Simulating bipedal walking using a

translating center of pressure

Karna Potwar¹ and Dongheui Lee^{1,2}

¹ Technical University of Munich (TUM)

² German Aerospace Center (DLR)

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Abstract

1

2 During walking, foot orientation and foot placement allow humans to stabilize their gait and to move forward. Consequently the upper body adapts 3 to the ground reaction force (GRF) transmitted through the feet. The foot-4 ground contact is often modeled as a fixed pivot in bipedal models for anal-5 ysis of locomotion. The fixed pivot models, however, cannot capture the 6 effect of shift in the pivot point from heel to toe. In this study, we propose 7 a novel bipedal model, called $SLIP_{COP}$, which employs a translating center 8 of pressure (COP) in a spring loaded inverted pendulum (SLIP) model. The 9 translating COP has two modes: one with a constant speed of translation and 10 the other as the weighted function of the GRF in the fore aft direction. We 11 use the relation between walking speed and touchdown (TD) angle as well 12 as walking speed and COP speed, from existing literature, to restrict steady 13 state solutions within the human walking domain. We find that with these 14 relations, SLIP_{COP} provides steady state solutions for very slow to very fast 15 walking speeds unlike SLIP. SLIP_{COP} for normal to very fast walking speed 16 shows good accuracy in estimating COM amplitude and swing stance ratio. 17 SLIP_{COP} is able to estimate the distance traveled by the COP during stance 18 with high precision. 19

20 1 Introduction

Walking is an efficient form of locomotion which allows humans to travel from 21 one place to another. Human walking has two facets, which are reducing cost of 22 locomotion and gait stabilization. Foot placement and orientation plays an im-23 portant role in stabilizing our gait. To improve our understanding of walking and 24 its underlying mechanism, human motion capturing and reductive modeling have 25 been widely used [1]. Human body is a complex redundant system and reductive 26 models or templates strip down the complex architecture of the body into simple 27 elements and allow a computationally inexpensive way to simulate locomotion, 28 with a certain degree of accuracy [1]. The earliest versions of such templates 29 include the sagittal plane based inverted pendulum model (IP) [2, 3] and spring-30 mass model [4, 5]. The IP model assumes incompressible legs with center of mass 31 (COM) vaulting over the legs, with a fixed foot. Due to its rigid legs, IP model 32 provides an incorrect representation of ground reaction force (GRF) pattern, com-33 pared to that observed in humans. To overcome this drawback of the IP model, 34 some templates include a springy leg to obtain more accurate estimates of walking 35 gait trajectories [6, 7, 8, 9, 10]. One such model is the sagittal plane based SLIP 36 model [10]. Due to SLIP's compliant legs it is able to generate the COM trajectory 37 and GRF pattern during walking to that observed in human walking. The steady 38

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

IP model and SLIP model assume a fixed pivot during stance as discussed 42 above, which is essentially the mean position traveled by COP during stance. In 43 a study to analyze treadmill walking¹, SLIP fairly estimates COM trajectory at 44 1m/s, while failing to estimate at other walking speeds [11]. They also men-45 tion that due to the fixed pivot of the SLIP, the model has to be simulated at a 46 steeper TD angle compared to human walking. This characteristic of SLIP might 47 be restricting its predictive capabilities at slower and faster walking speed. During 48 walking, COP of a particular foot travels approximately a distance of a foot length 49 and the COP progression velocity depends on speed of walking [12, 13, 14]. To 50 accommodate the mechanical consequences of COP progression, a few bipedal 51 models were developed for running [15, 16] and walking [17, 18]. Bullimore et 52 al. [16] take into consideration the change in TD and LO angles caused by COP 53 translation, called POFT (point of force translation), in the conventional spring 54 mass model for running. They use a constant velocity based COP progression 55 model without considering the acceleration of the COP during stance. They ob-56 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 57 model shows drastic decrease in spring stiffness. Lee et al. [17] use a translating 58 point of force application (PFA) in an IP model for walking and show that the 59 error in vertical displacement of the COM predicted by the model increases from 60 111% to 240%, as walking speed increases from 0.5 m/s to 2.5 m/s respectively. 61 However, these errors were considerably lesser compared to IP model with a fixed 62 pivot. Miff et al. [18] show that during walking vertical excursion of the trunk is 63 dependent on the foot rocker radius in the rocker based IP model. IP models in 64 the above studies consider only the single stance vaulting of the COM but not the 65 foot impact and double stance phase of walking. We need a bipedal model which 66 can simulate COP progression along with single/double stance, COM trajectory 67 and GRF patterns. 68

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

⁷⁶ other accelerated COP. We include the constant velocity COP progression model ⁷⁷ so as to compare our model with previous similar models. We make inter-model ⁷⁸ comparisons between SLIP and SLIP_{COP} for various walking parameters to an-⁷⁹ alyze the results qualitatively and quantitatively. Subsequently, we compare the ⁸⁰ two models with real walking scenarios to assess the quality of our solutions.

