

# Lecture 2: Central limit theorems for statistics of geometric structures

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# Probabilistic Analysis of Geometric Structures

- **Lecture 1:** Probabilistic analysis of Euclidean optimization problems
- **Lecture 2:** Central limit theorems for statistics of geometric structures
- **Lecture 3:** Limit theory for statistics of geometric structures via stabilizing score functions
- **Lecture 4:** Statistics of random polytopes
- **Lecture 5:** Rates of multivariate normal approximation for statistics of geometric structures

# Lecture 2: Central limit theorems for statistics of geometric structures

- **I Introduction**

- **II Models and Results**

Random packing models, Growth models, Gilbert Graph

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- **I Introduction**

- **II Models and Results**

Random packing models, Growth models, Gilbert Graph

- **III Key Idea**

Stability of difference operator

- **IV Central limit theorems**

Statement

Proof

- **V Large Deviation Principles**

# I Introduction

- Most models of physical systems involve particles which interact 'locally', inducing long-range interactions.
- We take our particles to be points, usually the realization of an i.i.d. collection of r.v.  $X_i, i \geq 1$  or a homogeneous Poisson point process  $\mathcal{P}_1$ .

# I Introduction

- Most models of physical systems involve particles which interact 'locally', inducing long-range interactions.
- We take our particles to be points, usually the realization of an i.i.d. collection of r.v.  $X_i, i \geq 1$  or a homogeneous Poisson point process  $\mathcal{P}_1$ .
- For ease of exposition, we consider  $Q_n := [-\frac{1}{2}n, \frac{1}{2}n]^d$  and let  $U_i, i \leq n$  be i.i.d. uniform on  $Q_n$ .
- We let  $H$  be a generic functional defined on finite point sets.
- We are interested in the behavior of the **Poisson** functional  $H(\mathcal{P}_1 \cap Q_n)$  and the **binomial** functional  $H(U_1, \dots, U_n)$ .

## Natural questions:

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$$\frac{H(U_1, \dots, U_n) - \mathbb{E} H(U_1, \dots, U_n)}{\sqrt{\text{Var} H(U_1, \dots, U_n)}} \xrightarrow{\mathcal{D}} N(0, 1)?$$



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3. (Probability bounds) Seek good bounds for

$$\mathbb{P}(H(U_1, \dots, U_n) \geq t).$$

## II Models and Results

a. **Packing Model.** Unit volume balls  $B_1, \dots, B_k$  arrive sequentially and uniformly at random in the cube  $Q_n := [-\frac{1}{2}n, \frac{1}{2}n]^d$ . Packing rules:

- Pack ball  $B_1$ .

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- Pack ball  $B_1$ .
- Pack ball  $B_i$ ,  $i > 1$ , if  $B_i$  does not overlap any ball in  $B_1, B_2, \dots, B_{i-1}$  which has already been packed.
- Picture for  $d = 1$  looks like this:



- **a. Packing Model (contd).** Fix  $k \in \{1, 2, \dots, \infty\}$ . If balls  $B_1, \dots, B_k$  have centers at points  $U_1, \dots, U_k \in Q_n$  with respective arrival times  $\tau_1, \dots, \tau_k$ , then let

$$H_n(U_1, \dots, U_k)$$

denote the number of packed (accepted) balls on the substrate  $Q_n$ .

- The random variable  $H_n(U_1, \dots, U_k)$  is the number of accepted particles in the random sequential adsorption (RSA) model.  $H_n(U_1, \dots, U_\infty)$  is ‘packing number’.

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**Rényi’s Thm:**  $d = 1 \Rightarrow$

$$\lim_{n \rightarrow \infty} \frac{\mathbb{E} H_n(U_1, \dots, U_\infty)}{n} = \int_0^\infty \exp(-2 \int_0^t \frac{1 - e^{-u}}{u} du) dt \sim 0.748.$$

**Dvoretzky + Robbins CLT:**  $d = 1 \Rightarrow$

$$\frac{H_n(U_1, \dots, U_\infty) - \mathbb{E} H_n(U_1, \dots, U_\infty)}{\sqrt{\text{Var} H_n(U_1, \dots, U_\infty)}} \xrightarrow{\mathcal{D}} N(0, 1).$$

- **a. Packing Model (contd)** What about the case  $d \geq 2$ ? Hurdles: lack of subadditivity, lack of independence over subsets of cube  $Q_n$ .

