

Geometry of Numbers Lecture 11

Reminder: $\mathcal{X}_n = \text{SL}_n(\mathbb{R}) / \text{SL}_n(\mathbb{Z})$ - space of lattices in \mathbb{R}^n of covolume 1.

$m_{\mathcal{X}_n}$ - the unique $\text{SL}_n(\mathbb{R})$ -inv. Borel prob. measure on \mathcal{X}_n .

Given F on \mathcal{X}_n , want to know the distribution of its values - is $F(L)$ close to $\mathbb{E}(F) \leftarrow$ average of F w.r.t. $m_{\mathcal{X}_n}$ for most L ?

Specifically for $F = \hat{f}$, $f \in L^1(\mathbb{R}^n)$,
eg. $f = \mathbb{1}_B$ $B \subset \mathbb{R}^n$ a ball.

(Recall: $\hat{f}: \mathcal{X}_n \rightarrow \mathbb{R}$, $\hat{f}(L) = \sum_{v \in L \setminus \{0\}} f(v)$,

$\hat{f} \in L^1(\mathcal{X}_n, m_{\mathcal{X}_n})$ whenever $f \in L^1(\mathbb{R}^n, \text{Vol})$,

and $\int_{\mathcal{X}_n} \hat{f} dm_{\mathcal{X}_n} = \int_{\mathbb{R}^n} f d\text{Vol}$ (SSFD).)

Q1 Can we compute higher moments of \hat{f} (thought of as a random variable on the prob. space $(\mathcal{X}_n, \mathbb{M}_{\mathcal{X}_n})$).

Recall: $\int \hat{f} d\mathbb{m}_{\mathcal{X}_n} = \mathbb{E}(X)$

when $X = \hat{f}$

For $k \geq 1$, the (raw) k-th moment of X

is $\mathbb{E}(X^k) = \int_{\mathcal{X}_n} (\hat{f})^k d\mathbb{m}_{\mathcal{X}_n}$

the centered k-th moment

$$\mathbb{E} \left((X - \mathbb{E}(X))^k \right) = \int_{\mathcal{X}_n} \left(\hat{f} - \int_{\mathbb{R}^n} \hat{f} d\text{Vol} \right)^k d\mathbb{m}_{\mathcal{X}_n}$$

The centered 2nd moment is the variance

$\text{Var}(X) = \sigma^2$ σ is the standard deviation.

Prop (Markov/Chebyshev inequality)

Suppose $k \geq 1$, $X \geq 0$, $a > 1$, and $\mathbb{E}(X^k) < \infty$.

$$\text{Then } \mathbb{P}(X > a \mathbb{E}(X)) \leq \frac{\mathbb{E}(X^k)}{(a \mathbb{E}(X))^k}.$$

If $k=1$, $\mathbb{P}(X > a \mathbb{E}(X)) \leq \frac{1}{a}$.

$$\begin{aligned} \text{PF: } \mathbb{E}(X^k) &\geq \mathbb{P}(X^k > a^k (\mathbb{E}(X))^k) a^k (\mathbb{E}(X))^k \\ &= \mathbb{P}(X > a \mathbb{E}(X)) (a \mathbb{E}(X))^k \end{aligned}$$

Moving terms, get the Prop.

Markov's inequality for $Y = (X - \mathbb{E}(X))^2$, $k=1$,

$\mathbb{E}(Y) = \text{Var}(X)$, for $a > 1$,

$$\mathbb{P}(Y > a \text{Var}(X)) \leq \frac{1}{a}.$$

Prop (Borel-Cantelli Lemma) If μ is a prob. measure, on a space X , A_1, A_2, \dots measurable sets

with $\sum_{n=1}^{\infty} \mu(A_n) < \infty$ then for a.e. $x \in X$,

$\exists n_0 = n_0(x) \forall n \geq n_0 \quad x \notin A_n.$

PC Define $A_{\infty} = \{x \in X : \text{there are inf. many } n \text{ for which } x \in A_n\}$

$$= \bigcap_m \bigcup_{n \geq m} A_n$$

$$\text{Then } \mu(A_{\infty}) = \lim_{m \rightarrow \infty} \mu\left(\bigcup_{n \geq m} A_n\right) \leq \lim_{m \rightarrow \infty} \sum_{n \geq m} \mu(A_n)$$

$$\xrightarrow{m \rightarrow \infty} 0$$

Thm (Rogers '55) If $f \in C_c(\mathbb{R}^n)$,

$k \in \mathbb{N}$, $k < n$, then $\hat{f} \in L^k(\mathcal{X}_n, \mu_{\mathcal{X}_n})$

(i.e. the raw k -th moment of \hat{f} is finite).

