



Introduction

- ▶ **Interactive robots benefit from perspective taking mechanisms**, allowing to infer what a user can see and to more accurately predict their behavior [1]
- ▶ For **self-referential understanding of actions** of others, assuming their visual perspective is essential

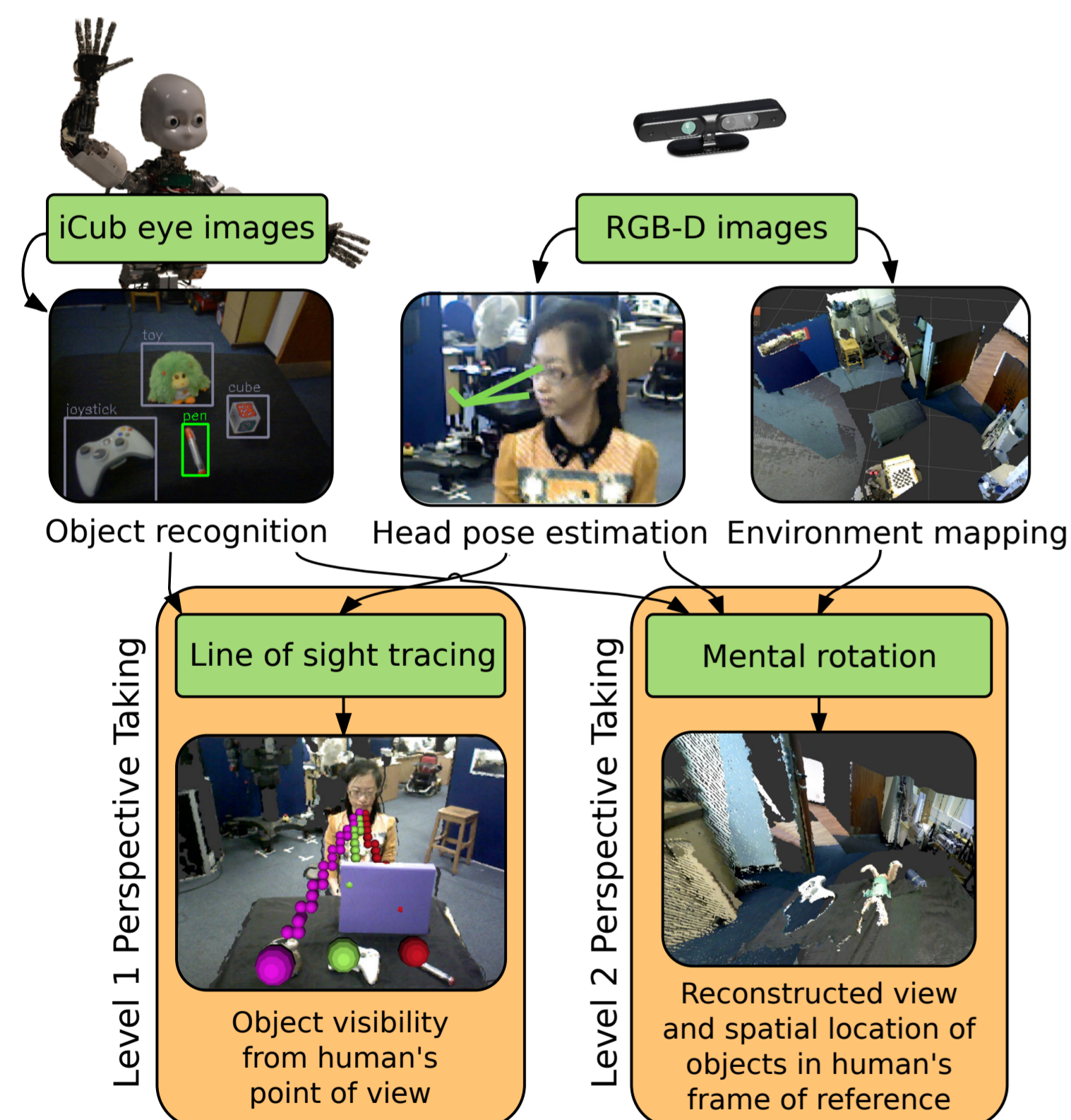


Figure 1: Overall flow of the proposed method

- ▶ Previous works [1, 2] are constrained to environments equipped with markers and/or motion capture systems
- ▶ Our goal: **estimate perceived world of humans** and the surroundings of the robot, **whilst not constraining the environment**

