

# Infinite Systems of Functional Equations and Gaussian Limiting Distributions

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joint work with

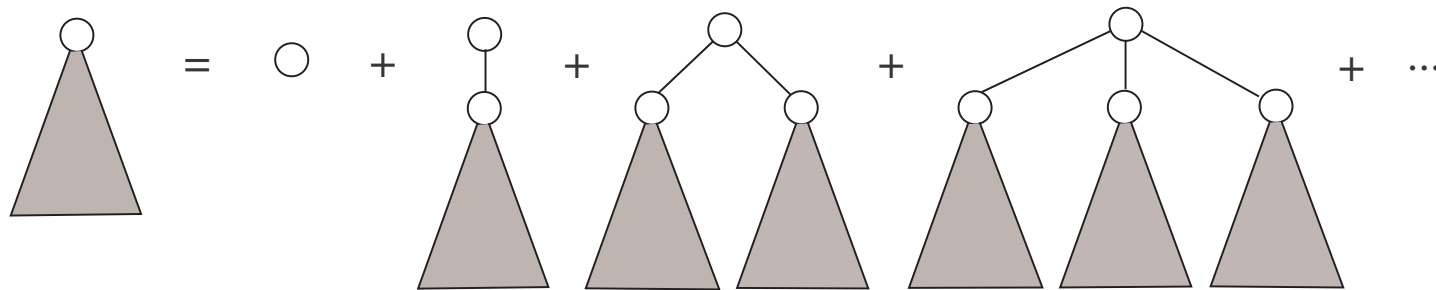
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# Outline

- Examples
- Preliminaries from functional analysis
- Some Probability Theory
- Implicit Functions
- Sketch of proofs
- Gaussian Limits
- Examples cont'd

# Example: Rooted Plane Trees



$$y(x) = x + xy(x) + xy(x)^2 + xy(x)^3 + \dots = \frac{x}{1 - y(x)}.$$

$\Rightarrow$

$$y(x) = \frac{1 - \sqrt{1 - 4x}}{2}$$

and

$$y_n = \frac{1}{n} \binom{2n-2}{n-1}.$$

# Example: Rooted Plane Trees

Let  $\mathbf{k} = (k_0, k_1, k_2, \dots)$  be a sequence of non-negative integers

$y_{n,\mathbf{k}}$  ... # trees of size  $n$  s.t.  $k_j$  vertices have exactly  $j$  successors

Let  $\mathbf{u} = (u_0, u_1, u_2, \dots)$  and  $\mathbf{u}^{\mathbf{k}} = u_0^{k_0} u_1^{k_1} u_2^{k_2} \dots$

Then  $y(x, \mathbf{u}) = \sum_{n,\mathbf{k}} y_{n,\mathbf{k}} x^n \mathbf{u}^{\mathbf{k}}$  satisfies

$$y(x, \mathbf{u}) = \underbrace{xu_0 + xu_1 y(x, \mathbf{u}) + xu_2 y(x, \mathbf{u})^2 + xu_3 y(x, \mathbf{u})^3 + \dots}_{F(x, y(x, \mathbf{u}), \mathbf{u})}$$

$\|\mathbf{u}\|_\infty$  bounded  $\rightsquigarrow$  analytic equation for  $y(x, \mathbf{u})$ .

# Example: Rooted Plane Trees

$X_n^{(j)}$  ... # nodes of out-degree  $j$  in random tree of size  $n$ .

Let  $\mathbf{X}_n = (X_n^{(0)}, X_n^{(1)}, X_n^{(2)}, \dots)$ ; then

$$\mathbb{E} \mathbf{u}^{\mathbf{X}_n} = \frac{1}{y_n} [x^n] y(x, \mathbf{u}),$$

Let  $\ell$  be the linear functional

$$\ell \cdot \mathbf{X}_n = \sum_{j \geq 0} \ell_j X_n^{(j)}.$$

Then

$$\mathbb{E} e^{it\ell \cdot \mathbf{X}_n} = \frac{1}{y_n} [x^n] y(x, e^{it\ell_0}, e^{it\ell_1}, \dots).$$

