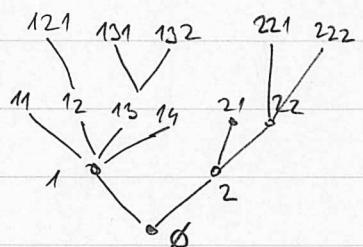


## Shape of large vertical G.W. trees.

We now understand sizes of clusters in ethical ER graphs.

To understand what is the typical slope of them we need to come back to Gw trees.

We start by introducing another ways how to encode finite tree by a stochastic process. We start with rooted, ordered trees and label them according to their genealogy as on the figure

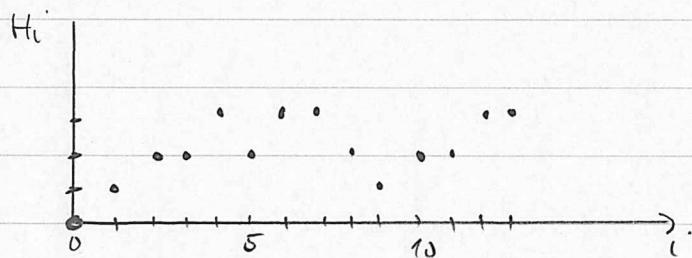


(4.51) Figure: See middle section labelled according to analogy.

Height function: Assume the tree has  $n$  vertices. List them as  $v_0, \dots, v_{n-1}$  in lexicographical order (w.r.t. their labels).

$$(4.52) \quad \text{Set} \quad H(i) = \text{dist}(v_i, \emptyset), \quad i \in \{0, \dots, n-1\}.$$

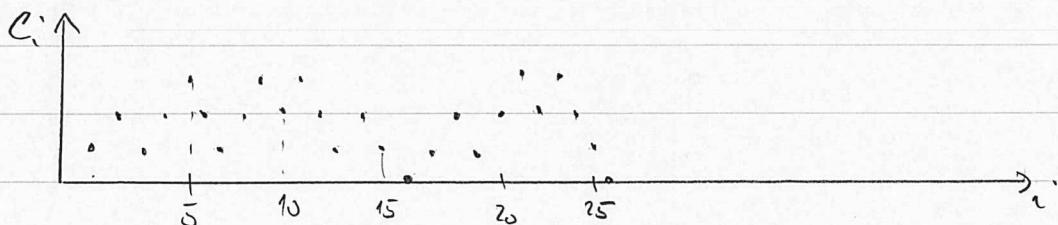
(4.53) Exercise: Show that the tree can be recovered from its height function.



(4.54) Figure: The height function of the tree from (4.51)

Contour function: Trace the "contour" of the tree from left to right, so that we pass every edge lower and record the distance from the root. We get

$$C(i) : i \in \{0, \dots, 2(n-1)\}.$$



(4.55) Figure: The contour function of (4.51)

(4.56) Exercise (easy): Show that the contour function determines the tree.

### Diskrete fixe Wahl (diskret fixe Wahl)

Let  $v_0, \dots, v_{m-1}$  be as above, and set  $X_i = \# \text{children of } v_i$ . Set

$$S(0) = 0, \quad S(i) = \sum_{j=0}^{i-1} (X_j - 1) \quad 1 \leq i \leq m.$$

This is the replacement path  $\vec{\phi}$  considered previously.

It is now clear that  $S$  encodes the shape of the tree. We show

(4.57) Lemma: For every fixed tree

$$H(i) = \#\left\{0 \leq j \leq i-1 : S(j) = \min_{j \leq k \leq i} S(k)\right\}.$$

Proof: By the same arguments as on page 10, for any subtree of the original tree, the value of  $X$  once we have finished replacing it is one less than its value when we visited the root of this subtree, whereas within the subtree  $S$  has at least value of its root. Now  $\text{desc}(v_i, \emptyset)$  is equal to the number of subtrees we began but not completed replacing before reaching  $v_i$ . The roots of these subtrees are leaves  $j < i$  with  $S(j) = \min\{S(k) : j \leq k \leq i\}$

]

Recall that if  $T$  is a GW tree, then  $S$  is a random walk stopped at hitting  $(-1)$ . Sometimes it is technically easier to deal with a sequence of iid GW trees and concatenate their height processes, use dyadic-fine walls. As on p.39, the first tree ends when  $S$  hits  $(-1)$ , the second when it hits  $(-2)$ , etc.