**LLN:**  $d \geq 2, \alpha \in (0, \infty)$ :

$$\lim_{n \rightarrow \infty} \frac{H_n(U_1, \dots, U_{[\alpha n]})}{n} = C(\alpha) \quad a.s.$$

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# II Models and Results

## b. Geometric Graph (Gilbert Graph)

**Def.** Given a finite point set  $\mathcal{X}$ ,  $r \in (0, \infty)$ , put

$$C_B(\mathcal{X}, r) := \bigcup_{x \in \mathcal{X}} B_r(x).$$

When  $\mathcal{X}$  is PPP we get the Boolean model. It gives rise to the geometric graph  $GG_r(\mathcal{X})$ : join two points  $x$  and  $y$  with an edge iff  $B_{r/2}(x) \cap B_{r/2}(y) \neq \emptyset$ .

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**Def.** Let  $H(\mathcal{X})$  be the number of edges in geometric graph  $GG_r(\mathcal{X})$ .

**CLT:**  $d \geq 2, r > 0$ :

$$\frac{H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)}{\sqrt{\text{Var} H(\mathcal{P}_1 \cap Q_n)}} \xrightarrow{\mathcal{D}} N(0, 1).$$

## c. Spatial Birth Growth Models

- The model: cells form at random locations  $U_1, \dots, U_k \in Q_n$  at times  $\tau_1, \dots, \tau_k$ , respectively.
- Initially the new cell around  $U_i$  takes the form of a ball of radius  $R_i \geq 0$  centered at  $U_i$ ; then the cell grows radially in all directions with constant speed  $v$ .

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- Initially the new cell around  $U_i$  takes the form of a ball of radius  $R_i \geq 0$  centered at  $U_i$ ; then the cell grows radially in all directions with constant speed  $v$ .
- New cells form only in the *uncovered space* in  $\mathbb{R}^d$ .
- This models crystal growth, cavitation.
- $H(U_1, \dots, U_n)$  is the volume of the region covered by the first  $n$  cells.

### c. Spatial Birth Growth Models (contd)

- $H(U_1, \dots, U_n)$  is the volume of the region covered by the first  $n$  cells.

**CLT:**  $\frac{H(U_1, \dots, U_n) - \mathbb{E} H(U_1, \dots, U_n)}{\sqrt{n}} \xrightarrow{\mathcal{D}} N(0, \sigma^2).$

Chiu, Quine,...

# III Key Ideas

## Recall:

$H$ : a function defined on point set of  $\mathbb{R}^d$ .

$$Q_n := [-\frac{1}{2}n, \frac{1}{2}n]^d$$

$U_i, 1 \leq i \leq n$ , i.i.d. uniform on  $Q_n$ .

**Goal:** Seek conditions on  $H$  yielding

$$\frac{H(U_1, \dots, U_n)}{n} \xrightarrow{P} \text{constant} \quad (LLN)$$

and

$$\frac{H(U_1, \dots, U_n) - \mathbb{E} H(U_1, \dots, U_n)}{\sqrt{\text{Var} H(U_1, \dots, U_n)}} \xrightarrow{\mathcal{D}} N(0, 1) \quad (CLT)$$

# III Key Ideas

- Write  $Q_n := \cup_{i=1}^n Q_{n,i}$ , where  $Q_{n,i}$  are disjoint sub-cubes of volume 1.
- Abbreviate  $\{U_1, \dots, U_n\}$  by  $\mathcal{U}_n$ .
- In general

$$H(\mathcal{U}_n) \neq \sum_{i=1}^n H(\mathcal{U}_n \cap Q_{n,i}),$$

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i.e.,  $H$  is NOT additive.

- If  $H$  were additive, then we could deduce LLN and CLT for  $H(\mathcal{U}_n)$  from the classical limit theorems.



# III Key Ideas

- Restriction of  $H$  to disjoint sets does not give independence
- $H(\mathcal{U}_n \cap A)$  and  $H(\mathcal{U}_n \cap B)$  are dependent in general!

# III Key Ideas

A key idea is to compare the functional  $H(\mathcal{U}_n)$  with  $H(\mathcal{U}_n \cup \{\mathbf{0}\})$ .

