

The Economic Value of Augmentative Exoskeleton Assistance

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The public must derive significant value from augmentative exoskeletons before they accept these technologies into their daily lives. While exoskeletons have already improved on millennia of evolution by aiding humans physiologically, particularly in the improvement of walking efficiency, it is unknown if people will perceive these benefits as valuable when compared to the costs of wearing these technologies. In this study, we quantified the economic value of the assistance from state-of-the-art ankle exoskeletons and compared it to the cost incurred from wearing the devices. We found that, while most participants found the assistance itself valuable, it did not significantly outweigh the cost of wearing the exoskeleton on average. These results suggest that modern augmentative exoskeletons may not yet provide significant value to the average person. This work also motivates the use of the economic value metric to directly guide exoskeleton design and control, and the need to identify why some individuals obtain substantial positive net value from exoskeletons to better maximize the translation of these technologies.

Introduction

Powered lower-limb exoskeletons have the potential to transform human mobility by extending the locomotor abilities of their wearers. These technologies augment lower-limb function by providing mechanical assistance to the joints of the legs in tandem with the human neuromotor system, and thus can make physically demanding tasks less challenging for both able-bodied and impaired individuals. Exoskeletons have been shown to reduce the caloric demands and muscular effort required for walking (1–9). Consequently, they may have beneficial implications for

21 users such as factory, military, and supply chain workers that perform physically demanding activities. In addition,
22 exoskeletons can reduce muscle activation, and thus may reduce joint loading (10–13), potentially extending the
23 physical capabilities of aging individuals. Rehabilitation-focused exoskeletons may also restore the mobility of
24 people with neuromotor deficits who face weakness and impairments in balance, coordination, and joint mechanics
25 following upper motor neuron disease (*e.g.* stroke). Exoskeletons can be used to assist the gait of these individuals,
26 with several commercially-available lower-limb exoskeletons having been approved by the U.S. Food and Drug
27 Administration.

28 Regardless of the exoskeleton’s use case, the metrics by which we assess exoskeletons drive their design, con-
29 trol, and potential impact. For rehabilitative applications, the design of these technologies has a clear physiological
30 objective: the restoration of impaired gait function. These exoskeletons are often considered successful if they
31 restore able-bodied kinematics or kinetics, which provide clear physiological goals on which to base exoskeleton
32 design and control decisions (8, 12, 14–16). However, for augmentative applications in which the user is typically
33 able-bodied, the metrics of success are less clear. Currently, augmentative exoskeletons are developed based on
34 their ability to meet a physiological objective; the “gold standard” for exoskeleton success is the reduction of
35 metabolic expenditure during locomotion (*i.e.* a reduction of the calories burned) (17). This objective is both
36 intuitively meaningful and objectively measurable (18, 19). Modern exoskeletons have reduced the metabolic rate
37 relative to unassisted walking by an average of approximately 14% (1–6, 20). This objective has led to the rise of
38 promising ‘human-in-the-loop’ (HILO) optimization techniques, which directly modulate exoskeleton assistance
39 based on the metabolic reductions experienced by the wearer (21–27). Recent work has established experimen-
40 tal infrastructure that illustrates the tight coupling between metrics of success and exoskeleton development. An
41 example of this infrastructure includes tethered emulator systems (7, 28), whose purpose is to inform exoskeleton
42 design and control based on their ability to reduce the metabolic expenditure of their wearer. Numerous studies
43 investigating the biomechanical underpinnings of metabolic cost reductions have also been conducted to find more
44 optimal exoskeleton assistance settings (20, 22, 29, 30). Finally, other metrics by which exoskeletons are developed
45 often relate back to metabolic rate. These metrics include net joint torque reduction (11) and muscle activation
46 reduction (10, 12), which are commonly used, easier-to-measure proxies for improvement in energetics.

47 Though the physiological benefits of exoskeletons have been demonstrated using metabolic metrics, these
48 benefits have not translated to wearer perception of enhanced endurance and strength. These perceptions are im-
49 portant, as for augmentative exoskeletons to reach their potential in society, users will need to voluntarily accept
50 these technologies into their lives. They thus must be developed to provide a perceivable ‘subjective’ benefit to

their wearer, in addition to objective assessment of their value. Our recent work has shown that during exoskeleton-assisted walking, the average user cannot yet perceive the benefit of most systems available today (31–33). That is, metabolic rate needed to be reduced by 23% ($N = 10$) before exoskeleton users could reliably perceive this improvement, (whereas most modern exoskeletons reduce the wearer’s metabolic rate by 14% compared to unassisted walking (17)). These results agree with prior studies that showed humans were relatively insensitive to small changes in exertion in other exercise contexts (34). Intuitively, if the user is unable to perceive the metabolic reduction provided by an exoskeleton, this value is unlikely to be incorporated into decision-making during exoskeleton use, translation, and adoption. Consequently, assessing and developing exoskeletons based on reductions in metabolic rate could result in systems that are not perceived as valuable by users, despite significant energetic benefits (3, 23, 30, 35–37).

These assessments of perception were measured using psychophysical techniques, which demonstrate that the human is a poor sensor of the physiological benefits from exoskeletons. An alternative method for measuring success in exoskeleton development is to quantify the perceived *economic value* provided to the wearer during its use. Economic value, measured in monetary currency (e.g. US dollars), is assigned by the wearer and can reflect the multifaceted nature of exoskeleton user experience. While exoskeletons can provide assistance that improves energetics, that assistance often comes at a cost to the wearer. Exoskeletons can add discomfort, weight, and audible noise, in addition to having aesthetic implications. If the user is able to assign economic value to the experience of exoskeleton use, they are able to inherently balance and quantify these trade offs. Therefore, exoskeletons that maximize economic value may have a greater likelihood for adoption and use. Prior work in management science has established that the perceived value of different technologies has a significant impact on user intent to adopt those technologies into their daily lives (38–42). When potential exoskeleton users, manufacturers, and others are weighing the choice to adopt or purchase an exoskeleton, the consciously perceived benefits must outweigh these costs. Thus, the perceived economic value of exoskeleton use is a potentially powerful metric for designing and controlling exoskeletons that quantifies meaningful, individualized benefits to wearers.

In this study, we introduce a tool for measuring the perceived economic value of exoskeleton use as a metric to evaluate their performance and user experience. We define and use this tool to quantify the economic value—termed *Marginal Value (MV)*—provided by bilateral ankle exoskeletons use during uphill walking. We leveraged the Vickrey second-price auction to measure participants’ “price to walk,” which was then aggregated and compared across conditions to obtain our economic value metric (*i.e.* MV). Our experiment revealed that while there was insignificant ($p = 0.24$) positive value of exoskeleton use across subjects, there was a large disparity between

81 subjects. Some subjects reported substantial value provided by the exoskeletons, which represents an opportunity
82 target these “responders” in future work. MV offers advantages over the more common metabolic rate metric
83 including its accessibility, in that it does not require specialized equipment, and its intuitiveness, as users and
84 manufacturers are more likely to understand the value of monetary currency over biomechanical quantities (*e.g.*
85 calories burned or muscle power). Our approach is also generalizable, and can be used to measure the value of not
86 only different types of exoskeleton assistance, but also various technologies, activities, or experimental conditions.

