

Characterization of an Algorithm for Autonomous, Closed-Loop Neuromodulation During Motor Rehabilitation

Neurorehabilitation and
Neural Repair
2024, Vol. 38(7) 493–505
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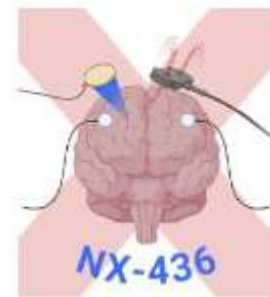
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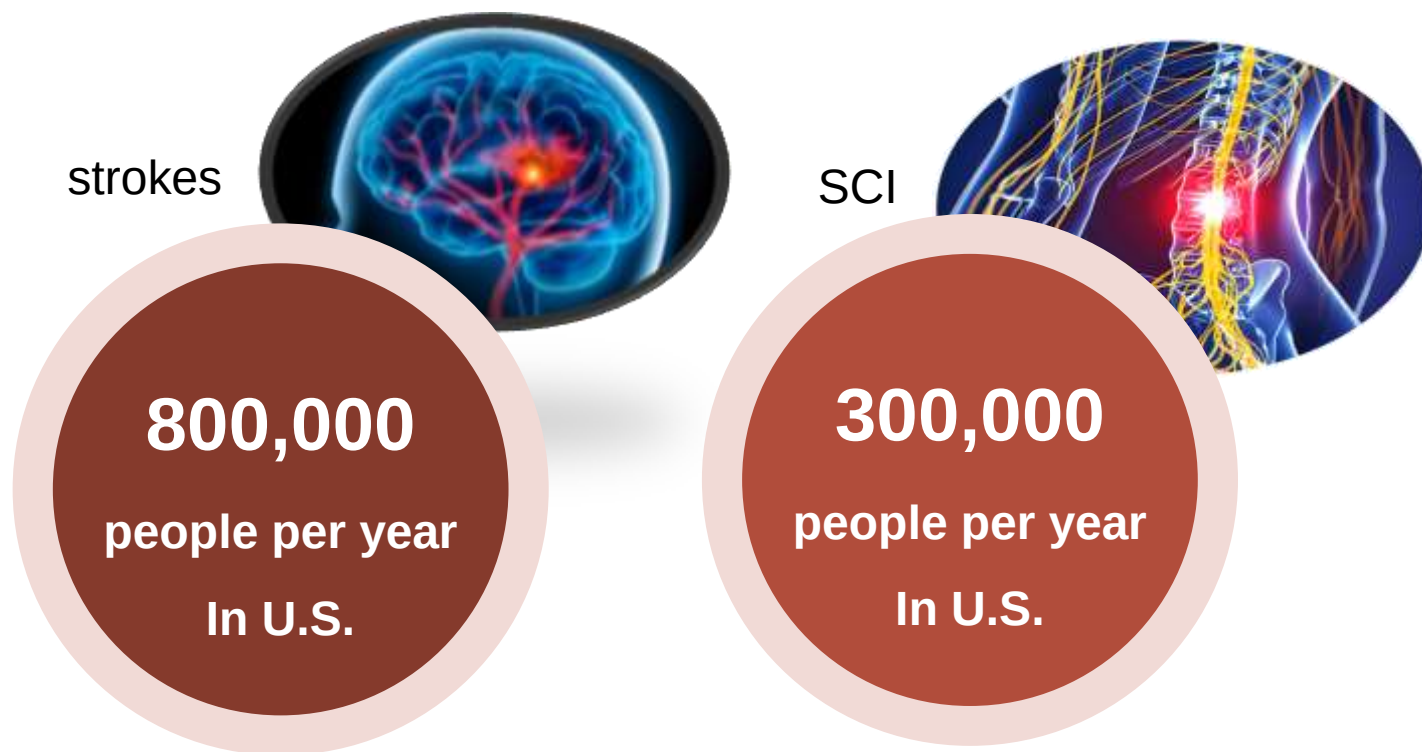
September 13th, 2024

Presenter: Leo Ganser, Penghui Du, Wenxin Che



Background - Clinical Problems

- **Stroke** and **spinal cord injury (SCI)** are leading causes of **upper limb motor impairments**, significantly affecting patients' quality of life.