(4.58) Exercise: Check that (4.57) holds also in this case!

We now explore critical GW trees. We assume that

$$(4.59) \quad \mathbb{E} X_i = 1 \text{ and } \text{Var } X_i = \sigma^2 < \infty$$

It is then obvious that

(4.60) Proposition (convergence of the dyadic fine walk). Under (4.59)  

$$\left( \frac{1}{\sqrt{n}} S(L_{n+1}) : t \geq 0 \right) \xrightarrow{n \rightarrow \infty} (\pm W(t) : t \geq 0)$$

where  $W$  is a standard Brownian motion.

(4.61) It is much more delicate to show that

(4.62) Theorem: (convergence of the height function) Under (4.59)

$$\left( \frac{1}{\sqrt{n}} H(L_{n+1}), t \geq 0 \right) \xrightarrow{n \rightarrow \infty} \left( \frac{2}{\sqrt{\pi}} B(t), t \geq 0 \right)$$

where  $B(t) = W(t) - \inf_{s \leq t} W(s)$  is a reflected BM.

Each excursion of  $H$  corresponds to one tree in "GW forest"; the length of this excursion is the total progeny of this tree. If one conditions on the total progeny being  $n$ , let  $n \rightarrow \infty$ , we should obtain an excursion of the height process.

This is indeed true but we will not show it.

(4.63) Theorem (Aldous 1991) Under (4.59), the height process of a critical GW tree conditioned to have total progeny  $n$ , satisfies  

$$\left( \frac{1}{\sqrt{n}} H(L_{n+1}), t \in [0, 1] \right) \xrightarrow{n \rightarrow \infty} \frac{2}{\sqrt{\pi}} (e(t) : 0 \leq t \leq 1),$$
  
 where  $e$  is a standard Brownian excursion.

(4.64) Remark: The law of  $\epsilon$  can formally be obtained by conditioning the standard BM  $W$  on  $\{W(1)=1, W(s)>0 \forall s \in (0,1)\}$ . As this event has probability 0, this is slightly unsound, see e.g. Revuz-Yor, Chapter XII. Otherwise we can obtain  $\epsilon$  by conditioning a SRW  $S$  on  $\{S_i > 0 \text{ } i \leq m, S_m = 0\}$  letting  $m \rightarrow \infty$  and rescaling as in (4.60).

Actually more is true

(4.65) Theorem: (Machkour-Mekdadum (2003)). Conditioned on the total peeling  $m$ , for which we have Sat. (4.59)

$$\left( n^{-1/2} S(\lfloor m \cdot \cdot \rfloor), n^{-1/2} H(\lfloor m \cdot \cdot \rfloor), n^{-1/2} C(\lfloor m \cdot \cdot \rfloor) \right) \xrightarrow{d} (0e, \frac{2}{\sqrt{\pi}}e, \frac{2}{\sqrt{\pi}}e).$$

This suggests the existence of a 'local tree' coded by Brownian excursion. We now construct it.

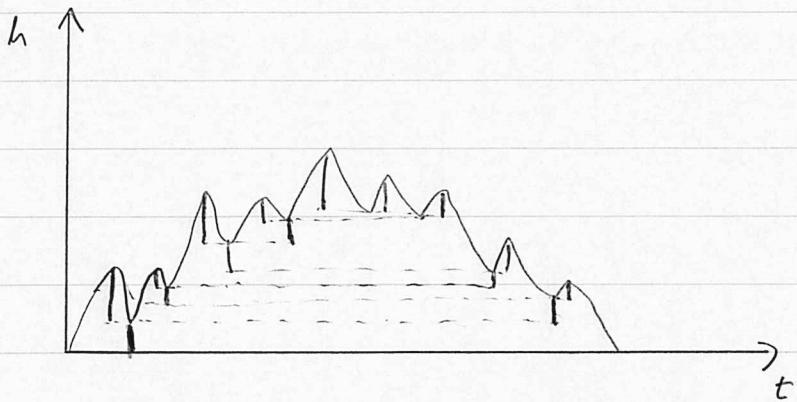
(4.66) Def: A compact metric space  $(\mathcal{T}, d)$  is a real tree if for all  $x, y \in \mathcal{T}$

(a) exists a unique shortest path  $f_{xy}$  from  $x$  to  $y$   
(and  $\text{length}(f_{xy}) = d(x, y)$ )

- (b)  $f_{xy}$  is the only non self-intersecting paths from  $x$  to  $y$ .
- Rooted real tree is a real tree with a distinguished vertex  $p$ . Elements of  $\mathcal{T}$  are called vertices of  $\mathcal{T}$ . A leaf is a vertex  $v$  such that  $v \notin f_{vw}$  for any  $w \neq v$ .