# Other Examples: Bipartite Planar Maps

Bouttier, Di Francesco, Guitter '04:

$x$  counts number of edges,  $z$  counts number of vertices

$u_j$  counts number of faces with boundary length  $2j$

Let  $R(x, z, \mathbf{u})$  denote the solution of

$$R = xz + x \sum_{j \geq 1} u_j \binom{2j-1}{j} R^j.$$

Then the g.f.  $M(x, z, \mathbf{u})$  of bipartite maps, satisfies

$$M_x = R$$

# Other Examples: Subcritical Graphs

$B(x)$  e.g.f. of 2-connected (labeled) graphs

$C(x)$  e.g.f. of connected graphs

Then

$$C'(x) = e^{B'(xC'(x))}.$$

Drmotá, Fusy, Kang, Kraus, Rue '11:

number of vertices of degree  $j$  satisfy a CLT  
(finite system of functional equations)

all  $j \geq 1$  simultaneously:

$B_r^\bullet(x, u_1, u_1, \dots)$  g.f. of rooted 2-connected graphs  
(root vertex has degree  $r$ ),

$C_j^\bullet(x, u_1, u_2, \dots)$  g.f. of rooted connected graphs.  
(root vertex has degree  $j$ ),  $C_0^\bullet(x, \mathbf{u}) = 1$

# Other Examples: Subcritical Graphs

Then

$$C_j^\bullet(x, \mathbf{u}) = \sum_{l_1+2l_2+\dots+jl_j=j} \prod_{r=1}^j \frac{B_r^\bullet(x, W_1, W_2, \dots)^{l_r}}{l_r!},$$

with

$$W_j = \sum_{i \geq 0} u_{i+j} C_i^\bullet(x, \mathbf{u})$$

The g.f. encoding the degree distribution is

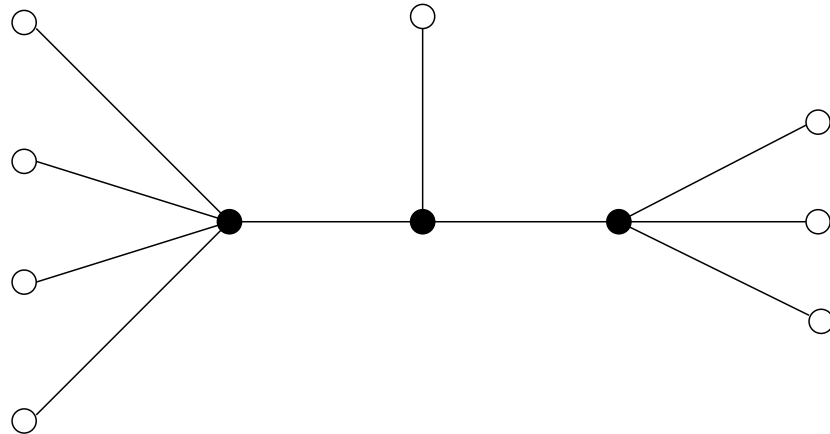
$$C^\bullet(x, \mathbf{u}) = \sum_{j \geq 0} C_j^\bullet(x, \mathbf{u}).$$



# Other Examples: Patterns in Trees

Chyzak, Drmota, Klausner, Kok '08:

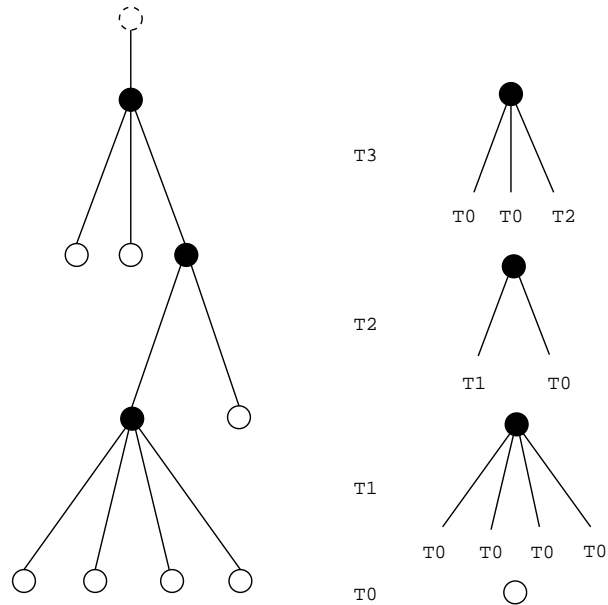
Count, e.g., occurrences of



in trees.

