## Limits of permutations and graphs avoiding substructures

Mathilde Bouvel Loria, CNRS and Univ. Lorraine (Nancy, France).

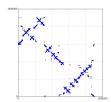
talk based on joint works with Frédérique Bassino, Jacopo Borga, Valentin Féray, Lucas Gerin, Michael Drmota, Mickaël Maazoun, Adeline Pierrot and Benedikt Stufler

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## (1/2)

#### The problem:

- Consider a class C of permutations or graphs defined by the avoidance of substructures (patterns or induced subgraphs).
- For any n, let  $\sigma_n$  or  $G_n$  be an object of size n in C, taken uniformly at random among objects of size n in C.
- We would like to describe the typical global behavior of  $\sigma_n$  or  $G_n$  as n tends to  $\infty$ , through its permuton or graphon limit.



Permutation matrix of a typical large permutation avoiding 2413 and 3142



Adjacency matrix of typical large graph with no induced  $P_4$ 

## (2/2)

#### The proof strategy:

- Permutons and graphons describe global limits of permutations and graphs. But permuton and graphon convergence are characterized by convergence of the densities of substructures.
- Using the substitution or modular decomposition, we can represent permutations or graphs by trees (decorated on their internal nodes).
- Substructures in permutations or graphs correspond to induced subtrees in these trees (subtrees induced by a set of leaves).
- We write functional equations for the generating functions counting decomposition trees, possibly with specified induced subtrees.
- Using analytic combinatorics, we derive the limiting densities of substructures in our permutations or graphs, proving permuton or graphon convergence.

#### A caveat

- Only some classes of permutations or graphs are amenable to the presented strategy:
   when the substitution/modular decomposition is "nice".
- These represent very few cases in the whole landscape of permutation classes/hereditary families of graphs.
- But it still covers quite many classes compared to what was previously known (especially in the permutation case).

The talk will mainly discuss the simplest case in the graph setting: the family of cographs, avoiding an induced  $P_4$  (the path on 4 vertices).

We will also discuss its permutation analogue: the family of separable permutations, avoiding the patterns 2413 and 3142, as well as hint at some generalizations.

# Induced substructures in permutations and graphs,

and a biased view of graphon and permuton convergence

## Induced subgraphs and hereditary families of graphs

• g is an induced subgraph of G when



In words, the subgraph of G = (V, E) induced by  $V' \subset V$  is the graph with vertex set V' and edge set  $E \cap (V' \times V')$ .

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- $\bullet$  A hereditary family of graphs is a set of graphs  $\mathcal C$  such that for every  $G \in C$ , if g is an induced subgraph of G, then  $g \in C$ .
- Examples include the families of cographs, comparability graphs, permutation graphs, circle graphs, parity graphs, ...
- Equivalently, hereditary families of graphs are characterized as sets of graphs whose induced subgraphs avoid a prescribed set (which may be finite or infinite).

## Densities of induced subgraphs

• **Definition:** The density of an induced subgraph g in G is

$$\mathsf{Dens}(g,G) = \mathbb{P}(\mathsf{SubGraph}_k(G) = g)$$

where k is the number of vertices of g and SubGraph $_k(G)$  is the (random) subgraph of G induced by a k-tuple of i.i.d. uniform random vertices of G.

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• Variant: The "injective density" is defined by

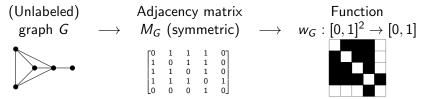
$$\mathsf{Dens}^{inj}(g,G) = \mathbb{P}(\mathsf{SubGraph}_k^{inj}(G) = g),$$

where SubGraph $_k^{inj}(G)$  is the (random) subgraph of G induced by a uniform random k-tuple of distinct vertices of G.

• Fact: For  $G_n$  a sequence of random graphs of size tending to infinity,  $\mathbb{E}[\mathsf{Dens}(g,G_n)] \to \Delta_g$  iff  $\mathbb{E}[\mathsf{Dens}^{inj}(g,G_n)] \to \Delta_g$ .

