

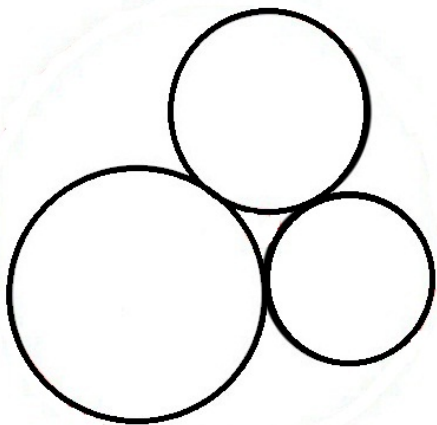
On longest paths and diameter in random Apollonian networks

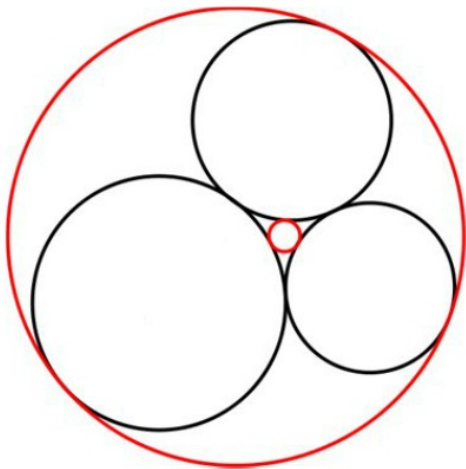
Abbas Mehrabian
amehrabi@cs.ubc.ca

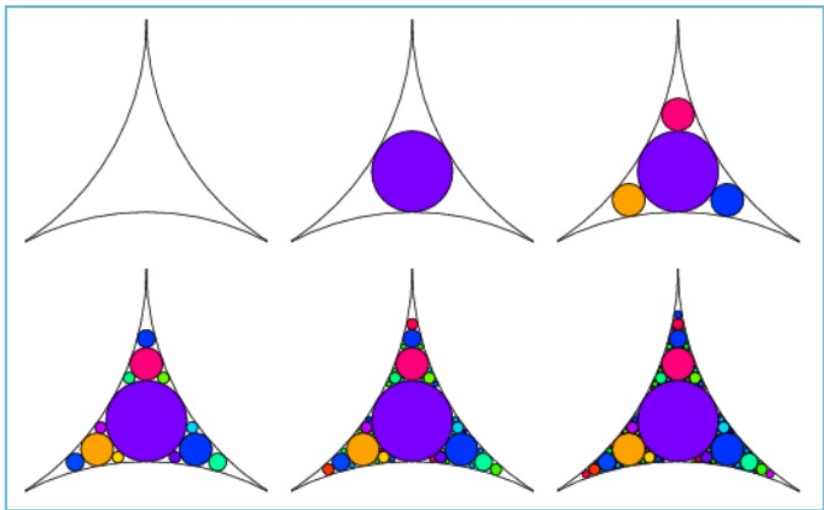
University of British Columbia

5 October 2016

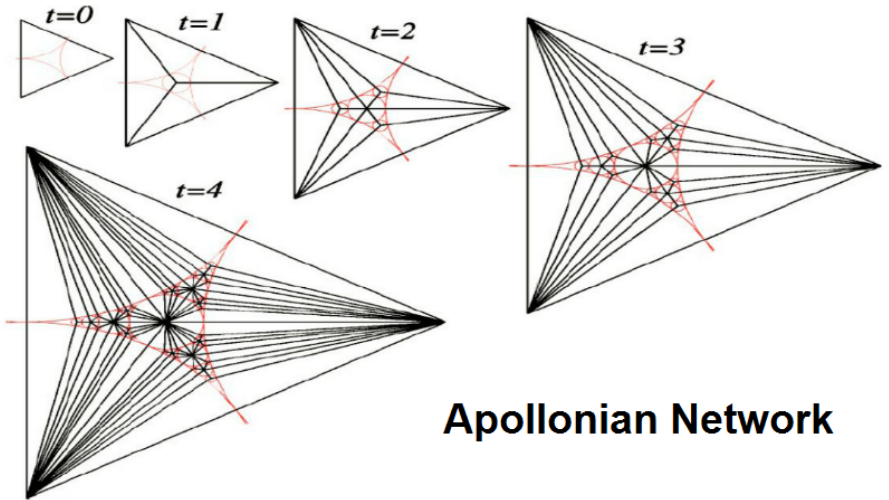
joint work with A. Collevocchio, E. Ebrahimzadeh, L. Farczadi, P. Gao,
C. Sato, N. Wormald, and J. Zung



