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1 **Quantifying the hip-ankle synergy in short-term maximal cycling**

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12

13 **Abstract**

14 Simulation studies have demonstrated that hip and ankle joints form a task-specific synergy
15 during downstrokes in maximal cycling to enable power produced by hip extensor muscles to
16 be transferred to the crank. Existence of the hip-ankle synergy has not been investigated
17 experimentally. Therefore, we sought to apply a modified vector coding technique to quantify
18 the strength of the hip-ankle moment synergy in the downstroke during short-term maximal
19 cycling at a pedalling rate of 135 rpm. Twelve track sprint cyclists performed 3 x 4 s seated
20 sprints at 135 rpm. Joint moments were calculated via inverse dynamics, using pedal forces
21 and limb kinematics. The hip-ankle moment synergy was quantified using a modified vector
22 coding method. Results showed, for 28.8% of the downstroke, hip and ankle moments were
23 in-phase, demonstrating the hip and ankle joints tend to work in synergy in the downstroke,
24 supporting findings from simulation studies of cycling. At a pedalling rate of 135 rpm the
25 hip-phase was most frequent (42.5%), significantly differing from the in- ($P = 0.044$), anti-
26 ($P < 0.001$), and ankle-phases ($P = 0.004$), demonstrating hip-dominant action. Our results
27 experimentally confirm the hip-ankle synergy, indicating an important mechanism that allows
28 production of crank power during maximal sprint cycling.

29 **Keywords:** joint moments, movement coordination, sprint cycling, vector coding.

30 **1 Introduction**

31 The goal of short-term maximal cycling is to maximise mechanical power output delivered to
32 the crank (van Soest & Casius, 2000). To achieve this, muscle and joint actions need to be
33 coordinated to facilitate energy transfer from muscles through body segments to deliver
34 maximum effective crank force (Raasch et al., 1997). Uni-articular hip and knee extensor
35 muscles (gluteus maximus and vastii) are the primary power producers in maximal cycling

36 (Dorel et al., 2012; Martin & Nichols, 2018; Raasch et al., 1997; van Ingen Schenau et al.,
37 1992). Simulation studies have demonstrated that hip extensor muscles (gluteus maximus)
38 produce energy in the downstroke, transferred to the limb (Fregly & Zajac, 1996; Raasch et
39 al., 1997). Additionally, ankle plantar-flexor muscles (gastrocnemius and soleus) need to be
40 co-excited with hip extensors to form a synergy to transfer this energy to the crank (Dorel et
41 al., 2012; Raasch et al., 1997). Without this co-excitation, simulations indicate that energy
42 produced by hip extensors would simply accelerate limbs (dorsiflexing the ankle and
43 hyperextending the knee), rather than being transferred into effective crank force (Raasch et
44 al., 1997). Knee extensors (vastii) are able to transfer most of the energy they generate
45 directly to the crank (Raasch et al., 1997). Martin and Nichols (2018) provided further
46 evidence for this functional coordination mechanism using simulated work loops. They
47 demonstrated that the ankle has a different role to knee and hip joints in maximal cycling -
48 acting to transfer - instead of maximising muscle power (Martin & Nichols, 2018). However,
49 existence of the hip-ankle synergy has not been verified experimentally. Hence, developing a
50 method to experimentally quantify the strength of this synergy in cycling performance is
51 important.

52 Vector coding can be used to quantify inter-segment, inter-joint and inter-limb coordination
53 (Bayne, 2020; Chang et al., 2008; Hamill et al., 2000; Wheat & Glazier, 2006). It has been
54 used to quantify inter-segment coordination between experienced and less experienced
55 runners (Hafer et al., 2019), anticipated and unanticipated sidestepping (Weir et al., 2019)
56 and during gait (Chang et al., 2008; Needham et al., 2014). Vector coding can identify and
57 quantify coordination differences between-participants and movements, providing insights
58 into coordination patterns not evident from joint or segment angle data alone (Needham et al.,
59 2014; Wheat & Glazier, 2006). Vector coding of joint moment data could provide a useful

60 methodology to quantify strength of hip-ankle joint moment synergies in short-term maximal
61 cycling.

62 This study aimed to apply a vector coding technique to quantify strength of hip-ankle
63 moment synergies in downstrokes during short-term maximal cycling at a pedalling rate of
64 135 rpm.

