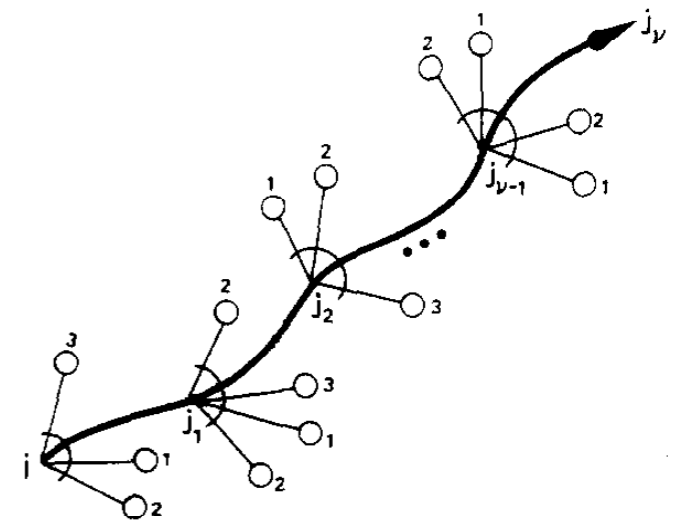
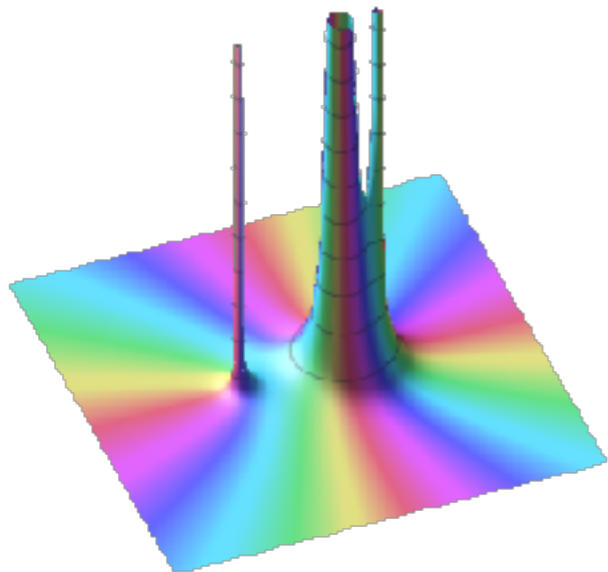


On (Effective) Analytic Combinatorics

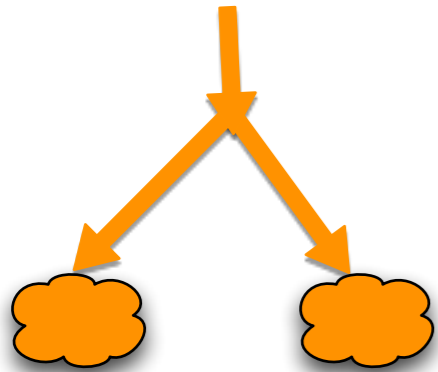
Bruno Salvy

Inria & ENS de Lyon

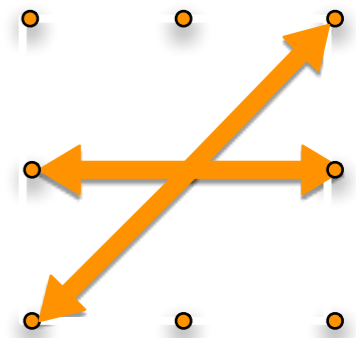
Séminaire Algorithmes
Dec. 2017



Combinatorics, Randomness and Analysis

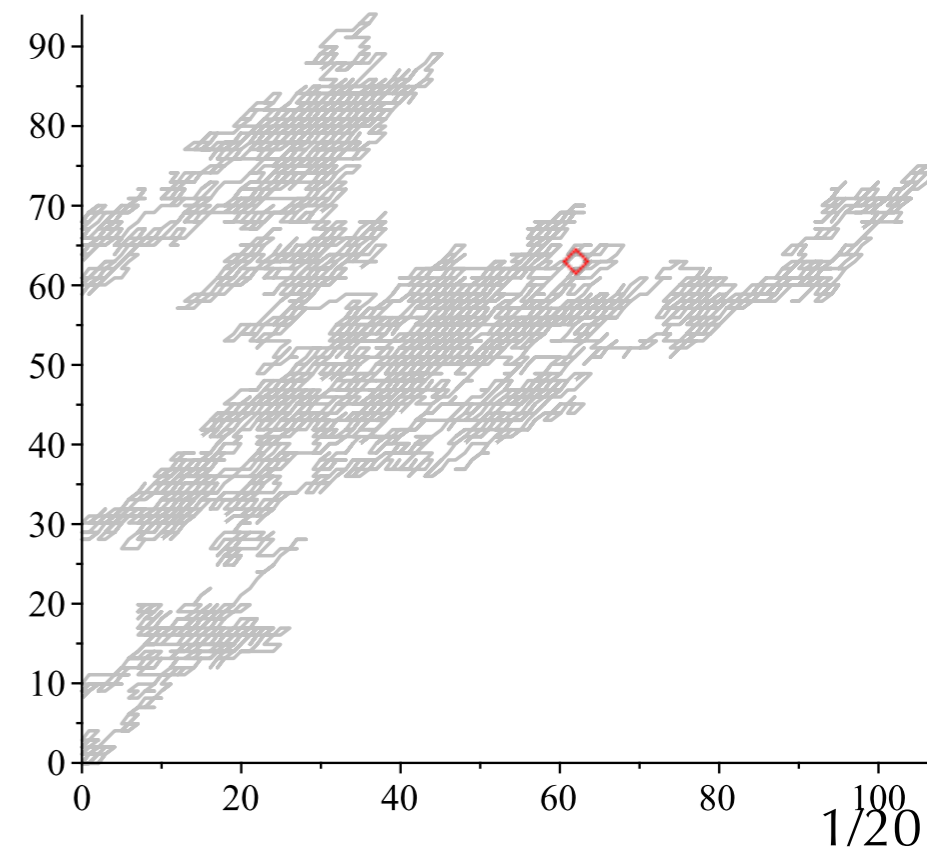


From **simple local** rules, a **global** structure arises.

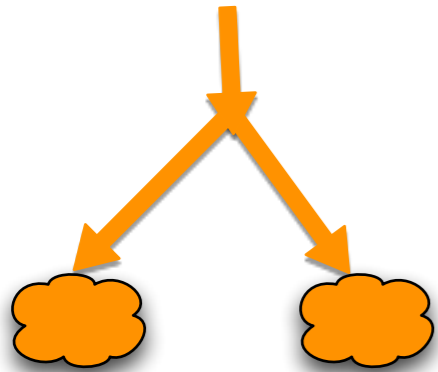


A quest for **universality** in random discrete structures:
➔ **probabilistic complexity** of structures and algorithms.

Quantitative results using **complex analysis**.



Combinatorics, Randomness and Analysis

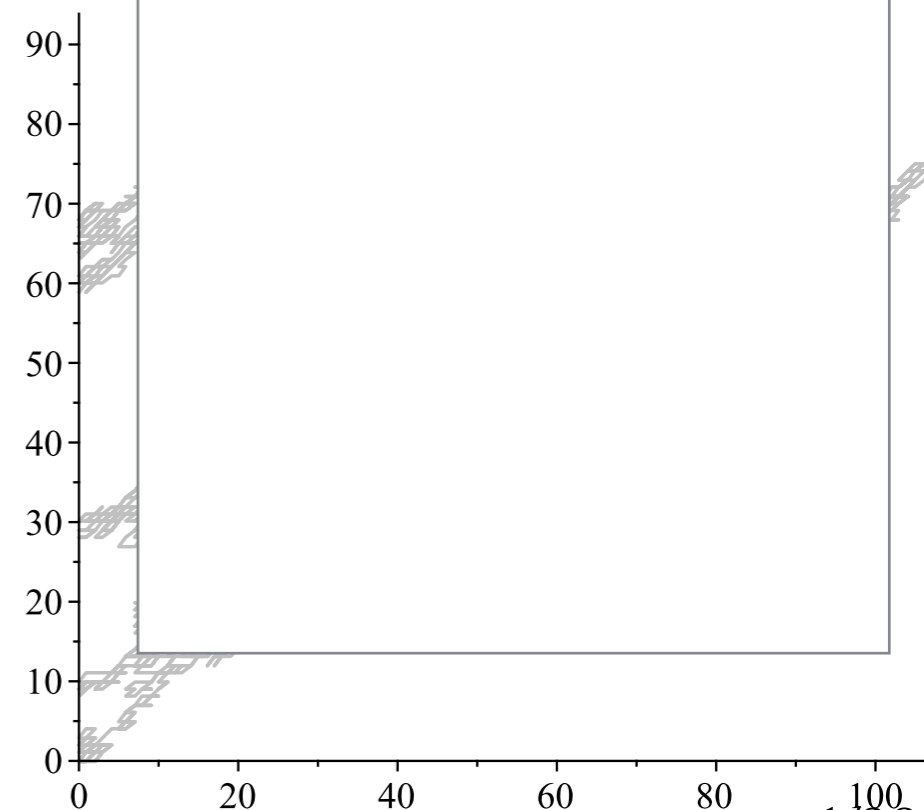


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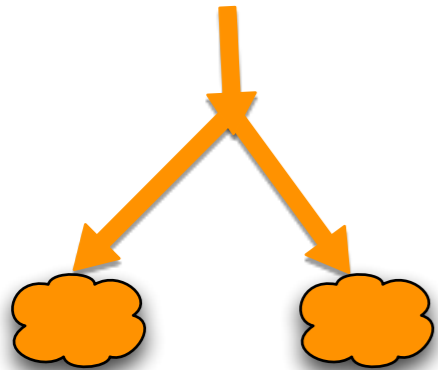
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Combinatorics, Randomness and Analysis



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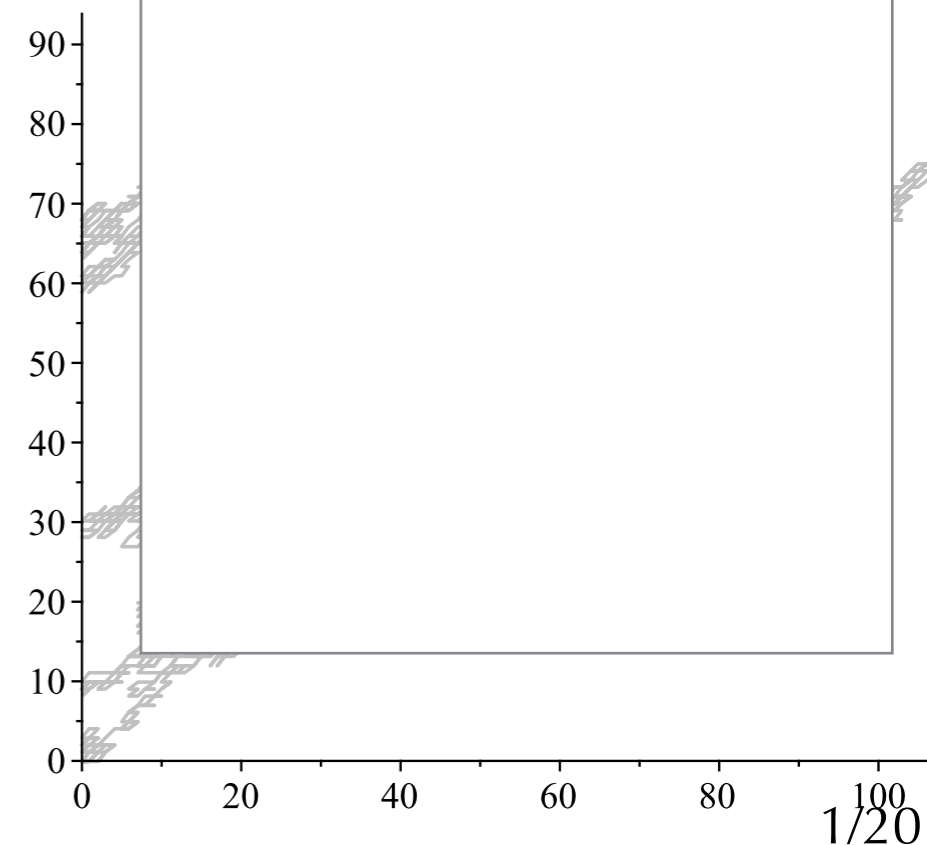
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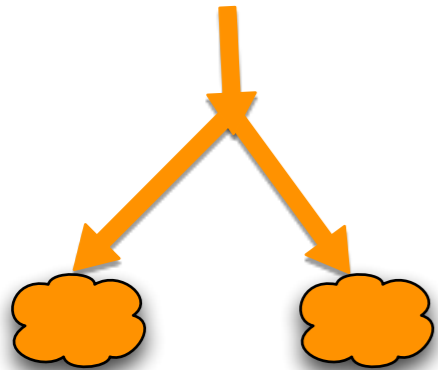
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For a tree of size n ,

- probability that a leaf is a child of the root?



Combinatorics, Randomness and Analysis



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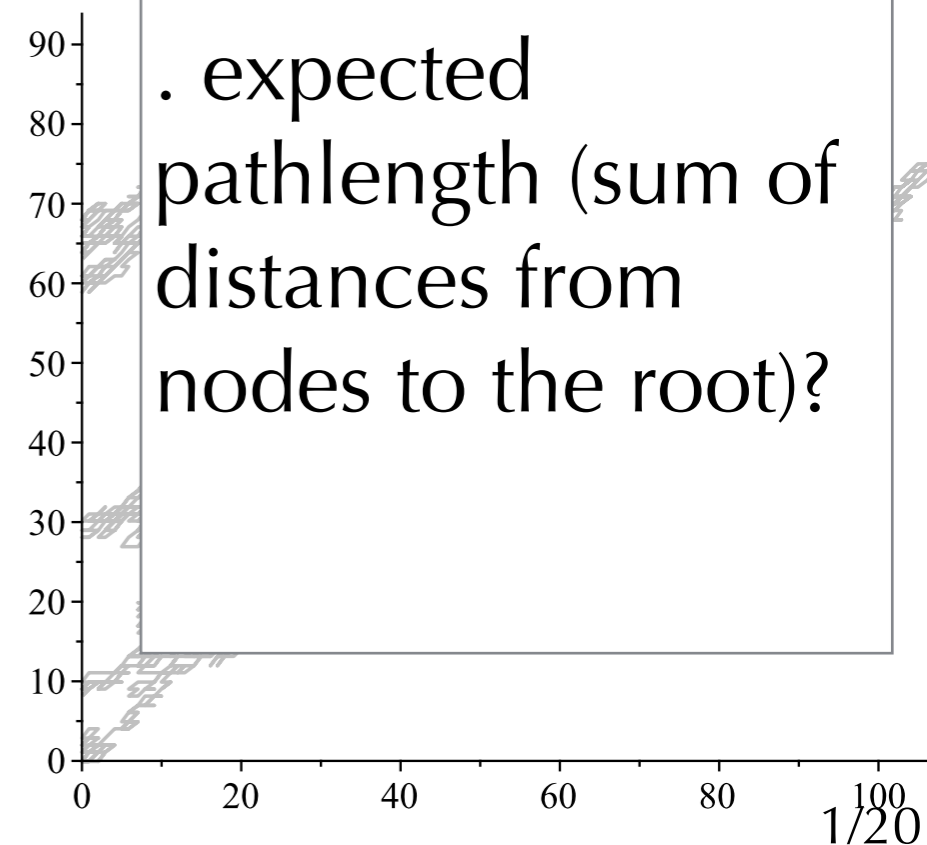
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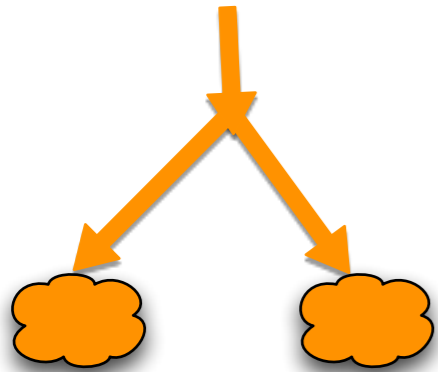
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Combinatorics, Randomness and Analysis



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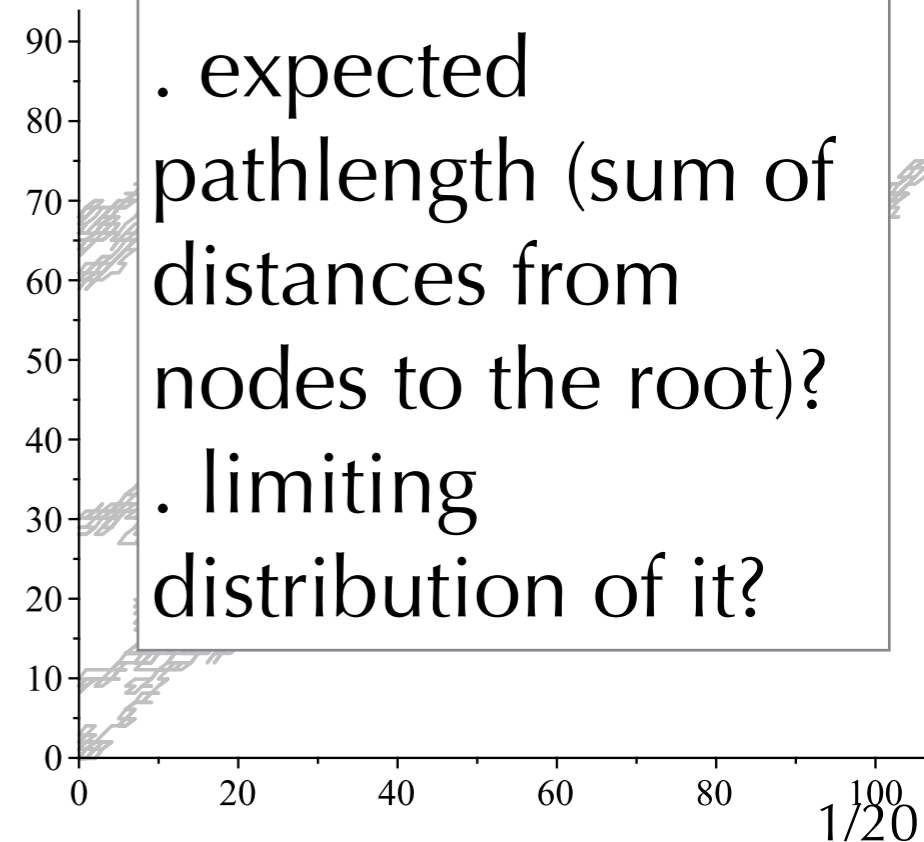
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. limiting distribution of it?

