Hypergraphon estimation error 2

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Notation and problem setting 1

Let E be a set of possible m-uniform hyperedges from n vertices without diagonal entries,

$$E = \{(i_1, \dots, i_m) \in [n]^m \colon |\{i_1, \dots, i_m\}| = m\}.$$

We denote an index of m-uniform hyperedges as $\omega = (\omega_1, \dots, \omega_m) \in [n]^m$ and a membership vector of m-vertices as $a = (a_1, \ldots, a_m) \in [k]^m$. Let $z : [n] \to [k]$ be a membership function. For a given membership function z and a membership vector $a \in [k]^m$, define $E_{Z^{-1}(a)}$ as a set of m-uniform hyperedges whose clustering group belongs to a i.e.,

$$E_{z^{-1}(a)} = \{ \omega \in E \colon z(\omega_{\ell}) = a_{\ell} \text{ for all } \ell \in [m] \}.$$

We define a block average on a set $E_{z^{-1}(a)}$ for a given membership function z, a membership vector a, and a tensor $\Theta \in ([n])^{\otimes m}$ as

$$\bar{\Theta}_a(z) = \frac{1}{|E_{z^{-1}(a)}|} \sum_{\omega \in E_{z^{-1}(a)}} \Theta_{\omega}.$$

Now we consider an undirected m-uniform hypergraph. The connectivity is encoded by an adjacency tensor $\{\mathcal{A}_{\omega}\}_{\omega\in E}$ which takes values in $\{0,1\}$. We assume that $\mathcal{A}_{\omega}\sim \text{Bernoulli}(\Theta_{\omega})$, where

$$\Theta_{\omega} = f(\xi_{\omega_1}, \dots, \xi_{\omega_m}), \text{ for all } \omega = (\omega_1, \dots, \omega_m) \in E,$$
 (0)

where $f:[0,1]^m \to [0,1]$ is a symmetric function called graphon such that $f(\xi_{\omega_1},\ldots,\xi_{\omega_m})=f(\xi_{\sigma(\omega_1)},\ldots,\xi_{\sigma(\omega_m)})$ for all permutation $\sigma: [n] \to [n]$. Conventionally, we set $\Theta_{\omega} = 0$ for all $\omega \in [n]^m \setminus E$. In addition, we further assume that a graphon f is α -Hölder smooth with a constant L.

Definition 1. A function $f: [0,1]^m \to [0,1]$ is a α -Hölder smooth with a constant L, denoted as $f \in \mathcal{H}(\alpha,L)$, if there exists a polynomial function $\mathcal{P}_{\lfloor \alpha \rfloor}(x)$ of degree $\lfloor \alpha \rfloor$ such that \mathbb{Q} : is the class of alpha-holder smooth

$$|f(x) - \mathcal{P}_{\lfloor \alpha \rfloor}(x'-x)| \le L ||x'-x||_{\alpha-\lfloor \alpha \rfloor}^{\alpha-\lfloor \alpha \rfloor}$$
, agger than the class of beta-smooth (1)?

where the norm $||x||_p^p := \sum_{i=1}^m |x_i|^p$ for $x \in \mathbb{R}^m$.

It follows from the standard embedding theorem for any $f \in \mathcal{H}(\alpha, L)$, Can we define the function class of interest, $\mathsf{H}(\mathsf{alpha}, \mathsf{L})$, using property(1) with $\mathsf{alpha} \mathsf{ln}[0,1]$? $|f(x) - f(y)| \leq M \|x - y\|_{\alpha \wedge 1}^{\alpha \wedge 1}$, (1)

where M > 0 is a global constant only depending on α and L. We only use the property (1) over the note for α -Hölder smoothness. How does M depend on L?

2 Probability matrix estimation For every integer k <=n, there exists Lemma 1. There exists z^* : $[n] \rightarrow [k]$, satisfying

Proof. Define $z^* : [n] \to [k]$ by

$$(z^*)^{-1}(\ell) = \left\{ i \in [n] : \xi_i \in \left[\frac{\ell - 1}{k}, \frac{\ell}{k} \right] \right\}, \quad \text{for each } \ell \in [k].$$

