

Subcritical branching processes in random environment

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and under the influence of

V.Afanasyev, Ch.Boeinghoff, J.Geiger and G.Kersting

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Galton-Watson processes in random environment

- Offspring generating functions $f_n(s) := \mathbf{E}s^{\xi^{(n)}}$ in generations $n = 0, 1, \dots$ are **RANDOM** and **I.I.D.**

$\Rightarrow f'_n(1) := \mathbf{E}\xi^{(n)}$ are **I.I.D. RANDOM** variables;

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$$Z_{n+1} = \sum_{j=1}^{Z_n} \xi_j^{(n)} \quad \text{or} \quad \mathbf{E}\left[s^{Z_{n+1}} \mid Z_n; f_0, f_1, \dots\right] = \left(f_n(s)\right)^{Z_n}.$$

- $\xi_j^{(n)} \stackrel{d}{=} \xi^{(n)}$ are i.i.d. **given** f_0, f_1, \dots

Associated random walk

- $X_i := \log f'_{i-1}(1)$
- $S_0 = 0, \quad S_n = \log f'_0(1) + \log f'_1(1) + \cdots + \log f'_{n-1}(1), \quad n \geq 1.$

$$\mathbf{E} \left[Z_n \mid f_0, f_1, \dots \right] =: \mathbf{E}_f \left[Z_n \right] = \prod_{j=0}^{n-1} f'_j(1) = e^{S_n}$$

Associated random walk

- $X_i := \log f'_{i-1}(1)$
- $S_0 = 0, \quad S_n = \log f'_0(1) + \log f'_1(1) + \cdots + \log f'_{n-1}(1), \quad n \geq 1.$

$$\mathbf{E}_f [Z_n] = e^{S_n}$$

Classification (Afanasyev, Geiger, Kersting, Vatutin (2005), Ann. of Probab.): A BPRE is called

- **super**critical if $\lim_{n \rightarrow \infty} S_n = +\infty$ with probability 1;
- **sub**critical if $\lim_{n \rightarrow \infty} S_n = -\infty$ with probability 1;
- critical if $\limsup_{n \rightarrow \infty} S_n = +\infty$ **and** $\liminf_{n \rightarrow \infty} S_n = -\infty$ (both with probability 1);

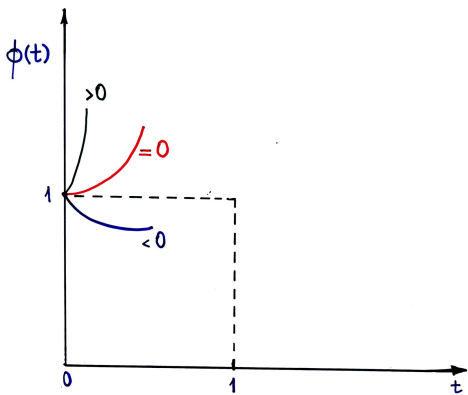
- $X_i := \log f'_{i-1}(1)$
- $S_0 = 0, \quad S_n = X_1 + X_2 + \cdots + X_n, \quad n \geq 1.$

We may express the classification in terms of the moment generating function:

$$\phi(t) = \mathbf{E}e^{tX} = \mathbf{E}e^{t \log f'(1)}.$$

Classification: A BPRE is called

- **super**critical if $\phi'(0) = \mathbf{E}X = \mathbf{E} \log f'(1) > 0,$
- critical if $\phi'(0) = \mathbf{E}X = \mathbf{E} \log f'(1) = 0$ or $\phi'(0)$ does not exist,
- **sub**critical if $\phi'(0) = \mathbf{E}X = \mathbf{E} \log f'(1) < 0.$



$$\phi(t) = E e^{tX}$$

Quenched approach:

The study the behavior of characteristics of a BPRE for **typical realizations** of the environment f_0, f_1, \dots .

For instance, if

$$T = \min\{n : Z_n = 0\}$$

then

$$\mathbf{P}(T > n | f_0, f_1, \dots) = \mathbf{P}_f(T > n) = \mathbf{P}_f(Z_n > 0)$$

is a random variable.

Annealed approach:

The study the behavior of characteristics of a BPRE performing **averaging over possible scenarios** f_0, f_1, \dots on the space of realizations of the environment:

$$\mathbf{P}(Z_n > 0) = \mathbf{E}[\mathbf{P}_f(Z_n > 0)]$$

is a number.

ANNEALED APPROACH
SUBCRITICAL PROCESSES

$$\mathbf{E}X = \mathbf{E} \log f'(1) < 0$$

The problems to be considered **under the annealed approach**:

- Survival probabilities $\mathbf{P}(Z_n > 0)$ as $n \rightarrow \infty$;
- The environments providing survival $\mathbf{P}(S_m \in dx | Z_n > 0)$ as $n \rightarrow \infty$;
- Conditional limit theorems $\mathbf{P}(Z_{nt} \in dx | Z_n > 0)$ as $n \rightarrow \infty$.

