Evaluating Arm Movement Imitation

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Abstract

In this paper, we present a metric for assessing the quality of arm movement imitation. We develop a joint-rotational-angle-based segmentation and comparison algorithm that rates pairwise similarity of arm movement trajectories on a scale of 1 - 10. We describe an empirical study designed to validate the algorithm we developed, by comparing it to human evaluation of imitation. The results provide evidence that the evaluation of the automated metric did not significantly differ from human evaluation.

1 Introduction

Imitation, the ability to mimic an observed behavior, appears to be an innate means of acquiring novel skills from observation. In humans, it is vital during development and remains an important aspect of social interaction and adaptation throughout life. From the standpoint of artificial intelligence, imitation can prove to be a fast and efficient form of transferring motor skills between robots or computers and humans.

Our research goal is to gain insight into the mechanisms underlying imitation and to explore the aspects and benefits of humans and machines interacting through the use of imitation. In this paper, we focus on the evaluation of motor imitation, specifically arm movement imitation. As part of our experiments, we gathered arm movement data from people imitating video stimuli, and we developed a metric for evaluating the quality of the imitation. The metric was compared to the evaluation of the same imitation by human observers, in order to gain insight into the efficacy of our method, as well as the mechanisms behind human evaluation of motor imitation.

The rest of this paper is organized as follows. In Section 2 of this paper we describe the previous research performed in the area of learning movement by imitation. Section 3 contains a detailed description of our evaluation metric and the components necessary to successfully perform trajectory comparison. In Section 4 we describe our experimental design and results. Section 5 provides a summary of the paper, as well as a presentation of ongoing and future research.

2 Related Work

Imitation is a powerful learning mechanism that has raised the interest of both behavioral sciences and robotics. Much research has been done on imitation in biological systems and, more recently, there has been a growing interest in imitation in synthetic systems, such as robots and autonomous agents. Here we present only a cursory overview of some activities relevant to the particular project described herein.

With the development of humanoid robots and other articulated systems, motor imitation has become a topic of interest to various robotics research groups (Atkeson and Schaal (1997); Billard and Matarić (1998); Ijspeert et al. (2001); Jenkins et al. (2000); Matarić (2001)).

Pomplun and Matarić presented a psychological study of human arm movement imitation, as well as an approach to evaluating imitation which is similar in structure to ours. The evaluation metric did not make a clear enough distinction between pairs of trajectories for the same stimulus and pairs of trajectories for different stimuli(Pomplun and Matarić (2001)).

In their earlier work, the same authors investigated the psychological differences between watching a movement with or without the intention to imitate, as well as the features people fixate on while trying to retain a sequence of movement (Matarić and Pomplun (1997, 1998)). Their findings were that people watching a movement fixated at the end points, and that the pupils of the subjects watching with the intent to imitate were more dilated than those of the rest of the subjects (Matarić and Pomplun (1997, 1998)).

Other studies compared human movements to the output of a computer vision systems, by using a simple mean square error metric. They demonstrated that in many cases, even pairs of movements with large MSE were still perceived as identical by human observers (Goncalves et al. (1995, 1998, 1999)).

Nehaniv and Dautenhahn (2001) have provided a formal framework for the problem of correspondence between dissimilar bodies. Within that framework, Alisandrakis et al. (2002) have explored multiple metrics, all of which have been specific to the particular agents and problems being considered.

Liu and Geiger (1999) and Sebastian et al. (2001) presented a framework for silhouette comparison, addressing the issue of identifying topological changes due to the original 3D scenarios and articulations. These methods did not prove to be extendable to 3D movement comparison.

The past research in imitation motivates the research of more sophisticated metrics evaluating the similarity of limb trajectories.

3 Approach

In order to evaluate arm movement imitation, we developed a metric for the distance between arm movement trajectories. Given two trajectories, our metric computes a similarity score on a 1 - 10 scale. Our aim was to develop a metric yielding higher values for more similar trajectories. The metric comprises of three techniques: segmentation, time scaling, and raw trajectory comparison. We describe each in turn in the next section.

Our approach is similar in structure to the work of Pomplun and Matarić. Both their algorithm and ours are based on the techniques of segmentation, time scaling and raw trajectory comparison. The techniques however differ significantly in their design.

We decided to apply the evaluation in joint-space, because, unlike Cartesian coordinates, joint angles are independent from the length of the subjects' limbs, and thus compensate for physical differences. We derived six rotational angles from Cartesian coordinates. The angles represented the roll, pitch, and yaw of the upper and lower arm respectively.

Our evaluation metric compares two arm movement trajectories, α and β , containing T_{α} and T_{β} samples for each of the J joint angles. The samples will be referred to as $\alpha_j(t)$ and $\beta_j(t)$. The metric encompasses three techniques: segmentation, time scaling, and raw trajectory comparison. We describe each in turn next.

3.1 Segmentation

The metric heavily relies on segmenting trajectories α and β into simpler trajectory-segments $\alpha^{1..S^{\alpha}}$ and $\beta^{1..S^{\beta}}$, and comparing each of them individually.