By the construction of z^* for $\xi_{\omega_{\ell}} \in [(a_{\ell-1}-1)/k, a_{\ell}/k]$,

$$|f(\xi_{\omega_{1}}, \dots, \xi_{\omega_{m}}) - \bar{\Theta}_{a}(z^{*})| = \left| f(\xi_{\omega_{1}}, \dots, \xi_{\omega_{m}}) - \frac{1}{|E_{z^{-1}(a)}|} \sum_{(\omega'_{1}, \dots, \omega'_{m}) \in E_{z^{-1}(a)}} f(\xi_{\omega'_{1}}, \dots, \xi_{\omega'_{m}}) \right|$$

$$\leq \frac{1}{|E_{z^{-1}(a)}|} \sum_{(\omega'_{1}, \dots, \omega'_{m}) \in E_{z^{-1}(a)}} |f(\xi_{\omega_{1}}, \dots, \xi_{\omega_{m}}) - f(\xi_{\omega'_{1}}, \dots, \xi_{\omega'_{m}})|$$

$$\leq \frac{1}{|E_{z^{-1}(a)}|} \sum_{(\omega_{1}, \dots, \omega_{m}) \in E_{z^{-1}(a)}} M \|(\xi_{\omega_{1}}, \dots, \xi_{\omega_{m}}) - (\xi_{\omega'_{1}}, \dots, \xi_{\omega'_{m}})\|_{\alpha \wedge 1}^{\alpha \wedge 1}$$

$$\leq CM \left(\frac{m}{k}\right)^{\alpha \wedge 1}.$$

Let $\tilde{\Theta}$ be a minimizer of the least square error from the adjacency tensor \mathcal{A} ,

$$\tilde{\Theta} = \underset{\Theta \in \mathcal{P}_k}{\operatorname{arg\,min}} \sum_{\omega \in E} (\mathcal{A}_{\omega} - \Theta_{\omega})^2,$$

where

$$\mathcal{P}_k = \{ \Theta \in ([0,1]^n)^{\otimes m} \colon \Theta = \mathcal{C} \times_2 \mathbf{M} \times_2 \cdots \times_m \mathbf{M}, \text{ with a}$$
membership matrix \mathbf{M} and a core tensor $\mathcal{C} \in ([0,1]^k)^{\otimes m} \}.$

We estimate the probability tensor by $\hat{\Theta} = \operatorname{cut}(\tilde{\Theta})$ such that

Suppose the true parameter, Theta admits the form (0) with f \in H(alpha, L)
$$\operatorname{cut}(\Theta_{\omega}) = \begin{cases} 0 & \text{if } \omega \in [n]^m \setminus E. \end{cases}$$

Notice $\|\mathcal{A} - \hat{\Theta}\|_F^2 \le \|\mathcal{A} - \Theta\|_F^2$ for any k block tensor $\Theta \in \mathrm{cut}(\mathcal{P}_k)$ Discuss which of the two terms dominates.

Theorem 2.1 (hypergraphon model). Let $\hat{\Theta}$ be the estimator from (will see depending on (alpha, $C_1, C_2 > 0$ such that,

Agree with intuition?

$$\frac{1}{n^m}\|\hat{\Theta}-\Theta\|_F^2 \leq C_1\left(n^{\frac{-2m(\alpha\wedge 1)}{m+2(\alpha\wedge 1)}} + \frac{\log n}{n^{m-1}}\right), \quad \ \ \frac{??}{\text{alpha}} > 1 \text{: second term}$$

with probability at least $1 - \exp\left(-C_2\left(n\log n + n^{\frac{m^2}{m+2(\alpha\wedge 1)}}\right)\right)$ uniformly over $f \in \mathcal{H}(\alpha, L)$.

alpha<=1:

Proof. First, we can find a block tensor Θ^* close to true Θ by Lemma 1. By triangular integral integral in the proof.

$$\|\hat{\Theta} - \Theta\|_F^2 \leq \underbrace{\|\hat{\Theta} - \Theta^*\|_F^2}_{(i)} + \underbrace{\|\Theta^* - \Theta\|_F^2}_{(ii)}.$$
 m=2: both term appear? m>=3: first term dominates.

