

## The implementation of the Neuroid in the Gate Control System leads to new ideas about pain processing

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**Abstract** *Introduction:* Several theories have been proposed to elucidate the mechanisms related with pain perception, among which, the Gate Control Theory (GCT) provides one of the most explicit explanations. This theory, as elegantly conceived, is unable to explain how the Frequency-Intensity (F-I) curves exhibited by A $\beta$ - and C-fibres influence pain processing. In this paper, a novel neuron-model known as the Neuroid, which emphasizes the functional rather the physiological character of nerve cells, was used as the main building block to replicate the Gate Control System (GCS). **Methods:** Two A $\beta$ -fibre models were built: one model that preserved the paradoxical relation between the activation threshold and the F-I curve slope, and one model based on the hypothetical average response across the receptive field. **Results:** The results suggest that the average response of the A $\beta$ -fibres does not increase monotonically but reaches a plateau for high intensity stimuli. In addition, it was seen that activation of C-fibres does not necessarily imply the activation of projection neurons and, therefore, the onset of pain sensation. Also, we observed that the activation of A $\beta$ -fibres may both, decrease and increase the activity of the projections neurons, an aspect which has not been directly described in previous works. **Conclusion:** Hypothetical implications arise as a consequence of the implementation of the Neuroid, specifically, about the correlation between the intensity of stimulation and the physiological pain threshold.

**Keywords** Gate control theory, Neuron-model, Frequency-Intensity curve, Pain modulation.

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## Introduction

Pain is currently conceived as a complex phenomenon that involves both physiological and psychological processes (Merskey and Bogduk, 1994). This duality demands an interdisciplinary approach. Different theories have been proposed to elucidate the mechanisms associated with pain perception, among which, Melzack and Wall's Gate Control Theory (1965) provides one of the most explicit explanations. This theory states that the impulses from nociceptive afferents are modulated by the activity of somatosensory afferents, such that the perceived pain sensation, may increase or decrease in intensity. This explains why when we get injured we immediately rub the area around the injury, and pain sensation decreases.

Due to its explicit nature, the Gate Control Theory (GCT) was successfully described in mathematical terms (Britton and Skevington, 1989). This led to the development and implementation of several computational models, specifically for the study of acute pain (Prince *et al.*, 2004). Many of such models have been developed at the cellular and molecular levels (Britton and Skevington, 1989; Britton *et al.* 1995, 1996; Xu *et al.*, 2008), as well as at the neural network level (Minamitani and Hagita, 1981; Haeri *et al.*, 2003). Nevertheless, although some implementations have included morphological and functional differences between primary sensory afferents (Agi *et al.*, 2010), the GCT, as elegantly conceived, does not explain how this differences influence the modulation of nociceptive impulses. It is widely accepted that large-diameter thickly-myelinated mechanoreceptors (A $\beta$ -fibres) respond to tactile stimuli and show high conduction velocities, whereas small-diameter thinly-myelinated/unmyelinated nociceptors (A $\delta$ - and C-fibres) are activated by high intensity stimuli and show lower conduction velocities. Several groups (Cain *et al.*, 2001; Slugg *et al.*, 2000; Van Hees and Gybels, 1981) have totalized the number of action potentials evoked as a function of mechanical stimulus intensity, but this characterization was restricted to thinly-myelinated/unmyelinated afferent neurons and no recent studies seem to be available to compare the frequency-intensity (F-I) characteristics (also referred as stimulus-response function) of mechanoreceptors and nociceptors. On the other hand, some studies (Levy and Strassman, 2002; Slugg *et al.*, 2000) demonstrate that F-I characteristics of A $\delta$ -fiber nociceptors exhibit relatively low thresholds and steep slopes, whereas C-fiber nociceptors exhibit high thresholds and flat slopes. In fact, the flatness observed in C-fiber responses suggested that they could not encode high intensity mechanical stimuli. Subsequent studies

show that the average response of C-fiber nociceptors across the receptive field does not reach a plateau, but increases monotonically with stimulus intensity (Slugg *et al.*, 2004). This leads to the assumption that individual and collective responses of A $\beta$ -fibres may also differ. Even though there is an inverse correlation between the activation threshold and the slope of the F-I curve (the higher the threshold, the lower the slope of the curve), the average response of A $\beta$ -fibres should reach a plateau rather than increase monotonically to a high intensity mechanical stimulation. Otherwise, although nociceptors were activated, we could not perceive pain since their response would always be obscured by the activity of tactile fibres.