87 **Background on Vickrey auctions**

88 Within the field of economics, the Vickrey second-price auction (43) provides a well-studied incentive structure for
89 quantifying the economic value of abstract concepts. In a seller’s Vickrey auction, participants compete via bidding
90 to sell a good. The winner of the auction is the participant that bids the lowest amount to sell the good. However,
91 the winner will earn the amount bid by the second-lowest bidder. The Vickrey auction is also “sealed-bid” in that
92 each participant’s bids are not revealed publicly. This second-price paradigm incentivizes subjects to bid their
93 true perceived value for the good (43–45) rather than attempt to guess (and slightly underbid) all other bidders’
94 lowest bids. Auction participants would be discouraged from bidding higher than what they believe the good is
95 truly worth, for fear of losing the auction to a competitor with a lower bid. Conversely, they will have no incentive
96 to underbid below their true value since they would risk selling for a “loss” if they won the auction. Thus, the
97 Vickrey auction provides a quantitative measurement for the value of arbitrary goods or services by measuring an
98 individual’s willingness to buy or sell (45, 46). Prior researchers have also used Vickrey auction metrics to measure
99 the value of abstract concepts or actions, such as food safety (47), GMO-free foods (48), the stigma resulting from
100 HIV (49), personally identifiable information (50), and smartphone battery life (51). In particular, Coursey *et al.*
101 employed the Vickrey auction to quantify the willingness of participants to endure performing an unpleasant task,
102 such as tasting a bitter liquid (52). These examples highlight the promise of the Vickrey auction to quantify the
103 perceived economic value provided by experiences, including exoskeleton use.

104 **Results**

105 Within each trial, participant bids invariably trended upwards as participants became more fatigued and this trend
106 was well represented by a first-order exponential. During each trial, participants either walked with normal walking
107 shoes (**“walking-no-exo”**), with the exoskeleton applying assistance (**“exo-powered”**), or with the exoskeleton
108 donned but applying no power (**“exo-powered-off”**). Across all three walking conditions, the average bid for

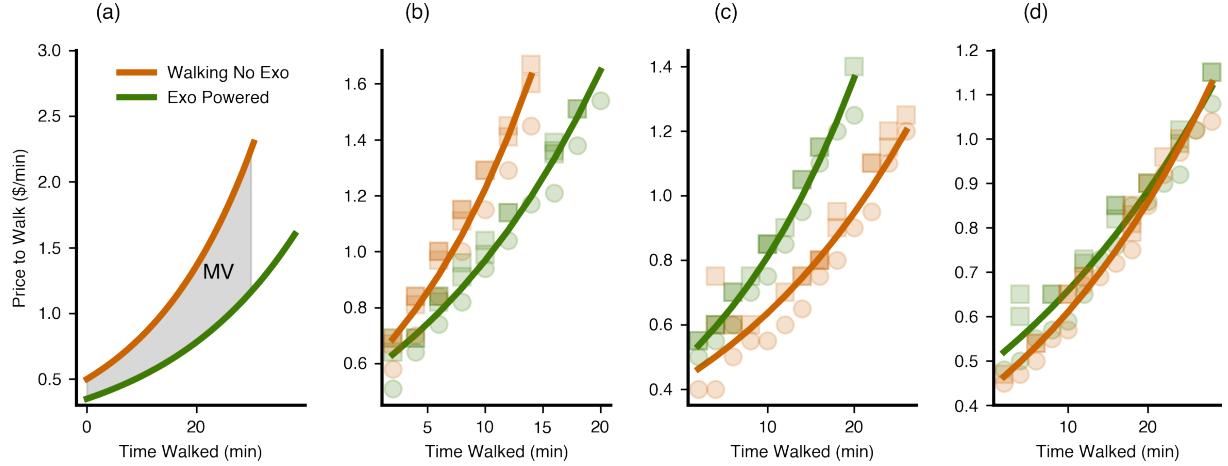


Figure 1: (a) Representative price to walk curves for the walking-no-exo condition (green) and the exo-powered condition (orange) from different subjects. The marginal value (MV, the value of the exoskeleton assistance) is given by the area between the curves. In actual trials, the price to walk curves 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.

each two-minute interval was \$0.75, with a standard deviation of \$0.39. The maximum bid was \$5.00, while the minimum bid was \$0.10. As participants continued to walk during each trial, their bids increased at varying rates (Fig. 1). The first-order exponential model exhibited an R^2 of 0.87 ± 0.11 , averaged over all conditions 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. These outcomes were a clear economic benefit from the exoskeleton's assistance (Fig. 1b, positive MEV), a clear economic penalty (Fig. 1c, negative MEV), and a negligible economic effect (Fig. 1d, near-zero MEV).

Across subjects, the MV of the exoskeleton plus its 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. We quantified these MVs for each participant (Fig. 2a) while using a commercial ankle exoskeleton, similar versions of which have been shown to reduce metabolic rate (36). The average inter-subject MV of exoskeleton use was 5.82%, with a standard deviation (SD) of 31.14% (N=16). The exoskeleton plus assistance thus provided only a small value benefit to the average participant. Using a t-test, the average MV of exoskeleton plus assistance (5.82%) was not significantly different from zero ($p = 0.24$). However, as denoted by the high standard devia-

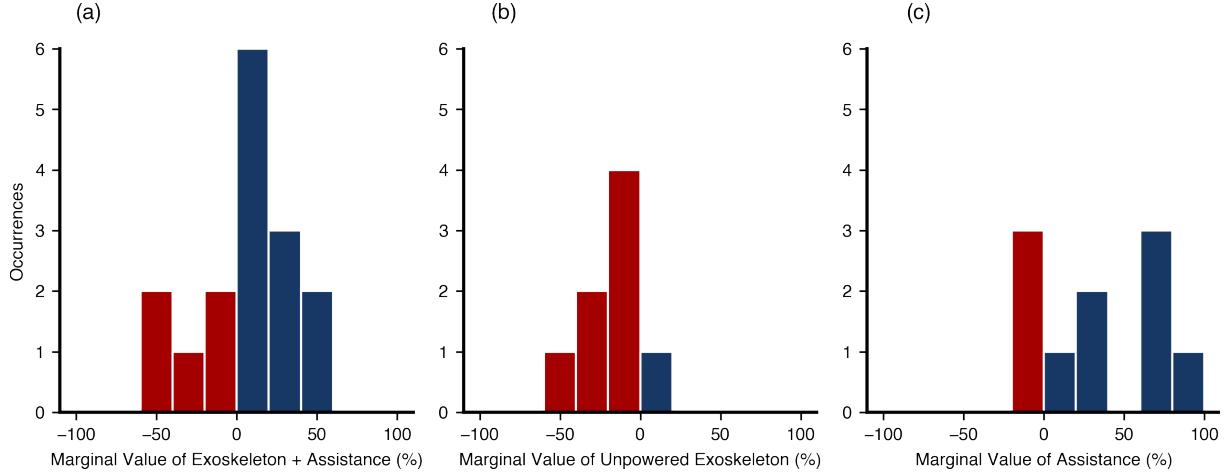


Figure 2: Histograms of the Marginal Values (MVs) for the different conditions. Positive MVs, indicating that value was added to the wearer are in blue, while negative MVs, indicating that costs were imposed on the wearer, are in red. (a) The MVs of exoskeleton plus assistance for all sixteen subjects. 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 plus its assistance was not significantly positive, the assistance alone conferred a significant benefit.

126 some participants received large benefits from the device's assistance, while others experienced an economic
 127 penalty from exoskeleton use. The average MV for the unpowered exoskeleton was -31.8% with a standard de-
 128 viation of 45.0% (N=13, 2b). Using the same t-test as in the exo-powered condition, we found this change to
 129 be significantly different from zero ($p = 0.03$). Additionally, the powered assistance alone from the exoskeleton
 130 provided a significant increase in value (mean: 33.8%, SD: 38.1%, $p = 0.01$, N=13, Fig. 2c).

