

The grip force dynamics of static grasp reveals a control hierarchy

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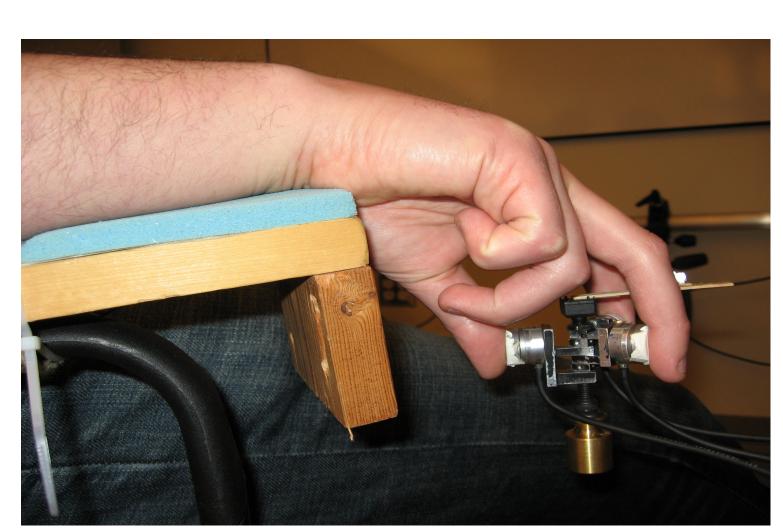




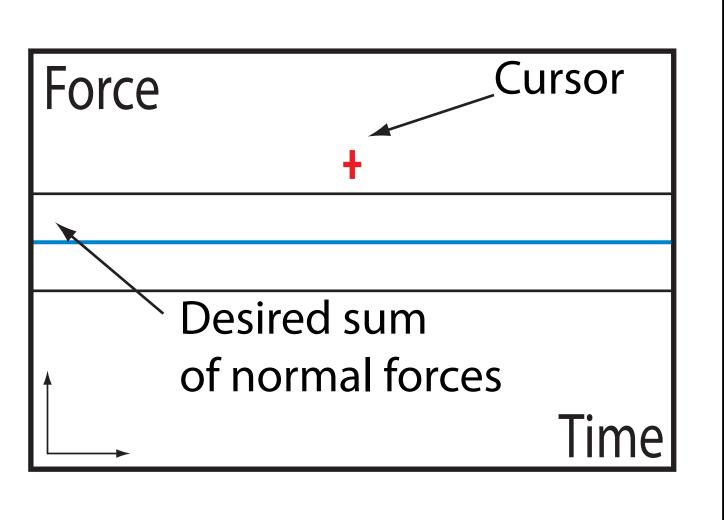
Abstract

- Simple static grasping task: holding against gravity a device with 3 load cells.
- Visual inspection of normal force trajectories reveals structured dynamics, rather than a noise pattern.
- Employed a fluctuation analysis used previously in postural studies [1].
- Detrended Fluctuation Analysis [4] reveals different control strategies over time; multiple scaling regions of displacement as time increases.