Subcritical processes: **five** different sub-cases:

- **strongly** subcritical, if

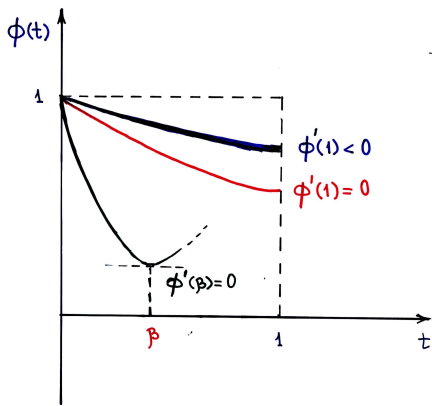
$$\phi'(1) = \mathbf{E} [Xe^X] < 0,$$

- **intermediately** subcritical, if

$$\phi'(1) = \mathbf{E} [Xe^X] = 0,$$

- **weakly** subcritical, if there exists $0 < \beta < 1$ such that

$$\phi'(\beta) = \mathbf{E} [Xe^{\beta X}] = 0.$$



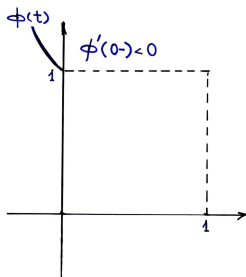
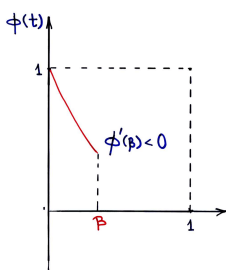
Subcritical processes ($\mathbf{E}X = \mathbf{E} \log f'(1) < 0$): **five** different sub-cases:

- **Non-Cramer** subcritical, if

$$\beta = \sup\{t \geq 0 : \mathbf{E}[e^{tX}] < \infty\} \in [0, 1)$$

and

$$\phi'(\beta) = \mathbf{E}[Xe^{\beta X}] < 0.$$

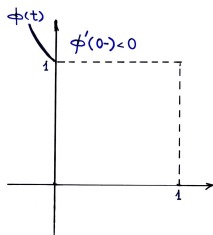
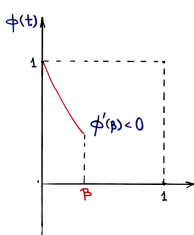


Two examples: the first $\beta \in (0, 1)$:

$$p_X(x) = \begin{cases} \frac{c}{(|x|+1)^4} & \text{if } x < 0; \\ \frac{c}{(|x|+1)^4} e^{-\beta x} & \text{if } x \geq 0. \end{cases}$$

The second $\beta = 0$:

$$p_X(x) = \begin{cases} \frac{c}{(|x|+1)^4} & \text{if } x < 0; \\ \frac{c}{(|x|+1)^5} & \text{if } x \geq 0. \end{cases}$$



Change of measure for ordinary RW meeting the **Cramer** condition:

$$\phi(t) = \mathbf{E}e^{tX} \text{ and } \phi'(\beta) = \mathbf{E}Xe^{\beta X} = 0$$

Change of measure

$$Y_n = \varphi(S_1, S_2, \dots, S_n)$$

⇒ New measure: $\hat{\mathbf{P}}_\beta, \beta > 0$ or $\hat{\mathbf{E}}_\beta$

$$\hat{\mathbf{E}}_\beta Y_n = \frac{\mathbf{E}Y_n e^{\beta S_n}}{\mathbf{E}e^{\beta S_n}} = \frac{\mathbf{E}Y_n e^{\beta S_n}}{\phi^n(\beta)}$$

with

$$\hat{\mathbf{E}}_\beta X = \frac{\mathbf{E}Xe^{\beta X}}{\mathbf{E}e^{\beta X}} = \frac{\phi'(\beta)}{\phi(\beta)} = 0$$

The change of measure for branching processes in RE

$$Y_n = \varphi(f_1, \dots, f_{n-1}; Z_0, \dots, Z_n)$$

We introduce a family $\hat{\mathbf{P}}_\beta, \beta > 0$ of probability measures, given by

$$\hat{\mathbf{E}}_\beta[Y_n] = \frac{\mathbf{E}[Y_n e^{\beta S_n}]}{\mathbf{E}[e^{\beta S_n}]} = \gamma^{-n} \mathbf{E}[Y_n e^{\beta S_n}]$$

