

1 **A 3D-printed passive exoskeleton for upper limb assistance in**
2 **children with motor disorders: proof of concept through an**
3 **electromyography-based assessment**

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45 **Abstract**

46 The rehabilitation of children with motor disorders is mainly focused on physical interventions.
47 Numerous studies have demonstrated the benefits of upper function using robotic exoskeletons.
48 However, there is still a gap between research and clinical practice, owing to the cost and
49 complexity of these devices. This study presents a proof of concept of a 3D-printed exoskeleton
50 for the upper limb, following a design that replicates the main characteristics of other effective
51 exoskeletons described in the literature. 3D-printing enables rapid prototyping, low cost, and
52 easy adjustment to the patient anthropometry. The 3D-printed exoskeleton, called POWERUP,
53 assists the user's movement by reducing the effect of gravity, thereby allowing them to perform
54 upper limb exercises. To validate the design, this study performed an electromyography-based
55 assessment of the assistive performance of POWERUP, focusing on the muscular response of
56 both the biceps and triceps during elbow flexion–extension movements in 11 healthy children.
57 The Muscle Activity Distribution (MAD) ~~is was the proposed~~ metric for the assessment. The
58 results show that (1) the exoskeleton correctly ~~assists-assisted~~ elbow flexion, and (2) the
59 proposed metric easily ~~identifies-identified~~ the exoskeleton configuration: statistically significant
60 differences ~~in the mean MAD value~~($p\text{-value} = 2.26 \cdot 10^{-7} < 0.001$) ~~and a large effect size~~
61 ~~(Cohen's $d = 3.78 > 0.8$) in the mean MAD value~~ were identified for both the biceps and triceps
62 when comparing the transparent mode (no assistance provided) with the assistive mode (~~anti-~~
63 ~~gravity effect~~). Therefore, this metric was proposed as a method for assessing the assistive

64 performance of exoskeletons. Further research is required to determine its usefulness for both the
65 evaluation of selective motor control (SMC) and the impact of robot-assisted therapies.

66

67 Introduction

68 Physical therapy in the rehabilitation of children with motor disabilities mainly focuses on motor
69 intervention. Bimanual training, constraint-induced movement therapy, fitness training, strength
70 training, and task-specific training, among others, have been shown to be effective in enhancing
71 baseline motor, sensory, and perceptual skills, and learning capabilities [1].

72 In this context, the robotic-assisted approaches aim at helping improve the motor control and
73 muscle strength of these patients [2]. An exoskeleton is an assistive and wearable technology that
74 helps people with motor disabilities or impairments to restore, improve, or at least maintain their
75 functional abilities. Although the idea of exoskeleton was dated back to the late 19th century, the
76 first prototype of a successful one (called Hardiman) was not developed until the 1960s and was
77 thought for military purposes [3]. Years later, Kazerooni *et al.* [4] developed (1990) an upper-
78 limb exoskeleton exploring the idea of physical human-robot interaction. Later (2003),
79 researchers from the University of Tsukuba presented the exoskeletal robotics suite HAL
80 (Hybrid Assistive Leg) originally developed to help disabled people in ADL (Activities of
81 Daily Living) [5]. In the last two decades, the use of upper-limb exoskeleton both for services
82 and rehabilitation has gained attention in the biomedical field as potential solutions for
83 physically weak or disabled people [6]. Devices such as InMotion® [73], Haptic Master® [84],
84 and Armeo Spring® [95 – 117] have been demonstrated to be effective complements to physical
85 therapy, especially for motor rehabilitation of the upper limbs. Moreover, in recent years, with
86 the rise in 3D design and printing, the development of low-cost exoskeletons inspired by these
87 commercial devices has increased remarkably [128 – 140].

88 POWERUP is a 3D-printing-based passive upper-limb exoskeleton [154] designed to assist
89 upper-limb movement in children with motor disabilities. The device could be used not only with
90 rehabilitation purposes, but also as a re-educational path in children with different temporary,
91 progressive, or permanent physical conditions implying postural and balance deficits as well as
92 difficulties in the movement and/or coordination of the upper limbs such as cerebral palsy (CP)
93 [16], juvenile arthritis [17], spina bifida [18] or muscular dystrophy [19], among others.

94 The design of POWERUP is partially inspired by the Wilmington Robotic Exoskeleton
95 (WREX) [2042 – 2244], which is a 4 degrees-of-freedom (DoF) mechanism with two rotations at
96 the shoulder and two rotations at the elbow that passively counterbalances the weight of the arm
97 using elastic bands [2345]. Moreover, POWERUP adds an extra DoF that allows
98 pronosupination of the elbow.

