

Limit theory for Lévy semistationary processes

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joint work with A. Basse-O'Connor and M. Podolskij

Conference on Ambit Fields and Related Topics

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In the last talk:

Limit theory for the power variation in the setting of infill asymptotics,

$$V(p)^n := \sum_{i=1}^n |\Delta_i^n X|^p, \quad \Delta_i^n X := X_{\frac{i}{n}} - X_{\frac{i-1}{n}}.$$

Process $(X_t)_{t \in \mathbb{R}}$ a stationary increment moving average of the form

$$X_t = \int_{-\infty}^t \{g(t-s) - g_0(-s)\} dL_s$$

Limiting behavior of $V(p)^n$ is divided into three different regimes depending on

- β – the Blumenthal-Gettoor index of the driving Lévy process L
- α – the power characterising the behavior of g at 0
- p – the power for the power variation.

In this talk:

- 1 Generalisation to Lévy semistationary processes:

$$X_t := \int_{-\infty}^t \{g(t-s) - g_0(-s)\} \sigma_s dL_s,$$

where σ is a predictable process.

- 2 Functional convergence
- 3 Applications and further generalisations.

Lévy semistationary (LSS) Processes, definition and assumptions

A **Lévy semistationary** (LSS) process is given as

$$X_t = \int_{-\infty}^t (g(t-s) - g_0(-s)) \sigma_s dL_s.$$

- $(L_t)_{t \in \mathbb{R}}$ is a Lévy process on the real line.

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- $(\sigma_t)_{t \in \mathbb{R}}$ is càdlàg and predictable process, not necessarily independent of L .
- If σ is stationary and independent of L , then X is stationary.
- X generally not a semimartingale, nor an infinitely divisible process.

Motivation: Ambit fields and relative intermittency

$$X_t = \int_{-\infty}^t (g(t-s) - g_0(-s)) \sigma_s dL_s.$$

- LSS processes are an important purely temporal subclass of **ambit fields**, a class of stochastic processes introduced for modelling velocities in turbulent flows (Barndorff-Nielsen and Schmiegel 2005).

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- The **relative intermittency** process $\tilde{\sigma}_t^{2+} = (\int_0^t |\sigma_s|^2 ds) / (\int_0^1 |\sigma_s|^2 ds)$, $t \in [0, 1]$ models energy dissipation and is important for application in physics.

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- Typically, σ^2 is modelled as (exponential of an) ambit process, e.g. (Hedevang and Schmiegel 2013).

Limit theory for Brownian semistationary processes

(Barndorff-Nielsen, Corcuera, Podolskij 2009,2011): Limit theory for **BSS processes**:

$$X_t = \int_{-\infty}^t g(t-s)\sigma_s dW_s,$$

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Denote $\tau_n^2 = \int_0^{1/n} g^2(x) dx + \int_0^\infty (g(1/n+x) - g(x))^2 dx$.

Theorem 2.1

It holds that

$$n^{-1} \tau_n^{-p} \sum_{i=1}^{[tn]} |\Delta_i^n X|^p \xrightarrow{\mathbb{P}} \mathbb{E}[|\mathcal{N}(0,1)|^p] \int_0^t |\sigma_s|^p ds.$$

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\Rightarrow Consistent estimation of relative intermittency

$(\int_0^t |\sigma_s|^2 ds) / (\int_0^1 |\sigma_s|^2 ds)$ possible, cf. (Barndorff-Nielsen, Pakkanen and Schmiegel 2015).

Pure jump LSS processes:

$$X_t = \int_{-\infty}^t g(t-s)\sigma_s dL_s,$$

where $(L_t)_{t \in \mathbb{R}}$ is a symmetric pure jump Lévy process with Lévy measure ν .

- $\beta \in [0, 2)$: Blumenthal-Gettoor index of L , defined as

$$\beta := \inf \left\{ r \geq 0 : \int_{-1}^1 |x|^r \nu(dx) < \infty \right\}.$$

- $\alpha > 0$: Behavior of g at 0:

$$\lim_{t \downarrow 0} |g(t)|/t^\alpha = c_0 \in (0, \infty)$$

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The limiting behavior of $V(p)^n$ depends on α, β and p . We obtain three different regimes with different limits and convergence rates.