## What is (informally) a graphon?

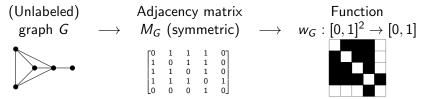
#### In the discrete setting:



The graphon  $W_G$  associated with G is the equivalence class of  $w_G$  under the action of permuting rows and columns of  $M_G$ .

## What is (informally) a graphon?

#### In the discrete setting:



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#### Continuous extension:

In general, a graphon is obtained as above, from a symmetric matrix M, possibly with a continuum of rows and columns, and with values in [0,1].

It is an equivalence class of symmetric functions from  $[0,1]^2 \to [0,1]$  under the action of permuting rows and columns of M.

## Subgraph densities in graphons

Fix g a graph with k vertices, unlabeled.

#### Reminder of the discrete case:

For a graph G,  $\mathsf{Dens}(g,G) = \mathbb{P}(\mathsf{SubGraph}_k(G) = g)$ , where  $\mathsf{SubGraph}_k(G)$  is the (random) subgraph of G induced by a k-tuple of i.i.d. uniform random vertices of G.

#### Continuous generalization:

```
For a graphon W, \mathsf{Dens}(g,W) = \mathbb{P}(\mathsf{Sample}_k(W) = g), where \mathsf{Sample}_k(W) is the (random) graph with k vertices v_1,\ldots,v_k such that v_i and v_j are connected with probability w(x_i,x_j), for x_1,\ldots,x_k i.i.d. uniform random variables in [0,1] and w:[0,1]^2 \to [0,1] a representative of W.
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**Remark**: For any graph G,  $Dens(g, W_G) = Dens(g, G)$ .

## Characterization of (deterministic) graphon convergence

#### (Non-)definition:

The space of graphons is (up to technicalities) metric, for the cut-distance (and in addition is compact).

So, it makes sense to study convergence of a sequence of graphons  $(W_n)_{n\geq 0}$  to a graphon W (for this cut-distance). We write  $W_n\to W$ .

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Typically,  $W_n = W_{G_n}$ , the graphon associated to a graph  $G_n$ , with the sequence of graphs  $(G_n)$  such that the size of  $G_n$  grows to infinity with n. In this case, we also write  $G_n \to W$ .

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#### Combinatorial characterization of convergence:

For  $(W_n)$  a sequence of graphons and W a graphon,  $W_n \to W$  iff for any (finite) graph g,  $\mathsf{Dens}(g,W_n) \to \mathsf{Dens}(g,W)$ .

## Characterization of graphon convergence: the random case

**Reminder**:  $G_n \to W$  iff  $Dens(g, G_n) \to Dens(g, W)$  for all g, for  $(G_n)$  a sequence of (deterministic) graphs and W a (deterministic) graphon.

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What if we take  $(G_n)$  random? (Dens $(g, G_n)$  being then a real r.v.)

Theorem [Diaconis-Janson, 2008]:

The distribution of a random graphon W is characterized by all expected subgraph densities  $\mathbb{E}[\mathsf{Dens}(g, W)]$  (for all g).

Theorem [Diaconis-Janson, 2008]:

Let  $(G_n)$  be a sequence of random graphs. TFAE:

- $G_n$  tends in distribution to some random graphon, W.
- For all g,  $\mathbb{E}[\mathsf{Dens}(g, \mathbf{G}_n)]$  converges to some value  $\Delta_g \in [0, 1]$ .

If this holds, in addition we have:

for all g,  $\mathbb{E}[\mathsf{Dens}(g, \mathbf{W})] = \Delta_g$ , so that  $(\Delta_g)_g$  characterizes  $\mathbf{W}$ .

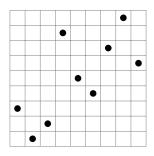
## Summary so far

#### **Graphs:**

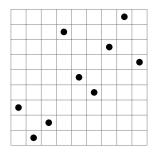
- Definition of induced subgraphs
- Definition of hereditary classes of graphs
- Notion of graphon as a "rescaled adjacency matrix"
- Combinatorial characterization of graphon convergence:
   by the convergence of the densities of induced subgraphs (in expectation in the random case)

