

EMYS: A SOCIAL ROBOT THAT
PLAYS “SUECA”

OUTLINE

1. Motivation
2. Goals
3. Background
4. Related Work
 - a. AI in Games
 - b. HRI
5. Proposed Architecture
6. Evaluation
7. Conclusion

1. MOTIVATION

MOTIVATION



MOTIVATION



MOTIVATION



MOTIVATION



2. GOALS

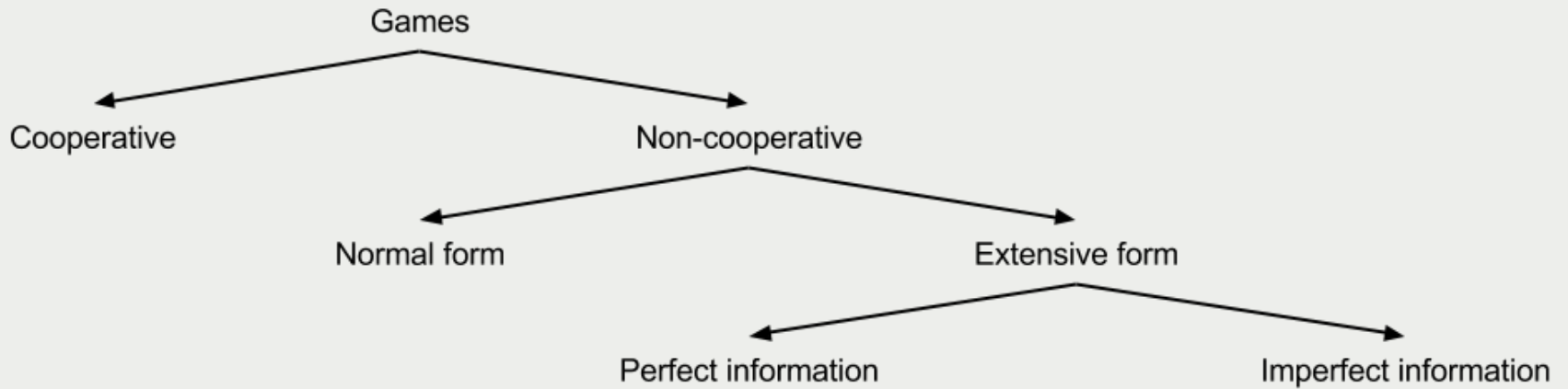
GOALS

*Integrate a social robot with aged humans
in a card game scenario*

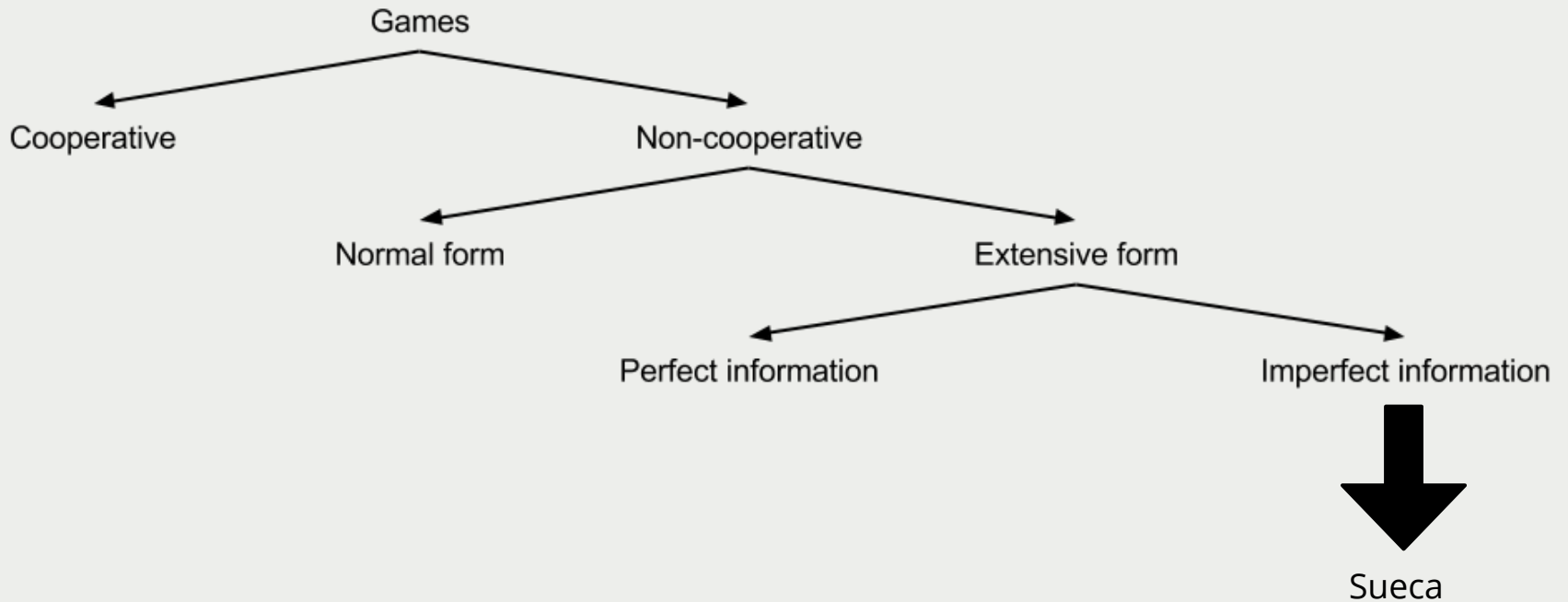
- Develop an agent that plays competently *Sueca*
- Develop a socially present embodied agent
- Evaluate the correctness of the system