99 The assessment and validation of the assistive performance of the exoskeleton are key points
100 before evaluating it with patients in clinical trials. However, there is a lack of well-established
101 methods and metrics for this purpose in the clinical practice. Nevertheless, in the industrial field,
102 the evaluation criteria and metrics are clear and well-classified [2446].

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103 Therefore, in this area, measurement of muscle activity (intensity of muscle contraction) using
104 surface electromyography (sEMG) is the most common technique for evaluating exoskeleton
105 assistance. The greater the amount of assistance provided to a muscle, the lower the level of
106 muscle activation [2446 – 2648]. To quantify this effect, it is important to compare the sEMG
107 registers of users conducting different tasks with and without the assistance of the exoskeleton.
108 The root mean square (RMS) of the sEMG signal, calculated using a moving window from the
109 raw register, is considered to provide the most insight on the amplitude of the EMG since it gives
110 a measure of the power of the signal. In fact, the peak value of the RMS in dynamic tasks, as
111 well as its time-average value in static ones, are metrics commonly used to carry out the
112 aforementioned comparison. a series of sEMG-based parameters obtained while users are
113 conducting a certain task, with and without the assistance of the exoskeleton, can be compared.
114 The most commonly used metric is the root mean square (RMS) of its peak value (for dynamic
115 tasks) or its time-averaged value (for static tasks)-[2446, 2749, 2820].

116 Several studies have extrapolated this method of validating assistive performance to assess
117 robotic devices designed for use in the clinical field. Wang et al. [294] evaluated their assistive
118 system for upper limb motion by comparing muscle fatigue (by means of the muscle activation
119 levels) in three healthy subjects lifting and holding a 1 kg object, both with and without the
120 exoskeleton. Xiao et al. [3022, 23] assessed the assistance of their cable-driven exoskeleton
121 using the RMS values obtained from the sEMG recordings of a series of muscles associated with
122 upper-limb movement in six healthy subjects. Wu et al. [3124] validated the different degrees of
123 assistance of an admittance-based patient-active control upper-limb exoskeleton by comparing
124 the RMS signals of a series of sEMG recordings in three healthy volunteers.

125 On the other hand, it is important to highlight that the sEMG is a very well-known and
126 commonly used technique in the diagnosis and in monitoring the evolution of patients suffering
127 from different motor impairments. In fact, ~~On the other hand, and, in addition to its diagnostic~~
128 uses, sEMG has become an important technique for both analyzing the movements and assessing
129 the motor function impairment of children with motor disabilities, as it provides crucial
130 information regarding muscle coordination [3225 – 3528]. Owing to sEMG, information
131 regarding muscle activation, myoelectric manifestation of muscle fatigue, and recruitment of
132 motor units can be obtained [3629]. Therefore, it can be said that sEMG is a well-known and
133 commonly used technique with patients suffering from different motor disabilities.

134 Considering that the sEMG is not only recording technique is well-known in the clinical field
135 and that sEMG analysis is a common technique to evaluate the performance of exoskeletons but
136 also a very well-known technique for diagnosis and follow-up of patients in the clinical field, we
137 hypothesized that :

- 138 • the sEMG analysis-The EMG analysis will make it possible to discriminate between the
139 transparent mode (no effect of the exoskeleton) and the assistive mode (anti-gravity
140 effect) of the POWERUP exoskeleton.
- 141 • The differences between the two modes of performance of the device will lead us to
142 describe, thus corroborating that if the exoskeleton correctly assists elbow flexion

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143 according to the scientific evidence (as stated before, the assistance provided to a muscle
144 should result in lower levels of muscle activation).

- 145 • ~~These results will make it possible to, and postulate ing that~~ this method ~~as is a~~
146 ~~particularly~~ suitable candidate for the validation of this type of robotic device.

147 Therefore, this study aimed to validate the assistance effect of a 3D-printed passive exoskeleton
148 for elbow flexion by performing an electromyography-based assessment in healthy children that
149 makes it possible to

150 easily distinguish between the transparent and the assistive mode of the device.

151 This study ~~foeuses~~ focused on elbow flexion–extension movements in healthy children, thus
152 laying the foundations and establishing normative reference values for subsequent validation in
153 children with motor disabilities.

