

Interacting Particle Systems: Almost sure uniqueness, pathwise duality, and the mean-field limit

Jan M. Swart

Lecture 3: The mean-field limit

A general set-up

Let (Λ, \sim) be a countable graph.

For each $k \geq 1$, let \mathcal{K}^k be the set of all words $i_1 \cdots i_k$, made from the alphabet Λ , such that:

$$(i) \ i_n \neq i_m \ \forall n \neq m \quad (ii) \ i_1 \sim i_2 \sim \cdots \sim i_k.$$

For each $g : S^k \rightarrow S$ and $i_1 \cdots i_k \in \mathcal{K}^k$, define $g_{i_1 \cdots i_k} : S^\Lambda \rightarrow S^\Lambda$ by

$$g_{i_1 \cdots i_k}(x)(j) := \begin{cases} g(x(i_1), \dots, x(i_k)) & \text{if } j = i_1, \\ x(j) & \text{otherwise.} \end{cases}$$

For $g : S^0 \rightarrow S$ and $i_1 \in \Lambda$, define $g_{i_1} : S^\Lambda \rightarrow S^\Lambda$ by

$$g_{i_1}(x)(j) := \begin{cases} g(\emptyset) & \text{if } j = i_1, \\ x(j) & \text{otherwise.} \end{cases}$$

Examples

Let $S := \{0, 1\}$ and define $\text{dth} : S^0 \rightarrow S$, $\text{bra} : S^2 \rightarrow S$, and $\text{cob} : S^3 \rightarrow S$ by

$$\text{dth}(\emptyset) := 0,$$

$$\text{bra}(x(1), x(2)) := x(1) \vee x(2),$$

$$\text{cob}(x(1), x(2), x(3)) := x(1) \vee (x(2) \wedge x(3)).$$

Then dth_{i_1} , $\text{bra}_{i_1 i_2}$, and $\text{cob}_{i_1 i_2 i_3}$ are the local maps defined in the first lecture.

A general set-up

- (i) Polish space S local state space.
- (ii) $(\Omega, \mathcal{B}, \mathbf{r})$ Polish space with Borel σ -field and finite measure: source of external randomness.
- (iii) $\kappa : \Omega \rightarrow \mathbb{N}$ measurable function.
- (iv) For each $\omega \in \Omega$, a measurable function $\gamma[\omega] : S^{\kappa(\omega)} \rightarrow S$.

We are interested in the interacting particle system that evolves as follows:

- ▶ We activate each site i with Poisson rate $|\mathbf{r}| := \mathbf{r}(\omega)$.
- ▶ We choose ω according to the law $|\mathbf{r}|^{-1} \mathbf{r}$.
- ▶ We uniformly choose $i = i_1 \sim \dots \sim i_{\kappa(\omega) \vee 1}$, all different, if this is possible.
- ▶ We apply the map $\gamma_{i_1 \dots i_{\kappa(\omega) \vee 1}}[\omega]$.

Examples

Let $S := \{0, 1\}$ and $\Omega = \{1, 2\}$. Then setting

$$\begin{aligned}\kappa(1) &:= 0, & \gamma[1] &:= \text{dth}, & \mathbf{r}(\{1\}) &:= 1, \\ \kappa(2) &:= 2, & \gamma[2] &:= \text{bra}, & \mathbf{r}(\{2\}) &:= \lambda,\end{aligned}$$

yields the contact process with infection rate λ . Similarly, setting

$$\begin{aligned}\kappa(1) &:= 0, & \gamma[1] &:= \text{dth}, & \mathbf{r}(\{1\}) &:= 1, \\ \kappa(2) &:= 3, & \gamma[2] &:= \text{cob}, & \mathbf{r}(\{2\}) &:= \lambda,\end{aligned}$$

yields the cooperative contact process with cooperative branching rate λ .

A general set-up

Our summability condition now reduces to

$$\int_{\Omega} \mathbf{r}(d\omega) \kappa(\omega) < \infty \quad (\text{SUM}).$$

Let:

$(\mathbb{X}_{s,u})_{s \leq u}$ stochastic flow constructed from a Poisson set π ,
 X_0 S^Λ -valued random variable, independent of π .

Assuming (SUM), the process $(X_t)_{t \geq 0}$ defined as

$$X_t := \mathbb{X}_{s,s+t}(X_0) \quad (t \geq 0)$$

is the interacting particle system with generator G .

The mean-field limit

Let (Λ_N, \sim) be the complete graph with N vertices.

Let $(X_t^N)_{t \geq 0}$ be the particle system on Λ_N with generator G .

