

# Lecture-11: Key Renewal Theorem

## 1 Key Renewal Theorem

**Theorem 1.1 (Key renewal theorem).** Consider a recurrent renewal process  $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$  with renewal function  $m(t)$ , and the common mean and the distribution of i.i.d. inter-renewal times being denoted by  $\mu$  and  $F$  respectively. For any directly Riemann integrable function  $z \in \mathbb{D}$ , we have

$$\lim_{t \rightarrow \infty} \int_0^t z(t-x) dm(x) = \begin{cases} \frac{1}{\mu} \int_0^\infty z(t) dt, & F \text{ is non-lattice,} \\ \frac{d}{\mu} \sum_{k \in \mathbb{Z}_+} z(t+kd), & F \text{ is lattice with period } d, \quad t = nd. \end{cases}$$

**Proposition 1.2 (Equivalence).** Blackwell's theorem and key renewal theorem are equivalent.

*Proof.* Let's assume key renewal theorem is true. We select  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  as a simple function with value unity on interval  $[0, a]$  for  $a \geq 0$  and zero elsewhere. That is,  $z(t) = \mathbb{1}_{[0, a]}(t)$  for any  $t \in \mathbb{R}_+$ . From Proposition A.3, it follows that  $z$  is directly Riemann integrable. Therefore, by Key Renewal Theorem, we have

$$\lim_{t \rightarrow \infty} [m(t) - m(t-a)] = \frac{a}{\mu}.$$

We defer the formal proof of converse for a later stage. We observe that, from Blackwell theorem, it follows

$$\lim_{t \rightarrow \infty} \frac{dm(t)}{dt} \stackrel{(a)}{=} \lim_{a \rightarrow 0} \lim_{t \rightarrow \infty} \frac{m(t+a) - m(t)}{a} = \frac{1}{\mu}.$$

where in (a) we can exchange the order of limits under certain regularity conditions. □

*Remark 1.* Key renewal theorem is very useful in computing the limiting value of some function  $g(t)$ , probability or expectation of an event at an arbitrary time  $t$ , for a renewal process. This value is computed by conditioning on the time of last renewal prior to time  $t$ .

**Corollary 1.3 (Delayed key renewal theorem).** Consider an aperiodic and recurrent delayed renewal process  $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$  with independent inter-arrival times  $X : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$  with first inter-renewal time distribution  $G$  and common inter-renewal time distribution  $F$  for  $(X_n : n \geq 2)$ . Let the renewal function be denoted by  $m_D(t)$  and means  $\mathbb{E}X_1 = \mu_G$  and  $\mathbb{E}X_2 = \mu_F$ . For any directly Riemann integrable function  $z \in \mathbb{D}$  and  $F$  non-lattice, we have

$$\lim_{t \rightarrow \infty} \int_0^t z(t-x) dm_D(x) = \frac{1}{\mu_F} \int_0^\infty z(t) dt.$$

*Remark 2.* Any kernel function  $K(t) = P\{Z_t \in A, X_1 > t\} \leq \bar{F}(t)$ , and hence is d.R.i. from Proposition A.3(b).

**Example 1.4 (Limiting distribution of regenerative process).** For a regenerative process  $Z$  over a delayed renewal process  $S$  with finite mean i.i.d. inter-arrival times, we have  $K_2(t) = P\{Z_{S_1+t} \in A, X_2 > t\} \leq \bar{F}(t)$  for any  $A \in \mathcal{B}(\mathbb{R})$ , and hence the kernel function  $K_2 \in \mathbb{D}$ . Applying Key Renewal Theorem to renewal function, we get the limiting probability of the event  $\{Z_t \in A\}$  as

$$\lim_{t \rightarrow \infty} P\{Z_t \in A\} = \lim_{t \rightarrow \infty} (m_D * K_2)(t) = \frac{1}{\mu_F} \int_0^\infty K_2(t) dt.$$

