

Predictive Gait Quality Measures Using Modular Neuromuscular Control Parameters in Chronic Post-stroke Individuals

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Research

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1 **Title: Predictive Gait Quality Measures using Modular Neuromuscular Control**
2 **Parameters in Chronic Post-Stroke Individuals**

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32 **Abstract**

33 **Background:** Recent evidence suggests that disinhibition and/or hyperexcitation of the
34 brainstem descending pathways and intraspinal motor network diffuse spastic synergistic
35 activation patterns after stroke. This results in simplified or merged muscle sets (i.e., muscle
36 modules or synergies) compared to non-impaired individuals and this leads to poor walking
37 performance. However, the causal relations of how these neuromuscular deficits influence gait
38 quality (e.g., symmetry or natural walking patterns) are still unclear. The objective of this
39 exploratory study was to investigate the relations of modular neuromuscular framework and gait
40 quality measures in chronic stroke individuals.

41 **Methods:** Sixteen chronic post-stroke individuals participated in this study. Full lower body
42 three-dimensional kinematics and electromyography (EMG) were concurrently measured during
43 overground walking at a comfortable speed. We classified subjects into two groups based on the
44 number of muscle modules and compared gait quality measures using a two-sampled t-test.
45 Then, a stepwise multiple regression was used to investigate the optimal combination of the
46 neuromuscular parameters to predict gait quality measures.

47 **Results:** Subjects who had a reduced number of muscle modules had greater asymmetry in the
48 kinematic parameters including limb length ($p < 0.01$), footpath area ($p < 0.01$), hip ($p < 0.05$) and
49 knee ($p < 0.01$) flexion/extensions, and hip abduction/adduction ($p < 0.01$). We also found that the
50 gait quality measures were predictable with the input variables from the modular neuromuscular
51 control framework including variability accounted for (*VAF*) information from the muscle
52 modules and area under the EMG envelope curves of the quadriceps (i.e., rectus femoris and
53 vastus lateralis) and tibialis anterior muscles with significant association (average $R^2 = 45.6\%$).

54 **Conclusions:** The results suggest that there exists a strong correlation between the
55 neuromuscular control framework and the gait quality measures. This study helps to understand
56 the underlying causality of disturbances in gait quality and provides insight for a more
57 comprehensive outcome measure to assess gait impairment after stroke.

58

59 **Key words:** muscle module, muscle synergy, gait quality, gait symmetry, stroke

60

61 **Background**

62 Effective gait recovery after a stroke involves improvements both in functional mobility and
63 quality of movement. Most of the clinical outcomes in gait studies have focused on functional
64 indices that provide a holistic picture of walking performance and recovery [1], [2]. On the other
65 hand, an increased number of recent studies accentuate the importance of monitoring detailed
66 gait quality to assess gait impairments [3]. Disturbances in gait quality are associated with an
67 increased risk of falls [4], greater energy expenditure [5], and long-term problems such as
68 learned non-use or use-dependent plasticity, musculoskeletal injuries, and pain [6], [7].

69 Symmetry is a common measure to characterize disturbances in gait quality. While
70 spatiotemporal symmetry (e.g., step length, step time) has been well charted to describe walking
71 [8], post-stroke individuals exhibit significant asymmetry in joint kinematics with greater inter-
72 individual variability than spatiotemporal measures [9]. Typical asymmetry in joint kinematics
73 includes reduced hip extension, knee flexion and ankle dorsi/plantarflexion, and knee
74 hyperextension on the impaired side [10]. Limb kinematics, related to the end-effector (i.e., foot)
75 motion in task space, has also been indicated as an important parameter for locomotor function
76 [3]. For instance, Shin et al., found that post-stroke individuals preferentially coordinated the
77 paretic side of limb function using limb kinematics by compensating joint kinematics during
78 walking [3].

79 Neuromuscular activity is crucial to execute biomechanical functions such as gait [11].
80 Previous studies have shown that muscle activity during walking can be grouped into sets of co-
81 excited muscles (also known as muscle modules or synergies) [12]. These modules may originate
82 from neural-adaptations caused by repetitive activities and may reduce the computational cost in
83 selecting strategies of motor coordination [13]. Previous studies have identified that well-

84 coordinated gait in healthy individuals can be produced by four group of modules [12]. Other
85 studies suggest that the concept of muscle modules can be used as an outcome measure to assess
86 motor recovery following therapeutic interventions [14].

87 Recent evidence suggests that disinhibition and/or hyperexcitation of the brainstem
88 descending pathways and intraspinal motor network diffuse spastic synergistic activation post-
89 stroke [15]. As a result, simplified or merged muscle sets compared to non-impaired individuals
90 are typically observed and lead to poor walking performance, for instance, reduced walking
91 speed with greater spatiotemporal asymmetry than those of healthy individuals [12]. While this
92 study discusses somewhat the influence of neuromuscular deficit on gait impairment, no
93 evidence exists to show how detailed gait quality measures are influenced by neuromuscular
94 deficits. Thus, exploring relations between gait quality measures and neuromuscular deficits,
95 such as merged muscle modules, may help to explain the underlying causality among
96 impairments and locomotor functions after stroke.

97 The objective of this exploratory study was to investigate the relations of modular
98 neuromuscular deficits and disturbances in gait quality measures (i.e., asymmetry) in terms of
99 spatiotemporal, limb and joint kinematic parameters in chronic post-stroke individuals (see Fig.
100 1). We measured lower body electromyography (EMG) activities and gait kinematics
101 concurrently during walking from 16 participants. We hypothesized that the post-stroke group
102 with a reduced number of muscle modules will exhibit greater asymmetry in gait quality
103 measures due to the loss of independence in motor activations [12]. Accordingly, we also
104 expected to predict the gait quality measures with the input modular neuromuscular control
105 framework assuming a causal relationship exists in these measures [16].

106

107 **Methods**

108 **Participants**

109 We recruited 16 individuals (6 left hemiparesis, 12 male, age: 62.9 ± 11.1 years) with chronic
110 stroke (> 6 months) to participate in this study approved by the Institutional Review Board of
111 Korea Advanced Institute of Science and Technology. Individuals ranged in age and impairment
112 level (see Table 1). The inclusion criteria were as follows: at least 6 months after stroke,
113 independent walking without falling regardless of walking speed, and over 70 points of Modified
114 Barthel index. The exclusion criteria were as follows: perceptual and cognitive dysfunction and
115 over 3 points on the Modified Ashworth Scale. Prior to the experiment, the experimenter
116 explained all the experimental procedures to each participant and obtained informed consent.