We are interested in the *empirical measure*

$$\mu_t^N := \frac{1}{N} \sum_{i=1}^N \delta_{X_t^N(i)} \quad (t \geq 0).$$

We will prove that in the limit $N \rightarrow \infty$, the process $(\mu_t^N)_{t \geq 0}$ solves the *mean-field equation*

$$\frac{\partial}{\partial t} \mu_t = \int_{\Omega} \mathbf{r}(d\omega) \{ \mathbf{T}_{\gamma[\omega]}(\mu_t) - \mu_t \} \quad (t \geq 0) \quad (\text{MEAN}),$$

where for any measurable $g : S^k \rightarrow S$, we define

$\mathbf{T}_g : \mathcal{P}(S) \rightarrow \mathcal{P}(S)$ by

$$\mathbf{T}_g(\mu) := \text{the law of } g(X_1, \dots, X_k),$$

where $(X_i)_{i \geq 1}$ are i.i.d. with law μ .

The mean-field contact process

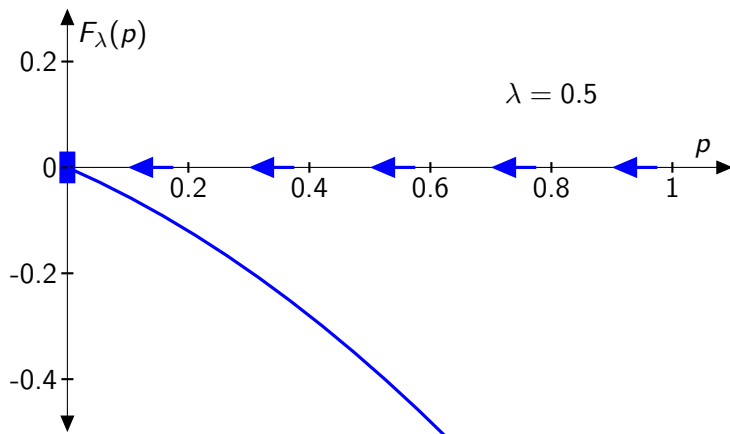
For the contact process, the mean-field equation takes the form

$$\frac{\partial}{\partial t} \mu_t = \lambda \{ \mathbf{T}_{\text{bra}}(\mu_t) - \mu_t \} + \{ \mathbf{T}_{\text{dth}}(\mu_t) - \mu_t \}.$$

Rewriting this in terms of $p_t := \mu_t(\{1\})$ yields

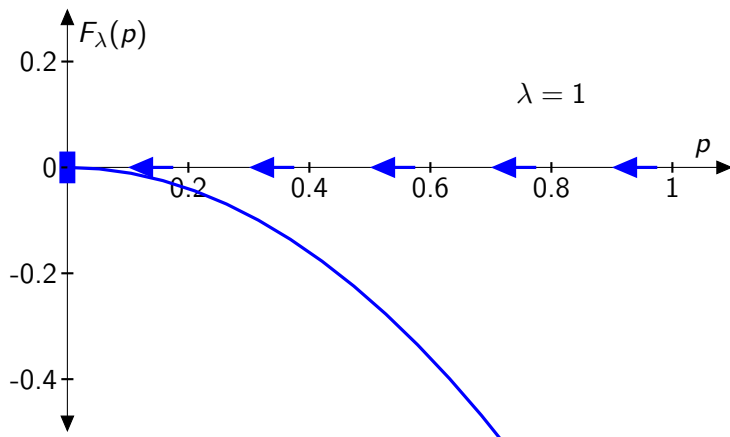
$$\frac{\partial}{\partial t} p_t = \lambda p_t (1 - p_t) - p_t =: F_\lambda(p_t) \quad (t \geq 0).$$

The mean-field contact process



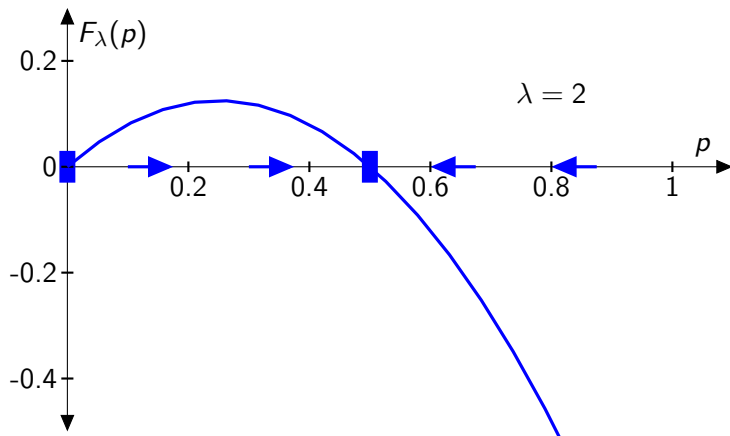
For $\lambda \leq 1$, the equation $\frac{\partial}{\partial t} p_t = F_\lambda(p_t)$ has a single, stable fixed point $p = 0$.