with

$$\gamma = \gamma_\beta = \mathbf{E}[e^{\beta X}]$$

or

$$\mathbf{E}[Y_n] = \gamma^n \hat{\mathbf{E}}_\beta[Y_n e^{-\beta S_n}].$$

It is necessary to select β in such a way for

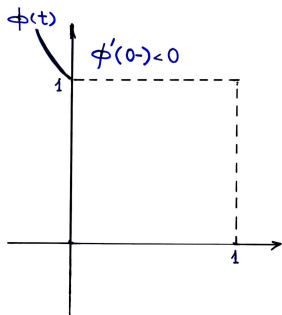
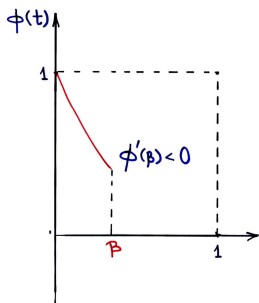
$$\hat{\mathbf{E}}_\beta[X] = \frac{\mathbf{E}[X e^{\beta X}]}{\mathbf{E}[e^{\beta X}]} = 0$$

- **Non-Cramer** subcritical, if

$$\beta = \sup\{t \geq 0 : \mathbf{E}[e^{tX}] < \infty\} \in [0, 1)$$

and

$$\phi'(\beta) = \mathbf{E}[Xe^{\beta X}] < 0.$$



Non-Cramer subcritical

$\beta = \sup\{t \geq 0 : \mathbf{E}[e^{tX}] < \infty\} \in [0, 1)$ and $\phi'(\beta) = \mathbf{E}[Xe^{\beta X}] < 0$.

The case $\beta = 0$:

$$\phi'(0-) = \mathbf{E}X = \mathbf{E} \log f'(1) = -a < 0$$

and

$$\phi(t) = \mathbf{E}e^{tX} = \infty$$

for all $t > 0$.

Non-Cramer subcritical

(V.+Xinghua Zheng) The case $\beta = 0$:

Hypothesis A.

$$\mathbf{E}X = -a < 0, \quad \mathbf{P}(X > x) \sim \frac{l(x)}{x^b}, b > 2.$$

Non-Cramer subcritical

(V.+Xinghua Zheng) The case $\beta = 0$:

Hypothesis A.

$$\mathbf{E}X = -a < 0, \quad \mathbf{P}(X > x) \sim \frac{l(x)}{x^b}, b > 2.$$

Let $f'(1) = \mathbf{E}_f[\xi] = e^X$.

Hypothesis B. For each $\lambda > 0$ as $y \rightarrow \infty$,

$$\mathcal{L} \left(\mathbf{E}_f \left[\exp \left\{ -\lambda \frac{Z_1}{y} \right\} \right] \mid f'(1) > y \right) \xrightarrow{d} \mathcal{L}(\gamma),$$

where γ is a random variable which is less than 1 with a positive probability and is **independent** of λ .

Hypothesis B. If

$$f(s) = \sum_{k=0}^{\infty} p_k(f) s^k$$

then the sequence of random distributions

$$\mathbf{P}_{y,f}(x) = \mathbf{P}_f \left(\frac{Z_1}{y} \leq x \right) = \sum_{0 \leq k \leq xy} p_k(f)$$

possesses the property: conditionally on $f'(1) > y$ for any $x > 0$

$$\mathbf{P}_f \left(\frac{Z_1}{y} \leq x \right) \xrightarrow{d} \gamma < 1$$

as $y \rightarrow \infty$. Thus, the limiting measure is

$$\delta_0 \gamma + (1 - \gamma) \delta_{\infty}.$$

$$\mathbf{E}X = -a < 0, \quad \mathbf{P}(X > x) \sim \frac{l(x)}{x^b}, b > 2.$$

If Hypothesis A and B are valid then

$$\mathbf{P}(Z_n > 0) \sim K\mathbf{P}(X > na) \sim K_1\mathbf{P}\left(\min_{0 \leq i \leq n} S_i \geq 0\right).$$

Let

$$U_n = \inf \{j : X_j > na\}$$

be **the moment of the first big jump.**

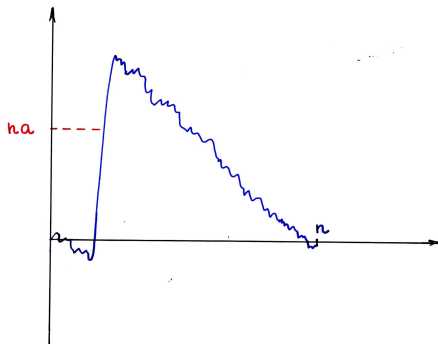
Then

$$\lim_{n \rightarrow \infty} \mathbf{P}(U_n = j | Z_n > 0) = p_j, \quad \sum_{j=1}^{\infty} p_j = 1.$$

In fact, we have **only one big jump exceeding na** , and

$$\lim_{M \rightarrow \infty} \lim_{n \rightarrow \infty} \mathbf{P}(S_{U_n} > na - M\sqrt{n} | Z_n > 0) = 1.$$

