

Towards connectivities of random geometric complexes

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Joint work with:

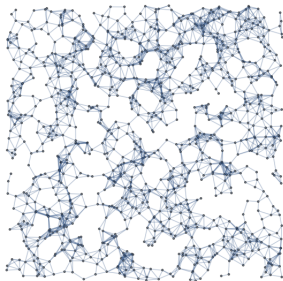
Srikanth K. Iyer, Indian Institute of Science, Bangalore

Leiden, June 2018.

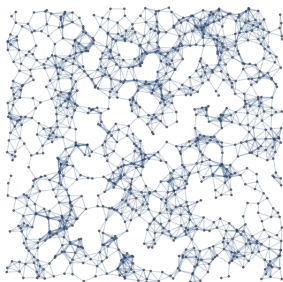


Random geometric graphs

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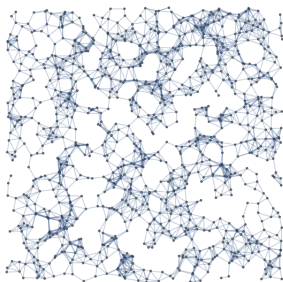


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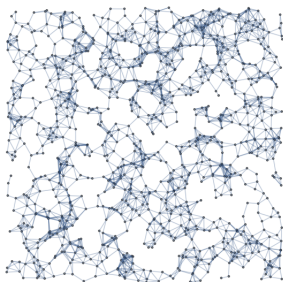
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- ▶ $\mathcal{P}_n = \{X_1, \dots, X_{N_n}\}$ - X_i i.i.d. uniform and $N_n \sim Poi(n)$.
- ▶ **Random geometric graph** : $G(\mathcal{P}_n, r)$: Vertices, $V = \mathcal{P}_n$,
Edges : $x_i \sim x_j$ if $0 < |x_i - x_j| \leq 2r$, $r > 0$.

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- ▶ **Threshold for isolated nodes :** $J_{n,0}$. For $w(n) \rightarrow \infty$,

$$P(J_{n,0} = 0) \rightarrow \begin{cases} 0 & \text{if } n\theta_d 2^d r_n^d = \log n - w(n) \\ 1 & \text{if } n\theta_d 2^d r_n^d = \log n + w(n) \end{cases}$$

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- ▶ G_0^U : Vertices - $F^0 \subset V$. Edges $\subset F^1$ i.e., the usual graph.

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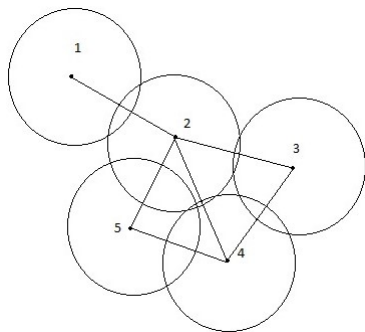
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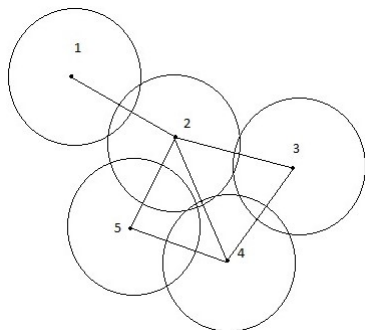
$$C(\mathcal{X}, r) \subset R(\mathcal{X}, r) \subset C(\mathcal{X}, \sqrt{2}r).$$

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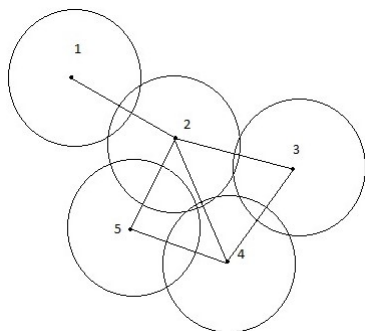


