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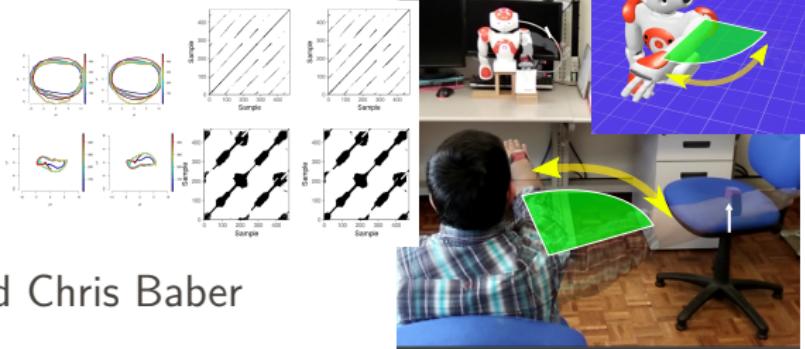
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Quantifying the Inherent Chaos of Human Movement Variability

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Madrid, Spain, 4-7 June 2018



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Movement Variability
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RSS
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RQA
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Experiment
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Results
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Conclusions and Future Work
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OVERVIEW

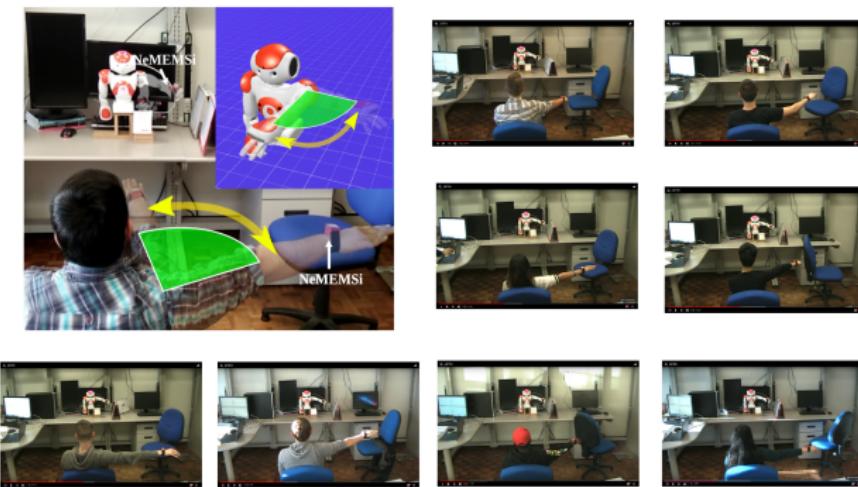
1. Movement Variability
2. RSS
3. RQA
4. Experiment
5. Results
6. Conclusions and Future Work

MOVEMENT VARIABILITY

WHAT IS MOVEMENT VARIABILITY?

- * MOVEMENT VARIABILITY (MV) is defined as the variations that occur in motor performance across multiple repetitions of a task.
- * MV is also considered as an inherent feature within and between each person's movement.

MV IN THE CONTEXT OF HUMAN-ROBOT INTERACTION



Work in progress (Xochicale M. et al. 2018)

Figure 1: MV within and between each person's movement.

RSS

RECONSTRUCTED STATE SPACE (RSS) THEOREM

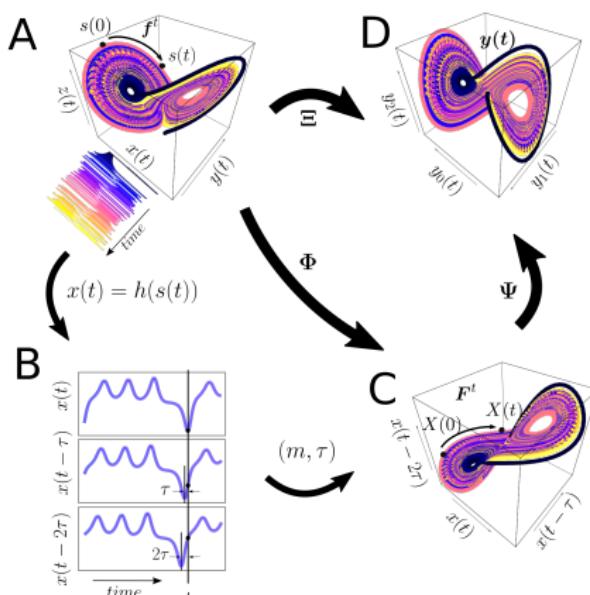


Figure adapted from (Casdagli et al. 1991, Uzal et al. 2011)

