

# Convergence and limits of finite trees

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# limit of (dense) graphs

Lovász, Szegedy at al.

- random  $k$ -sample of vertices of a fixed graph  $G$

↓ induces

- random graph  $R$  on  $k$  (labeled) vertices

↓

- subgraph densities:  $d(K, G) = \Pr(R=K)$

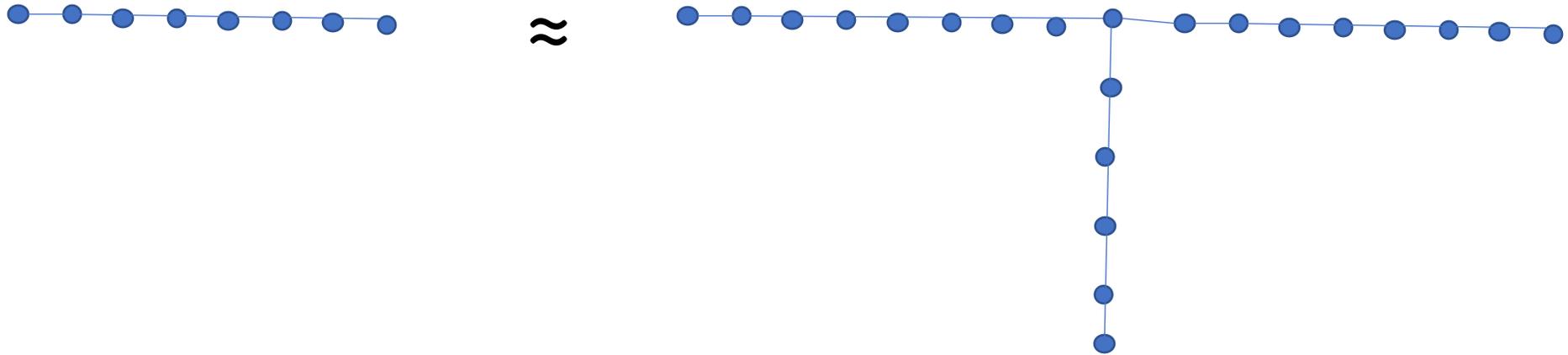
- $(G_i)$  is a convergent sequence if  $d(K, G_i)$  is convergent for every  $K$

What is the limit? Need  $G_i \rightarrow G^*$  with  $d(K, G_i) \rightarrow d(K, G^*)$

Graphons:  $G^*: [0,1]^2 \rightarrow [0,1]$  symmetric, measurable (not unique)

# limit of trees

- trees are sparse  $\rightarrow$  random samples induce the empty graph
- sparse graph limits (**Benjamini-Schramm**):  
small neighborhood of a random point  
too **local**



- we want **global** limits instead

# limit of trees

- trees are sparse  $\rightarrow$  random samples induce the empty graph
- consider them as a dense structure:

graph distance

need: normalization  
here: with diameter

- random  $k$ -sample of vertices of a fixed tree  $T$

distances

- random  $k$  by  $k$  real matrix  $M_k$  – distributed as  $\tau_k(T)$

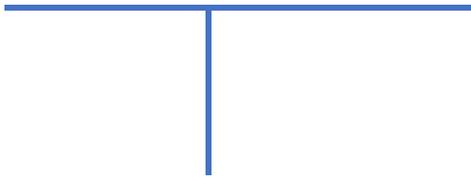
- convergence in distribution

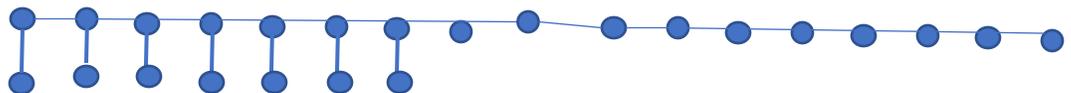
$(T_i)$  is a convergent sequence if  $\tau_k(T_i)$  weakly converges for all  $k$

# limit of trees

intuitive examples

- paths  $(P_n)$  — clearly convergent  
limit: unit interval:  uniform  
for sampling we need **distance** and **distribution** on it