# Other Examples: Patterns in Trees

Splitting the pattern yields partition of the class of trees



⇒ system of functional equations

infinitely many patterns simultaneously

⇒ infinite system of functional equations

# Preliminaries

We consider linear operators  $T : B_1 \rightarrow B_2$ , where  $B_1$  and  $B_2$  are Banach spaces

$U$  is the open unit ball in  $B_1$

$T : B_1 \rightarrow B_2$  compact  $\Leftrightarrow$  closure of  $T(U)$  is compact in  $B_2$   
 $\Leftrightarrow$  every bounded  $(x_n)_{n \in \mathbb{N}}$  has subsequence  
 $(x_{n_i})_{i \in \mathbb{N}}$  s.t.  $(Tx_{n_i})_{i \in \mathbb{N}}$  converges in  $B_2$

$A : B \rightarrow B$  bounded operator;

$$r(A) = \sup_{\lambda \in \sigma(A)} |\lambda|$$

denotes its spectral radius,  $\sigma(A)$  its spectrum.

# Preliminaries

Let  $1 \leq p < \infty$ :  $\ell^p$  ... set of  $\mathbb{C}$ -valued sequences  $(t_n)_{n \in \mathbb{N}}$   
s.t.  $\|(t_n)\|_p^p := \sum_{n=1}^{\infty} |t_n|^p < \infty$ .

In  $\ell^p$ :

bounded linear operator  $A \Leftrightarrow$  infinite dimensional matrix  $(a_{ij})_{1 \leq i, j < \infty}$   
(Not true for  $\ell^\infty$ !)

**A positive**, if  $a_{ij} \geq 0$

**A irreducible**, if  $A$  is positive and for all  $(i, j)$  there is an  $n > 0$   
s.t.  $(A^n)_{ij} > 0$

Dual space of  $\ell^p$  is  $\ell^q$  with  $\frac{1}{p} + \frac{1}{q} = 1$ .

# Some Probability Theory

$\mathbf{X}$  ... random variable from  $(\Omega, \mathcal{F}, \mathbb{P})$  to a Banach space  $B$

$P$  ... law of  $\mathbf{X}$

$\ell^*$  continuous functional

$\implies \ell^*(\mathbf{X})$  determine the distribution of  $\mathbf{X}$

$\mathbf{X}_n \xrightarrow{w} \mathbf{X}$  if  $P_n \xrightarrow{w} P$ , i.e., if

$$\int_B f \, dP_n \rightarrow \int_B f \, dP$$

for every bounded continuous real function  $f$ .

A set  $\Pi$  of probability measures is **tight** if for each  $\varepsilon > 0$  there is a compact set  $K$  s.t.  $P(K) > 1 - \varepsilon$  for every  $P \in \Pi$ .

# Some Probability Theory

## Theorem 1

$$\mathbf{X}_n \xrightarrow{w} \mathbf{X}$$

if  $\{P_n : n \in \mathbb{N}\}$  is tight and

$$\mathbb{E} \left[ e^{it\ell^*(\mathbf{X}_n)} \right] \rightarrow \mathbb{E} \left[ e^{it\ell^*(\mathbf{X})} \right]$$

for all  $\ell^* \in B^*$ ,

**X Gaussian** if  $\ell^*(\mathbf{X})$  is Gaussian variable for all  $\ell^* \in B^*$ .

Then  $\mathbb{E}\mathbf{X} = \mathbf{y} \in B$  s.t.

$$\ell^*(\mathbf{y}) = \mathbb{E}(\ell^*(\mathbf{X})).$$

for all  $\ell^* \in B^*$ .