The change in the direction of joint rotations is a natural indicator of transitions between successive segments in a trajectory. Our approach thus relies on using change in direction of joint rotations for segmenting trajectories α and β .

One challenge of this approach is finding the frame where the direction of joint rotations really changed, and was not just the result of noise. To address this challenge, we used three parameters, ϵ , δ , and γ . δ represents the minimum difference between two values required to consider the values distinct. ϵ represents the minimum number of ensuing frames that need to be consistent with the change in direction of rotation in order for the variation not to be considered noise. A frame f is considered consistent with the change in the direction of rotation of joint j if, at frame f, j's rotation continues to occur in the changed direction within error δ . γ represents the minimum duration of a segment, in frames.

Instead of estimating appropriate values for ϵ , δ , and γ , the segmentation algorithm performs a parameter space search. It systematically performs segmentations with different combinations of integer values for the three parameters. ϵ ranges from 1 to 8, δ from 1 to 15, and γ from 3 to 8. Empirical investigations found that expanding these intervals does not yield significantly better results. For each combination of segmentations, a distance metric is computed as illustrated in Figure 1, and the segmentation of α and β yielding the highest similarity score is used as the final imitation score.

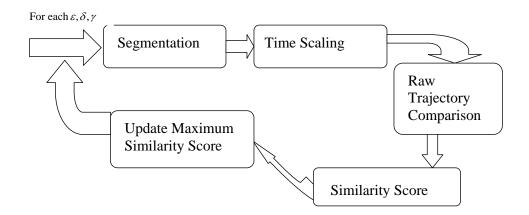


Figure 1: Execution of Evaluation Algorithm

3.2 Time Scaling

In order to derive a metric that is invariant to the absolute speed of movement, time scaling is performed for each pair of corresponding trajectory-segments. The shorter of two trajectory segments, say α^i , is expanded to contain as many samples as its counterpart, β^i . In order to perform this expansion, we noted that the values of rotational joint angles form a continuous function with respect to time. It is impossible for a rotational joint angle to skip from a value θ_1 to a value θ_2 without taking at least all other values in between.

Formally, let α^i be the continuous function corresponding to trajectory segment α^i . Then, it is reasonable to assume that if $\alpha^i_j(t) = \theta_1$ and $\alpha^i_j(t+1) = \theta_2$, then $\hat{\alpha}^i_j(t) = \theta_1$, $\hat{\alpha}^i_j(t+1) = \theta_2$, and $\hat{\alpha}^i_j$ takes all the values in between θ_1 and θ_2 during the time interval [t..t+1]. Additionally, because the time interval is very short, it is practical to assume that α^i_j was monotone on this interval.

 α^i can thus be approximated on each interval [t..t+1] by the following formula:

$$\begin{aligned} \hat{\alpha}_j^i(t) &= \theta_1 \\ \hat{\alpha}_j^i(t) &= \theta_2 \\ \hat{\alpha}_j^i(t_x) &= \theta_1 + (t_x - t) * (\theta_2 - \theta_1), \\ \text{where } t_x \in [t..t + 1] \end{aligned}$$

 α_i^i is expanded using $\hat{\alpha}_i^i$ in the following way:

$$\alpha_j^i(t_{new}) = \hat{\alpha_j^i}(\frac{t_{new} * T_{\gamma}}{T_{\omega}})$$

3.3 Raw Trajectory Comparison

After identifying and time-scaling two corresponding trajectory segments, α^i and β^i , a distance metric is applied in order to determine how similar the two segments are. The distance metric records the number of joint angle samples that were significantly different from each other:

$$d(\alpha^{i},\beta^{i}) = \sum_{t=0}^{T_{\alpha^{i},\beta^{i}}} \sum_{j=1}^{J} |\alpha_{j}^{i}(t) - \beta_{j}^{i}(t)| > \epsilon.$$

Two angles are significantly different from each other if their difference exceeds a threshold, ϵ . We determined empirically that an appropriate value for the angle-difference threshold, ϵ , is 25 degrees. Angle differences of 25 degrees or less are not identified by human observers comparing two movements performed by different persons.

3.4 Rating the Movement

After the above three techniques are applied to the movements being compared, an imitation score, R, is determined from the percentage of correct joint angle samples across all segments:

$$R = min(10, round(10 * \frac{\sum_{s=0}^{min(S_{\alpha}, S_{\beta})} d(\alpha^{s}, \beta^{s}) + Rsample)}{Nsamples}) + 1),$$

where *Nsamples* is the total number of joint angle samples, and *Rsamples* is the number of joint angle samples of the trajectory segments with no counterpart (if the movements contain a different number of trajectory-segments).

4 Evaluation

To evaluate the metric we developed, we designed and performed an empirical study which compared the metric to human evaluation of imitation. The study was performed at USC's Interaction Lab, part of the USC Robotics Lab.

4.1 Experimental Design

The purpose of this experiment was to test an evaluation metric developed to compare two arm movements. The metric rates the similarity of two arm movements on a scale from 1 - 10.

