

The Economic Value of Augmentative Exoskeletons and their Assistance

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Understanding the success of augmentative exoskeletons is critical to their impact on society. Many exoskeletons are developed to address impairments stemming from neuromotor pathology, which provides guidance for understanding their abilities; however, for augmentative exoskeletons that assist able-bodied users, a clear metric of success remains an open question. In this study, we leverage an approach from economics--the Vickrey second-price auction--to quantify the economic value added by lower-limb exoskeletons during uphill walking. In our protocol, participants describe the monetary compensation needed to continue walking uphill for consecutive bouts of two minutes. To incentivize truthful descriptions of participant's "price to walk," their compensation amounts were provided as bids in Vickrey auctions before each bout. We compared these data across different conditions to determine the marginal value of the exoskeleton weight, assistance, and weight + assistance. Our approach found that the total value of the exoskeleton and its assistance was modest (\$3.40/hr, SD: \$3.35/hr); while most participants found the assistance itself valuable (\$19.76/hr, SD: \$19.43/hr), this value was nearly offset by the cost of wearing the exoskeleton weight (-\$18.59/hr, SD: \$18.28/hr). Our approach and results demonstrate that economic value can be a powerful tool to develop exoskeletons that maximize user benefit.

22

23 **Introduction**

24 Powered lower-limb exoskeletons have the potential to transform human mobility by extending the locomotor
25 abilities of their wearers. These technologies augment lower-limb function by providing mechanical assistance
26 to the joints of the legs in tandem with the human neuromotor system, and thus can make physically demanding
27 tasks less challenging for both able-bodied and impaired individuals. Assistance provided by modern exoskeletons
28 have been shown to reduce the caloric demands and muscular effort required for walking (1–9). Consequently,
29 exoskeletons may have beneficial implications for recreational users and workers in factory, military, or supply
30 chain environments. In addition, exoskeletons can reduce muscle activation, and thus may reduce joint loading
31 (10–13), potentially extending the physical capabilities of aging individuals. Rehabilitation-focused exoskeletons
32 may also restore the mobility of people with neuromotor deficits who face weakness and impairments in balance,
33 coordination, and joint mechanics following upper motor neuron disease (*e.g.* stroke). Exoskeletons can be used
34 to assist the gait of these individuals, with several commercially-available lower-limb exoskeletons having been
35 approved by the U.S. Food and Drug Administration.

36 The metrics by which we assess exoskeletons drive their design, control, and potential impact. For rehabilita-
37 tive applications, the design of these technologies has a clear physiological objective: the restoration of impaired
38 gait function. These exoskeletons are often considered successful if they restore able-bodied kinematics or kinetics,
39 which provide clear physiological goals on which to base exoskeleton design and control decisions (8, 12, 14–16).
40 However, for augmentative applications in which the user is typically able-bodied, the metrics of success are
41 less clear. Currently, augmentative exoskeletons are developed based on their ability to meet a physiological
42 objective; the “gold standard” for exoskeleton success is the reduction of metabolic expenditure during locomo-
43 tion (*i.e.* a reduction of the calories burned) (17). This objective is both intuitively meaningful and objectively
44 measurable (18, 19). Modern exoskeletons have reduced the metabolic rate relative to unassisted walking by an
45 average of approximately 14% (1–6, 20). This objective has led to the rise of promising ‘human-in-the-loop’
46 (HILO) optimization techniques, which directly modulate exoskeleton assistance based on the metabolic reduc-
47 tions experienced by the wearer (21–27). Recent work has established experimental infrastructure that illustrates
48 the tight coupling between metrics of success and exoskeleton development. An example of this infrastructure
49 includes tethered emulator systems (7, 28), whose purpose is to inform exoskeleton design and control based on
50 their ability to reduce the metabolic expenditure of their wearer. Numerous studies investigating the biomechanical

51 underpinnings of metabolic cost reductions have also been conducted to find more optimal exoskeleton assistance
52 settings (20, 22, 29, 30). Finally, other metrics include net-joint torque reduction (11) and muscle activation reduc-
53 tion (10, 12), which are commonly used, easier-to-measure proxies for improvement in energetics.

54 Though the physiological benefits of exoskeletons have been demonstrated using metabolic metrics, these ben-
55 efits have not widely translated to wearer perception of enhanced endurance and strength. These perceptions are
56 important, as for augmentative exoskeletons to reach their potential in society, users will need to voluntarily accept
57 these technologies into their lives. They thus must be developed to provide a perceivable benefit to their wearer, in
58 addition to objective assessment of their impact. Our recent work has shown that during exoskeleton-assisted walk-
59 ing, the average user cannot yet perceive the benefit of most systems available today (31–33). That is, the metabolic
60 rate needed to be reduced by 23% (N = 10) before exoskeleton users could reliably perceive this improvement
61 (whereas most modern exoskeletons reduce the wearer’s metabolic rate by 14% compared to unassisted walk-
62 ing (17)). These results agree with prior studies that showed humans were relatively insensitive to small changes
63 in exertion in other exercise contexts (34). Intuitively, if the user is unable to perceive the metabolic reduction
64 provided by an exoskeleton, this value may be difficult to incorporate into decision-making during exoskeleton
65 design, translation, and adoption. Consequently, assessing and developing exoskeletons based on reductions in
66 metabolic rate could result in systems that are not perceived as valuable by users, despite significant energetic
67 benefits (3, 23, 30, 35–37).

68 An alternative method for measuring success in exoskeleton development is to quantify the perceived *eco-*
69 *nomic value* provided to the wearer during its use. Economic value, measured in monetary currency (*e.g.* US
70 dollars), is assigned by the wearer and can reflect the multifaceted nature of exoskeleton user experience. Al-
71 though exoskeletons can provide assistance that improves energetics, that assistance often comes at a cost to the
72 wearer. Exoskeletons can add discomfort, weight, and audible noise, in addition to having aesthetic implications.
73 While exoskeletons may potentially have universal positive value, the heterogeneity of the metabolic response
74 to exoskeleton assistance, coupled with the known variety of responses to new innovations within the social sci-
75 ences (38), could also imply a wide range of valuations for wearing an exoskeleton. If the user is able to assign
76 economic value to the experience of exoskeleton use, they are able to inherently balance and quantify these trade
77 offs. Thus, we posit that exoskeletons that maximize economic value may have a greater likelihood for adoption
78 and use. Prior work in management science has established that the perceived value of different technologies
79 has a significant impact on user intent to adopt those technologies into their daily lives (39–43). When potential
80 exoskeleton users, manufacturers, and others are weighing the choice to adopt or purchase an exoskeleton, the con-

81 sciously perceived benefits must outweigh these costs. Thus, the perceived economic value of exoskeleton use is a
82 potentially powerful metric for designing and controlling exoskeletons that quantifies meaningful, individualized
83 benefits to wearers.

84 In this study, we introduce a tool for measuring the perceived economic value of exoskeleton use as a metric
85 to evaluate their performance and user experience. We define and use this tool to quantify the economic value—
86 termed *Marginal Value* (*MV*)—provided by bilateral ankle exoskeleton use during uphill walking. We leveraged the
87 Vickrey second-price auction to measure participants’ “price to walk,” which was then aggregated and compared
88 across conditions to obtain our economic value metric (*i.e.* *MV*). Our experiment revealed that while there was
89 insignificant positive value of exoskeleton use across subjects, there was a large disparity between subjects. Some
90 subjects reported substantial value provided by the exoskeletons, which represents an opportunity target these
91 “responders” in future work. The near-zero net *MV* from the exoskeleton stems from two competing effects: the
92 *MV* of the powered assistance alone was substantially positive, but it was counteracted by the cost from wearing
93 the device itself. The use of *MV* offers advantages over the more common metabolic rate metric, including its
94 accessibility, in that it does not require specialized equipment, and its intuitiveness, as users and manufacturers
95 are more likely to understand the value of monetary currency over biomechanical quantities (*e.g.* calories burned
96 or muscle power). Our approach is also generalizable, and can be used to measure the value of not only different
97 types of exoskeleton assistance, but also various technologies, activities, or experimental conditions.

98 **Background on Vickrey auctions**

99 Within the field of economics, the Vickrey second-price auction (44) provides a well-studied incentive structure for
100 quantifying the economic value of abstract concepts. In a seller’s Vickrey auction, participants compete via bidding
101 to sell a good. The winner of the auction is the participant that bids the lowest amount to sell the good. However,
102 the winner will earn the amount bid by the second-lowest bidder. The Vickrey auction is also “sealed-bid” in that
103 each participant’s bids are not revealed publicly. This second-price paradigm incentivizes subjects to bid their true
104 perceived value for the good (44–46) rather than attempt to guess (and slightly underbid) all other bidders’ lowest
105 bids. Auction participants would be discouraged from bidding higher than what they believe the good is truly
106 worth, for fear of losing the auction to a competitor with a lower bid, and the second-price nature breaks the link
107 between the auction winner and their actual bid. Conversely, they will have no incentive to underbid below their
108 true value since they would risk selling for a “loss” if they won the auction. Thus, the Vickrey auction provides
109 a quantitative measurement for the value of arbitrary goods or services by measuring an individual’s willingness

110 to buy or sell (46, 47). The specific incentive structure is thus less prone to biases when quantifying the value of
 111 a good when compared to alternative methods such as direct feedback (48, 49). Prior researchers have also used
 112 Vickrey auction metrics to measure the value of abstract concepts or actions, such as food safety (50), GMO-free
 113 foods (51), the stigma resulting from HIV (52), personally identifiable information (48), and smartphone battery
 114 life (53). In particular, Coursey *et al.* employed the Vickrey auction to quantify the willingness of participants to
 115 endure performing an unpleasant task, such as tasting a bitter liquid (54). These examples highlight the promise of
 116 the Vickrey auction to quantify the perceived economic value provided by experiences, including exoskeleton use.