Figure 2: Reconstructed State Space

THE METHOD OF UNIFORM TIME-DELAY EMBEDDING (UTDE)

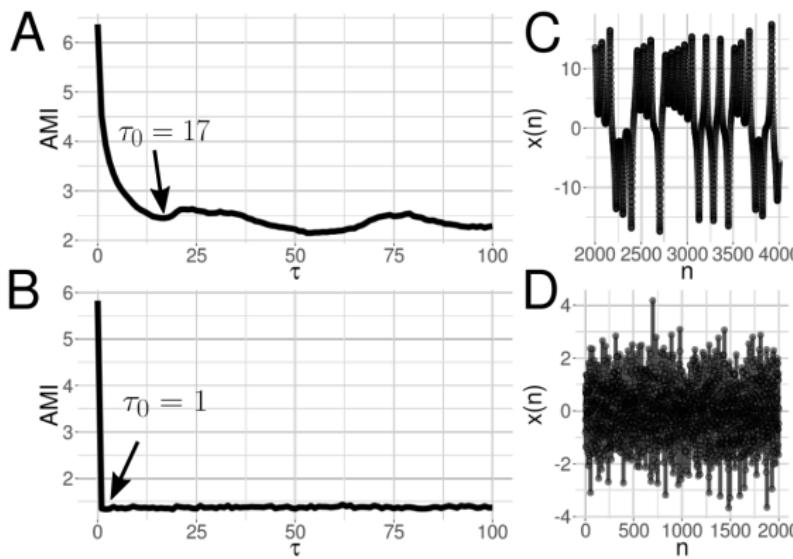
For a given discrete time series $x(n) = [x(1), x(2), \dots, x(N)]$, a reconstructed state space matrix is defined as

$$\mathbf{X}_\tau^m = \begin{pmatrix} \tilde{x}(n) \\ \tilde{x}(n - \tau) \\ \vdots \\ \tilde{x}(n - (m - 1)\tau) \end{pmatrix}^T$$

where m is the **embedding dimension** and τ is the **embedding delay**.

Dimensions of \mathbf{X}_τ^m are $(m, (N - (m - 1)\tau))$ and sample length for $\tilde{x}(n - i\tau)$, where $0 \leq i \leq (m - 1)$, is $N - (m - 1)\tau$.

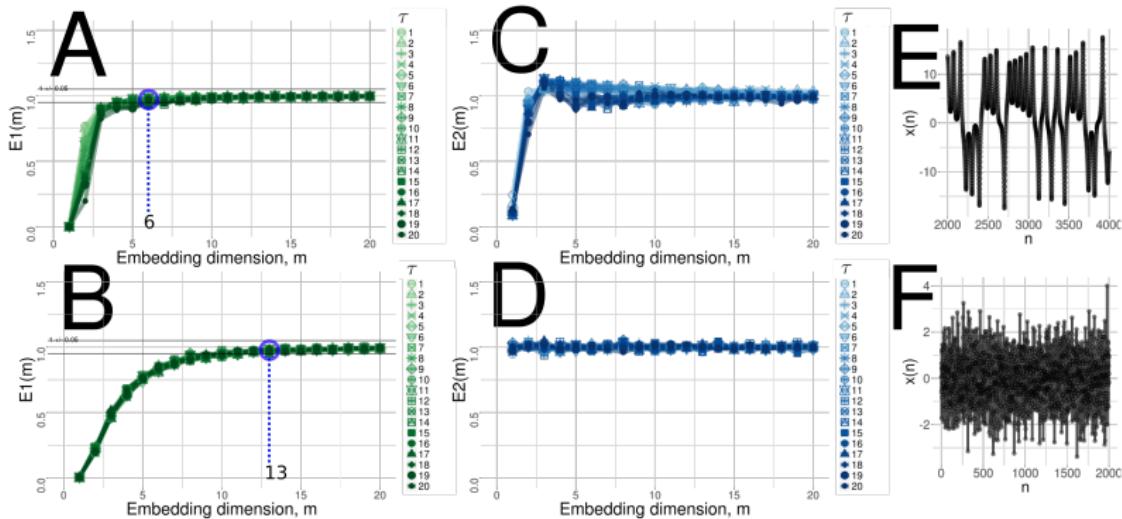
AVERAGE MUTUAL INFORMATION (AMI)