### Locating real lines:

Assume real  $h: [0, \infty) \rightarrow [0, \infty)$  is continuous with compact support.  $h$  will play the rôle of centre of real line.



(4.67) Figure:  $h$  and corresponding real line

Use  $h$  to define a distance  $d_h$  on  $\text{supp } h$ .

$$(4.68) \quad d_h(x, y) = h(x) + h(y) - 2 \inf \{h(z) : z \in [x, y]\}, \quad x \leq y$$

Let  $\sim$  be an equivalence relation

$$x \sim y \iff d_h(x, y) = 0,$$

and take the quotient  $T_h = \text{supp } h / \sim$

Then  $(T_h, d_h)$  is a real line. We always take the equivalence class of 0 to be the look  $\emptyset$ .

(4.69) Definition Let  $\epsilon$  be the standard Brownian motion.  
The random line  $T_\epsilon$  is called Aldous' continuum random line (CRT).

### Measuring distance between metric spaces.

Let  $(M, \delta)$  be a metric space. The Hausdorff distance between two compact subsets  $K, K'$  is

$$(4.70) \quad d_H(K, K') = \inf \{ \epsilon > 0 : K \subset F_\epsilon(K'), K' \subset F_\epsilon(K) \}$$

with  $F_\epsilon(K) = \{y \in M : \delta(y, K) \leq \epsilon\}$ .

(46)

To measure the distance between two compact metric spaces  $(X, d), (X', d')$ , the idea is to embed them isometrically into a single larger metric space and use  $d_H$ .

(4.71) Def (Gromov-Hausdorff distance)

$$d_{\text{GH}}(X, X') = \inf \{ d_H(g(X), g'(X')) \},$$

where the infimum is taken over all choices of metric spaces  $(Y, \delta)$  and isometric embeddings  $g: X \rightarrow Y, g': X' \rightarrow Y$ .

For bounded spaces, we set

$$d_{\text{GH}}(X, X') = \inf \{ d_H(g(X), g'(X')) \vee \delta(g(p), g'(p)) \}$$

This seems complicated, but fortunately for real trees  $T_h, T_{h'}$  with coding functions  $h, h'$  we have

(4.72) Lemma:  $d_{\text{GH}}(T_h, T_{h'}) \leq 2 \|h - h'\|_\infty$ .

From (4.72) and (4.65) we can then deduce

(4.73) Theorem (Aldous) Let  $T_m$  be the critical GW tree conditioned to have size  $m$ , satisfying (4.59), endowed with the usual distance. Then

$$\frac{T}{T_m} T_m \xrightarrow[m \rightarrow \infty]{\text{GH}} \mathbb{T}_{2c}.$$

We will not prove all statements made here but it is useful to key

(4.74) Exercise: Let  $T_m$  be GW with offspring distribution

$$P(X=k) = 2^{-(k+1)}, \text{ conditioned on } |T_m|=m.$$

Show that  $C$  is a SRW conditioned to hit 0 for the first time at  $2(m-1)$ .

We now come back and show the consequence of the height function of the critical GW first.

### Proof of Theorem (4.62)

We use the so called "ladder time decomposition of RW".

Recall that  $S$  is a random walk whose steps  $X_i - 1 =: Y_i$  have mean 0 and variance  $\sigma^2$ , in particular they take values in the set  $\{-1, 0, 1, 2, \dots\}$ . We define

$$(4.75) \quad T = \inf \{k \geq 1 : S(k) \geq 0\}$$

to be the first (real) ascending ladder time.

$S$  is recurrent, so  $T$  is a.s. finite.

If  $Y_0 \geq 0$ , then  $T=1$  and  $S(T) = Y_0$ .

On the other hand, if  $Y_0 = -1$ , then  $S$  might do several excursions going below -1 and eventually return there several times, but finally  $S$  leaves -1 (maybe by going downwards) and reaches  $\{0, 1, \dots\}$  without hitting -1 again. By strong Markov property, the number of excursions from  $\{-1\}$  to  $\{-1\}$  before hitting  $\{0, 1, \dots\}$  is geometrically distributed with parameter  $P(Y_0 > 0)$ .