Non-Cramer subcritical $\beta = 0$



Non-Cramer subcritical $\beta = 0$

Moreover, for any $\varepsilon \in (0, 1)$

$$\mathcal{L} \left(\frac{Z_{nt}}{Z_{U_n} \exp(S_{nt} - S_{U_n})}, \varepsilon \leq t \leq 1 \mid Z_n > 0 \right) \implies \mathcal{L}(1, \varepsilon \leq t \leq 1),$$

and

$$\mathcal{L} \left(\frac{1}{\sigma\sqrt{n}} \left(\log \frac{Z_{nt}}{Z_{U_n}} + nta \right), \varepsilon \leq t \leq 1 \mid Z_n > 0 \right) \implies \mathcal{L}(B_t, \varepsilon \leq t \leq 1),$$

where B_t is a Brownian motion.

Non-Cramer subcritical $\beta \in (0, 1)$

$\beta = \sup\{t \geq 0 : \mathbf{E}[e^{tX}] < \infty\} \in [0, 1)$ and $\phi'(\beta) = \mathbf{E}[Xe^{\beta X}] < 0$.

Hypothesis A1. The distribution of X has density

$$p_X(x) = \frac{l_0(x)}{x^{b+1}} e^{-\beta x},$$

where $l_0(x)$ is a function slowly varying at infinity, $b > 2$, $\beta \in (0, 1)$.

Hypothesis A1. The distribution of X has density

$$p_X(x) = \frac{l_0(x)}{x^{b+1}} e^{-\beta x},$$

where $l_0(x)$ is a function slowly varying at infinity, $b > 2$, $\beta \in (0, 1)$.

Hypothesis B1. There exists a random function

$$g(\lambda), \lambda \in [0, \infty), 0 < g(\lambda) < 1$$

for all $\lambda > 0$, and $\lim_{\lambda \rightarrow \infty} g(\lambda) = 0$ such that

$$\mathcal{L} \left(\mathbf{E}_f \left[\exp \left\{ -\lambda \frac{Z_1}{y} \right\} \mid f'(1) = y \right] \right) \xrightarrow{d} \mathcal{L}(g(\lambda)).$$

Hypothesis B1. If

$$f(s) = \sum_{k=0}^{\infty} p_k(f) s^k$$

then the sequence of random distributions

$$\mathbf{P}_{y,f}(x) = \mathbf{P}_f \left(\frac{Z_1}{y} \leq x \right) = \sum_{0 \leq k \leq xy} p_k(f)$$

possesses the property: conditionally on $f'(1) = y$ for any $x > 0$

$$\mathbf{P}_f \left(\frac{Z_1}{y} \leq x \right) \xrightarrow{d} \mathbf{P}_\omega (\xi \leq x)$$

as $y \rightarrow \infty$ and the random distribution $\mathbf{P}_\omega (\xi \leq x)$ is proper and nondegenerate with probability 1.

(V.+ Vincent Bansaye) The case $\beta \in (0, 1)$:

Let

$$a = -\frac{\phi'(\beta)}{\phi(\beta)} > 0.$$

If Hypothesis A1 and B1 are valid then

$$\mathbf{P}(Z_n > 0) \sim C_1 \mathbf{P}\left(\min_{0 \leq k \leq n} S_k \geq 0\right) \sim C \phi^n(\beta) \frac{l_0(n)}{(an)^{b+1}}.$$

Let

$$U_n = \inf \{j : X_j > na(1 - \varepsilon)\}, \quad \varepsilon \in (0, 1).$$

be **the moment of the first big jump.**

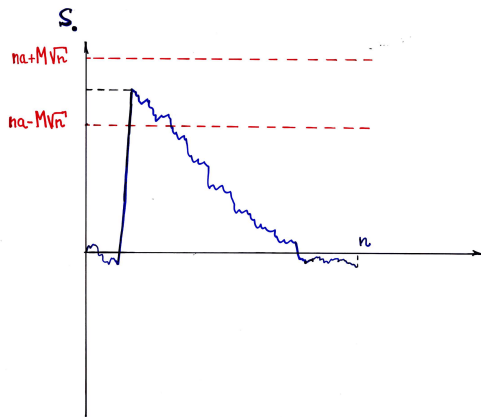
Then for **any** $\varepsilon \in (0, 1)$

$$\lim_{n \rightarrow \infty} \mathbf{P}(U_n = j | Z_n > 0) = p_j, \quad \sum_{j=1}^{\infty} p_j = 1.$$

and

$$\lim_{M \rightarrow \infty} \lim_{n \rightarrow \infty} \mathbf{P}(S_{U_n} \in [na - M\sqrt{n}, na + M\sqrt{n}] | Z_n > 0) = 1.$$

Non-Cramer subcritical $\beta > 0$



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

$$\lim_{n \rightarrow \infty} \mathbf{E}[s^{Z_n} | Z_n > 0] = F(s) = \sum_{k=1}^{\infty} p_k s^k,$$

with $\sum_{k=1}^{\infty} p_k = 1$.

