EXTREMAL INDEX, HITTING TIME STATISTICS AND PERIODICITY

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ABSTRACT. We give conditions to prove the existence of an Extremal Index for general stationary stochastic processes by detecting the presence of one or more underlying periodic phenomena. This theory, besides giving general useful tools to identify the extremal index, is also tailored to dynamical systems. In fact, we apply this idea to analyse the possible Extreme Value Laws for the stochastic process generated by observations taken along dynamical orbits with respect to various measures. As in the authors' previous works on this topic, the analogy of these laws in the context of hitting time statistics is explained and exploited extensively.

1. INTRODUCTION

The study of extreme or rare events is of great importance in a wide variety of fields and is often tied in with risk assessment. This explains why Extreme Value Laws (EVL) and the estimation of the tail distribution of the maximum of a large number of observations has drawn much attention and become a highly developed subject.

In many practical situations, such as in the analysis of financial markets or climate phenomena, time series can be modelled by a dynamical system which describes its time evolution. The recurrence effect introduced by Poincaré, which is present in chaotic systems, is the starting point for a deeper analysis of the limit distribution of the elapsed time until the occurrence of a rare event, which is usually referred to as Hitting Time Statistics (HTS) and Return Time Statistics (RTS).

In [FFT10], we established the connection between the existence of EVL and HTS/RTS for stochastic processes arising from discrete time chaotic dynamical systems. This general link allowed us to obtain results of EVL using tools from HTS/RTS and the other way around (this was applied in cases where the extremal index was 1, which is the most classical setting).

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The extremal index (EI) $\theta \in [0, 1]$ is a measure of clustering of extreme events, the lower the index, the higher the degree of clustering. In this paper, we give general conditions to prove the existence of an extremal index $0 < \theta < 1$, which can be applied to any stationary stochastic process. Although our results apply to general stationary stochastic processes, we will be particularly interested in the case where the stochastic process arises from a discrete time dynamical system. This setup will provide not only a huge diversity of examples, but also a motivation for the conditions we propose as well, as a better understanding of their implications. Namely, motivated by the study of stochastic processes arising from chaotic dynamical systems, we associate the extremal index to the occurrence of periodic phenomena. We will illustrate these results by applying them to time series provided by dynamical systems as well as to an Autoregressive (AR) process introduced by Chernick and two Maximum Moving Averages (MMA) processes.

From a probabilist's perspective, the main interest of our results is likely to follow from the facts that:

- we give conditions to prove the existence of an extremal index, that, to our knowledge, are the weakest of the kind and which can, roughly speaking, be reduced to checking periodicity and mixing;
- we interpret the extremal index as a result of underlying periodic phenomena; and
- we provide a new class of examples of stochastic processes (arising from chaotic dynamical systems) which present an extremal index that can take any value $\theta \in (0, 1]$.

On the other hand, the main interest of our paper from a dynamicist' viewpoint is likely to be our use of tools from EVL to prove the existence of HTS/RTS around periodic points, and which apply to balls rather than cylinders. To our knowledge this is the first result of HTS/RTS different from the standard exponential which applies to balls. We do this for non-uniformly hyperbolic systems such as Rychlik maps and the full quadratic map (also known as the quadratic Chebyshev polynomial), where invariant measures are absolutely continuous w.r.t. Lebesgue or are, more generally, equilibrium states w.r.t. certain potentials. Moreover, for the full shift equipped with the Bernoulli measure, we show that an extremal index different from 1 can only occur at periodic points (we believe that this last result holds in greater generality, but do not prove that here). Hence, this result raises the following:

Question. Is it possible to prove the existence of an EI in (0,1) without some sort of periodicity?

In a more concrete formulation:

Question. For stationary stochastic processes arising from chaotic dynamical systems, is it possible to prove the existence of an EI in (0,1), either for EVL or HTS/RTS around non-periodic points?

Throughout this paper the notation $A(u) \sim B(u)$, for u approaching u_0 , means that $\lim_{u\to u_0} \frac{A(u)}{B(u)} = 1$. When u = n and $u_0 = \infty$ we will just write $A(n) \sim B(n)$. The notation

A(n) = o(n) means that $\lim_{n\to\infty} \frac{A(n)}{n} \to 0$. Also, let [x] denote the integer part of the positive real number x and for a set A, the notation A^c will denote the complement of the set A.

1.1. Extreme Value Theory for both independent and dependent stochastic processes. From here on, the sequence X_0, X_1, X_2, \ldots will always denote a stationary stochastic process with marginal distribution function (d.f.) F, *i.e.*, $F(x) = \mathbb{P}(X_0 \leq x)$. Let

$$\bar{F} = 1 - F$$

and u_F denote the right endpoint of the d.f. F, *i.e.*, $u_F = \sup\{x : F(x) < 1\}$. We have an *exceedance* of the level $u \in \mathbb{R}$ at time $j \in \mathbb{N}$ if the event $\{X_j > u\}$ occurs. Define a new sequence of random variables (r.v.) M_1, M_2, \ldots given by

$$M_n = \max\{X_0, \dots, X_{n-1}\}$$

Definition 1. We say that we have an *Extreme Value Law* (EVL) for M_n if there is a non-degenerate d.f. $H : \mathbb{R} \to [0, 1]$ with H(0) = 0 and, for every $\tau > 0$, there exists a sequence of levels $u_n = u_n(\tau)$, $n = 1, 2, \ldots$, such that

$$n\mathbb{P}(X_0 > u_n) \to \tau, \text{ as } n \to \infty,$$
 (1.1)

and for which the following holds:

$$\mathbb{P}(M_n \le u_n) \to \overline{H}(\tau), \text{ as } n \to \infty.$$
 (1.2)

In the case X_0, X_1, X_2, \ldots are independent and identically distributed (i.i.d.) r.v. then since $\mathbb{P}(M_n \leq u_n) = (F(u_n))^n$ we have

$$\log(\mathbb{P}(M_n \le u_n)) = n \log(1 - \mathbb{P}(X_0 > u_n)) \sim -n\mathbb{P}(X_0 > u_n),$$

which implies that if (1.1) holds, then (1.2) holds with $\bar{H}(\tau) = e^{-\tau}$ and vice versa (see [LLR83, Theorem 1.5.1]).

When X_0, X_1, X_2, \ldots are not independent but satisfy some mixing condition $D(u_n)$ introduced by Leadbetter in [L73] then something can still be said about H. Let F_{i_1,\ldots,i_n} denote the joint d.f. of X_{i_1}, \ldots, X_{i_n} , and set $F_{i_1,\ldots,i_n}(u) = F_{i_1,\ldots,i_n}(u,\ldots,u)$.

Condition $(D(u_n))$. We say that $D(u_n)$ holds for the sequence X_0, X_1, \ldots if for any integers $i_1 < \ldots < i_p$ and $j_1 < \ldots < j_k$ for which $j_1 - i_p > m$, and any large $n \in \mathbb{N}$,

$$\left|F_{i_1,\dots,i_p,j_1,\dots,j_k}(u_n) - F_{i_1,\dots,i_p}(u_n)F_{j_1,\dots,j_k}(u_n)\right| \le \gamma(n,m),$$

where $\gamma(n, m_n) \xrightarrow[n \to \infty]{} 0$, for some sequence $m_n = o(n)$.

If $D(u_n)$ holds for X_0, X_1, \ldots and the limit (1.2) exists for some $\tau > 0$ then there exists $0 \leq \theta \leq 1$ such that $\bar{H}(\tau) = e^{-\theta\tau}$ for all $\tau > 0$ (see [L83, Theorem 2.2] or [LLR83, Theorem 3.7.1]).

Definition 2. We say that X_0, X_1, \ldots has an *Extremal Index* (EI) $0 \le \theta \le 1$ if we have an EVL for M_n with $\bar{H}(\tau) = e^{-\theta\tau}$ for all $\tau > 0$. The notion of the EI was latent in the work of Loynes [L65] but was established formally by Leadbetter in [L83]. It gives a measure of the strength of the dependence of X_0, X_1, \ldots , so that $\theta = 1$ indicates that the process has practically no memory while $\theta = 0$, conversely, reveals extremely long memory. Another way of looking at the EI is that it gives some indication on how much exceedances of high levels have a tendency to "cluster". Namely, for $\theta > 0$ this interpretation of the EI is that θ^{-1} is the mean number of exceedances of a high level in a cluster of large observations, *i.e.*, is the "mean size of the clusters".

Remark 1. The sequences of real numbers $u_n = u_n(\tau)$, n = 1, 2, ..., are usually taken to be one parameter linear families such as $u_n = a_n y + b_n$, where $y \in \mathbb{R}$ and $a_n > 0$, for all $n \in \mathbb{N}$. Observe that τ depends on y through u_n and, in fact, in the i.i.d. case, depending on the tail of the marginal d.f. F, we have that $\tau = \tau(y)$ is of one of the following three types (for some $\alpha > 0$):

 $\tau_1(y) = e^{-y}$ for $y \in \mathbb{R}$, $\tau_2(y) = y^{-\alpha}$ for y > 0 and $\tau_3(y) = (-y)^{\alpha}$ for $y \le 0$.

1.2. Hitting and return time statistics. Consider a deterministic discrete time dynamical system $(\mathcal{X}, \mathcal{B}, \mu, f)$, where \mathcal{X} is topological space, \mathcal{B} is the Borel σ -algebra, $f : \mathcal{X} \to \mathcal{X}$ is a measurable map and μ is an *f*-invariant probability measure, *i.e.*, $\mu(f^{-1}(B)) = \mu(B)$, for all $B \in \mathcal{B}$. One can think of $f : \mathcal{X} \to \mathcal{X}$ as the evolution law that establishes how time affects the transitions from one state in \mathcal{X} to another.

Consider now a set $A \in \mathcal{B}$ and a new r.v. that we refer to as first hitting time to A and denote by $r_A : \mathcal{X} \to \mathbb{N} \cup \{+\infty\}$ where

$$r_A(x) = \min\{j \in \mathbb{N} \cup \{+\infty\} : f^j(x) \in A\}.$$

Given a sequence of sets $\{U_n\}_{n\in\mathbb{N}}$ so that $\mu(U_n) \to 0$ we consider the sequence of r.v. r_{U_1}, r_{U_2}, \ldots If under suitable normalisation r_{U_n} converges in distribution to some nondegenerate d.f. G we say that the system has *Hitting Time Statistics* (HTS) for $\{U_n\}_{n\in\mathbb{N}}$. For systems with 'good mixing properties', G is the standard exponential d.f., in which case, we say that we have *exponential HTS*.

We say that the system has HTS G to balls at ζ if for any sequence $(\delta_n)_{n \in \mathbb{N}} \subset \mathbb{R}^+$ such that $\delta_n \to 0$ as $n \to \infty$ we have HTS G for $(U_n)_n = (B_{\delta_n}(\zeta))_n$.

Let \mathcal{P}_0 denote a partition of \mathcal{X} . We define the corresponding pullback partition $\mathcal{P}_n = \bigvee_{i=0}^{n-1} f^{-i}(\mathcal{P}_0)$, where \vee denotes the join of partitions. We refer to the elements of the partition \mathcal{P}_n as cylinders of order n. For every $\zeta \in \mathcal{X}$, we denote by $Z_n[\zeta]$ the cylinder of order n that contains ζ . For some $\zeta \in \mathcal{X}$ this cylinder may not be unique, but we can make an arbitrary choice, so that $Z_n[\zeta]$ is well defined. We say that the system has HTS G to cylinders at ζ if we have HTS G for $U_n = Z_n(\zeta)$.

Let μ_A denote the conditional measure on $A \in \mathcal{B}$, *i.e.*, $\mu_A := \frac{\mu|_A}{\mu(A)}$. Instead of starting somewhere in the whole space \mathcal{X} , we may want to start in U_n and study the fluctuations of the normalised return time to U_n as n goes to infinity, *i.e.*, for each n, we look at the random variables r_{U_n} as being defined in the probability space $(U_n, \mathcal{B} \cap U_n, \mu_{U_n})$ and wonder if, under some normalisation, they converge in distribution to some non-degenerate d.f. G, in which case, we say that the system has *Return Time Statistics* (RTS) \tilde{G} for $\{U_n\}_{n\in\mathbb{N}}$. The existence of exponential HTS is equivalent to the existence of exponential RTS. In fact, according to the Main Theorem in [HLV05], a system has HTS G if and only if it has RTS \tilde{G} and

$$G(t) = \int_0^t (1 - \tilde{G}(s)) \, ds.$$
(1.3)

Regarding normalising sequences to obtain HTS/RTS, we recall Kac's Lemma, which states that the expected value of r_A with respect to μ_A is $\int_A r_A d\mu_A = 1/\mu(A)$. So in studying the fluctuations of r_A on A, the relevant normalising factor should be $1/\mu(A)$.

Definition 3. Given a sequence of sets $(U_n)_{n \in \mathbb{N}}$ so that $\mu(U_n) \to 0$, the system has *HTS* G for $(U_n)_{n \in \mathbb{N}}$ if for all $t \ge 0$

$$\mu\left(r_{U_n} \le \frac{t}{\mu(U_n)}\right) \to G(t) \text{ as } n \to \infty,$$
(1.4)

and the system has RTS \tilde{G} for $(U_n)_{n \in \mathbb{N}}$ if for all $t \geq 0$

$$\mu_{U_n}\left(r_{U_n} \le \frac{t}{\mu(U_n)}\right) \to \tilde{G}(t) \text{ as } n \to \infty.$$
(1.5)

The theory of HTS/RTS laws is now a well developed theory, applied first to cylinders and hyperbolic dynamics, and then extended to balls and also to non-uniformly hyperbolic systems. We refer to [C00] and [S09] for very nice reviews as well as many references on the subject. (See also [AG01], where the focus is more towards a finer analysis of uniformly hyperbolic systems.) Since the early papers [P91, H93], several different approaches have been used to prove HTS/RTS: from the analysis of adapted Perron-Frobenius operators as in [H93], the use of inducing schemes as in [BSTV03], to the relation between recurrence rates and dimension as explained in [S09, Section 4].

For many mixing systems it is known that the HTS/RTS are standard exponential around almost every point. Among these systems we note the following: Markov chains [P91], Axiom A diffeomorphisms [H93], uniformly expanding maps of the interval [C96], 1dimensional non-uniformly expanding maps [HSV99, BSTV03, BV03, BT09], partially hyperbolic dynamical systems [D04], toral automorphisms [DGS04], higher dimensional non-uniformly hyperbolic systems (including Hénon maps) [CC10].

However, even for systems with good mixing properties, it is known at least since [H93] that at some special (periodic) points, similar distributions for the HTS/RTS (for cylinders) with an exponential parameter $0 < \theta < 1$ (*i.e.*, $1 - G(t) = e^{-\theta t}$) apply. This subject was studied, also in the cylinder context, in [HV09], where the sequence of successive returns to neighbourhoods of these points was proved to converge to a compound Poisson process.

1.3. The connection between EVL and HTS/RTS. We start by explaining what we mean by stochastic processes arising from discrete time dynamical systems. Take a system $(\mathcal{X}, \mathcal{B}, \mu, f)$ and consider the time series X_0, X_1, X_2, \ldots arising from such a system simply by evaluating a given random variable $\varphi : \mathcal{X} \to \mathbb{R} \cup \{\pm \infty\}$ along the orbits of the system, or in other words, the time evolution given by successive iterations by f:

$$X_n = \varphi \circ f^n$$
, for each $n \in \mathbb{N}$. (1.6)

Clearly, X_0, X_1, \ldots defined in this way is not an independent sequence. However, f-invariance of μ guarantees that this stochastic process is stationary. We assume that φ achieves a global maximum at $\zeta \in \mathcal{X}$ and the event $\{x \in \mathcal{X} : \varphi(x) > u\} = \{X_0 > u\}$ corresponds to a topological ball "centred" at ζ . EVLs for the partial maximum of such sequences have been proved directly in the recent papers [Co01, FF08, FFT10, HNT10, GNH10]. We highlight the pioneer work of Collet [Co01] for the innovative ideas introduced, some of which were used in the latter including the present paper. The dynamical systems covered in these papers include non-uniformly hyperbolic 1-dimensional maps (in all of them), higher dimensional non-uniformly expanding maps in [FFT10], suspension flows in [HNT10], billiards and Lozi maps in [GNH10].