117

118 **Experimental Setup and Data Collection**

119 Full lower body three-dimensional kinematics and EMG data were collected concurrently
120 during overground walking from each participant. Gait kinematics were acquired using the
121 VICON Motion Capture System (MX T-series Vicon Motion Systems Ltd, Oxford, UK),
122 consisting of eight cameras at 100Hz. Surface EMG data were amplified and measured at
123 2000Hz with Delsys Trigno (Delsys, Inc., Natick, MA) from 16 bilateral electrodes placed on the
124 muscles including extensor hallucis longus (*EHL*), tibialis anterior (*TA*), soleus (*SO*),
125 gastrocnemius (*GA*), vastus lateralis (*VA*), rectus femoris (*RF*), semitendinosus (*SM*), and biceps
126 femoris (*BF*).

127 Each participant completed 4-6 trials of walking back and forth on a plain six meters
128 walkway at a comfortable speed. Participants did not use any of assistive devices such as a cane,
129 walker or ankle foot orthosis (AFO) during the recording sessions. All participants wore a

130 harness without any body weight support for safety to catch them if needed to prevent possible
131 falls.

132

133 **Kinematic Data Analysis**

134 Visual 3D v6 Professional (C-Motion, Inc., Germantown, MD) software was used to extract
135 three-axis joint angle trajectories at pelvis and bilateral hip, knee and ankle from the marker data
136 for each trial. Custom software was written in MATLAB (Mathworks, Inc. R2016a, Natick, MA)
137 to calculate features and outcomes. The joint angle trajectories and EMG data were truncated and
138 normalized into 100% of the gait cycle from heel strike to heel strike events for each trial [17].
139 An example of the recorded gait kinematics data of bilateral joint angle trajectories of a single
140 gait cycle in the sagittal plane is illustrated in Fig. 2.

141

142 **Modular Neuromuscular Control Parameters**

143 **Area under the EMG envelope**

144 The selected EMG signals from each participant were high-pass filtered at 40Hz with a zero-
145 lag fourth-order Butterworth filter, demeaned, rectified, and low-pass filtered with a zero-lag
146 fourth-order Butterworth filter at 10Hz, resulting in the EMG envelope [18]. For each muscle,
147 the filtered signal was normalized to its peak value then the area under the EMG envelope curve
148 was calculated and used to represent the neuromuscular indicator as $A_{i,j}$ where i can be *EHL*,
149 *TA*, *SO*, *GA*, *VA*, *RF*, *SM*, or *BF*, and j can be the unaffected (*US*) or affected (*AS*) side. An
150 example area under the EMG envelope curve at *TA* of a participant (P1) is depicted in Fig. 3.

151 **Muscle Modules**

152 The processed EMG signals were decomposed into muscle group weightings and activation
153 timing patterns using nonnegative matrix factorization (NNMF) as previously described in [12].
154 The NNMF determined the minimum number of muscle modules based on a reconstruction
155 quality criterion: variability accounted for (VAF) $\geq 90\%$ [12]. We additionally executed the
156 NNMF algorithm five times, considering that one to five modules were needed for the EMG
157 reconstruction. In this work, the parameters to represent the muscle modules included: number of
158 muscle modules at each side (MM_{US} and MM_{AS}), total number of muscle modules at both sides
159 (i.e., $MM_{Total} = MM_{US} + MM_{AS}$), value at $VAF \geq 90\%$ (VAF_{US} and VAF_{AS}), and VAF values
160 with one to five modules ($VAF_{US,k}$ and $VAF_{AS,k}$, where $k = 1, \dots, 5$).

161

162 **Gait Performance Measure**

163 **Gait Quality Measures**

164 We categorized the gait quality measures into spatiotemporal, limb, and joint kinematic
165 domains [3]. The parameters were extracted by imposing the average joint kinematics data of a
166 single gait cycle into a lower body model. Spatiotemporal parameters included step length (SL)
167 and step time (ST), defined as the linear distance between right and left feet, and the duration of
168 each step, respectively. The parameters of limb kinematics incorporate leg extension angle (LEA)
169 [19], limb length (LL) [20], and footpath area (FPA) [21] defined as the angle between a line
170 from hip to the foot and vertical before toe-off, the range of linear distance between hip and the
171 foot, and the area under the foot pattern from hip sagittal plain during gait cycle, respectively.
172 The parameters of joint kinematics were defined as the range of motion (RoM) of selected joints
173 including all rotations of hip, knee flex/extension, and ankle dorsi/plantarflexion.

174 The symmetry index metric [22] was used to evaluate the gait quality given by

175
$$SI_n = \frac{US_n - AS_n}{0.5(US_n + AS_n)} \quad (1)$$

176 where US_n and AS_n are the n^{th} gait parameter of the unaffected and affected sides, respectively,
177 and n can be the aforementioned spatiotemporal, limb and joint kinematic parameters. The value
178 is always between -2 to 2, and a positive (or negative) value indicates $US > AS$ (or vice versa)
179 [22]. Note that the symmetry index, $SI_n = 0$, when perfect symmetry.

180 **Functional Gait Measure**

181 For the functional gait measure, we selected gait speed (GS) because speed is a well-
182 accepted indicator of gait performance after stroke [23].

183

184 **Statistical Analysis**

185 Subjects were classified into two groups based on the criterion of total number of muscle
186 modules at both sides equal or less than four (i.e., $MM_{Total} \leq 4$), which was determined by
187 rounding the average total number of muscle modules ($MM_{Average} = 4.37$) of all subjects. A
188 two-sided, two-sample t-test was used to test the difference of gait performance measures
189 between the two groups with a significance level of $\alpha < 0.05$.

190 Second, a stepwise multiple regression analysis was used to investigate which combination
191 of the neuromuscular parameters was best associated with gait quality measures with a
192 significance level of $\alpha < 0.05$. A total of 32 aforementioned independent variables including the
193 modular neuromuscular control parameters and a functional gait measure (i.e., speed) were
194 selected. The dependent variables were the gait quality measures including symmetry index of
195 spatiotemporal, limb and joint kinematic parameters. A preliminary analysis was conducted
196 using a linear regression on each considered independent variable and each dependent variable to
197 minimize the number of independent variables and to simplify the final model as possible. Only

198 those independent variable candidates with p -value ≤ 0.05 were selected as input for the
199 stepwise multiple regression analysis.

200

201 **Results**

202 **Classification based on Number of Muscle Modules**

203 The average total number of modules (MM_{Total}) was 4.37 with a minimum and maximum of
204 two and six modules, respectively. We classified subjects whose total number of muscle modules
205 were equal or less than four (i.e., $MM_{Total} \leq 4$) into Group 1 and the rest of the subjects (i.e.,
206 $MM_{Total} > 4$) into Group 2. A total seven of subjects were classified into Group 1 (subject ID: 3,
207 4, 5, 6, 8, 10, and 13), and nine subjects were included into Group 2 (subject ID: 1, 2, 7, 9, 11,
208 12, 14, 15, and 16). All results are summarized in Table 2.

