

when changing the number of available degrees of freedom

Enrico Chiovetto^{(1)*}, Meghan Huber^{(2,3)*}, Ludovic Righetti⁽³⁾, Stefan Schaal⁽³⁾, Dagmar Sternad^{(3,4)**}, Martin Giese^{(1)**}

(1) Section for Computational Sensorimotorics, Department of Cognitive Neurology, Hertie Institute for Clinical Brain Research, Centre for Integrative Neuroscience, University Clinic Tübingen, Tübingen, Germany; (2) Department of Bioengineering, Northeastern University, Boston, Massachusetts; (3) Autonomous Motion Department, Max Planck Institute for Intelligent Systems, Tübingen, Germany; (4) Departments of Biology, Electrical and Computer Engineering, and Physics, Boston, Massachusetts; Equal *first and **last author contributions

Introduction

- How the central nervous system controls the excessive number of degrees of freedom (dof) to accomplish complex movements remains still an open question.
- A possible solution to the problem might be provided by a modular architecture that is composed from invariant control modules (motor primitives or synergies), which are linearly combined to generate the desired motor output [1,2,3,4,5].
- Although many studies have focused on the identification of such primitives, less attention has been given to adaptive mechanisms that allow the system to deal flexibly with varying constraints or situations when specific dof become unavailable [6,7]. A previous study from our groups have shown that freezing of dof improves task performance [8].
- In this study, we investigated how complex motor coordination patterns vary during a highly redundant whole-body task, in both constrained and unconstrained conditions.

Experiments

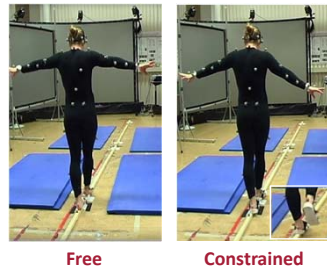
Experiment 1 (arms constrained)

- 9 participants were asked to walk on a narrow beam (3.5cm wide, 5m long) at a self-selected speed.
- Three sessions (free, constrained, free).
- In each session, participants performed as many trials as necessary to complete 20 successful trials.
- A trial was deemed successful, if the participant remained on the beam for its entire length.
- In the constrained condition elbow and wrist joints were fixated by rigid tubes.



Experiment 2 (feet constrained)

- 7 participants were asked to perform the same walking task as in experiments 1, but in the second session the feet, instead of the arms, were constrained.
- In the constrained condition the flexion of the feet was prevented by having participants wearing special sandals with rigid soles.



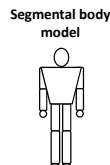
Analysis

Angular momentum (x-axis direction)

- Motion of a 14-segment rigid body model was fit to the 3D motion capture data.
- Angular momenta (AM) about the x-axis (beam direction) around the body's COM were calculated for each segment i .
- For each subject, the contribution of each link to the total angular momentum was computed.

$$\mathbf{L}_i = (\mathbf{r}_{CM}^i - \mathbf{r}_{CM}) \times m_i (\mathbf{v}_{CM}^i - \mathbf{v}_{CM}) + \mathbf{I}^i \boldsymbol{\omega}_i$$

\mathbf{L}_i = angular momentum
 M = total mass
 \mathbf{r}_{CM} = COM position
 \mathbf{v}_{CM} = COM velocity
 \mathbf{r}_{CM}^i = COM position i -th segment
 \mathbf{v}_{CM}^i = COM velocity i -th segment
 m_i = mass i -th segment
 \mathbf{I}^i = moment of inertia i -th segment
 $\boldsymbol{\omega}_i$ = angular velocity i -th segment



PCA

- Principal component analysis (PCA,[9]) was performed on the matrix \mathbf{L} , each row \mathbf{L}_i being the angular momentum contribution provided the i -th segment.
- PCA on the covariance matrix.
- PCA factorization: $\mathbf{L} = \mathbf{W}\mathbf{H}$

each column of \mathbf{W} being a principal component (PC). Each PC consists of a 14-component vector, corresponding to the 14 body segments of the human model. $\mathbf{L} \in \mathbb{R}^{14 \times T}$, $\mathbf{W} \in \mathbb{R}^{14 \times N}$, $\mathbf{H} \in \mathbb{R}^{N \times T}$, with $N < T$.

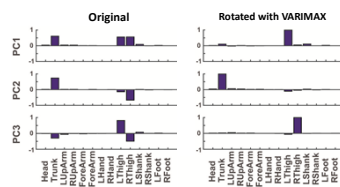
VARIMAX rotation

- Only N principal components (PCs), each explaining at least 5% of variance, were retained for each participant. PCs were rotated using VARIMAX rotation [10], as done in Factor Analysis, to improve sparseness of components.
- VARIMAX maximizes the sum of the variances of the squared loadings (squared correlations between variables and PCs):

$$V = \sum (w_{ij}^2 - \bar{w}_{ij}^2)^2$$

- with w_{ij}^2 being the squared loading of the i -th variable on the j -th PC and \bar{w}_{ij}^2 being the mean of the squared loadings. VARIMAX keeps the components orthogonal to each other.

Representative example of VARIMAX rotation of the PCs

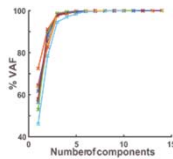


The main outcome of VARIMAX is sparsification of the PCs.

