Multidimensional Joint Coupling: A Case Study Visualisation Approach to Movement Co-ordination and Variability

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Abstract

A case study visualisation approach to examining the coordination and variability of multiple interacting segments is presented using a whole-body gymnastic skill as the task example. One elite male gymnast performed 10 trials of 10 longswings whilst three-dimensional locations of joint centres were tracked using a motion analysis system. Segment angles were used to define coupling between the arms and trunk, trunk and thighs and thighs and shanks. Rectified continuous relative phase profiles for each interacting couple for 80 longswings were produced. Graphical representations of coordination couplings are presented that include the traditional single coupling, followed by the relational dynamics of two couplings and finally three couplings simultaneously plotted. This method highlights the power of visualisation of movement dynamics and identifies properties of the global interacting segmental couplings that a more formal analysis may not reveal. Visualisation precedes and informs the appropriate qualitative and quantitative analysis of the dynamics.

Key words: Motor behaviour, Dynamical Systems, Coordination, Visualisation, Coaching

1 1. Introduction

Understanding the learning and performance of motor skills requires the analysis of not only the outcome of the action for each attempt, in addition, the organisation of the individual joint motions and their coordination. This is particularly the case for the whole body motions of sport skills that reflect the challenge of Bernstein's problem (Bernstein, 1967): namely, how the many degrees of freedom (DF) of the system are organised so as to master the redundancy of the system. Bernstein gave emphasis to the joint motion DF, but even at this macroscopic level the problem for analysis is the challenge of a multivariate system (Bernstein, 1967).

Biomechanics, with its emphasis on the measurement of the kinematics and kinetics of
human movement, has investigated the motions of the individual DF in action. Coordinative
structure theory (Kugler, Kelso and Turvey, 1980) and coordination dynamics (Haken, Kelso
and Bunz, 1985) have emphasised the coordination between the individual DF in movement
and action. Importantly, it is the combination of measurement levels (task outcome, DF motion,
DF coordination) in the context of their redundancies within and between levels of analysis that
reflects what Saltzman and Kelso called task dynamics (Saltzman and Kelso, 1987).

16 There has been considerable progress in understanding the coordination and control of 17 bimanual coordination tasks that is anchored in the principles of the HKB model (Haken, Kelso 18 and Bunz, 1985; Turvey, 1990). There has been less progress in investigating Bernstein's 19 (1967) problem in the multivariate multi-joint movement task context. This problem is reflected 20 in the range of examples apparent in the learning and performance of whole-body actions and 21 sport skills. It is our view that progress in the multivariate (multiple beyond 2 DF case) can 22 only be addressed by a multivariate system (network) oriented approach to the problem that 23 goes beyond the bivariate bimanual case.

Movement science has become increasingly interested in the problem of coordination, control and skill. As a result there has been an increased use of multivariate statistics, nonlinear

dynamics and network analyses to the problems of system decomposition. For example, the 26 linear statistics of principal component analysis (Daffertshofer, Lamoth, Meijer & Beek, 2004; 27 28 Lamoth, Daffertshofer, Huys and Beek, 2009), canonical analysis (Ivanovic and Ivanovic, 29 2011; Kakebeeke et al., 2014) and cluster analysis (Sailer, Engert, Dietrich and Straube, 2000) 30 have been used to decompose the multivariate relations in movement and posture tasks. 31 Stergiou (2004) has introduced some analytic tools from nonlinear methods to the analysis of 32 human movement. There have also been nonlinear network machine learning approaches to 33 motor control through support vector machines (Chow, Davids, Button and Rein, 2008).

34 The purpose of this paper is to re-emphasise the power of the visualisation of movement dynamics to explore and understand the coordination problem at the individual case study level 35 36 in sports biomechanics. The visualisation of complex movement sequences is shown as a 37 precursor to the use of formal quantification methods. We follow Abraham and Shaw (1984) 38 who provided a unique contribution in formalising a set of strategies for the visualisation of the 39 geometry of behaviour that was grounded in dynamics. This visualisation framework is not a 40 replacement for the formal mathematics of dynamics but an adjunct to it that reflects the power 41 of visual representations to understanding and describing the motion of dynamical systems. A 42 visualisation approach to the movement coordination problem can provide a qualitative 43 (topological) and quantitative description of movement and its constraints (McGinnis and 44 Newell, 1982).