The change of measure for branching processes in RE

$$Y_n = \varphi(f_1, \dots, f_{n-1}; Z_0, \dots, Z_n)$$

We introduce a family $\hat{\mathbf{P}}_\beta, \beta > 0$ of probability measures, given by

$$\hat{\mathbf{E}}_\beta[Y_n] = \frac{\mathbf{E}[Y_n e^{\beta S_n}]}{\mathbf{E}[e^{\beta S_n}]} = \gamma^{-n} \mathbf{E}[Y_n e^{\beta S_n}]$$

with

$$\gamma = \gamma_\beta = \mathbf{E}[e^{\beta X}]$$

or

$$\mathbf{E}[Y_n] = \gamma^n \hat{\mathbf{E}}_\beta[Y_n e^{-\beta S_n}].$$

It is necessary to select β in such a way for

$$\hat{\mathbf{E}}_\beta[X] = \frac{\mathbf{E}[X e^{\beta X}]}{\mathbf{E}[e^{\beta X}]} = 0$$

$$\mathbf{E}[Y_n] = \gamma^n \hat{\mathbf{E}}_\beta[Y_n e^{-\beta S_n}]$$

Thus, for

$$\gamma = \gamma_\beta = \mathbf{E}[e^{\beta X}]$$

$$\mathbf{P}(Z_n > 0) = \mathbf{E}[\mathbf{P}_f(Z_n > 0)] = \gamma^n \hat{\mathbf{E}}_\beta[\mathbf{P}_f(Z_n > 0) e^{-\beta S_n}].$$

One can show that under mild conditions

$$\mathbf{P}_f(Z_n > 0) \asymp e^{\min_{0 \leq k \leq n} S_k}.$$

Thus,

$$\mathbf{P}(Z_n > 0) \asymp \gamma^n \hat{\mathbf{E}}_\beta[e^{\min_{0 \leq k \leq n} S_k - \beta S_n}].$$

Intermediately subcritical (Afanasyev, Boeinghoff, Kersting+V.):

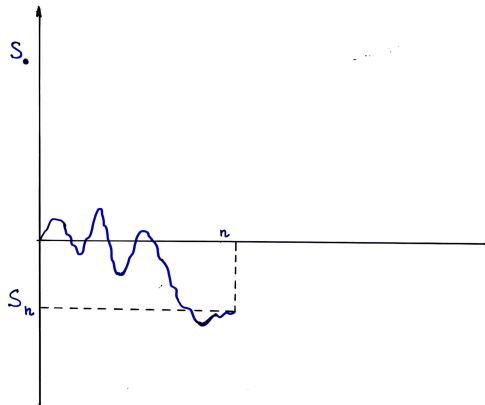
$$\frac{\phi'(1)}{\phi(1)} = \frac{\mathbf{E}[Xe^X]}{\mathbf{E}[e^X]} = \hat{\mathbf{E}}_1[X] = 0,$$

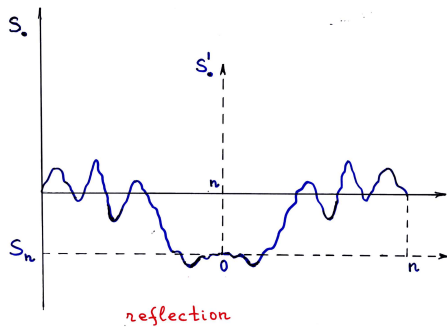
and

$$\frac{\mathbf{E}[X^2e^X]}{\mathbf{E}[e^X]} = \hat{\mathbf{E}}_1[X^2] \in (0, \infty).$$

Then one should take $\beta = 1$, the random walk is recurrent under $\hat{\mathbf{P}}_1$

$$\hat{\mathbf{E}}_1[e^{\min_{0 \leq k \leq n} S_k - S_n}] \asymp \hat{\mathbf{P}}_1(\min_{0 \leq k \leq n} S_k \approx S_n).$$





Intermediately subcritical:

$$\frac{\phi'(1)}{\phi(1)} = \frac{\mathbf{E}[Xe^X]}{\mathbf{E}[e^X]} = \hat{\mathbf{E}}_1[X] = 0,$$

and

$$\frac{\mathbf{E}[X^2e^X]}{\mathbf{E}[e^X]} = \hat{\mathbf{E}}_1[X^2] \in (0, \infty).$$

