

Poisson Process

SI252 Reinforcement Learning

School of Information Science and Technology
ShanghaiTech University

March 25, 2020

Outline

1 Poisson Distribution

2 Poisson Process

Poisson Distribution

Definition

An r.v. X has the **Poisson distribution** with parameter λ if the PMF of X is

$$P(X = k) = \frac{e^{-\lambda} \lambda^k}{k!}, \quad k = 0, 1, 2, \dots$$

We write this as $X \sim \text{Pois}(\lambda)$.

Example: Poisson Expectation & Variance

Example (Poisson Expectation & Variance)

Consider an r.v. $X \sim \text{Pois}(\lambda)$, find $\mathbb{E}(X)$ and $\text{Var}(X)$.

Example: Poisson Expectation & Variance

(Solution)

Mean:

$$\begin{aligned} E(X) &= \sum_{k=0}^{\infty} k \cdot \frac{e^{-\lambda} \lambda^k}{k!} = e^{-\lambda} \sum_{k=0}^{\infty} k \frac{\lambda^k}{k!} = e^{-\lambda} \sum_{k=1}^{\infty} k \frac{\lambda^k}{k!} \\ &= \lambda e^{-\lambda} \sum_{k=0}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} = \lambda e^{-\lambda} e^{\lambda} = \lambda \end{aligned}$$

Variance: $\text{Var}(X) = E(X^2) - (E(X))^2$

$$E(X^2) = \sum_{k=0}^{\infty} k^2 \cdot \frac{e^{-\lambda} \lambda^k}{k!} = e^{-\lambda} \left[\sum_{k=1}^{\infty} k^2 \frac{\lambda^k}{k!} \right]$$

$$\sum_{k=0}^{\infty} \frac{\lambda^k}{k!} = e^{\lambda} \Rightarrow \sum_{k=0}^{\infty} k \frac{\lambda^{k-1}}{(k-1)!} = e^{\lambda} \Rightarrow \sum_{k=1}^{\infty} k \frac{\lambda^k}{k!} = \lambda e^{\lambda}$$

Example: Poisson Expectation & Variance (Solution)

$$\sum_{k=1}^{\infty} k^2 \frac{\lambda^{k-1}}{k!} = e^{\lambda} + \lambda e^{\lambda} = (\lambda + 1) e^{\lambda}$$

$$\Rightarrow \sum_{k=1}^{\infty} k^2 \frac{\lambda^k}{k!} = e^{\lambda} \lambda (\lambda + 1)$$

$$E(X^2) = e^{-\lambda} \left(\sum_{k=1}^{\infty} k^2 \frac{\lambda^k}{k!} \right) = \lambda (\lambda + 1)$$

$$\text{Var}(X) = \lambda(\lambda + 1) - \lambda^2 = \lambda$$

Poisson Approximation

Theorem (Poisson Approximation)

Let A_1, A_2, \dots, A_n be events with $p_j = P(A_j)$, where n is large, the p_j are small, and the A_j are independent or weakly dependent. Let

$$X = \sum_{j=1}^n I(A_j) \quad I: \text{indicator}$$

count how many of the A_j occur. Then X is approximately $\text{Pois}(\lambda)$, with $\lambda = \sum_{j=1}^n p_j$.

Example: Birthday Problem Revisited

Example (Birthday Problem Revisited)

There are m people in a room. Assume each person's birthday is equally likely to be any of the 365 days of the year (we exclude February 29), and that people's birthdays are independent (we assume there are no twins in the room). What is the probability that two or more people in the group have the same birthday?