154 **Materials & Methods**

155 **Upper limb exoskeleton**

156 The POWERUP exoskeleton (Figure 1) is a 5 DoF upper limb orthosis without
157 electromechanical actuators and is fabricated using 3D-printing technology. During the design
158 process, a series of expert pediatric physiotherapists from the Instituto de Rehabilitación
159 Funcional La Salle (Madrid, Spain) helped establish the clinical and functional criteria that the
160 device should meet. The present wearable prototype was designed for children aged between 6
161 and 15 years (arm and forearm lengths between 40 and 60 cm) [37@]. Its function is to allow
162 elbow and shoulder movements by providing stability as well as anti-gravity weight support
163 when needed, by keeping the wrist and hand in the neutral position.

164 POWERUP comprises four structural modules (shoulder/back, arm, forearm, and hand). All of
165 ~~them~~ the modules allow the physiotherapist to easily adjust (width, height, and length) the device
166 to adapt to the user's anthropometry using telescopic bars, anchors, and elastic straps. In addition,
167 this modular design allows for a quick and intuitive assembly. The exoskeleton allows flexion–
168 extension and internal–external rotation of the shoulder, flexion–extension, internal–external
169 rotation, and pronosupination of the elbow. The hand module has a horizontal surface on which
170 the hand rests and the wrist is fastened with straps while resting in a neutral position in the
171 coronal plane. This surface rotates within a concentric sliding ring mechanism, allowing
172 pronation and supination of the elbow.

173 The device was manufactured by 3D-printing (Crealty CR-5 Pro® and BQ Witbox 2® printers)
174 with a PLA+ 1.75 mm filament (a thermoplastic monomer derived from renewable and organic
175 sources), which makes it a lightweight, low-cost, and easily reproducible solution. All the pieces
176 were printed with a 0.2 mm layer height, while the rest of the printing parameters were adjusted
177 for each piece. Thus, bar-type parts with holes were printed with the CR-5 printer, with
178 minimum support on the holes, with a 20% fill density, and adding extra fixation to the hot bed
179 owing to the use of spray adhesive and a raft, as some pieces require more than 24 h to be
180 printed. In this case, the parameters according to the printer test were as following: 207°C with a
181 60°C hot bed, 96% flux, and Bowden extruder with 6–30 mm/s retraction speed. Witbox 2 was
182

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183 used for more complex pieces with circular parts, with 60° supports at a minimum of 25% infill
184 density, and spray adhesive to increase fixation. On this occasion, the parameters according to
185 the printer test were as following: 204°C, 98% flux, and direct drive with 1–30 mm/s retraction
186 speed. The overall printing time for the entire device was approximately 170 h.

187 The POWERUP assistive mode provides anti-gravity weight support, helping the children lift
188 and maintain the weight of their arm and enhance elbow flexion movement. The assistance was
189 achieved by placing elastic bands around protruding lugs at the ends of the forearm segments, as
190 depicted in Figure 2. The use of conventional materials, 3D-printing and common elastic bands
191 makes it possible to replicate the exoskeleton easily.

192 The POWERUP exoskeleton was anchored to an external metal-rolling frame (Figure 3). This
193 structure can be easily moved and adjusted according to the user's shoulder height. Once placed
194 in a convenient position, the wheels were locked to prevent unintentional displacement during
195 training sessions.

196 **Surface electromyography**

197 The measurement of the assistance effect ~~will fall~~[will be assessed](#) on the [analysis-analyses](#) of
198 muscle activity (thanks to sEMG recordings) of the biceps brachii (agonist in elbow flexion and
199 antagonist in elbow extension) and triceps brachii (agonist in elbow extension and antagonist in
200 elbow flexion).

201 The root mean square (RMS) signal is proportional to the number of muscle fibers activated in a
202 particular muscle (recruitment) [381, 392]. On the other hand, the Muscle Activity Distribution
203 (MAD), which is the proposed metric for the POWERUP assessment, is defined as the
204 percentage of muscle activity of a specific muscle over the total amount of muscle activity
205 recorded for a particular movement. This parameter was chosen because (1) it makes it possible
206 to compare the contractile activity between agonist (or synergist) and antagonist muscles and/or
207 to study muscle overactivation/inhibition [4033]; and (2) it can be explored as an alternative for
208 comparison among different subjects instead of the metrics obtained from the normalized RMS,
209 which usually requires the attainment of maximum voluntary contraction (MVC), which can be
210 relatively easy to obtain in healthy children but not in children with motor disabilities for
211 obvious reasons.