65 **2 Methods**

66 **2.1 Participants**

67 Twelve, competitively-experienced, track sprint cyclists, at U-23 international level (n=5),
68 Master's international and national levels (n=4), or Junior national level (n=3) participated in
69 this study. Participants were varied in sex, age and anthropometrics (4 males and 8 females,
70 age: 24.1 ± 13.8 yr, body mass: 68.2 ± 11.1 kg, height: 1.70 ± 0.07 m), and were similar in
71 cycling performance level (flying 200 m personal best: 11.61 ± 0.90 s). Participants were
72 provided with study details and gave written informed consent. The study was approved by
73 the Sheffield Hallam University Research Ethics Committee.

74 **2.2 Experimental protocol**

75 An isokinetic ergometer was set up to replicate each participant's track bicycle position.
76 Riders undertook their typical warm-up on the ergometer at self-selected pedalling rates and
77 resistance for at least 10 minutes, followed by one 4 s familiarisation sprint at 135 rpm.
78 Riders then undertook 3 x 4 s seated sprints at 135 rpm with 4 minutes recovery between
79 efforts. A pedalling rate of 135 rpm was chosen as this is representative of the pedalling rate
80 during the flying 200 m event in track cycling and within an optimal pedalling rate range for
81 track sprint cyclists (Dorel et al., 2005; Kordi et al., 2020).

82 **Isokinetic ergometer**

83 A SRM cycle ergometer frame and flywheel (Julich, Germany) were used to construct an
84 isokinetic ergometer (Burnie et al., 2020). The modified ergometer flywheel was driven by a
85 2.2-kW AC induction motor (ABB Ltd, Warrington, UK), controlled by a frequency inverter
86 equipped with a braking resistor (Model: Altivar ATV312 HU22, Schneider Electric Ltd,
87 London, UK) (Burnie et al., 2020). This set-up enabled participants to start their bouts at the
88 target pedalling rate, rather than expending energy in accelerating the flywheel. The
89 ergometer was fitted with force pedals (Model ICS4, Sensix, Poitiers, France) and a crank
90 encoder (Model LM13, RLS, Komenda, Slovenia), sampling data at 200 Hz.

91 **2.3 Kinematic and kinetic data acquisition**

92 Two-dimensional kinematic data of each participant's left side were recorded at 100 Hz using
93 one high speed video camera with infra-red ring lights (Model: UI-522xRE-M, IDS,
94 Obersulm, Germany) (Burnie et al., 2020). Reflective markers were placed on the pedal
95 spindle, lateral malleolus, lateral femoral condyle and greater trochanter. Kinematics and
96 kinetics on the ergometer were recorded by CrankCam software (CSER, SHU, Sheffield,
97 UK), which synchronised the camera and pedal force data and was used for data processing
98 (Burnie et al., 2020).

99 **2.4 Data processing**

100 All kinetic and kinematic data were filtered using a Butterworth fourth order (zero lag) low
101 pass filter, with a cut off frequency of 14 Hz. Instantaneous left crank power was calculated
102 from the product of the left crank torque and crank angular velocity. The average left crank
103 power was calculated by averaging the instantaneous left crank power over a complete pedal
104 revolution. Joint moments were calculated via inverse dynamics (Elftman, 1939), using pedal

105 forces, limb kinematics, and body segment parameters (de Leva, 1996). Joint extension
106 moments were defined as positive.

107 Data were analysed using a custom Matlab (R2017a, MathWorks, Cambridge, UK) script.
108 Each sprint lasted for 4 s, revealing six complete crank revolutions. Joint moments were
109 resampled to 100 data points around the crank cycle and the mean value at each time point
110 was calculated to obtain a single, ensemble-averaged time series for each trial. Owing to
111 technical problems for two participants only data from two instead of three sprints were
112 collected.

113 **Quantifying hip-ankle joint synergy**

114 To quantify hip-ankle joint coordination and strength of hip-ankle joint synergies a vector
115 coding method was applied to joint moment-moment diagrams (Chang et al., 2008). These
116 were selected as the most appropriate variables to evidence if net hip and ankle joint
117 moments act in synergy during the downstroke (Fregly & Zajac, 1996). Coupling angles (γ_i)
118 were calculated from hip-ankle moment diagrams (Figure 1) for each crank-cycle data point
119 for all revolutions of each participant's sprints (Chang et al., 2008). The coupling angle is
120 defined as the orientation of the vector (relative to the right horizontal) between two adjacent
121 points on the moment-moment plot, Figure 2. Since coupling angles are directional in nature,
122 mean coupling angles were computed using circular statistics (Batschelet, 1981).