117 Results

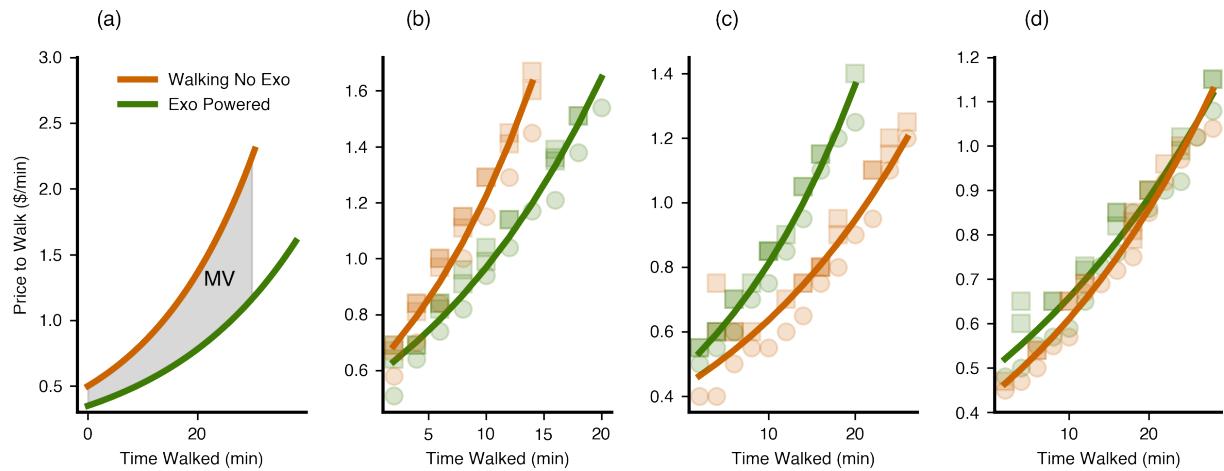


Figure 1: (a) Generic examples of price to walk curves for the walking-no-exo condition (green) and the exo-powered condition (orange) from different subjects. The marginal value (MV, the value of the exoskeleton assistance) is given by the area between the curves. In actual trials, the price to walk curves are estimated by fitting first order exponentials to subject bids. Circles denote winning bids, while squares denote losing bids. The participant in (b) shows a clear benefit from the device, the one in (c) shows a clear detriment, and the one in (d) is more ambiguous.

118 Within each trial, participant bids invariably trended upwards as participants became more fatigued and this
 119 trend was well represented by a first-order exponential. During each trial, participants either walked with nor-
 120 mal walking shoes ('walking-no-exo'), with the exoskeleton applying assistance ('exo-powered'), or with the
 121 exoskeleton donned but applying no power ('exo-powered-off'). Across all three walking conditions, the average
 122 bid for each two-minute interval was \$0.75, with a standard deviation of \$0.39. The maximum bid was \$5.00,
 123 while the minimum bid was \$0.10. As participants continued to walk during each trial, their bids increased at
 124 varying rates (Fig. 1). The first-order exponential model exhibited an R^2 of 0.87 ± 0.11 , averaged over all con-

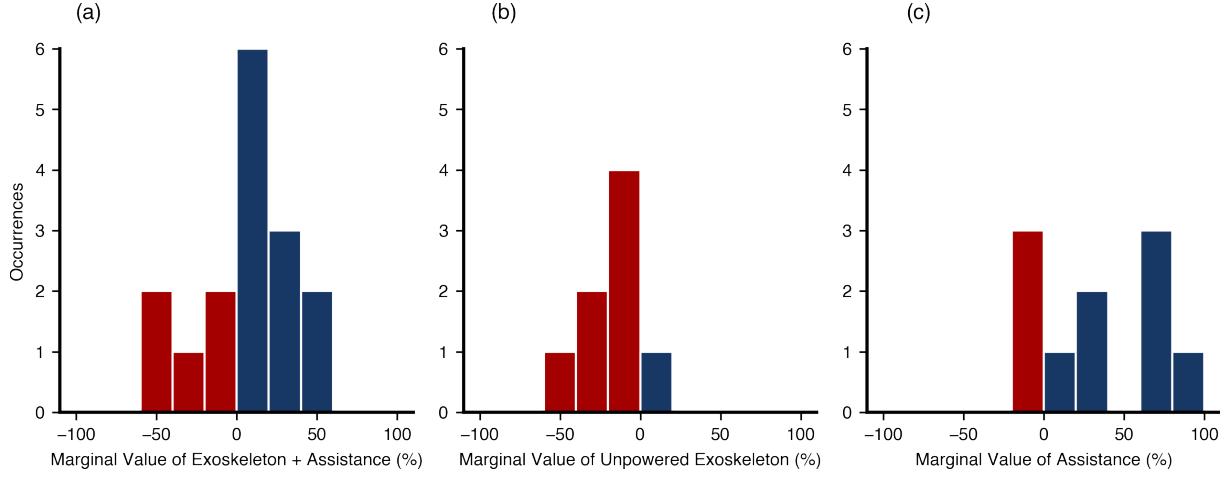


Figure 2: Histograms of the Marginal Values (MVs) for the different conditions. Positive MVs, indicating that value was added to the wearer are in blue, while negative MVs, indicating that costs were imposed on the wearer, are in red. (a) The MVs of exoskeleton + assistance for all sixteen subjects. The average MV was 5.81%, with an SD of 31.1%. (b) The MVs of the unpowered exoskeleton for the twelve subjects who participated in this condition. Aside from one subject, all participants experienced notable economic costs from this condition, reflected by the average MV being significantly negative (average: -31.8%, SD: 45.0%). (c) The MVs for the exoskeleton assistance alone (average: 33.8%, SD: 38.1%). While the average MV of the exoskeleton + assistance was not significantly positive, the assistance alone conferred a significant benefit.

ditions experienced by the sixteen participants. These first-order responses denote the user's price to walk curves for each condition (the curves for the walking-no-exo and exo-powered conditions are shown in Fig. 1a). The area between the walking-no-exo and exo-powered curves denotes the Marginal Value (MV) of the exoskeleton and its assistance, which is the value obtained by the participant from the exoskeleton's use. Participant price-to-walk curves broadly demonstrated three potential outcomes; namely, a clear economic benefit from the exoskeleton's assistance (higher walking-no-exo curve than exo-powered curve, Fig. 1b, positive MV), a clear economic penalty (lower walking-no-exo curve than exo-powered curve, Fig. 1c, negative MV), and a negligible economic effect (similar walking-no-exo and exo-powered curves, Fig. 1d, near-zero MV).

Our approach was able to quantify the intuitive effects of added mass and assistance. Across subjects, the MV of the exoskeleton + assistance was positive but not significant, while the MV of the unpowered exoskeleton was significantly negative, and the MV of the assistance itself was significantly positive. The average inter-subject MV of exoskeleton use was 5.82%, with a standard deviation (SD) of 31.14% (N=16, Fig. 2a). The exoskeleton + assistance thus provided only a small value benefit to the average participant. Using a t-test, the average MV of exoskeleton + assistance (5.82%) was not significantly different from zero ($p = 0.24$). However, as denoted by the high standard deviation, some participants received large benefits from the device's assistance, while others

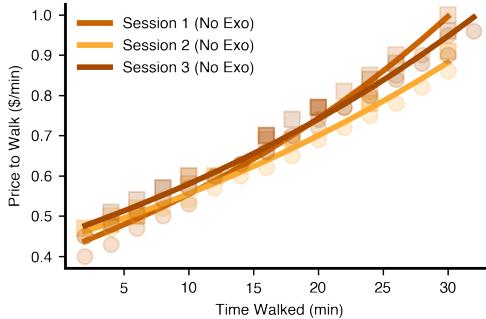


Figure 3: Representative price-to-walk curves for a single 'repeater' subject across different days. The agreement across the curves demonstrates the test re-test reliability of our approach. These sessions were completed over a seven day span.

experienced an economic penalty from exoskeleton use. The average MV for the unpowered exoskeleton was -31.8% with a standard deviation of 45.0% (N=13, 2b). Using the same t-test as in the exo-powered condition, we found this change to be significantly different from zero ($p = 0.03$). Additionally, the powered assistance alone from the exoskeleton provided a significant increase in value (mean: 33.8%, SD: 38.1%, $p = 0.01$, N=13, Fig. 2c).

The integral of each price to walk curve yields the cumulative price to walk for each condition (Fig. 10). The average cumulative price for the walking-no-exo condition was \$29.23 ± \$28.11 (Fig. 10a), for the exo-powered condition \$25.38 ± \$13.87 (Fig. 10b), and for the exo-powered-off \$49.68 ± \$54.34 (Fig. 10c). Changes in participant value due to the different walking conditions can be measured by comparing the associated cumulative prices.

We also characterized the repeatability of our measurements with a subset of participants to verify that the changes in price to walk detected due to the changing walking conditions were attributable to the difference in perceived difficulty between conditions and not due to participant day-to-day variability (a representative repeater trial is shown in Fig. 3). For the four subjects that repeated the walking-no-exo condition at least once, the average intra-subject standard deviation of the cumulative price for the walking-no-exo condition was 3.38%, expressed as a percentage of each subject's average walking-no-exo cumulative price. This quantity represents the day-to-day fluctuation in the value demanded to walk, in addition to any user or experimentally derived noise.