There are several useful consequences to the visualisation of the movement coordination and control problem. First, it requires one to select, *a priori*, the relevant dimensions and variables on which to represent the motion of the system. This is not necessarily a straightforward decision particularly when there are multiple DF, but it forces an explicit determination of the coordinate frame. Second, the selection of the coordinate dimensions can occur in a theory driven or looser 'warm feeling' way about the organisation of the system.

This contrasts with the multivariate statistic approach that allows the formal assumptions of the technique to find the respective relations in the data set independent of a large degree of theoretical bias. Third, the visualisation strategy can allow one 'to see' relations in the data that may not be apparent from the more formal analysis techniques with groups of individuals. There will still be limitations arising particularly when there are multiple DF, but visualisation can provide an approach to dynamical 'data snooping' that can lead to insights prior to the selection and use of formal analysis techniques.

58 2. Methods

59 2.1. Participants

A priori approval was gained from Cardiff Metropolitan University's Research Ethics
Committee and written informed consent obtained from the participant who was a male,
international level gymnast (competiting at World Championship level) : age, 24 years; mass,
69 kg; and stature, 1.67 m. The participant was healthy and without injury at the time of data
collection.

65 *2.2. Data collection*

Kinematic data (200 Hz) were collected using an automated 3D motion capture system
(CODAmotion, Charnwood Dynamics Ltd, Leicester, UK). Two CX1 scanners provided a field
of view exceeding 2.5 m around the centre of the bar. The CODA system was aligned according
to the manufacturer's guidelines.

Seven active markers were placed on the lateral aspect of the participant's right side at the estimated centre of rotation for the: glenohumeral joint, mid forearm, lateral epicondyle of the elbow, greater trochanter, femoral condyle, lateral malleolus, fifth metatarsophalageal and the centre of the underside of the bar. Whole-body mass and height were measured using laboratory weighing scales (Avery Berkel Ltd, model ED01) and a stadiometer (Holtain, Ltd.), respectively. The inertia parameters of the gymnast were customised from a database of 30

male gymnasts generated from direct measurements using Yeadon's inertia model (Yeadon,
1990). The gymnast's height and mass were combined with limb length scaling derived from
the coordinate video data.

The gymnast performed 10 sets of 10 consecutive longswings on a standard competition high bar (Continental Sports, Huddersfield, UK). Sufficient rest was provided between each set so that the gymnast did not become fatigued.

82 Raw marker data in the horisontal and vertical directions were identified from CODA 83 output and all subsequent analysis took place using customised code written in MathCAD 13TM 84 (MathSoft Engineering & Education, Inc., Surrey, UK). Coordinate data were filtered using a 85 Butterworth low-pass digital filter with cut-off frequency set to 6 Hz that was determined based 86 on Winter's residual analysis (Winter, 1990). Segment angles were defined with respect to the 87 right horisontal and angular velocities determined using a variation of Ridder's divided difference method (Press, Flannery, Teukolsky and Vetterline, 1992). The angular orientation 88 89 of the gymnast about the bar was described by a circle angle. A circle angle was defined by a 90 vector from the mass centre to bar with respect to the horisontal, where a circle angle of 90° 91 and 450° reports the CM of the gymnast above the bar (in handstand)

92 -----Insert Figure 1 here-----

93 2.3 Data Analysis

Continuous relative phase (CRP) was determined using the methods of Hamill, van Emmerik, Heiderscheit and Li (1999). CRP was calculated for pairs of segments: Arm-Trunk '(shoulder), Trunk-Thigh (hip) and Thigh-Shank (knee). Phase plane portraits were normalised to ± 1 of the maximum rectified angle (horizontal axis) and angular velocity (vertical axis). The phase angle (ϕ) was calculated as the arctangent of the angular velocity over the angular displacement profile within the range $0^{\circ} \le \phi \le 360^{\circ}$ and then rectified differences in phase angles of each pair of segments provided the CRP profiles according to the methods of Hamill et al. (1999). The initiating and ending swings in each set were ignored leaving the middle eight
swings from each set of ten for analysis resulting in 80 swings for each participant (10 sets of
eight consecutive swings). In order to provide intra-trial comparisons between swings, data
were interpolated in 1° increments of the circle angle about the bar.

Given our interest in the coordination and variability of the highbar longswing we
established a series of coordinate frames to visualise the couplings and variability of the
movement sequence.