# Implicit Functions

Set  $\mathbf{u} = e^{\mathbf{v}} := (e^{v_1}, e^{v_2}, \dots)$  and fix  $p, r$  with  $1 \leq p < \infty$ ,  $1 \leq r \leq \infty$ .

**Theorem 2** Let  $\mathbf{F} : \mathbb{C} \times \ell^p \times \ell^r \rightarrow \ell^p$ ,  $(x, \mathbf{y}, \mathbf{v}) \mapsto \mathbf{F}(x, \mathbf{y}, \mathbf{v})$  s.t.:

$$\frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, \mathbf{y}, \mathbf{0}) = A(x, \mathbf{y}) + \alpha(x, \mathbf{y}) \mathbf{I}_p,$$

$\alpha(x, \mathbf{y})$  is analytic and positive,  $A$  irred.,  $A^n$  compact for some  $n$ ;

Let  $\mathbf{y} = \mathbf{y}(x, \mathbf{v})$  be the unique solution of 
$$\begin{cases} \mathbf{y} = \mathbf{F}(x, \mathbf{y}, \mathbf{v}), \\ \mathbf{y}(0, \mathbf{v}) = \mathbf{0}. \end{cases}$$

For  $\mathbf{y}(x, \mathbf{0})$  assume rad. of conv.  $x_0 > 0$  and that  $\mathbf{y}_0 := \mathbf{y}(x_0, \mathbf{0})$  exists. Then

$$\mathbf{y}(x, \mathbf{v}) = \mathbf{g}(x, \mathbf{v}) - \mathbf{h}(x, \mathbf{v}) \sqrt{1 - \frac{x}{x_0(\mathbf{v})}}$$

Under aperiodicity conditions we have also  $\Delta$ -analyticity.

# Sketch of the Proof

First, let  $v = 0$ .

Implicit fctn theorem + Banach fixed-point theorem:

$$\mathbf{y}^{(n+1)}(x, \mathbf{0}) = \mathbf{F}(x, \mathbf{y}^{(n)}(x, \mathbf{0}), \mathbf{0})$$

converges uniformly to the unique solution  $\mathbf{y}(x, \mathbf{0})$ .

$\mathbf{F}$  positive  $\implies \mathbf{y}(x, \mathbf{0})$  positive.



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converges uniformly to the unique solution  $\mathbf{y}(x, \mathbf{0})$ .

$\mathbf{F}$  positive  $\implies \mathbf{y}(x, \mathbf{0})$  positive.

Next:

$$y_0 = \mathbf{F}(x_0, y_0, \mathbf{0}), \quad r \left( \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x_0, y_0, \mathbf{0}) \right) = 1.$$

We use that  $x \mapsto r \left( \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, \mathbf{y}(x, \mathbf{0}), \mathbf{0}) \right)$  is strictly increasing

and that  $r \left( \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, \mathbf{y}(x, \mathbf{0}), \mathbf{0}) \right) < 1$  for  $x < x_0$

follows from  $\left( I - \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, \mathbf{y}(x, \mathbf{0}), \mathbf{0}) \right) \frac{\partial \mathbf{y}}{\partial x}(x, \mathbf{0}) = \frac{\partial \mathbf{F}}{\partial x}(x, \mathbf{y}(x, \mathbf{0}), \mathbf{0})$

# Sketch of the Proof

Third step: Let  $\mathbf{y} = (y_1, \bar{y})$  and  $\mathbf{F} = (F_1, \bar{F})^t$ .

$$\begin{aligned}y_1 &= F_1(x, y_1, \bar{y}, \mathbf{0}), \\ \bar{y} &= \bar{F}(x, y_1, \bar{y}, \mathbf{0}).\end{aligned}$$

Jacobian operators:

$$A = \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x_0, y_0, \mathbf{0}) \quad B = \frac{\partial \bar{F}}{\partial \bar{y}}(x_0, y_{01}, \bar{y}_0, \mathbf{0}).$$

# Sketch of the Proof

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Jacobian operators:

$$A = \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x_0, y_0, \mathbf{0}) \quad B = \frac{\partial \bar{F}}{\partial \bar{y}}(x_0, y_{01}, \bar{y}_0, \mathbf{0}).$$

Note:

$$r(B) < r(A) = 1.$$

Insert solution of second equation into first one:

$$y_1 = F_1(x, y_1, \bar{y}(x, y_1, \mathbf{0}), \mathbf{0}) =: G(x, y_1).$$

Equation for  $y_1 = y_1(x, \mathbf{0})$ .