In this paper, we replicate the Gate Control System (GCS) using the Neuroid (Argüello *et al.*, 2012) to offer some explanations about how the noxious information is processed at the dorsal horn, considering the differences between primary sensory afferents regarding to their F-I curves.

## Methods

### A "customizable" Neuron-model

In a recent work, a neuron-model based on the mathematical description of the signal transmitted along the neuron's axon (i.e. the train of action potentials), the Neuroid, was developed (Argüello *et al.*, 2012). Let  $\delta(t)$  be a single spike, if the resulting depolarization  $s(t)$  exceeds the activation threshold  $umbr$ , then the signal propagating along the axon,  $y(t)$ , is given by

$$y(t) = \sum_{n=0}^{\infty} \delta \left( t - \frac{n\beta T}{s(t) - umbr} \right), \quad (1)$$

where  $T$  is the time between two consecutive spikes, and  $\beta$  is the reciprocal of the Neuroid's F-I curve slope. This frequency-modulated impulse train is "demodulated" into a graded signal,  $nt\_out(t)$ , which resembles the input signal,  $s(t)$ , but prolongs over a finite period of time after the last spike, to model somehow the temporal and concentration profiles of neurotransmitter in the synaptic cleft ( $nt\_out(t) \approx s(t)$ ). As described by Argüello *et al.* (2012), two other parameters were required to adjust the amplitude of the output signal ( $K_r$ ), and prevent indefinite growth ( $maxcount$ ). The simulation was performed in LabVIEW (version 10.0 running on Acer Aspire One) by three algorithms, to emphasize the computational (functional) rather than physiological neuron-model, known as the Neuroid.

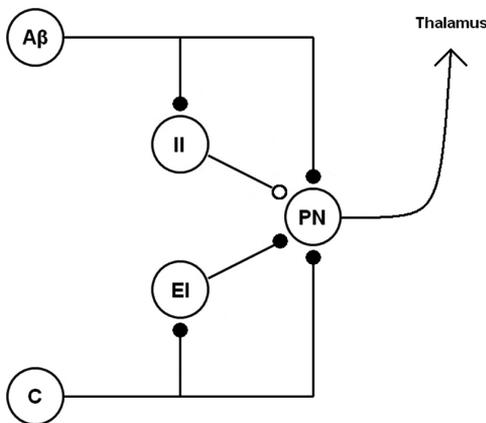
### Modeling the gate control system

The architecture proposed by Melzack and Wall (1982), without the descending control pathway, was modeled using a customized Neuroid network. As shown in

Figure 1, the projection neuron (PN) received signals from one large-diameter neuron ( $A\beta$ ), one small-diameter neuron (C), one excitatory interneuron (EI), and one inhibitory interneuron (II). For both,  $A\beta$ - and C-fibres, we used the median mechanical thresholds reported by Cain *et al.*, (2001). To get the F-I curve slope, we found the equation of the line ( $f(i) = mi + b$ ) that best fits the experimental results obtained by Slugg *et al.* (2000) and by Prescott and De Koninck (2002), for C-fiber and lamina I tonic interneurons, respectively, such that it was possible to correlate each experimental equation with the F-I curve of the Neuroid, which is shown as follows

$$f_N(i) = \frac{i}{\beta T} - \frac{umbr}{\beta T} \tag{2}$$

Thus, with  $T = 2$  ms (the minimum value of time between two consecutive spikes in LabVIEW environment), we correlated the slope of the experimental line equation (denoted as  $m$ ) with  $1/\beta T$ , such that we could get the value of  $\beta$ . For interneurons, the activation threshold was set as the minimum current value for which was obtained a measurable response. The mechanical stimulation range used by Slugg *et al.* (2000) was extended to 200 mN, since values for mechanical cutaneous pain threshold varies between 70 and 130 mN, as has been reported in clinical studies of healthy patients (Rolke *et al.*, 2006; Wienemann and Chantelau, 2012). The range of electrical stimulation used in the characterization of the lamina I tonic interneurons (0-300 pA), as well as the one used by Slugg *et al.* (2000), was normalized. The parameters obtained for lamina I interneurons were also used to configure the projection neuron.



