

# Limit theorems for regularly varying functions of Markov chains

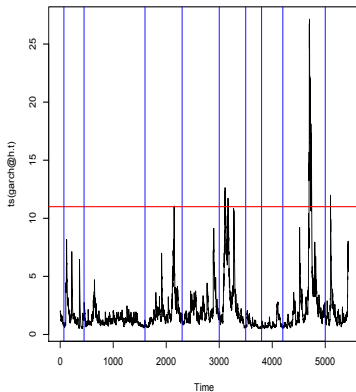
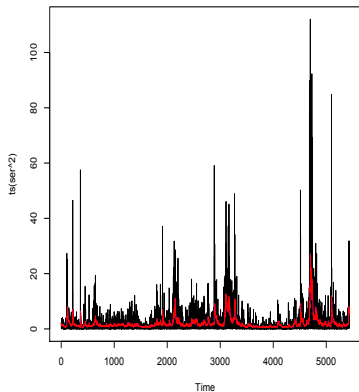
In collaboration with T. Mikosch

Olivier Wintenberger  
olivier.wintenberger@upmc.fr

LSTA, University Pierre et Marie Curie.

**WQMIF, Zagreb, June 6, 2014**

# Illustrations



# Motivation: characterization of limit theorems

## lid case

The  $\alpha$ -stable limit in CLT, the large deviations of the partial sums, the ruin probabilities... are characterized by the regular variations of the margins.

## Clusters of extremes

For regularly varying processes, there are clusters of extremes.  
How do the clusters modify the limit characteristics?

## 1 Markov chains

- Regular variation, splitting scheme and drift condition
- Regular variation of cycles

## 2 Limit theorems for functions of Markov chains

- Central Limit Theorem
- Large deviations and ruin probabilities

## 3 Markov chains with extremal linear behavior

## 1 Markov chains

- Regular variation, splitting scheme and drift condition
- Regular variation of cycles

## 2 Limit theorems for functions of Markov chains

- Central Limit Theorem
- Large deviations and ruin probabilities

## 3 Markov chains with extremal linear behavior

# Regularly varying sequences

## Regularly varying condition of order $\alpha > 0$

A stationary sequence  $(X_t)$  is regularly varying if a non-null Radon measure  $\mu_d$  is such that

$$(RV_\alpha) \quad n \mathbb{P}(a_n^{-1}(X_1, \dots, X_d) \in \cdot) \xrightarrow{v} \mu_d(\cdot),$$

where  $(a_n)$  satisfies  $n \mathbb{P}(|X| > a_n) \rightarrow 1$  and  $\mu_d(tA) = t^{-\alpha} \mu_d(A)$ ,  $t > 0$ .

## Definition (Basrak & Segers, 2009)

It is equivalent to the existence of the **spectral tail process**  $(\Theta_t)$  defined for  $k \geq 0$ ,

$$\mathbb{P}(|X_0|^{-1}(X_0, \dots, X_k) \in \cdot \mid |X_0| > x) \xrightarrow{w} \mathbb{P}((\Theta_0, \dots, \Theta_k) \in \cdot), \quad x \rightarrow \infty.$$

# Regeneration of Markov chains with an accessible atom (Doebelin, 1939)

## Definition

$(\Phi_t)$  is a Markov chain of kernel  $P$  on  $\mathbb{R}^d$  and  $A \in \mathcal{B}(\mathbb{R}^d)$ .

- $A$  is an atom if  $\exists$  a measure  $\nu$  on  $\mathcal{B}(\mathbb{R}^d)$  st  $P(x, B) = \nu(B)$  for all  $x \in A$ .
- $A$  is accessible, i.e.  $\sum_k P^k(x, A) > 0$  for all  $x \in \mathbb{R}^d$ ,

Let  $(\tau_A(j))_{j \geq 1}$  visiting times to the set  $A$ , i.e.

$\tau_A(1) = \tau_A = \min\{k > 0 : X_k \in A\}$  and  $\tau_A(j+1) = \min\{k > \tau_A(j) : X_k \in A\}$ .