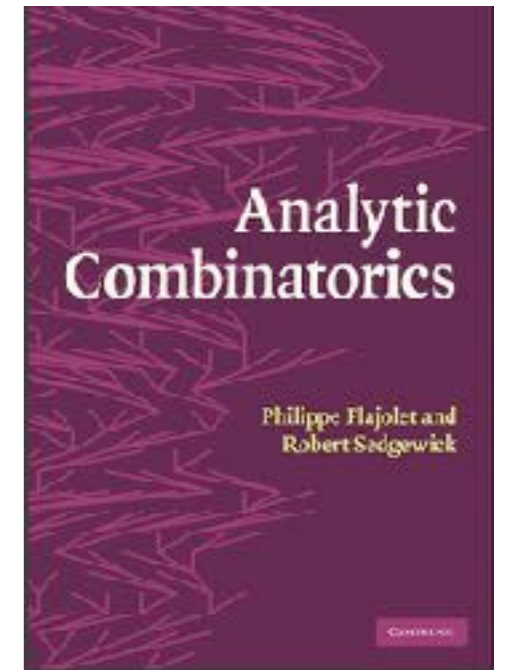


Philippe Flajolet, the Father of Analytic Combinatorics

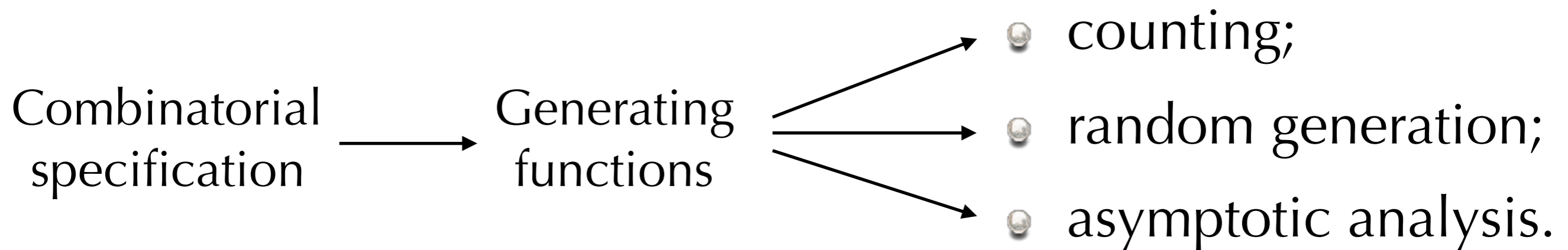


1948-2011

Planned complete works span
7 volumes of approx 600 pp. each.



2009



If you can specify, you can analyze.

Constructible Structures

Language: $1, \mathcal{L}, +, \times, \text{SEQ}, \text{SET}, \text{CYC}$
and recursion.

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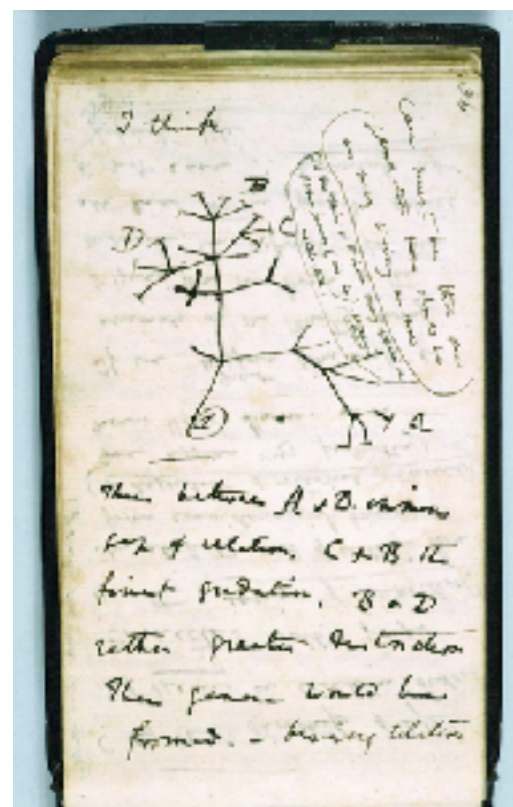
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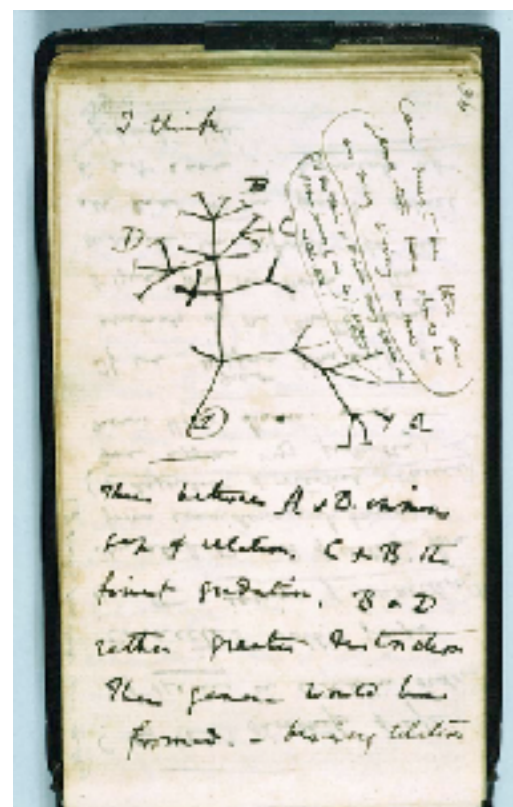


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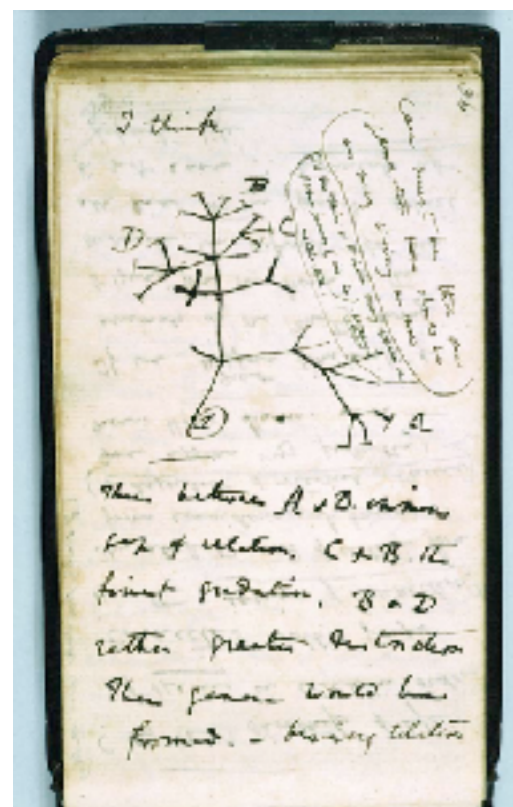
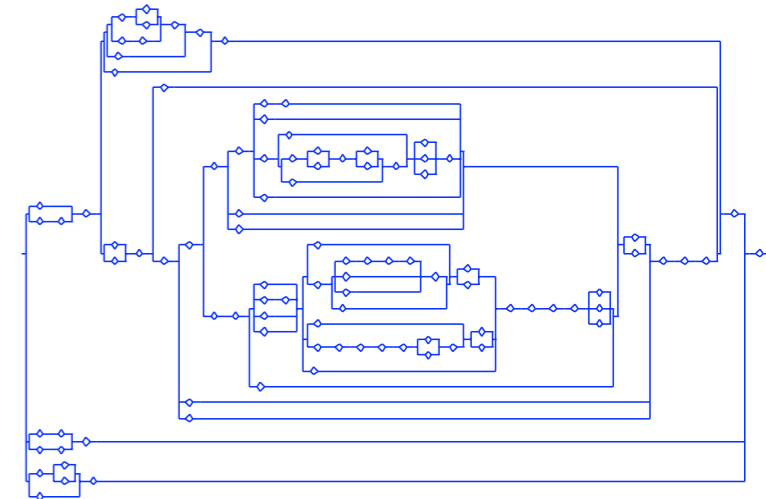


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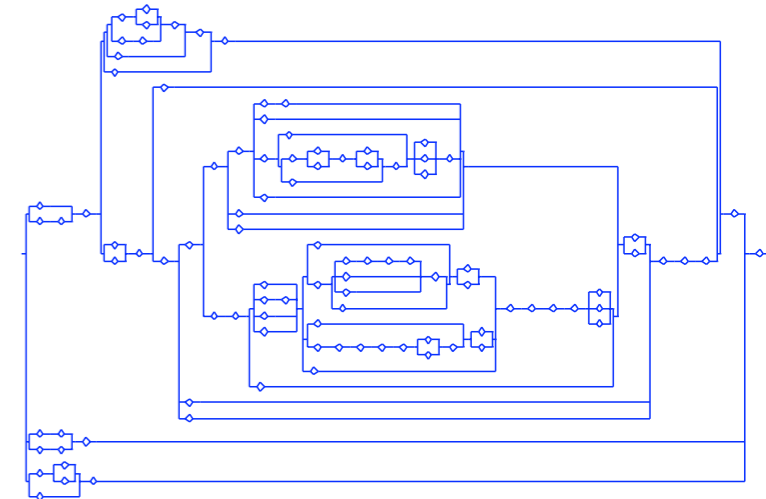


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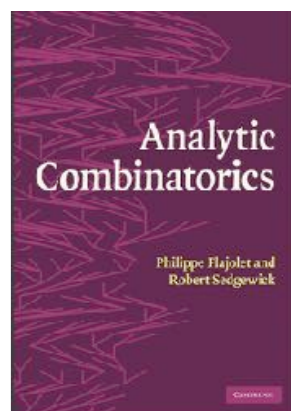
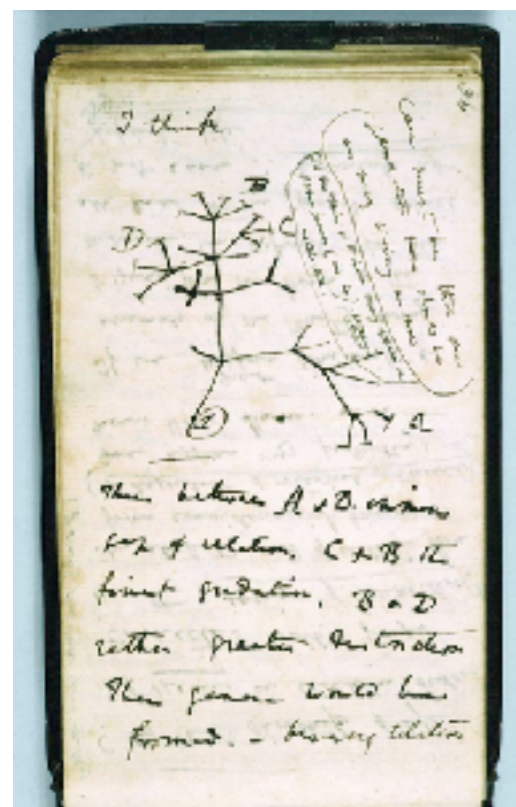


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- ...hundreds of examples in “the purple book”.



I. Enumeration and Generating Functions

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$$F(z) = \sum_{n=0}^{\infty} f_n z^n$$

Example: Binary Trees

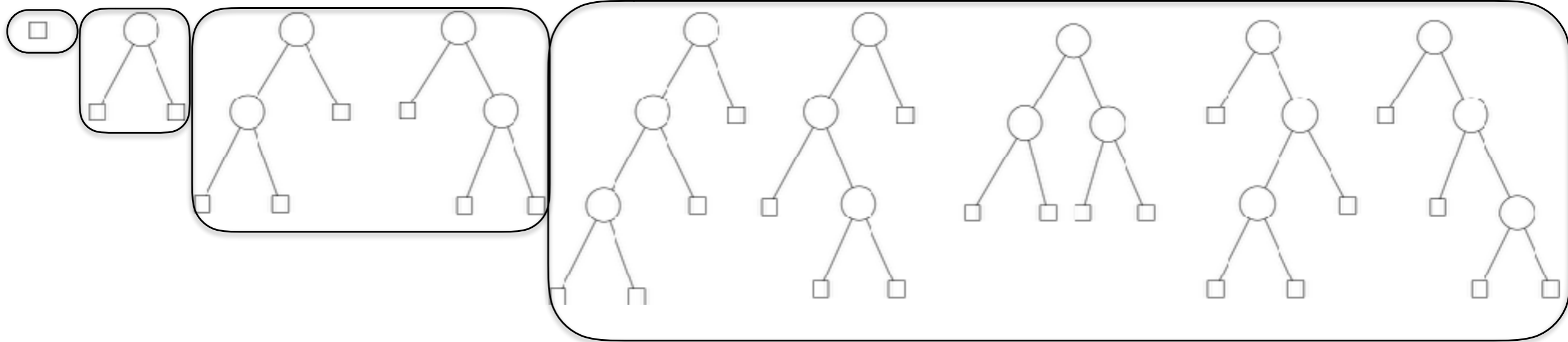
$$\mathcal{B} = 1 + \mathcal{L} \times \mathcal{B} \times \mathcal{B}$$

$$B(z) = \sum_{n=0}^{\infty} B_n z^n$$

number of binary trees with n nodes
(Catalan)

$$B_0 = B_1 = 1 \quad B_2 = 2$$

$$B_3 = 5$$



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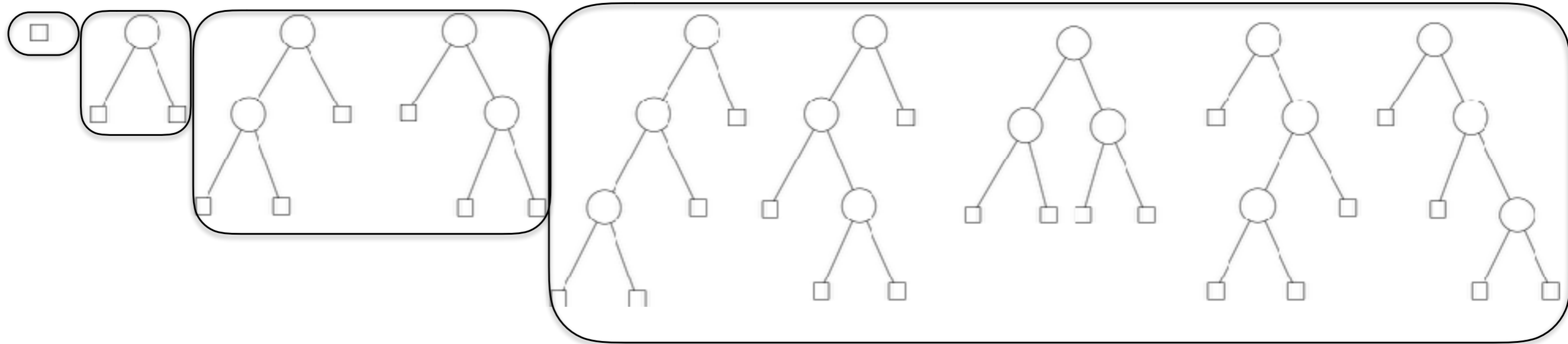
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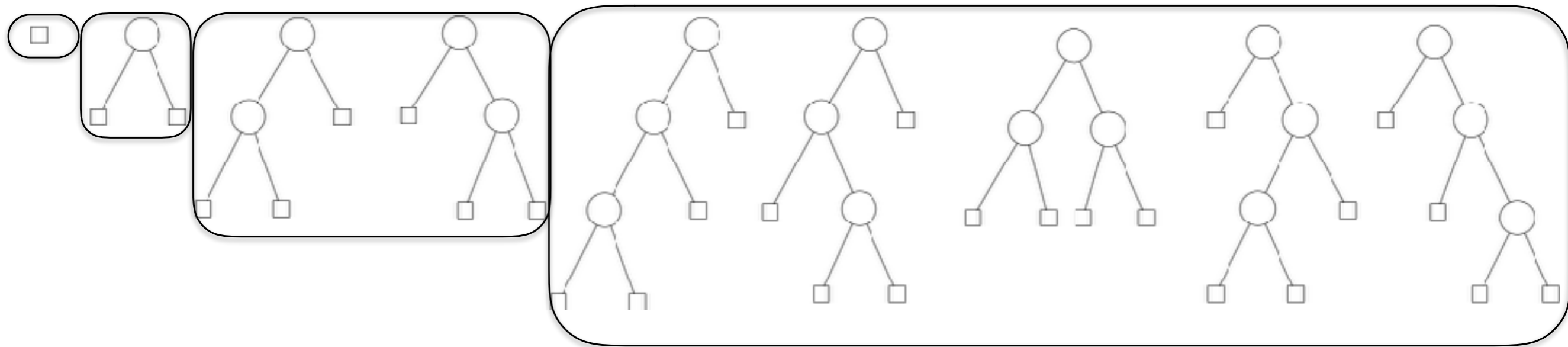
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$$\text{For } n \geq 1, \quad B_n = \sum_{k=0}^{n-1} B_k B_{n-k-1} \Rightarrow B_n z^n = \sum_{k=0}^{n-1} z (B_k z^k) (B_{n-k-1} z^{n-k-1})$$