Experiments and Data processing

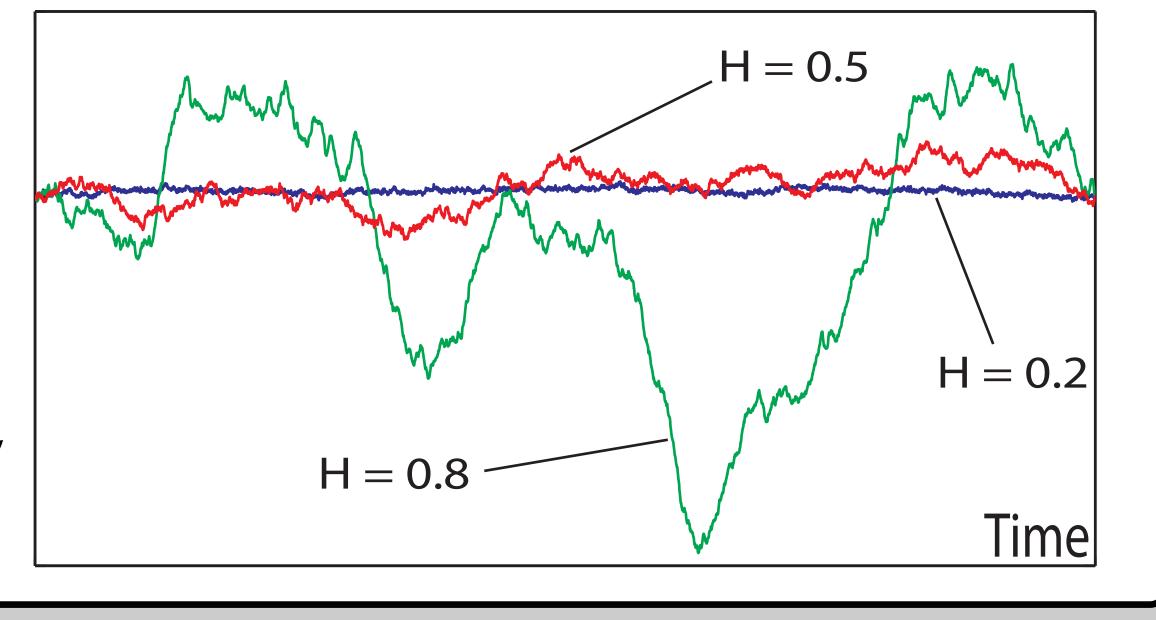


- 5 consenting male adult subjects (ages: 24 44 years).
- Task: Holding a device (see left) statically for 2 min with 3 fingers of dominant hand [5].
- Fingertip contact with 6-axis force transducer (ATI Nano 17) sampling at 400 Hz.
- Repeated with 3 weights: 50 g, 100 g, 200 g, with and without visual feedback.
- Feedback (see right): keep sum of normal forces constant.
- Retained first min, downsampled to 100 Hz, then filtered.



Hurst exponent

- Characterizes the tendency of a time series to stay close to the mean or deviate away from it, or, the scaling of the average displacement over time:
- H < 0.5: Negative correlations, tendency towards mean (see right, blue line).
- H > 0.5 : Positive correlations, tendency away from mean (see right, red line).



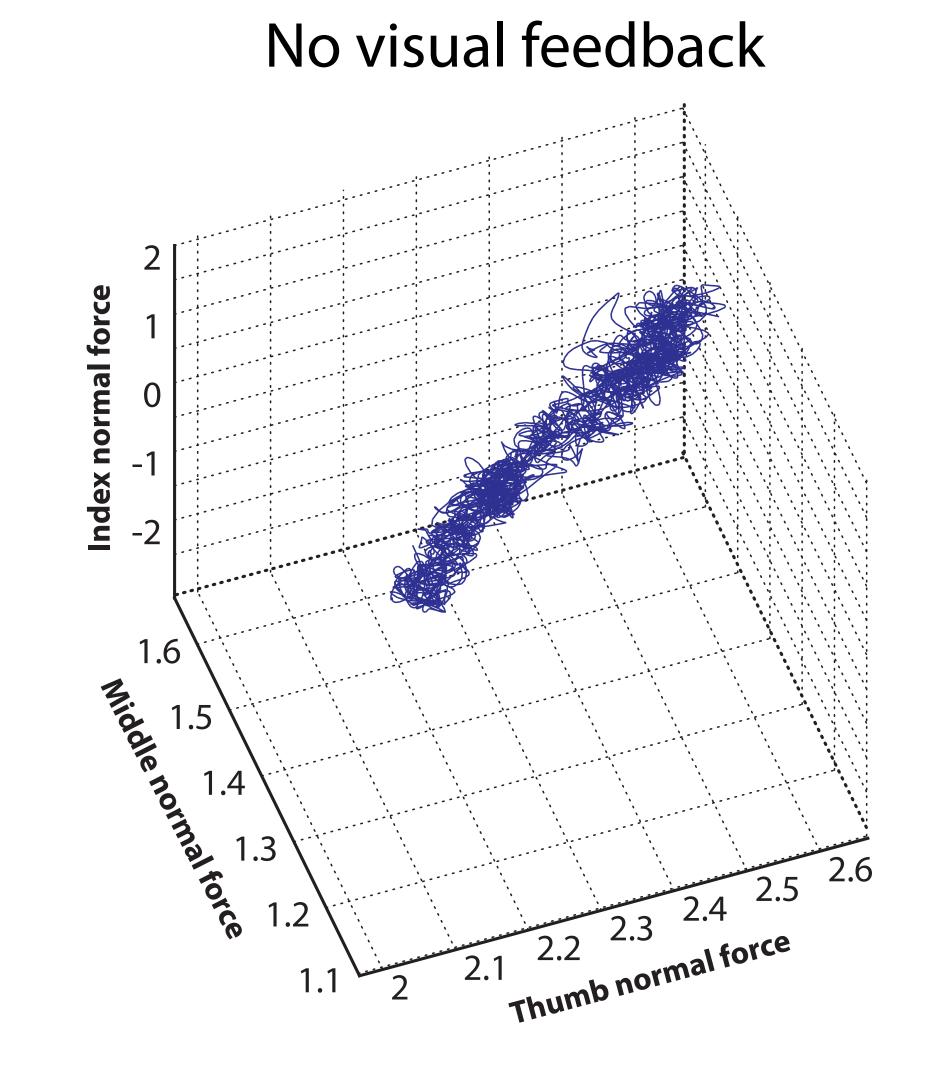
Detrended Fluctuation Analysis

- More refined approach [4] for computing Hurst exponent, which is less sensitive to trends, and detects differential scaling in small vs. large fluctuations.
- Long-range correlations give rise to power-law scaling, which can be quantified as follows:

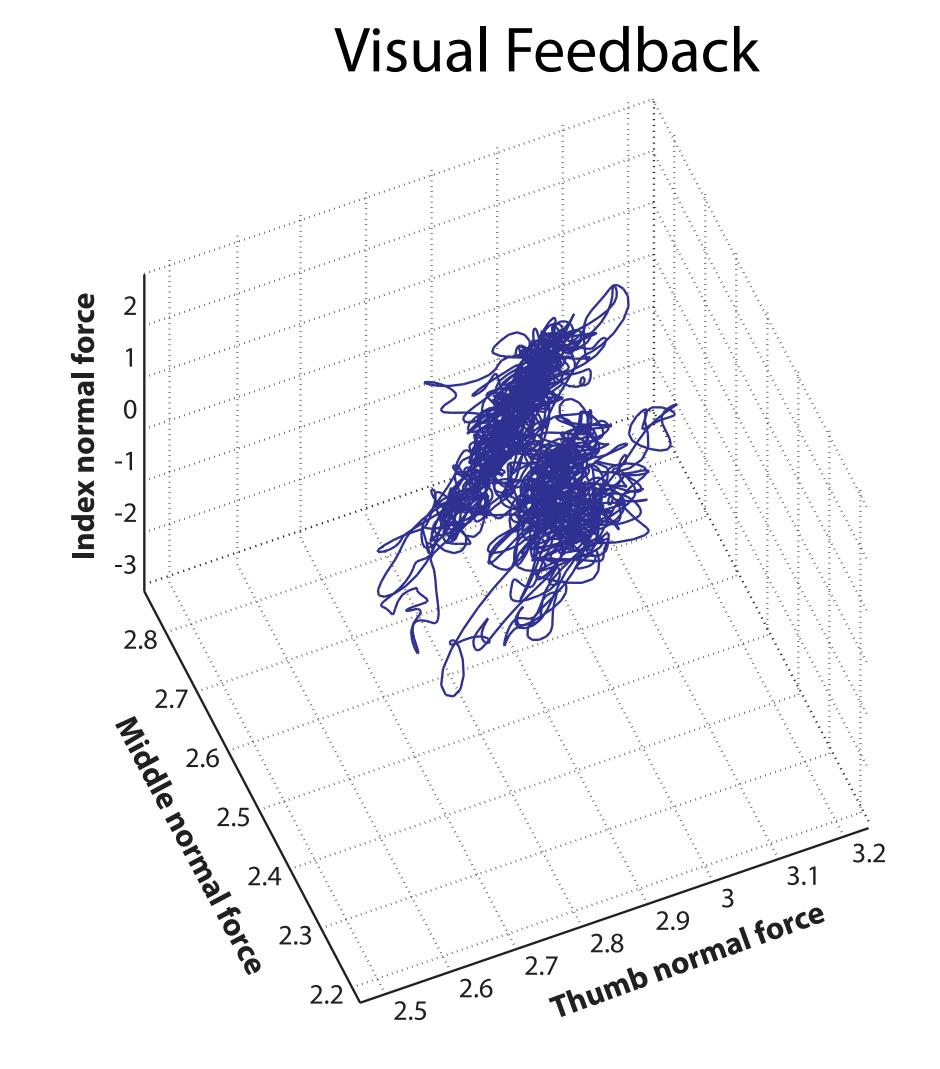
$$F^p(v,s) = \frac{1}{s} \sum_{i=1}^s \left(Y\left[(v-1)s + i \right] - y_p(i) \right)^2 \qquad F_q(s) = \left(\frac{1}{N} \sum_{v=1}^N \left(F^p(v,s) \right)^{\frac{q}{2}} \right)^{\frac{1}{q}}$$

where F(v,s) is the squared mismatch between the v-th segment of data Y, of length s, and the polynomial fit y(i) to it, of order p. The exponent q weights fluctuations according to their size. Doing this for all length-s segments and all s=1,...,N and averaging, we arrive at the **scaling function** with Hurst exponent h(q): $F_q(s) \sim s^{h(q)}$