Then $\beta = 1$, the random walk is recurrent under $\hat{\mathbf{P}}_1$

$$\hat{\mathbf{P}}_1\left(\min_{0 \leq k \leq n} S_k \approx S_n\right) \approx \hat{\mathbf{P}}_1\left(\min_{0 \leq k \leq n} S'_k \geq -a\right) \sim \frac{C}{\sqrt{n}}, \quad a > 0.$$

Intermediately subcritical:

Hence,

$$\mathbf{P}(Z_n > 0) \asymp \gamma^n \hat{\mathbf{E}}_1[e^{\min_{0 \leq k \leq n} S_k - S_n}] \asymp \gamma^n \hat{\mathbf{P}}_1(\min_{0 \leq k \leq n} S_k \approx S_n) \sim C \frac{\gamma^n}{\sqrt{n}}.$$

Intermediately subcritical:

Trajectories of $\{S_n, n \geq 0\}$ that provide survival:

Let $B^* = (B_t^*, 0 \leq t \leq 1)$ denote a Brownian motion on $[0, 1]$ conditioned to attain its minimum at time $t = 1$. Then

$$\mathcal{L} \left(\frac{1}{\sigma\sqrt{n}} S_{nt}, 0 \leq t \leq 1 | Z_n > 0 \right) \Rightarrow \mathcal{L} (B_t^*, 0 \leq t \leq 1).$$

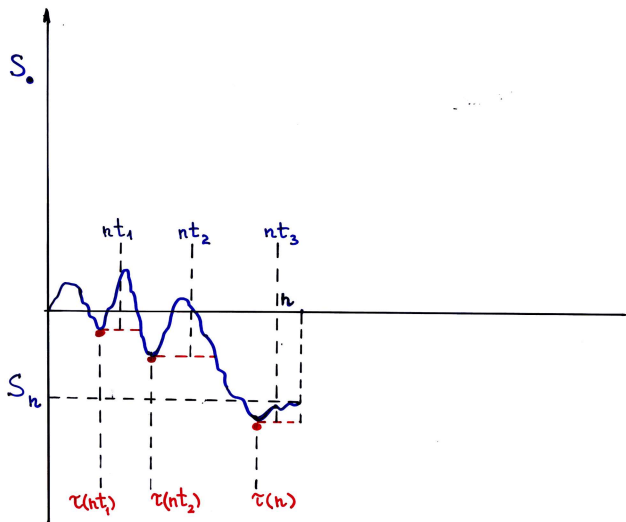
Intermediately subcritical

Let $\tau(m)$ be the moment at which the minimal value of the random walk $\{S_k, 0 \leq k \leq m\}$ is attained.

Assume that for $0 < t_1 < \dots < t_k < 1$ the **random moments**

$$\tau(nt_1) < \tau(nt_2) < \dots < \tau(nt_k)$$

are increasing.



Intermediately subcritical: $\mathbf{E} [Xe^X] = 0$

Let $\tau(m)$ be the moment at which the minimal value of the random walk $\{S_k, 0 \leq k \leq m\}$ is attained.

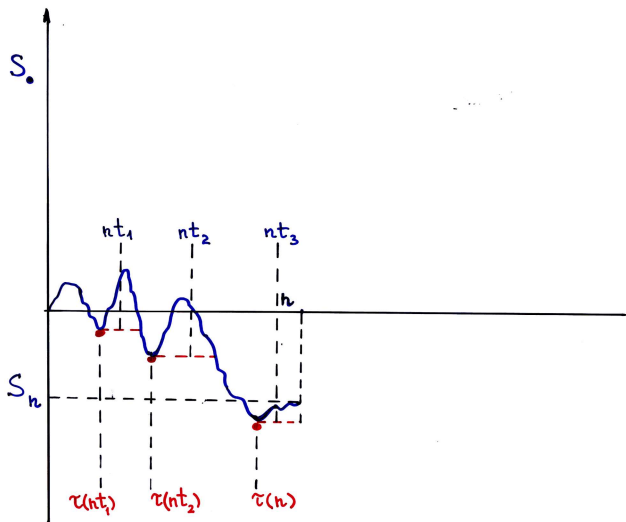
Assume that for $0 < t_1 < \dots < t_k < 1$ the **random moments**

$$\tau(nt_1) < \tau(nt_2) < \dots < \tau(nt_k).$$

are increasing. Then

$$\mathcal{L} \left(\frac{Z_{nt_1}}{e^{S_{nt_1} - S_{\tau(nt_1)}}}, \dots, \frac{Z_{nt_k}}{e^{S_{nt_k} - S_{\tau(nt_k)}}} \mid Z_n > 0 \right) \xrightarrow{d} \mathcal{L}(V_1, \dots, V_k)$$

where V_1, V_2, \dots are i.i.d. copies of a strictly positive random variable V .