Since we have already shown the error bound of (ii) in Lemma 1, we bound the error from (i). From the definition of $\hat{\Theta}$, we have

$$\|\hat{\Theta} - \mathcal{A}\|_F^2 \le \|\Theta^* - \mathcal{A}\|_F^2. \tag{4}$$

Combining (4) with the fact

$$\begin{split} \|\hat{\Theta} - \mathcal{A}\|_F^2 &= \|\hat{\Theta} - \Theta^* + \Theta^* - \mathcal{A}\|_F^2 \\ &= \|\hat{\Theta} - \Theta^*\|_F^2 + \|\Theta^* - \mathcal{A}\|_F + 2\langle \hat{\Theta} - \Theta^*, \Theta^* - \mathcal{A} \rangle, \end{split}$$

yields

$$\begin{split} \|\hat{\Theta} - \Theta^*\|_F^2 &\leq 2\langle \hat{\Theta} - \Theta^*, \mathcal{A} - \Theta^* \rangle \\ &= 2\left(\langle \hat{\Theta} - \Theta^*, \mathcal{A} - \Theta \rangle + \langle \hat{\Theta} - \Theta^*, \Theta - \Theta^* \rangle\right) \\ &\leq 2\|\hat{\Theta} - \Theta^*\|_F \left(\left\langle \frac{\hat{\Theta} - \Theta^*}{\|\hat{\Theta} - \Theta^*\|_F}, \mathcal{A} - \Theta \right\rangle + \|\Theta - \Theta^*\|_F\right). \end{split}$$

Let $\mathcal{M} = \{M : M \text{ is the collection of membership matrices}\}$. Then,

$$\left\langle \frac{\hat{\Theta} - \Theta^*}{\|\hat{\Theta} - \Theta^*\|_F}, \mathcal{A} - \Theta \right\rangle \leq \sup_{\Theta' \in \mathcal{P}_k} \sup_{\Theta'' \in \mathcal{P}_k} \left\langle \frac{\operatorname{cut}(\Theta') - \operatorname{cut}(\Theta'')}{\|\operatorname{cut}(\Theta') - \operatorname{cut}(\Theta'')\|_F}, \mathcal{A} - \Theta \right\rangle \\
\leq \sup_{\mathbf{M}, \mathbf{M}' \in \mathcal{M}} \sup_{\mathcal{C}, \mathcal{C}' \in ([0,1]^n) \otimes^m} \left\langle \frac{\operatorname{cut}(\Theta(\mathbf{M}, \mathcal{C})) - \operatorname{cut}(\Theta(\mathbf{M}', \mathcal{C}'))}{\|\operatorname{cut}(\Theta(\mathbf{M}, \mathcal{C})) - \operatorname{cut}(\Theta(\mathbf{M}', \mathcal{C}'))\|_F}, \mathcal{A} - \Theta \right\rangle.$$

Notice that $A - \Theta$ is sub-Gaussian with proxy parameter $\sigma^2 = 1/4$. By union bound and the property of sub-Gaussian, we have, for any t > 0.

$$\mathbb{P}\left(\sup_{\boldsymbol{M},\boldsymbol{M}'\in\mathcal{M}}\sup_{\mathcal{C},\mathcal{C}'\in([0,1]^n)\otimes^m}\left|\left\langle\frac{\operatorname{cut}(\Theta(\boldsymbol{M},\mathcal{C}))-\operatorname{cut}(\Theta(\boldsymbol{M}',\mathcal{C}'))}{\|\operatorname{cut}(\Theta(\boldsymbol{M},\mathcal{C}))-\operatorname{cut}(\Theta(\boldsymbol{M}',\mathcal{C}'))\|_F},\mathcal{A}-\Theta\right\rangle\right|\geq t\right) \\
\leq \sum_{\boldsymbol{M},\boldsymbol{M}'\in\mathcal{M}}\mathbb{P}\left(\sup_{\mathcal{C},\mathcal{C}'\in([0,1]^n)\otimes^m}\left|\left\langle\frac{\operatorname{cut}(\Theta(\boldsymbol{M},\mathcal{C}))-\operatorname{cut}(\Theta(\boldsymbol{M}',\mathcal{C}'))}{\|\operatorname{cut}(\Theta(\boldsymbol{M},\mathcal{C}))-\operatorname{cut}(\Theta(\boldsymbol{M}',\mathcal{C}'))\|_F},\mathcal{A}-\Theta\right\rangle\right|\left(\frac{m}{k}\right)^{\alpha\Delta 1}\geq t\right) \\
\leq |\mathcal{M}|^2C_1^{k^m}\exp\left(-C_2t^2\right) \\
= \exp\left(2n\log k + C_1k^m - C_2t^2\right),$$

where $C_1, C_2 > 0$ are universal constants. The secibd line follows from Phillippe Rigollet [2015] and the fact that $\Theta = \Theta(\mathbf{M}, \cdot)$ lies in a linear space of dimension k^m . Choosing $t = C\sqrt{n \log k + k^m}$ yields

(ii)
$$\leq C_1 \left(\left(\frac{m}{k} \right)^{2(\alpha \wedge 1)} + \left(\frac{k}{n} \right)^m + \frac{\log k}{n^{m-1}} \right),$$

with probability at least $1 - \exp\left(-C_2(n\log k + k^m)\right)$. Combinations of two error bounds in (3) and setting $k = \left\lceil n^{\frac{m}{m+2(\alpha \wedge 1)}} \right\rceil$, completes the theorem.