Let's start by comparing  $H(\mathcal{P}_1 \cap Q_n)$  with  $H((\mathcal{P}_1 \cap Q_n) \cup \{\mathbf{0}\})$ .

In other words, what happens to  $H$  when we insert an extra point at the origin into homogenous rate 1 Poisson input  $\mathcal{P}_1$ ? How does  $H$  change?

# III Key Ideas

Consider this situation:

$$H\left(\mathcal{P}_1 \cap Q_n\right) - H\left(\mathcal{P}_1 \cap Q\right) = H\left(\mathcal{P}_1 \cap Q\right) - H\left(\mathcal{P}_1 \cap Q\right).$$

This picture says that for the fixed deterministic square  $Q$  we have

$$\begin{aligned} & H((\mathcal{P}_1 \cap Q_n) \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q_n) \\ &= H((\mathcal{P}_1 \cap Q) \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q). \end{aligned}$$

# III Key Ideas: Stability of Difference Operators

The above phenomena rarely happens. Fortunately, a slightly weaker one does. Let  $\mathcal{P}_1$  be unit intensity PPP on  $\mathbb{R}^d$ .

**Def.** Say that  $H$  **stabilizes** on  $\mathcal{P}_1$  if there is a cube  $Q$ ,  $\text{diam}(Q) < \infty$  a.s., such that

$$\begin{aligned} \lim_{n \rightarrow \infty} H((\mathcal{P}_1 \cap Q_n) \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q_n) \\ = H((\mathcal{P}_1 \cap Q) \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q). \end{aligned}$$

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This condition says that the ‘add-one cost’ does not propagate far; it is confined to  $Q$ .

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**Def.**  $D_0 H(\mathcal{P}_1) := H(\mathcal{P}_1 \cup \{\mathbf{0}\}) - H(\mathcal{P}_1)$ .

$D_0$  is called the **first order difference operator**.

Stabilization says that the first order difference operator has a behavior which is determined by local data.

# III Key Ideas: Stability of Difference Operators

Which functionals  $H$  stabilize in the above sense?

Consider the nearest neighbors graph (put an edge between every point and its nearest neighbor). Let  $H(\mathcal{X})$  be the total edge length of the nearest neighbor graph on  $\mathcal{X}$ .



# IV General CLT and Variance Asymptotics

**Theorem.** Let  $H$  be a functional on points in  $\mathbb{R}^d$  which satisfies:

(i) translation invariance, i.e.,  $H(\mathcal{X} + y) = H(\mathcal{X})$ ,  $y \in \mathbb{R}^d$ ,

(ii)  $(2 + \epsilon)$  bounded increments:

$$\sup_{Q \subset \mathbb{R}^d, Q \text{ a cube}} \mathbb{E} |H((\mathcal{P}_1 \cap Q) \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q)|^{2+\epsilon} < \infty,$$

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Then

$$\frac{H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)}{\sqrt{n}} \xrightarrow{\mathcal{D}} N(0, \sigma^2)$$
$$\frac{\text{Var} H(\mathcal{P}_1 \cap Q_n)}{n} \rightarrow \sigma^2.$$

# IV General CLT and Variance Asymptotics

## Applications

We can show that the following functionals stabilize and satisfy the CLT and variance asymptotics:

- a. The number of balls accepted in the RSA packing model,
- b. The volume of the occupied region in spatial birth-growth models,

# IV General CLT and Variance Asymptotics

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- a. The number of balls accepted in the RSA packing model,
- b. The volume of the occupied region in spatial birth-growth models,
- c. The number of edges in the random geometric graph with parameter  $r$ ,
- d. Total edge length of nearest neighbors graph.

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**Limitations:**

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## Limitations:

- no formula for  $\sigma^2$ , no rate of normal convergence, does not address non-uniform input.

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## Limitations:

- no formula for  $\sigma^2$ , no rate of normal convergence, does not address non-uniform input.
- we are unable to show that  $L_{TSP}$  stabilizes. Same for  $L_{MM}, L_{SMST}$ .