81 **2 Method**

We simulate the two models, SLIP and SLIP_{COP}, as seen in Fig. 1. The position 82 and velocity of right and left COP are denoted as $[f_r, \dot{f}_r]$ and $[f_l, \dot{f}_l]$ respectively. 83 For SLIP, \dot{f}_r and \dot{f}_l are always 0 due to a fixed pivot. Like the conventional SLIP, 84 SLIP_{COP} consists of a COM attached with two springy legs. The legs are con-85 sidered massless and a swinging leg can be ignored. As illustrated in Fig. 2, 86 COM state at apex is described by $[x, \dot{x}, y, \dot{y}]$ and at TD the leg makes an angle of 87 θ_o . Both models are simulated in the sagittal plane. We non-dimensionalize the 88 equations of motion to develop generic models catering to humans with different 89 anthropometric measurements [19, 20, 21]. Force experienced by the COM before 90 non-dimensionalization in the forward and vertical direction is given as 91

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

$$F_{y_d} = m\ddot{y}_d = F_{rd}\sin\alpha + F_{ld}\sin\beta - mg \tag{2}$$

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 nondimensionalizing 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

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

$$F_{y} = \ddot{y} = \tilde{F}_{r} \sin \alpha + \tilde{F}_{l} \sin \beta - 1 \tag{4}$$

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 TD, the leg angle reorients to θ_o and at lift off (LO) occurs when the GRF becomes 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

$$\ddot{f}_r = \mu F_r \cos \alpha \tag{5}$$

$$\ddot{f}_l = \mu F_l \cos \beta \tag{6}$$

Especially, two modes of this translating COP model are considered: one considering effect of a constant COP speed during stance ($\mu = 0$) and the other as 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.

111 3 Results

Firstly, we compare COM, COP and GRF trajectories for the two models (SLIP, 112 $SLIP_{COP}$) for a given set of optimized parameters (see Table 1) to assess the qual-113 itative nature of the solutions. Fig. 4a & b are GRF and COP trajectories for 114 individual legs. GRF pattern in vertical and horizontal direction resemble that of 115 experimental walking[10, 11, 23]. SLIP_{COP} shows a lower vertical GRF value at 116 mid-stance (F_v), for both of the COP modalities ($\mu = 0$ and 1) compared to SLIP. 117 At $\mu = 0$ the COP translates with constant speed and at $\mu = 1$ the speed results 118 in a U-shape profile which correlates to the horizontal GRF (F_x) (Fig. 4b). The 119 shape of the COP speed trajectory for $\mu = 1$ resembles the COP speed trajectory 120 of human walking. As seen in Fig. 4c, SLIP_{COP} has higher COM amplitude and 121

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 $\dot{x}_i = 0.335$ 125 and $\theta_o = 76.33^\circ$. We compare the steady state solutions, at the mentioned speed 126 and TD angle, obtained by varying $y_i \in (\sin(\theta_o), 1)$. We discard solutions at 127 $y_i = sin(\theta_o)$ and $y_i = 1$. Because at $y_i = sin \theta_o$ the apex height will be equal to 128 the COM height at TD, which is physically impossible. And at $y_i = 1$, the system 129 will be under free fall as the leg will be at its natural uncompressed length sug-130 gesting no foot contact with the ground. As seen in Fig. 5a for SLIP, increase in 131 y_i leads to increase in \hat{k} , from 22.75 to 29.54. SLIP_{COP} for both COP modalities 132 shows a considerably lower and constant stiffness for all values of y_i . We ex-133 pected lower stiffness for $SLIP_{COP}$ because of the leg lengthening that occurs due 134 to a virtual pivot point generated as shown in Fig. 3. During walking, stride length 135 is approximately twice the value of step length as seen in Fig. 5c & e. A higher 136 value for step lengths for SLIP_{COP} is observed: e.g. a value of 0.46 at $y_i = 0.99$ 137 with $\mu = 1$. For walking, cadence c and step length s_l are related to walking speed 138 as $v = (c)(s_l)$ [18]. We see the effect of this hyperbolic relation between cadence 139 c and step length s_l in the plots Fig. 5c & d. The swing/stance duration ratio 140