The mean-field contact process



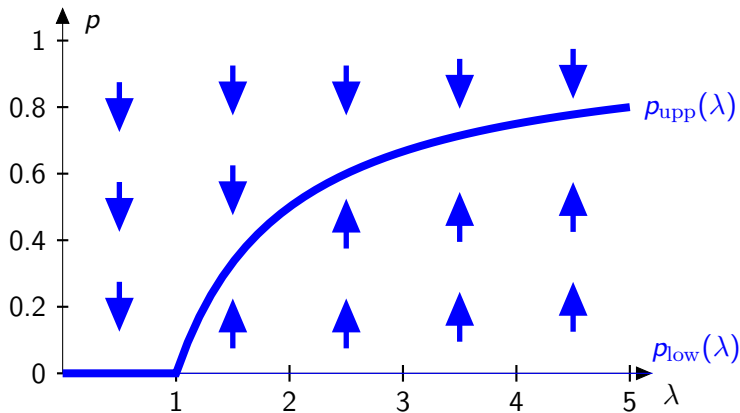
For $\lambda \leq 1$, the equation $\frac{\partial}{\partial t} p_t = F_\lambda(p_t)$ has a single, stable fixed point $p = 0$.

The mean-field contact process



For $\lambda > 1$, the fixed point $p = 0$ becomes unstable and a new stable fixed point $p(\lambda) = 1 - 1/\lambda$ appears.

The mean-field contact process



Fixed points and their domains of attraction as a function of λ .

The mean-field cooperative contact process

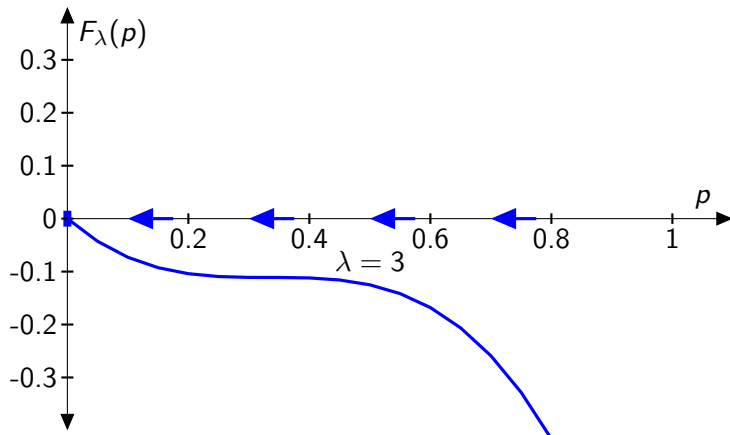
For the cooperative contact process, the mean-field equation takes the form

$$\frac{\partial}{\partial t} \mu_t = \lambda \{ \mathbf{T}_{\text{cob}}(\mu_t) - \mu_t \} + \{ \mathbf{T}_{\text{dth}}(\mu_t) - \mu_t \}.$$

Rewriting this in terms of $p_t := \mu_t(\{1\})$ yields

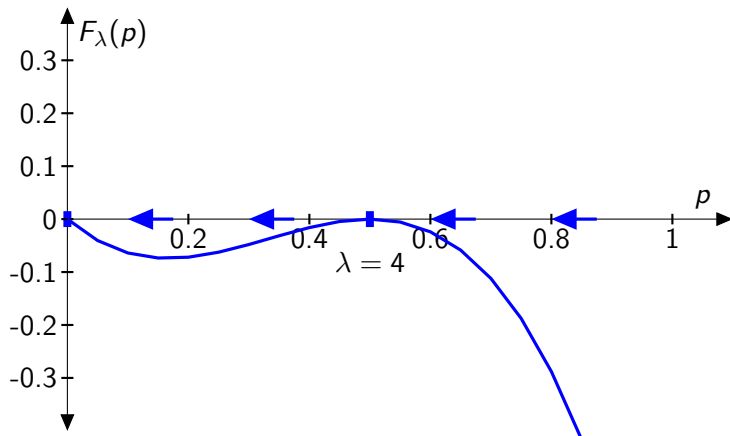
$$\frac{\partial}{\partial t} p_t = \lambda p_t^2 (1 - p_t) - p_t =: F_\lambda(p_t) \quad (t \geq 0).$$