In [FFT10], we formally established the link between EVL and HTS/RTS (for balls) of stochastic processes given by (1.6). Essentially, we proved that if such time series have an EVL H then the system has HTS H for balls "centred" at ζ and vice versa. Recall that having HTS H is equivalent to say that the system has RTS \tilde{H} , where H and \tilde{H} are related by (1.3). This was based on the elementary observation that for stochastic processes given by (1.6) we have:

$$\{M_n \le u\} = \{r_{\{X_0 > u\}} > n\}.$$
(1.7)

We exploited this connection to prove EVL using tools from HTS/RTS and the other way around. In [FFT10a], we carried the connection further to include the cases where the invariant measure μ may not be absolutely continuous with respect to Lebesgue measure and also to understand HTS/RTS for cylinders rather than balls in terms of EVL. To achieve the latter we introduced the notion of *cylinder EVL* which essentially requires that the limits (1.1) and (1.2) exist only for particular time subsequences $\{\omega_j\}_{j\in\mathbb{N}}$ of $\{n\}_{n\in\mathbb{N}}$ (see Section 5).

Hence, under the conditions of [FFT10, Theorem 2], when X_0, X_1, X_2, \ldots has an EI $\theta < 1$ then we have HTS for balls G given by

$$G(\tau) = 1 - \mathrm{e}^{-\theta\tau}.\tag{1.8}$$

Using (1.7) plus the integral relation (1.3) and arguing as in the proof of [FFT10, Theorem 2], we have RTS for balls \tilde{G} that can be written as:

$$\tilde{G}(\tau) = \lim_{n \to \infty} \mu(M_n > u_n | X_0 > u_n) = (1 - \theta) + \theta(1 - e^{-\theta\tau}),$$
(1.9)

or in other words: the return time law is the convex combination of a Dirac law at zero and an exponential law of average θ^{-1} where the weight is the EI θ itself.

As a consequence of this relation new light can be brought to the work of Galves and Schmitt [GS90] who introduced a short correction factor λ in order to get exponential HTS, that was then studied later in great detail by Abadi *et al.* [AG01, A04, A06, AV08, AVe09, AS10], and which, in case of being convergent, can be seen as the EI. This means that we now have two different perspectives from which to look at the aforementioned papers, as well as to look at the work of those who developed the probabilistic theory of the EI such as [L83, LLR83, O87, LR88, HHL88, LN89, CHM91]. Just to give an example of the advantage of realising this connection, we observe that O'Brien's formula for the EI in [O87], which is widely used in the estimation of the EI, can be easily derived from formula (1.9) for the RTS.

1.4. Structure of the paper. The paper is organised as follows. In Section 2 we give conditions to prove the existence of an EI for general stochastic processes; initially this is applied to 'first order' clustering behaviour and then later to higher order clustering. In Section 3 we give a very general introduction to the dynamical systems and accompanying measures we will be studying. We also explain the link between EVL and HTS and state general theorems for those laws in this context. In Section 4 we give some concrete examples of dynamical systems, measures and observables yielding EVLs with EI in (0, 1). These examples are so-called Rychlik systems as well as the full quadratic map. Section 5 is a short section explaining the relevant conditions required to guarantee an EVL for returns to cylinders rather than balls, while Section 6 shows that in that context we can completely characterise all possible EVLs for simple dynamical systems. Finally in the appendix we show how our conditions apply to various types of random variables not necessarily produced by a dynamical system, namely, two MMA processes and one AR(1) introduced by Chernick in [C81].

2. Extremal index and periodicity

In this section we give conditions that can be applied to any stationary stochastic process and which allow to prove the existence of an EI by realising the presence of one or more underlying periodic phenomenon.

2.1. Existence of an EI due to the presence of periodic phenomena. The first condition imposes some type periodic behaviour of period $p \in \mathbb{N}$ plus a summability requirement. For that reason we shall denote it by $SP_{p,\theta}$ which stands for Summable Periodicity of period p. To state the condition we will use a sequence of levels $(u_n)_n$ as in (1.1).

Condition (SP_{*p*, θ}(u_n)). We say that X_0, X_1, X_2, \ldots satisfies condition SP_{*p*, θ}(u_n) for $p \in \mathbb{N}$ and $\theta \in [0, 1]$ if

$$\lim_{n \to \infty} \sup_{1 \le j < p} \mathbb{P}(X_j > u_n | X_0 > u_n) = 0 \quad \text{and} \quad \lim_{n \to \infty} \mathbb{P}(X_p > u_n | X_0 > u_n) \to (1 - \theta) \quad (2.1)$$

and moreover

$$\lim_{n \to \infty} \sum_{i=0}^{\left[\frac{n-1}{p}\right]} \mathbb{P}(X_0 > u_n, X_p > u_n, X_{2p} > u_n, \dots, X_{ip} > u_n) = 0.$$
(2.2)

Condition (2.1), when $\theta < 1$, imposes some sort of periodicity of period p among the exceedances of high levels u_n , since if at some point the process exceeds the high level u_n , then, regardless of how high u_n is, there is always a strictly positive probability of another exceedance occurring at the (finite) time p. In fact, if the process is generated by a deterministic dynamical system $f : \mathcal{X} \to \mathcal{X}$ as in (1.6) and f is continuous then (2.1) implies that ζ is a periodic point of period p, *i.e.*, $f^p(\zeta) = \zeta$.

We also state a stronger condition, which is often simpler to check than $SP_{p,\theta}(u_n)$ and which requires, besides the periodicity, some type of Markov behaviour that which immediately implies the summability condition (2.2). We call it $MP_{p,\theta}$ which stands for Markovian Periodicity. We will check this condition rather than $SP_{p,\theta}(u_n)$ in the applications presented in Sections 3 and 4 as well as in Appendix C.

Condition (MP_{*p*, θ}(u_n)). We say that X_0, X_1, X_2, \ldots satisfies the condition MP_{*p*, θ}(u_n) for $p \in \mathbb{N}$ and $\theta \in [0, 1]$ if

$$\lim_{n \to \infty} \sup_{1 \le j < p} \mathbb{P}(X_j > u_n | X_0 > u_n) = 0 \text{ and}$$

$$\lim_{n \to \infty} \sup_i \frac{\mathbb{P}(X_p > u_n, X_{2p} > u_n, \dots, X_{ip} > u_n | X_0 > u_n)}{(1 - \theta)^i} = 1.$$
(2.3)

Note that if besides condition (2.1) the stationary stochastic process satisfies the following Markovian property:

$$\mathbb{P}(X_{ip} > u | X_{(i-1)p} > u, \dots, X_0 > u) = \mathbb{P}(X_{ip} > u | X_{(i-1)p} > u), \text{ for all } i \in \mathbb{N},$$
(2.4)

then it can easily be seen by an induction argument that condition $MP_{p,\theta}(u_n)$ holds. Assuming that $SP_{p,\theta}(u_n)$ holds, for $i, s, \ell \in \mathbb{N} \cup \{0\}$, we define the events:

$$Q_{p,i}(u) := \{X_i > u, X_{i+p} \le u\}, \ Q_{p,i}^*(u) := \{X_i > u\} \setminus Q_{p,i}(u) \text{ and } \mathscr{Q}_{p,s,\ell}(u) = \bigcap_{i=s}^{s+\ell-1} Q_{p,i}^c(u).$$

Assuming $\theta < 1$, by (2.1), we know that the stochastic process has some underlying periodic behaviour such that the occurrence of an exceedance of an high level u_n at time *i* leads to another exceedance at time i + p, with probability approximately $(1 - \theta)$. Therefore,

- $Q_{p,i}^*(u_n)$ corresponds exactly to the realisations of the process with an exceedance of u_n , at time *i*, which were "*captured*" by the underlying periodic phenomena; and
- $Q_{p,i}(u_n)$ corresponds to those realisations with an exceedance of u_n , at time *i*, but that manage to "escape" the periodic behaviour.

Hence, if $Q_{p,i}^*(u_n)$ occurs, then we say we have a *capture* at time *i* while, if $Q_{p,i}(u_n)$ occurs, then we say we have an *escape* at time *i*. In this terminology, the event $\mathscr{Q}_{p,s,\ell}(u_n)$ corresponds to the realisations for which no escapes occur between time *s* and $s + \ell - 1$. Note that for either a capture or an escape to occur at time *i*, an exceedance must occur at that time.

Note that if condition $SP_{p,\theta}(u_n)$ holds we must have:

$$n\mathbb{P}(Q_{p,0}^*(u_n)) = n\mathbb{P}(X_0 > u_n, X_p > u_n) = n\mathbb{P}(X_0 > u_n)\mathbb{P}(X_p > u_n | X_0 > u_n) \xrightarrow[n \to \infty]{} \tau(1-\theta)$$

and consequently conclude that under $SP_{p,\theta}(u_n)$, we have

$$n\mathbb{P}(Q_{p,0}(u_n)) \to \theta \tau, \quad \text{as } n \to \infty.$$
 (2.5)

Condition $(D^p(u_n))$. We say that $D^p(u_n)$ holds for the sequence X_0, X_1, X_2, \ldots if for any integers ℓ, t and n

$$\left|\mathbb{P}\left(Q_{p,0}(u_n) \cap \mathscr{Q}_{p,t,\ell}(u_n)\right) - \mathbb{P}(Q_{p,0}(u_n))\mathbb{P}(\mathscr{Q}_{p,0,\ell}(u_n))\right| \le \gamma(n,t),$$

where $\gamma(n,t)$ is nonincreasing in t for each n and $n\gamma(n,t_n) \to 0$ as $n \to \infty$ for some sequence $t_n = o(n)$.

This condition requires some sort of mixing by demanding that an escape at time 0 is an event which gets more and more independent from an event corresponding no escapes during some period, as the time gap between these two events gets larger and larger. It is in this condition that the main advantage of our approach to prove the EI lies. This is because in all the approaches we are aware of (see for example [L83, O87, HHL88, LN89, CHM91]), some condition like $D(u_n)$ from Leadbetter [L73] is used. Some are slightly weaker like AIM (u_n) from [O87] or $\Delta(u_n)$ in [LR98], but they all have a uniform bound on the "independence" of two events separated by a time gap, where both these events may depend on an arbitrarily large number of r.v.s of the sequence X_0, X_1, \ldots In contrast, in our condition $D^p(u_n)$, the first event $Q_{p,0}(u_n)$ depends only on the r.v.s X_0 and X_p and this proves to be crucial when applying it to stochastic processes arising from dynamical systems as explained in Subsection 3.3.

Assuming $D^p(u_n)$ holds let $(k_n)_{n\in\mathbb{N}}$ be a sequence of integers such that

$$k_n \to \infty$$
 and $k_n t_n = o(n)$. (2.6)

Condition $(D'_p(u_n))$. We say that $D'_p(u_n)$ holds for the sequence X_0, X_1, X_2, \ldots if there exists a sequence $\{k_n\}_{n\in\mathbb{N}}$ satisfying (2.6) and such that

$$\lim_{n \to \infty} n \sum_{j=1}^{[n/k_n]} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) = 0.$$
(2.7)

This last condition is very similar to Leadbetter's $D'(u_n)$ from [L83], except that instead of preventing the clustering of exceedances it prevents the clustering of escapes by requiring that they should appear scattered fairly evenly through the time interval from 0 to n-1.

Our main result is this section is the following:

Theorem 1. Let $(u_n)_{n\in\mathbb{N}}$ be such that $n\mathbb{P}(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Consider a stationary stochastic process X_0, X_1, X_2, \ldots satisfying $SP_{p,\theta}(u_n)$ for some $p \in \mathbb{N}$, and $\theta \in (0, 1)$. Assume further that conditions $D^p(u_n)$ and $D'_p(u_n)$ hold. Then

$$\lim_{n \to \infty} \mathbb{P}(M_n \le u_n) = \lim_{n \to \infty} \mathbb{P}(\mathscr{Q}_{p,0,n}(u_n)) = e^{-\theta\tau}.$$
(2.8)

Theorem 1 and in particular formula (2.8) allow us to paint the following picture: In (1.9) we are concerned with the distribution of M_n given that an exceedence of level u_n has occurred at time 0. The underlying periodic phenomena and in particular the capture incidents are responsible for the appearance of the Dirac term in (1.9) for the distribution of RTS with a weight given by the probability of a capture occurring, given that an exceedance has occurred, which is $1 - \theta$. On the other hand, the escapes are responsible for the appearance of the exponential term in (1.9), again with a weight given by the probability of an escape occurring, given that an exceedance has occurred, which is θ . However, the distribution of M_n , where we assume nothing about exceedences at time 0, is equal to the one of the HTS, which, as can be seen in (1.8), only sees the exponential term or in other words the escape component. Formula (2.8) is then saying that computing the distribution of M_n can be reduced to computing the distribution of the escapes.

Remark 2. If we enrich the process and the statistics by considering either multiple returns or multiple exceedances we can study Exceedance Point Processes or Hitting Times Point Processes as in [FFT10, Section 3]. One can then adapt the proof of Theorem 1 to obtain a result analogous to [FFT10, Theorem 5], thus obtaining the convergence of the Exceedance Point Process to a compound Poisson Process. In order to achieve this, we would have to change $D^p(u_n)$ in the the same way that $D_2(u_n)$ was changed to $D_3(u_n)$ in [FFT10]. Essentially, we would need to replace exceedances in $D_3(u_n)$ from [FFT10] by escapes, and adapt the proof of [FFT10, Theorem 5] to cope with this exchange of roles between exceedances and escapes.

We start the proof of Theorem 1 with the following two simple observations.

Lemma 2.1. For any integers $p, l \in \mathbb{N}, s \in \mathbb{N} \cup \{0\}$ and real numbers 0 < u < v we have

$$\sum_{j=s}^{s+\ell-1} \mathbb{P}(Q_{p,j}(u)) \ge \mathbb{P}(\mathscr{Q}_{p,s,\ell}^c(u)) \ge \sum_{j=s}^{s+\ell-1} \mathbb{P}(Q_{p,j}(u)) - \sum_{j=s}^{s+\ell-1} \sum_{i>j}^{s+\ell-1} \mathbb{P}(Q_{p,j}(u) \cap Q_{p,i}(u))$$

Proof. This is a straightforward consequence of the formula for the probability of a multiple union on events. See for example first theorem of Chapter 4 in [F50]. \Box

Lemma 2.2. Assume that t, r, m, ℓ, s are nonnegative integers and u > 0 is a positive real number. Then, we have

$$0 \le \mathbb{P}(\mathscr{Q}_{p,s,\ell}(u)) - \mathbb{P}(\mathscr{Q}_{p,s,\ell+t}(u)) \le t \cdot \mathbb{P}(Q_{p,0}(u))$$
(2.9)

and

$$\left| \mathbb{P}(\mathscr{Q}_{p,0,s+t+m}(u)) - \mathbb{P}(\mathscr{Q}_{p,0,m}(u)) + \sum_{j=0}^{s-1} \mathbb{P}\left(Q_{p,0}(u) \cap \mathscr{Q}_{p,s+t-j,m}(u)\right) \right|$$

$$\leq 2s \sum_{j=1}^{s-1} \mathbb{P}(Q_{p,0}(u) \cap Q_{p,j}(u)\}) + t \mathbb{P}(Q_{p,0}(u)).$$
(2.10)

The proof of this lemma can easily be done by following the proof of [FF08a, Lemma 3.2] or [Co01, Proposition 3.2] with minor adjustments.

Proof of Theorem 1. We split the proof in two parts. The first is devoted to showing the second equality in (2.8), leaving the first equality for the second part of the proof.