#### **Next: permutations**

A permutation  $\sigma = \sigma(1) \dots \sigma(n)$  of size n is a bijection from  $\{1, 2, \dots, n\}$  to itself, which we represent by its permutation matrix, or diagram.

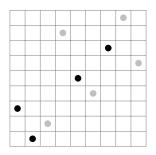


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A permutation  $\pi$  of size k is a pattern of a permutation  $\sigma$  of size n if there exist  $1 \leq i_1 < \ldots < i_k \leq n$  such that  $\sigma(i_1) \ldots \sigma(i_k)$  is in the same relative order  $(\equiv)$  as  $\pi$ .

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Example: 2134 is a pattern of **31**28**5**4**7**96 since  $3157 \equiv 2134$ .

Permutation classes are sets of permutations closed downwards for the pattern partial order relation.

They are equivalently characterized by the avoidance of patterns.

#### Permutons

A permuton is a probability measure on the unit square with uniform marginals (or projections),

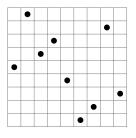
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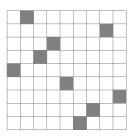
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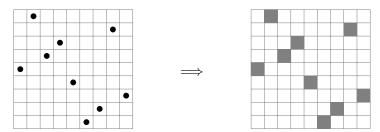


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Informally, permuton can represent permutations of finite size, but also "permutations of infinite size".

## Permuton convergence

We say that a sequence of permutations  $(\sigma_n)$  converges to a permuton  $\mu$  when the sequence of permutons  $(\mu_{\sigma_n})$  converges to  $\mu$  (for the weak convergence of measures).

This extends to sequences of random permutations  $(\sigma_n)$ , converging (in distribution) to a (a priori random) permuton  $\mu$ .

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**Theorem:** Permuton convergence is characterized by the convergence of frequencies of all patterns (in expectation in the random setting).

- $\widetilde{\text{occ}}(\pi, \sigma)$  = frequency of occurrence of the pattern  $\pi$  in  $\sigma$
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- $\widetilde{\operatorname{occ}}(\pi,\mu)$  = frequency of occurrence of the pattern  $\pi$  in  $\mu$
- $(\sigma_n)$  converges in distribution to some permuton  $\mu \Leftrightarrow$  for every pattern  $\pi$ ,  $\mathbb{E}[\widetilde{\text{occ}}(\pi, \sigma_n)]$  converges to some value  $\Delta_{\pi} \in [0, 1]$ .
- ullet If this holds,  $\mathbb{E}[\widetilde{
  m occ}(\pi,\mu)]=\Delta_\pi$  for all  $\pi$  and  $(\Delta_\pi)_\pi$  characterizes  $\mu.$

## Summary so far

#### **Graphs:**

- Definition of induced subgraphs
- Definition of hereditary classes of graphs
- Notion of graphon as a "rescaled adjacency matrix"
- Combinatorial characterization of graphon convergence:
   by the convergence of the densities of induced subgraphs (in expectation in the random case)

#### Permutations:

- Definition of patterns
- Definition of permutation classes
- Notion of permuton as a "rescaled diagram"
- Combinatorial characterization of permuton convergence: by the convergence of the frequencies of patterns (in expectation in the random case)

## **Decomposition trees**

## Modular decomposition of graphs

• A module in a graph G = (V, E) is a set  $S \subseteq V$  of vertices which cannot be distinguished by vertices outside of S:

for every 
$$v \in V \setminus S$$
, either  $\{v, s\} \in E$  for all  $s \in S$  or  $\{v, s\} \notin E$  for all  $s \in S$ 

- ullet Given a partition of V into modules, G can be described
  - the subgraph induced keeping exactly one vertex in each module (sometimes called quotient)
  - the subgraph induced by each module (sometimes called factors)



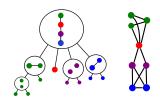


