

# Sharp Tail Bounds for Functionals of Markov Chains

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- 3 Probabilistic study based on renewal theory
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- 4 The regenerative method
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  - Tail bounds through the regenerative approach

# Overall purpose

- **(Bennett, '62)** Let  $Y_1, \dots, Y_n$  be i.i.d. r.v.'s such that  $\mathbb{E}[Y_i] = 0$ ,  $\mathbb{E}[Y_i^2] = \sigma^2 < \infty$  and  $|Y_i| \leq M$  a.s. .

$$\mathbb{P}\left\{\sum_{i=1}^n Y_i \geq x\right\} \leq \exp\left(-\frac{n\sigma^2}{M^2} H\left(\frac{Mx}{n\sigma^2}\right)\right),$$

where  $H(x) = (1+x)\log(1+x) - x$ .

- **(Fuk & Nagaev, '71)** Relaxing the boundedness constraint:  
 $\forall M, x > 0$ ,

$$\mathbb{P}\left\{\sum_{i=1}^n Y_i \geq x\right\} \leq \exp\left(-\frac{n\sigma_M^2}{M^2} H\left(\frac{Mx}{n\sigma_M^2}\right)\right) + n\mathbb{P}\{Y_1 > M\},$$

with  $\sigma_M^2 = \mathbb{E}[Y_i^2 \mathbb{I}\{|Y_i| \leq M\}]$ .

- **Our goal:** extension to instantaneous functionals of a Markov chain, i.e.  $Y_i = f(X_i)$ , by means of the **regenerative method**.

- **The Markov property:**  $X = (X_n)_{n \in \mathbb{N}}$  is a chain with state space  $(E, \mathcal{E})$ , trans. prob.  $\Pi(x, dy)$  and initial distr.  $\nu$  iff  $X_0 \sim \nu$  and

$$\mathbb{P}(X_{n+1} \in dx \mid X_0, \dots, X_n) = \Pi(X_n, dx).$$

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- **Strong version:**

- "Shift operator"  $\theta$ :  $X_{n+1} = X_n \circ \theta$ ,
- $\mathcal{F}_n = \sigma(X_0, \dots, X_n)$ ,  $\tau$  stopp. time,  $H = H(x_1, x_2, \dots)$  bounded,

$$\mathbb{E}_\nu[H \circ \theta^\tau \mid \mathcal{F}_\tau] = \mathbb{E}_{X_\tau}[H] \text{ on } \{\tau < \infty\}.$$

- **Communication/ Stochastic stability**

- **Irreducibility:**  $\Phi / \forall B \in \mathcal{E},$

$$\Phi(B) > 0 \Rightarrow \forall x, \sum_{n \geq 1} \Pi_n(x, B) > 0,$$

$\Rightarrow$  existence of a maximal irreducibility measure  $\Psi$ .

- **Periodicity:**  $X$   $\Psi$ -irreduc.,  $\exists d' \geq 1, D_1, \dots, D_{d'}, \Psi(D_i) > 0,$   
 $D_i \cap D_j = \emptyset$  if  $i \neq j$  and

- 1  $\Pi(x, D_{i+1}) = 1, x \in D_i,$

- 2  $\Psi(\cup\{D_i\}^c) = 0,$

$\Rightarrow$  period =  $GCD\{d' \geq 1 / 1. \text{ and } 2.\}$ , aperiodic when = 1.

- **Positive recurrence:**  $\exists!$  prob.  $\mu / \mu(dy) = \int_{x \in E} \mu(dx) \Pi(x, dy)$  and

" $\Pi_n(x, dy) = \mathbb{P}(X_n \in dy \mid X_0 = x) \rightarrow \mu(dy)$ " as  $n \rightarrow \infty$ .

# Regenerative Markov chains

- A set  $A \in \mathcal{E}$  is an **accessible atom** if  $\Psi(A) > 0$  and

$$\forall (x, y) \in A^2, \quad \Pi(x, \cdot) = \Pi(y, \cdot).$$

- "Dividing sample paths into regeneration cycles"**

Suppose  $A$  is Harris recurrent. Let  $\mathbb{E}_A[\cdot] = \mathbb{E}[\cdot \mid X_0 \in A]$  and