Example: Birthday Problem Revisited (Solution)

$A_{ij} = \{ \text{person } i \text{ and } j \text{ have the same birthday} \}$

$X \triangleq$ number of birthday matches

$$X = \sum_{i \neq j} I(A_{ij}) \approx \text{Pois} \left(\frac{1}{365} \binom{m}{2} \right)$$

$$\begin{aligned} P_{ij} &= P_r(A_{ij}) = \sum_{d=1}^{365} P_r(i\text{'s birthday is } d) P_r(j\text{'s birthday is } d) \\ &= \sum_{d=1}^{365} \frac{1}{365} \times \frac{1}{365} = \frac{1}{365} \quad \left(\lambda = \sum_{i \neq j} P_{ij} \right) \end{aligned}$$

$$\underline{P(X \geq 1)} = 1 - P(X=0) \approx 1 - e^{-\lambda}$$

$$m = 23 \quad 1 - e^{-\lambda} \approx 0.50002$$

True value : 0.507

Sum of Independent Poissons

$$\begin{aligned}P(X+Y=k) &= \sum_{j=0}^k P(X+Y=k | X=j) P(X=j) \quad (\text{LOTP}) \\&= \sum_{j=0}^k P(Y=k-j | X=j) P(X=j) \\&= \sum_{j=0}^k P(Y=k-j) P(X=j)\end{aligned}$$

Theorem (Sum of Independent Poissons)

If $X \sim \text{Pois}(\lambda_1)$, $Y \sim \text{Pois}(\lambda_2)$, and X is independent of Y , then $X + Y \sim \text{Pois}(\lambda_1 + \lambda_2)$.

$$\begin{aligned}&= \sum_{j=0}^k \frac{e^{-\lambda_2} \lambda_2^{k-j}}{(k-j)!} \cdot \frac{e^{-\lambda_1} \lambda_1^j}{j!} \\&= \frac{e^{-(\lambda_1+\lambda_2)}}{k!} \sum_{j=0}^k \binom{k}{j} \lambda_1^j \lambda_2^{k-j} \\&= \frac{e^{-(\lambda_1+\lambda_2)}}{k!} (\lambda_1 + \lambda_2)^k \quad \Rightarrow \quad X+Y \sim \text{Pois}(\lambda_1 + \lambda_2) \\&\quad \forall k=0, 1, 2, \dots\end{aligned}$$

Poisson Given A Sum of Poissons

$$\begin{aligned} P(X=k | X+Y=n) &= \frac{P(X+Y=n | X=k) P(X=k)}{P(X+Y=n)} \\ &= \frac{P(Y=n-k) P(X=k)}{P(X+Y=n)} \end{aligned}$$

$$X+Y \sim \text{Pois}(\lambda_1 + \lambda_2)$$

Theorem (Poisson Given A Sum of Poissons)

If $X \sim \text{Pois}(\lambda_1)$, $Y \sim \text{Pois}(\lambda_2)$, and X is independent of Y , then the conditional distribution of X given $X + Y = n$ is $\text{Bin}(n, \lambda_1/(\lambda_1 + \lambda_2))$.

$$\begin{aligned} &= \frac{\frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!} \cdot \frac{e^{-\lambda_1} \lambda_1^k}{k!}}{\frac{e^{-(\lambda_1 + \lambda_2)} (\lambda_1 + \lambda_2)^n}{n!}} \\ &= \binom{n}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-k} \end{aligned}$$

$\text{Bin}(n, \frac{\lambda_1}{\lambda_1 + \lambda_2})$
PMF

Poisson Approximation to Binomial

Theorem (Poisson Approximation to Binomial)

If $X \sim \text{Bin}(n, p)$ and we let $n \rightarrow \infty$ and $p \rightarrow 0$ such that $\lambda = np$ remains fixed, then the PMF of X converges to the $\text{Pois}(\lambda)$ PMF. More generally, the same conclusion holds if $n \rightarrow \infty$ and $p \rightarrow 0$ in such a way that np converges to a constant λ .

Poisson Approximation to Binomial (Proof)

$\lambda = np$ is fixed while $n \rightarrow \infty$, $p \rightarrow 0$.

$$\begin{aligned} P(X=k) &= \binom{n}{k} p^k (1-p)^{n-k} \\ X \sim \text{Bin}(n, p) &= \frac{n(n-1) \cdots (n-k+1)}{k!} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-k} \\ \lambda = np &= \frac{\lambda^k}{k!} \underbrace{\frac{n(n-1) \cdots (n-k+1)}{n^k}}_{\rightarrow 1} \underbrace{\left(1 - \frac{\lambda}{n}\right)^n}_{\rightarrow e^{-\lambda}} \underbrace{\left(1 - \frac{\lambda}{n}\right)^{-k}}_{\rightarrow 1} \end{aligned}$$