Work in progress (Xochicale M. et al. 2018)

Figure 3: (A, B) AMI values for (C) chaotic and (D) noise time series.

FALSE NEAREST NEIGHBOURS (FNN)



Work in progress (Xochicale M. et al 2018)

Figure 4: (A,B) $E_1(m)$ and (C, D) $E_2(m)$ values for (E) chaotic and (F) random time series

RQA

RECURRENCE PLOTS

$\mathbf{R}_{i,j}^m(\epsilon)$ is two dimensional plot of $N \times N$ square matrix defined by

$$\mathbf{R}_{i,j}^m(\epsilon) = \Theta(\epsilon_i - \|X(i) - X(j)\|), \quad X(i) \in \mathbb{R}^m, \quad i, j = 1, \dots, N \quad (1)$$

where N is the number of considered states of $X(i)$, ϵ is a threshold distance, $\|\cdot\|$ a norm, and $\Theta(\cdot)$ is the Heaviside function.

RECURRENCE PLOTS

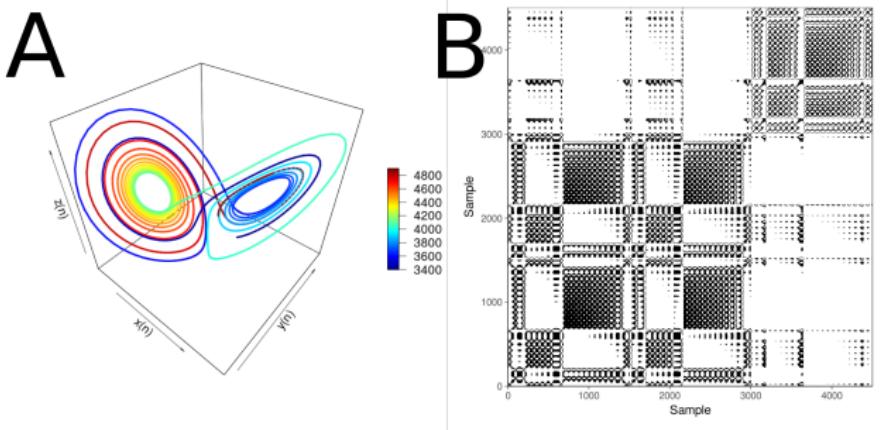


Figure is adapted from (Marwan et al. 2007).

Figure 5: (A) State space for Lorenz systems, and (B) Recurrence plots with no embeddings and $\epsilon = 5$

