

Consider a Poisson process with rate λ and let each event be classified as a type I or type II event with probability p and $1-p$ respectively. The classification is independent of the events in the Poisson process. Let $N_1(t)$ and $N_2(t)$ count the number of type I and type II customers.

Graphically

Proposition 5.2. $\{N_1(t), t \geq 0\}$ and $\{N_2(t), t \geq 0\}$ are both Poisson processes with rates $p\lambda$ and $(1-p)\lambda$ respectively. The two processes are independent.

mar. 6-10:13

Proof. Will verify that $\{N_1(t), t \geq 0\}$ is a Poisson process with rate $p\lambda$ by checking the definition of a Poisson process.

(i) $N_1(0) = 0$ since $N(0) = N_1(0) + N_2(0)$

(ii) $N_1(t)$ has stationary and independent increments: Conditional on the number of events in an interval, one finds the distribution of type I events. This will only depend on the length of the interval and is independent of what happens in other intervals.

(iii) $P(N_1(t) = 1) = P(N_1(t) = 1 | N(t) = 1) \cdot P(N(t) = 1) + P(N_1(t) = 1 | N(t) \geq 2) \cdot P(N(t) \geq 2)$
 $= p \cdot [p\lambda t + o(t)] + o(t) = p\lambda t + o(t)$

(iv) $P(N_1(t) \geq 2) \leq P(N(t) \geq 2) = o(t)$

Similarly, $\{N_2(t), t \geq 0\}$ is a Poisson process with rate $(1-p)\lambda$.

mar. 6-10:25

Independence? Events of type I in an interval is independent of events in non-overlapping intervals, and therefore also of whether the events are of type I or II.

Example: Call to a bank, 30 calls per hour

type I: questions about accounts 80%

type II: loans

The number of questions about loan in 3 hours is Poisson distributed with $3 \cdot 30 \cdot 0.8 = 18$

mar. 6-10:35

Example: Company strategy for accepting offers

- Offers arrive as a Poisson process, rate λ
- Value of random offer, $X \sim f(x)$, cdf $F(x)$
- Strategy: Accept or reject and wait for new offer
- cost of rate c per unsold items
- accept if offer large than critical value y

Value of y maximizing return?
 Maximal expected return?

Fix y , offers greater than y arrive at rate $\lambda(1-F(y))$ (Proposition 5.2)

$R(y)$ return.

$$E[R(y)] = E[\text{accepting offer}] - cE[\text{time to accept}]$$

$$= E[X | X > y] - c \cdot \frac{1}{\lambda(1-F(y))}$$

mar. 6-10:41

$$= \int_y^\infty x f(x) dx - \frac{c}{\lambda(1-F(y))}$$

$$= \int_y^\infty \frac{x \cdot f(x)}{1-F(y)} dx - \frac{c}{\lambda(1-F(y))} \cdot \frac{1}{1-F(y)} \left[\int_y^\infty x f(x) dx - \frac{c}{\lambda} \right]$$

First order conditions

$$\frac{\partial}{\partial y} R(y) = \frac{c}{\lambda(1-F(y))^2} (-f(y)) \left[\int_y^\infty x f(x) dx - \frac{c}{\lambda} \right] - \frac{1}{\lambda(1-F(y))} y f(y) = 0$$

or $y [1-F(y)] = \int_y^\infty x f(x) dx - \frac{c}{\lambda}$

$y \int_y^\infty f(x) dx = \int_y^\infty x f(x) dx - \frac{c}{\lambda}$

so $\int_y^\infty (x-y) f(x) dx = \frac{c}{\lambda}$

which has a unique solution y^*

mar. 6-10:49

$$E[R(y^*)] = \frac{1}{1-F(y^*)} \left[\int_{y^*}^\infty (x-y^*) f(x) dx - \frac{c}{\lambda} \right]$$

$$= \frac{1}{1-F(y^*)} \cdot \left[\int_{y^*}^\infty (x-y^*) f(x) dx + y^* \int_{y^*}^\infty f(x) dx - \frac{c}{\lambda} \right]$$

$$= y^*$$

Optimal expected return = optimal critical value.

mar. 6-10:59

Example

• Particles at each node
 • Moves according to P_{ij} at each event
 • At time 0 number in state $i, i=1, \dots, n \sim PD(X_i)$
 $N_j(t)$ is number initially at state i and state j at time t .
 $N_j(t) \sim \text{Poisson}(\lambda_i P_{ij}^n)$
 and $\sum_{j=1}^m N_j(t) \sim \text{Poisson}(\sum_{i=1}^m \lambda_i P_{ij}^n)$

mar. 6-11:21

Example, coupon collector's problem

- m different coupons
- new coupon type j with probability p_j
 $\sum_{j=1}^m p_j = 1$
- N number coupons needed for a full collection.