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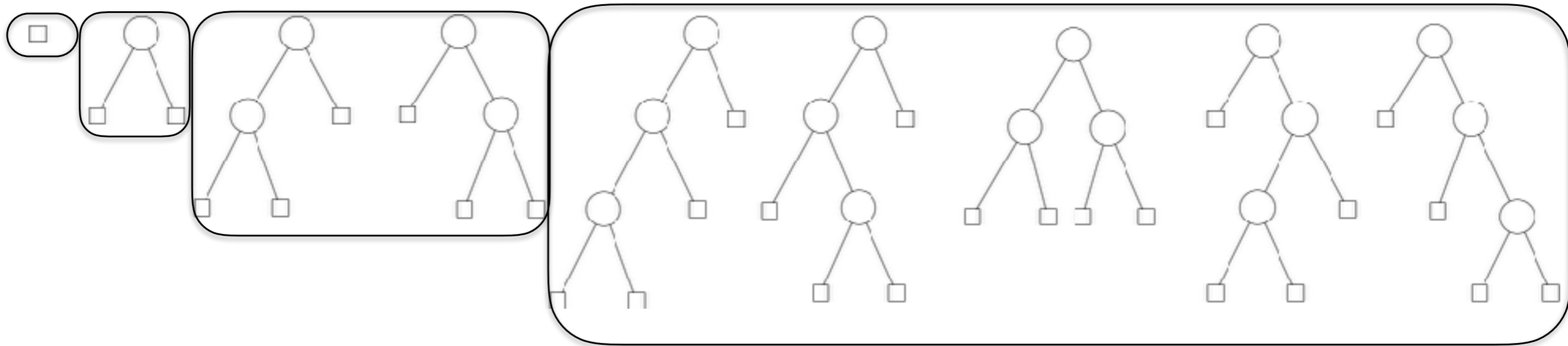
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$$B(z) = 1 + zB(z)^2$$

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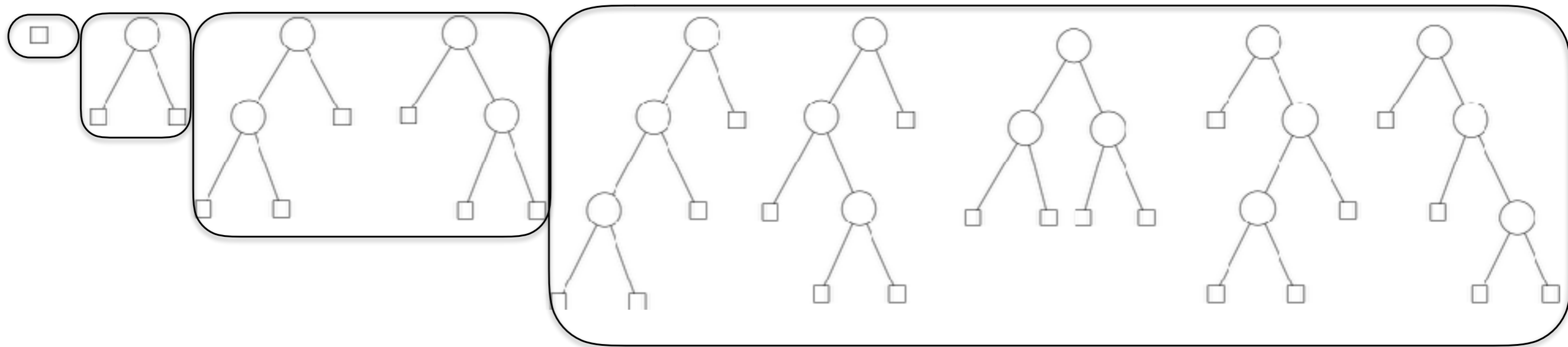
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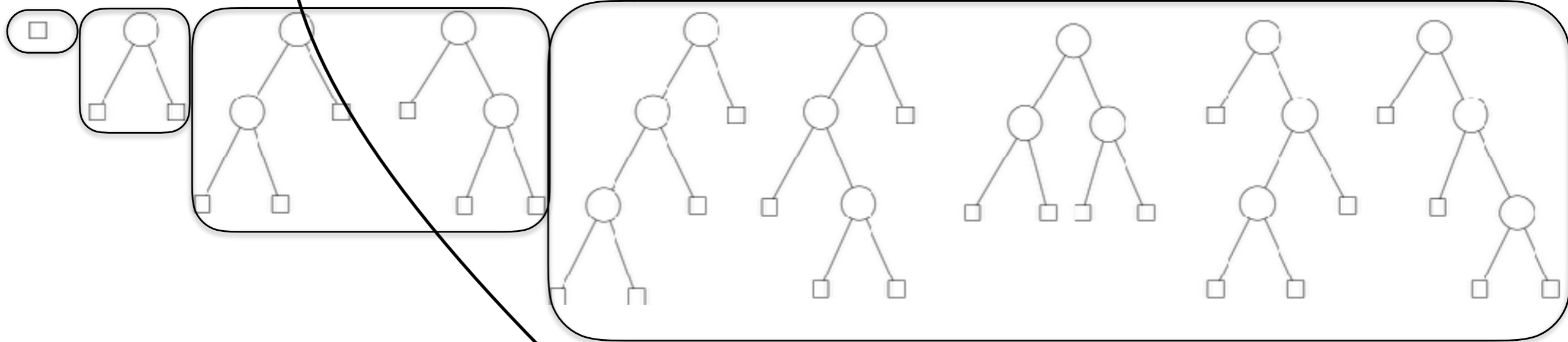
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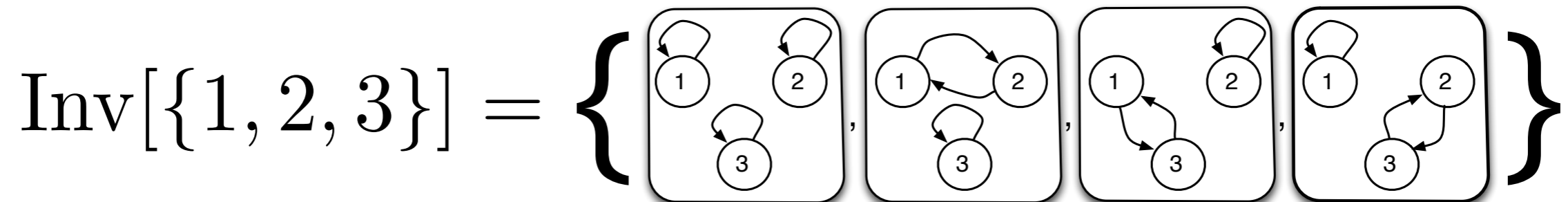
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What do we count?



4 involutions;

3 of them permuted by $\mathfrak{S}_3 \rightarrow 2$ unlabelled structures.

Exponential generating series (EGF):

$$F(z) = \sum_{n=0}^{\infty} f_n \frac{z^n}{n!}, \quad f_n = \text{nb. labelled structs of size } n.$$

$$\text{Inv}_3(z) = \frac{2}{3} z^3$$

Ordinary generating series (OGF):

$$\tilde{F}(z) = \sum_{n=0}^{\infty} \tilde{f}_n z^n, \quad \tilde{f}_n = \text{nb. unlabelled structs of size } n.$$

$$\widetilde{\text{Inv}}_3(z) = 2z^3$$

A Dictionary for Generating Series

Language: $1, \mathcal{L}, +, \times, \text{SEQ}, \text{SET}, \text{CYC}$ and recursion.

describes *constructible structures*

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Structure	EGF	OGF
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$\text{SEQ}(\mathcal{A})$	$\frac{1}{1 - A(z)}$	$\frac{1}{1 - \tilde{A}(z)}$

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$\text{SEQ}(\mathcal{A})$	$\frac{1}{1 - A(z)}$	$\frac{1}{1 - \tilde{A}(z)}$
$\text{SET}(\mathcal{A})$	$\exp(A(z))$	$\exp\left(\sum \tilde{A}(z^i)/i\right)$

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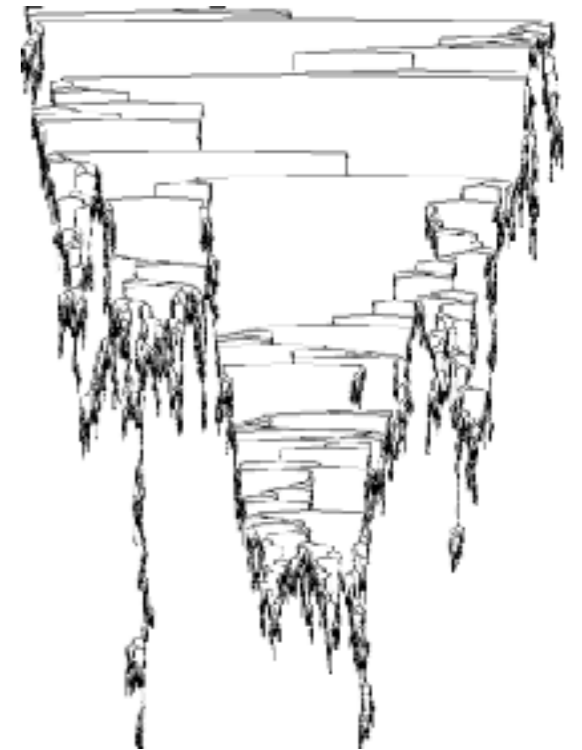
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$\text{SET}(\mathcal{A})$	$\exp(A(z))$	$\exp\left(\sum \tilde{A}(z^i)/i\right)$
$\text{CYC}(\mathcal{A})$	$\log \frac{1}{1 - A(z)}$	$\sum \frac{\phi(i)}{i} \log \frac{1}{1 - \tilde{A}(z^i)}$

Examples

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Binary trees: $\mathcal{B} = 1 + \mathcal{L} \times \mathcal{B} \times \mathcal{B}$

$$\longrightarrow B(z) = 1 + zB(z)^2 = \tilde{B}(z)$$



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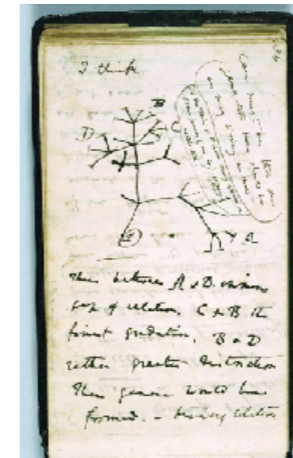
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Cayley trees: $\mathcal{T} = \mathcal{L} \times \text{SET}(\mathcal{T})$

$$\longrightarrow T(z) = z \exp(T(z));$$

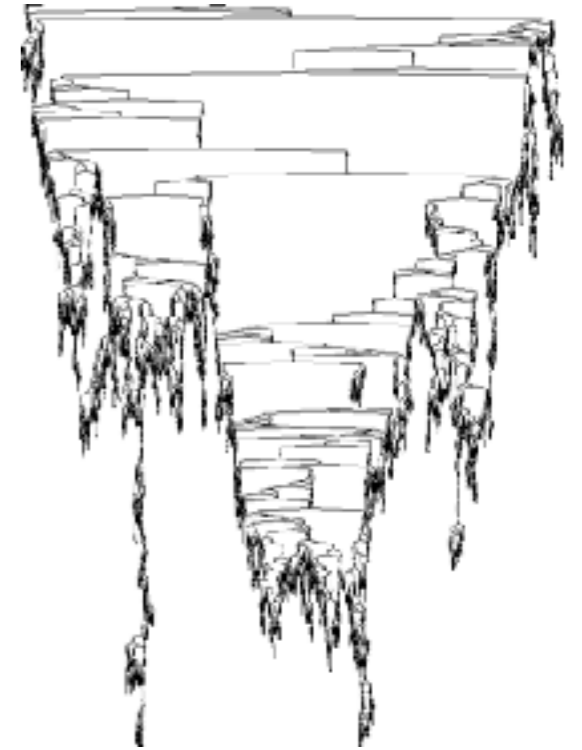
$$\longrightarrow \tilde{T}(z) = z \exp\left(\tilde{T}(z) + \frac{1}{2}\tilde{T}(z^2) + \frac{1}{3}\tilde{T}(z^3) + \dots\right)$$



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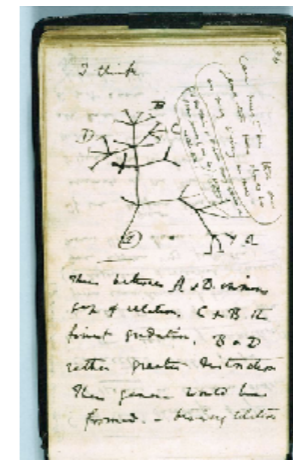
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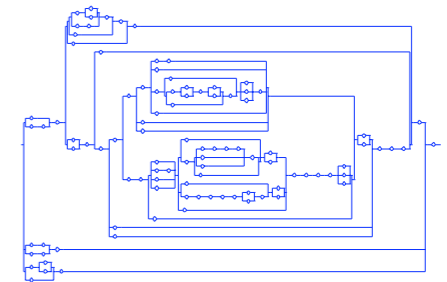
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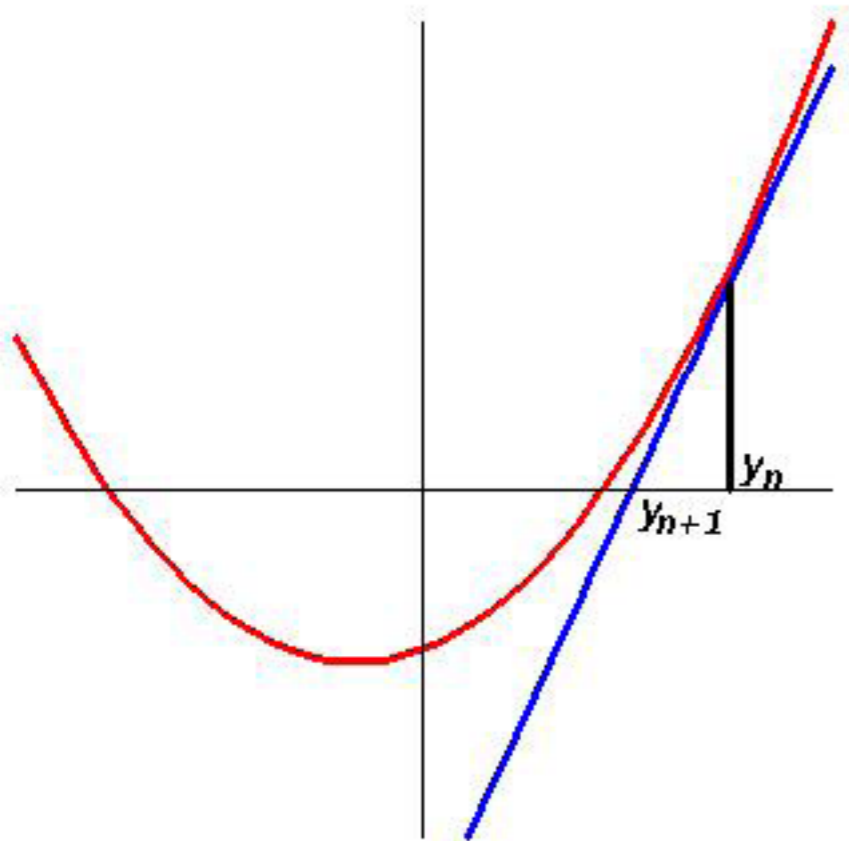
Series-parallel graphs:

$$\mathcal{G} = \mathcal{Z} + \mathcal{S} + \mathcal{P}, \mathcal{S} = \text{SEQ}_{>0}(\mathcal{Z} + \mathcal{P}), \mathcal{P} = \text{SET}_{>0}(\mathcal{Z} + \mathcal{S})$$



$$\longrightarrow \left\{ G(z) = z + S(z) + P(z), S(z) = \frac{1}{1 - z - P(z)} - 1, P(z) = e^{z + S(z)} - 1 \right\}$$

II. Newton Iteration and Fast Enumeration



$$y^3 + a^2y - 2a^3 + axy - x^3 = 0. \quad y = a - \frac{x}{4} + \frac{x^2}{64a} + \frac{111x^3}{512a^2} + \frac{509x^4}{16384a^3} \quad \&c.$$