# Sketch of the Proof

4th step: Find singularity of solution  $y_1$ .

$$\begin{aligned}y_1 &= G(x, y_1), \\ 1 &= G_{y_1}(x, y_1),\end{aligned}$$

with  $G_x(x_0, y_{01}) \neq 0$  and  $G_{y_1 y_1}(x_0, y_{01}) \neq 0$ .

$\implies$

$$y_1(x, 0) = g_1(x, 0) - h_1(x, 0) \sqrt{1 - \frac{x}{x_0}}$$

Insert this solution into  $\bar{y}$

$\implies$

$$\bar{y}(x, y_1(x, 0), 0) = \bar{g}(x, 0) - \bar{h}(x, 0) \sqrt{1 - \frac{x}{x_0}}.$$

# Sketch of the Proof

5th step: analyticity in  $\mathbf{v}$ . From

$$r \left( \frac{\partial \bar{\mathbf{F}}}{\partial \bar{\mathbf{y}}} (x_0, (y_0)_1, \bar{y}_0, \mathbf{0}) \right) < 1.$$

we get unique solution  $\bar{y}(x, y_1, \mathbf{v})$  of the functional equation

$$\bar{y} = \bar{\mathbf{F}}(x, y_1, \bar{y}, \mathbf{v})$$

near  $(x_0, (y_0)_1, \mathbf{0})$ .

$\implies$

$$\begin{aligned} y_1 &= G(x, y_1, \mathbf{v}), \\ 1 &= G_{y_1}(x, y_1, \mathbf{v}), \end{aligned}$$

has positive solution  $(x_0, (y_0)_1, \mathbf{0})$ . Furthermore,

$$\det \begin{pmatrix} -G_x & 1 - G_{y_1} \\ -G_{y_1, x} & -G_{y_1, y_1} \end{pmatrix} = G_x \cdot G_{y_1 y_1} \neq 0,$$

$\implies$  there exist  $x_0(\mathbf{v})$  and  $y_1(\mathbf{v})$  near  $\mathbf{0}$ .

# Gaussian Limits

$$y(x, \mathbf{v}) = G(x, y(x, \mathbf{v}), \mathbf{v}) = \sum_{n=0}^{\infty} c_n(\mathbf{v}) x^n,$$

$\mathbf{X}_n$  corresponding  $\ell^p$ -valued random variable then

$$\mathbb{E} \left[ e^{it\ell \cdot \mathbf{X}_n} \right] = \frac{c_n(it\ell)}{c_n(\mathbf{0})}$$

for all  $\ell \in \ell^q$ .

In applications:

$$G(x, y(x, \mathbf{v}), \mathbf{v}) = \sum_{n=0}^{\infty} \sum_{\mathbf{m} \in \ell^p} c_{n,\mathbf{m}} e^{\mathbf{m} \cdot \mathbf{v}} x^n = \sum_{n=0}^{\infty} \sum_{\mathbf{m} \in \ell^p} c_{n,\mathbf{m}} \mathbf{u}^{\mathbf{m}} x^n,$$

where  $c_{n,\mathbf{m}}$  denotes  $\#$  objects of size  $n$  and characteristic  $\mathbf{m}$ .