123 Mean coupling angles for each participant were categorised into four coordination phases: in-
124 phase, anti-phase, hip-phase and ankle-phase, based on proposals of Chang et al. (2008)
125 (Figure 2). When coupling angle values are 45° and 225° (a positive diagonal), components
126 are in-phase: both hip and ankle moments are increasing or decreasing at similar rates, i.e.,
127 hip and ankle joints are working synergetically (Arnold et al., 2017). Conversely, when

128 coupling angles are 135° and 315° (a negative diagonal), components are anti-phase. For
129 example, when hip moments are increasing, ankle moments are decreasing. When coupling
130 angles are parallel to the horizontal (0° and 180°), ankle moments are changing, but not hip
131 moments – ankle-phase. When coupling angles are parallel to the vertical (90° and 270°), hip
132 moments are changing, but not ankle moments – hip-phase. Since coupling angles rarely lie
133 precisely on these angles, the unit circle was split into 45° bins used by Chang et al. (2008)
134 (Figure 2). Frequencies within which mean coupling angles lay in these coordination patterns,
135 during the downstroke (defined between crank angles of 0 to 180°) were calculated for each
136 participant, using the following equation: Frequency of coordination phase (%) = (Number of
137 occurrences of coordination phase/51) \times 100, (note there are 51 data points in the
138 downstroke). This process was repeated to calculate group mean coupling angles for sprints
139 at 135 rpm (Figure 3). Strength of hip-ankle synergies were quantified by the frequency of in-
140 phase coordination pattern between hip and ankle moments in downstrokes.

141 **2.5 Statistical analysis**

142 Differences between frequencies of coordination phases were assessed using a Friedman test,
143 with post-hoc Wilcoxon matched pairs using IBM SPSS Statistics Version 28 (IBM UK Ltd,
144 Portsmouth, UK).

145 **3 Results**

146 Average left crank power over a complete revolution for sprints at 135 rpm was 494.1 ± 91.2
147 W. Hip and ankle moments were in-phase for 28.8% of the downstroke, with the hip-phase
148 the most frequent coordination phase (42.5%) (Figure 4). A Friedman test ($\chi^2 = 19.3$, $P <$
149 0.0005) indicated that coordination phase frequencies differed across the four coordination
150 phases. Post-hoc Wilcoxon matched pairs indicated that in-phase was significantly different

151 to anti-phase ($P = 0.004$), hip-phase ($P = 0.044$) and ankle phase ($P = 0.017$). Hip phases
152 significantly differed to anti-phases ($P < 0.001$), and ankle phases ($P = 0.004$).

153 **4 Discussion**

154 We demonstrated that a vector coding method can be used to quantify strength of hip-ankle
155 joint moment synergies during downstrokes in cycling. Data imply a tendency for hip and
156 ankle joints to work in synergy in the downstroke during short-term maximal cycling at a
157 pedalling rate of 135 rpm.

158 Findings revealing a weak synergy between hip and ankle joints in downstrokes at 135 rpm,
159 support conclusions of simulation studies suggesting that hip and ankle joints need to work in
160 synergy to transfer energy produced by hip extensors to the crank (Fregly & Zajac, 1996;
161 Raasch et al., 1997). Fregly and Zajac (1996) modelled steady state pedalling at 75 rpm and
162 Raasch et al. (1997) modelled the acceleration phase with pedalling rate increasing from 80
163 to 120 rpm through a revolution. Our results suggest that hip-ankle synergies may not be as
164 strong at pedalling rates higher than previously modelled. The strength of hip-ankle synergies
165 depend on time available in downstrokes to coordinate joint actions. At 75 rpm downstrokes
166 last 0.40 seconds (pedalling rate used in Fregly and Zajac (1996) study), compared to 0.22
167 seconds at 135 rpm. The suggestion is that, as task complexity increases (e.g., due to changes
168 in pedalling rate from 75 to 135 rpm), strength of hip-ankle synergies reduce, implying it is
169 more challenging to coordinate joint moments at higher pedalling rates. This observation
170 supports previous findings suggesting that, as task complexity increases, differences in
171 coordination and coordination variability emerge (Weir et al., 2019).