**Example 1.5 (Limiting distribution of age and excess time).** For a delayed renewal process  $S$  with finite mean independent inter-renewal times such that the distribution of first renewal time is  $G$ , and the distribution of subsequent renewal times are identically  $F$ . Denoting the associated counting process by  $N_D$  and renewal function  $m_D$ , we can write the limiting probability distribution of age as  $F_e(x) \triangleq \lim_{t \rightarrow \infty} P\{A(t) \leq x\}$ . We can write the complementary distribution as

$$\bar{F}_e(x) = \lim_{t \rightarrow \infty} P\{A(t) \geq x\} = \lim_{t \rightarrow \infty} \int_0^t dm_D(t-y) \mathbb{1}_{\{y \geq x\}} \bar{F}(y) = \frac{1}{\mu_F} \int_x^\infty \bar{F}(y) dy.$$

**Example 1.6 (Limiting on probability of alternating renewal process).** Consider an alternating renewal process  $W$  with random on and off time sequence  $Z$  and  $Y$  respectively, such that  $(Z, Y)$  is *i.i.d.*. We denote the distribution of on and off times by non-lattice functions  $H$  and  $G$  respectively. If  $\mathbb{E}Z_n$  and  $\mathbb{E}Y_n$  are finite, then applying Key renewal theorem to the limiting probability of alternating process being on, we get

$$\lim_{t \rightarrow \infty} P(t) = \lim_{t \rightarrow \infty} (m * H)(t) = \frac{\mathbb{E}Z_n}{\mathbb{E}Z_n + \mathbb{E}Y_n}.$$

## A Directly Riemann Integrable

For each scalar  $h > 0$  and natural number  $n \in \mathbb{N}$ , we can define intervals  $I_n(h) \triangleq [(n-1)h, nh)$ , such that the collection  $(I_n(h), n \in \mathbb{N})$  partitions the positive real-line  $\mathbb{R}_+$ . For any function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  be a function bounded over finite intervals, we can denote the infimum and supremum of  $z$  in the interval  $I_n$  as

$$\underline{z}_n(h) \triangleq \inf\{z(t) : t \in I_n(h)\} \quad \bar{z}_n(h) \triangleq \sup\{z(t) : t \in I_n(h)\}.$$

We can define functions  $\underline{z}_h, \bar{z}_h : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  such that  $\underline{z}_h(t) \triangleq \sum_{n \in \mathbb{N}} \underline{z}_n(h) \mathbb{1}_{I_n(h)}(t)$  and  $\bar{z}_h(t) \triangleq \sum_{n \in \mathbb{N}} \bar{z}_n(h) \mathbb{1}_{I_n(h)}(t)$  for all  $t \in \mathbb{R}_+$ . From the definition, we have  $\underline{z}_h \leq z \leq \bar{z}_h$  for all  $h \geq 0$ . The infinite sums of infimum and supremums over all the intervals  $(I_n(h), n \in \mathbb{N})$  are denoted by

$$\int_{t \in \mathbb{R}_+} \underline{z}_h(t) dt = h \sum_{n \in \mathbb{N}} \bar{z}_h(n), \quad \int_{t \in \mathbb{R}_+} \bar{z}_h(t) dt = h \sum_{n \in \mathbb{N}} \underline{z}_h(n).$$

*Remark 3.* Since  $\underline{z}_h \leq z \leq \bar{z}_h$ , we observe that  $\int_{t \in \mathbb{R}_+} \underline{z}_h(t) dt \leq \int_{t \in \mathbb{R}_+} z(t) dt \leq \int_{t \in \mathbb{R}_+} \bar{z}_h(t) dt$ . If both left and right limits exist and are equal, then the integral value  $\int_{t \in \mathbb{R}_+} z(t) dt$  is equal to the limit.

**Definition A.1 (directly Riemann integrable (d.R.i.)).** A function  $z : \mathbb{R}_+ \mapsto \mathbb{R}_+$  is **directly Riemann integrable** and denoted by  $z \in \mathbb{D}$  if the partial sums obtained by summing the infimum and supremum of  $h$ , taken over intervals obtained by partitioning the positive axis, are finite and both converge to the same limit, for all finite positive interval lengths. That is,

$$\lim_{h \rightarrow 0} \int_{t \in \mathbb{R}_+} \underline{z}_h(t) dt = \lim_{h \rightarrow 0} \int_{t \in \mathbb{R}_+} \bar{z}_h(t) dt.$$

The limit is denoted by  $\int_{t \in \mathbb{R}_+} z(t) dt$ .