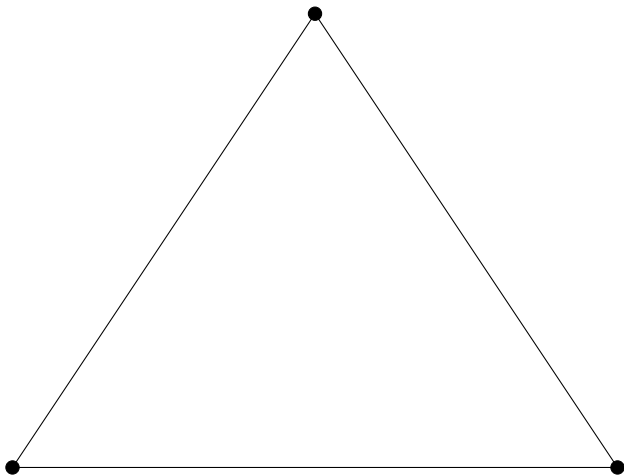




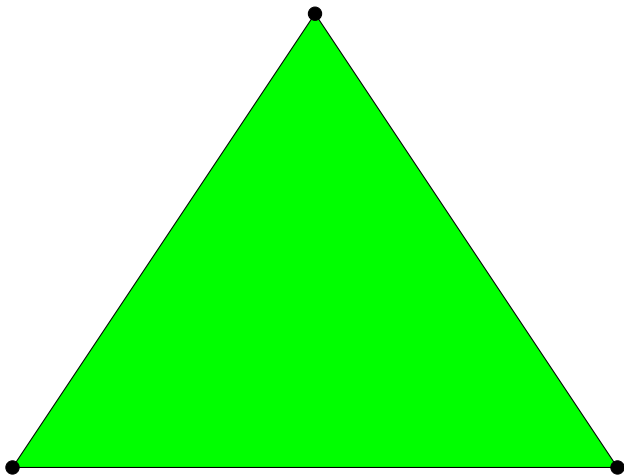
Apollonian Gasket



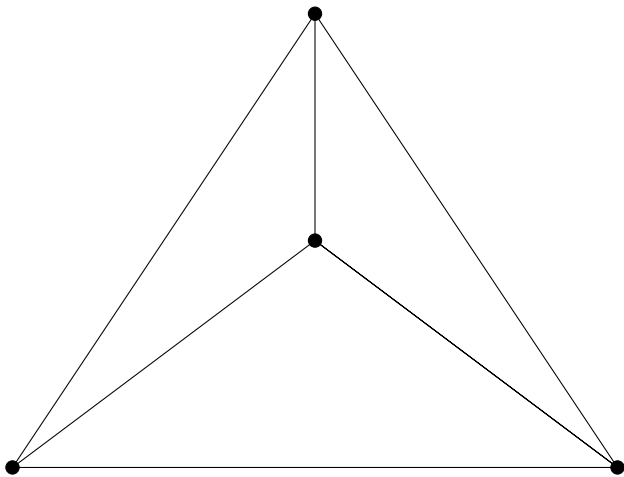
Apollonian Network



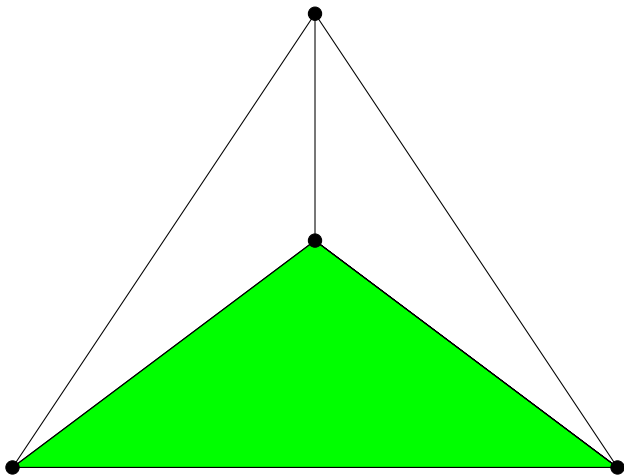
$t = 0$



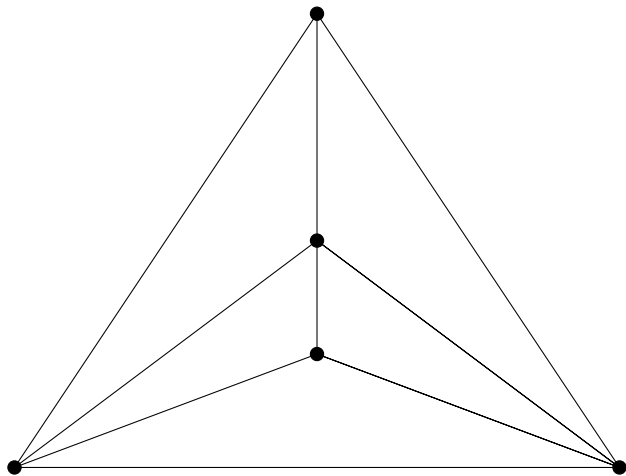
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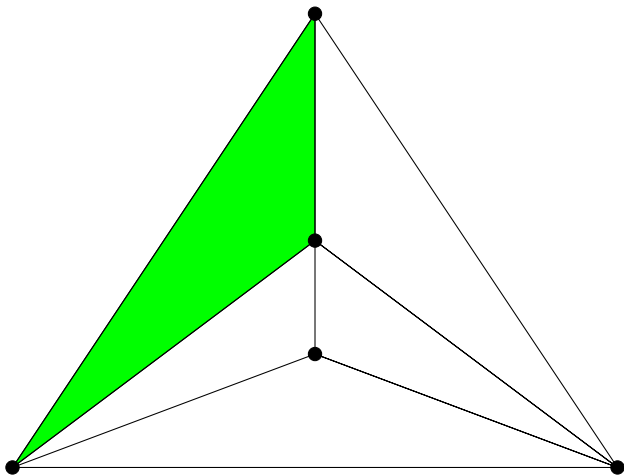
$t = 1$



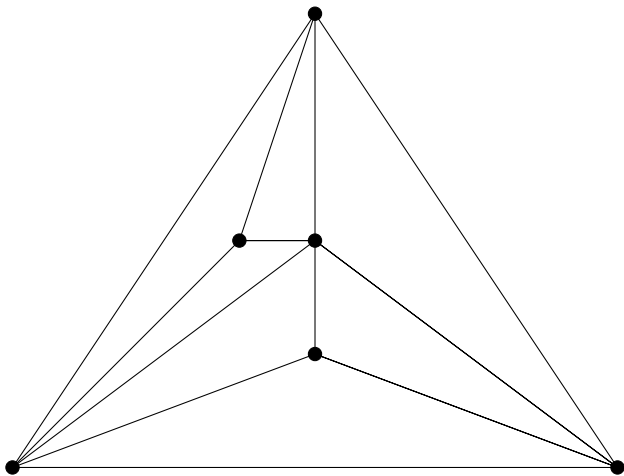
$t = 1$



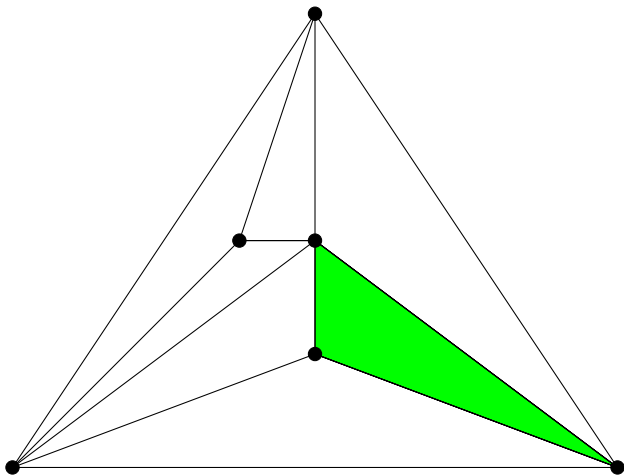
$$t = 2$$



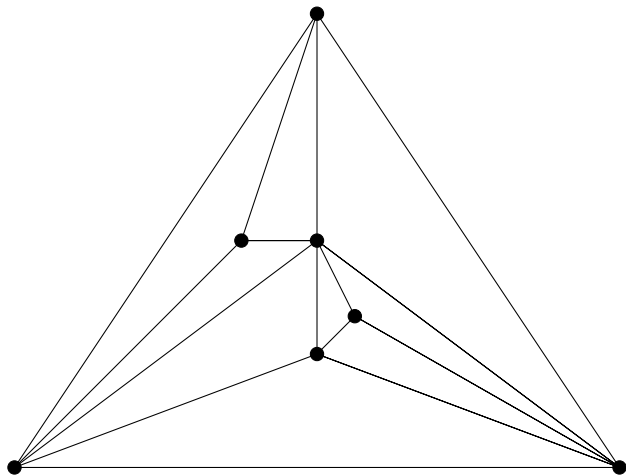
$t = 2$



$t = 3$



$t = 3$



$t = 4$

After t steps,

- ✓ a random triangulated plane graph
- ✓ $n = t + 3$ vertices
- ✓ $3t + 3$ edges
- ✓ $2t + 1$ faces

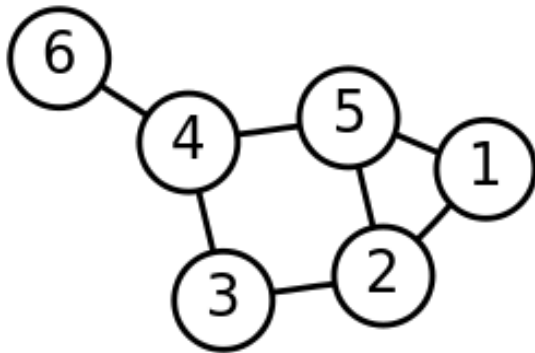
called a **Random Apollonian Network (RAN)**.

Zhou, Yan, Wang'05: generating power-law **planar** graphs.

Theorem (Frieze and Tsourakakis'12)

For any fixed k , the fraction of vertices with degree k is concentrated around k^{-3} .

The Diameter of a Graph



Diameter = 3

Diameters of RANs

Theorem (Albenque and Marckert'08; Frieze and Tsourakakis'12)

With high probability (asymptotically almost surely),

$$0.54 \log n < \text{diameter} < 7.1 \log n$$

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A similar result was proved independently by Cooper, Frieze, Uehara'13 and Kolossváry, Komjáty, Vágó'13.

Longest paths in RANs

\mathcal{L}_n := length of a longest path (self-avoiding walk)

Frieze and Tsourakakis'12 Is $\mathcal{L}_n = \Omega(n)$ whp?