Intermediately subcritical:

$$\mathcal{L} \left(\frac{Z_{nt_1}}{e^{S_{nt_1} - S_{\tau(nt_1)}}}, \dots, \frac{Z_{nt_k}}{e^{S_{nt_k} - S_{\tau(nt_k)}}} \mid Z_n > 0 \right) \xrightarrow{d} \mathcal{L}(V_1, \dots, V_k)$$

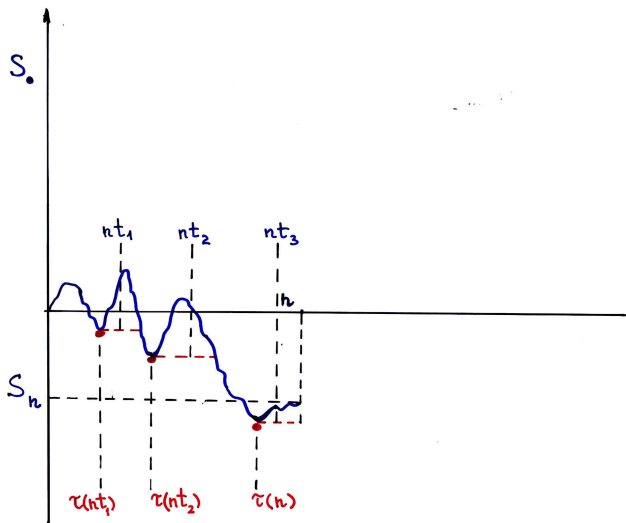
where V_1, V_2, \dots are i.i.d. copies of a strictly positive random variable V .

In addition,

$$\left((Z_{\tau(nt_1)}, \dots, Z_{\tau(nt_k)}) \mid Z_n > 0 \right) \xrightarrow{d} (W_1, \dots, W_k)$$

where W_1, W_2, \dots are i.i.d. copies of some strictly positive **integer-valued** random variable W .

BOTTLENECKS!!



Weakly subcritical (Afanasyev, Boeinghoff, Kersting+V.):

There exists $0 < \beta < 1$ such that

$$\mathbf{E} [X e^{\beta X}] = 0, \quad \mathbf{E} [X^2 e^{\beta X}] \in (0, \infty).$$

Then $\gamma := \mathbf{E} [e^{\beta X}]$.

Weakly subcritical (Afanasyev, Boeinghoff, Kersting+V.):

There exists $0 < \beta < 1$ such that

$$\mathbf{E} [X e^{\beta X}] = 0, \quad \mathbf{E} [X^2 e^{\beta X}] \in (0, \infty).$$

Then $\gamma := \mathbf{E} [e^{\beta X}]$ and

$$\mathbf{P} (Z_n > 0) = \gamma^n \hat{\mathbf{E}}_{\beta} [\mathbf{P}_f(Z_n > 0) e^{-\beta S_n}] \asymp \gamma^n \hat{\mathbf{E}}_{\beta} [e^{\min(S_0, S_1, \dots, S_n) - \beta S_n}]$$

Weakly subcritical (Afanasyev, Boeinghoff, Kersting+V.):

There exists $0 < \beta < 1$ such that

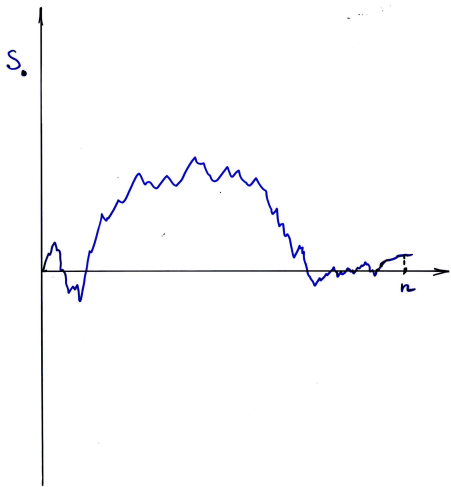
$$\mathbf{E} [X e^{\beta X}] = 0, \quad \mathbf{E} [X^2 e^{\beta X}] \in (0, \infty).$$

Then $\gamma := \mathbf{E} [e^{\beta X}]$ and

$$\mathbf{P} (Z_n > 0) = \gamma^n \hat{\mathbf{E}}_\beta [\mathbf{P}_f(Z_n > 0) e^{-\beta S_n}] \asymp \gamma^n \hat{\mathbf{E}}_\beta [e^{\min(S_0, S_1, \dots, S_n) - \beta S_n}]$$

and

$$\hat{\mathbf{E}}_\beta [e^{\min(S_0, S_1, \dots, S_n) - \beta S_n}] \asymp \hat{\mathbf{P}}_\beta (\min(S_0, S_1, \dots, S_n) \approx 0, S_n \approx 0) \asymp n^{-3/2}.$$