 Repeating this construction inside the modules, we obtain a modular decomposition tree of G (which is rooted, non planar, with internal vertices labeled by quotient graphs, and leaves corresponding to V).

### Modular decomposition trees

- The trivial modules of G are  $\emptyset$ , V, and  $\{v\}$  for any  $v \in V$ .
- A graph *G* is prime if it contains no non-trivial module.

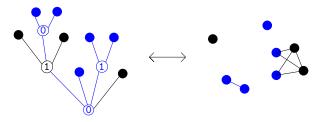
**Theorem:** Every graph has a unique modular decomposition tree whose vertices are either cliques (denoted 1), or independent sets (denoted 0), or prime graphs (denoted P), and with no 0-0 nor 1-1 edges. We call it canonical and denote it  $\mathcal{T}(G)$ .



T(G) is obtained considering recursively the quotients resulting from the partition of V into maximal modules different from V (in the prime case, with special cases for cliques and independent sets).

### Induced subgraphs in decomposition trees

- Let G be a graph. Let S be a subset of its vertices.
- Consider the subgraph of G induced by S.
- Let T(G) be its canonical modular decomposition tree.



**Fact:** A decomposition tree for the induced subgraph of G corresponding to S is obtained considering the subtree of T(G) induced by the set of leaves corresponding to S.

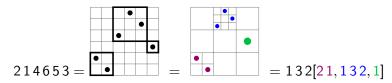
**Remark:** The induced tree is not necessarily the canonical tree of the induced subgraph (e.g. it may contain 0-0 edges).

### Substitution decomposition of permutations

Analogue of module: An interval of a permutation  $\sigma$  is an interval [i,j] of integers such that  $\{\sigma(i),\ldots,\sigma(j)\}$  is also an interval.



**Quotient and factors:** Given a partition of the points of  $\sigma$  into intervals, we can describe  $\sigma$  as a quotient permutation whose points have been inflated by factor permutations.



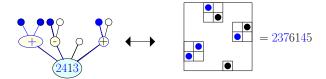
Analogue of prime graph: A simple permutation is a permutation  $\sigma$  whose only intervals are the trivial ones (those of size 0, 1, and  $|\sigma|$ ).

### Substitution decomposition trees, and patterns

**Theorem:** Every permutation admits a unique (substitution) decomposition tree whose internal vertices are labeled by  $\oplus$  (for increasing permutations),  $\ominus$  (for decreasing permutations) and simple permutations, with no  $\oplus - \oplus$  nor  $\ominus - \ominus$  edges. We call it canonical and denote it  $T(\sigma)$ .

#### Remarks:

- The leaves of  $T(\sigma)$  correspond to the elements of  $\sigma$ .
- Patterns in  $\sigma$  correspond to subtrees of  $T(\sigma)$  induced by leaves (although not giving canonical trees in general).

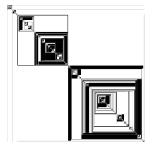


### A summary highlighting similarities and differences

- Using substitution/modular decomposition, permutations and graphs can be encoded by "decorated" trees.
- However, the trees associated with permutations are planar, while those associated with graphs are non-planar.
- Patterns/induced subgraphs correspond to subtrees induced by leaves.
- For a permutation  $\sigma$ , consider its inversion graph G. Up to planarity and adapting the decorations, it holds that  $T(\sigma) = T(G)$ .
- However, the number of permutations realizing a certain (permutation) graph G varies; as a consequence, the uniform distribution on a set of permutations does not translate to the uniform distribution on the set of corresponding graphs.

**Next:** How to use these trees to prove permuton/graphon convergence for uniform permutations/graphs avoiding substructures.

## **Graphon limit of (labeled) cographs**



### Cographs and their modular decomposition trees

- Cographs are defined by the avoidance of  $P_4$  (induced path on 4 vertices, which is the smallest prime graph).