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Theorem (EFGMSWZ'13)

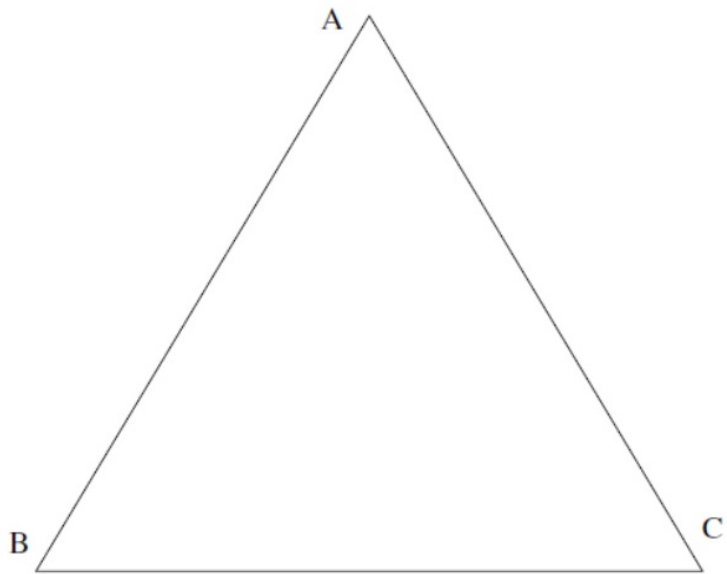
We have

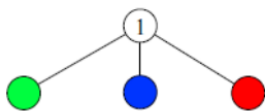
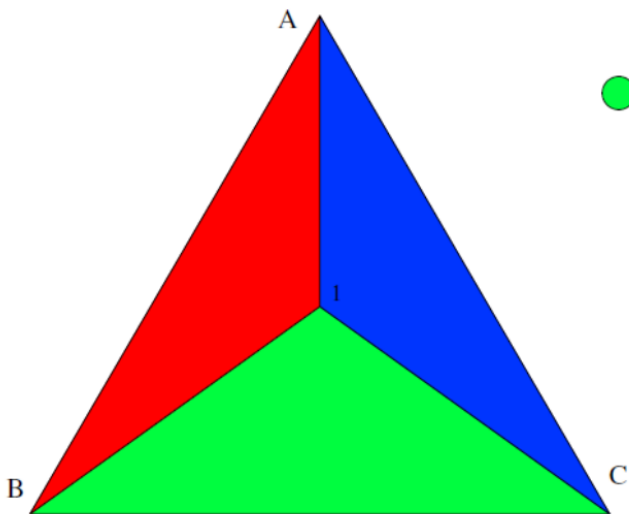
$$\mathcal{L}_n > n^{0.63}$$

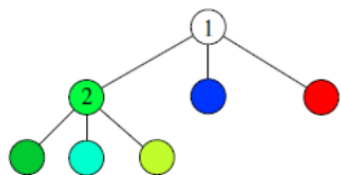
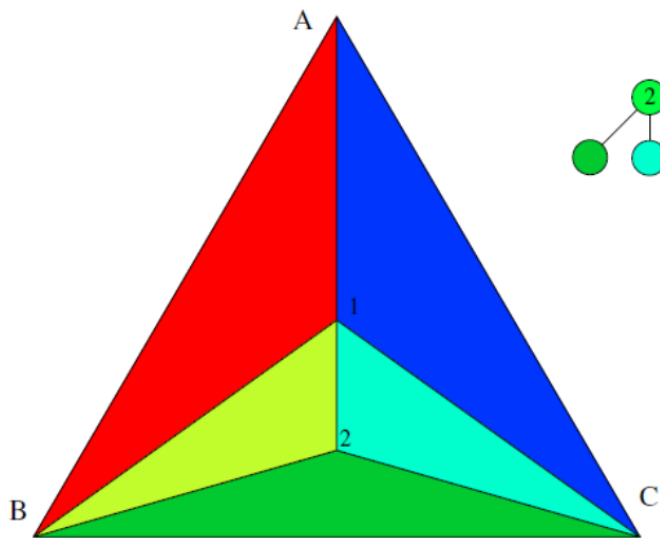
and

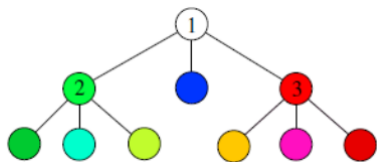
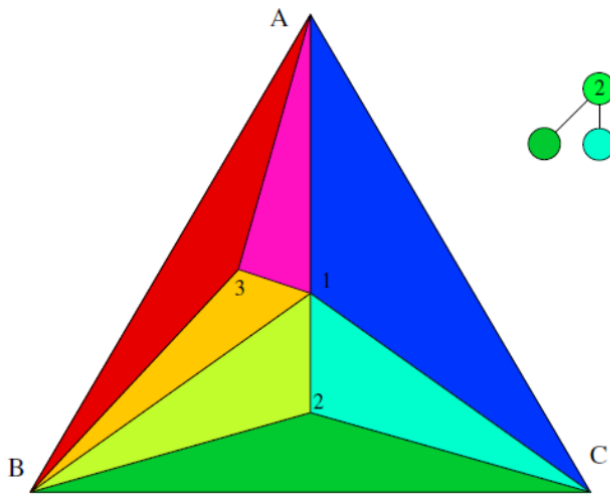
$$\mathbb{E}[\mathcal{L}_n] = \Omega(n^{0.88})$$

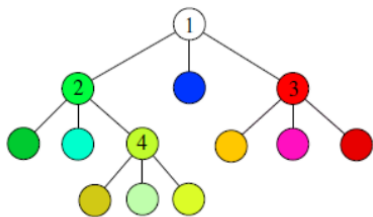
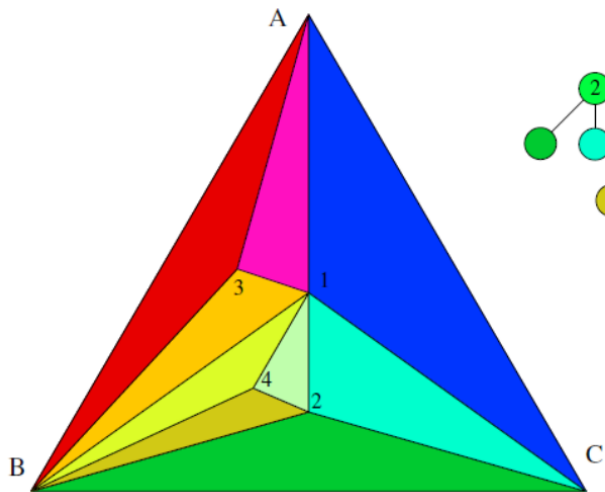
Δ -tree of a RAN

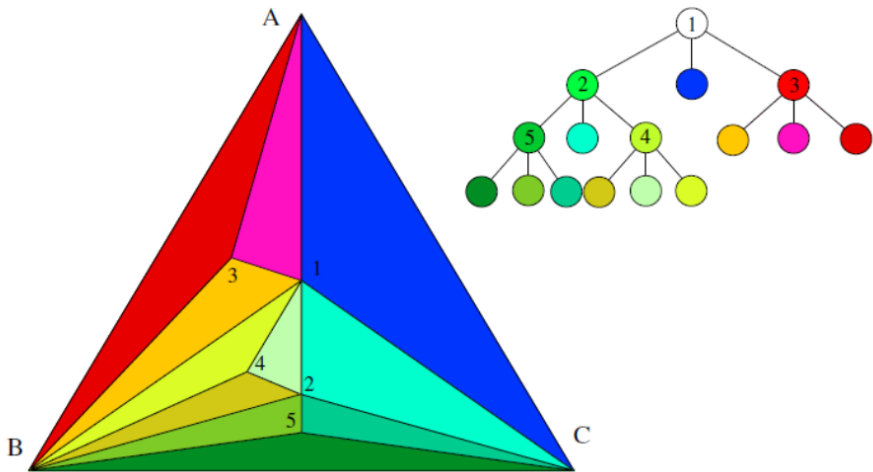










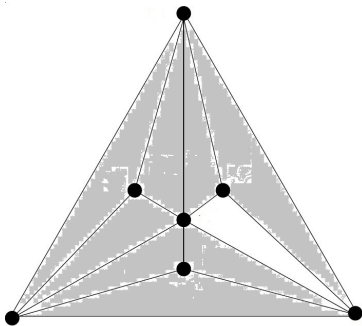


In each step, a random leaf gives birth to three children.
 This is called a **random (recursive) ternary tree**.