- 3 equal paths from common root — also convergent  
limit: 3 intervals:   
**distance**: each has length  $\frac{1}{2}$   
**distribution**: uniform

- combs  — convergent

limit: unit interval:   
**distribution**: not uniform

- stars  $(K_{1,n})$  — clearly convergent  
almost every distance = diameter

many equivalent  
definitions

# measured real trees

- Real tree = complete metric space  $(X, d)$ 
  - +  
 $\forall x, y \exists$  isometric embedding of length  $d(x, y)$  interval connecting them
  - +  
 $\forall$  intermediate point of embedding separates  $x$  from  $y$
- Measured real tree = real tree + probability Borel measure  
(a metric measure space)
- sampling makes sense,  
so does the distribution  $\tau_k(T)$  for a measured real tree  $T$
- $(T_i) \rightarrow T$  iff  $\forall k: \tau_k(T_i) \rightarrow \tau_k(T)$

# Ideally we had

1. Each convergent sequence of finite trees **has a limit measured real tree**.
2. **Every measured real tree** (of diameter  $\leq 1$ ) is a limit of finite trees.
3. The limit is **unique** up to isomorphism.

## PROBLEMS

- Measure 0 branch of a real tree has **no effect** on sampling  
Let's forbid measure 0 branches.

makes all such  
real trees  
separable

- What about the sequence of stars??

**do not converge to a separable real tree**

Let's add "distance" parameter from tree to sample  
the distribution is on  $T \times [0, \infty)$

$$d((x,a),(y,b)) = d(x,y) + a + b$$

for stars:  $T = \{p\}$ , the distribution concentrates on  $(p, 1/2)$

# The limit object: dendron

- Dendron:  $D = (T, \mu)$ , where
  - $T = (T, d)$  is a (separable) real tree
  - $\mu$  is a probability Borel measure on  $T \times [0, \infty)$
  - such that  $\mu(B \times [0, \infty)) > 0$  for any branch  $B$  of  $T$
- Sampling:  $k$  i.i.d.  $(x_i, a_i) \sim \mu$
- Distance matrix  $M_k = (d_{ij})$  with  $d_{ij} = d(x_i, x_j) + a_i + a_j$ ,  $d_{ii} = 0$
- $\tau_k(D)$  is the distribution of  $M_k$
- We write  $(T_i) \rightarrow D$  iff  $\forall k: \tau_k(T_i) \rightarrow \tau_k(D)$

a marked metric  
measure space

Depperschmidt, Greven,  
Pfaffelhuber

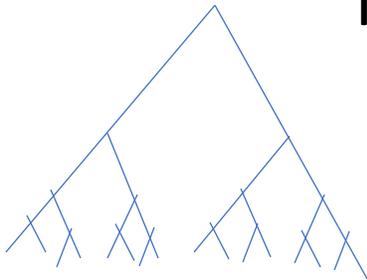
# Results

1. Each convergent sequence of finite trees **has a limit dendron**.
2. **Every dendron** (of essential diameter  $\leq 1$ ) is a limit of finite trees.
3. The limit is **unique** up to isomorphism.
4. The space of dendrons of essential diameter  $\leq 1$  is compact.

# More examples

Svante Janson

- Stars  $(K_{1,i}) \rightarrow (T, \mu)$ , where  $T = \{p\}$ ,  $\mu$  concentrates on  $(p, 1/2)$
- complete binary tree of depth  $i$ : same limit as for stars
- comp. binary tree with edges on level  $j$  replaced by paths of length  $2^{i-j}$ :  
limit = fractal tree with uniform distribution on Cantor set



- many paths from a single high degree vertex  
(every single path has  $o(n)$  vertices)  
limit = single point tree, but tricky distribution on  $[0, 1/2]$

# Random tree limits

Svante Janson

- Conditioned critical Galton-Watson trees (finite variance)

↓ in distribution  
with slightly different scaling

Aldous's Continuum Random Tree

↑ in distribution  
with slightly different scaling

- Random tree on  $n$  labelled vertices
- Conditioned subcritical Galton-Watson trees (+mild conditions)