- Equivalently, cographs are all graphs whose modular decomposition trees involve only 0 (indep. set) and 1 (clique) nodes (no prime node). We call cotrees their modular decomposition trees.
- Therefore, labeled<sup>1</sup> cographs can be described from the combinatorial specification:

Indeed, via its canonical modular decomposition tree, a cograph correspond to a tree of  $\mathcal{L}$  with a label 0 or 1 on the root (propagating labels alternating between 0 and 1 along each edge).

 $<sup>^{1}</sup>$ meaning vertices are labeled by the integers from 1 to n; in the unlabeled case, we need to consider Multiset, and hence later Polya operators for the GF

## Expressing $\mathbb{E}[\mathsf{Dens}^{inj}(g, \boldsymbol{G}_n)]$

#### Notation:

Let  $G_n$  be a uniform random labeled cograph with n vertices.

#### Reminder:

Knowing  $\mathbb{E}[\mathsf{Dens}^{inj}(g, \mathbf{G}_n)]$  for all g characterizes the graphon limit of  $\mathbf{G}_n$ .

#### Notation:

for all n, and all  $k \leq n$ ,

 $\mathbf{t}^{(n)}$  is a uniform random labeled canonical cotree of size n, and  $\mathbf{t}^{(n)}$  is the subtree of  $\mathbf{t}^{(n)}$  induced by a uniform k tuple of distinct k

 $\boldsymbol{t}_k^{(n)}$  is the subtree of  $\boldsymbol{t}^{(n)}$  induced by a uniform k-tuple of distinct leaves.

#### Observation:

For any cograph g, we have:

$$\mathbb{E}[\mathsf{Dens}^{inj}(g, \mathbf{G}_n)] = \mathbb{P}(\mathsf{SubGraph}_k^{inj}(\mathbf{G}_n) = g) = \sum \mathbb{P}(\mathbf{t}_k^{(n)} = t_0),$$

where the sum runs over all cotrees  $t_0$  corresponding to g.

# Expressing $\mathbb{P}(\boldsymbol{t}_k^{(n)}=t_0)$

**Observation**: 
$$\mathbb{P}(\boldsymbol{t}_{k}^{(n)} = t_0) = \frac{n![z^n]M_{t_0}(z)}{n![z^n]M(z) \times n(n-1)...(n-k+1)}$$
, where

- $\bullet$   $\mathcal{M}$  is the set of labeled canonical cotrees
- for any cotree  $t_0$  with k leaves,  $\mathcal{M}_{t_0}$  is the set of labeled canonical cotrees with a marked k-tuple of distinct leaves, which induce  $t_0$ ,
- M(z) and  $M_{t_0}(z)$  are the corresponding exponential generating functions
- as usual,  $[z^n]F(z)$  denotes the coefficient of  $z^n$  in the generating function F(z)

**Next:** Use symbolic and analytic combinatorics to compute the asymptotic behavior of the numerator and the denominator in

$$\frac{n![z^n]M_{t_0}(z)}{n![z^n]M(z)\times n(n-1)\dots(n-k+1)}.$$

### Estimating the denominator

- Recall that  $\mathcal{L} = \bullet \ \uplus \ \mathsf{Set}_{\geq 2}(\mathcal{L})$ .
- Let L(z) be the exponential generating function of  $\mathcal{L}$ .
- From [Flajolet-Sedgewick] (rather a variant on trees counted by leaves), L(z) satisfies  $L(z) = z + \exp(L(z)) 1 L(z)$ .
- The generating function of cographs is M(z) = 2L(z) z.
- L(z) and M(z) have the same radius of convergence  $\rho = 2\log(2) 1$  and are  $\Delta$ -analytic.
- Near  $z = \rho$ ,  $L(z) = \log(2) \sqrt{\rho}\sqrt{1 z/\rho} + \mathcal{O}(1 z/\rho)$ and  $M(z) = 1 - 2\sqrt{\rho}\sqrt{1 - z/\rho} + \mathcal{O}(1 - z/\rho)$ .
- From the transfer theorem,  $n(n-1)\dots(n-k+1)[z^n]M(z) \underset{n\to+\infty}{\sim} \frac{n^{k-3/2}}{\rho^{n-1/2}\sqrt{\pi}}.$

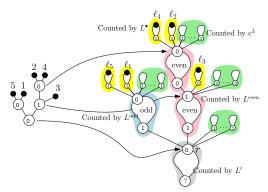
### Estimating the numerator

**Prop.**: If  $t_0$  with k leaves has  $n_v$  internal vertices,  $n_{=}$  edges of the form 0-0 or 1-1, and  $n_{\neq}$  edges of the form 0-1 or 1-0, then

$$M_{t_0} = (L')(\exp(L))^{n_v} (L^{\bullet})^k (L^{\operatorname{odd}})^{n_{=}} (L^{\operatorname{even}})^{n_{\neq}},$$

these series being variations on L(z) whose singular behavior results from that of L(z).

#### Proof:



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Corollary: Like before, we obtain

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Proof:



Corollary: Like before, we obtain