Discussion

In this work, we use the Vickrey second-price auction as a method to capture the economic value or detriment provided by exoskeletons and their assistance. As part of our broader goal of emphasizing the user's role in the

159 design, control, and evaluation of exoskeletons (32, 55), the intent of this work is to quantify the success of these
160 technologies through a user-centered metric which encompasses the different aspects of the user experience, in-
161 cluding exertion, weight, comfort, and assistance. To this end, the wearer specified their “price to walk” during a
162 series of Vickrey auctions from which we quantified the economic value of different conditions. Our approach is
163 particularly relevant for assessing the success of augmentative exoskeletons, which do not have a clear, clinically-
164 relevant, biomechanical or physiological objective (as in the case of prostheses or orthoses). Our work is motivated
165 by the belief that obtaining accessible and relevant metrics of success is critical to the successful development and
166 adoption of exoskeletons in the real world. This motivation is supported by a history of strong interconnection
167 between the assessment of exoskeletons and their design and control architectures (21–27). Our strategy quanti-
168 fies success using monetary currency, as opposed to biomechanical or physiological quantities that may be more
169 difficult for users, manufacturers, and those outside of the exoskeleton research field to interpret.

170 Our results showed that the economic value provided by the exoskeleton assistance was comparable to the
171 cost incurred by wearing the unpowered system. We obtained these results by quantifying the marginal value
172 (MV) between two conditions, and by varying which two conditions were compared, we isolated the value of
173 the exoskeleton + assistance, the assistance alone, and the detriment of the unpowered exoskeleton. The average
174 cumulative price of walking uphill for 30 minutes without the exoskeleton was \$29.23 (\$58.46/hr SD: \$57.49/hr).
175 The MV of wearing the unpowered exoskeleton—that was not providing assistance—was -31.8%, which translates
176 to a monetary cost of \$18.59/hr (SD: \$18.28/hr) for wearing the unpowered system. When assistance was applied
177 by the exoskeleton, the MV increased to just above zero (\$3.40/hr, SD:\$3.35/hr) when compared to not wearing
178 the exoskeleton. The marginal value added by the assistance alone was 33.8%, which translates to an added value
179 of \$19.76/hr (SD: \$19.43/hr). Thus, the assistance applied by the exoskeleton offset the cost of wearing the system,
180 and the net benefit was modest but positive. These results suggest that modern augmentative ankle exoskeletons
181 may not provide a substantial benefit to their wearers during short-term uphill walking. This is particularly sur-
182 prising because of the lightweight, refined design of the exoskeletons used, and the high physiological demands of
183 the uphill walking task (56, 57), which we chose to increase the observed value of the exoskeleton (*i.e.* to improve
184 the signal-to-noise ratio in our measurement of economic value).

185 Our approach and metrics can be used to guide the development of exoskeletons in new ways. Exoskeleton
186 controllers can potentially be directly engineered to maximize the MV added (in dollars) from the assistance of the
187 exoskeleton during walking. Similarly, exoskeletons can themselves be designed to minimize the economic cost
188 they impose on the wearer, which includes not only weight, but also other factors such as inertia and discomfort.

189 This information can be obtained using the MV of the unpowered exoskeleton when compared to the walking-no-
190 exo condition. Additionally, this study suggests that exoskeleton development may be improved by identifying the
191 users who innately receive greater economic value from wearing the system. We found that nine participants ob-
192 tained substantial positive economic value from wearing the system ($27.8\% \pm 16.3\%$, corresponding to \$16.25/hr
193 $\pm \$9.53/\text{hr}$), even though the average across all participants was lower. Such “responders” may be more willing to
194 adopt exoskeletons in the face of potential drawbacks, including cost and added mass, among others. Understand-
195 ing why some users obtain this value would enable the targeting of these individuals to maximize the translation
196 and impact of augmentative exoskeletons. For example, models such as the diffusion of innovation (DOI) theory
197 within the social sciences categorize individuals based on their willingness to adopt a new innovation (38), such
198 as exoskeletons. It is possible that the individuals who obtained positive value within our study fall into the “early
199 adopter” category, and are more willing to find value in exoskeleton use. Future investigations could focus on the
200 separation of these individuals *a priori* based on the social characteristics within DOI theory, or based on other
201 human factors, such as fitness level or perceptual abilities (32). Furthermore, investigations of whether users can
202 be “trained” to receive a greater economic value from the exoskeleton are promising avenues of future study.

203 Participant results were consistent when repeated across days, supporting the quality of our measurements.
204 Four participants repeated the walking-no-exo condition across several days, and the inter-day variance of the
205 cumulative prices was low (\$0.67, Table 4). This consistency supports the ability for our approach to quantify
206 perceived changes in value, with potentially minimal effect from inter-day confounding factors. Using this inter-
207 day variance, we estimated the Minimum Detectable Change (MDC), defined as the minimum change in value not
208 caused by chance. The MDC for the MV measurement was 9.4% (95% confidence interval, difference between
209 two measurements, standard deviation 3.4%). To identify which subjects had a noticeable change in value, we
210 compared each participant’s MV against the inter-day variability’s 95% confidence interval ($\pm 9.4\%$); nine of 16
211 subjects had MVs exceeding this threshold in the positive direction, while four subjects exceeded it in the negative
212 direction.

213 Our auction strategy avoids several potential limitations of more direct methods to quantify value, such as a
214 Likert scale test, direct feedback, or auctions without a real monetary payout (58). The locomotion task imple-
215 mented in our approach does not necessitate extreme fatigue and can remain relatively short, avoiding confounding
216 factors (*e.g.* boredom or opportunity cost). Additionally, the incentive structure of the auction links participant bids
217 to a specific consequence (*e.g.* walking uphill), and are thus less prone to biases, such as self-enhancement or social
218 desirability, while still remaining intuitive (48, 49).

219 Individual bids and, subsequently, cumulative prices varied widely across subjects. Since participants were
220 able to set their own bids, this added inherent variability across subjects. This variation may have resulted from
221 each participant having different internal valuations of their time, which could have been driven by differing so-
222 cioeconomic status, athleticism, opportunity cost, or other factors. This large discrepancy motivated the creation
223 of the marginal value (MV) metric, which normalizes by each participant’s ‘baseline’ cumulative price from the
224 walking-no-exo condition, expressing the change as a percent of the walking-no-exo condition.

225 We anticipate that the conclusions drawn in this study extend to other augmentative ankle exoskeletons, al-
226 though further study is needed to quantify the value from other technologies. The ankle exoskeletons from this
227 work were developed for commercial use, and represent a “best-in-class” technology; the system’s refined design
228 is lightweight, untethered, and provides substantial net-positive energy during each stride (13.4 ± 2.9 J). We ex-
229 pect that other, similar exoskeletons would have comparable marginal values, and thus may also not provide a
230 substantial economic benefit to their wearers. Additionally, although the task of uphill walking does not represent
231 all possible uses for augmentative exoskeletons, it enables an opportunity to quantify value provided by the ex-
232 oskeleton during an intuitive application where it can provide substantial benefit. We expected that the increased
233 energetic difficulty of uphill walking (56, 57) would enable the exoskeleton to more readily demonstrate its value
234 to the wearer. Furthermore, we chose the uphill walking task to reduce experimental duration as the greater dif-
235 ficulty would cause participant bids to rise more quickly. We expect that for less strenuous tasks, exoskeletons
236 will show reduced marginal value; this hypothesis is supported by the results of a separate study we conducted
237 (see Supplementary Information) in which the MVs of the exoskeleton + assistance were lower for level-ground
238 walking when compared to uphill walking.

239 The changes in exoskeleton conditions caused intuitive changes in value, supporting the validity of our eco-
240 nomic value metric. That is, we believe the different exoskeleton conditions (walking-no-exo, exo-powered-off,
241 and exo-powered) likely drove the changes in value measured, and these changes agree with the biomechanical
242 demands of the conditions. The MV of the exo-powered-off condition was strongly negative, denoting a cost to
243 wearing the system. This result is expected, since without assistance, wearing an exoskeleton is akin to wearing
244 shank weights during locomotion, which would necessitate greater mechanical work from the triceps surae and
245 cause an upstream increase in (by ~ 16.8 W based on the results from (36, 59)). Additionally, when the assistance
246 was added, the value increased to just above zero, indicating the assistance was useful in offsetting the challenge
247 of wearing the unpowered system during uphill walking.

248 The MV for the exo-powered condition may shift with repeated sessions, as the participants adapt to the assis-

249 tance. Prior work has found that training sessions across multiple days yield greater reductions in metabolic rate
250 in naive users (3). Accordingly, the MV for the exoskeleton may adjust as well while the user adapts to the expe-
251 rience of wearing an exoskeleton. For this initial investigation, we sought to understand the immediate economic
252 value obtained from a first-time experience of exoskeleton use, analogous to a user assessing an exoskeleton when
253 making the decision to adopt (*i.e.* a “test drive”). To calculate the MV, we thus conducted a single exo-powered
254 session with a brief adaptation period of a few minutes. Future work is needed to understand any adaptation of
255 value that may occur over time, which would have implications in the longer-term value of these technologies.