Markerless Perception

- ▶ **Self-exploration phase**: build a map of the environment using RGB-D images and visual SLAM
 - ▷ Reason about parts of the environment which are currently out of sight, but visited previously
- ▶ **Object recognition using deep learning** on iCub eye images [3]

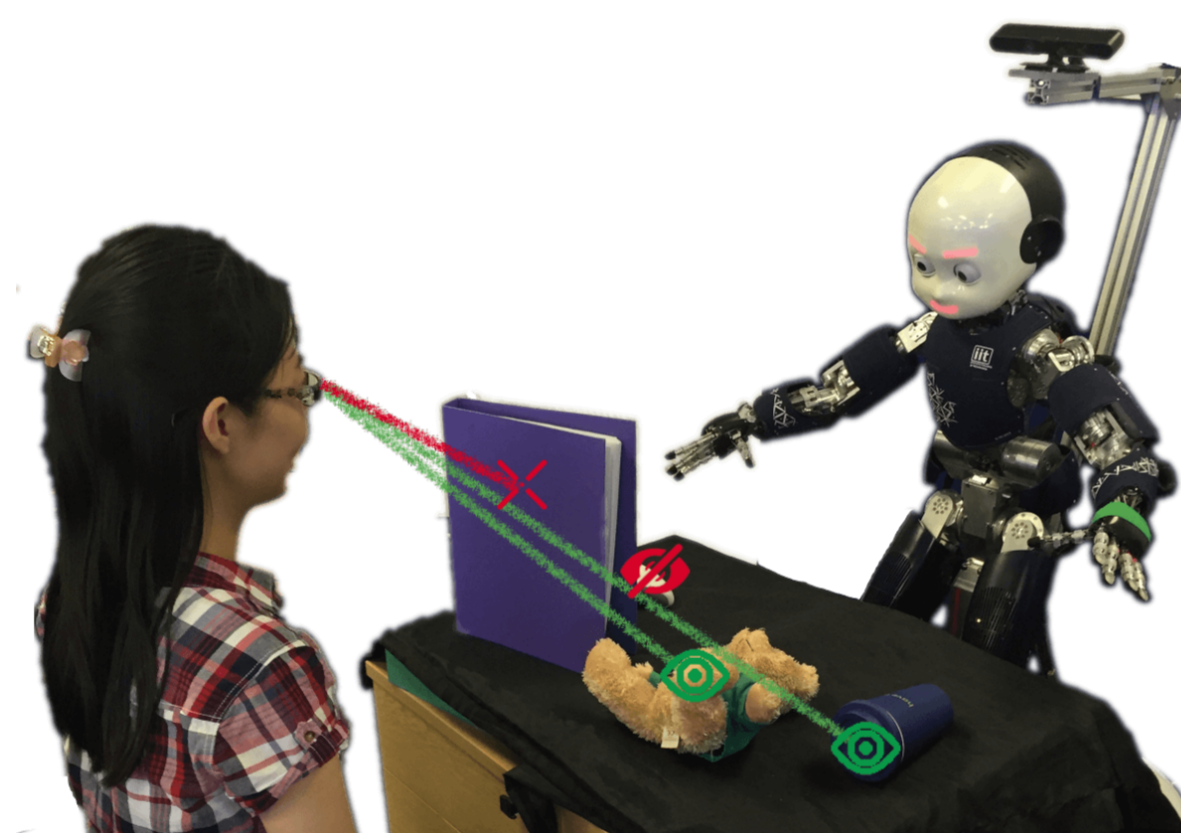


Figure 2: Typical setup

- ▶ To reason about the human's field of view, we **improve performance** of a state of the art **head pose estimator** [4]
 - ▷ We propose new method which normalizes the input so it becomes more similar to the training data