209 A two-sided, two-sample t-test was used to evaluate the significant difference of each gait
210 parameter (i.e., symmetry index) between the two groups. Among all the gait parameters,
211 symmetry indices of LL ($p < 0.01$) and FPA ($p < 0.01$) from limb kinematics and hip flex/extension
212 ($p < 0.05$), hip abd/adduction ($p < 0.01$), and knee flex/extension ($p < 0.01$) from joint kinematics
213 revealed a significant difference between the two groups. While gait speed ($p = 0.07$) and ankle
214 dorsi/plantarflexion ($p = 0.06$) revealed a considerably strong difference, these parameters
215 including other parameters did not reveal a significant difference between the two groups (see
216 Table 3).

217 We additionally fitted a linear regression model on the symmetry index of those significant
218 parameters including LL ($p < 0.05$), FPA ($p < 0.05$), hip flex/extension ($p = 0.10$), hip abd/adduction
219 ($p < 0.05$), and knee flex/extension ($p < 0.05$) across the total number of modules. All parameters

220 except for hip flex/extension still revealed a significant trend in the slope of the linear regression
221 models (See Fig. 4).

222

223 **Stepwise Regression to Predict Gait Features**

224 **Prediction of Spatiotemporal Parameters**

225 For *SL*, the stepwise regression analysis selected VAF_{AS} ($\beta_1 = -10.29$) and $A_{RF,AS}$ ($\beta_2 =$
226 14.01) as independent variables. The model revealed a statistically significant relationship ($F =$
227 8.26 , $p < 0.001$) and accounted for approximately 56% of the variance of *SL* ($R^2 = 0.56$,
228 $Adjusted R^2 = 0.49$). In contrast, none of the independent variable candidates had a p -value $<$
229 0.05 for *ST* (see Table 4). The visualization of the regression model with selected independent
230 variables is shown in Fig. 5 (top left).

231 **Prediction of Limb Kinematic Parameters**

232 For *LEA*, the stepwise regression analysis selected $A_{RF,AS}$ ($\beta_1 = 15.18$) as an independent
233 variable, was statistically significant ($F = 9.16$, $p < 0.001$), and accounted for approximately
234 40% of the variance ($R^2 = 0.40$, $Adjusted R^2 = 0.35$). For *LL* and *FPA*, both stepwise
235 regression analyses selected $VAF_{AS,3}$ ($\beta_1 = 17.82$ for *LL* and $\beta_1 = 20.52$ for *FPA*) as an
236 independent variable. The fitted models were statistically significant ($F = 5.96$, $p < 0.05$ for *LL*
237 and $F = 7.73$, $p < 0.05$ for *FPA*) and accounted for approximately 30% and 36% of the
238 variance of *LL* and *FPA*, respectively ($R^2 = 0.30$, $Adjusted R^2 = 0.25$ for *LL* and $R^2 = 0.36$,
239 $Adjusted R^2 = 0.31$ for *FPA*). The results are summarized in Table 4 with the visualization of
240 the regression models with selected independent variables shown in Fig. 5 (top right three).

241 **Prediction of Joint Kinematic Parameters**

242 For the hip flex/extension, the stepwise regression analysis selected $VAF_{US,1}$ ($\beta_1 = 2.27$)
243 and $A_{RF,US}$ ($\beta_2 = 4.91$) as the independent variables and were statistically significant ($F = 13.0$,
244 $p < 0.001$) accounting for approximately 67% of the variance ($R^2 = 0.67$, *Adjusted R*² =
245 0.62). For the hip abd/adduction, the stepwise regression analysis selected $VAF_{US,1}$ ($\beta_1 = 2.56$)
246 and $A_{VL,AS}$ ($\beta_2 = 1.76$) as the independent variables and were statistically significant ($F = 8.39$,
247 $p < 0.01$) accounting for approximately 56% of the variance ($R^2 = 0.56$, *Adjusted R*² = 0.50).
248 For the hip int/external rotation, the stepwise regression analysis selected $A_{TA,US}$ ($\beta_1 = -2.25$)
249 as the independent variable and was statistically significant ($F = 8.39$, $p < 0.01$) accounting for
250 approximately 39% of the variance ($R^2 = 0.39$, *Adjusted R*² = 0.35). For the knee
251 flex/extension, the stepwise regression analysis selected $VAF_{AS,3}$ ($\beta_1 = 19.24$) as the
252 independent variable. The fitted model was statistically significant ($F = 9.91$, $p < 0.01$) and
253 accounted for approximately 41% of the variance of the knee flex/extension ($R^2 = 0.41$,
254 *Adjusted R*² = 0.37). Finally, for the ankle dorsi/plantarflexion, none of the independent
255 variable candidates had a p -value < 0.05 . (See Table 4). The visualization of the regression
256 models with selected independent variables is shown in Fig. 5 (bottom row).

257

258 **Discussion**

259 The major goal of this study was to delineate the causal relations of modular neuromuscular
260 control parameters and quality of movement during gait after stroke. The main findings were as
261 follows: first, post-stroke individuals with a reduced number of muscle modules exhibit greater
262 asymmetry in gait quality measures, particularly in kinematics level. Second, the gait quality
263 measures were predictable with modular neuromuscular control parameters extracted from the
264 EMG data. Specifically, those parameters were variability accounted for (*VAF*) information from

265 muscle modules (i.e., VA_{AS} , $VA_{AS,3}$, and $VA_{US,1}$) and area under the EMG envelope curves
266 from rectus femoris, vastus lateralis, and tibialis anterior muscles (i.e., $A_{RF,AS}$, $A_{RF,US}$, $A_{VL,AS}$,
267 and $A_{TA,US}$). To our knowledge, this work is a novel in the fact that it integrates detailed gait
268 quality measures with the neuromuscular control framework. The results in this study offer
269 preliminary evidence justifying that the modular neuromuscular framework can be a useful
270 predictor of gait quality measures and help to understand the underlying causality of disturbances
271 in gait quality after stroke.

272 The analysis of the muscle modules has become a more popular tool to describe the
273 neuromotor control of multi-limb movement such as gait after stroke [12]. For the analysis, many
274 previous studies classified muscle modules into paretic and non-paretic sides [12], [24], [25].
275 While this is a reasonable separation given that most stroke populations exhibit a hemiplegic
276 gait, our rationale was that walking is essentially the performance of an inter-coordinated
277 behavior between both legs. Thus, we analyzed the data from a different perspective of
278 classifying participant groups based on the total number of muscle modules of both sides. Our
279 data revealed that there is a significant difference in gait quality measures between the groups,
280 indicating a link between the total number of modules and the gait quality. This result is
281 consistent with a previous study that found merged muscle module is critical to poor walking
282 performance [25]. This also justifies the investigation of therapeutic interventions that can
283 increase the number of modules to improve the gait quality after stroke [14].