↓ in distribution  
no scaling

$D=(T,\mu)$ , with  $T=\{p\}$ ,  $\mu =$  geometric distribution

# Sampling dense graphs versus trees

## An important difference

- Let  $G_1$  and  $G_2$  be two independent samples from the random graph  $G(n, 1/2)$ .  
 $G_1$  and  $G_2$  are very **close in sampling**: a random  $k$ -sample is close to  $G(k, 1/2)$   
(small cut distance too)  
 $G_1$  and  $G_2$  are very **far in edit distance**  
(need to add/delete cca  $n^2/4$  edges to make  $G_1$  isomorphic to  $G_2$ )
- If  $T_1$  and  $T_2$  are two  $n$  vertex trees that are **close in sampling** distance,  
then there exists a bijection between the vertex sets with **small distortion**  
in the graph distance

# The proof technique ultraproducts

See also:

proof of the **Hypergraph Regularity Lemma** by **Elek and Szegedy**

Advantages:

- Very “clean” finite-infinite transitions
- Conceptually straight forward
- Compact proofs of qualitative statements (like existence, unicity, etc.)

Disadvantages:

- Technical details
- Absolutely no quantitative bounds

# The proof technique ultraproducts

- Fix ultrafilter  $\omega$  on the positive integers: the family of **LARGE** sets
- ultralimit:  $\lim_{\omega} x_i = x$  if every neighborhood of  $x$  contains a **LARGE** subsequence of  $x_i$ .  
**ALWAYS** exists and unique in a compact Hausdorff space
- ultraproduct:  $\prod_{\omega} A_i$ : combines a sequence of structures  $A_i$ , keeps properties held by a **LARGE** subsequence

$A_i$  can be

- sets
- metric spaces, **van den Dries, Wilkie**
- probability measure spaces, **Elek, Szegedy**
- (marked) metric measure spaces, etc.

## Proof sketch

# Existence

- $T_i$  are finite trees, therefore also measured real trees
- $T^* := \prod_{\omega} T_i$ : also a real tree, but ugly (not separable)  
also a probability measure space but ugly (not a Borel measure)
- “Prune”  $T^*$ : remove branches with measure 0.  
 $T := \text{core of } T^* = \text{what remains after pruning}$
- $T$  is nice (separable)
- Some part of the probability distribution may be outside  $T$
- Replace  $x \in T^*$  with  $(p, a) \in T \times [0, \infty)$ , where  $p = \text{projection of } x \text{ to } T$   
 $a = \text{distance of } x \text{ from } T$
- Dist. on  $T^*$  is transformed to  $\mu$  on  $T \times [0, \infty)$
- $D = (T, \mu)$  is the ultralimit dendron
- $\lim_{\omega} \tau_k(T_i) = \tau_k(D)$
- If  $(T_i)$  is convergent, then  $\tau_k(T_i) \rightarrow \tau_k(D)$ ,  $T_i \rightarrow D$ , Q.E.D.

Proof sketch

# Uniqueness

- $T$  is a measured real tree (works for dendrons too)
- $(x_1, \dots, x_k)$  is a  $k$ -sample from  $T$ ,  
 $T_k$  is subtree spanned by these points.
- $d(x_i, x_j)$  is enough to reconstruct  $T_k$  upto isomorphism.
- The core of  $\prod_{\omega} T_k$  is isomorphic to  $T$  almost surely.
- $T$  is reconstructed from  $\tau_k(T)$ , therefore the limit is unique.      Q.E.D.

Proof sketch

# Finite trees are dense among dendrons (of essential diameter $\leq 1$ )

- $T$  is a measured real tree (works for dendrons too)
- We saw  $T_k \rightarrow T$ , where  $T_k$  is a real tree spanned by  $k$  points
- So finitely spanned real trees are dense
- It is easy to approximate a finitely spanned real tree with finite trees.  
(if the essential diameter is  $\leq 1$ )

Q.E.D.

# Missing numeric results

- Trees close in sampling  $\Rightarrow \exists$  vertex-bijection with small distortion
- A  $k$ -sample is close in sampling to the full tree with high probability

THANK  
YOU