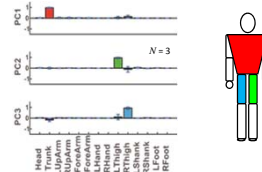
Results

Experiment 1

Session 1 (free, control)

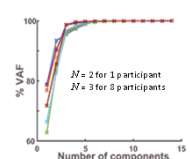


- Identified PCs result very sparse after VARIMAX rotation.
- They are associated mainly with one single segment (either trunk or left/right thigh).

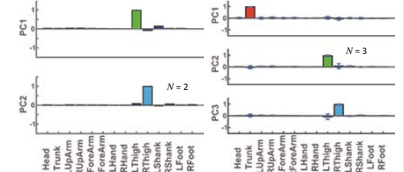


Experiment 2

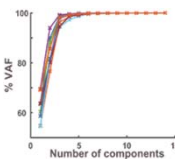
Session 1 (free, control)



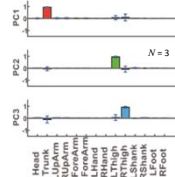
$N = 2$ for 1 participant
 $N = 3$ for 3 participants



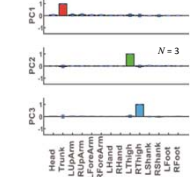
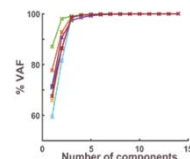
Session 2 (arms constrained)



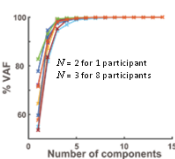
- In both the constrained sessions three PCs were always considered.
- "Elbows" in the % VAF curves seem to be qualitatively more pronounced for some subjects with respect the first session.



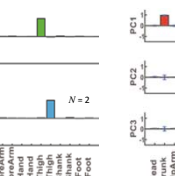
Session 2 (feet constrained)



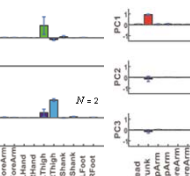
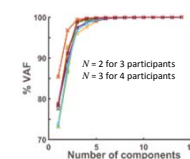
Session 3 (free, control)



- In the third session a decrease of the number of PC occurred for some participants (increased coordination).
- In all sessions, for $N = 2$ the PCs are usually associated with the thighs.
- For only one subjects in session 3 one component resulted associated with the trunk and one with the right thigh.



Session 3 (free, control)



Conclusions

- Only few PCs are needed to account for the majority of the AM variation along the walking direction.
- Despite of their large kinematic variability along the frontal plane [8], arms do not contribute substantially to the total whole-body AM. Rather, AM seems determined by the segments that are the most proximal to the whole-body COM.
- Freezing the dof did not cause significant differences in the low-dimensional organization of the AM.

References

- Flash T and Hochner B. Motor primitives in vertebrates and invertebrates. *Curr Opin Neurobiol*, vol. 15, no. 6, pp. 660-666, 2005.
- Chiovetto E and Giese MA. Kinematics of the coordination of pointing during locomotion. *PLoS ONE* 8: e79555. doi:10.1371/journal.pone.0079555, 2013.
- ig W and Giese MA. Modeling of movement sequences based on hierarchical spatial-temporal correspondence of movement primitives. *Biologically Motivated Computer Vision*. Springer Berlin Heidelberg, 2002.
- Hogan N and Sternad D. Dynamic primitives of motor behavior. *Biological Cybernetics* 106.11-12: 727-739, 2012.
- Schaal S, Kotosaka S and Sternad D. Nonlinear dynamical systems as movement primitives. *IEEE International Conference on Humanoid Robotics*. 2000.
- Verwejen B, Whiting HTA and Newell KM. Free (2) ing degrees of freedom in skill acquisition. *Journal of motor behavior* 24.1: 133-142, 1992.
- Bernstein NA. *The Coordination and Regulation of Movement*. London: Pergamon Press, 1967, pp. 1967.
- Huber M, Chiovetto E, Righetti L, Schaal S, Giese MA and Sternad D. From Humans to Robots and Back: Role of Arm Movement in Medio-lateral Balance Control. *Annual Meeting of the Neural Control of Movement*, Charleston, South Carolina, 2015.
- Jolliffe, Ian. *Principal component analysis*. John Wiley & Sons, Ltd, 2002.
- Abdi, Hervé. *Factor rotations in factor analyses: Encyclopedia for Research Methods for the Social Sciences*. Sage: Thousand Oaks, CA: 792-795, 2003.

Acknowledgements

Enrico Chiovetto and Martin Giese are supported by the following research grants: Korobot FP7-611909, CogIMon H2020 ICT-644727, DFG Gi 305/4-1, DFG GZ: 1258/15-1, HBP FP7-604102, 8MBF, FKZ: 01GQ1002A, ABC PITN-GA-011-290011.
 Meghan E. Huber is supported by NEU Graduate School of Engineering, The Mathworks, and Max Planck Institute for Intelligent Systems, and Dagmar Sternad supported by NIH R01-HD04563, NSF DMS-0928587, and Visiting Professorship at Max Planck Institute for Intelligent Systems.
 Ludovic Righetti is supported by the Max-Planck-Society, and Stefan Schaal is supported by National Science Foundation grants IIS-1205249, IIS-1017134, CNS-0960061, EECs-0926052, the DARPA program on Autonomous Robotic Manipulation, the Océ of Naval Research, the Okawa Foundation, and the Max-Planck-Society.