

# Appendix A

## Discrete Renewal Theory: Basic (and a Few Less Basic) Facts and Estimates

### A.1 A Crash Course on Renewal Theory

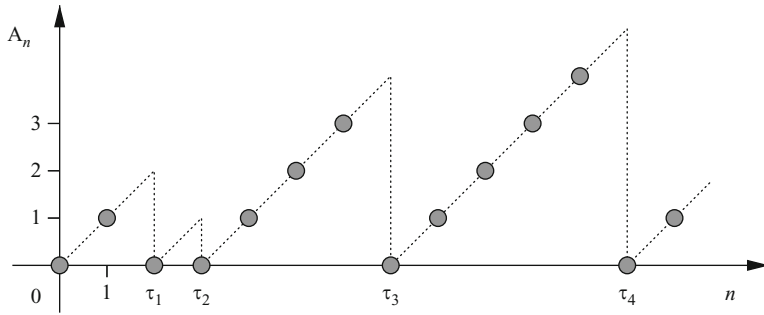
#### A.1.1 Renewal and Markov Chains

We start by working in a general (discrete) framework, that is we choose a discrete probability density  $K(\cdot)$  on  $\mathbb{N} \cup \{\infty\}$  ( $K(\infty) < 1$  to avoid trivialities) and we introduce  $\tau := \{\tau_j\}_{j=0,1,\dots}$  as the sequence of partial sums of an IID sequence of  $K(\cdot)$  distributed variables, that we call *inter-arrival variables*. We call  $\tau$   $K(\cdot)$ -renewal and we stress that  $\tau_0 = 0$  unless explicitly stated. We also freely switch from looking at  $\tau$  as a sequence of random variables and as a (random) subset of  $\mathbb{N} \cup \{0\}$  (*point process*): note that we do not include infinity in this case, because it is always the case that either  $\tau$  is a finite set (when  $\tau_n = \infty$  for some  $n$ ) or  $\tau$  contains infinitely many points (but not  $\infty$ ). The point process notational convention is rather practical and compact: for example  $\{\text{there exists } j \text{ such that } \tau_j = n\}$  shrinks down to  $\{n \in \tau\}$ . We say that  $\tau$  is *persistent* when  $|\tau| = \infty$  ( $|\tau|$  is the number of points in  $\tau$ ); otherwise we say that it is *terminating*. Of course  $\tau$  is persistent if and only if  $K(\infty) = 0$ .

Renewal processes enjoy the renewal property, i.e. if  $A \subset \mathcal{P}(\{0, 1, \dots, n\})$  and  $B \subset \mathcal{P}(\{n+1, n+2, \dots\})$  we have

$$\mathbf{P}(\tau \cap [0, n] \in A, n \in \tau, \tau \cap [n+1, \infty) \in B) = \mathbf{P}(\tau \cap [0, n] \in A, n \in \tau) \mathbf{P}(\tau + n \in B). \quad (\text{A.1})$$

There is natural link between renewal processes and Markov chains (see [1, Chap. I] for a quick self-contained review on Markov chains and for all basic notions). In fact, by the strong Markov property the sequence of successive returns of a Markov chain to a fixed (recurrent or transient) state is a renewal process. But this works also in the opposite direction: any renewal process is the return time sequence of a suitable Markov chain. In fact if we define  $A_n := A_n(\tau) := n - \sup\{\tau_k : \tau_k \leq n\}$ , then the sequence  $A := \{A_n\}_{n=0,1,\dots}$  is a Markov chain called



**Fig. A.1** The A process (large gray dots) associated to the renewal  $\tau$ . In this case  $\tau_1 = 2$ ,  $\tau_2 = 3$ ,  $\tau_3 = 7$ ,  $\tau_4 = 12$  and  $\tau_5 > 13$

*backward recurrence time*. Note that  $A_n \in \mathbb{N} \cup \{0\}$  is the time elapsed since the last renewal when looking from  $n$ , see Fig. A.1. The probability transition from  $A_n = i$  to  $A_{n+1} = j$  is non-zero only if  $j = i + 1$  or  $j = 0$  and the probability that the process moves up is  $(\bar{K}(i+1) + K(\infty)) / (\bar{K}(i) + K(\infty))$ , independently of  $A_0, A_1, \dots, A_{n-1}$  (we have used  $\bar{K}(i) := \sum_{n>i} K(n)$  and in this sum  $K(\infty)$  is not included).

We say that  $K(\cdot)$  has period  $p \in \mathbb{N}$  if  $\{n : K(n) > 0\}$  is contained in  $\{pn : n \in \mathbb{N}\}$  and if  $p$  is the largest number with this property. If  $p = 1$  we say that  $K(\cdot)$  is aperiodic. The aperiodicity of  $K(\cdot)$  implies the aperiodicity of the Markov chain A. Note that when  $K(\infty) = 0$ , the state space of A is  $\{0, 1, \dots, \sup\{n : K(n) > 0\}\} \setminus \{\infty\}$  and that A is irreducible (that is all states communicate in a finite number of steps, with positive probability).

For ease of notation set  $m_K := \sum_{n \in \mathbb{N} \cup \{\infty\}} nK(n) \in [1, \infty]$ . Of course  $m_K = \infty$  may arise also when  $K(\infty) = 0$ : in this case A is a recurrent Markov chain, but it is immediate to see that it is a null recurrent chain, since  $\tau_1$  coincides with  $\inf\{n > 0 : A_n = 0\}$ . On the other hand, A is clearly positive recurrent if  $m_K < \infty$ . We will therefore say that  $\tau$  is positive (respectively null) persistent if A is.

### A.1.2 The Renewal Theorem

We state now the Theorem.

**Theorem A.1.** *If  $K(\cdot)$  is aperiodic then  $\lim_{n \rightarrow \infty} \mathbf{P}(n \in \tau) = 1/m_K$ , with  $1/\infty = 0$ .*

We direct the reader to [1, Chap. I, Theorem 2.2] for a proof, which is based on the fundamental formula (direct consequence of the renewal property), often called *renewal equation* (and  $n \mapsto \mathbf{P}(n \in \tau)$  is the *renewal function*), that says

$$\mathbf{P}(n \in \tau) = \mathbf{1}_{n=0} + \sum_{k=0}^n \mathbf{P}(k \in \tau) K(n-k). \quad (\text{A.2})$$

One way to extract information from (A.2) is to pass to Laplace transform: for  $s > 0$

$$\sum_{n=0}^{\infty} \exp(-sn) \mathbf{P}(n \in \tau) = \left( 1 - \sum_{n=1}^{\infty} \exp(-sn) K(n) \right)^{-1}. \quad (\text{A.3})$$

It helps the intuition to note that Theorem A.1 can be proven also by looking at the ergodic properties of the backward recurrence time process  $A$  [1, Sect. VII.2]. In particular one sees that the condition  $m_K < \infty$  is precisely the condition for  $A$  to be positive recurrent. This implies directly the existence of a unique invariant probability measure which we can write explicitly:

$$p_A(n) = \frac{1}{m_K} \sum_{j \geq n+1} K(j), \quad n = 0, 1, \dots \quad (\text{A.4})$$

By the Ergodic Theorem for irreducible aperiodic Markov chains one also has that

$$\lim_{N \rightarrow \infty} \mathbf{P}(A_N = n) = p_A(n). \quad (\text{A.5})$$

If  $A = \{A_n\}_n$  is redefined so that  $A_0$  is distributed according to  $p_A(\cdot)$ , so that  $A$  is stationary, then  $\{n : A_n = 0\}$  is a renewal process translated by a random quantity (independent of the renewal). In other terms,  $\tau_0$  is no longer degenerate, but it has its own distribution (on  $\mathbb{N} \cup \{0\}$ ). We call such a process a *delayed renewal* and when  $\tau_0$  has distribution  $p_A(\cdot)$  [cf. (A.4)] we call *stationary renewal* such a delayed renewal (note that also a delayed renewal enjoys the renewal property).