- the behavior at  $\rho$  of  $M_{t_0}(z)$ ,
- and the asymptotic estimate of  $[z^n]M_{t_0}(z)$ .

More precisely, we have

$$[z^n]M_{t_0}(z) \underset{n \to +\infty}{\sim} \frac{(k-1)!}{(2k-2)!} \frac{n^{k-3/2}}{\rho^{n-1/2}\sqrt{\pi}},$$

if  $t_0$  is binary (which implies  $n_v = k - 1$  and  $n_= + n_{\neq} = k - 2$ ).

### Conclusion of the proof

#### Notation (reminder):

- $t^{(n)}$ : uniform random labeled canonical cotree of size n
- $t_k^{(n)}$ : subtree of  $t^{(n)}$  induced by a uniform k-tuple of distinct leaves
- $t_0$ : cotree with k leaves

What we proved: If  $t_0$  is binary, then  $\lim_{n\to\infty} \mathbb{P}(\boldsymbol{t}_k^{(n)}=t_0)=\frac{(k-1)!}{(2k-2)!}$ .

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#### Remark/reminder:

Summing over all  $t_0$  encoding a cograph g, this gives  $\lim_{n\to\infty} \mathbb{E}[\mathsf{Dens}(g, \mathbf{G}_n)]$ . These quantities characterize the graphon limit of cographs.

### The Brownian cographon

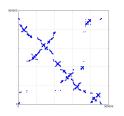
- Starting from a Brownian excursion, whose local minima receive unbiased decorations by 0 and 1, we can build the Brownian cographon of parameter 1/2, denoted  $\mathbf{W}^{1/2}$ .
- ullet We can compute  $\Delta_g = \mathbb{E}[\mathsf{Dens}(g, oldsymbol{W}^{1/2})]$  for all cographs g.

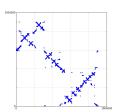
But this is a story for another time...

- We observe that  $\lim_{n\to\infty}\mathbb{E}[\mathsf{Dens}(g,\textbf{\textit{G}}_n)]=\Delta_g$  for all g.
- Therefore, not only do we prove existence of a graphon limit for uniform random cographs, but we also provide a construction of this limit.
- The limiting graphon is a genuinely random and fractal object.



## Permuton limit of separable permutations





### Decomposition trees of separable permutations

- Separable permutations are those avoiding the patterns 2413 and 3142 (the two smallest simple permutations).
- Equivalently, it is the family of all permutations whose decomposition trees involve only ⊕ and ⊖ nodes (no simple permutations).
- Separable permutations are therefore the permutation analogue of cographs.
- From a combinatorial specification for the decomposition trees of separable permutations, and using analytic combinatorics as before, we obtain the limiting behavior of  $\mathbb{E}[\widetilde{occ}(\pi,\sigma_n)]$  for  $\pi$  any pattern, and  $\sigma_n$  a uniform random separable permutation of size n. These quantities characterize the permuton limit of separable permutations.
- Again, we have an explicit construction of the limiting object  $\mu^{1/2}$  (called the Brownian separable permuton of parameter 1/2) from a Brownian excursion with decorations.

# More classes of permutations

### Transposing the proof strategy to a more general setting

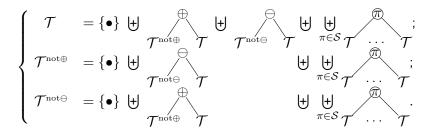
#### Idea of the method:

- Assume that you know a combinatorial specification for the decomposition trees of permutations in some class C.
- It translates into a system of equations for the GF of C.
- We can in addition "track patterns" in these equations.
- IF the method of analytic combinatorics goes through, we obtain convergence to a certain permuton, as for separable permutations.

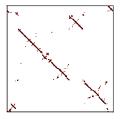
#### Some results:

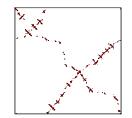
- Convergence to Brownian separable permutons of parameters  $p \in [0,1]$  for substitution-closed classes, under some analytic condition on the GF of the simple permutations in the class.
- Dichotomy for classes for which a specification is known (in particular: whenever they contain finitely many simple permutations): (random) Brownian permutons VS (deterministic) X-permutons.