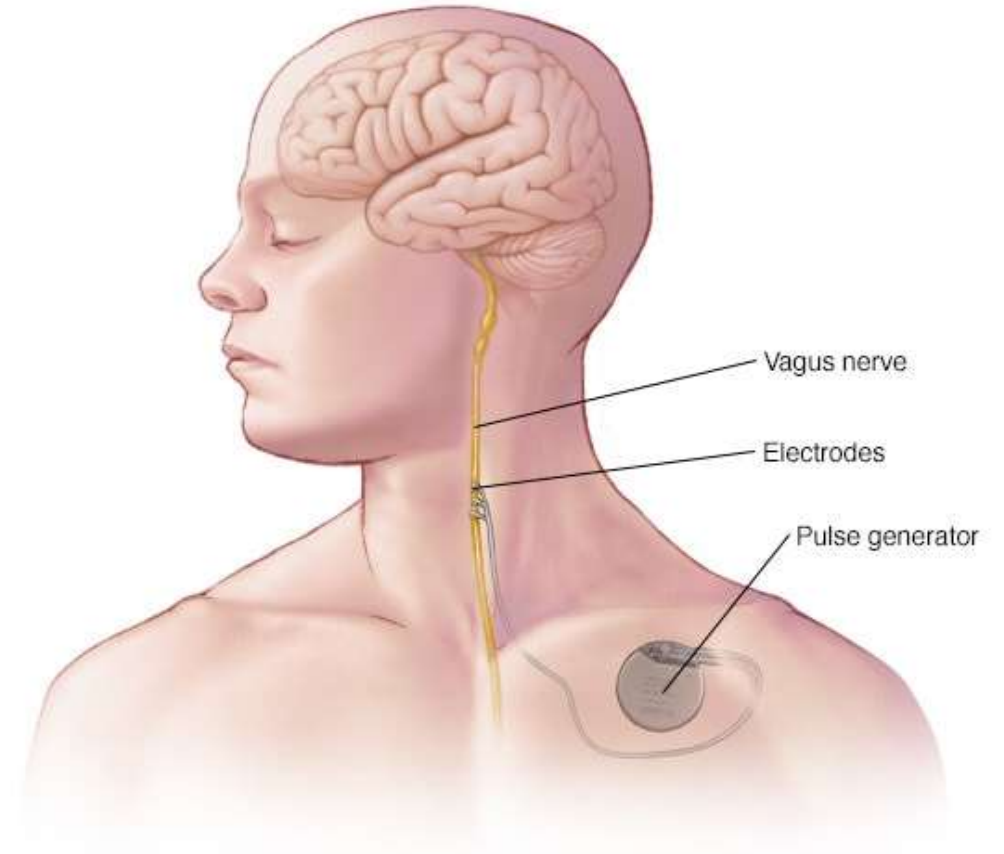


- There are approximately 800 000 strokes each year and over 300 000 people live with the effects of spinal cord injury (SCI)^{1,2}.
- Many survivors are left with **long-term upper limb hemiparesis**, which can lead to disability³.



Background - Vagus Nerve Stimulation in Rehabilitation

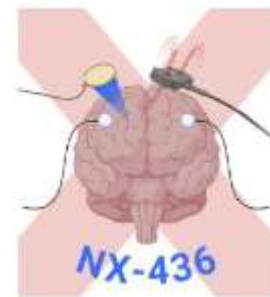
- Vagus Nerve Stimulation (VNS), when combined with rehabilitation exercises, has emerged as a promising approach for treating upper extremity motor deficits following stroke or SCI^{4,5,6}.
- **Mechanism:** VNS induces a rapid release of neuromodulators, enhancing synaptic plasticity within active motor networks, which is crucial for motor recovery.





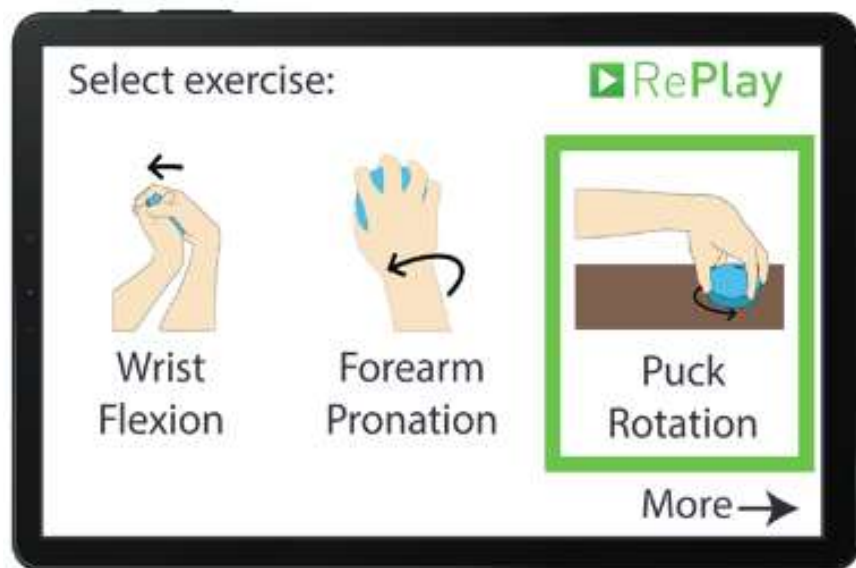
Background - Current Practice & Limitations

- 1 Currently, VNS is delivered through an implanted device, and a therapist **manually** triggers the stimulation during exercises by pressing a button.
- 2 **Timing** is crucial: Studies in animal models and human trials have demonstrated that **immediate** VNS yields significantly better recovery, compared to delayed VNS⁷.
- 3 **Stimulating during the best movements** is crucial: VNS delivered during rehabilitation fail to be effective if not concurrent with the “**best**” movements⁸.