We will often need information on the number of renewal points up to a certain time  $n$ , that is on  $|\tau \cap (0, n]|$ . The fact that

$$\lim_{n \rightarrow \infty} \frac{1}{n} |\tau \cap (0, n]| = \frac{1}{m_K}, \quad (\text{A.6})$$

both in the almost sure and in the  $L^1$  sense (as a matter of fact, in  $L^p$  for any  $p \geq 1$ ), is a direct consequence of Kolmogorov law of large numbers.

### A.1.3 Beyond the Renewal Theorem

Theorem A.1 does apply when  $m_K = \infty$ , but in this case it calls for refinements. We state here results in this direction in the framework in which we typically work: that is, we assume (2.30).

First of all, the transient case is covered by the following result.

**Theorem A.2.** *If  $K(\infty) > 0$  then*

$$\mathbf{P}(N \in \tau) \stackrel{N \rightarrow \infty}{\sim} \frac{K(N)}{(K(\infty))^2}. \quad (\text{A.7})$$

A proof can be found in [7, A.5.2].

For the null recurrent case we have of course  $\alpha \in (0, 1]$ . The following results can be found in [2, Theorem 8.7.3 and Theorem 8.7.5].

**Theorem A.3.** *For  $\alpha \in (0, 1)$  and  $K(\infty) = 0$  we have that*

$$\mathbf{E}[\tau \cap (0, N)] = \sum_{n=1}^N \mathbf{P}(n \in \tau) \stackrel{N \rightarrow \infty}{\sim} \frac{\sin(\pi\alpha)}{c_K \pi} N^\alpha. \quad (\text{A.8})$$

For  $\alpha = 1$  and  $K(\infty) = 0$  we have that  $\sum_{n=1}^N \mathbf{P}(n \in \tau) \sim N/(c_K \log N)$ .

Theorem A.3 gives *integral* bounds and it can be obtained by standard Tauberian arguments [2] applied to (A.3). Getting *local* estimates is substantially harder: sharp local estimates however have been obtained and this is the content of the next theorem.

**Theorem A.4.** *For  $\alpha \in (0, 1)$  and  $K(\infty) = 0$  we have*

$$\mathbf{P}(N \in \tau) \stackrel{N \rightarrow \infty}{\sim} \frac{\alpha \sin(\pi\alpha)}{c_K \pi} \frac{1}{N^{1-\alpha}}. \quad (\text{A.9})$$

For  $\alpha = 1$  instead

$$\mathbf{P}(N \in \tau) \stackrel{N \rightarrow \infty}{\sim} \frac{1}{c_K \log N}. \quad (\text{A.10})$$

Formula (A.9) is due to Doney [4, Theorem B], that completed a partial result of [6]. The second result instead, that is (A.10), can be found in [2, Theorem 8.7.5].

Of course, from (A.9) one directly extracts the existence for  $\alpha \in (0, 1)$  of  $C = C(K(\cdot)) > 0$  such that for every  $N \in \mathbb{N}$

$$\frac{1}{C} \leq N^{1-\alpha} \mathbf{P}(N \in \tau) \leq C. \quad (\text{A.11})$$

### A.1.4 Convergence of Renewal and Point Processes

A renewal or, more generally, a discrete point process can be seen as a random variable  $\tau : \Omega \rightarrow \mathcal{P}(\mathbb{N} \cup \{0\}) = 2^{\mathbb{N} \cup \{0\}}$ , with  $(\Omega, \mathcal{F}, \mathbf{P})$  a generic probability space and  $\mathcal{P}(\mathbb{N} \cup \{0\})$  is equipped with the  $\sigma$ -algebra  $\mathcal{G} := \sigma(\cup_{n \in \mathbb{N}} \mathcal{G}_n)$ , where  $\{\mathcal{G}_n\}_{n=0,1,\dots}$  is the natural filtration of the process  $A = A(\tau)$  defined in Sect. A.1. It is convenient to introduce a metric space for which  $\mathcal{G}$  is the Borel  $\sigma$ -algebra. For this we introduce the semi-metric  $d_n$  on  $\mathcal{P}(\mathbb{N} \cup \{0\})$  by setting  $d_n(B_1, B_2) = \mathbf{1}_{B_1 \cap [0, n] \neq B_2 \cap [0, n]}$ . Then one directly verifies that  $\mathcal{G}$  is the Borel  $\sigma$ -algebra for the metric  $d(B_1, B_2) := \sum_n 2^{-n} d_n(B_1, B_2)$ . We introduce the notation

$$\mathcal{P}_0 := (\mathcal{P}(\mathbb{N} \cup \{0\}), d), \quad (\text{A.12})$$

for the metric space. In general we do not deal only with renewal processes: the typical case we consider is the one of a sequence of measures  $\{\mathbf{P}_N\}_N$  on  $\mathcal{P}_0$  converging to a limit (that happens to be a renewal process). In any case the convergence in law of  $\{\mathbf{P}_N\}_N$  to a limit  $\mathbf{P}_\infty$  means simply if  $B \in \mathcal{G}_n$  for some  $n$  then  $\lim_N \mathbf{P}_N(B) = \mathbf{P}_\infty(B)$ .

## A.2 Some Pinning Oriented Renewal Issues

### A.2.1 On Boundary Effects

Next is a bound on boundary effects: we just state the inequality that we use, but also the reciprocal bound can be proven (with a different constant) along the same line.

**Lemma A.5.** Assume (2.30). For every  $K(\cdot)$ -renewal  $\tau$  there exists  $C_{bc} > 0$  such that

$$\mathbf{E}[F_n(\tau) | 2n \in \tau] \leq C_{bc} \mathbf{E}[F_n(\tau)], \quad (\text{A.13})$$

for every  $n \in \mathbb{N}$  and every  $F_n : \mathcal{P}(\mathbb{N}) \rightarrow [0, \infty)$  ( $\mathcal{P}(\mathbb{N})$  is the set of subsets of  $\mathbb{N}$ ) which is measurable with respect to the  $\sigma$ -algebra generated by  $\{\{A \in \mathcal{P}(\mathbb{N}) : j \in A\} : j = 1, 2, \dots, n\}$  (that is  $F_n(A) = F_n(A \cap (0, n])$  for every  $A \subset \mathbb{N}$ ).