Theorem (Basse-O'Connor, H. and Podolskij)

(i): Assume that L is a $S\beta S$ process with $\beta \in (0, 2)$. If $\alpha \in (0, 1 - 1/\beta)$ and $p < \beta$, we obtain

$$n^{p(\alpha+1/\beta)-1} V(p)^n \xrightarrow{\mathbb{P}} m_p \int_0^1 |\sigma_t|^p dt.$$

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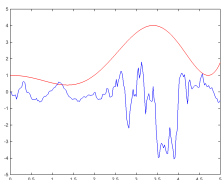
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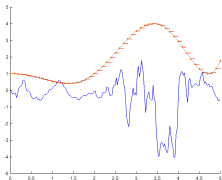
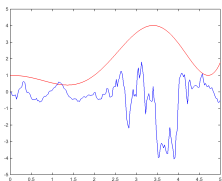
$$n^{p(\alpha+1/\beta)-1} V(p)^n \xrightarrow{\mathbb{P}} m_p.$$

Proof: Bernstein's blocking technique



$$X_t = \int_{-\infty}^t g(t-s)\sigma_s dL_s$$

$$V(p)^n = \sum_{i=1}^n |\Delta_i^n X|^p, \quad \Delta_i^n X = X_{\frac{i}{n}} - X_{\frac{i-1}{n}}$$



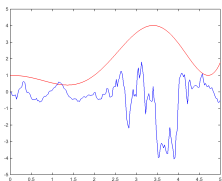
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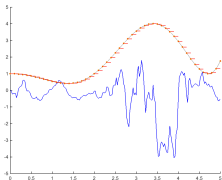
$$\tilde{V}(\rho)^n = \sum_{i=1}^n |\sigma_{\frac{i-1}{n}} \Delta_i^n Y|^\rho$$

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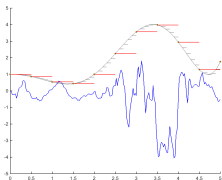
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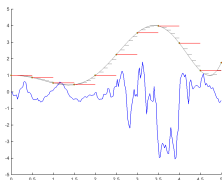
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Step 2: Introduce second block size $1/l$.

$$\tilde{V}(p)^{n,l} = \sum_{j=1}^l |\sigma_{\frac{j-1}{l}}|^p \left(\sum_{\substack{i \in [\frac{j-1}{l}, \frac{j}{l}]} |\Delta_i^n Y|^p \right)$$

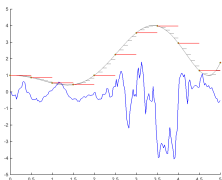


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It holds for all $\varepsilon > 0$ that

$$\lim_{l \rightarrow \infty} \limsup_{n \rightarrow \infty} \mathbb{P}(|n^{p(\alpha+1/\beta)-1}(\tilde{V}(p)^{l,n} - \tilde{V}(p)^n)| > \varepsilon) = 0$$



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Applying the limit theorem for constant σ we obtain

$$n^{p(\alpha+1/\beta)-1} \tilde{V}(p)^{l,n} \xrightarrow[n \rightarrow \infty]{\mathbb{P}} \sum_{j=1}^l |\sigma_{\frac{j-1}{l}}|^p \frac{m_p}{l} \xrightarrow[l \rightarrow \infty]{\mathbb{P}} m_p \int_0^1 |\sigma_t|^p dt.$$

Important ingredient: For asymptotic equivalence of $V(p)^n$, $\tilde{V}(p)^n$ and $\tilde{V}(p)^{l,n}$ we need the following isometry of the integral mapping.

Theorem (Kwapień, Woyczyński)

Let L be a symmetric β -stable Lévy process. There are positive constant c, C such that for all predictable F that are integrable w.r.t. L

$$c\mathbb{E}\left[\int_{\mathbb{R}} |F_s|^\beta ds\right] \leq \left\| \int_{\mathbb{R}} F_s dL_s \right\|_{\beta,\infty}^\beta \leq C\mathbb{E}\left[\int_{\mathbb{R}} |F_s|^\beta ds\right],$$

where $\|\cdot\|_{\beta,\infty}$ denotes the weak $L^\beta(\Omega)$ -norm.

The weak L^β -norm satisfy $\|X\|_{\beta'} \leq \|X\|_{\beta,\infty} \leq \|X\|_\beta$ for all $\beta' < \beta$.