Proof outline for
length of the longest paths $< n^{0.999999996}$

Upper Bound for Longest Path

The Main Idea



Claim: A path cannot contain internal vertices of all 9 faces.

Regions

vertices

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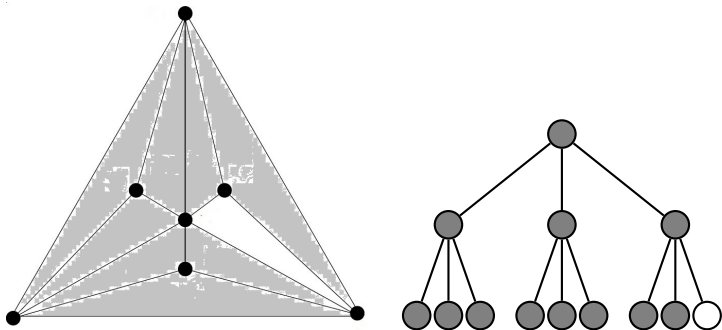
•

0

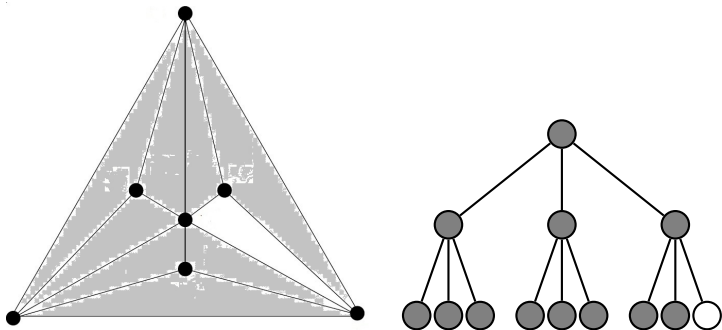
≥ 16

≤ 14

•



If we colour those nodes of Δ -tree which a path goes inside, each coloured node can have at most 8 black grandchildren.



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New goal: bound the total number of coloured nodes in a random ternary tree.

Simplified goal

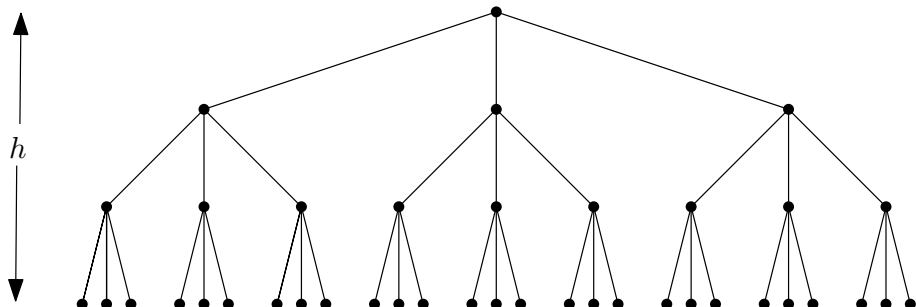
Simplified goal: any binary subtree of a random n -vertex ternary tree has size $\leq n^{0.9999}$

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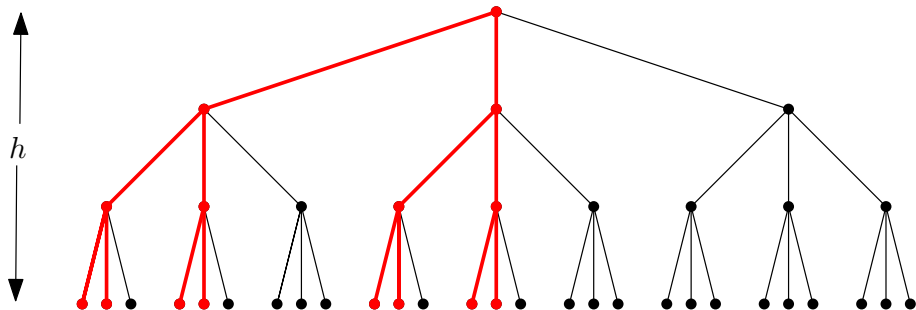
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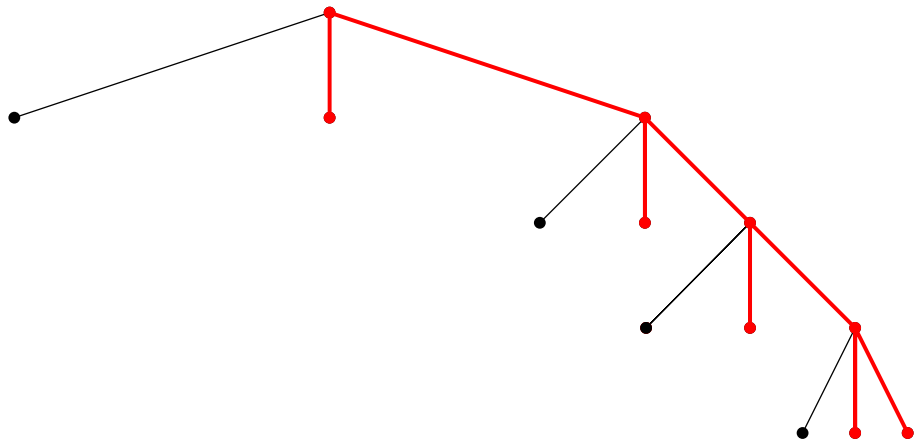


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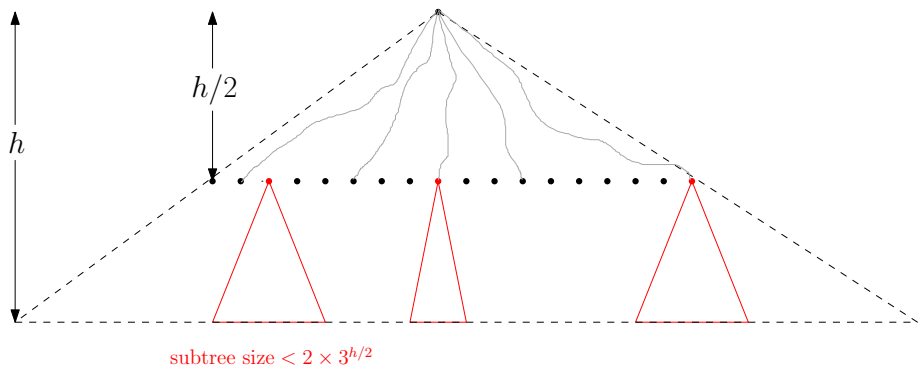
$$\text{size of binary subtree} = 2^{h+1} - 1 \leq 2 \times (3^h)^{\log_3 2} < 2 \times n^{0.64}$$

