Simulating bipedal walking using a translating center of pressure

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Abstract

During walking, foot orientation and foot placement allow humans to stabilize their gait and to move forward. Consequently the upper body adapts to the ground reaction force (GRF) transmitted through the feet. The foot-ground contact is often modeled as a fixed pivot in bipedal models for analysis of locomotion. The fixed pivot models, however, cannot capture the effect of shift in the pivot point from heel to toe. In this study, we propose a novel bipedal model, called $SLIP_{COP}$, which employs a translating center of pressure (COP) in a spring loaded inverted pendulum (SLIP) model. The translating COP has two modes: one with a constant speed of translation and the other as the weighted function of the GRF in the fore aft direction. We use the relation between walking speed and touchdown (TD) angle as well as walking speed and COP speed, from existing literature, to restrict steady state solutions within the human walking domain. We find that with these relations, $SLIP_{COP}$ provides steady state solutions for very slow to very fast walking speeds unlike SLIP. $SLIP_{COP}$ for normal to very fast walking speed shows good accuracy in estimating COM amplitude and swing stance ratio. $SLIP_{COP}$ is able to estimate the distance traveled by the COP during stance with high precision.

1 Introduction

Walking is an efficient form of locomotion which allows humans to travel from one place to another. Human walking has two facets, which are reducing cost of locomotion and gait stabilization. Foot placement and orientation plays an important role in stabilizing our gait. To improve our understanding of walking and its underlying mechanism, human motion capturing and reductive modeling have been widely used [1]. Human body is a complex redundant system and reductive models or templates strip down the complex architecture of the body into simple elements and allow a computationally inexpensive way to simulate locomotion, with a certain degree of accuracy [1]. The earliest versions of such templates include the sagittal plane based inverted pendulum model (IP) [2, 3] and spring-mass model [4, 5]. The IP model assumes incompressible legs with center of mass (COM) vaulting over the legs, with a fixed foot. Due to its rigid legs, IP model provides an incorrect representation of ground reaction force (GRF) pattern, compared to that observed in humans. To overcome this drawback of the IP model, some templates include a springy leg to obtain more accurate estimates of walking gait trajectories [6, 7, 8, 9, 10]. One such model is the sagittal plane based SLIP model [10]. Due to SLIP's compliant legs it is able to generate the COM trajectory and GRF pattern during walking to that observed in human walking. The steady

39 state solutions of SLIP also suggests that walking is one of the many domains of
40 different locomotion patterns generated. Hence, it is important to narrow down
41 the parameters in a bipedal model pertaining only to the domain of walking.

42 IP model and SLIP model assume a fixed pivot during stance as discussed
43 above, which is essentially the mean position traveled by COP during stance. In
44 a study to analyze treadmill walking¹, SLIP fairly estimates COM trajectory at
45 1m/s, while failing to estimate at other walking speeds [11]. They also men-
46 tion that due to the fixed pivot of the SLIP, the model has to be simulated at a
47 steeper TD angle compared to human walking. This characteristic of SLIP might
48 be restricting its predictive capabilities at slower and faster walking speed. During
49 walking, COP of a particular foot travels approximately a distance of a foot length
50 and the COP progression velocity depends on speed of walking [12, 13, 14]. To
51 accommodate the mechanical consequences of COP progression, a few bipedal
52 models were developed for running [15, 16] and walking [17, 18]. Bullimore et
53 al. [16] take into consideration the change in TD and LO angles caused by COP
54 translation, called POFT (point of force translation), in the conventional spring
55 mass model for running. They use a constant velocity based COP progression
56 model without considering the acceleration of the COP during stance. They ob-

¹Subjects walked at speeds of (0.52,1.04,1.55,2.07 and 2.59) m/s in this experiment

serve similar COM trajectories and GRF patterns as in human running but the model shows drastic decrease in spring stiffness. Lee et al. [17] use a translating point of force application (PFA) in an IP model for walking and show that the error in vertical displacement of the COM predicted by the model increases from 111% to 240%, as walking speed increases from 0.5 m/s to 2.5 m/s respectively. However, these errors were considerably lesser compared to IP model with a fixed pivot. Miff et al. [18] show that during walking vertical excursion of the trunk is dependent on the foot rocker radius in the rocker based IP model. IP models in the above studies consider only the single stance vaulting of the COM but not the foot impact and double stance phase of walking. We need a bipedal model which can simulate COP progression along with single/double stance, COM trajectory and GRF patterns.