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**If the add-one cost  $D_0 H(\mathcal{P}_1) := H(\mathcal{P}_1 \cup \{\mathbf{0}\}) - H(\mathcal{P}_1)$  is non-degenerate, then  $\sigma^2 > 0$ .**

# IV General CLT and Variance Asymptotics

**Proof of CLT.** The proof involves these ingredients:

1. A stability lemma for stabilizing functionals,
2. Expressing  $H$  as a sum of martingale differences,
3. McLeish CLT (this result says that to prove a CLT it is enough to prove an  $L^1$  WLLN).

## IV General CLT and Variance Asymptotics

If  $H$  is stabilizing and trans. invariant, then there is a r.v.  $\Delta_0$  such that

$$\lim_{n \rightarrow \infty} [H((\mathcal{P}_1 \cap Q_n) \cup \mathbf{0}) - H(P_1 \cap Q_n)] = \Delta_0 \quad a.s.$$

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Thus, if  $H$  is stabilizing and trans. invariant, then  $\forall x, y \in \mathbb{R}^d$

$$H((\mathcal{P}_1 \cap Q_n) \cup x) - H(\mathcal{P}_1 \cap Q_n) \rightarrow \Delta_x \text{ a.s.}$$

$$H((\mathcal{P}_1 \cap Q_n) \cup y) - H(\mathcal{P}_1 \cap Q_n) \rightarrow \Delta_y \text{ a.s.}$$

Consequently,

$$H((\mathcal{P}_1 \cap Q_n) \cup x) - H((\mathcal{P}_1 \cap Q_n) \cup y) \rightarrow \Delta_x - \Delta_y \text{ a.s.}$$

i.e, replacing  $x$  by  $y$  produces a change in  $H$  which converges a.s.

# IV General CLT and Variance Asymptotics

## Stability lemma for stabilizing functionals.

**Def.**  $\forall x \in \mathbb{Z}^d$  let  $C(x)$  be unit volume cube with center  $x$ .

**Def.**  $\mathcal{P}'_1$  an independent copy of  $\mathcal{P}_1$ .  $\forall x \in \mathbb{Z}^d$  let

$$\mathcal{P}_{C(x)} := \mathcal{P}_1 \setminus (\mathcal{P}_1 \cap C(x)) \cup (\mathcal{P}'_1 \cap C(x)).$$

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**Stability Lemma.** Let  $H$  be stabilizing and trans. invariant. Then  $\forall x \in \mathbb{Z}^d$  there is a r.v.  $\Delta_{C(x)}$  such that as  $n \rightarrow \infty$

$$H(\mathcal{P}_1 \cap Q_n) - H(\mathcal{P}_{C(x)} \cap Q_n) \rightarrow \Delta_{C(x)}.$$

# IV General CLT and Variance Asymptotics

## Martingale Differences

- $X_i, 1 \leq i \leq n$ , independent r.v.
- $\mathcal{F}_i := \sigma(X_1, \dots, X_i)$ .
- Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .
- Then

$$f(X_1, \dots, X_n) - \mathbb{E} f(X_1, \dots, X_n) = \sum_{i=1}^n D_i,$$

where

# IV General CLT and Variance Asymptotics

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$$\begin{aligned} D_i &:= \mathbb{E} (f(X_1, \dots, X_n) | \mathcal{F}_i) - \mathbb{E} (f(X_1, \dots, X_n) | \mathcal{F}_{i-1}) \\ &= \mathbb{E} (f(X_1, \dots, X_n) | \mathcal{F}_i) - \mathbb{E} (f(X_1, \dots, X'_i, \dots, X_n) | \mathcal{F}_i). \end{aligned}$$

Here  $X'_i$  is a copy of  $X_i$ .



# IV General CLT and Variance Asymptotics

## Martingale Differences

- $X_i, 1 \leq i \leq n$ , independent r.v.
- $\mathcal{F}_i := \sigma(X_1, \dots, X_i)$ .
- Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .
- Then

$$f(X_1, \dots, X_n) - \mathbb{E} f(X_1, \dots, X_n) = \sum_{i=1}^n D_i,$$

where

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Here  $X'_i$  is a copy of  $X_i$ .

