

- def: Poisson process $\equiv$
- 1) Involves measurements of times when events occur.
 - 2) The past timing pattern of events has no influence on timing of future events; no memory; independence of event times.
 - 3) The probability density that an event will occur at a particular time is the same for all times; no clock; uniform likelihood of event times.
 - 4) Events occur at some average rate; the ave # of events in time Δt is proportional to the duration of Δt .

ex: Leaves falling from a tree in autumn when there is no wind. The times when leaves hit the ground approximates a Poisson process.

Note: in many situations we can argue that a process is influenced by variables that cause event times to be dependent on one another, (such as one falling leaf knocking off another one), but if the dependence is weak we may still gain advantage by assuming our process is a Poisson process.

Note: even if conditions for a process change, (a wind blows more leaves off), we may be able to assume a Poisson process for an interval of time, (while the wind is steady).

def: rate of Poisson process $\equiv \lambda \equiv$ ave # events per unit time

tool: mean number of events per unit time for Poisson process $= \lambda$

tool: $E\{\# \text{ events in time } t\} = \lambda t \equiv \mu$

tool: $P\{n \text{ events in time } t\} = \frac{e^{-\lambda t} (\lambda t)^n}{n!}$

def: " $\equiv$ Poisson distribution

check: $\sum_{n=0}^{\infty} P\{n \text{ events in time } t\} = \sum_{n=0}^{\infty} \frac{e^{-\lambda t} (\lambda t)^n}{n!}$

sum must = 1 since total prob will equal 1 $= e^{-\lambda t} \sum_{n=0}^{\infty} \frac{(\lambda t)^n}{n!}$

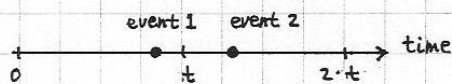
Recall Taylor series for e^x

is $\frac{1}{0!} + \frac{x}{1!} + \frac{x^2}{2!} + \dots + \frac{x^n}{n!} + \dots$

$= e^{-\lambda t} e^{\lambda t} = 1 \checkmark$

check: We independence for the number of events in adjacent intervals. For two intervals of length t side-by-side we have

$$P\{2 \text{ events in time } 2 \cdot t\} = \frac{e^{-2\lambda t} (2\lambda t)^2}{2!}$$



Given independence, we have

$$P\{2 \text{ events in time } 2 \cdot t\} = (P\{2 \text{ events in time } t\} \cdot P\{0 \text{ events in time } t\})$$

$$+ (P\{1 \text{ event in time } t\} \cdot P\{1 \text{ event in time } t\})$$

$$+ (P\{0 \text{ events in time } t\} \cdot P\{2 \text{ events in time } t\})$$

and sum probs of scenarios.

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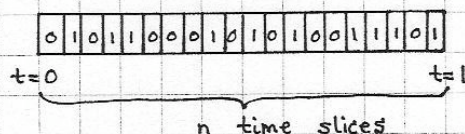
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So we have

$$\begin{aligned}
 & \frac{e^{-\lambda t} (\lambda t)^2}{2!} \cdot \frac{e^{-\lambda t} (\lambda t)^0}{0!} \\
 & + \frac{e^{-\lambda t} (\lambda t)^1}{1!} \cdot \frac{e^{-\lambda t} (\lambda t)^1}{1!} \\
 & + \frac{e^{-\lambda t} (\lambda t)^0}{0!} \cdot \frac{e^{-\lambda t} (\lambda t)^2}{2!} \\
 & = e^{-2\lambda t} \left[\frac{(\lambda t)^2}{2} \cdot 1 + \frac{\lambda t}{1} \cdot \frac{\lambda t}{1} + 1 \cdot \frac{(\lambda t)^2}{2} \right] \\
 & = e^{-2\lambda t} \frac{2(\lambda t)^2}{2} \\
 & = \frac{e^{-2\lambda t} (2\lambda t)^2}{2!} \\
 & = P\{2 \text{ events in time } 2 \cdot t\} \quad \checkmark
 \end{aligned}$$

tool: Poisson distribution = $\lim_{n \rightarrow \infty}$ (Binomial distribution)
 with rate $\lambda = \mu$ with $np = \lambda = \mu$

justification:



Divide one unit of time into n slices.