The objective of this study is to check, if addition of a COP progression model would improve SLIP model's performance at slower and faster walking speeds. Instead of using a predefined leg stiffness [11], we optimize our spring stiffness. We use the relation between TD angle, walking speed and COP speed obtained from existing literature so as to make model-experiment comparison. We develop a generic model called $SLIP_{COP}$ (Fig. 1) with COP progression considering two modes of COP translation during stance: one with a constant COP speed and the

other accelerated COP. We include the constant velocity COP progression model so as to compare our model with previous similar models. We make inter-model comparisons between SLIP and SLIP_{COP} for various walking parameters to analyze the results qualitatively and quantitatively. Subsequently, we compare the two models with real walking scenarios to assess the quality of our solutions.

2 Method

We simulate the two models, SLIP and SLIP_{COP}, as seen in Fig. 1. The position and velocity of right and left COP are denoted as $[f_r, \dot{f}_r]$ and $[f_l, \dot{f}_l]$ respectively. For SLIP, $\dot{f}_r$ and $\dot{f}_l$ are always 0 due to a fixed pivot. Like the conventional SLIP, SLIP_{COP} consists of a COM attached with two springy legs. The legs are considered massless and a swinging leg can be ignored. As illustrated in Fig. 2, COM state at apex is described by $[x, \dot{x}, y, \dot{y}]$ and at TD the leg makes an angle of θ_o . Both models are simulated in the sagittal plane. We non-dimensionalize the equations of motion to develop generic models catering to humans with different anthropometric measurements [19, 20, 21]. Force experienced by the COM before non-dimensionalization in the forward and vertical direction is given as

$$F_{xd} = m\ddot{x}_d = F_{rd} \cos \alpha + F_{ld} \cos \beta \quad (1)$$

$$F_{yd} = m\ddot{y}_d = F_{rd} \sin \alpha + F_{ld} \sin \beta - mg \quad (2)$$

where $F_{rd} = k(L_o - L_{rd})$ is the GRF in the right leg, m is the mass, k is the spring stiffness, g acceleration due to gravity, $L_{rd} = \sqrt{((x_d - f_{rd})^2 + (y_d)^2)}$ is the length of the right leg in stance, L_o is the uncompressed leg length, the subscript l and r refer left and right leg, and the subscript d means dimensionalized. Upon non-dimensionalizing eqns.(1)(2), the time-dependent terms are divided by $\sqrt{\frac{L_o}{g}}$, distance terms by uncompressed leg length L_o and divide the equations throughout by mg [20, 10]. After non-dimensionalization the force experienced by the COM is given as

$$F_x = \ddot{x} = \tilde{F}_r \cos \alpha + \tilde{F}_l \cos \beta \quad (3)$$

$$F_y = \ddot{y} = \tilde{F}_r \sin \alpha + \tilde{F}_l \sin \beta - 1 \quad (4)$$

92 where $\tilde{F}_r = \tilde{k}(1 - L_r)$ (see Fig. 1), $\tilde{k} = \frac{kL_o}{mg}$ is the relative stiffness of the legs. At
 93 TD, the leg angle reorients to θ_o and at lift off (LO) occurs when the GRF becomes
 94 0.

95 2.1 Gait parameter relations

In order to restrict our model's solution search within the walking domain, we use the relation between walking speed, TD angle and COP speed obtained through existing literature [11, 12] (see Table 2). The lower and upper bound for the COP progression velocity are the minimum and maximum speed of the COP during experimental walking. The COP model during stance is described as the function of the GRF in the fore-aft direction as

$$\ddot{f}_r = \mu F_r \cos \alpha \quad (5)$$

$$\ddot{f}_l = \mu F_l \cos \beta \quad (6)$$

96 Especially, two modes of this translating COP model are considered: one con-
97 sidering effect of a constant COP speed during stance ($\mu = 0$) and the other as
98 weighted function of the GRF during stance ($\mu = 1$).