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

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

$$a = 0.054v + 0.002 \quad [18] \tag{7}$$

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

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

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

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

provided measures of ratios of different body parts with respect to human height. 154 Such a ratio for foot measure is shown in eq.(10). Swing/stance ratio is calculated 155 by dividing the swing time of a particular leg by its stance time during a gait 156 cycle. To compare the models with experiment data of adults we dimensionalize 157 the parameters. Walking speed is dimensionalized by multiplying \dot{x} with $\sqrt{gl_o}$, 158 where $l_0=1$ m is the uncompressed leg length. In Fig. 5, we compared solutions 159 at every apex y_i at \dot{x}_i f 0.335 and $\theta_o = 76.33^o$. In Fig. 6 we calculate the average 160 of all limit cycle solutions for all dimensionalized apex speeds to make model-161 experiment comparison. 162

¹⁶³ One of the objectives of our study was to 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].

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

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

189 4 Discussion

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

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

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

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

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

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

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

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

As both sides of the eqn.(13) are positive with $y_i < L_{VPP}$, this implies $k > k_{VPP}$. 243 To reduce the vertical displacement occurring due to reduced spring stiffness, 244 Bullimore et al. [16] added a constraint on the vertical movement of the COM. 245 Constraining the vertical displacement for our models resulted a difficulty to find 246 limit cycle solutions and hence we relaxed this constraint. Lee et al. [17] showed 247 that with increasing walking speeds the error in COM vertical displacement in-248 creases when compared to experimental data. We observe a decrease in error for 249 COM vertical displacement, at $\mu = 0$ and 1, with increasing walking speeds which 250 shows the effectiveness of our bipedal model. The IP model in the above studies 251 was simulated only for single stance which could have limited its predictive nature 252 unlike the SLIP and SLIP_{COP}. 253

One of the characteristics of walking is the relation between cadence and step length represented by eqn.(8). We obtain quite low errors for both models for walking speed using eqn.(8) as seen in Fig. 6c. We observe an increase in step length and decrease in cadence for $SLIP_{COP}$ which was expected in our study. With COP progression the distance between consecutive heel strikes increases 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 260 stance time which is also responsible for decrease in cadence because is defined as 261 steps per min. One more factor that is characteristic of a progressive COP model 262 is the generation of a virtual pivot point as discussed above (Fig. 3). To the best 263 of our knowledge, there exists no study with SLIP model that has estimated the 264 VPP factor γ . The average γ value with our proposed SLIP_{COP} is around 1.4 with 265 a 20% mean error, where the range of γ is between 1.33 and 2.1. To put the value 266 of γ into perspective, we evaluate the distance travelled by the COP during stance. 267 As COP travels approximately a foot length [12], we evaluated the COP distance 268 D_{COP} for our speed range. The accelerated COP modality shows a considerably 269 lower error values than the constant velocity modality for very slow to normal 270 walking speeds. One of our objectives was to differentiate between the two COP 271 modalities $\mu = 0$ and 1. With D_{COP} we see that for very slow to slow speeds $\mu = 1$ 272 provides better estimation and for normal walking speeds $\mu = 0$ is better. Overall 273 the accelerated model shows lower error value for D_{COP} for majority of speeds 274 suggesting its reliability over the constant velocity modality. 275

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

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

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

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
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307 Author contributions

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

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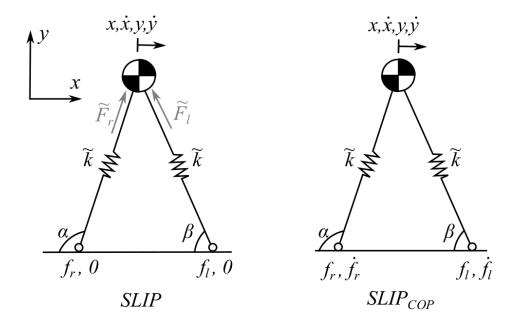


Figure 1: Diagram showing the two models, (Left) SLIP and (Right) SLIP with translating COP (SLIP $_{COP}$) with COM and COP coordinates in the sagittal plane. Subscripts r and l stand for right and left leg respectively.