*Proof.* The various constants  $c_1, c_2, \dots$  appearing below depend only on  $K(\cdot)$ . Set  $X_n = X_n(\tau) := \max\{j = 0, 1, \dots, n : j \in \tau\}$ . By the measurability properties of  $F_n(\cdot)$  we see that  $F_n(\tau) = F_n(\tau \cap (0, j])$  if  $X_n(\tau) = j$  (read  $(0, 0]$  as  $\emptyset$ ) and therefore

$$\mathbf{E}[F_n(\tau) | 2n \in \tau] = \sum_{j=0}^n \mathbf{E}[F_n(\tau \cap (0, j]) | X_n = j] \mathbf{P}(X_n = j | 2n \in \tau), \quad (\text{A.14})$$

and therefore it is sufficient to show that

$$\mathbf{P}(X_n = j | 2n \in \tau) \leq C_{bc} \mathbf{P}(X_n = j). \quad (\text{A.15})$$

For this we write

$$\begin{aligned} \mathbf{P}(X_n = j, 2n \in \tau) &= \mathbf{P}(j \in \tau) \sum_{m=n+1}^{2n} K(m-j) \mathbf{P}(2n-m \in \tau) \\ &= \mathbf{P}(j \in \tau) \left( \sum_{m=n+1}^{\lfloor 3n/2 \rfloor} \dots + \sum_{m=\lfloor 3n/2 \rfloor + 1}^{2n} \dots \right) =: T_1 + T_2. \end{aligned} \quad (\text{A.16})$$

For what concerns  $T_1$  we use the fact that  $2n - m \geq n/2$  so that  $\mathbf{P}(2n - m \in \tau) \leq c_1 \mathbf{P}(2n \in \tau)$  and

$$T_1 \leq c_1 \mathbf{P}(2n \in \tau) \mathbf{P}(j \in \tau) \sum_{m=n+1}^{\lfloor 3n/2 \rfloor} K(m - j). \quad (\text{A.17})$$

For  $T_2$  we use that  $\mathbf{P}(j \in \tau) \leq c_2(j + 1)^{\alpha-1}$  for every  $j$  and that  $m - j \geq \lfloor n/2 \rfloor$ , which implies  $K(m - j) \leq c_3/n^{1+\alpha}$ , so that

$$T_2 \leq c_2 c_3 \mathbf{P}(j \in \tau) n^{-1-\alpha} \sum_{m=\lfloor 3n/2 \rfloor + 1}^{2n} (2n - m + 1)^{\alpha-1} \leq c_4 \mathbf{P}(j \in \tau) n^{-1}, \quad (\text{A.18})$$

for every  $n \in \mathbb{N}$ . Since we have  $\mathbf{P}(2n \in \tau) \geq c_5 n^{\alpha-1}$  and  $\sum_{m=\lfloor 3n/2 \rfloor + 1}^{\infty} K(m - j) \geq c_6 n^{-\alpha}$  (for  $j \leq n$ ), from (A.18) we get that

$$T_2 \leq \frac{c_4}{c_5 c_6} \mathbf{P}(2n \in \tau) \mathbf{P}(j \in \tau) \sum_{m=\lfloor 3n/2 \rfloor + 1}^{\infty} K(m - j). \quad (\text{A.19})$$

By putting (A.17) and (A.19) together we have that

$$\begin{aligned} & \mathbf{P}(X_n = j, 2n \in \tau) \\ & \leq c_7 \mathbf{P}(2n \in \tau) \mathbf{P}(j \in \tau) \sum_{m=n+1}^{\infty} K(m - j) = c_7 \mathbf{P}(2n \in \tau) \mathbf{P}(X_n = j), \end{aligned} \quad (\text{A.20})$$

which completes the proof with  $C_{bc} = c_7$ .  $\square$

## A.2.2 Two Scaling Results for Renewal Processes

The first result is applied in Chap. 6 when  $\alpha \in (1/2, 1)$ , but it plays a central role also for the case  $\alpha = 1/2$ , since it is used in Proposition A.7 below, that, in turn, is used in Chap. 6.

**Proposition A.6.** *For every  $K(\cdot)$ -renewal  $\tau$  with  $K(\cdot)$  as in (2.30) and  $\alpha \in (0, 1)$  we have*

$$\mathcal{L}\text{-}\lim_{n \rightarrow \infty} \frac{1}{n^\alpha} |\tau \cap (0, n]| = \frac{Y_\alpha}{c_K}, \quad (\text{A.21})$$

where  $Y_\alpha$  is a random variable that depends only on  $\alpha$  with the property  $\mathbf{P}(Y_\alpha > 0) = 1$ . In particular  $Y_{1/2} = |Z|/\sqrt{2\pi}$ , with  $Z \sim \mathcal{N}(0, 1)$ .

*Proof.* This is treated for example in [5]. The point is simply that  $|\tau \cap (0, n]| < m$  is equivalent to  $\tau_m > n$  and  $\tau_m$  is a sum of  $m$  IID variables and the question is therefore

an issue of domain of stability of stable laws. In [5, XI.5, p. 373] it is proven that for every  $x > 0$

$$\lim_{n \rightarrow \infty} \mathbf{P} \left( \bar{K}(n) | \tau \cap (0, n] | \geq \frac{2 - \alpha}{\alpha} \frac{1}{x^\alpha} \right) = G_\alpha(x), \quad (\text{A.22})$$

where  $G_\alpha(\cdot)$  is the distribution function of the one sided stable distribution satisfying  $\lim_{x \rightarrow \infty} x^\alpha (1 - G_\alpha(x)) = (2 - \alpha)/\alpha$ , characterized by the Laplace transform

$$\int_0^\infty \exp(-\lambda x) dG_\alpha(x) = \exp(-c_\alpha \lambda^\alpha) \quad \text{with } c_\alpha := \frac{(2 - \alpha)\Gamma(1 - \alpha)}{\alpha}, \quad (\text{A.23})$$

for  $\lambda > 0$ . Such stable laws are treated in detail for example in [5, XIII.6, Theorem 1], where (A.23) is proven along with the fact that  $\lim_{x \searrow 0} G_\alpha(x) = 0$ , so that  $(0, \infty)$  is of full measure under this distribution (for completeness: in this limit  $G_\alpha(x) = o(\exp(-cx^{-\alpha}))$ , with  $c = \alpha/((2 - \alpha)\Gamma(1 - \alpha))$ ). By a change of variable and by using  $\bar{K}(n) \sim (c_K/\alpha)n^{-\alpha}$  in (A.22) we see that

$$\lim_{n \rightarrow \infty} \mathbf{P} \left( \frac{|\tau \cap (0, n]|}{n^\alpha} \geq y \right) = G_\alpha \left( \left( \frac{2 - \alpha}{c_K} \right)^{\frac{1}{\alpha}} y^{-\frac{1}{\alpha}} \right) =: 1 - F_\alpha(y), \quad (\text{A.24})$$