RECURRENCE PLOT PATTERNS

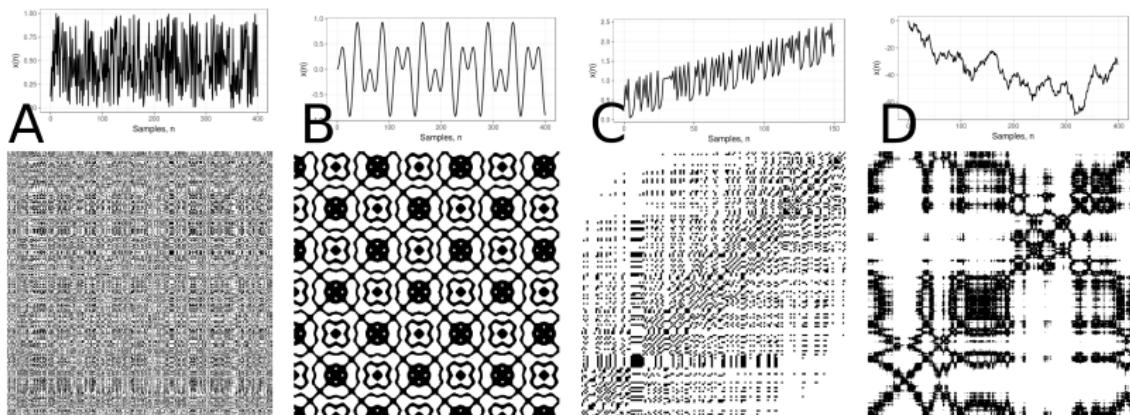


Figure is adapted from (Marwan et al. 2007)

Figure 6: Recurrence plots for (A) uniformly distributed noise, (B) super-positioned harmonic oscillation, (C) drift logistic map with a linear increase term, and (D) disrupted brownian motion.

RECURRENCE QUANTIFICATION ANALYSIS (RQA)

REC enumerates the black dots in the RP.

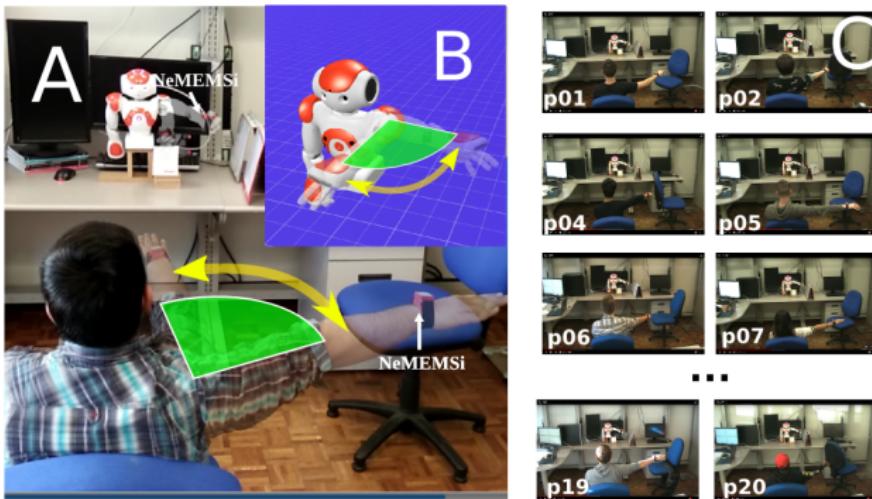
DET fraction of recurrence points that form diagonal lines.
(interpreted as the predictability where, for example, periodic signals show longer diagonal lines than chaotic ones.)

RATIO is the ratio of DET to REC.
(useful to discover dynamic transitions).

LAM computes the recurrence points in the vertical lines
(analogous to DET).

EXPERIMENT

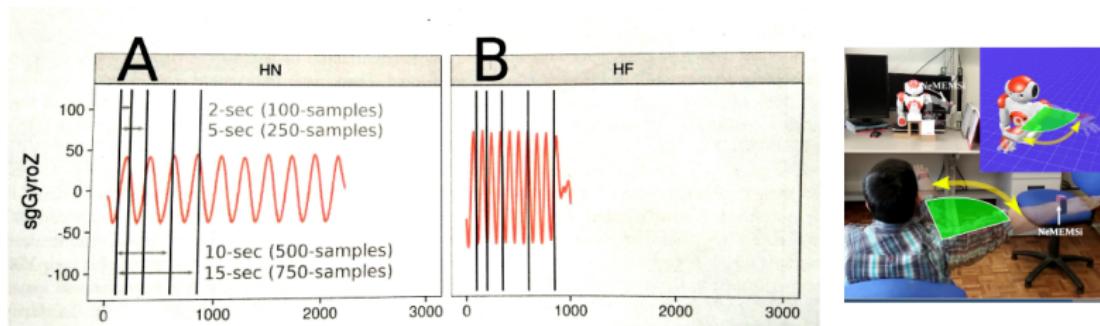
HUMAN-HUMANOID IMITATION ACTIVITIES



Work in progress (Xochicale M. et al. 2018)

Figure 7: (A) Front-to-Front Human-Humanoid Imitation of Horizontal Movements, (B) NAO, humanoid robot, and (C) 20 right-handed healthy participants (age: mean=19.8, SD=1.39).