99 We obtain steady state solutions of the two models by optimizing their param-
100 eters to generate a limit cycle. To generate a limit cycle we consider the apex
101 to apex state errors. The state of the model at apex is completely described by
102 its relative horizontal distance between COM and COP denoted by $(x - f)$, hor-
103 izontal velocity $\dot{x}_i$, apex height y_i , vertical velocity $\dot{y}_i$, and COP velocity $\dot{f}_i$. To
104 obtain a limit cycle, we calculate the stride to stride error for consecutive apex

Algorithm 1 Algorithm to obtain a limit cycle

To Optimize: $[\tilde{k}, \dot{f}_r, \dot{f}_l]$

Constraints: $5 < \tilde{k} < 80, f_{LB} < \dot{f}_r, \dot{f}_l < f_{UB}$

if SLIP **then**

fixed pivot

else if SLIP_{COP} **then**

translating pivot, $\mu = 0$ or $\mu = 1, \dot{f} > 0$

for μ **do**

for $\dot{x} = \dot{x}_{min} : \dot{x}_{max}$ **do**

Estimate $\theta_o, \dot{f}_{LB}, \dot{f}_i, \dot{f}_{UB}$ from Table. 2

for $y_o = \sin(\theta_o) : 1$ **do**

Optimize Parameters $[\tilde{k}, \dot{f}_r, \dot{f}_l]$

Solve eqns.(3)(4)(5)(6)

Evaluate error (e) between apex states

$e = [\dot{x}_{i+1} - \dot{x}_i; x_{i+1} - f_{l(i+1)}; y_{i+1} - y_i;$

$\dot{y}_{i+1} - \dot{y}_i; \dot{f}_{l(i+1)} - \dot{f}_{r(i)}]$

end for

end for

end for

end if

states (i and $i + 1$) using a 5-dimensional nonlinear Poincaré return map [22]. The initial apex state and final apex state of the model are given as $[x, \dot{x}, y, \dot{y}, f_r, \dot{f}_r]_i$ and $[x, \dot{x}, y, \dot{y}, f_l, \dot{f}_l]_{i+1}$ respectively. For a given set of $x_i, \dot{x}_i, y_i, \dot{y}_i$, we optimize relative stiffness $\tilde{k}$, right COP speed $\dot{f}_r$ and left COP speed $\dot{f}_l$ to get a limit cycle as shown in the Algorithm 1. At the start of simulation, the foot is placed at the origin with the COM directly above it.

3 Results

Firstly, we compare COM, COP and GRF trajectories for the two models (SLIP, SLIP_{COP}) for a given set of optimized parameters (see Table 1) to assess the qualitative nature of the solutions. Fig. 4a & b are GRF and COP trajectories for individual legs. GRF pattern in vertical and horizontal direction resemble that of experimental walking[10, 11, 23]. SLIP_{COP} shows a lower vertical GRF value at mid-stance (F_y), for both of the COP modalities ($\mu = 0$ and 1) compared to SLIP. At $\mu = 0$ the COP translates with constant speed and at $\mu = 1$ the speed results in a U-shape profile which correlates to the horizontal GRF (F_x) (Fig. 4b). The shape of the COP speed trajectory for $\mu = 1$ resembles the COP speed trajectory of human walking. As seen in Fig. 4c, SLIP_{COP} has higher COM amplitude and

horizontal distance travelled than SLIP. This result, as expected, is a consequence of COP progression. A higher gait distance for $\mu = 1$ is due to a larger average COP speed (see Fig. 4b) compared to at $\mu = 0$.

We make inter-model comparisons at a non-dimensionalized speed $\dot{x}_i = 0.335$ and $\theta_o = 76.33^\circ$. We compare the steady state solutions, at the mentioned speed and TD angle, obtained by varying $y_i \in (\sin(\theta_o), 1)$. We discard solutions at $y_i = \sin(\theta_o)$ and $y_i = 1$. Because at $y_i = \sin \theta_o$ the apex height will be equal to the COM height at TD, which is physically impossible. And at $y_i = 1$, the system will be under free fall as the leg will be at its natural uncompressed length suggesting no foot contact with the ground. As seen in Fig. 5a for SLIP, increase in y_i leads to increase in $\tilde{k}$, from 22.75 to 29.54. $SLIP_{COP}$ for both COP modalities shows a considerably lower and constant stiffness for all values of y_i . We expected lower stiffness for $SLIP_{COP}$ because of the leg lengthening that occurs due to a virtual pivot point generated as shown in Fig. 3. During walking, stride length is approximately twice the value of step length as seen in Fig. 5c & e. A higher value for step lengths for $SLIP_{COP}$ is observed: e.g. a value of 0.46 at $y_i = 0.99$ with $\mu = 1$. For walking, cadence c and step length s_l are related to walking speed as $v = (c)(s_l)$ [18]. We see the effect of this hyperbolic relation between cadence c and step length s_l in the plots Fig. 5c & d. The swing/stance duration ratio

141 is around 0.4 for walking, and SLIP achieves this ratio as y_i approaches 1 (see
 142 Fig. 5f). SLIP_{COP} shows a reduced swing/stance duration time which occurs due
 143 to its increased stance time. This increase in stance time occurs as a result of the
 144 COP progression which we expected [16].