Write a 0 if no event occurs in a given slice.

" " 1 " an event does occur " " " "

We expect to have $\#1's = \lambda$

So $P\{1 \text{ in a slice}\} = \frac{\lambda}{n} \equiv p$ or $np = \lambda$

$$P\{\lambda \text{ events in } n \text{ slices}\} = \binom{n}{\lambda} p^{\lambda} (1-p)^{n-\lambda}$$

= Binomial distribution

The difference between Poisson and the Binomial is that there could be 2 events in a slice for Poisson.

If we take $n \rightarrow \infty$, the probability of this approaches zero, and the Binomial approaches the Poisson dist.

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tool: $\lim_{\lambda \rightarrow 0}$ (Poisson distribution for $n=1$ event in time $t=1$)

= Binomial distribution for success in one trial
for $p = \lambda$

$$= \binom{1}{1} p^1 (1-p)^0 = p = \lambda$$

pf: Poisson distribution for $n=1$ event in time $t=1$

$$= \frac{e^{-\lambda t} (\lambda t)^1}{1!} \Big|_{t=1} = e^{-\lambda} \lambda$$

$$\begin{aligned} \text{Now } \lim_{\lambda \rightarrow 0} e^{-\lambda} \lambda &= \lim_{\lambda \rightarrow 0} \left(\frac{1 - \lambda + \frac{\lambda^2}{2!} - \dots}{0!} \right) \lambda \\ &= \lim_{\lambda \rightarrow 0} \left(\frac{\lambda - \frac{\lambda^2}{1!} + \frac{\lambda^3}{2!} - \dots}{0!} \right) \\ &= \lim_{\lambda \rightarrow 0} \lambda \quad \begin{array}{l} \text{acts like } \lambda \\ \text{as } \lambda \rightarrow 0 \\ \text{because higher} \\ \text{order terms small} \end{array} \end{aligned}$$

Note: A more precise statement is that
the Poisson distribution behaves
asymptotically like λ as $\lambda \rightarrow 0$, (for $n=1$ event)

tool: probability density of intervals between events
for Poisson process $\equiv p(\Delta t) = \lambda e^{-\lambda \Delta t} \equiv \text{exp dist}$
where λ = rate of process $\equiv$ ave events per unit time.

$$\begin{aligned} \text{pf: } p(\Delta t) &= \frac{d}{d\Delta t} P\{\text{interval} \leq \Delta t\} = \frac{d}{d\Delta t} (1 - P\{\text{0 events in } \Delta t\}) \\ &= \frac{d}{d\Delta t} \left(\frac{(1 - e^{-\lambda \Delta t}) (\lambda \Delta t)^0}{0!} \right) = \frac{d}{d\Delta t} (1 - e^{-\lambda \Delta t}) \\ p(\Delta t) &= \lambda e^{-\lambda \Delta t} \end{aligned}$$

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tool: probability density of interval Δt_n required for n events to occur in Poisson process with rate λ

$$p(\Delta t_n) = \frac{\lambda^n}{(n-1)!} (\Delta t_n)^{n-1} e^{-\lambda \Delta t_n} \equiv \text{gamma dist}$$

$\alpha = n, \beta = \frac{1}{\lambda}$

pf: (for $n=2$)

$$p(\Delta t_2) = \frac{d}{d\Delta t_2} P\{\text{interval for 2 events} \leq \Delta t_2\}$$

$$= \frac{d}{d\Delta t_2} [1 - (P\{0 \text{ events in } \Delta t_2\} + P\{1 \text{ event in } \Delta t_2\})]$$

$$= \frac{d}{d\Delta t_2} [1 - (e^{-\lambda \Delta t_2} \frac{(\lambda \Delta t_2)^0}{0!} + e^{-\lambda \Delta t_2} \frac{(\lambda \Delta t_2)^1}{1!})]$$