HORIZONTAL ARM MOVEMENTS

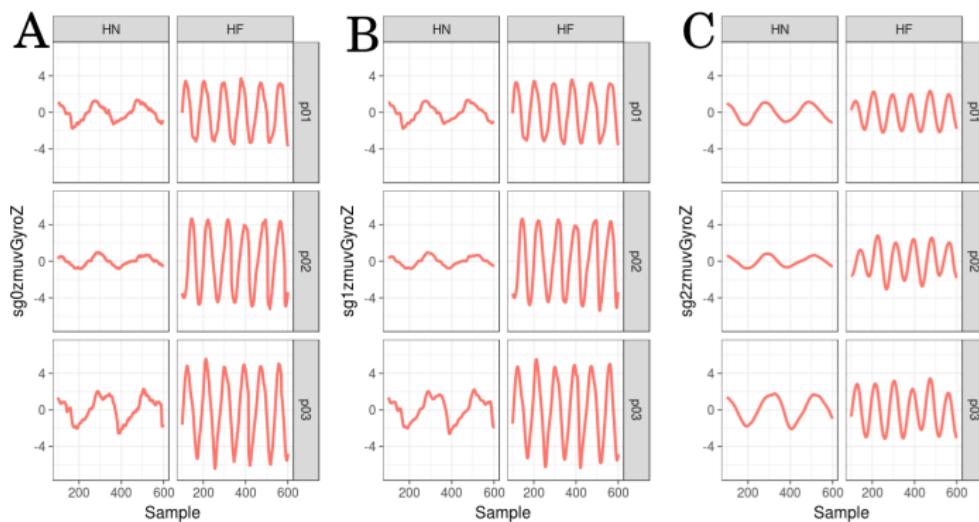


Work in progress (Xochicale M. et al. 2018)

Figure 8: Arm Movements for two speed conditions: Horizontal Normal (HN) and Horizontal Faster (HF) with different window lengths.

RESULTS

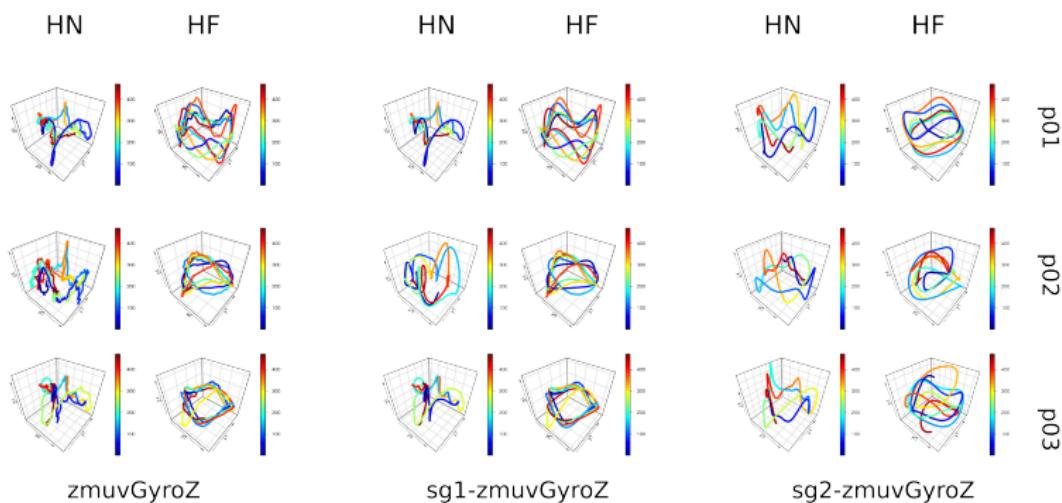
FROM RAW TO SMOOTHED TIME SERIES



Work in progress (Xochicale M. et al. 2018)