# Gaussian Limits

**Corollary**  $y = y(x, \mathbf{v})$  as before,  $G : (\mathbb{C}, \ell^p, \ell^r) \rightarrow \mathbb{C}$  analytic at  $(x_0(\mathbf{0}), y(x_0(\mathbf{0}), \mathbf{0}), \mathbf{0})$ ,

$$\frac{\partial G}{\partial \mathbf{y}}(x_0(\mathbf{0}), y(x_0(\mathbf{0}), \mathbf{0}), \mathbf{0}) \neq \mathbf{0}.$$

Then

$$G(x, y(x, \mathbf{v}), \mathbf{v}) = \bar{g}(x, \mathbf{v}) - \bar{h}(x, \mathbf{v}) \sqrt{1 - \frac{x}{x_0(\mathbf{v})}}$$

Moreover,

$$[x^n]G(x, y(x, \mathbf{v}), \mathbf{v}) \sim \frac{\bar{h}(x_0(\mathbf{v}), \mathbf{v})}{2\sqrt{\pi}} x_0(\mathbf{v})^{-n} n^{-3/2}$$

uniformly for  $\mathbf{v}$  in a neighborhood of  $\mathbf{0}$ .

# Gaussian Limits

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$$\frac{\partial G}{\partial \mathbf{y}}(x_0(\mathbf{0}), y(x_0(\mathbf{0}), \mathbf{0}), \mathbf{0}) \neq \mathbf{0}.$$

Then

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Moreover,

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uniformly for  $\mathbf{v}$  in a neighborhood of  $\mathbf{0}$ .

Proof:

$$G(x, \mathbf{y}, \mathbf{v}) = G(x, \mathbf{g}(x, \mathbf{v}), \mathbf{v}) + \sum_{n \geq 1} \frac{A_n(x, \mathbf{v})}{n!} \left( (\mathbf{y} - \mathbf{g}(x, \mathbf{v}))^n \right).$$



# Gaussian Limits

**Theorem 3**  $1 \leq p < \infty$ ,  $\mathbf{X}_n$  is a sequence of  $\ell^p$ -valued random variables as above,  $\ell \in \ell^q$ .

Then  $\ell \cdot \mathbb{E}\mathbf{X}_n = \mu_\ell n + O(1)$  with  $\mu_\ell = -\frac{\partial x_0}{\partial \mathbf{v}}(\mathbf{0}) \cdot \ell / x_0$  and

$$\ell \cdot \left( \frac{\mathbf{X}_n - \mathbb{E}\mathbf{X}_n}{\sqrt{n}} \right) \xrightarrow{w} \mathcal{N}(\mathbf{0}, \ell^T B \ell)$$

where  $B \in L(\ell^q, \ell^p)$  is given by the matrix

$$\frac{1}{x_0^2} \left( \frac{\partial x_0}{\partial v_i}(\mathbf{0}) \cdot \frac{\partial x_0}{\partial v_j}(\mathbf{0})^T \right)_{1 \leq i, j < \infty} - \frac{1}{x_0} \left( \frac{\partial^2 x_0}{\partial v_i \partial v_j}(\mathbf{0}) \right)_{1 \leq i, j < \infty}.$$

**Corollary** If in addition the set of laws of  $(\mathbf{X}_n - \mathbb{E}\mathbf{X}_n) / \sqrt{n}$  is tight then there exists a centered Gaussian random variable  $\mathbf{X}$  such that

$$\frac{\mathbf{X}_n - \mathbb{E}\mathbf{X}_n}{\sqrt{n}} \xrightarrow{w} \mathbf{X},$$

# Tightness

**Theorem 4** *Single functional equation  $y = F(x, y, \mathbf{v})$  where  $F : B \times U \times V \rightarrow \mathbb{C}$  is a positive and analytic and  $B \times U \times V \subseteq \mathbb{C}^2 \times \ell^2$  such that there exist positive real  $(x_0, y_0) \in B \times U$  with  $y_0 = F(x_0, y_0, \mathbf{0})$  and  $1 = F_y(x_0, y_0, \mathbf{0})$  such that  $F_x(x_0, y_0, \mathbf{0}) \neq 0$  and  $F_{yy}(x_0, y_0, \mathbf{0}) \neq 0$ .*

*Furthermore assume that the corresponding r.v.'s  $X_n^{(j)}$  fulfil  $X_n^{(j)} = 0$  for  $j > cn$  and that:*

$$\sum_{j \geq 0} F_{v_j} < \infty, \quad \sum_{j \geq 0} F_{y v_j}^2 < \infty, \quad \sum_{j \geq 0} F_{v_j v_j} < \infty,$$

$$F_{x v_j} = o(1), \quad F_{x v_j v_j} = o(1), \quad F_{y y v_j} = o(1), \quad F_{y y v_j v_j} = o(1),$$

$$F_{x x v_j} = O(1), \quad F_{x y v_j} = O(1), \quad F_{x y y v_j} = O(1), \quad F_{y y y v_j} = O(1),$$

