Supplementary Material



Detecting actions of social fruit flies

Actions overview

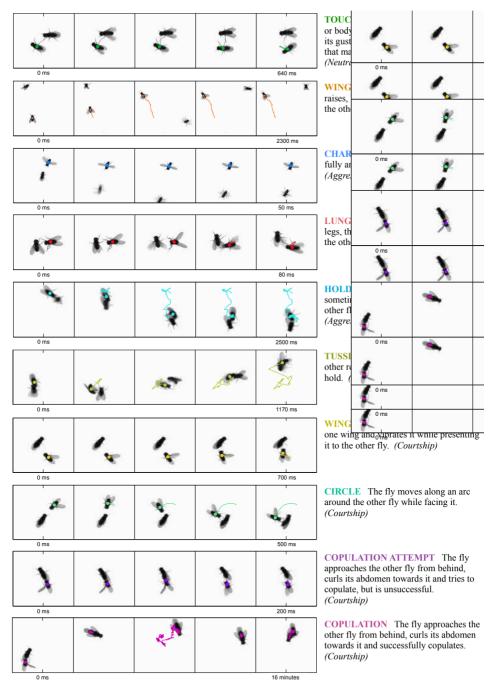


Figure 1: Action examples and descriptions.

Experimental setup

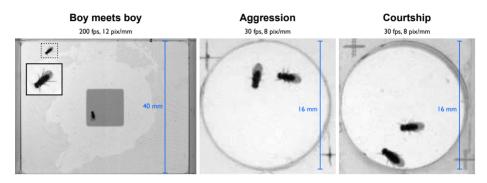


Figure 2: Boy meets boy has high temporal and spatial resolution videos and a large chamber with a food patch, Aggression and Courtship have lower resolution and a much smaller chamber with uniform food surface. The Courtship experiments contain one male and one female fly, the others have two male flies.

Bout vs. frame wise performance

A predicted output can have high precision-recall when measured on a per frame basis, but low when measured on a per bout basis, and vise versa. Figure 3 shows examples where this is the case:

- i) Under/over segmented bouts \rightarrow lower bout than frame wise performance
- ii) Short missed/false detections \rightarrow lower bout than frame wise performance
- iii) Under/over estimated duration \rightarrow lower frame than bout wise performance
- iv) Offset bout boundaries \rightarrow lower frame than bout wise performance





Figure 3: Examples of frame-bout performance discrepancies. Red squares denote missed/false detections, depending on whether A or B is ground truth.

Human vs. human

We trained novice annotators to learn to detect actions in the Fly-vs-Fly dataset, by showing them a subset of annotated movies, having them annotate another subset and providing them with feedback such that they could adjust their detection criteria. Once trained, they re-annotated a large portion of the test data, enough to give an idea about the difficulty of detecting each action. Overall, the trained annotators achieved best performance on the Courtship sub-dataset, which they described as being easier to annotate than the other two sub-datasets, with actions seemingly less ambiguous. Figure 4 shows the bout- and frame wise precision-recall for each action in Fly-vs-Fly, and Figure 5 explains bout-frame performance discrepancies. The human performance is a good indicator for what to expect from automatic detection algorithms; we do not expect perfection, due to action ambiguity and imperfections in ground truth annotations, but ideally they should achieve at least as good performance as humans.

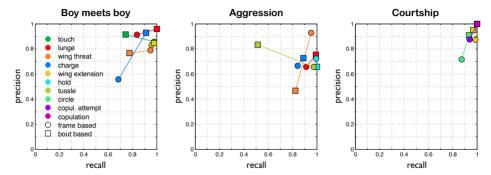


Figure 4: Human performance measured in terms of frame based (circles) and bout based (squares) precision-recall.

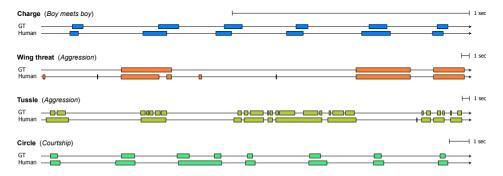


Figure 5: Segmentation samples of actions with high frame-bout performance discrepancy. Here GT refers to experts and Human to trained annotators.

Method performance per action

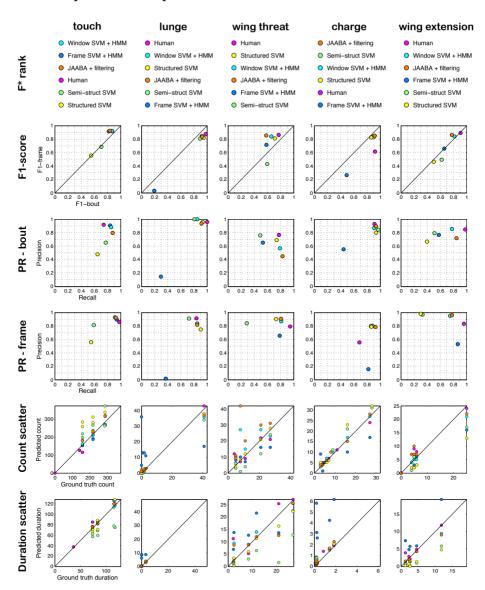


Figure 6: Results: Boy meets boy

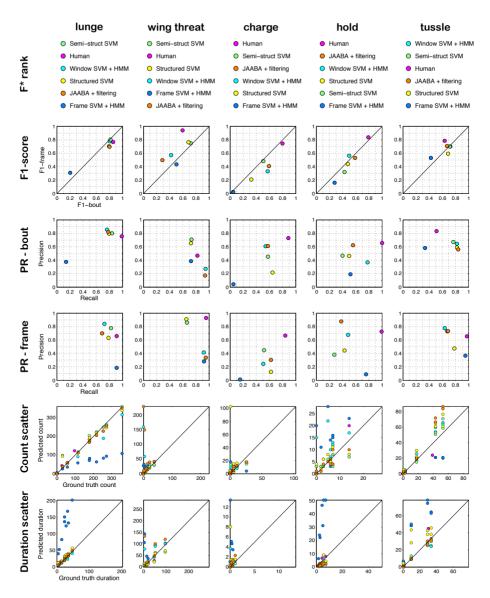


Figure 7: Results: Aggression

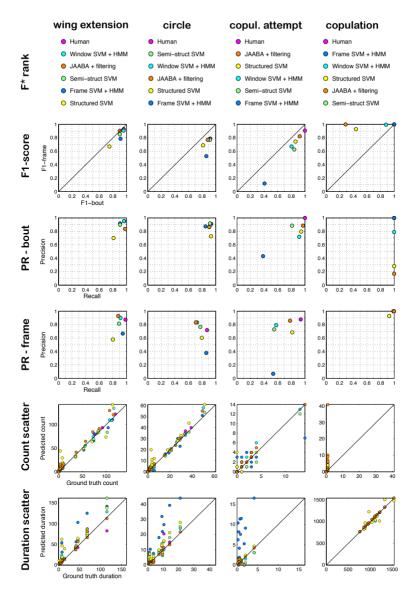


Figure 8: Results: Courtship