3. BACKGROUND

BACKGROUND

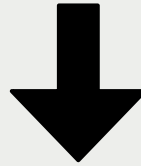


BACKGROUND



BACKGROUND

Hidden information?



Information Set!

BACKGROUND

Monte-Carlo Tree Search

1. Selection
2. Expansion
3. Simulation
4. Backpropagation

4. RELATED WORK

4.1. AI IN GAMES

AI IN GAMES

Solving hidden information games...

- Monte-Carlo Methods
- Nash-Equilibrium Strategy
- Belief distributions

AI IN GAMES

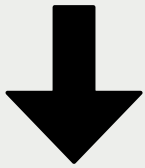
Solving hidden information games...

- Monte-Carlo Methods
- ~~Nash Equilibrium Strategy~~
- Belief distributions

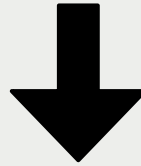
AI IN GAMES

MONTE-CARLO METHODS

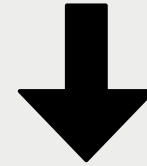
How to deal with hidden information?



PIMC



ISMCTS



IIMC

AI IN GAMES

MONTE-CARLO METHODS

PIMC

Domains	Pros / Cons	Hidden Information
Bridge Skat	<ul style="list-style-type: none">● Simpler to implement● Strategy fusion● Non-locality	Determinization

AI IN GAMES

MONTE-CARLO METHODS

ISMCTS

Domains	Pros / Cons	Hidden Information
Dou Dizhu	<ul style="list-style-type: none">● Computational Budget● Strategy fusion (less)● Non-locality● Harder to implement	Information Set

AI IN GAMES

MONTE-CARLO METHODS

IIMC

Domains	Pros / Cons	Hidden Information
Skat	<ul style="list-style-type: none">● Player Module● Strategy fusion (less)● Non-locality● Harder to implement	Recursive Determ.

AI IN GAMES

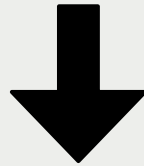
BELIEF DISTRIBUTIONS

Domain	Technique	Goal	Suitable
Skat	Determine the winning probability of a hand	Improve the bidding	N
Skat	Fastest-cut-first heuristic	Order moves	Y
Skat	Considering similar states equally	Reduce tree exploration	Y
Skat	Calculate the mistake rate of each player	Improve the bidding	~
Poker	Opponent model	Improve MCTS policies	Y