$+ a + p = y.$	$+y^3$ $+axy$ $+x^2y$ $-x^3$ $-2a^3$	$+a^3 + 3a^2p + 3ap^2 + p^3$ $+a^2x + axp$ $+x^3 + a^2p$ $-x^3$ $-2a^3$
$-\frac{1}{4}x + q = p.$	$+p^3$ $+3ap^2$ $+xsp$ $+4a^2p$ $+a^2x$ $-x^3$	$-\frac{1}{64}x^3 + \frac{1}{16}x^2q - \frac{1}{4}xq^2 + q^3$ $+\frac{1}{16}ax^2 - \frac{1}{4}axq + 3aq^2$ $-\frac{1}{4}ax^2 + axq$ $-a^2x + 4a^2q$ $+a^2x$ $-x^3$
$+\frac{x^2}{64a} + r = q.$	$+q^3$ $-\frac{1}{4}xq^2$ $+3aq^2$ $+\frac{1}{16}x^2q$ $-\frac{1}{4}axq$ $+4a^2q$ $-\frac{6}{64}x^3$ $-\frac{1}{16}ax^2$	$*$ $*$ $+\frac{3x^4}{4096a} * + \frac{1}{16}x^2r + 3ar^2$ $+\frac{3x^4}{1024a} * + \frac{1}{16}x^2r$ $-\frac{1}{16}x^3 - \frac{1}{4}axr$ $+\frac{1}{16}ax^2 + 4a^2r$ $-\frac{6}{64}x^3$ $-\frac{1}{16}ax^2$
$* + 4a^2 - \frac{1}{4}ax + \frac{1}{16}x^2 + \frac{111}{128}x^3 - \frac{15x^4}{4096a} + \left(+ \frac{131x^3}{512a^2} + \frac{509x^4}{16384a^3} \right) \quad \&c.$		

1671

Numerical Newton Iteration

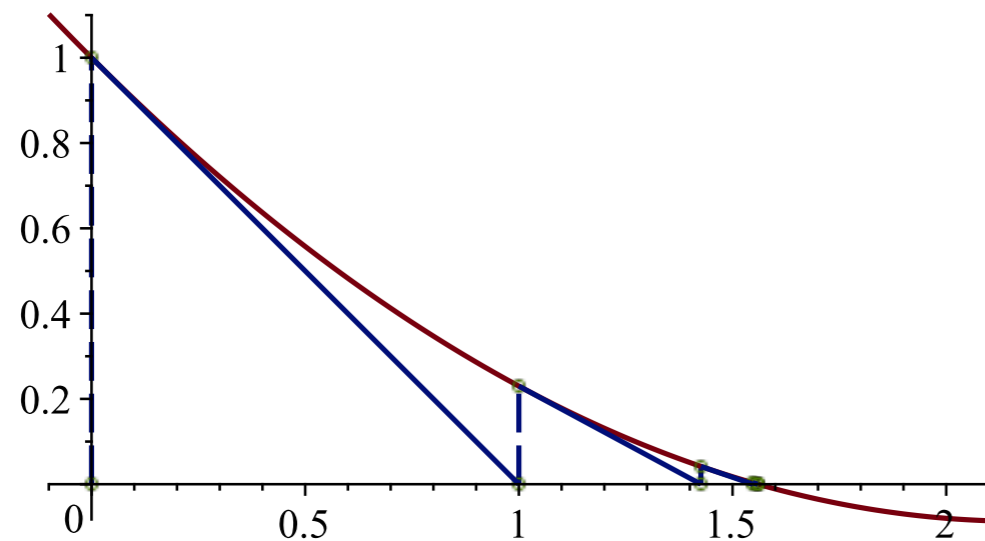
To solve $\phi(y) = 0$, iterate

$$y^{[n+1]} = y^{[n]} + u^{[n]},$$

with $\phi(y^{[n]}) + \phi'(y^{[n]})u^{[n]} = 0$.

$$\phi(y) = 1 + zy^2 - y$$

$$z = 0.23$$



Numerical Newton Iteration

To solve $\phi(y) = 0$, iterate

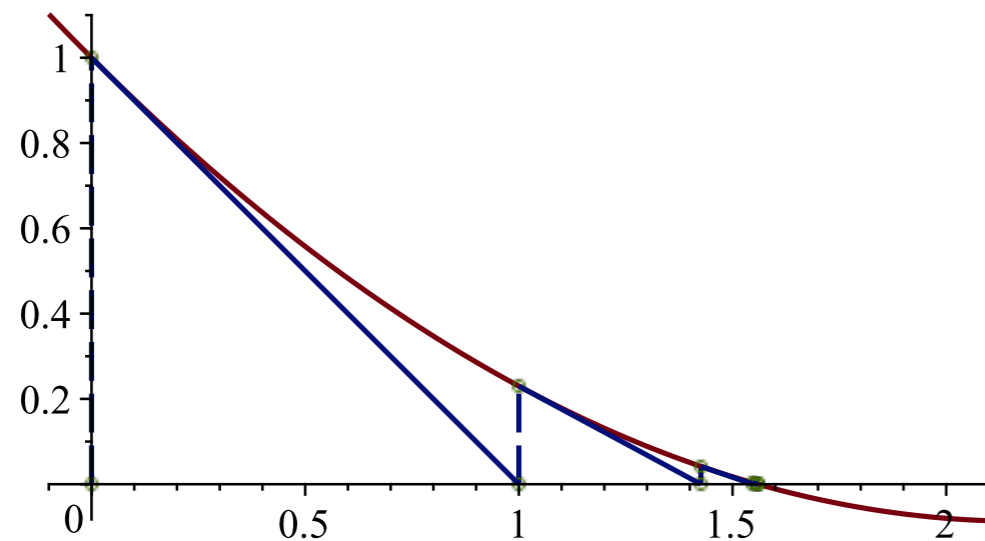
$$y^{[n+1]} = y^{[n]} + u^{[n]},$$

with $\phi(y^{[n]}) + \phi'(y^{[n]})u^{[n]} = 0$.

$$\phi(y) = 1 + zy^2 - y$$

$$y^{[n+1]} = \mathcal{N}(y^{[n]}) = y^{[n]} + \frac{1 + zy^{[n]2} - y^{[n]}}{1 - 2zy^{[n]}}$$

$$z = 0.23$$



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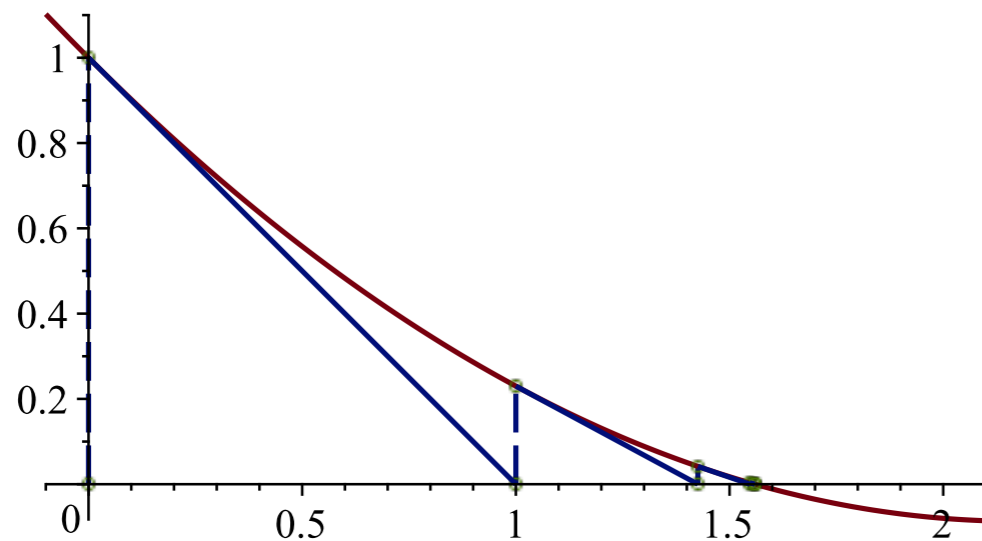
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$$z = 0.23$$



Quadratic convergence

$$y^{[0]} = 0,$$

$$y^{[1]} = 1.0000000000000000,$$

$$y^{[2]} \simeq 1.4259259259259259,$$

$$y^{[3]} \simeq 1.5471933181836303,$$

$$y^{[4]} \simeq 1.5589256602748822,$$

$$y^{[5]} \simeq 1.5590375713926592,$$

$$y^{[6]} \simeq 1.5590375815769151$$

Newton Iteration for Power Series

Same
Newton
Iteration

$$y^{[n+1]} = \mathcal{N}(y^{[n]}) = y^{[n]} + \frac{1 + zy^{[n]^2} - y^{[n]}}{1 - 2zy^{[n]}}$$

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$$y^{[0]} = 0$$

$$y^{[1]} = 1$$

$$y^{[2]} = 1 + z + 2z^2 + 4z^3 + 8z^4 + 16z^5 + 32z^6 + 64z^7 + \dots$$

$$y^{[3]} = 1 + z + 2z^2 + 5z^3 + 14z^4 + 42z^5 + 132z^6 + 428z^7 + \dots$$

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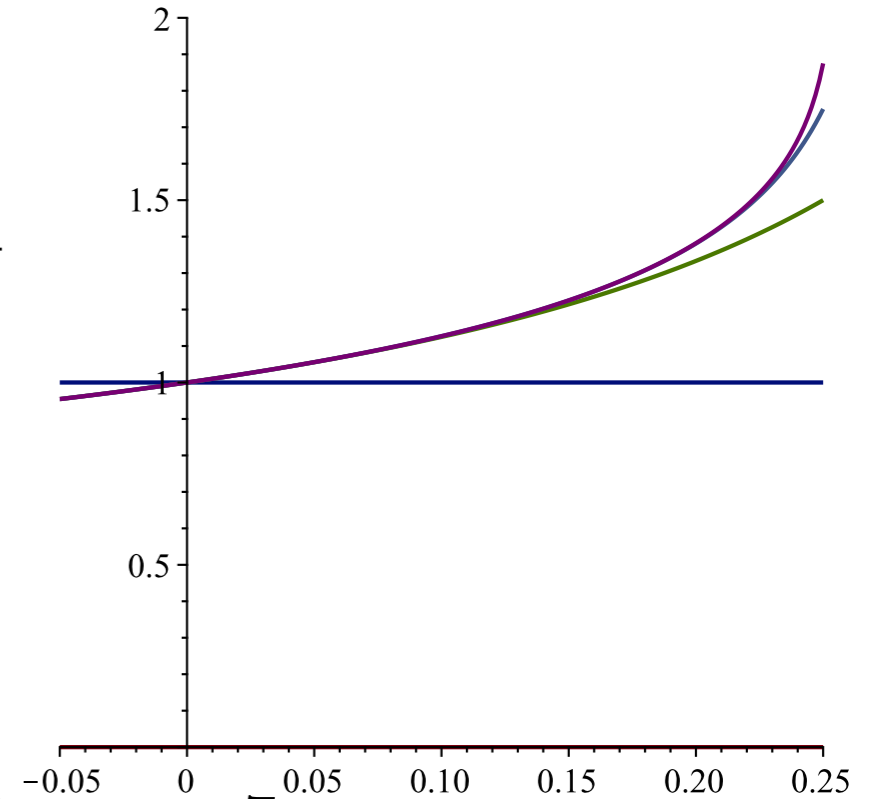
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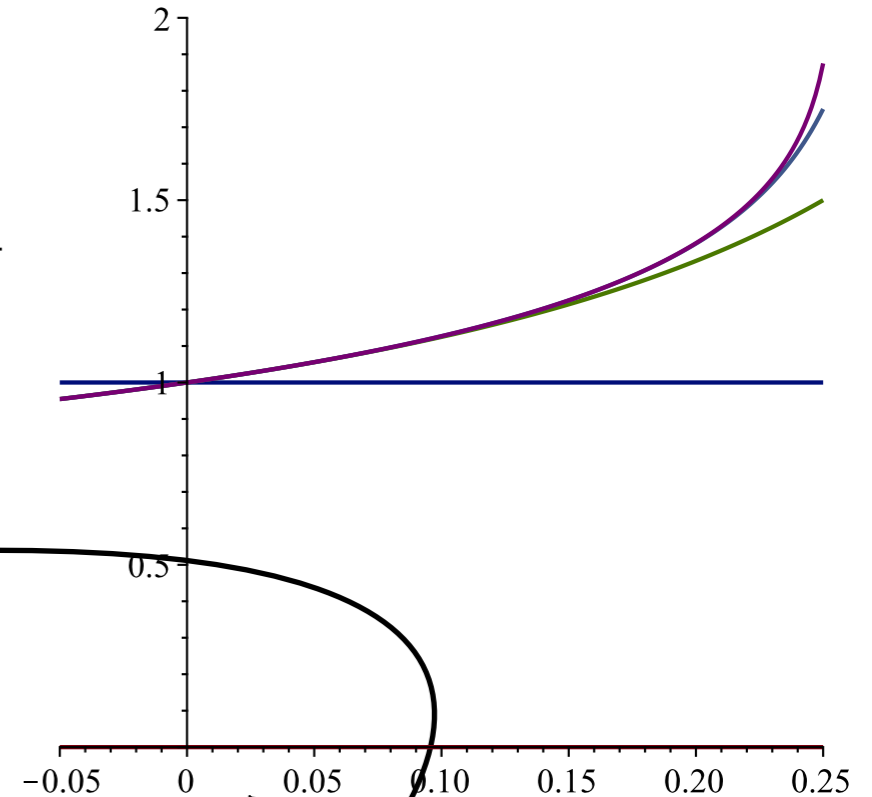
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Proving numerical convergence requires control over the tails

$$y^{[0]} = 0$$

$$y^{[1]} = 1$$

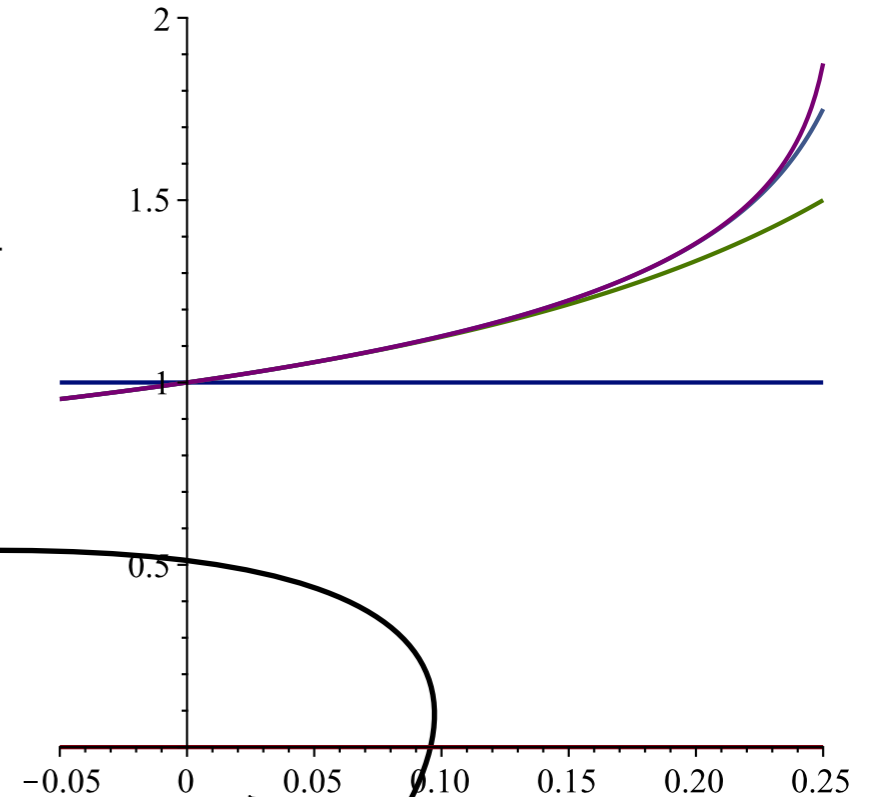
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On power series: $y - y^{[\infty]} = O(z^m) \implies \mathcal{N}(y) - y^{[\infty]} = O(z^{2m(+1)})$