256 The use of simulated bidding agents may have influenced the results but we expect the overarching conclusions
257 of the study would not be affected. In our auction protocol, we used simulated bidding agents (“robo-bidders”)
258 to model human behavior, rather than implementing our approach with multiple human subjects. The intent for
259 this choice was to reduce the logistical challenges of our approach, which would otherwise have required multiple
260 treadmills and exoskeletons. To mitigate any effect of the simulated nature of the other participants, the human
261 participant was informed that the robo-bidders were humans participating in the experiment in remote locations.
262 The behavior of the robo-bidders—defined by a parameter that governs the rise and fall of their bids—was derived
263 from pilot data obtained from human participants. To reduce the likelihood that the human participants could infer
264 the robo-bidder behavior model, the robo-bidder bids were corrupted with noise. The use of robo-bidders also
265 enabled us to standardize the interaction between the human participant and the other auction participants. Any
266 series of auctions would naturally establish an equilibrium between the participants; thus, by using robo-bidders,
267 we were able to control for this equilibrium, which strengthens our ability to compare across subjects. In addition,
268 the use of robo-bidders enables our results to more readily be compared across other researchers, institutions, and
269 exoskeletons that are assessed using comparable methods. The code to implement the robo-bidders is provided in
270 [TBD].

271 During the experimental protocol, participants responded with bids as instructed, but we are unable to know
272 for certain that their bids were truly honest (*i.e.* truthfully reflecting their internal sense of value). Our study relies
273 on participants honestly reporting their bids, which is theoretically guaranteed as an optimal strategy by the nature
274 of the Vickrey second-price auction for rational actors (44). In addition to the natural structure of the Vickrey
275 auction, which incentivizes truthful bidding, we minimized the potential influence of dishonest bidding by setting
276 the auction interval to two minutes (as opposed to a shorter duration). We chose the two minute duration to increase
277 the effort required, and thus minimize potentially ‘dishonest’ exploratory bidding, which would corrupt the price
278 to walk curves. Additionally, subjects continuously received verbal instructions to always bid honestly as the

279 best strategy, which has been demonstrated to increase the likelihood of honest bidding (60). Finally, participants
280 received the expected monetary compensation that resulted from their winning bids, which similarly incentivizes
281 truthfulness.

282 Conclusion

283 We have developed a method to quantify the economic value of augmentative exoskeletons, and used these methods
284 to assess the value provided by bilateral ankle exoskeletons during uphill walking. Our results underscore the
285 challenge of developing exoskeletons that provide a clear, meaningful benefit when augmenting the healthy human
286 neuromotor system. The value of the assistance provided by the exoskeletons was modest, and just offset the
287 cost of wearing the unpowered system. Our results also suggest that more work is needed to identify why some
288 participants received substantially greater value from the exoskeleton assistance, which could be used to identify
289 or train individuals for maximizing the real-world impact of these technologies. Finally, the economic value metric
290 we have developed can be readily used to compare different design and control strategies to develop exoskeletons
291 that are maximally valuable to their wearers.

292 Methods

293 Participants

294 In this study, sixteen able-bodied participants ($N = 16$, 5 female, 9 male; age = 26.3 ± 4.61 years; mass = 77.1 ± 13.4 kg, Table. 1) walked using bilateral ankle exoskeletons on a treadmill. All participants provided written
295 informed consent before participation. The study protocol was approved and overseen by the Institutional Review
296 Board of the University of Michigan Medical School. Participants had no prior experience walking with the
297 bilateral ankle exoskeletons featured in this study.

299 Experimental Protocol

300 Exoskeleton Apparatus

301 Our approach quantifies the value of bilateral ankle exoskeletons that were designed to improve the energetic
302 efficiency of human walking (ExoBoot, Dephy Inc. Maynard MA, Fig. 4). The commercially-available system
303 utilizes an onboard brushless electric motor and flat cable transmission (for a mean transmission ratio $\sim 15:1$,
304 see Supplementary Fig. 6.) to apply ankle assistance during walking. The exoskeletons have a single powered
305 degree of freedom (dorsi-plantar flexion) and a passive, unactuated degree of freedom (inversion-eversion). The

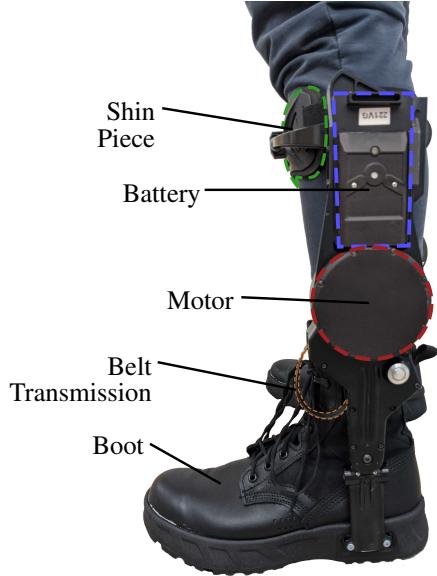


Figure 4: The Dephy ExoBoot used in this experiment. An electric motor applies plantarflexion torques at the ankle via a belt transmission.

306 transmission is unidirectional, which enables the system to apply plantar flexion assistance torque but it cannot
 307 provide dorsiflexion assistance. Each side of the exoskeleton applies a torque profile (Fig. 5a) that provides a burst
 308 of positive power (Fig. 5b) during the terminal stance phase of gait, augmenting the propulsive effort provided by
 309 the triceps surae. The average energy provided by the exoskeleton during the gait cycle is $13.4 \pm 2.9\text{J}$. During
 310 swing phase, the exoskeleton is able to add slack to the belt drive, thereby preventing any unwanted resistance to the
 311 foot; this capability stems from the unidirectional nature of the design. Gait progression, inferred from heel-strike
 312 timing events, was used to schedule how assistance was provided during each step. Similar exoskeletons have
 313 been shown to lower the user's metabolic expenditure during walking as well as reduce the biomechanical power
 314 requirements at the ankle and other joints of the legs (1, 29, 36). The walking assistance controller was developed
 315 by Dephy Inc. as part of their commercial system. As one of the first commercially available exoskeletons, the
 316 Dephy Exoboot was chosen for its ease of use, robustness, and representation of the state-of-the-art.

317 Vickrey Auction Protocol

318 The Vickrey auction (44) is a powerful economic tool for determining the true value placed on goods or actions. In
 319 this type of auction, participants compete to purchase (or sell) a good or item. For each participant, the auction's

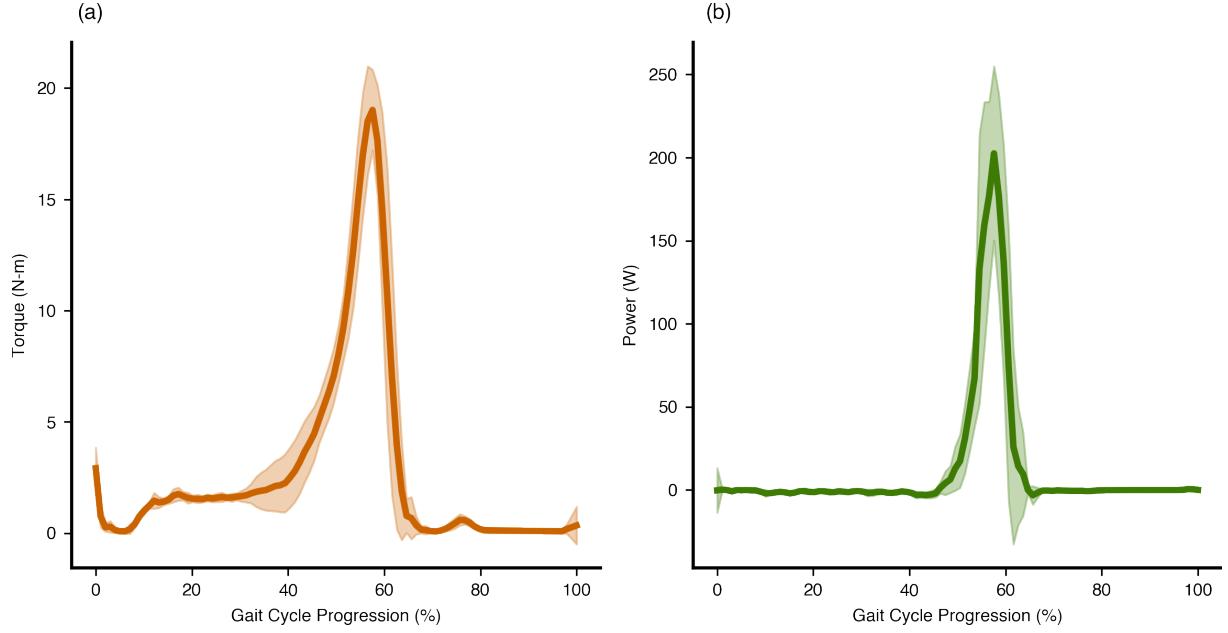


Figure 5: (a) The torque profile applied by the bilateral ankle exoskeletons during the uphill walking task. The torque applied stems from the proprietary control strategy implemented by Dephy Inc and represents the state of the art. The mean profile is shown in solid orange, with a single standard deviation shown by the shaded region. Torque was quantified by using the measured q-axis motor current and the experimentally-derived, q-axis torque constant to calculate motor torque (61), and then multiplying that value by the instantaneous transmission ratio. Gait cycle progression is defined as beginning and ending at sequential ipsilateral heel-strokes. (b) The power applied by the exoskeletons at the ankle during the walking trial. Power was calculated by multiplying the current-derived torque profiles in (a) by measured ankle angular velocities. The mean profile is shown in solid green, with a single standard deviation shown by the shaded region. The average energy provided by the exoskeleton, obtained by integrating the power curves over time is 13.4 ± 2.9 J.