Simplified goal: any binary subtree of a random n -vertex ternary tree has size $\leq n^{0.9999}$ (with probability $\rightarrow 1$ as $n \rightarrow \infty$).



size of binary tree = $n - o(n)$

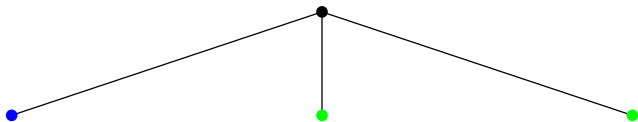
The strategy



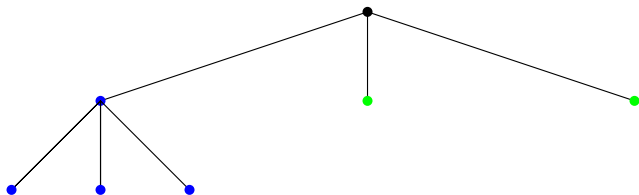
size of any binary subtree

$$< 2 \times 3^{h/2} + 2^{h/2} \times 2 \times 3^{h/2} < O(3^{\log_3 2 \times h/2 + h/2}) < O(3^{0.82h}) < n^{0.83}$$

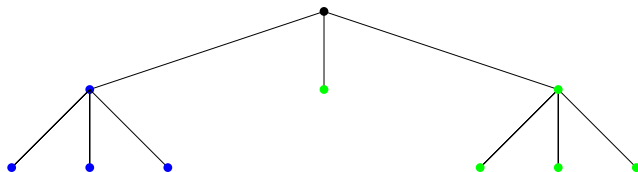
Analyzing subtree sizes



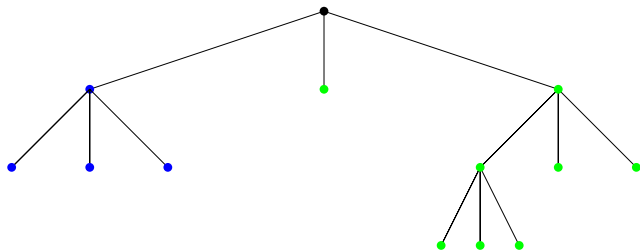
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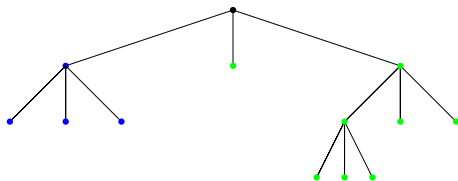
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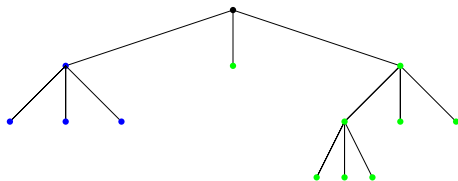
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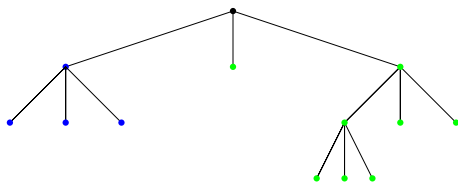


Analyzing subtree sizes



Growth rule: Start with one blue two green. In each step, choose a uniformly random leaf, and increase number of leaves of that colour by 2.

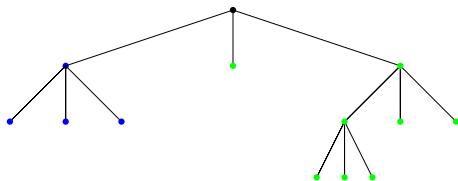
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This is exactly an Eggenberger-Polya (1923) urn!

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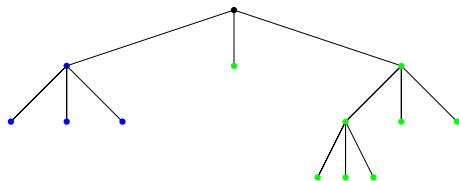


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$$\frac{\text{blue vertices}}{n-1} \xrightarrow[n \rightarrow \infty]{\text{in distribution}} \text{Beta}(1/2, 1)$$

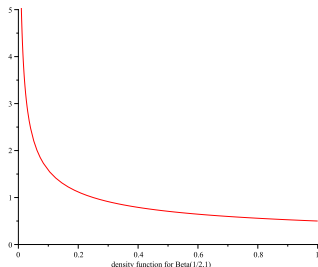
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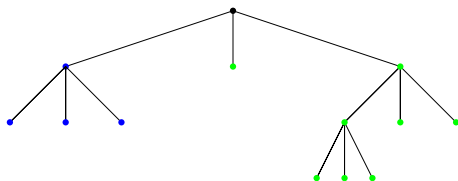
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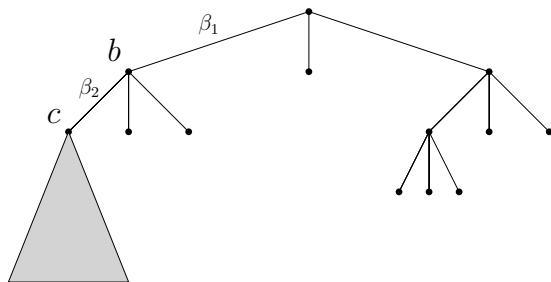
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Draws from an E-P urn are exchangeable, so by de Finetti's theorem,

$$\text{blue vertices} \sim \text{Binomial}(n - 1, \text{Beta}(1/2, 1))$$

analyzing subtree sizes



Suppose $\beta_1, \beta_2 \sim \text{Beta}(1/2, 1)$ independent.

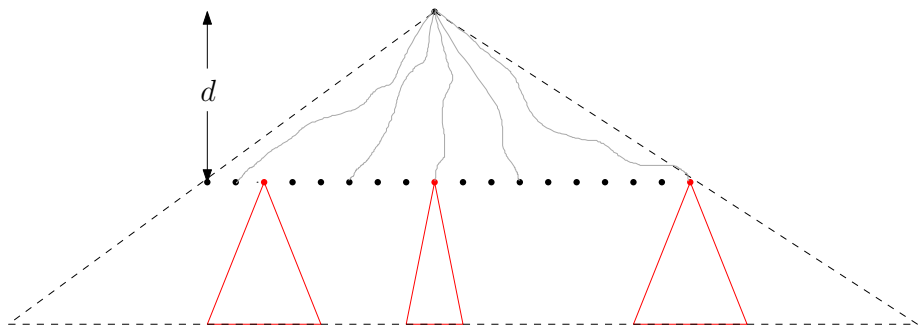
Size of subtree rooted at $b \sim \text{Bin}(n - 1, \beta_1)$