172 At a pedalling rate of 135 rpm, the hip-phase is the most frequent coordination phase in
173 downstrokes, suggesting that sprints at 135 rpm display a hip-dominant coordination pattern.

174 A much greater contribution from hip extension power to crank power at higher pedalling
175 rates has been observed, proposing an optimal pedalling rate for maximum hip extension
176 power at around 150 rpm, whereas the optimal pedalling rate for knee extension and flexion
177 power was lower (McDaniel et al., 2014). Our finding that hip-phase coordination is
178 dominant at higher pedalling rates elaborates on the notion that the role of the hip becomes
179 more important at higher pedalling rates, from a power production and coordination
180 perspective.

181 This study demonstrated the application of vector coding to quantify hip-ankle moment
182 synergies in maximal cycling. Further research is required to investigate the presence of hip-
183 ankle moment synergies in downstrokes at different pedalling rates and power outputs.

184 **5 Conclusion**

185 A modified vector coding technique can be used to quantify strength of hip-ankle moment
186 synergies in cycling downstrokes. Hip and ankle joints tend to work in synergy in the
187 downstroke during short-term maximal cycling, supporting findings of previous cycling
188 simulation studies, highlighting it as a key biomechanical feature of maximal cycling. Data
189 suggest that this method could be used to assess cyclists' pedalling techniques and to monitor
190 effects of training or equipment interventions on coordination patterns.