The mean-field cooperative contact process



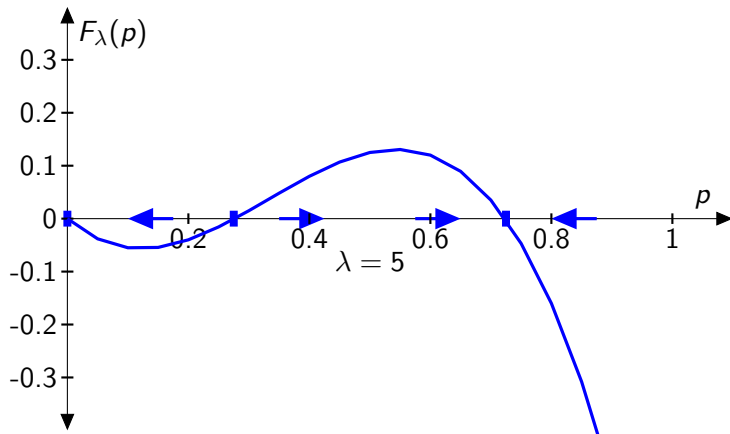
For $\lambda < 4$, the equation $\frac{\partial}{\partial t} p_t = F_\lambda(p_t)$ has a single, stable fixed point $p = 0$.

The mean-field cooperative contact process



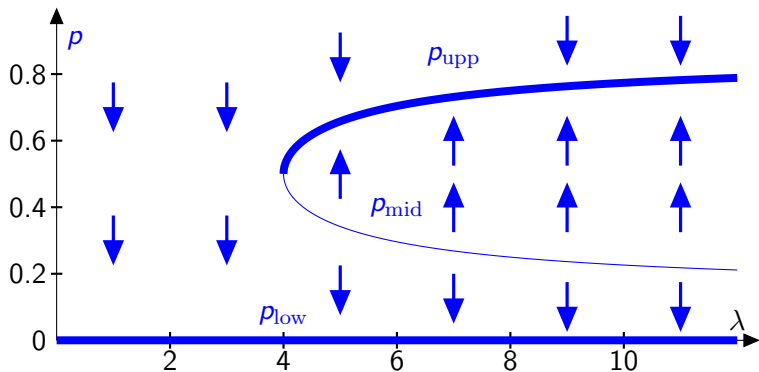
For $\lambda = 4$, a second fixed point appears at $p = 0.5$.

The mean-field cooperative contact process



For $\lambda > 4$, there are two stable fixed points and one unstable fixed point, which separates the domains of attraction of the other two.

The mean-field cooperative contact process



Fixed points of $\frac{\partial}{\partial t} p_t = F_\lambda(p_t)$ for different values of λ .

The abstract setting

Using the notation $|\mathbf{r}| := \mathbf{r}(\omega)$ and

$$\mathbf{T}(\mu) := |\mathbf{r}|^{-1} \int_{\Omega} \mathbf{r}(d\omega) \mathbf{T}_{\gamma[\omega]}(\mu),$$

we can rewrite the mean-field equation as

$$\frac{\partial}{\partial t} \mu_t = |\mathbf{r}| \{ \mathbf{T}(\mu_t) - \mu_t \} \quad (t \geq 0) \quad (\text{MEAN}).$$

Recall

$$\int_{\Omega} \mathbf{r}(d\omega) \kappa(\omega) < \infty \quad (\text{SUM}).$$

Theorem [Mach, Sturm, S. '20] Under the condition (SUM), the mean-field equation (MEAN) has a unique solution for each initial state $\mu_0 \in \mathcal{P}(S)$.

The abstract setting

We define a (nonlinear) semigroup $(\mathbf{T}_t)_{t \geq 0}$ of operators acting on probability measures by

$$\mathbf{T}_t(\mu) := \mu_t \quad \text{where } (\mu_t)_{t \geq 0} \text{ solves (MEAN) with } \mu_0 = \mu.$$

This is a sort of continuous-time version of the discrete evolution $\mu \mapsto \mathbf{T}(\mu) \mapsto \mathbf{T}^2(\mu) \mapsto \dots$.

Assuming that, for all $k \geq 0$ and $x \in S^k$,

$$\mathbf{r}(\{\omega : \kappa(\omega) = k, \gamma[\omega] \text{ is discontinuous at } x\}) = 0 \quad (\text{CONT}),$$

one can show that the operators \mathbf{T}^n and \mathbf{T}_t are continuous w.r.t. weak convergence.

The abstract setting

Let d be any metric that generates the topology of weak convergence and let $\|\cdot\|$ denote the total variation norm.