Letting $n \rightarrow \infty$ with k fixed:

$$\frac{n(n-1) \cdots (n-k+1)}{n^k} \rightarrow 1$$

$$\left(1 - \frac{\lambda}{n}\right)^n \rightarrow e^{-\lambda}$$

$$\left(1 - \frac{\lambda}{n}\right)^{-k} \rightarrow 1$$

$$P(X=k) \rightarrow e^{-\lambda} \frac{\lambda^k}{k!} \quad \forall k=0,1,2,\dots$$

Pois (λ)

Example: Visitors to A Website

$X \triangleq$ number of visitors $X \sim \text{Bin}(n, p)$ $n = 10^6$ $p = 2 \times 10^{-6}$

$$\text{Pois}(np) \Rightarrow \text{Pois}(2)$$

Example (Visitors to A Website)

The owner of a certain website is studying the distribution of the number of visitors to the site. Every day, a million people independently decide whether to visit the site, with probability $p = 2 \times 10^{-6}$ of visiting. Give a good approximation for the probability of getting at least three visitors on a particular day.

$$P(X \geq 3) = 1 - P(X=0) - P(X=1) - P(X=2) \approx 1 - 5e^{-2}$$

Outline

1 Poisson Distribution

2 Poisson Process

Definition (Poisson Process-Definition 1)

A Poisson process with parameter λ is a counting process $(N_t)_{t \geq 0}$ with the following properties:

- 1 $N_0 = 0$.
- 2 For all $t > 0$, N_t has a Poisson distribution with parameter λt .
- 3 (Stationary increments) For all $s, t > 0$, $N_{t+s} - N_s$ has the same distribution as N_t . That is,

$$P(N_{t+s} - N_s = k) = P(N_t = k) = \frac{e^{-\lambda t} (\lambda t)^k}{k!}, \quad \text{for } k = 0, 1, \dots$$

- 4 (Independent increments) For $0 \leq q < r \leq s < t$, $N_t - N_s$ and $N_r - N_q$ are independent random variables.

Example

$N_t \triangleq$ number of texts in an interval of t hours

$$\underbrace{N_t \sim \text{Pois}(\lambda t)}_{(N_t)_t} \quad \lambda = 10$$

Example

Joe receives text messages starting at 10 a.m. at the rate of 10 texts per hour according to a Poisson process. Find the probability that he will receive exactly 18 texts by noon and 70 texts by 5 p.m. $\lambda = 10$

12:00

$$\begin{aligned} P(N_2 = 18, N_7 = 70) &= P(N_2 = 18, N_7 - N_2 = 52) \\ &= P(N_2 = 18) P(N_7 - N_2 = 52) \\ &= P(N_2 = 18) P(N_5 = 52) \\ &= 0.0095 \end{aligned}$$

Definition (Translated Poisson Process)

Let $(N_t)_{t \geq 0}$ be a Poisson process with parameter λ . For $s > 0$, let $\tilde{N}_t = N_{t+s} - N_s$, for $t \geq 0$. Then we have

- $(\tilde{N}_t)_{t \geq 0}$ is called "Translated Poisson Process".
- $(\tilde{N}_t)_{t \geq 0}$ is a Poisson process with parameter λ .

Example

$N_t \triangleq$ number of arrivals in the first t hours

$$(N_t)_{t \geq 0} \quad \lambda = 100$$

Example

$$\lambda = 100$$

On election day, people arrive at a voting center according to a Poisson process. On average, 100 voters arrive every hour. If 150 people arrive during the first hour, what is the probability that at most 350 people arrive before the third hour?

$$\begin{aligned} P(N_3 \leq 350 \mid N_1 = 150) &= P(N_3 - N_1 \leq 200 \mid N_1 = 150) \\ &= P(N_3 - N_1 \leq 200) = P(N_2 \leq 200) \\ &= \sum_{k=0}^{200} P(N_2 = k) = \sum_{k=0}^{200} \frac{e^{-100(2)} (100(2))^k}{k!} \end{aligned}$$

Definition (Poisson Process-Definition 2)

Let X_1, X_2, \dots be a sequence of i.i.d. exponential random variables with parameter λ . For $t > 0$, let $X_i \sim \text{Expo}(\lambda)$

$$N_t = \max\{n : X_1 + \dots + X_n \leq t\},$$

with $N_0 = 0$. Then, $(N_t)_{t \geq 0}$ defines a Poisson process with parameter λ .

