

Comparison-based sorting

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Overview

Introduction

Counting

Information theory

Game theory

Sorting

Given a list $(l_i)_{i=1}^n$ and an ordering \prec , find permutation π s.t.

$$l_{\pi_i} \prec l_{\pi_{i+1}}, \quad \forall i \in \mathbb{N}, 1 \leq i \leq n - 1.$$

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Comparison-based: only access l through queries against oracle

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Assume algorithm *functionally pure*, i.e. deterministic.

A simple counting argument

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So $2^k \geq n!$, or $k \geq \Theta(n \log n)$.

A fatal flaw

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Possible responses still 2^k , but possible *queries* $\binom{n}{2}^k$ for

$$2^k \binom{n}{2}^k = [n(n-1)]^k \leq n^{2k},$$

resulting in the disappointing trivial bound $k \geq \Theta(n)$.

An information-theoretic perspective?

Distribution X over permutations

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So we need $\Theta(n \log n)$ queries.

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Even so, this is sufficient to show no algorithm is $o(n \log n)$.

A different perspective

We can view finding a *lower bound* on the *worst-case* as

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Instead, *adaptively* choose ℓ dynamically.

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The best \mathcal{A} can possibly do is split the set evenly.

This means \mathcal{A} must do $\Omega(n \log n)$ queries on this list.