

EE-239AS hw3 Cong Peng 904760493

1.(a) $\Pr(N=n) \sim \text{Poisson}(\lambda s) = \frac{(\lambda s)^n e^{-\lambda s}}{n!}$

$$\Pr(M=m | N=n) = \binom{n}{m} (1-p)^m p^{n-m}$$

$$\begin{aligned}\Pr(M=m) &= \Pr(M=m | N=n) \Pr(N=n) \\ &= \frac{((1-p)\lambda s)^m e^{-(1-p)\lambda s}}{m!}\end{aligned}$$

Hence, $M \sim \text{Poisson}((1-p)\lambda s)$

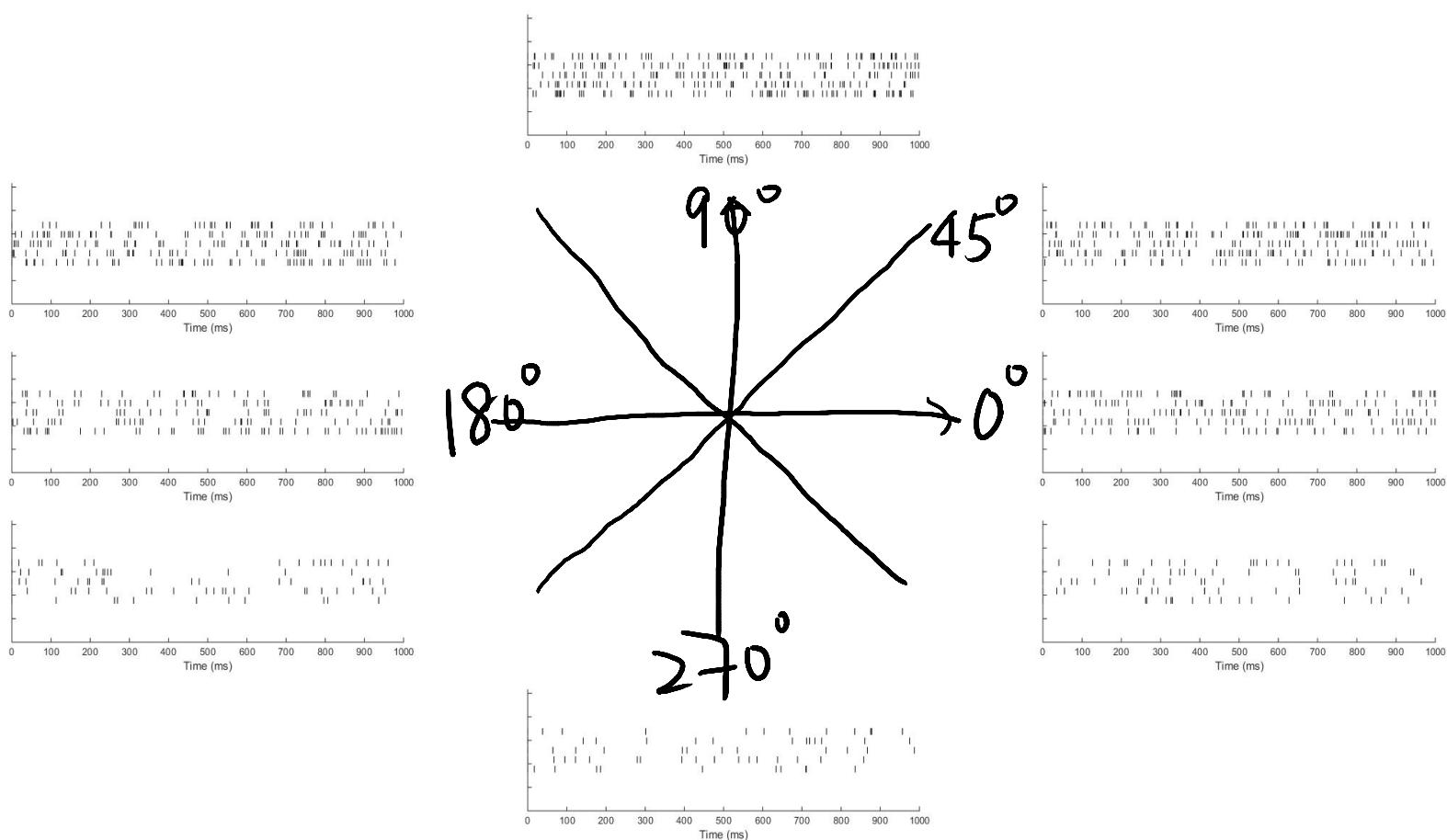
(b) rate: $(1-p)\lambda$

(c) Similarly, $N-M$ (# of spikes dropped within a T second) is also a Poisson distribution at a rate of $p\lambda$

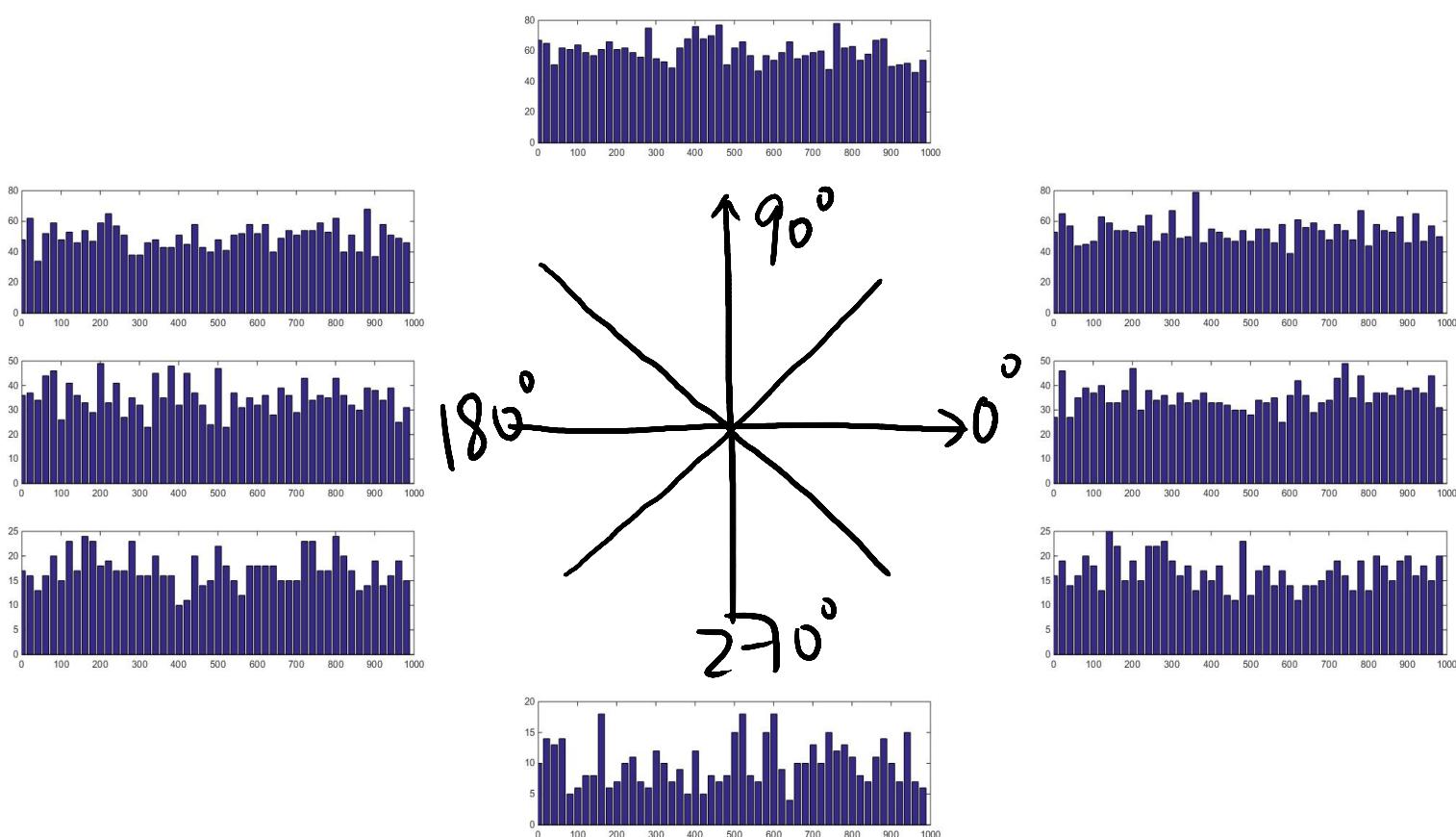
$$\Pr(N-M=r) = \frac{(p\lambda s)^r e^{-p\lambda s}}{r!}$$