Size of subtree rooted at $c \sim \text{Bin}(\text{size of } b, \beta_2)$

$$\sim \text{Bin}(\text{Bin}(n - 1, \beta_1), \beta_2) \preccurlyeq \text{Bin}(n, \beta_1 \beta_2)$$

wrap up

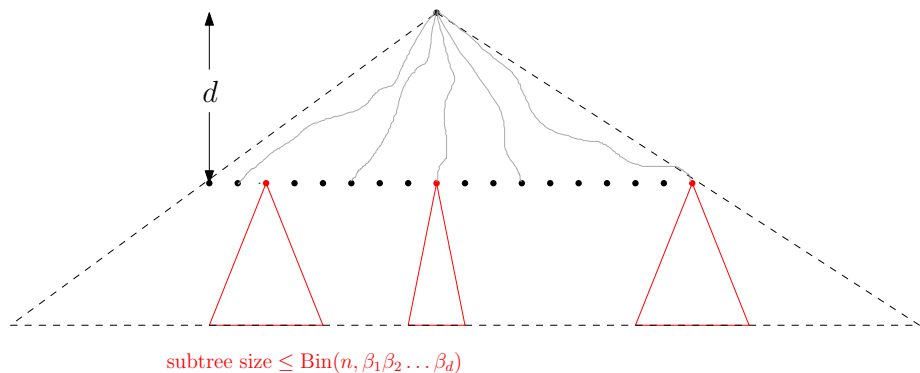
goal: any binary subtree of random n -vertex ternary tree has size $\leq n^{0.9999}$



subtree size $\leq \text{Bin}(n, \beta_1 \beta_2 \dots \beta_d)$

wrap up

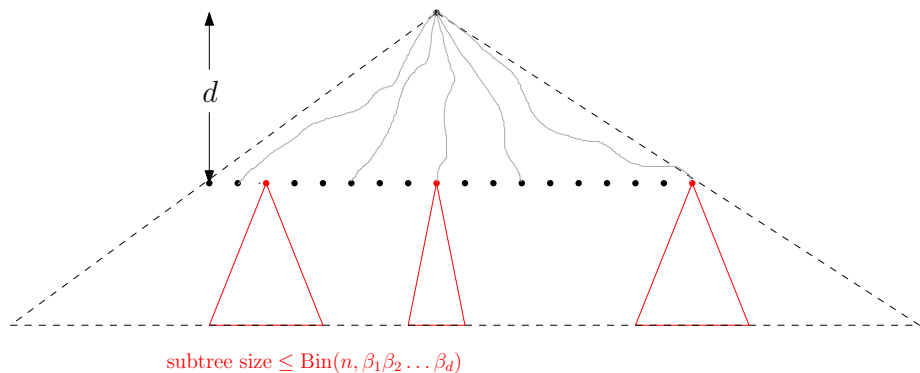
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For suitable d , each red subtree size is sharply concentrated around $n/3^d$.

wrap up

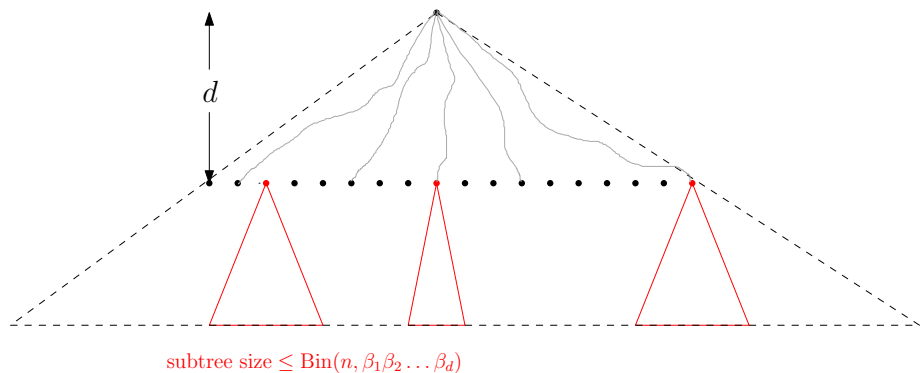
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For suitable d , each red subtree size is sharply concentrated around $n/3^d$.
Apply union bound over 3^d nodes gives uniform bound for all red subtrees

wrap up

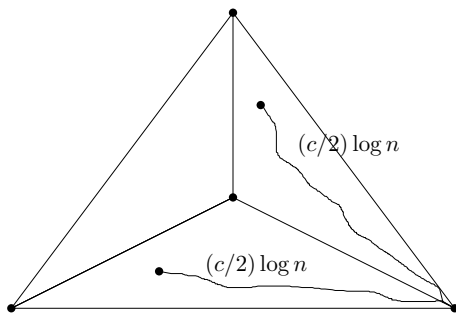
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For suitable d , each red subtree size is sharply concentrated around $n/3^d$.
Apply union bound over 3^d nodes gives uniform bound for all red subtrees
Size of any binary subtree $\leq 2 \times 3^d + 2^d \times \text{uniform bound} \leq n^{0.99986}$

$\frac{\text{diameter}}{\log n} \rightarrow c \approx 1.668$ Proof Outline for
in probability

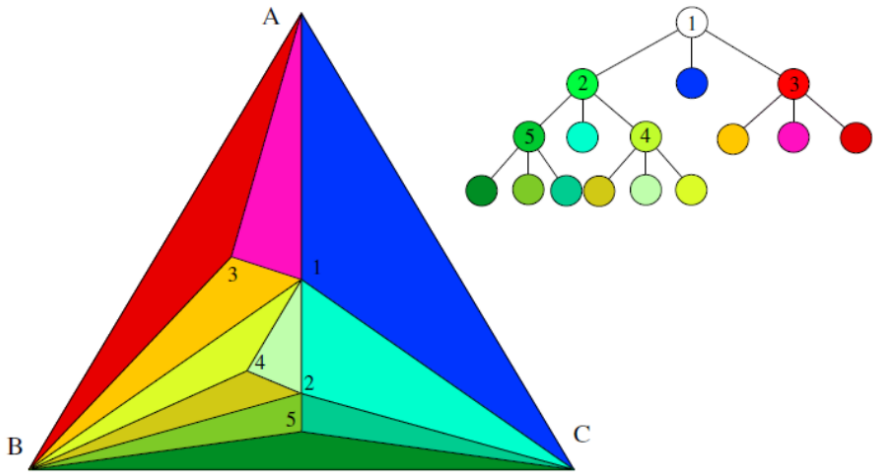
Radius



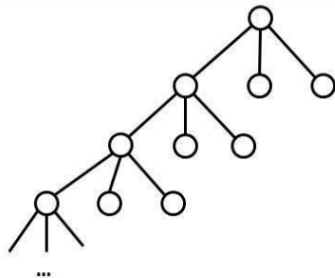
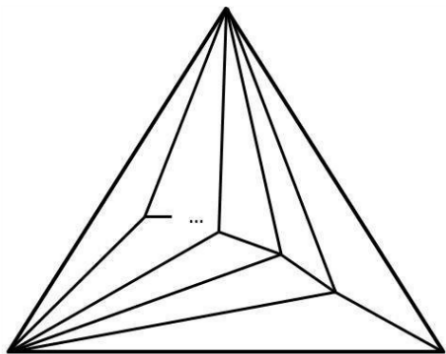
Radius : max distance between a vertex and the boundary

Lemma

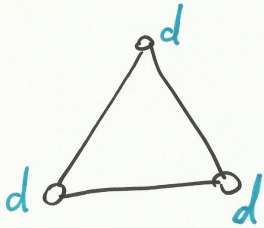
*If radius / $\log n \rightarrow c/2$ in probability,
then diameter / $\log n \rightarrow c$ in probability.*



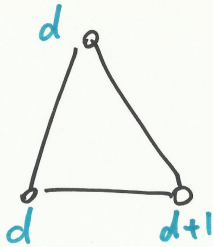
distance in graph \leq distance in tree



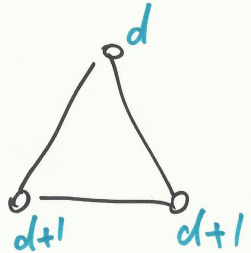
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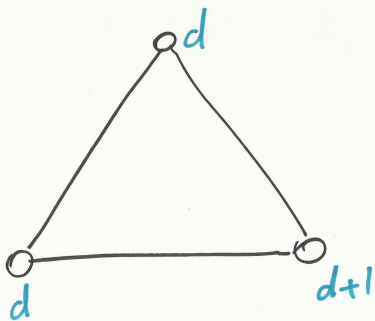
Type 1