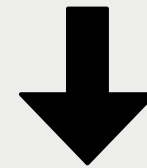
4.2. HRI

HRI

*Integrate a social robot with aged humans
in a card game scenario*



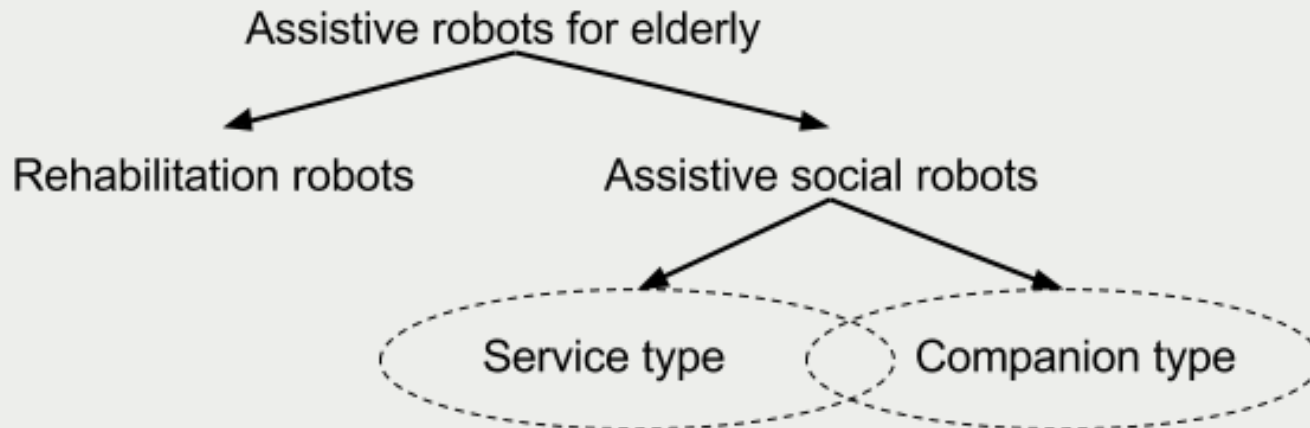
Robots in elderly care



Social robots in games

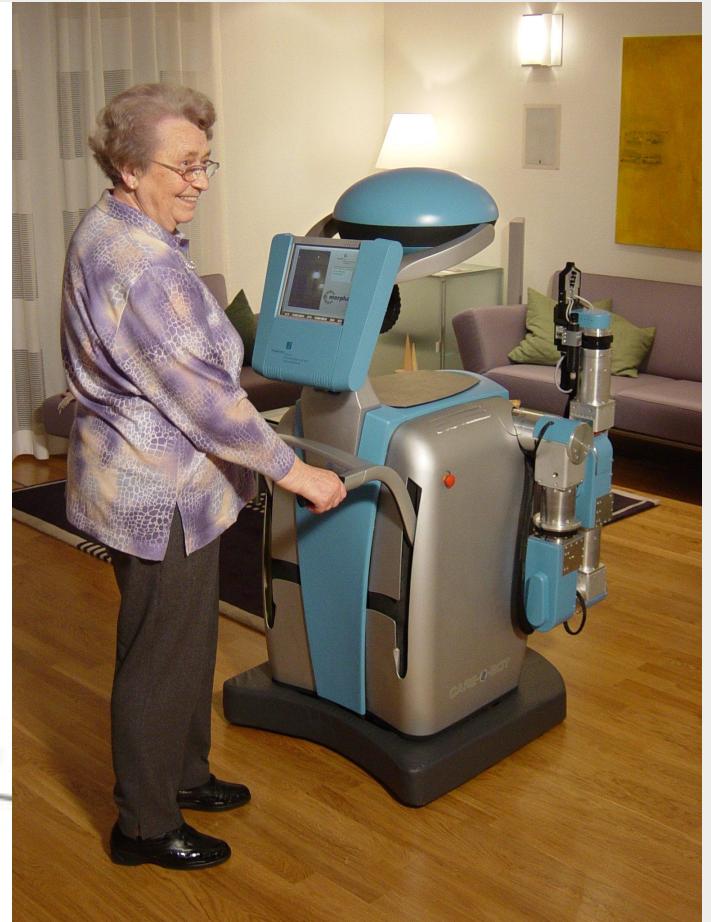
HRI

ROBOTS IN ELDERLY CARE



HRI

ROBOTS IN ELDERLY CARE



HRI

ROBOTS IN ELDERLY CARE



HRI

SOCIAL ROBOTS IN GAMES



- Children tutor
- Careful advices
- Long-term interactions

HRI

SOCIAL ROBOTS IN GAMES

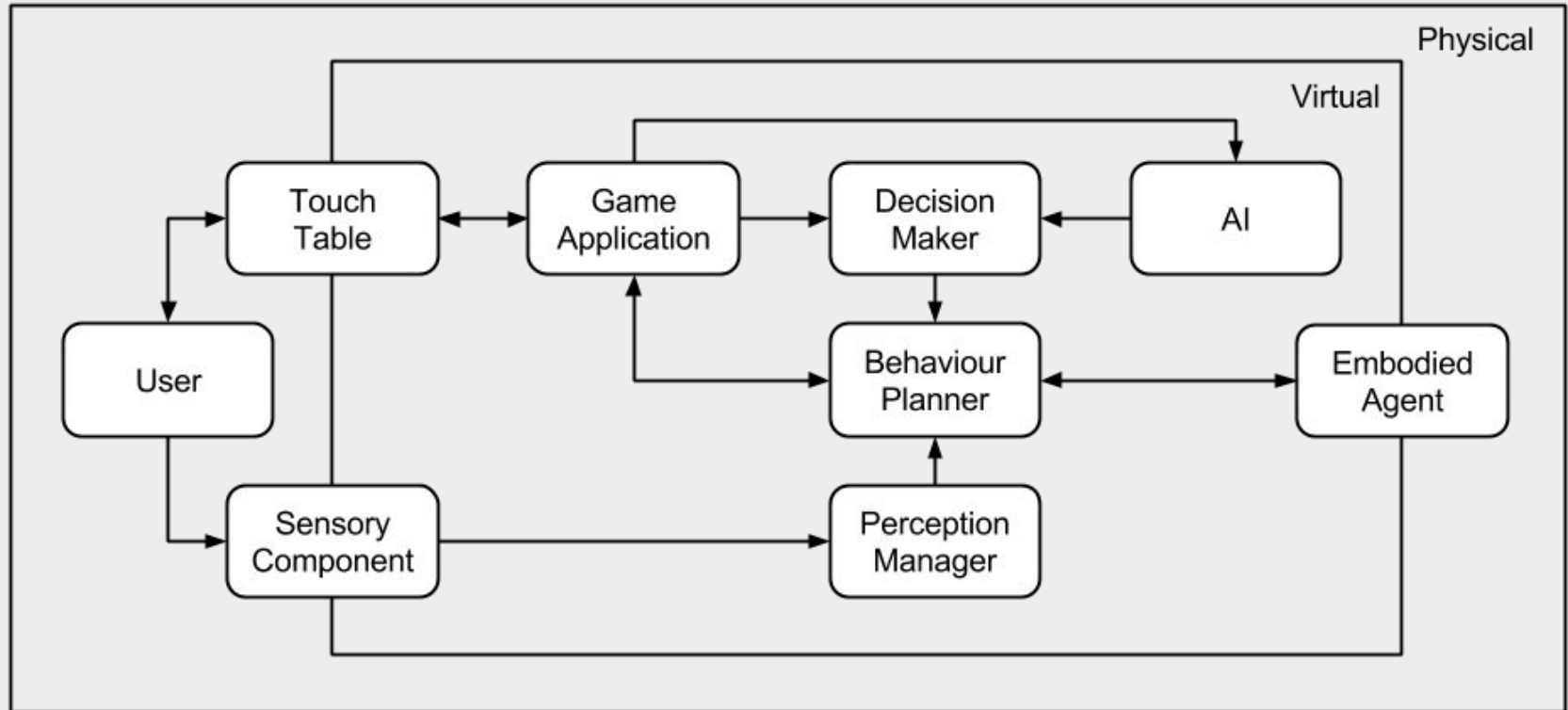
Improved social presence:

- Topology of speeches
- Relevance value of a move
- Power of a player
- Simulation of roles
- Luck perception



5. PROPOSED ARCHITECTURE

PROPOSED ARCHITECTURE



PROPOSED ARCHITECTURE

AI MODULE

- PIMC
- Opponent model
 - Cards' predictions
 - Actions' predictions



How to collect data?

- Ask for it
- Collect it! (it requires a platform)

6. EVALUATION

EVALUATION

Develop an agent that plays competently Sueca

Performance measures

- Game points
 - Offline pre-computation time
-
- These measures will be compared to different parametrizations and a naive approach
 - University community will test it

EVALUATION

Develop a socially present embodied agent

Two conditions

- Few or nonexistent social behaviours
 - Several behaviours from the game state
-
- The elderly will test it
 - Godspeed - participants' perception of the robot
 - Networked Minds - presence perception

THANK YOU!