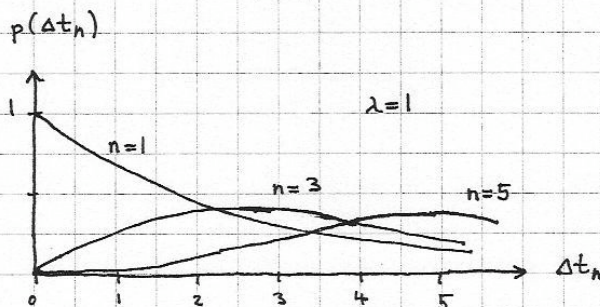
$$= \frac{d}{d\Delta t_2} [1 - e^{-\lambda \Delta t_2} (1 + \lambda \Delta t_2)]$$

$$= -[-\lambda e^{-\lambda \Delta t_2} (1 + \lambda \Delta t_2) + e^{-\lambda \Delta t_2} \lambda]$$

$$= e^{-\lambda \Delta t_2} [\lambda (1 + \lambda \Delta t_2) - \lambda]$$

$$= e^{-\lambda \Delta t_2} \lambda^2 \Delta t_2$$

$$= \frac{\lambda^2}{(2-1)!} (\Delta t_2)^{2-1} e^{-\lambda \Delta t_2} \quad \checkmark$$



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tool: $E\{\Delta t_n\} = \frac{n}{\lambda}$ for Poisson process of rate λ

Note: The proof is by direct calculation, (tedious but straightforward).

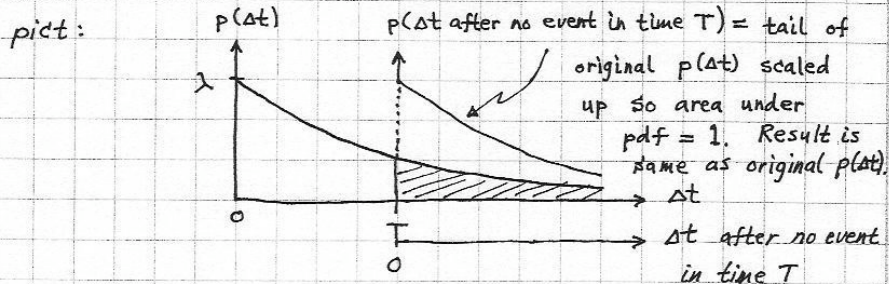
tool: $\sigma_{\Delta t_n}^2 = \frac{n}{\lambda^2}$ for Poisson process of rate λ

Note: Again, found by direct calculation.

tool: $\int x^n e^{ax} dx = e^{ax} \left[\frac{x^n}{a} - \frac{n x^{n-1}}{a^2} + \frac{n(n-1)x^{n-2}}{a^3} - \dots \right.$
 where $n \geq 0$
 $\left. + (-1)^{n-1} \frac{n! x}{a^n} + (-1)^n \frac{n!}{a^{n+1}} \right]$

ref: Tables of Integrals and other Mathematical Data, Herbert Bristol Dwight, MacMillan: NY, 1961.

tool: If no event occurs for time T in a Poisson process, the density function for the interval, Δt , to the next event is the same as the original $p(\Delta t)$, namely $p(\Delta t) = \lambda e^{-\lambda \Delta t}$. Thus, the process is memoryless. We may also use this observation to derive $p(\Delta t)$ to begin with.



ex: An astronomer and a climatologist have determined that an asteroid large enough to cause mass extinctions hits the earth on average once every 10 million years.

