varepsilon, \alpha \in (0, 1)$ such
 80 that*

- 81 ■ \mathcal{P} is α -erasure-resiliently ε -testable;
- 82 ■ \mathcal{P} is not $(\alpha, \alpha + \varepsilon(1 - \alpha))$ -tolerantly testable.

2 Conclusions

To summarize, we solve an open question proposed by Dixit et al. [3] and prove that tolerant testing is harder than erasure-resilient testing.

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