$$\tau_A(1) = \tau_A = \inf\{k \geq 1 : X_k \in A\},$$

$$\tau_A(j) = \inf\{k \geq 1 + \tau_A(j-1) : X_k \in A\}, \quad j > 1.$$

Strong Markov  $\Rightarrow \{\tau_A(j)\}_{j \geq 1}$  is a (possibly delayed) **renewal process**

$$\dots, \underbrace{X_{1+\tau_A(1)}, \dots, X_{\tau_A(2)}}_{\text{regenerative segment}}, \dots, \underbrace{X_{1+\tau_A(j)}, \dots, X_{\tau_A(j+1)}}_{\text{regenerative segment}}, \dots$$

i.i.d. "regenerative segments" of random size

# General Harris Markov chains - Pseudo-renewal



- A meas. set  $S \subset E$  is **small** if  $\exists m \geq 1, \delta > 0, \Phi$  proba. s.t.  $\Phi(S) = 1$

$$\forall x \in E, \Pi_m(x, dy) \geq \delta \mathbb{I}_{\{x \in S\}} \Phi(dy).$$

- **The Nummelin technique** ( $m = 1$ ). Construct a bivar. chain  $(X, Y)$  with state sp.  $E \times \{0, 1\}$  by **randomizing each time  $X_n$  hits  $S$** :
  - if  $x \notin S$ ,  $\Pi^*((x, y), B \times \{1\}) = \delta \Pi(x, B)$ ,
  - if  $x \in S$ , then

$$\Pi^*((x, 0), B \times \{1\}) = \delta \frac{\Pi(x, B) - \delta \Phi(B)}{1 - \delta},$$

$$\Pi^*((x, 1), B \times \{1\}) = \delta \Phi(B),$$

$\Rightarrow S \times \{1\}$  is an atom for the *split chain*.

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## Theorem (Kac's theorem)

*The chain  $X$  is positive recurrent iff  $\alpha = \mathbb{E}_A[\tau_A] < \infty$ .*

*If  $X$  is positive recurrent, then  $\mu$  is the Pitman's **occupation measure**:*

$$\mu(B) = \frac{1}{\mathbb{E}_A[\tau_A]} \mathbb{E}_A \left[ \sum_{i=1}^{\tau_A} \mathbb{I}_{\{X_i \in B\}} \right].$$

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- Let  $f : E \rightarrow \mathbb{R}$   $\mu$ -integrable

$$\mu(f) = \int_{x \in E} f(x) \mu(dx) = \frac{1}{\mathbb{E}_A[\tau_A]} \mathbb{E}_A \left[ \sum_{i=1}^{\tau_A} f(X_i) \right].$$

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$$\sum_{i=1}^n f(X_i) = \sum_{i=1}^{\tau_A} f(X_i) + \sum_{j=1}^{l_n-1} S_j(f) + \sum_{i=1+\tau_A(l_n)}^n f(X_i),$$

with  $S_j(f) = \sum_{i=1+\tau_A(j)}^{\tau_A(j+1)} f(X_i)$ . The  $S_j(f)$ 's are **i.i.d. r.v.'s**:

$$\Rightarrow \hat{\mu}_n = \frac{1}{n} \sum_{i=1}^n f(X_i) \rightarrow \mu(f) = \alpha^{-1} \mathbb{E}[S_1(f)] \text{ as } n \rightarrow \infty.$$

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- **Central Limit Theorem** - Suppose that  $\mathbb{E}_A[(\sum_{i=1}^{\tau_A} f(X_i))^2] < \infty$ ,

$$\sqrt{n}(\hat{\mu}_n - \mu(f)) \Rightarrow \mathcal{N}(0, \sigma_f^2) \text{ as } n \rightarrow \infty,$$

with  $\sigma_f^2 = \alpha^{-1} \mathbb{E}_A[(\sum_{i=1}^{\tau_A} \{f(X_i) - \mu(f)\})^2] \neq \text{Var}_\mu[f(X)]$ .



# Non asymptotic bounds and second order results

- Given a sample path of length  $n$ , form  $l_n + 1$  blocks:

$$\mathcal{B}_0 = (X_1, \dots, X_{\tau_A}), \mathcal{B}_1 = (X_{1+\tau_A(1)}, \dots, X_{\tau_A(2)}), \dots$$

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  - Partition** the space according to all possible ways for  $X$  to regenerate

$$U_{r,l,m} = \left\{ \tau_A = r, \sum_{j=1}^{m-1} s_j = n - r - l, \tau_A(m+1) - \tau_A(m) > l \right\}.$$

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- Sum up** the results obtained so as to **identify** the global bound/limit.