and the asymptotic properties of  $G_\alpha(x)$  mentioned just above directly yield that  $F_\alpha(y)$  tends to 0 as  $y \searrow 0$  (more precisely:  $F_\alpha(y) \sim c_K y/\alpha$ ) and that  $\lim_{y \rightarrow \infty} F_\alpha(y) = 1$  (more precisely,  $1 - F_\alpha(y) = o(\exp(-cy))$  for a  $c > 0$ ). These facts suffice to conclude the converge in law that we claim in the statement and that the limit variable  $Y_\alpha$  is a.s. positive. A number of further properties of  $Y_\alpha$  can be derived by exploiting the properties of the stable distribution  $G_\alpha(\cdot)$ , for example that  $G_\alpha(t) = \int_0^t g_\alpha(s) ds$  for a suitable probability density  $g_\alpha(\cdot)$  (this follows immediately from the fact that the characteristic function  $\psi_\alpha(t) = \int_0^\infty \exp(itx) dG_\alpha(x) = \exp(-c_\alpha t^\alpha (\cos(\pi\alpha/2) + i \sin(\pi\alpha/2)))$  for  $t \geq 0$ , so that  $|\psi_\alpha(t)| = \exp(-c_\alpha |t|^\alpha \cos(\pi\alpha/2))$  and therefore  $\int_{\mathbb{R}} |\psi(t)| dt < \infty$ ). However, it seems impossible to express stable densities in a closed form [5, p. 581], with the notable exception of  $\alpha = 1/2$  for which we can use that if the random variable  $X$  has density  $f_X(x) = x^{-3/2} (2\sqrt{\pi})^{-1} \exp(-1/(4x)) \mathbf{1}_{x>0}$  we have

$$\mathbf{E}[\exp(-\lambda X)] = \exp(-\sqrt{\lambda}), \quad \text{for } \lambda > 0. \quad (\text{A.25})$$

A straightforward, but rather painful, constant tracking exercise [via (A.23) and (A.24)] leads to  $Y_{1/2} = |Z|/\sqrt{2\pi}$ .  $\square$

**Proposition A.7.** *For every  $K(\cdot)$ -renewal  $\tau$  with  $K(\cdot)$  as in (2.30) and  $\alpha = 1/2$  we have*

$$\mathcal{L}\text{-}\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n} \log n} \sum_{1 \leq i < j \leq n} \frac{\delta_i \delta_j}{\sqrt{j-i}} = \frac{|Z|}{(2\pi)^{3/2} c_K^2}, \quad (\text{A.26})$$

where  $Z \sim \mathcal{N}(0, 1)$ .

*Proof.* We introduce the notation

$$Y_n^{(i)} := \sum_{j=i+1}^n \frac{\delta_j}{\sqrt{j-i}}, \quad (\text{A.27})$$

that allows writing

$$\frac{1}{\sqrt{n} \log n} \sum_{1 \leq i < j \leq n} \frac{\delta_i \delta_j}{\sqrt{j-i}} = \frac{1}{\sqrt{n} \log n} \sum_{i=1}^{n-1} \delta_i Y_n^{(i)} =: X_n. \quad (\text{A.28})$$

Note that, by the renewal property of  $\tau$ ,  $Y_n^{(i)}$  (under  $\mathbf{P}(\cdot | \delta_i = 1)$ ) is distributed like  $Y_{n-i} := Y_{n-i}^{(0)}$  (under  $\mathbf{P}$ ). The first step in the proof is observing that, by (A.11), we have

$$\mathbf{E} \left[ \frac{1}{\sqrt{n} \log n} \sum_{i=(1-\varepsilon)n}^{n-1} \delta_i Y_n^{(i)} \right] = \frac{1}{\log n \sqrt{n}} \sum_{i=(1-\varepsilon)n}^{n-1} \sum_{j=i+1}^n \frac{\mathbf{P}(i \in \tau) \mathbf{P}(j-i \in \tau)}{\sqrt{j-i}} = O(\varepsilon), \quad (\text{A.29})$$

uniformly in  $n$ : we have introduced the short-cut convention (that we will keep throughout this proof) that summing from  $(1-\varepsilon)n$  means summing from  $\lfloor (1-\varepsilon)n \rfloor + 1$  and, just below, summing up to  $(1-\varepsilon)n$  means up to  $\lfloor (1-\varepsilon)n \rfloor$ . What (A.29) is telling us is that we can focus on studying  $X_{n,\varepsilon}$ , defined as  $X_n$ , but stopping the sum over  $i$  at  $(1-\varepsilon)n$ :

$$X_{n,\varepsilon} := \frac{1}{\sqrt{n} \log n} \sum_{i=1}^{(1-\varepsilon)n} \delta_i Y_n^{(i)}. \quad (\text{A.30})$$

At this point we use that

$$\lim_{n \rightarrow \infty} \frac{Y_n}{\log n} = \frac{1}{2\pi c_K} =: \widehat{c}_K, \quad (\text{A.31})$$

in  $L^2(\mathbf{P})$  (and hence in  $L^1(\mathbf{P})$ ). We postpone the proof of (A.31) and observe that, since the normalization is the logarithm of  $n$ , it implies that for every  $\varepsilon \in (0, 1)$

$$\lim_{n \rightarrow \infty} \sup_{q \in [\varepsilon, 1]} \mathbf{E} \left[ \left| \frac{1}{\log n} \sum_{j=1}^{\lfloor qn \rfloor} \frac{\delta_j}{\sqrt{j}} - \widehat{c}_K \right| \right] = 0. \quad (\text{A.32})$$

Let us write

$$R_n := X_{n,\varepsilon} - \frac{\widehat{c}_K}{\sqrt{n}} \sum_{i=1}^{(1-\varepsilon)n} \delta_i \quad (\text{A.33})$$



and note that  $n^{-1/2} \sum_{i=1}^{(1-\varepsilon)n} \delta_i$  converges in law toward  $\sqrt{(1-\varepsilon)/(2\pi c_K^2)} |Z|$  by Proposition A.6. It suffices therefore to show that for every  $\varepsilon \in (0, 1)$  we have  $\lim_{n \rightarrow \infty} \mathbf{E}[|R_n|] = 0$ . And in fact

$$\begin{aligned} \mathbf{E}[|R_n|] &\leq \frac{1}{\sqrt{n}} \sum_{i=1}^{(1-\varepsilon)n} \mathbf{E}[\delta_i] \mathbf{E} \left[ \left| \frac{Y_n^{(i)}}{\log n} - \hat{c}_K \right| \middle| \delta_i = 1 \right] \\ &= \frac{1}{\sqrt{n}} \sum_{i=1}^{(1-\varepsilon)n} \mathbf{E}[\delta_i] \mathbf{E} \left[ \left| \frac{Y_{n-i}}{\log n} - \hat{c}_K \right| \right] \xrightarrow{n \rightarrow \infty} 0, \quad (\text{A.34}) \end{aligned}$$

where in the last step we have used (A.32) and (A.11).

We are therefore left with proving (A.31). This result has been established in [3, Theorem 6] in the case in which  $\tau$  is given by the successive returns to zero of a centered, aperiodic and irreducible random walk on  $\mathbb{Z}$  with bounded increment variance. Note that, by well established local limit theorems, for such a class of random walks we have (A.9). In [3] it is proven more, namely that (A.31) holds also almost surely and this is extracted from the estimate  $\text{var}_{\mathbf{P}}(Y_n) = O(\log n)$ . What we are going to do is simply to re-obtain such a bound, by repeating the steps in [3] and using (A.9)–(A.11), for the general renewal processes that we consider (and one can verify that almost sure convergence comes as a bonus, but we will not use it).