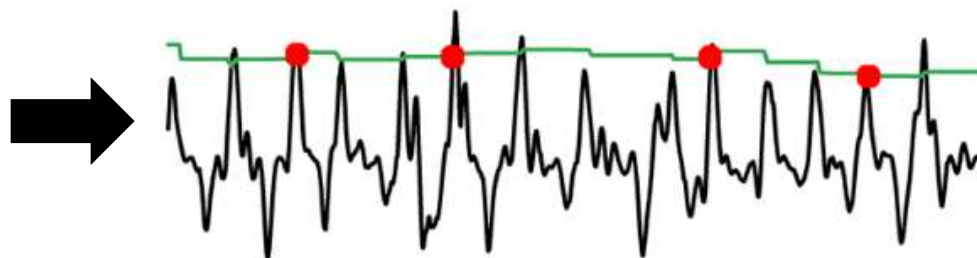


Background - Overview of Proposed Solution

- They developed three algorithms to **automate** VNS delivery, replicating the therapist's role in timing the stimulation, and analyzed which approach was the most effective.



Signal recording during rehabilitation movements



Signal processing and algorithm analysis



Trigger VNS at proper timing



Methods – Subjects and Devices



32 Patients

14 with stroke
18 with cervical spinal
cord injury (SCI)

Device	Game	Exercises
 FIMI	 Rep It Out	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation
	 Breakout	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation
	 Traffic Light	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation
	 Speed Runner	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation
 TAPDC	 Fruit Ninja	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation
	 Typing	Reach: Front, Back, Diagonal, Up, Down, Left, Right, Shoulder: Flexion, Extension, Abduction, Adduction, Internal Rotation, External Rotation, Wrist: Flexion, Extension, Deviation, Supination, Pronation, Thumb: Opposition, Adduction, Abduction, Extension, Flexion, Extension, Rotation

Performing tasks on RePlay⁹

A game-based, unsupervised system for
rehabilitation



Vagus Nerve Stimulation using ReStore¹⁰

An implantable wireless nerve stimulator system with no
battery. Can be programmed wirelessly to deliver stimulation.

> “Most commercially available nerve stimulators include a
battery and wire leads which often require subsequent
surgeries to address failures in these components”

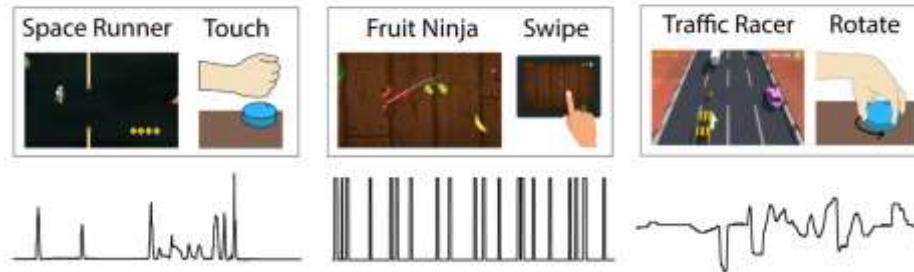


Methods – Rehabilitation Tasks

9 miniaturized sensors: (Record angle, force, and distance)

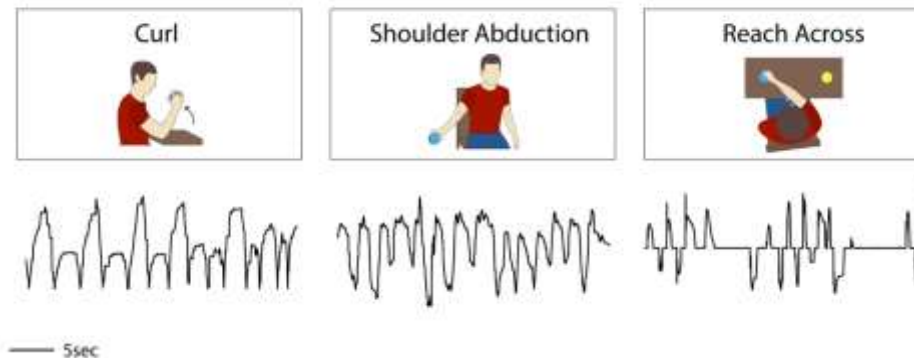
- FitMi handheld motion controller (Flint Rehab, California) – For repetition-based exercises
- ReCheck system – For playing games on an Android tablet¹¹.
- Tablet's touchscreen – For fruit ninja.