284 One thing to note is that most participants in our data had a relatively severe gait impairment
285 with slow walking speed (average speed: 0.29 ± 0.13 m/s, see Table 1) [26]. Nevertheless, we
286 found significant differences in gait quality measures between the groups and those differences
287 were mostly observed in the kinematic parameters (see Table 3). However, none of the

288 parameters from spatiotemporal characteristics and functional gait measure (i.e., speed) revealed
289 a significant difference between the groups. These results correspond with previous research that
290 found post-stroke individuals exhibit significant asymmetry in joint kinematics with greater
291 inter-individual variability than spatiotemporal characteristics [9]. Given that significance was
292 observed between the groups within a severe population in our data, we speculate to find a more
293 distinct trend with a larger sample size with various severity levels including mild to moderately
294 impaired stroke population.

295 A recent study attempted to combine parameters from muscle modules and gait analysis
296 (i.e., spatiotemporal and joint kinematics) to predict functional outcomes such as walking speed
297 [24]. While this study successfully predicted locomotor function with combined biomechanical
298 and neuromuscular measures, the relations between those measures, which would more likely to
299 have direct causal relations, were not analyzed. Thus, we used the stepwise multiple regression
300 approach to find the optimal linear regression model that predicts the gait quality measures with
301 the input modular neuromuscular control parameters. We found that all symmetry indices of gait
302 quality measures, except for step time and ankle dorsi/plantarflexion, were predictable with the
303 neuromuscular modular control parameters with significant associations (see Table 4). Common
304 predictors included quadriceps muscles (i.e., rectus femoris and vastus lateralis) and information
305 from variability accounted for (*VAF*) in the regression models. The inclusion of quadriceps were
306 expected as these muscles have long been believed to be key contributors to hip and knee motion
307 during gait [27]. *VAF*s were also commonly selected because these parameters are critical
308 information determining the number of muscle modules [24]. On the other hand, it is unclear
309 why the tibialis anterior muscle appeared to be negatively associated with hip internal/external
310 rotation given that this muscle is known as ankle dorsiflexor. This result may be related to

311 abnormal compensatory coupling between irrelevant muscles due to neurological impairment
312 [28]. However, further research is needed where the EMG measures at other locations such as
313 hip lateral rotator group muscles are added to investigate this connection, which may indeed be
314 epiphenomenal. Overall, these results suggest causal relations exist between the neuromuscular
315 control framework and the gait quality measures.

316 One possible therapeutic application of this causal relationship would be using the concept
317 of neuromuscular and gait quality parameters as outcome measures to assess gait impairment
318 after stroke. An example of previous research would be a locomotor intervention study that
319 applied muscle modules as pre- and post-therapy outcome measures [14], [29]. These studies
320 found increase in number of muscle modules and improvements in quality of modular
321 organization (i.e., timing and compositions) as well as clinical measures such as gait speed after
322 a locomotor rehabilitation therapy. This indicates that therapeutic interventions can change or
323 improve the neuromuscular control framework, but the influence of these interventions on
324 detailed movement quality was not reported in these studies. Thus, we expect the application of
325 the neuromuscular and gait quality parameters as outcome measures will fill the gap and provide
326 a more comprehensive characterization of gait recovery post-stroke.

327 The purpose of this proof-of-concept study was to provide initial evidence of causal
328 relations between gait quality and neuromuscular parameters for chronic post-stroke individuals.
329 However, this study was limited to a small sample size of 16 patients with a restricted range of
330 impairment level; only severely impaired individuals with a slow walking speed participated in
331 this work. However, our data still showed clear initial results on the relative importance of
332 several predictors, providing the significant causal relationships exist between neuromuscular
333 control framework and gait quality measures. We expect this methodology can be potentially

334 used for gait training research or in clinical practice to understand impairments related to gait
335 function.

336

337 **Conclusions**

338 The purpose of this study was to investigate the influence of modular neuromuscular control
339 framework on gait quality measures. We observed that chronic post-stroke individuals with a
340 reduced number of muscle modules exhibit a greater deficit in gait quality measures, particularly
341 in kinematics level. We also found that the gait quality measures are predictive with the input
342 modular neuromuscular control variables. We conclude that there exists causal relations between
343 the neuromuscular control framework and gait quality measures. These promising results justify
344 further research with a larger post-stroke population and expanded range of impairment level for
345 more reliable generalization.

346

347 **List of abbreviations**

348 AFO: ankle foot orthosis; *AS*: affected side; *BF*: biceps femoris; *EHL*: extensor hallucis longus;
349 EMG: electromyography; *FPA*: footpath area; *GA*: gastrocnemius; *GS*: gait speed; *LEA*: leg
350 extension angle; *LL*: limb length; *MM*: muscle module; NNMF: nonnegative matrix
351 factorization; *RF*: rectus femoris; RoM: range of motion; *SI*: symmetry index; *SL*: step length;
352 *SM*: semitendinosus; *SO*: soleus; *ST*: step time; *TA*: tibialis anterior; *US*: unaffected side; *VA*:
353 vastus lateralis; *VAF*: variability accounted for

354

355 **Ethics approval and consent to participate**

356 This study was approved by Institutional Review Board of the Korea Advanced Institute of
357 Science and Technology (KH2019-05). All subjects gave written informed consent prior to data
358 collection.

359

360 **Consent for publication**

361 Participants gave written informed consent to data treatment in this research study and
362 permission to publish anonymous data and results.

363

364 **Availability of data and materials**

365 The data collected in this study are available from the corresponding author on reasonable
366 request.

367

368 **Competing interests**

369 The authors declare that they have no competing interests.

370

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374

375 **Author's contribution**

376 SYS drafted the complete manuscript, performed data post-processing and statistical data
377 analysis. YK participated in conducting experimental data collection, recruited subjects and

378 managed IRB approvals. AJ and HSP conceived the whole idea, managed IRB approvals and
379 helped in finalizing the manuscript. All authors read and approved the final manuscript.

380

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463

464 **Figure Legends**

465 **Figure 1.** Overview of the study.

466

467 **Figure 2.** Example joint angle trajectories of a single gait cycle (heel strike to heel strike) in
468 sagittal plane, hip flexion/extension (left), knee flexion/extension (middle), and ankle
469 dorsi/plantarflexion (right).

