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# Neuroengineering

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we focus on single neuron

#### 5- NEURAL ENCODING

### Learning objectives of the lesson

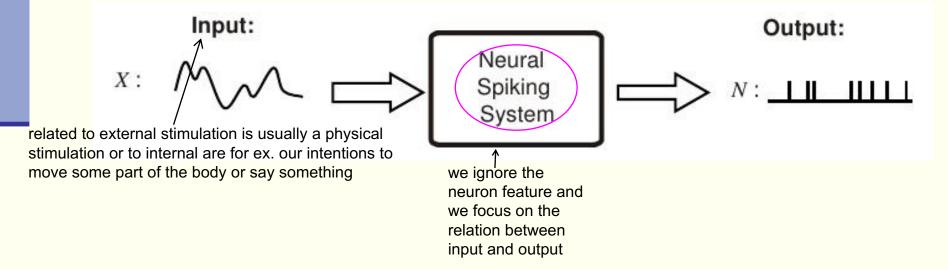
- 1. Understand the definition of neural encoding and decoding
- 2. Describe the rate-coding hypothesis
- 3. Define the neural response function and the spike-count firing rate
- 4. Illustrate the experimental procedure used to build a tuning curve
- 5. Interpret a given a tuning curve, and specifically:
  - a. Describe the neuron behavior as a response to stimulus properties
  - **b.** Compare the neural response to specific values of the stimulus *s*

#### Introduction

it's the process in which we are able to carachterize a neuron as a function linking inputs and outputs to the neuron

<u>Neural (En/De)coding</u>: measuring and characterizing how an external (physical, e.g., light or sound intensity) or internal (e.g. the direction of a planned movement) input received by a neuron is translated into a sequence of action potentials (output):

using a detailed model for each neuron is not feasible because each single model will be very complicated and impossible to use

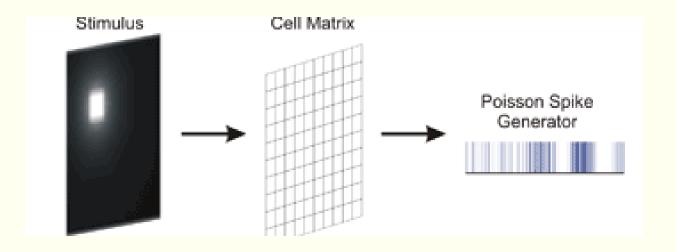


## Neural encoding

From stimulus to response

Neural response = f(stimulus)

- *Aim:* describing how neurons react to different stimuli and trying to **predict their response to a new one** 

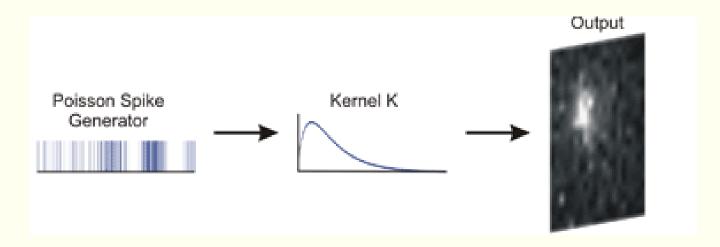


## **Neural decoding**

From neural response to the stimulus that induced it

Stimulus =  $f^{-1}$ (neural response)

- Aim: recognize the stimulus (or its properties) that induced the spike train response

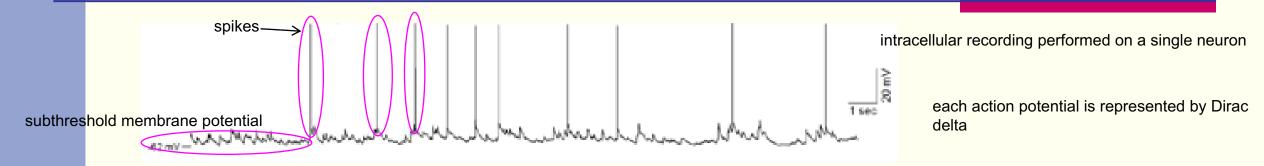


### Rate-coding hypothesis

it's just an hypothesis but supported in several cases

- The rate-coding hypothesis suggests that spike temporal unit frequency (or rate) is the fundamental mechanism of coding information
- e.g.: the number of action potentials from cutaneous nerve fibers in the leg of a cat was proportional to the pressure applied to the footpad (Adrian and Zotterman, 1926)

## Spike trains and firing rate



- Each spike is like a Dirac  $\delta$  function
- For *n* spikes,  $t_i$  is the time in which the i<sup>th</sup> spike occurred:

we use this to model the sequence

$$i=1,2,...,n$$
 and  $0 \le t_i \le T$  ( $T=$  trial duration)