Moreover Rogers obtained explicit bounds

for the k -th moment.

Theorem (Eskin-Margulis-Mozes '98) If $p < n$
 $1 \leq p \in \mathbb{R}$, $f \in C_c(\mathbb{R}^n)$, then $\hat{f} \in L^p(\mathcal{X}_n, m_{\mathcal{X}_n})$.

OTOH there are $f \in C_c(\mathbb{R}^n)$ for
 which $\hat{f} \notin L^n(\mathcal{X}_n, m_{\mathcal{X}_n})$.

We will give a sketch of proof of the
 Rogers bound (more detailed for $k=2$).

Let $k < n$, $k_i \in \mathbb{N}$, and write

$$\mathbb{R}^{kn} \cong M_{k \times n}(\mathbb{R}) \cong \left\{ (u_1, \dots, u_k) : u_i \in \mathbb{R}^n \text{ for } i=1, \dots, k \right\}$$

Theorem (Siegel '45) Let $f \in L^1(\mathbb{R}^{kn}, \text{Vol})$

Define $\hat{f}^{n,k} : \mathcal{X}_n \rightarrow \mathbb{R}$

$$\hat{f}^{n,k} = \sum_{\substack{u_1, \dots, u_k \in L^1 \text{ not} \\ \text{lin. ind.}}} f(u_1, \dots, u_k)$$

(what we denoted by \hat{f} is $\hat{f}^{n,1}$).

note: u_1, \dots, u_k
 linearly
 independent

$$\text{Then } \int_{\mathcal{X}_n} \widehat{f} d\mu_{\mathcal{X}_n} = \int_{\mathbb{R}^{kn}} f d\text{Vol} \quad (\text{SSF2})_{n,k}$$

(for $k=1$ we recover (SSF2)).

Sketch of proof of Siegel's theorem:

$$\text{Denote } \mathbb{R}_{\text{l.i.}}^{kn} = \{ (u_1, \dots, u_k) \in \mathbb{R}^{kn} : u_i \text{ lin. ind.} \}$$

$\mathbb{R}_{\text{l.i.}}^{kn}$ is an open dense subset of \mathbb{R}^{kn}

$$\mathbb{R}_{\text{l.i.}}^{kn} = \mathbb{R}^{kn} \setminus \bigcap_{i_1 < \dots < i_k} \{ \det \text{ of } k \times k \text{ minor } u_{i_1}, \dots, u_{i_k} \text{ is } 0 \}$$

So enough to prove $(\text{SSF2})_{n,k}$ for

$$f \in C_c(\mathbb{R}_{\text{l.i.}}^{kn}).$$

This uses the lin. ind requirement in def of \widehat{f}

$SL_n(\mathbb{R})$ acts transitively on $\mathbb{R}_{\text{l.i.}}^{kn}$,

$$\text{Stab}_{SL_n(\mathbb{R})}(e_1, \dots, e_k) = \left\{ \left(\begin{array}{c|c} I_k & * \\ \hline 0 & * \end{array} \right) \right\}$$

So $\mathbb{R}_{\text{l.i.}}^{kn}$ admits a unique $SL_n(\mathbb{R})$ -inv.

Radon measure (up to scaling), namely restriction of Vol on \mathbb{R}^{kn} .

* Show that for $f \in C_c(\mathbb{R}_{(i.i.)}^{kn})$

$$\widehat{f} \in L^1(\mathcal{X}_n, m_{\mathcal{X}_n}).$$

We will not do this, this is the main technical part of the proof.

Assuming *, the map $C_c(\mathbb{R}_{(i.i.)}^{kn}) \rightarrow \mathbb{R}$

$$f \longmapsto \int_{\mathcal{X}_n} \widehat{f} \, d m_{\mathcal{X}_n}.$$

is a continuous positive linear functional on

$C_c(\mathbb{R}_{(i.i.)}^{kn})$ so defines a Radon measure μ on $\mathbb{R}_{(i.i.)}^{kn}$.