145 The reliability of the two models is tested by making model-experiment com-
 146 parisons. In particular, we compared the mean error in between model and exper-
 147 iment data for the following parameters: vertical COM amplitude a , swing/stance
 148 ratio, walking speed v , virtual pivot point (VPP) length factor γ , COP speed and
 149 distance (D_{COP}) travelled as shown in Fig. 6. We use the following equations from
 150 existing literature for experimental walking.

$$a = 0.054v + 0.002 \quad [18] \quad (7)$$

$$v = (c)(s_l) \quad [18] \quad (8)$$

$$L_{VPP} = \gamma L \quad [18] \quad (9)$$

$$D_{COP} = 0.152h \quad [23] \quad (10)$$

151 where c is cadence, s_l is step length, L_{VPP} is distance between COM and virtual
 152 pivot point, γ is the VPP factor usually around 1.8, D_{COP} is distance travelled
 153 by the COP during stance and h is the height of the human. Winter et al. [23]

154 provided measures of ratios of different body parts with respect to human height.
 155 Such a ratio for foot measure is shown in eq.(10). Swing/stance ratio is calculated
 156 by dividing the swing time of a particular leg by its stance time during a gait
 157 cycle. To compare the models with experiment data of adults we dimensionalize
 158 the parameters. Walking speed is dimensionalized by multiplying $\dot{x}$ with $\sqrt{gl_o}$,
 159 where $l_o=1\text{m}$ is the uncompressed leg length. In Fig. 5, we compared solutions
 160 at every apex y_i at $\dot{x}_i$ 0.335 and $\theta_o = 76.33^\circ$. In Fig. 6 we calculate the average
 161 of all limit cycle solutions for all dimensionalized apex speeds to make model-
 162 experiment comparison.

163 One of the objectives of our study was to see if the relation among TD angle,
 164 walking speed and COP speed provides steady state solutions for the given adult
 165 speed range². We get steady state solutions for very slow to slow walking speeds
 166 for SLIP and very slow to very fast walking speeds for SLIP_{COP}. For SLIP_{COP}
 167 with both μ values, we see a larger error for COM amplitude and swing/stance
 168 ratio at very slow to normal walking speeds, compared to SLIP. But the error
 169 decreases as the walking speed increases. SLIP estimates COM amplitude much

²We classify walking speeds as very slow (0.7-1.12 m/s), slow (1.12-1.31 m/s), normal (1.31-1.58 m/s), fast (1.58-1.76 m/s) and very fast (1.76-2.19 m/s) based on the classification provided by [24].

170 better than $SLIP_{COP}$ at lower speed ranges because its optimized spring stiffness
 171 values lie close to human leg stiffness. Fig. 6b illustrates the error of swing/stance
 172 ratio. At lower walking speed, both models perform similarly. As walking speed
 173 increases, the errors for both models decrease. At the normal walking speeds SLIP
 174 provides less error than $SLIP_{COP}$; at 1.36 m/s, 25.72% error for SLIP, 55.17% for
 175 $SLIP_{COP}$ ($\mu = 0$), and 61.13% for $SLIP_{COP}$ ($\mu = 1$). For higher walking speeds,
 176 $SLIP_{COP}$ outperforms (error below 25%) SLIP, which fails to find a solution.

177 The concept of virtual pivot point (VPP) is illustrated in Fig. 3 and expressed
 178 in eq.(9). In Fig. 6d the VPP factor (γ) is compared with the physiologically
 179 measured value of 1.8 provided by [18]. As expected, SLIP model provides γ
 180 which is equal to 1 under all scenarios because of its fixed pivot. $SLIP_{COP}$ ($\mu =$
 181 0, 1) provides more accurate estimates of γ with approximately 20% error. VPP
 182 is a good metric to measure the effectiveness of our COP progression model. The
 183 error remains constant at 20% for most of the speed range but decreases to 10% at
 184 very fast speed ran (see Fig. 6e). D_{COP} is calculated, for an adult with an average
 185 height of 1.7 meters [11] with an uncompressed leg length, $L_o = 1m$. In Fig. 6f
 186 we observe, $SLIP_{COP}$ shows quite a low error except at very slow and very fast
 187 walking speeds: the least error of 4% at 0.75 m/s with $\mu = 1$ and 7% at 1.4m/s
 188 with $\mu = 0$.