Stage One: The CRP profiles of the Arm-Trunk (shoulder), Trunk-Thigh (hip), Thigh-Shank (knee) for all 80 swings were plotted separately against circle angle. These 1-DF coordinative structures include both intra-trial and intra-participant variation. The trajectories of interacting segments represent a local state space of this skill.

Stage Two: The CRP profiles of the shoulder, hip, and knee were then averaged across the 80 trials to produce a mean CRP profile for each coupling. Arm-Trunk (shoulder), Trunk-Thigh (hip), Thigh-Shank (knee), respectively, were then plotted against circle angle to provide a 2 DF coordinative structure. Plotting both these 2-DF coordinative structures simultaneously against circle angle provided a more complete view of the coordinate interactions and a more global representation of the state space.

Stage Three: The average CRP profiles of each individual couple were plotted simultaneously i.e. Arm-Trunk (shoulder), Trunk-Thigh (hip), Thigh-Shank (knee) (x,y,z). In order to maintain context to the direction and magnitude, the circle angle was depicted through a colour code (green = 90-180 deg, yellow = 180-270 deg, blue = 270-360 deg, red = 360-450 deg). This representation of combined segments provides a 3-DF coordinative structure that represents a macroscopic view of the functional coordinative system. Varibility was reprensted as a ciontuiius standard standard deviation was employed.

125 **3. Results**

126 Single couplings for each interacting segment are displayed in Figure 2. This 127 visualisation shows the distinctive similarities in phase relationships within the individual 128 couplings in terms of phase and magnitude. Specifically, the shoulders showed an out-of-phase 129 peak in the latter stages of the circle, hips a double peak in the first and third quarters of the 130 circle and the knee, which was most variable, showed an anti-phase peak in the first third of the 131 skill.

132 -----Insert Figure 2 here-----

The combined couplings of the hip-knee and hip-Shoulder, respectively, are shown in Figure 3a. These coupling trajectories represent a 2-DF coordinative structure and, thus, a more global profile of the interacting couplings. Combining these couplings (hip-knee and hipshoulder) against circle angle (Figure 3b) provides an identification of the anti-phase locations in relation to the circle angle, specifically, the couples of the hip-knee and hip-shoulder dominating in the first and final quarter of the circle, respectively.

139 -----Insert Figure 3a and Figure 3b here-----

140 Simultaneously transposing the three couplings of the shoulder (arms-trunk), hip (trunk-141 thigh) and knee (thigh-shank) together provides a graphical representation of this interacting 142 system (Figure 4). The visualisation of the multiple segments during this specific task provides 143 a macroscopic coordination perspective. Figure 4 also shows this coordination structure with 144 variability through the inclusion of the standard devistion across the 80 swings.

145 ------Insert Figure 4 here-----

With the three couplings occupying the orthogonal axis, the angular position of the gymnast about the bar is described through the inclusion of colour coding (Figure 4). The angular position of the gymnast about the bar was defined by the mass centre as a vector angle projected from the centre of the bar hence providing context to this 3-DF coordinative structure.
Figure 4 defines the macroscopic coordinative structure and its variability showing anti-phase
couples at the hip and knee during the first (90-180 deg) and second (180-270 deg) quadrant.

152 It is evident from the visual depiction that the variability increases from first to the 153 second quadrant peaking at 225 deg then, interestingly, reducing as the gymnast moves to the 154 lower vertical. During quadrant three, the shoulder and knee couples are inversely related, as 155 the former increases its out of phase pattern whilst the latter become more in phase. During this third quadrant the variability stays relatively stable. In the 4th quadrant (360-450 deg) the 156 157 shoulder couple moves out of phase followed by the hips and knees. The variability of the 158 coordinative structure changes as a function of the segmental interaction and the angular 159 position of the gymnast.

160 4. **Discussion**

161 The visualisation technique presented here simultaneously transposes three CRP 162 couplings to produce a 3D plot that provides a holistic representation of the movement 163 coordination pattern and its variability (Figure 4). The 3D visualisation of the movement 164 dynamics reveals specific patterns of coordination and variability employed to successfully 165 perform this motor skill. This visual representation of movement behaviour provides an 166 opportunity to examine the interaction of the segments of this system and its attractors 167 (Abraham and Shaw, 1984).