470

471 **Figure 3.** Example area under the EMG envelope curve of tibialis anterior muscle from
472 participant P1. A.U. in the y-axis indicates arbitrary unit.

473

474 **Figure 4.** Symmetry index (*SI*) of significant gait parameters including limb length (*LL*),
475 footpath area (*FPA*), hip flex/extension (*Hip FE*), hip abd/adduction (*Hip AA*) and knee
476 flex/extension (*Knee FE*) with increased total number of muscle modules at both sides
477 (MM_{Total}). Linear regression models (bottom, right) were fitted on each gait parameter. The
478 shaded area indicates 95% confidence interval.

479

480 **Figure 5.** Regression models with selected independent variables (see Table 3) from stepwise
481 multiple regression. Note that a plane in three-dimensional space is illustrated with two
482 independent variables for step length, hip flex/extension and hip abd/adduction.

483

484 **Table 1.** Demographics and clinical characteristics of the participants.

ID	Age (years)	Sex	Height (cm)	Weight (kg)	Side affected	Time after stroke (years)	Type of lesion	FMA	FIM	Speed (m/s)
P1	59	M	170	69	R	4	Ischemic	21	11	0.29
P2	61	M	169	72	L	3	Ischemic	16	9	0.27
P3	70	M	158	69	R	2	Ischemic	12	12	0.16
P4	55	F	158	60	L	11	Hemorrhage	22	14	0.24
P5	59	M	176.5	66	R	3	Ischemic	13	8	0.21
P6	73	M	170	74	Both	21	Hemorrhage	22	3	0.18
P7	75	F	153	49	R	5	Ischemic	25	12	0.20
P8	75	M	164	62	R	15	Ischemic	18	13	0.31
P9	68	M	172	83	L	10	Ischemic	26	13	0.46
P10	36	M	173	80	L	1	Ischemic	17	7	0.15
P11	77	M	169	58	R	25	Ischemic	10	12	0.26
P12	55	M	177	84	L	13	Ischemic	15	14	0.37
P13	47	F	163	57	R	4	Hemorrhage	19	13	0.29
P14	64	M	172	81	R	27	Ischemic	31	14	0.49
P15	62	F	156	57	L	15	Hemorrhage	16	11	0.13
P16	70	M	171	83	R	18	Ischemic	27	12	0.58

485 M – male, F – female, R – right, L – left, FMA – Fugl-Meyer assessment (max: 34), FIM – Functional Independent Measure
 486 (locomotion, max: 14)

487

488 **Table 2.** Classification of the subject groups based on the total number of muscle modules at
 489 both sides

Group	MM_{Total}	MM_{US}	MM_{AS}	Subject ID	Number of Subjects
Group 1	2	1	1	P6	1
	3	2	1	P4, P10	2
	4	2	2	P3, P5, P8, P13	4
Group 2	5	2	3	P7	1
	5	3	2	P1, P2, P9, P11, P12, P14, P15	7
	6	3	3	P16	1

490 MM – number of muscle modules, US – unaffected side, AS – affected side

491

492 **Table 3.** Results of the two-sided, two-sample t-tests of gait performance measures between two
 493 groups

Gait Features	SI Parameters [%]	Group1	Group2	95% CI	p-value
Functional	Speed	0.22±0.06	0.34±0.15	[-0.25, 0.01]	0.07
Spatiotemporal characteristics	Step length	-0.34±0.34	-0.44±0.32	[-0.27, 0.45]	0.59
	Step duration	-0.19±0.15	-0.15±0.08	[-0.16, 0.09]	0.58
Limb kinematics	Leg extension angle	-0.05±0.27	-0.16±0.35	[-0.23, 0.45]	0.50
	Limb length	0.84±0.10	0.38±0.35	[0.17, 0.76]	< 0.01
	Footpath area	1.04±0.13	0.51±0.36	[0.22, 0.84]	< 0.01
Joint kinematics	Hip flex/ex	0.67±0.27	0.38±0.18	[0.04, 0.53]	< 0.05
	Hip abd/add	0.31±0.25	-0.02±0.16	[0.11, 0.55]	< 0.01
	Hip int/ext ro.	-0.05±0.14	-0.02±0.23	[-0.25, 0.18]	0.75
	Knee flex/ex	0.56±0.23	0.15±0.23	[0.15, 0.65]	< 0.01
	Ankle dorsi/plantar	0.37±0.36	0.05±0.25	[-0.01, 0.64]	0.06

494 SI – symmetry index, flex/ex – flexion/extension, abd/add – abduction/adduction, int/ext ro. – internal/external rotation,
 495 dorsi/plantar – dorsi/plantarflexion, CI – confidence interval

496

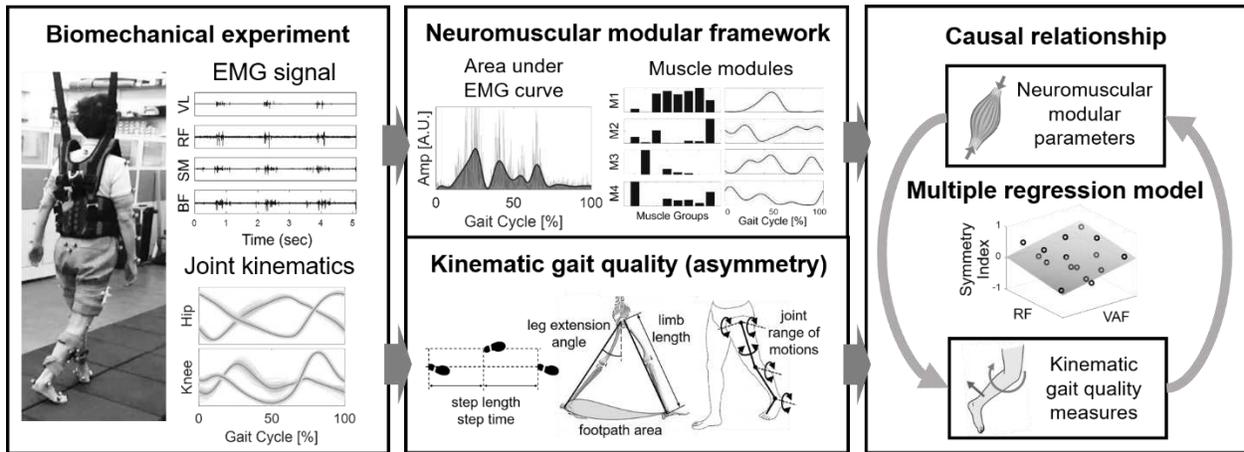
497 **Table 4.** Results of the stepwise multiple regression models with the dependent variable of
 498 symmetry ratio of gait quality measures and independent variable of modular neuromuscular
 499 parameters