Let $\ell = \ell_n = [n/k_n]$ and $k = k_n$ be as in Condition $D'_p(u_n)$. We begin by replacing $\mathbb{P}(\mathscr{Q}_{0,n}(u_n))$ by $\mathbb{P}(\mathscr{Q}_{0,k(\ell+t)}(u_n))$ for some t > 1. By (2.9) of Lemma 2.2 and the fact that $Q_{p,0}(u_n) \subset \{X_0 > u_n\}$, we have

$$\left|\mathbb{P}(\mathscr{Q}_{0,n}(u_n)) - \mathbb{P}(\mathscr{Q}_{0,k(\ell+t)}(u_n))\right| \le kt\mathbb{P}(X_0 > u_n).$$
(2.11)

We now estimate recursively $\mathbb{P}(\mathscr{Q}_{p,0,i(\ell+t)}(u_n))$ for $i = 0, \ldots, k$. Using (2.10) of Lemma 2.2 and stationarity, we have for any $1 \leq i \leq k$

$$\left|\mathbb{P}(\mathscr{Q}_{p,0,i(\ell+t)}(u_n)) - \left(1 - \ell \mathbb{P}(Q_{p,0}(u_n))\right)\mathbb{P}(\mathscr{Q}_{p,0,(i-1)(\ell+t)}(u_n))\right| \le \Gamma_{n,i},$$

where

$$\Gamma_{n,i} = \left| \ell \mathbb{P}(Q_{p,0}(u_n)) \mathbb{P}(\mathscr{Q}_{p,0,(i-1)(\ell+t)}(u_n)) - \sum_{j=0}^{\ell-1} \mathbb{P}\left(Q_{p,j}(u_n) \cap \mathscr{Q}_{p,\ell+t,(i-1)(\ell+t)}(u_n)\right) \right| \\ + t \mathbb{P}(X_0 > u_n) + 2\ell \sum_{j=1}^{\ell-1} \mathbb{P}\left(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)\right).$$

Using stationarity, $D^p(u_n)$ and, in particular, that $\gamma(n,t)$ is nonincreasing in t for each n we conclude

$$\Gamma_{n,i} \leq \sum_{j=0}^{\ell-1} \left| \mathbb{P}(Q_{p,0}(u_n)) \mathbb{P}(\mathscr{Q}_{p,0,(i-1)(\ell+t)}(u_n)) - \mathbb{P}\left(Q_{p,0}(u_n) \cap \mathscr{Q}_{p,\ell+t-j,(i-1)(\ell+t)}(u_n)\right) \right| \\
+ t \mathbb{P}(X_0 > u_n) + 2\ell \sum_{j=1}^{\ell-1} \mathbb{P}\left(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)\right) \\
\leq \ell \gamma(n,t) + t \mathbb{P}(X_0 > u_n) + 2\ell \sum_{j=1}^{\ell-1} \mathbb{P}\left(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)\right).$$

Define $\Upsilon_n = \ell \gamma(n,t) + t \mathbb{P}(X_0 > u_n) + 2\ell \sum_{j=1}^{\ell-1} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n))$. Then for every $1 < i \leq k$ we have

$$\left|\mathbb{P}(\mathscr{Q}_{p,0,i(\ell+t)}(u_n)) - \left(1 - \ell \mathbb{P}(Q_{p,0}(u_n))\right) \mathbb{P}(\mathscr{Q}_{p,0,(i-1)(\ell+t)}(u_n))\right| \leq \Upsilon_n$$

and for i = 1

$$\left|\mathbb{P}(\mathscr{Q}_{p,0,\ell+t}(u_n)) - \left(1 - \ell \mathbb{P}(Q_{p,0}(u_n))\right)\right| \leq \Upsilon_n$$

Since $n\mathbb{P}(X > u_n) \to \tau$, as $n \to \infty$, by (2.5), it follows that $n\mathbb{P}(Q_{p,0}(u_n)) \to \theta\tau$. Hence, if k and n are large enough we have $\ell \mathbb{P}(Q_{p,0}(u_n)) < 2$, which implies that $|1 - \ell \mathbb{P}(Q_{p,0}(u_n))| < 1$. Then, a simple inductive argument allows us to conclude

$$\left| \mathbb{P}(\mathscr{Q}_{p,0,k(\ell+t)}(u_n)) - \left(1 - \ell \mathbb{P}(Q_{p,0}(u_n))\right)^k \right| \le k \Upsilon_n.$$

Recalling (2.11), we have

$$\left|\mathbb{P}(\mathscr{Q}_{p,0,n}(u_n)) - \left(1 - \ell \mathbb{P}(Q_{p,0}(u_n))\right)^k\right| \le kt \mathbb{P}(Q_{p,0}(u_n)) + k\Upsilon_n.$$

Since by (2.5), it follows that $n\mathbb{P}(Q_{p,0}(u_n)) \to \theta\tau$, as $n \to \infty$, for some $\tau \ge 0$, we have

$$\lim_{n \to \infty} \left(1 - \left[\frac{n}{k} \right] \mathbb{P}(Q_{p,0}(u_n)) \right)^k = e^{-\theta \tau}.$$

It is now clear that, the second equality in (2.8) holds if

$$\lim_{n \to \infty} (kt \mathbb{P}(Q_{p,0}(u_n)) + k\Upsilon_n = 0,$$

that is

$$\lim_{n \to \infty} 2kt \mathbb{P}(Q_{p,0}(u_n)) + n\gamma(n,t) + 2n \sum_{j=1}^{\ell} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) = 0.$$
(2.12)

Assume that $t = t_n$ where $t_n = o(n)$ is given by Condition $D^p(u_n)$. Then, by (2.6), we have $\lim_{n\to\infty} kt_n \mathbb{P}(Q_{p,0}(u_n)) = 0$, since $n\mathbb{P}(Q_{p,0}(u_n)) \to \theta\tau \ge 0$. Finally, we use $D^p(u_n)$ and $D'_p(u_n)$ to obtain that the two remaining terms in (2.12) also go to 0.

Now, we need to show that the first equality in (2.8) holds. First observe that

$$\{M_n \le u_n\} \subset \mathscr{Q}_{p,0,n}(u_n).$$

Next, note that if $\mathscr{Q}_{p,0,n}(u_n) \setminus \{M_n \leq u_n\}$ occurs, then we may define $i = \inf\{j \in \{0, 1, \ldots, n-1\} : X_j > u_n\}$ and $s_i = [\frac{n-1-i}{p}]$. But since $\mathscr{Q}_{p,0,n}(u_n)$ does occur, then for all $j = 1, \ldots, s_i$ we must have $X_{i+jp} > u_n$, otherwise, there would exist $j_i = \min\{j \in \{1, \ldots, s_i\} : X_j \leq u_n\}$ and $Q_{p,i+(j_i-1)p}(u_n)$ would occur, which contradicts the the occurrence of $\mathscr{Q}_{p,0,n}(u_n)$. This means that

$$\mathscr{Q}_{p,0,n}(u_n) \setminus \{M_n \le u_n\} \subset \bigcup_{i=0}^{n-1} \{X_i > u_n, X_{i+p} > u_n, \dots, X_{i+s_ip} > u_n\}.$$

It follows by $SP_{p,\theta}(u_n)$ and stationarity that

$$\mathbb{P}(\mathscr{Q}_{p,0,n}(u_n)) - \mathbb{P}(\{M_n \le u_n\}) \le \sum_{i=0}^{n-1} \mathbb{P}(X_i > u_n, X_{i+p} > u_n, \dots, X_{i+s_ip} > u_n)$$
$$\le p \sum_{i=0}^{[n/p]} \mathbb{P}(X_0 > u_n, X_p > u_n, X_{2p} > u_n, \dots, X_{ip} > u_n) \xrightarrow[n \to \infty]{} 0.$$

2.2. Existence of an EI due to multiple underlying periodic phenomena. It may happen that the escapes themselves form clusters which means that $D'_p(u_n)$ does not hold. This occurs if, for example, for some $1 \leq j \leq [n/k_n]$ we have $n\mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) \rightarrow \alpha > 0$. Let p_2 be the smallest such j. Then, since $\mathbb{P}(Q_{p,0}(u_n))/n \sim \theta \tau$, we have that (2.1) holds if we replace exceedances by escapes and p by p_2 . Therefore there is a second underlying periodic phenomena which leads to the notion of escapes of second order. This motivates the introduction of similar conditions to $SP_{p,\theta}$, $D^p(u_n)$, $D'_p(u_n)$, where the role of the exceedances is replaced by escapes, in order to obtain a statement like Theorem 1, where the distribution of the maximum would be equal to the distribution of these escapes of second order. Since it may also happen that these escapes of second order also form clusters, we may have to repeat the process all over again. Hence, we establish a hierarchy of escapes in the following way.

Given the sequences $(p_i)_{i\in\mathbb{N}}$ and $(\theta_i)_{i\in\mathbb{N}}$, with $p_i \in \mathbb{N}$ and $\theta_i \in (0, 1)$ for all $i \in \mathbb{N}$, let $\mathbf{p}_i = (p_1, p_2, \dots, p_i), \ \Theta_i = (\theta_1, \theta_2, \dots, \theta_i)$. For each $j \in \mathbb{N}$ and $u \in \mathbb{R}$, assuming that $Q_{\mathbf{p}_{i-1},j}^{(i-1)}(u)$ is already defined we define the *escape of order i* as

$$Q_{\mathbf{p}_{i},j}^{(i)}(u) = Q_{\mathbf{p}_{i-1},j}^{(i-1)}(u) \cap \left(Q_{\mathbf{p}_{i-1},j+p_{i}}^{(i-1)}(u)\right)^{c}.$$

We set $Q_{\mathbf{p}_{1},j}^{(1)}(u) = Q_{p_{1},j}(u)$ and in the case i = 0 we can consider that $Q_{0,j}^{(0)}(u) = \{X_{j} > u\}$. For $i, s, \ell \in \mathbb{N} \cup \{0\}$, let $\mathscr{Q}_{\mathbf{p}_{i},s,\ell}^{(i)}(u) = \bigcap_{j=s}^{s+\ell-1} \left(Q_{\mathbf{p}_{i},j}^{(i)}(u)\right)^{c}$. Now we restate conditions $SP_{p,\theta}$, $D^{p}(u_{n}), D'_{p}(u_{n})$ with respect to the escapes of order $i \in \mathbb{N}$.

Condition $(SP_{\mathbf{p}_i,\Theta_i}^{(i)}(u_n))$. We say that X_0, X_1, X_2, \ldots satisfies condition $SP_{\mathbf{p}_i,\Theta_i}^{(i)}(u_n)$, for $\mathbf{p}_i \in \mathbb{N}^i$ and $\Theta_i \in (0,1)^i$ defined as above, if

$$\lim_{n \to \infty} \sup_{1 \le j < p_i} \mathbb{P}\left(Q_{\mathbf{p}_{i-1},j}^{(i-1)}(u_n) \middle| Q_{\mathbf{p}_{i-1},0}^{(i-1)}(u_n)\right) = 0 \text{ and} \\ \lim_{n \to \infty} \mathbb{P}\left(Q_{\mathbf{p}_{i-1},p_i}^{(i-1)}(u_n) \middle| Q_{\mathbf{p}_{i-1},0}^{(i-1)}(u_n)\right) \to (1 - \theta_i)$$

$$(2.13)$$

and moreover

$$\lim_{n \to \infty} \sum_{j=0}^{\left[\frac{n-1}{p_i}\right]} \mathbb{P}\left(Q_{\mathbf{p}_{i-1},0}^{(i-1)}(u_n), Q_{\mathbf{p}_{i-1},p_i}^{(i-1)}(u_n), Q_{\mathbf{p}_{i-1},2p_i}^{(i-1)}(u_n), \dots, Q_{\mathbf{p}_{i-1},jp_i}^{(i-1)}(u_n)\right) = 0.$$
(2.14)

Condition $(D^{\mathbf{p}_i}(u_n))$. We say that $D^{\mathbf{p}_i}(u_n)$ holds for the sequence X_0, X_1, X_2, \ldots if for any integers ℓ, t and n

$$\left| \mathbb{P}\left(Q_{\mathbf{p}_{i},0}^{(i)}(u_{n}) \cap \mathscr{Q}_{\mathbf{p}_{i},t,\ell}^{(i)}(u_{n}) \right) - \mathbb{P}\left(Q_{\mathbf{p}_{i},j}^{(i)}(u_{n}) \right) \mathbb{P}\left(\mathscr{Q}_{\mathbf{p}_{i},0,\ell}^{(i)}(u_{n}) \right) \right| \leq \gamma(n,t),$$

where $\gamma(n,t)$ is nonincreasing in t for each n and $n\gamma(n,t_n) \to 0$ as $n \to \infty$ for some sequence $t_n = o(n)$.

Note that, as before, the first event $Q_{\mathbf{p}_i,0}^{(i)}(u_n)$ depends only on a finite number of r.v., namely, $X_0, X_{p_1}, X_{p_1+p_2}, \ldots, X_{p_1+p_2+\cdots+p_i}$.

Condition $(D'_{\mathbf{p}_i}(u_n))$. We say that $D'_{\mathbf{p}_i}(u_n)$ holds for the sequence X_0, X_1, X_2, \ldots if there exists a sequence $\{k_n\}_{n\in\mathbb{N}}$ satisfying (2.6) and such that

$$\lim_{n \to \infty} n \sum_{j=1}^{\lfloor n/k_n \rfloor} \mathbb{P}\left(Q_{\mathbf{p}_i,0}^{(i)}(u_n) \cap Q_{\mathbf{p}_i,j}^{(i)}(u_n)\right) = 0.$$

Observe that condition $D'_{\mathbf{p}_i}(u_n)$ gets weaker and weaker as *i* increases, which means that every time a new underlying periodic phenomenon is found, there is a higher chance that escapes of the next order satisfy D'.

The next result generalises Theorem 1, which corresponds exactly to the case i = 1, to the case of higher order escapes. We stated these theorems separately since Theorem 1 contains the essential ideas required for Theorem 2 and, moreover, shows the influence of the periodic behaviour in a more transparent way.

Theorem 2. Let $(u_n)_{n\in\mathbb{N}}$ be such that $n\mathbb{P}(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Consider a stationary stochastic process X_0, X_1, X_2, \ldots satisfying conditions $SP_{\mathbf{p}_j,\Theta_j}^{(j)}(u_n)$ for all $1 \le j \le i$. Assume further that conditions $D^{\mathbf{p}_i}(u_n)$ and $D'_{\mathbf{p}_i}(u_n)$ hold. Then

$$\lim_{n \to \infty} \mathbb{P}(M_n \le u_n) = \lim_{n \to \infty} \mathbb{P}\left(\mathscr{Q}_{p,0,n}^{(1)}(u_n)\right) = \dots = \lim_{n \to \infty} \mathbb{P}\left(\mathscr{Q}_{p,0,n}^{(i)}(u_n)\right) = e^{-\theta\tau}, \quad (2.15)$$

where $\theta = \theta_1 \theta_2 \cdots \theta_i$.

We notice that in the particular case i = 2 then θ_2 corresponds to the upcrossings index η in [F06].

Proof. The proof of the last equality in (2.15) is basically done as in Theorem 1 simply by replacing everything by its corresponding i version.

The proof of the *j*-th equality, with $1 \leq j \leq i$, in (2.15) follows as the proof of the first equality in (2.8) except that instead of (2.2) we use (2.14) of the corresponding condition $\operatorname{SP}_{\mathbf{p}_{j},\Theta_{j}}^{(j)}$.