Claim: $\forall f \in C_c(\mathbb{R}_{(i.i.)}^{kn}) \forall g \in S_n(\mathbb{R}),$

$$\widehat{n, k}(f \circ g) = \left(\widehat{n, k} f \right) \circ g$$

$$\widehat{n, k}(f \circ g)(L) = \sum_{\substack{u_1, \dots, u_k \in L \text{ dof} \\ \text{c.i.}}} f \circ g(u_1, \dots, u_k) = \sum_{\substack{u_1, \dots, u_k \in L \text{ dof} \\ \text{c.i.}}} f(gu_1, \dots, gu_k)$$

$$= \sum_{\substack{v_1, \dots, v_k \in gL \text{ dof} \\ \text{c.i.}}} f(v_1, \dots, v_k) = \left(\widehat{n, k} f \right)(gL)$$

\uparrow
 $v_i = gu_i$

This proves claim.

It follows that μ defined above is $SL_n(\mathbb{R})$ -inv.,
because $f \circ g \in SL_n(\mathbb{R})$

$$\int_{\mathbb{R}^{nk} \text{ c.i.}} (f \circ g) d\mu = \int_{\mathcal{X}_n} \widehat{n, k} f \circ g d m_{\mathcal{X}_n} \stackrel{\text{claim}}{=} \int_{\mathcal{X}_n} \widehat{n, k} f d m_{\mathcal{X}_n} \stackrel{\text{def}}{=} \int_{\mathbb{R}^{nk} \text{ c.i.}} f d\mu$$

$$\int_{\mathcal{X}_n} \widehat{n, k} f \circ g d m_{\mathcal{X}_n} \stackrel{\text{inv.}}{=} \int_{\mathcal{X}_n} \widehat{n, k} f d m_{\mathcal{X}_n} = \int_{\mathbb{R}^{nk} \text{ c.i.}} f d\mu$$

$m_{\mathcal{X}_n}$ is $SL_n(\mathbb{R})$ -inv.

By uniqueness, $\exists c > 0$ s.t. $\mu = c \text{Vol}$

$$\int_{\mathcal{X}_n} \hat{f}^{n,k} d\mu_{\mathcal{X}_n} = c \int_{\mathbb{R}^{nk}} f d\text{Vol} \quad (**)$$

Remains to show $c=1$, for this use

$$f_R = \frac{\mathbb{1}_{B_R}}{\text{Vol}(B_R)} \quad B_R = [-R, R]^{kn}$$

As we did for $k=1$, not hard to show

$$\forall L \in \mathcal{X}_n, \quad \hat{f}_R(L) \xrightarrow{R \rightarrow \infty} 1.$$

Using the dominated convergence theorem (need to justify), and plugging (**), get $c=1$.

Cor For $f \in L^1(\mathbb{R}^n)$, $\hat{f}(L) = \int_{\mathbb{R}^n} f(x) \chi_L(x) dx$

satisfies $\hat{f} \in L^k(\mathcal{X}_n, \mu_{\mathcal{X}_n})$, for all $k \in \mathbb{N}$, $k < n$.

Pf Given $f \in L^1(\mathbb{R}^n)$, define $f_i \in L^1(\mathbb{R}^{kn})$
 by $f_i(u_1, \dots, u_k) = f(u_1) f(u_2) \dots f(u_k)$.

$$\int_{\mathcal{X}_n} (\hat{f})^k d\mu_{\mathcal{X}_n}(L) = \int_{\mathcal{X}_n} \left(\sum_{u \in L \text{ dot}} f(u) \right)^k d\mu_{\mathcal{X}_n}(L) =$$

$$= \int_{\mathcal{X}_n} \left(\sum_{u_1, \dots, u_k \in L \text{ dot}} f(u_1) \dots f(u_k) \right) d\mu_{\mathcal{X}_n}(L) = \int_{\mathcal{X}_n} \sum_{\substack{u_1, \dots, u_k \\ \in L \text{ dot}}} f_i(u_1, \dots, u_k) d\mu_{\mathcal{X}_n}(L)$$

$$\neq \int_{\mathcal{X}_n} \sum_{i=1}^k \hat{f}_i d\mu_{\mathcal{X}_n} < \infty$$

This is a mistake, def of \hat{f}
 involves lin. ind. vectors. Need to
 separate into independent and dependent
 k -tuples, and work inductively on k , see
 next proof.

Want bounds on $\int_{\mathcal{X}_n} (\hat{f})^k d\mu_{\mathcal{X}_n}$.