Definition 2 \Rightarrow Definition 1

Assume Definition 2.

X_1, X_2, \dots a sequence of i.i.d. r.v.s $X_i \sim \text{Expo}(\lambda)$

$$S_n \triangleq X_1 + \dots + X_n \quad S_0 \triangleq 0$$

$$N_t = \max \{n : S_n \leq t\}$$

We need to show that N_t has a Poisson distribution with para. λt .

$$\underline{N_t = k \Leftrightarrow S_k \leq t < S_{k+1}} \quad S_{k+1} = S_k + X_{k+1}$$

S_k is independent of X_{k+1} . $S_k \sim \text{Gamma}(k, \lambda)$ $X_{k+1} \sim \text{Expo}(\lambda)$

$$f_{S_k, X_{k+1}}(s, x) = f_{S_k}(s) f_{X_{k+1}}(x) = \frac{\lambda^k s^{k-1} e^{-\lambda s}}{(k-1)!} \lambda e^{-\lambda x} \quad s, x > 0$$

Definition 2 \Rightarrow Definition 1

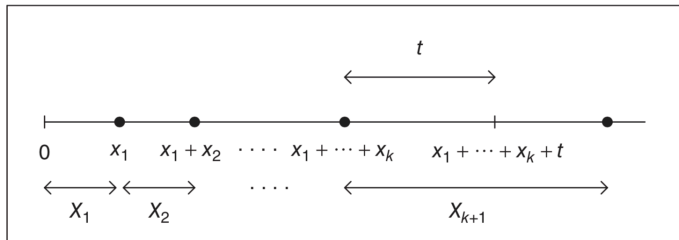
$\forall k \geq 0$

$$\begin{aligned} P(N_t = k) &= P(S_k \leq t < S_k + X_{k+1}) \\ &= P(S_k \leq t, X_{k+1} > t - S_k) \\ (\text{LOTP}) \quad &= \int_0^\infty \int_0^\infty P(S_k \leq t, X_{k+1} > t - S_k \mid S_k = s, X_{k+1} = x) f_{S_k, X_{k+1}}(s, x) dx ds \\ &= \int_0^t \int_{t-s}^\infty f_{S_k, X_{k+1}}(s, x) dx ds \\ &= \frac{e^{-\lambda t} (\lambda t)^k}{k!} \Rightarrow N_t \sim \text{Pois}(\lambda t) \end{aligned}$$

Definition 1 \Rightarrow Definition 2

Assume Definition 1. $N_t \sim \text{Pois}(\lambda t)$

We need to show X_1, X_2, \dots is a sequence of i.i.d. r.v.s
 $X_i \sim \text{Expo}(\lambda)$



For X_1 : $P(X_1 > t) = P(N_t = 0) = e^{-\lambda t}$ $X_1 \sim \text{Expo}(\lambda)$

For X_2 : $P(X_2 > t | X_1 = s) = P(\text{no arrival during interval } (s, s+t] | X_1 = s)$
 $\quad \quad \quad = P(N_{s+t} - N_s = 0 | X_1 = s)$
 $\quad \quad \quad = P(N_{s+t} - N_s = 0) = P(N_t = 0) = e^{-\lambda t}$

$X_2 \sim \text{Expo}(\lambda)$

Definition 1 \Rightarrow Definition 2

$$\begin{aligned} & P(X_{k+1} > t \mid X_1 = x_1, X_2 = x_2, \dots, X_k = x_k) \\ &= P(\text{no arrival during interval } (x_1 + \dots + x_k, x_1 + \dots + x_k + t] \mid \dots) \\ &= P(N_{x_1 + \dots + x_k + t} - N_{x_1 + \dots + x_k} = 0 \mid \dots) \\ &= P(N_{x_1 + \dots + x_k + t} - N_{x_1 + \dots + x_k} = 0) \\ &= P(N_t = 0) = e^{-\lambda t} \end{aligned}$$

