Prospectus

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1 Introduction

I've been working on several problem involving networks, graphs, random matrices, inspired by the Van Vu's and Dan Spielman's courses I was taking during my second year.

With Harry I started working on the problem of spam detection based on Kyng et al. (2015) and Zhou et al. (2007). At the same tine I was exploring general network theory (Davis-Kahan, random matrix theory, spectral theory).

In particular reading Vu (2007) on behavior of spectral norm of a random matrix.

I made some progress with spam but found myself more interested in the underlying theory. Started working on spectral theory more intensely.

As described in the section 2 we made a lot of progress, read a lot of literature, eventually discovered that state of art Soshnikov. Unfortunately Ssoshnikov's papers rely heavily on the first paper by Sinai and Soshnikov (1998a) in the sequence. We found what appear to be a serious error. Reluctantly, decided to put it on hold.

Still wanted to work on networks. Yihong suggested a problem in RDS, a technique that appears to be much used in the social science literature despite the lack of theory. Since February I have been exploring with simulations which appear promising. I want to concentrate on this.

2 Vu

Vu (2005) studied random $n \times n$ symmetric matrix A with independent bounded entries ξ_{ij} with bounded variance:

$$\mathbb{E}\xi_{ij} = 0,\ var\ \xi_{i,j} = \sigma^2,\ |\xi_{ij}| < K$$

Vu uses Wigner method that involves calculating the trace of powers of the matrix:

$$\lambda(A)^k = \lambda(A^k) \leq \sum_{i=1}^{n} \lambda_i(A^k) = trace(A^k) = \sum_{i_0=1}^{n} \sum_{i_1=1}^{n} ... \sum_{i_{k-1}=1}^{n} \xi_{i_0 i_1} \xi_{i_1 i_2} ... \xi_{i_{k-1} i_0}, \quad (1)$$

where k is even.

Need to calculate number of summands which is a combinatorics problem. To handle it Vu used a counting scheme adapted from Furedi and Komlos. We had trouble understanding the method, so split the task: David looked at the original method used by Furedi and Komlos, while I was going through Vu's explantion of his scheme. We figured it out. (`https://elidechse.github.io/VanVu.pdf`My work on Van Vu)

Vu cited (Vu, 2007, remark 1.5) one of the Sinai and Soshnikov papers (Soshnikov (1999)) noting that theirs was the best result of that time for the case of symmetrically distributed entries. I did a literature search and found a more recent paper that improved the Vu's bound. In fact Soshnikov has written lots of papers on the topic:

- In Sinai and Soshnikov (1998a) and Sinai and Soshnikov (1998b) they prve CLT for Wigner (symmetric with symetrically distributed entries?) matrices and developed the technique to count the number of summands

- In Soshnikov (1999) they proved the behaviour of the the k largest eigenvalues of a Wigner matrix
- In Péché and Soshnikov (2007) the key result that helped achieve the bound was proved: they showed that the trace expectation of a power of a matrix is close to the trace expectation of the standard Wigner matrix.

Additionally:

- In Soshnikov et al. (2004) they studies the behaviour of the largest eigenvalue of a matrix with entries distribution having heavt tails
- In Soshnikov and Fyodorov (2005) they studied a rectangular matrix with Cauchy entries and showed that its largest eigenvalues distribution does not follow Tracy-Widom law.
- In **?**, they proved a lower bound for a random symmetric matrix with iid entries.
- etc.

We decided we needed to go to the original paper (Sinai and Soshnikov (1998a)) and the accompanied one (Sinai and Soshnikov (1998b)). We had a trouble with one step of the argument (Sinai and Soshnikov, 1998a, (2.3)), then I eventually came up with a counterexmaple (`https://elidechse.github.io/Soshnikov.pdf`My work on Soshnikov) to this assertion.

3 RDS

In some social and health studies it is difficult to reach the population of interest because of the nature of the study: if it is related to such aspects as drugs, sex, sexual orientation, HIV that are either very personal or stigmatized in the society then usual methods of sampling do not work any more. What researchers use instead is a method called Respondent Driven Sampling (RDS): they need to reach only to several people of interest and then use them to recruit new participants for the study among their acquaintances.

In practice (cite something) everything starts with several initial participants called *seeds*. The they recruit someone to participate through a system of coupons and are paid for each recruited friend, then these new participans are being surveyed and given coupons and in turn recruit someonle else and so on until we reach the sample size. With human beings tend to cheat we cannot alow sampling with replacement: as all recruitments a re paid for then it's too tempting to agree with someone to get recruited as much as possible, which would make the sample far from random walk.

The method was introduced in (Heckathorn (1997)) and now the paper has been cited almost 3000 times. The estimator proposed in Volz and Heckathorn (2008) is used in studies about HIV (Szwarcwald et al. (2011), Johnston et al. (2010)), racial studies (Wejnert (2010)), injection drug users (Wejnert et al. (2012)), etc

Two main assumptions: sampling with replacment, uniform random recruitment among the friends. The first assumption is always violated by the above note, but claim that with small sample sizes sampling with replacement is the same as sampling without replacment. Pretend that the second one is met.

Unfortunately there is a huge gap between proved theoretical results and results that are hoped for and believed in practice. In practice people believe that seeds are drawn according to some stationary distribution and that even with a small size compared to the population size sample the estimator is close to the truth. (cite!!!)

On the other hand most theoretical results imply a recruitment scheme when each participant recruits only one of his friends and thus the whole recruitment sample is considered to be an irreducible Markov chain, that starts from a stationary distribution.

Guntuboyina et al. (2012) gave an example when assuming differential sampling node degree distributions were completely different and as the estimator heavely rely on the degree of nodes it was shown to be bad.

Rohe (2015) was actually considering a recruitment scheme when each participant recruits m of his friends and showed ...

Goel and Salganik (2010) showed bad behavior of variance of the VH estimator (??)

With simulations I compared the behavior of the VH estimator in ideal circumstances on different graphs: it's always close to the truth and is never worse than other estimators I tried. So the question is: can get something with similar good behavior but with working assumptions?

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