4.1.1 Participants

The subject pool consisted of 6 subjects: 1 male and 5 female. Four of the participants were students and two were faculty members. The handedness of the subjects was not taken into account. All subjects were instructed to perform the movements with the right arm, regardless of arm dominance.

4.1.2 Stimuli

The stimuli consisted of 6 video clips, ranging from 4 seconds to 9 seconds in duration. The movements of one of the authors, Alexandra Constantin, were videotaped to be shown as stimuli to the experimental subjects. She participated in a special session to produce reference data for comparison.

Part of the movements contained smooth, circular trajectories. Others were more linear in nature. All movements started with a straightened arm and typically involved movements of the upper-arm and the lower-arm. The stimuli involved no external objects or recognized patterns, so that they could be thought of as goal-independent; the only goal the subjects had was to imitate the arm movement as accurately as possible.

4.1.3 Apparatus

Stimuli videos were presented on a 15" computer monitor. Subjects' arm movements were tracked using a custom-made motion capture system developed in the USC Interaction Lab. The system is comprised of a set of small, lightweight sensors. Each sensor provides its own global orientation and is physically and computationally independent, requiring only external communication. Orientation information from sensors is communicated via wireless to host computer for processing (Miller et al. (2004)).

4.2 Evaluation of the Data

4.2.1 Method

The subject and the experimenter entered a room equipped with a computer and two motion tracking sensors. The two sensors were strapped on the subject's arm, one above the elbow and one above the wrist. The sensor suit was then turned on by the experimenter, making sure that the sensors were initiated correctly.

Prior to each recording trial, one of the 6 video stimuli was shown to the subject on the computer monitor. The trials were divided into two conditions: "rehearsal" and "non rehearsal". In the "rehearsal" condition, the subject was asked to imitate the movement while watching the video. In the "non-rehearsal" condition, the subject was asked to imitate the movement right after having watched the video.

To initiate the imitation process in either condition, the experimenter gave the verbal signal "ready". The subject then straightened his/her arm. Next, the subject was given the instruction "imitate", at which point s/he started imitating the movement shown in the video (concurrently or just prior the imitation). When finished imitating, the subject said "done". The movements of the subject were recorded, through the motion tracking sensors, during the time period between the "imitate" and "done" signals.

Each subject was presented with 6 stimuli, 3 in the "rehearsal" condition and 3 in the "non-rehearsal" condition. The order the stimuli were presented in was randomized across subjects. Each stimulus was shown and imitated three times in succession.

After all movements had been captured, the automated evaluation metric was applied on pairs of files corresponding to the same movement.

Next, the metric results were compared to human evaluation. Human evaluation was obtained as described below.

The execution of the movements was videotaped and observed by three unbiased viewers. The viewers rated the similarity of the movements on a scale of 1-10 for each of the imitation trials, with higher scores corresponding to more similar movements.

Finally, the scores of the three evaluators of the same experiments were compared to those computed by the automated metric.

4.3 **Results and Discussion**

Our results indicate that, given the variance in human rating, the scores given by the human viewers did not significantly differ from those given by the automated metric. The analysis of variance(ANOVA) test revealed no significant effect of the raters on the imitation scores, F = 1.47, p = 0.22. The mean score given by the automated metric was 7.7; the mean scores given by the three human viewers were 7.4, 7.6, and 7.5 respectively. A distribution of the scores given by each rater is presented in Figure 2.

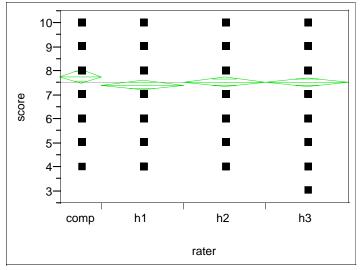


Figure2: Mean Diamond Plot of Scores Given by Each Rater

In addition to validating the auto-mated imitation metric, we also acquired additional data about human imitation. Specifically, the subjects' performance significantly improved from the first to the third imitation of the same stimulus, as shown in Figure 3. In addition to this we discovered that when the subject executed the movement while viewing the stimulus their performance did not significantly differ from when they executed the movement subsequent to viewing the stimulus.

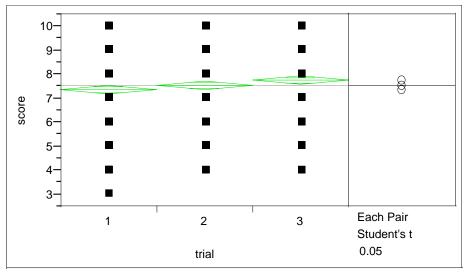


Figure 3: Mean Diamond Plot of Scores Received after Each Consequent Imitation (trial)

Given the variance in human ratings, evaluating imitation seems to be very subjective. Human observers varied in their evaluation even when presented with the same pair of movements at different times.

5 Conclusions and Future Work

We have described an evaluation metric for human movement imitation. The first results have provided encouraging feedback on the use of the metric for evaluating arm movement imitation. Future work could test the accuracy of the metric in evaluating the imitation of more limbs, or complete body movements.

