

Scientific Computing — Point Processes

Jan Benda

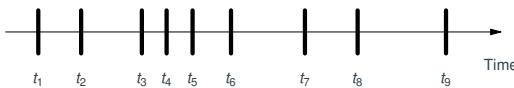
Neuroethology

WS 14/15



Point processes

Point process



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.

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Point processes

Homogeneous Poisson process

Interval statistics

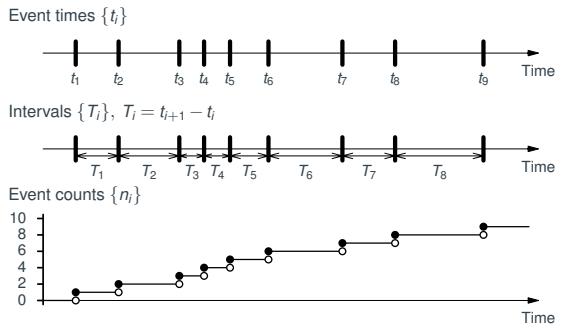
Count statistics

Integrate-and-fire models

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Point processes

Point process



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Homogeneous Poisson process

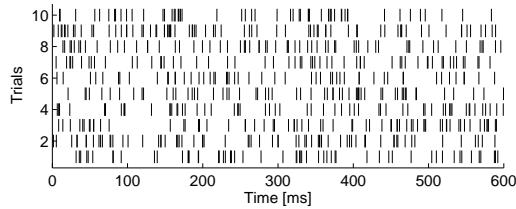
Homogeneous Poisson process

The probability $p(t)\delta t$ of an event occurring at time t is independent of t and independent of any previous event (independent of event history).

The probability P for an event occurring within a time bin of width Δt is

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

for a Poisson process with rate λ .



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Interval statistics

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: $r = \frac{1}{\mu_{ISI}}$

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Interval statistics

(Interspike) interval statistics

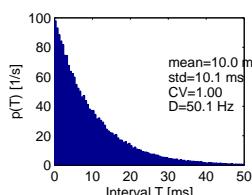
- Histogram $p(T)$ of intervals T . Normalized to $\int_0^\infty p(T) dT = 1$

$$\text{Mean interval } \mu_{ISI} = \langle T \rangle = \frac{1}{n} \sum_{i=1}^n T_i$$

$$\text{Variance of intervals } \sigma_{ISI}^2 = \langle (T - \langle T \rangle)^2 \rangle$$

$$\text{Coefficient of variation } CV_{ISI} = \frac{\sigma_{ISI}}{\mu_{ISI}}$$

$$\text{Diffusion coefficient } D_{ISI} = \frac{\sigma_{ISI}^2}{2\mu_{ISI}}$$



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Interval statistics

Interval statistics of homogeneous Poisson process

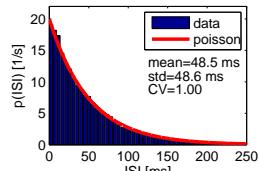
- Exponential distribution of intervals T : $p(T) = \lambda e^{-\lambda T}$

$$\text{Mean interval } \mu_{ISI} = \frac{1}{\lambda}$$

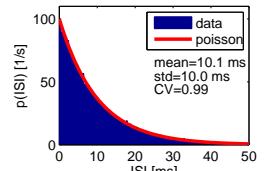
$$\text{Variance of intervals } \sigma_{ISI}^2 = \frac{1}{\lambda^2}$$

$$\text{Coefficient of variation } CV_{ISI} = 1$$

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



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

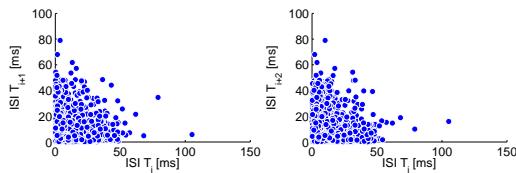


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Interval return maps

Scatter plot between succeeding intervals separated by lag k .