(4.76) Exercise: (a) Use above considerations to show that for  $k \geq 0$

$$P(S(T) = k) = P(Y_0 = k) + P(Y_0 = -1) \cdot P(S(T) = k+1 \mid S(T) > 0)$$

(b) Use (a) to show that for  $k \geq 0$

$$P(S(T) = k) = \sum_{j=0}^{\infty} \left( \frac{P(Y_0 = -1)}{P(S(T) > 0)} \right)^j P(Y_0 = k+j)$$

(c) Observe that  $\sum_{k=0}^{\infty} \sum_{j=k}^{\infty} P(Y_0 = j) = E[Y_0 + 1] = 1$ .

Deduce from this and (b) that

$$P(S(T) > 0) = P(Y_0 = -1)$$

And thus  $P(S(T) = k) = \sum_{j=0}^{k-1} P(Y_0 = j)$ ,  $k \geq 0$ .

Hence also  $E[S(T)] = \frac{12k}{2}$ .

We come back to (4.62). We write

$$(4.77) \quad M(n) = \sup_{k \leq n} S(k), \quad m(n) = \inf_{k \leq n} S(k)$$

and introduce a time reversal walk

$$(4.78) \quad \hat{S}^n(k) = S(n) - S(n-k), \quad k \in \{0, \dots, n\}.$$

It is easy to see that the increments of  $\hat{S}^n$  have the same distribution as the increments of  $S$ , i.e.

$$(4.79) \quad (\hat{S}_k^n, k \leq n) \stackrel{d}{=} (S_k, k \leq n)$$

By (4.57)

$$(4.80) \quad \begin{aligned} H(n) &= \#\{0 \leq k < n : S(k) = \min_{0 \leq j \leq n} S(j)\} \\ &= \#\{1 \leq i \leq n : \hat{S}^n(i) = \sup_{0 \leq k \leq i} \hat{S}^n(k)\} \end{aligned}$$

By analogy define

$$(4.81) \quad J(n) = \#\{1 \leq i \leq n : S(i) = M(i)\}$$

and observe that

$$(4.82) \quad \sup_{0 \leq k \leq n} \hat{S}^n(k) = S(n) - \inf_{0 \leq k \leq n} S(k) = S(n) - m(n).$$

From (4.79) - (4.82) we get that for every  $n$  fixed

$$(4.83) \quad (M(n), J(n)) \stackrel{d}{=} (S(n) - m(n), H(n)).$$

We now clarify

$$(4.84) \quad \text{Claim: } \frac{H_n}{S(n) - m(n)} \xrightarrow{\mathbb{P}} \frac{2}{\pi^2}.$$

By (4.60), for  $t_1 \leq \dots \leq t_m$

$$(4.85) \quad \begin{aligned} \frac{1}{t_m} (S(L_{[t_1, 1]}) - m(L_{[t_1, 1]}), \dots, S(L_{[t_m, 1]}) - m(L_{[t_m, 1]})) \\ \xrightarrow{d} \pi (W(t_1) - \inf_{s \leq t_1} W(s), \dots, W(t_m) - \inf_{s \leq t_m} W(s)) \end{aligned}$$

Hence, by (4.84).

$$\frac{1}{t_m} (H(L_{[t_1, 1]}), \dots, H(L_{[t_m, 1]})) \xrightarrow{d} \frac{2}{\pi} (W(t_1) - \inf_{s \leq t_1} W(s), \dots, W(t_m) - \inf_{s \leq t_m} W(s))$$

proving (4.62) at least at the level of finite dimensional distributions. The proof of the tightness is omitted.  $\square$

It remains to show (4.84).

Proof of (4.84): Set  $T_0 = 0$ ,

$$(4.86) \quad T_i = \inf \{ k > T_{i-1} : S(k) = M(k) \}, \quad i \geq 1.$$

be the sequence of successive weak ascending ladder times (observe  $T_1 = T$ ). Then,

$$M(n) = \sum_{\substack{k \geq 1 \\ T_k \leq n}} (M(T_k) - M(T_{k-1})) \stackrel{(4.81)}{=} \sum_{k=1}^{J(n)} (M(T_k) - M(T_{k-1}))$$

$$= \sum_{k=1}^{J(n)} (S(T_k) - S(T_{k-1}))$$

Since  $J(n) \nearrow \infty$  as  $n \nearrow \infty$ , by strong LLN and (4.76)

$$(4.88) \quad \frac{M(n)}{J(n)} \xrightarrow{n \rightarrow \infty} E[S(T_1)] = \frac{\sigma^2}{2} \text{ a.s.}$$

Together with (4.83) it then implies (4.84)  $\square$ .