Regarding the formula for the extremal index θ observe that it follows by an easy induction argument from the fact that $SP_{\mathbf{p}_{i},\Theta_{j}}^{(j)}(u_{n})$ holds for all $1 \leq j \leq i$. In fact, as always, let

 $(u_n)_{n\in\mathbb{N}}$ be a sequence of levels such that $n(1 - F(u_n)) = n\mathbb{P}(X_0 > u_n) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Assuming by induction that $n\mathbb{P}(Q_{\mathbf{p}_{j-1},0}^{(j-1)}(u_n)) \to \theta_1\theta_2\dots\theta_{j-1}\tau$, as $n \to \infty$, by (2.13) of condition $\mathrm{SP}_{\mathbf{p}_j,\Theta_j}^{(j)}(u_n)$ we must have:

$$n\mathbb{P}\left(Q_{\mathbf{p}_{j},0}^{(j)}(u_{n})\right) = n\mathbb{P}\left(Q_{\mathbf{p}_{j-1},0}^{(j-1)}(u_{n}) \cap \left(Q_{\mathbf{p}_{j-1},p_{j}}^{(j-1)}(u_{n})\right)^{c}\right)$$
$$= n\mathbb{P}\left(Q_{\mathbf{p}_{j-1},0}^{(j-1)}(u_{n})\right)\mathbb{P}\left(\left(Q_{\mathbf{p}_{j-1},p_{j}}^{(j-1)}(u_{n})\right)^{c} \middle| Q_{\mathbf{p}_{j-1},0}^{(j-1)}(u_{n})\right) \xrightarrow[n \to \infty]{} \theta_{1}\theta_{2} \dots \theta_{j-1}\tau\theta_{j}.$$

Since, by (2.5), we have $n\mathbb{P}(Q_{\mathbf{p}_1,0}^{(1)}(u_n)) \to \theta_1 \tau$, as $n \to \infty$, the result follows at once. \Box

When comparing Theorem 2 with similar results in the literature, particularly the most similar in [LN89, CHM91] and [F06], we highlight the following advantages: the interpretation of the EI is explicitly motivated by the existence of underlying periodic phenomena; and the fact that our conditions are weaker, especially because our condition $D^{\mathbf{p}_i}(u_n)$ is much weaker than $D(u_n)$. In fact, as we explain in greater depth in Section 3.3, if we had to check $D(u_n)$ for stochastic processes arising from dynamical systems we could only get HTS/RTS for cylinders (see definition in Section 5) instead of balls, which we do obtain in Corollaries 4 and 6. In terms of EVL, this means that we would get cylinder EVL with convergence only for certain subsequences ω_n of time $n \in \mathbb{N}$, which contrasts with our results in Theorems 3 and 5.

Regarding applications of Theorems 1 and 2, we mention that for the examples of stochastic processes that besides $D(u_n)$ also satisfy $D''(u_n)$ from [LN89], then Theorem 1 can be used to prove the existence of an EI. While for the examples we know of stochastic processes that, for some $k \ge 2$, satisfy $D^{(k)}(u_n)$ from [CHM91] instead, then eventually Theorem 2 can be used for the same purpose.

Besides the applications to stochastic processes coming from dynamical systems given in Section 3, for which $MP_{p,\theta}$ is shown to hold, we give two examples in the appendix, one of Maximum Moving Average sequences and one of an Autoregressive process, to which the results of this section also apply. While these examples are not novel, they illustrate how to check conditions $SP_{p,\theta}$, $MP_{p,\theta}$, $D^p(u_n)$, $D'_p(u_n)$, $SP_{\mathbf{p}_i,\Theta_i}^{(i)}$, $D^{\mathbf{p}_i}(u_n)$ and $D'_{\mathbf{p}_i}(u_n)$ in different, more classical, settings. The Maximum Moving Average in Appendix A satisfies $SP_{p,\theta}$ with p = 2, and $\theta = 1/2$, while the one in Appendix B satisfies $SP_{\mathbf{p}_i,\Theta_i}^{(i)}$, with i = 1, 2, $\mathbf{p}_2 = (p_1, p_2) = (1, 3)$, $\Theta_2 = (2/3, 1/2)$. The Autoregressive process of order 1 (AR(1)), introduced in [C81] and considered in Appendix C, is shown to satisfy $MP_{p,\theta}$ with p = 1and $\theta = 1 - 1/r$.

3. The general theory for sequences generated by dynamical systems

In this section, we set out the general theory of the extremal index in the context of a discrete time dynamical system $(\mathcal{X}, \mathcal{B}, \mu, f)$, where \mathcal{X} is a Riemannian manifold, \mathcal{B} is the Borel σ -algebra, $f : \mathcal{X} \to \mathcal{X}$ is a measurable map and μ an f-invariant probability measure.

We will initially show that $MP_{p,\theta}(u_n)$ can be proved for quite general systems, and later, in Section 4, give specific examples where we can also prove $D_p(u_n)$ and $D'_p(u_n)$ and thus apply Theorem 1.

We consider a Riemannian metric on \mathcal{X} that we denote by 'dist' and for any $\zeta \in \mathcal{X}$ we define the ball of radius $\delta > 0$ around ζ , as $B_{\delta}(\zeta) = \{x \in \mathcal{X} : \operatorname{dist}(x,\zeta) < \delta\}$. Let Leb denote a normalised volume form defined on \mathcal{B} that we call Lebesgue measure.

We suppose that the stochastic processes X_0, X_1, X_2, \ldots defined by (1.6) are such that the r.v. $\varphi : \mathcal{X} \to \mathbb{R} \cup \{\pm \infty\}$ achieves a global maximum at $\zeta \in \mathcal{X}$ (we allow $\varphi(\zeta) = +\infty$).

In order to study the statistical properties of the system, the invariant probability measure μ and its properties play a crucial role. First, we want the measure to provide relevant information about the system. This is achieved, for example, by requiring that the measure is 'physical' or even more generally an 'equilibrium state'. We will emphasise the first kind of measures here due to their importance in the study of the statistical properties of dynamical systems.

A measure μ is said to be *physical* if the Lebesgue measure of the set of points \mathcal{U} (called basin of μ), for which the law of large numbers holds for any stochastic process defined as in (1.6) for any continuous r.v. $\psi : \mathcal{X} \to \mathbb{R}$, is positive. In other words, if the set of points x such that

$$\frac{1}{n}\sum_{i=0}^{n-1}\psi(f^i(x)) \to \int \psi d\mu \tag{3.1}$$

has positive Lebesgue measure. For example, μ is a physical measure if it is absolutely continuous with respect to Lebesgue, in which case we write $\mu \ll \text{Leb}$, and *ergodic*, which simply means that (3.1) holds μ -a.e. Note that these measures do provide a nice picture of the statistical behaviour of the system, since describing how the time averages $\frac{1}{n}\sum_{i=0}^{n-1}\psi(f^i(x))$ of any continuous function ψ behave, reduces to compute the spacial average $\int \psi d\mu$ simply by integrating ψ against the measure μ . Moreover, this works on a "physically observable" set \mathcal{U} of positive Lebesgue measure.

More generally, we can study the statistical properties of a system through the following class of measures, known as equilibrium states. For good introductions to this topic see for example [Bo75, W82, K98].

Let $f: \mathcal{X} \to \mathcal{X}$ be a measurable function as above, and define

 $\mathcal{M}_f := \{ f - \text{invariant Borel probability measures on } \mathcal{X} \}.$

I.e., for $\mu \in \mathcal{M}_f$, $\mu(\mathcal{X}) = 1$ and for any Borel measurable set A, $\mu(f^{-1}(A)) = \mu(A)$. Then for a measurable potential $\phi : \mathcal{X} \to \mathbb{R}$, we define the *pressure* of (\mathcal{X}, f, ϕ) to be

$$P(\phi) := \sup_{\mu \in \mathcal{M}_f} \left\{ h(\mu) + \int \phi \ d\mu : -\int \phi \ d\mu < \infty \right\},$$

where $h(\mu)$ denotes the metric entropy of the measure μ , see [W82] for details. If, for $\mu \in \mathcal{M}_f$, $h(\mu) + \int \phi \ d\mu = P(\phi)$ then we say that μ is an *equilibrium state* for (\mathcal{X}, f, ϕ) .

The absolutely continuous measures given above can often be shown to be particular examples of equilibrium states. This is explained in more depth in Section 3.2.

3.1. Measures absolutely continuous with respect to Lebesgue. In this subsection, we assume that the measure μ is absolutely continuous with respect to Lebesgue. Besides, we assume that ζ is a *repelling p-periodic point*, which means that $f^p(\zeta) = \zeta$, f^p is differentiable at ζ and $0 < |\det D(f^{-p})(\zeta)| < 1$. Moreover, we also assume that ζ is a Lebesgue density point with $0 < \frac{d\mu}{d\text{Leb}}(\zeta) < \infty$ and the observable $\varphi : \mathcal{X} \to \mathbb{R} \cup \{+\infty\}$ is of the form

$$\varphi(x) = g(\operatorname{dist}(x,\zeta)), \qquad (3.2)$$

where the function $g : [0, +\infty) \to \mathbb{R} \cup \{+\infty\}$ is such that 0 is a global maximum (g(0)) may be $+\infty$; g is a strictly decreasing bijection $g : V \to W$ in a neighbourhood V of 0; and has one of the following three types of behaviour:

Type 1: there exists some strictly positive function $\hat{\pi}: W \to \mathbb{R}$ such that for all $y \in \mathbb{R}$

$$\lim_{s \to g_1(0)} \frac{g_1^{-1}(s + y\hat{\pi}(s))}{g_1^{-1}(s)} = e^{-y};$$
(3.3)

Type 2: $g_2(0) = +\infty$ and there exists $\beta > 0$ such that for all y > 0

$$\lim_{s \to +\infty} \frac{g_2^{-1}(sy)}{g_2^{-1}(s)} = y^{-\beta};$$
(3.4)

Type 3: $g_3(0) = D < +\infty$ and there exists $\gamma > 0$ such that for all y > 0

$$\lim_{s \to 0} \frac{g_3^{-1}(D - sy)}{g_3^{-1}(D - s)} = y^{\gamma}.$$
(3.5)

Examples of each one of the three types are as follows: $g_1(x) = -\log x$ (in this case (3.3) is easily verified with $\hat{\pi} \equiv 1$), $g_2(x) = x^{-1/\alpha}$ for some $\alpha > 0$ (condition (3.4) is verified with $\beta = \alpha$) and $g_3(x) = D - x^{1/\alpha}$ for some $D \in \mathbb{R}$ and $\alpha > 0$ (condition (3.5) is verified with $\gamma = \alpha$).

Remark 3. Recall that the d.f. F is given by $F(u) = \mu(X_0 \leq u)$ and $u_F = \sup\{y : F(y) < 1\}$. Observe that if at time $j \in \mathbb{N}$ we have an exceedance of the level u (sufficiently large), i.e., $X_j(x) > u$, then we have an entrance of the orbit of x into the ball $B_{g^{-1}(u)}(\zeta)$ of radius $g^{-1}(u)$ around ζ , at time j. This means that the behaviour of the tail of F, *i.e.*, the behaviour of 1 - F(u) as $u \to u_F$ is determined by g^{-1} , if we assume that Lebesgue's Differentiation Theorem holds for ζ , since in that case $1 - F(u) \sim \rho(\zeta)|B_{g^{-1}(u)}(\zeta)|$, where $\rho(\zeta) = \frac{d\mu}{d\text{Leb}}(\zeta)$. From classical Extreme Value Theory we know that the behaviour of the tail determines the limit law for partial maximums of i.i.d. sequences and vice-versa. The above conditions are just the translation in terms of the shape of g^{-1} , of the sufficient and necessary conditions on the tail of F of [LLR83, Theorem 1.6.2], in order to exist a non-degenerate limit distribution for \hat{M}_n .

Recall that X_0, X_1, X_2, \ldots is given by (1.6) for observables of the type (3.2), which means the event $\{X_0 > u\}$ corresponds to a ball centred at ζ . Suppose that $p \in \mathbb{N}$ and consider as before

$$Q_{p,0}^*(u) := \{x : \varphi(x) > u\} \cap f^{-p}(\{x : \varphi(x) > u\}) \text{ and } Q_{p,0}(u) := \{x : \varphi(x) > u\} \setminus Q^*(u).$$

The set $Q_p^*(u) := Q_{p,0}^*(u)$ corresponds to a ball, while $Q_p(u) := Q_{p,0}(u)$ corresponds to an annulus, both centred at ζ . For all $i \in \mathbb{N}$, set

$$Q_p^{*i}(u) := \bigcap_{j=0}^{i-1} Q_{p,j}^*(u).$$

Given the special structure of these dynamically defined stochastic processes, observe that for all $i \in \mathbb{N}$ we have $Q_{p,i}(u) = f^{-i}(Q_p(u)) = \{X_i \in Q_p(u)\}, Q_{p,i}^*(u) = f^{-i}(Q_p^*(u)) = \{X_i \in Q_p^*(u)\}$ and $Q_p^{*i}(u) = \bigcap_{j=0}^i f^{-jp}(\{x : \varphi(x) > u\}).$

We will provide some conditions which guarantee an Extreme Value Law with a given extremal index. We will give some systems which satisfy these conditions in Section 4.

Theorem 3. Suppose that ζ is a repelling periodic point of period p, with $\theta = \theta(\zeta) = 1 - |\det D(f^{-p})(\zeta)| \in (0,1)$. Let $(u_n)_{n \in \mathbb{N}}$ be such that $n\mu(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Assume further that conditions $D^p(u_n)$ and $D'_p(u_n)$ hold. Then

$$\lim_{n \to \infty} \mu(M_n \le u_n) = \lim_{n \to \infty} \mu(\mathscr{Q}_{p,0,n}(u_n)) = e^{-\theta\tau}.$$

Proof. To prove Theorem 3 we only need to show property $MP_{p,\theta}(u_n)$ and apply Theorem 1. Since ζ is a repelling periodic point, by the Mean Value Theorem we have

$$\operatorname{Leb}(Q^*(u)) \sim (1-\theta) \operatorname{Leb}(\{x : \varphi(x) > u\})$$

for u close to u_F . By induction, we get

$$\operatorname{Leb}(Q_p^{*i}(u)) = \operatorname{Leb}\left(\bigcap_{j=0}^i f^{-jp}(\{x : \varphi(x) > u\})\right) \sim (1-\theta)^i \operatorname{Leb}(\{x : \varphi(x) > u\}),$$

for u close to u_F . Consequently, using the fact that ζ is a Lebesgue density point, we have for $i \in \mathbb{N}$,

$$\mu(X_p > u, \dots, X_{ip} > u | X_0 > u) = \frac{\mu(Q_p^{*i}(u))}{\mu(X_0 > u)} \sim \frac{\operatorname{Leb}(Q_p^{*i}(u))}{\operatorname{Leb}(\{x : \phi(x) > u\})} \sim (1 - \theta)^i.$$

So replacing u with $(u_n)_n$, summing over i and letting $n \to \infty$, we have MP_{p, θ}, as required.

The relation between EVL and HTS established in [FFT10] allows us to obtain the following: **Corollary 4.** Suppose that ζ is a repelling periodic point of period p, with $\theta = \theta(\zeta) = 1 - |\det D(f^{-p})(\zeta)| \in (0,1)$. Let $(u_n)_{n \in \mathbb{N}}$ be such that $n\mu(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Assume further that conditions $D^p(u_n)$ and $D'_p(u_n)$ hold. Then we have Hitting Time Statistics to balls at ζ ,

$$\lim_{n \to \infty} \mu \left(r_{B_{\delta_n}(\zeta)} < \frac{t}{\mu(B_{\delta_n}(\zeta))} \right) = 1 - e^{-\theta t}, \tag{3.6}$$

and Return Time Statistics to balls at ζ ,

$$\lim_{n \to \infty} \mu_{B_{\delta_n}(\zeta)} \left(r_{B_{\delta_n}(\zeta)} < \frac{t}{\mu(B_{\delta_n}(\zeta))} \right) = (1 - \theta) + \theta(1 - e^{-\theta t}), \tag{3.7}$$

for all sequences $\delta_n \to 0$, as $n \to \infty$.

Proof. The limit (3.6) is an immediate consequence of Theorem 3 and [FFT10, Theorem 2]. The limit (3.7) derives from (3.6) and the integral formula in [HLV05, Main Theorem] that relates the HTS and RTS distributions.