Integration theory (Kwapień & Woyczyński, 1993), part 1

- Extension of the integration theory w.r.t. Lévy bases established in (Rajput & Rosiński 1989) towards predictable integrands

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- Decoupling inequalities approach

F is L -integrable (K & W) $\Leftrightarrow F(\omega)$ is L -integrable (R & R),
for almost all ω .
 $\Leftrightarrow \Phi_{L,0}(F) < \infty$, almost surely

Here, $\Phi_{L,0}$ is the functional

$$\Phi_{L,0}(F) := \int_{\mathbb{R}^2} |F_s x|^2 \wedge 1 \nu(dx) ds.$$

(Recall that L is a symmetric Lévy process)

Theorem (Basse-O'Connor, H. and Podolskij)

(ii) For $p \geq 1$, $\alpha > 1 - 1/(\beta \vee p)$ it holds that

$$n^{-1+p} V(p)^n \xrightarrow{\mathbb{P}} \int_0^1 |F_u|^p du$$

where

$$F_u = \int_{-\infty}^u g'(u-s) \sigma_s dL_s \quad \text{a.s.} \quad \text{and} \quad \int_0^1 |F_u|^p du < \infty \quad \text{a.s.}$$

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- For $\alpha > 1 - 1/(\beta \vee p)$, the sample paths of X are almost surely absolutely continuous with derivative F , cf. (Braverman and Samorodnitsky 1998).

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- For $\alpha > 1 - 1/(\beta \vee p)$, the sample paths of X are almost surely absolutely continuous with derivative F , cf. (Braverman and Samorodnitsky 1998).
- \Rightarrow By mean value theorem: $n^{-1} \sum_{i=1}^n |n\Delta_i^n X|^p \approx n^{-1} \sum_{i=1}^n |F_{\frac{i-1}{n}}|^p$, for large n .

Theorem (Basse-O'Connor, H. and Podolskij)

Assume that $\alpha < 1 - 1/p$, $p > \beta$ and $p \geq 1$. We obtain the \mathcal{F} -stable convergence

$$n^{\alpha p} V(p)^n \xrightarrow{\mathcal{L}\text{-s}} |c_0|^p \sum_{m: T_m \in [0,1]} |\Delta L_{T_m} \sigma_{T_m}|^p Z_m.$$

Here, $(T_m)_{m \geq 1}$ is a sequence of stopping times exhausting the jumps of $(L_t)_{t \geq 0}$, and

$$Z_m = \sum_{l=0}^{\infty} |(l + U_m)^\alpha - (l + U_m - 1)_+^\alpha|^p,$$

where $(U_m)_{m \geq 1}$ is a sequence of independent and uniform $[0, 1]$ -distributed random variables, defined on an extension of the original probability space, independent of L and σ .

Note that Z_m is finite since $(\alpha - 1)p < -1$.

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- In which sense do we get convergence of $n^\gamma V(p)^n$ to a limiting process in $L^0(\Omega, \mathbb{D}([0, 1]))$, where γ is the convergence rate established in the last section?

Theorem (Basse-O'Connor, H. and Podolskij)

(i'): Assume that L is a S β S process with $\beta \in (0, 2)$. If $\alpha \in (0, 1 - 1/\beta)$ and $p < \beta$, we obtain

$$n^{p(\alpha+1/\beta)-1} V(p)_t^n \xrightarrow{\text{u.c.p.}} m_p \int_0^t |\sigma_s|^p ds.$$

- $Z^n \xrightarrow{\text{u.c.p.}} Z$ ('uniformly on compacts in probability') if for all $C > 0$ and for all $\varepsilon > 0$

$$\mathbb{P}\left(\sup_{t \in [0, C]} |Z_t^n - Z_t| > \varepsilon\right) \rightarrow 0.$$

Theorem (Basse-O'Connor, H. and Podolskij)

(ii') For $p \geq 1$, $\alpha > 1 - 1/(\beta \vee p)$ it holds that

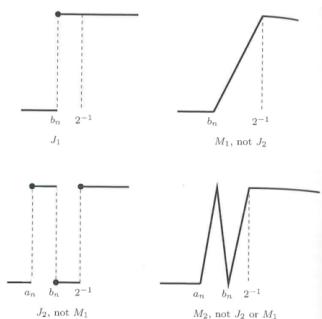
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Examples for convergence towards a function with a single jump in the different topologies.

Source: W.Whitt, Stochastic-Process Limits

Figure 11.2. Four candidate sequences of functions $\{x_n : n \geq 1\}$ that might converge to $x = I_{(1/2, 1]}$ in $D([0, 1], \mathbb{R})$, where $a_n = 2^{-1} - 2n^{-1}$ and $b_n = 2^{-1} - n^{-1}$.