320 structure is designed such that the optimal strategy for obtaining the item is to truthfully represent their internal
 321 value with their bid (*e.g.* to bid an amount equal to the true worth of the item). This optimality stems from the
 322 *second-price* nature of the auction (44–46), in which the winner is the participant who bids the highest (or lowest,
 323 as in the selling implementation used in this study). However, rather than paying the highest bid, the winner of the
 324 auction instead pays the *second-highest* bid. Participants do not know the value of competing bids (sealed-bid). In
 325 theory, bidders are disincentivized to bid less than their true value, as they run the risk of not winning the auction
 326 (not acquiring the good). Similarly, participants should not bid more than their true value, which otherwise could
 327 cause them to pay more than the value of the item. The second-price nature breaks the link between the auction
 328 winner and their specific bid. The inverse is also true for the case of selling an item (second-lowest bid is paid,
 329 lowest bid wins); thus, the incentive structure that elicits truthful bidding still holds in the seller’s auction. Due to
 330 the presence of this optimal strategy, the Vickrey auction provides a method for quantifying the value of abstract

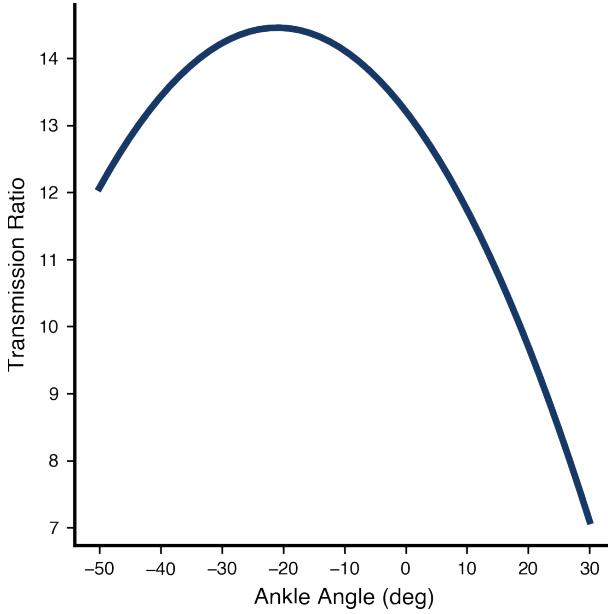


Figure 6: The transmission ratio curve of the exoskeleton device used in this experiment. Positive ankle angles denote dorsiflexion and negative angles denote plantarflexion. The curve was modeled as a second-order polynomial, and was obtained by taking the derivative of the best-fit third order polynomial that relates motor angle to ankle angle.

331 concepts (46, 47). In our protocol, we use the Vickrey auction sequentially to repeatedly sample individuals'
 332 valuations of their time during uphill walking. The participants bids across the sequential auctions denotes the
 333 wearer's "price-to-walk" curve.

334 **Walking Protocol**

335 Participants received guidance on the Vickrey auction protocol prior to the experiment. During this time, par-
 336 ticipants read a lay explanation of the Vickrey auction. Subsequently, participants underwent two mock Vickrey
 337 auctions, where the optimal strategy of truthful bidding was explained. Subjects were also repeatedly informed of
 338 the optimality of the honest-bid strategy to improve the number of honest bids provided (60). In the first mock
 339 auction, participants were presented with a miscellaneous office supply item and told to bid honestly on it as if
 340 they were competing to purchase it in a real Vickrey auction. Each participant wrote down their bid on an index
 341 card, while the researcher wrote down some artificial competing bids. The participant then revealed their bid, and
 342 the experimenter explained which bid won (the highest bid) and what the cost would be (the second-highest bid).
 343 Next, participants were told to imagine they were participating in the actual study, in which they would compete
 344 to sell their walking time in exchange for monetary compensation. Participants were prompted to consider bidding

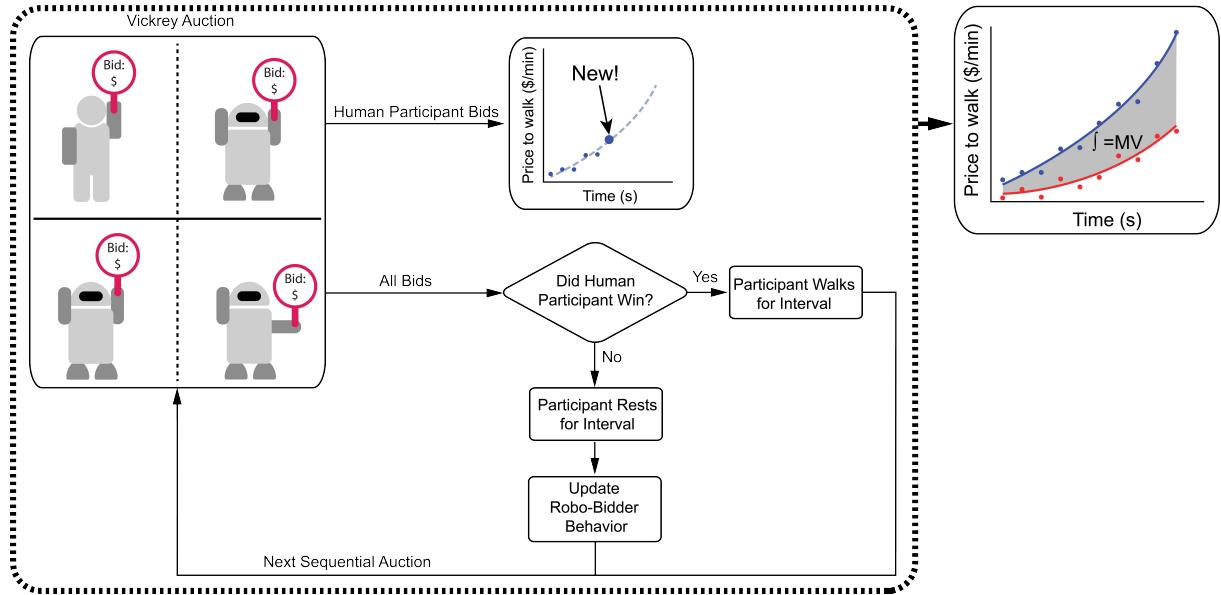


Figure 7: The experimental Vickrey auction protocol. In this protocol, human participants compete against computerized bidders (depicted in the 2×2 grid) in sequential Vickrey auctions for the prize of monetary compensation in exchange for undergoing a strenuous locomotion task. A human participant, shown in the upper-left quadrant, competes against computerized ('robo-') bidders, represented by the robots in the remaining three quadrants. Every two minutes, the human participants verbalize their bids to their experimenter, which are recorded and compared to the bids from the computerized opponents; the lowest bid is the winner of the auction, and the payout is the second-lowest bid. The second-price nature of the auction elicits truthful bids/prices from the participants. Should the human win, they accrue the payout and walk for the two-minute interval; should they lose, they instead miss out on the payout and rest until the next interval. Regardless of whether the human won or suffered a loss, their bid is aggregated to form their price to walk curve. This process repeats for roughly an hour. As the human walks more, they become more fatigued, causing their price to walk to increase; a trend well-modeled by a first-order exponential according to the observed data. Among other conditions, participants competed while wearing normal shoes ('no-exo' condition) and while receiving powered assistance from the exoskeleton ('exo-powered'). The integral of the difference between the price-to-walk curves for the no-exo and exo-powered conditions is the value added from the device's assistance.

345 based on an hourly wage, although they were also told that any truthful bid would be accepted. Again, participants
346 wrote down their bids while the researcher aggregated the competing bids. As in the previous mock auction, the
347 participants revealed their bids, and the experimenter walked them through which bid won and the payout.

348 Following familiarization with the exoskeleton and auction mechanics, participants competed in sequential
349 Vickrey auctions to sell their time while walking uphill. The protocol was organized into a set of Vickrey auctions
350 experienced in series, each lasting two minutes. If the participant won the auction, they would walk for the two
351 minute interval, and they would rest if they lost the auction. During walking, participants walked on a split-belt
352 treadmill with a 10° incline (Fig. 7). Unlike in the mock auctions, the participants would accrue and receive
353 monetary compensation according to the Vickrey auction protocol. At the end of each interval, the winning bid
354 (second-lowest bid) was revealed. The sequential auctions lasted a randomly-specified duration between 50 and
355 70 minutes, with a mean duration of 60 minutes; participants walked on average a total of 31.3 minutes ± 11.5
356 minutes. The duration uncertainty was added to discourage subjects from waiting until their bidding competitors
357 were “exhausted” to inflate payouts. The two-minute interval duration was chosen to minimize subject exploratory
358 behavior. In total, participants experienced approximately 25 - 35 Vickrey auctions in series. By aggregating
359 each subject’s bids across the series auctions, the resulting price to walk curve captured the participants’ valuation
360 of their time to complete the experiment, which can be used to provide insight into the value provided by the
361 exoskeleton in this application.

362 Our protocol was broken up across several days to reduce the effect of fatigue from the experiments. The order
363 of the testing was randomized across participants. On one day, subjects completed the Vickrey auction protocol
364 with normal shoes (*i.e.* the walking-no-exo condition); these data establish their baseline price to walk curve and
365 cumulative price. On a different day, subjects instead walked with powered exoskeleton assistance (*i.e.* the exo-
366 powered condition). Participants were not explicitly told that the exoskeleton would provide assistance; instead,
367 they were told that the exoskeleton was applying a randomized torque profile that would either help or hinder their
368 locomotion. This was done to reduce the chance that participants would experience a placebo effect and perceive
369 the exoskeleton as valuable due to information provided by the experimenter. By comparing the price to walk
370 curves from the walking-no-exo and exo-powered conditions, we obtained a measurement for the value provided
371 by the exoskeleton in this task. If undergoing the exo-powered condition on the second day, subjects were given
372 time to re-familiarize and experience the exoskeleton’s assistance.