Theorem 2.2 (stochastic block model). Let $\hat{\Theta}$ be the estimator from (2). Suppose true probability tensor $\Theta \in \text{cut}(\mathcal{P}_k)$ for fixed block size k. Then, there exists two constants $C_1, C_2 > 0$, such that

$$\frac{1}{n^m}\|\hat{\Theta} - \Theta\|_F^2 \leq C_1 \left(\frac{k}{n}\right)^m + \frac{\max\limits_{\log k} \log k}{n^{m-1}},$$

with probability at Second temp $(-C_2(n \log k + k^m))$. In particular, suppose $k \approx n^{\delta}$ for some $\delta \in [0,1]$.

fix a mode: assign n entries to k groups -> how many possible configurations? # $k^n -> error \sim log (k^n) \sim n log k$ in total m modes: -> m n log k

First term: m modes in total, each mode has k groups. how many blocks -> $k^m -$

Then, the convergence rate becomes

Perform the similar discussion for

$$\left(\frac{k}{n}\right)^m + \frac{\log k}{n^{m-1}} \asymp \begin{cases} n^{-m} & k = 1, \\ n^{-m+1} & \delta = 0, k \ge 2, \\ n^{-m+1} \log(n) & \delta \in (0, 1/m], \\ n^{-m(1-\delta)} & \delta \in (1/m, 1]. \end{cases}$$

Proof. By similar way in the proof of Theorem 2.1, we have

$$\begin{split} \|\hat{\Theta} - \Theta\|_F^2 &\leq 2\langle \hat{\Theta} - \Theta, \mathcal{A} - \Theta \rangle \\ &= 2\|\hat{\Theta} - \Theta\|_F \left\langle \frac{\hat{\Theta} - \Theta}{\|\hat{\Theta} - \Theta\|_F}, \mathcal{A} - \Theta \right\rangle \\ &\leq \sup_{\Theta' \in \mathcal{P}_b} \sup_{\Theta'' \in \mathcal{P}_b} \left\langle \frac{\operatorname{cut}(\Theta') - \operatorname{cut}(\Theta'')}{\|\operatorname{cut}(\Theta') - \operatorname{cut}(\Theta'')\|_F}, \mathcal{A} - \Theta \right\rangle. \end{split}$$

Notice the last inequality holds because $\Theta \in \text{cut}(\mathcal{P}_k)$. Therefore, we have the result following the proof of Theorem 2.1.

3 Hypergraphon estimation

For a given probability tensor Θ , define the empirical hypergraphon $f_{\Theta} \colon [0,1]^m \to [0,1]$ as the following piecewise constant function:

$$\tilde{f}_{\Theta}(x_1,\ldots,x_m) = \Theta_{\lfloor x_1\rfloor,\ldots,\lfloor x_m\rfloor}.$$

For any hypergraphon estimator \hat{f} , we define the squared error

[Adaptivity]

What if the underling function f is not holder smooth?

E.g. true graphon is k-piecewise constant

Conjecture: ~ m k \over sqrt(n)

$$\delta^{2}(\hat{f}, f) := \inf_{\tau \in \mathcal{T}} \int_{(0,1)^{m}} |f(\tau(x)) - \hat{f}(x)|^{2} dx,$$

where \mathcal{T} is the set of all measure-preserving bijection $\tau \colon [0,1] \to [0,1]$.

Our goal is to construct the upper bound of error $\mathbb{E}\left[\delta^2(f_{\hat{\Theta}},f)\right]$. By triangular inequality, we have

$$\mathbb{E}\left[\delta^2(f_{\hat{\Theta}},f)\right] \leq \underbrace{\frac{2}{n^m}\mathbb{E}\|\hat{\Theta} - \Theta\|_F^2}_{(\mathrm{i})} + 2\underbrace{\mathbb{E}\left[\delta^2(f_{\Theta},f)\right]}_{(\mathrm{ii})}.$$

(i) directly follows from Thm 2.1

Currently, I only derived the upper bound of (ii).