The proof goes as follows: by (A.9) one directly sees that  $\lim_{n \rightarrow \infty} \mathbf{E}[Y_n / \log n] = \hat{c}_K$ , therefore we are done if we show that  $\lim_{n \rightarrow \infty} \text{var}_{\mathbf{P}}(Y_n / \log n) = 0$ . So we start by observing that

$$\text{var}_{\mathbf{P}}(Y_n) = \sum_{i,j} \frac{\mathbf{E}[\delta_i \delta_j] - \mathbf{E}[\delta_i] \mathbf{E}[\delta_j]}{\sqrt{i} \sqrt{j}} = 2 \sum_{i=1}^{n-1} \sum_{j=i+1}^n \frac{\mathbf{E}[\delta_i \delta_j] - \mathbf{E}[\delta_i] \mathbf{E}[\delta_j]}{\sqrt{i} \sqrt{j}} + O(1), \quad (\text{A.35})$$

by (A.11). Now we compute

$$\begin{aligned} \sum_{i=1}^{n-1} \sum_{j=i+1}^n \frac{\mathbf{E}[\delta_i \delta_j] - \mathbf{E}[\delta_i] \mathbf{E}[\delta_j]}{\sqrt{i} \sqrt{j}} &= \sum_{i=1}^{n-1} \frac{\mathbf{E}[\delta_i]}{\sqrt{i}} \left[ \sum_{j=1}^{n-i} \frac{\mathbf{E}[\delta_j]}{\sqrt{j+i}} - \sum_{j=i+1}^n \frac{\mathbf{E}[\delta_j]}{\sqrt{j}} \right] \\ &\leq \sum_{i=1}^{n-1} \frac{\mathbf{E}[\delta_i]}{\sqrt{i}} \left[ \sum_{j=1}^{n-i} \frac{\mathbf{E}[\delta_j]}{\sqrt{j+i}} - \sum_{j=i+1}^n \frac{\mathbf{E}[\delta_j]}{\sqrt{j+i}} \right] \\ &\leq \sum_{i=1}^{n-1} \frac{\mathbf{E}[\delta_i]}{\sqrt{i}} \sum_{j=1}^i \frac{\mathbf{E}[\delta_j]}{\sqrt{j+i}} \leq \sum_{i=1}^{n-1} \frac{\mathbf{E}[\delta_i]}{i} \sum_{j=1}^i \mathbf{E}[\delta_j] \\ &\leq C^2 \sum_{i=1}^{n-1} \frac{1}{i^{3/2}} \sum_{j=1}^i \frac{1}{j^{1/2}} = O(\log n), \quad (\text{A.36}) \end{aligned}$$

where, in the last line, we have used (A.11). In view of (A.35), we have obtained  $\text{var}_{\mathbf{p}}(Y_n) = O(\log n)$  so that the proof (A.31) is complete and, with it, the proof of Lemma A.7.  $\square$

### A.2.3 On the Derivatives of the Free Energy Near Criticality

We have seen that, assuming (2.30) and  $\sum_n K(n) = 1$ , for  $\alpha \in (0, 1)$  we have

$$F(h) \stackrel{h \searrow 0}{\sim} c h^{1/\alpha} =: F_{\text{cr}}(h), \quad (\text{A.37})$$

where  $c = (\alpha / (c_K \Gamma(1 - \alpha)))^{1/\alpha} > 0$  (cf. Theorem 2.10). The subscript *cr* is used to indicate that the function captures the leading critical behavior. Recall that  $F(\cdot)$  is real analytic except at the origin. Here we prove that:

**Proposition A.8.** *For  $\alpha \in (0, 1)$  and  $1/\alpha \notin \mathbb{N}$  we have that for every  $j \in \mathbb{N}$*

$$\left( \frac{d}{dh} \right)^j F(h) \stackrel{h \searrow 0}{\sim} \left( \frac{d}{dh} \right)^j F_{\text{cr}}(h) = c h^{-j+1/\alpha} \prod_{i=1}^j \left( \frac{1}{\alpha} - i + 1 \right). \quad (\text{A.38})$$

If  $1/\alpha \in \mathbb{N}$  then (A.38) holds for  $j \leq 1/\alpha$ .

This result largely suffices for our purposes, but let us point out that generalizing the statement to  $j > 1/\alpha$  when  $1/\alpha \in \mathbb{N}$  requires more on  $K(\cdot)$  than (2.30).

*Proof.* Let us start by setting up some notation:

$$\Psi(x) \stackrel{x \geq 0}{=} 1 - \sum_{n=1}^{\infty} K(n) \exp(-nx) \stackrel{x \searrow 0}{\sim} c_K \frac{\Gamma(1 - \alpha)}{\alpha} x^{\alpha} =: \Psi_{\text{cr}}(x). \quad (\text{A.39})$$

Let us recall that the relation defining  $F(h)$  for  $h > 0$  is

$$\Psi(F(h)) = 1 - \exp(-h) \stackrel{h \searrow 0}{\sim} h. \quad (\text{A.40})$$

This formula has the important companion:

$$\Psi_{\text{cr}}(F_{\text{cr}}(h)) = h. \quad (\text{A.41})$$

In the sequel we use the notation  $f^{(j)}(h) := (d/dh)^j f(h)$  and we point out that a standard Riemann sum approximation yields for  $j \in \mathbb{N}$ :

$$\begin{aligned} \Psi^{(j)}(x) &= (-1)^{j+1} \sum_n n^j K(n) \exp(-nx) \stackrel{x \searrow 0}{\sim} (-1)^{j+1} \Gamma(j - \alpha) c_K x^{\alpha-j} \\ &= \left( \prod_{i=1}^{j-1} (\alpha - i) \right) \Gamma(1 - \alpha) c_K x^{\alpha-j} = \Psi_{\text{cr}}^{(j)}(x), \end{aligned} \quad (\text{A.42})$$

with the convention  $\prod_{i=1}^0 (\alpha - i) = 1$ .

Let us now compute the asymptotic behavior of  $F'(h)$  and of  $F''(h)$ : this will serve the double purpose of getting acquainted with the general case and of serving to verify the first step in the induction argument for the general case. First of all from the relation (A.40) defining  $F(h)$  we have

$$\frac{d}{dh} \Psi(F(h)) = \Psi'(F(h)) F'(h) = \exp(-h) \stackrel{h \searrow 0}{\sim} 1, \quad (\text{A.43})$$

so that

$$F'(h) \stackrel{h \searrow 0}{\sim} \frac{1}{\Psi'(F(h))} \sim \frac{1}{\Psi'_{\text{cr}}(F_{\text{cr}}(h))} = F'_{\text{cr}}(h), \quad (\text{A.44})$$

where the last equality follows by using (A.37) and (A.42) or (more easily!) by taking the derivative of (A.41). For what concerns  $F''(h)$  we compute and use once again the relation (A.40) to get

$$\left( \frac{d}{dh} \right)^2 \Psi(F(h)) = \Psi''(F(h)) (F'(h))^2 + \Psi'(F(h)) F''(h) \stackrel{h \searrow 0}{\sim} -1. \quad (\text{A.45})$$