Our ongoing work focuses on providing instructive feedback encouraging the learner to work on the least accurate parts of his/her performance with the help of repeated, amplified, and focused demonstrations. Towards this end, we are working on replacing the video stimuli with a 3D skeletal animation.

The successful implementation of this work could lead to an application capable of teaching humans how to perform certain movements, demonstrated either through a graphical interface or by a humanoid robot.

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Appendix – Data Table

This table contains the data collected in our empirical study. Six subjects participated in the study. They imitated six different movements. Each movement was imitated in three subsequent trials. The movements were repeated in either the "rehearsal" (r) or non-rehearsal (nr) condition, as described in the experimental design section. The imitations were evaluated by four different rankers: the automated metric ("comp"), and three human observers ("h1"-"h3").

Rater	Subject	Movement	Trial	Condition		Rater	Subject	Movement	Trial	Condition	Score
comp	1	1	1	r	9	h3	1	3	2	nr	8
h1	1	1	1	r	9	comp	1	3	3	nr	10
h1	1	1	1	r	9	h1	1	3	3	nr	8
h2	1	1	1	r	9	h1	1	3	3	nr	10
h2	1	1	1	r	9	h2	1	3	3	nr	9
h3	1	1	1	r	8	h2	1	3	3	nr	10
h3	1	1	1	r	10	h3	1	3	3	nr	8
comp	1	1	2	r	10	h3	1	3	3	nr	10
h1	1	1	2	r	10	comp	1	4	1	nr	10
h1	1	1	2	r	8	h1	1	4	1	nr	10
h2	1	1	2	r	10 9	h1	1	4 4	1	nr nr	10 9
h2	1	1	2	r r	9	h2 h2	1	4 4	1	nr	9
h3	1	1	2	r	7	h2 h3	1	4	1	nr	9
h3	1	1	3	r	9	h3	1	4	1	nr	9
comp h1	1	1	3	r	8	comp	1	4	2	nr	10
h1	1	1	3	r	9	h1	1	4	2	nr	10
h1 h2	1	1	3	r	8	h1	1	4	2	nr	9
h2	1	1	3	r	10	h1 h2	1	4	2	nr	10
h3	1	1	3	r	8	h2	1	4	2	nr	9
h3	1	1	3	r	9	h2	1	4	2	nr	10
comp	1	2	1	r	9	h3	1	4	2	nr	9
h1	1	2	1	r	9	comp	1	4	3	nr	10
h1	1	2	1	r	9	h1	1	4	3	nr	9
h2	1	2	1	r	10	h1	1	4	3	nr	9
h2	1	2	1	r	9	h2	1	4	3	nr	10
h3	1	2	1	r	10	h2	1	4	3	nr	10
h3	1	2	1	r	9	h3	1	4	3	nr	9
comp	1	2	2	r	9	h3	1	4	3	nr	9
h1	1	2	2	r	9	comp	1	5	1	r	6
h1	1	2	2	r	8	h1	1	5	1	r	6
h2	1	2	2	r	10	h1	1	5	1	r	6
h2	1	2	2	r	9	h2	1	5	1	r	7
h3	1	2	2	r	10	h2	1	5	1	r	6
h3	1	2	2	r	9	h3	1	5	1	r	5
comp	1	2	3	r	9	h3	1	5	1	r	7
h1	1	2	3	r	8	comp	1	5	2	r	6
h1	1	2	3	r	8	h1	1	5	2	r	7
h2	1	2	3	r	9	h1	1	5	2	r	6
h2	1	2	3	r	9 9	h2	1	5	2	r	8
h3 h3	1	2	3	r r	9	h2 h3	1	5	2	r r	8
	1	3	1	nr	9 10	h3 h3	1	5	2	r	8
comp h1	1	3	1	nr	8	comp	1	5	3	r	7
h1	1	3	1	nr	9	h1	1	5	3	r	7
h1 h2	1	3	1	nr	9	h1	1	5	3	r	7
h2	1	3	1	nr	9	h1 h2	1	5	3	r	7
h2 h3	1	3	1	nr	10	h2	1	5	3	r	7
h3	1	3	1	nr	10	h2	1	5	3	r	7
comp	1	3	2	nr	9	h3	1	5	3	r	7
h1	1	3	2	nr	8	comp	1	6	1	nr	10
h1	1	3	2	nr	9	h1	1	6	1	nr	7
h2	1	3	2	nr	9	h1	1	6	1	nr	7
h2	1	3	2	nr	9	h2	1	6	1	nr	7
h3	1	3	2	nr	8	h2	1	6	1	nr	7