Seek an analogous decomposition for  $H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)$  in terms of martingale differences. But first:

## IV General CLT and Variance Asymptotics

**Thm (McLeish):** Let  $D_i, i \geq 1$ , be martingale differences such that:

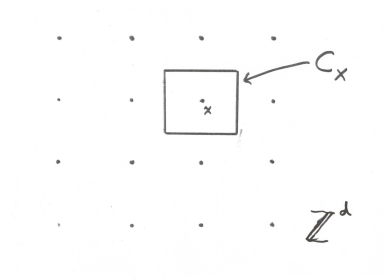
- $\sup_n \mathbb{E} \max_{i \leq n} \left( \frac{D_i^2}{n} \right) < \infty,$
- $\frac{1}{\sqrt{n}} \max_{i \leq n} |D_i| \xrightarrow{P} 0,$
- $\frac{1}{n} \sum_{i=1}^n D_i^2 \xrightarrow{L^1} \sigma^2.$

Then as  $n \rightarrow \infty$  we have

$$\frac{\sum_{i=1}^n D_i}{\sqrt{n}} \xrightarrow{\mathcal{D}} N(0, \sigma^2).$$

# IV General CLT and Variance Asymptotics

**Martingale differences for  $H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)$**



- $\mathbb{Z}^d$  ordered with dictionary ordering  $\prec$ .
- $\forall x \in \mathbb{Z}^d$  let  $\mathcal{F}_x := \sigma(\mathcal{P}_1 \cap \cup_{y \preceq x} C_y)$ .

# IV General CLT and Variance Asymptotics

**Martingale differences for  $H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)$  :**

- $Q_n := [-\frac{1}{2}n, \frac{1}{2}n]^d$
- Write  $\mathbb{Z}^d \cap Q_n = \{x_1, x_2, \dots, x_n\}$ ,  $x_{i-1} \prec x_i$ .

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$$\begin{aligned} D_i &:= \mathbb{E}(H(\mathcal{P}_1 \cap Q_n) | \mathcal{F}_{x_i}) - \mathbb{E}(H(\mathcal{P}_1 \cap Q_n) | \mathcal{F}_{x_{i-1}}) \\ &= \mathbb{E}(H(\mathcal{P}_1 \cap Q_n) | \mathcal{F}_{x_i}) - \mathbb{E}(H(\mathcal{P}_{C(x_i)} \cap Q_n) | \mathcal{F}_{x_i}) \\ &= \mathbb{E}([H(\mathcal{P}_1 \cap Q_n) - H(\mathcal{P}_{C(x_i)} \cap Q_n)] \mid \mathcal{F}_{x_i}) \end{aligned}$$

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To apply the McLeish CLT we need to show WLLN in  $L^1$  sense:

$$\frac{1}{n} \sum_{i=1}^n D_i^2 \xrightarrow{L^1} \sigma^2.$$

# IV General CLT and Variance Asymptotics

**Proof that**  $\frac{1}{n} \sum_{i=1}^n D_i^2 \xrightarrow{L^1} \sigma^2$ . **(\*\*)**

(i)  $\forall x \in \mathbb{Z}^d \cap Q_n$ , stability lemma implies for all  $1 \leq i \leq n$ :

$$[H(\mathcal{P}_1 \cap Q_n) - H(\mathcal{P}_{C(x_i)} \cap Q_n)] \rightarrow \Delta_{C(x_i)} \quad a.s.$$

$(2 + \epsilon)$ -bounded increments condition implies uniformly in  $x_i \in \mathbb{Z}^d \cap Q_n$

$$D_{x_i}^2 := (\mathbb{E}([\dots]) | \mathcal{F}_{x_i}))^2 \xrightarrow{L^1} \mathbb{E}(\Delta_{C(x_i)} | \mathcal{F}_{x_i})^2.$$

## IV General CLT and Variance Asymptotics

**Proof that**  $\frac{1}{n} \sum_{i=1}^n D_i^2 \xrightarrow{L^1} \sigma^2$ . **(\*\*)**

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(ii) Ergodic thm implies

$$\frac{1}{n} \sum_{x_i \in Q_n} (\mathbb{E}(\Delta_{C(x_i)} | \mathcal{F}_{x_i}))^2 \xrightarrow{L^1} \sigma^2 := \mathbb{E}(\mathbb{E}(\Delta_{C(0)} | \mathcal{F}_0))^2.$$

(iii) Combining (i) and (ii) we deduce the desired convergence (\*\*).