Newton Iteration for Power Series

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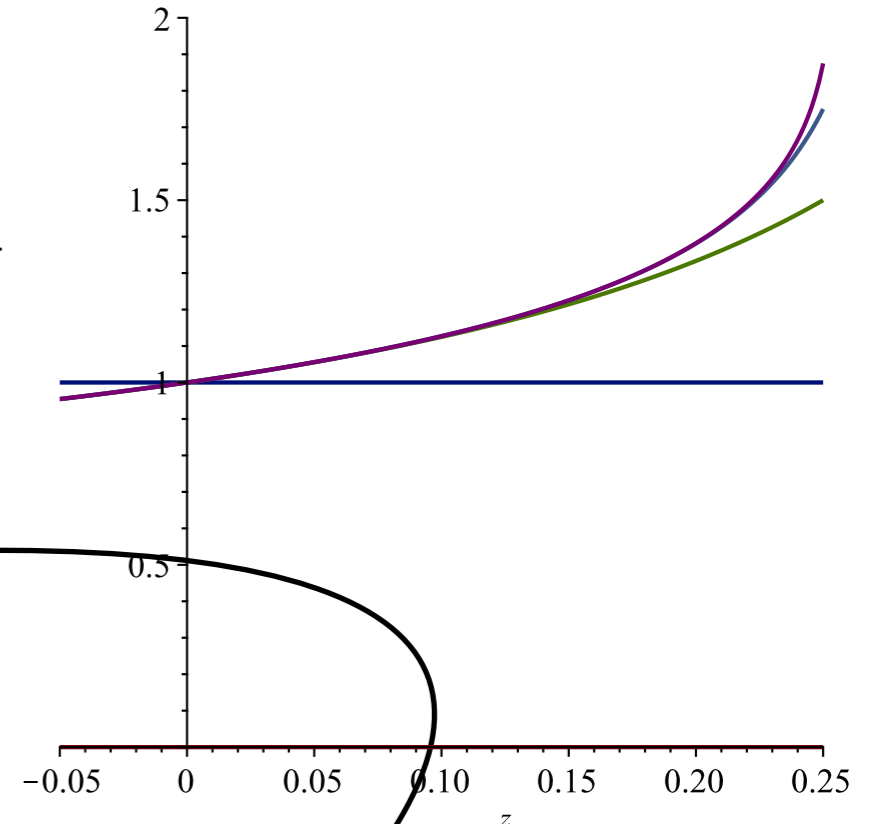
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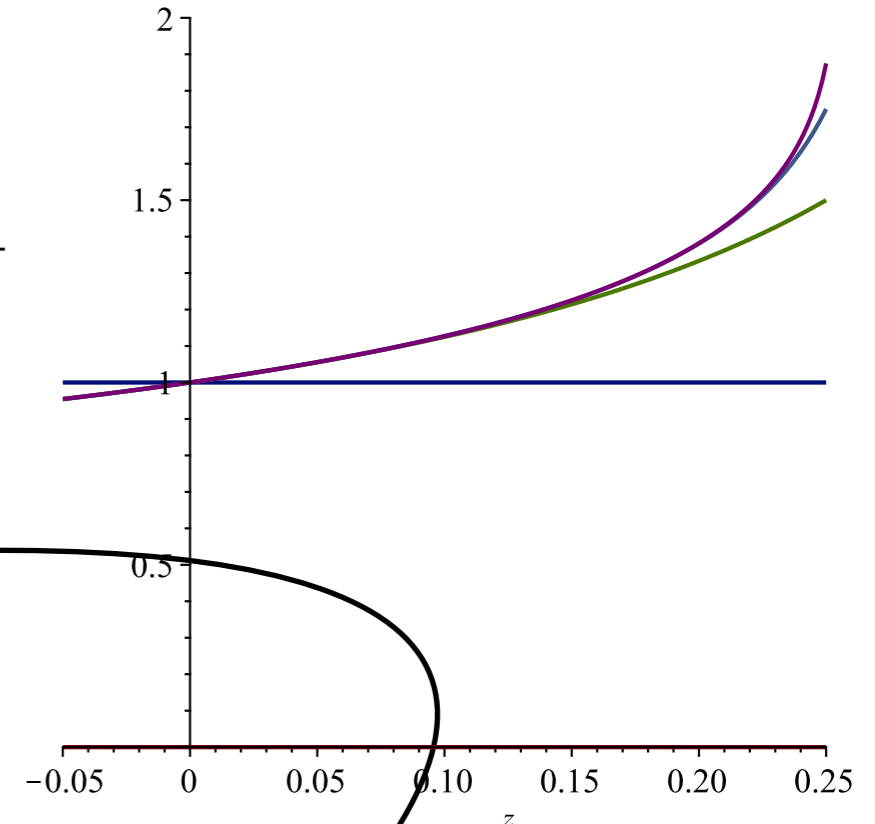
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Proving numerical
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On power series: $y - y^{[\infty]} = O(z^m) \implies \mathcal{N}(y) - y^{[\infty]} = O(z^{2m(+1)})$

```
Expand(N) = {
  res=Expand(N/2);
  a=phi(res); b=phi'(res);
  u=Solve(a+bx,x);
  return res+u; }
```

Newton Iteration for Power Series

Same
Newton
Iteration

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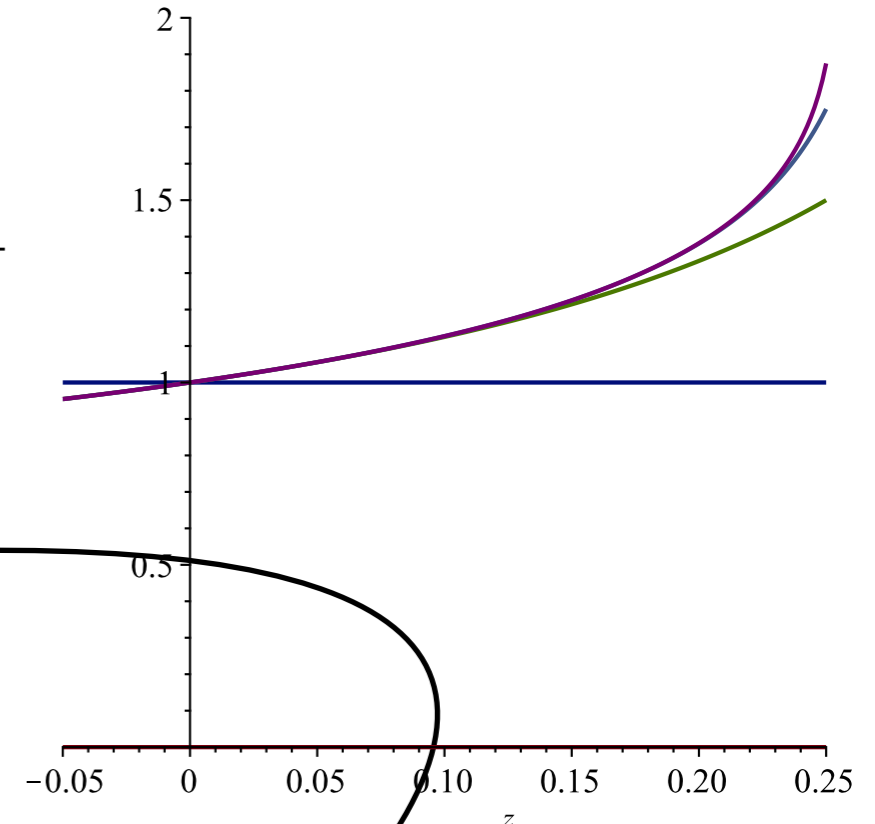
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On power series: $y - y^{[\infty]} = O(z^m) \implies \mathcal{N}(y) - y^{[\infty]} = O(z^{2m+1})$

```
Expand(N) = {
  res=Expand(N/2);
  a=phi(res); b=phi'(res);
  u=Solve(a+bx,x);
  return res+u; }
```

$\text{Cost}(N) \leq ct \times \text{Cost}(\text{last step})$

Fast Enumeration

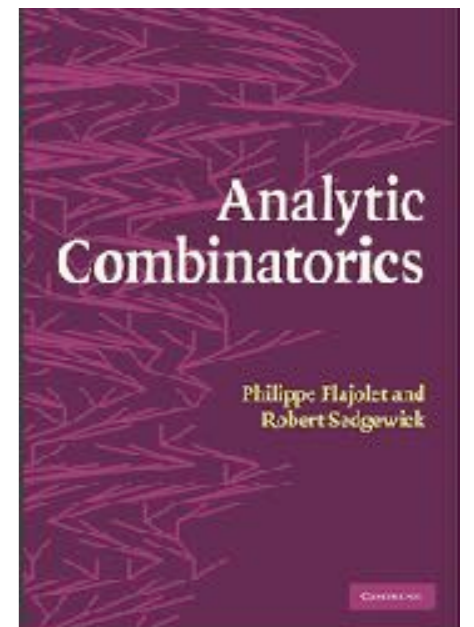
Thm. First N coefficients of GFs of all *constructible* structures:

1. arithmetic complexity $O(N \log N)$ (both ogf & egf);

2. bit complexity

$O(N^2 \log^2 N \log \log N)$ (ogf); $O(N^2 \log^3 N \log \log N)$ (egf).

Ingredients: Newton iteration & FFT.



Fast Enumeration

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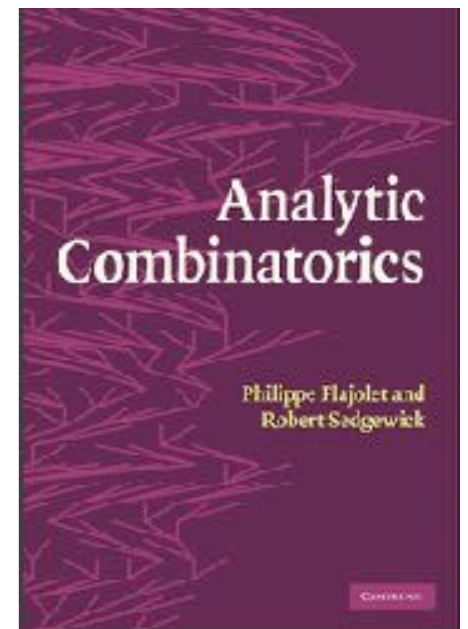
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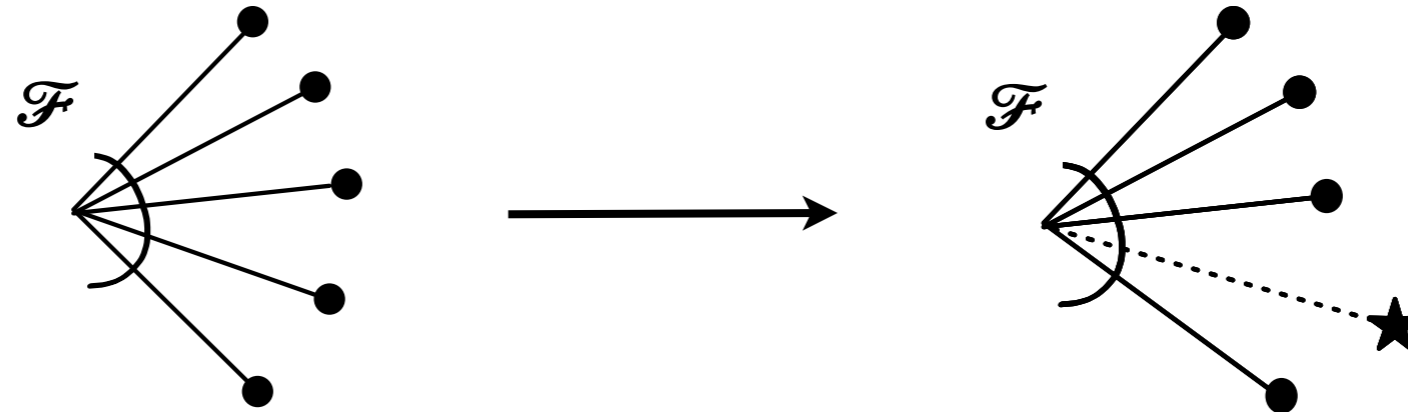
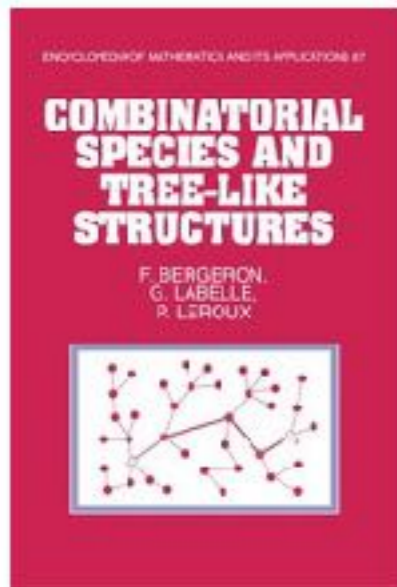
```
> with(NewtonGF):  
> BinTrees := {B = Union(Epsilon, Prod(Z, B, B))}:  
> GFSeries(BinTrees, labelled, z, 21)[2];  
B = 1 + z + 2 z2 + 5 z3 + 14 z4 + 42 z5 + 132 z6 + 429 z7 + 1430 z8  
+ 4862 z9 + 16796 z10 + 58786 z11 + 208012 z12 + 742900 z13  
+ 2674440 z14 + 9694845 z15 + 35357670 z16 + 129644790 z17  
+ 477638700 z18 + 1767263190 z19 + 6564120420 z20 + O(z21)
```



Demo
NewtonGF

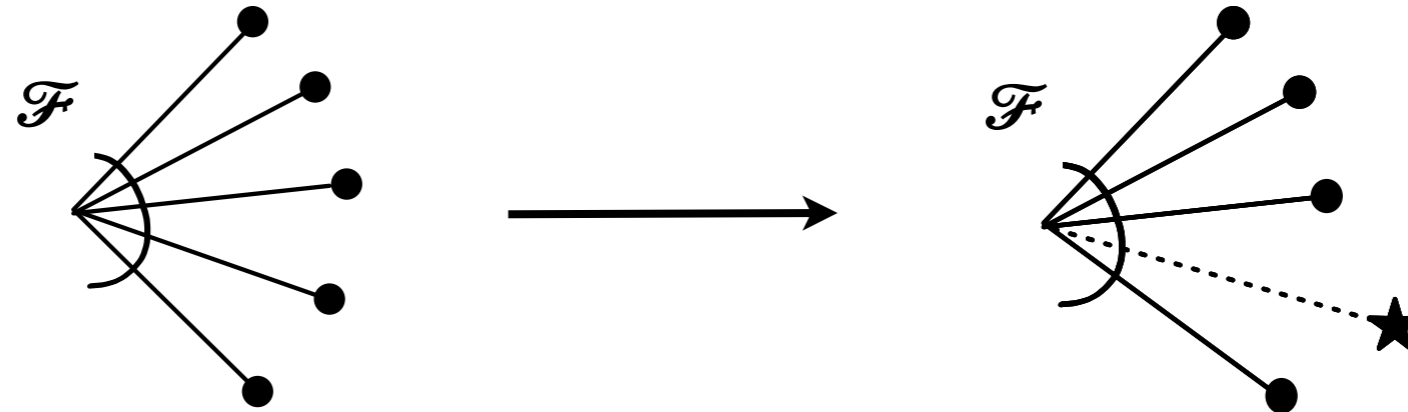
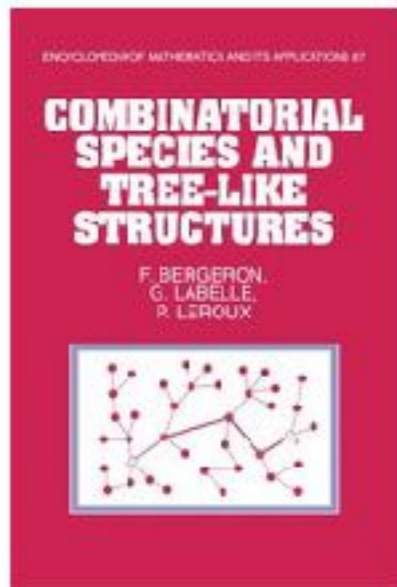
Derivative of Combinatorial Structures

(origin: species theory)



Derivative of Combinatorial Structures

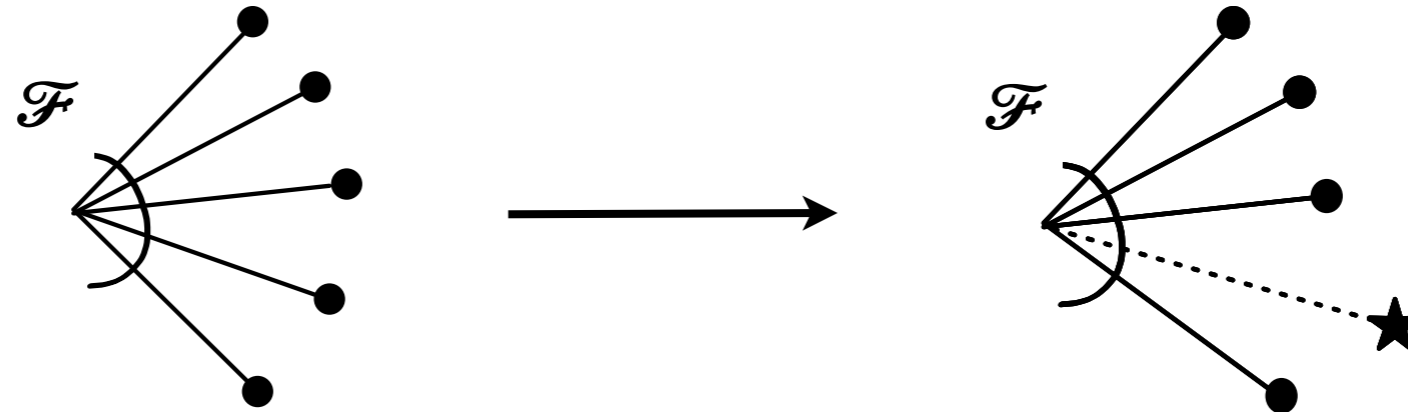
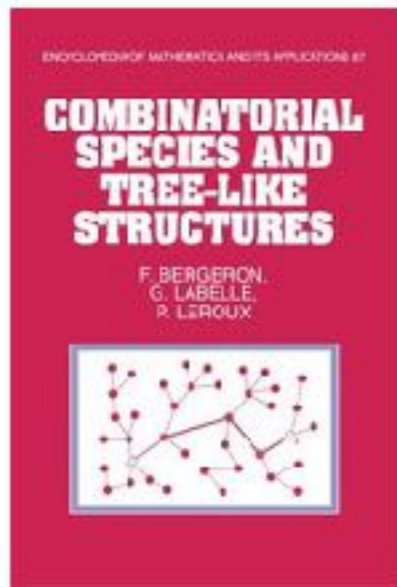
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• $(\mathcal{F} + \mathcal{G})' = \mathcal{F}' + \mathcal{G}'; (\mathcal{F} \times \mathcal{G})' = \mathcal{F}' \times \mathcal{G} + \mathcal{F} \times \mathcal{G}';$