		Movement		Condition				Movement		Condition	
h3 h3	1	6	1	nr nr	6 9	h1 h1	2	3	3	r	9 7
comp	1	6	2	nr	9	h1 h2	2	3	3	r	8
h1	1	6	2	nr	8	h2	2	3	3	r	7
h1	1	6	2	nr	8	h3	2	3	3	r	8
h2	1	6	2	nr	8	h3	2	3	3	r	7
h2	1	6	2	nr	8	comp	2	4 4	1	nr	7 7
h3 h3	1	6	2	nr nr	8	h1 h1	2	4	1	nr nr	7
comp	1	6	3	nr	10	h1 h2	2	4	1	nr	7
h1	1	6	3	nr	8	h2	2	4	1	nr	7
h1	1	6	3	nr	9	h3	2	4	1	nr	8
h2	1	6	3	nr	7	h3	2	4	1	nr	7
h2	1	6	3	nr nr	8	comp	2	4	2	nr	9 8
h3 h3	1	6	3	nr	8	h1 h1	2	4	2	nr	7
comp	2	1	1	r	10	h2	2	4	2	nr	7
h1	2	1	1	r	8	h2	2	4	2	nr	7
h1	2	1	1	r	8	h3	2	4	2	nr	7
h2	2	1	1	r	8	h3	2	4	2	nr	8
h2	2	1	1	r	7 8	comp	2	4 4	3	nr	9 10
h3 h3	2	1	1	r	8	h1 h1	2	4	3	nr	7
comp	2	1	2	r	10	h1 h2	2	4	3	nr	7
h1	2	1	2	r	9	h2	2	4	3	nr	7
h1	2	1	2	r	9	h3	2	4	3	nr	7
h2	2	1	2	r	8	h3	2	4	3	nr	5
h2	2	1	2	r r	8 9	comp	2	5	1	nr	6 4
h3 h3	2	1	2	r	7	h1 h1	2	5	1	nr	4
comp	2	1	3	r	10	h1 h2	2	5	1	nr	5
h1	2	1	3	r	8	h2	2	5	1	nr	4
h1	2	1	3	r	8	h3	2	5	1	nr	4
h2	2	1	3	r	8	h3	2	5	1	nr	5
h2	2	1	3	r	6 7	comp	2	5	2	nr	5
h3 h3	2	1	3	r	6	h1 h1	2	5	2	nr nr	5
comp	2	2	1	nr	8	h1 h2	2	5	2	nr	6
h1	2	2	1	nr	7	h2	2	5	2	nr	5
h1	2	2	1	nr	7	h3	2	5	2	nr	5
h2	2	2	1	nr	6	h3	2	5	2	nr	6
h2	2	2 2	1	nr	5 6	comp	2	5	3	nr	5 4
h3 h3	2	2	1	nr nr	6	h1 h1	2	5	3	nr nr	5
comp	2	2	2	nr	9	h1 h2	2	5	3	nr	6
h1	2	2	2	nr	7	h2	2	5	3	nr	5
h1	2	2	2	nr	7	h3	2	5	3	nr	6
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h2 h3	2	2 2	2	nr nr	6 8	comp h1	2	6 6	1	r r	7 8
h3 h3	2	2	2	nr	7	h1 h1	2	6	1	r	8
comp	2	2	3	nr	8	h2	2	6	1	r	7
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h2	2	3 3	1	r	8 7	h3	2	6 6	2	r	6 7
h2 h3	2	3	1	r	9	comp h1	2	6	3	r r	8
h3 h3	2	3	1	r	8	h1 h1	2	6	3	r	8
comp	2	3	2	r	7	h2	2	6	3	r	7
h1	2	3	2	r	5	h2	2	6	3	r	7
h1	2	3	2	r	6	h3	2	6	3	r	8
h2	2	3	2	r	6	h3	2	6	3	r	9
h2 h3	2	3	2	r	6 5	comp b1	3	1	1	nr nr	8
h3 h3	2	3	2	r	5	h1 h1	3	1	1	nr	8
comp	2	3	3	r	10	h1 h2	3	1	1	nr	9
comp	-	1 -				112		1 -			, í