Gait Features	SI parameters [%]	Multiple regression model	F-value	p-value	R^2	adjusted R^2
Spatiotemporal characteristics	Step length	$SI_{SL} = 8.422 - 10.29VAF_{AS} + 14.01A_{RF,AS}$	8.26	< 0.001	0.56	0.49
	Step time	-	-	-	-	-
Limb kinematics	Leg extension angle	$SI_{LEA} = -0.98 + 15.18A_{RF,AS}$	9.16	< 0.001	0.40	0.35
	Limb length	$SI_{LL} = -16.52 + 17.82VAF_{AS,3}$	5.96	< 0.05	0.30	0.25
	Footpath area	$SI_{FPA} = -18.98 + 20.52VAF_{AS,3}$	7.73	< 0.05	0.36	0.31
Joint kinematics	Hip flex/ex	$SI_{Hip,FE} = -1.82 + 2.27VAF_{US,1} + 4.91A_{RF,US}$	13.0	< 0.001	0.67	0.62
	Hip abd/add	$SI_{Hip,AA} = -2.19 + 2.56VAF_{US,1} + 1.76A_{VL,AS}$	8.39	< 0.01	0.56	0.50
	Hip int/ext ro.	$SI_{Hip,IE} = 0.46 - 2.25A_{TA,US}$	8.39	< 0.01	0.39	0.35
	Knee flex/ex	$SI_{Knee,FE} = -18.1 + 19.24VAF_{AS,3}$	9.91	< 0.01	0.41	0.37
	Ankle dorsi/plantar	-	-	-	-	-

500 SL – step length, LEA – leg extension angle, LL – limb length, FPA – footpath area, Hip,FE – hip flexion/extension, Hip,AA – hip
 501 abduction/adduction, Hip,IE – hip internal/external rotation, $Knee,FE$ – knee flexion/extension, VAF – variability accounted for,
 502 A – area under EMG envelope curve, RF – rectus femoris, VL – vastus lateralis, TA – tibialis anterior

503

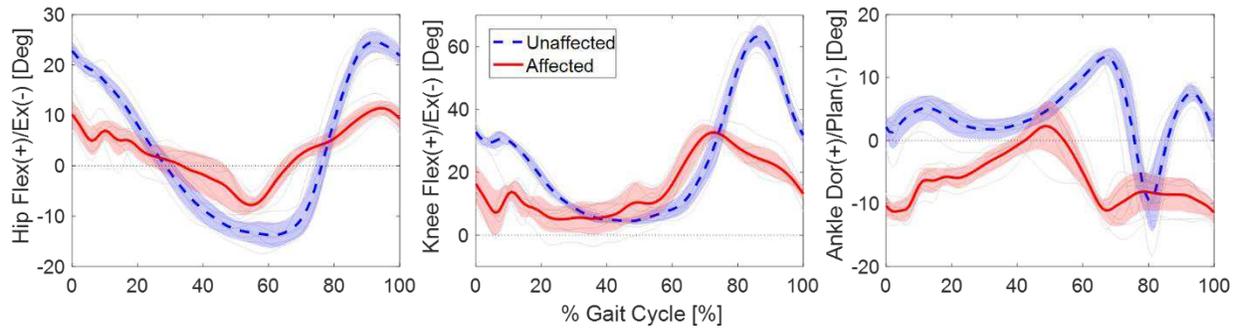
504 **Fig 1.**



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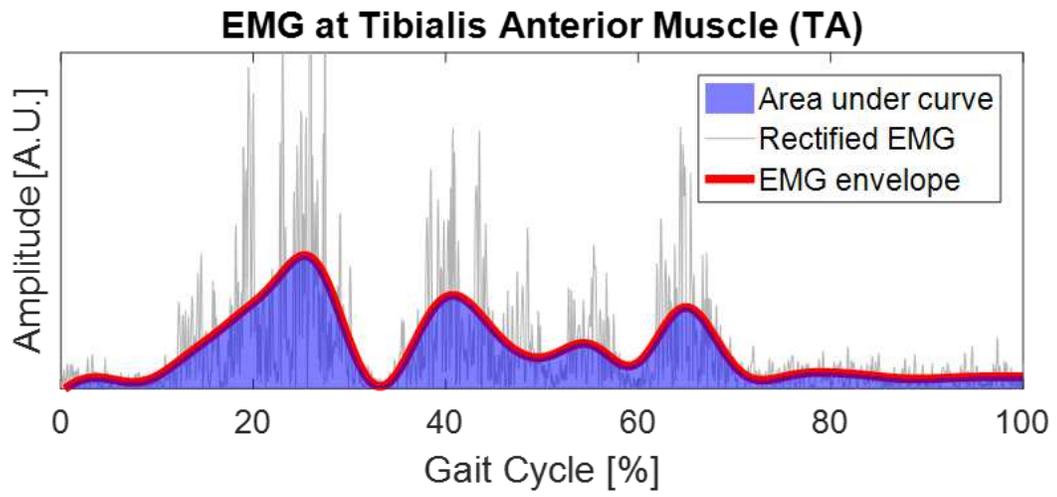
507 **Fig 2.**



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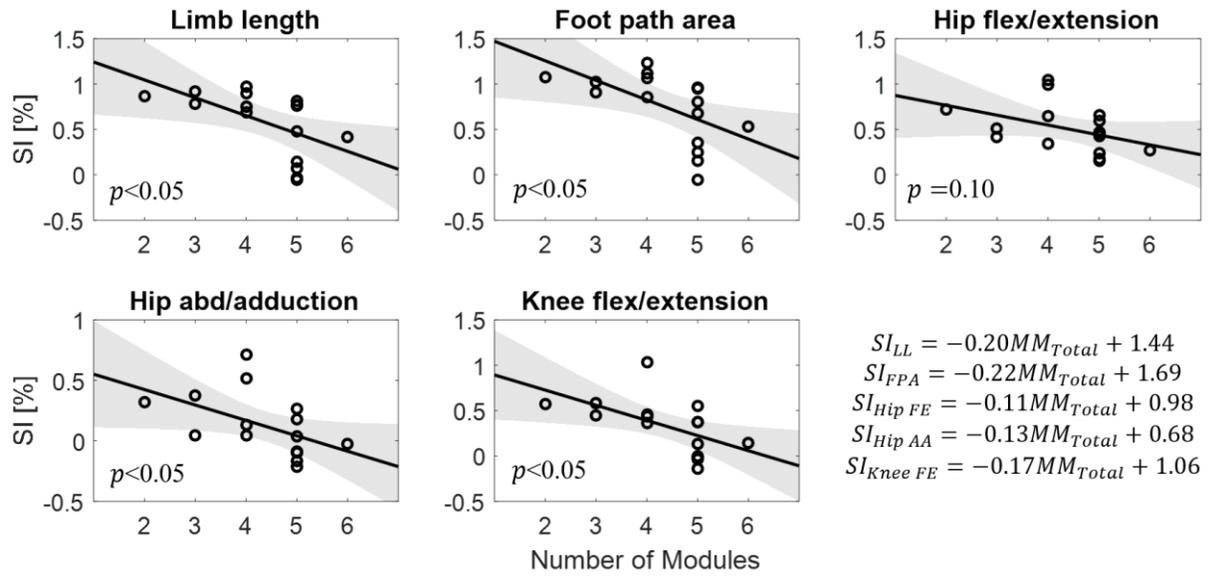
510 Fig 3.