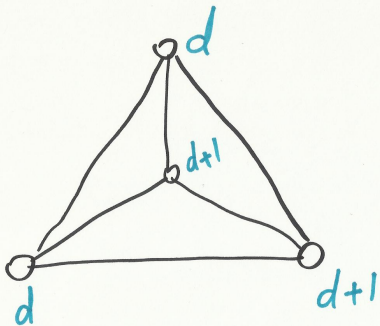
Type 2



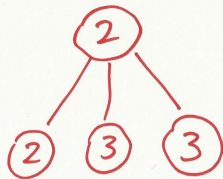
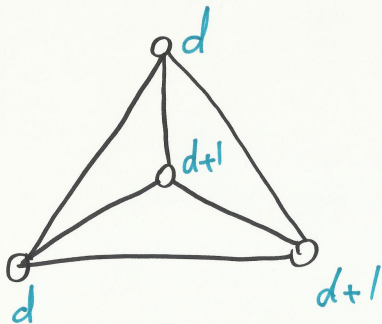
Type 3



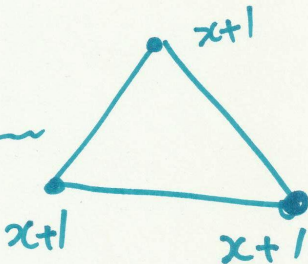
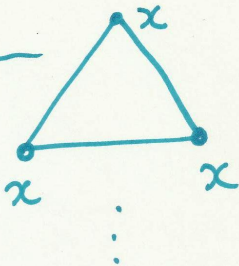
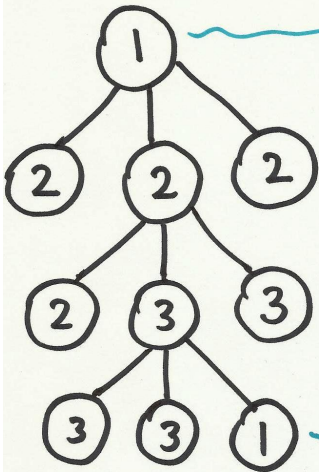
Type 2



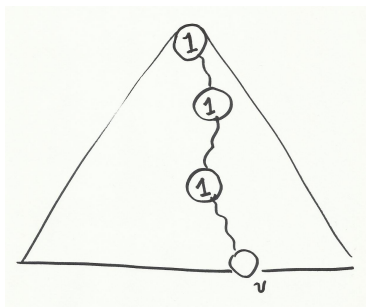
Type 2



Type 2

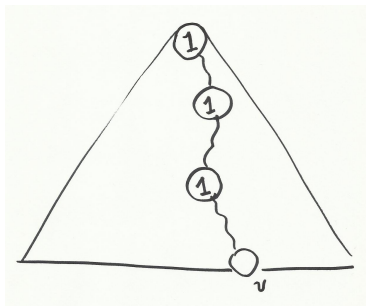


Crucial Observation



Distance of a vertex to the boundary (in graph) equals number of type-1 nodes on path of the corresponding node to the root (in tree)

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New goal: bound the largest number of type-1 nodes in any root-to-leaf path

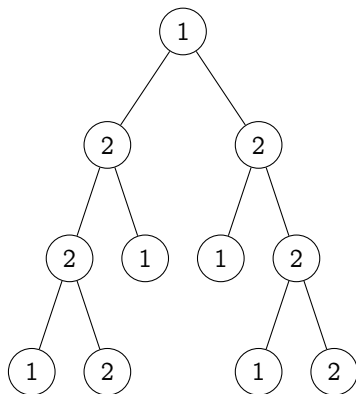
Simplified problem

Consider binary trees for simplicity:

a type-1 node has two type-2 children,

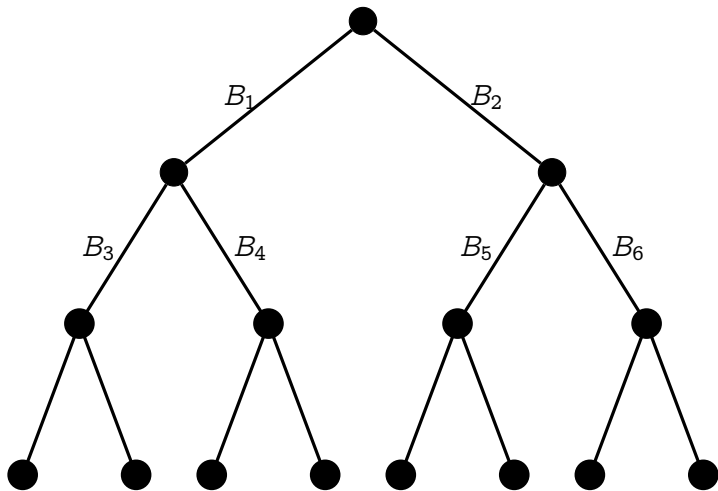
a type-2 node has one type-1 child and one type-2 child.

In every step a random leaf gives birth. After n steps, what's the largest number of type-1 nodes in any root-to-leaf path?!



The theorem of Broutin and Devroye

Infinite binary tree:

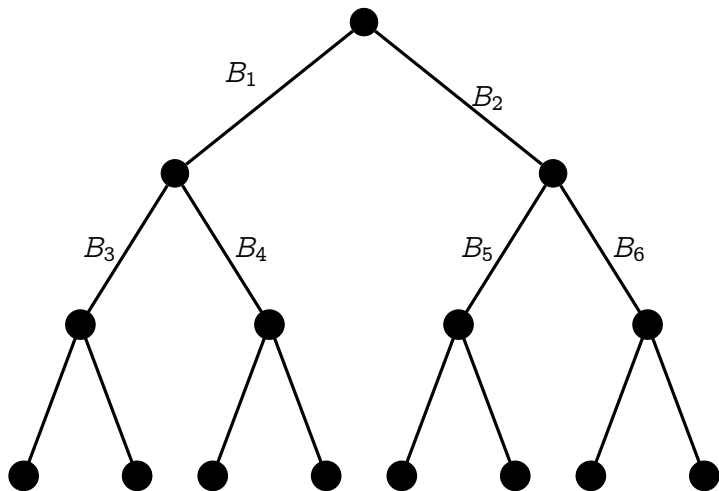


Theorem (Broutin and Devroye'06)

Assume:

- ✓ *All birth times have the same distribution.*
- ✓ *One-level offsprings of distinct vertices are mutually independent.*

Infinite binary tree:



$B_1 \sim B_2 \sim \dots \sim B_6$ and $B_1 \perp B_3, B_4, B_5, B_6$ etc.

Theorem (Broutin and Devroye'06)

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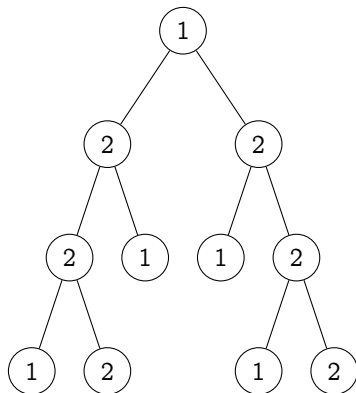
Then, *height* of tree at time t is whp asymptotic to ρt ,
 $\rho :=$ unique solution to

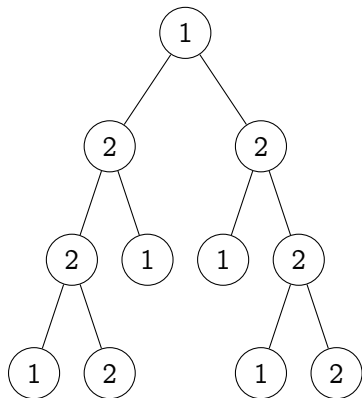
$$\sup_{\lambda \leq 0} \{\lambda/\rho - \log(\mathbb{E}[\exp(\lambda E)])\} = \log 2 .$$

Back to random Apollonian networks...