2. Homogeneous Poisson process

(a) Spike trains

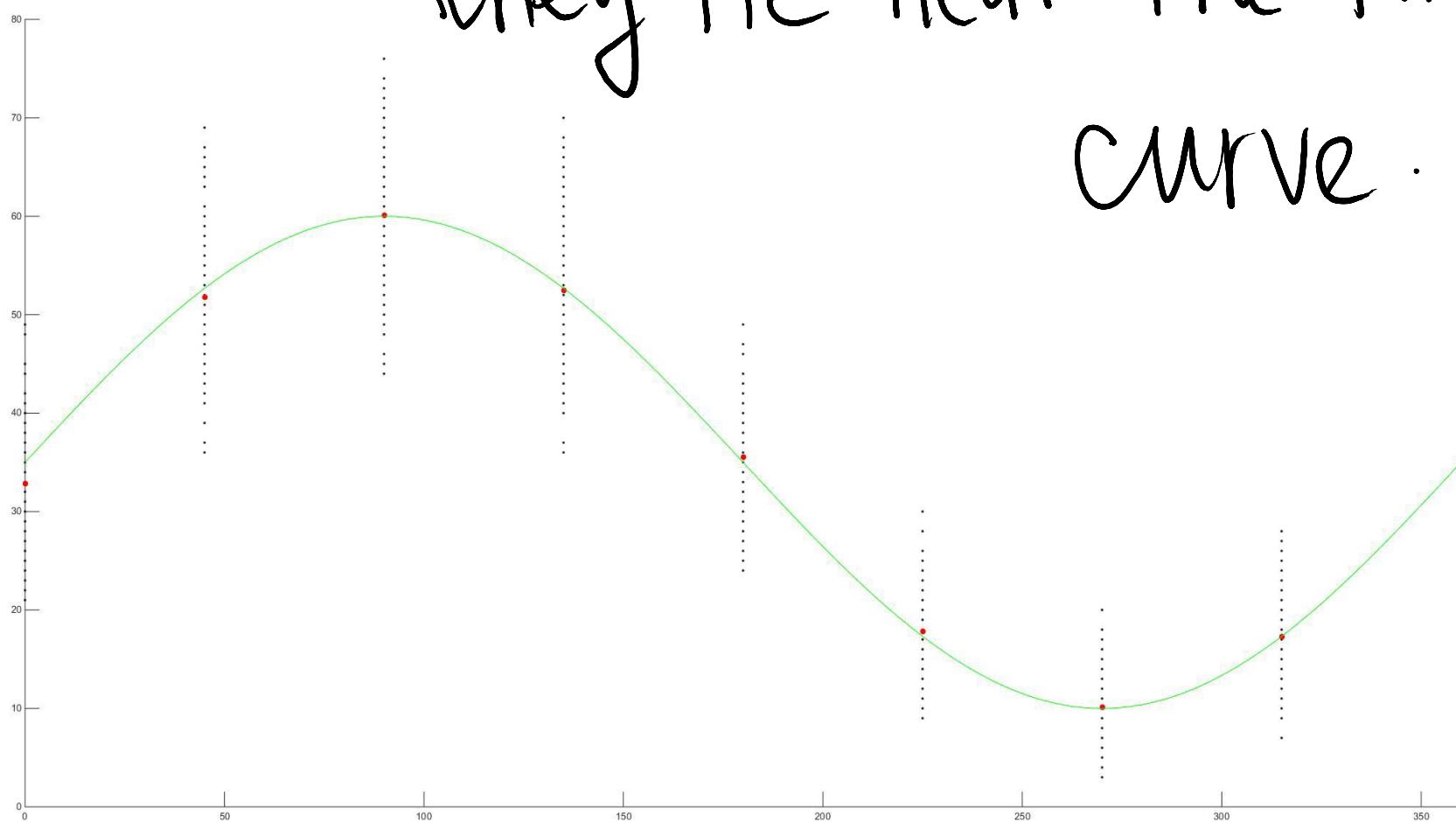


(b) Spike histogram



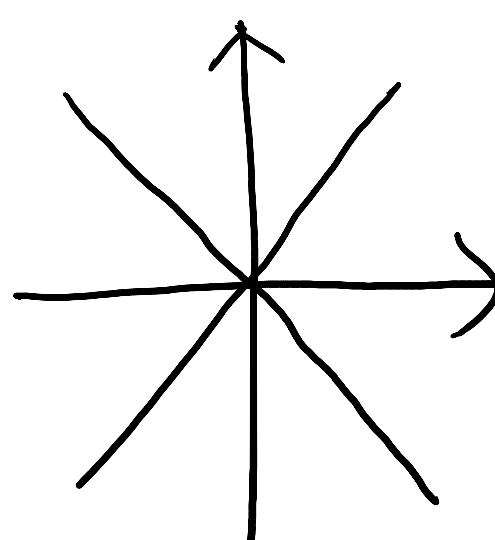
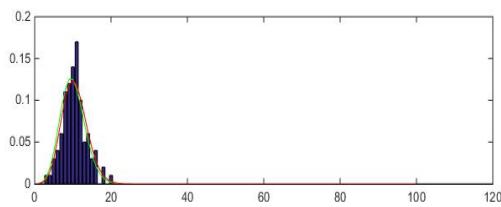
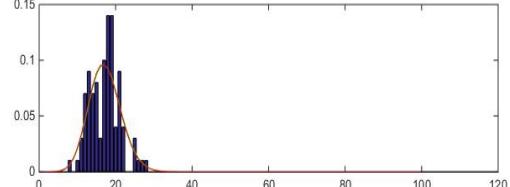
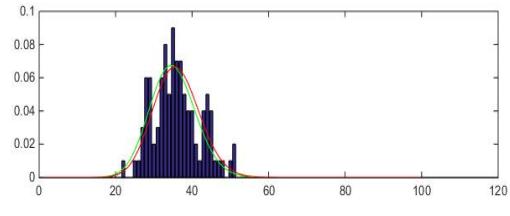
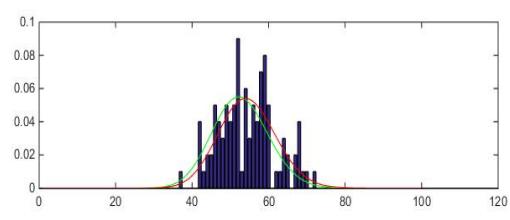
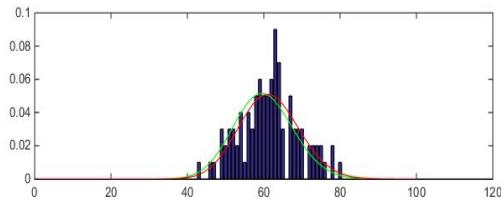
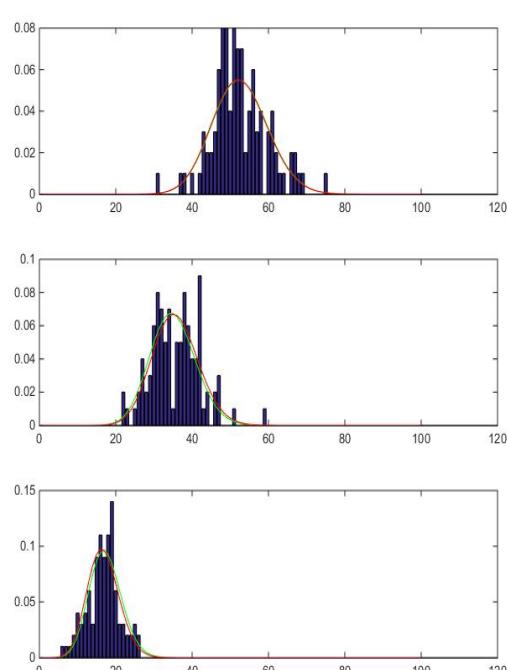
(c) Tuning curve

they lie near the tuning
curve.