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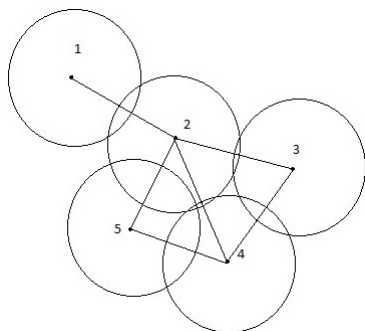
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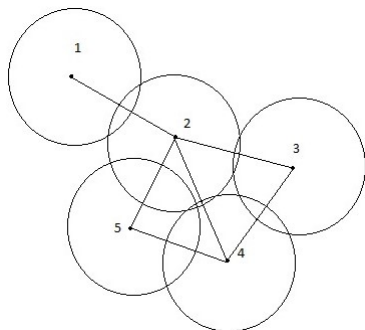
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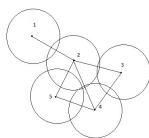


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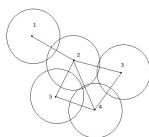


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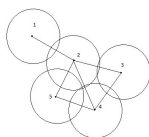


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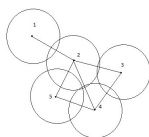


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- ▶ Combinatorial structure matters for up/down connectivity !

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- ▶ Topological data analysis : Carlsson, 2014 ; Adler, 2015 ; Bobrowski-Kahle, 2017.

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- ▶ k -faces $\equiv (k + 1)$ -cliques in RGG.
- ▶ $G_k^*(X_n(r))$: $*$ = U/D i.e., up / down.
 $X_n(r) := X(\mathcal{P}_n, r), X = C/R$ i.e., Čech / Vietoris-Rips.

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- ▶ Also threshold for coverage - ([\[P. Hall, 1988\]](#))

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- ▶ No finite components exist above vanishing threshold !

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- ▶ Why not Connectivity thresholds ? Can there be two infinite components ?

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- ▶ For fixed " $r \in (\log n - \log \log n, \log n)$ ", no isolated edge but infinitely many appear and disappear very quickly !

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- ▶ Rate of $c_n \rightarrow m_k \Rightarrow$ 2nd order term = $k \log \log n$ or $(k - 2) \log \log n$ or $(k - 1) \log \log n$.

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- ▶ Lower bounds involve second-moments and some more Palm calculus.

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- ▶ **Key geometric Lemma:** If $\mathbf{x}, \mathbf{y} \in \mathcal{X}^{(k+1)}$ are isolated k -faces ($(k+1)$ -cliques) in $R(\mathbf{x} \cup \mathbf{y} \cup \mathcal{P}, 1)$ then for some $\epsilon, \beta > 0$, either $|Q(\mathbf{x}) \vee Q(\mathbf{y})| \geq m_k + \epsilon$ or $|Q(\mathbf{x}) \setminus Q(\mathbf{y})| \geq \beta$.

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- ▶ Equivalent to show $\lambda_2(G(\mathcal{P}_n \cap B_1(O), 1)) > 1/2$.

What more ?

- ▶ **Normalized Laplacian:** $L = I - D^{-1/2}AD^{-1/2}$
- ▶ **Garland's method:** Garland '73, Ballman-Świątkowski '97. : If maximal cliques in a graph G are atleast of order $(k + 2)$ and $\lambda_2(Ik_\sigma) > 1 - 1/k$ for all k -cliques σ , then $\beta_{k-1}(R(G)) = 0$.
- ▶ Used by Kahle '14 to show threshold for $R(G(n, p))$ - Erdős-Rényi clique complex.
- ▶ In our case for $k = 1$: To show that w.h.p., for every $X \in \mathcal{P}_n$, $\lambda_2(G(\mathcal{P}_n \cap B_{2r_n}(X), 2r_n)) > 1/2$.
- ▶ Equivalent to show $\lambda_2(G(\mathcal{P}_n \cap B_1(O), 1)) > 1/2$.
- ▶ "Bobrowski-D.Y." $\lambda_2(G(\mathcal{P}_n \cap B_1(O), 1)) \rightarrow 1/2$ a.s. !!!

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