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

136 We also characterized the repeatability of our measurements with a subset of participants to verify that the
 137 changes in price to walk detected due to the changing walking conditions were attributable to the difference in
 138 perceived difficulty between conditions and not due to participant day-to-day variability (a representative repeater
 139 trial is shown in Fig. 3). For the four subjects that repeated the walking-no-exo condition at least once, the average
 140 intra-subject standard deviation of the cumulative price for the walking-no-exo condition was 3.38%, expressed as

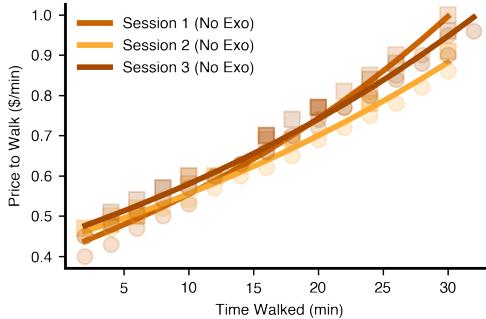


Figure 3: Representative price-to-walk curves for a single 'repeater' subject across different days. The agreement across the curves demonstrates the test re-test reliability of our approach. These sessions were completed over a seven day span.

¹⁴¹ a percentage of each subject's average walking-no-exo cumulative price. This quantity represents the day-to-day
¹⁴² fluctuation in the value demanded to walk, in addition to any user or experimentally derived noise.

¹⁴³ Discussion

¹⁴⁴ In this work, we use the Vickrey second-price auction as a method to capture the economic value or detriment
¹⁴⁵ provided by exoskeletons and their assistance. As part of our broader goal of emphasizing the user's role in the
¹⁴⁶ design, control, and evaluation of exoskeletons (32, 53), the intent of this work is to quantify the success of these
¹⁴⁷ technologies through a user-centered metric which encompasses the different aspects of the user experience, in-
¹⁴⁸ cluding exertion, weight, comfort, and assistance. To this end, the wearer specified their "price to walk" during a
¹⁴⁹ series of Vickrey auctions from which we quantified the economic value of different conditions. Our approach is
¹⁵⁰ particularly relevant for assessing the success of augmentative exoskeletons, which do not have a clear, clinically-
¹⁵¹ relevant, biomechanical or physiological objective (as in the case of prostheses or orthoses). Our work is motivated
¹⁵² by the belief that obtaining accessible and relevant metrics of success is critical to the successful development and
¹⁵³ adoption of exoskeletons in the real world. This motivation is supported by a history of strong interconnection
¹⁵⁴ between the assessment of exoskeletons and their design and control architectures (21–27). Our strategy quanti-
¹⁵⁵ fies success using monetary currency, as opposed to biomechanical or physiological quantities that may be more
¹⁵⁶ difficult for users, manufacturers, and those outside of the exoskeleton research field to interpret.

¹⁵⁷ Our results showed that the economic value provided by the exoskeleton assistance was comparable to the
¹⁵⁸ cost incurred by wearing the unpowered system. We obtained these results by quantifying the marginal value
¹⁵⁹ (MV) between two conditions, and by varying which two conditions were compared, we isolated the value of

160 the exoskeleton + assistance, the assistance alone, and the detriment of the unpowered exoskeleton. The average
161 cumulative cost of walking uphill for 30 minutes without the exoskeleton was \$29.23 (\$58.46/hr). The MV of
162 wearing the unpowered exoskeleton—that was not providing assistance—was -31.8%, which translates to a mon-
163 etary cost of \$18.59/hr for wearing the unpowered system. When assistance was applied by the exoskeleton, the
164 MV increased to just above zero (\$3.40/hr) when compared to not wearing the exoskeleton. The marginal value
165 added by the assistance alone was 33.8%, which translates to an added value of \$19.76/hr. Thus, the assistance
166 applied by the exoskeleton offset the cost of wearing the system, and the net benefit was modest but positive.
167 These results suggest that modern augmentative ankle exoskeletons may not provide a substantial benefit to their
168 wearers during short-term uphill walking. This is particularly surprising because of the lightweight, refined design
169 of the exoskeletons used, and the high physiological demands of the uphill walking task (54, 55), which we chose
170 to increase the observed value of the exoskeleton (*i.e.* to improve the signal-to-noise ratio in our measurement of
171 economic value).

172 Our approach and metrics can be used to guide the development of exoskeletons in new ways. Exoskeleton
173 controllers can potentially be directly engineered to maximize the value added (in dollars) during assisted walking.
174 That is, the MV obtained from the comparison of the unpowered exoskeleton and exoskeleton + assistance quanti-
175 fies this value. Similarly, exoskeletons can themselves be designed to minimize the economic cost they impose on
176 the wearer, which includes not only weight, but also other factors such as inertia and discomfort. This information
177 can be obtained using the MV of the unpowered exoskeleton when compared to the walking-no-exo condition. Ad-
178 ditionally, this study suggests that exoskeleton development may be improved by identifying the users who innately
179 receive greater economic value from wearing the system. We found that nine participants obtained substantial pos-
180 itive economic value from wearing the system ($27.8\% \pm 16.3\%$, corresponding to $\$16.25/hr \pm \$9.53/hr$), even
181 though the average across all participants was lower. Such “responders” may be more willing to adopt exoskeletons
182 in the face of potential drawbacks, including cost and added mass, among others. Future investigations could focus
183 on the separation of these individuals *a priori* based on human factors, such as fitness level or perceptual abili-
184 ties (32). Furthermore, investigations of whether users can be “trained” to receive a greater economic value from
185 the exoskeleton are promising avenues of future study. Understanding why some users obtain this value would
186 enable the targeting of these individuals to maximize the translation and impact of augmentative exoskeletons.

187 Participant results were consistent when repeated across days, supporting the quality of our measurements.
188 Four participants repeated the walking-no-exo condition across several days, and the inter-day variance of the
189 cumulative prices was low (\$0.67, Table 4). This consistency supports the ability for our approach to quantify

perceived changes in value, with potentially minimal effect from inter-day confounding factors. Using this inter-day variance, we estimated the Minimum Detectable Change (MDC), defined as the minimum change in value not caused by chance. The MDC for the MV measurement was 9.4% (95% confidence interval, difference between two measurements, standard deviation 3.4%). To identify which subjects had a noticeable change in value, we compared each participant's MV against the inter-day variability's 95% confidence interval ($\pm 9.4\%$); nine of 16 subjects had MVs exceeding this threshold in the positive direction, while four subjects exceeded it in the negative direction. An extension of this result is that we expect participants with a net-negative MV would rather pay money than wear an assistive ankle exoskeleton during uphill walking.

Our auction strategy avoids several potential limitations of more direct methods to quantify value, such as a Likert scale test, direct feedback, or auctions without a real monetary payout (56). The locomotion task implemented in our approach does not necessitate extreme fatigue and can remain relatively short, avoiding confounding factors (e.g. boredom or opportunity cost). Additionally, the incentive structure of the auction links participant bids to a specific consequence (e.g. walking uphill), and are thus less prone to biases, such as self-enhancement or social desirability, while still remaining intuitive (50, 57).

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

We anticipate that the conclusions drawn in this study extend to other augmentative ankle exoskeletons, although further study is needed to quantify the value from other technologies. The ankle exoskeletons from this work were developed for commercial use, and represent a "best-in-class" technology; the system's refined design is lightweight, untethered, and provides substantial net-positive energy during each stride. We expect that other, similar exoskeletons would have comparable marginal values, and thus may also not provide a substantial economic benefit to their wearers. Additionally, although the task of uphill walking does not represent all possible uses for augmentative exoskeletons, it enables an opportunity to quantify value provided by the exoskeleton during an application where it can provide substantial benefit. We expected that the increased energetic difficulty of uphill walking (54, 55) would enable the exoskeleton to more readily demonstrate its value to the wearer. Furthermore, we chose the uphill walking task to reduce experimental duration as the greater difficulty would cause participant

220 bids to rise more quickly. We expect that for less strenuous tasks, exoskeletons will show reduced marginal value;
221 this hypothesis is supported by the results of a separate study we conducted (see Supplementary Information) in
222 which the MVs of the exoskeleton + assistance were lower for level-ground walking when compared to uphill
223 walking.