## Block-moment conditions

- (i)  $\alpha = \mathbb{E}_A[\tau_A] < \infty$  and  $0 < \sigma_A^2 = \mathbb{E}_A[\tau_A^2] < \infty$ ,
- (ii)  $\mathbb{E}_A[S_A(f)] = 0$  and  $0 < \sigma_f^2 = \mathbb{E}_A[S_A(f)^2] < \infty$ .

## Notations

$$\tilde{\sigma}_f^2 = \text{Var}_A(S_A(f)\mathbb{I}_{\{|S_A(f)| \leq M_1\}}),$$

$$\tilde{\sigma}_A^2 = \text{Var}_A(\tau_A\mathbb{I}_{\{\tau_A \leq M_2\}}),$$

$$\tilde{\rho} = (\tilde{\sigma}_f \tilde{\sigma}_A)^{-1} \text{Cov}_A(S_A(f)\mathbb{I}_{\{|S_A(f)| \leq M_1\}}, \tau_A\mathbb{I}_{\{\tau_A \leq M_2\}}),$$

$$\tilde{\sigma} = \sqrt{\tilde{\sigma}_f^2 \tilde{\sigma}_A^2 / (\tilde{\sigma}_f^2 + \tilde{\sigma}_A^2)},$$

## Theorem (Bertail & Cléménçon, 2007)

Under the previous assumptions, for any vector  $M = (M_1, M_2) \in \mathbb{R}_+^{*2}$  of euclidian norm  $\|M\| = (M_1^2 + M_2^2)^{1/2}$ , there exists a constant  $C_M$  depending on  $\sigma_A^2$  and  $\sigma_f^2$  such that , for any  $n \geq \alpha$ ,

$$\mathbb{P}_\nu\left(\sum_{j=1}^{I_n-1} S_j(f) \geq x\right) \leq C_M \exp\left\{-\frac{n(1+|\tilde{\rho}|)\tilde{\sigma}^2}{2\|M\|^2} H\left(\frac{\|M\|\sqrt{2}}{(1+|\tilde{\rho}|)\tilde{\sigma}\tilde{\sigma}_f} x\right)\right\} \\ + (n-1)\mathbb{P}_A(|S_A(f)| \geq M_1) + (n-1)\mathbb{P}_A(\tau_A \geq M_2),$$

- For  $A = E$ ,  $f$  bounded and  $M$  properly chosen, one recovers Bennett's inequality (1962).

# Sketch of Proof - Step 1 "Truncation"

- Let

$$\tilde{S}_j(f) = S_j(f)\mathbb{I}\{|S_j(f)| \leq M_1\} \text{ and } \tilde{s}_j(f) = s_j(f)\mathbb{I}\{|s_j| \leq M_2\}$$

- From the union bound, get:

$$\begin{aligned} \mathbb{P}_\nu\left(\sum_{j=1}^{l_n-1} S_j(f) \geq x\right) &\leq \mathbb{P}_\nu\left(\left\{\sum_{j=1}^{l_n-1} \tilde{S}_j(f) \geq x\right\} \cap \{s_j = \tilde{s}_j, 1 \leq j \leq l_n - 1\}\right) \\ &\quad + (n-1)\mathbb{P}_A(\{|S_A(f)| > M_1\} \cup \{\tau_A > M_2\}) \end{aligned}$$

- Set

$$P_{\nu,n}(x) = \mathbb{P}_\nu\left(\left\{\sum_{j=1}^{l_n-1} \tilde{S}_j(f) \geq x\right\} \cap \{s_j = \tilde{s}_j, 1 \leq j \leq l_n - 1\}\right)$$



## Step 2 "Partitioning"

- consider the collection of events: for  $1 \leq r$ ,  $l \leq n$  and  $2 \leq m \leq n$ ,

$$U_{r,l,m} = \left\{ \tau_A = r, \sum_{j=1}^m s_j = n - r - l, s_{m+1} > l \right\},$$

- By the formula of total probability and the strong Markov property:

$$P_{\nu,n}(x) = \sum_{m=2}^n \sum_{r=1}^n \sum_{l=1}^n \mathbb{P}_{\nu}(\tau_A = r) \times p_{r,l,m}(x) \times \mathbb{P}_A(\tau_A > l),$$

where  $\tilde{\alpha} = \mathbb{E}[\tilde{s}_j]$

$$p_{r,l,m}(x) = \mathbb{P}\left( \sum_{1 \leq j \leq m} \tilde{S}_j(f) \geq x, \sum_{1 \leq j \leq m} \{\tilde{s}_j - \mathbb{E}[\tilde{s}_j]\} = n - r - l - m\tilde{\alpha} \right)$$