Therefore,

$$\mathbf{P}(Z_n > 0) \asymp \gamma^n \hat{\mathbf{E}}_\beta \left[e^{\min(S_0, S_1, \dots, S_n) - \beta S_n} \right] \sim \gamma^n \frac{C}{n^{3/2}}.$$

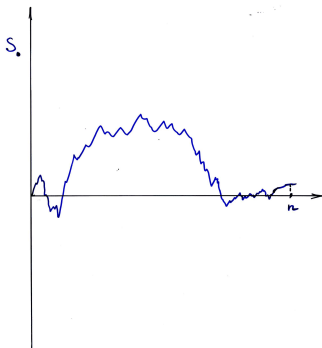
In fact, it follows from **Borovkov**³ that

$$\mathbf{P}(Z_n > 0) \sim \gamma^n \frac{C}{n^{3/2}} \sim C_2 \mathbf{P} \left(\min_{0 \leq k \leq n} S_k \geq 0 \right).$$

Environment providing survival

$$\mathcal{L} \left(\frac{S_{nt}}{\sigma\sqrt{n}}, 0 \leq t \leq 1 \mid Z_n > 0 \right) \Rightarrow \mathcal{L} (B_t^0, 0 \leq t \leq 1),$$

where $B_t^0, 0 \leq t \leq 1$ is the Brownian excursion.



Limit theorems

$$\mathcal{L}(Z_n | Z_n > 0) \rightarrow \mathcal{L}(\zeta).$$

where ζ is a proper integer-valued r.v.

For each $\varepsilon \in (0, 1/2)$

$$\mathcal{L}\left(\frac{Z_{nt}}{e^{S_{nt}}}, \varepsilon \leq t \leq 1 - \varepsilon | Z_n > 0\right) \Rightarrow \mathcal{L}(W_t, \varepsilon \leq t \leq 1 - \varepsilon),$$

where $W_t = W, 0 < t < 1$ a.s. and $\mathbf{P}(0 < W < \infty) = 1$.

Strongly subcritical (Afanasyev, Geiger, Kersting+V.):

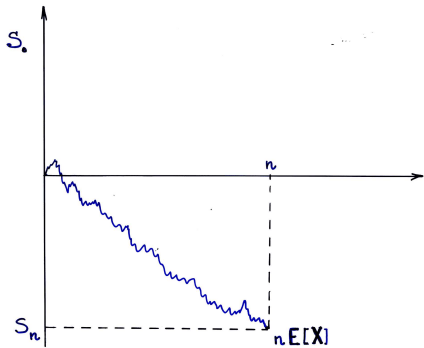
$$\frac{\phi'(1)}{\phi(1)} = \mathbf{E} [Xe^X] = \hat{\mathbf{E}}_1[X] < 0.$$

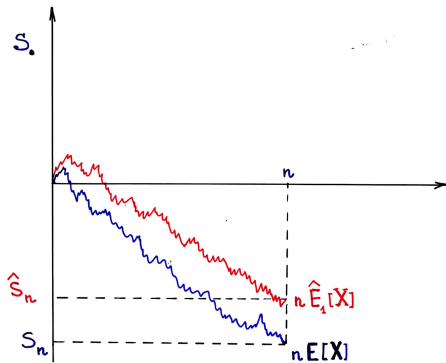
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Then $\beta = 1$

$$\hat{\mathbf{E}}_1[e^{\min_{0 \leq k \leq n} S_k - S_n}] \asymp \hat{\mathbf{P}}_1(\min_{0 \leq k \leq n} S_k \approx S_n) \approx C \in (0, \infty).$$

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Hence,

$$\mathbf{P}(Z_n > 0) \sim C\gamma^n.$$

For every $\varepsilon > 0$

$$\lim_{n \rightarrow \infty} \mathbf{P} \left(\sup_{0 \leq t \leq 1} \left| \frac{1}{n} S_{nt} - t \hat{\mathbf{E}}[X] \right| > \varepsilon \mid Z_n > 0 \right) = 0.$$

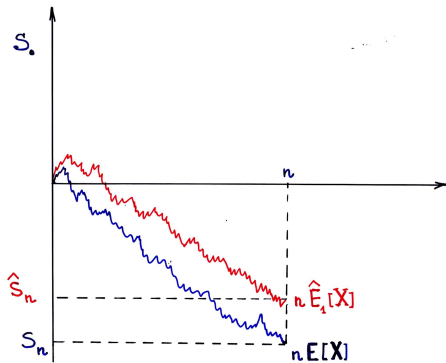
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and for $0 < t_1 < \dots < t_k < 1$

$$\mathbf{P}((Z_{nt_1}, \dots, Z_{nt_k}) \mid Z_n > 0) \xrightarrow{d} (W_1, \dots, W_k)$$

where W_1, W_2, \dots are **I.I.D.** copies of some strictly positive integer-valued random variable W .



THANKS!