3.2. Equilibrium states. We gave the notions of pressure and equilibrium states at the beginning of this section. Here we will introduce further notions in order to generalise Theorem 3 to general equilibrium states. A measure m is called a ϕ -conformal measure if $m(\mathcal{X}) = 1$ and if whenever $f: A \to f(A)$ is a bijection, for a Borel set A, then

$$m(f(A)) = \int_A e^{-\phi} \, dm.$$

Note that for example for a smooth map interval map f, Lebesgue measure is ϕ -conformal for $\phi(x) := -\log |Df(x)|$. Moreover, if for example f is a topologically transitive quadratic interval map then as in Ledrappier [Le81], any physical measure μ with $h(\mu) > 0$ is an equilibrium state for ϕ . That this also holds for the even simpler case of piecewise smooth uniformly expanding maps follows from Section 4.1.

We define

$$S_n\phi(x) := \phi(x) + \dots + \phi \circ f^{n-1}(x).$$

In the following proposition we will assume that for a potential ϕ , we have $P(\phi) = 0$. Note that if $P(\phi) = p$ for $p \in (-\infty, \infty)$ then we can replace ϕ by $\phi - p$ to obtain $P(\phi - p) = 0$. Clearly any equilibrium state for ϕ is an equilibrium state for $\phi - p$ and vice versa. Recall that we are assuming that $f : \mathcal{X} \to \mathcal{X}$ is a measurable map of a Riemannian manifold which is differentiable at a periodic point $\zeta \in \mathcal{X}$.

Again, we consider that the stochastic processes X_0, X_1, X_2, \ldots defined by (1.6) are such that the r.v. $\varphi : \mathcal{X} \to \mathbb{R} \cup \{\pm \infty\}$ achieves a global maximum at $\zeta \in \mathcal{X}$. However, in order to still be able of establishing the connection between EVL and HTS in this setting, where the invariant measure may present a more irregular behaviour than when it is absolutely continuous, we need to taylor the observable φ to cope with this lack of regularity as in [FFT10a]. Essentially, this means that we need to replace dist (x, ζ) in (3.2) with $\mu(B_{\operatorname{dist}(x,\zeta)}(\zeta))$. So throughout this section the stochastic processes X_0, X_1, X_2, \ldots defined by (1.6) is such that the r.v. $\varphi : \mathcal{X} \to \mathbb{R} \cup \{\pm \infty\}$ is given by

$$\varphi(x) = g(\mu(B_{\operatorname{dist}(x,\zeta)}(\zeta))), \qquad (3.8)$$

where g is as in Section 3.1.

Since for this application, we would like a sequence $(u_n)_n$ such that $\lim_{n\to\infty} n\mu(\{X_0 > u_n\}) = \tau$, it is useful to assume that the measure μ has some continuity: otherwise, 'jumps' in the size of balls around ζ may prevent us from finding such a sequence. To deal with this issue, in [FFT10a], we defined a function \hbar for small $\eta \ge 0$ and given by

$$\hbar(\eta) = \mu(B_{\eta}(\zeta)). \tag{3.9}$$

We required that \hbar is continuous on η . For example, if \mathcal{X} is an interval and μ a Borel probability with no atoms, *i.e.*, points with positive μ measure, then \hbar is continuous.

Theorem 5. Let $\phi : \mathcal{X} \to \mathbb{R}$ be a continuous potential with $P(\phi) = 0$, a conformal measure m_{ϕ} and an equilibrium state $\mu_{\phi} \ll m_{\phi}$. Suppose that the observable φ is as in (3.8) and that $\zeta \in \mathcal{X}$ is a repelling periodic point of period p such that $0 < \frac{d\mu_{\phi}}{dm_{\phi}}(\zeta) < \infty$ and for all u sufficiently close to u_F ,

$$\sum_{j=0}^{\infty} \sup\{|S_p\phi(x) - S_p\phi(\zeta)| : x \in \{\varphi > u\} \cap f^{-jp}(\{\varphi > u\})\} < \infty.$$
(3.10)

Let $(u_n)_{n\in\mathbb{N}}$ be such that $n\mu(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Assume further that conditions $D^p(u_n)$ and $D'_p(u_n)$ hold. Then

$$\lim_{n \to \infty} \mu(M_n \le u_n) = \lim_{n \to \infty} \mu(\mathscr{Q}_{p,0,n}(u_n)) = e^{-\theta\tau},$$

where $\theta = 1 - e^{S_p \phi(\zeta)}$.

Condition (3.10) is to control the distortion of ϕ on small scales. It follows for example from a Hölder condition on ϕ .

Remark 4. If $\zeta \in \mathcal{X}$ is a repelling *p*-periodic point in the support of a ϕ -conformal measure m_{ϕ} as in Theorem 5, then $S_p\phi(\zeta)$ must be non-positive. We can show this by taking a very small set A around ζ such that $f^p : A \to f^p(A)$ is a bijection and such that $A \subset f^p(A)$, then

$$m_{\phi}(f^{p}(A)) = \int_{A} e^{-S_{p}\phi} dm_{\phi} \approx e^{-S_{p}\phi(\zeta)} m_{\phi}(A).$$

So if $S_p \phi(\zeta) > 0$ then $m_{\phi}(f^p(A)) < m_{\phi}(A)$, which is impossible.

Using the theory developed in [FFT10a], an analogue of Corollary 4 holds for equilibrium states, namely:

Corollary 6. Under the conditions of Theorem 5 we have Hitting Time Statistics to balls at ζ ,

$$\lim_{n \to \infty} \mu \left(r_{B_{\delta_n}(\zeta)} < \frac{t}{\mu(B_{\delta_n}(\zeta))} \right) = 1 - e^{-\theta t},$$

and Return Time Statistics to balls at ζ ,

$$\lim_{n \to \infty} \mu_{B_{\delta_n}(\zeta)} \left(r_{B_{\delta_n}(\zeta)} < \frac{t}{\mu(B_{\delta_n}(\zeta))} \right) = (1 - \theta) + \theta (1 - e^{-\theta t}),$$

for all sequences $\delta_n \to 0$, as $n \to \infty$.

The proof of Theorem 5 follows almost immediately from the following lemma.

Lemma 3.1. Let $\phi : \mathcal{X} \to \mathbb{R}$ be a potential which is continuous at ζ , $f(\zeta), \ldots, f^{p-1}(\zeta)$ as in Theorem 5 with $P(\phi) = 0$. If ϕ has a conformal measure m_{ϕ} then $\frac{m_{\phi}(Q_p^*(u))}{m_{\phi}(\{\phi > u\})} \to e^{S_p\phi(\zeta)}$ as $u \to u_F$. Moreover, if (3.10) holds then

$$m_{\phi}(Q_{p}^{*i}(u)) \sim m_{\phi}(\{\varphi > u\})e^{S_{ip}\phi(\zeta)} = m_{\phi}(\{\varphi > u\})(1-\theta)^{i}$$

for $\theta = 1 - e^{S_p \phi(\zeta)}$.

Proof. The continuity of ϕ implies that for any $\varepsilon > 0$, for all u sufficiently close to u_F , $e^{|S_p\phi(x)-S_p\phi(y)|} < (1+\varepsilon)$ for $x, y \in \{\varphi > u\}$. Using this and conformality, for u close enough to u_F ,

$$\frac{m_{\phi}\left(Q_{p}^{*}(u)\right)}{m_{\phi}(\{\varphi > u\})} = \frac{m_{\phi}\left(\{\varphi > u\} \cap f^{-p}(\{\varphi > u\})\right)}{m_{\phi}(\{\varphi > u\})} \sim \frac{m_{\phi}\left(f^{p}\{\varphi > u\}\right) \cap \{\varphi > u\})}{m_{\phi}(f^{p}\{\varphi > u\})} = \frac{m_{\phi}\left(\{\varphi > u\}\right)}{m_{\phi}(f^{p}\{\varphi > u\})} = \frac{m_{\phi}\left(\{\varphi > u\}\right)}{\int_{\{\varphi > u\}} e^{-S_{p}\phi} dm_{\phi}} \sim e^{S_{p}\phi(\zeta)},$$

proving the first part of the lemma.

For the second part of the lemma, note that (3.10) implies that for any $\varepsilon > 0$ we can choose u so close to u_F that

$$\sum_{j=0}^{\infty} \sup\{|S_p\phi(x) - S_p\phi(\zeta)| : x \in f^{-jp}(\{\varphi > u\})\} < \varepsilon.$$

So as in the proof of Theorem 3, we inductively obtain

$$m_{\phi}(Q_p^{*i}(u)) = m_{\phi}\left(\bigcap_{j=0}^{i} f^{-jp}(\{\varphi > u\})\right) \sim e^{S_{ip}\phi(\zeta)}m_{\phi}(\{\varphi > u\})$$

as required.

Proof of Theorem 5. The proof is reduced to apply Theorem 1 after checking that $MP_{p,\theta}(u_n)$ holds with $\theta = 1 - e^{S_p\phi(\zeta)}$ for any sequence $(u_n)_n$ with $u_n \to u_F$ as $n \to \infty$, which follows using Lemma 3.1 and the same ideas as those in the proof of Theorem 3.

Before giving specific examples of dynamical systems satisfying $D_p(u_n)$ and $D'_p(u_n)$, in the next subsection we discuss general conditions which imply those conditions.

3.3. On the roles of $D^p(u_n)$ and $D'_p(u_n)$ for stochastic processes arising from dynamical systems. Theorems 3 and 5 and Corollaries 4 and 6 assert that the existence of limiting laws of rare events with an extremal index for stochastic processes arising from dynamical systems as in (1.6) for observables given by (3.2) or (3.8) centred at repelling periodic points depends on the good mixing properties of the system both at long range $(D^p(u_n))$ and short range $(D'_p(u_n))$.

In general terms the Condition $D^p(u_n)$ follows from sufficiently fast (e.g. polynomial) decay of correlations of the dynamical system. This is where $D^p(u_n)$ are seen to be much more useful then Leadbetter's $D(u_n)$. While $D(u_n)$ usually follows only from strong uniform mixing, like α -mixing (see [B05] for definition), and even then only at certain subsequences, which means most of the time the final result holds only for cylinders, $D^p(u_n)$ follows from decay of correlations which is much weaker and allows to obtain the result for balls, instead.

Just to give an idea of how simple it is to check $D^p(u_n)$ for systems with sufficiently fast decay of correlations, assume that for all $\phi, \psi : M \to \mathbb{R}$ with bounded variation, there are $C, \alpha > 0$ independent of ϕ, ψ and n such that

$$\left| \int \phi \cdot (\psi \circ f^t) d\mu - \int \phi d\mu \int \psi d\mu \right| \le C \operatorname{Var}(\phi) \|\psi\|_{\infty} \varrho(t), \quad \forall t \in \mathbb{N}_0,$$
(3.11)

where $\operatorname{Var}(\phi)$ denotes the total variation of ϕ (see Section 4.1 for more details) and $n\varrho(t_n) \to 0$, as $n \to \infty$ for some $t_n = o(n)$. Take $\phi = \mathbf{1}_{Q_p(u_n)}, \ \psi = \mathbf{1}_{\mathscr{Q}_{p,t,\ell}(u_n)}$, let C' > 0 be such that $\operatorname{Var}(\mathbf{1}_{Q_p(u_n)}) \leq C'$, for all $n \in \mathbb{N}$ and set c = CC'. Then (3.11) implies that Condition $D^p(u_n)$ holds with $\gamma(n,t) = \gamma(t) := c\varrho(t)$ and for the sequence t_n such that $n\varrho(t_n) \to 0$, as $n \to \infty$. Observe that the existence of such C' > 0 derives from the fact that $Q_p(u_n)$ depends only on X_0 and X_p . This is why we cannot apply the same argument to prove $D(u_n)$ directly from Leadbetter, since we would have to take ϕ to be the indicator function over an event depending on an arbitrarily large number of r.v. X_0, X_1, \ldots , which could imply the variation to be unbounded.

It could happen that decay of correlations is only available for Hölder continuous functions against L^{∞} ones, instead. This means that we cannot use immediately the test function $\phi = \mathbf{1}_{Q_p(u_n)}$, as we did before. However, proceeding as in [FFT10, Lemma 6.1], if we use a suitable Hölder approximation one can still prove $D^p(u_n)$.

Rates of decay of correlations are nowadays well known for many chaotic systems. Examples of these include Hyperbolic or uniformly expanding systems as well as the non Hyperbolic or non-uniformly expanding admitting, for example, inducing schemes with a well behaved return time function. In fact, in two remarkable papers Lai-Sang Young showed that the rates of decay of correlations of the original system are intimately connected with the recurrence rates of the respective induced map. This means that, basically, for all the above mentioned systems $D^p(u_n)$ can easily be checked. In short, in order to prove the existence of EVL or HTS/RTS with an extremal index around repelling periodic points for systems with sufficiently fast decay of correlations is reduced to proving $D'_p(u_n)$. Usually, this requires a more closed analysis of the dynamics around the periodic points. Below, we will show that $D'_p(u_n)$ holds for Rychlik maps and for the full quadratic map, which are chaotic systems with exponential decay of correlations, and in this way obtain the existence of an extremal index different of 1. Up to our knowledge, these are the first limiting laws of rare event with an extremal index different of 1 to be proven for balls rather than cylinders.

4. Examples of dynamical systems and observables with extremal index in (0, 1)

We begin this section by introducing a particularly well-behaved class of interval maps and measures for which $MP_{p,\theta}(u_n)$, $D_p(u_n)$ and $D'_p(u_n)$ hold at periodic points. This class is more general than the class of piecewise smooth uniformly hyperbolic interval maps. We then go on to consider a particular example of a non-uniformly hyperbolic dynamical system - the 'full quadratic map'.

4.1. **Rychlik systems.** We will introduce a class of dynamical systems considered by Rychlik in [R83]. This class includes, for example, piecewise C^2 uniformly expanding maps of the unit interval with the relevant physical measures. We first need some definitions.

Definition 4. Given a potential $\psi: Y \to \mathbb{R}$ on an interval Y, the *variation* of ψ is defined as

$$\operatorname{Var}(\psi) := \sup \left\{ \sum_{i=0}^{n-1} |\psi(x_{i+1}) - \psi(x_i)| \right\},\$$

where the supremum is taken over all finite ordered sequences $(x_i)_{i=0}^n \subset Y$.

We use the norm $\|\psi\|_{BV} = \sup |\psi| + \operatorname{Var}(\psi)$, which makes $BV := \{\psi : Y \to \mathbb{R} : \|\psi\|_{BV} < \infty\}$ into a Banach space.

Definition 5 (Rychlik system). (Y, f, ϕ) is a *Rychlik system* if Y is an interval, $f : \bigcup_{i \in \mathbb{N}} Y_i \to Y$ is a function continuous on each Y_i , and $\phi : Y \to [-\infty, \infty)$ is a potential such that

- (1) there exists a neighbourhood $Z \supset Y$ and for each *i* a neighbourhood $Z_i \supset Y_i$ such that $f|_{Y_i}$ can be extended to a diffeomorphism $f: Z_i \xrightarrow{\text{onto}} Z$;
- (2) Var $e^{\phi} < +\infty, \phi = -\infty$ on $Y \setminus \bigcup_i Y_i$ and $P(\phi) = 0;$
- (3) there is a ϕ -conformal measure m_{ϕ} on Y;
- (4) (f, ϕ) is expanding: $\sup \phi(x) < 0$.

Proposition 1. Suppose that (Y, f, ϕ) is a topologically mixing Rychlik system, ϕ is Hölder continuous on each Z_i , and μ is the corresponding equilibrium state. Suppose that ζ is a

repelling periodic point of period p, with $\theta = \theta(\zeta) = 1 - e^{S_p \phi(\zeta)} \in (0, 1)$. Let $(u_n)_{n \in \mathbb{N}}$ be such that $n\mu(X > u_n) = n(1 - F(u_n)) \to \tau$, as $n \to \infty$, for some $\tau \ge 0$. Then $D^p(u_n)$ and $D'_p(u_n)$ hold and we have a Extreme Value Law with extremal index θ .

The idea behind the proof is that it takes a point in $Q_p(u_n)$ at least something of the order $\log n$ iterations to return to $Q_p(u_n)$. Then after $\log n$ iterates, the decay of correlations estimates take over to give $D'_p(u_n)$.

We first give a theorem and a lemma.