224 The changes in exoskeleton conditions caused intuitive changes in value, supporting the validity of our eco-
225 nomic value metric. That is, the different exoskeleton conditions (walking-no-exo, exo-powered-off, and exo-
226 powered) likely drove the changes in value measured, and these changes agree with the biomechanical demands
227 of the different conditions. The MV of the exo-powered-off condition was strongly negative, denoting a cost to
228 wearing the system. This result is expected, since without assistance, wearing an exoskeleton is akin to wearing
229 shank weights during locomotion, which would increase task difficulty by increasing the metabolic cost (by \sim 16.8
230 W based on the results in (36, 58)). Additionally, when the assistance was added, the value increased to just above
231 zero, indicating the assistance was useful in offsetting the challenge of wearing the unpowered system during uphill
232 walking.

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

241 The use of simulated bidding agents may have influenced the results but we expect the overarching conclusions
242 of the study would not be affected. In our auction protocol, we used simulated bidding agents (“robo-bidders”)
243 to model human behavior, rather than implementing our approach with multiple human subjects. The intent for
244 this choice was to reduce the logistical challenges of our approach, which would otherwise have required multiple
245 treadmills and exoskeletons. To mitigate any effect of the simulated nature of the other participants, the human
246 participant was informed that the robo-bidders were humans participating in the experiment in remote locations.
247 The behavior of the robo-bidders—defined by a parameter that governs the rise and fall of their bids—was derived
248 from pilot data obtained from human participants. To reduce the likelihood that the human participants could infer
249 the robo-bidder behavior model, the robo-bidder bids were corrupted with noise. The use of robo-bidders also

250 enabled us to standardize the interaction between the human participant and the other auction participants. Any
251 series of auctions would naturally establish an equilibrium between the participants; thus, by using robo-bidders,
252 we were able to control for this equilibrium, which strengthens our ability to compare across subjects. In addition,
253 the use of robo-bidders enables our results to more readily be compared across other researchers, institutions, and
254 exoskeletons that are assessed using comparable methods.

255 During the experimental protocol, participants responded with bids as instructed, but we are unable to know
256 for certain that their bids were truly honest (*i.e.* truthfully reflecting their internal sense of value). Our study relies
257 on participants honestly reporting their bids, which is theoretically guaranteed as an optimal strategy by the nature
258 of the Vickrey second-price auction for rational actors (43). In addition to the natural structure of the Vickrey
259 auction, which incentivizes truthful bidding, we minimized the potential influence of dishonest bidding by setting
260 the auction interval to two minutes (as opposed to a shorter duration). We chose the two minute duration to increase
261 the effort required, and thus minimize potentially 'dishonest' exploratory bidding, which would corrupt the price
262 to walk curves. Additionally, subjects continuously received verbal instructions to always bid honestly as the
263 best strategy, which has been demonstrated to increase the likelihood of honest bidding (59). Finally, participants
264 received the expected monetary compensation that resulted from their winning bids, which similarly incentivizes
265 truthfulness.

266 Conclusion

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

276 **Methods**

277 **Participants**

278 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 informed consent before participation. The study protocol was approved and overseen by the Institutional Review Board of the University of Michigan Medical School. Participants had no prior experience walking with the bilateral ankle exoskeletons featured in this study.

283 **Experimental Protocol**

284 **Exoskeleton Apparatus**

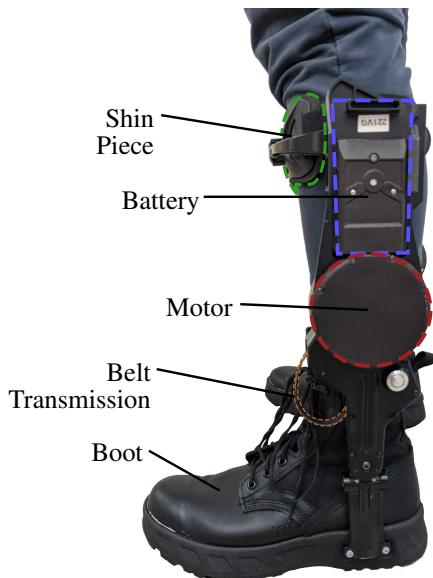


Figure 4: The Dephy ExoBoot used in this experiment. An electric motor applies plantarflexion torques at the ankle via a belt transmission.

285 Our approach quantifies the value of bilateral ankle exoskeletons that were designed to improve the energetic
286 efficiency of human walking (ExoBoot, Dephy Inc. Maynard MA, Fig. 4). The commercially-available system
287 utilizes an onboard brushless electric motor and flat cable transmission (for a mean transmission ratio $\sim 15:1$,
288 see Supplementary Fig. 6.) to apply ankle assistance during walking. The exoskeletons have a single powered

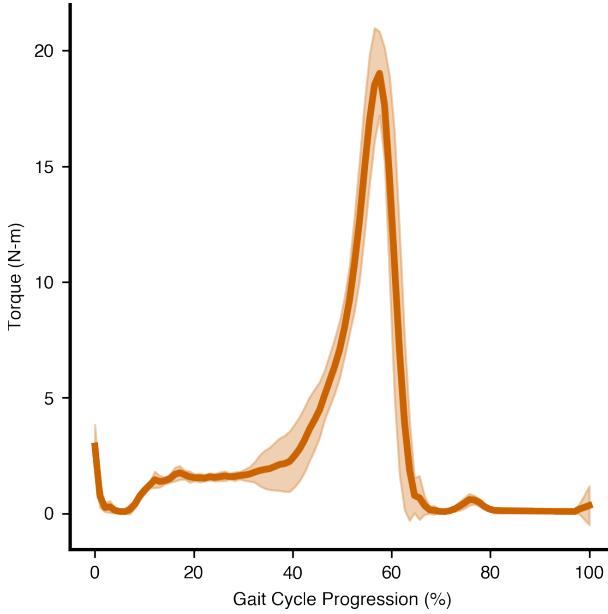


Figure 5: The torque profile applied by the bilateral ankle exoskeletons during the uphill walking task. The torque applied stems from the proprietary control strategy implemented by Dephy Inc and represents the state of the art. The mean profile is shown in solid green, with a single standard deviation shown by the shaded region. Gait cycle progression is defined as beginning and ending at sequential ipsilateral heel-strokes.

289 degree of freedom (dorsi-plantar flexion) and a passive, unactuated degree of freedom (inversion-eversion). The
 290 transmission is unidirectional, which enables the system to apply plantar flexion assistance torque but it cannot
 291 provide dorsiflexion assistance. Each side of the exoskeleton applies a torque profile (Fig. 5) that provides a burst
 292 of positive power during the terminal stance phase of walking, augmenting the propulsive effort provided by the
 293 triceps surae. During swing phase, the exoskeleton is able to add slack to the belt drive, thereby preventing any
 294 unwanted resistance to the foot; this capability stems from the unidirectional nature of the design. Gait progression,
 295 inferred from heel-strike events, was used to schedule how assistance was provided during each step. Similar
 296 exoskeletons have been shown to lower the user's metabolic expenditure during walking as well as reduce the
 297 biomechanical power requirements at the ankle and other joints of the legs (29, 36). The walking assistance
 298 controller was developed by Dephy Inc. as part of their commercial system. As one of the first commercially
 299 available exoskeletons, the Dephy Exoboot was chosen for its ease of use, robustness, and representation of the
 300 state-of-the-art.