Figure 9: (A) Normalised, (B) sgolay($p=5, n=25$), and (C) sgolay($p=5, n=159$)

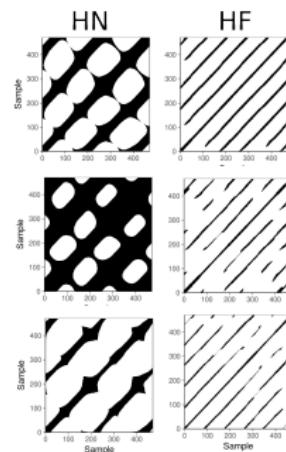
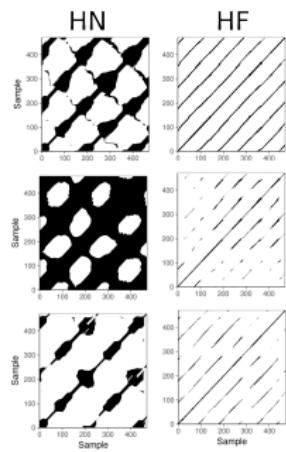
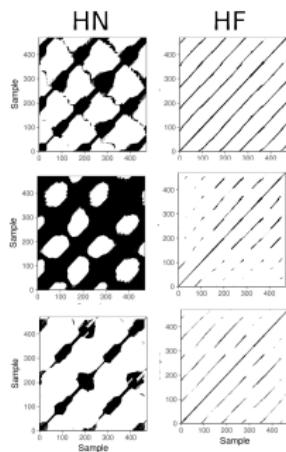
RECONSTRUCTED STATE SPACES



Work in progress (Xochicale M. et al. 2018)

Figure 10: RSS computed with ($m = 7, \tau = 5$)

RECURRENCE PLOTS

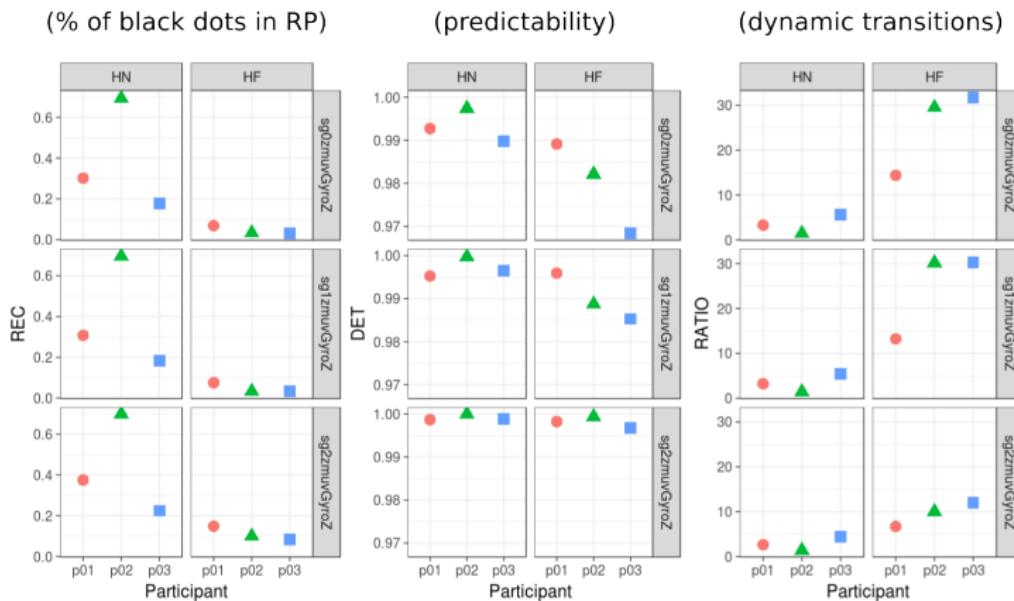


p01 p02 p03

Work in progress (Xochicale M. et al. 2018)