Theorem 7 ([R83]). Suppose that (Y, f, ϕ) is a topologically mixing Rychlik system. Then there exists an equilibrium state $\mu_{\phi} = hm_{\phi}$ where $h \in BV$ is strictly positive and m_{ϕ} and μ_{ϕ} are non-atomic and (Y, f, μ_{ϕ}) has exponential decay of correlations, i.e., there exists C > 0 and $\gamma \in (0, 1)$ such that

$$\left| \int \psi \circ f^n \cdot \upsilon \, d\mu_\phi - \int \psi \, d\mu_\phi \int \upsilon \, d\mu_\phi \right| \le C \|\psi\|_{L^1(m_\phi)} \|\upsilon\|_{BV} \gamma^n$$

for any $\phi \in L^1(m_{\phi})$ and $\upsilon \in BV$.

Lemma 4.1. There exists C > 0 such that

$$\mu\left(\left\{x \in Q_p(u_n) : r_{Q_p(u_n)} = k\right\}\right) \le \mu(Q_p(u_n))\left(Ce^{-\beta k} + \mu(Q_p(u_n))\right)$$

Proof. As in the proof of [BSTV03, Theorem 3.2], there exist $\beta, C' > 0$ such that $\mu\left(\left\{x \in Q_p(u_n) : r_{Q_p(u_n)} = k\right\}\right) \leq C'm(Q_p(u_n)) \left\|\mathbf{1}_{Q_p(u_n)} - \mu(Q_p(u_n))\right\|_{BV} e^{-\beta k} + \mu(Q_p(u_n))^2.$

Since we have assumed, as above, that $\frac{d\mu}{dm} \in BV$ and is strictly positive, and since $\|1_{Q_p(u_n)} - \mu(Q_p(u_n))\|_{BV} \leq 5$ there is C > 0 as required. \Box

Proof of Proposition 1. First observe that the non-atomicity of μ , given by Theorem 7, implies that we can indeed find a suitable sequence $(u_n)_n$ as in (1.1), see the discussion around (3.9). Moreover, condition $D^p(u_n)$ follows from Theorem 7 as in Section 3.3.

To prove $D'_p(u_n)$, first let $\overline{U} \ni \zeta$ denote a domain such that $x \in U$ implies $d(f^p(x), \zeta) > d(x, \zeta)$. In order for a point in $Q_p(u_n)$ to return to $Q_p(u_n)$ at time $k \in \mathbb{N}$, there must be some time $\ell \leq k/p$ such that image $f^{\ell p}(Q_p(u_n))$ must have only just escaped from the domain U. Therefore we must have $\mu(f^{(\ell-1)p}(Q_p^*(u_n))) \geq C\mu(U)$ for some C > 0 which depends only on U and ζ . Since $\mu(Q_p(u_n)) \sim \frac{\tau \theta}{n}$, $e^{S_p \phi(\zeta)} \in (0, 1)$ and

$$m_{\phi}(f^{(\ell-1)p}(Q_p(u_n))) = \int_{Q_p(u_n)} e^{S_{(\ell-1)p}\phi} dm_{\phi}$$

we must have ℓ , and therefore k, greater than $B \log n$ for some B > 0, depending on C, U and $\frac{d\mu}{dm}$.

Using this and Lemma 4.1,

$$n \sum_{j=1}^{[n/k_n]} \mathbb{P}(\{X_0 \in Q_p(u_n)\} \cap \{X_j \in Q_p(u_n)\}) \approx n \sum_{j=B \log n}^{[n/k]} \mathbb{P}(\{X_0 \in Q_p(u_n)\} \cap \{X_j \in Q_p(u_n)\})$$

$$\leq n \left([n/k] - B \log n\right) \mu(Q_p(u_n))^2 + n \mu(Q_p(u_n)) e^{-B\beta \log n} \sum_{j=1}^{[n/k] - B \log n} e^{-\beta j}$$

$$\leq \frac{(n \mu(Q_p(u_n)))^2}{k} + C_\beta n \mu(Q_p(u_n)) n^{-B\beta}$$
where $C_\beta := \sum_{j=0}^{\infty} e^{-j\beta}$. Since $n \mu(Q_p(u_n) \to \tau \theta$ as $n \to \infty$ we have for some $C > 0$

$$\lim_{n \to \infty} n \sum_{j=1}^{\lfloor n/k_n \rfloor} \mathbb{P}(\{X_0 \in Q_p(u_n)\}) \cap \{X_j \in Q_p(u_n)\}) \le \lim_{n \to \infty} C \frac{(\tau \theta)^2}{k_n} = 0,$$

as required.

We give a short list of some of the simplest examples of Rychlik systems:

- Given $m \in \{2, 3, \ldots\}$, let $f: x \mapsto mx \mod 1$ and $\phi \equiv -\log m$. Then $m_{\phi} = \mu_{\phi} = \text{Leb}$.
- Let $f: x \mapsto 2x \mod 1$ and for $\alpha \in (0, 1)$, let

$$\phi(x) := \begin{cases} -\log \alpha & \text{if } x \in (0, 1/2) \\ -\log(1-\alpha) & \text{if } x \in (1/2, 1) \end{cases}$$

(and $\phi = -\infty$ elsewhere). Then $m_{\phi} = \mu_{\phi}$ is the $(\alpha, 1 - \alpha)$ -Bernoulli measure on [0, 1].

• Let $f: (0,1] \to (0,1]$ and $\phi: (-\infty,0)$ be defined as $f(x) = 2^k(x-2^{-k})$ and $\phi(x) := -k \log 2$ for $x \in (2^{-k}, 2^{-k+1}]$. Then $m_{\phi} = \mu_{\phi} = \text{Leb}$.

4.2. Existence of an Extremal Index for stochastic processes arising from the full quadratic map. In this section we give an example of a system which is non-uniformly hyperbolic, but nevertheless has an observable yielding an EI in (0, 1).

Let $\mathcal{X} = [-1, 1]$ equipped with the usual metric and Lebesgue measure defined on the Borelean sets of the interval. Let $f : [-1, 1] \rightarrow [-1, 1]$ be given by $f(x) = 1 - 2x^2$. This map is known as the full quadratic map or the quadratic Chebyshev polynomial. It is well known that there exists an invariant measure μ absolutely continuous with respect to Lebesgue with density given by

$$\frac{d\mu}{d\text{Leb}}(x) = \frac{1}{\pi} \frac{1}{\sqrt{(1-x)(1+x)}}.$$

We will consider the fixed point $\zeta = -1$ and the observable $\varphi : [-1,1] \to \mathbb{R}$ given by $\varphi(x) = -x$ which achieves the maximum value 1 at $\zeta = -1$. Notice that φ can be written as $\varphi(x) = g(\operatorname{dist}(x,\zeta))$ as in 3.2, simply by taking $g : [0,\infty) \to \mathbb{R}$ defined by g(y) = 1 - y.

Clearly, since $f'(\zeta) = 4 > 1$, then ζ is a repelling periodic point with p = 1. Let F denote the d.f. of X_0 . We have that for s close to 0, the tail of the d.f. may be written as:

$$1 - F(1 - s) = \int_{-1}^{-1+s} \frac{1}{\pi} \frac{1}{\sqrt{(1 - x)(1 + x)}} dx = \frac{1}{2} + \frac{1}{\pi} \arcsin(-1 + s) \sim \frac{\sqrt{2}}{\pi} \sqrt{s}$$

This implies that the level u_n , which is such that $n(1 - F(u_n)) \to \tau \ge 0$, as $n \to \infty$, may be written as

$$u_n \sim 1 - \frac{(\pi\tau)^2}{2} \frac{1}{n^2}.$$
 (4.1)

This system has exponential decay of correlations which, in light of the discussion in Subsection 3.3, implies that Condition $D^1(u_n)$ holds. In fact, from [KN92, Y92] one has that for all $\phi, \psi : M \to \mathbb{R}$ with bounded variation, there is $C, \alpha > 0$ independent of ϕ, ψ and t such that

$$\left| \int \phi \cdot (\psi \circ f^t) \, d\mu - \int \phi \, d\mu \int \psi \, d\mu \right| \le C \operatorname{Var}(\phi) \|\psi\|_{\infty} \mathrm{e}^{-\alpha t}, \quad \forall t \in \mathbb{N}_0, \tag{4.2}$$

where $\operatorname{Var}(\phi)$ denotes the total variation of ϕ . So, again, we take $\phi = \mathbf{1}_{Q_{1,0}(u_n)}$ and $\psi = \mathbf{1}_{\mathcal{Q}_{1,t,\ell}(u_n)}$, then (4.2) implies that Condition $D^1(u_n)$ holds with $\gamma(n,t) = \gamma(t) := 2Ce^{-\alpha t}$ and for the sequence $t_n = \sqrt{n}$, for example.

To check Condition $D'_1(u_n)$ we have to look at the particular behaviour of the system and estimate the probability of starting in a neighbourhood of ζ and returning in relatively few iterates. The idea is to observe that since the critical orbit ends in ζ , one can just estimate the probability of starting close to the critical point and returning close to it. This can easily be done by using the estimates in [FF08, Section 6] where $D'(u_n)$ was proved for Benedicks-Carleson maps (which include the example in hand) for observables achieving a maximum either at the critical point or at its image. Note that, in here, $D'(u_n)$, which imposes some significant memory loss for relatively fast returns, cannot hold because ζ is a fixed point. Nevertheless, what we will prove is that the set points that manage to escape from a tight vicinity of ζ , *i.e.*, $Q_{1,0}(u_n)$, only return after having a significant memory loss.

As in [FF08, Section 6], we start by computing a turning instant T which splits the time interval $0, \ldots, [n/k]$ of the sum in 2.7 into two parts. We compute $T \in \mathbb{N}$ such that for every j > T we have $2Ce^{-\alpha j} < \frac{1}{n^3}$. For n large enough it suffices to take

$$T = \left[\frac{4\log n}{\alpha}\right]$$

From (4.2) with $\phi = \psi = \mathbf{1}_{Q_1(u_n)}$ and for j > T one easily gets

$$\left|\mu(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) - \mu(Q_{p,0}(u_n))^2\right| \le \frac{1}{n^3},$$

which implies that for some C > 0 we have

$$n\sum_{j=T}^{[n/k_n]} \mu(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) \le n[n/k_n] \left(\mu(Q_{p,0}(u_n))^2 + \frac{1}{n^3} \right)$$
$$\le \frac{n^2}{k_n} \mu(X_0 > u_n)^2 + \frac{n^2}{k_n n^3} \le C \frac{\tau^2}{k_n} \xrightarrow[n \to \infty]{} 0.$$

Thus, we are left with the piece of history from time 0 to T to analyse. For that purpose, we start by studying the pre-images of a small interval in the vicinity of ζ , namely, A(s) = [-1, -1 + s] for s close to 0. We have $f^{-1}(A(s)) = A_1(s) \cup A_2(s)$, where $A_1(s) = [-1, -\sqrt{1 - s/2}]$ is a small neighbourhood of -1 and $A_2(s) = [\sqrt{1 - s/2}, 1]$ is a small neighbourhood of 1. Also, the pre-image of any small neighbourhood of 1 is a small neighbourhood of the critical point 0, in particular:

$$f^{-1}(A_2(s)) = \left[-\sqrt{1/2 - 1/2\sqrt{1 - s/2}}, \sqrt{1/2 - 1/2\sqrt{1 - s/2}}\right].$$

Moreover, for s close to 0, we may write

$$\sqrt{\frac{1}{2} - \frac{1}{2}\sqrt{1 - \frac{s}{2}}} \sim \frac{\sqrt{2}}{4}\sqrt{s}.$$
(4.3)

Now, observe that $Q_1(u_n) \subset A(1-u_n)$ and if you are to enter $A(1-u_n)$ then either you are in $A_1(1-u_n) \subset A(1-u_n)$ or in $A_2(1-u_n)$. Since, by definition of $Q_1(u_n)$, if you start in a point of $Q_1(u_n)$, then you immediately leave $A(1-u_n)$ in the next iterate, this means that the only way you can return to $Q_1(u_n)$ is if you enter $A_2(1-u_n)$, which implies that you must enter the neighbourhood of the critical point $f^{-1}(A_2(1-u_n))$ first. Hence,

$$Q_1(u_n) \cap f^{-j}(Q_1(u_n)) \subset A(1-u_n) \cap f^{-j+2}(f^{-1}(A_2(1-u_n))).$$

Using the symmetry of the map and of the invariant density plus the invariance of μ , we have

$$\mu \left(A(1-u_n) \cap f^{-j+2}(f^{-1}(A_2(1-u_n))) \right) = 2\mu \left(A_2(1-u_n) \cap f^{-j+1}(f^{-1}(A_2(1-u_n))) \right)$$
$$= 2\mu \left(f^{-1}(A_2(1-u_n)) \cap f^{-j}(f^{-1}(A_2(1-u_n))) \right).$$

This means, that we only need to study the probability of starting in the neighbourhood of the critical point and returning after j iterates and for that we use the computations in [FF08, Section 6]. The threshold Θ defined in [FF08, Equation (6.1)], in here, is given by

$$\Theta = \Theta(n) = \left[-\log\left(\sqrt{1/2 - 1/2\sqrt{1 - (1 - u_n)/2}}\right) \right].$$

Using (4.1) and (4.3) we may write

$$\Theta(n) \sim -\log\left(\frac{\sqrt{2}}{4}\sqrt{1-u_n}\right) \sim \log n,$$

which implies the existence of $C_1 > 0$ such that $2T/\Theta \leq C_1$ for all $n \in \mathbb{N}$. We are now in condition of using the final computations of [FF08, Section 6] to get that

$$\mu\left(f^{-1}(A_2(1-u_n))\cap f^{-j}(f^{-1}(A_2(1-u_n)))\right) \le \operatorname{const} T^{C_1+1} \mathrm{e}^{-(1-7\beta)2\Theta}$$

where $0 < \beta < 0.01$. Finally, since by definition of Θ , (4.1) and (4.3), we have $e^{-\Theta} \leq const 1/n$, it follows that

$$n\sum_{j=1}^{T} \mu(Q_1(u_n) \cap f^{-j}(Q_1(u_n))) \le \operatorname{const} n\sum_{j=1}^{T} T^{C_1+1} e^{-2(1-7\beta)\Theta}$$
$$\le \operatorname{const} \frac{n(\log n)^{C_1+2}}{n^{2(1-7\beta)}} \xrightarrow[n \to \infty]{} 0$$

5. EVL/HTS FOR CYLINDERS

Many results on HTS for dynamical systems were initially proved for HTS to dynamically defined cylinders, which is usually a more straightforward problem to study. Indeed many results which are known about the statistics of hits to cylinders are not known for balls. Therefore one of our goals in [FFT10a] was to extend the results of [FFT10] to this setting. In this short section we outline this theory and in Section 6 we will apply it to the problem of EVLs with non-trivial EI.

Many dynamical systems (\mathcal{X}, f) come with a natural partition \mathcal{P}_1 , for example this might be the collection of maximal sets on which f is locally homeomorphic. The dynamically defined *n*-cylinders are then $\mathcal{P}_n := \bigvee_{i=0}^{n-1} f^{-i}(\mathcal{P}_1)$. For $x \in \mathcal{X}$, let $\mathbb{Z}_n[x]$ denote an element of \mathcal{P}_n containing x. Note that in principle there may be more than one choice of cylinder, but in the cases we consider we can make an arbitrary choice.