Theorem [Mach, Sturm, S. '20] Assume (SUM) and at least one of the following conditions:

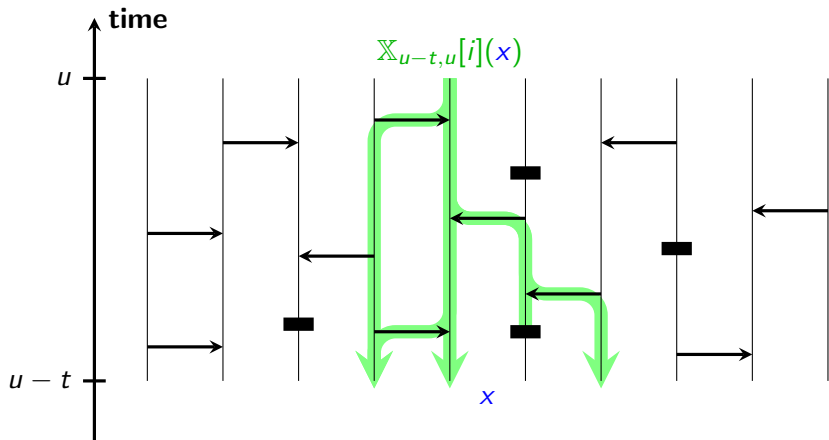
- (i) $\mathbb{P}[d(\mu_0^N, \mu_0) \geq \varepsilon] \xrightarrow{N \rightarrow \infty} 0$ for all $\varepsilon > 0$, and (CONT) holds.
- (ii) $\|\mathbb{E}[(\mu_0^N)^{\otimes n}] - \mu_0^{\otimes n}\| \xrightarrow{N \rightarrow \infty} 0$ for all $n \geq 1$.

Then

$$\mathbb{P}\left[\sup_{0 \leq t \leq T} d(\mu_t^N, \mathbf{T}_t(\mu_0)) \geq \varepsilon\right] \xrightarrow{N \rightarrow \infty} 0 \quad (\varepsilon > 0, T < \infty).$$

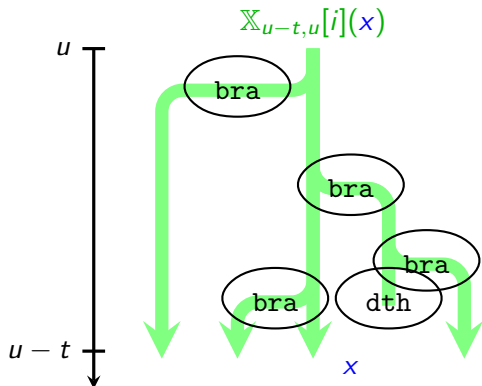
Proof By a probabilistic representation of the semigroup $(\mathbf{T}_t)_{t \geq 0}$.

The backward in time process



Recall the “backward in time” Markov process
 $(\mathcal{R}(\mathbb{X}_{u-t,u}[i]), \mathbb{X}_{u-t,u}[i])_{t \geq 0}$.

The backward in time process



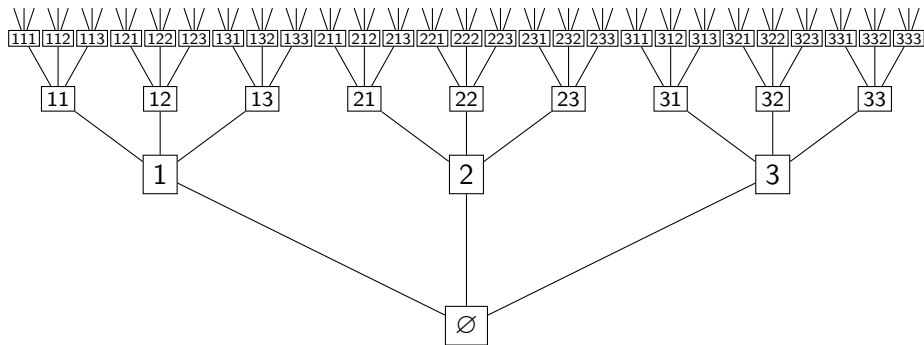
In the mean-field limit, the “backward in time” process becomes a random tree with maps attached to its nodes.

Aim Develop a stochastic representation for the nonlinear semigroup $(\mathbf{T}_t)_{t \geq 0}$, and also for the discrete-time evolution maps \mathbf{T}^n , in terms of a random tree with maps attached to its nodes.

Aldous & Banyopadyay (2005) (discrete time),
Mach, Sturm & S. (2020) (continuous time).

