

# 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  $\epsilon$ -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  $\epsilon$ -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  $\rho$ , 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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