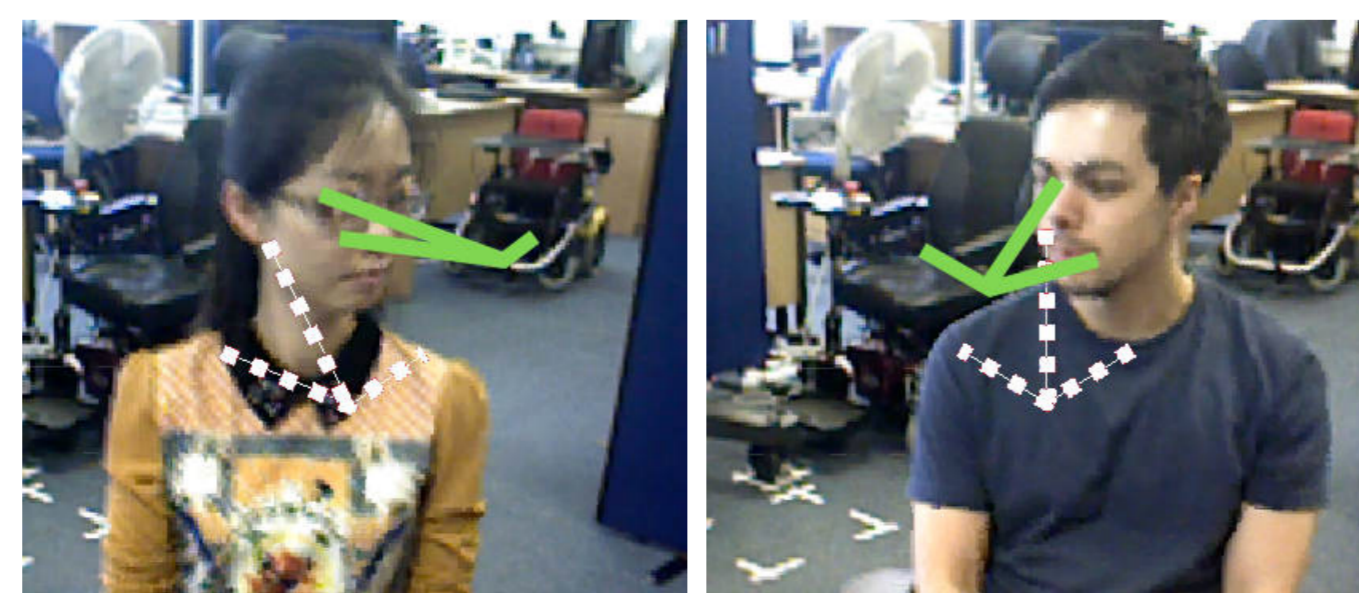


Figure 3: Comparison of the normalized head pose algorithm (green) and original method (white, dotted)

Perspective Taking

- ▶ Following the literature [5], we distinguish level 1 and level 2 perspective taking
- ▶ **Level 1 perspective taking**:
 - ▷ find whether human can see an object
 - ▷ we propose using **line of sight tracing**