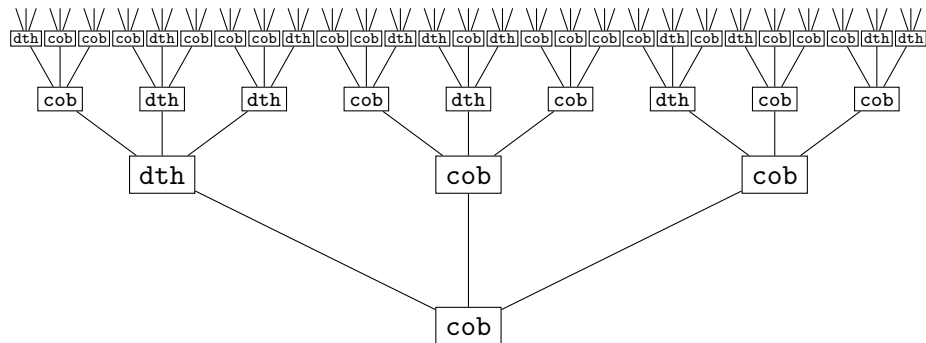
Fix $d \in \mathbb{N}_+ \cup \{\infty\}$ such that $\kappa_i(\omega) \leq d$ for all $\omega \in \Omega$. Let $\mathbb{T} = \mathbb{T}^d$ denote the space of all words $\mathbf{i} = i_1 \cdots i_n$ made from the alphabet $\{1, \dots, d\}$ (if $d < \infty$) resp. \mathbb{N}_+ (if $d = \infty$).

A recursive tree representation



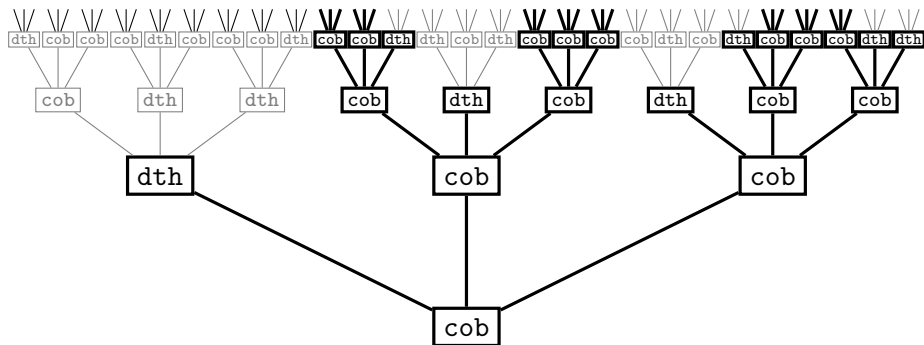
We view $\mathbb{T} = \mathbb{T}^d$ as a tree with root \emptyset , the word of length zero.

A recursive tree representation



We attach i.i.d. $(\omega_i)_{i \in \mathbb{T}}$ with law $|\mathbf{r}|^{-1} \mathbf{r}$ to each node, which translate into maps $(\gamma[\omega_i])_{i \in \mathbb{T}}$.

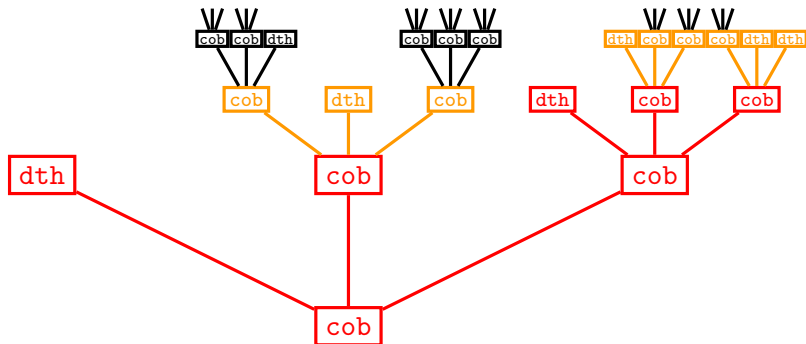
A recursive tree representation



Let \mathbb{S} be the random subtree of \mathbb{T} defined as

$$\mathbb{S} := \{i_1 \cdots i_n \in \mathbb{T} : i_m \leq \kappa(\omega_{i_1 \dots i_{m-1}}) \forall 1 \leq m \leq n\}.$$