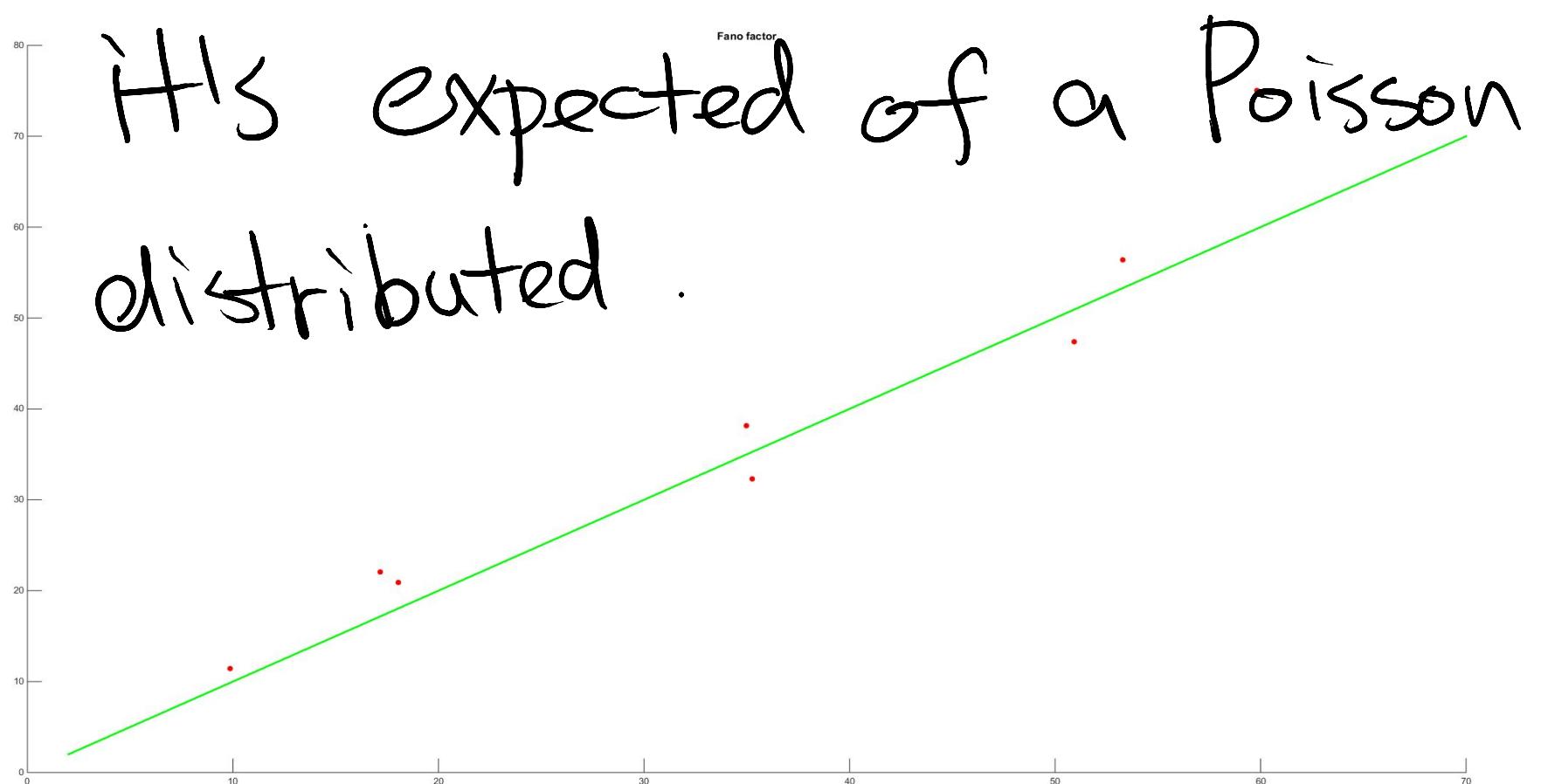


(d) Count distribution

it's Poisson
distributed

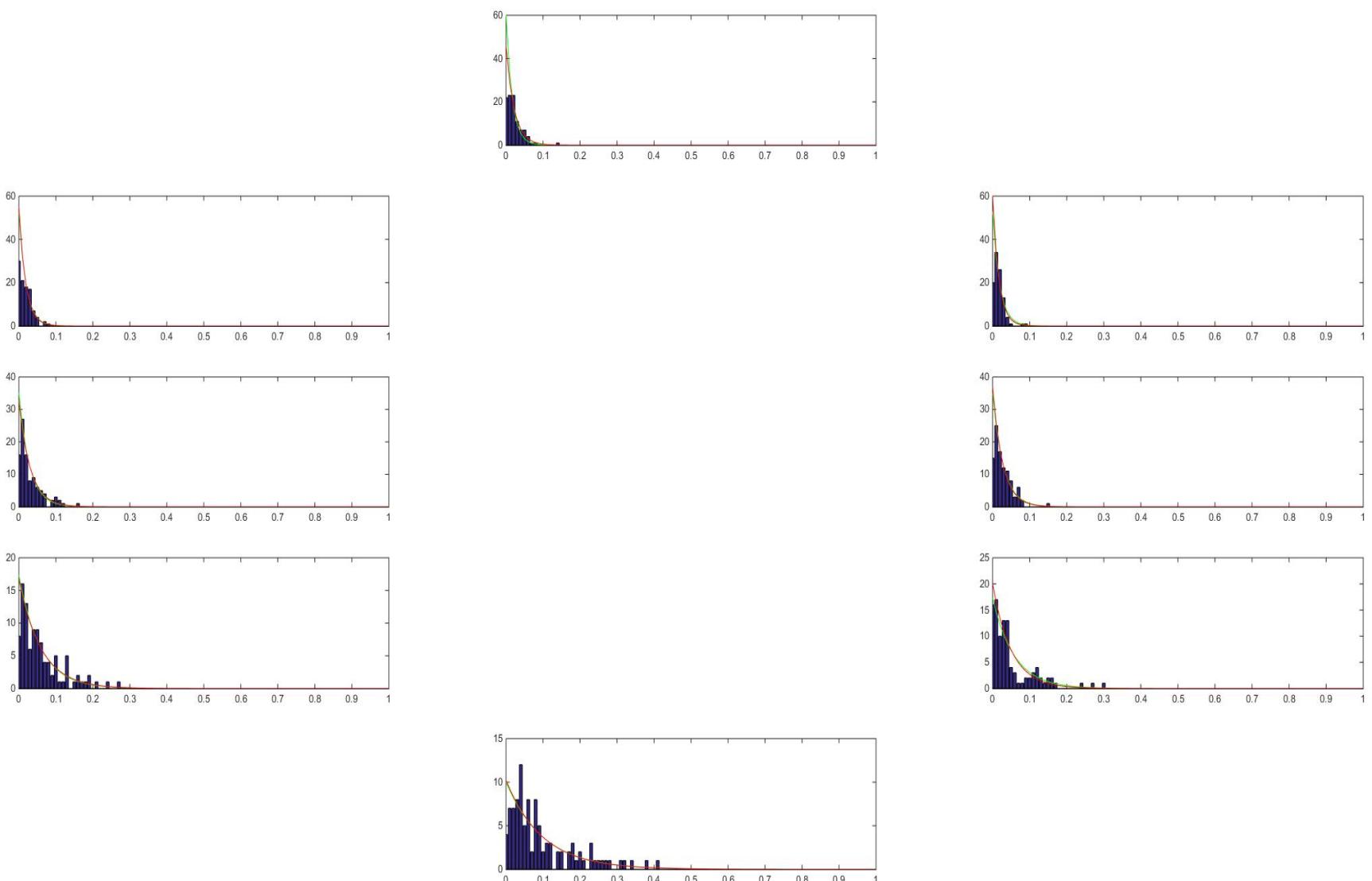


(e) Fano Factor

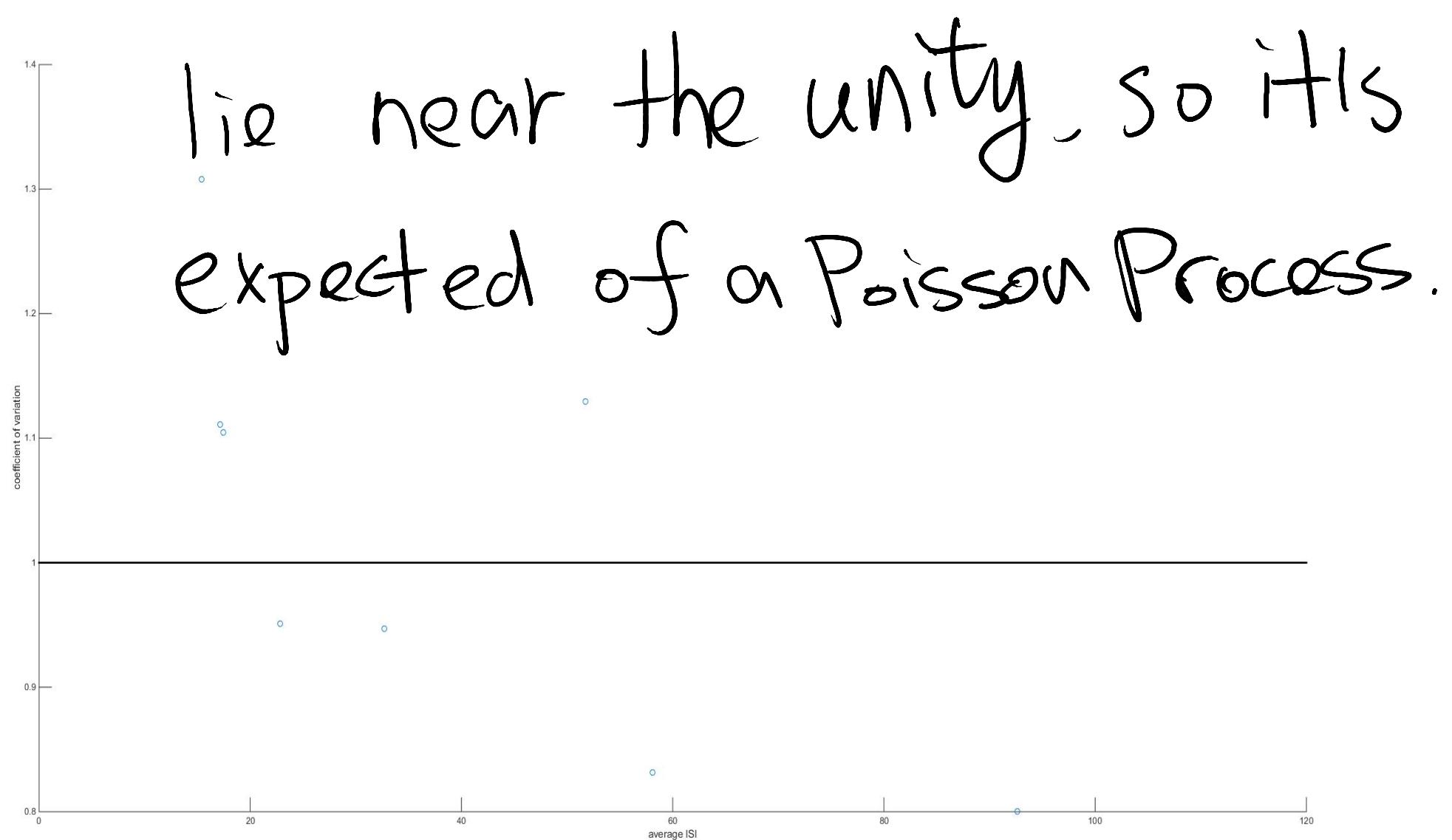


(f) ISI distribution

it's exponential distributed .

