

# Tolerant Junta Testing and the Connection to Submodular Optimization and Function Isomorphism

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  - ▶ Connections to hardness of approximation.

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**Theorem ([FKRSS04, CG04, Bla09]).**

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However, in many practical scenarios the function is not exactly a  $k$ -junta but close to such.

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The proof uses a techniques from **submodular optimization**

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

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### Theorem 3.

There is an algorithm that given query access to  $f$  and  $g$  and any  $\epsilon > 0$  satisfies the following:

- ▶ If  $f$  and  $g$  are  $\epsilon/c$ -close to **isomorphic**, then the algorithm accepts with high probability.
- ▶ If  $f$  and  $g$  are  $\epsilon$ -far from **isomorphic**, then the algorithm rejects with high probability.

The query complexity of the algorithm is  $O(2^{k^*/2}/\epsilon)$  where  $k^*$  is the smallest  $k$  such that either  $f$  or  $g$  are  $\epsilon/c$ -close to a junta.

**For more details, come talk to me during the poster session**

**Thanks**