A recursive tree representation

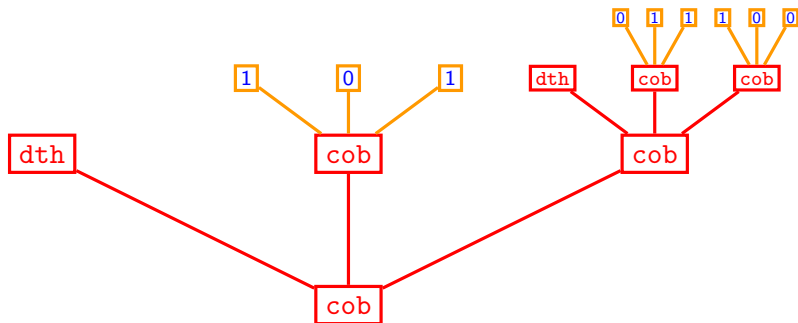


For any rooted subtree $\mathcal{U} \subset \mathcal{S}$, let

$$\nabla \mathcal{U} := \{i_1 \cdots i_n \in \mathcal{S} : i_1 \cdots i_{n-1} \in \mathcal{U}, i_1 \cdots i_n \notin \mathcal{U}\}$$

denote the boundary of \mathcal{U} relative to \mathcal{S} .

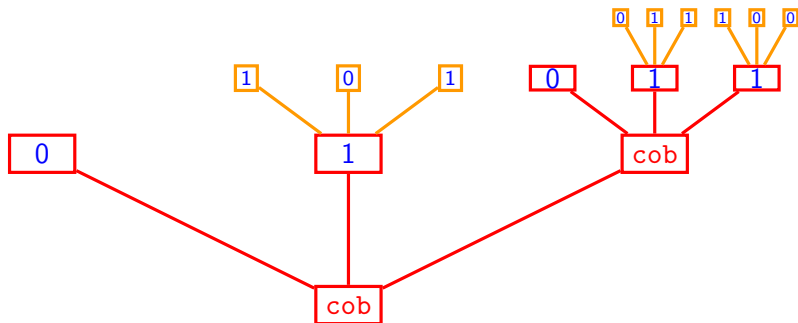
A recursive tree representation



Given $(X_i)_{i \in \nabla \mathbb{U}}$, we inductively define $(X_i)_{i \in \mathbb{U}}$ by

$$X_i = \gamma[\omega_i](X_{i_1}, \dots, X_{i_{\kappa(\omega)}}) \quad (i \in \mathbb{U}).$$

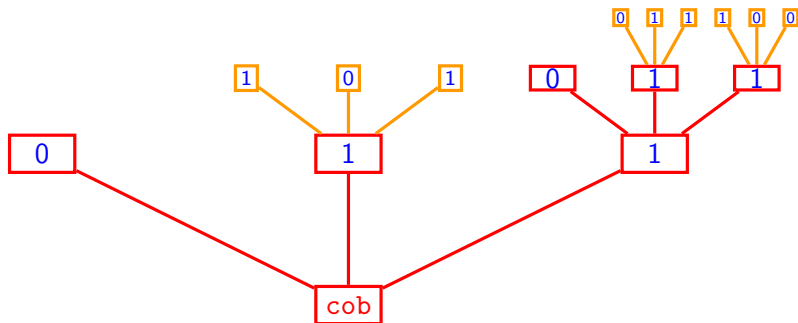
A recursive tree representation



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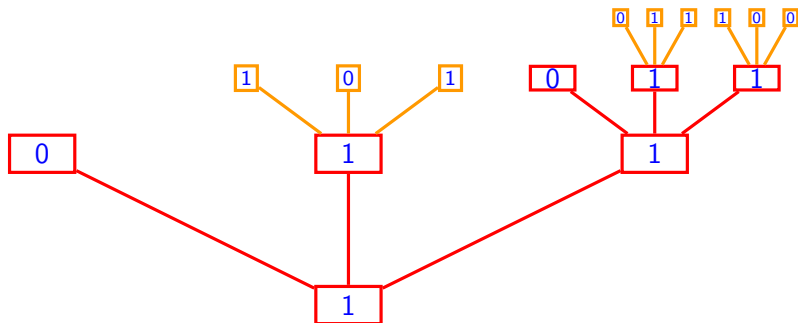
A recursive tree representation



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A recursive tree representation



Given $(X_i)_{i \in \nabla \mathbb{U}}$, we inductively define $(X_i)_{i \in \mathbb{U}}$ by

$$X_i = \gamma[\omega_i](X_{i_1}, \dots, X_{i_{\kappa(\omega)}}) \quad (i \in \mathbb{U}).$$

A recursive tree representation

Define $\Gamma_{\mathbb{U}} : S^{\nabla\mathbb{U}} \rightarrow S$ by $\Gamma_{\mathbb{U}}((X_{\mathbf{i}})_{\mathbf{i} \in \nabla\mathbb{U}}) := X_{\emptyset}$.