X_{k+1} is independent of x_1, \dots, x_k $X_{k+1} \sim \text{Expo}(\lambda)$

$\forall k \geq 2$

x_1, x_2, \dots i.i.d. $\text{Expo}(\lambda)$ (Definition 2)

Conditional Counts

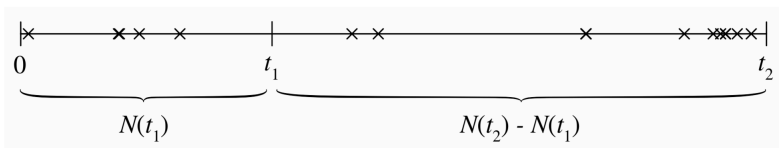
Theorem (Conditional Counts) $N(t) = N_t$

Let $\{N(t), t > 0\}$ be a Poisson process with rate λ , and let $t_1 < t_2$. Then the conditional distribution of $N(t_1)$ given $N(t_2) = n$ is

$$N(t_1) | N(t_2) = n \sim \text{Bin} \left(n, \frac{t_1}{t_2} \right).$$

Conditional Counts

$$N(t_1) \mid \{N(t_2) = n\} \sim \text{Bin}\left(n, \frac{t_1}{t_2}\right)$$



$$X \triangleq N(t_1) \sim \text{Pois}(\lambda t_1) \quad Y \triangleq N(t_2) - N(t_1) \sim \text{Pois}(\lambda(t_2 - t_1))$$

$$\begin{aligned} N(t_1) \mid \{N(t_2) = n\} &= X \mid \{X + Y = n\} \sim \text{Bin}\left(n, \frac{\lambda t_1}{\lambda t_1 + \lambda(t_2 - t_1)}\right) \\ &= \text{Bin}\left(n, \frac{t_1}{t_2}\right) \end{aligned}$$

Arrival Times & Uniform Distribution

Theorem (Arrival Times & Uniform Distribution)

Let S_1, S_2, \dots , be the arrival times of a Poisson process with parameter λ . Conditional on $N_t = n$, the joint distribution of (S_1, \dots, S_n) is the distribution of the order statistics of n i.i.d. uniform random variables on $[0, t]$. That is, the joint density function of S_1, \dots, S_n is

$$f(s_1, \dots, s_n) = \frac{n!}{t^n}, \quad \text{for } 0 < s_1 < \dots < s_n < t.$$

Equivalently, let U_1, \dots, U_n be an i.i.d. sequence of random variables uniformly distributed on $[0, t]$. Then, conditional on $N_t = n$,

$$(S_1, \dots, S_n) \quad \text{and} \quad (U_{(1)}, \dots, U_{(n)})$$

have the same distribution.

Arrival Times & Uniform Distribution (Proof)

$$f_{s_1, \dots, s_n} (s_1, \dots, s_n | N_t = n)$$

$$= \lim_{\varepsilon_1 \rightarrow 0} \dots \lim_{\varepsilon_n \rightarrow 0} \frac{P(s_1 \leq S_1 \leq s_1 + \varepsilon_1, \dots, s_n \leq S_n \leq s_n + \varepsilon_n | N_t = n)}{\varepsilon_1 \dots \varepsilon_n}$$

$$0 < s_1 < \dots < s_n < t$$

$$P(s_1 \leq S_1 \leq s_1 + \varepsilon_1, \dots, s_n \leq S_n \leq s_n + \varepsilon_n, N_t = n)$$

$$P(N_t = n)$$

$$= \frac{P(\text{one arrival in each interval } [s_k, s_k + \varepsilon_k] \text{ and no arrival in rest intervals})}{P(N_t = n)}$$

$$= \frac{P(N_{s_1 + \varepsilon_1} - N_{s_1} = 1) \dots P(N_{s_n + \varepsilon_n} - N_{s_n} = 1) P(N_{t - \varepsilon_1 - \dots - \varepsilon_n} = 0)}{P(N_t = n)}$$