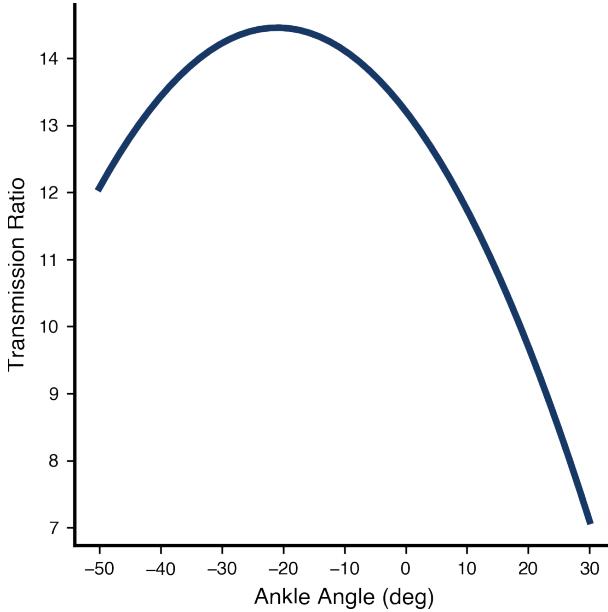


Figure 6: The transmission ratio curve of the exoskeleton device used in this experiment. Positive ankle angles denote dorsiflexion and negative angles denote plantarflexion. The curve was modeled as a second-order polynomial, and was obtained by taking the derivative of the best-fit third order polynomial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

301 **Vickrey Auction Protocol**

302 The Vickrey auction (43) is a powerful economic tool for determining the true value placed on goods or actions. In
 303 this type of auction, participants compete to purchase (or sell) a good or item. For each participant, the auction's
 304 structure is designed such that the optimal strategy for obtaining the item is to truthfully represent their internal
 305 value with their bid (*e.g.* to bid an amount equal to the true worth of the item). This optimality stems from the
 306 *second-price* nature of the auction (43–45), in which the winner is the participant who bids the highest (or lowest,
 307 as in the selling implementation used in this study). However, rather than paying the highest bid, the winner of the
 308 auction instead pays the *second-highest* bid. Participants do not know the value of competing bids (sealed-bid). In
 309 theory, bidders are disincentivized to bid less than their true value, as they run the risk of not winning the auction
 310 (not acquiring the good). Similarly, participants should not bid more than their true value, which otherwise could
 311 cause them to pay more than the value of the item. The inverse is also true for the case of selling an item (second-
 312 lowest bid is paid, lowest bid wins); thus, the incentive structure that elicits truthful bidding still holds in the seller's
 313 auction. Due to the presence of this optimal strategy, the Vickrey auction provides a method for quantifying the
 314 value of abstract concepts (45, 46). In our protocol, we use the Vickrey auction sequentially to repeatedly sample

315 individuals' valuations of their time during uphill walking. The participants bids across the sequential auctions
 316 denotes the wearer's "price to walk" curve.

317 **Walking Protocol**

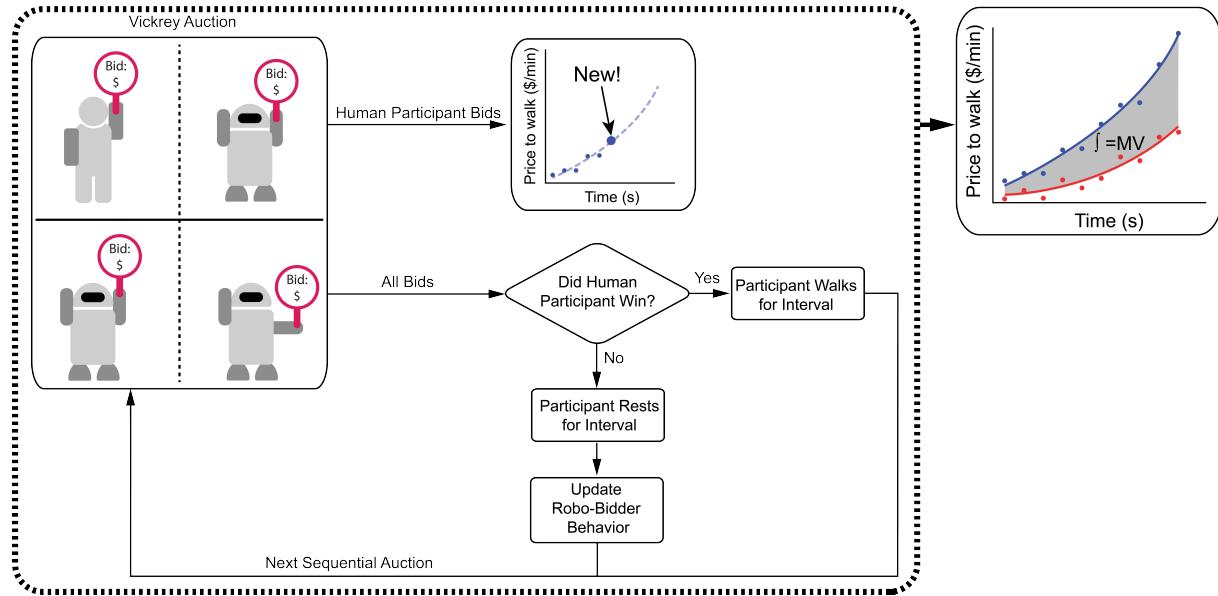


Figure 7: The experimental Vickrey auction protocol. In this protocol, human participants compete against computerized bidders (depicted in the 2×2 grid) in sequential Vickrey auctions for the prize of monetary compensation in exchange for undergoing a strenuous locomotion task. A human participant, 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. 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, depicted in the upper row) and while receiving powered assistance from the exoskeleton ('exo-powered', depicted in the lower row). In this example depicting a single auction, the presence of the exoskeleton added value to the human and thus reduced their price to walk; this then led to them bidding lower than the robot, rendering them the victor and causing the robot to suffer a loss. The integral of the difference between the prices-to-walk for the no-exo and exo-powered conditions is the value added from the device's assistance.

318 Participants received guidance on the Vickrey auction protocol prior to the experiment. During this time,
 319 participants read a lay explanation of the Vickrey auction. Subsequently, participants underwent two mock Vickrey
 320 auctions, where the optimal strategy of truthful bidding was explained. Subjects were also repeatedly informed of

321 the optimality of the honest-bid strategy to improve the number of honest bids provided (59). In the first mock
322 auction, participants were presented with a miscellaneous office supply item and told to bid honestly on it as if they
323 were competing to purchase it in a real Vickrey auction. Each participant wrote down their bid on an index card,
324 while the researcher wrote down some artificial competing bids. The participant then revealed their bid, and the
325 experimenter explained which bid won (the highest bid) and what the payout would be (the second-highest bid).
326 Next, participants were told to imagine they were participating in the actual study, in which they would compete
327 to sell their walking time in exchange for monetary compensation. Participants were prompted to consider bidding
328 based on an hourly wage, although they were also told that any truthful bid would be accepted. Again, participants
329 wrote down their bids while the researcher aggregated the competing bids. As in the previous mock auction, the
330 participants revealed their bids, and the experimenter walked them through which bid won and the payout.

331 Following familiarization with the exoskeleton and auction mechanics, participants competed in sequential
332 Vickrey auctions to sell their time while walking uphill. The protocol was organized into a set of Vickrey auctions
333 experienced in series, each lasting two minutes. If the participant won the auction, they would walk for the two
334 minute interval, and they would rest if they lost the auction. During walking, participants walked on a split-belt
335 treadmill with a 10° incline (Fig. 7). Unlike in the mock auctions, the participants would accrue and receive
336 monetary compensation according to the Vickrey auction protocol. At the end of each interval, the winning bid
337 (second-lowest bid) was revealed. The sequential auctions lasted a randomly-specified duration between 50 and
338 70 minutes, with a mean duration of 60 minutes; participants walked on average a total of 31.3 minutes *pm* 11.5
339 minutes. The duration uncertainty was added to discourage subjects from waiting until their bidding competitors
340 were “exhausted” to inflate payouts. The two-minute interval duration was chosen to minimize subject exploratory
341 behavior. In total, participants experienced approximately 25 - 35 Vickrey auctions in series. By aggregating
342 each subject’s bids across the series auctions, the resulting price to walk curve captured the participants’ valuation
343 of their time to complete the experiment, which can be used to provide insight into the value provided by the
344 exoskeleton in this application.