ullet Their specification adds some simple permutations to that of separable permutations. We denote by  ${\cal S}$  the set of allowed simple permutations.



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- Limit permutons are (biased) Brownian separable permutons.



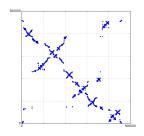


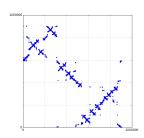


simulations of  $\mu_p$  for p = 0.2, 0.45, 0.5.

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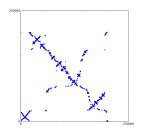
Example 1: Separable permutations, *i.e.*  $S = \emptyset$ ,  $\Rightarrow p = 0.5$ 

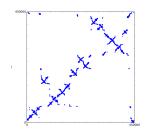




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Example 2: 
$$S = \{2413, 3142, 24153\}, \Rightarrow p = 0.5$$





- Their specification adds some simple permutations to that of separable permutations. We denote by  $\mathcal S$  the set of allowed simple permutations.
- Limit permutons are (biased) Brownian separable permutons.

Example 3: 
$$S = Av(321) \cap \{Simples\}$$
 (infinite),  $\Rightarrow p \in [0.577, 0.622]$ 



### Brownian case of the dichotomy

When the specification contains a product of critical families.

 $\Rightarrow$  The limiting permuton of the class is a biased Brownian separable permuton (of parameter p possibly 0 or 1).

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Example 1: Av(132), with critical families in blue.

The limit is the Brownian separable permuton of parameter p = 0.

### Brownian case of the dichotomy

When the specification contains a product of critical families.

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Example 2: Av(2413, 31452, 41253, 41352, 531246), with critical families in blue.

$$\begin{cases} \mathcal{T}_0 = & \{ \bullet \} \ \, \uplus \ \, \oplus [\mathcal{T}_1, \mathcal{T}_0] \ \, \uplus \ \, \ominus [\mathcal{T}_2, \mathcal{T}_0] \ \, \uplus \ \, 3142[\mathcal{T}_0, \mathcal{T}_3, \mathcal{T}_3, \mathcal{T}_0] \\ \mathcal{T}_1 = & \{ \bullet \} \ \, \uplus \ \, \ominus [\mathcal{T}_2, \mathcal{T}_0] \ \, \uplus \ \, 3142[\mathcal{T}_0, \mathcal{T}_3, \mathcal{T}_3, \mathcal{T}_0] \\ \mathcal{T}_2 = & \{ \bullet \} \ \, \uplus \ \, \ominus [\mathcal{T}_1, \mathcal{T}_0] \ \, \uplus \ \, 3142[\mathcal{T}_0, \mathcal{T}_3, \mathcal{T}_3, \mathcal{T}_0] \\ \mathcal{T}_3 = & \{ \bullet \} \ \, \uplus \ \, \ominus [\mathcal{T}_4, \mathcal{T}_3] \\ \mathcal{T}_4 = & \{ \bullet \} \end{cases}$$

The limit is the Brownian separable permuton of parameter  $p \approx 0.4748692376...$  (only real root of a certain polynomial of degree 9).

When the specification contains no product of critical families.

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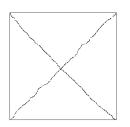
Example 1: Av(2413, 3142, 2143, 3412), a.k.a. the X-class, with critical families in blue.

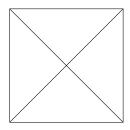
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 \begin{cases} \mathcal{T}_0 = & \{ \bullet \} \uplus \oplus [\mathcal{T}_1, \mathcal{T}_2] \uplus \oplus [\mathcal{T}_1, \mathcal{T}_3] \uplus \oplus [\mathcal{T}_4, \mathcal{T}_2] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_5] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_6] \uplus \ominus [\mathcal{T}_7, \mathcal{T}_5] \\ \mathcal{T}_1 = & \{ \bullet \} \\ \mathcal{T}_2 = & \{ \bullet \} \uplus \oplus [\mathcal{T}_1, \mathcal{T}_2] \\ \mathcal{T}_3 = & \oplus [\mathcal{T}_1, \mathcal{T}_3] \uplus \oplus [\mathcal{T}_4, \mathcal{T}_2] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_5] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_6] \uplus \ominus [\mathcal{T}_7, \mathcal{T}_5] \\ \mathcal{T}_4 = & \ominus [\mathcal{T}_1, \mathcal{T}_5] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_6] \uplus \ominus [\mathcal{T}_7, \mathcal{T}_5] \\ \mathcal{T}_5 = & \{ \bullet \} \uplus \ominus [\mathcal{T}_1, \mathcal{T}_5] \\ \mathcal{T}_6 = & \oplus [\mathcal{T}_1, \mathcal{T}_2] \uplus \oplus [\mathcal{T}_1, \mathcal{T}_3] \uplus \oplus [\mathcal{T}_4, \mathcal{T}_2] \uplus \ominus [\mathcal{T}_1, \mathcal{T}_6] \uplus \ominus [\mathcal{T}_7, \mathcal{T}_5] \\ \mathcal{T}_7 = & \oplus [\mathcal{T}_1, \mathcal{T}_2] \uplus \oplus [\mathcal{T}_1, \mathcal{T}_3] \uplus \oplus [\mathcal{T}_4, \mathcal{T}_2]. \end{cases}