*all derivatives evaluated at  $(x_0, y_0, \mathbf{0})$ .  $\implies$  tightness.*

# Sketch of Proof

By a theorem of Grenander (1963) tightness follows from

$$\lim_{N \rightarrow \infty} \sup_{n \geq 1} \mathbb{E} \left[ \sum_{j > N} \frac{(X_n^{(j)} - \mathbb{E}X_n^{(j)})^2}{n} \right] = 0.$$

The assumption  $X_n^{(j)} = 0$  if  $j > cn$  implies that this reduces to

$$\lim_{N \rightarrow \infty} \sup_{n \geq 1} \sum_{N < j \leq cn} \sigma_{n,j}^2 = 0,$$

where  $\sigma_{n,j}^2 = \text{Var}(X_n^{(j)} - \mathbb{E}X_n^{(j)}) / \sqrt{n}$ .

Tightness follows from  $\sigma_{n,j}^2 = \sigma_j^2 + \tau_j/n + O(n^{-2})$  if

$$\sum_{j \geq 0} \sigma_j^2 < \infty, \quad \text{and} \quad \tau_j = o(1) \quad (j \rightarrow \infty),$$

uniformly in  $j$ .

# Examples cont'd: Subcritical Graphs

Recall

$$C_j^\bullet(x, \mathbf{u}) = \sum_{l_1+2l_2+\dots+jl_j=j} \prod_{r=1}^j \frac{B_r^\bullet(x, W_1, W_2, \dots)^{l_r}}{l_r!},$$

with

$$W_j = \sum_{i \geq 0} u_{i+j} C_i^\bullet(x, \mathbf{u})$$

The g.f. of interest is:  $C^\bullet(x, \mathbf{u}) = \sum_{j \geq 0} C_j^\bullet(x, \mathbf{u})$ .

# Examples cont'd: Subcritical Graphs

Recall

$$C_j^\bullet(x, \mathbf{u}) = \sum_{l_1+2l_2+\dots+jl_j=j} \prod_{r=1}^j \frac{B_r^\bullet(x, W_1, W_2, \dots)^{l_r}}{l_r!},$$

with

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The g.f. of interest is:  $C^\bullet(x, \mathbf{u}) = \sum_{j \geq 0} C_j^\bullet(x, \mathbf{u})$ .

Only difficulty:

$$\frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, \mathbf{y}, \mathbf{0}) = A(x, \mathbf{y}) + \alpha(x, \mathbf{y}) \mathbf{I}_p,$$

$\alpha$  analytic and positive,  $A$  irreducible,  $A^n$  compact for some  $n$ ;

# Examples cont'd: Subcritical Graphs

**Lemma 1** *Let  $H(x, y, w)$  be a positive function, suppose that  $y(x)$  has finite radius of convergence  $x_0$  (so that  $H(x, y, 1)$  is analytic at  $(x_0, y_0)$ ) and satisfies  $y(x) = H(x, y(x), 1)$ .*

*Consider the system*

$$y_j(x, \mathbf{u}) = F_j(x, \mathbf{y}(x, \mathbf{u}), \mathbf{u})$$

*and assume*

$$F_i(x, \mathbf{y}, \mathbf{1}) = [w^i] H \left( x, \sum_j y_j, w \right).$$

*Then we have  $y(x) = \sum_i y_i(x, \mathbf{1})$  and the operator  $A = \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, y, \mathbf{1})$  is compact.*

# Examples cont'd: Subcritical Graphs

**Lemma 2** *Let  $H(x, y, w)$  be a positive function, suppose that  $y(x)$  has finite radius of convergence  $x_0$  (so that  $H(x, y, 1)$  is analytic at  $(x_0, y_0)$ ) and satisfies  $y(x) = H(x, y(x), 1)$ .*

