

t -Scrambling Permutations

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Permutation Patterns in Paris, July 4 2013

Other Collaborators and Titles

Basic Definitions and Equivalences

Bounds

Thresholds

Current and Future Work

Collaborators and Titles

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- ▶ Shattering n permutations in any t positions (also Statistics and learning theory);
- ▶ t -covering arrays (combinatorial design theory);
- ▶ The phrase “scrambling” has been used before in the permutations context, as pointed out by a referee of our submitted paper.

VC Dimension

DEFINITION: A class \mathcal{F} of subsets of a set X is said to **shatter** a subset $A = \{a_1, \dots, a_t\} \subseteq X$ if

$$\forall S \subseteq A, \exists F \in \mathcal{F} \text{ such that } A \cap F = S,$$

or equivalently, if

$$|\{A \cap F\} : F \in \mathcal{F}| = 2^t.$$

DEFINITION: The **VC dimension** of \mathcal{F} , $\text{VC}(\mathcal{F})$, is the cardinality of the smallest subset not shattered by \mathcal{F} . If all subsets of finite size are shattered by \mathcal{F} , then the $\text{VC}(\mathcal{F}) = \infty$.

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- ▶ Many authors define the VC dimension to be the size of the largest shattered set, in this case our values would be 1 and ∞ respectively.
- ▶ Often, the two numbers are off by one, as in Example 1.

Covering Arrays

- ▶ A $k \times n$ array with entries from the alphabet $\{0, 1, \dots, q - 1\}$ is said to be a (t, q, n, k) -covering array, or briefly a t -covering array, if for each of the $\binom{n}{t}$ choices of t columns, each of the q^t q -ary words of length t can be found at least once among the rows of the selected columns.

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- ▶ If $q = 2$, we can interpret any row as the characteristic vector of a subset of $[n]$ – by making a correspondence between the positions where the row has ones, and the set of those positions.
- ▶ We thus have the following alternative formulation of covering arrays: A family \mathcal{F} of subsets of $[n]$ is a t -covering array if for each $\{a_1, \dots, a_t\} \subset [n]$,

$$|\{\{a_1, \dots, a_t\} \cap F\} : F \in \mathcal{F}\}| = 2^t.$$

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- ▶ We say that a $k \times n$ rectangular array of k permutations on $[n]$ is t -scrambling if for each set of t columns, each of the $t!$ permutations on $[t]$ may be found in an order isomorphic fashion at least once among the rows of the selected columns.

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- ▶ There are clear parallels with VC, shattering, t -covering arrays etc, but the original terminology is “scrambling”.

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- ▶ How much larger can such an array be?
- ▶ They provided superexponential bounds on the size of the extremal such array.

Known Results

- ▶ Let $k = m(t, n)$ be the size (i.e. number of rows) of the *smallest* array that is t -scrambling, i.e., all $t!$ perms are present in any set of t columns, i.e. arrays with VC dimension $\geq t + 1$. Then

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- ▶ Lower bounds of Füredi (1996) were improved by Radhakrishnan (2003).
- ▶ We were able to improve the Spencer 1972 bound using the Lovász Local Lemma:
- ▶ (deGraaf, G, Koch, Lan, 2013+):
 $m(n, t) \leq \frac{(t-1) \lg n}{\lg(t!/(t!-1))}$; $t \geq 4$. Furthermore, a log log result holds:

Improved Upper Bounds (deGraaf, G, Koch, Lan, 2013+)

- ▶ If $m(n, t, \lambda)$ is the smallest number of rows so that for any choice of t columns, each of the $t!$ permutations are present at least λ times among the rows of the selected columns, then....

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- ▶ We conjecture that the t can be replaced by $t - 1$.

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- ▶ E_i is dependent on at most d other E_j , with
- ▶ $ep(d + 1) < 1$, then
- ▶ $\mathbb{P}(\text{none of the } E_i \text{ occur}) = \mathbb{P}(\cap E_i^C) > 0$

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- ▶ $\mathbb{P}(E_i) \leq t!(t! - 1/t!)^k := p$.
- ▶ $d = O(n^{t-1})$ (why?)
- ▶ This, on simplification, yields the required bound;
 $\mathbb{P}(\cap E_i^C) > 0$ means that a construction exists, which yields an upper bound for the minimum size of a scrambling array.

Key Results of this talk

Theorem

Let $t = 3$. Then, for $\phi(n)$ growing to infinity arbitrarily slowly we have

$$k \leq (3\lg n - \phi(n)) / \lg(6/5) \Rightarrow \mathbb{P}(\text{array is } 3\text{-scrambling}) \rightarrow 0; n \rightarrow \infty$$

and

$$k \geq (3\lg n + \phi(n)) / \lg(6/5) \Rightarrow \mathbb{P}(\text{array is } 3\text{-scrambling}) \rightarrow 1; n \rightarrow \infty.$$

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- ▶ $\mathbb{P}(X \geq 1) \leq \mathbb{E}(X) \leq \binom{n}{3} \cdot 6 \cdot (5/6)^k \rightarrow 0$ if $k \geq (3 \lg n + \phi(n)) / \lg(6/5)$.

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- ▶ Thus the chance that the array is 3-scrambling tends to one.

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- ▶ by Chebychev's inequality, and an intricate correlation analysis reveals that this quantity tends to zero with n if $k \leq (3\lg n - \phi(n))/\lg(6/5).$
- ▶ Thus, the chance that the array is scrambling tends to zero!

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- ▶ Several techniques are being explored, including packing permutations, various statistics for, e.g., 312-avoiding permutations, etc.
- ▶ With Yuan and Koch, we are investigating similar questions for t -covering arrays, in which we are to shatter sets ($q = 2$) and words/multisets, $q \geq 3$.