# Martingale differences for $H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n :)$

**Thm (McLeish):** Let  $D_i, i \geq 1$ , be martingale differences such that:

$$\cdot \sup_n \mathbb{E} \max_{i \leq n} \left( \frac{D_i^2}{n} \right) < \infty, \frac{1}{\sqrt{n}} \max_{i \leq n} |D_i| \xrightarrow{P} 0, \frac{1}{n} \sum_{i=1}^n D_i^2 \xrightarrow{L^1} \sigma^2.$$

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We showed the third condition. Thus we have

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**Thm (CLT).** Let  $H$  be a functional on points in  $\mathbb{R}^d$  which satisfies:

- (i) translation invariance, i.e.,  $H(\mathcal{X} + y) = H(\mathcal{X})$ ,  $y \in \mathbb{R}^d$ ,
- (ii) bounded increments:  $\sup_Q \mathbb{E} |H(\mathcal{P}_1 \cap Q \cup \{\mathbf{0}\}) - H(\mathcal{P}_1 \cap Q)|^{2+\epsilon} < \infty$ ,
- (iii) stability of ‘add-one cost’.

Then

$$\frac{H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)}{\sqrt{n}} \xrightarrow{\mathcal{D}} N(0, \sigma^2).$$

## IV General CLT and Variance Asymptotics

### General CLT for $H(\mathcal{P}_1 \cap Q_n)$ .

Under certain moment conditions on  $H$  we have

$$\sup_{x \in \mathbb{R}} \left| \mathbb{P} \left( \frac{H(\mathcal{P}_1 \cap Q_n) - \mathbb{E} H(\mathcal{P}_1 \cap Q_n)}{\sqrt{\text{Var} H(\mathcal{P}_1 \cap Q_n)}} \leq x \right) - \mathbb{P}(N(0, 1) \leq x) \right| \leq \sum_{i=1}^5 \tau_i.$$

The  $\tau_i$  are integrals of sums and products of first order difference operators for  $H$ :

$$D_x H(\mathcal{P}_1) := H(\mathcal{P}_1 \cup x) - H(\mathcal{P}_1)$$

as well as second order difference operators

$$\begin{aligned} D_{x_1, x_2}^{(2)} H(\mathcal{P}_1) &:= D_{x_1} D_{x_2} (H(\mathcal{P}_1)) \\ &= H(\mathcal{P}_1 \cup \{x_1, x_2\}) - H(\mathcal{P}_1 \cup \{x_2\}) - H(\mathcal{P}_1 \cup \{x_1\}) + H(\mathcal{P}_1). \end{aligned}$$

The results are an improvement because they provide rates of convergence. The unwieldy terms  $\tau_i$  can only be simplified when the functional  $H$  is a stabilizing functional. We discuss this in the next lecture.

# V Large Deviation Principles

**Example.**  $X_i, i \geq 1$ , i.i.d.  $N(0, 1)$ .  $S(n) = \sum_{i=1}^n X_i$ ;  $\frac{S(n)}{\sqrt{n}} \stackrel{\mathcal{D}}{=} N(0, 1)$ .

Thus for all Borel  $A \subset \mathbb{R}$  we have

$$\mathbb{P}\left(\frac{S(n)}{\sqrt{n}} \in A\right) = \frac{1}{\sqrt{2\pi}} \int_A \exp\left(-\frac{x^2}{2}\right) dx.$$

So

$$\mathbb{P}\left(\left|\frac{S(n)}{n}\right| \geq t\right) = \frac{2}{\sqrt{2\pi}} \int_{t\sqrt{n}}^{\infty} \exp\left(-\frac{x^2}{2}\right) dx \sim \sqrt{\frac{2}{\pi}} \frac{1}{t\sqrt{n}} \exp\left(-\frac{t^2 n}{2}\right),$$

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which gives as  $n \rightarrow \infty$

$$\frac{\log \mathbb{P}\left(\left|\frac{S(n)}{n}\right| \geq t\right)}{n} \rightarrow \frac{-t^2}{2}.$$

- With small probability,  $\left|\frac{S(n)}{n}\right|$  takes on relatively large values.
- $\frac{-t^2}{2}$  is the rate function.