7 Tasks: **A** Game-based exercises



Game-based exercises.
Less boring and easy to
do at home

B Repetition-based exercises



Traditional, well
studied exercises



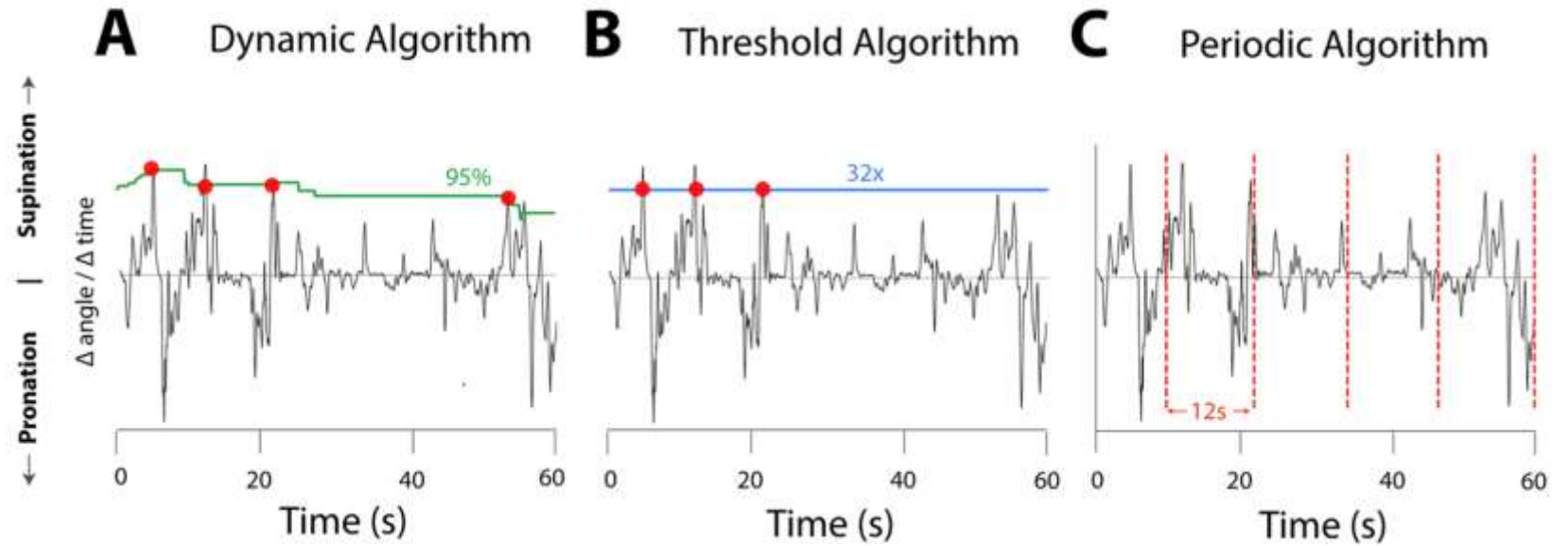
Methods – Picking Movements & Deliver Stimulus

Manual Stimulation

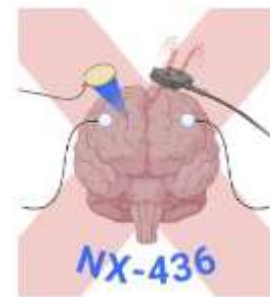


Delivered by
trained therapist

Automatic Stimulation



Minimum Inter-Stimulus-Interval (ISI): 5 seconds



Methods – Signal Preprocessing

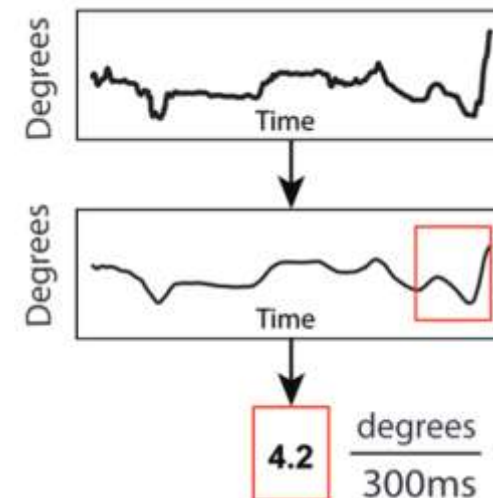
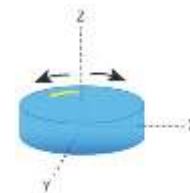
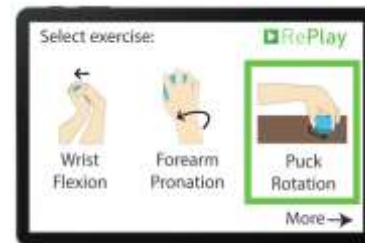
Movement selection and capture

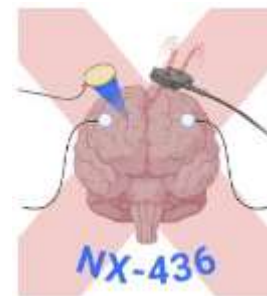
Smooth the recorded movement signals
(300ms kernel. Large movements ~300ms to complete)

Calculate the temporal gradients.