By (A.42) and (A.44) we see that

$$\Psi''(F(h)) (F'(h))^2 \stackrel{h \searrow 0}{\sim} \frac{c}{h}, \quad (\text{A.46})$$

with  $c \neq 0$ , so that

$$\Psi''(F(h)) (F'(h))^2 \stackrel{h \searrow 0}{\sim} -\Psi'(F(h)) F''(h), \quad (\text{A.47})$$

from which we extract the asymptotic relation of  $F''(h) \sim F''_{\text{cr}}(h)$  by using again (A.42) and (A.44), that is by using the asymptotic behavior of  $F$ ,  $F'$ ,  $\Psi'$  and  $\Psi''$ . Note however that, once again, there is a much cheaper way to go from (A.47) to  $F''(h) \sim F''_{\text{cr}}(h)$ : by taking two derivatives of (A.41) one obtains (A.47) with the subscripts  $\text{cr}$  added (six subscripts in total) and with  $\sim$  replaced by  $=$ , but we already know that we can add the subscripts in (A.47) without altering the validity of the statement to all the functions except a priori  $F''$ , and this implies  $F''(h) \sim F''_{\text{cr}}(h)$ .

We have therefore established the claim for  $j = 1$  and  $2$ , but explicit expressions for the  $j$ th derivative of the composition of two functions are rather involved for arbitrary  $j$ . To get the result we want we will go around this point (much in the spirit of the alternative approach used twice above) by observing that

$$\begin{aligned} \left( \frac{d}{dh} \right)^j \Psi(F(h)) &= \Psi^{(1)}(F(h)) F^{(j)}(h) \\ &+ P_j \left( F^{(1)}(h), \dots, F^{(j-1)}(h), \Psi^{(1)}(F(h)), \dots, \Psi^{(j)}(F(h)) \right), \end{aligned} \quad (\text{A.48})$$

where  $P_j$  is a polynomial: actually  $P_1(x)$  is just zero [cf. (A.43)] and  $P_2(x, y, z) = zx^2$  [cf. (A.45)], and that (A.48) has the companion

$$\begin{aligned} \left(\frac{d}{dh}\right)^j \Psi_{\text{cr}}(F_{\text{cr}}(h)) &= \Psi_{\text{cr}}^{(1)}(F_{\text{cr}}(h)) F_{\text{cr}}^{(j)}(h) \\ &+ P_j \left( F_{\text{cr}}^{(1)}(h), \dots, F_{\text{cr}}^{(j-1)}(h), \Psi_{\text{cr}}^{(1)}(F_{\text{cr}}(h)), \dots, \Psi_{\text{cr}}^{(j)}(F_{\text{cr}}(h)) \right). \end{aligned} \quad (\text{A.49})$$

But the expression in (A.49) is equal to one if  $j = 1$  and it is equal to zero if  $j \geq 2$ , while the analogous expression without the  $\text{cr}$  subscripts takes the value one, or minus one, as  $h \searrow 0$ . Therefore in this limit

$$\Psi^{(1)}(F(h)) - \Psi_{\text{cr}}^{(1)}(F_{\text{cr}}(h)) = O(1). \quad (\text{A.50})$$

For simplicity below we write  $P_j(F^{(1)}, \dots)$  and  $P_j(F_{\text{cr}}^{(1)}, \dots)$  for the more cumbersome complete expressions.

We start now the induction procedure that consists in obtaining that  $F^{(j)}(h) \sim F_{\text{cr}}^{(j)}(h)$  for all  $j < \hat{n}$ , with  $\hat{n} = 1 + 1/\alpha$  if  $1/\alpha \in \mathbb{N}$  and  $\hat{n} = \infty$  otherwise, knowing that  $F^{(k)}(h) \sim F_{\text{cr}}^{(k)}(h)$  for  $k = 0, 1, \dots, j-1$ . Note that this follows if we can show that

$$P_j(F^{(1)}(h), \dots) \stackrel{h \searrow 0}{\sim} P_j(F_{\text{cr}}^{(1)}(h), \dots), \quad (\text{A.51})$$

and that these expressions diverge in this limit. This is because by using (A.50) and the fact that the asymptotic behaviors of the functions in the right-hand sides of (A.48) and (A.49) are all known, except for  $F^{(j)}(h)$  that is then necessarily the same as the one of  $F_{\text{cr}}^{(j)}(h)$  and we are done.

Let us therefore establish (A.51) and the fact that the quantities in it diverge. That they diverge can be established by observing that for  $2 \leq j < \hat{n}$

$$P_j(F_{\text{cr}}^{(1)}(h), \dots) = -\Psi_{\text{cr}}^{(1)}(F_{\text{cr}}(h)) F_{\text{cr}}^{(j)}(h) = c h^{1-j}, \quad (\text{A.52})$$

where  $c \neq 0$  is an explicit constant (note that if  $1/\alpha \in \mathbb{N}$  then  $F_{\text{cr}}^{(j)}(h) = 0$  for  $j > 1/\alpha$ ). For what concerns (A.51) we start by observing that the leading behavior of each of the term constituting  $P_j(F^{(1)}, \dots)$  ( $P_j$  is of course a sum of monomials) is the same of the corresponding term in  $P_j(F_{\text{cr}}^{(1)}(h), \dots)$ . All these monomial terms are of the same order (in a strong sense: any ratio converge to a non-zero value): this can be either verified on the expression with or without the subscript  $\text{cr}$  (in a rather illogical way we verify it for the quantities without subscript: the proof is slightly more involved, since we need to use the induction assumption, but

formulas are lighter without subscripts). In fact in taking the derivative that builds  $P_j(\dots)$  we repeat two types of operations:

1. Taking a derivative of  $\Psi^k(F(h))$  for  $k \leq j$  and for this we have

$$\begin{aligned} \frac{d}{dh} \Psi^k(F(h)) &= \Psi^{k+1}(F(h)) F'(h) \stackrel{h \searrow 0}{\sim} -(j+1-\alpha) \Psi^k(F(h)) \frac{F'(h)}{F(h)} \\ &\stackrel{h \searrow 0}{\sim} - \left( \frac{j-1+\alpha}{\alpha} \right) \frac{\Psi^k(F(h))}{h}, \end{aligned} \quad (\text{A.53})$$

that is such a derivative makes the term more singular (by a factor  $1/h$ ).

2. Taking derivatives of  $F^{(k)}(h)$  for  $k = 1, \dots, j-2$  (by definition of  $P_j$ , the derivative of  $F^{(j-1)}(h)$  does not enter  $P_j$ ): but the asymptotic behaviors of  $F^{(k)}(h)$  and  $F^{(k+1)}(h)$  are in the induction assumption and one directly verifies that  $hF^{(k+1)}(h)/F^{(k)}(h)$  tends to a non-zero constant (recall that  $k < j < \hat{n}$ ). Once again, such a derivative asymptotically just introduces a multiplicative  $1/h$  factor (times a non-zero constant).