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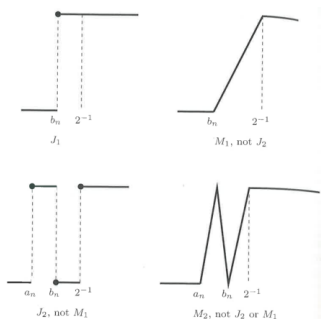


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(Avram & Taqqu 1998): Functional convergence of sums of moving averages w.r.t. M_1 but not J_1 topology.

Theorem (Basse-O'Connor, H. and Podolskij)

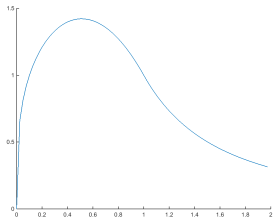
Assume that $\alpha < 1 - 1/p$, $p > \beta$ and $p \geq 1$. We obtain the functional \mathcal{F} -stable convergence

$$n^{\alpha p} V^n(p)_t \xrightarrow{\mathcal{L}_{M_1} - s} |c_0|^p \sum_{m: T_m \in [0, t]} |\Delta L_{T_m} \sigma_{T_m}|^p Z_m,$$

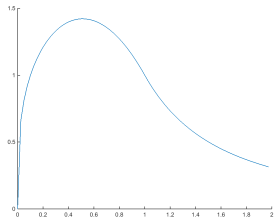
where

$$Z_m = \sum_{l=0}^{\infty} |(l + U_m)^\alpha - (l + U_m - 1)_+^\alpha|^p, \quad U_m \sim \mathcal{U}([0, 1]).$$

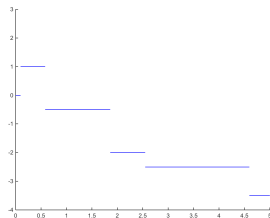
The stable convergence in law does also hold with respect to the M_2 topology, but not with respect to the J_1 or J_2 topology.



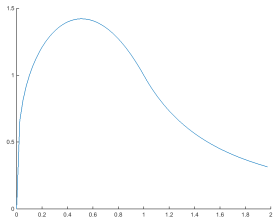
$$g(x) \sim x^\alpha, \text{ as } x \rightarrow 0.$$



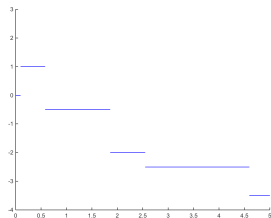
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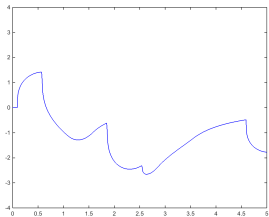
Lévy process L .



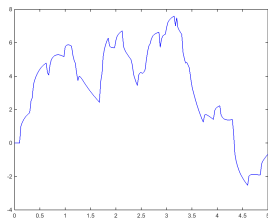
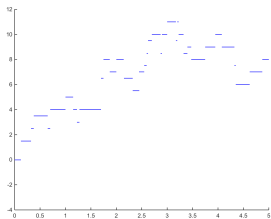
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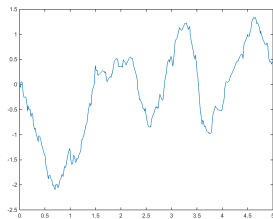
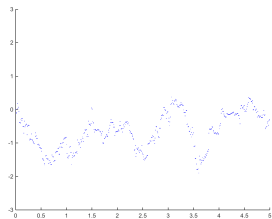
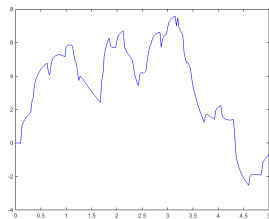
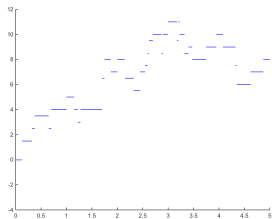


Lévy process L .



- $X_t = \int_{-\infty}^t g(t-s)\sigma_s dL_s$, $\sigma \equiv 1$.
- Jump times of Lévy process govern asymptotic behavior of $V(\rho)^n$.





Let L be compound Poisson process with a jump at time $T \in ((i_0 - 1)/n, i_0/n]$, then

$$\Delta_{i_0}^n X \approx c_0(i_0/n - T)^\alpha \sigma_T \Delta L_T,$$

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Lemma 1

For an absolutely continuous random variable Z with differentiable density we have the \mathcal{F} -stable convergence

$$\{nZ\} \xrightarrow{\mathcal{L}\text{-s}} U,$$

where $U \sim \mathcal{U}([0, 1])$ is independent of Z . Here, $\{x\} = x - [x]$ denotes the fractional part of x .