For a real function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}$ , we can define the positive and negative parts by  $z^+, z^- : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  such that for all  $t \in \mathbb{R}_+$   $z^+(t) \triangleq z(t) \vee 0$ , and  $z^-(t) \triangleq -(z(t) \wedge 0)$ . If both  $z^+, z^- \in \mathbb{D}$ , then  $z \in \mathbb{D}$  and the limit is

$$\int_{\mathbb{R}_+} z(t) dt \triangleq \int_{\mathbb{R}_+} z^+(t) dt - \int_{\mathbb{R}_+} z^-(t) dt.$$

*Remark 4.* We compare the definitions of directly Riemann integrable and Riemann integrable functions. For a finite positive  $M$ , a function  $z : [0, M] \rightarrow \mathbb{R}$  is Riemann integrable if

$$\lim_{h \rightarrow 0} \int_0^M \underline{z}_h(t) dt = \lim_{h \rightarrow 0} h \int_0^M \underline{z}_h(t) dt.$$

In this case, the limit is the value of the integral  $\int_0^M z(t)dt$ . For a function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}$ ,

$$\int_{t \in \mathbb{R}_+} z(t)dt = \lim_{M \rightarrow \infty} \int_0^M z(t)dt,$$

if the limit exists. For many functions, this limit may not exist.

*Remark 5.* A directly Riemann integrable function over  $\mathbb{R}_+$  is also Riemann integrable, but the converse need not be true. For instance, for  $E_n \triangleq \left[ n - \frac{1}{2n^2}, n + \frac{1}{2n^2} \right]$  for each  $n \in \mathbb{N}$ , consider the following Riemann integrable function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$

$$z(t) = \sum_{n \in \mathbb{N}} \mathbb{1}_{E_n}(t), \quad t \in \mathbb{R}_+.$$

We observe that  $z$  is Riemann integrable, but  $\int_{t \in \mathbb{R}_+} \bar{z}_h(t)dt$  is always infinite for every  $h > 0$ .

**Proposition A.2 (Necessary conditions for d.R.i.).** *If a function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is directly Riemann integrable, then  $z$  is bounded and continuous a.e.*

**Proposition A.3 (Sufficient conditions for d.R.i.).** *A function  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is directly Riemann integrable, if any of the following conditions hold.*

- (a)  $z$  is monotone non-increasing, and Lebesgue integrable.
- (b)  $z$  is bounded above by a directly Riemann integrable function.
- (c)  $z$  has bounded support.
- (d)  $\int_{t \in \mathbb{R}_+} \bar{z}_h dt$  is bounded for some  $h > 0$ .

**Proposition A.4 (Tail Property).** *If  $z : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is directly Riemann integrable and has bounded integral value, then  $\lim_{t \rightarrow \infty} z(t) = 0$ .*

**Corollary A.5.** *Any distribution  $F : \mathbb{R}_+ \rightarrow [0, 1]$  with finite mean  $\mu$ , the complementary distribution function  $\bar{F}$  is d.R.i.*

*Proof.* Since  $\bar{F}$  is monotonically non-increasing and its Lebesgue integration is  $\int_{\mathbb{R}_+} \bar{F}(t)dt = \mu$ , the result follows from Proposition A.3(a).  $\square$

## B Chebyshev's sum inequality

**Lemma B.1.** *Let  $f : \mathbb{R} \rightarrow \mathbb{R}_+$  and  $g : \mathbb{R} \rightarrow \mathbb{R}_+$  be arbitrary functions with the same monotonicity. For any random variable  $X$ , functions  $f(X)$  and  $g(X)$  are positive and*

$$\mathbb{E}[f(X)g(X)] \geq \mathbb{E}[f(X)]\mathbb{E}[g(X)].$$

*Proof.* Let  $Y$  be a random variable independent of  $X$  and with the same distribution. Then,

$$(f(X) - f(Y))(g(X) - g(Y)) \geq 0.$$

Taking expectation on both sides the result follows.  $\square$