Poisson process $\lambda = 100$ Hz:



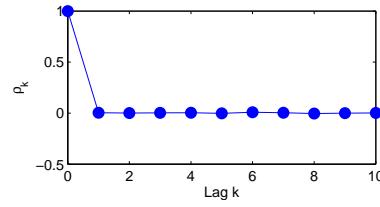
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Serial interval correlations

Correlation coefficients between succeeding intervals separated by lag k :

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

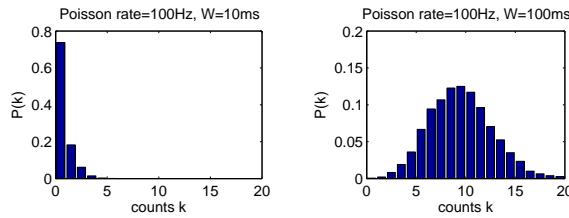
- $\rho_0 = 1$ (correlation of each interval with itself).
- Poisson process: $\rho_k = 0$ for $k > 0$ (renewal process!)



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Count statistics

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

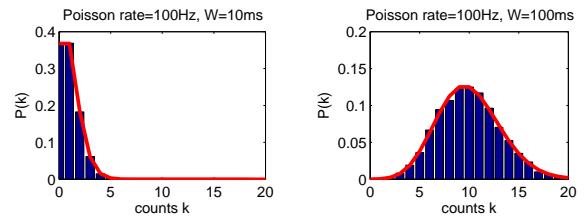


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Count statistics of Poisson process

Poisson distribution:

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



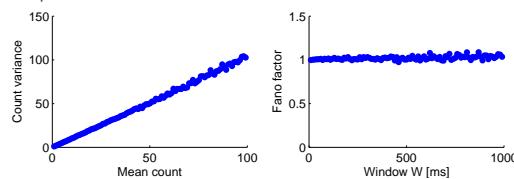
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Count statistics — Fano factor

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:



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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 !)
- $D\xi$: additive Gaussian white noise of strength D
- θ : firing threshold (here 10 mV)
- V_{reset} : reset potential (here 0 mV)

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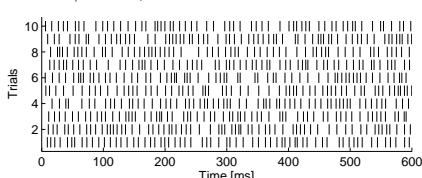
Integrate-and-fire models

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

Euler integration:

$$\begin{aligned} \frac{dV}{dt} &\approx \frac{V_{i+1} - V_i}{\Delta t} \\ \Rightarrow V_{i+1} &= V_i + \Delta t \frac{-V_i + RI + \sqrt{2D\Delta t}N_i}{\tau} \end{aligned}$$

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



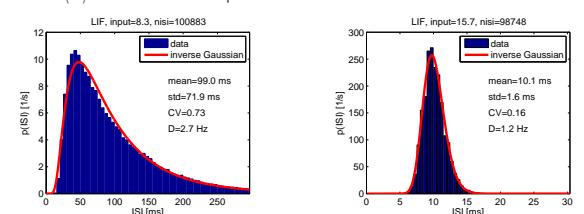
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Interval statistics of LIF

Interval distribution approaches Inverse Gaussian for large I :

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

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



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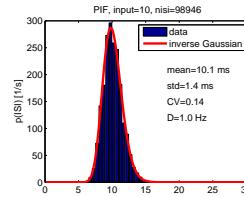
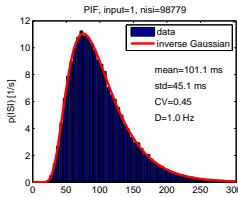
Interval statistics of PIF

For the perfect integrate-and-fire (PIF)

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

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

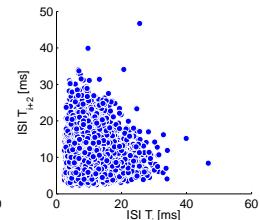
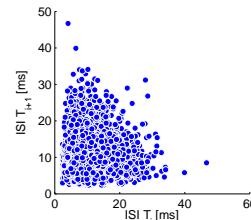
the Inverse Gaussian describes exactly the interspike interval distribution.



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Interval return map of LIF

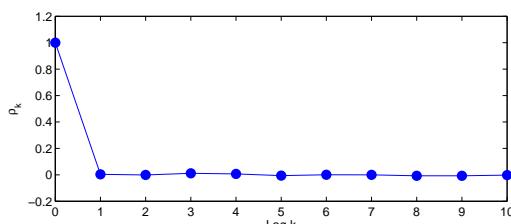
LIF $I = 15.7$:



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Serial correlations of LIF

LIF $I = 15.7$:

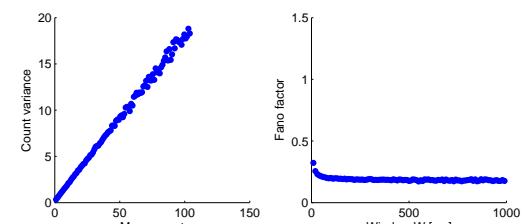


Integrate-and-fire driven with white noise are still renewal processes!