Assuming the arrivals of these asteroids follow a Poisson process, answer the following questions:

- If $X \equiv$ interval between large asteroids, find $E\{X\}$ and σ_X^2
- Find $P\{X \geq 10 \text{ million years}\}$. X as in (a). Is it $1/2$?
- Find $P\{2 \text{ large asteroids hit earth in next 1 million years}\}$.
- Find the probability density function for the interval Δt_3 for 3 large asteroids to hit earth. Locate the peak value as well.

soln: a) $E\{X\} = \frac{n=1}{\lambda = 1/10^7 \text{ yrs}} = 10^7 \text{ yrs} \equiv \mu_X$

$$\sigma_X^2 = \frac{n=1}{\lambda^2} = 10^{14} \text{ yrs}^2$$

So $\sigma_X = 10^7 \text{ yrs} = \mu_X$. (Always, if $n=1$)

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b) $p_X(x) = \lambda e^{-\lambda x}$ exponential distribution
when $n=1$. Here $\lambda = \frac{1}{10^7}$

$$\begin{aligned}
 P\{X \geq 10 \text{ million yrs}\} &= \int_{10^7}^{\infty} p_X(x) dx \\
 &= \int_{10^7 = \frac{1}{\lambda}}^{\infty} \lambda e^{-\lambda x} dx = \lambda \left[\frac{e^{-\lambda x}}{-\lambda} \right]_{\frac{1}{\lambda}}^{\infty} \\
 &= e^{-1}
 \end{aligned}$$

$$P\{X \geq 10 \text{ million yrs}\} = \frac{1}{e} \approx \frac{1}{3}$$

So, more often than not, the interval between asteroids is less than 10^7 yrs, yet the average interval is 10^7 yrs.

Why? There are a few very long intervals that increase the average.

c) $P\{2 \text{ large asteroids in } 1 \text{ million yrs}\}$

$$\begin{aligned}
 &= \int_0^{10^6} p(\Delta t_2) d\Delta t_2 \\
 &= \int_0^{10^6} \frac{\lambda^2 (\Delta t_2)^{2-1} e^{-\lambda \Delta t_2}}{(2-1)!} d\Delta t_2 \\
 &= \int_0^{10^6} \frac{\lambda^2 x}{1!} e^{-\lambda x} dx \quad \text{use } x \text{ for } \Delta t_2 \\
 &= \dots \quad \text{Wait, This is the hard way!}
 \end{aligned}$$

Use Poisson distribution instead:

$$P\{2 \text{ events in } t=10^6\} = \frac{e^{-\lambda t} (\lambda t)^2}{2!}$$

where $\lambda = \frac{1}{10^7}$ and $\lambda t = \frac{1}{10}$

$$P\{2 \text{ large asteroids in } 10^6 \text{ yrs}\} = \frac{e^{-1/10}}{200} \approx \frac{1}{200}$$

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$$d) \quad p(\Delta t_3) = \frac{\lambda^3}{(3-1)!} (\Delta t_3)^{3-1} e^{-\lambda \Delta t_3}$$

$$\text{where } \lambda = \frac{1}{10^7 \text{ yrs}}$$

$$p(\Delta t_3) = \lambda^3 \Delta t_3^2 e^{-\lambda \Delta t_3}$$

$$p(\Delta t_3) = \frac{\Delta t_3^2}{10^{21}} e^{-\Delta t_3 / 10^7}$$

To find max (or peak) use time scale of 1 time unit = 10^7 yrs so $\lambda = 1$.

$$\text{Then } p(\Delta t_3) = \Delta t_3^2 e^{-\Delta t_3}$$

$$\text{where } \Delta t_3 = \text{time} / 10^7 \text{ yrs}$$

$$\text{max } p(\Delta t_3) \text{ when } \frac{d}{d\Delta t_3} p(\Delta t_3) = 0.$$

Use x for Δt_3 .

$$\frac{d}{dx} x^2 e^{-x} = 0 = 2x e^{-x} - x^2 e^{-x}$$

$$\therefore 2 - x = 0 \Rightarrow x = 2 \text{ or } \Delta t_3 = 2 \cdot 10^7 \text{ yrs}$$

So $p(\Delta t_3)$ peaks at 20 million yrs = 2μ .