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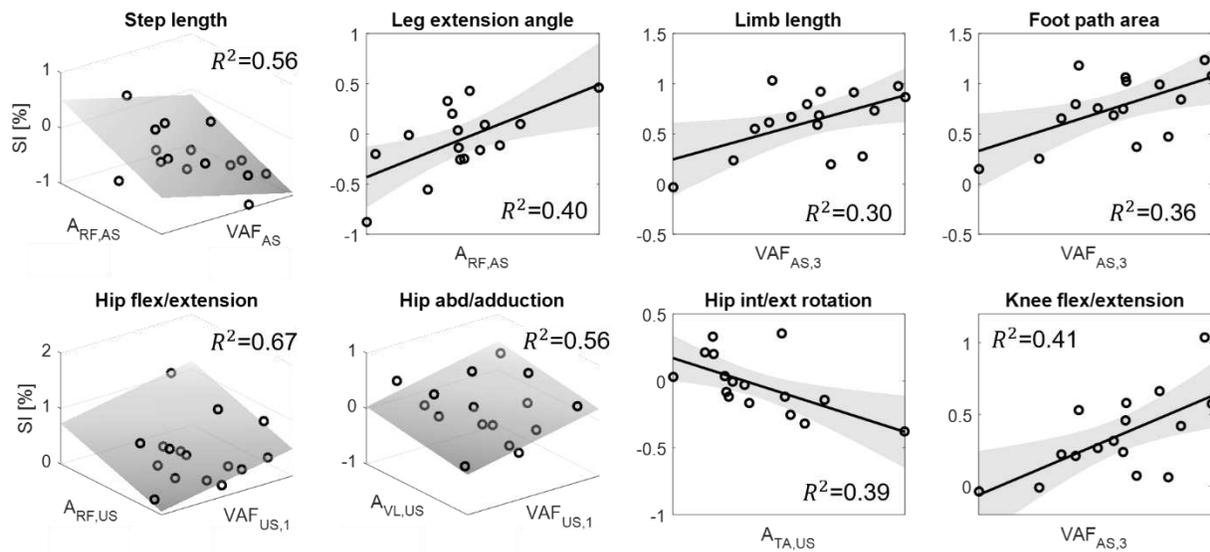
513 **Fig 4.**



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Figures

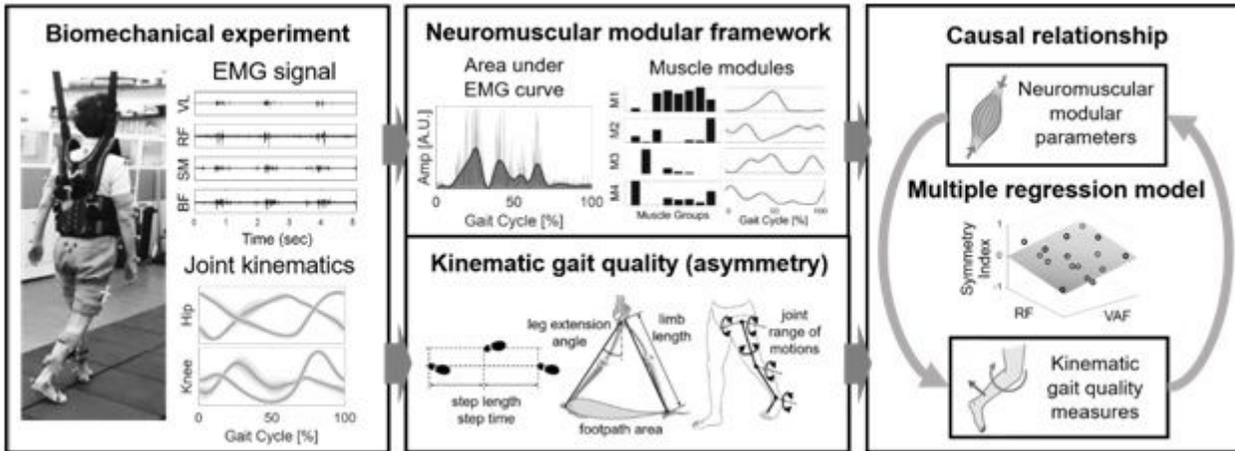


Figure 1

Overview of the study.

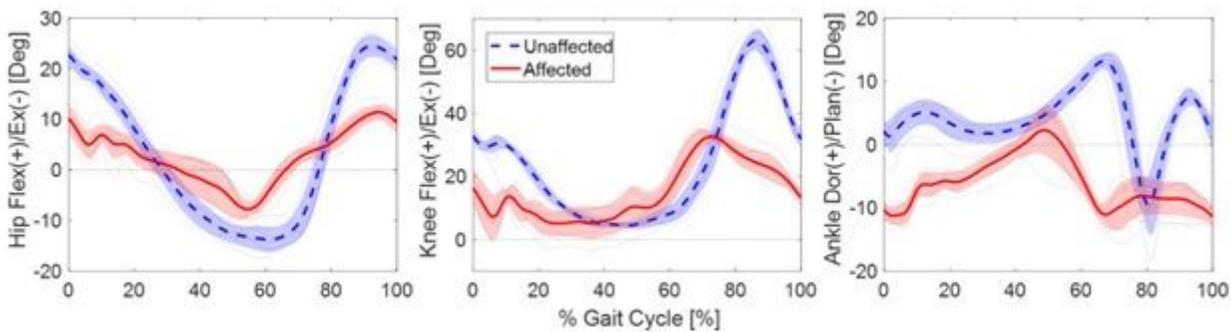


Figure 2

Example joint angle trajectories of a single gait cycle (heel strike to heel strike) in sagittal plane, hip flexion/extension (left), knee flexion/extension (middle), and ankle dorsi/plantarflexion (right).