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Count statistics of LIF

LIF $I = 15.7$:



Fano factor is not one!

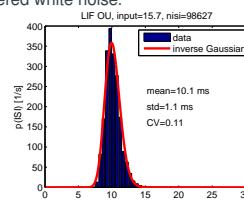
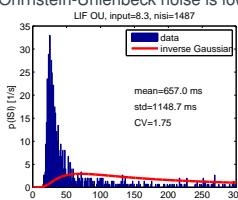
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Interval statistics of LIF with OU noise

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

$$\tau_{OU} \frac{dU}{dt} = -U + D_s^{\xi}$$

Ohrnstein-Uhlenbeck noise is lowpass filtered white noise.



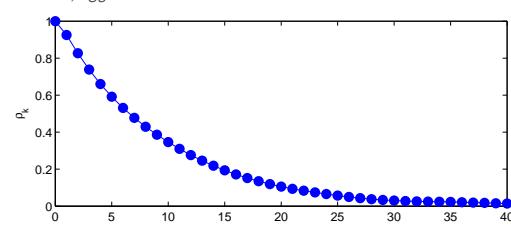
More peaky than the inverse Gaussian!

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Serial correlations of LIF with OU noise

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

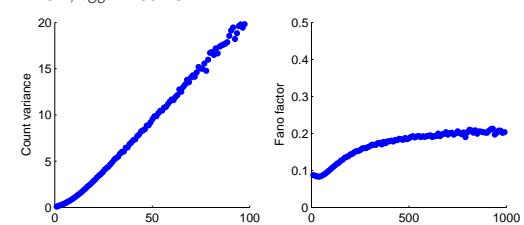


OU-noise introduces positive interval correlations!

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Count statistics of LIF with OU noise

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



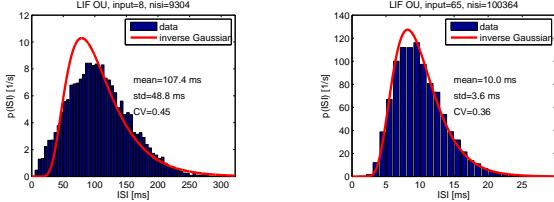
Fano factor increases with count window duration.

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Interval statistics of LIF with adaptation

$$\begin{aligned}\tau \frac{dV}{dt} &= -V - A + RI + D\zeta \\ \tau_{adapt} \frac{dA}{dt} &= -A\end{aligned}$$

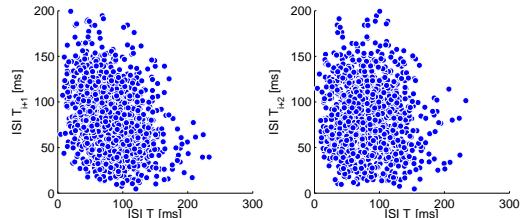
Adaptation A with time constant τ_{adapt} and increment ΔA at spike.



Similar to LIF with white noise.

Interval return map of LIF with adaptation

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

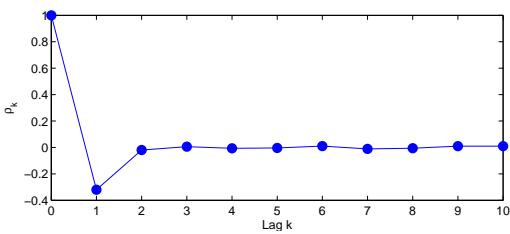


Negative correlation at lag one.

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Serial correlations of LIF with adaptation

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

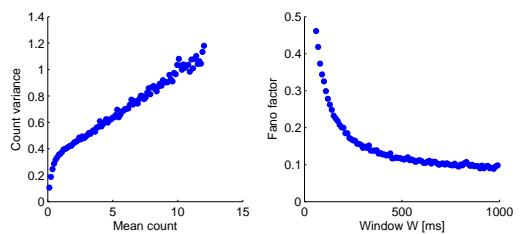


Adaptation with white noise introduces negative interval correlations!

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Count statistics of LIF with adaptation

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



Fano factor decreases with count window duration.

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