		Movement		Condition				Movement		Condition	
h2 h3	3	1	1	nr nr	7 8	comp h1	3	4 4	3	r r	8 9
h3	3	1	1	nr	8	h1	3	4	3	r	8
comp	3	1	2	nr	8	h2	3	4	3	r	10
h1	3	1	2	nr	8	h2	3	4	3	r	10
h1	3	1	2	nr	8	h3	3	4 4	3	r	9 8
h2 h2	3	1	2	nr nr	8	h3 comp	3	5	1	r r	5
h3	3	1	2	nr	8	h1	3	5	1	r	7
h3	3	1	2	nr	9	h1	3	5	1	r	5
comp	3	1	3	nr	8	h2	3	5	1	r	6
h1	3	1	3	nr nr	8	h2 h3	3	5	1	r r	5
h1 h2	3	1	3	nr	9	h3	3	5	1	r	6
h2	3	1	3	nr	8	comp	3	5	2	r	5
h3	3	1	3	nr	8	h1	3	5	2	r	6
h3	3	1	3	nr	9	h1	3	5	2	r	5
comp h1	3	2 2	1	r r	8 7	h2 h2	3	5	2	r r	6 6
h1	3	2	1	r	8	h2	3	5	2	r	7
h2	3	2	1	r	9	h3	3	5	2	r	6
h2	3	2	1	r	8	comp	3	5	3	r	6
h3	3	2 2	1	r	9	h1	3	5	3	r	7
h3 comp	3	2	1 2	r r	9 7	h1 h2	3	5	3	r r	6 7
h1	3	2	2	r	6	h2	3	5	3	r	5
h1	3	2	2	r	7	h3	3	5	3	r	7
h2	3	2	2	r	8	h3	3	5	3	r	6
h2	3	2 2	2	r	7 8	comp	3	6 6	1	nr	7 8
h3 h3	3	2	2	r r	7	h1 h1	3	6	1	nr nr	8
comp	3	2	3	r	7	h2	3	6	1	nr	7
h1	3	2	3	r	6	h2	3	6	1	nr	8
h1	3	2	3	r	7	h3	3	6	1	nr	9
h2	3	2 2	3	r r	8	h3	3	6 6	1	nr nr	8 7
h2 h3	3	2	3	r	8	comp h1	3	6	2	nr	7
h3	3	2	3	r	7	h1	3	6	2	nr	8
comp	3	3	1	nr	6	h2	3	6	2	nr	6
h1	3	3	1	nr	8	h2	3	6	2	nr	8
h1 h2	3	3	1	nr	9	h3 h3	3	6 6	2	nr nr	8 7
h2	3	3	1	nr	7	comp	3	6	3	nr	7
h3	3	3	1	nr	8	h1	3	6	3	nr	7
h3	3	3	1	nr	7	h1	3	6	3	nr	8
comp	3	3	2	nr nr	6 8	h2	3	6 6	3	nr nr	6 9
h1 h1	3	3	2	nr	8	h2 h3	3	6	3	nr	7
h2	3	3	2	nr	8	h3	3	6	3	nr	8
h2	3	3	2	nr	8	comp	4	1	1	r	7
h3	3	3	2	nr	9 8	h1	4	1	1	r	7
h3 comp	3	3 3	3	nr	8 6	h1 h2	4	1	1	r r	8
h1	3	3	3	nr	8	h2	4	1	1	r	8
h1	3	3	3	nr	8	h3	4	1	1	r	8
h2	3	3	3	nr	7	h3	4	1	1	r	7
h2	3	3 3	3	nr	8 9	comp	4	1	2	r	9 8
h3 h3	3	3	3	nr nr	8	h1 h1	4	1	2	r r	8
comp	3	4	1	r	7	h2	4	1	2	r	9
h1	3	4	1	r	7	h2	4	1	2	r	8
h1	3	4	1	r	8	h3	4	1	2	r	7
h2	3	4 4	1	r r	8	h3	4	1	2	r r	8 9
h2 h3	3	4	1	r	8 9	comp h1	4	1	3	r	8
h3	3	4	1	r	8	h1	4	1	3	r	8
comp	3	4	2	r	6	h2	4	1	3	r	10
h1	3	4	2	r	8	h2	4	1	3	r	8
h1	3	4	2	r	8 9	h3	4	1	3	r	8 7
1.0	3	4		r		h3	4	1	3	r	
h2 h2	3	4	2	r	9	comp		2		pr	×
h2 h2 h3	3	4 4	2	r r	9 9	comp h1	4	2 2	1	nr nr	8 6