If we wish to deal with HTS/EVL to dynamically defined cylinders $Z_n[\zeta]$ around a point ζ , we replace the sets $(U_n)_n$ with $(Z_n[\zeta])_n$ in (1.4). In this case we chose our observable φ to be of the form

$$\varphi = g_i \circ \psi, \tag{5.1}$$

where g_i is one of the three forms given above and $\psi(x) := \mu(\mathbb{Z}_n[\zeta])$ where *n* is maximal such that $x \in \mathbb{Z}_n[\zeta]$. Moreover, we select a subsequence of the time *n*, which we denote by $(\omega_n)_{n \in \mathbb{N}}$ and such that

$$\omega_n \mu(X_0 > u_n) \xrightarrow[n \to \infty]{} \tau > 0, \tag{5.2}$$

and for every $n \in \mathbb{N}$, $\tau \geq 0$, where u_n is taken to be such that

$$\{X_0 > u_n\} = Z_n[\zeta].$$
(5.3)

We achieve this, for example, by letting

$$\omega_n = \omega_n(\tau) = [\tau (\mu(X_0 > u_n))^{-1}].$$
(5.4)

Finally, we say that we have a cylinder EVL H for the maximum if for any sequence $(u_n)_{n \in \mathbb{N}}$ such that (5.3) holds and for ω_n defined in (5.4), the limit (5.2) holds and

$$\mu\left(M_{\omega_n} \le u_n\right) \to \bar{H}(\tau),\tag{5.5}$$

for some non-degenerate d.f. H, as $n \to \infty$. The cylinder HTS is defined analogously. The equivalence between these two perspectives was given in [FFT10a, Theorem 3]. We also showed in that paper that the following two conditions imply that (5.5) holds with $\bar{H}(\tau) = e^{-\tau}$.

Condition $(D(u_n, \omega_n))$. We say that $D(u_n, \omega_n)$ holds for the sequence X_0, X_1, \ldots if for any integers ℓ, t and n

$$|\mu(\{X_0 > u_n\} \cap \{\max\{X_t, \dots, X_{t+\ell-1}\} \le u_n\}) - \mu(\{X_0 > u_n\})\mu(\{M_\ell \le u_n\})| \le \gamma(n, t),$$

where $\gamma(n,t)$ is nonincreasing in t for each n and $\omega_n \gamma(n,t_n) \to 0$ as $n \to \infty$ for some sequence $t_n = o(\omega_n)$;

Condition $(D'(u_n, \omega_n))$. We say that $D'(u_n, \omega_n)$ holds for the sequence X_0, X_1, \ldots if

$$\lim_{k \to \infty} \limsup_{n \to \infty} \omega_n \sum_{j=1}^{\lfloor \omega_n / k \rfloor} \mu(\{X_0 > u_n\} \cap \{X_j > u_n\}) = 0.$$
(5.6)

Remark 5. We say a system is Φ -mixing, if for an n-cylinder U and a measurable set V,

$$\left|\mu(U \cap f^{-j}(V)) - \mu(U)\mu(V)\right| \le \Phi(j)\mu(U)\mu(V)$$

where $\Phi(j)$ decreases to 0 monotonically in j. This holds in the Axiom A case: see Haydn and Vaienti [HV09] for example. In [HV09, Section 3] they showed that Φ -mixing dynamical systems give rise to Poisson HTS around periodic points with a parameter, interpreted in the current paper as the EI. The Poisson law is for the number of returns to asymptotically small cylinders. Our results imply theirs for the *first* hitting time. As can be seen from our examples, we do not require our systems to have such good mixing properties and moreover our results also apply to balls.

6. DICHOTOMY FOR UNIFORMLY EXPANDING MAPS

In this section we will prove that for a simple class of dynamical systems periodic points are the only points which can generate a cylinder EVL with EI in (0, 1). Therefore we understand all the cylinder EVLs for this system.

We assume that the dynamics is $f : x \mapsto 2x \mod 1$ on the unit interval I = [0, 1]Let $\alpha \in (0, 1/2]$ and μ be the $(\alpha, 1 - \alpha)$ -Bernoulli measure. This is thus a Rychlik system as in Section 4.1. Moreover, for $\alpha = 1/2$ the measure is Lebesgue. While this system has stronger mixing properties, we will only actually use the fact that for our system, for n-cylinders U, V,

$$\left|\mu(U \cap f^{-j}(V)) - \mu(U)\mu(V)\right| \le \Phi(j)\mu(U) \text{ where } \sum_{j} \Phi(j) < \infty.$$
(6.1)

(Observe that this is a weaker assumption than Φ -mixing.)

Proposition 2. Suppose that (I, f, μ) and the observable φ is as in (5.1). If $\zeta \in I$ is non-periodic then $D'(u_n, \omega_n)$ and $D(u_n, \omega_n)$ hold. Hence there is an EVL with EI equal to 1.

- *Remark* 6. This proposition allied to the cylinder version of Proposition 1, completely characterises the possible cylinder EVLs for this system.
 - In the proof of the proposition, the only properties we need for our dynamical system are that it is Markov, that (6.1) holds and the measure of *n*-cylinders decay exponentially in *n*.
 - We would expect a similar proposition to be true for balls also. However we strongly use the cylinder structure of the system (I, f) in our proof. It may be possible to approximate the balls by cylinders, but and since *n*-cylinders can not all be assumed to be symmetric about ζ , in the usual metric on *I*, this may not be straightforward.

Before proving the proposition, we will discuss the symbolic structure of our dynamical system and then prove a lemma.

First we recall that the system (I, f) has a natural coding: $x \in I$ can be given the code $x_0x_1...$ where $x_i = 0$ if $f^i(x) \in [0, 1/2)$ and $x_i = 1$ if $f^i(x) \in [1/2, 1)$. Then the dynamics is semi-conjugate to the full shift on two symbols $(\{0, 1\}^{\mathbb{N}_0}, \sigma)$, where $\sigma(x_0x_1...) = x_1x_2...$ for $x_i \in \{0, 1\}$. Notice that the points x where there is a problem in the conjugacy are precisely the points which map onto the fixed point at zero. Following the proofs below it is easy to see that Proposition 2 follows almost immediately in this case. In fact this is also the situation in which the cylinder $Z_n[x]$ is not well defined.

Now let $(p_i)_i$ be the sequence of integers such that whenever $p_i \leq n < p_{i+1}$ the time the orbit of ζ takes to visit $Z_n[\zeta]$ is at least p_i . (We will sometimes denote *i* such that $p_{i_n} \leq n < p_{i_{n+1}}$ by i_n .) For example, suppose that the first 153 symbols representing ζ are

In this case $p_0 = 1$, $p_1 = 15$ and $p_2 = 153$.

Letting $a_n \in \mathbb{N}$ be maximal such that $a_n p_i \leq n$, we can also interpret $(p_i)_i$ as the first times $r = p_i$ when $\mathbb{Z}_r[\zeta]$ contains no periodic point of period less than r. So for $p_i \leq n < p_{i+1}$ where $j = a_n p_i + q_n$, for some $0 \leq q_n < p_i$, the coding for ζ up to time n must consist of the block $\zeta_0 \dots \zeta_{p_i-1}$ repeated a_n times followed by the block $\zeta_0 \dots \zeta_{q_n-1}$.

Remark 7. The periodic structure of cylinders was considered in [AVe09], see particularly Section 3 (note that there they are interested in first returns/hitting times of the *whole* cylinder to itself, which is slightly different to what we look at here). They considered the p_i blocks $\zeta_0 \ldots \zeta_{p_i-1}$ as being '*i*-period' blocks and the block $\zeta_{a_np_i} \ldots \zeta_{a_np_i+q_n}$ as an '*i*-rest' block. In the example in (6.2) the relevant blocks have 1-period 15 and the 1-rest period is 2.

A key point in the proof of Proposition 2 is that the assumption that ζ is not periodic implies that $p_{i_n} \to \infty$ as $n \to \infty$. The following lemma explains how this affects short term returns to *n*-cylinders.

Lemma 6.1. For $p_i \leq n < p_{i+1}$ as above, if, for $j \leq n$, there is a cylinder $Z_{n+j} \subset Z_n[\zeta]$ such that $f^j(Z_{n+j}) \subset Z_n[\zeta]$ then

(a) there is only one such cylinder in $Z_n[\zeta]$ with the same return time j;

(b) there exists $0 \le k \le a_n$ such that $j = kp_i$.

Proof. Suppose that Z_{n+j} is as in the lemma. Since $Z_{n+j} \subset Z_n[\zeta]$, the coding for Z_{n+j} must be of the form

$$\zeta_0 \ldots \zeta_{n-1} \alpha_0 \ldots \alpha_{j-1}$$

for some $\alpha_0 \ldots \alpha_{j-1} \in \{0,1\}^j$. Moreover the fact that $j \leq n$ means that the coding for \mathbb{Z}_{n+j} must be

$$\zeta_0\ldots\zeta_{j-1}\zeta_0\ldots\zeta_{n-1}.$$

In particular there is only one possible code for such a cylinder, determined only by j and by the first n entries in the code for ζ and the first part of the lemma follows. The second part also follows immediately from the periodic structure of the code for ζ . (The proof can also be seen from the setup described in [AVe09, Section 3].)

Proof of Proposition 2. The fact that $D(u_n, \omega_n)$ holds follows from Theorem 7 as in Section 3.3.

To prove $D'(u_n, \omega_n)$, we first estimate the first *n* terms in the sum (5.6). We note that for our system and Z_{n+j} as in Lemma 6.1

$$\mu(\mathbf{Z}_{n+j}) \le \mu(\mathbf{Z}_n[\zeta])\vartheta^{j}$$

for $\vartheta = 1/\alpha$. Then for a_n maximal such that $a_n p_{i_n} \leq n$ and $\omega_n = \omega_n(\tau)$

$$\omega_n \sum_{j=1}^n \mu \left(x \in \mathbf{Z}_n[\zeta] : f^j(x) \in \mathbf{Z}_n[\zeta] \right) = \omega_n \sum_{j=p_{i_n}}^n \mu \left(x \in \mathbf{Z}_n[\zeta] : f^j(x) \in \mathbf{Z}_n[\zeta] \right)$$
$$= \omega_n \sum_{k=1}^{a_n} \mu \left(x \in \mathbf{Z}_n[\zeta] : f^{kp_i}(x) \in \mathbf{Z}_n[\zeta] \right)$$
$$\lesssim \omega_n \mu(\mathbf{Z}_n[\zeta]) \sum_{k=1}^\infty \vartheta^{kp_i} \lesssim \frac{\tau \vartheta^{p_i}}{1 - \vartheta}.$$

Therefore using the mixing condition,

$$\omega_n \sum_{j=1}^{\lfloor n/k \rfloor} \mu(\{X_0 > u_n\} \cap \{X_j > u_n\}) = \omega_n \sum_{j=1}^n \mu\left(x \in Z_n[\zeta] : f^j(x) \in Z_n[\zeta]\right) + \omega_n \sum_{j=n+1}^{\lfloor \omega_n/k \rfloor} \mu(x \in Z_n[\zeta] : f^j(x) \in Z_n[\zeta]) \lesssim \frac{\tau \vartheta^{p_i}}{1 - \vartheta} + \omega_n \sum_{j=n+1}^{\lfloor \omega_n/k \rfloor} \Phi(j) \mu(Z_n[\zeta]) + \mu(Z_n[\zeta])^2 \leq \tau \left(\frac{\vartheta^{p_i}}{1 - \vartheta} + \frac{\tau}{k} + \sum_{j=n+1}^\infty \Phi(j)\right).$$

Since $\Phi(j)$ is summable,

$$\sum_{j=n+1}^{\infty} \Phi(j) \to 0 \text{ as } n \to \infty.$$

Moreover, as $n \to \infty$, our assumption that ζ is not periodic implies that $p_{i_n} \to \infty$ as $n \to \infty$. Therefore,

$$\limsup_{n \to \infty} \omega_n \sum_{j=1}^{\lfloor \omega_n/k \rfloor} \mu(\{X_0 > u_n\} \cap \{X_j > u_n\}) \le \frac{\tau^2}{k},$$

 $D'(u_n, \omega_n)$ follows by taking $k \to \infty$. The existence of an EVL with EI equal to 1 follows from [FFT10a, Section 5].

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APPENDIX A. A MAXIMUM MOVING AVERAGE PROCESS WITH PERIOD 2

In all sections of the appendix we show how our theory applies to some classical stochastic processes.

Let $Y_{-2}, Y_{-1}, Y_0, Y_1, \ldots$ be a sequence of i.i.d. random variables with common d.f. G. We require that for all $\tau \ge 0$ there exists a sequence $\{v_n\}_{n\in\mathbb{N}}$ such that $n(1-G(v_n)) \to \tau$, as $n \to \infty$. We define a Maximum Moving Average process X_0, X_1, \ldots based on the previous sequence in the following way: for each $n \in \mathbb{N}_0$ set

$$X_n = \max\{Y_{n-2}, Y_n\}.$$

Note that by definition of the sequence $\{u_n\}_{n\in\mathbb{N}}$, we must have $\mathbb{P}(X_0 > u_n) \to \tau \ge 0$, as $n \to \infty$. For simplicity let $\alpha_n = \mathbb{P}(Y_0 \le u_n)$. Then, since $\mathbb{P}(X_0 > u_n) = 2(1-\alpha_n)-(1-\alpha_n)^2$ we must have that $\alpha_n \to 1$ and $n(1-\alpha_n) \to \tau/2 \ge 0$, as $n \to \infty$.

We claim that the 2-dependent process X_0, X_1, \ldots satisfies $SP_{p,\theta}$, with $p = 2, \theta = 1/2, D^p(u_n)$ and $D'_p(u_n)$. Applying Theorem 1 it follows that X_0, X_1, \ldots has an EI given by $\theta = 1/2$.

We start by verifying $SP_{p,\theta}$. Since $\mathbb{P}(X_0 > u_n) = 2(1-\alpha_n) - (1-\alpha_n)^2$ and $\mathbb{P}(X_1 \le u_n, X_0 > u_n) = 2\alpha_n^2(1-\alpha_n) - \alpha_n^2(1-\alpha_n)^2$, it follows

$$\mathbb{P}(X_1 > u_n | X_0 > u_n) = 1 - \frac{2\alpha_n^2 (1 - \alpha_n) - \alpha_n^2 (1 - \alpha_n)^2}{2(1 - \alpha_n) - (1 - \alpha_n)^2} \xrightarrow[n \to \infty]{} 0.$$

On the other hand since $\mathbb{P}(X_2 \leq u_n, X_0 > u_n) = \alpha_n^2(1 - \alpha_n)$ we have

$$\mathbb{P}(X_2 > u_n | X_0 > u_n) = 1 - \frac{\alpha_n^2 (1 - \alpha_n)}{2(1 - \alpha_n) - (1 - \alpha_n)^2} \xrightarrow[n \to \infty]{} \frac{1}{2},$$

which means that (2.1) is satisfied with p = 2 and $\theta = 1/2$. Condition (2.2) follows from the fact that X_0, X_1, \ldots is 2-dependent. In fact, since for all $i \in \mathbb{N}$

$$\mathbb{P}(X_0 > u_n, X_2 > u_n, \dots, X_{2(2i+1)} > u_n) \leq \mathbb{P}(X_0 > u_n, X_2 > u_n, \dots, X_{2(2i)} > u_n) \\
\leq \mathbb{P}(X_0 > u_n, X_4 > u_n, \dots, X_{4i} > u_n) \leq 2^i (1 - \alpha_n)^i$$

and $(1 - \alpha_n) \to 0$, it follows that there exists C > 0 such that

$$\sum_{i=1}^{n} \mathbb{P}(X_0 > u_n, X_2 > u_n, \dots, X_{2i}) \le 2 \sum_{i=1}^{[n/2]} 2^i (1 - \alpha_n)^i \le C(1 - \alpha_n) \sum_{i=1}^{\infty} (1/2)^i \xrightarrow[n \to \infty]{} 0.$$

Condition $D^p(u_n)$ follows trivially from the fact that the process is 2-dependent.

We are left now with condition $D'_p(u_n)$. Recall that, in this case, $Q_{p,i}(u_n) = \{X_i > u_n, X_{i+2} > u_n\}$. It is easy to check that $\mathbb{P}(Q_{p,0} \cap Q_{p,i}) = (1 - \alpha_n)^2 \alpha_n^4$ for all $i \in \mathbb{N}$, except for i = 2 and i = 4 for which such probability is 0. Hence,

$$\sum_{i=1}^{[n/k_n]} n \mathbb{P}(Q_{p,0} \cap Q_{p,i}) \le [n/k_n] n (1 - \alpha_n)^2 \alpha_n^4 \xrightarrow[n \to \infty]{} 0,$$

because $[n/k_n](1-\alpha_n) \to 0$, $\alpha_n^4 \to 1$ and $n(1-\alpha_n) \to \tau/2 \ge 0$, as $n \to \infty$.