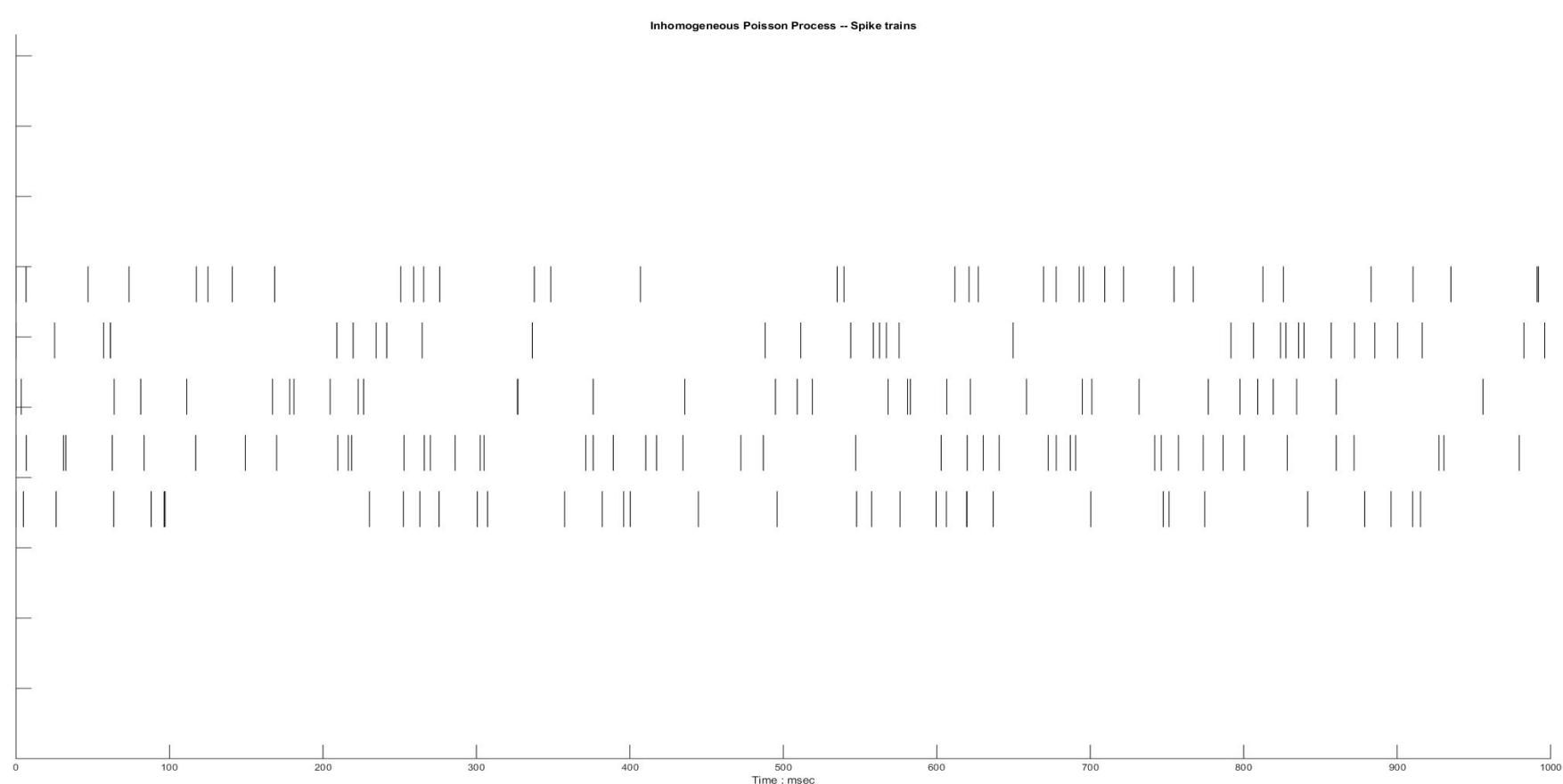


(g) Coefficient of variation

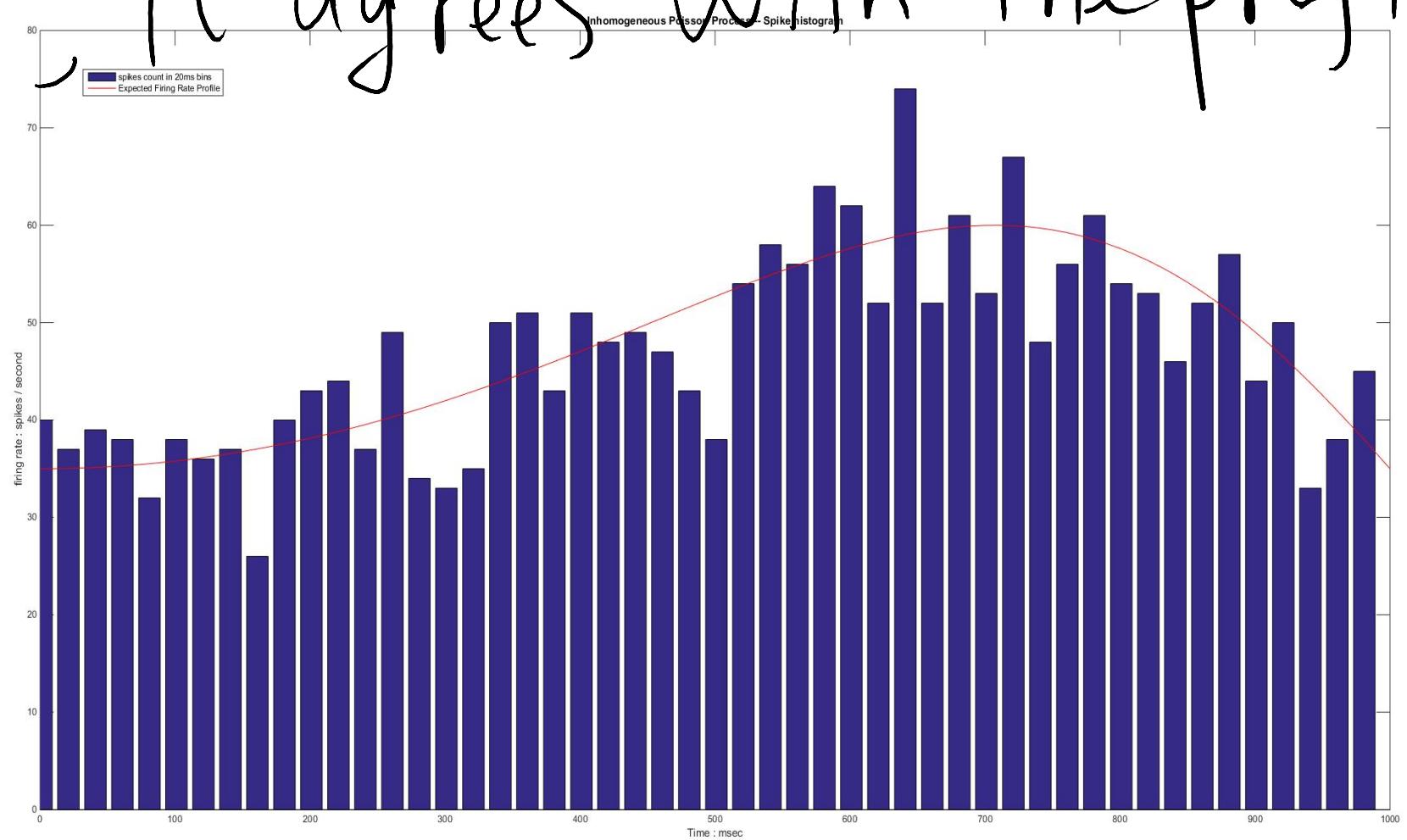


3. Inhomogenous Poisson Process

