

DESCRIPTION OF WORK

1. *Graph colouring is hard on average for Polynomial Calculus and Nullstellensatz*

With Jonas Conneryd, Susanna de Rezende, Jakob Norström, Kilian Risse
FOCS 2023

Description. This work proves that Polynomial Calculus, hence also Nullstellensatz, requires linear degree to refute the 3-colorability of sparse random graphs and random regular graphs. This gives optimal, exponential size lower bounds via the known size-degree relation.

The proof constructs an Alekhovich-Razborov pseudo-reduction operator, based on the Romero-Tunçel technique which was originally designed for large-girth graphs. We extend this technique to rid of the large-girth assumption, which enables us to deal with general sparse expanders in an elegant and simple way.

2. *SoS lower bound for exact planted clique*

CCC 2021

Description. This work proves Sum-of-Squares (SoS) degree lower bounds for the Exact Planted Clique problem on random graphs $G(n, \frac{1}{2})$. The algorithms' task is to refute the existence of a size ω clique in a sample of random graphs, where ω is so large that with probability say $> 99.99\%$ the sample cannot contain such a large clique. Here, by 'exact' we mean the SoS algorithms are allowed to reason with polynomial identities generated by axioms including $\sum_{i=1}^n x_i = \omega$, the one about the size of the (nonexistent) clique. This is a canonical and feasible modeling of the SDP algorithms, while previous lower bound method have to weaken this axiom into an approximate one. We overcome this shortcoming by proving a SoS degree lower bound in this setting in the form $d = \Omega(\frac{\epsilon^2 \log n}{\log \log n})$ if $\omega = O(n^{\frac{1}{2}-\epsilon})$, which is almost optimal in d and ω .

One motivation of the work is to further develop the average-case SoS lower bound techniques. We define the pseudo-expectation operator $\tilde{E}(\cdot)$ differently than the popular pseudo-calibration method; the cost is, the resulting moment matrix is complicated and is quite *rigid* in a sense, so that the factorization method breaks down. We use a two-step decomposition to deal with it: 1. An Hadamard product with Johnson schemes, inspired by [Feige-Krauthgamer03]; and 2. A relativized factorization [Barak et al.16].

The task then reduces to studying a special family of matrices, which we term the factorial Hankel matrices. The decomposition argument relies on somewhat intricate combinatorial transforms, whose applicability depends on the choice of $\tilde{E}(\cdot)$ in, again, a very rigid way. A conceptual question remains: is there a clear a priori rationale, in either analytic or combinatorial terms, that justifies the apparently contingent choice we made here?

3. *On CDCL-based proof systems with the ordered decision strategy*

With Nathan Mull and Alexander Razborov

SAT 2020, SICOMP 2022

Description. In this work, we prove that conflict-driven clause learning (CDCL) SAT-solvers with the *ordered decision strategy* and *DECISION learning scheme*, are equivalent to ordered resolution. We also prove that if replacing this learning scheme with its opposite, which stops backtracking right after the first non-conflict clause, then the solvers become equivalent to general resolution. This is among the first theoretical studies of the interplay between specific decision strategies and clause learning.

For both results, we allow nondeterminism in the solver’s ability to perform unit propagation, conflict analysis, and restarts, in a way similar to previous works in the literature. To aid the presentation of our results, and possibly future research, we define a model and language for discussing CDCL-based proof systems that allow for succinct and precise theorem statements.

4. *Large clique is hard on average for resolution*

CSR 2021

Description. The main result is a $2^{\Omega(k^{1-o(1)})}$ resolution size lower bounds for the k -Clique problem on suitable random graphs, for $k < n^{1/3}$. This complements the result in [Beame-Impagliazzo-Subharwal07] which is for $k > n^{5/6}$. The proof uses the classical bottleneck counting/random restriction framework plus a variant of clause width, which is defined using neighborhood density that appeared in [Beyesdorff-Galesi-Lauria13, Atserias-Bonacina-De Rezende-Lauria-Nordström-Razborov18].

Work in Preparation

1. *Hardness condensation and applications*

With Duri Janett and Jakob Norström

2. *SoS lower bounds for Non-Gaussian Component Analysis*

With Ilias Diakonikolas, Sushrut Karmalkar, and Aaron Potechin