## Regeneration cycles

- 1  $N_A(t) = \#\{j \geq 1 : \tau_A(j) \leq t\}$ ,  $t \geq 0$ , is a renewal process,
- 2 The cycles  $(\Phi_{\tau_A(t)+1}, \dots, \Phi_{\tau_A(t+1)})$  are iid.

# Irreducible Markov chain and Nummelin scheme

Definition (Minorization condition, Meyn and Tweedie, 1993)

$\exists \delta > 0$ ,  $C \in \mathcal{B}(\mathbb{R}^d)$  and a distribution  $\nu$  on  $C$  such that

$$(MC_k) \quad P^k(x, B) \geq \delta \nu(B), \quad x \in C, \quad B \in \mathcal{B}(\mathbb{R}^d).$$

$(MC_1)$  is called the strongly aperiodic case.

If  $P$  is an irreducible aperiodic Markov chain then it satisfies  $(MC_k)$  for some  $k \in \mathbb{N}$ .

Nummelin splitting scheme

Under  $(MC_1)$  an enlargement of  $(\Phi_t)$  on  $\mathbb{R}^d \times \{0, 1\} \subset \mathbb{R}^{d+1}$  possesses an accessible atom  $A = C \times \{1\} \implies$  **the enlarged Markov chain regenerates.**



# Main assumptions

Assume that  $(\Phi_t)$  (possibly enlarged) possesses an accessible atom  $A$ , the existence of its invariant measure  $\pi$  and  $\Phi_0 \sim \pi$ .

Assume the existence of  $f$  such that:

- 1 There exist constants  $\beta \in (0, 1)$ ,  $b > 0$  such that for any  $y$ ,

$$(DC_p) \quad \mathbb{E}(|f(\Phi_1)|^p \mid \Phi_0 = y) \leq \beta |f(y)|^p + b 1_A(y).$$

- 2  $(X_t = f(\Phi_t))$  satisfies  $(RV_\alpha)$  with index  $\alpha > 0$  and spectral tail process  $(\Theta_t)$ .

## Remarks

- 1 it is absolutely  $(\beta-)$ mixing with exponential rate,
- 2  $\sup_{x \in A} \mathbb{E}_x(\kappa^{\tau_A})$  for some  $\kappa > 1$ .
- 3  $(DC_p) \implies (DC_{p'})$  for  $0 < p' \leq p$ .

# The cluster index

Under  $(RV_\alpha)$  denote  $b_k(\pm) = \lim_{n \rightarrow \infty} n \mathbb{P}(\pm S_k > a_n)$ ,  $k \geq 1$ .

## Theorem (4)

Assume  $(\mathbf{RV}_\alpha)$  for some  $\alpha > 0$  and  $(\mathbf{DC}_p)$  for positive  $p \in (\alpha - 1, \alpha)$ . Then the limits (called cluster indices)

$$\begin{aligned} b_\pm : &= \lim_{k \rightarrow \infty} (b_{k+1}(\pm) - b_k(\pm)) \\ &= \lim_{k \rightarrow \infty} \mathbb{E} \left[ \left( \sum_{t=0}^k \Theta_t \right)_\pm^\alpha - \left( \sum_{t=1}^k \Theta_t \right)_\pm^\alpha \right] \\ &= \mathbb{E} \left[ \left( \sum_{t=0}^{\infty} \Theta_t \right)_\pm^\alpha - \left( \sum_{t=1}^{\infty} \Theta_t \right)_\pm^\alpha \right] \end{aligned}$$

exist and are finite. Here  $(\Theta_t)$  is the spectral tail process of  $(X_t)$ .

## Other indices

The extremal index  $0 < \theta = \mathbb{E} \left[ \left( \sup_{t \geq 0} \Theta_t \right)_+^\alpha - \left( \sup_{t \geq 1} \Theta_t \right)_+^\alpha \right] \leq \mathbb{E}[(\Theta_0)_+^\alpha]$ .

Under  $(RV_\alpha)$  denote  $\tilde{b}_k = \lim_{n \rightarrow \infty} n \mathbb{P}(\sup_{t \geq k} S_t > a_n)$ ,  $k \geq 1$ .

