

Discrete mathematics and algorithms

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Main topics

- Discrete optimization
- Limits of combinatorial structures
- Random discrete structures
- Classification by multifractal analysis techniques

Discrete optimization

- **Frédéric Meunier**: Optimization techniques for bike-sharing systems
- **Éric Colin de Verdière, Frédéric Meunier**: Embedded T -joins

Bike-sharing systems

Abound in optimization problems.

- ★ Station location
- ★ Fleet dimensioning
- ★ Inventory setting
- ★ Rebalancing incentives
- ★ Bike repositioning

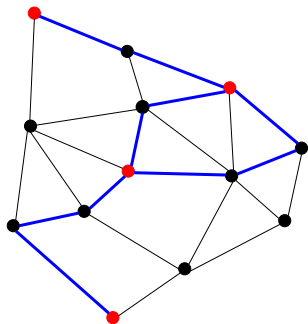
Except the first problem, they all have **static/dynamic** versions.

Contribution:

- Efficient algorithm for the 1-truck **static** repositioning problem
- Survey

Project: Study of the 1-truck **dynamic** repositioning problem

Embedded T -joins



Graph $G = (V, E)$, subset $T \subseteq V$

T -join = subgraph whose odd degree vertices are the elements in T

Useful for various combinatorial optimization problems, e.g., postman tour, max-cut.

Can we compute efficiently optimal T -joins of an embedded graph, with a homological constraint?

Limits of combinatorial structures

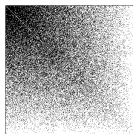
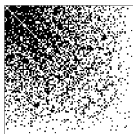
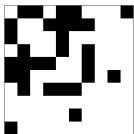
- **Jean-François Delmas**: Statistical approximation of large networks by graphons.
- **Xavier Goaoc, Alfredo Hubard**: Combinatorial limits of order types of finite point sets

Graphons and large networks

Graphon W : symmetric measurable function $[0, 1]^2 \rightarrow [0, 1]$

Random graph $G_n(W)$ sampled from W :

- vertex set: $[n]$
- ij : edge of $G_n(W)$ with probability $W(X_i, X_j)$, where $(X_i: i \in \mathbb{N}^*)$ are uniformly iid on $[0, 1]$.



Convergence and fluctuations

Normalized degree function of W : $D(x) = \int_0^1 W(x, y) dy$

Empirical cumulative distribution function of $G_n(W)$:

$$\Pi_n(y) = \frac{1}{n} \sum_{i=1}^n \mathbf{1}_{\{D_i^{(n)} \leq y\}}$$

where $D_i^{(n)}$ is the normalized degree of vertex i in $G_n(W)$.

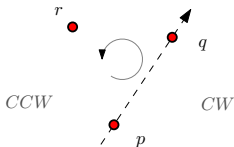
Theorem (Delmas, Dhersin, Sciaudeau 2018)

If D is increasing, we have

- $\sup_y |\Pi_n(y) - D^{-1}(y)| \xrightarrow[n \rightarrow +\infty]{a.s.} 0$ (almost sure convergence)
- $(\sqrt{n}(\Pi_n(y) - D^{-1}(y)))_y \xrightarrow[n \rightarrow +\infty]{(fdd)} \chi$
(convergence of finite-dimensional distributions)
where $\chi = (\chi_y)_y$ is a centered Gaussian process.

From geometry to combinatorics and back

Combinatorial limits (Razborov + Lovász et al.) of order types of finite point sets



Q: Can measures in the plane be to limits of order types what graphons are to dense graphs?

A: NO in general, yet, we observe some rigidity.

Results

Theorem (Goaoc, Hubard, de Joannis de Verclos, Sereni, Volec, 2018+)

Two absolutely continuous measures in the plane induce the same limit of order types if and only if one is the push-forward of the other one by a projective transformation.

Provide first non trivial results on the cases $k = 5, 6, +\infty$ of (Sylvester, Erdős):

$$\inf_{\mu} \mathbb{P}_{\mu}(k : \text{points form a convex polygon})$$

Random discrete structures

Matthieu Fradelizi, Xavier Goaoc, Alfredo Hubbard: Random polytopes

Bárány-Larman theorem

Given a measure μ on \mathbb{R}^d , we define

- ★ a random polytope

$$P_n^\mu = \text{conv}(x_1, \dots, x_n),$$

where (x_i) are uniformly independently drawn from μ ;

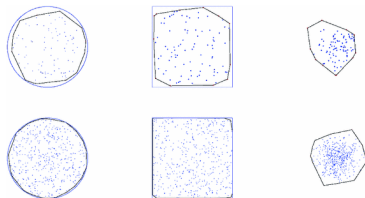
- ★ the **wet parts**

$$W_t^\mu = \{x \in \mathbb{R}^d : \exists \text{ a halfspace } h \ni x \text{ s.t. } \mu(h) \leq t\}.$$

Theorem (Bárány-Larman 1988)

If μ is uniform on a convex body, then

$$\frac{n}{e} \text{Vol}(W_{1/n}^\mu) \leq \mathbb{E}(|V(P_n^\mu)|) \leq 2n \text{Vol}(W_{c/n}^\mu).$$



Random polytopes and wet parts

Ongoing project: to what extent can this be generalized to any measure?

Theorem (Bárány, Fradelizi, Goaoc, Hubard, Rote 2018+)

For any measure μ , we have

$$\frac{n}{e} \operatorname{Vol}(W_{1/n}^{\mu}) \leq \mathbb{E}(|V(P_n^{\mu})|) \leq 2n \operatorname{Vol}(W_{c \log n/n}^{\mu}),$$

and both inequalities are sharp.

(Proof uses ε -nets.)

There are still many questions left:

- ♣ does the B-L bound hold for log-concave measures?
- ♣ which behavior is more typical among integers?

Multifractal analysis

Stéphane Jaffard: Classification by multifractal analysis techniques

Multifractal spectrum

Let $f \in L_{loc}^{\infty}(\mathbb{R}^d)$.

$f \in C^{\alpha}(x_0)$ if there exists a polynomial P such that, for r small enough, $\sup_{B(x_0, r)} |f(x) - P(x - x_0)| \leq Cr^{\alpha}$.

$$h_f(x_0) = \sup\{\alpha : f \in C^{\alpha}(x_0)\}.$$

Multifractal spectrum of f :

$$D_f(H) = \dim_B \{x : h_f(x) = H\},$$

where \dim_B is the fractal dimension.

Multifractal analysis of paintings: Van Gogh challenge