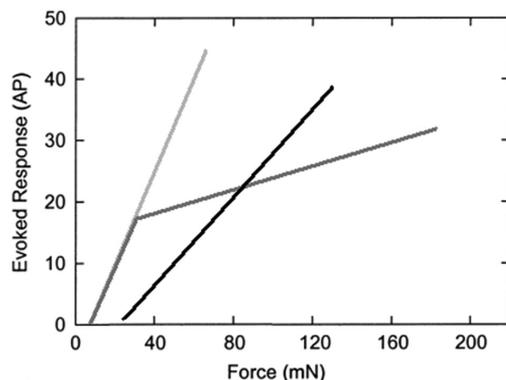
**Figure 1.** The architecture of the Gate Control System used for modeling purposes. The descending control pathway was omitted. Filled circle: excitatory influence. Empty circle: inhibitory influence.  $A\beta$ :  $A\beta$ -fibre. C: C-fibre. EI: excitatory interneuron. II: inhibitory interneuron. PN: projection neuron.

For the  $A\beta$ -fibre, two models were built: one first model that preserved the paradoxical relation between the activation threshold and the slope of the F-I curve, and another one based on the hypothetical average response of mechanoreceptors across the receptive field. The value of  $\beta$  was arbitrarily assumed to guarantee that the F-I curve of the  $A\beta$ -fibre model had a steeper slope. For the second model, a change in the slope value would occur at 35 mN; then this line would intersect the F-I curve of the C-fibre at 70 mN (see Figure 2). Thus, the F-I curve could reach a “plateau” at mechanical intensities below the mechanical pain threshold. We adjusted the values of  $K_r$  and  $maxcount$ , such that  $nt\_out(t) \approx s(t)$  for each single Neuroid, in agreement with the results obtained by Argüello *et al.* (2012). The values of the parameters are summarized in Table 1.

To allow the interconnection of two or more Neuroids and, at the same time, emulate the excitatory and inhibitory influences depicted in the GCS, we built an additional block that allowed us to obtain a linear combination of multiple inputs, each of which would be multiplied by a synaptic weight ( $W > 0$  for excitatory influence, and  $W < 0$  for inhibitory influence). Since the projection neuron receives three terminals with positive sign and only one with negative sign, we thought we could balance excitatory and inhibitory influences as a function of the number of inputs as follows:

$$n_e W_e + n_i W_i = 0, \tag{3}$$

where  $n_e$  and  $n_i$  represent the total number of excitatory and inhibitory inputs, respectively, and  $W_e$  and  $W_i$  their corresponding synaptic weights. As far we know, the inhibitory neurons represent almost 1/3 of the total population of neurons in laminae I-III (Todd and Sullivan, 1990), but direct measurements



**Figure 2.** Frequency-Intensity curves for modeling different primary sensory afferents. Black: C-fiber model. Light gray: Paradoxical model of  $A\beta$ -fibre. Dark gray: average response model of  $A\beta$ -fibre.

**Table 1.** Settings for the 5-Neuroid array that replicates the Gate Control System.

Neuron type	Parameters				
	$umbr$	$T$ (ms)	$\beta = 1/mT$	$K_r$	$maxcount$ (ms)
A $\beta$ -Fibre	0.011 <sup>(1)</sup>	2 <sup>(3)</sup>	8.3	2.4	32 <sup>(3)</sup>
C-Fibre	0.122 <sup>(1)</sup>	2 <sup>(3)</sup>	79.34 <sup>(4)</sup>	9.1	32 <sup>(3)</sup>
Excitatory Interneuron	0.067 <sup>(2)</sup>	2 <sup>(3)</sup>	11.25 <sup>(5)</sup>	3.1	32 <sup>(3)</sup>
Inhibitory Interneuron	0.067 <sup>(2)</sup>	2 <sup>(3)</sup>	11.25 <sup>(5)</sup>	3.1	32 <sup>(3)</sup>
Projection Neuron	0.067 <sup>(2)</sup>	2 <sup>(3)</sup>	11.25 <sup>(5)</sup>	3.1	32 <sup>(3)</sup>

<sup>(1)</sup>: Normalized value from Cain *et al.* (2001) (0-200 mN). <sup>(2)</sup>: Normalized value from Prescott and De Koninck (2002) (0-300 pA). <sup>(3)</sup>: Based on the results obtained by Argüello *et al.* (2012). <sup>(4)</sup>: Based on the results obtained by Slugg *et al.* (2000). <sup>(5)</sup>: Based on the results obtained by Prescott and De Koninck (2002).

of the excitatory inhibitory ratio at the dorsal horn are unavailable, so that we had to carry out several simulations in order to empirically find the value of  $W_i$ , assuming  $W_e = 1$ .