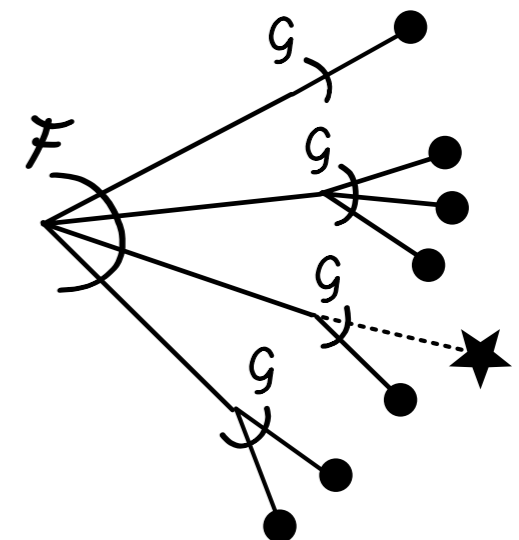
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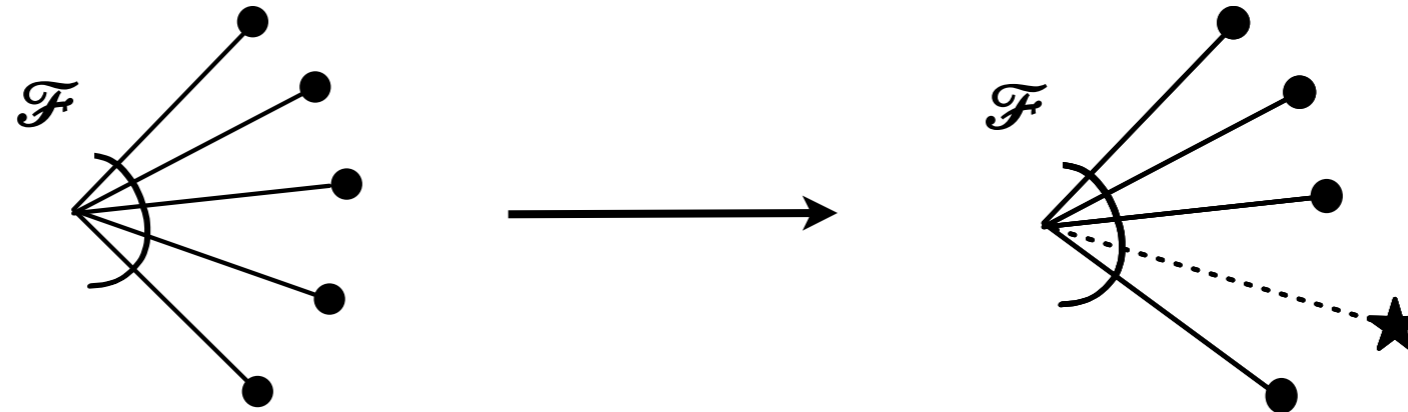
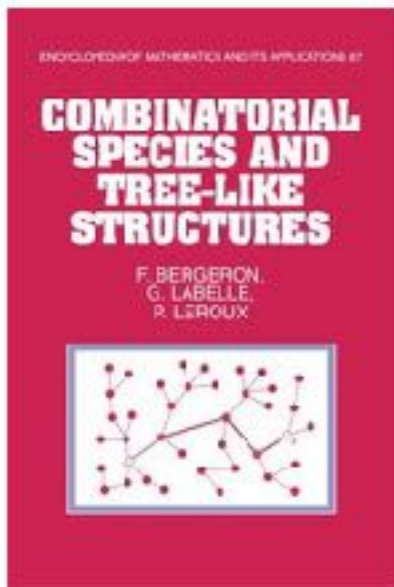
$$\bullet (\mathcal{F} + \mathcal{G})' = \mathcal{F}' + \mathcal{G}'; (\mathcal{F} \times \mathcal{G})' = \mathcal{F}' \times \mathcal{G} + \mathcal{F} \times \mathcal{G}';$$

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Derivative of Combinatorial Structures

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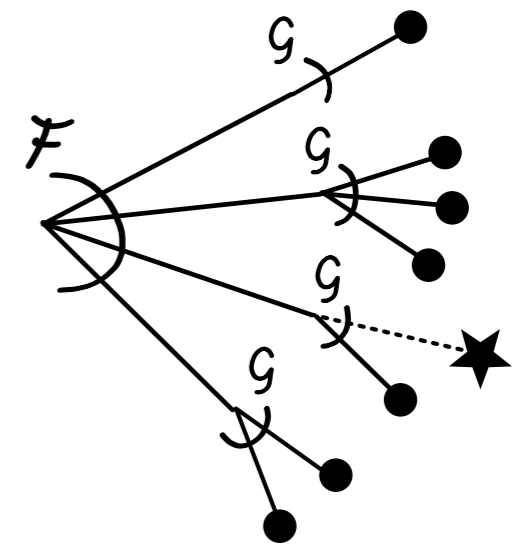


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- $\mathcal{F}(\mathcal{G})' = \mathcal{F}'(\mathcal{G}) \times \mathcal{G}';$

- $0' = 1' = 0; \mathcal{L}' = 1;$

$\text{SET}' = \text{SET}; \text{CYC}' = \text{SEQ}; \text{SEQ}' = \text{SEQ} \times \text{SEQ}.$



Combinatorial Newton Iteration

$$\mathcal{Y} = 1 + \mathcal{Z} \times \mathcal{Y} \times \mathcal{Y} =: \mathcal{H}(\mathcal{Z}, \mathcal{Y})$$

$$\mathcal{Y}_0 = \emptyset \quad \mathcal{Y}_1 = \circ$$

$$\mathcal{Y}_2 = \left[\begin{array}{c} \circ \\ + \\ \begin{array}{c} \bullet \\ / \backslash \\ \circ \quad \circ \end{array} \end{array} \right] + \left[\begin{array}{c} \bullet \\ / \backslash \\ \circ \quad \bullet \\ / \backslash \\ \circ \quad \circ \end{array} \right] + \left[\begin{array}{c} \bullet \\ / \backslash \\ \circ \quad \bullet \\ / \backslash \\ \bullet \quad \circ \end{array} \right] + \dots + \left[\begin{array}{c} \bullet \\ / \backslash \\ \bullet \quad \bullet \quad \bullet \quad \bullet \\ / \backslash \quad / \backslash \\ \circ \quad \circ \quad \circ \quad \circ \end{array} \right] + \dots$$

②

$$\mathcal{Y}_3 = \left[\mathcal{Y}_2 + \left[\begin{array}{c} \bullet \\ / \backslash \\ \bullet \quad \bullet \\ / \backslash \\ \circ \quad \circ \end{array} \right] + \dots + \left[\begin{array}{c} \bullet \\ / \backslash \\ \bullet \quad \bullet \\ / \backslash \\ \bullet \quad \bullet \\ / \backslash \\ \circ \quad \circ \end{array} \right] + \dots \right] + \left[\begin{array}{c} \bullet \\ / \backslash \\ \bullet \quad \bullet \quad \bullet \quad \bullet \\ / \backslash \quad / \backslash \\ \bullet \quad \bullet \quad \bullet \quad \bullet \\ / \backslash \quad / \backslash \\ \circ \quad \circ \quad \circ \quad \circ \end{array} \right] + \dots$$

⑥

Combinatorial Newton Iteration

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$$\mathcal{Y}^{[n+1]} = \mathcal{Y}^{[n]} + \text{SEQ}(\mathcal{Z} \cdot \mathcal{Y}^{[n]} \cdot \star + \mathcal{Z} \cdot \star \cdot \mathcal{Y}^{[n]}) \cdot ((1 + \mathcal{Z} \cdot \mathcal{Y}^{[n]} \cdot \mathcal{Y}^{[n]}) \setminus \mathcal{Y}^{[n]}).$$

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$$\left(\text{Id} - \frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]}) \right)^{-1}$$

Combinatorial Newton Iteration

$$\mathcal{Y} = 1 + \mathcal{Z} \times \mathcal{Y} \times \mathcal{Y} =: \mathcal{H}(\mathcal{Z}, \mathcal{Y})$$

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$$\mathcal{Y}_2 = \boxed{\text{Diagram 1}} + \text{Diagram 2} + \text{Diagram 3} + \dots + \text{Diagram 4} + \dots$$

$\left(\text{Id} - \frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]}) \right)^{-1}$

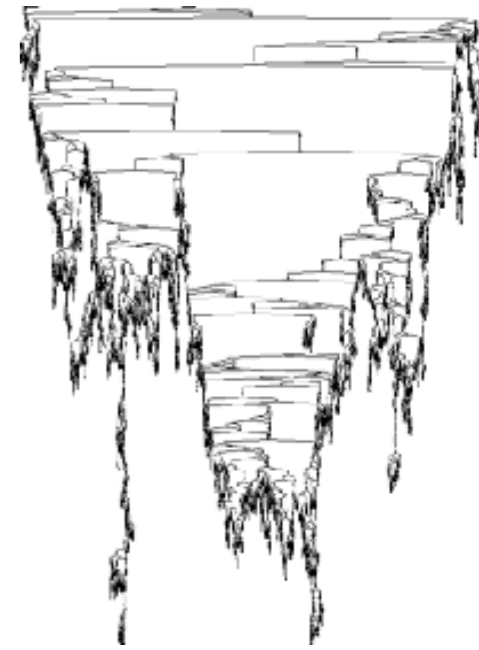
$$\mathcal{Y}_3 = \boxed{\mathcal{Y}_2 + \text{Diagram 5} + \dots + \text{Diagram 6} + \dots} + \text{Diagram 7} + \dots$$

Ccl: numerical convergence of the Newton iteration starting from 0.

III. Random Generation

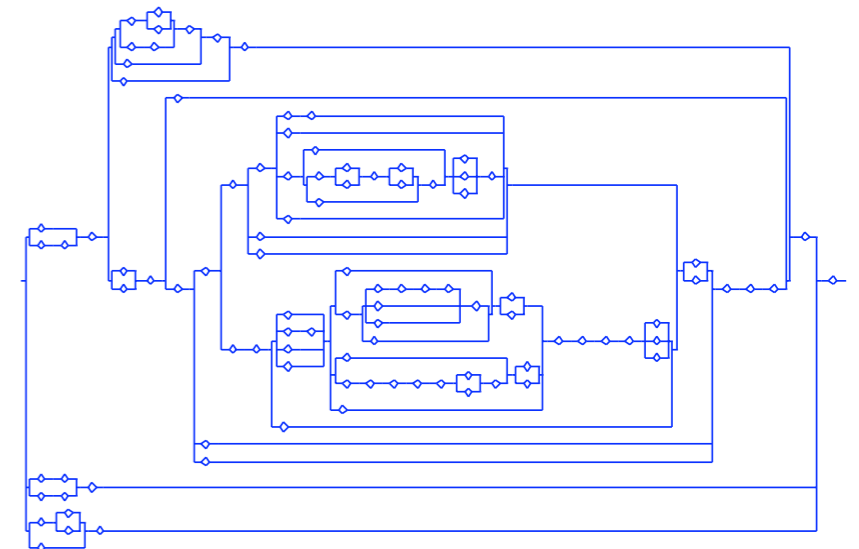
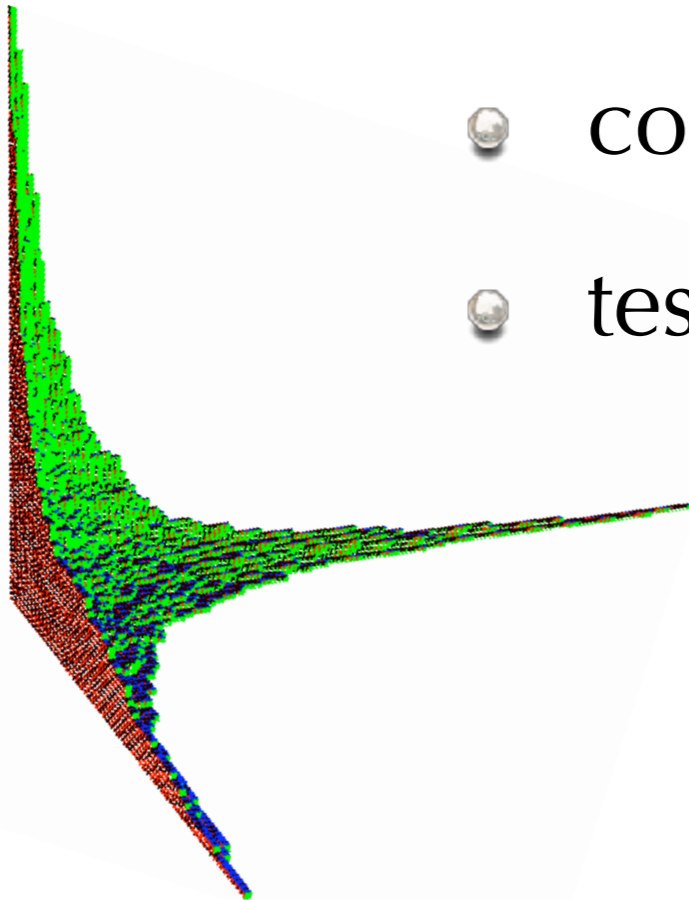


Why Random Generation?



Simulation in the discrete world; helps

- evaluate parameters;
- compare/refine models;
- test software.



Boltzmann Sampling

$$\text{Proba}(t) = \frac{x^{|t|}}{T(x)} \quad \text{with} \quad T(x) = \sum_{t \in \mathcal{T}} x^{|t|} = \sum_n T_n x^n$$

x is a parameter of the algorithm

Boltzmann Sampling

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```
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  if  $\mathcal{H} = 1$  return 1;  
  if  $\mathcal{H} = \mathcal{L}$  return  $\mathcal{L}$ ;  
  if  $\mathcal{H} = \mathcal{F} \times \mathcal{G}$  {  
    Generate( $f \in \mathcal{F}$ );  
    Generate( $g \in \mathcal{G}$ );  
    return (f, g); }  
  if  $\mathcal{H} = \mathcal{F} + \mathcal{G}$  {  
    b := Bernoulli( $F(x)/H(x)$ );  
    if b=1 return Generate( $f \in \mathcal{F}$ );  
    else return Generate( $g \in \mathcal{G}$ );  
  }  
  if  $\mathcal{H} = \text{Set}(\mathcal{F})$  {  
    ...  
  }
```

Boltzmann Sampling

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  if  $\mathcal{H} = \text{Set}(\mathcal{F})$  {  
    ...
```

F(x), H(x) by Newton iteration

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  }  
  if  $\mathcal{H} = \text{Set}(\mathcal{F})$  {  
    ...  
  }
```

$F(x), H(x)$ by Newton iteration

Example: binary trees

$$\mathcal{B} = 1 + \mathcal{L} \times \mathcal{B} \times \mathcal{B}$$

with $1/B(.23) \approx 1/1.559 \approx .6414$

Coin: 0,0,1,0,1,1,1,1,0,1,...

Boltzmann Sampling

$$\text{Proba}(t) = \frac{x^{|t|}}{T(x)} \quad \text{with} \quad T(x) = \sum_{t \in \mathcal{T}} x^{|t|} = \sum_n T_n x^n$$

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```
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$F(x), H(x)$ by Newton iteration

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$$\mathcal{B} = 1 + \mathcal{L} \times \mathcal{B} \times \mathcal{B}$$

with $1/B(.23) \approx 1/1.559 \approx .6414$

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  if ℋ = ℒ return ℒ;  
  if ℋ = ℱ × ℊ {  
    Generate(f ∈ ℱ);  
    Generate(g ∈ ℊ);  
    return (f, g); }  
  if ℋ = ℱ + ℊ {  
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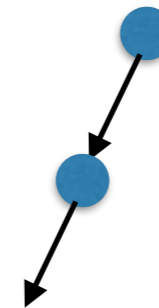
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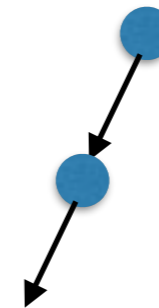
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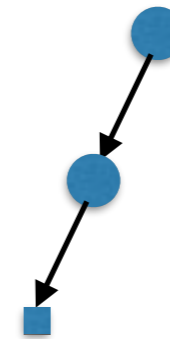
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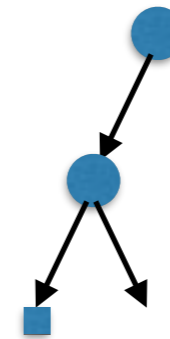
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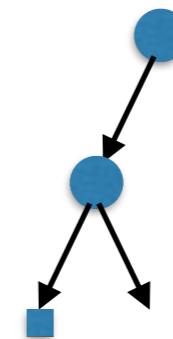
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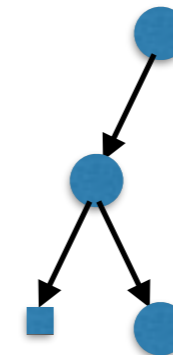
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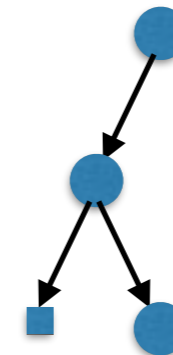
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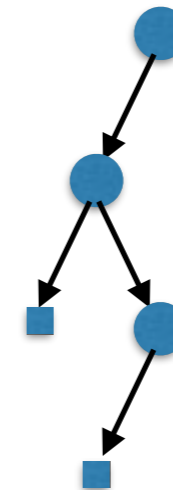
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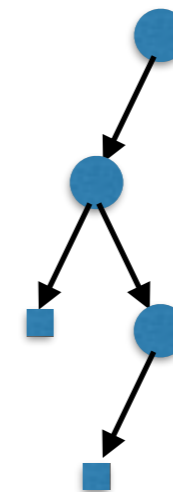
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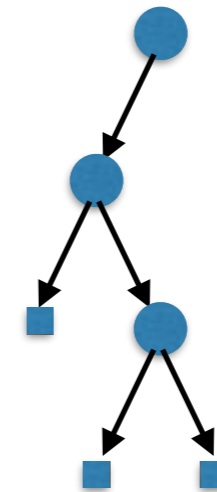
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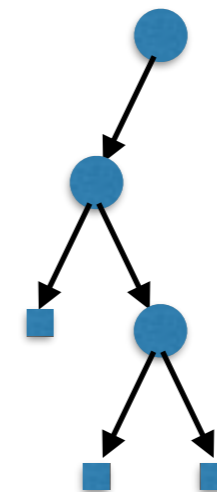
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  if  $\mathcal{H} = \mathcal{F} \times \mathcal{G}$  {  
    Generate( $f \in \mathcal{F}$ );  
    Generate( $g \in \mathcal{G}$ );  
    return ( $f, g$ ); }  
  if  $\mathcal{H} = \mathcal{F} + \mathcal{G}$  {  
     $b := \text{Bernoulli}(F(x)/H(x))$ ;  
    if  $b = 1$  return Generate( $f \in \mathcal{F}$ );  
    else return Generate( $g \in \mathcal{G}$ );  
  }  
  if  $\mathcal{H} = \text{Set}(\mathcal{F})$  {  
    ...  
  }
```

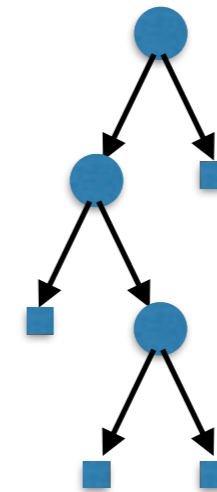
$F(x), H(x)$ by Newton iteration

Example: binary trees

$$\mathcal{B} = 1 + \mathcal{L} \times \mathcal{B} \times \mathcal{B}$$

with $1/B(.23) \approx 1/1.559 \approx .6414$

Coin: 0,0,1,0,1,1,1,1,0,1,...