### Shape of critical ER-graph clusters:

We close this chapter by exploring how the ideas of the previous section can be extended to understand the clusters of the critical ER graph. It is based on paper by Aldous-Berry, Broutin and Goldschmidt.

First we should however come back to Aldous (1997) paper. We have seen in (4.16) that rescaled sizes of clusters converge to excursion length of the process  $B^\theta$  of (4.15).

This was proved by replacing cluster by depth-first type walk, i.e. essentially by looking at trees contained in the clusters.

One can ask how much the clusters differ from trees.

The first natural question is how many additional edges they contain. We thus define

$$(4.89) \quad S_m(i) = \# \text{edges in } i^{\text{th}} \text{ largest cluster} - \# \text{vertices} + 1$$

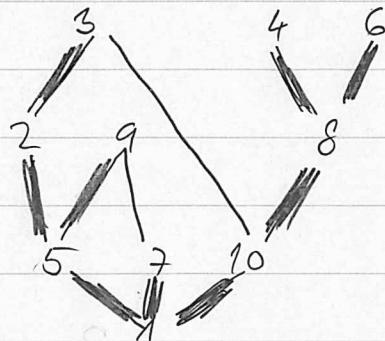
to be the number of "surplus" edges in  $i^{\text{th}}$  largest cluster.

To understand  $S_m$  we come back to the explanation,

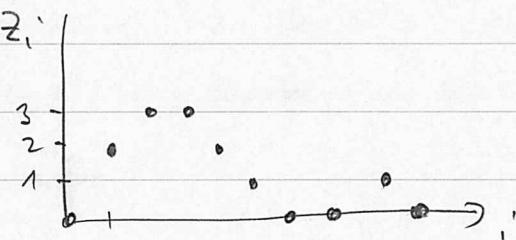
(50)

and recall that the exploration checks the state of all possible edges between active and non-active surfaces, but never checks edges between active surfaces.

(4.90) Example Consider graph



It's depth-first walk visits nodes  $1, 5, 2, 3, 9, 7, 10, 8, 4, 6$  in this order.  $S_i$ , the number of active surfaces at the end of the step  $i$  is  $S_i = Z_i + 1$  with  $Z_i$  as



which is roughly the "area" under the graph of  $\mathbb{Z}$ .

as consequence, in  $ER(m, \frac{1}{m} + \frac{\theta}{m^{4/3}})$ , given a compound and its depth-first walk, the surplus is a

- (4.92)  $Bin(P, \frac{1}{m} + \frac{\theta}{m^{4/3}})$  random variable. Taking now the scaling of  $\mathbb{Z}$  from (4.25) one obtains the following improvement of (4.16)

- (4.93) Theorem (Aldous): Let  $\mathcal{B}^\theta$  be as in (4.15) and consider Poisson point process with rate  $1/m [0, \infty)^2$ . Let  $j_1 \geq j_2 \geq \dots$  be the ordered length of excursions of  $\mathcal{B}^\theta$  and let  $s_i$  be the # of points of the Poisson process under the excursion corresponding to  $j_i$ . Then
- $$(m^{-2/3} C_m(i), S_m(i))_{i \geq 1} \xrightarrow[m \rightarrow \infty]{} (j_i, s_i)_{i \geq 1}.$$

To understand the shape one observes first:

- (4.94) Claim: A compound of  $G(n, p)$  conditioned to have  $m$  vertices and  $s$  surplus edges is a uniform connected graph with  $m$  vertices and  $m+s-1$  edges.

For  $n=0$  Aldous (1993), Le Gall (2005) proved that

- (4.95) Thm: Let  $T_m$  be a uniformly chosen tree on  $m$  labelled vertices, viewed as metric space. Then

$$\frac{1}{T_m} T_m \xrightarrow[GH]{d} \mathbb{T}_{\text{eu}} \quad \text{as } m \rightarrow \infty.$$

Here  $\mathbb{T}_{\text{eu}}$  is the CRT.

Surplus explained on the blackboard....