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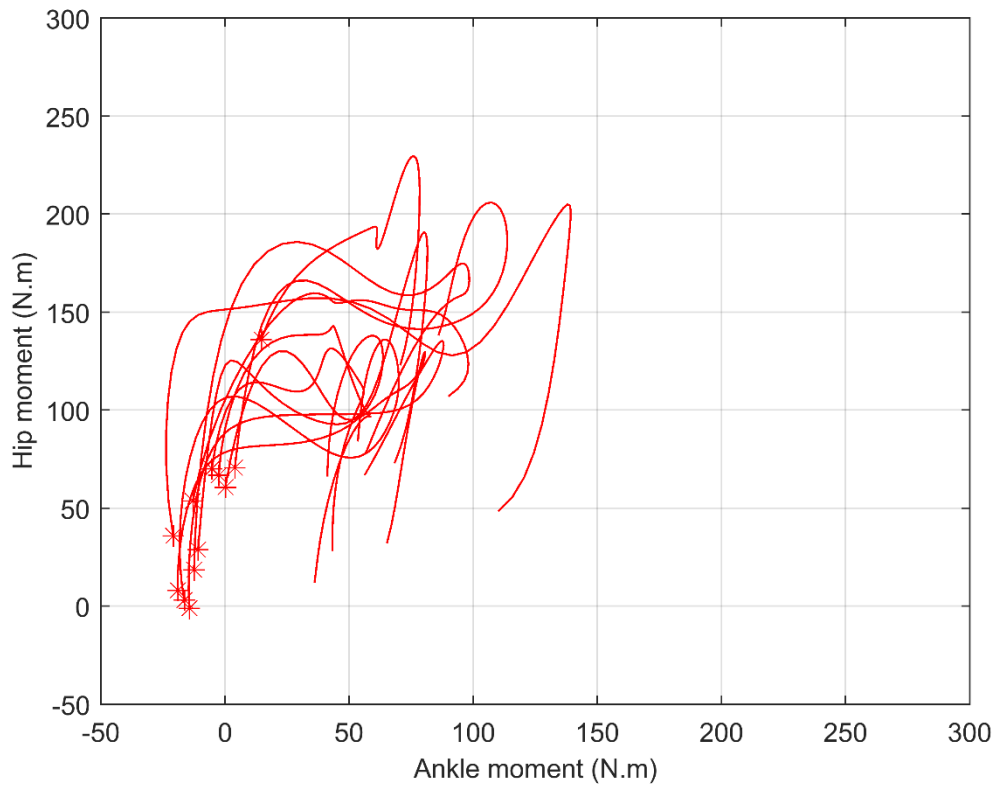
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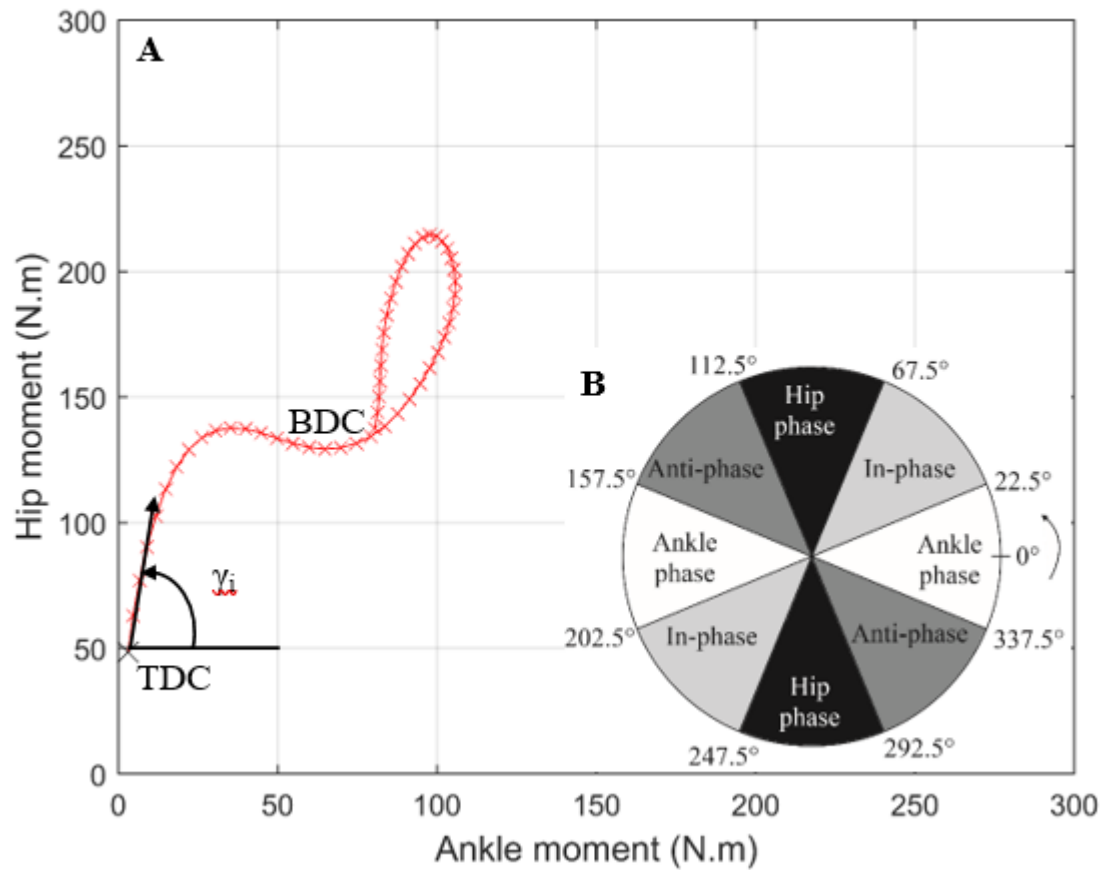
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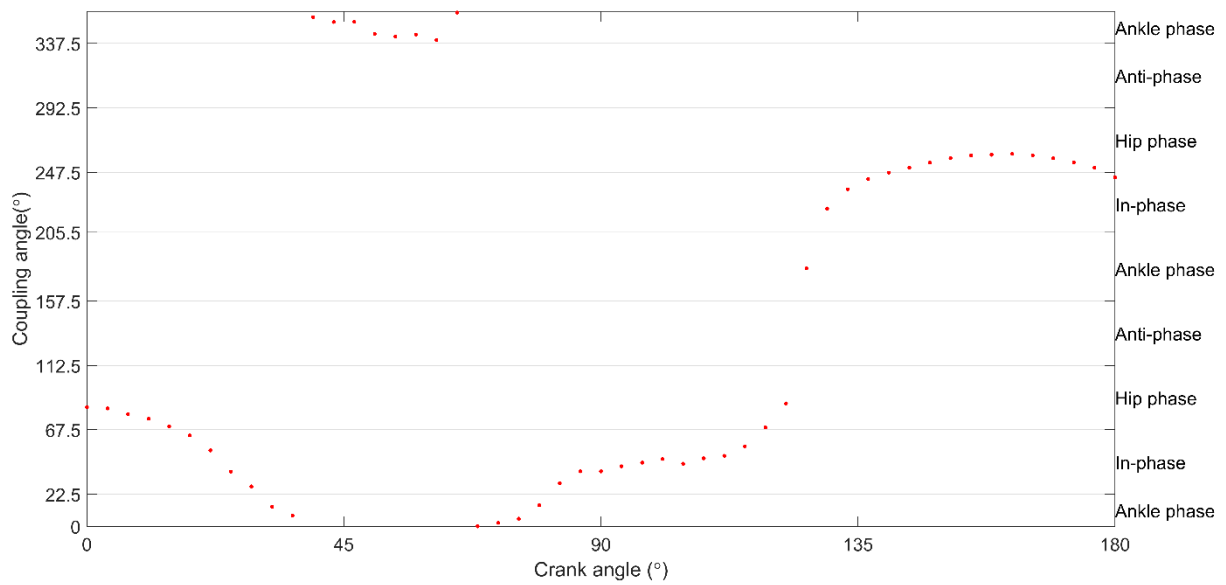
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250 **Figure 1: Mean hip-ankle moment plots for sprints at 135 rpm for the downstroke for**
251 **each participant, with * indicating top dead centre (TDC).**