$E(N) = ?$
 N_j number of coupons needed to get a coupon of type j
 N_j geometric distributed, $N_j \sim \text{Geo}(p_j)$
 so $N = \max(N_j)$ but N_j 's dependent.

Alternative: Coupons are collected at times following a Poisson process, rate $\lambda = 1$
 $N_j(t), j=1, \dots, m$ associated counting process for m types
 X_j first event process $N_j(t)$
 $X = \max_{1 \leq j \leq m} (X_j)$ time to a complete collection

mar. 6-11:27

$P(X < t) = P(\max X_j < t) = P(\text{all } X_j < t)$
 $= \prod_{j=1}^m (1 - e^{-\lambda_j t})$
 and $E[X] = \int_0^\infty x f(x) dx = \int_0^\infty \int_0^x \underbrace{I[x > t]}_{P(x > t)} dt dx$
 $= \int_0^\infty \int_0^\infty \underbrace{I[x > t]}_{P(x > t)} dx dt$
 $= \int_0^\infty [1 - \prod_{j=1}^m (1 - e^{-\lambda_j t})] dt$

Let T_1, T_2, \dots be interarrival times in process counting coupons
 $X = T_1 + \dots + T_N$
 N independent of $T_1, T_2, \dots \sim \text{Exp}(\lambda)$
 $\lambda = 1$.

mar. 6-11:34

$E[X|N] = N \cdot E[X] = N$
 $E[X] = E[E[X|N]] = E(N)$

Let $I_i = \begin{cases} 1 & \text{only 1 coupon of type } i \text{ in collection} \\ 0 & \text{else} \end{cases}$

$E[\sum_{i=1}^m I_i] = \sum_{i=1}^m P(I_i = 1) = ?$
 Define S_i second time a coupon type i appear
 Then $I_i = 1 \Leftrightarrow X_j < S_i \forall j \neq i$
 S_i is gamma(2, p_i)

$P(I_i = 1) = \int_0^\infty P(X_j < S_i \forall j \neq i | S_i = x) \cdot p_i \cdot x \cdot e^{-p_i x} dx$
 $= \int_0^\infty P(X_j < x \forall j \neq i) \cdot p_i^2 x e^{-p_i x} dx$

mar. 6-11:40

$= \int_0^\infty \prod_{j \neq i} (1 - e^{-p_j x}) \cdot p_i^2 x e^{-p_i x} dx$
 $= \int_0^\infty x \cdot \prod_{j=1}^m (1 - e^{-p_j x}) \cdot p_i^2 \cdot \frac{e^{-p_i x}}{1 - e^{-p_i x}} dx$

$E(I_i)$

mar. 6-11:49

Consider two independent Poisson processes rates λ_1 and λ_2
 Let S_n^i be time of n th event in i th process

Then $P(S_1^1 < S_1^2) = \frac{\lambda_1}{\lambda_1 + \lambda_2}$

But $P(S_2^1 < S_2^2) = ?$ i.e. two events in process 1 before first event process 2.

Use memoryless property. The processes start over again after first arrival of in process 1. so therefore $P(S_2^1 < S_2^2) = \frac{\lambda_1}{\lambda_1 + \lambda_2} \cdot \frac{\lambda_1}{\lambda_1 + \lambda_2} = \frac{\lambda_1^2}{(\lambda_1 + \lambda_2)^2}$

n events of type 1 before m events of type 2
 \Leftrightarrow n or more successes

$P(S_n^1 < S_m^2) = \sum_{k=n}^{n+m-1} \binom{n+m-1}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n+m-1-k}$
 $= \sum_{k=n}^{n+m-1} \binom{n+m-1}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n+m-1-k}$

mar. 6-11:53