

Lecture-08: Distribution and renewal functions

1 Convolution of distribution functions

Definition 1.1. For two distribution functions $F, G : \mathbb{R} \rightarrow [0, 1]$ the convolution of F and G is a distribution function $F * G : \mathbb{R} \rightarrow [0, 1]$ defined as

$$(F * G)(x) \triangleq \int_{y \in \mathbb{R}} F(x - y) dG(y), \quad x \in \mathbb{R}.$$

Lemma 1.2. Let $F, G \in [0, 1]^{\mathbb{R}}$, then the convolution $F * G$ is a distribution function.

Proof. It suffices to show that the function $(F * G)$ is

- (a) right continuous, i.e. $\lim_{x_n \downarrow x} (F * G)(x_n)$ exists,
 - (b) non-decreasing, i.e. $(F * G)(z) \geq (F * G)(x)$ for all $z \geq x$,
 - (c) having left limit of zero and right limit of unity, i.e. $\lim_{x \rightarrow -\infty} (F * G)(x) = 0, \lim_{x \rightarrow \infty} (F * G)(x) = 1$.
- Part (a) and (c) can be verified by exchanging limit and integration using Monotone convergence theorem. Part (b) can be verified from monotonicity of integration. \square

Lemma 1.3. Convolution is a symmetric and bi-linear operator.

Proof. We note that for any $F, G \in [0, 1]^{\mathbb{R}}$, we have $(F * G) \triangleq F * G$.

- (a) **Symmetry.** Let $F, G \in [0, 1]^{\mathbb{R}}$ be distribution functions. Then, it suffices to show that $F * G = G * F$. To this end, we observe from Fubini's theorem for nonnegative functions, exchanging order of integration, and change of variables, that

$$\int_{y \in \mathbb{R}} F(x - y) dG(y) = \int_{y \in \mathbb{R}} \int_{u \leq x - y} dF(u) dG(y) = \int_{u \in \mathbb{R}} dF(u) \int_{y \leq x - u} dG(y) = \int_{u \in \mathbb{R}} dF(u) G(x - u).$$

- (b) **Bilinearity.** It suffices to show for any two finite sets of distribution functions $(F_i \in [0, 1]^{\mathbb{R}} : i \in [n])$ and $(G_j \in [0, 1]^{\mathbb{R}} : j \in [m])$ and vectors $\alpha \in \mathbb{R}^n, \beta \in \mathbb{R}^m$, we have

$$\left(\sum_{i \in [n]} \alpha_i F_i \right) * \left(\sum_{j \in [m]} \beta_j G_j \right) = \sum_{i \in [n]} \sum_{j \in [m]} \alpha_i \beta_j (F_i * G_j).$$

This follows from the linearity of integration in its arguments. \square

Lemma 1.4. Let X and Y be two independent random variables defined on the probability space (Ω, \mathcal{F}, P) with distribution functions F and G respectively, then the distribution of $X + Y$ is given by $F * G$.

Proof. The distribution function of sum $X + Y$ is denoted by $H \in [0, 1]^{\mathbb{R}}$ where $H(z) \triangleq \mathbb{E} \mathbb{1}_{\{X+Y \leq z\}}$ for any $z \in \mathbb{R}$. It follows from the tower property of conditional expectation and independence of X and Y that

$$H(z) = \mathbb{E}[\mathbb{E}[\mathbb{1}_{\{X+Y \leq z\}} | \sigma(Y)]] = \mathbb{E}[F(z - Y)] = \int_{y \in \mathbb{R}_+} F(z - y) dG(y).$$

\square

Definition 1.5. Consider a real valued random walk $S : \Omega \rightarrow \mathbb{R}^{\mathbb{N}}$ with *i.i.d.* step size sequence $X : \Omega \rightarrow \mathbb{R}^{\mathbb{N}}$, defined as $S_n \triangleq \sum_{i=1}^n X_i$ for all $n \in \mathbb{N}$. We denote the distribution of X_1 by F and the distribution of S_n by F_n , i.e. $F_n(t) \triangleq P\{S_n \leq t\}$ for all $t \in \mathbb{R}$.

Remark 1. The distribution F_n is computed inductively as $F_n = F_{n-1} * F$ for all $n \geq 2$, where $F_1 = F$.

