

ECE C143A/C243A, Spring 2023

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Homework #3

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Due Friday, 5 May 2023, uploaded to Gradescope.

Covers material up to Poisson Processes III.

100 points total.

1. (13 points) You are recording the activity of a neuron, which is spiking according to a Poisson process with rate λ . At some point during your experiment, the recording equipment breaks down and begins dropping spikes randomly with probability p .
 - (a) (10 points) Let the random variable M be the number of recorded spikes with the broken equipment. Show that the distribution of M is $\text{Poisson}((1-p)\lambda s)$. (Hint: If N is a random variable denoting the number of actual spikes, what is $\Pr(M = m | N = n)$?)
 - (b) (1 points) What is the rate of the Poisson process in part (a)?
 - (c) (2 points) What is the distribution on the number of spikes dropped within a τ second interval?
2. (35 points) Homogeneous Poisson process
Complete the Jupyter notebook `hw3p2.ipynb`. Print your notebook code and output to a pdf as your submission for this question.
3. (22 points) Inhomogeneous Poisson process
Complete the Jupyter notebook `hw3p3.ipynb`. Print your notebook code and output to a pdf as your submission for this question.
4. (30 points) Real neural data
Complete the Jupyter notebook `hw3p4.ipynb`. Print your notebook code and output to a pdf as your submission for this question.

12) neuron spiking according to Poisson process with rate λ .

→ recording equip. breaks, starts dropping spikes randomly with probability p

if N is a RV denoting the actual # of spikes, what is $\Pr(M=m | N=n)$

RV # of recorded spikes

$$P(N=n) = \frac{(\lambda s)^n e^{-\lambda s}}{n!}$$

$$P(M=m | N=n) = \frac{n!}{(n-m)! m!} (1-p)^m p^{n-m}$$

$$P(M=m | N=n) = \frac{P(M=m, N=n)}{P(N=n)}$$

$$P(M=m | N=n) P(N=n) = \underline{P(M=m, N=n)}$$

$P(M=m)$ = sum across $P(M=m, N=n)$

$$P(M=m) = \sum_{n=0}^{\infty} P(M=m, N=n)$$

$$P(M=m) = \sum_{n=0}^{\infty} P(M=m | N=n) P(N=n)$$

$$= \sum_{n=0}^{\infty} \frac{n!}{(n-m)! m!} (1-p)^m p^{n-m} \cdot \frac{(\lambda s)^n e^{-\lambda s}}{n!}$$

$$= \sum_{n=0}^{\infty} \frac{1}{(n-m)! m!} (1-p)^m p^{n-m} (\lambda s)^n e^{-\lambda s}$$

$$= \sum_{n=0}^{\infty} \frac{(\lambda s)^n e^{-\lambda s}}{(n-m)! m!} (1-p)^m p^{n-m}$$

I need:
$$\frac{[\lambda s (1-p)]^m e^{-\lambda s (1-p)}}{m!}$$

sep.

$$\frac{[\lambda s (1-p)]^m e^{-\lambda s (1-p)}}{m!} \cdot \sum_{n=0}^{\infty} \frac{p^{n-m} (\lambda s)^{n-m} e^{-\lambda s p}}{(n-m)!}$$



sum over poisson
distribution = 1

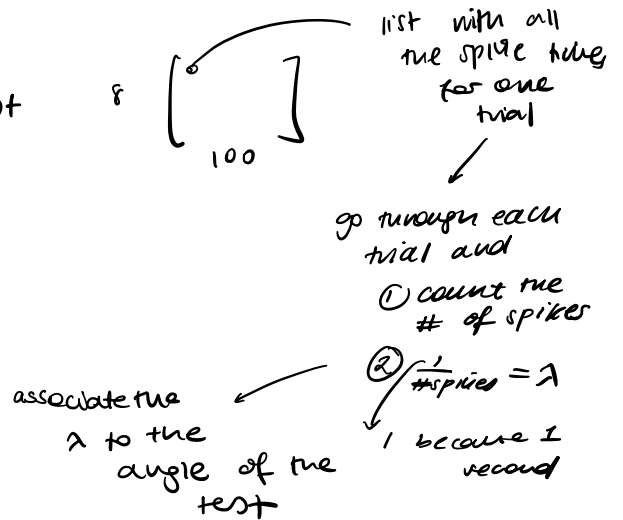
$$P(N=m) = \frac{\lambda s [1-p]^m e^{-\lambda s (1-p)}}{m!}$$

QED

b) $(1-p)\lambda$

c) $\sim \text{poisson}(\rho\lambda\tau)$

2c) There should be 800 points on plot



2d) for each reach angle, plot normalised distribution

↓
count in a class
divided by the total
of observations



1. need total # of spikes

Spike-counts → each element in spike counts has total spike count for that angle and trial

totalspikes needs to be a $\begin{bmatrix} 0_1 \\ 0_2 \\ \vdots \\ 0_8 \end{bmatrix}$ such that you can normalise based on that value

row, column

8 x 1 →

totalspikes[con] = add spike counts.