## Step 2 "Partitioning"

- Set  $\lambda_{r,l,m} = (n - r - l - \tilde{\alpha}m)/(\tilde{\sigma}_A\sqrt{m})$  and denote by  $\tilde{p}_{r,l,m}(x)$  the probability

$$\mathbb{P}\left(\frac{\sum_{j=1}^m\{\tilde{S}_j(f) - \mathbb{E}[\tilde{S}_j(f)]\}}{\tilde{\sigma}_f\sqrt{m}} \geq \frac{x}{\tilde{\sigma}_f\sqrt{m}}, \frac{\sum_{j=1}^m\{\tilde{s}_j - \mathbb{E}[\tilde{s}_j]\}}{\tilde{\sigma}_A\sqrt{m}} = \lambda_{r,l,m}\right)$$

- Since  $\mathbb{E}[\tilde{S}_j(f)] \leq 0$ , we have  $p_{r,l,m}(x) \leq \tilde{p}_{r,l,m}(x)$
- From limit theorems for 1-lattice random vectors (Dubinskaite, '82):

$$\tilde{p}_{r,l,m}(x) \sim m^{-1/2}h \int_{t=x}^{\infty} \phi_{\Sigma}(t/\sqrt{m}, \lambda_{r,l,m})dt$$

with  $\Sigma = \text{Var}((\tilde{\sigma}_f^{-1}\tilde{S}_j(f), \tilde{\sigma}_A^{-1}\tilde{s}_j))$

- Set  $\tilde{\rho} = (\tilde{\sigma}_f \tilde{\sigma}_A)^{-1} \text{Cov}(\tilde{S}_j(f), \tilde{s}_j)$

## Step 3 Exponential inequality for 1-lattice bounded r.v.'s

### Lemma

Let  $(S_j^*, L_j^*)_{1 \leq j \leq m}$  be i.i.d. centered and square integrable bivariate r.v.'s with covariance  $\Sigma = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$ . Assume that the  $L_j^*$ 's are lattice r.v.'s with minimal span  $h > 0$  and there exists finite constants  $B_1$  and  $B_2$  such that  $|S_j^*| \leq B_1$  and  $|L_j^*| \leq B_2$  for  $j = 1, \dots, m$ . Set  $B^2 = B_1^2 + B_2^2$ , then there exists a universal constant  $c$  s.t. for all  $m \geq 1$  and  $y \geq 0$ ,

$$\mathbb{P}\left(m^{-\frac{1}{2}} \sum_{j=1}^m S_j^* \geq y, m^{-\frac{1}{2}} \sum_{j=1}^m L_j^* = \lambda\right) \leq e^{\frac{3}{2}} \left(\frac{h}{2\pi\sqrt{m}} + 4c\frac{B}{m}\right) \exp\left\{-\frac{y^2 + \lambda^2}{4B^2}\right\}$$

## Step 3 - Application: control of $\tilde{p}_{r,l,m}(x)$

- In this case, the  $L_j^{*}$ 's are lattice with minimal span  $h = \tilde{\sigma}_A^{-1}$

$$\begin{aligned} \tilde{p}_{r,l,m}(x) &\leq e^{\frac{3}{2}} \left( \frac{1}{2\pi\tilde{\sigma}_A\sqrt{m}} + 8c \frac{\|M\|}{\tilde{\sigma}_m} \right) \\ &\times e^{-\frac{m(1+|\tilde{\rho}|)\tilde{\sigma}^2}{2\|M\|^2} \left\{ H\left(\frac{\|M\|_x\sqrt{2}}{(1+|\tilde{\rho}|)\tilde{\sigma}_f\tilde{\sigma}_m}\right) + H\left(\frac{\|M\|\lambda_{r,l,m}\sqrt{2}}{(1+|\tilde{\rho}|)\tilde{\sigma}\sqrt{m}}\right) \right\}}. \end{aligned}$$