Figure 11: Recurrence Plots computed with ($m = 7$, $\tau = 5$, $\epsilon = 1$)

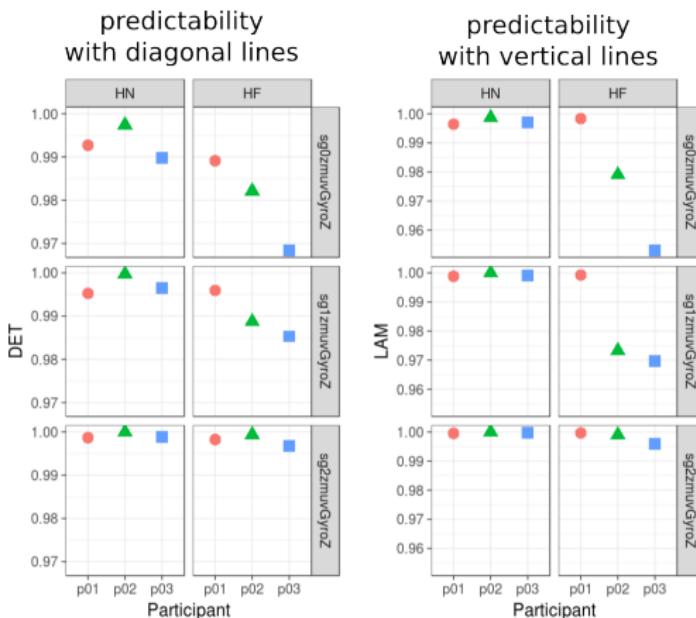
RQA WITH REC, DET & RADIO



Work in progress (Xochicale M. et al. 2018)

Figure 12: REC & RATIO

RQA WITH DET & LAM



Work in progress (Xochicale M. et al. 2018)

Figure 13: REC

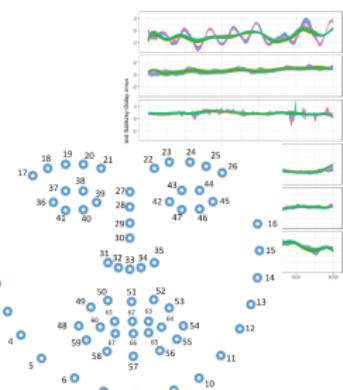
CONCLUSIONS AND FUTURE WORK

CONCLUSIONS

- (+) RSS and RQA can potentially be used to quantify arm movement variability. However,
- (-) RSS and RQA are sensitive to window length, embedding parameters or thresholds.

FUTURE WORK

- Test other techniques of Nonlinear Dynamics, e.g. Lyapunov Exponents, Poincare Maps.
- Use of Convolutional Neural Networks for automatic identification of Movement Variability

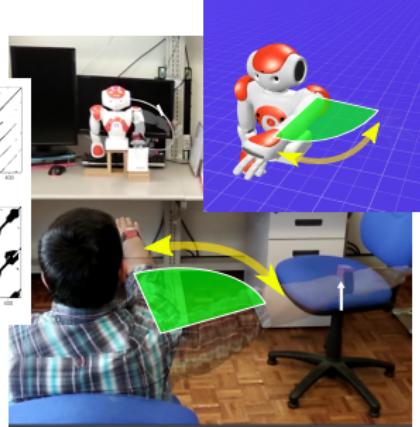
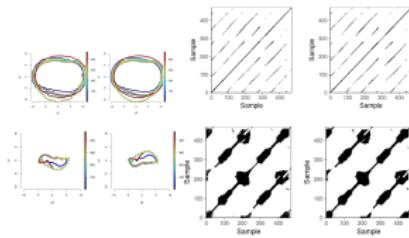


Work in progress (Xochicale M. et al. 2018)

Figure 14: Pilot experiment for quantification of MV for complex movements and changes in facial expressions.

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GitHub repo (2018), <https://github.com/mxochicale/emmov-pilotstudy> [Q]



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