*Consider the system*

$$y_j(x, \mathbf{u}) = F_j(x, \mathbf{y}(x, \mathbf{u}), \mathbf{u})$$

*and assume*

$$F_i(x, \mathbf{y}, \mathbf{1}) = [w^i] H \left( x, \sum_j y_j, w \right).$$

*Then we have  $y(x) = \sum_i y_i(x, \mathbf{1})$  and the operator  $A = \frac{\partial \mathbf{F}}{\partial \mathbf{y}}(x, y, \mathbf{1})$  is compact.*

*Proof.*  $y(x) = \sum_i y_i(x, \mathbf{1})$  is obvious.

$\frac{\partial F_i}{\partial y_j} = a_i(x, \sum_\ell y_\ell) \cdot y_j$ , thus  $A$  has rank 1 and is compact.

# Examples cont'd: Subcritical Graphs

In our example:  $C'(x) = C^\bullet(x, \mathbf{1})$ ,  $B'(x) = B^\bullet(x, \mathbf{1})$

Set  $y_j(x, \mathbf{u}) = C_j^\bullet(x, \mathbf{u})$ ; then  $y = \sum_j y_j$ .

Choose

$$H(x, y, w) = \sum_j y_j w^j = \exp \left( \sum_k B_k^\bullet(xy, \mathbf{1}) w^k \right)$$
$$H(x, C'(x), w) = \exp \left( \sum_k B_k^\bullet(xC'(x), \mathbf{1}) w^k \right)$$

Then

$$H(x, y, 1) = e^{B'(xy)}$$

which corresponds to  $C'(x) = e^{B'(xC'(x))}$ .



# Examples cont'd: Random Walks on Groups

$\Gamma = \Gamma_1 * \Gamma_2 * \cdots := (\cup_i (\Gamma_i \setminus \{1\}))^*$  free product of groups

random walk on  $\Gamma$ :  $S_n = \xi_1 \xi_2 \cdots \xi_n$

where  $\xi_i$  are i.i.d with common distribution

$$P\{\xi_1 = \alpha\} = p_i q_\alpha \text{ if } \alpha \in \Gamma_i \setminus \{1\} \quad \text{and } P\{\xi_1 = \varepsilon\} = p_0$$

where  $p_i > 0$ ,  $i \in \mathbb{N}$ , and  $q_\alpha > 0$  and  $\{n : P\{S_n = \varepsilon\} > 0\} \not\subseteq U < \Gamma$

**Theorem 5 (Lalley '02)** *There exist  $C > 0$  and  $1 < R < \infty$  such that  $P\{S_n = \varepsilon\} \sim CR^{-n}n^{-3/2}$ .*

Proof based on the generating functions for first hitting times:

$\tau_\alpha = \min\{n \geq 0 : S_n = \alpha\}$  ( $\alpha \in \Gamma_i$ ) and

$$y_{i;\alpha}(x) = \sum_{n \geq 1} P\{\tau_\alpha = n\} x^n,$$

# Examples cont'd: Random Walks on Groups

Lalley showed

$$y_{i;\alpha}(x) = x \left( p_i q_\alpha + p_0 y_{i;\alpha}(x) + \sum_{\gamma \in \Gamma_i \setminus \{\alpha\}} p_i q_\gamma y_{i;\gamma^{-1}\alpha}(x) + \sum_{j \neq i} \sum_{\gamma \in \Gamma_i} p_j q_\gamma y_{j;\gamma^{-1}}(x) y_{i;\alpha}(x) \right).$$

Furthermore, he showed:  $\mathbf{y}(x) \in \ell^1$  and

$$\mathcal{J}(x) = \mathcal{K}(x) + \mathcal{L}(x) + \frac{G(x) - 1}{xG(x)} \mathbf{I}_1,$$

where

$$G(x) = \sum_{n \geq 1} P\{S_n = \varepsilon\} x^n = \left( 1 - p_0 x - \sum_{\alpha \neq \varepsilon} p_\alpha x y_{\alpha^{-1}}(x) \right)^{-1},$$

**MERCI!**

**THANK YOU!**