$\Gamma_{\mathbb{U}}$ is the concatenation of the maps $(\gamma[\omega_{\mathbf{i}}])_{\mathbf{i} \in \mathbb{U}}$ according to the tree structure of \mathbb{U} .

Let $|i_1 \cdots i_n| := n$ denote the length of a word \mathbf{i} and set

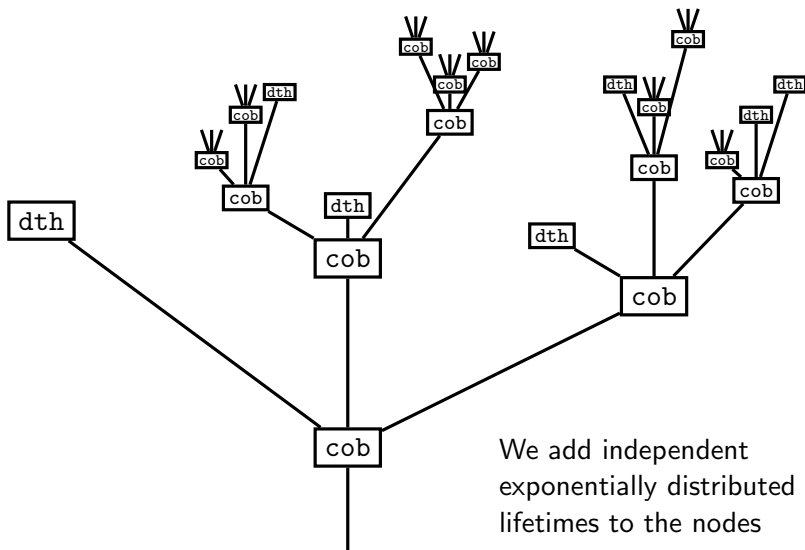
$$\mathbb{S}_{(n)} := \{\mathbf{i} \in \mathbb{S} : |\mathbf{i}| < n\} \quad \text{and} \quad \nabla\mathbb{S}_{(n)} = \{\mathbf{i} \in \mathbb{S} : |\mathbf{i}| = n\}.$$

Aldous and Bandyopadhyay (2005) proved that

$$\mathbf{T}^n(\mu) := \text{the law of } \Gamma_{\mathbb{S}_{(n)}}((X_{\mathbf{i}})_{\mathbf{i} \in \nabla\mathbb{S}_{(n)}}),$$

with $(X_{\mathbf{i}})_{\mathbf{i} \in \nabla\mathbb{S}_{(n)}}$ i.i.d. with law μ and independent of $(\omega_{\mathbf{i}})_{\mathbf{i} \in \mathbb{S}_{(n)}}$.

A recursive tree representation



We add independent exponentially distributed lifetimes to the nodes

A recursive tree representation

Let $(\sigma_{\mathbf{i}})_{\mathbf{i} \in \mathbb{T}}$ be i.i.d. exponentially distributed with mean $|\mathbf{r}|^{-1}$, independent of $(\omega_{\mathbf{i}})_{\mathbf{i} \in \mathbb{T}}$, and set

$$\tau_{\mathbf{i}}^* := \sum_{m=1}^{n-1} \sigma_{i_1 \dots i_m} \quad \text{and} \quad \tau_{\mathbf{i}}^\dagger := \tau_{\mathbf{i}}^* + \sigma_{\mathbf{i}} \quad (\mathbf{i} = i_1 \dots i_n),$$
$$\mathbb{S}_t := \{\mathbf{i} \in \mathbb{S} : \tau_{\mathbf{i}}^\dagger \leq t\} \quad \text{and} \quad \nabla \mathbb{S}_t = \{\mathbf{i} \in \mathbb{S} : \tau_{\mathbf{i}}^* \leq t < \tau_{\mathbf{i}}^\dagger\}.$$

Let \mathcal{F}_t be the filtration

$$\mathcal{F}_t := \sigma(\nabla \mathbb{S}_t, (\omega_{\mathbf{i}}, \sigma_{\mathbf{i}})_{\mathbf{i} \in \mathbb{S}_t}) \quad (t \geq 0).$$

Theorem [Mach, Sturm, S. '20]

$$\mathbf{T}_t(\mu) := \text{the law of } \Gamma_{\mathbb{S}_t}((X_{\mathbf{i}})_{\mathbf{i} \in \nabla \mathbb{S}_t}),$$

where $(X_{\mathbf{i}})_{\mathbf{i} \in \nabla \mathbb{S}_t}$ are i.i.d. with law μ and independent of \mathcal{F}_t .

A recursive tree representation