$$\text{totalspikes}[k] = \sum_{n=0}^{99} \text{spike counts}[k, n]$$

$$\text{totalspikes}[0] = \sum_{n=0}^{99} \text{spike counts}[0, n]$$

spike counts[con, rep]

8 x 100

↓
so I need to go column by column.

first row all columns.

$$\text{totalspikes}[0] = \text{np.sum}(\text{spikecounts}[0,:])$$

\uparrow \uparrow
con con

$$\therefore \text{totalspikes}[con] = \text{np.sum}(\text{spikecounts}[con,:])$$

↗ this did not work.

scipy.stats.poisson → "fitting" a POISSON distribution to empirical distribution → trying to find the poisson distribution that best describes the empirical distr.

↓

find the PARAMETERS

↙

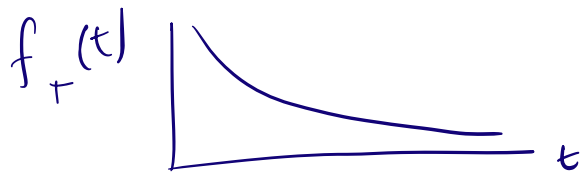
in this case mean

2e) given there are 8 points, one for each angle, I have to take mean and variance of spikes-count[con]

variance →

2f). for each angle, plot normalised dist. of ISIs

I have an $8 \times \begin{bmatrix} \\ 100 \end{bmatrix}$ with spike times → each has its own trial with spike times



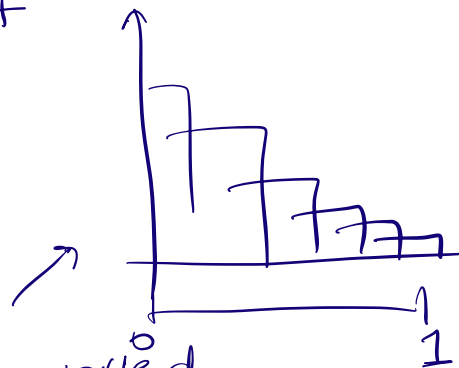
$$f_+(t) = \begin{cases} \lambda e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{else} \end{cases}$$

np.diff ()

ISI

This worked with isi-test

right
now
they
are all plotted



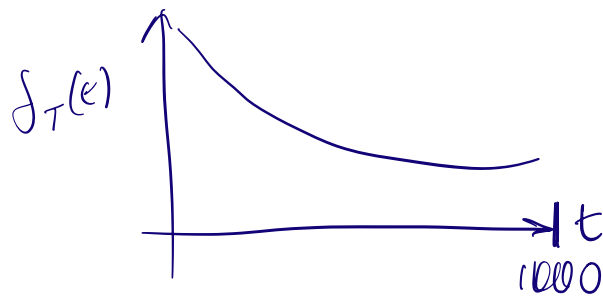
separately. I need to
take mean of each
bin

yes, so
should
be 1000

is it in
ms?

1 because
each trial lasts

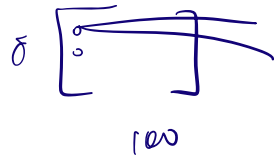
1 second so that's
the max value mi
can take



not working

new approach

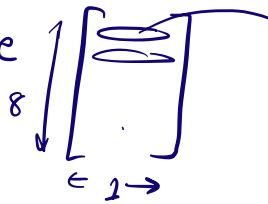
unpack all the



for each row (i.e. each angle).

then plot mat as a histogram

spikes unpacked will be



here this will be a list of all the spike times for ϕ_2

then I will do

$isis_unpacked_8[1]$

for con in range (num - cons):

$isis_unpacked = np.diff(spikes_unpacked[con])$

How to do this:

new vector = []

for con in range (num - cons):

for rep in (num trials):

new vector [con] += spike times [con, rep]

↑ does not work

/

use .exte

2g) spikes_unpacked $\begin{bmatrix} \circ \\ \vdots \\ 7 \end{bmatrix}$ list of all the spike times for that reach angle

isis_array $\begin{bmatrix} \circ \\ \vdots \\ 7 \end{bmatrix}$ list of all ISI's for that reach angle.

```
mean_isi = np.zeros(num_reach)
std_isi = np.zeros(num_reach)
coef_v = np.zeros(num_reach)
```

```
for con in range(num_reach):
    mean_isi[con] = np.mean(isis_array[con])
    std_isi[con] = np.std(isis_array[con])
    coef_v[con] = std_isi[con] / mean_isi[con]
```

```
plt.plot(mean_isi, coef_v)
```