### Theorem (Under the hypothesis of the Theorem 4)

*The limit (called cluster index)*

$$\begin{aligned} \tilde{b} &:= \lim_{k \rightarrow \infty} (\tilde{b}_{k+1} - \tilde{b}_k) \\ &= \mathbb{E} \left[ \left( \sup_{k \geq 0} \sum_{t=0}^k \Theta_t \right)_\pm^\alpha - \left( \sup_{k \geq 1} \sum_{t=1}^k \Theta_t \right)_\pm^\alpha \right] \end{aligned}$$

*exist and is finite. Here  $(\Theta_t)$  is the spectral tail process of  $(X_t)$ .*

# Regular variation of cycles

## Theorem (Under the hypothesis of the Theorem 4)

Assume  $(RV_\alpha)$  with  $\alpha > 0$  and  $(DC_p)$  with  $(\alpha - 1)_+ < p < \alpha$  and  $b_\pm \neq 0$  then

$$\mathbb{P}_A\left(\sup_{1 \leq i \leq \tau_A} f(\Phi_i) > x\right) \sim_{x \rightarrow \infty} \theta \mathbb{E}_A(\tau_A) \mathbb{P}(|X| > x),$$

$$\mathbb{P}_A\left(\pm \sum_{i=1}^{\tau_A} f(\Phi_i) > x\right) \sim_{x \rightarrow \infty} b_\pm \mathbb{E}_A(\tau_A) \mathbb{P}(|X| > x),$$

$$\mathbb{P}_A\left(\sup_{1 \leq i \leq \tau_A} \sum_{j=1}^i f(\Phi_j) > x\right) \sim_{x \rightarrow \infty} \tilde{b} \mathbb{E}_A(\tau_A) \mathbb{P}(|X| > x),$$

## Remarks

- 1 We always have  $\mathbb{E}_A(\tau_A) \mathbb{P}(X > x) = \mathbb{E}_A[\sum_{i=1}^{\tau_A} \mathbf{1}_{f(\Phi_i) > x}]$ ,
- 2 If  $\tau_A$  was independent of  $(X_t)$  then  $\mathbb{P}_A(S_A(1) > x) \sim_{x \rightarrow \infty} \mathbb{E}_A(\tau_A) \mathbb{P}(X > x)$ .

## 1 Markov chains

- Regular variation, splitting scheme and drift condition
- Regular variation of cycles

## 2 Limit theorems for functions of Markov chains

- Central Limit Theorem
- Large deviations and ruin probabilities

## 3 Markov chains with extremal linear behavior

# Stable Central Limit Theorem

## Theorem (Under the hypothesis of the Theorem 4)

If  $0 < \alpha < 2$ ,  $\alpha \neq 1$  and  $X$  is centered if  $1 < \alpha < 2$  then  $a_n^{-1}S_n \xrightarrow{d} \xi_\alpha$ , with the characteristic function  $\xi_\alpha$  given by  $\exp(-|x|^\alpha \chi_\alpha(x, b_+, b_-))$ , where

$$\chi_\alpha(x, b_+, b_-) = \frac{\Gamma(2 - \alpha)}{1 - \alpha} \left( (b_+ + b_-) \cos\left(\frac{\pi\alpha}{2}\right) - i \operatorname{sgn}(x)(b_+ - b_-) \sin\left(\frac{\pi\alpha}{2}\right) \right).$$

# Precise large deviations and ruin probabilities

## Theorem (Under the hypothesis of the Theorem 4)

If  $0 < \alpha < 1$  then  $\lim_{n \rightarrow \infty} \sup_{x \geq b_n} \left| \frac{\mathbb{P}(\pm S_n > x)}{n \mathbb{P}(|X| > x)} - b_{\pm} \right| = 0$ . If  $\alpha > 1$  and  $X$  is centered then  $\lim_{n \rightarrow \infty} \sup_{b_n \leq x \leq c_n} \left| \frac{\mathbb{P}(\pm S_n > x)}{n \mathbb{P}(|X| > x)} - b_{\pm} \right| = 0$  with  $\sqrt{n} = o(b_n)$  if  $\alpha > 2$ ,  $n^{1/\alpha} L(n) = o(b_n)$  otherwise and  $\mathbb{P}(\tau_A > n) = o(n \mathbb{P}(|X| > c_n))$ .

