## 1 The moment method

Last time we saw that by finding an upper bound for  $\text{Etr}(M^4)$ , we would obtain  $||M|| \leq n^{3/4}$  with high probability. Let's see what will happen if we consider  $\text{Etr}(M^6)$ .

Again,

$$\mathbf{E}\mathrm{tr}(M^{6}) = \sum_{1 \le i_{1}, \dots, i_{6} \le n} \mathbf{E}\xi_{i_{1}, i_{2}} \cdots \xi_{i_{5}, i_{6}}\xi_{i_{6}, i_{1}},$$

a sum over cycles of length 6 in  $\{1, 2, ..., n\}$ . As before, we just need to consider those cycles in which each edge occurs at least twice. We can divide the cycles into 4 classes:

- 1. There are 3 distinct edges, each occurring twice. This type of cycle give a contribution 1, and such cycles have at most 4 vertices, which give a total contribution  $O(n^4)$ .
- 2. There are 2 distinct edges, one of them occurring 4 times and the other occurs 2 times. The number of vertices involved is at most 3, so the total contribution is  $O(K^2n^3)$ .
- 3. There are 2 distinct edges and each of them occurs three times. This contributes  $O(K^2n^3)$  again.
- 4. There is only 1 edge occurs 6 times. The contribution is  $O(K^4n^2)$ .

This implies

$$\operatorname{Etr}(M^6) \le O(n^4) + O(K^2 n^3) + O(K^4 n^2).$$

If  $K = O(\sqrt{n})$ , this gives  $\mathbf{E}tr(M^6) \leq O(n^4)$ . Recall that

$$||M|| \le \operatorname{tr}(M^6)^{1/6}.$$

Using similar argument as before we see that  $||M|| \leq O(n^{2/3})$  with high probability.

## 1.1 *k*-th moment computation

Let's consider  $\mathbf{E}$ tr $(M^k)$ , where k is an even integer. Again,

$$\mathbf{E}\mathrm{tr}(M^k) = \sum_{1 \le i_1, \dots, i_k \le n} \mathbf{E}\xi_{i_1, i_2} \cdots \xi_{i_k, i_1},$$

where the sum if taken over all cycles of length k in  $\{1, 2, ..., n\}$ . The only non-vanishing expectation are those for which each edge occurs at least twice. In particular, there are at most k/2 distinct edges, and thus at most k/2 + 1 vertices.

We would like to divide the cycles into various classes. For instance, let's again consider k = 4. In this case, we have terms like

$$\mathbf{E}\xi_{i_1,i_2}\xi_{i_2,i_3}\xi_{i_3,i_4}\xi_{i_4,i_1}.$$

Recall that we have several classes, one of which is  $i_1 = i_3$ ,  $i_2 = i_4$ , but  $i_1 \neq i_2$ . In this case, all the edges  $\{i_1, i_2\}$ ,  $\{i_2, i_3\}$ ,  $\{i_3, i_4\}$  and  $\{i_4, i_1\}$  are the same. We will write  $1 \sim 2 \sim 3 \sim 4$ , because all the four terms are the same.

Let's do one more example, again with k = 4. Consider the case  $i_1 = i_3$ , but  $i_2, i_4$  are distinct from each other and from  $i_1$ . In this case, the edge  $\{i_1, i_2\}$  will be the same as  $\{i_2, i_3\}$ , and the edge  $\{i_3, i_4\}$  will be the same as  $\{i_4, i_1\}$ . We can write  $1 \sim 2$  and  $3 \sim 4$  in this case, because the first and the second terms are the same, and the third and the fourth terms are the same. However,  $2 \not\sim 3$ .

In general, we divide the cycles into various classes, depending on which edges are equal to each other. More precisely, a class is an equivalence relation  $\sim$  on a set of k labels, say  $\{1, \ldots, k\}$ , where each class has at least two elements. We associate a cycle of k edges  $\{i_1, i_2\}, \ldots, \{i_k, i_1\}$  to a class when we have  $\{i_j, i_{j+1}\} = \{i_{j'}, i_{j'+1}\}$  if and only if  $j \sim j'$ . For instance, in the first example we saw, the case  $i_1 = i_3$ ,  $i_2 = i_4$ , but  $i_1 \neq i_2$  is associated to the class  $1 \sim 2 \sim 3 \sim 4$ .

How many classes are there? Well, we need to assign up to k/2 labels to k edges, so a crude upper bound would be  $(k/2)^k$ .

Now consider a given class of cycle. it has j edges (and hence at most j+1 many vertices), where  $1 \leq j \leq k/2$ , with multiplicity  $a_1, \ldots, a_j$ . Note that  $a_i \geq 2$  (because each edge appears at least twice so that the expectation will not vanish) and  $\sum_{i=1}^{j} a_i = k$ . Therefore,

$$\mathbf{E}\xi_{i_1,i_2}\cdots\xi_{i_k,i_1} \le K^{a_1-2}K^{a_2-2}\cdots K^{a_j-2} = K^{k-2j},$$

and thus

$$\begin{aligned} \mathbf{E} \mathrm{tr}(M^k) &\leq (k/2)^k K^{k-2j} O(n^{j+1}) \\ &\leq (k/2)^k \max\{n^{k/2+1}, n^2 K^{k-2}\} \\ &= (k/2)^k n^{k/2+1} \max\{1, (K/\sqrt{n})^{k-2}\}. \end{aligned}$$

By  $||M||^k \leq \operatorname{tr}(M^k)$  again, we have

$$\mathbf{E} \|M\|^{k} \le (k/2)^{k} n^{k/2+1} \max\{1, (K/\sqrt{n})^{k-2}\}.$$

So by Markov's inequality, we have

$$\mathbf{P}(\|M\| \ge \lambda) \le \lambda^{-k} (k/2)^k n^{k/2+1} \max\{1, (K/\sqrt{n})^{k-2}\}$$

for all  $\lambda > 0$ . We will continue next time.

## References

- [1] Ghosh, Koushik. Talk on 2/9.
- Tao, Terence. Topics in random matrix theory. Graduate Studies in Mathematics, 132. American Mathematical Society, Providence, RI, 2012. x+282 pp. ISBN: 978-0-8218-7430-1