		Movement		Condition				Movement		Condition	
h2	4	2 2	1	nr nr	7 7	h3	4	5	2	nr nr	6 5
h2 h3	4	2	1	nr	7	comp h1	4	5	3	nr	7
h3	4	2	1	nr	7	h1	4	5	3	nr	6
comp	4	2	2	nr	7	h2	4	5	3	nr	6
h1	4	2	2	nr	7	h2	4	5	3	nr	7
h1	4	2	2	nr	7	h3	4	5	3	nr	7
h2	4	2 2	2	nr	8	h3	4	5 6	3	nr	6 7
h2 h3	4	2	2	nr nr	8	comp h1	4	6	1	r r	6
h3	4	2	2	nr	7	h1	4	6	1	r	5
comp	4	2	3	nr	7	h2	4	6	1	r	5
h1	4	2	3	nr	6	h2	4	6	1	r	6
h1	4	2	3	nr	7	h3	4	6	1	r	6
h2	4	2	3	nr nr	8 9	h3	4	6	1 2	r r	4 8
h2 h3	4	2	3	nr	8	comp h1	4	6	2	r	6
h3	4	2	3	nr	7	h1	4	6	2	r	5
comp	4	3	1	r	5	h2	4	6	2	r	4
h1	4	3	1	r	8	h2	4	6	2	r	4
h1	4	3	1	r	8	h3	4	6	2	r	5
h2	4	3	1	r	8	h3	4	6 6	2	r r	5 7
h2 h3	4	3	1	r	8	comp h1	4	6	3	r	7
h3	4	3	1	r	8	h1	4	6	3	r	7
comp	4	3	2	r	8	h2	4	6	3	r	6
h1	4	3	2	r	9	h2	4	6	3	r	5
h1	4	3	2	r	9	h3	4	6	3	r	6
h2	4	3	2	r	8	h3	4	6	3	r	9
h2	4	3	2	r	8	comp	5	1	1	nr nr	10 9
h3 h3	4	3	2	r	9	h1 h1	5	1	1	nr	9
comp	4	3	3	r	9	h2	5	1	1	nr	9
h1	4	3	3	r	9	h2	5	1	1	nr	9
h1	4	3	3	r	9	h3	5	1	1	nr	9
h2	4	3	3	r	9 8	h3	5	1	1 2	nr	9
h2 h3	4	3	3	r	8 10	comp h1	5	1	2	nr nr	8
h3	4	3	3	r	9	h1	5	1	2	nr	9
comp	4	4	1	nr	9	h2	5	1	2	nr	9
h1	4	4	1	nr	9	h2	5	1	2	nr	10
h1	4	4	1	nr	9	h3	5	1	2	nr	9
h2	4	4	1	nr nr	9 8	h3	5	1	2	nr	10 9
h2 h3	4	4	1	nr	10	comp h1	5	1	3	nr nr	8
h3	4	4	1	nr	9	h1	5	1	3	nr	9
comp	4	4	2	nr	9	h2	5	1	3	nr	10
h1	4	4	2	nr	9	h2	5	1	3	nr	9
h1	4	4	2	nr	9 10	h3	5 5	1	3	nr	10 9
h2 h2	4	4 4	2	nr nr	8	h3 comp	5	2	3	nr nr	8
n2 h3	4	4	2	nr	9	comp h1	5	2	1	nr	7
h3	4	4	2	nr	9	h1	5	2	1	nr	7
comp	4	4	3	nr	9	h2	5	2	1	nr	8
h1	4	4	3	nr	8	h2	5	2	1	nr	7
h1 h2	4	4	3	nr nr	9 9	h3 h3	5 5	2 2	1	nr nr	7 6
h2 h2	4	4	3	nr	9	h3 comp	5	2	2	nr	0
h3	4	4	3	nr	10	h1	5	2	2	nr	8
h3	4	4	3	nr	9	h1	5	2	2	nr	8
comp	4	5	1	nr	6	h2	5	2	2	nr	7
h1	4	5	1	nr	7	h2	5	2	2	nr	8
h1	4	5 5	1	nr	6 5	h3	5 5	2 2	2	nr	8 7
h2 h2	4	5	1	nr nr	5	h3 comp	5	2 2	2	nr nr	10
n2 h3	4	5	1	nr	5	comp h1	5	2	3	nr	8
h3	4	5	1	nr	6	h1	5	2	3	nr	9
comp	4	5	2	nr	5	h2	5	2	3	nr	8
h1	4	5	2	nr	7	h2	5	2	3	nr	9
h1	4	5	2	nr	6	h3	5	2	3	nr	8
h2 h2	4	5 5	2	nr nr	6 7	h3	5 5	2 3	3	nr r	9 8
h2 h3	4	5	2	nr	6	comp	5	5	1	1	0
LU LU	, r	-	-		5		5	1	1	1	