If we now recall that in the starting step ( $j = 2$ ) of the induction we had just two terms, cf. (A.45), and each of order  $1/h$  [cf. (A.45)] we see that  $P_j(F^{(1)}(h), \dots)$  (and  $P_j(F_{\text{cr}}^{(1)}(h), \dots)$ ) is a sum of terms of order  $h^{1-j}$ , so that (A.52) is telling us that the asymptotic behavior of  $P_j(F_{\text{cr}}^{(1)}(h), \dots)$  is of the same order of each of the monomial terms that constitute it (and it is not the result of the cancellation between the leading orders of larger terms). Therefore (A.51) holds for  $j < \hat{n}$  and the proof is complete.  $\square$

## References

1. S. Asmussen, *Applied Probability and Queues*, 2nd edn. (Springer, New York, 2003)
2. N.H. Bingham, C.M. Goldie, J.L. Teugels, *Regular Variation* (Cambridge University Press, Cambridge, 1987)
3. K.L. Chung, P. Erdős, Probability limit theorems assuming only the first moment I. Mem. Am. Math. Soc. **6**, 1–19 (1951), paper 3
4. R.A. Doney, One-sided local large deviation and renewal theorems in the case of infinite mean. Probab. Theory Relat. Fields **107**, 451–465 (1997)
5. W. Feller, *An Introduction to Probability Theory and Its Applications*, vol. II, 2nd edn. (Wiley, New York, 1971)
6. A. Garsia, J. Lamperti, A discrete renewal theorem with infinite mean. Comment. Math. Helv. **37**, 221–234 (1963)
7. G. Giacomin, *Random Polymer Models* (Imperial College Press, London, 2007)

# Index

- annealed bound, [32](#)
- annealed model, [33](#), [42](#), [63](#)
  
- backward recurrence time, [114](#)
- boundary condition, [6](#), [35](#), [117](#)
  
- change of measure estimates, [63](#), [64](#), [74](#), [88](#)
- charge, [29](#)
- charge distribution, [30](#)
- coarse graining, [41](#), [64](#), [68](#), [76](#), [91](#), [98](#)
- concentration properties, [36](#), [39](#), [102](#), [106](#), [108](#)
- contact density, [10](#)
- copolymer, [60](#), [109](#)
- correlation length, [19](#), [38](#), [56](#), [64](#), [76](#), [110](#)
- critical behavior, [18](#), [41](#), [42](#), [56](#), [111](#), [122](#)
- critical curve, [33](#)
- critical exponent, [43](#), [56](#)
- critical point, [19](#), [32](#), [34](#), [42](#), [56](#), [63](#)
  
- defect, [21](#)
- delocalization, [10](#), [32](#), [101](#)
- diluted Ising model, [41](#), [55](#), [58](#)
- directed polymer, [21](#)
- disordered pinning model, [29](#)
  
- entropy-energy competition, [35](#)
  
- fractional moment estimates, [64](#), [65](#), [87](#)
- free energy, [8](#), [11](#), [15](#), [19](#), [30](#), [32](#), [122](#)
- free pinning model, [6](#)
  
- Harris criterion, [41](#), [55](#)
- homogeneous pinning model, [14](#)
  
- inter-arrival law, [7](#), [14](#), [113](#)
- interface, [21](#)
- irrelevant disorder, [41](#), [46](#), [58](#), [110](#)
- Ising model, [21](#), [41](#), [55](#)
  
- localization, [10](#), [32](#), [101](#)
  
- marginal disorder, [42](#), [58](#)
- Markov chain, [113](#)
  
- null persistent renewal, [114](#)
  
- order of phase transition, [19](#)
  
- partition function, [6](#), [15](#), [29](#)
- path properties, [11](#), [17](#), [101](#)
- persistent renewal, [7](#), [45](#), [113](#)
- phase transition, [11](#), [18](#), [19](#), [56](#)
- pinning model, [5](#), [14](#), [29](#)
- Poland-Scheraga model, [24](#)
- polymer pinning, [21](#)
- positive persistent renewal, [7](#), [114](#)
- pure model, [41](#)
  
- quenched disorder, [29](#)
  
- random walk pinning model, [5](#)
- rare stretch strategy, [51](#)
- relative entropy, [54](#), [108](#)
- relevant disorder, [42](#), [58](#), [111](#)

- renewal function, [7](#)
- renewal process, [7](#), [113](#)
- renewal property, [113](#)
- renewal theorem, [114](#)
- renormalization, [41](#)
- replica, [44](#)
- self-averaging property, [32](#)
- smoothing inequality, [51](#), [60](#)
- super-additive property, [30](#), [34](#), [35](#), [48](#), [54](#)
- terminating renewal, [7](#), [44](#), [113](#)

# List of participants

## 40<sup>th</sup> Probability Summer School, Saint-Flour, France July 4–17, 2010

### Lecturers

Franco FLANDOLI	Università di Pisa, Italy
Giambattista GIACOMIN	Université Paris Diderot, France
Takashi KUMAGAI	Kyoto University, Japan

### Participants

Sergio ALMADA	Georgia Inst. Technology, Atlanta, USA
Marek ARENDARCZYK	Univ. Wroclaw, Poland
David BARBATO	Univ. Padova, Italy
David BELIUS	ETH Zurich, Switzerland
Pierre BERTIN	Univ. Paris 6 et 7, F
Luigi Amedeo BIANCHI	Scuola Normale Superiore, Pisa, Italy
Thomas BOUILLOC	Univ. Nice Sophia Antipolis, F
Omar BOUKHADRA	U. Provence, F & U. Constantine, Algeria
Charles-Edouard BREHIER	ENS Cachan, Antenne Bretagne, Rennes, F
Elisabetta CANDELLERO	Graz Univ. Technology, Austria
Francesco CARAVENNA	Univ. Padova, Italy
Reda CHHAIBI	Univ. Pierre et Marie Curie, Paris, F
Mirko D'OVIDIO	Sapienza Univ. Roma, Italy
Latifa DEBBI	Univ. Setif, Algeria
François DELARUE	Univ. Nice Sophia Antipolis, F
Francisco DELGADO	Univ. Barcelona, Spain
Aurélien DEYA	Univ. Henri Poincaré, Nancy, F
Hacène DJELLOUT	Univ. Blaise Pascal, Clermont-Ferrand, F
Aurélien EBERHARDT	Univ. Strasbourg, F
François EZANNO	Univ. de Provence, Marseille, F
Ennio FEDRIZZI	Univ. Paris Diderot, F
Matthieu FELSINGER	Univ. Bielefeld, Germany
Robert FITZNER	Eindhoven Univ. Technology, NL
Elena ISSOGLIO	Friedrich Schiller Univ., Jena, Germany