The longswing on high bar is a representative task to combine knowledge of the underlying biomechanical determinants with the self-organisation (coordinative structures) of the athlete as a system (Irwin and Kerwin, 2007a, 2007b; Williams, Irwin, Kerwin and Newell, 2015a, 2015b). The visual representations provide insight into the nature and structure of this system although this is not to be confused with the somewhat predictable orbiting oscillations of the performance-based mechanics of these skills (e.g. Irwin and Kerwin, 2005). Indeed, the

174 visual representations revealed properties of movement organisation that were masked, or not 175 apparent, in the more standard biomechanical analyses. The coordinative structure that emerges 176 from this 3D visualisation demonstrates that, within the action work space, there are clear 177 phases where all three couplings merge towards a global in-phase relationship (final stages of 178 the 2nd quadrant) before a clear hip dominated anti-phase coordinative pattern arises (3rd 179 quadrant). These global interactions are accompanied by a reduction in coordination variability 180 during this key transitional phase as the performer moves through the lower part of the circle. 181 These observations serve two purposes: 1) they highlight the overall segmental interactions and 182 variability i.e. the holistic nature of this skill; and 2) the importance of this key phase under the 183 bar that is supported by the classic biomechanics of this skill.

184 The results of this study highlight similarities in segmental coupling within a participant 185 across the CRP profiles (Figure 2). The knee coupling showed the highest level of variation 186 across the 80 trials. Combining the couplings (Figure 3a and 3b) gave an indication of the 187 changes in the coupling within the system but also the location in terms of the angular position 188 of the gymnast on the bar. These findings concur with the functional characteristics of the skill 189 (Irwin, Exell, Manning and Kerwin, 2014). The assimilation of the three couplings (Figure 4) 190 displays changes in the coordinative structure and its variability during the four quadrants of 191 the longswing.

During the first quartile (90-180), hip coordination dominates being increasingly outof-phase. In the second quartile (180-270), the hip and knee couplings tend to out-of-phase in line with the occurrence of the functional phase (Irwin and Kerwin, 2005). During the third quadrant (270-360), the hip is again dominant with an out-of-phase coordination pattern, with less than five degree change in the knee and shoulder couplings. In the final quartile, the shoulder couplings peaks at 40 degrees while the hip coupling tends to be in-phase culminating in the shoulder returing to in-phase. Throughout the 3D plot, the relatively planar trajectories of the 2nd and 4th quartile highlight that two couples tend to out-of-phase at any one time. These
changes in coordination trajectories can only be observed in 3D plots of this nature, highlighting
the kinematic sequencing that determines a successful performance.

202 Interestingly the variability of hip and shoulder is highest during the 2nd and 4th 203 quadrants and signifies the position of a control mechanism supporting successful completion 204 of the skill. Variability is high as the couplings tend to be out-of-phase. Maintaining the stability 205 of the coupling and low variability in quartile 3 (270-360), as in Figure 4, links to the concept 206 of end point variability as the gymnast performs the functional phase of the skill (Irwin & 207 Kerwin, 2005). These three dimensional representations of the coupling variability provide a 208 ungiue visualisation of the magnitude and structure of the variability in all three couplings 209 simultaniously.

210 The continuous standard deviation shown in Figure 4 provides an indication of the 3D 211 structure and nature of coordination variability. It is interesting to note that the current literature 212 provides opposing views on the topic of inter-particiant variability. For example, Wilson, 213 Simpson, van Emmerik and Hamill (2008) reported high levels of coordination variability for 214 key aspects of elite triple jump technique; whereas, Irwin and Kerwin (2007a, 2007b) found 215 low levels of coordination variability for elite gymnasts performing longswings. More 216 generally, coordination variability has been shown here to be a task and a variable specific 217 problem. Strategies used to perform the gymnastics skill in this study may be expected to be 218 more consistent due to their delimited nature through the constraints-to-action concept (Newell, 219 1986, pp. 341-360). This is highlighted throughout the literature in line with the nonlinear 220 dynamics systems approach (van Emmerik, Ducharme, Amado and Hamill, 2016). In addition, 221 the inclusion of the continuous nature of coordination variability across the cycle highlights 222 discrepancies that would not be provided by focusing on specific regions.