345 Our protocol was broken up across several days to reduce the effect of fatigue from the experiments. The order
346 of the testing was randomized across participants. On one day, subjects completed the Vickrey auction protocol
347 with normal shoes (*i.e.* the walking-no-exo condition); these data establish their baseline price to walk curve and
348 cumulative price. On a different day, subjects instead walked with powered exoskeleton assistance (*i.e.* the exo-
349 powered condition). Participants were not explicitly told that the exoskeleton would provide assistance; instead,
350 they were told that the exoskeleton was applying a randomized torque profile that would either help or hinder their

351 locomotion. This was done to reduce the chance that participants would experience a placebo effect and perceive
352 the exoskeleton as valuable due to information provided by the experimenter. By comparing the price to walk
353 curves from the walking-no-exo and exo-powered conditions, we obtained a measurement for the value provided
354 by the exoskeleton in this task. If undergoing the exo-powered condition on the second day, subjects were given
355 time to re-familiarize and experience the exoskeleton's assistance.

356 In addition, each subject was randomly assigned to additional experimental sessions to investigate other at-
357 tributes our approach. Participants were randomly assigned to groups that either quantified the inter-day variabil-
358 ity of the price-to-walk measurements, or investigated the additional condition of walking with the exoskeleton
359 without assistance (*i.e.* the exo-powered-off condition). Due to the number of experimental sessions required,
360 participants were assigned to only one of these groups. Of the sixteen total participants, ten participants walked
361 in the exo-powered-off condition, while four participants repeated the walking-no-exo condition at least once (the
362 remaining two subjects were not able to participate in the additional sessions). Within the experimental group that
363 repeated the walking-no-exo condition, three participants repeated the condition twice more, while the last partic-
364 ipant repeated the condition once. The intent of investigating the exo-powered-off condition was to support that
365 the price-to-walk curves captured the cost of wearing the additional exoskeleton weight, when no assistance was
366 provided. The goal of repeating the walking-no-exo condition was to provide insight into the test re-test variability
367 of the price to walk curves and how they may be affected by inter-day confounding factors.

368 **Robo-bidders**

369 For a live Vickrey auction, multiple participants are needed; however, this constraint adds practical challenges in
370 coordination and logistics. To this end, we utilized computerized bidding agents (robo-bidders) as a substitute for
371 other human participants in the Vickrey auctions. The mechanics of the robo-bidders are modeled on the bidding
372 behavior of three pilot subjects. The robo-bidders bid with prices following a first-order exponential function of
373 the number of intervals spent walking. This model captured the effect of fatigue and increased their bids as the
374 robo-bidder won the auction. If the robo-bidder did not win the auction (meaning the human participant won),
375 their bid remained constant. Gaussian white noise (zero mean, standard deviation \$0.01) was added to corrupt the
376 robo-bidder bids, to reduce the likelihood that the human participant would intuit the robo-bidder model. Having
377 robo-bidders instead of human auction participants reduced the logistical difficulty of executing the Vickrey auction
378 experiments over time, while still replicating the behavior of a human participant. Naturally, the robo-bidder bids
379 competed in parallel with the human subject and approached an equilibrium behavior, which affected the total

walking time of the human participant. In the application of auctions in series, an equilibrium would also likely be established between multiple human participants. The human participants were not initially made aware of the fact that they were competing against computerized opponents. They instead were told that they were competing in live Vickrey auctions against other humans located remotely, with the experimenters communicating live. This was done to prevent the participants from attempting to learn and exploit the price to walk curves of the robo-bidders in an attempt to maximize profits beyond the strategy of honest bidding. Participants were debriefed on the true nature of the robo-bidders at the conclusion of their participation in the experiment.

Analysis

Marginal Value

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

$$Y(t) = k \cdot e^{(b \cdot c \cdot t)}, \quad (1)$$

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 factor equal to the participant's win-rate, and t is time. The scaling factor is necessary to control for the different durations of walking experienced by the participants (*i.e.* variations in the number of auctions won). After correcting for this win-rate, these curves are integrated and subtracted, yielding the area between the two curves. This area represents the marginal value added—in US dollars—from one exoskeleton condition compared to another.

The expression for the MV is as follows:

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

where k_1, b_1 correspond to the first exoskeleton condition in the comparison, k_2, b_2 correspond to the second condition, and t_1, t_2 are the bounds of the time domain for the integrals—0 and 30 minutes respectively, which roughly corresponds to the average time each participant walked in each trial. The MV is then equivalent to the value added or removed by the exoskeleton during a continuous thirty minute, uphill walking task. We commonly normalized the MV by the cumulative price from the walking-no-exo condition to obtain a percentage change.

⁴⁰³ However, this metric can be converted to compare any two experimental conditions or other candidate control
⁴⁰⁴ 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 |

405 1 Supplementary Information

406 1.1 Supplementary Text

407 1.1.1 Level-ground study

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

⁴¹⁷ **1.2 Supplementary Tables**

⁴¹⁸ **1.3 Supplementary Figures**

⁴¹⁹ **Declarations**

⁴²⁰ **Author's contributions**

⁴²¹ RLM, GCT, DM, and EJR designed the study; RLM performed the experiment and analyzed data with input
⁴²² from GCT, DM, and EJR.; RLM, GCT, and EJR wrote the paper. All authors approved the final version of the
⁴²³ manuscript.

⁴²⁴ **Availability of data and materials**

⁴²⁵ All data needed to evaluate the conclusions of the paper are available in the paper or the Supplementary Materials.

⁴²⁶ **Funding**

⁴²⁷ Not applicable

⁴²⁸ **Competing interests**

⁴²⁹ The authors declare that they have no competing interests.

⁴³⁰ **Ethics approval and consent to participate**

⁴³¹ All experiments were carried out with informed consent at the University of Michigan, with approval from the
⁴³² Institutional Review Board of the University of Michigan Medical School (IRBMED).