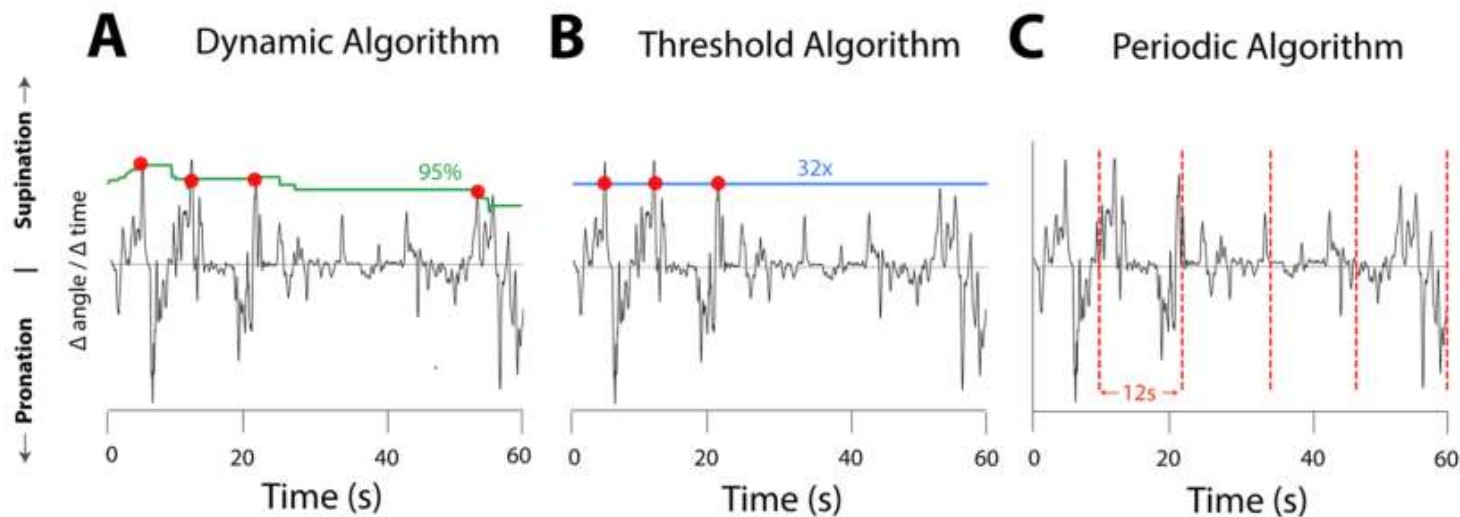
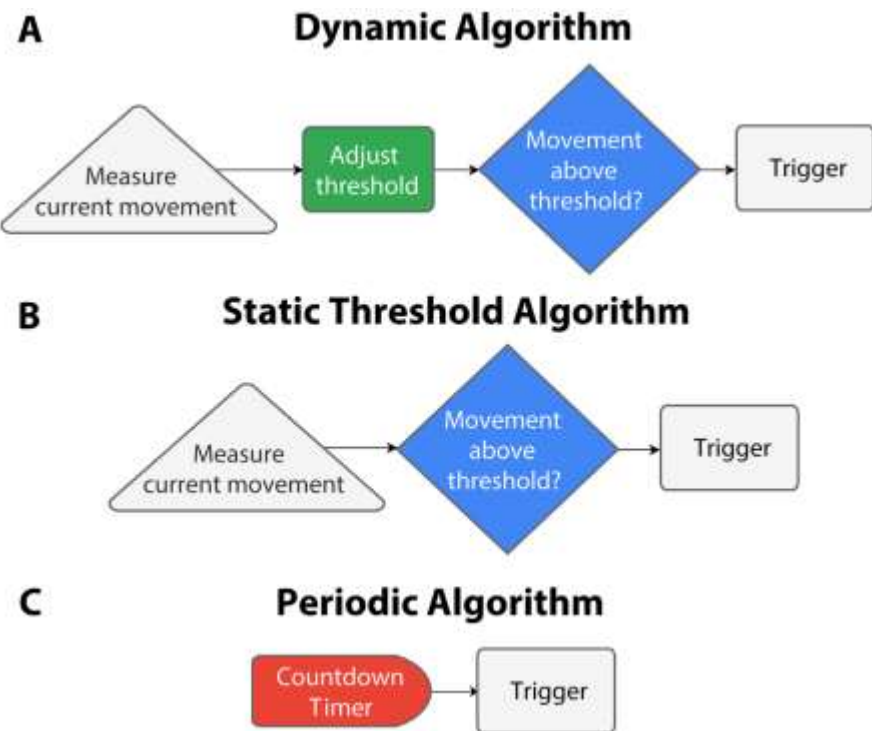
Calculate mean gradient over 300ms time window
=> A single value for average rate of change
(of angle / force / distance)

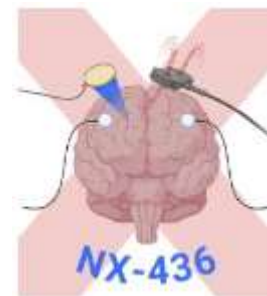
Exclude signals below a threshold to separate
movement from noise