		Movement		Condition				Movement		Condition	
h1	5 5	3	1	r	9 9	h2	5	6 6	2	r r	8
h1 h2	5	3	1	r	9	h3 h3	5	6	2	r	8
h2	5	3	1	r	9	comp	5	6	3	r	9
h3	5	3	1	r	8	h1	5	6	3	r	9
h3	5	3	1	r	8	h1	5	6	3	r	8
comp	5	3	2	r	7	h2	5	6	3	r	9
h1	5	3	2	r	9	h2	5	6	3	r	9
h1 h2	5	3	2	r	9 10	h3 h3	5	6 6	3	r r	8 9
h2	5	3	2	r	9	comp	6	1	1	r	10
h2	5	3	2	r	10	h1	6	1	1	r	7
h3	5	3	2	r	8	h1	6	1	1	r	6
comp	5	3	3	r	9	h2	6	1	1	r	7
h1	5	3	3	r	9	h2	6	1	1	r	6
h1	5 5	3	3	r	9 10	h3	6 6	1	1	r	76
h2 h2	5	3	3	r	9	h3	6	1	2	r r	9
h2 h3	5	3	3	r	10	comp h1	6	1	2	r	8
h3	5	3	3	r	10	h1	6	1	2	r	7
comp	5	4	1	r	9	h2	6	1	2	r	7
h1	5	4	1	r	9	h2	6	1	2	r	7
h1	5	4	1	r	9	h3	6	1	2	r	7
h2	5	4	1	r	8	h3	6	1	2	r	6
h2 h3	5 5	4	1	r	8 9	comp h1	6 6	1	3	r r	10 8
h3 h3	5	4 4	1	r	9 10	h1 h1	6	1	3	r	7
comp	5	4	2	r	8	h2	6	1	3	r	7
h1	5	4	2	r	8	h2	6	1	3	r	7
h1	5	4	2	r	9	h3	6	1	3	r	6
h2	5	4	2	r	9	h3	6	1	3	r	7
h2	5	4 4	2	r	9 8	comp	6 6	2 2	1	r	10 8
h3 h3	5	4	2	r	8	h1 h1	6	2	1	r r	8 6
comp	5	4	3	r	8	h2	6	2	1	r	7
h1	5	4	3	r	9	h2	6	2	1	r	6
h1	5	4	3	r	9	h3	6	2	1	r	7
h2	5	4	3	r	9	h3	6	2	1	r	7
h2	5	4	3	r	9	comp	6	2	2	r	10
h3	5 5	4	3	r	10 8	h1	6 6	2 2	2	r	76
h3 comp	5	5	1	r nr	6	h1 h2	6	2	2	r r	7
h1	5	5	1	nr	6	h2	6	2	2	r	7
h1	5	5	1	nr	5	h3	6	2	2	r	6
h2	5	5	1	nr	7	h3	6	2	2	r	7
h2	5	5	1	nr	5	comp	6	2	3	r	10
h3	5	5	1	nr	7	h1	6	2	3	r	8 7
h3	5 5	5	1 2	nr nr	6 5	h1 h2	6 6	2 2	3	r r	6
comp h1	5	5	2	nr	6	h2 h2	6	2	3	r	7
h1	5	5	2	nr	6	h3	6	2	3	r	7
h2	5	5	2	nr	8	h3	6	2	3	r	5
h2	5	5	2	nr	6	comp	6	3	1	nr	7
h3	5	5	2	nr	7	h1	6	3	1	nr	5
h3	5 5	5	2	nr	5	h1	6	3	1	nr	6
comp h1	5	5	3	nr nr	6 5	h2 h2	6 6	3	1	nr nr	6 6
h1	5	5	3	nr	6	h2 h3	6	3	1	nr	7
h2	5	5	3	nr	8	h3	6	3	1	nr	8
h2	5	5	3	nr	6	comp	6	3	2	nr	4
h3	5	5	3	nr	6	h1	6	3	2	nr	7
h3	5	5	3	nr	7	h1	6	3	2	nr	7
comp	5	6	1	r	9	h2	6	3	2	nr	7
h1	5 5	6 6	1	r r	7 7	h2 h3	6 6	3 3	2	nr nr	7 7
h1 h2	5	6	1	r	7	h3 h3	6	3	2	nr	7
h2	5	6	1	r	6	comp	6	3	3	nr	5
h3	5	6	1	r	7	h1	6	3	3	nr	8
h3	5	6	1	r	8	h1	6	3	3	nr	7
comp	5	6	2	r	8	h2	6	3	3	nr	7
h1	5	6	2	r	8	h2	6	3	3	nr	7
h1	5	6	2	r	8	h3	6	3	3	nr	8
h2	5	6	2	r	8	h3	6	3	3	nr	6

Rater	Subject	Movement	Trial	Condition	Score
comp	6	4	1	nr	8
h1	6	4	1	nr	7
h1	6	4	1	nr	8
h2	6	4	1	nr	7
h2	6	4	1	nr	6
h3	6	4	1	nr	7
h3	6	4	1	nr	6
comp	6	4	2	nr	7
h1	6	4	2	nr	7
h1	6	4	2	nr	7
h2	6	4	2	nr	8
h2	6	4	2	nr	7
h3	6	4	2	nr	8
h3	6	4	2	nr	7
comp	6	4	3	nr	7
h1	6	4	3	nr	8
h1	6	4	3	nr	7
h2	6	4	3	nr	7
h2	6	4	3	nr	8
h3	6	4	3	nr	8
h3	6	4	3	nr	6
comp	6	5	1	nr	5
h1	6	5	1	nr	4
h1	6	5	1	nr	4
h2	6	5	1	nr	5
h2	6	5	1	nr	4
h3	6	5	1	nr	3
h3	6	5	1	nr	4
comp	6	5	2	nr	5
h1	6	5	2	nr	5
h1	6	5	2	nr	5
h2	6	5	2	nr	6
h2	6	5	2	nr	5

Rater	Subject	Movement	Trial	Condition	Score
h3	6	5	2	nr	5
h3	6	5	2	nr	5
comp	6	5	3	nr	6
h1	6	5	3	nr	4
h1	6	5	3	nr	4
h2	6	5	3	nr	6
h2	6	5	3	nr	6
h3	6	5	3	nr	4
h3	6	5	3	nr	5
comp	6	6	1	r	8
h1	6	6	1	r	7
h1	6	6	1	r	6
h2	6	6	1	r	7
h2	6	6	1	r	6
h3	6	6	1	r	6
h3	6	6	1	r	6
comp	6	6	2	r	8
h1	6	6	2	r	6
h1	6	6	2	r	7
h2	6	6	2	r	8
h2	6	6	2	r	8
h3	6	6	2	r	7
h3	6	6	2	r	6
comp	6	6	3	r	7
h1	6	6	3	r	6
h1	6	6	3	r	6
h2	6	6	3	r	7
h2	6	6	3	r	8
h3	6	6	3	r	6
h3	6	6	3	r	9