# Identifying Kinematic Cues for Action Style Recognition

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**Background**: Recognition of emotional states from other's actions is one of key capability for smooth social interaction.

**Approach**: a computational-theory-level analysis kinematic features for recognition of emotional attributes in human actions represented as pointlight display.

### Background

#### Emotion coordinates social interactions

- To provide information to peers about surrounding environment
- Elicit complementary and similar emotions
- To be an incentive promoting social relationship

#### Easy to "read" others' attribute through their actions.

- Identity (Cutting & Kozlowski, 1977; Troje, Westhoff & Lavrov, 2005)
- Gender (Kozlowski & Cutting, 1977; Troje, 2002)
- Emotions (Dittrich et al., 1996; Pollick, 2001; Atkinson, 2009; Hobson & Lee, 1999)
- **Dynamics properties** (the weight of lifted object: Bingham, 1987)

**Results**: emotional attributes in could be identified by covariance of velocity profiles distributed among multiple body parts. Suggestion: the action styles may be mediated by an information channel parallel to action types per se instead of hierarchical manner.

## **Model outline**



#### But not trivial

- Different development recognition different types of emotions (Boone & Cunningham, 1998)
- ASD patients can recognize action types (lifting, pushing), but not style of actions (sad, happy, fear, harsh, gentle) (Moore, et al., 1997; Hubert et al., 2006), can detect coherent random-dot motions, but less with human-biological motion (but evidences are divergent: Kaiser & Shiffrar, 2009).

## **Computational processes (hypothesized)**

#### Dynamic cues

- Duration of actions (Pollick et al., 2001)
- Velocity (DeMeijer, 1989)
- Acceleration (2<sup>nd</sup> order derivatives) (Chang & Troje, 2008; 2009)
- Jerk or (3<sup>rd</sup> order derivatives) (Cook et al., 2009)

#### Structural cues

Body structure (Troje, 2002)

### **<u>4 levels of recognition of bio. Motion</u>** (Troje, 2008)



### **Emotion recognition from action**



, Sparse Logistic Regression (Hierarchical Bayes)





Fig 2: (a) Point-light actor (no link in the model and behavioral test), (a) A temporal profile of right-handelbow-shoulder joint angle (solid line) and its velocity profile (black) in 5 repeating knock, lift, and throw actions.



**Fig 1**: A hypothesized computational process of recognition of biological motion

### **Behavioral study**

- 10 subjects were asked to judge action and emotion types of presented point-light actors.
- *Stimuli*: Action-emotion stimuli were sampled from the biological motion library (Ma et pairwise combinations of 3 actions (knock, lift, and throw) and 3 2006). Nine al., emotions (angry, happy, sad) were sampled from each of 3 selected actors. This yielded 27 video clips in total.

### **Results & Discussion**

- Nested models with incrementally higher order features were compared (*0-DOF models*).
- Velocity model: only velocity profile
- Acceleration model: velocity + acceleration profile,
- *Jerk* model: velocity + acceleration + jerk (third order derivative)

Acceleration or Jerk model fits better than Velocity model Velocity (LL=-93.931/ $R^2$ =0.810) < Acceleration (LL=-90.051/ $R^2$ =0.890) < Jerk (LL=-89.116, -R<sup>2</sup>=0.900) (but not A<J) Action-specific recognizer does not improve fitting

 $W_{jL}$ : weights for the dimension *j* given class index k (normally varying). : hyper parameter (precision) of  $\Lambda_{ik}$ variability of weights.

**Fig 3**: The sparse logistic regression linking the velocity features to given emotion/action class.









#### Fig 5. The variance/covariance in profile significantly velocity relevant to each emotion attribute mapped on a body scheme. The white and gray cell indicates variance/covariance of effective velocity acceleration, and respectively. No lower triangle cells due to its presented were symmetricity. The bottom right showed the number of for each effective dimensions emotion attribute.

Fig 4. The response patterns for each emotion type in human subjects (upper panel) and the velocity model (bottom panel).



### Conclusions

**Results**: Emotional attributes in actions as well as action types could be identified by covariance of velocity profiles among multiple body parts.

Implication: Since, despite different velocity profiles in different actions, these features for emotional attributes were found commonly in multiple different actions, it suggests that the action styles may be mediated by an information channel parallel to action types per se.

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