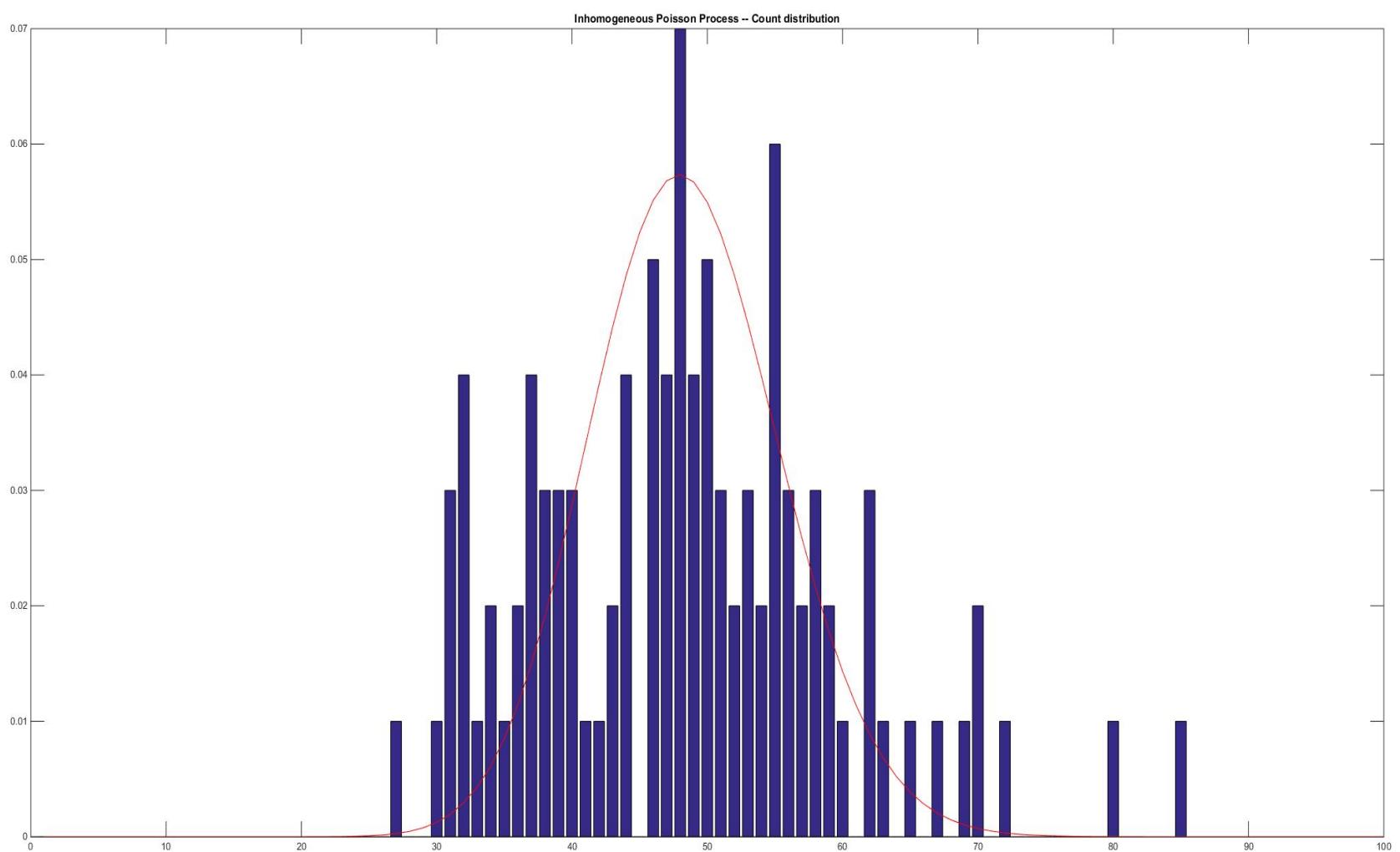
(a) Spike Trains



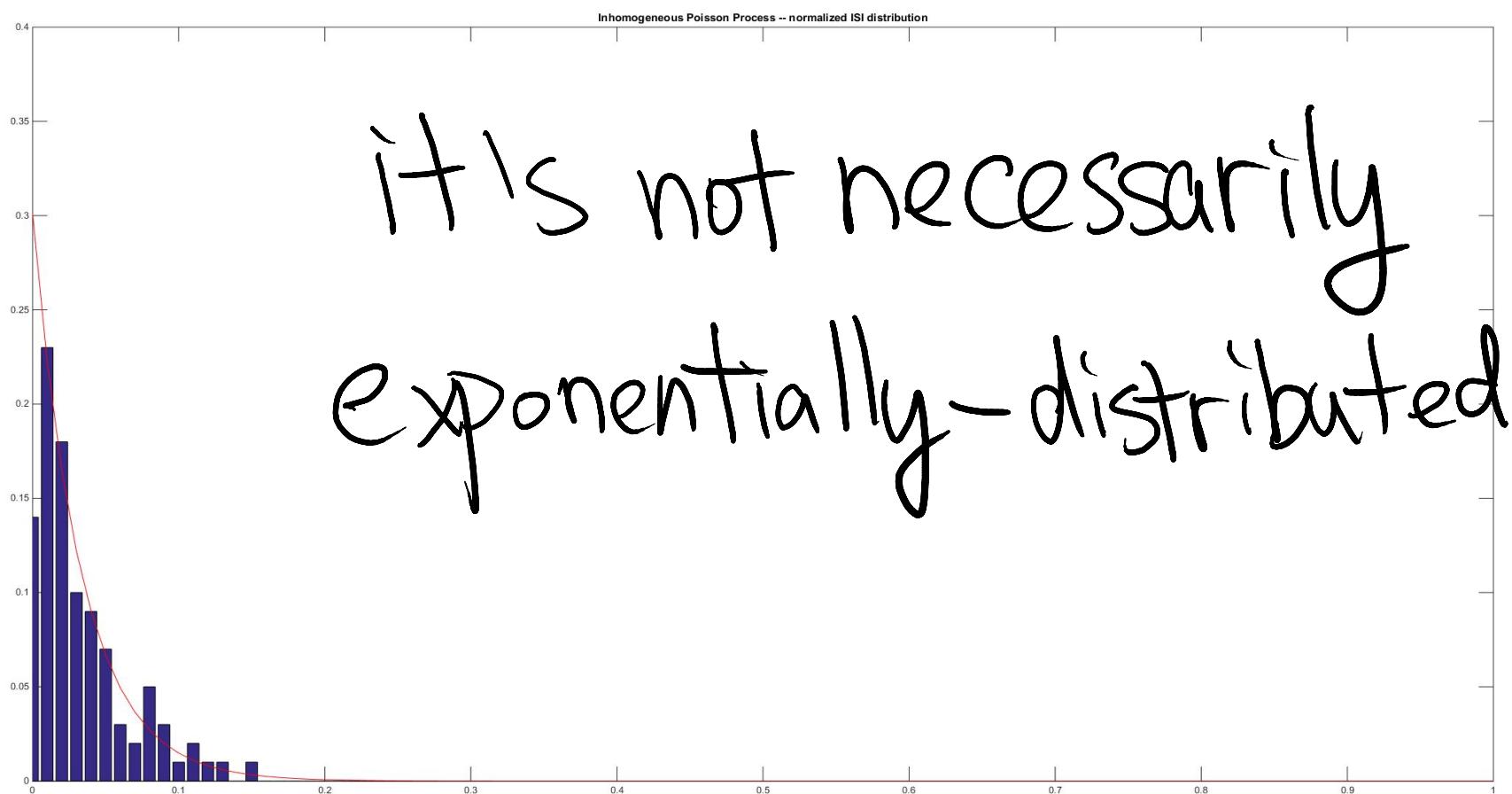
(b) Spike histogram
yes, it agrees with the profile