Thm (Rogers second moment bound '85) $\forall n \exists C$

$\forall f \in C_c(\mathbb{R}^n), f \geq 0$, $\hat{f}(L) = \sum_{x \in L \text{ dot}} f(x)$ $\hat{f}: \mathcal{X}_n \rightarrow \mathbb{R}$,

satisfies $\hat{f} \in L^2(\mathcal{X}_n, \mu_{\mathcal{X}_n})$ and

$$\text{Var}(\hat{f}) = \int_{\mathcal{X}_n} \left(\hat{f} - \int_{\mathcal{X}_n} \hat{f} d\mu_{\mathcal{X}_n} \right)^2 d\mu_{\mathcal{X}_n} \leq C \|f\|_{\infty}^2 \int_{\mathcal{X}_n} |f| d\mu_{\mathcal{X}_n}$$

Pf: $\text{Var}(X) = \mathbb{E}((X - \mathbb{E}(X))^2) = \mathbb{E}(X^2) - (\mathbb{E}(X))^2$

So we compute $\int_{\mathcal{X}_n} (\hat{f})^2 d\mu_{\mathcal{X}_n}$, using

(SSF2)_{n,k} with $k=2$.

Define $f_i(x,y) = f(x)f(y)$ $f_i: \mathbb{R}^{2n} \rightarrow \mathbb{R}$

$$\int_{\mathcal{X}_n} (\hat{f})^2 d\mu_{\mathcal{X}_n} = \int_{\mathcal{X}_n} \left(\sum_{u \in L \setminus \text{dot}} f(u) \right) \left(\sum_{v \in L \setminus \text{dot}} f(v) \right) d\mu_{\mathcal{X}_n}$$

$$= \int_{\mathcal{X}_n} \left[\left(\sum_{u,v \in L \setminus \text{dot}} f(u)f(v) \right) + \left(\sum_{u,v \in L \setminus \text{dot}} f(u)f(v) \right) \right] d\mu_{\mathcal{X}_n}$$

i.i. lin. dep.

$$\stackrel{\text{①}}{=} \int_{\mathcal{X}_n} n^2 \hat{f}_1 d\mu_{\mathcal{X}_n} + \int_{\mathcal{X}_n} \left(\sum_{w \in L} \sum_{\substack{k_1, k_2 \in Z \setminus \text{dot} \\ \text{primitive } k_1 > 0}} f(k_1 w) f(k_2 w) \right) d\mu_{\mathcal{X}_n}$$

uses corrected form of $n^2 \hat{f}$

divide into two cases, $k_1 \neq k_2$ and $k_1 = k_2$, they are identical

$$\stackrel{\text{②}}{\leq} \int_{\mathbb{R}^{2n}} f_1 d\text{Vol} + 2 \int_{\mathcal{X}_n} \sum_{\substack{k_1, k_2 \in Z \\ |k_1| \leq |k_2|}} \sum_{\substack{w \in L \\ \text{primitive}}} f(k_1 w) f(k_2 w) d\mu_{\mathcal{X}_n}$$

(SSF)_{2,n}

P

divide into cases, $k_2 > 0$ and $k_2 < 0$. Write $k = k_2$ and bound $f(kw) \leq \|f\|_\infty$

$$\leq \left(\int_{\mathbb{R}^n} f dVol \right)^2 + 4 \|f\|_\infty \int_{\mathbb{Z}^n} \sum_{k \in \mathbb{N}} k \sum_{w \in L} f(kw) d\mu_{\mathbb{Z}^n}$$

WEL primitive

Using $f_k(x) = f(kx)$ and $\hat{f}^{\wedge P} = \sum_{w \in L} f(w)$ primitive

$$\Rightarrow \left(\int_{\mathbb{R}^n} f dVol \right)^2 + 4 \|f\|_\infty \sum_{k \in \mathbb{N}} k \int_{\mathbb{Z}^n} \hat{f}_k^{\wedge P} d\mu_{\mathbb{Z}^n}$$

$$= \left(\int_{\mathbb{R}^n} f dVol \right)^2 + 4 \|f\|_\infty \sum_{k \in \mathbb{N}} \frac{k}{\zeta(n)} \int_{\mathbb{R}^n} f_k(x) dVol(x)$$

