Fluctuations of random 2-SAT

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INTRODUCTION

- 2-CNF is a boolean formula, written as $(l_1 \lor l_2) \land \ldots \land (l_{2m-1} \lor l_{2m})$, where l_i is either a variable or a negation of a variable.
- Satisfiable assignment maps each variable to a value, such that in each clause there is at least one true literal.
- Consequently, random 2-CNF consists of *n* variables and $m = \lfloor \frac{dn}{2} \rfloor$ uniformly cho-

PROOF IDEA

Two main questions are:

- 1. How to show assymptotic normality?
- 2. How to calculate the variance?

First part is just an application of a martingale central limit theorem, however the variance calculation is a bit more tricky.

For $M \in \{0, ..., m\}$ let us generate a pair of random 2-CNFs $(\Phi_1^{(M)}, \Phi_2^{(M)})$. First we generate M random clauses and add them to both formulas. Second we generate two independent sets of m - M

sen clauses. Here d can be viewed as an average number of clauses which contain any chosen variable.

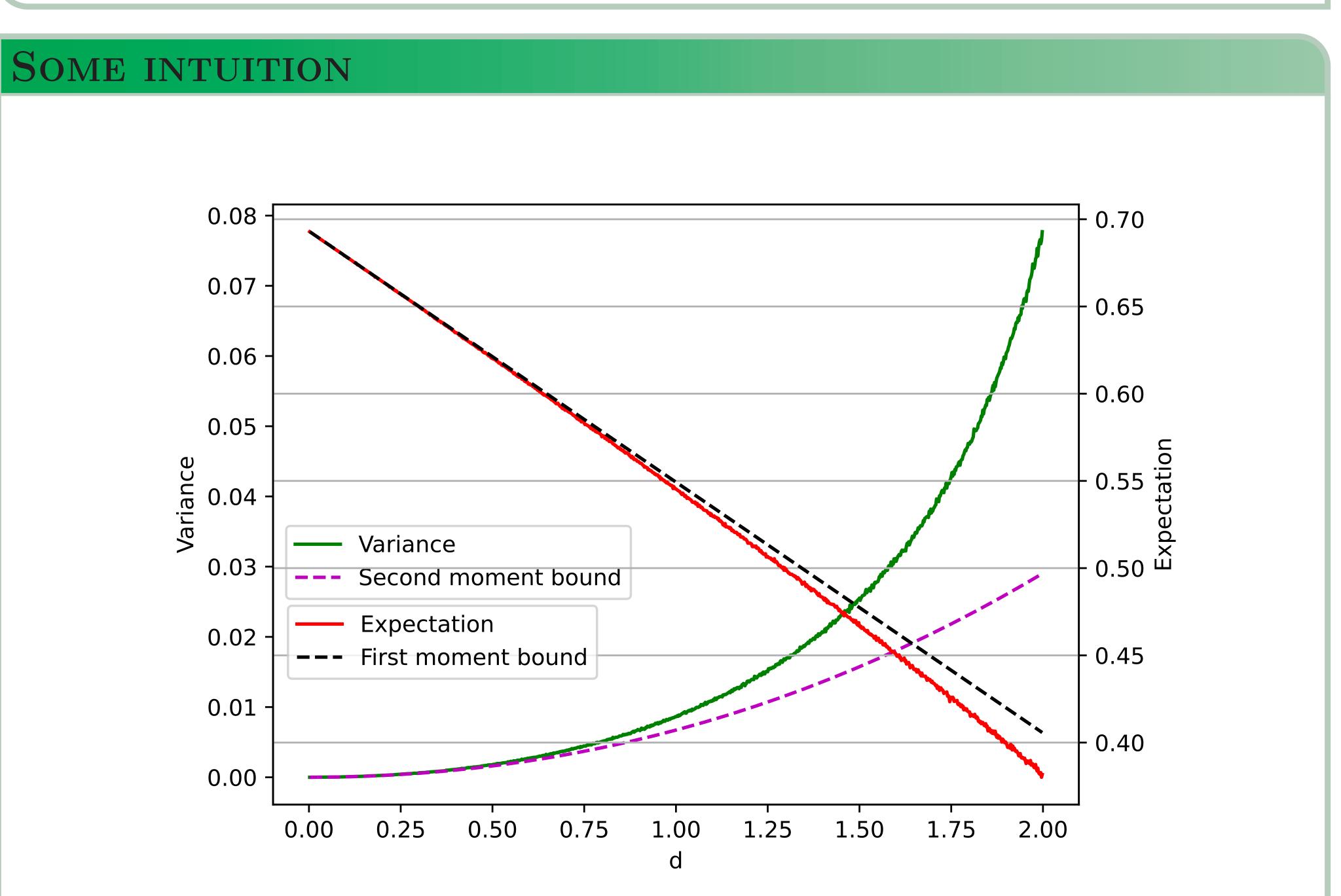
KNOWN RESULTS

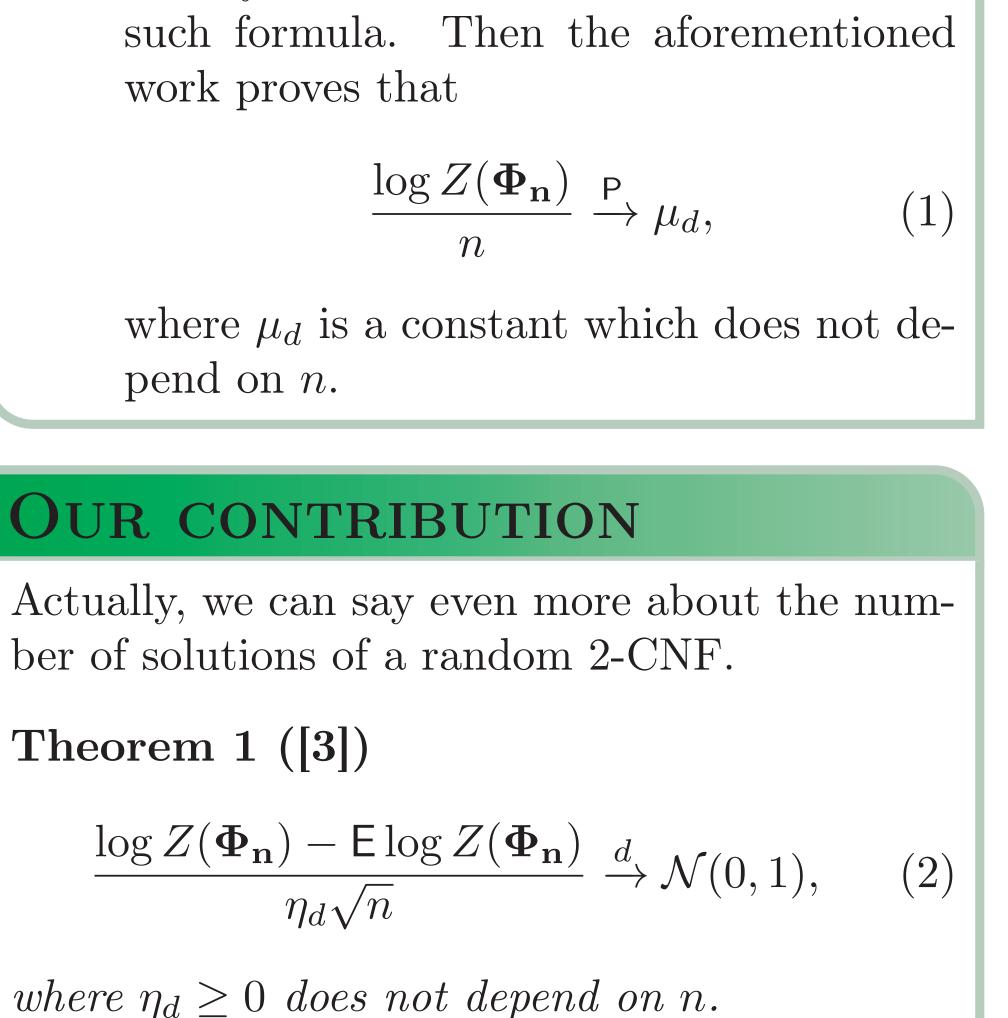
1 The well-known result [1] states that d = 2is a threshold for satisfability of a random 2-CNF.

In other words, for any $\varepsilon > 0$ probability that random 2-CNF is satisfiable tends to one if $d < 2 - \varepsilon$ and to zero if $d < 2 + \varepsilon$, as $n \to +\infty$.