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Since $nT \in ((i_0 - 1), i_0]$, we obtain

$$|\Delta_{i_0}^n X|^p \stackrel{d}{\approx} n^{-\alpha p} |c_0 \sigma_T \Delta L_T|^p |U|^{\alpha p},$$

$$|\Delta_i^n X|^p \stackrel{d}{\approx} n^{-\alpha p} |c_0 \sigma_T \Delta L_T|^p (U + i - i_0)^\alpha - (U + i - i_0 - 1)^\alpha |^p.$$

Extension to general L

Idea: Let $a > 0$ and let $L^{>a}$ be the truncated Lévy process

$$L_t^{>a} - L_s^{>a} = \sum_{u \in (s, t]} \Delta L_u 1_{\{|\Delta L_u| > a\}},$$

and $L_t^{\leq a} = L_t - L_t^{>a}$. Let

$$X_t^{>a} = \int_{-\infty}^t g(t-s) \sigma_s dL_s^{>a}, \quad X_t^{\leq a} = \int_{-\infty}^t g(t-s) \sigma_s dL_s^{\leq a}.$$

Claim: The error in the power variation caused by replacing X by $X^{>a}$ becomes negligible for $a \rightarrow 0$. More precisely, we show that $\limsup_{n \rightarrow \infty} \mathbb{P}[n^{p\alpha} V(X^{\leq a}, p)_t^n > \varepsilon] \rightarrow 0$, as $a \rightarrow 0$ for all $\varepsilon > 0$.

Integration Theory (Kwapień & Woyczyński), part 2

Define for $p \geq 1$ and for a predictable process F the functional

$$\Phi_{L,p}(F) = \int_{\mathbb{R}^2} |F_s x|^2 1_{\{|F_s x| \leq 1\}} + |F_s x|^p 1_{\{|F_s x| > 1\}} \nu(dx) ds.$$

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Moreover, on the linear space of F with $\Phi_{L,p}(F) < \infty$ almost surely, introduce the random (quasi-)norm

$$\|F\|_{p,L} := \inf\{\lambda \geq 0 : \Phi_{p,L}(F/\lambda) \leq 1\}.$$

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There are positive constants c, C such that we obtain for all F with $\Phi_{L,p}(F) < \infty$

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For deterministic integrands the result was shown in (Rajput & Rosiński 1989).

Application: Estimation of α and β

Three regimes:

Thm (i): $\alpha < 1 - 1/\beta, p < \beta.$ $n^{-1+p(\alpha+1/\beta)} V(p)^n$ converges

Thm (ii): $\alpha > 1 - 1/p.$ $n^{p-1} V(p)^n$ converges

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$$S_{\alpha,\beta}(n, p) := -\frac{\log V(p)^n}{\log n}$$

$$S_{\alpha,\beta}(n, p) \xrightarrow{\mathbb{P}} S_{\alpha,\beta}(p) := \begin{cases} \alpha p: & \alpha < 1 - 1/p \text{ and } p > \beta \\ p(\alpha + 1/\beta) - 1: & \alpha < 1 - 1/\beta \text{ and } p < \beta \\ p - 1: & \alpha > 1 - 1/\max(p, \beta) \end{cases}$$

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$$(\hat{\alpha}_n, \hat{\beta}_n) := \operatorname{argmin}_{\alpha > 0, \alpha + 1/\beta \in (1/2, 1)} \int_1^2 (S_{\alpha,\beta}(n, p) - S_{\alpha,\beta}(p))^2 dp.$$

In the context of Theorem (i), that is for β -stable driving Lévy process, and $\alpha < 1 - 1/\beta$, $p < \beta$, we obtain

$$\frac{\sum_{i=1}^{\lfloor tn \rfloor} |\Delta_i^n X|^p}{\sum_{i=1}^n |\Delta_i^n X|^p} \xrightarrow{\mathbb{P}} \frac{\int_0^t |\sigma_s|^p ds}{\int_0^1 |\sigma_s|^p ds}, \quad t \in (0, 1).$$

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The right hand side is the **relative intermittency** and plays an important role in turbulence applications.

We allow for k th order increments:

$$\Delta_{i,k}^n X := \sum_{j=0}^k (-1)^j \binom{k}{j} X_{(i-j)/n}, \quad i \geq k.$$

The limiting behavior of $V_n(X, p; k)$ depends on the interplay between α, β, p and k .

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