373 In addition, each subject was randomly assigned to additional experimental sessions to investigate other at-
374 tributes our approach. Participants were randomly assigned to groups that either quantified the inter-day variabil-

375 ity of the price-to-walk measurements, or investigated the additional condition of walking with the exoskeleton
 376 without assistance (*i.e.* the exo-powered-off condition). Due to the number of experimental sessions required,
 377 participants were assigned to only one of these groups. Of the sixteen total participants, ten participants walked
 378 in the exo-powered-off condition, while four participants repeated the walking-no-exo condition at least once (the
 379 remaining two subjects were not able to participate in the additional sessions). Within the experimental group that
 380 repeated the walking-no-exo condition, three participants repeated the condition twice more, while the last partic-
 381 ipant repeated the condition once. The intent of investigating the exo-powered-off condition was to support that
 382 the price-to-walk curves captured the cost of wearing the additional exoskeleton weight, when no assistance was
 383 provided. The goal of repeating the walking-no-exo condition was to provide insight into the test re-test variability
 384 of the price to walk curves and how they may be affected by inter-day confounding factors.

385 Robo-bidders

386 For a live Vickrey auction, multiple participants are needed; however, this constraint adds practical challenges in
 387 coordination and logistics. To this end, we utilized computerized bidding agents (robo-bidders) as a substitute for
 388 other human participants in the Vickrey auctions. The mechanics of the robo-bidders are modeled on the bidding
 389 behavior of three pilot subjects. The robo-bidders bid with prices following a first-order exponential function of
 390 the number of intervals spent walking. This model captured the effect of fatigue and increased their bids as the
 391 robo-bidder won the auction. If the robo-bidder did not win the auction (meaning the human participant won), their
 392 bid remained constant. Gaussian white noise (zero mean, standard deviation \$0.01) was added to corrupt the robo-
 393 bidder bids, to reduce the likelihood that the human participant would intuit the robo-bidder model. Robo-bidder
 394 behavior at time t_k during simulated walking was governed by the following equation:

$$y_k(t) = k \cdot e^{(b \cdot t_k)} + \sigma, \quad (1)$$

395 where y_k was the robo-bidder's price, k was the initial price, b was the rate at which the price increases, t is time, and
 396 σ was drawn from normal distribution $\mathcal{N}(0, 0.01)$. Parameters k and b were set differently for each robo-bidder;
 397 full implementation details can be found in [TBD]. Having robo-bidders instead of human auction participants
 398 reduced the logistical difficulty of executing the Vickrey auction experiments over time, while still replicating
 399 modeled behavior of a human participant. Naturally, the robo-bidder bids competed in parallel with the human
 400 subject and approached an equilibrium behavior, which affected the total walking time of the human participant.
 401 In the application of auctions in series, an equilibrium would also likely be established between multiple human

402 participants. The human participants were not initially made aware of the fact that they were competing against
 403 computerized opponents. They instead were told that they were competing in live Vickrey auctions against other
 404 humans located remotely, with the experimenters communicating live. This was done to prevent the participants
 405 from attempting to learn and exploit the price to walk curves of the robo-bidders in an attempt to maximize profits
 406 beyond the strategy of honest bidding. Participants were debriefed on the true nature of the robo-bidders at the
 407 conclusion of their participation in the experiment.

408 Analysis

409 Marginal Value

410 Our outcome metric is the marginal value (MV) that stems from difference in two price to walk curves obtained
 411 for different experimental conditions. Each price to walk curve was fit with a first-order exponential response of
 412 the form:

$$Y(t) = k \cdot e^{(b \cdot c \cdot t)}, \quad (2)$$

413 where Y is the participant's price, k is the initial price, b is the rate at which the price increases, c is a scaling
 414 factor equal to the participant's win-rate, and t is time. The scaling factor is necessary to control for the different
 415 durations of walking experienced by the participants (*i.e.* variations in the number of auctions won). After cor-
 416 recting for this win-rate, these curves are integrated and subtracted, yielding the area between the two curves. This
 417 area represents the marginal value added—in US dollars—from one exoskeleton condition compared to another.
 418 The expression for the MV is as follows:

$$MV = \overbrace{\int_{t_1}^{t_2} k_1 \cdot e^{(b_1 \cdot t)} dt}^{\text{exo condition 1}} - \overbrace{\int_{t_1}^{t_2} k_2 \cdot e^{(b_2 \cdot t)} dt}^{\text{exo condition 2}}, \quad (3)$$

419 where k_1, b_1 correspond to the first exoskeleton condition in the comparison, k_2, b_2 correspond to the second
 420 condition, and t_1, t_2 are the bounds of the time domain for the integrals—0 and 30 minutes respectively, which
 421 roughly corresponds to the average time each participant walked in each trial. The MV is then equivalent to the
 422 value added or removed by the exoskeleton during a continuous thirty minute, uphill walking task. We commonly
 423 normalized the MV by the cumulative price from the walking-no-exo condition to obtain a percentage change.
 424 However, this metric can be converted to compare any two experimental conditions or other candidate control
 425 strategies that researchers wish to evaluate in terms of economic value.

Table 1: Participant characteristics.

| Participant | Gender | Weight (kg) | Age | Height (cm) |
|-------------|--------|-------------|-----|-------------|
| 1 | F | 83.64 | 24 | 160.02 |
| 2 | M | 99.2 | 27 | 176 |
| 3 | M | 84 | 22 | 186 |
| 4 | M | 90.7 | 25 | 182.9 |
| 5 | M | 80.7 | 26 | 177.8 |
| 6 | F | 72.58 | 29 | 170.18 |
| 7 | M | 80 | 34 | 180.3 |
| 8 | M | 74.8 | 34 | 74.8 |
| 9 | M | 68.03 | 25 | 170.18 |
| 10 | M | 70.3 | 22 | 185.4 |
| 11 | M | 67.13 | 25 | 167.6 |
| 12 | M | 104.3 | 36 | 193 |
| 13 | F | 74.84 | 23 | 177.8 |
| 14 | M | 70.3 | 21 | 172.7 |
| 15 | M | 61.2 | 23 | 163 |
| 16 | F | 52.4 | 25 | 162.6 |

426 1 Supplementary Information

427 1.1 Supplementary Text

428 1.1.1 Level-ground study

429 We conducted a separate, smaller study in which two participants (Subjects 2 and 7) underwent the main study's
 430 protocol, but on level-ground instead of at a 10° incline. This was done to investigate how their MVs changed when
 431 the walking was made easier and more representative of exoskeleton use in the community. Both participants' MVs
 432 diminished when the task was level-ground walking—Subject 2's MEV dropped from 36.35% to -28.10%, while
 433 Subject 7's MEV dropped from 26.68% to 12.62% (Fig. 13). These results suggest that for everyday mobility tasks,
 434 which are likely to be more similar in difficulty to level-ground walking than uphill walking, energetic exoskeleton
 435 assistance may have less value to potential adopters. Furthermore, augmentative exoskeletons may thus provide
 436 maximal value during highly strenuous tasks. Future work should investigate if these results are consistent across
 437 more subjects and different walking conditions.

Table 2: Powered condition cumulative prices and marginal values.

| Subject | MV (%) | Exo Powered (\$) | Walking No Exo (\$) | | |
|---------|--------|------------------|---------------------|-------|-------|
| | | | Day 1 | Day 2 | Day 3 |
| 1 | 1.07 | 13.34 | 13.48 | | |
| 2 | 36.35 | 42.36 | 66.55 | | |
| 3 | 26.14 | 15.46 | 20.93 | | |
| 4 | -13.85 | 17.66 | 15.09 | 15.94 | |
| 5 | -8.00 | 25.03 | 23.16 | | |
| 6 | 2.45 | 26.88 | 27.55 | | |
| 7 | 49.27 | 36.53 | 72.02 | | |
| 8 | 55.83 | 62.77 | 142.09 | | |
| 9 | 11.70 | 11.40 | 12.91 | | |
| 10 | 11.47 | 11.18 | 12.63 | | |
| 11 | 14.86 | 17.36 | 20.64 | 19.81 | 20.71 |
| 12 | -59.43 | 30.99 | 19.44 | | |
| 13 | -32.35 | 35.16 | 26.57 | | |
| 14 | 18.04 | 11.51 | 14.04 | | |
| 15 | -46.56 | 33.42 | 21.97 | 23.30 | 23.14 |
| 16 | 26.08 | 14.86 | 19.24 | 20.23 | 20.85 |

Table 3: Powered-off condition cumulative prices and marginal values.

| Subject | MV (%) | Exo Powered Off (\$) | Walking No Exo (\$) |
|---------|---------|----------------------|---------------------|
| 1 | -0.53 | 13.56 | 13.48 |
| 2 | -26.44 | 84.15 | 66.55 |
| 3 | -3.53 | 21.31 | 20.59 |
| 4 | -0.10 | 15.10 | 15.09 |
| 5 | -7.40 | 24.88 | 23.16 |
| 6 | 17.44 | 22.75 | 27.55 |
| 7 | -40.51 | 199.66 | 142.09 |
| 9 | -26.64 | 16.34 | 12.91 |
| 12 | -119.94 | 42.76 | 19.44 |
| 13 | -111.95 | 56.31 | 26.57 |

Table 4: Cumulative prices for participants who repeated the walking-no-exo condition

| Subject | Average Cumulative Price (\$) | St. Dev. Cumulative Price (\$) | St. Dev. Cumulative Price (% of Average) | Prices (\$) | | |
|---------|-------------------------------|--------------------------------|------------------------------------------|-------------|-------|-------|
| | | | | Day 1 | Day 2 | Day 3 |
| 4 | 15.52 | 0.43 | 2.75 | 15.09 | 15.94 | |
| 11 | 20.39 | 0.41 | 2.01 | 20.64 | 19.81 | 20.71 |
| 15 | 22.80 | 0.59 | 2.59 | 21.97 | 23.30 | 23.14 |
| 16 | 20.11 | 0.66 | 3.30 | 19.24 | 20.23 | 20.85 |

⁴³⁸ **1.2 Supplementary Tables**

⁴³⁹ **1.3 Supplementary Figures**

⁴⁴⁰ **Declarations**

⁴⁴¹ **Author's contributions**

⁴⁴² RLM, GCT, DM, and EJR designed the study; RLM performed the experiment and analyzed data with input from
⁴⁴³ GCT, DM, and EJR.; RLM, GCT, and EJR wrote the paper. EJR directed the study and all authors approved the
⁴⁴⁴ final version of the manuscript.