2 Another question is how much solution does the random 2-CNF have below the satisfability threshold (e.g. d < 2). The first-order approximation was given by Coja-Oghlan et al. [2]. By Φ_n define a random formula *n* variables and $m = \lfloor \frac{dn}{2} \rfloor$ and by *Z* the number of the solutions of such formula. Then the aforementioned clauses and add them to corresponding formulas. It is easy to see that $(\Phi_1^{(0)}, \Phi_2^{(0)})$ are two independent random formulas, and $(\Phi_1^{(M)}, \Phi_2^{(M)})$ is a pair of copies of a random 2-CNF. Therefore,

$$\operatorname{Var} \log Z(\mathbf{\Phi}_{\mathbf{n}}) = \mathsf{E} \left[\log Z(\mathbf{\Phi}_{1}^{(m)}) \log Z(\mathbf{\Phi}_{2}^{(m)}) \right] - \mathsf{E} \left[\log Z(\mathbf{\Phi}_{1}^{(0)}) \log Z(\mathbf{\Phi}_{2}^{(0)}) \right]$$
(3)
$$= \sum_{M=0}^{m-1} \mathsf{E} \left[\log Z(\mathbf{\Phi}_{1}^{(M+1)}) \log Z(\mathbf{\Phi}_{2}^{(M+1)}) \right] - \mathsf{E} \left[\log Z(\mathbf{\Phi}_{1}^{(M)}) \log Z(\mathbf{\Phi}_{2}^{(M)}) \right].$$
(4)

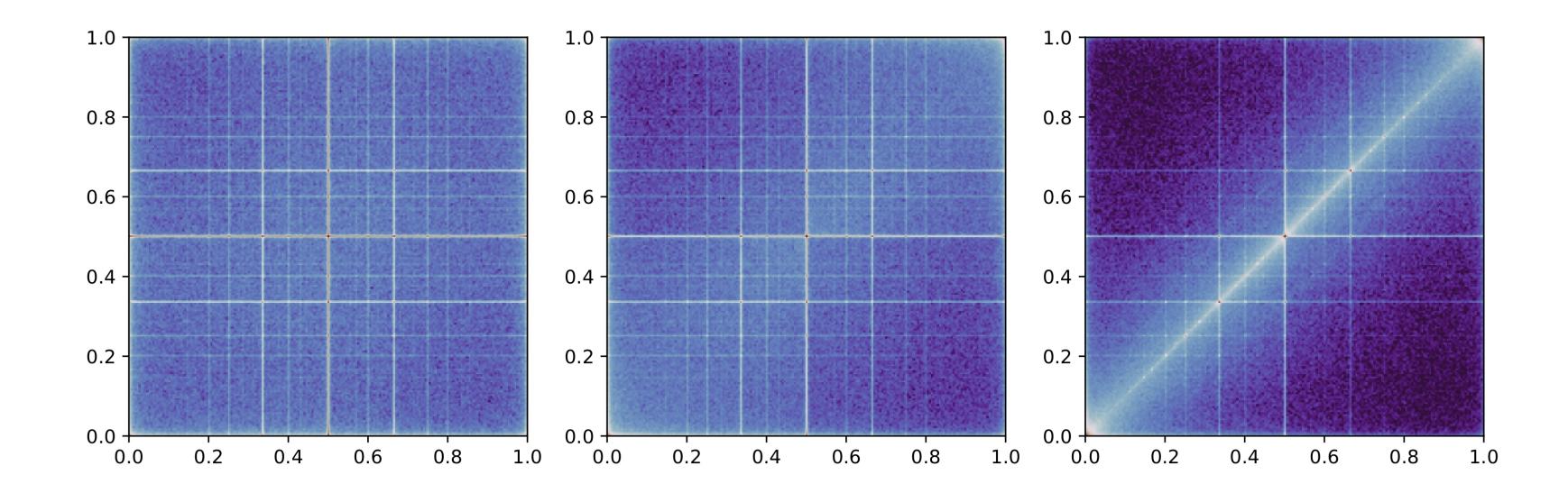




REFERENCES

[1] Andreas Goerdt. A threshold for unsatisfiabil-

Figure 1: Plots of μ_d and η_d as well as their first/second moment estimates;



- ity. Journal of Computer and System Sciences, 53(3):469–486, 1996.
- [2] D. Achlioptas, A. Coja-Oghlan, M. Hahn-Klimroth, J. Lee, N. Müller, M. Penschuck, and G. Zhou. The number of satisfying assignments of random 2-sat formulas. *Random Structures & Algorithms*, 58(4):609– 647, 2021.
- [3] A. Chatterjee, A. Coja-Oghlan, N. Müller, C. Riddlesden, M. Rolvien, P. Zakharov, and H. Zhu. The number of random 2-sat solutions is asymptotically log-normal, 2024.

Figure 2: Marginal distribution on two correlated formulas for d = 1.9 and M = 0.1m, 0.5m, 0.9m

To understand the second figure we should generate a pair of random formulas and take a uniformly random pair of satisfying assignments. The joint distribution of any of n coordinates can be viewed on the heatmaps: almost independet forlumas on the left and highly correlated formulas on the right.