212 The sEMG signals and proposed metrics were obtained using an ultralight wearable device
213 (mDurance™) that integrates a three-dimensional inertial sensor with a two-channel
214 electromyograph. The device is controlled using an Android™ mobile device via Bluetooth. This
215 application enabled the acquisition of two simultaneous sEMG signals with a sampling rate of
216 1024 Hz. Each sEMG record was stored in a device cloud service. The recordings were reloaded
217 (and/or downloaded) for subsequent analysis. A cloud service makes it possible to obtain and
218 visualize the time-dependent evolution of the recruitment of muscle fibers in terms of RMS,
219 enabling the observation of the co-activation patterns for the recorded muscles. The MAD can
220 also be visualized using the corresponding bar plots (Figure 4).

221 For obtaining the RMS signal, first, a fourth order Butterworth bandpass filter with a cut-off
222 frequency at 20–450 Hz is applied and, second, the resulting signal is smoothed using a window

223 size of 0.25 s. For a specific muscle, the MAD (%) is calculated as the ratio between the result of
224 the time integral of the RMS signal for this muscle over the sum of all the results of the time
225 integrals for each one of the muscles measured in this record [3441].

226 The sEMG recordings were made with solid gel 45 × 42 mm general-purpose disposable
227 Ag/AgCl electrodes that were placed according to the recommendations for sensor locations in
228 arm muscles from the [Surface Electromyography for the Non-Invasive Assessment of Muscles](#)
229 [\(SENIAM\) SENIAM project](#)¹ ([Surface Electromyography for the Non-Invasive Assessment of](#)
230 [Muscles \(SENIAM\) project](#), on the line between the acromion and the fossa cubit [at 1/3 from the](#)
231 [fossa cubit](#) to record biceps brachii activity, and at 50% on the line between the posterior crista
232 of the acromion and the olecranon at two finger widths medial to the line for recording triceps
233 brachii activity (Figure 5) [42].

234 Participants

235 A total of 11 typically developed children (8 males and 3 females, aged 9–10 years old, with
236 average ± standard deviation height, weight, and [Body-body Mass-mass Index-index](#) (BMI) of
237 32.0 ± 4.2 kg/m², 138.9 ± 4.6 cm, and 16.5 ± 1.6, respectively), who were randomly selected
238 from a class of students from the Colegio CEU San Pablo Montepíncipe in Madrid (Spain),
239 [participated in the study](#).

240 [The expected effect size, according to preliminary tests should be large. With a Cohen's d = 1,](#)
241 [the Power Analysis \('pwr' library, R statistical computing\) resulted in a sample size of 9.9 so](#)
242 [that, a sample of 10-11 subjects should be enough to corroborate the hypothesis of this proof of](#)
243 [concept](#).

244 All children, their parents or legal guardians, and the head office of Colegio San Pablo CEU
245 Montepíncipe provided [written informed](#) consent to participate in the study. Ethical approval
246 was obtained from the Research Ethics Committee of the San Pablo CEU University
247 (561/21/53).

248 Data acquisition protocol

249 The participants were first informed about the experiment and received clear explanations of the
250 protocol established for the movements they should perform. Each participant sat next to the
251 exoskeleton mounted on the platform. All participants were asked to sit comfortably with their
252 backs straight, trying not to change their trunk position during the experiment. Subsequently, a
253 physiotherapist placed both the sEMG electrodes and the exoskeleton, properly aligned and
254 fixed, on the dominant arm of the child.

255 Once the wearable electromyography device was properly configured and ready to record the
256 sEMG signals, each child was asked to perform two tests. The first test (Test 1) consisted of
257 recording the sEMG signal of elbow flexion–extension movements when the children wore the
258 exoskeleton, but no elastic bands were added (transparent mode). Test 1 was performed to
259 establish a reference [to be compared with the against which to compare](#) exoskeleton assistance
260 settings. Test 2 consisted of measuring the sEMG signal of elbow flexion–extension movements
261 with the children wearing the exoskeleton configured in its assistive mode. The elastic bands

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¹ http://seniam.org/arm_location.htm

262 were attached to assist elbow flexion (Figure 6), thus resisting its extension. No bands were used
263 on the upper arm segment of the exoskeleton in either of the tests to work only with pure elbow
264 flexion–extension movements, keeping the upper arm stable in its position.