Lemma 2. Suppose $f \in \mathcal{H}(\alpha, L)$. Then

Name this Lemma. (Agnostic error)
$$\mathbb{E}\left[\delta^2(f_\Theta,f)\right] \leq \frac{C}{n^{\alpha\wedge 1}},$$

for some constant C only depending on constants m and L.

Proof. By triangular inequality, we have

Make the dependence on m explicit -> Our goal is to investigate the impact of higher-order (m) compared to m=2.

$$\mathbb{E}\left[\delta^{2}(f_{\Theta}, f)\right] \leq 2\mathbb{E}\left[\delta^{2}(f_{\Theta}, f_{\Theta'})\right] + 2\mathbb{E}\left[\delta^{2}(f_{\Theta'}, f)\right],$$

where $\Theta' \in ([0,1]^n)^{\otimes m}$ such that $\Theta'_{\omega} = f(\xi_{\omega_1}, \dots, \xi_{\omega_m})$ for all $\omega \in [n]^m$. Notice $\Theta'_{\omega} = \Theta_{\omega}$ for $\omega \in E$ but

 $\Theta_{\omega} = 0$ for $\omega \in [n]^m \setminus E$. By definition of Θ' ,

$$\mathbb{E}\left[\delta^2(f_{\Theta}, f_{\Theta'})\right] = \int_{[0,1]^m} |f_{\Theta}(x) - f_{\Theta'}(x)| dx < \frac{C}{n},$$

for some C>0 only depending on m. This is because $|f_{\Theta}(x)-f_{\Theta'}(x)|=0$ outside of a set of measure $\left(n^m-\binom{n}{m}n!\right)/n^m=C/n$. Hence it suffices to prove that

$$\mathbb{E}\left[\delta^2(f_{\Theta'}, f)\right] \le \frac{C}{n^{\alpha \wedge 1}}.$$

We have

$$\delta^{2}(f_{\Theta'}, f) = \inf_{\tau \in \mathcal{T}} \sum_{i_{1}, \dots, i_{m} = 1}^{n} \int_{(i_{1} - 1)/n}^{i_{1}/n} \cdots \int_{(i_{m} - 1)/n}^{i_{m}/n} |f(\tau(x_{1}), \dots, \tau(x_{m})) - \Theta'_{i_{1}, \dots, i_{m}}|^{2} dx_{1} \cdots dx_{m}$$

The infimum over all measure-preserving bijection is smaller than the minimum over the subclass of measure-preserving bijection τ such that

$$\int_{(i_1-1)/n}^{i_1/n} \cdots \int_{(i_m-1)/n}^{i_m/n} f(\tau(x_1), \dots, \tau(x_m)) dx_1 \cdots dx_m = \int_{(\sigma(i_1)-1)/n}^{\sigma(i_1)/n} \cdots \int_{(\sigma(i_m)-1)/n}^{\sigma(i_m)/n} f(x_1, \dots, x_m) dx_1 \cdots dx_m$$

for some permutation σ . For $x \in \prod_{\ell=1}^m [(\sigma(i_\ell) - 1)/n, \sigma(i_\ell)/n]$,

$$|f(x_{1},...,x_{m}) - f(\xi_{1},...,\xi_{m})|^{2} \leq 2 \left| f(x_{1},...,x_{m}) - f\left(\frac{\sigma(i_{1})}{n+1},...,\frac{\sigma(i_{m})}{n+1}\right) \right|^{2}$$

$$+ 2 \left| f\left(\frac{\sigma(i_{1})}{n+1},...,\frac{\sigma(i_{m})}{n+1}\right) - f(\xi_{(\sigma(i_{1}))},...,\xi_{(\sigma(i_{m}))}) \right|^{2}$$

$$+ 2 \left| f(\xi_{(\sigma(i_{1}))},...,\xi_{(\sigma(i_{m}))}) - f(\xi_{i_{1}},...,\xi_{i_{m}}) \right|^{2},$$

$$(5)$$

where $\xi_{(\ell)}$ denotes the ℓ -th largest element of the set $\{\xi_1, \ldots, \xi_n\}$. Choose random permutation σ such that $\xi_{\sigma^{-1}(1)} \leq \xi_{\sigma^{-1}(2)} \cdots \leq \xi_{\sigma^{-1}(n)}$. Then the third summand in (5) is 0 almost surely.