Once built the GCS, we simulated the tactile somatosensory process by simultaneously stimulating both primary afferent neurons. The response of the projection neuron to progressive increases of the input signal (5, 10, 20, 40, 60, 80, 110, 140, 170 and 200 mN) was plotted over a 10 seconds window, as well as the A $\beta$ - and C-fibre responses. Using  $W_e = 1$ , we varied  $W_i$  to observe any change in the response of the projection neuron. Two additional simulations were performed to compare the activity developed by the projection neuron under two different conditions of stimulation: an increasing nociceptor input (same values showed above) with no stimulation on the A $\beta$ -fibre, and a constant C-fibre input (80 mN) with variable A $\beta$ -fibre input (5, 10, 20, 30, 40, 60, 80, 100, 120 and 140 mN). The first simulation was repeated for each A $\beta$ -fibre model, whereas the others were performed using only the average response model.

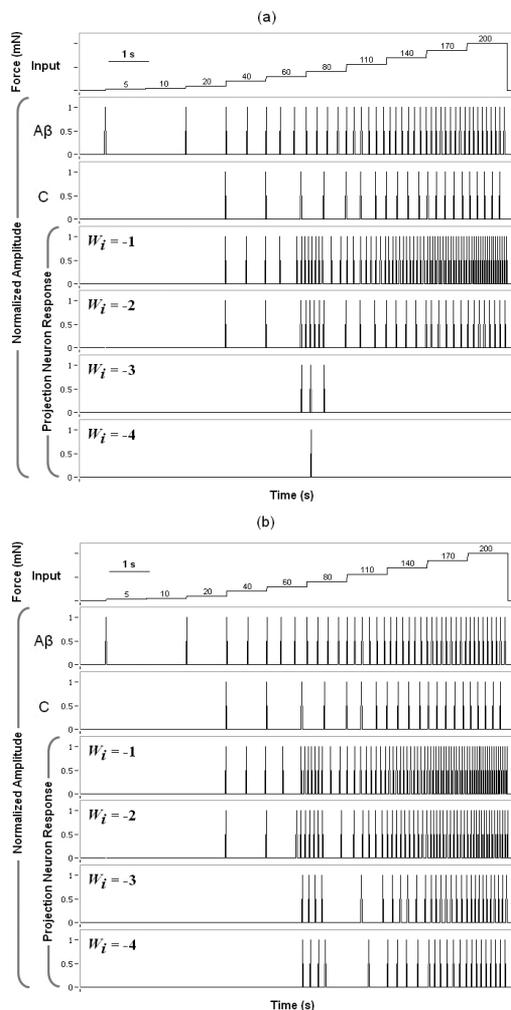
## Results

### *The F-I Relation of the A $\beta$ -fibres could modulate pain perception*

As illustrated in Figure 3a, for  $W_i \leq -3$ , the projection neuron showed no significant response to progressive increases of the input signal from 0 to 200 mN, when the A $\beta$ -fibre model that preserved the inverse correlation between threshold and slope was used in the GCS. In contrast, when we used the average response model, the projection neuron kept on firing for mechanical intensities  $\geq 80$  mN (see Figure 3b).

### *Nociceptors activation and the modulation of nociceptive information*

Using the average response A $\beta$ -fibre model and  $W_i = -3$ , with no stimulation on the A $\beta$ -fibre, the C-fibre



**Figure 3.** Response of the projection neuron to progressive increases of the input signal using the paradoxical model of A $\beta$ -fibre (a), and the average response model (b).

responded to mechanical stimuli  $\geq 40$  mN, whereas the projection neuron started firing in a non-linear fashion at 60 mN or greater, as shown in Figure 4. On the other hand, for constant C-fibre input, we

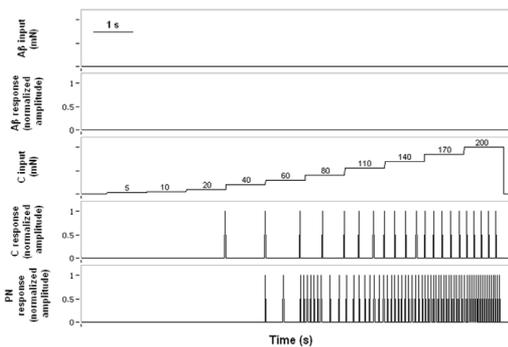
progressively increased the stimulation intensity on the A $\beta$ -fibre, such that we could modulate the response of the projection neuron. Figure 5 shows a slight increase in the firing frequency of the projection neuron, for mechanical intensities between 5 and 10 mN on the A $\beta$ -fibre. When the stimulus amplitude was increased above 40 mN, the firing frequency of the projection neuron decreased substantially. Finally, for stimuli greater than 100 mN, there was no response in the projection neuron.