⁴⁴⁵ **Availability of data and materials**

⁴⁴⁶ All data needed to evaluate the conclusions of the paper are available in the paper or the Supplementary Materials.

⁴⁴⁷ **Funding**

⁴⁴⁸ Not applicable

⁴⁴⁹ **Competing interests**

⁴⁵⁰ The authors declare that they have no competing interests.

⁴⁵¹ **Ethics approval and consent to participate**

⁴⁵² All experiments were carried out with informed consent at the University of Michigan, with approval from the
⁴⁵³ Institutional Review Board of the University of Michigan Medical School (IRBMED).

⁴⁵⁴ **Consent for publication**

⁴⁵⁵ Not applicable

⁴⁵⁶ **Acknowledgements**

⁴⁵⁷ Not applicable

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532 **Acknowledgments**

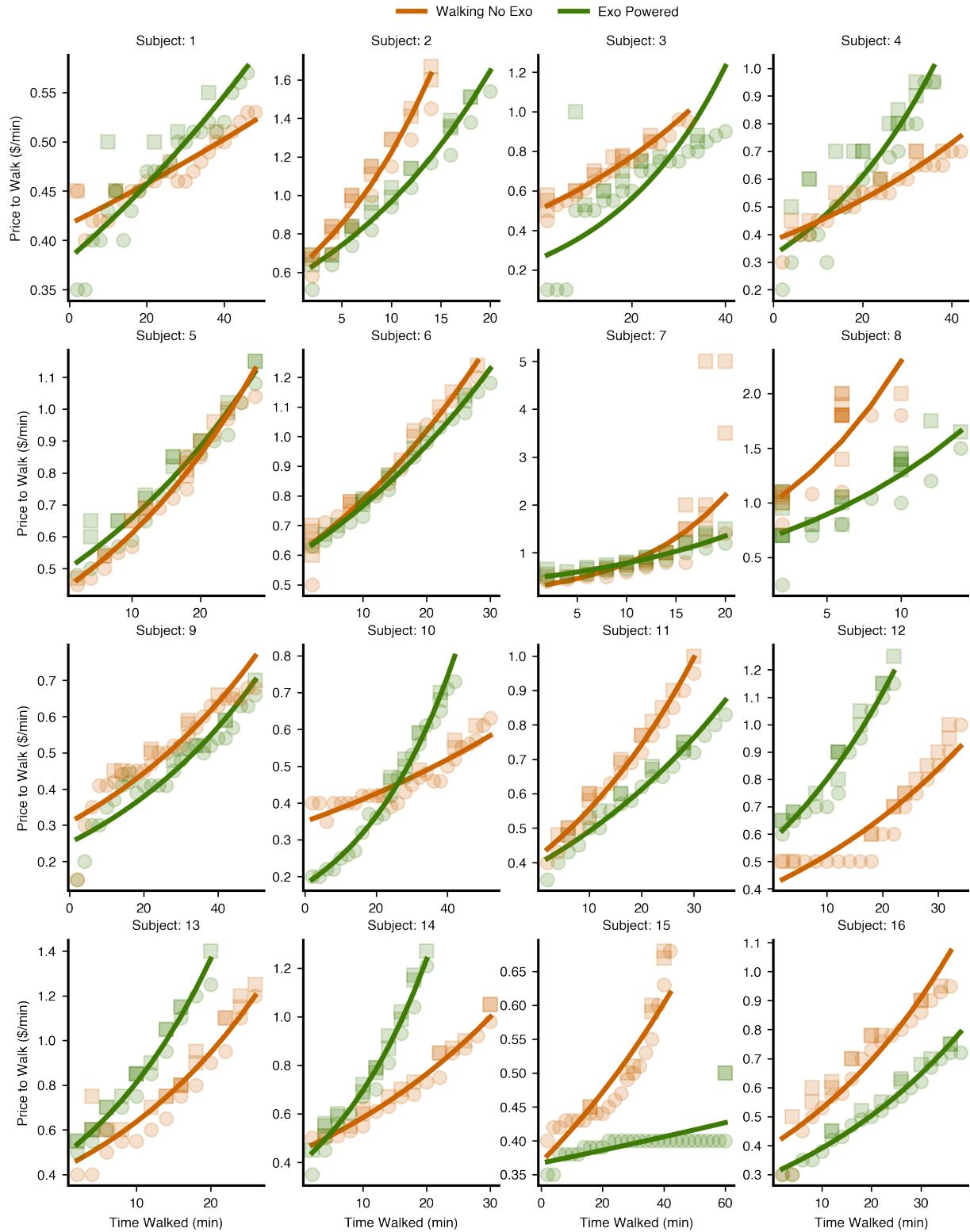


Figure 8: The price to walk curves for the exo-powered (orange) and walking-no-exo (teal) from all sixteen subjects. Circles denote winning bids, while squares denote losing bids.

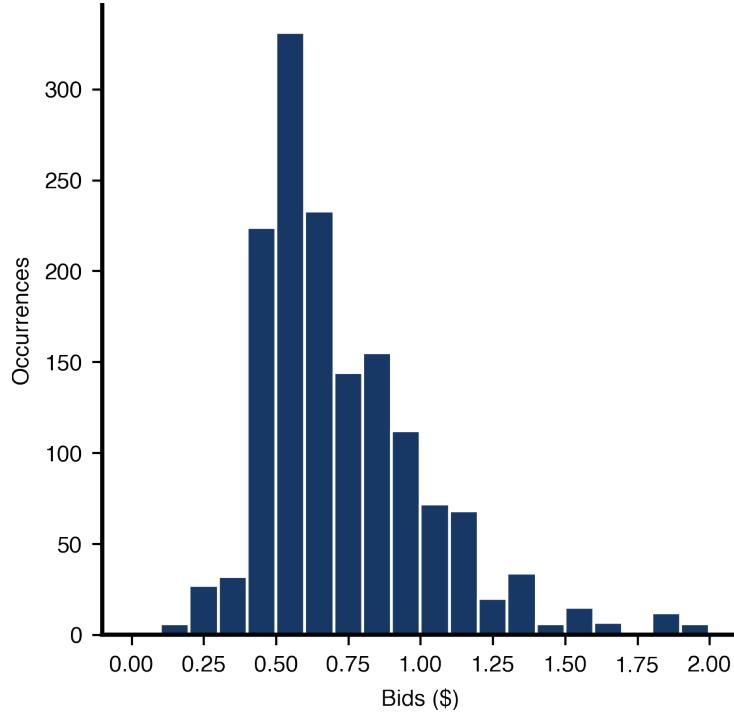


Figure 9: A histogram of all the subject bids for the two-minute intervals across all tested conditions. The average bid was \$0.75, with a standard deviation of \$0.39.

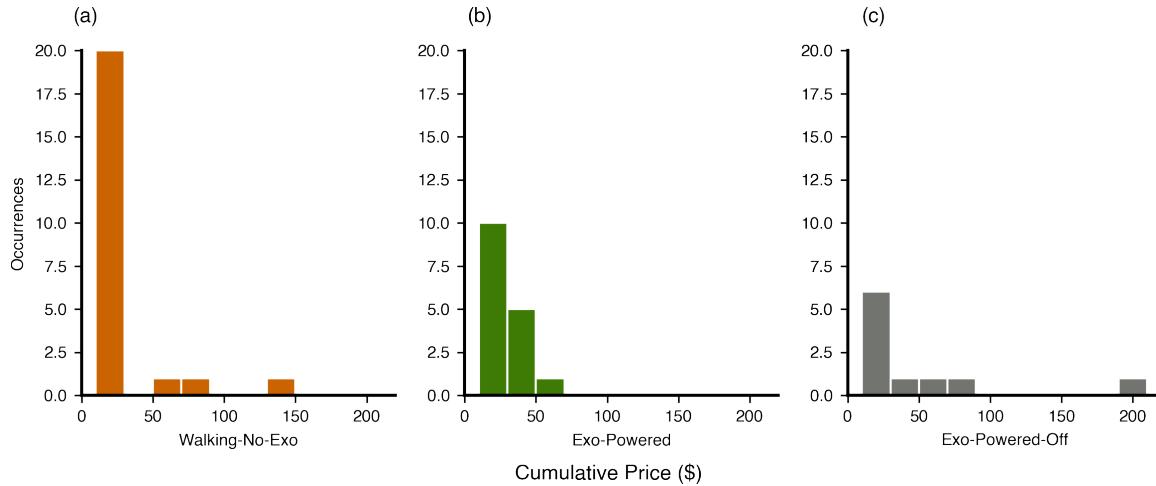


Figure 10: Histograms of the cumulative costs for (a) the walking-no-exo condition, (b) the exo-powered condition, and (c) the exo-powered off condition.