265 Each of the two tests (Test 1 and Test 2) started with an sEMG recording of 5 s while the subject
266 remained relaxed and without performing any voluntary movements and/or contractions to check
267 whether the arm was really at rest (no observable increase in basal tone). After the first 5 s, and
268 considering that the fiber recruitment depends on the contraction strength and speed of the
269 exercise, the time evolution of each test was controlled as follows: the subject ~~had~~was to flex the
270 elbow fully (trying to touch the shoulder) and maintain this position for 3 s. Then, the elbow was
271 fully extended and maintained for 3 s. This holding flexion—extension movement was repeated
272 three times for each test. This protocol aims to (1) minimize the muscle fatigue that occurs when
273 an intense effort is maintained over time and (2) favor the ordered recruitment of the motor units
274 (from the smallest to the largest) that generally occurs in controlled movements that do not
275 require intense contraction strength and/or fast performance velocities [3543].

276 Data analysis

277 The analysis focused on (1) corroborating the initial ~~hypothesis~~hypotheses, demonstrating, that
278 ~~is,~~ that the sEMG analysis based on the, and more specifically, the analysis of the proposed
279 metric (~~the~~ MAD), discriminates between the transparent and assistive modes of the POWER
280 UP exoskeleton, and (2) studying if the extent to which the differences in the MAD, comparing
281 Test 1 and Test 2 for both biceps and triceps, are statistically significant. The analysis was
282 conducted using RStudio, an integrated development environment (IDE) for R, which is a
283 programming language for statistical computing and graphics. To determine whether the samples
284 met the assumptions to conduct the corresponding paired t-test analyses ($p < 0.05$), the presence
285 of significant outliers was checked by visualizing the data using boxplots, the normality of the
286 variables was checked using the Shapiro–Wilk test and by visual inspection using Q-Q plots with
287 0.95 confidence intervals. Cohen’s d was used to determine the effect size. Graphical
288 visualization of the results was performed using the corresponding bar plots.

289

290 Results

291 The data analysis previously described and conducted to compare the results of the two tests
292 (Test 1 and Test 2, transparent and assistive modes, respectively) yielded the results described
293 below.

294 Figure 7 shows the boxplots of the MAD (%) for both biceps and triceps and for each of the
295 considered tests. The boxplots show that the MAD in the biceps brachii decreased during
296 exoskeleton-assisted elbow flexion (Test 2) in comparison with the transparent mode (Test 1).
297 Conversely, the MAD in the triceps brachii increased during exoskeleton-assisted elbow flexion
298 (Test 2) in comparison with the transparent mode (Test 1), confirming the initial hypothesis. In
299 addition, the antagonistic roles of the biceps brachii and triceps brachii in elbow flexion and
300 extension movements were clearly shown. No significant outliers are identified.

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301 The following statistical analysis studies on the extent to which these identified differences
302 between Tests 1 and 2 are statistically significant.
303 The Shapiro–Wilk tests yielded $p = 0.52 (> 0.05)$ and, in addition, the visual inspection of the
304 subsequent Q-Q plots made it possible to affirm that the points were located within the
305 previously stated 0.95 confidence intervals. Therefore, there was no evidence of a lack of
306 normality in the tested data groups.

307 Figure 8 shows the bar plots of the mean values \pm standard deviations of MAD (%) for both the
308 biceps brachii and triceps brachii in Tests 1 and 2, as well as the results from the corresponding
309 paired t-tests: $p = 2.26 \cdot 10^{-7} < 0.001$ [***], $df = 10$. Cohen’s $d = 3.78 (> 0.8$, large effect size).
310 Table 1 summarizes the results from the comparative analysis between Test 1 and Test 2.
311

312 Discussion

313 It is well known that robotic-based interventions involve multiple variables, which is why it is
314 very important to have reliable metrics not only to efficiently configure each rehabilitation
315 session, but also to assess the real impact of these kinds of treatments in processes such as skill
316 acquisition, generalization of motor skills to functional activities, and retention (persistence of
317 the acquired skills) [4436].

318 In this context, this paper describes the design and validation of a 3D-printed exoskeleton for
319 upper limb assistance. Our approach involves replicating the functionality of effective
320 exoskeletons previously described in the literature by using more accessible materials, 3D-
321 printing and elastic bands. To validate our device, we performed an electromyography-based
322 assessment of the assistive performance of the exoskeleton for elbow flexion in healthy children
323 as a proof of concept.