## Theorem (Under the hypothesis of the Theorem 4)

Assume that  $(X_t)$  is regularly varying with index  $\alpha > 1$  and centered. Then we have for any  $\rho > 0$ ,

$$\mathbb{P}\left(\sup_{t \geq 0} (S_t - \rho t) > x\right) \sim \frac{\tilde{b}_x \mathbb{P}(|X| > x)}{(\alpha - 1)\rho}, \quad x \rightarrow \infty.$$

Consequence: if  $b_+ \neq \tilde{b}$  the functional CLT cannot hold.

## 1 Markov chains

- Regular variation, splitting scheme and drift condition
- Regular variation of cycles

## 2 Limit theorems for functions of Markov chains

- Central Limit Theorem
- Large deviations and ruin probabilities

## 3 Markov chains with extremal linear behavior



# The AR(1) model

## Definition (AR(1) model)

The AR(1) model is the solution of  $X_t = \phi X_{t-1} + Z_t$ ,  $|\phi| < 1$  with  $(Z_t)$  is an iid regularly varying sequence of order  $\alpha > 0$ .

## Proposition

We have  $(X_t) \in RV_\alpha$  and the conclusions of the theorems hold with  $\Theta_t = \phi^t$ ,  $t > 0$ .

# Markov chains with extremal linear behavior (Kesten, 1974, Goldie, 1991, Segers, 2007, Mirek, 2011)

Assume  $(A, B)$  is absolutely continuous on  $\mathbb{R}^+ \times \mathbb{R}$  with  $\mathbb{E}A^\alpha = 1$ ,  $X_t = \Psi_t(X_{t-1})$  with iid iterated Lipschitz functions  $\Psi_t$  with negative top Lyapunov exponent and

$$A_t X_{t-1} - |B_t| \leq X_t \leq A_t X_{t-1} + |B_t|.$$

## Proposition

We have  $(X_t) \in RV_\alpha$  and the conclusions of the theorems hold with

$$\Theta_t = \prod_{i=1}^t A_i, \quad t \geq 0.$$

# The GARCH(1,1) model

## Definition (Bollerslev, 1986)

The GARCH(1,1) model  $(X_t)$  is the solution of  $X_t = \sigma_t Z_t$ ,  $t \in \mathbb{Z}$  with  $(Z_t)$  is an iid mean zero and unit variance sequence of random variables and  $(\sigma_t^2)$  satisfies the stochastic recurrence equation

$$\sigma_t^2 = \alpha_0 + (\alpha_1 Z_{t-1}^2 + \beta_1) \sigma_{t-1}^2, \quad t \in \mathbb{Z}.$$

## Proposition

If  $X_0 \in RV_\alpha$  then we have

$$\mathbb{P}(|X_0|^{-1}(X_0, \dots, X_t) \in \cdot \mid |X_0| > x) \rightarrow \frac{1}{\mathbb{E}|Z_0|^\alpha} \mathbb{E} \left[ |Z_0|^\alpha \mathbf{1}_{(Z_0, Z_1 \Pi_1^{0.5}, \dots, Z_t \Pi_t^{0.5}) \in |Z_0| \cdot} \right],$$

where  $\Pi_t = A_1 \cdots A_t$  with  $A_t = \alpha_1 Z_{t-1}^2 + \beta_1$ .

# Conclusion

- Cluster indices  $b_{\pm}$ ,  $\tilde{b}$  together with  $\theta$  determine the limit theorems of dependent and regularly varying variables,
- Under the hypothesis of the theorems

$$\mathbb{P}(S_n > x) \sim_{n \rightarrow \infty} b_+ n \mathbb{P}(X > x) \quad \text{for } b_n \leq x \leq c_n$$

with  $b_+/\theta \gg 1$ . Consequences in risk management....

- Inference of the cluster indices  $b_{\pm}$ ,  $\tilde{b}$ ...

Thank you for your attention!