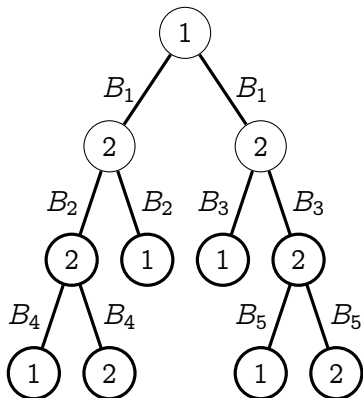
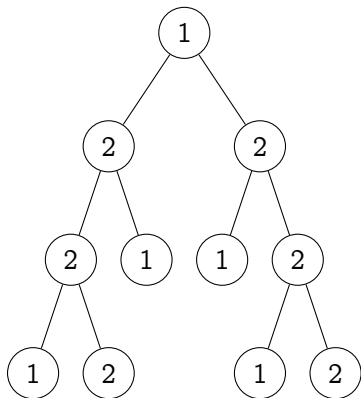
Simplified problem

A type-1 node has two type-2 children,
 a type-2 node has one type-1 child and one type-2 child.
 In every step a random leaf gives birth.
 After n steps, what's the largest number of type-1 nodes in any root-to-leaf path?!

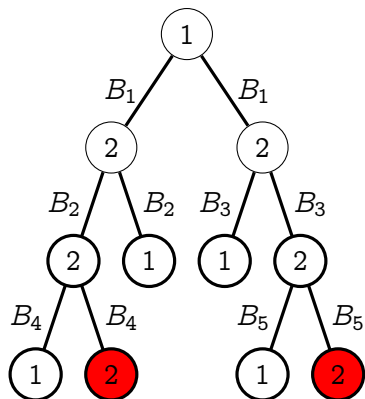
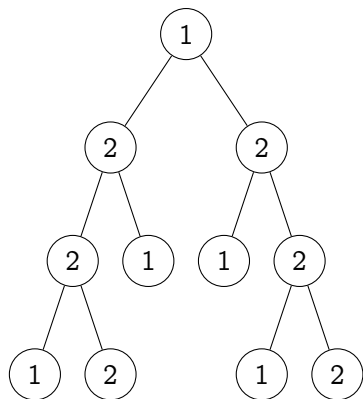


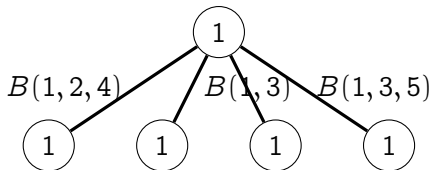
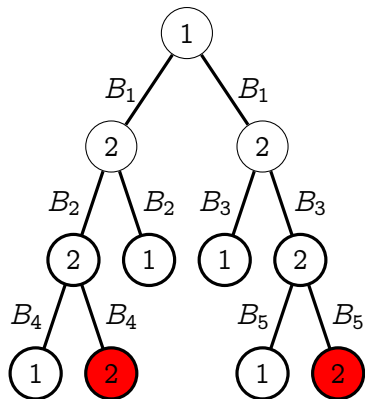
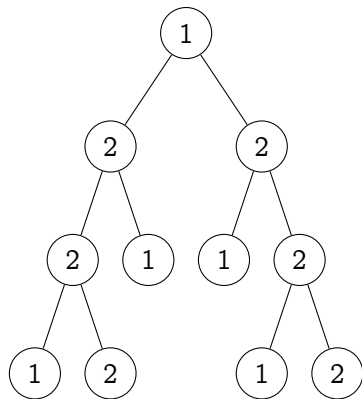


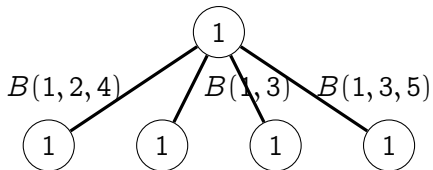
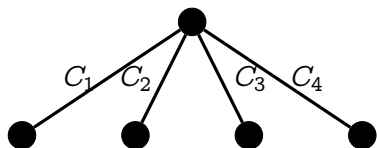
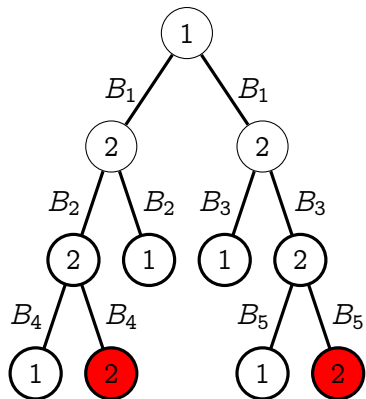
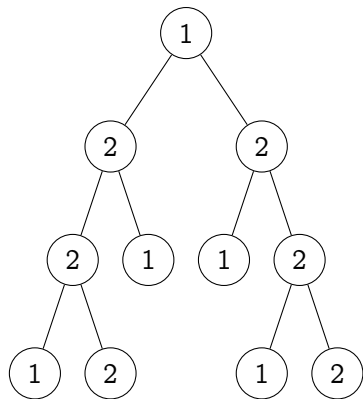
Use exponential birth times !



Use exponential birth times !







- ✓ For every fixed cut-off threshold k , we stochastically sandwich 1-height of our typed tree between heights of B&D-friendly trees.
- ✓ As $k \rightarrow \infty$, lower and upper bounds converge to $(c/2) \log n$.

Theorem (EFGMSWZ'13)

$$f(x) := \frac{12x^3}{1-2x} - \frac{6x^3}{1-x},$$

$y :=$ unique solution to

$$x(x-1)f'(x) = f(x) \log f(x), \quad x \in (0, 1/2),$$

$$c := (1 - y^{-1}) / \log f(y) \approx 1.668$$

Then for every fixed $\varepsilon > 0$,

$$\mathbb{P}[(1 - \varepsilon)c \log n \leq \text{diameter of a RAN} \leq (1 + \varepsilon)c \log n] \rightarrow 1$$



Eggenberger-Pólya Urn

Theorem (Eggenberger and Pólya 1923)

Start: g green, r red balls.

In each step:

- ✓ *pick a random ball and return it to the urn;*
- ✓ *add s balls of the same colour.*

After n draws:

g_n : green balls

t_n : number of balls

For any $\alpha \in [0, 1]$

$$\begin{aligned} \lim_{n \rightarrow \infty} \mathbb{P} \left[\frac{g_n}{t_n} < \alpha \right] &= \frac{\Gamma((g+r)/s)}{\Gamma(g/s)\Gamma(r/s)} \int_0^\alpha x^{\frac{g}{s}-1} (1-x)^{\frac{r}{s}-1} dx \\ &= \mathbb{P}[\text{Beta}(g/s, r/s) < \alpha] \end{aligned}$$

Broutin-Devroye's Theorem

Theorem (Broutin and Devroye 2006)

E := a positive random variable

b := a positive integer

T_∞ := an infinite b -ary tree.

Label the edges of T_∞ randomly,

- ① The label of every edge is distributed like E .
- ② For vertices u and v , edges going down from u and v are independent.

H_t := height of the subtree containing nodes whose sum of labels on their path to root $\leq t$.

Then $\frac{H_t}{t} \rightarrow \rho$ in probability

ρ := unique solution to

$$\sup\{\lambda/\rho - \log(\mathbb{E}[\exp(\lambda E)]) : \lambda \leq 0\} = \log b .$$