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The limit is the centered X-permuton.

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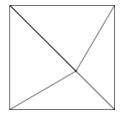
Example 2: Av(2413, 3142, 2143, 34512), with critical families in blue.

When the specification contains no product of critical families.

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Example 2: Av(2413, 3142, 2143, 34512), with critical families in blue.





The limit is a non-centered X-permuton.

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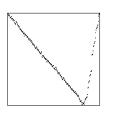
Example 3: Av(2413, 1243, 2341, 531642, 41352), with critical families in blue.

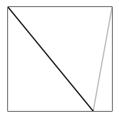
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The limit is a degenerate X-permuton.

# **Concluding remarks**

### Towards more classes of graphs

- As in the permutation case, we can extend the study of cographs to families of graphs whose modular decomposition trees are described by a combinatorial specification.
- Our analytic approach can only work with GF having positive radius of convergence. This was necessary in permutation classes, but is an additional requirement for hereditary graph classes.
- This is the PhD topic of Théo Lenoir, who started working in September 2021, supervised by Frédérique Bassino and Lucas Gerin.
- The classes that he studied are the P<sub>4</sub>-blah graphs, where blah ∈ {reducible, sparse, lite, extensible, tidy}.
   All converge to the Brownian cograpon.
- Recall that cographs are  $P_4$ -free graphs.

### An alternative proof strategy

Consider a class of graphs/permutations C, seen as trees.

For any n, let  $G_n$  be a uniform random object of size n in C.

**Goal:** Describe the graphon/permuton limit of  $G_n$  as  $n \to \infty$ .

#### The strategy I presented (in part):

- Build the limiting graphon/permuton from a Brownian excursion
- Compute the densities  $(\Delta_g)_g$  of substructures in it.
- $(\Delta_g)_g$  characterizes the graphon/permuton limit
- Use a combinatorial specification for  $\mathcal C$  and analytic combinatorics to compute, for any g, the limiting behavior for  $n \to \infty$  of the density of g in  $G_n$ . Observe that it coincides with  $\Delta_g$ .

#### The "random trees" strategy

• Compute densities of induced subtrees directly on the random tree of  $G_n$ , using techniques from the random trees literature.

### A nice consequence of permuton/graphon limits

#### Results:

- The size of the largest independent set of a uniform random cograph is sublinear.
  - (hence  $P_4$  does not have the asymptotic linear Erdős-Hajnal property, answering a question of Kang, McDiarmid, Reed and Scott in 2014)
- The length of the longest increasing subsequence of a uniform random separable permutations is sublinear.

#### Main proof ingredients:

- Convergence to the Brownian cographon
- ullet The independence number of the Brownian cographon  $oldsymbol{W}^{1/2}$  is 0

**Bonus**: The sublinearity result applies to all classes with graphon/permuton limit  $W^p$  or a Brownian separable permuton.

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Thank you for being there!