

TOLERANT JUNTA TESTING AND THE CONNECTION TO SUBMODULAR OPTIMIZATION AND FUNCTION ISOMORPHISM

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WHAT IS A K -JUNTA, AND WHY SHOULD WE CARE?



Object of interest: Boolean function $f: \{0, 1\}^n \rightarrow \{0, 1\}$. Might want to *learn, approximate, manipulate* f – but n is **huge**. This will take time, and resources.

Hope: many **irrelevant features**. What if f actually only depended on $k \ll n$ variables? We then could try to “pay” k instead of n everywhere!

Goal: given blackbox access to f and a parameter k , **find out** if the function is a k -variate function “in disguise.”

Now, even that may not be enough: we want to be **robust**. If our function only *mostly* depends on k variables, that should be good enough! I.e., we want to be able to **tolerate a little bit of noise**.

JUNTAS, TESTING, AND TOLERANCE

Definition. A Boolean function $f: \{0, 1\}^n \rightarrow \{0, 1\}$ is said to be a k -junta if there exists a set $T \subseteq [n]$ of size at most k , such that $f(x) = f(y)$ for every two assignments $x, y \in \{0, 1\}^n$ that satisfy $x_i = y_i$ for every $i \in T$.

We want to detect juntas efficiently, to avoid insane running times depending on n whenever possible. And this can be done:

Theorem ([3, 4, 5, 6]). Testing whether a Boolean function $f: \{0, 1\}^n \rightarrow \{0, 1\}$ is a k -junta has query complexity $\tilde{O}(k/\epsilon)$, **independent of n** .

But what about this robustness we would like to obtain? Can we test efficiently whether a function is *close* to a junta?

Definition. A tolerant testing algorithm for a property \mathcal{P} is a probabilistic algorithm \mathcal{T} that gets two input parameters $\epsilon_1, \epsilon_2 \in [0, 1]$ with $\epsilon_1 < \epsilon_2$, and oracle access to a function $f: \{0, 1\}^n \rightarrow \{0, 1\}$; and outputs a binary verdict that satisfies the following two conditions.

- If $\text{dist}(f, \mathcal{P}) \leq \epsilon_1$, then \mathcal{T} accepts with probability at least $2/3$.
- If $\text{dist}(f, \mathcal{P}) > \epsilon_2$, then \mathcal{T} rejects with probability at least $2/3$.

Case $\epsilon_1 = 0$: “usual” testing. But being tolerant is harder – and sometimes **much** harder [2]. *Is it the case here?*

SUMMARY OF RESULTS

We give two (incomparable) results for **tolerant testing of k -juntas**, each with query complexity *independent of n* .

Theorem. There exists an algorithm that, given query access to $f: \{0, 1\}^n \rightarrow \{0, 1\}$ and parameters $k \geq 1$ and $\epsilon \in (0, 1)$, satisfies the following.

- If f is $\epsilon/10$ -close to some k -junta, then the algorithm accepts with probability at least $2/3$.
- If f is ϵ -far from every $2k$ -junta, then the algorithm rejects with probability at least $2/3$.

The query complexity of the algorithm is $\text{poly}(k, \frac{1}{\epsilon})$.

Exploits a connection to **submodular minimization**: approximate minimization of a (noisy) submodular function under a cardinality constraint. Yields an *efficient* algorithm for our testing problem – with a small catch.

Our second algorithm does not include that relaxation of the soundness condition, but features a **tradeoff** between tolerance and query complexity:

Theorem. There exists an algorithm that, given query access to $f: \{0, 1\}^n \rightarrow \{0, 1\}$ and parameters $k \geq 1$, $\epsilon \in (0, 1)$ and $\rho \in (0, 1)$, satisfies the following.

- If f is $\rho\epsilon/16$ -close to some k -junta, then the algorithm accepts with high constant probability.
- If f is ϵ -far from every k -junta, then the algorithm rejects with high constant probability.

The query complexity of the algorithm is $O\left(\frac{k \log k}{\epsilon \rho (1-\rho)^k}\right)$.

Retrieves weakly tolerant results of Fischer et al. [7] for $\rho = \Theta(1/k)$, and tolerant tester with query complexity $O(2^k/\epsilon)$ for $\rho = \Omega(1)$. Setting ρ , this can also be leveraged to obtain the following:

Application: “instance-by-instance” (tolerant) isomorphism testing of $f, g: \{0, 1\}^n \rightarrow \{0, 1\}$. **“Why pay n if there is a better parameter $k = k(f, g)$?”**

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