Remark 2. For a renewal sequence $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ with *i.i.d.* inter renewal time sequence $X : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ having a common distribution $F \in [0, 1]^{\mathbb{R}_+}$, the distribution function of the n th renewal instant is the n -fold convolution F_n of the distribution function F .

Example 1.6 (Poisson process). Consider a renewal sequence $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ with *i.i.d.* inter renewal times $X : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ having a common exponential distribution $F \in [0, 1]^{\mathbb{R}_+}$ defined as $F(x) \triangleq 1 - e^{-\lambda x}$ for any $x \in \mathbb{R}_+$ and rate $\lambda \in \mathbb{R}_+$. We will show by induction that the distribution of n th renewal instant at any time $t \in \mathbb{R}_+$ is

$$F_n(t) \triangleq \int_0^t \frac{\lambda^n s^{n-1}}{(n-1)!} e^{-\lambda s} ds.$$

We first observe that the statement holds true for base case of $n = 1$, since $F_1(t) = \int_0^t \lambda e^{-\lambda s} ds = 1 - e^{-\lambda t} = F(t)$ for all $t \in \mathbb{R}_+$. We assume that the hypothesis holds true for step $n - 1$, and compute $F_n = F_{n-1} * F$ written as

$$\begin{aligned} F_n(t) &= \int_{s=0}^t F(t-s) dF_{n-1}(s) = \lambda^{n-1} \int_0^t (e^{-\lambda s} - e^{-\lambda t}) d \frac{s^{n-1}}{(n-1)!} \\ &= (e^{-\lambda s} - e^{-\lambda t}) \frac{s^{n-1}}{(n-1)!} \Big|_{s=0}^t + \int_0^t \frac{\lambda^n s^{n-1}}{(n-1)!} e^{-\lambda s} ds. \end{aligned}$$

The equality in the second line of the above equation follows from the integration by parts.

Corollary 1.7. The distribution function of n th arrival instant S_n for delayed renewal sequence is $G * F_{n-1}$.

Corollary 1.8. The distribution function of counting process $N^D : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{R}_+}$ for the delayed renewal sequence is

$$P \{N_t^D = n\} = P \{S_n \leq t\} - P \{S_{n+1} \leq t\} = (G * F_{n-1})(t) - (G * F_n)(t).$$

2 Renewal functions

Definition 2.1. Mean of the counting process $N : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{R}_+}$ is called the **renewal function** denoted by $m : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ defined by $m_t \triangleq \mathbb{E}[N_t]$ for all $t \in \mathbb{R}_+$.

Proposition 2.2. Renewal function $m \in \mathbb{R}_+^{\mathbb{R}_+}$ for a renewal process $N : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{R}_+}$ is $m_t = \sum_{n \in \mathbb{N}} F_n(t)$ for all $t \in \mathbb{R}_+$, where the distribution of renewal instant S_n is denoted by $F_n \in [0, 1]^{\mathbb{R}_+}$ for each $n \in \mathbb{N}$.

Proof. Using the inverse relationship between counting process and the arrival instants, we can write

$$m_t = \mathbb{E}[N_t] = \sum_{n \in \mathbb{N}} P \{N_t \geq n\} = \sum_{n \in \mathbb{N}} P \{S_n \leq t\} = \sum_{n \in \mathbb{N}} F_n(t).$$

For the second inequality in the above equation, we observe that $\mathbb{E}N_t = \mathbb{E} \sum_{m \in \mathbb{N}} \sum_{n=1}^m \mathbb{1}_{\{N_t=m\}}$. Switching the order of summations and using monotone convergence theorem to exchange expectation and infinite sum, we obtain $\mathbb{E}N_t = \mathbb{E} \sum_{n \in \mathbb{N}} \sum_{m \geq n} \mathbb{1}_{\{N_t=m\}} = \mathbb{E} \sum_{n \in \mathbb{N}} \mathbb{1}_{\{N_t \geq n\}} = \mathbb{E} \sum_{n \in \mathbb{N}} P \{N_t \geq n\}$. \square

