

# Extreme-value theory and the stochastic exit problem

7<sup>th</sup> Workshop on Random Dynamical Systems, Bielefeld, 12/12/14

Refs: NB, B Gentz, SIAM J Math Anal 46: 310-352 (2014) [arXiv: 1208.2557]  
 NB [arXiv: 1403.7393]

## 1. Extreme-value theory

$X_1, X_2, \dots$   $\mathbb{R}$ -valued i.i.d. r.v.

$$S_n = \sum_{i=1}^n X_i$$

limit thm:  $\lim_{n \rightarrow \infty} \frac{S_n - b_n}{a_n} \stackrel{d}{=} Y$   
 $(a_n > 0, b_n \in \mathbb{R})$

ex: CLT:  $X_i \in L^2$   $\begin{cases} b_n = n \mathbb{E}(X_i) \\ a_n = \sqrt{n \text{Var}(X_i)} \\ Y \sim N(0, 1) \end{cases}$

Cauchy:  $\frac{S_n}{n} \stackrel{d}{=} X_1$

$\frac{\sum_{i=1}^n Y_i - b_n}{a_n} \stackrel{d}{=} Y \Rightarrow Y \text{ stable}$   
 ex:  $\mathbb{E}(e^{itY}) = e^{-|t|^\alpha}$

Thm: [Fréchet '27, Fisher & Tippett '28, Gnedenko '43]

$F(t) \neq 1_{\{t \geq c\}}$ ,  $F \in D(\Phi)$

$\Rightarrow \Phi \in \{\Phi_\alpha, \Psi_\alpha, \Lambda\}$  where  $\begin{cases} \Phi_\alpha(t) = e^{-t^{-\alpha}} 1_{\{t > 0\}} & (\alpha > 0) \\ \Psi_\alpha(t) = e^{-(t-\mu)^\alpha} 1_{\{t \leq 0\}} + 1_{\{t > 0\}} & \text{Weibull} \\ \Lambda(t) = e^{-e^{-t}} & \text{Gumbel} \end{cases}$

Let  $R(t) := 1 - F(t) = \mathbb{P}\{X_1 > t\}$

Lemma:  $F \in D(\Phi) \Leftrightarrow \exists a_n > 0, b_n: \lim_{n \rightarrow \infty} n R(a_n t + b_n) = -\log \Phi(t)$   
 $\forall t: \Phi(t) > 0$

Gumbel: possible choice  $b_n = \inf \{t: F(t) > 1 - \frac{1}{n}\}$   $a_n = \inf \{t: F(t) + b_n > 1 - \frac{1}{n}e\}$   
 Ex:  $N \in D(\Lambda)$   $\hookrightarrow R(b_n) = n$

Thm:  $t_0 := \inf \{t: F(t) = 1\} \in \mathbb{R} \cup \{\infty\}$

$F \in D(\Lambda) \Leftrightarrow \exists A(z), \lim_{z \rightarrow t_0^-} A(z) = 0: \lim_{z \rightarrow t_0^-} \frac{R(z[1+A(z)t])}{R(z)} = -\log(\Lambda(t)) = e^{-t}$

possible choice:  $A(b_n) = a_n/b_n \quad \forall n \quad (z = b_n)$

$= \mathbb{P}\{X_1 > z[1+A(z)t] \mid X_1 > z\}$  residual lifetime

## 2. Stochastic exit problem in dim 1

$$dx_t = -V'(x_t)dt + \sigma dW_t$$

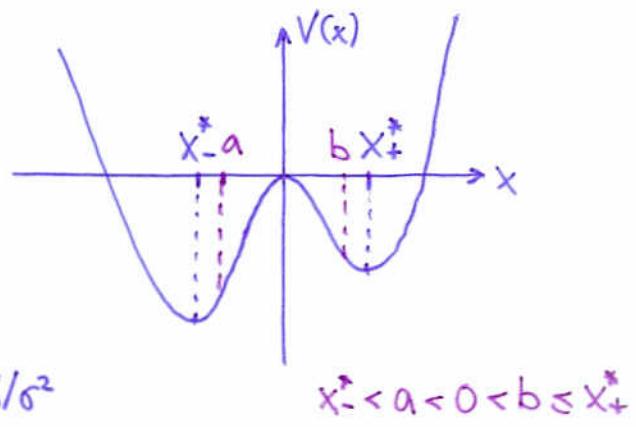
$$\lambda = |V''(0)|$$

$$\tau_x = \inf \{t > 0 : x_t = x\}$$

Well known:

$$\lim_{\delta \rightarrow 0} \mathbb{P}^a \{\tau_b > t | \mathbb{E}^a[\tau_b]\} = e^{-t}$$