Appendix B. A Maximum Moving Average process with two underlying periodic phenomena of periods 1 and 3

As before, let $Y_{-2}, Y_{-1}, Y_0, Y_1, \ldots$ be a sequence of i.i.d. random variables as in Appendix A. This time, we define a Maximum Moving Average process X_0, X_1, \ldots in the following way: for each $n \in \mathbb{N}_0$ set

$$X_n = \max\{Y_{n-3}, Y_{n-2}, Y_n\}.$$

This example appears also in [F06, Section 3]. Letting $\alpha_n = \mathbb{P}(Y_0 \leq u_n)$, since $\mathbb{P}(X_0 > u_n) = 3(1-\alpha_n)-3(1-\alpha_n)^2+(1-\alpha_n)^3$ we must have that $\alpha_n \to 1$ and $n(1-\alpha_n) \to \tau/3 \geq 0$, as $n \to \infty$.

We claim that the 4-dependent process X_0, X_1, \ldots satisfies $SP_{\mathbf{p}_i,\Theta_i}^{(i)}$, with i = 1, 2, $\mathbf{p}_2 = (p_1, p_2) = (1, 3)$, $\Theta_2 = (2/3, 1/2)$, $D^{\mathbf{p}_2}(u_n)$ and $D'_{\mathbf{p}_2}(u_n)$. Applying Theorem 2 it follows that X_0, X_1, \ldots has an EI given by $\theta = 2/3.1/2 = 1/3$.

We start by verifying $SP_{p,\theta}$ with $p_1 = 1$ and $\theta_1 = 2/3$. Since $\mathbb{P}(X_0 > u_n) = 3(1 - \alpha_n) - 3(1 - \alpha_n)^2 + (1 - \alpha_n)^3$ and $\mathbb{P}(X_1 \le u_n, X_0 > u_n) = 2\alpha_n^3(1 - \alpha_n) - \alpha_n^3(1 - \alpha_n)^2$, it follows

$$\mathbb{P}(X_1 > u_n | X_0 > u_n) = 1 - \frac{2\alpha_n^3 (1 - \alpha_n) - \alpha_n^3 (1 - \alpha_n)^2}{3(1 - \alpha_n) - 3(1 - \alpha_n)^2 + (1 - \alpha_n)^3} \xrightarrow[n \to \infty]{} 1/3,$$

which means that (2.1) is satisfied with $p_1 = 1$ and $\theta_1 = 2/3$. Condition (2.2) follows from the fact that X_0, X_1, \ldots is 4-dependent just as in Appendix A. This time we cannot apply Theorem 1 because D'_{p_1} does not hold since

$$\mathbb{P}(Q_{p_1,0}(u_n) \cap Q_{p_1,3}(u_n)) = \mathbb{P}(Q_{\mathbf{p}_1,0}^{(1)}(u_n) \cap Q_{\mathbf{p}_1,3}^{(1)}(u_n)) = (1 - \alpha_n)\alpha_n^5,$$

which means that for $\tau > 0$ we have $n\mathbb{P}(Q_{p_1,0}(u_n) \cap Q_{p_1,3}(u_n)) \to \tau/3$, which, in turn, reveals the presence of another periodic phenomena of period 3. In fact, since $\mathbb{P}(Q_{p_1,0}(u_n)) = 2(1-\alpha_n)\alpha_n^2 - (1-\alpha_n)^2\alpha_n^2$, we have

$$\mathbb{P}(Q_{p_{1},1}(u_{n})|Q_{p_{1},0}(u_{n})) = 0$$

$$\mathbb{P}(Q_{p_{1},2}(u_{n})|Q_{p_{1},0}(u_{n})) = \frac{(1-\alpha_{n})^{2}\alpha_{n}^{5}}{2(1-\alpha_{n})\alpha_{n}^{2} - (1-\alpha_{n})^{2}\alpha_{n}^{2}} \xrightarrow[n \to \infty]{} 0$$

$$\mathbb{P}(Q_{p_{1},3}(u_{n})|Q_{p_{1},0}(u_{n})) = \frac{(1-\alpha_{n})\alpha_{n}^{5}}{2(1-\alpha_{n})\alpha_{n}^{2} - (1-\alpha_{n})^{2}\alpha_{n}^{2}} \xrightarrow[n \to \infty]{} 1/2$$

which implies that (2.13) holds when i = 2. Besides, using the 4-dependence of the process we can show that, for i = 2, condition (2.14) holds just as in the proof of (2.2) in Appendix A.

As before, the fact that X_0, X_1, \ldots is 4-dependent clearly implies that condition $D^{\mathbf{p}_2}(u_n)$ holds.

We are left with $D'_{\mathbf{p}_2}(u_n)$. To verify it we observe that

$$\mathbb{P}\left(Q_{\mathbf{p}_{2},0}^{(2)}(u_{n}) \cap Q_{\mathbf{p}_{2},1}^{(2)}(u_{n})\right) = 0$$
$$\mathbb{P}\left(Q_{\mathbf{p}_{2},0}^{(2)}(u_{n}) \cap Q_{\mathbf{p}_{2},2}^{(2)}(u_{n})\right) = (1 - \alpha_{n})^{3}\alpha_{n}^{5}(1 + \alpha_{n})$$
$$\mathbb{P}\left(Q_{\mathbf{p}_{2},0}^{(2)}(u_{n}) \cap Q_{\mathbf{p}_{2},3}^{(2)}(u_{n})\right) = 0,$$

and for all $j \ge 4$ we have

$$\mathbb{P}\left(Q_{\mathbf{p}_{2},0}^{(2)}(u_{n}) \cap Q_{\mathbf{p}_{2},j}^{(2)}(u_{n})\right) \leq \mathbb{P}\left((Y_{-3} > u_{n} \lor Y_{0} > u_{n}), (Y_{j-2} > u_{n} \lor Y_{j} > u_{n})\right) \leq 4(1-\alpha_{n})^{2}.$$

All these together give

$$n\sum_{j=1}^{[n/k_n]} \mathbb{P}\left(Q_{\mathbf{p}_2,0}^{(2)}(u_n) \cap Q_{\mathbf{p}_2,j}^{(2)}(u_n)\right) \le 2n(1-\alpha_n)^3 + [n/k_n]n4(1-\alpha_n)^2 \xrightarrow[n \to \infty]{} 0.$$

APPENDIX C. AN AUTOREGRESSIVE PROCESS DUE TO CHERNICK

In this section we consider a stationary first-order autoregressive process with uniform marginal distributions introduced by Chernick [C81] which illustrates the existence of an Extremal Index different from 1.

Let $r \in \mathbb{N} \setminus \{1\}$ and $\epsilon_1, \epsilon_2, \ldots$ be a sequence of i.i.d. random variables with uniform discrete distribution on $\{0, 1/r, 2/r, \ldots, (r-1)/r\}$, *i.e.*, $\mathbb{P}(\epsilon_1 = k/r) = 1/r$ for all $k = 0, 1, \ldots, r-1$. The uniform AR(1) process is defined recursively as follows:

$$X_n = \frac{1}{r} X_{n-1} + \epsilon_n,$$

where ϵ_n is independent of X_{n-1} and X_0 is uniformly distributed in [0, 1]. It is simple to check that X_0, X_1, \ldots forms a stationary stochastic process such that each X_n is uniformly distributed on [0, 1].

In [C81, Theorem 3.1], Chernick shows that this process satisfies $D(u_n)$ from Leadbetter but $D'(u_n)$ fails. Besides, in [C81, Theorem 4.1], using a direct approach, he shows that the partial maxima has a EVL of type III with an extremal index equal to 1 - 1/r.

We will show that machinery we developed can be applied to this process and obtain the same result as Chernick [C81, Theorem 4.1] simply by checking the conditions of Theorem 1.

C.1. Verification of $MP_{p,\theta}$. The proof of condition $MP_{p,\theta}$ relies on the following property of the process:

Lemma C.1. For all u > (r-1)/r and $n \in \mathbb{N}$, if $X_{n-1} > u$ then $X_n > u$ if and only if $\epsilon_n = (r-1)/r$.

This means that the probability of having an exceedance of any high level u, given that you have just had an exceedance, is 1/r, which makes it a periodic phenomenon of period p = 1 (in the sense of condition 2.1).

Proof of Lemma C.1. Fix u > (r-1)/r, $n \in \mathbb{N}$ and assume that $X_{n-1} > u$.

First we show that if $\epsilon_n = (r-1)/r$ then $X_n > u$. To see this observe that since r > 1 we have:

$$1 - X_n = 1 - \epsilon_n - \frac{1}{r}X_{n-1} = \frac{1}{r} - \frac{X_{n-1}}{r} < \frac{1}{r}(1-u) < 1 - u,$$

which gives $X_n > u$.

Finally, if $\epsilon_n \leq (r-2)/r$, we have $X_n = 1/rX_{n-1} + \epsilon_n \leq 1/r + (r-2)/r \leq (r-1)/r < u$. \Box

Letting u > (r-1)/r, $i \in \mathbb{N}$, using Lemma C.1 and the facts that the ϵ_n 's are a sequence of iid random variables and each ϵ_n is independent of X_{n-1} we have:

$$\mathbb{P}(X_1 > u, \dots, X_i > u | X_0 > u) = \frac{\mathbb{P}(X_0 > u, X_1 > u, \dots, X_i > u)}{\mathbb{P}(X_0 > u)} \\
= \frac{\mathbb{P}(X_0 > u, \epsilon_1 = (r-1)/r, \dots, \epsilon_i = (r-1)/r)}{\mathbb{P}(X_0 > u)} \\
= \frac{\mathbb{P}(X_0 > u)\mathbb{P}(\epsilon_1 = (r-1)/r) \dots \mathbb{P}(\epsilon_i = (r-1)/r)}{\mathbb{P}(X_0 > u)} \\
= (1/r)^i$$

Hence, we have that for all $i \in \mathbb{N}$

 $\lim_{u \to 1} \mathbb{P}\left(X_1 > u, \dots, X_i > u | X_0 > u\right) = (1/r)^i,$

which means that $MP_{p,\theta}$ holds with p = 1 and $\theta = 1 - 1/r$.

The validity of $D^p(u_n)$ follows from a trivial adaption of the proof of $D(u_n)$ in [C81, Theorem 3.1] and, in fact, $D^p(u_n)$ holds with $\gamma(n,t) = \frac{(1/r)^{t-1}}{1-1/r}$.

C.2. Verification of $D'_p(u_n)$. We start by computing a turning instant t^* which splits the time interval $0, \ldots, [n/k]$ of the sum in (2.7) into two parts. The idea behind this splitting is that the dependence between the random variables X_j in the second part, *i.e.*, with $j > t^*$ and X_0 is negligible because of the fast (exponential) decay of $\gamma(n, t)$ in t. This will leave us to analyse the time interval $0, \ldots, t^*$.

Let $\rho = 1/r$. We compute $t^* \in \mathbb{N}$ such that for every $j > t^*$ we have

$$\frac{\rho^{j-1}}{1-\rho} < \frac{1}{n^3}.$$

Since $\frac{\rho^{j-1}}{1-\rho} < \frac{1}{n^3} \Rightarrow j > \frac{3\log n}{\log r} + \frac{\log(1-\rho)}{\log \rho} + 1$, we take, for *n* sufficiently large,

$$t^* = \left[\frac{4\log n}{\log r}\right].$$

From the expression for $\gamma(n, t)$, one easily gets

$$\left|\mathbb{P}\left(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)\right) - \mathbb{P}\left(Q_{p,0}(u_n)\right)^2\right| \le \frac{\rho^{j-1}}{1-\rho},$$

which implies that since $k = k_n \to \infty$ as $n \to \infty$ we have

$$n\sum_{j=t^*+1}^{[n/k]} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) \le n[n/k] \left(\mathbb{P}(Q_{p,0}(u_n))^2 + \frac{1}{n^3} \right) \le \frac{n^2}{k} \mathbb{P}(X_0 > u_n)^2 + \frac{n^2}{kn^3} \xrightarrow[n \to \infty]{} 0.$$

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Thus, we are left with the piece of history from time 0 to t^* to analyse. Recall that $u_n = 1 - \tau/n$ so that $n\mathbb{P}(X_0 > u_n) \to \tau \ge 0$, as $n \to \infty$. Observe that $Q_{p,0}(u_n)$ occurs if and only if $X_0 > u_n$ and $X_1 \le u_n$, which, for n large enough, will only happen if $\epsilon_1 < (r-1)/r$ which means that $X_1 \le (r-1)/r$. Besides, for $Q_{p,j}(u_n)$ to occur we must have $X_j > u_n$. Since $X_j = \epsilon_j + \rho \epsilon_{j-1} + \ldots + \rho^{j-1} X_1$, we have that, for very large n, there exists $\varsigma = \varsigma(n)$ such that $\epsilon_j = \epsilon_{j-1} = \ldots = \epsilon_{j-\varsigma} = (r-1)/r$, otherwise X_j cannot exceed the level u_n . Next, we compute a lower bound for ς .

$$\frac{r-1}{r}\left(1+\frac{1}{r}+\ldots+\frac{1}{r^{\varsigma}}\right) \ge 1-\frac{\tau}{n} \Rightarrow 1-\frac{1}{r^{\varsigma+1}} \ge 1-\frac{\tau}{n} \Rightarrow \varsigma+1 \ge \frac{\log n}{\log r} - \frac{\log \tau}{\log r}$$

Hence, we set

$$\varsigma = \left[\frac{\log n}{\log r} - \frac{\log \tau}{\log r}\right] - 1.$$

For $j - \varsigma > 1$ and using that $\epsilon_{j-\varsigma}, \ldots, \epsilon_j$ are independent random variables which are also independent from X_0 and X_1 , we have

$$\mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) \leq \mathbb{P}(Q_{p,0}(u_n) \cap \{X_j > u_n\})$$

$$\leq \mathbb{P}(Q_{p,0}(u_n) \cap \{\epsilon_j = \epsilon_{j-1} = \dots = \epsilon_{j-\varsigma} = (r-1)/r\})$$

$$\leq \mathbb{P}(Q_{p,0}(u_n))\mathbb{P}(\epsilon_{j-\varsigma} = (r-1)/r) \dots \mathbb{P}(\epsilon_j = (r-1)/r)$$

$$\leq \mathbb{P}(Q_{p,0}(u_n))\frac{1}{r^{\varsigma}}.$$

If $j - \varsigma \leq 1$ then $\mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) = 0.$

Observe that the occurrence of $Q_{p,j}(u_n)$ implies an exceedance of u_n at time j followed by the occurrence of the event $\epsilon_{j+1} < (r-1)/r$, which, in turn, implies that we have to wait at least a period of length ς before another exceedance of u_n occurs: This means that there can only occur at most a $[t^*/\varsigma] + 1$ number of $Q_{p,j}$ events with $j = 1, \ldots, t^*$. Hence, we have

$$n\sum_{j=1}^{t^*} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) \le n([t^*/\varsigma] + 1)\mathbb{P}(X_0 > u_n)\frac{1}{r^\varsigma}.$$

Finally, since there exists some constant C > 0 such that $[t^*/\varsigma] + 1 \leq \frac{\frac{\log n}{\log r} - \frac{\log \tau}{\log r}}{\frac{4\log n}{\log r} - 1} + 1 \leq C$, for all $n \in \mathbb{N}$; $n\mathbb{P}(X_0 > u_n) \to \tau$, as $n \to \infty$; $(1/r)^{\varsigma} \leq (1/r)^{\frac{\log n}{\log r} - \frac{\log \tau}{\log r}} \to 0$, as $n \to \infty$; we have

$$\lim_{n \to \infty} n \sum_{j=1}^{t^*} \mathbb{P}(Q_{p,0}(u_n) \cap Q_{p,j}(u_n)) = 0$$

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