Example 2.3 (Poisson process). Consider the renewal sequence $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ with *i.i.d.* inter renewal times $X : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ having a common exponential distribution $F \in [0, 1]^{\mathbb{R}_+}$ defined as $F(x) \triangleq 1 - e^{-\lambda x}$ for any $x \in \mathbb{R}_+$ and rate $\lambda \in \mathbb{R}_+$. The associate renewal function with this renewal sequence is

$$m_t = \sum_{n \in \mathbb{N}} F_n(t) = \int_0^t \lambda \left(e^{-\lambda s} \sum_{n \in \mathbb{Z}_+} \frac{(\lambda s)^n}{n!} \right) ds = \int_0^t \lambda ds = \lambda t.$$

Corollary 2.4. The renewal function $m^D : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ for a delayed renewal process $N^D : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{R}_+}$ with distribution $G \in [0, 1]^{\mathbb{R}_+}$ for the first inter renewal time X_1 and common distribution $F \in [0, 1]^{\mathbb{R}_+}$ for inter renewal times X_n for $n \geq 2$, is given by $m^D = G + G * m$.

Proof. We can write the renewal function for the delayed renewal process as $m_t^D = \mathbb{E}N_t^D = \sum_{n \in \mathbb{N}} (G * F_{n-1})(t) = G(t) + (G * m)(t)$. \square

Remark 3. If $G = F$, then we have the identity $m = F + F * m$.

3 Laplace transform of distribution functions and renewal functions

Definition 3.1. The Laplace transform $\mathcal{L} : [0, 1]^{\mathbb{R}} \rightarrow \mathbb{C}^{\mathbb{C}}$ for a distribution function $F \in [0, 1]^{\mathbb{R}}$ is a map $\mathcal{L}_F \in \mathbb{C}^{\mathbb{C}}$ defined for all $s \in \mathbb{C}$ such that $|\mathcal{L}_F(s)| < \infty$, as

$$\mathcal{L}_F(s) \triangleq \int_{y \in \mathbb{R}} e^{-sy} dF(y).$$

Remark 4. If $X : \Omega \rightarrow \mathbb{R}$ is a random variable with distribution function F , then $\mathcal{L}_F(s) = \mathbb{E}e^{-sX}$.

Lemma 3.2. The Laplace transform of convolution of two distribution functions is product of Laplace transform of individual distribution functions, i.e. if $F, G \in [0, 1]^{\mathbb{R}}$ are distribution functions, then $\mathcal{L}_{F * G} = \mathcal{L}_F \mathcal{L}_G$.

Proof. Let $F, G \in [0, 1]^{\mathbb{R}}$ be two distribution functions, then $d(F * G)(x) = \int_{y \in \mathbb{R}} dF(x - y) dG(y)$ from exchange of limits and integration using monotone convergence theorem. Therefore,

$$\mathcal{L}_{F * G}(s) = \int_{x \in \mathbb{R}} e^{-sx} \int_{y \in \mathbb{R}} dF(x - y) dG(y) = \int_{y \in \mathbb{R}} e^{-sy} dG(y) \int_{x - y \in \mathbb{R}} e^{-s(x - y)} dF(x - y) = \mathcal{L}_F \mathcal{L}_G.$$

Alternatively, consider two independent random variables X, Y with distributions F, G respectively. Then the distribution of $X + Y$ is $F * G$ and $\mathcal{L}_{F * G}(s) = \mathbb{E}e^{-s(X + Y)} = \mathbb{E}e^{-sX} \mathbb{E}e^{-sY} = (\mathcal{L}_F \mathcal{L}_G)(s)$ for any $s \in \mathbb{C}$ such that $|\mathcal{L}_F(s)| |\mathcal{L}_G(s)| < \infty$. \square

Remark 5. Consider a renewal sequence $S : \Omega \rightarrow \mathbb{R}_+^{\mathbb{N}}$ with i.i.d. inter renewal time sequence $X : \Omega \rightarrow \mathbb{R}^{\mathbb{N}}$ having a common distribution function F . The Laplace transform of the distribution of n th renewal instant S_n is $\mathcal{L}_{F_n} = \mathcal{L}_F^n$.

Corollary 3.3. The Laplace transform of the renewal function $m \in \mathbb{R}_+^{\mathbb{R}^+}$ is given by $\mathcal{L}_m = \frac{\mathcal{L}_F}{1 - \mathcal{L}_F}$ defined for each $s \in \mathbb{C}$ such that $|\mathcal{L}_F(s)| < 1$.

