Scientific Computing — Point Processes

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Neuroethology

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A point process is a stochastic (or random) process that generates a sequence of events at times $\{t_i\}, t_i \in \mathbb{R}$.

For each point process there is an underlying continuous-valued process evolving in time. The associated point process occurs when the underlying continuous process crosses a threshold. Examples:

- Spikes/heartbeat: generated by the dynamics of the membrane potential of neurons/heart cells.
- Earth quakes: generated by the pressure dynamics between the tectonic plates on either side of a geological fault line.
- Onset of cricket/frogs/birds/... songs: generated by the dynamics of the state of a nervous system.

Point process



Homogeneous Poisson process

The probability $p(t)\delta t$ of an event occuring at time *t* is independent of *t* and independent of any previous event (independent of event history). The probability *P* for an event occuring within a time bin of width Δt is

$$P = \lambda \cdot \Delta t$$

for a Poisson process with rate λ .



Rate

Rate of events r ("spikes per time") measured in Hertz.

- Number of events N per observation time W: $r = \frac{N}{W}$
- Without boundary effects: $r = \frac{N-1}{t_N-t_1}$
- Inverse interval: $r = \frac{1}{\mu_{ISI}}$

(Interspike) interval statistics

- Histogram p(T) of intervals T. Normalized to $\int_0^{\infty} p(T) dT = 1$
- Mean interval $\mu_{ISI} = \langle T \rangle = \frac{1}{n} \sum_{i=1}^{n} T_i$
- Variance of intervals $\sigma_{ISI}^2 = \langle (T \langle T \rangle)^2 \rangle$
- Coefficient of variation $CV_{ISI} = \frac{\sigma_{ISI}}{\mu_{ISI}}$
- Diffusion coefficient $D_{ISI} = \frac{\sigma_{ISI}^2}{2\mu_{ISI}^3}$



Interval statistics of homogeneous Poisson process

- Exponential distribution of intervals $T: p(T) = \lambda e^{-\lambda T}$
- Mean interval $\mu_{ISI} = \frac{1}{\lambda}$
- Variance of intervals $\sigma_{ISI}^2 = \frac{1}{\lambda^2}$
- Coefficient of variation CV_{ISI} = 1

Poisson spike trains, rate=20 Hz, nisi=102



Poisson spike trains, rate=100 Hz, nisi=49



Interval return maps

Scatter plot between succeeding intervals separated by lag k.



Serial interval correlations

Correlation coefficients between succeeding intervals separated by lag k:

$$\rho_{k} = \frac{\langle (T_{i+k} - \langle T \rangle)(T_{i} - \langle T \rangle) \rangle}{\langle (T_{i} - \langle T \rangle)^{2} \rangle} = \frac{\operatorname{cov}(T_{i+k}, T_{i})}{\operatorname{var}(T_{i})}$$

• $\rho_0 = 1$ (correlation of each interval with itself).

Poisson process: ρ_k = 0 for k > 0 (renewal process!)



Count statistics

Histogram of number of events N (counts) within observation window of duration W.



Count statistics of Poisson process

Poisson distribution:

$$P(k) = \frac{(\lambda W)^k e^{\lambda W}}{k!}$$



Count statistics — Fano factor

Statistics of number of events N within observation window of duration W.

- Mean count: $\mu_N = \langle N \rangle$
- Count variance: $\sigma_N^2 = \langle (N \langle N \rangle)^2 \rangle$
- Fano factor (variance divided by mean): $F = \frac{\sigma_N^2}{\mu_N}$
- Poisson process: F = 1

Poisson process $\lambda = 100$ Hz:



Integrate-and-fire models

Leaky integrate-and-fire model (LIF):

$$\tau \frac{dV}{dt} = -V + RI + D\xi$$

Whenever membrane potential V(t) crosses the firing threshold θ , a spike is emitted and V(t) is reset to V_{reset} .

- τ: membrane time constant (typically 10 ms)
- R: input resistance (here 1 mV (!))
- Dξ: additive Gaussian white noise of strength D

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- θ: firing threshold (here 10 mV)
- V_{reset}: reset potential (here 0 mV)

Integrate-and-fire models

Discretization with time step Δt : $V(t) \rightarrow V_i$, $t_i = i\Delta t$. Euler integration:

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$$\frac{dV}{dt} \approx \frac{V_{i+1} - V_i}{\Delta t}$$

$$\Rightarrow V_{i+1} = V_i + \Delta t \frac{-V_i + RI_i + \sqrt{2D\Delta t}N_i}{\tau}$$

 N_i are normally distributed random numbers (Gaussian with zero mean and unit variance) — the $\sqrt{\Delta t}$ is for white noise.



Interval statistics of LIF

Interval distribution approaches Inverse Gaussian for large I:

$$p(T) = rac{1}{\sqrt{4\pi DT^3}} \exp\left[-rac{(T - \langle T \rangle)^2}{4DT \langle T \rangle^2}
ight]$$

where $\langle T \rangle$ is the mean interspike interval and *D* is the diffusion coefficient.



Interval statistics of PIF

For the perfect integrate-and-fire (PIF)

$$\tau \frac{dV}{dt} = RI + D\xi$$

(the canonical model or supra-threshold firing on a limit cycle)

the Inverse Gaussian describes exactly the interspike interval distribution.



Interval return map of LIF



Serial correlations of LIF



Integrate-and-fire driven with white noise are still renewal processes!

Count statistics of LIF



Interval statistics of LIF with OU noise



Ohrnstein-Uhlenbeck noise is lowpass filtered white noise.



More peaky than the inverse Gaussian!

Interval return map of LIF with OU noise

LIF I = 15.7, $\tau_{OU} = 100$ ms:



Serial correlations of LIF with OU noise

LIF I = 15.7, $\tau_{OU} = 100$ ms:



OU-noise introduces positive interval correlations!

Count statistics of LIF with OU noise

LIF I = 15.7, $\tau_{OU} = 100$ ms:



Fano factor increases with count window duration.

Interval statistics of LIF with adaptation

$$\tau \frac{dV}{dt} = -V - A + RI + D\xi$$

$$\tau_{adapt} \frac{dA}{dt} = -A$$

Adaptation A with time constant τ_{adapt} and increment ΔA at spike.



Similar to LIF with white noise.

Interval return map of LIF with adaptation

LIF I = 10, $\tau_{adapt} = 100$ ms:



Negative correlation at lag one.

Serial correlations of LIF with adaptation

LIF I = 10, $\tau_{adapt} = 100$ ms:



Adaptation with white noise introduces negative interval correlations!

Count statistics of LIF with adaptation

LIF I = 10, $\tau_{adapt} = 100$ ms:



Fano factor decreases with count window duration.