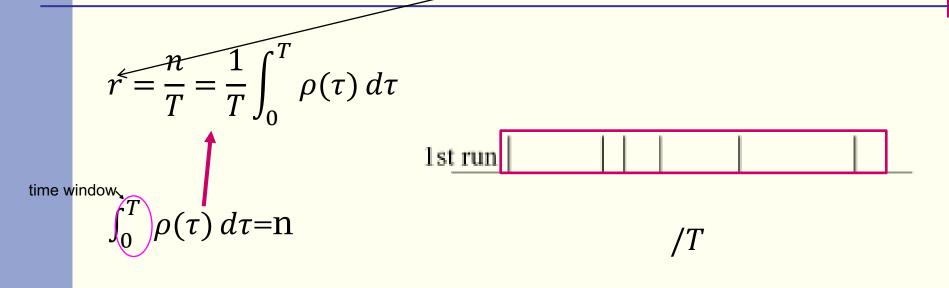
• The spike train can be thus written as:

mathematical description 
$$\rho(t) = \sum_{i=1}^{n} \delta(t-t_i)$$

 $\rho(t)$  is the neural response function

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important property because there is a theory according to which the way in which the neurons encode their info is in the frequency with which they produce a train of action potential(spike train).



r(t): firing rate

r: spike-count firing rate <r>: average firing rate

The spike-count firing rate is the time average of the neural response

function over the duration of the trial—number of events/temporal duration of the time window we're recording

$$n/T = Hz$$

e.g.: 
$$T = 1 \text{ s}$$
  $n = 6$ 

firing rate 
$$\longrightarrow$$
  $r = 6 Hz$ 

## Stimulus-response link

- Spike sequences reflect both the intrinsic dynamics of the neuron and the temporal characteristics of the stimulus
- We can use the spike-count firing rate when the stimulus is stationary (otherwise we need a time-resolved definition for the spike rate)
- Many neurons respond to the same given stimulus → we should examine the relationships of these firing patterns to each other across the population of responding cells (neural networks)

### Stimulus-response link

number of spikes change

- Neural responses can vary across repetitions (trials) even when the same stimulus is presented repeatedly → We cannot describe the timing of each spike deterministically

→ we use a probabilistic approach (<u>average firing rate <r></u>

across trials) average number of spikes more accurate, robust and meaningful for our  $\langle r \rangle = \frac{\langle n \rangle}{T} = \frac{1}{T} \int_{0}^{T} r(t) dt$ ri = n/Te.g.:  $r_1 = 6 Hz$ 1st run considering T=1s 2nd  $r_2 = 4 Hz$  $r_3 = 5 Hz$ 3rd < r > = 5 Hz< r > = < n > /T = [(6+4+5)/3] / 1 = 5 Hz

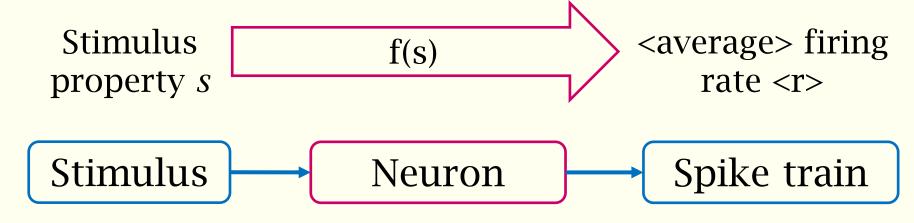
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Tuning curves it's a function that links stimulus property s(input) with firing rate(output). So the output is not the sequence of spikes but the property of firing rate the output is not the sequence of spikes but the property of firing rate

if we have many trials we use the average firing rate as output

A stimulus is characterized by many properties (e.g.: shape, orientation, contrast, movement, ...)

We focus on a single property *s* and we assume it's stationary along each trial (repetition of the stimulation)



$$\langle r \rangle = f(s)$$

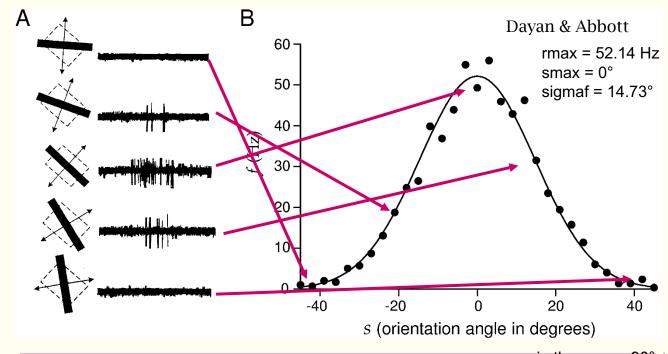
Tuning curve of the neuronal response

you remove the cell from the subject and analize them out of the brain

#### Examples of tuning curves - 1

means that you measure the cellular activity by keeping it in its natural position(so in the brain)

- Primary visual cortex in an animal subject (monkey)
- Visual stimulation (bar with different orientations)
- *s* = orientation angle of the bar (degrees)



$$f(s) = r_{max}e^{-\frac{1}{2}(\frac{s-s_{max}}{\sigma_f})^{\frac{\text{in the range -90°}}{2} + 90°}}$$