Corollary 3.4. The Laplace transform of the renewal function m^D for the delayed renewal process is $\mathcal{L}_{m^D} = \frac{\mathcal{L}_G}{1 - \mathcal{L}_F}$ defined for each $s \in \mathbb{C}$ such that $|\mathcal{L}_F(s)| < 1$.

Example 3.5 (Poisson process). The Laplace transform of an exponential distribution $F \in [0, 1]^{\mathbb{R}_+}$ defined as $F(x) \triangleq 1 - e^{-\lambda x}$ for $x \in \mathbb{R}_+$ and rate $\lambda \in \mathbb{R}_+$ is given by $\mathcal{L}_F(s) = \frac{\lambda}{\lambda + s}$ for $\Re(s) > -\lambda$. Consider a renewal sequence S with i.i.d. inter renewal times having the common exponential distribution F . The Laplace transform for the distribution F_n of n th renewal instant is given by $\mathcal{L}_{F_n}(s) = \left(1 + \frac{s}{\lambda}\right)^{-n}$ for all $s \in \mathbb{C}$ such that $-\Re(s) < \lambda$. The Laplace transform for the renewal function for renewal sequence S is

$$\mathcal{L}_m(s) = \frac{\mathcal{L}_F(s)}{1 - \mathcal{L}_F(s)} = \frac{\lambda}{s} \text{ for all } s \in \{r \in \mathbb{C} : -\Re(r) < \lambda < |\lambda + r|\}.$$

That is, if $s = \sigma + j\omega$ for $\sigma, \omega \in \mathbb{R}$, then $-\sigma < \lambda$ and $\lambda^2 < (\lambda + \sigma)^2 + \omega^2$. We observe that it suffices that $\sigma > 0$ or $\omega > \lambda$.

Exercise 3.6. Invert the Laplace transform $\mathcal{L}_{F_n}(s) = \left(1 + \frac{s}{\lambda}\right)^{-n}$ in the region of convergence $\{s \in \mathbb{C} : -\Re(r) < \lambda < |\lambda + s|\}$ to obtain the distribution function F_n for the n th arrival instant of a Poisson process with rate λ .

Proposition 3.7. For renewal sequence S with i.i.d. inter renewal times having positive common mean $\mathbb{E}X_1 > 0$, the renewal function is bounded for all finite times.

Proof. Since $\mathbb{E}X_1 > 0$ and $X_1 \geq 0$, it follows that $P\{X_1 = 0\} < 1$. From the continuity of probability, there exists $\alpha > 0$ and $\beta \in (0, 1)$, such that $P\{X_n \geq \alpha\} = \beta$. We define a map $g_\alpha : \mathbb{R}_+ \rightarrow \{0, \alpha\}$ such that $g_\alpha(x) \triangleq \alpha \mathbb{1}_{\{x \geq \alpha\}}$ for all $x \in \mathbb{R}_+$. We observe that $g_\alpha(x) \leq x$. We can define bivariate random sequence $\bar{X} : \Omega \rightarrow \{0, \alpha\}^{\mathbb{N}}$ where $\bar{X}_n \triangleq g_\alpha(X_n)$ for all $n \in \mathbb{N}$. It follows that \bar{X} is i.i.d. with probability mass function $P\{\bar{X}_1 = 0\} = P\{X_1 < \alpha\} = 1 - \beta$ and $P\{\bar{X}_1 = \alpha\} = \beta$. Further, since $g_\alpha(x) \leq x$, it follows that

$\bar{X}_n \leq X_n$. It follows that $\bar{S}_n \triangleq \sum_{i=1}^n \bar{X}_i \leq \sum_{i=1}^n X_i = S_n$, and hence $\{S_n \leq t\} \subseteq \{\bar{S}_n \leq t\}$ for each $n \in \mathbb{N}$ and $t \in \mathbb{R}_+$. Let $\bar{N} : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{R}_+}$ denote the renewal counting process with *i.i.d.* inter arrival time sequence $\bar{X} : \Omega \rightarrow \{0, \alpha\}^{\mathbb{N}}$ with arrivals $\bar{S} : \Omega \rightarrow \{d\alpha : d \in \mathbb{Z}_+\}^{\mathbb{N}}$ at integer multiples of α . Then, it follows that for all sample paths and all times $t \in \mathbb{R}_+$,

$$N_t = \sum_{n \in \mathbb{N}} \mathbb{1}_{\{S_n \leq t\}} \leq \sum_{n \in \mathbb{N}} \mathbb{1}_{\{\bar{S}_n \leq t\}} = \bar{N}_t.$$