189 4 Discussion

190 Through this study we compared the effect of adding a translating COP to the
 191 conventional SLIP model. The motivation behind using a translating COP was to
 192 simulate the effects of the heel-to-toe pivoting of the foot during stance phase in
 193 human walking. One of the objectives of our study was to improve the predictive
 194 capabilities of SLIP for a wider speed range² when comparing with experimental
 195 data. Utilizing experimental data (of the relation among walking speed, TD angle,
 196 and COP speed) enables us to obtain limit cycle solutions at these speed ranges.
 197 The relation between TD angle and walking speed also suggests that as the walk-
 198 ing speed approaches 0, TD angle approaches 90 degrees (erect standing), which
 199 can be considered as decent validation of our walking speed and TD angle rela-
 200 tion. This shows that as the walking speed approaches a lower value step length
 201 approaches 0. Lipfert et al. [11] showed that to obtain similar walking dynamics
 202 at a given walking speed, the SLIP model was simulated at a steeper angle be-
 203 cause of premature lift off at the correct TD angle. With the relation between TD
 204 angle and walking this limitation is overcome. On the other hand, provision of
 205 COP speed-walking speed relation leads to increase in stance time. When com-
 206 paring the 2 models at similar optimized state variables, we see an increase in gait
 207 distance for SLIP_{COP} because of increase in its stance time. To comment upon

208 gait distance estimation of the 2 models we dimensionalize the result in Fig. 4c
 209 with a leg length $L_o = 1m$ and speed of 1m/s so as to compare to a previous
 210 model-experiment study using SLIP [11]. Upon comparing, the COM trajectories
 211 in Fig. 4c, we observe that $SLIP_{COP}$ estimates the gait distance with an error of
 212 0.05 m and SLIP with an error of 0.25 m [11]. One of the reasons for this un-
 213 derestimation by SLIP could be its fixed pivot point. Although we used constant
 214 stiffness in our models, we preferred optimizing the spring stiffness rather than
 215 using predefined leg stiffness [11]. This was done because the relation among TD
 216 angle, walking speed, and COP speed could affect the optimal value of stiffness.
 217 We understand that the human walking gait is a consequence of the stabilization
 218 occurring at the foot. This has led us to put more emphasis on the relation among
 219 TD angle, walking speed, and COP speed rather than on leg stiffness as done by
 220 Lipfert and colleagues. Although human muscular strength determines the flex-
 221 ion and extension of our lower limbs during walking, it is difficult to measure this
 222 strength just by observation. Through inverse dynamics we can utilize observable
 223 kinematic and dynamic characteristics to understand more about the functioning
 224 of human walking. The COP speed trajectory for the accelerated COP modality
 225 shows a similar trend in COP speed as shown in the study by Cornwall and col-
 226 leagues [25], with high speed at initial contact phase, lower speed at mid stance

227 and higher speed at TD. This U-shape speed profile correlates with the horizontal
228 COM acceleration F_x because COM decelerates in the first half of stance phase
229 and then accelerates in the next half. This suggests that our weighted function
230 approximates the acceleration of the COP quite well.

231 In Fig. 4a & c we see the effects of a translating COP which leads to higher
232 variation in vertical GRF and COM amplitude respectively, compared to SLIP.
233 Such behavior was also observed in the IP model [17] and POFT model [16],
234 where the addition of a translating COP increases the vertical displacement signif-
235 icantly. Bullimore et al. [16] also mention that due to a translating COP the stance
236 time for a leg increases which we also observe in $SLIP_{COP}$. They also mentioned
237 that addition of a COP progression model decreases the spring stiffness of the
238 model. The decrease in spring stiffness can be explained by eqns.(11)(12)(13). In
239 Fig. 4a, F_y at mid stance (apex) is lower for $SLIP_{COP}$ than SLIP. At apex the COM
240 experiences centripetal acceleration due to its weight and spring force. Hence,
241 upon referring Fig. 2, 3 and Table 2, the force balance for SLIP and $SLIP_{COP}$ at
242 apex is

$$m \frac{\dot{x}_{id}^2}{y_i} = k(L_o - y_{id}) - mg \quad (11)$$

$$m \frac{\dot{x}_{id}^2}{L_{VPP}} = k_{VPP}(L_o - y_{id}) - mg \quad (12)$$

Subtracting eqns.(12) from (11) we get,

$$m \frac{\dot{x}_{id}^2}{y_i} - m \frac{\dot{x}_{id}^2}{L_{VPP}} = k(L_o - y_{id}) - k_{VPP}(L_o - y_{id}) \quad (13)$$