Methods – Automatic Stimulation





Methods – Signal Preprocessing

Dynamic algorithm

Receive Movement
Magnitude Samples

4.2

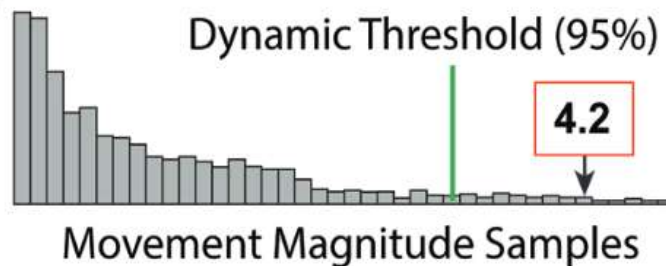
degrees
300ms

Receive movement samples, each in units of the specific exercise metric (degrees or grams) per last 300 milliseconds of movement



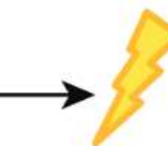
Movement Comparison

Number of Samples



Keep a distribution (movement buffer) of up to 3000 of the most recent movement samples

Triggering Decision

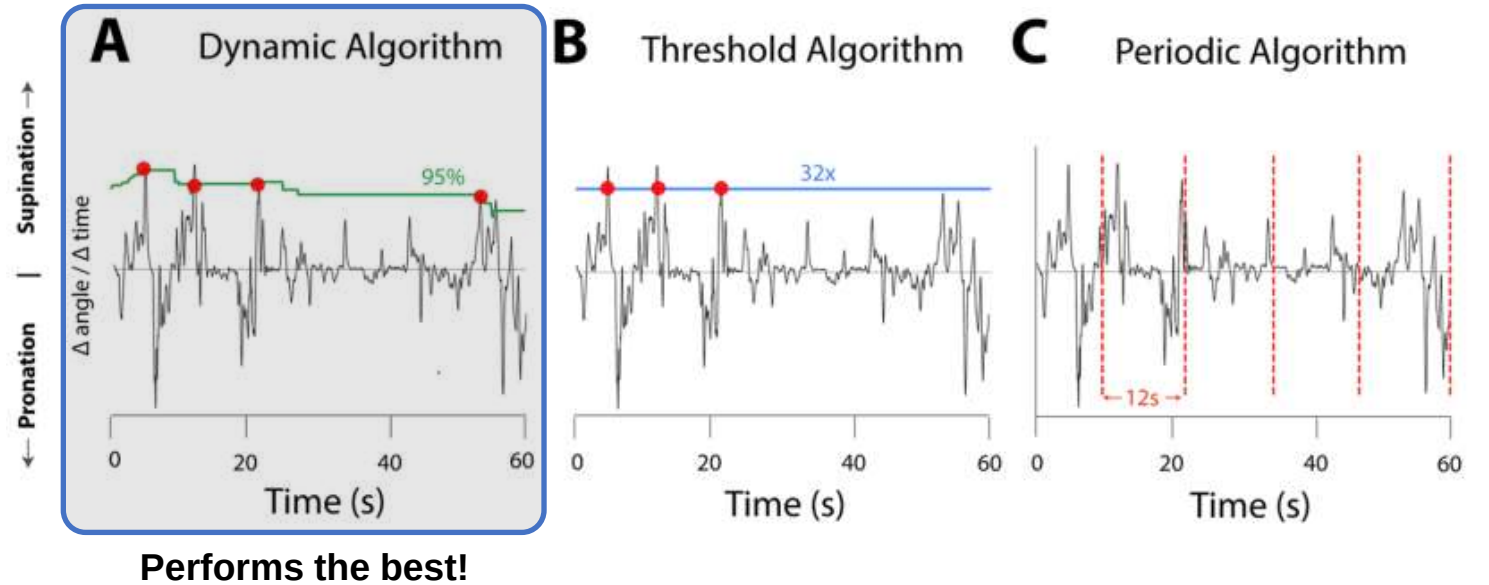
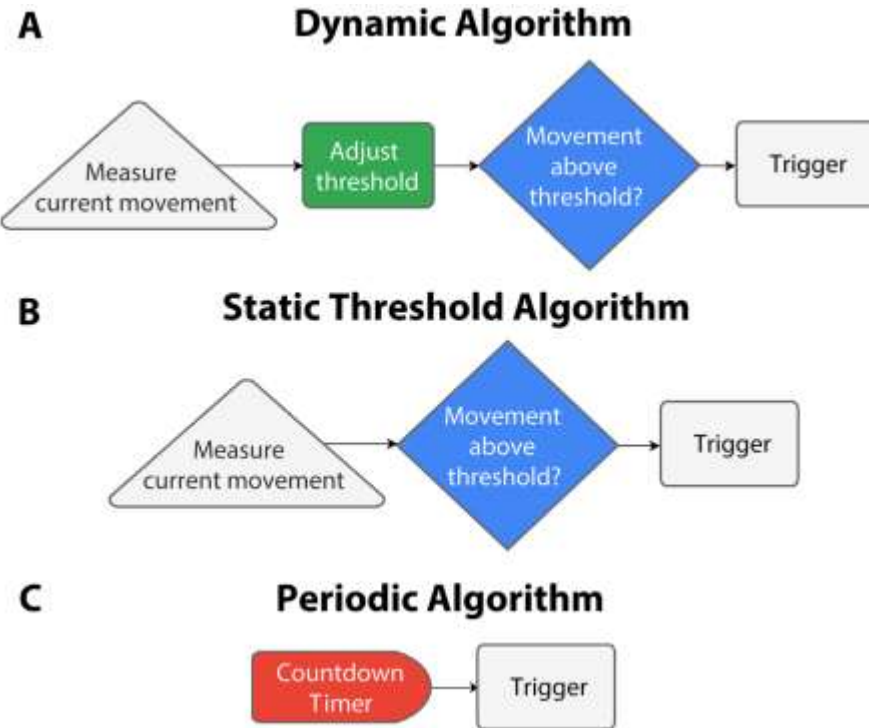