## Discussion

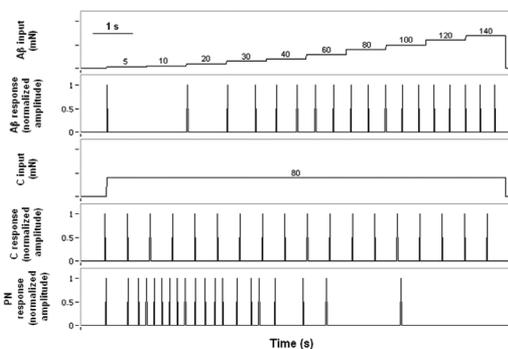
As established by Melzack and Wall's Gate Control Theory, pain perception is modulated by a gate mechanism found in the spinal cord, whose activation is conditioned by the activity of tactile afferents, at a peripheral level, as well as by cognitive and emotional processes that occur in the structures of the central

nervous system (Melzack and Wall, 1965). The descending control pathway was omitted because we wanted to emphasize the peripheral control mechanism. The results from the first two simulations suggest that the average response of the A $\beta$ -fibres across the receptive field does not increase monotonically at high amplitudes of stimulation, but it reaches a plateau, as assumed for the construction of the second model of A $\beta$ -fibre. Otherwise, it would be impossible to perceive pain at all, even in response to stimuli capable of inflicting real damage. But, how could we justify this hypothesis to the paradoxical relationship between the activation threshold and the slope of the F-I curve? First, by definition, the mechanoreceptors respond to low-intensity mechanical stimuli, but also respond to high intensity stimuli as do the nociceptors; the difference is that only the latter encode noxious stimuli (Devor, 2009). Physiologically, there would be no reason why mechanoreceptors had to encode high intensity stimuli, since that is the purpose of nociceptors; therefore, the F-I slope curve of the A $\beta$ -fibre should be flatter for high intensity stimuli. Second, the refractory period (or the amount of time it takes for a neuron to fire a second potential action under stimulation) may limit the firing frequency to a maximum value, no matter how above the threshold is the stimulus amplitude. This maximum firing frequency may be different between mechanoreceptors and nociceptors since they express different families of sodium channels, which influences the temporal profiles of the action potentials evoked by each type of fiber (Fang *et al.*, 2005). Third, the recruitment of rapidly adapting mechanoreceptors as a consequence of the increase in the stimulation intensity, could approximate the average response across the receptive field to the one proposed in this paper, since the responses of these mechanoreceptors may statistically cancel the responses exhibited by slowly adapting mechanoreceptors.

From the third simulation, we observed that the activation of nociceptors does not necessarily imply an activation of projection neuron, and therefore, the onset of pain sensation. This is consistent with experimental evidence that demonstrates the fundamental difference between the threshold of pain perception and the activation threshold of nociceptive afferents (Van Hees and Gybels, 1981). It is worth noting that the projection neuron start firing at 60 mN or greater, when no stimulus was applied on the A $\beta$ -fibre (average response model), unlike what was observed in the previous simulation, when both primary afferents were stimulated simultaneously (the neuron projection intensities responded to 80 mN or greater), suggesting the existence of a range of intensities within which the mechanoreceptors could modulate the information



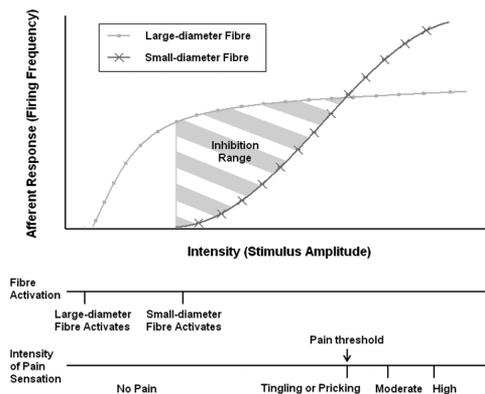
**Figure 4.** With no stimulation on the A $\beta$ -fibre, the C-fibre responded to intensities greater or equal than 40 mN, but the projection neuron started firing at 60 mN or greater, suggesting that nociceptors activation does not necessarily lead to pain experience.