243 As both sides of the eqn.(13) are positive with $y_i < L_{VPP}$, this implies $k > k_{VPP}$.

244 To reduce the vertical displacement occurring due to reduced spring stiffness,
245 Bullimore et al. [16] added a constraint on the vertical movement of the COM.
246 Constraining the vertical displacement for our models resulted a difficulty to find
247 limit cycle solutions and hence we relaxed this constraint. Lee et al. [17] showed
248 that with increasing walking speeds the error in COM vertical displacement in-
249 creases when compared to experimental data. We observe a decrease in error for
250 COM vertical displacement, at $\mu = 0$ and 1, with increasing walking speeds which
251 shows the effectiveness of our bipedal model. The IP model in the above studies
252 was simulated only for single stance which could have limited its predictive nature
253 unlike the SLIP and SLIP_{COP}.

254 One of the characteristics of walking is the relation between cadence and
255 step length represented by eqn.(8). We obtain quite low errors for both mod-
256 els for walking speed using eqn.(8) as seen in Fig. 6c. We observe an increase
257 in step length and decrease in cadence for SLIP_{COP} which was expected in our
258 study. With COP progression the distance between consecutive heel strikes in-
259 creases subsequently increasing the step length. This in turn reduces cadence (re-

fer eqn.(8)). As discussed before, due to COP progression we have an increase in stance time which is also responsible for decrease in cadence because is defined as steps per min. One more factor that is characteristic of a progressive COP model is the generation of a virtual pivot point as discussed above (Fig. 3). To the best of our knowledge, there exists no study with SLIP model that has estimated the VPP factor γ . The average γ value with our proposed $SLIP_{COP}$ is around 1.4 with a 20% mean error, where the range of γ is between 1.33 and 2.1. To put the value of γ into perspective, we evaluate the distance travelled by the COP during stance. As COP travels approximately a foot length [12], we evaluated the COP distance D_{COP} for our speed range. The accelerated COP modality shows a considerably lower error values than the constant velocity modality for very slow to normal walking speeds. One of our objectives was to differentiate between the two COP modalities $\mu = 0$ and 1. With D_{COP} we see that for very slow to slow speeds $\mu = 1$ provides better estimation and for normal walking speeds $\mu = 0$ is better. Overall the accelerated model shows lower error value for D_{COP} for majority of speeds suggesting its reliability over the constant velocity modality.

We propose a bipedal spring mass model utilizing the COP translation observed during human walking. We compare this model with the SLIP model with respect to human walking data. We observe that the SLIP and $SLIP_{COP}$ show

279 pretty high error estimates for COM vertical amplitude and swing/stance ratio at
 280 very slow to slow walking speeds. At normal to very fast walking speeds, we see
 281 the benefits of the $SLIP_{COP}$ as it not only provides limit cycle solutions for these
 282 speed zones but also considerably decreases error in predicting COM amplitude
 283 and swing/stance duration ratio. $SLIP_{COP}$ is able to reproduce a symmetrical COP
 284 speed profile in the fore-aft direction. The distance traveled by the COP for the
 285 two COP progression modes at normal walking speeds concurs with distance trav-
 286 eled by the COP during human walking and can be considered as a substitute for
 287 an ankle based walking model. This pilot study on using a translating COP based
 288 $SLIP$ model takes into consideration the fact that COP movement is closely related
 289 to the GRF force in the horizontal direction.

290 In the future work, this model can be further developed by utilizing actual COP
 291 data from human walking, which could enhance our models capabilities from the
 292 point of view of simulating slow to normal walking speeds. This study will be
 293 also undertaken from the point of view of assessing gaits in people with move-
 294 ment disorder such as Cerebral Palsy, Stroke and Parkinson's. People with such
 295 movement disorders often portray unequal strength in their legs. This affects their
 296 walking style, foot placement consequently affecting their COP dynamics. De-
 297 veloping our COP model towards adapting it to assess these movement disorders

298 would help us understand the difference between healthy and impaired walking
299 styles.