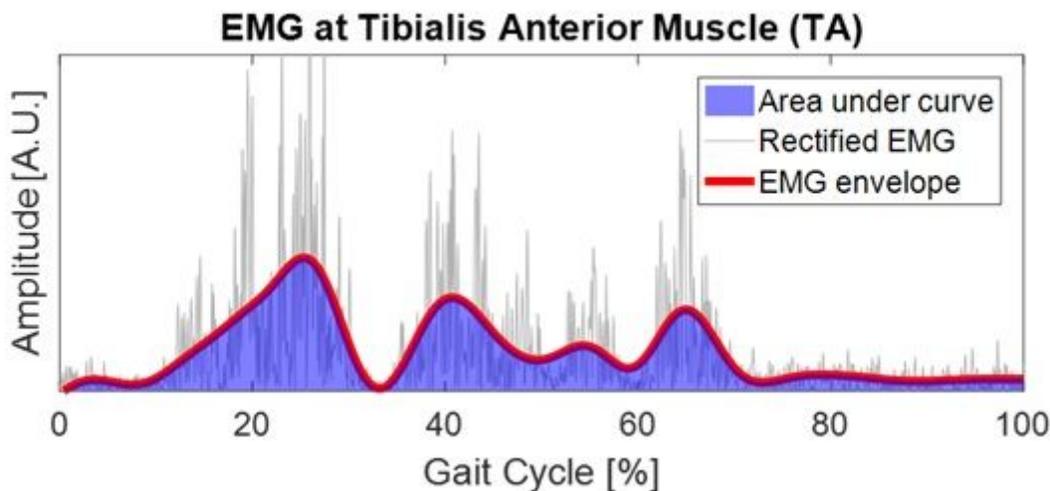


Figure 3

Example area under the EMG envelope curve of tibialis anterior muscle from participant P1. A.U. in the y-axis indicates arbitrary unit.

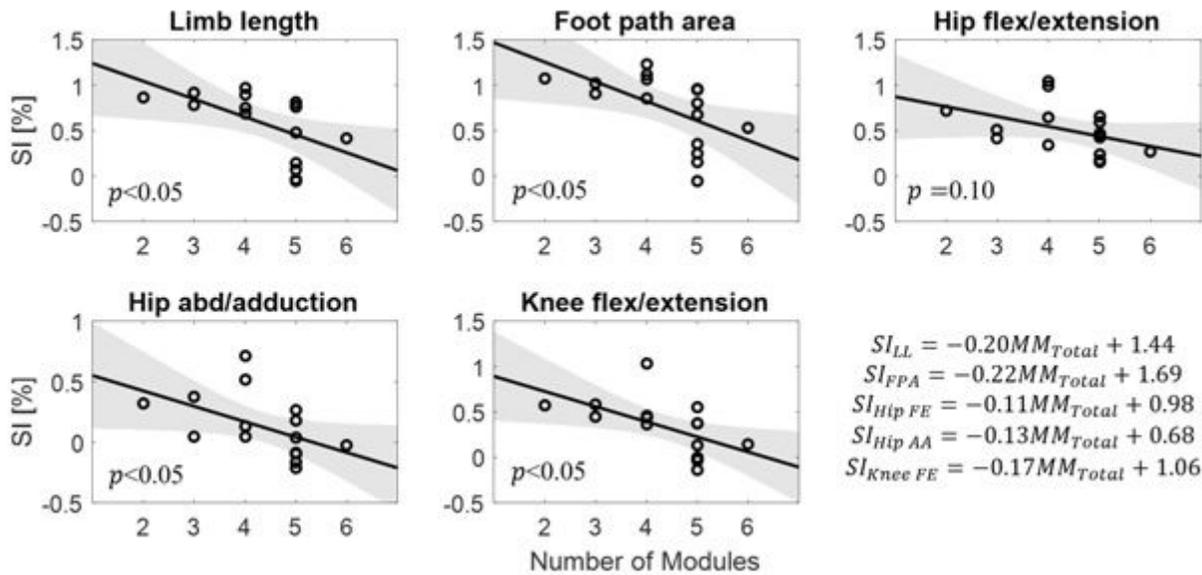


Figure 4

Symmetry index (SI) of significant gait parameters including limb length (LL), footpath area (FPA), hip flex/extension (Hip FE), hip abd/adduction (Hip AA) and knee flex/extension (Knee FE) with increased total number of muscle modules at both sides (MM_{Total}). Linear regression models (bottom, right) were fitted on each gait parameter. The shaded area indicates 95% confidence interval.

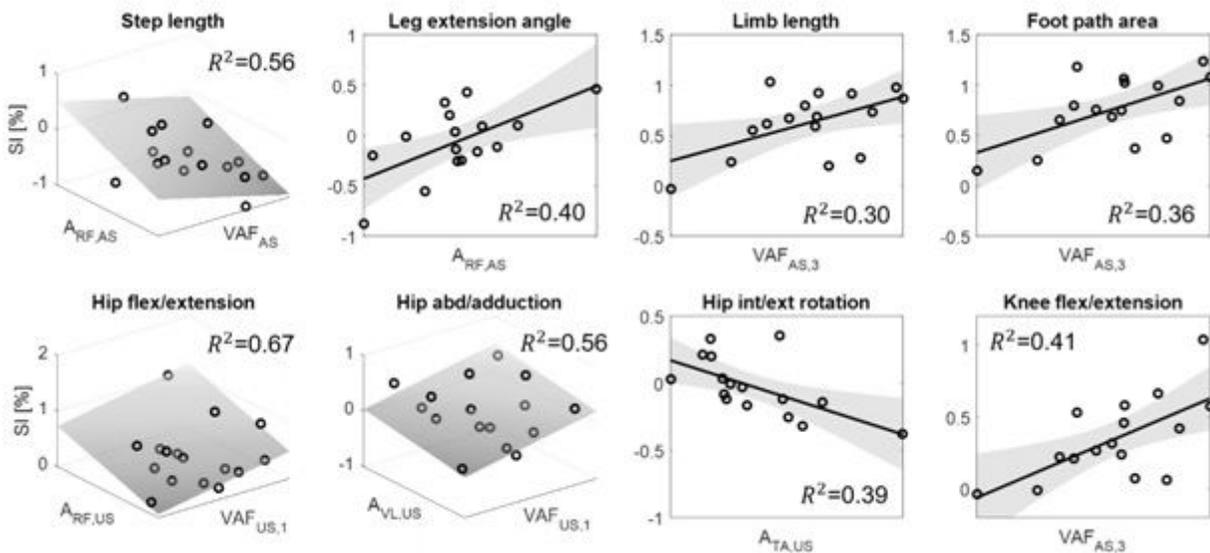


Figure 5

Regression models with selected independent variables (see Table 3) from stepwise multiple regression. Note that a plane in three-dimensional space is illustrated with two independent variables for step length, hip flex/extension and hip abd/adduction.