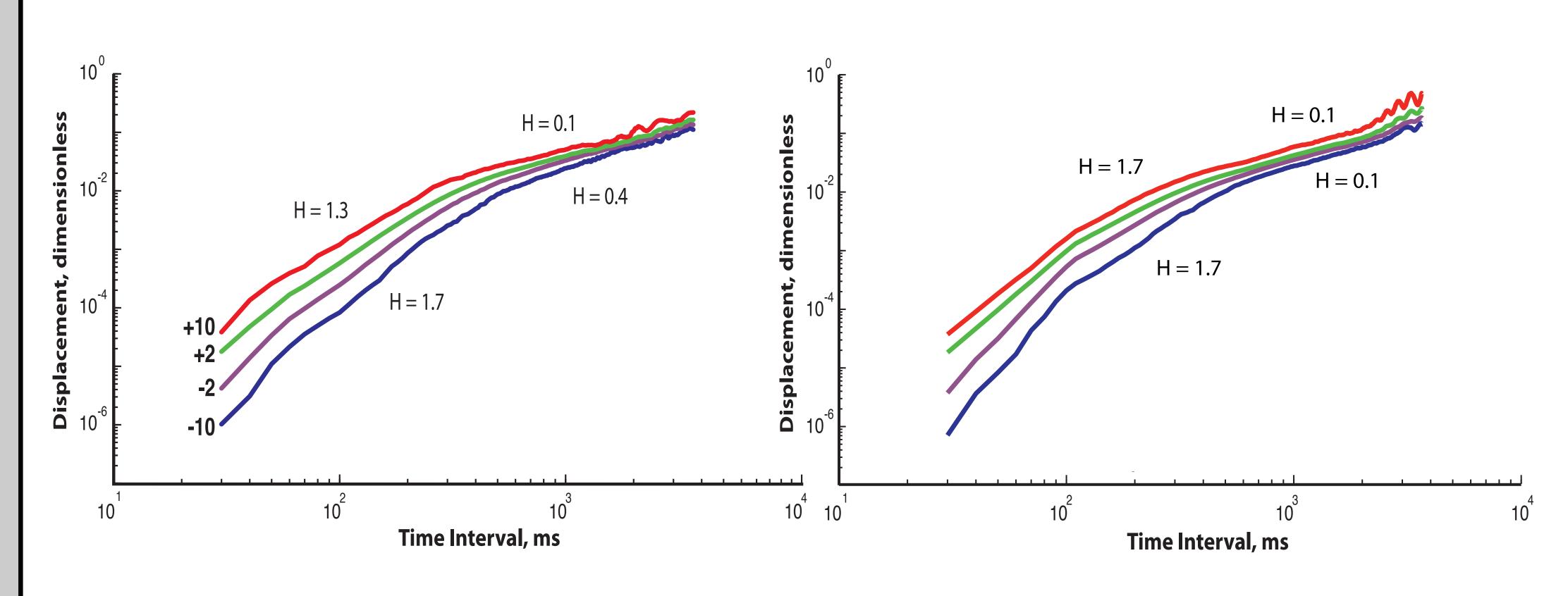
Results







50 g weight. Main axis of variance is [0 -1 1].



DFA plot exhibits multiple scaling regions, and a switch from positive to negative correlations at approx. 400-800 ms.

Conclusion

- The observed dynamics cannot be solely attributed to noise, but strongly suggest stabilizing control activity.
- The Detrended Fluctuation analysis reveals at least two distinct time scales, with **switching** between correlation ranges occurring at 400-800 ms.
- At **short time scales**, the time series tends to move away from the mean, at **long time scales**, it tends to return, indicative of **stabilization**.
- Large fluctuations are more negatively correlated, contributing more to stability.
- These results indicate different control strategies at different time scales, possibly **intermittency** [2], or an effect of **sensorimotor delays** [1].

References and Acknowledgements

- 1. Collins, J.J., De Luca, C.J. (1994). Random Walking during Quiet standing. Phys. Rev. Lett., 73, 764.
- 2. Cabrera, J.L., Milton, J.G.(2002). On-off Intermittency in a Human Balancing Task. Phys. Rev. Lett., 89, 15.
- 3. Kantz, H., Schreiber T. (2003). Nonlinear Time Series Analysis (2nd ed.). Cambridge University Press.
- 3. Kantz, A., Schreiber I. (2003). Nonlinear Time Series Analysis (2nd ed.). Cambridge University Pres.
- 4. Kantelhardt, J.W., Zschiegner, S.A., Koscielny-Bunde, E., Havlin, S., Bunde, A., Stanley, H.E. (2002). Multifractal detrended fluctuation analysis of nonstationary time series. Physica A, 316, 1-4, 87.
- 5. Oldfield, R.C. (1971). The assessment and analysis of handedness: The Edinburgh inventory. Neuropsychologia, 9, 97-113.
- This work is funded in part by grants NIH R01 050520, NSF EFRI-0836042 to Francisco J. Valero-Cuevas.