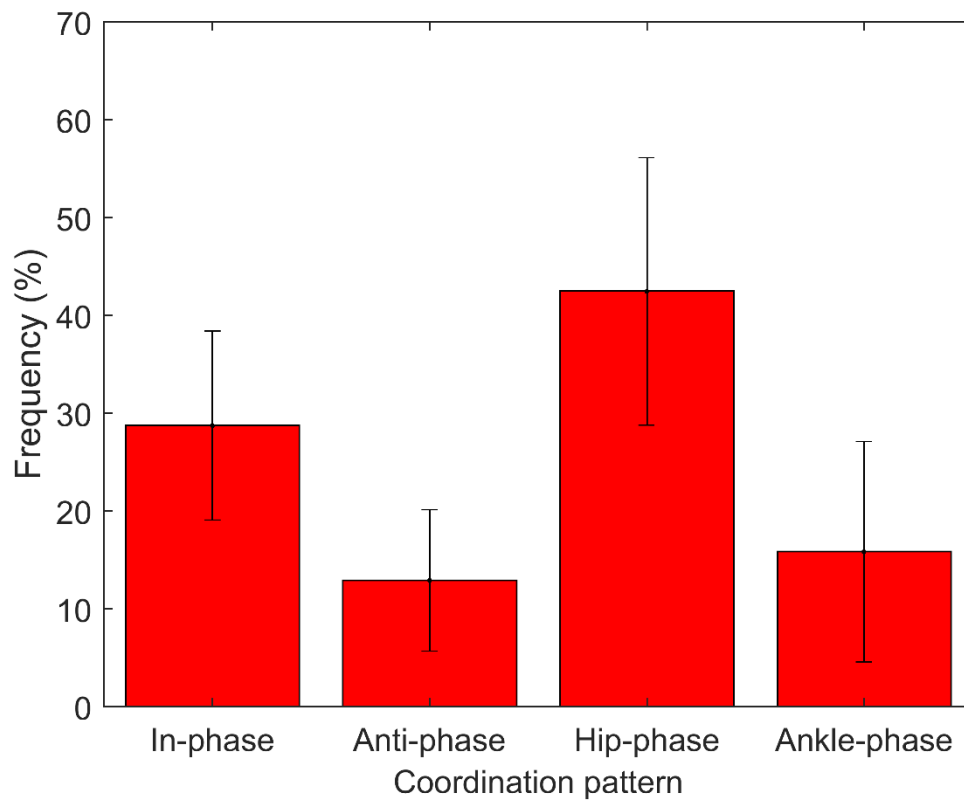
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253 **Figure 2: A: An illustration of the calculation for a coupling angle (γ_i) from hip-ankle**
 254 **moment plot (one downstroke during a sprint). Top dead centre (TDC) = 0°, and**
 255 **bottom dead centre (BDC) = 180°. B: The coordination pattern classification system for**
 256 **the coupling angle (γ_i)**



257

258 **Figure 3: Mean coupling angle for hip-ankle moments for sprints at 135 rpm for the**
 259 **downstroke.**



260

261 **Figure 4: Hip-ankle moment coordination patterns during downstroke phase of the**
 262 **crank cycle for sprints at 135 rpm.**