300 **List of abbreviations**

Abbreviations

GRF	Ground reaction force
SLIP	Spring loaded inverted pendulum
COM	Center of mass
COP	Center of pressure
TD	Touch down
IP	Inverted pendulum
POFT	Point of force translation
LO	Lift off
VPP	Virtual pivot point

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304 **Competing interests**

305 We declare that this manuscript is original and has not been published before. We
306 know of no conflict of interest associated with this publication.

307 **Author contributions**

308 Karna Potwar conducted the research, dealing with generating the hypothesis as
309 well as simulating and analyzing the bipedal walking model. He also documented
310 and edited the manuscript. Dongheui Lee supervised this research and critiqued
311 the manuscript.

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References

- [1] R. J. Full and D. E. Koditschek, “Templates and anchors: neuromechanical hypotheses of legged locomotion on land,” *Journal of experimental biology*, vol. 202, no. 23, pp. 3325–3332, 1999.
- [2] R. M. Alexander, “Mechanics of bipedal locomotion,” in *Zoology*. Elsevier, 1976, pp. 493–504.
- [3] S. Mochon and T. A. McMahon, “Ballistic walking,” *Journal of biomechanics*, vol. 13, no. 1, pp. 49–57, 1980.
- [4] R. Blickhan, “The spring-mass model for running and hopping,” *Journal of biomechanics*, vol. 22, no. 11-12, pp. 1217–1227, 1989.
- [5] T. A. McMahon and G. C. Cheng, “The mechanics of running: how does stiffness couple with speed?” *Journal of biomechanics*, vol. 23, pp. 65–78, 1990.
- [6] S. Siegler, R. Seliktar, and W. Hyman, “Simulation of human gait with the aid of a simple mechanical model,” *Journal of biomechanics*, vol. 15, no. 6, pp. 415–425, 1982.

- 332 [7] M. Van Gurp, H. Schamhardt, and A. Crowe, “The ground reaction force
333 pattern from the hindlimb of the horse simulated by a spring model,” *Cells*
334 *Tissues Organs*, vol. 129, no. 1, pp. 31–33, 1987.
- 335 [8] M. G. Pandy, “Simple and complex models for studying muscle function
336 in walking,” *Philosophical Transactions of the Royal Society of London B:*
337 *Biological Sciences*, vol. 358, no. 1437, pp. 1501–1509, 2003.
- 338 [9] F. E. Zajac, R. R. Neptune, and S. A. Kautz, “Biomechanics and muscle
339 coordination of human walking: part ii: lessons from dynamical simulations
340 and clinical implications,” *Gait & posture*, vol. 17, no. 1, pp. 1–17, 2003.
- 341 [10] H. Geyer, A. Seyfarth, and R. Blickhan, “Compliant leg behaviour explains
342 basic dynamics of walking and running,” *Proceedings of the Royal Society*
343 *of London B: Biological Sciences*, vol. 273, no. 1603, pp. 2861–2867, 2006.
- 344 [11] S. W. Lipfert, M. Günther, D. Renjewski, S. Grimmer, and A. Seyfarth, “A
345 model-experiment comparison of system dynamics for human walking and
346 running,” *Journal of Theoretical Biology*, vol. 292, pp. 11–17, 2012.
- 347 [12] M.-C. Chiu, H.-C. Wu, and L.-Y. Chang, “Gait speed and gender effects
348 on center of pressure progression during normal walking,” *Gait & posture*,
349 vol. 37, no. 1, pp. 43–48, 2013.

- 350 [13] S. Fuchioka, A. Iwata, Y. Higuchi, M. Miyake, S. Kanda, and T. Nishiyama,
351 “The forward velocity of the center of pressure in the midfoot is a major pre-
352 dictor of gait speed in older adults,” *International Journal of Gerontology*,
353 vol. 9, no. 2, pp. 119–122, 2015.
- 354 [14] M. A. Sharbafi and A. Seyfarth, *Bioinspired Legged Locomotion: Models*,
355 *Concepts, Control and Applications*. Butterworth-Heinemann, 2017.
- 356 [15] D. Maykranz, S. Grimmer, S. Lipfert, and A. Seyfarth, “Foot function in
357 spring mass running,” in *Autonome Mobile Systeme*. Springer, 2009, pp.
358 81–88.
- 359 [16] S. R. Bullimore and J. F. Burn, “Consequences of forward translation of the
360 point of force application for the mechanics of running,” *Journal of theoret-*
361 *ical biology*, vol. 238, no. 1, pp. 211–219, 2006.
- 362 [17] C. R. Lee and C. T. Farley, “Determinants of the center of mass trajectory
363 in human walking and running,” *Journal of experimental biology*, vol. 201,
364 no. 21, pp. 2935–2944, 1998.
- 365 [18] S. Miff, S. A. Gard, and D. Childress, “The effect of step length, cadence,
366 and walking speed on the trunk’s vertical excursion,” in *Proceedings of the*
367 *22nd Annual International Conference of the IEEE Engineering in Medicine*