For the first summand in (5), notice $(\sigma(i_1)/(n+1), \ldots, \sigma(i_m)/(n+1)) \in \prod_{\ell=1}^m [(\sigma(i_\ell)-1)/n, \sigma(i_\ell)/n]$. From (1), we obtain

$$\left| f(x_1, \dots, x_m) - f\left(\frac{\sigma(i_1)}{n+1}, \dots, \frac{\sigma(i_m)}{n+1}\right) \right|^2 \le m^2 M^2 \left(\frac{1}{n}\right)^{2(\alpha \wedge 1)}.$$

Integrating and taking expectation on the first summand yields,

$$\mathbb{E}\left[\sum_{i_1,\dots,i_m=1}^n \int_{(\sigma(i_1)-1)/n}^{\sigma(i_1)/n} \cdots \int_{(\sigma(i_m)-1)/n}^{\sigma(i_m)/n} \left| f(x_1,\dots,x_m) - f\left(\frac{\sigma(i_1)}{n+1},\dots\frac{\sigma(i_m)}{n+1}\right) \right|^2 dx_1 \cdots dx_m \right] \le m^2 M^2 \left(\frac{1}{n}\right)^{2(\alpha \wedge 1)}.$$

$$\tag{6}$$

With (1), the second summand on (5) is bounded,

$$\left| f\left(\frac{\sigma(i_1)}{n+1}, \dots \frac{\sigma(i_m)}{n+1}\right) - f(\xi_{(\sigma(i_1))}, \dots, \xi_{(\sigma(i_m))}) \right|^2 \le \left(M \sum_{\ell=1}^m \left| \frac{\sigma(i_\ell)}{n+1} - \xi_{(\sigma(i_\ell))} \right|^{\alpha \wedge 1} \right)^2$$

$$\le 2M^2 \sum_{\ell=1}^m \left| \frac{\sigma(i_\ell)}{n+1} - \xi_{(\sigma(i_\ell))} \right|^{2(\alpha \wedge 1)}.$$

Perfect, you are almost halfway done! will discuss with you tomorrow the interpretation (and a natural next question)

Integrating and taking expectation on the second summand yields,

$$\mathbb{E}\left[\sum_{i_{1},\dots,i_{m}=1}^{n} \int_{(\sigma(i_{1})-1)/n}^{\sigma(i_{1})/n} \cdots \int_{(\sigma(i_{m})-1)/n}^{\sigma(i_{m})/n} \left| f\left(\frac{\sigma(i_{1})}{n+1},\dots \frac{\sigma(i_{m})}{n+1}\right) - f(\xi_{(\sigma(i_{1}))},\dots,\xi_{(\sigma(i_{m}))}) \right|^{2} dx_{1} \cdots dx_{m} \right] \\
\leq \left[\sum_{i_{1},\dots,i_{m}=1}^{n} \int_{(\sigma(i_{1})-1)/n}^{\sigma(i_{1})/n} \cdots \int_{(\sigma(i_{m})-1)/n}^{\sigma(i_{m})/n} 2M^{2} \sum_{\ell=1}^{m} \left| \frac{\sigma(i_{\ell})}{n+1} - \xi_{(\sigma(i_{\ell}))} \right|^{2(\alpha\wedge1)} dx_{1} \cdots dx_{m} \right] \\
= 2mM^{2} \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^{n} \left| \frac{i}{n+1} - \xi_{(i)} \right|^{2(\alpha\wedge1)} \right] \\
\leq 2mM^{2} \max_{i=1,\dots,n} \mathbb{E}\left[\left| \frac{i}{n+1} - \xi_{(i)} \right|^{2(\alpha\wedge1)} \right] \\
\leq 2mM^{2} \max_{i=1,\dots,n} \left[\operatorname{Var}(\xi_{(i)}) \right]^{\alpha\wedge1} \leq C\left(\frac{1}{n}\right)^{\alpha\wedge1}, \tag{7}$$

where we have used $\mathbb{E}(\xi_{(\ell)}) = \ell/(n+1)$, $\operatorname{Var}(\xi_{(\ell)}) \leq C/n$ and Jensen's inequality. Combining (6) and (7) proves the lemma.

References

Jan-Christian Hitter Phillippe Rigollet. High dimensional statistics. Lecture notes for course 18S997, 2015.