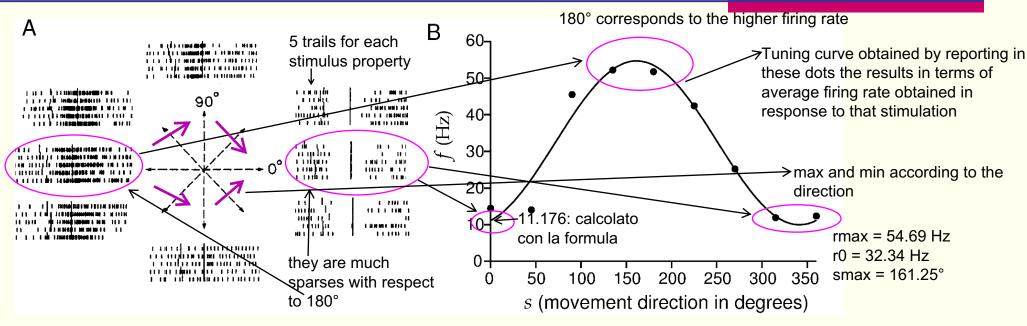
 $s_{max}$ =angle evoking the maximum response  $r_{max}$   $\sigma_f$ =amplitude of the tuning curve

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#### Examples of tuning curves - 2

these are the results of experiments performed by making the trained monkey to reach in different directions

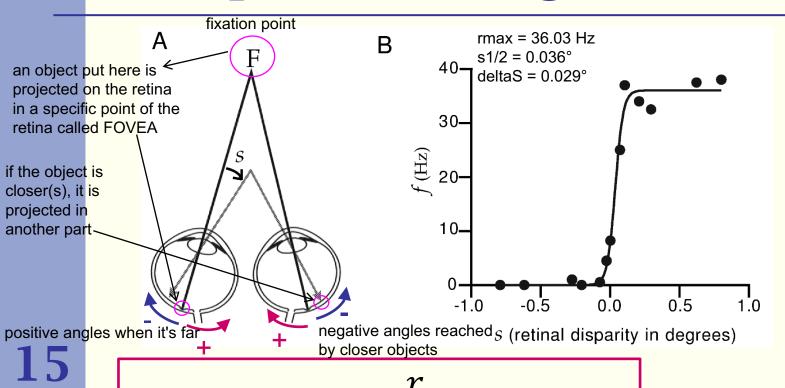


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- A neuron in the <u>primary motor cortex</u> of a monkey trained to reach in different <u>directions</u>
- The firing rate of the cell is correlated with the direction of arm movement s (in degrees)

$$f(s) = r_0 + (r_{max} - r_0)\cos(s - s_{max})$$

s=movement direction (degrees)  $s_{max}$ =angle evoking the max response  $r_{max}$   $r_0$ =offset (to avoid negative firing rate)
or background firing rate

#### Examples of tuning curves - 3



 $r_{max}$ controls how quickly  $1 + exp[(s_{1/2} - s)/\Delta_s]$ the firing rate moves from 0 to the max. So it's small

s=binocular retinal disparity (degrees)

logistic or →

sigmoidal

function

• Primary visual cortex neuron (cat) reacting to retinal disparity (a difference in the retinal location of an image between the two eyes)

- F: fixation point
- The neuron responds only to positive  $s \rightarrow far$  objects

rom 0 to the max. So t's small (far-tuned neuron)
If deltaS is negative, the firing rate is a monotonically decreasing function of s

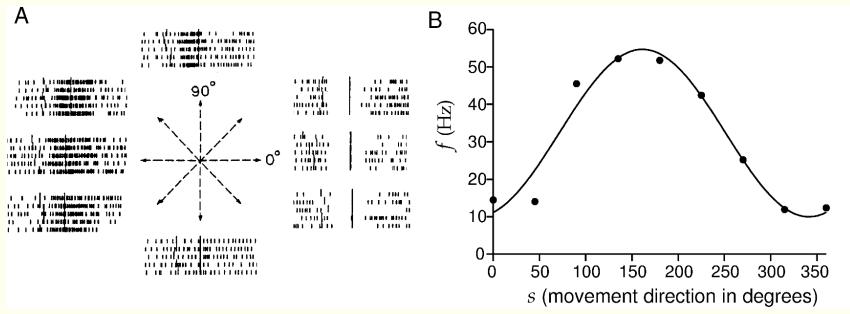
 $S_{1/2}$ =disparity inducing a response equal to  $\frac{1}{2}$  of the maximum  $r_{max}$  $\Delta_s$ =controls how quickly the firing rate increases as a function of s

#### References

- Dayan & Abbott:
  - Chapter 1.1 (From Stimulus to Response)
  - Chapter 1.2 (Spike Trains and Firing Rates, Tuning Curves)

#### **Self-evaluation**

• Given the following tuning curve:



- Is the neural response for a movement direction of 90 degrees greater than for 180 degrees?
- Will I build a different tuning curve for each trial?
- Which firing rate can I expect when the movement direction is 250 degrees?
- If the measured firing rate is 55Hz, can I «guess» which was the movement direction that produced that response?