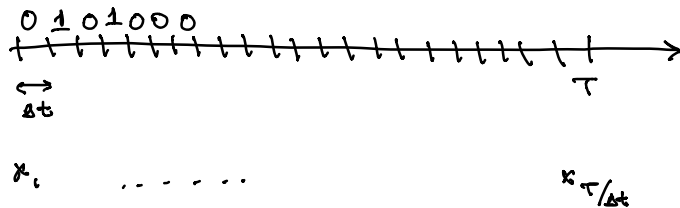


Computing Fano factor for a Poisson process



$$\begin{cases} \Pr(X_j = 1) = r \Delta t \\ \Pr(X_j = 0) = (1 - r \Delta t) \end{cases}$$

$$\langle X_j \rangle = 1 \cdot r \Delta t + 0 \cdot (1 - r \Delta t) = r \Delta t$$

$$\begin{aligned} \text{Var}(X_j) &= \langle (X_j - \langle X_j \rangle)^2 \rangle = (1 - r \Delta t)^2 \cdot r \Delta t + (r \Delta t)^2 (1 - r \Delta t) \\ &= (1 - 2r \Delta t + (r \Delta t)^2) \cdot r \Delta t + (r \Delta t)^2 - (r \Delta t)^3 \\ &= r \Delta t - (r \Delta t)^2 \approx r \Delta t, \text{ small } \Delta t \end{aligned}$$

Spikes count $S = \sum_{j=1}^{T/\Delta t} x_j$

$$\langle S \rangle = \left\langle \sum_{j=1}^{T/\Delta t} x_j \right\rangle = \sum_{j=1}^{T/\Delta t} \langle x_j \rangle = \sum_{j=1}^{T/\Delta t} r \Delta t = r T$$

Recall from reading:

• x_i and x_j are indep. if $P(x_i, x_j) = P(x_i) P(x_j)$

• point: x_i, x_j are indep. $\forall i \neq j$ for Poisson process.

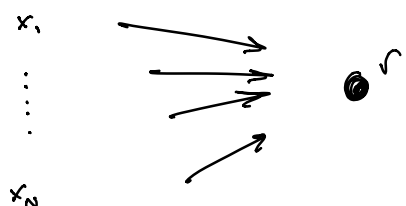
• Fact: $\text{var}(x_i + x_j) = \text{var}(x_i) + \text{var}(x_j)$, if x_i, x_j indep.

$$\rightarrow \text{Var}(S) = \sum_{j=1}^{T/\Delta t} \text{var}(x_j) = \frac{T}{\Delta t} (r\Delta t - (r\Delta t)^2) = rT - r^2 T \Delta t$$

$$\text{Fano Factor} = F_T = \frac{\text{Var}}{\text{Mean}} = \frac{rT - r^2 T \Delta t}{rT} \approx 1 \quad \text{as } \Delta t \rightarrow 0$$

Balanced regime of firing in the cortex.

Q: How is it that neurons produce this poisson firing?



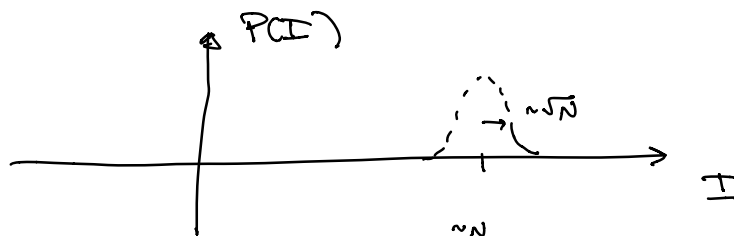
Total input to a neuron

$$I = \sum_{j=1}^N x_j$$

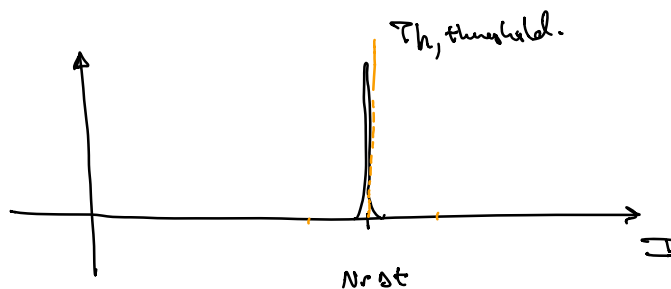
, $N \approx 10^3$

$$\langle I \rangle = N r_{st}$$

$$\text{Var}(I) = N r_{st} \quad ; \quad \text{std}(I) = \sqrt{N r_{st}}$$



large N limit

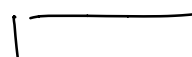


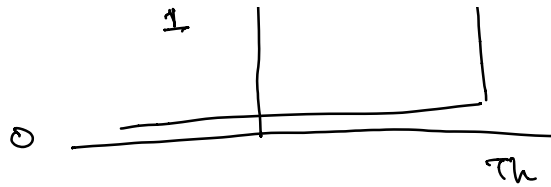
World's simplest spiking neuron model:

spike ($r=1$) if $I > \tau$
 no spike if $I < \tau$

Plot:

$P(\text{spike})$





only extremely finely tuned neuron, on scale of mean inputs, would produce noisy spike w/ prob. not 0 or 1.

Solution: Positive and negative inputs.

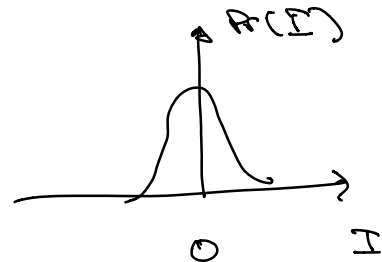
$$i = 1 \dots n/2 \quad \begin{cases} x_i = 1 & \text{w/ prob } r \Delta t \\ 0 & \text{w/ prob } 1 - r \Delta t \end{cases}$$

Excitatory inputs.

$$i = \frac{N}{2} + 1 \dots N \quad \begin{cases} x_i = -1 \\ 0 \end{cases}$$

Inhibitory inputs

Then: $\langle I \rangle = 0$
 $\text{std}(I) = \sqrt{N r \Delta t}$



→ No fine tuning needed for good adaptation firing.

Softky + Koch 1993

Shadlen + Newsome 1998

Chaotic Dynamics: van Vreeswijk + Sompolinsky, 1996.

Final calculation ... for shadler paper.

Say single neuron responses = $r_i + \epsilon r_c$, common part.

Sum over N cells.

$$\sum_i (r_i + r_c) = \sum_{i=1}^N r_i + N \epsilon r_c$$

$$\text{Var} \left(\sum_{i=1}^N r_i \right) = N \text{var}(r_i)$$

$$\text{Var} (N \epsilon r_c) = N^2 \epsilon^2 \text{var}(r_c)$$

$$\text{Fraction of var from common signal} = \frac{N^2 \epsilon^2 \text{var}(r_c)}{N^2 \epsilon^2 \text{var}(r_c) + N \text{var}(r_i)} \xrightarrow{N \rightarrow \infty} 1$$