

EVALUATING THE INFLUENCE OF AUTOMATIC ATTENTIONAL MECHANISMS IN HUMAN-ROBOT INTERACTION Pablo Lanillos, João Filipe Ferreira, Jorge Dias

EVALUATING HUMAN ROBOT INTERACTION (HRI)

The hypothesis is that using automatic attentional mechanisms as a basic block for posteriorly build a robotic cognitive intelligence is the correct approach and, in order to experimentally prove it, we propose a systematic evaluation method.



Turing test for social skills





Social Compliance Score (SCS) as an unique value to evaluate the robot behavior. X is the SCS of using attentional process and Y is the SCS of using a tailor -made solution, then we state:

 $H0: X > Y \rightarrow attention mechanisms better than tailor - made for HRI$

OBSERVERS



ROBOT CLASS DEPENDING ON ITS CAPABILITIES

- robots capable of gaze-shifts and fixation (e.g. active head) Α
- robots with speech capabilities. B1 speech recognition and B2 able to talk В
- robots with arms and pointing or showing capabilities С
- robots with emotional capabilities D







CRITERIA

Evaluation bridge conventional objective measures from robotics with subjective criteria from human studies in psychology and neuroscience.

items	ID	Robot C
Qualitative Criteria in Reciprocal Social Interaction		
Engaging and Joint Attention		
Interlocutor detection error	RSI1	A
Robot engaging delay after human IJA	RSI2	A
Disengaging delay after human pointing or gazing	RSI3	Α
Object search latency	RSI4	Α
Acknowledging joint attention period or cycle	RSI5	Α
#Engages	RSI6	Α
#Disengages	RSI7	Α
#Robot initiating joint attention (IJA)	RSI8	Α
#Pointing events	RSI9	С
#Showing events	RSI10	С
Looking at people frequency	RSI11	Α
Object detection success	RSI12	Α
Joint attention error	RSI13	Α
Turn taking coherence (IJA switching)	RSI15	Α
Social-emotional reciprocity		1
Expressions towards interlocutors rate	RSI16	D
Frequency of contact (touching)	RSI17	C / CD
Restricted, Repetitive Behaviors and Interests		, ,
Contingent stimulus attention error	RB1	A / B2 /
Time to attend novel contingent multimodal stimulus	RB2	A / B /
Head movements per second	RB3	A
Social delay error	RB4	A / B /
Qualitative impairments Communication and Language	ge	
Utterances rate per interlocutor	CL1	B1
Name hearing error	CL2	B1
Frequency of vocalization directed to others	CL3	B2
HRI performance metrics		'
Reciprocal Social Interaction	Ro	-
Behaviors and Interests	Bo	-
Communication and Language	Co	-

Subjective

items	ID	Robot Class		
Qualitative Criteria in Reciprocal Social Interaction				
Engaging and Joint Attention				
Unusual eye contact	JA1	А		
Low frequency of looking at people	JA2	A		
Spontaneous initiation of joint attention	JA3	A / B2		
Pointing	JA4	С		
Showing	JA5	С		
Looking at object	JA6	A		
Acknowledging joint attention	JA7	A / B2 / C		
Social reward				
Shared enjoyment	SR1	D		
Social-emotional reciprocity				
Facial expressions directed to others	ER1	D		
Quality of social overtures	ER2	D		
Use of other's body to communicate	ER3	C / CD		
Restricted, Repetitive Behaviors and Interests				
Repetitive interests and behaviors	RB1	A / B / C / D		
Unusual sensory behaviors or interest	RB2	A / B / C / D		
Overactivity	RB3	A / B / C / D		
Negative behavior and distraction	RB4	A/B/C/D		
Slow behavior	RB5	A / B / C / D		
Qualitative impairments Communication and Language				
Conventional gestures	CL1	A / C		
Response to name	CL2	B1		
Frequency of vocalization directed to others	CL3	B2		
Immediate echoing	CL4	B2		
Speech abnormalities	CL5	B2		
HRI performance metrics				
Reciprocal Social Interaction	R_s	-		
Behaviors and Interests	B_s	-		
Communication and Language	C_s	-		

SOCIAL COMPLIANCE SCORE

Objective (O) = $\{R_o, B_o, C_o\}$

Subjective (S) = $\{R_S, B_S, C_S\}$

One-Class classification

[Tax 2001, Chandola 2009]

Anomaly detection

 $SCS = P(human | [O S], \Theta)$

Reciprocal social interaction Behaviors and Interests Communication and language



normal behavior unusual behavior

parameterized behavior







EXPECTED BEHAVIOR



Subjective scores should be correlated with the objective measures.

The expected behavior can:

- Be learned from human data
- Be used as an error function criteria by comparison with the real robot behavior

REFERENCES

Tax, D. M. J. (2001), 'One-class classification: Concept learning in the absence of counter-examples', PhD thesis, Technische Universiteit Delft.

Chandola, V., Banerjee, A., Kumar, V. (2009). 'Anomaly detection: A survey'. ACM Comput. Surv.





FCOMP-01-0124-FEDER-028914 PTDC/EEI-AUT/3010/2012