# V Large Deviation Principles

## Cramer's LDP

- $X_i, i \geq 1$ , i.i.d.  $S(n) = \sum_{i=1}^n X_i$ ;  $\mathbb{E} X_1 = \mu$ .
- log m.g.f.

$$\Lambda(\lambda) := \log \mathbb{E} \exp(\lambda X_1) \geq \lambda \mu.$$



# V Large Deviation Principles

## Cramer's LDP

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- log m.g.f.

$$\Lambda(\lambda) := \log \mathbb{E} \exp(\lambda X_1) \geq \lambda \mu.$$

- $\Lambda(\lambda)$  is convex. Fenchel-Legendre transform (convex dual) of  $\Lambda(\lambda)$  is

$$\Lambda^*(x) = \sup_{\lambda \in \mathbb{R}} (\lambda x - \Lambda(\lambda)) \geq 0.$$

- The convex dual is quadratic if  $X_1$  is  $N(\mu, \sigma^2)$ .

## Cramer's LDP

· Assume  $\Lambda(\lambda) < \infty$ . Then

(a) For all closed  $F \subseteq \mathbb{R}$  we have

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}\left(\frac{S(n)}{n} \in F\right) \leq - \inf_{x \in F} \Lambda^*(x).$$

(b) For all open  $O \subseteq \mathbb{R}$  we have

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}\left(\frac{S(n)}{n} \in O\right) \geq - \inf_{x \in O} \Lambda^*(x).$$

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The sequence  $\frac{S(n)}{n}$  satisfies the LDP with rate function  $\Lambda^*$ .

- Can one show that statistics of geometric structures also satisfy a LDP?

## LDP for the TSP

- Notation:  $X(n) := L_{TSP}(\mathcal{P}_1 \cap Q_n)$ .
- The limit of the log m.g.f. for the sequence  $X(n), n \geq 1$ , is

$$\Lambda(t) := \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{E} \exp(tX(n)).$$

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- It is a fact that the limit exists.
- Convex dual of  $\Lambda(t)$  is

$$\Lambda^*(x) := \sup_{t \in \mathbb{R}} (tx - \Lambda(t)).$$

- Now we can state an LDP for  $X(n) := L_{TSP}(\mathcal{P}_1 \cap Q_n)$ .

# V Large Deviation Principles

## Donsker-Varadhan LDP for the TSP

- $X(n) := L_{TSP}(\mathcal{P}_1 \cap Q_n)$  and  $\frac{X(n)}{n} \rightarrow \gamma_{TSP}$  a.s. (Lecture 1)
- $\Lambda(t) := \lim_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{E} \exp(tX(n)) < \infty$  and

(a) For all closed  $F \subseteq \mathbb{R}$  we have

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}\left(\frac{X(n)}{n} \in F\right) \leq - \inf_{x \in F} \Lambda^*(x).$$

(b) For all open  $O \subseteq \mathbb{R}$  we have

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}\left(\frac{X(n)}{n} \in O\right) \geq - \inf_{x \in O} \Lambda^*(x).$$

The sequence  $\frac{X(n)}{n}$  satisfies the LDP with rate function  $\Lambda^*$  and

$$\Lambda^*(x) = 0 \Leftrightarrow x = \gamma_{TSP} \quad \text{and} \quad \Lambda^*(x) \leq C(x - \gamma_{TSP})^2.$$

See Seppalainen + Y.

## LDP for the TSP

### Remark.

$\frac{X(n)}{n}$  converges exponentially to  $\gamma_{TSP}$ , i.e.,  $\forall \epsilon > 0$  there is  $M := M(\epsilon) > 0$  such that  $\forall n$

$$\mathbb{P}\left(\left|\frac{X(n)}{n} - \gamma_{TSP}\right| > \epsilon\right) \leq \exp(-Mn).$$

## Final thoughts: LDP for the TSP

How does one prove an LDP for TSP and other statistics of large geometric structures?

**Key.** The sequence  $X(n), n \geq 1$ , satisfies near additivity in the sense that  $\forall \epsilon > 0, \forall C > 0$ , there is  $n_0$  such that for  $n \geq n_0$  and for all integers  $m$  we have

$$\mathbb{P} \left( \left| X(nm^d) - \sum_{i=1}^{m^d} X_i(n) \right| \geq \epsilon nm^d \right) \leq \exp(-Cnm^d)$$

where  $X_i(n)$  are i.i.d. copies of  $X(n)$ .

Near additivity says that except on a set with exponentially small probability,  $X(nm^d)$  can be expressed as a sum of i.i.d. random variables with a suitably small error.



**THANK YOU**