⁴³³ **Consent for publication**

⁴³⁴ Not applicable

⁴³⁵ **Acknowledgements**

⁴³⁶ Not applicable

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509 **Acknowledgments**

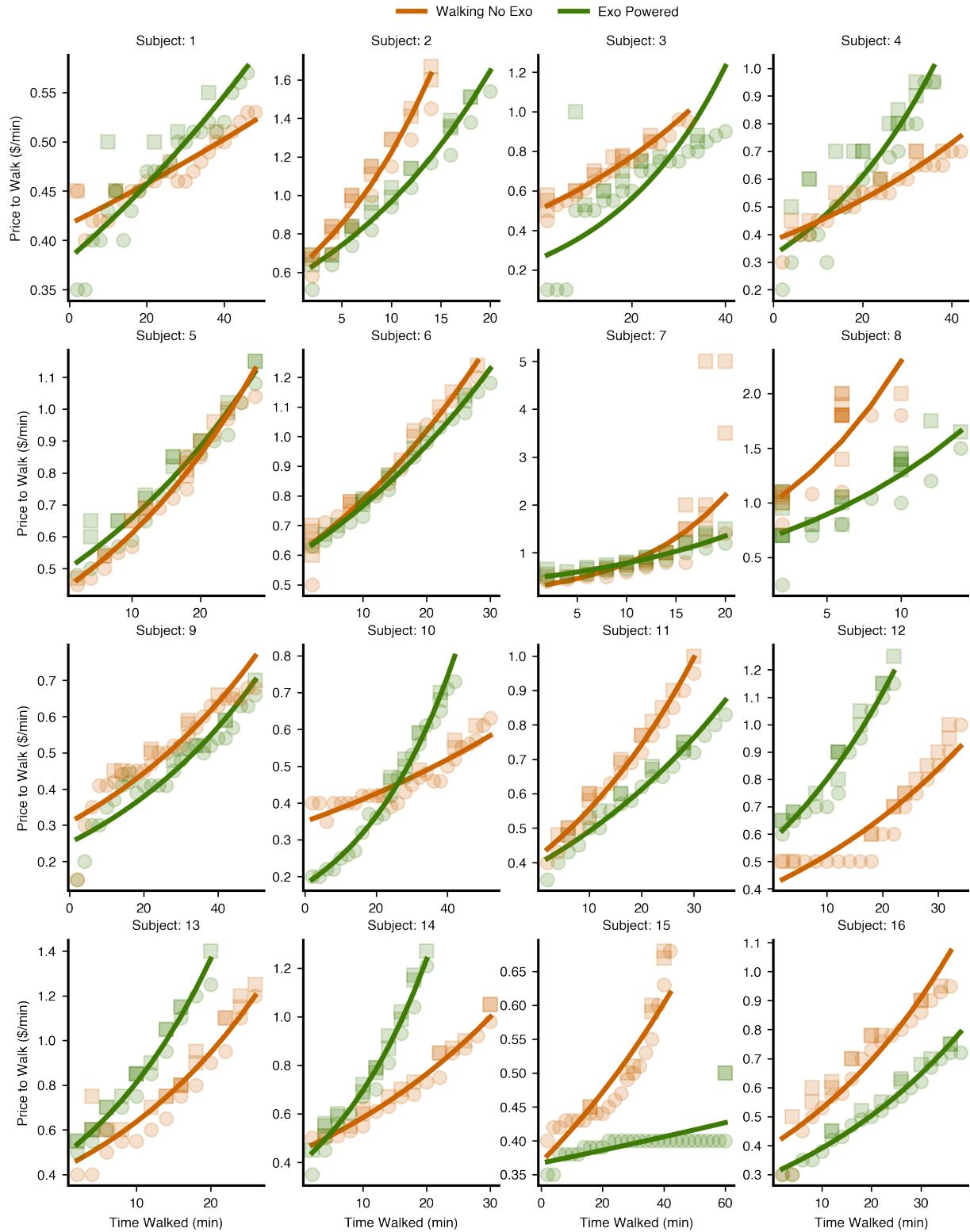


Figure 8: The price to walk curves for the exo-powered (orange) and walking-no-exo (teal) from all sixteen subjects. Circles denote winning bids, while squares denote losing bids.

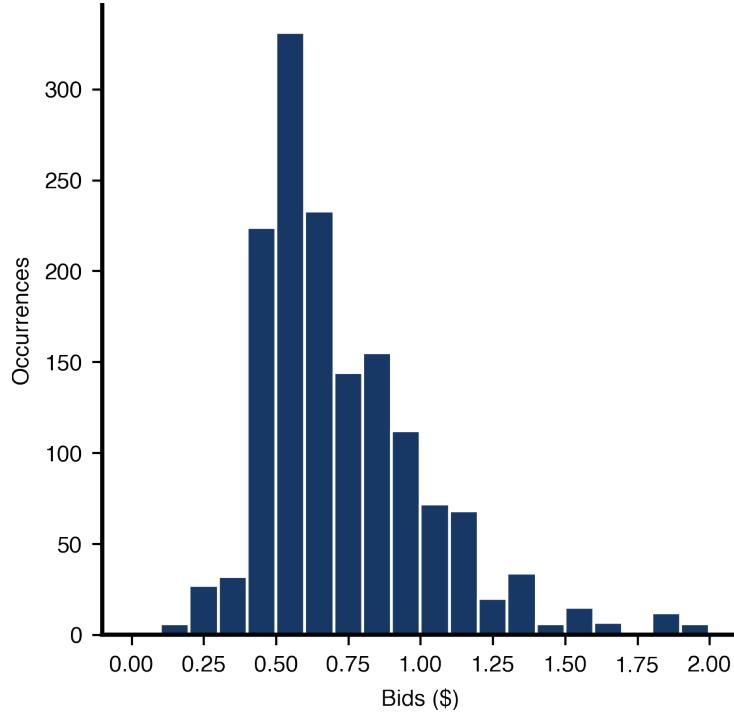


Figure 9: A histogram of all the subject bids for the two-minute intervals across all tested conditions. The average bid was \$0.75, with a standard deviation of \$0.39.

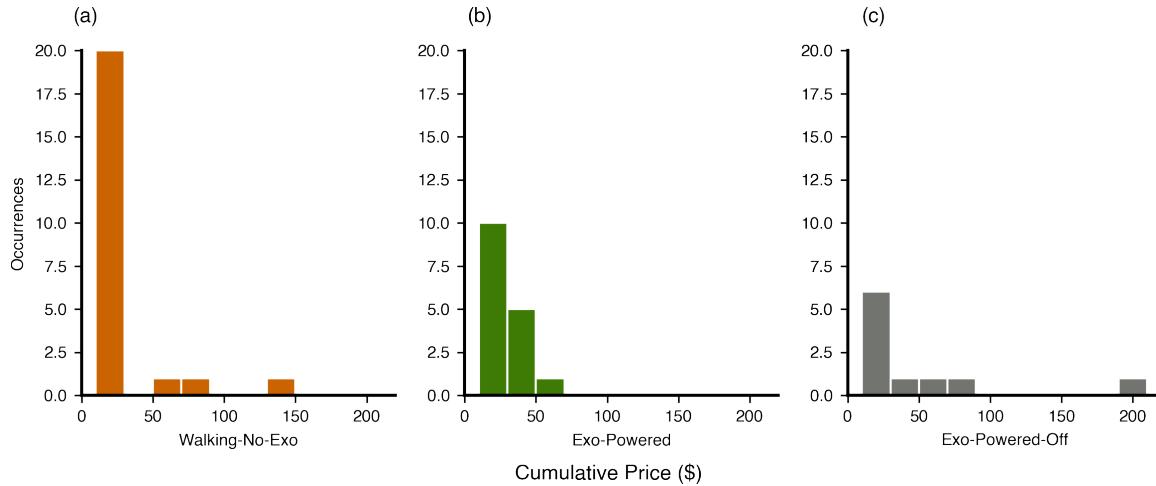


Figure 10: Histograms of the cumulative costs for (a) the walking-no-exo condition, (b) the exo-powered condition, and (c) the exo-powered off condition.