## Step 4 - Control of the sum

- We deduce that

$$P_{n,\nu}(x) \leq e^{3/2} \exp \left\{ -\frac{n(1 + |\tilde{\rho}|)\tilde{\sigma}^2}{2\|M\|^2} H \left( \frac{\|M\|_x \sqrt{2}}{(1 + |\tilde{\rho}|)\tilde{\sigma}_f \tilde{\sigma}_n} \right) \right\} \Gamma_n, \quad (1)$$

where

$$\Gamma_n = \sum_{m=2}^n \sum_{r=1}^n \sum_{l=1}^n \mathbb{P}_\nu(\tau_A = r) \mathbb{P}_A(\tau_A > l) \left( \frac{c_1}{\sqrt{m}} + \frac{c_2}{m} \right) \gamma_{r,l,m},$$

with

$$\gamma_{r,l,m} \leq \exp \left\{ -\frac{\lambda_{r,l,m}^2}{2(1 + |\tilde{\rho}| + \frac{|\lambda_{r,l,m}| \|M\| \sqrt{2}}{3\tilde{\sigma}\sqrt{m}})} \right\}.$$

## Step 4 - Control of the sum

- Recall that  $\lambda_{r,l,m} = (n - r - l - \tilde{\alpha}m)/\tilde{\sigma}_A\sqrt{m}$  and consider the subdivision

$$a_{n,m} = (n - \tilde{\alpha}m)/(\tilde{\sigma}_A\sqrt{m}), \text{ for } 1 \leq m \leq n.$$

- Split the sum into two terms

$$\begin{aligned} \sum_{r,l} \sum_m &= \sum_{r+l > \tilde{\sigma}_A} \sum_m + \sum_{r+l \leq \tilde{\sigma}_A} \sum_m \\ &\leq \text{constant} + \text{Riemann sum} \end{aligned}$$

- etc.*

# Proof of the Lemma (Sketch of)

- Talagrand '95 "The missing factor in Hoeffding's inequalities"
- Exponential change of probability measure:  $\mathcal{S}_m = m^{-1/2} \sum_{i \leq m} S_j^*$ ,  $\mathcal{L}_m = m^{-1/2} \sum_{i \leq m} L_j^*$  and  $\psi_m(u) = \log \mathbb{E}[\exp\{\langle u, (\mathcal{S}_m, \mathcal{L}_m) \rangle\}]$

$$d\mathbb{P}_{u,m} = \exp\{\langle u, (\mathcal{S}_m, \mathcal{L}_m) \rangle - \psi_m(u)\} d\mathbb{P}$$

- One has

$$\begin{aligned} \mathbb{P}(\mathcal{S}_m \geq y, \mathcal{L}_m = \lambda) &= e^{-\langle u, (y, \lambda) \rangle - \psi_m(u)} \\ &\times \mathbb{E}_{u,m}[e^{\langle u, (\mathcal{S}_m - y, \mathcal{L}_m - \lambda) \rangle} \mathbb{I}_{\{\mathcal{S}_m \geq y, \mathcal{L}_m = \lambda\}}] \end{aligned}$$

- Choose  $u = u^*$  such that  $\nabla \psi_m(u^*) = (y, \lambda)$ :

$$u^* = \arg \sup_{u \in \mathbb{R}_+ \times \mathbb{R}} \{\langle u, (y, \lambda) \rangle - \psi_m(u)\}$$

# Proof of the Lemma (Sketch of)

- The quantity  $\mathbb{E}_{u,m}[e^{-\langle u, (S_m - y, \mathcal{L}_m - \lambda) \rangle} \mathbb{I}_{\{S_m \geq y, \mathcal{L}_m = \lambda\}}]$  is bounded by

$$u_1 \int_{s=0}^{\infty} e^{-u_1 s} \mathbb{P}_{u,m}(S_m - y \geq 0, L_m - \lambda = 0) - \mathbb{P}_{u,m}(S_m - y \geq s, L_m - \lambda = 0) ds$$

- Use a local Berry-Esseen bound (Bolthausen '80)

$$|\mathbb{P}_{u,m}(S_m - y \geq s, \mathcal{L}_m - \lambda = 0) - \frac{h}{\sqrt{m}} \int_s^{\infty} \phi_{W_u}(t, 0) dt| \leq C_u m^{-1},$$

where  $W_u = \text{Var}_{u,m}[(S_m, \mathcal{L}_m)]$  and  $\phi_{W_u}$  is the density of the bivariate gaussian distribution with covariance matrix  $W_u$