223 The macroscopic approach visualised in Figure 4 builds on previous single degree of 224 freedom approaches to co-ordination (i.e Figure 2). The current approach demonstrates how 225 adding, simultaneously, multidimensional segmental couplings and their variability across an 226 entire movement gives an indication of the dynamic nature of the coordinative structures 227 throughout the whole movement. It provides a visual representation of three joint couplings 228 capturing the interaction of all degrees of freedom involved in this skill. This information may 229 not be forthcoming in a two degree of freedom analyses. The 3D structure draws on the 230 visualisation work of Abraham and Shaw (1984) and aligns with the theories of nonlinear 231 dynamics.

232 This visualisation analysis provides an adjunct to quantitative approaches rather than a 233 replacement of them. This method provides a useful tool to describe and observe the global 234 nature of skills and guide the priority and flow of their analysis. In addition, it highlights the 235 need to begin to quantify the coordination of human movement between more than two 236 segments. Finally, the individual strategies employed for these skills may provide some 237 explanation of the differences in the amplitude and frequency of variability. The individual 238 nature of biological systems suggests that participants meet the task demands more effectively 239 with a unique individually characteristic organisation.

Coordinate frames other than those presented could have been selected to visualise aspects of the longswing dynamics. One candidate provides time on one dimension to observe the evolving coordination and movement variability. Any of the possible coordinate frames of reference for representing the coordination and variability of the movement could be implemented to provide information feedback to the gymnast (McGinnis and Newell, 1982).

245 **5.** Conclusion and Implications

In this paper, the power of visualising the coordination and variability of themovement pattern of the highbar longswing has been demonstrated. The longswing is a

sports skill that has had considerable study in recent years from a biomechanical 248 249 perspective and how this relates to the skill level of the gymnast (Williams, Irwin, Kerwin and Newell, 2015a, 2015b; Williams, Irwin, Kerwin, Hamill, van Emmerik and Newell, 250 251 2016). The longswing is a multivariate (multiple DF) task, but it is also rather tightly constrained by the rules of gymnastics and the physics of the body in motion. 252 253 Nevertheless, it is anticipated that the visualisation approach developed here offers 254 generality to the study of movement coordination and control in a broad range of 255 movement skills. Furthermore, and importantly, we show that the visualisation of the 256 relative motions of the longswing reveals properties of control that have not been 257 forthcoming from traditional multivariate statistical analysis of biomechanics data.

258 Here, we were guided by our interest in visualising the coordination and variability of 259 several joint space DF as a coordinative structure in the performance of a gymnast. The 260 visualisation of the qualitative and quantitative properties of the movement dynamics are revealing and provide insights to guide more formal analysis of coordination and variability. 261 262 The strength of our macroscopic approach is that it adds to the understanding of the complex nature of human movement as a dynamic system. The visualisation approach outlined here 263 264 precedes and informs the selection of the appropriate quantitative analysis of the dynamics and 265 hence has direct implications.

266 **Conflict of interest statement**

267 There are no conflicts of interest associated with this research.

268 Authors' contributions

GI, GW, JH and KN drafted the manuscript; GI and GW performed literature searches and helped to draft and revise the manuscript. All authors developed the methods, results and discussion. All authors have read and approved the final version of the manuscript, and agree with the order of presentation of the authors.

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344

345 **Figure Captions**

- 346
- 347348 Figure 1. Angular position of the gymnast.

Figure 2. Continuous relative phase profiles for the Arm-Trunk, Trunk-Thigh and ThighShank couplings for each looped bar longswing from 0-360° (x), for each of 80 trials (y)
showing amplitude of CRP (z).

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Figure 3a. LEFT: Combined Continuous relative profiles for the Trunk-Thigh (hips) and
 Thigh-Shank (knee) RIGHT: Combined Continuous relative profiles for the Trunk-Thigh
 (hip) and Arm-Trunk (shoulder)

Figure 3b Combined Continuous relative profiles for the Trunk-Thigh (hip) and Thigh-Shank
(knee) with the Combined Continuous relative profiles for the Trunk-Thigh (hips) and ArmTrunk (shoulder)

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362 Figure 4 Continuous relative profiles and associated SD of the Trunk-Thigh (hips), Thigh-

363 Shank (knees) and Arm-Trunk (shoulders) plotted simultaneously. Axis: x = shoulder, z =

knees, y = hips. Angular position of the gymnasts denoted via colour coding.

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