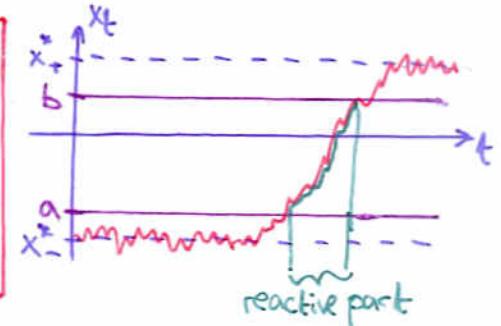
$$\mathbb{E}^a[\tau_b] \sim C e^{2[V(0) - V(x_-)]/\delta^2}$$



Thm A: [Cérou, Guyader, Lelièvre, Malrieu, ALEA 2013]

$$\lim_{\delta \rightarrow 0} \text{Law}(\lambda \tau_b - 2 \log \delta | \tau_b < \tau_a)$$

$$= \text{Law}(\underbrace{Z}_{\text{Gumbel}} + \underbrace{T(x_0, b)}_{\text{deterministic}})$$



Proof based on Doob's h-transform + exact computation

Yuri Bakhtin's approach [Stoch Dyn 2014 & arXiv: 1307.7060]

$$dx_t = \lambda x_t dt + \sigma dW_t \Rightarrow x_t = e^{\lambda t} \tilde{x}_t$$

$$\tilde{x}_t = x_0 + \sigma \int_0^t e^{-\lambda s} dW_s \stackrel{L}{=} x_0 + \sigma \underbrace{\tilde{W}_{(1-e^{-2\lambda t})/2\lambda}}_{\sim \sqrt{\frac{1-e^{-2\lambda t}}{2\lambda}} N}$$

Reflection principle:  $\mathbb{P}\{\tau_0 < t\} = 2 \mathbb{P}\{\tilde{x}_t > 0\}$

$$\mathbb{P}\{\tau_0 < t | \tau_0 < \infty\} = \frac{\mathbb{P}\{\tau_0 < t\}}{\mathbb{P}\{\tau_0 < \infty\}} = \frac{2 \mathbb{P}\{\tilde{x}_t > 0\}}{2 \mathbb{P}\{\tilde{x}_\infty > 0\}} = \mathbb{P}\{\tilde{x}_t > 0 | \tilde{x}_\infty > 0\}$$

$$\mathbb{P}\{\tau_0 < t + \frac{1}{\lambda} \log \delta | \tau_0 < \infty\} = \mathbb{P}\{\tilde{x}_{t + \frac{1}{\lambda} \log \delta} > 0 | \tilde{x}_\infty > 0\}$$

$$= \mathbb{P}\{N > \frac{|x_0|}{\delta} \sqrt{\frac{2\lambda}{1-\delta^2 e^{-2\lambda t}}} | N > \frac{|x_0|}{\delta} \sqrt{2\lambda}\}$$

$$[-\log \Lambda(e^{-x}) = \Lambda(x)] \quad \xrightarrow{\delta \rightarrow 0} \exp\{-x_0^2 \lambda e^{-2\lambda t}\} = \mathbb{P}\left\{\frac{Z}{2} + \frac{\log(x_0^2 \lambda)}{2\lambda} < t\right\}$$

Thm B: [Day, Bakhtin]  $\lim_{\delta \rightarrow 0} \text{Law}(\lambda \tau_0 - 1 \log \delta | \tau_0 < \tau_a) = \text{Law}\left(\frac{Z}{2} + \frac{\log(x_0^2 \lambda)}{2\lambda}\right)$