Boltzmann Sampling

$$\text{Proba}(t) = \frac{x^{|t|}}{T(x)} \quad \text{with} \quad T(x) = \sum_{t \in \mathcal{T}} x^{|t|} = \sum_n T_n x^n$$

x is a parameter of the algorithm

Expected size:

$$\frac{xT'(x)}{T(x)}$$

```

Generate( $h \in \mathcal{H}$ ) = {
  if  $\mathcal{H} = 1$  return 1;
  if  $\mathcal{H} = \mathcal{L}$  return  $\mathcal{L}$ ;
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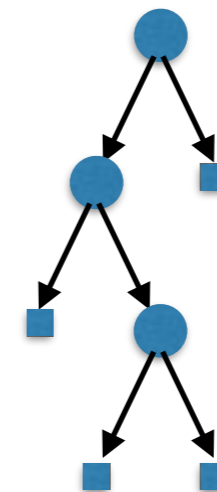
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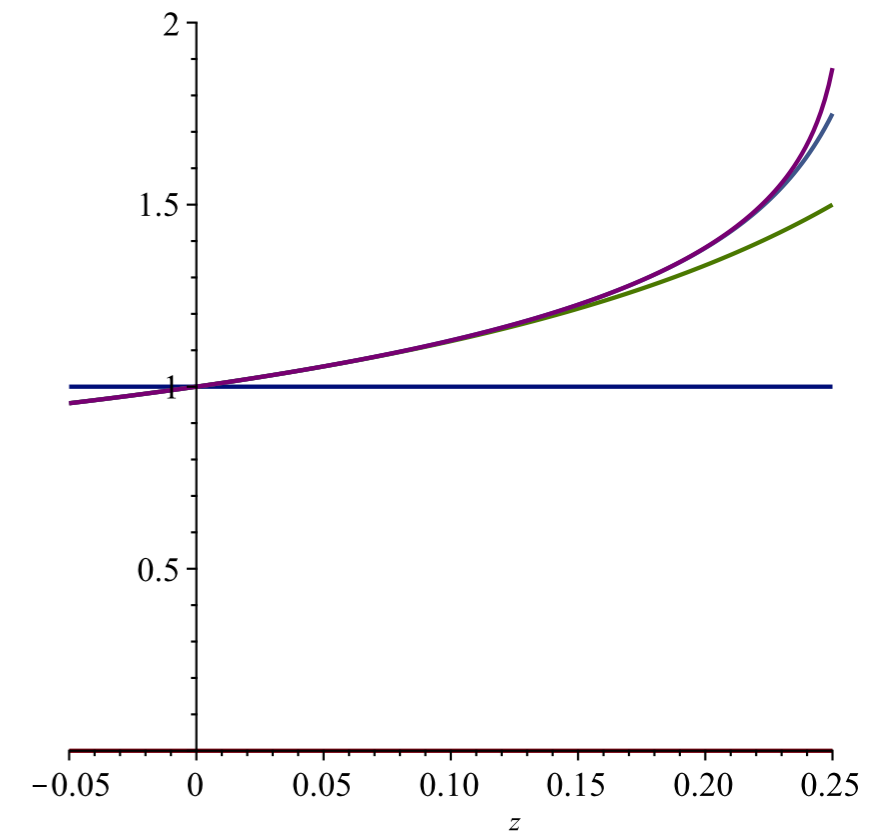


Exp. size

$$\sim \frac{1}{2\sqrt{1-4x}}$$

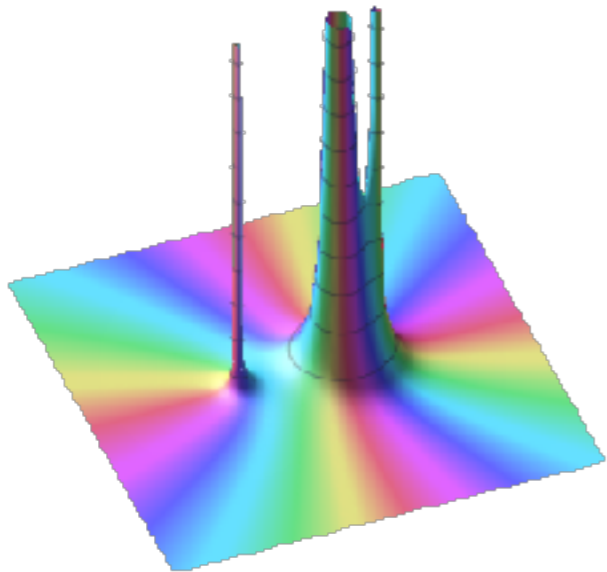
Example: XML-trees

Grammar	File size	#eq.	Newton
rss	9.5k	16	0.02s.
Relax NG	124k	114	0.10s.
XSLT	168k	122	0.12s.
XHTML Basic	284k	96	0.32s.
SVG	6.3M	232	0.23s.
OpenDocument	2.8M	814	0.34s.
DocBook	11M	977	23s.



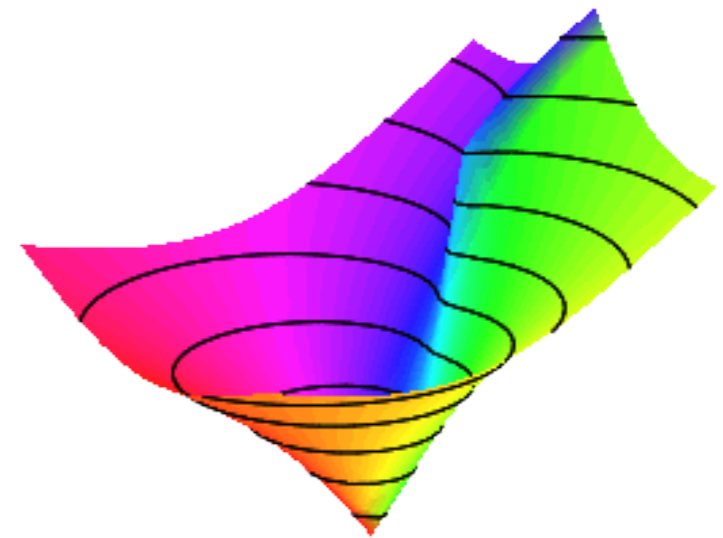
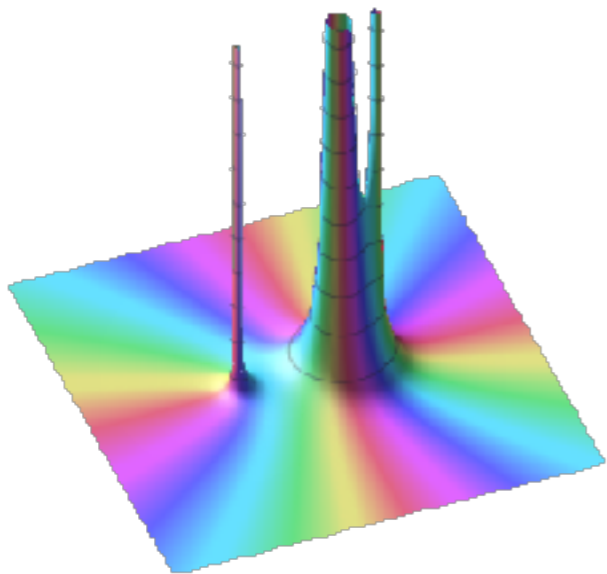
Time for x s.t. $E(\text{size})=10,000$

IV. Asymptotic Analysis



$$F(z) = \sum_{n=0}^{\infty} f_n z^n \longrightarrow f_n \sim \dots, \quad n \rightarrow \infty$$

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Singularity Analysis

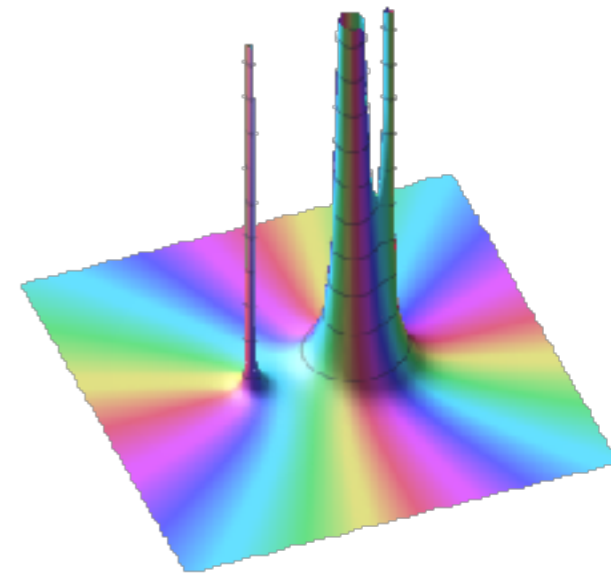
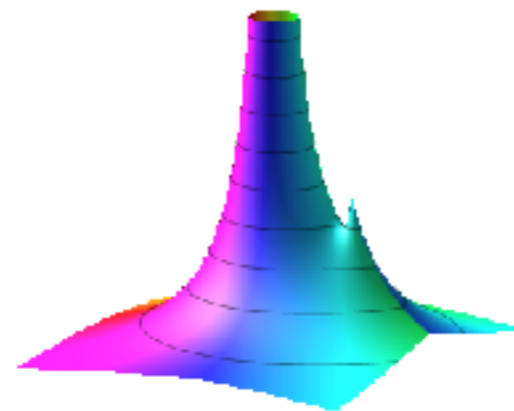
counts the number
of objects of size n

$$(a_n) \mapsto A(z) := \sum_{n \geq 0} a_n z^n$$

captures some
structure

A 3-Step Method:

$$a_n = \frac{1}{2\pi i} \oint \frac{A(z)}{z^{n+1}} dz$$



Singularity Analysis

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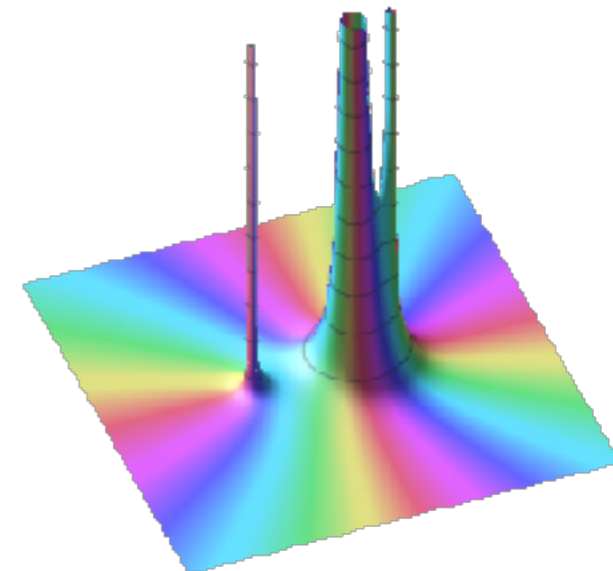
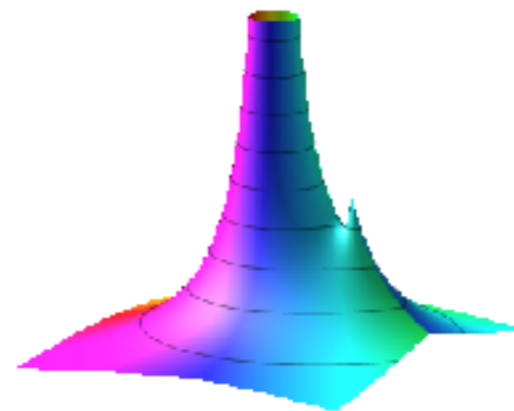
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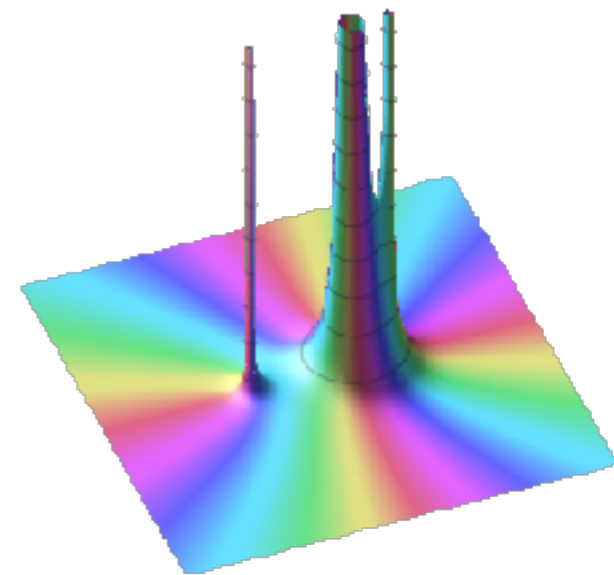
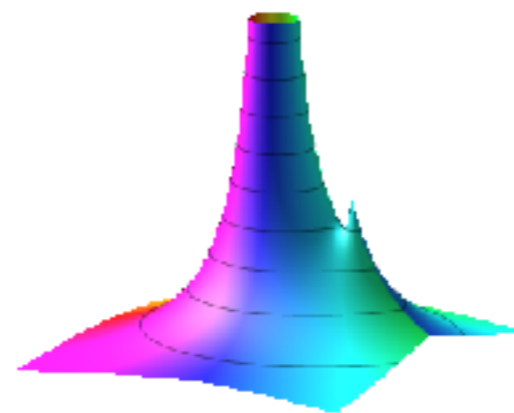
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$$A(z) \underset{z \rightarrow \rho}{\sim} c \left(1 - \frac{z}{\rho}\right)^\alpha \log^m \frac{1}{1 - \frac{z}{\rho}}$$

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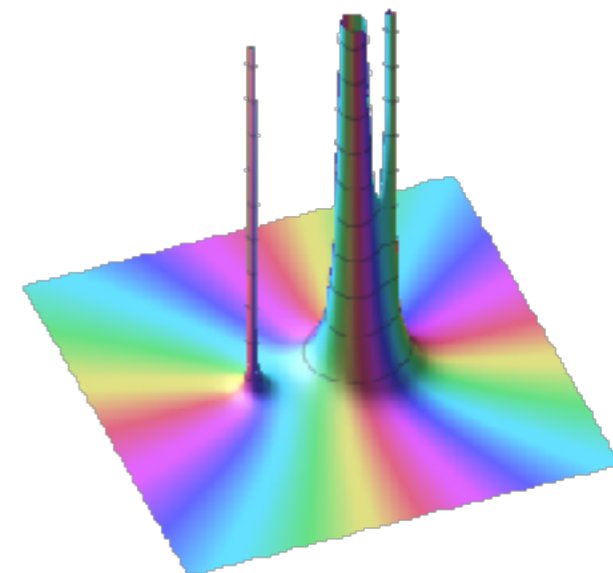
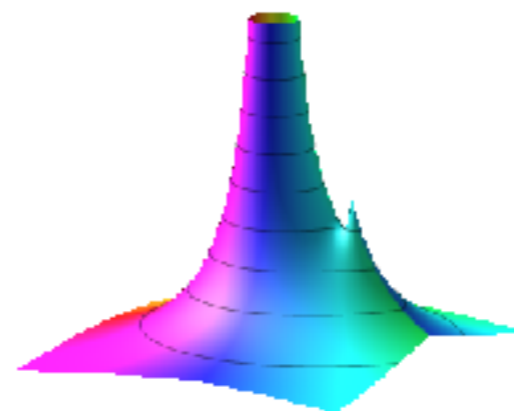
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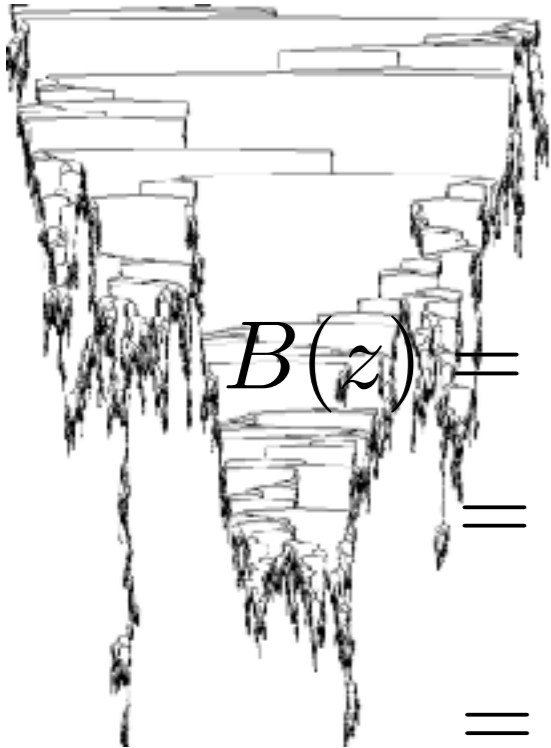


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full asymptotic expansion available

Example: Binary Trees (Again!)



$$\begin{aligned} B(z) &= 1 + zB(z)^2 \\ &= 1 + z + 2z^2 + 5z^3 + \dots \\ &= \frac{1 - \sqrt{1 - 4z}}{2z} \end{aligned}$$

1. sing. en $z = 1/4$

2. local behaviour:

$$B(z) \underset{z \rightarrow 1/4}{=} 2 - 2(1 - 4z)^{1/2} + \dots$$

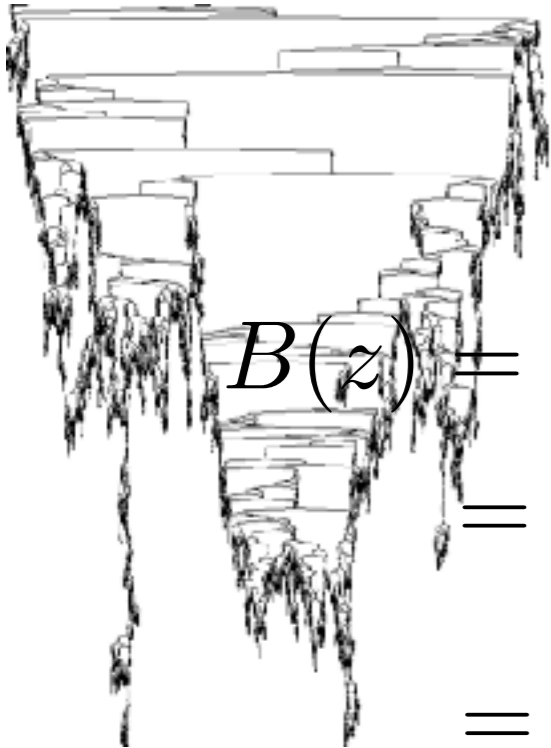
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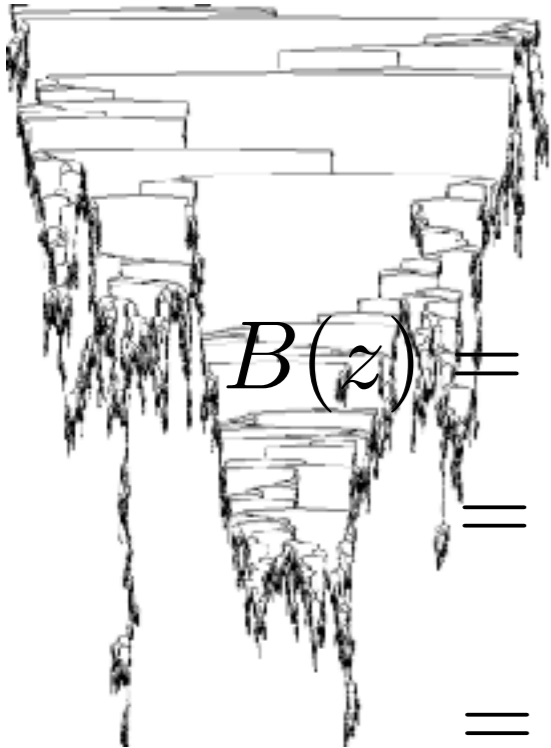