(SSF 1) for $\hat{f}^{\wedge P}$

$$= \left(\int_{\mathbb{R}^n} f dVol \right)^2 + \frac{4}{\zeta(n)} \|f\|_\infty \sum_{k \in \mathbb{N}} \frac{k}{k^n} \int_{\mathbb{R}^n} f dVol$$

$kx = y$
 $dVol(x) = \frac{1}{k^n} dVol(y)$

$$= \left(\int_{\mathbb{R}^n} f dVol \right)^2 + \frac{4 \zeta(n-1)}{\zeta(n)} \|f\|_\infty \int_{\mathbb{R}^n} f dVol.$$

Subtracting $\left(\int_{\mathbb{Z}^n} \hat{f} d\mu_{\mathbb{Z}^n} \right)^2 = \left(\int_{\mathbb{R}^n} f dVol \right)^2$ from both sides, get conclusion desired, with $C = \frac{4 \zeta(n-1)}{\zeta(n)}$

Application to counting lattice points in growing sets (following Schmidt '60)

Recall (ex. 3) $\forall L \in \mathcal{X}_n$, any bounded convex set S with non-empty interior, rS is the dilation of S by factor r

$$\#(L \cap rS) = \underset{\substack{\uparrow \\ \text{main term}}}{\text{Vol}(rS)} + \underset{\substack{\uparrow \\ \text{error term}}}{O(r^{n-1})} \text{ as } r \rightarrow \infty,$$

where implicit const. depends on L and on S .

$$= \text{Vol}(rS) + O\left(\left(\text{Vol}(rS)\right)^{1-\frac{1}{n}}\right).$$

Q: can error term be improved?

for which choices of L, S , maybe for more general sequences of growing bodies.

Conj (Götze '98) For $m_{\mathcal{X}_n}$ -a.e. $L \subset \mathbb{R}^n$,
for $B = S = B(0, 1)$ euclidean ball, $\forall \epsilon > 0$

$$\begin{aligned} \#(L \cap rS) &= \text{Vol}(rS) + O\left(r^{\frac{n}{2} - \frac{1}{2} + \varepsilon}\right) \\ &= \text{Vol}(rS) + O\left(\text{Vol}(rS)^{\frac{1}{2} - \frac{1}{2n} + \varepsilon}\right) \end{aligned}$$

where implicit const. depends on L, S, ε .

Let $\{\Omega_r : r > 0\}$ be a collection of Doml sets. We say it is an unbounded ordered family if:

(i) $r_1 \leq r_2 \Rightarrow \Omega_{r_1} \subset \Omega_{r_2}$

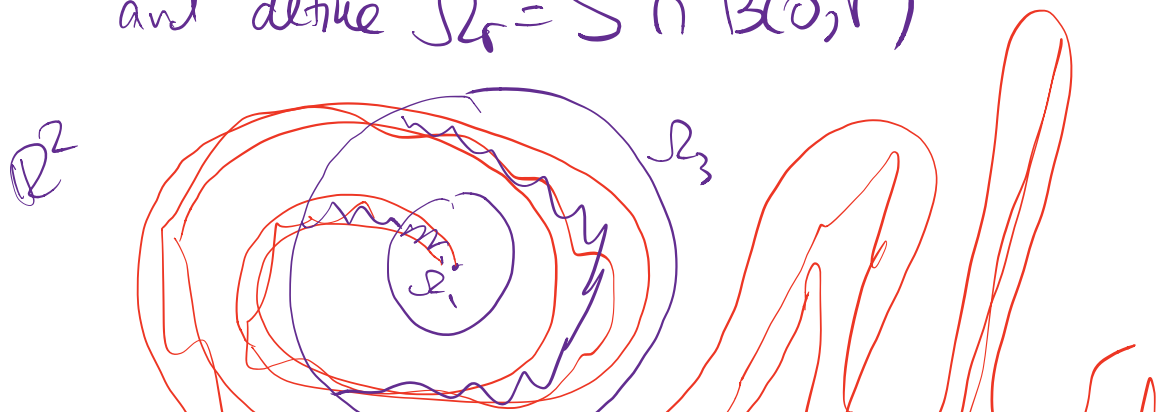
(ii) $\text{Vol}(\Omega_r) < \infty$ and $\text{Vol}(\Omega_r) \xrightarrow[r \rightarrow \infty]{} \infty$.

(iii) $\forall V > 0 \exists r$ s.t. $\text{Vol}(\Omega_r) = V$.