324 Our results provide observable evidence (Figure 7) of the change (according to the initial
325 hypothesis) that occurs in the MAD for the biceps and triceps when the exoskeleton assists
326 elbow flexion (Test 2) compared to the transparent mode (Test 1). In fact, the statistical analysis
327 affirms that there are statistically significant differences in the MAD when comparing Tests 1
328 and 2 for both biceps and triceps (Figure 8) as the assistance provided to the arm in Test 2 clearly
329 results in lower levels of muscle activation for the biceps. These results are in line with the ones
330 recently published by F.V. dos Anjos *et al.* [45], who studied the changes in the distribution of
331 muscle activity when using a passive trunk exoskeleton using high-density sEMG and conclude
332 that the assistive effect of the exoskeleton decreases the average RMS amplitude, implying a
333 decrement in the percentage of muscle activity of the low back muscles for both static and
334 dynamic tasks.

335 Therefore, based on these results, it can be stated that (1) the exoskeleton correctly assists
336 assisted elbow flexion, reducing the effect of gravity and, consequently, enabling more upper
337 limb exercises; and (2) the sEMG and, more specifically, the MAD metric makes-made it
338 possible to determine the configuration in which the exoskeleton is operating and distinguish
339 perfectly between the exoskeleton assisting elbow flexion and the exoskeleton in its transparent

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340 mode (without assistance). Therefore, the results lead to the conclusion that MAD can be
341 considered a reliable metric for the validation of the assistive performance of the exoskeleton.
342 In this context and as the differences between the two studied configurations are remarkable
343 according to our results, the MAD could be studied, in a subsequent analysis, as a metric to
344 quantify different levels of assistance (the multilevel quantification and classification could be
345 based on both the number of elastic bands used in each case and the magnitude of the proposed
346 parameter, the MAD). In addition, this parameter could also be studied to be proposed as an
347 objective metric to evaluate selective motor control (SMC), as the results clearly show that the
348 differences in the activation of the biceps brachii and triceps brachii as agonist muscles in elbow
349 flexion–extension can be quantified. Therefore, it is expected that the co-activation of synergist
350 muscles in this or other movements or functional activities can also be measured and quantified.
351 Previous studies have already demonstrated the reliability of sEMG in extracting relevant
352 parameters related to SMC and spasticity [4637, 4738].

353 Raouafi *et al.* [4839] proposed an upper limb motor function index after principal component
354 analysis of kinematics, electromyography, and inertial measurements, which can detect deviation
355 from the upper limb motor function of a typically developing group of children. Therefore, the
356 proposed metric can play a key role in assessing the actual impact of robot-assisted therapies.
357 These results open the door to identifying new markers that can quantify the level of skill
358 acquisition, generalization of motor skills to functional activities, and retention of previously
359 acquired skills.

360 Finally, the limitations of this study should be considered. It is important to note that our sample
361 is composed of healthy children with relatively homogeneous physical and motor characteristics
362 so the results cannot be directly extrapolated (although they can serve as a reference) to children
363 with different kinds of motor impairments as, in this case, much more heterogeneity in the
364 physical and motor development conditions is expected. In addition, it is important to highlight
365 that although in this proof of concept, a relatively small sample was enough to obtain statistically
366 significant results with a large effect size (again possibly due to the homogeneity of the sample),
367 it is expected to need a larger sample for the validation of the exoskeleton in children with motor
368 diseases.

370 Conclusions

371 POWERUP is a low-cost and easy-to-use passive exoskeleton for upper limb whose assistive
372 performance (tested for assisting elbow flexion in a sample of healthy children as a proof of
373 concept) can be easily validated through an electromyography-based assessment.

374 According to the results, there ~~are~~ were statistically significant differences in our sample in the
375 Muscle Activity Distribution (MAD, %) for both biceps and triceps between the transparent
376 mode (no assistance provided) and the assistive mode. It can be stated that (1) the exoskeleton
377 correctly ~~assists~~ assisted elbow flexion, reducing the effect of gravity and, (2) the MAD can be
378 considered a reliable metric for the validation of the assistive performance of the exoskeleton.

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379 Overall, it can be said that sEMG ~~is-was~~ a powerful tool for both (1) the assessment of the
380 assistance capacity in elbow flexion of the POWERUP exoskeleton tested in healthy children as
381 a proof of concept and, by extension, of other similar upper limb exoskeletons, and (2) it offers
382 an interesting metric, the MAD, to be more deeply studied for both the evaluation of the SMC
383 and the impact of robot-assisted therapies in children with motor disabilities.

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388 and collaboration in the experiments.

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