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Probability that a leaf is a child of the root?

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Demo.

Parameters



Parameters

- Equations over combinatorial structures + parameters

Ex.: path length in binary trees



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- Equations over combinatorial structures + **parameters**
- **Multivariate** generating series

$$F(z, u) = \sum_{n,k} f_{n,k} u^k z^n$$

Ex.: path length in binary trees

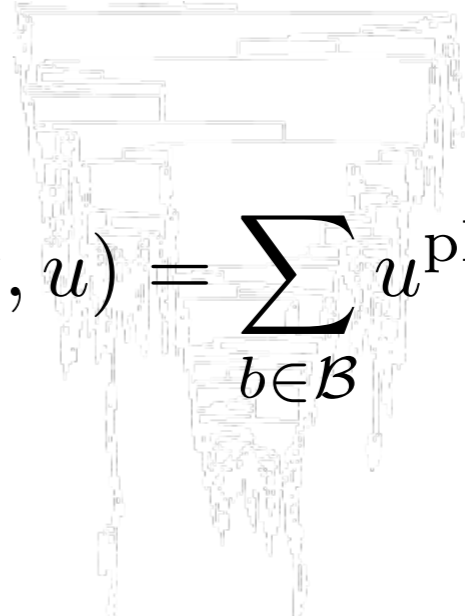


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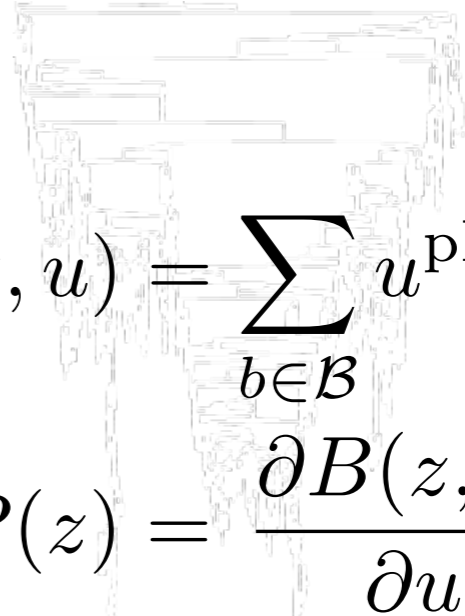

$$B(z, u) = \sum_{b \in \mathcal{B}} u^{\text{pl}(b)} z^{|b|}$$

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$$P(z) = \left. \frac{\partial B(z, u)}{\partial u} \right|_{u=1}$$

Parameters

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$$F(z, u) = \sum_{n,k} f_{n,k} u^k z^n$$

- Complex analysis

$$f_{n,k} \sim \dots, n \rightarrow \infty$$

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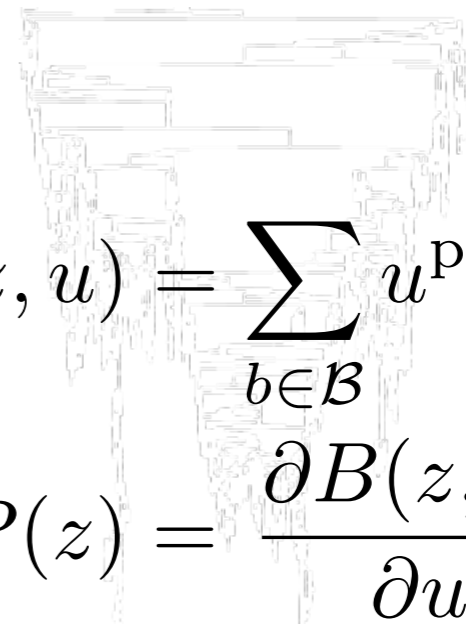
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$$\frac{P_n}{nB_n} \sim \sqrt{\pi n}$$

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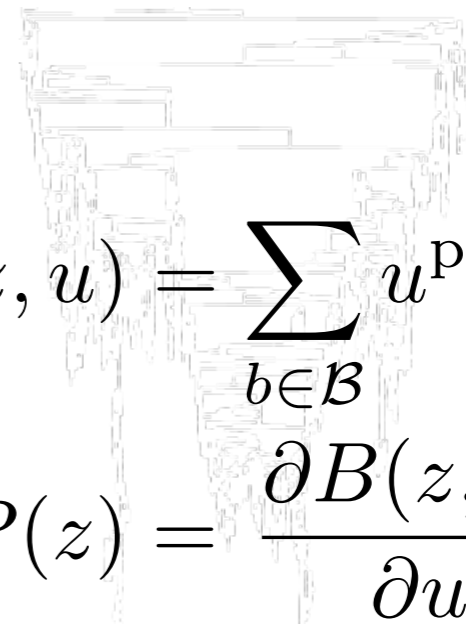
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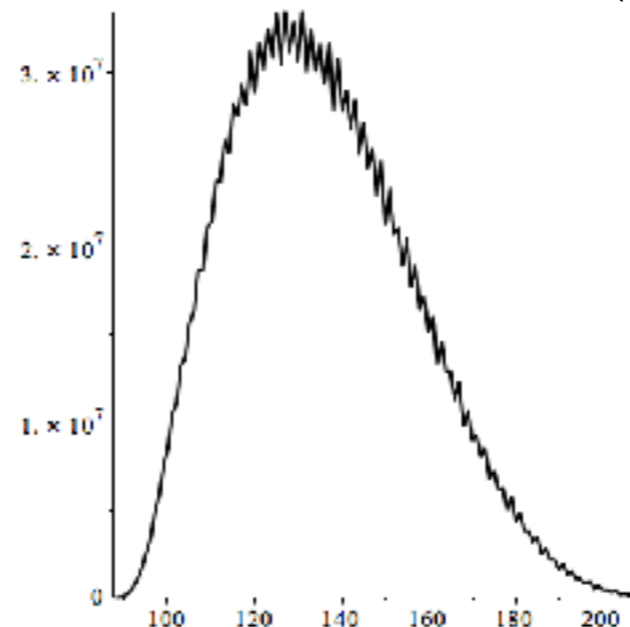
- **Limiting distribution**

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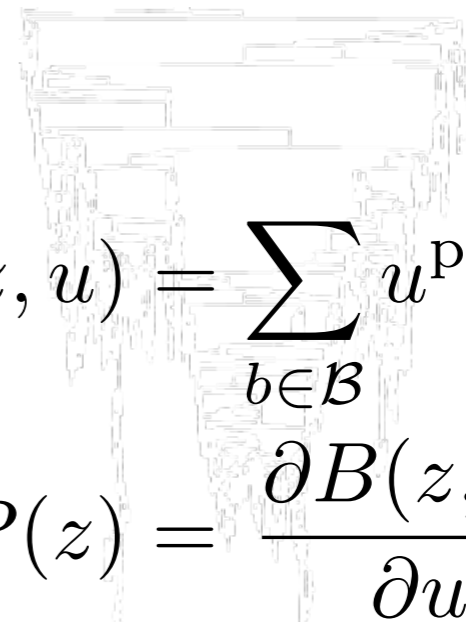
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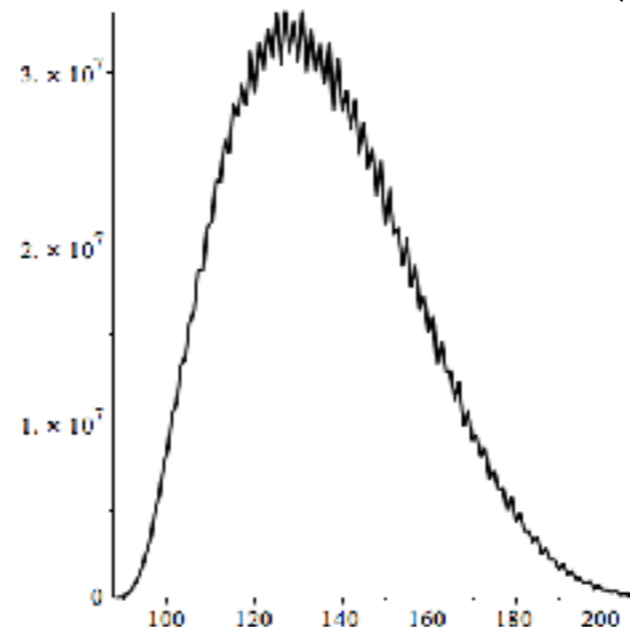
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Demo.

If you can specify it, you can analyze it!

Permutations

Mappings

Words

Strings

Urns

Trees

Languages

Integers

Compositions

Partitions

...

+ *parameters*

Universality phenomena:

Ex.: # trees of various types $\longrightarrow K \rho^n n^{-3/2} \longrightarrow$ pathlength in $n^{1/2}$

Summary & Conclusion

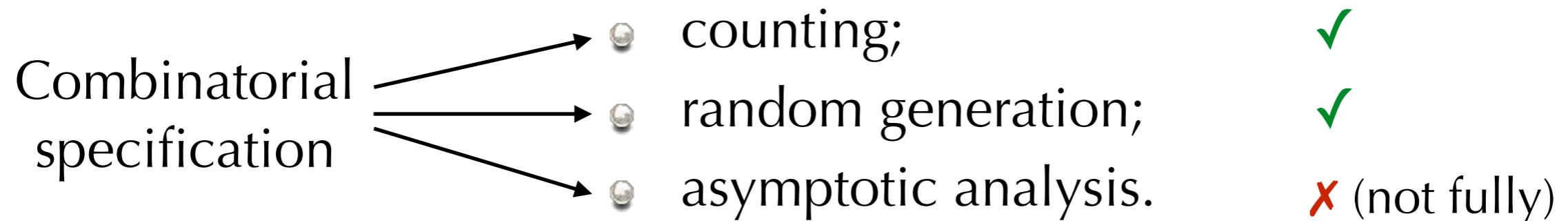
Summary & Conclusion

What we have:

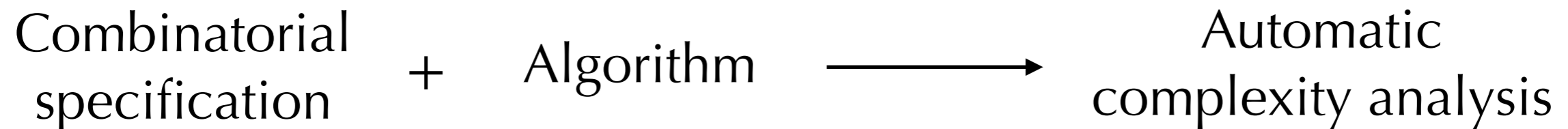
		Automated?
Combinatorial specification	● counting;	✓
	● random generation;	✓
	● asymptotic analysis.	✗ (not fully)

Summary & Conclusion

What we have:

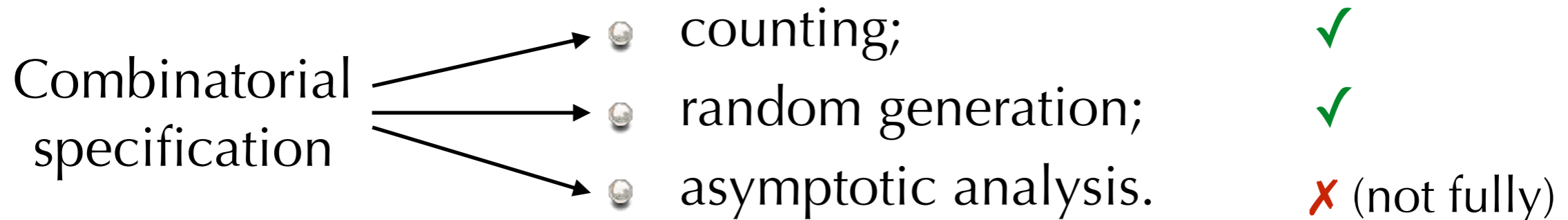


Ultimate (dream) goal:

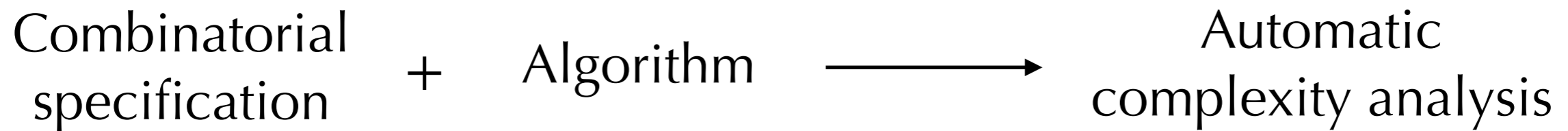


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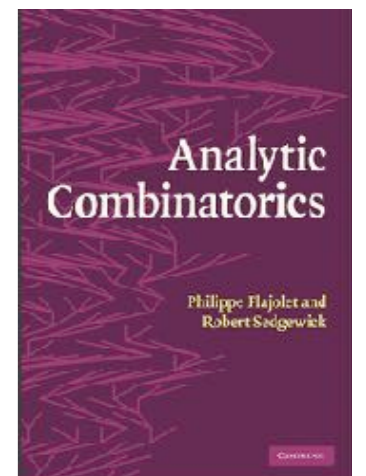


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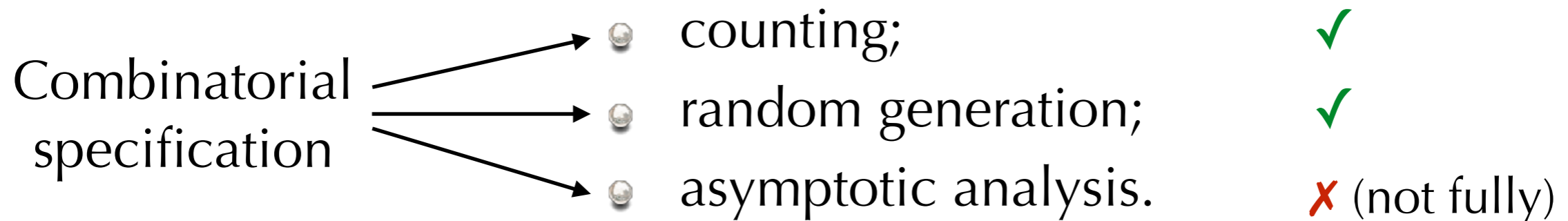
Where to learn more:

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