Examples * $\Omega_r = B(0, r)$

* For any $S \subset \mathbb{R}^n$ $\text{Vol}(S) = \infty$,

and define $\Omega_r = S \cap B(0, r)$





Thm (Schmidt '60) Let $\{\Omega_r: r>0\}$ be an unbounded ordered family in \mathbb{R}^n . Then for every $\varepsilon>0$, for $m_{\mathcal{X}_n}$ -a.e. $L \in \mathcal{X}_n$,

$$(*) \#(L \cap \Omega_r) = \text{Vol}(\Omega_r) + O\left(\text{Vol}(\Omega_r)^{\frac{1}{2}+\varepsilon}\right)$$

(implicit const depends on L and on ε).

Remark Schmidt's result does not attain the bound $\frac{1}{2} - \frac{1}{2n} + \varepsilon$ in Conj, but it's the best result known.

For the proof we will use some lemmas and notation.

By (iii), for each $N \in \mathbb{N}$ $\exists r_N$ s.t.

$S_N = \Omega_{r_N}$ satisfies $\text{Vol}(S_N) = N$.

Let $S_{N_1} = \mathbb{1} S_{N_2}$, for $N_1 < N_2$ let

$$N_1 \rho_{N_2} = \rho_{N_2} - \rho_{N_1} = \mathbb{1}_{S_{N_2} \setminus S_{N_1}}$$

sets are nested.

Lemma (dyadic decomposition) let $T \in \mathbb{N}$

$$\text{and let } K_T = \left\{ (N_1, N_2) \in \mathbb{N} \times \mathbb{N} : 0 \leq N_1 < N_2 \leq 2^T \right\}$$

$$\exists t \in \mathbb{N}^{<T}, 2^t | N_1, N_2 = N_1 + 2^t$$

$$\text{Then } \sum_{(N_1, N_2) \in K_T} \text{Var}(\widehat{\rho}_{N_1, N_2}) \leq 2(T+1)2^T.$$

Example $T=4 \quad z=4 \quad N_1=0 \quad N_2=2^T$

$t=3 \quad N_1=0, N_2=8 \quad N_1=8, N_2=16$

$t=2 \quad (N_1, N_2) \in \{(0, 4), (4, 8), (8, 12), (12, 16)\}$

⋮

$t=0 \quad \{(k, k+1) : k \in \{0, \dots, 15\}\}$

$$\underline{\text{pf}} \quad \text{Var}(\widehat{\rho}_{N_1, N_2}) \leq 2 \|\rho_{N_1, N_2}\| \int_{S_{N_1, N_2}} \widehat{\rho}_{N_1, N_2} d\mu_{\mathcal{X}_n}$$

$$= 2 \text{Vol}(S_{N_2} \setminus S_{N_1}) = 2(N_2 - N_1) \quad (1)$$

(note N_1, N_2 not cts, can derive from Rogers 2nd moment bound and an approximation argument).

Each value of $N_2 - N_1 = 2^t$, for $t \in \{0, \dots, T\}$, occurs 2^{T-t} times, and therefore

$$(2) \quad \sum_{(N_1, N_2) \in \mathcal{K}_T} (N_2 - N_1) = \sum_{0 \leq t \leq T} 2^t 2^{T-t} = (T+1)2^T.$$

Combining (1) and (2) gives the conclusion.

Lemma 2 Fix $\varepsilon > 0$, let C be as in Rogers 2nd moment bound. For all $T \in \mathbb{N}$, there is $\text{Bad}_T \subset \mathcal{X}_n$ with

$$M_{\mathcal{X}_n}(\text{Bad}_T) \leq \frac{C}{(\log 2)^{T-1}} \varepsilon$$

s.t. for every every $N \leq 2^T$, and every

$L \notin \text{Bad}_T$,

$$(3) \quad \left(\hat{\int}_N(L) - N \right)^2 \leq T(T+1)2^T (\log(2)T-1)^{1+\varepsilon}$$

Pf of Thm assuming Lemma 2

$$\sum_{T \in \mathbb{N}} \frac{1}{(\log(2)T-1)^{1+\varepsilon}} < \infty$$

By Borel-Cantelli, for m_X a.e. L ,

$\exists T_0 = T_0(L)$ s.t. $\forall T \geq T_0, L \notin \text{Bad}_T$.

Let $N \geq N_0 = 2^{T_0}$, and let T s.t.