(c) Count distribution
yes, it's Poisson distributed.

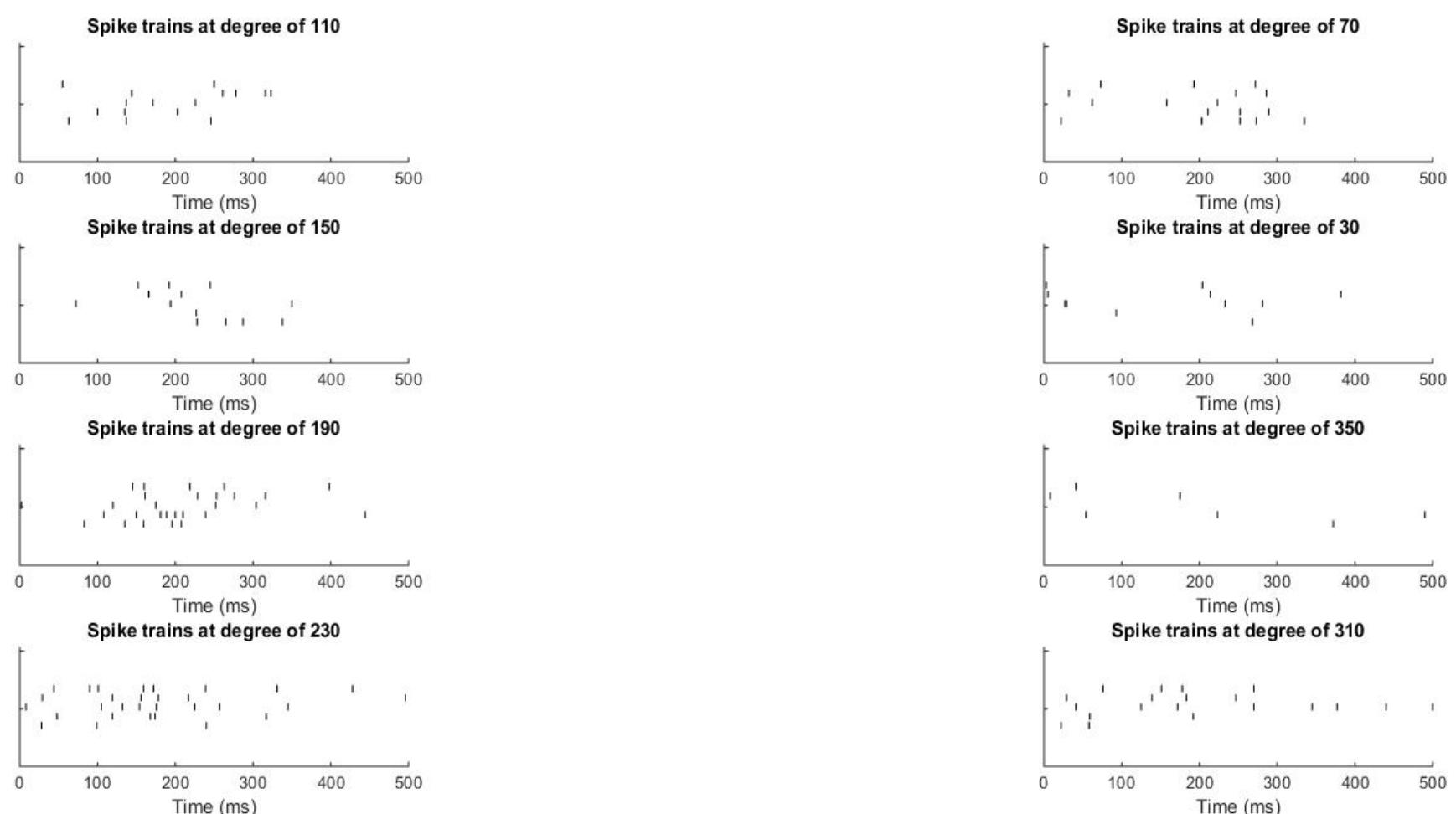


(d) ISI distribution

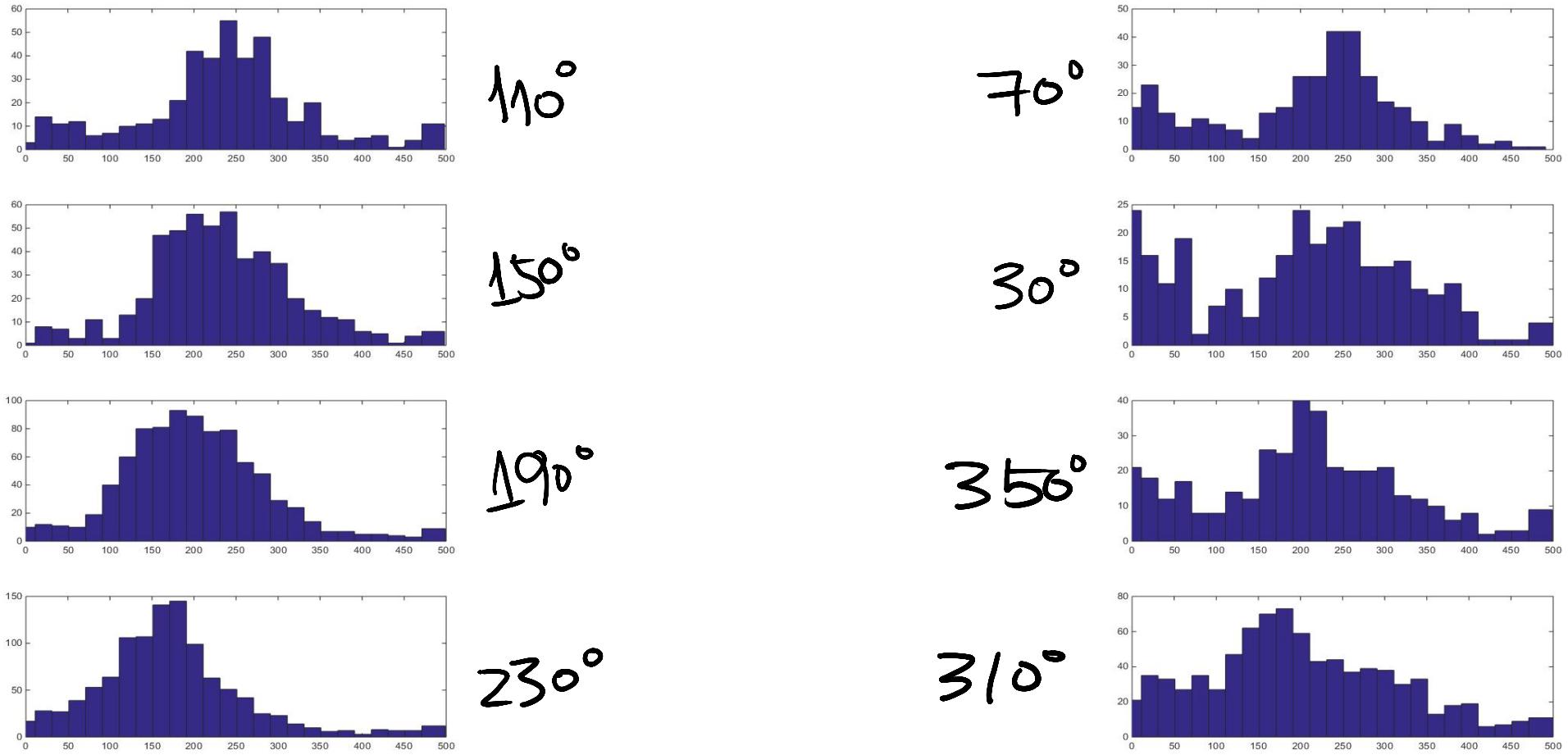