Thm C: [Day]  $x_0 = 0, a < 0 < b: \lim_{\delta \rightarrow 0} \text{Law}(\lambda \tau_b - 1 \log \delta | \tau_b < \tau_a) = \text{Law}\left(\Theta + \frac{\log(b^2 \lambda)}{2\lambda}\right)$   
 $\Theta = -\log(N)$

idea:  $|x_t| \stackrel{d}{=} \sigma \sqrt{\frac{1-e^{-2\lambda t}}{2\lambda}} |N| e^{\lambda t} \Rightarrow b \approx \sigma \frac{1}{\sqrt{2\lambda}} |N| e^{\lambda t}$

Thm B & Thm C  $\Rightarrow$  Thm A because  $\frac{1}{2} Z + \Theta \stackrel{d}{=} Z + \frac{1}{2} \log(2)$

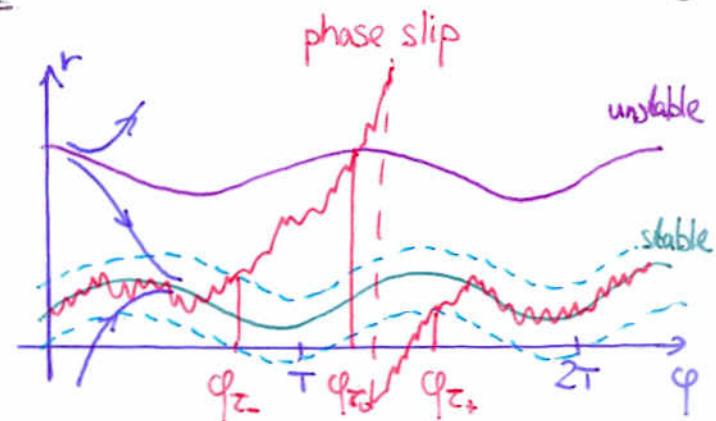
### 3. Stochastic exit problem in dim 2

③

$$\begin{cases} dr_t = f_r(r_t, \varphi_t) dt + \sigma g_r(r_t, \varphi_t) dW_t \\ d\varphi_t = f_\varphi(r_t, \varphi_t) dt + \sigma g_\varphi(r_t, \varphi_t) dW_t \end{cases}$$

Ex:  $r =$  phase diff between 2 oscillators  
 $\varphi =$  average phase

Q: Law of  $\varphi_{z_0}, \varphi_{z_-}, \varphi_{z_+}$ ?



⚠ quasipotential const on unst. orbit

Thm: [B&Gantz, SIAM J Math Anal 2014]

$$\lim_{m \rightarrow \infty} \left( \lim_{\delta \rightarrow 0} \text{Law}(\Theta(\varphi_{z_0}) - \log \delta) - \lambda T Y_m^\delta \right) = \text{Law}\left(\frac{Z}{2} - \frac{\log(2)}{2}\right)$$

\*  $\lambda T$ : instability of orbit (Lyapunov exponent  $\times$  period)

\*  $\Theta$ : explicit parametrisation of orbit,  $\Theta(\varphi+1) = \Theta(\varphi) + \lambda T$

\*  $Y_m^\delta$ :  $\lim_{n \rightarrow \infty} P\{Y_m^\delta = n+1 | Y_m^\delta = n\} = e^{-I_m/\delta^2}$   $I_m = I_\infty + O(e^{-2m\lambda T})$

"asymptotically geometric"  $Y_\infty^\delta = \#$  of optimal path

Thm: [NB 2014]

For appropriate def of  $Z_\pm$

$$\begin{cases} \lim_{\delta \rightarrow 0} \text{Law}(\Theta(\varphi_{z_0}) - \Theta(\varphi_{z_-}) - \log \delta) = \text{Law}\left(\frac{Z}{2} - \frac{\log 2}{2} + c_1\right) \\ \lim_{\delta \rightarrow 0} \text{Law}(\Theta(\varphi_{z_+}) - \Theta(\varphi_{z_0}) - \log \delta) = \text{Law}(\Theta + c_2) \\ \lim_{\delta \rightarrow 0} \text{Law}(\Theta(\varphi_{z_+}) - \Theta(\varphi_{z_-}) - 2 \log \delta) = \text{Law}(Z + c_1 + c_2) \end{cases}$$

and  $\text{Law}(\Theta(\varphi_{z_0^{nm}}) - \Theta(\varphi_{z_0^n}) - Y_m^\delta) \rightarrow \text{Logistic}$   
(density  $\sim \frac{1}{\cosh^2(t)}$ )