$2^{T-1} \leq N < 2^T$. By Lemma 2,

$$\left(\hat{\int}_N(L) - N \right)^2 \leq T(T+1)2^T (\log(2)T-1)^{1+\varepsilon} = O\left(N (\log N)^{3+\varepsilon}\right)$$

(4)

For any $\rho > 0$, let N be such that $N \in \mathbb{N}$,

$\text{Vol}(\Omega_r) \in [N, N+1)$. Let $r_1 \leq r < r_2$ s.t.

$$\text{Vol}(\Omega_{r_1}) = N, \text{Vol}(\Omega_{r_2}) = N+1$$

$S_N = \Omega_{r_1} \subset \Omega_r \subset \Omega_{r_2} = S_{N+1}$. Then

$$(5) \quad \#(\Omega_{r_1} \cap L) - (N+1) \leq \\ \leq \#(\Omega_r \cap L) - \text{Vol}(\Omega_r) \leq \#(S_{N+1} \cap L) - N$$

Taking square root in (4), both sides of

$$(5) \text{ are } O(N^{\frac{1}{2} + \epsilon}).$$

$$\text{So } \# \Omega_{r_1} \cap L - \text{Vol}(\Omega_r) \leq C N^{\frac{1}{2} + \epsilon}$$

and the opposite inequality is similar.

This proves thm, modulo lemma 2.

Pf of Lemma 2 Define

$$\text{Bad}_T = \left\{ L \in \mathcal{X}_n : \sum_{(N_1, N_2) \in \mathcal{K}_T} (N_1 \hat{P}_{N_2}(L) - (N_2 - N_1))^2 > (T+1) 2^T (\log 2)^{HT} \right\}$$

$$\text{let } X = \sum_{(N_1, N_2) \in K_T} \left(\widehat{P}_{N_1, N_2}(L) - (N_2 - N_1) \right)^2,$$

$$\text{so } \mathbb{E}(X) = \sum_{(N_1, N_2) \in K_T} \text{Var} \left(\widehat{P}_{N_1, N_2} \right)$$

$$\text{let } a = \frac{1}{c} (\log(2)T - 1)^{1+\varepsilon}$$

$$\text{Then } \text{Bad}_T = \left\{ L \in \mathcal{X}_n : X > c(T+1)2^T \cdot a \right\}$$

$$\subset \left\{ L \in \mathcal{X}_n : X > a \mathbb{E}(X) \right\}$$

Lemma 1

By the Markov inequality,

$$m_{\mathcal{X}}(\text{Bad}_T) < \frac{1}{a} = \frac{c}{(\log(2)T - 1)^{1+\varepsilon}}$$

To prove (3), assume $N \leq 2^T$ and $L \notin \text{Bad}_T$.

Express $[0, N)$ as a union of intervals $[N_1, N_2)$

where (N_1, N_2) , that is

$$[0, N) = \bigsqcup_{(N_1, N_2) \in \mathcal{I}_N} [N_1, N_2), \text{ where } \mathcal{I}_N \subset K_T, \#\mathcal{I}_N \leq T.$$

This possible by writing N in dyadic expansion, for example (taking $T=4$ as in previous example) $13 = 8 + 4 + 1$,

$$\text{so } [0, 13) = [0, 8) \sqcup [8, 12) \sqcup [12, 13).$$

$$\text{Then } \hat{P}_N = \sum_{(N_1, N_2) \in \mathcal{I}_N} \widehat{P}_{N_1, N_2},$$

$$\hat{P}_N(L) - N = \sum_{(N_1, N_2) \in \mathcal{I}_N} (\widehat{P}_{N_1, N_2}(L) - (N_2 - N_1))$$

$$\leq \sum_{(N_1, N_2) \in \mathcal{I}_N} (\widehat{P}_{N_1, N_2}(L) - (N_2 - N_1)^2)^{\frac{1}{2}} \sqrt{T}$$

Cauchy-Schwarz

$$\sum_{i=1}^m a_i = \sum_{i=1}^m a_i \cdot 1 \leq \left(\sum_{i=1}^m a_i^2 \right)^{\frac{1}{2}} \cdot \sqrt{m}$$

$$\leq \sqrt{T} \left((TH) 2^T (\log_2(T-1))^{1+\varepsilon} \right)^{\frac{1}{2}}$$

def of Bad T

Now square both sides to obtain (3).