Trigger VNS

Trigger VNS if current movement is above dynamic threshold and minimum ISI has passed



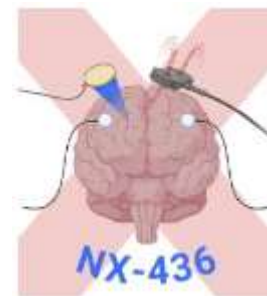
Methods – Automatic Stimulation





Results – Algorithm comparaison

Algorithms	Selectivity	Consistence over different exercices session	Complexity
Dynamic	97%	100% of sessions had VNS → Adjustable	high
Static	64%	29% → No VNS 43% → < 95% selectivity	medium
Periodic at 12 sec interval	34%	NA	very low



Results – Algorithm comparison

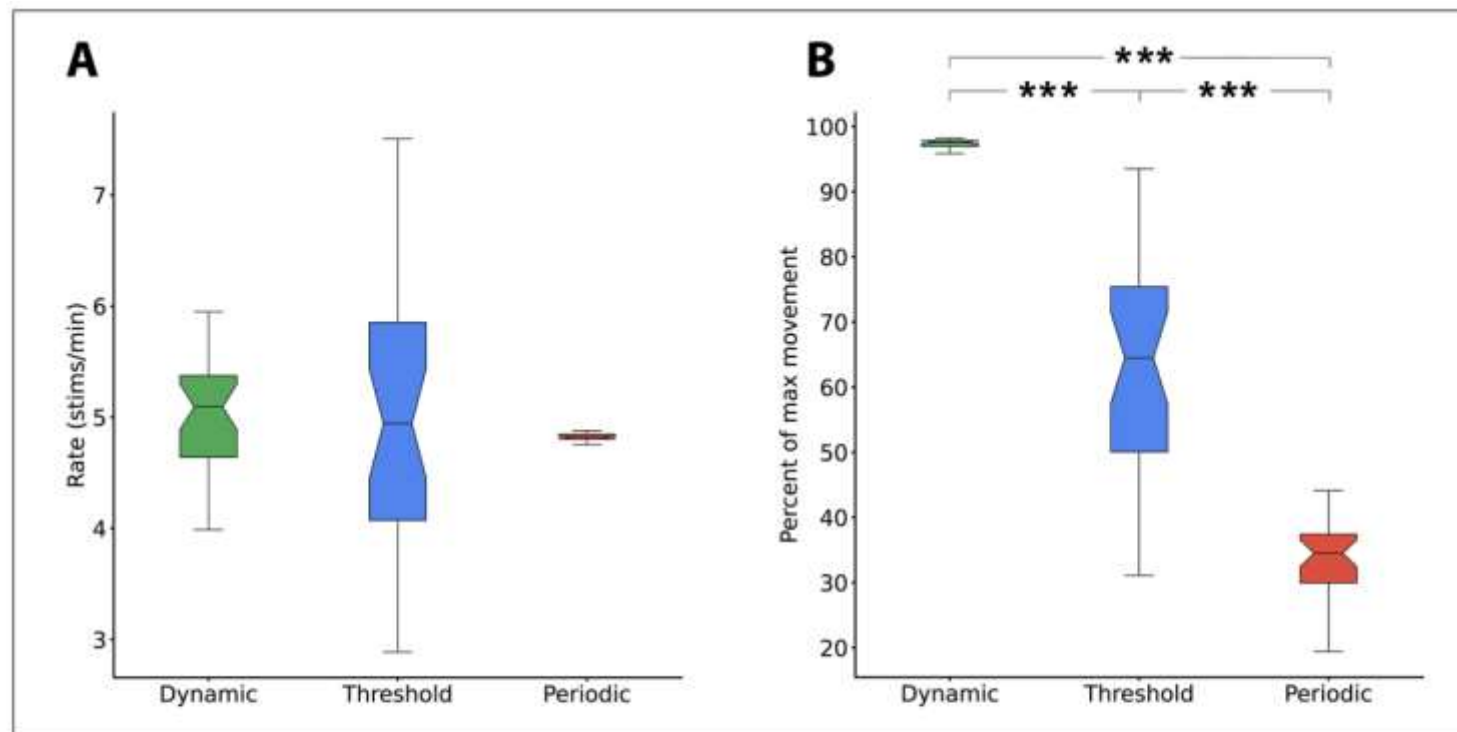
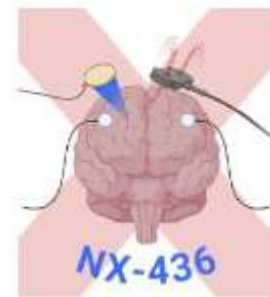