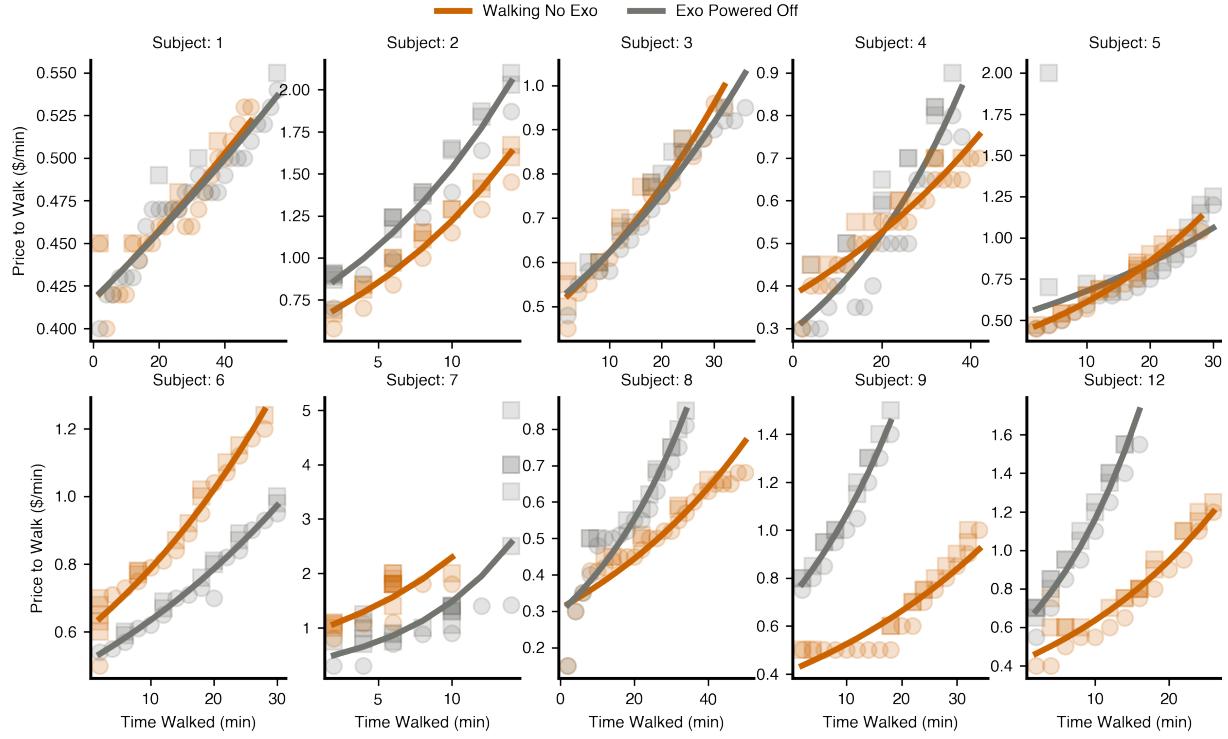


Figure 11: The price to walk curves for the exo-powered-off (gray) and walking-no-exo (teal) for the ten subjects who completed the former condition. Circles denote winning bids, while squares denote losing bids.

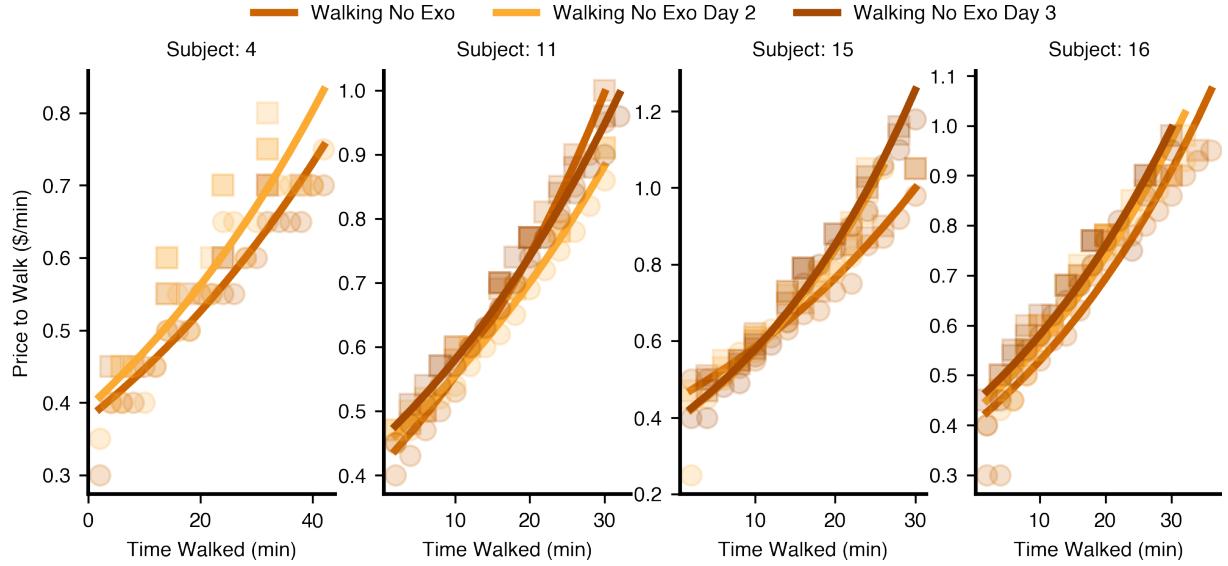


Figure 12: The price to walk curves for the walking-no-exo conditions across different days for the four subjects who repeated the walking-no-exo condition. Circles denote winning bids, while squares denote losing bids. Subject 4 only repeated the walking-no-exo condition once, while the rest of the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

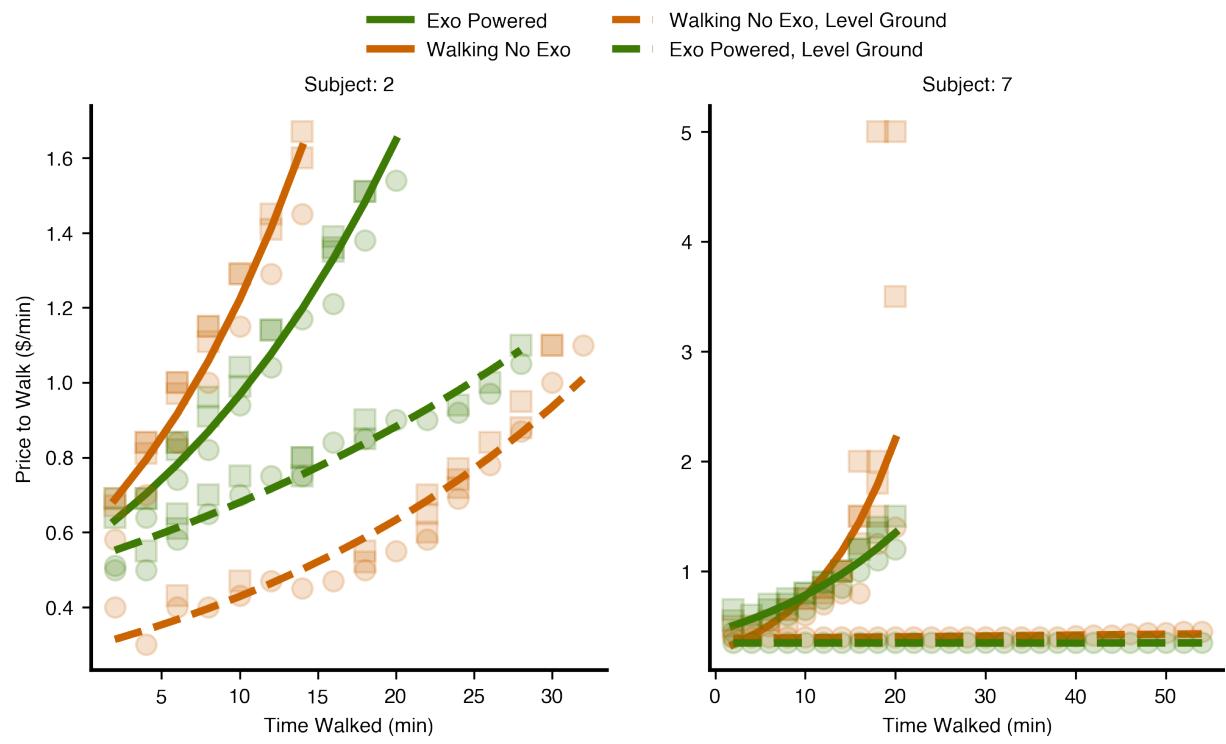


Figure 13: The price to walk curves for two subjects (2 and 7) that repeated the walking-no-exo and exo-powered conditions on a level ground condition. The solid lines denote the standard 10 degree incline task featured in this work; the dashed lines are the level ground condition. Both participants'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