Shuai JING	Univ. Bretagne Occidentale, Brest, F
Yasmina KHELOUFI	Univ. Setif, Algeria
Konrad KOLESKO	Univ. Wroclaw, Poland
Noemi KURT	TU Berlin, Germany
Kazumasa KUWADA	Ochanomizu Univ., Tokyo, Japan
Mateusz KWASNICKI	Wroclaw Univ. Technology, Poland
Clément LAURENT	Univ. de Provence, Marseille, F
Qian LIN	Univ. Bretagne Occidentale, Brest, F
Arnaud LIONNET	Univ. Oxford, UK
J.-A. LOPEZ-MIMBELA	CIMAT, Guanajuato, Mexico
Eric LUÇON	Univ. Pierre et Marie Curie, Paris, F
Camille MALE	ENS Lyon, F
Mario MAURELLI	Univ. Pisa, Italy
Francesco MORANDIN	Univ. Parma, Italy
Jean-Christophe MOURRAT	Univ. de Provence, Marseille, F
Mikhail NEKLYUDOV	Univ. York, UK
Eyal NEUMANN	Technion Inst. Technology, Haifa, Israel
Harald OBERHAUSER	TU Berlin, Germany
Jean PICARD	Univ. Blaise Pascal, Clermont-Ferrand, F
K. PIETRUSKA-PALUBA	Univ. Warsaw, Poland
Marco ROMITO	Univ. Firenze, Italy
Erwan SAINT LOUBERT BIÉ	Univ. Blaise Pascal, Clermont-Ferrand, F
Martin SAUER	TU Darmstadt, Germany
Georg SCHOECHTEL	TU Darmstadt, Germany
Laurent SERLET	Univ. Blaise Pascal, Clermont-Ferrand, F
Yuhao SHEN	Univ. Pierre et Marie Curie, Paris, F
Mykhaylo SHKOLNIKOV	Stanford Univ., USA
Damien SIMON	Univ. Pierre et Marie Curie, Paris, F
Julien SOHIER	Univ. Paris Diderot, F
Philippe SOSOE	Princeton Univ., USA
Andrzej STOS	Univ. Blaise Pascal, Clermont-Ferrand, F
E. TODOROVA KOLKOVSKA	CIMAT, Guanajuato, Mexico
Dario VINCENZI	Univ. Nice Sophia Antipolis, F
Jing WANG	Purdue Univ., West Lafayette, USA
Frédérique WATBLED	Univ. Bretagne-Sud, Vannes, F
Hendrik WEBER	Univ. Warwick, UK
Lihu XU	Eindhoven Univ. Technology, NL
Ramon XULVI-BRUNET	Harvard Univ., Cambridge, MA, USA
Danyu YANG	Oxford Univ., UK
Lorenzo ZAMBOTTI	Univ. Pierre et Marie Curie, Paris, F

Edited by J.-M. Morel, B. Teissier; P.K. Maini

**Editorial Policy** (for the publication of monographs)

1. Lecture Notes aim to report new developments in all areas of mathematics and their applications - quickly, informally and at a high level. Mathematical texts analysing new developments in modelling and numerical simulation are welcome.  
Monograph manuscripts should be reasonably self-contained and rounded off. Thus they may, and often will, present not only results of the author but also related work by other people. They may be based on specialised lecture courses. Furthermore, the manuscripts should provide sufficient motivation, examples and applications. This clearly distinguishes Lecture Notes from journal articles or technical reports which normally are very concise. Articles intended for a journal but too long to be accepted by most journals, usually do not have this “lecture notes” character. For similar reasons it is unusual for doctoral theses to be accepted for the Lecture Notes series, though habilitation theses may be appropriate.
2. Manuscripts should be submitted either online at [www.editorialmanager.com/lnm](http://www.editorialmanager.com/lnm) to Springer’s mathematics editorial in Heidelberg, or to one of the series editors. In general, manuscripts will be sent out to 2 external referees for evaluation. If a decision cannot yet be reached on the basis of the first 2 reports, further referees may be contacted: The author will be informed of this. A final decision to publish can be made only on the basis of the complete manuscript, however a refereeing process leading to a preliminary decision can be based on a pre-final or incomplete manuscript. The strict minimum amount of material that will be considered should include a detailed outline describing the planned contents of each chapter, a bibliography and several sample chapters.  
Authors should be aware that incomplete or insufficiently close to final manuscripts almost always result in longer refereeing times and nevertheless unclear referees’ recommendations, making further refereeing of a final draft necessary.  
Authors should also be aware that parallel submission of their manuscript to another publisher while under consideration for LNM will in general lead to immediate rejection.
3. Manuscripts should in general be submitted in English. Final manuscripts should contain at least 100 pages of mathematical text and should always include
  - a table of contents;
  - an informative introduction, with adequate motivation and perhaps some historical remarks: it should be accessible to a reader not intimately familiar with the topic treated;
  - a subject index: as a rule this is genuinely helpful for the reader.

For evaluation purposes, manuscripts may be submitted in print or electronic form (print form is still preferred by most referees), in the latter case preferably as pdf- or zipped psfiles. Lecture Notes volumes are, as a rule, printed digitally from the authors’ files. To ensure best results, authors are asked to use the LaTeX2e style files available from Springer’s web-server at:

[ftp://ftp.springer.de/pub/tex/latex/svmonot1/](http://ftp.springer.de/pub/tex/latex/svmonot1/) (for monographs) and  
[ftp://ftp.springer.de/pub/tex/latex/svmult1/](http://ftp.springer.de/pub/tex/latex/svmult1/) (for summer schools/tutorials).

Additional technical instructions, if necessary, are available on request from [lnm@springer.com](mailto:lnm@springer.com).

4. Careful preparation of the manuscripts will help keep production time short besides ensuring satisfactory appearance of the finished book in print and online. After acceptance of the manuscript authors will be asked to prepare the final LaTeX source files and also the corresponding dvi-, pdf- or zipped ps-file. The LaTeX source files are essential for producing the full-text online version of the book (see <http://www.springerlink.com/openurl.asp?genre=journal&issn=0075-8434> for the existing online volumes of LNM). The actual production of a Lecture Notes volume takes approximately 12 weeks.
5. Authors receive a total of 50 free copies of their volume, but no royalties. They are entitled to a discount of 33.3 % on the price of Springer books purchased for their personal use, if ordering directly from Springer.
6. Commitment to publish is made by letter of intent rather than by signing a formal contract. Springer-Verlag secures the copyright for each volume. Authors are free to reuse material contained in their LNM volumes in later publications: a brief written (or e-mail) request for formal permission is sufficient.

**Addresses:**

Professor J.-M. Morel, CMLA,  
École Normale Supérieure de Cachan,  
61 Avenue du Président Wilson, 94235 Cachan Cedex, France  
E-mail: [morel@cmla.ens-cachan.fr](mailto:morel@cmla.ens-cachan.fr)

Professor B. Teissier, Institut Mathématique de Jussieu,  
UMR 7586 du CNRS, Équipe “Géométrie et Dynamique”,  
175 rue du Chevaleret  
75013 Paris, France  
E-mail: [teissier@math.jussieu.fr](mailto:teissier@math.jussieu.fr)

*For the “Mathematical Biosciences Subseries” of LNM:*

Professor P. K. Maini, Center for Mathematical Biology,  
Mathematical Institute, 24-29 St Giles,  
Oxford OX1 3LP, UK  
E-mail : [maini@maths.ox.ac.uk](mailto:maini@maths.ox.ac.uk)

Springer, Mathematics Editorial, Tiergartenstr. 17,  
69121 Heidelberg, Germany,  
Tel.: +49 (6221) 487-8259

Fax: +49 (6221) 4876-8259  
E-mail: [lnm@springer.com](mailto:lnm@springer.com)