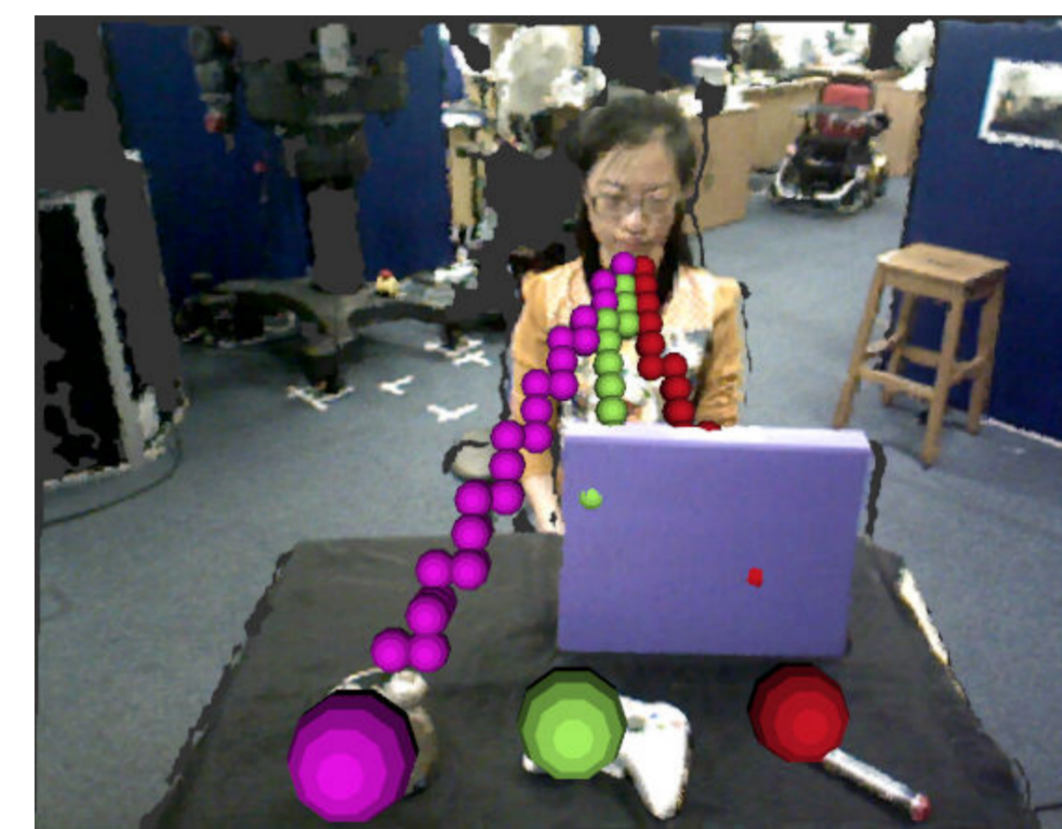


Figure 4: Line of sight tracing

- ▶ **Level 2 perspective taking**:
 - ▷ estimate what world looks like to human
 - ▷ **mentally rotate** environment map in the frame of reference of the human
 - ▷ **left-right judgments** using the same spatial reasoning algorithm as from the robots perspective

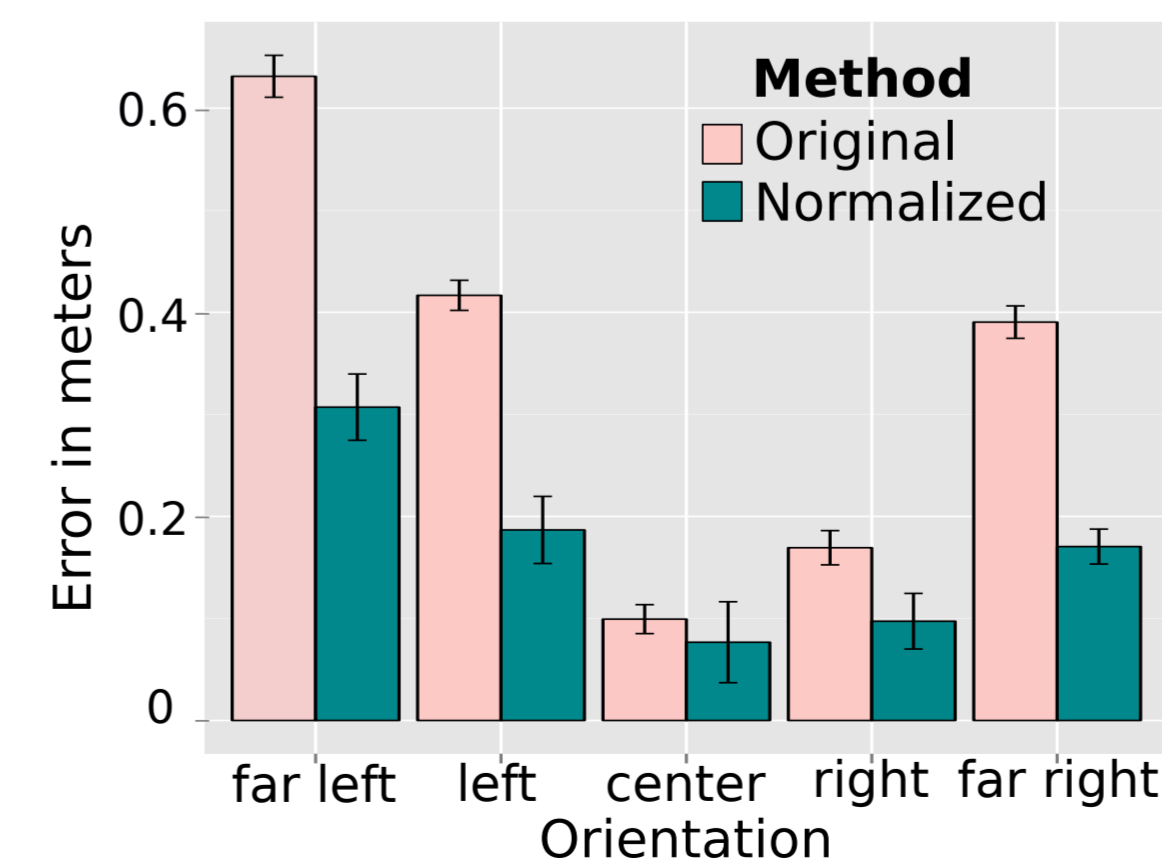


Figure 5: Horizontal error for spatial perspective taking using original / proposed head pose estimation

Conclusions

- ▶ Improvements in key parts of the perspective taking pipeline allow system to work in markerless setups
- ▶ Validation in several experiments (head pose estimation + line of sight + mental transformation) using iCub
- ▶ Extend application scenarios of popular head pose estimation algorithm

Future work

- ▶ More accurate gaze estimates: take human's eye movements into account
- ▶ Investigate developmental process of perspective taking and relationship to joint attention

Acknowledgements

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References

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