4. Real Neural data

(a) Spike trains

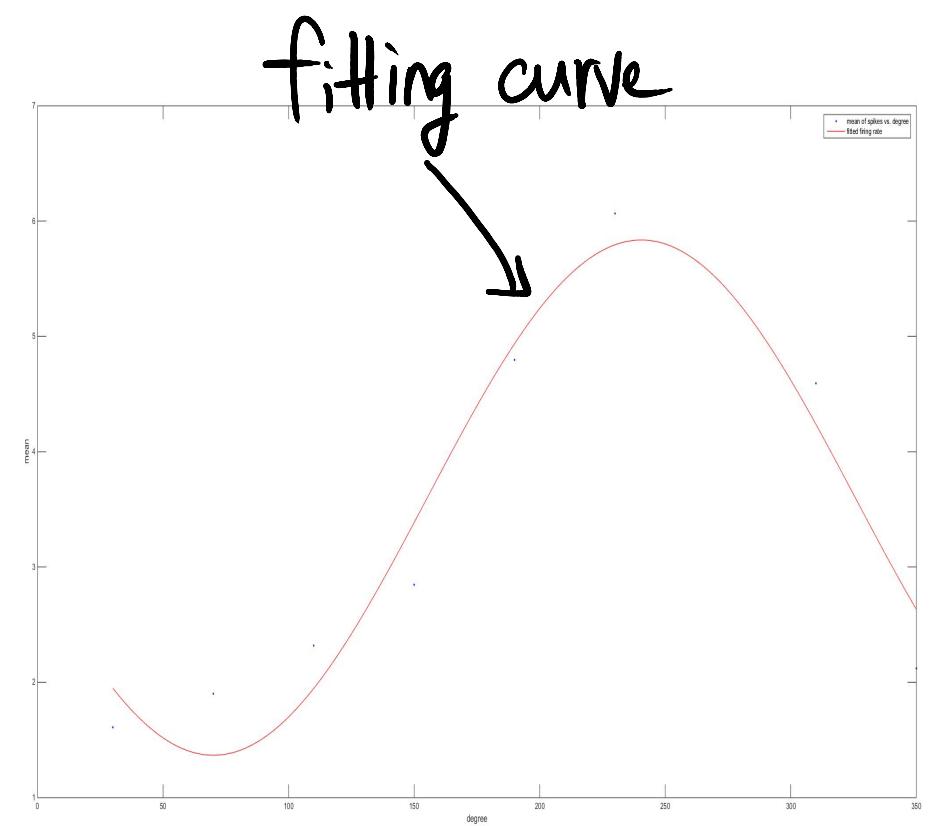
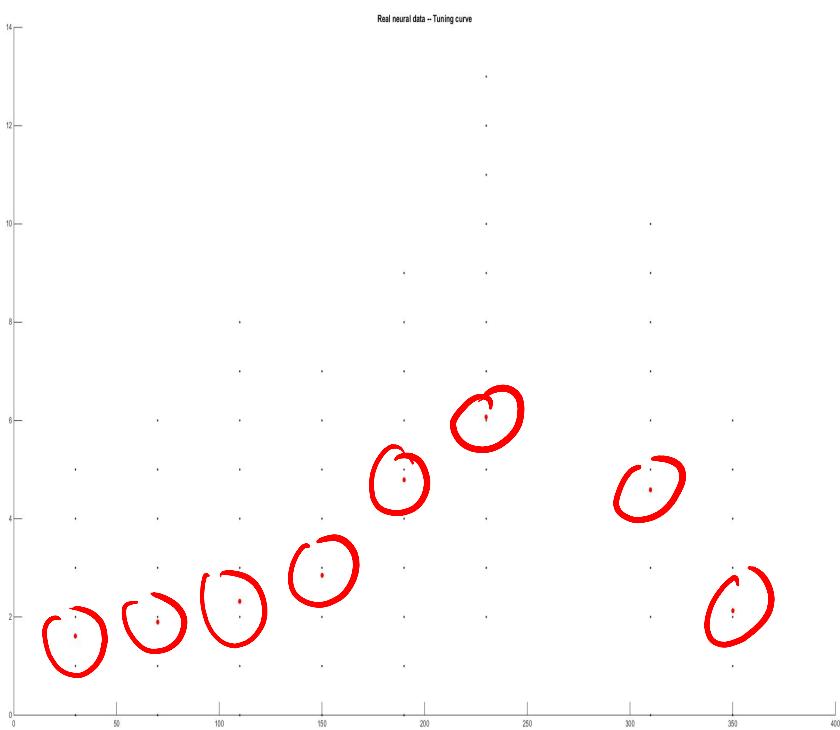