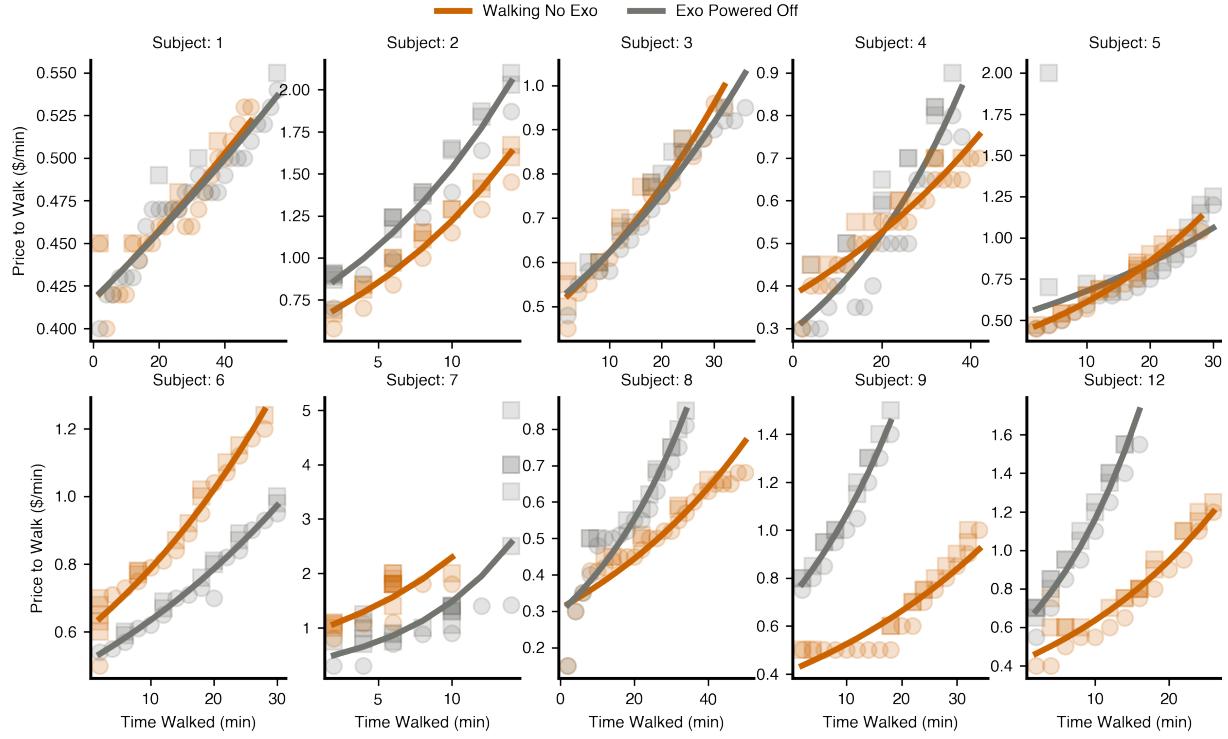


Figure 11: The price to walk curves for the exo-powered-off (gray) and walking-no-exo (teal) for the ten subjects who completed the former condition. Circles denote winning bids, while squares denote losing bids.

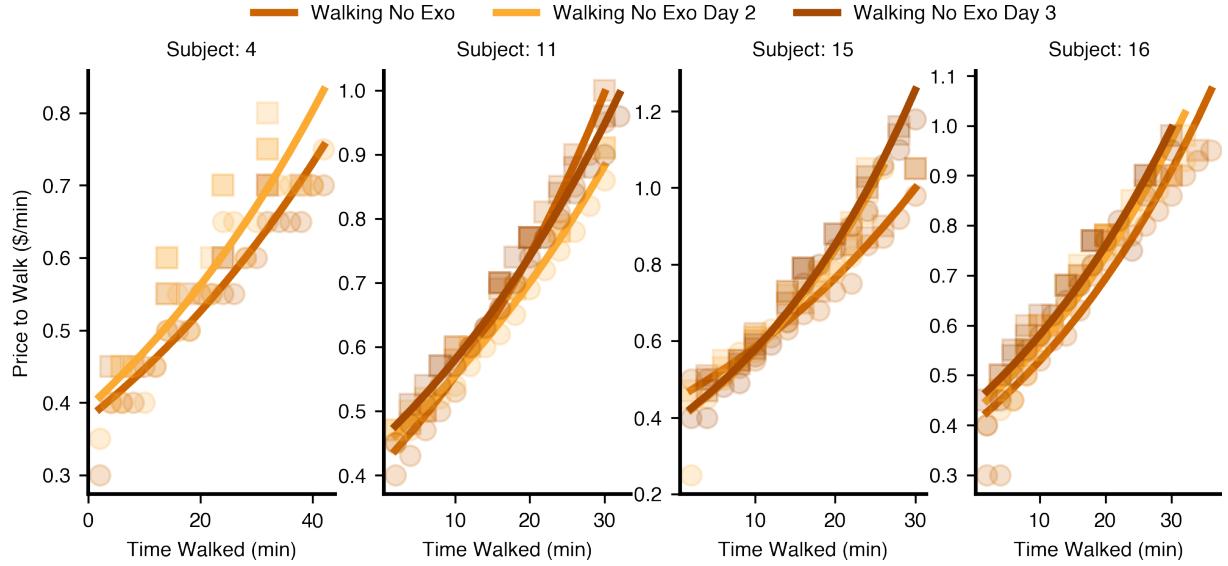


Figure 12: The price to walk curves for the walking-no-exo conditions across different days for the four subjects who repeated the walking-no-exo condition. Circles denote winning bids, while squares denote losing bids. Subject 4 only repeated the walking-no-exo condition once, while the rest of the participants repeated it twice.

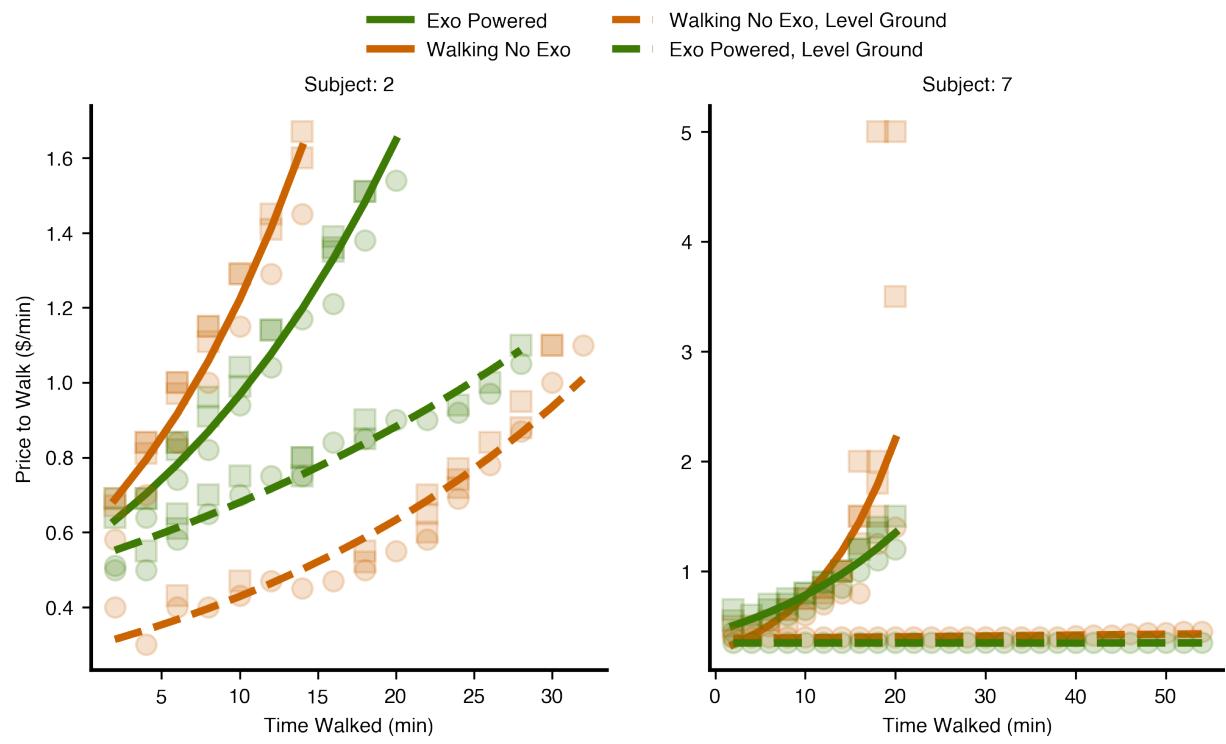


Figure 13: The price to walk curves for two subjects (2 and 7) that repeated the walking-no-exo and exo-powered conditions on a level ground condition. The solid lines denote the standard 10° incline task featured in this work; the dashed lines are the level ground condition. Both participants' MEVs decreased during the level ground condition.