Hence, it follows that $m_t \triangleq \mathbb{E}N_t \leq \mathbb{E}\bar{N}_t$, and to show finiteness of m_t it suffices to show that renewal function $\bar{m}_t \triangleq \mathbb{E}\bar{N}_t$ associated with quantized inter renewal time is finite at any time $t \in \mathbb{R}_+$.

We observe that $\{\bar{N}_0 = n_0\} = \{\bar{N}_0 = n_0, \bar{N}_\alpha \geq 1\} = \cap_{i=1}^{n_0} \{\bar{X}_i = 0\} \cap \{\bar{X}_{n_0+1} = \alpha\}$. We define $M_i \triangleq \sum_{j=0}^i n_j$ and observe that $\{\bar{S}_n - \bar{S}_k = 0\} = \cap_{i=k+1}^n \{\bar{X}_i = 0\}$. Hence, we write

$$\begin{aligned} \{\bar{N}_0 = n_0, \bar{N}_\alpha = n_1\} &= \{\bar{N}_0 = n_0, \bar{N}_\alpha = n_1, \bar{N}_{2\alpha} \geq 1\} = \{\bar{S}_{M_0} = 0, \bar{S}_{M_0+1} = \alpha\} \cap \{\bar{S}_{M_1} = \alpha, \bar{S}_{M_1+1} = 2\alpha\} \\ &= \bigcap_{i=1}^{M_0} \{\bar{X}_i = 0\} \cap \{\bar{X}_{M_0+1} = \alpha\} \bigcap_{i=M_0+2}^{M_1} \{\bar{X}_i = 0\} \cap \{\bar{X}_{M_1+1} = \alpha\}. \end{aligned}$$

We can write the joint event of number of arrivals n_i at each arrival instant in $i\alpha$ for $i \in \{0, \dots, k-1\}$, as

$$\begin{aligned} \bigcap_{i=0}^{k-1} \{\bar{N}_{i\alpha} = n_i\} &= \bigcap_{i=0}^{k-1} \{\bar{N}_{i\alpha} = n_i\} \cap \{\bar{N}_{k\alpha} \geq 1\} = \bigcap_{i=0}^{k-1} \{\bar{S}_{M_i} = i\alpha, \bar{S}_{M_i+1} = (i+1)\alpha\} \\ &= \bigcap_{i=1}^{M_0} \{\bar{X}_i = 0\} \cap \{\bar{X}_{M_0+1} = \alpha\} \bigcap_{j=1}^{k-1} \left(\bigcap_{i=M_{j-1}+2}^{M_j} \{\bar{X}_i = 0\} \cap \{\bar{X}_{M_j+1} = \alpha\} \right). \end{aligned}$$

It follows that the joint distribution of number of arrivals at first k arrival instants is

$$P \left(\bigcap_{i=0}^{k-1} \{\bar{N}_{i\alpha} = n_i\} \right) = (1 - \beta) \prod_{i=0}^{k-1} (\beta)(1 - \beta)^{n_i-1}.$$

It follows that the number of arrivals is independent at each arrival instant $k\alpha$ and geometrically distributed over \mathbb{N} with mean $1/\beta$ for $k \in \mathbb{N}$ and over \mathbb{Z}_+ with mean $(1 - \beta)/\beta$ for $k = 0$ respectively. Thus, for all $t \geq 0$,

$$\mathbb{E}N_t \leq \mathbb{E}\bar{N}_t \leq \frac{\lceil \frac{t}{\alpha} \rceil}{\beta} \leq \frac{\frac{t}{\alpha} + 1}{\beta} < \infty.$$

□

Corollary 3.8. *For delayed renewal sequence with $\mathbb{E}X_2 > 0$, the renewal function is bounded at all finite times.*