Figure 5. The dynamic algorithm yields the most robust selection of large movements and a reliable triggering interval. (A) The periodic algorithm produced consistent triggers every 12 seconds for a rate of 5 stims/minute. Variance of the periodic triggering rate seen here is from gameplay times not divisible by 12. The static threshold algorithm produced a median triggering rate of 4.94 (IQR = 1.78) stimulations per minute with considerable variance. The dynamic algorithm produced a median triggering rate of 5.09 (IQR = 0.74) stimulations per minute with moderate variance. No significant differences between median rates were observed. The notched boxes in each plot represent 1160 exercise sessions and include all controllers, games, and participants. Notches represent a 90% CI of the median. (B) The quality of the movements selected by the algorithms are represented as the percent of maximum movements during each exercise. The periodic algorithm triggered VNS on 34.50% (IQR = 7.47) of maximum movement. The threshold algorithm triggered VNS on 64.50% (IQR = 25.38) of maximum movement. The selective algorithm triggered VNS on 97.61% (IQR = 0.88) of maximum movement.



Results – Dynamic Algorithm VS Therapist

- The dynamic algorithm performs really well compared to the periodic algorithm and the therapist.
- Surprisingly, this result also shows that the therapist is not doing so much better than the periodic algorithm

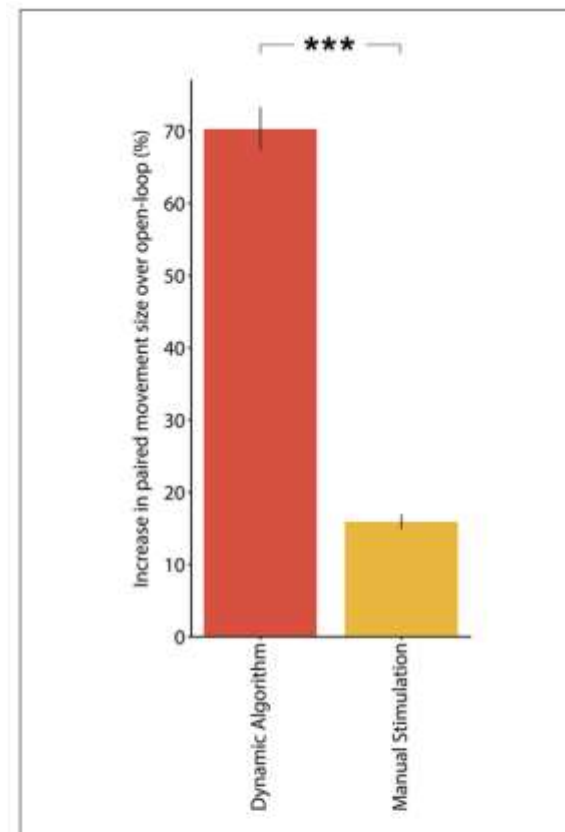
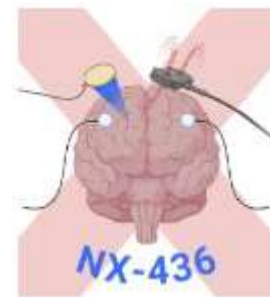


Figure 7. The dynamic algorithm selects larger movements than a trained observer. During upper limb physical therapy with RePlay and ReCheck, the dynamic algorithm triggered stimulation on movements that were $54.38 \pm 2.97\%$ larger than movements selected by a trained physical therapist (unpaired 2-tailed t-test, $P = 1.77 \times 10^{-74}$). We individually normalized the movement data by calculating the average paired peak size within ± 1 second of periodic stimulations at 12 second intervals throughout the therapy session. The dynamic algorithm and the periodic algorithm were applied in post-hoc analysis and the manual stimulations were conducted in real time.



Discussion

Limitations:

- Very simple algorithm
- A lot of data for exercise sessions but only 32 patients, 1160 exercises sessions
- Maybe some bias by not presenting an other relevant algorithm
- Sensor in single dimension → Not very complex movements



Open question

- ❖ How good are therapists evaluating the best movements?
Or how good are the sensors at evaluating the best movement
- ❖ To what extent, telerehabilitation will improve the recovery process ?
- ❖ Getting approval for this unsupervised device ?

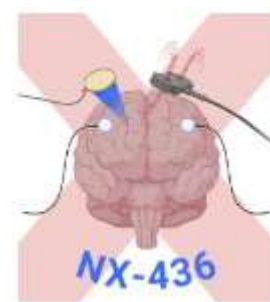


Take home message

What is the novelty ? **The ALGORITHM**

→ No more manual VNS

→ First step toward telerehabilitation



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