$$= \frac{P(N_{\varepsilon_1} = 1) \dots P(N_{\varepsilon_n} = 1) P(N_{t - \varepsilon_1 - \dots - \varepsilon_n} = 0)}{P(N_t = n)} \quad N_t \sim \text{Pois}(\lambda t)$$

$$= \frac{\lambda^n}{t^n} \varepsilon_1 \dots \varepsilon_n$$

Arrival Times & Uniform Distribution (Proof)

$$\begin{aligned} & \underbrace{f_{s_1, \dots, s_n} (s_1, \dots, s_n \mid N_t = n)} \\ &= \lim_{\varepsilon_1 \rightarrow 0} \dots \lim_{\varepsilon_n \rightarrow 0} \frac{\frac{n!}{t^n} \varepsilon_1 \dots \varepsilon_n}{\varepsilon_1 \dots \varepsilon_n} = \underbrace{\frac{n!}{t^n}} \end{aligned}$$

$$f_{v_1, \dots, v_n} (u_1, \dots, u_n) = \frac{1}{t^n} \quad v_i \sim \text{Unif} (0, t)$$

$$f_{v_{(1)}, \dots, v_{(n)}} (u_1, \dots, u_n) = \underbrace{\frac{1}{t^n} \cdot n!}$$

Arrival Times & Uniform Distribution (Proof)

Example

Example

Students enter a campus building according to a Poisson process $(N_t)_{t \geq 0}$ with parameter λ . The times spent by each student in the building are i.i.d. random variables with continuous cumulative distribution function $F(t)$. Find the probability mass function of the number of students in the building at time t , assuming there are no students in the building at time 0.

$$N_0 = 0$$

Example (Solution)

$B_t \triangleq$ number of students in the building at time t .

By LOTP:

$$\begin{aligned} \underline{P(B_t = k)} &= \sum_{n=k}^{\infty} P(B_t = k | N_t = n) P(N_t = n) && (N_t)_t \text{ Poisson process} \\ &= \sum_{n=k}^{\infty} \underline{P(B_t = k | N_t = n)} \frac{e^{-\lambda t} (\lambda t)^n}{n!} && N_t \sim \text{Poi}(\lambda t) \end{aligned}$$

Assume that n students enter the building by time t ,
with arrival times s_1, \dots, s_n .

$z_k \triangleq$ length of time spent in the building by the k -th student

Students leave the building at times $s_1 + z_1, \dots, s_n + z_n$

$$\begin{aligned} P(B_t = k | N_t = n) &= P(k \text{ of the } s_1 + z_1, \dots, s_n + z_n \text{ exceed } t | N_t = n) \\ &= P(k \text{ of the } \underline{U_{(1)} + z_1, \dots, U_{(n)} + z_n \text{ exceed } t}) \end{aligned}$$

$$U_k \sim \text{Unif}[0, t]$$

Example (Solution)

$$\frac{(v_{(1)}, \dots, v_{(n)})}{(v_1, \dots, v_n)} \quad p(B_t = k \mid N_t = n) = p(\underbrace{k \text{ of the } v_1 + z_1, \dots, v_n + z_n \text{ exceed } t})$$

$$B_t \mid \{N_t = n\} \sim \text{Bin}(n, p)$$

$$\begin{aligned} p &= P(v_1 + z_1 > t) = P(z_1 > t - v_1) \\ &= \int_0^t P(z_1 > t - v_1 \mid v_1 = s) \underbrace{f_{v_1}(s)}_{f_{v_1}(s)} ds \\ &= \frac{1}{t} \int_0^t (1 - P(z_1 \leq t - s)) ds \\ &= \frac{1}{t} \int_0^t (1 - F(t - s)) ds = \frac{1}{t} \int_0^t (1 - F(s)) ds \end{aligned}$$

$$\underline{P(B_t = k \mid N_t = n) = \binom{n}{k} p^k (1-p)^{n-k}}$$

Example (Solution)

$$P(B_t = k) = \frac{e^{-\lambda p t} (\lambda p t)^k}{k!}$$

$$B_t \sim \text{Pois}(\lambda p t)$$