- 368 *and Biology Society (Cat. No. 00CH37143)*, vol. 1. IEEE, 2000, pp. 155–
369 158.
- 370 [19] R. Alexander, “Optimization and gaits in the locomotion of vertebrates,”
371 *Physiological reviews*, vol. 69, no. 4, pp. 1199–1227, 1989.
- 372 [20] S. Shen and J. Seipel, “Effective leg stiffness of animal running and the co-
373 optimization of energetic cost and stability,” *Journal of theoretical biology*,
374 vol. 451, pp. 57–66, 2018.
- 375 [21] Z. Shen and J. Seipel, “A fundamental mechanism of legged locomotion with
376 hip torque and leg damping,” *Bioinspiration & biomimetics*, vol. 7, no. 4, p.
377 046010, 2012.
- 378 [22] S. Strogatz, *Nonlinear dynamics and chaos: with application to physics*,
379 *biology, chemistry and engineering*. CRC press, 2018.
- 380 [23] D. A. Winter, *Biomechanics and motor control of human movement*. John
381 Wiley & Sons, 2009.
- 382 [24] R. Drillis, “Objective recording and biomechanics of pathological gait,” *An-*
383 *nals of the New York Academy of Sciences*, vol. 74, no. 1, pp. 86–109, 1958.

- 384 [25] M. W. Cornwall and T. G. McPoil, “Velocity of the center of pressure during
385 walking,” *Journal of the American Podiatric Medical Association*, vol. 90,
386 no. 7, pp. 334–338, 2000.

Figure 1: Diagram showing the two models, (Left) SLIP and (Right) SLIP with translating COP (SLIP_{COP}) with COM and COP coordinates in the sagittal plane.

Subscripts r and l stand for right and left leg respectively.

Figure 2: A limit cycle of the translating COP model. The model starts at the apex i and attains the consecutive apex $i + 1$, while the COP translates along the ground.

Figure 3: Translation of the COP in $SLIP_{COP}$ leads to a virtual pivot point (VPP)

under the surface. l_v is the extended length where $l_v=1.8l$ [18] during walking .

Figure 4: GRF, COP speed and COM trajectory plotted for SLIP and SLIP_{COP} ($\mu = 0$ and $\mu = 1$) with parameters in Table 1.

Figure 5: Plotting temporal and distance variables for different values of y_i as at an apex speed $\dot{x}_i$ of 0.335 and θ_o of 76.33° .

Figure 6: Mean errors of the obtained model parameters (amplitude, swing/stance duration ratio, average walking speed, VPP factor, average COP speed and COP distance) with respect to experimental human data.

Parameters	SLIP	SLIP _{COP}	
		$\mu = 0$	$\mu = 1$
$\dot{x}_i$	0.335	0.335	0.335
y_i	0.986	0.986	0.986
θ_o	76.33	76.33	76.33
$\dot{f}_i$	0	0.105	0.087
$\tilde{k}$	28.67	13.39	11.48

Table 1: Simulation parameters for results in Figure 5. $x_i = \dot{y}_i = 0$

Fitted Parameter	Equation
Touchdown angle, θ_o	$\theta_o = 0.36\dot{x}_i^3 + 0.25\dot{x}_i^2 - 0.84\dot{x}_i + 1.57$
COP velocity lower bound, $\dot{f}_{LB}$	$\dot{f}_{LB} = 0.21\dot{x} - 0.002$
COP velocity upper bound, $\dot{f}_{UB}$	$\dot{f}_{UB} = 0.74\dot{x} + 0.22$

Table 2: Parameter relations obtained from physiological data[11, 12]. Subscript

LB and UB stand for lower bound and upper bound respectively.