**Figure 5.** The response of the projection neuron was modulated by the progressive increases of the A $\beta$ -fibre. For intensities between 5 and 10 mN, there was a slight increase in the firing frequency of the projection neuron, but when the intensity increased above 40 mN, the firing frequency decreased substantially. This suggests that the activation of A $\beta$ -fibres may both, increase and decrease the response of the projection neuron.

conveyed by the nociceptors. As originally stated in the GCT, the large-diameter thickly-myelinated fibres exert excitatory as well as inhibitory influence on spinal neurons transmitting nociceptive information, whereas the influence exerted by the small-diameter thinly-myelinated/unmyelinated fibres is exclusively excitatory, as observed after carrying out the fourth simulation. As shown in Figure 5, the activation of the fibres that encode stimuli in a range of non-noxious intensities could not only reduce the activity of the projection neuron, but also increase it for a small range of intensities (5-10 mN). This is an aspect that has not been directly described in previous implementations of biophysically accurate neuron models for the study of pain (Agi *et al.*, 2010; Britton *et al.*, 1996; Xu *et al.*, 2008), which only emphasize the inhibitory action exerted by A $\beta$ -fibres. Another aspect worth mentioning is the fact that, even though the F-I curve of the Neuroid is described as a linear relationship (see (2) in Methods), the response of the projection neuron stimulation to progressive increases was not linear. These nonlinearities are a hallmark of neural networks and become more evident through the implementation of much simpler neural models, like the Neuroid. In fact, it has been showed that the collective properties of neural networks often prevail over the individual properties of single units, regardless of the level of description that was used to build the main building block (Hopfield, 1882, 1984). This would allow us to focus not only on the most relevant computational aspects of nerve cells, but also on the collective capabilities of “pain networks” as processing units, composed of multiple subunits.

Through the implementation of the Neuroid, we formulated several hypotheses about how the F-I exhibited by A $\beta$ - and C-fibres influences the processing of the noxious information at the dorsal horn. These assumptions were made based on the absence of studies that make possible to compare the F-I curves of mechanoreceptors and nociceptors, under the same mechanical stimulation protocol. Results from simulation confirmed the original postulates of the GCT, but taking into account the functional differences between the two types of primary afferent fibres. All of this led us to propose a possible explanation for pain processing at the dorsal horn. As shown in Figure 6, for low intensity stimuli, the small-diameter fibres remain inactive and no pain sensation will be perceived. When the intensity increases enough to activate the nociceptors, the large-diameter fibres start reaching their maximum firing frequency, but the balance in activity between small- and large-diameter afferents is still in favor of a non-painful response. Only when the activity of nociceptors exceeds the inhibitory action exerted by the non-noxious fibres,



**Figure 6.** For an increasing stimulation, unpleasant sensations are experienced only when the activity of noxious afferents exceeds the one that still exert the non-noxious afferents (the shadowed area represents the range of intensities within the A $\beta$ -fibres are able to modulate the information conveyed by the C-fibres) which eventually reaches its maximum value.

the pain is consciously experienced. Since we cannot actually stimulate large- and small-diameter afferents separately, if the sensory energy increases enough to cause tissue damage, then the nociceptor’s activity should rise quickly above the one that still exert the non-noxious fibres, which in turn loses significance once the stimulus becomes harmful. Thus, we could associate the alterations in physiological pain threshold to the variations in the F-I curves belonging to the neurons involved in pain pathways. It has been reported that the F-I curves of mechanical C-fibres experience significant changes in both, inflammatory and neuropathic states (Ahlgren *et al.*, 1997). Therefore, by using a standardized protocol of stimulation, it would be possible to detect some pathologies by exploring the variability of these thresholds, such that this could lead to novel approaches for treatment of these painful conditions.

We used the Neuroid for replicating the Gate Control System of Pain. From its implementation, which emphasized the different F-I curves exhibited by A $\beta$ - and C-fibres, we believe that is possible to achieve a greater understanding of how sensory information is encoded, including those associated with pain perception. This not only supports the utilization of modeling for the study of pain, but also shows that future implementations can generate new findings leading to a hypothesis that can be tested experimentally.

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