(b) Spike histogram

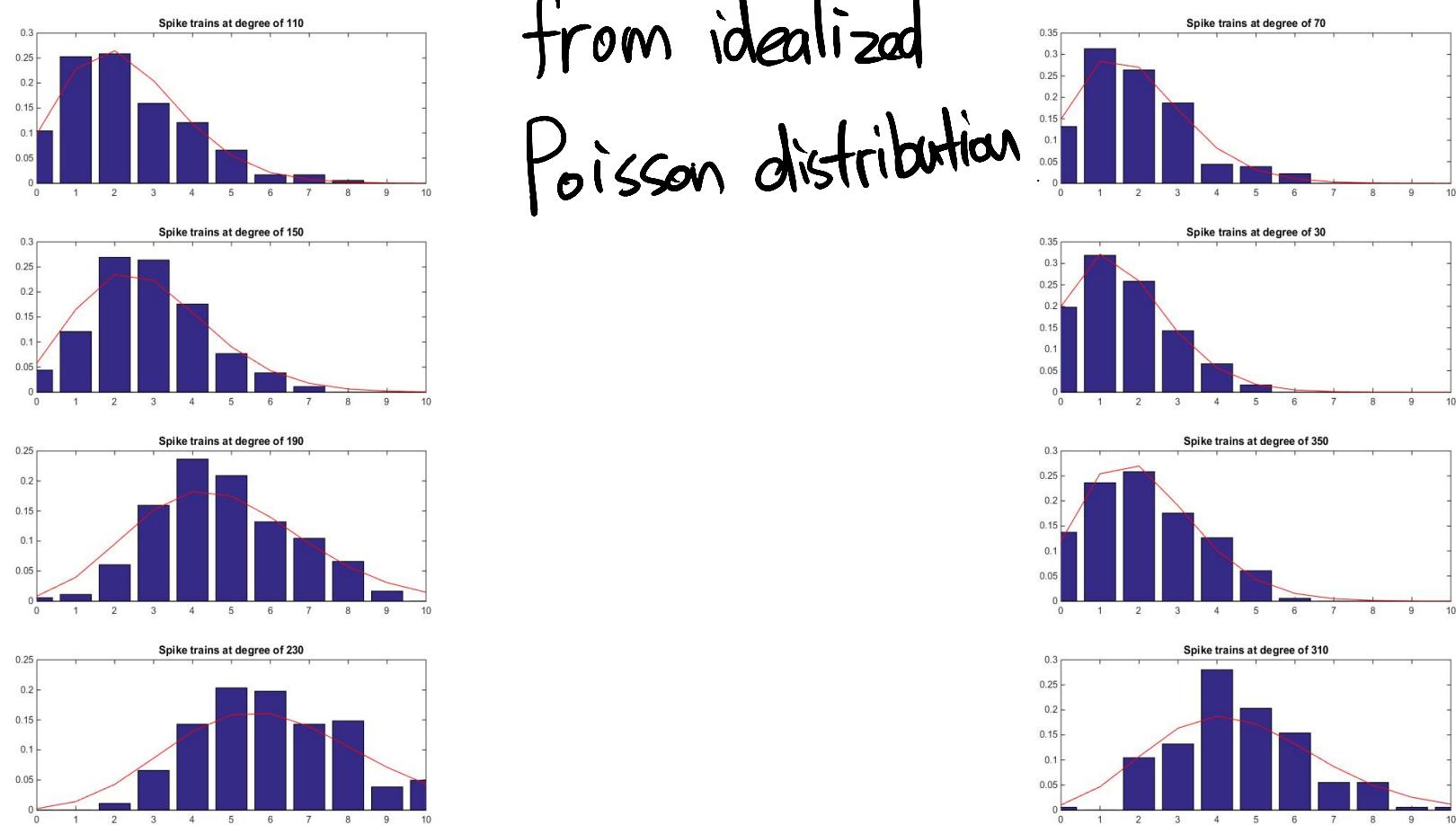


(c) Tuning curve

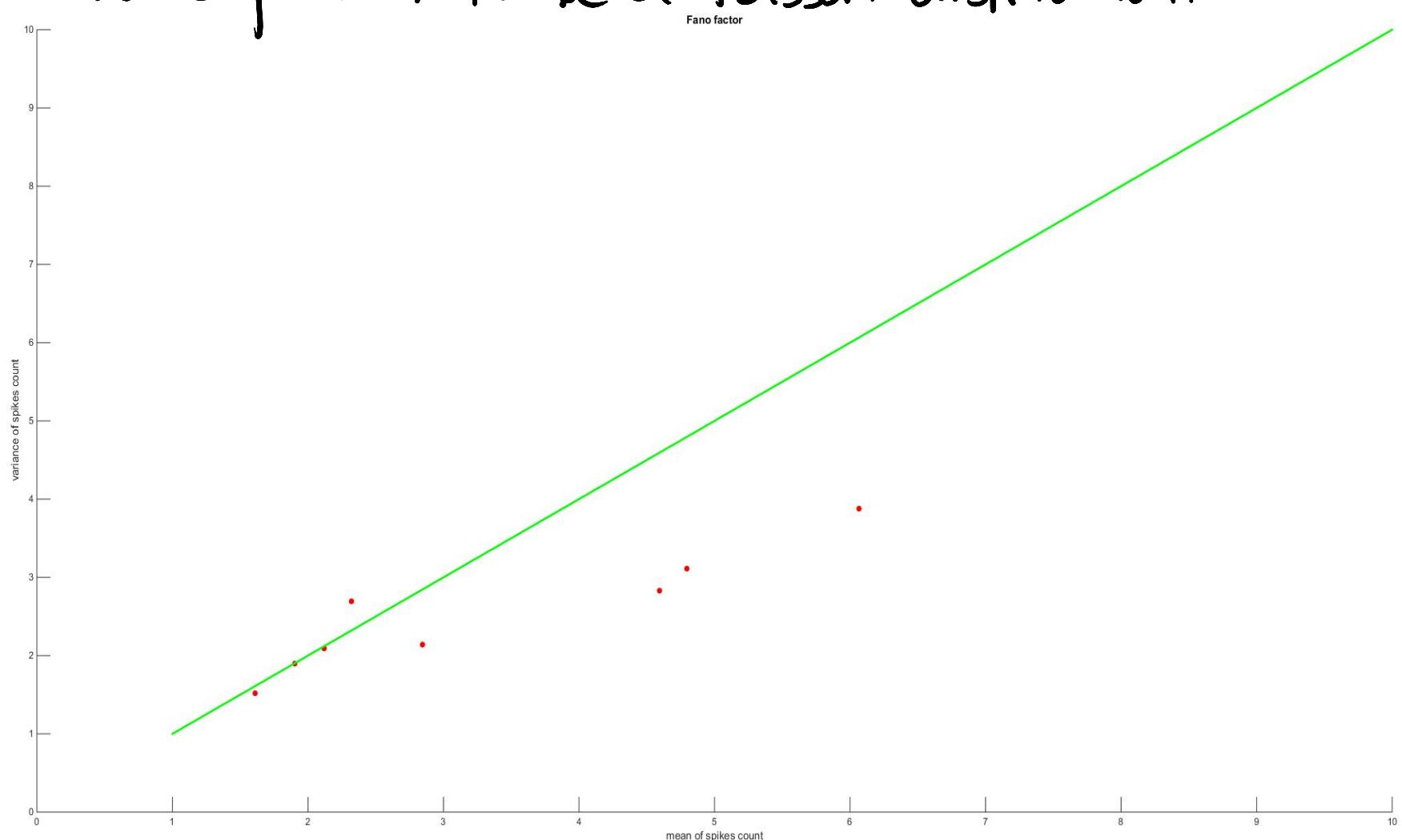
red points — mean of spikes



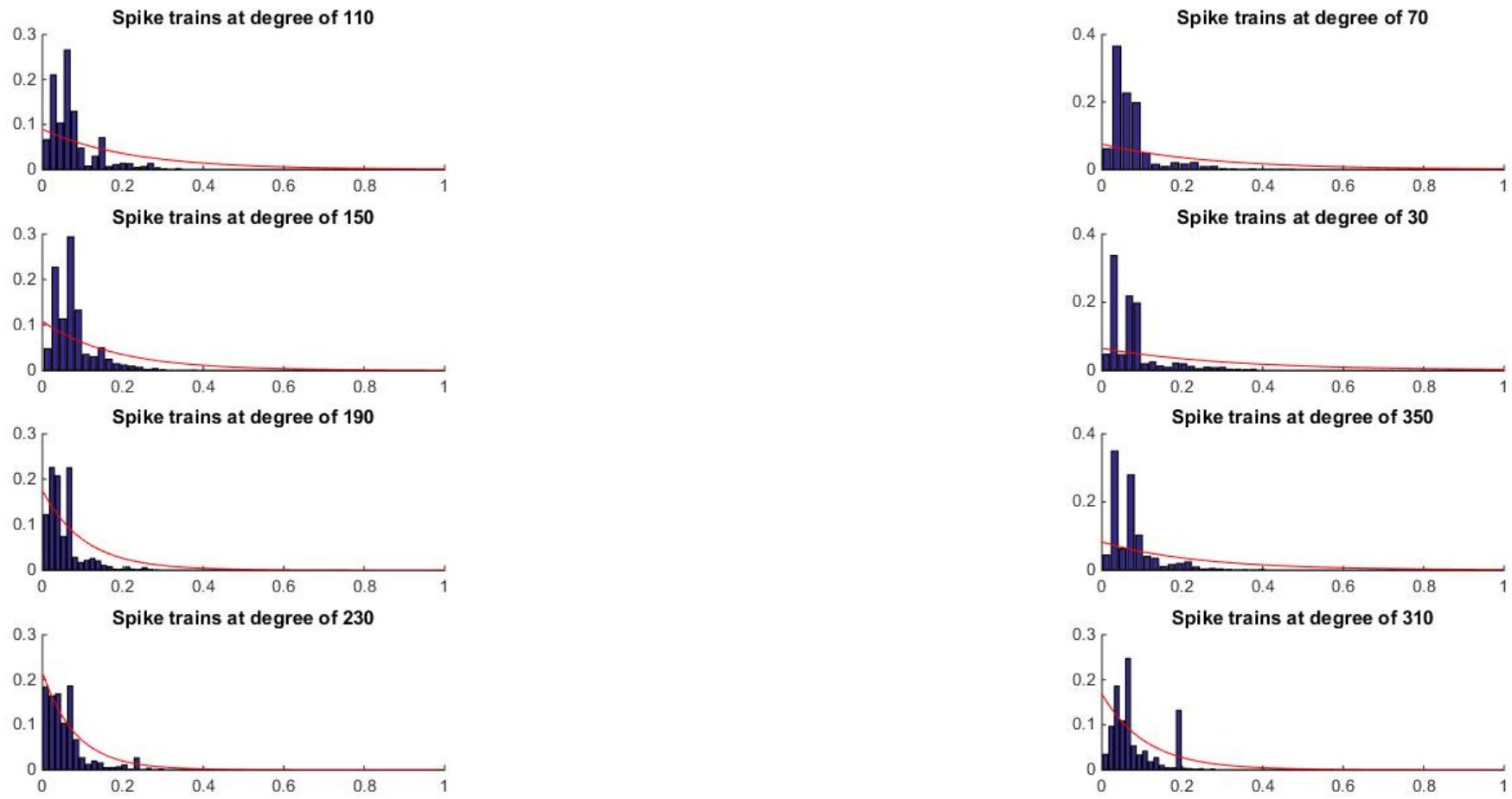
(d) Count distribution due to refractory period, it differs from idealized Poisson distribution



(e) Fano factor they lie near 45° diagonal, and it is expected to be a Poisson distribution.



(f) ISI distribution



Q: Why empirical distributions differ from the idealized exponential distribution?

- ① Firing rate varies over time
- ② Because it is a real neural data analysis, we have to consider refractory period.