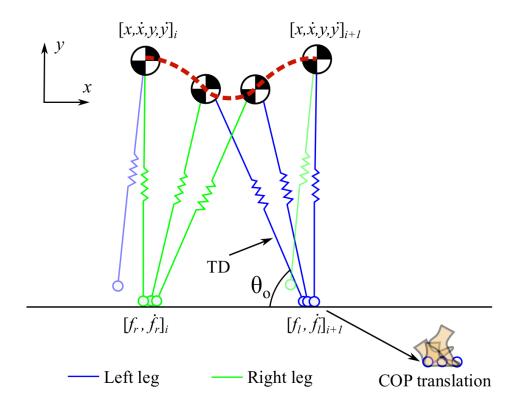


Figure 2: A limit cycle of the translating COP model. The model starts at the apex i and attains the consecutive apex i + 1, while the COP translates along the ground.

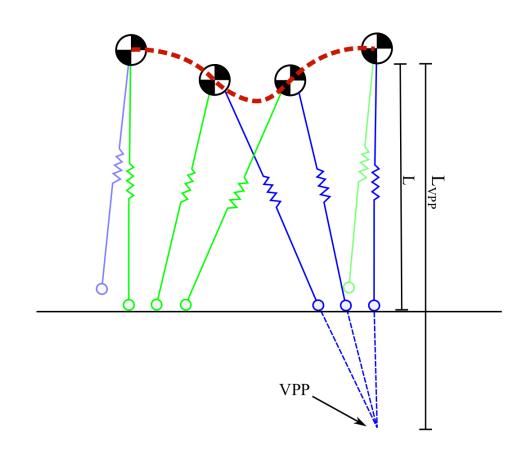


Figure 3: Translation of the COP in $SLIP_{COP}$ leads to a virtual pivot point (VPP) under the surface. l_v is the extended length where $l_v=1.8l$ [18] during walking.

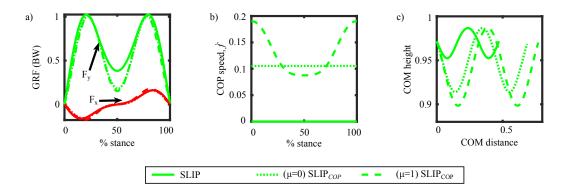


Figure 4: GRF, COP speed and COM trajectory plotted for SLIP and SLIP_{COP} $(\mu = 0 \text{ and } \mu = 1)$ with parameters in Table 1.

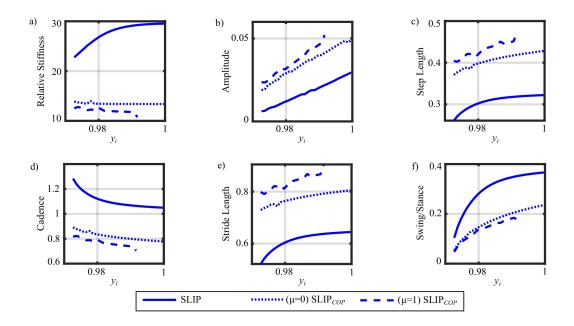


Figure 5: Plotting temporal and distance variables for different values of y_i as at an apex speed \dot{x}_i of 0.335 and θ_o of 76.33°.

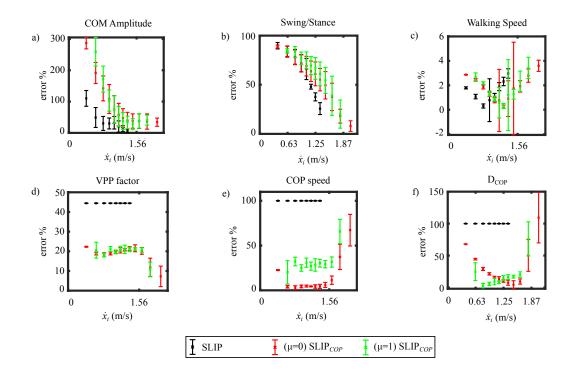


Figure 6: Mean errors of the obtained model parameters (amplitude, swing/stance duration ratio, average walking speed, VPP factor, average COP speed and COP distance) with respect to experimental human data.

Demonstration		SLIP _{COP}		
Parameters	SLIP	$\mu = 0$	$\mu = 1$	
$\dot{x_i}$	0.335	0.335	0.335	
yi	0.986	0.986	0.986	
$ heta_o$	76.33	76.33	76.33	
$\dot{f_i}$	0	0.105	0.087	
Ĩ	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.