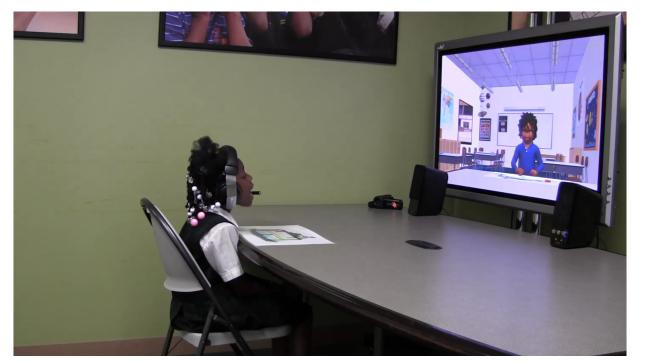
Designing Bots, Virtual Humans, & Other Systems that Hold up their End of the Conversation

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Carnegie Mellon University & ISIR / La Sorbonne Chaire Blaise Pascale & Chaire Sorbonne Mai 2018

The ineffable quality of Rapport in learning



Children who report more rapport are more likely to learn from the virtual peer

Search Engine vs. Conversation

Justine: "OK Google, I love Manchester United"

Google: Manchester United Football Club is a professional football club

Based in Old Trafford, Greater Manchester, England, that competes

in the Premier League, the top flight of English Football

Justine: "I love Manchester United"

Friend.: "No way! Arsenal wipes the floor with those Red Devils!"

Socially-Aware Robot Asst: "No way! Arsenal wipes the floor with those Red Devils!"

Motivation for Socially-Aware Bots

- 1. People pursue *multiple conversational goals* in every conversation & expect the same from their interlocutors. To put people at ease, and increase relationship strength, we must understand the *propositional*, *interactional* & *interpersonal* functions of conversation.
- 2. People change interaction styles over time. We must increasingly *manage long-term interactions* with people by changing interaction style in a way that evokes increasing loyalty, rapport and trust.

Rapport improves task performance

Surveys

- Survey respondents gave higher quality answers if they felt rapport with interviewer (Berg (1989)

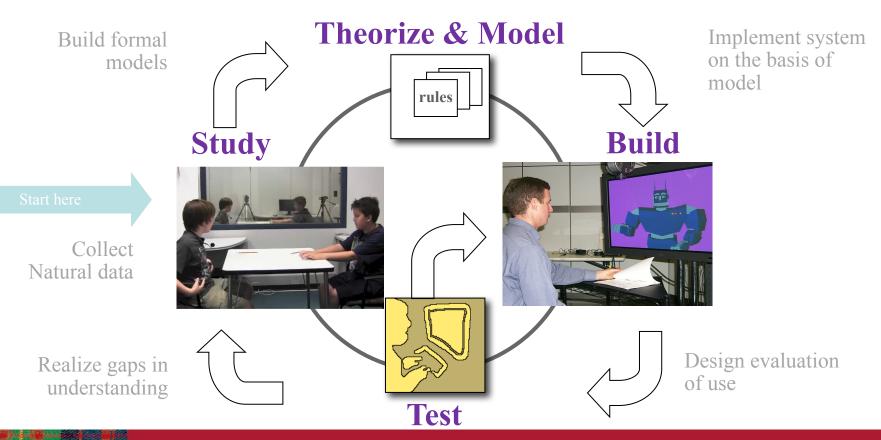
Health

– Physicians who build rapport during trial interviews enroll more participants (Albrecht *et al.*, 1999).

Sales

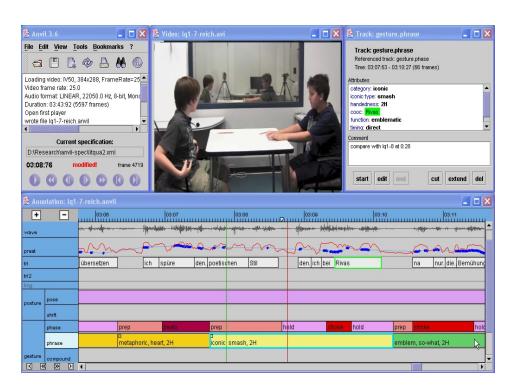
- Rapport with sales staff leads to increased likelihood of purchasing goods/service (Brooks, 1989).
- Customers show increased trust and disclosure when rapport is maintained with sales staff (LaBahn, 1996).

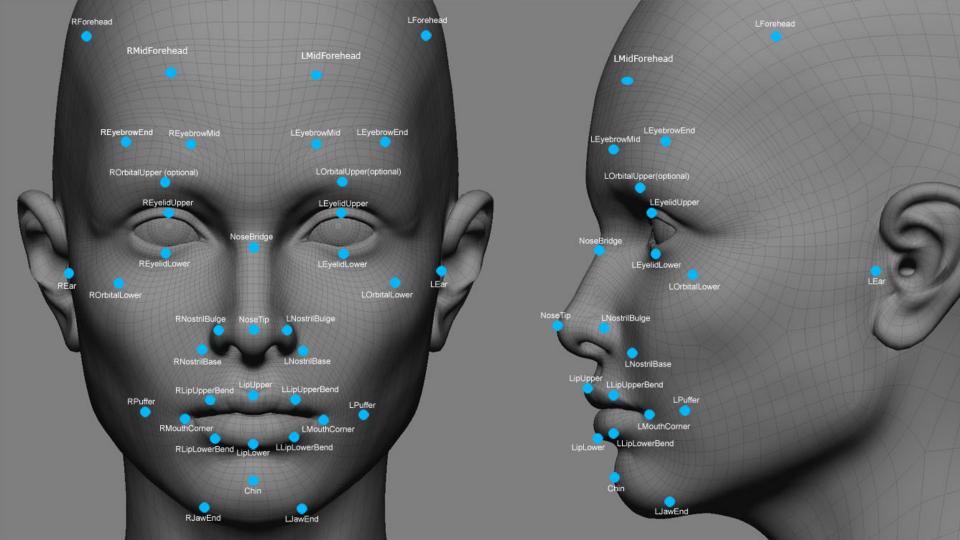
Methodology





Analyze









Analysis of Rapport

Thin-slice (Ambady & Rosenthal, 1992) judgments of every 30 second video segment) presented to 3 annotators in randomized order

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- "Consensus" measured by Intra-Class Correlation (single measure): 0.37
- "Consistency" measured by Cronbach alpha: 0.68
- Inverse-based bias correction (Kruger et al., 2014) was used mitigate rater bias & pick single rapport rating for each 30 second video segment.

IRR



Carnegie Mellon University

Evaluate

Total Interactions:

250+ sessions

Current Dataset:

120 sessions (totally 10+ hours)

- Mean duration = approx. 5 min.
- SD = 69.00 seconds



SARA, 00:13.7, I'm so glad to have the opportunity to meet you in person, 00:16.7

User, *00:18.7*, hi how are you [doing, *00:19.2*]

SARA, 00:19.3, [I'm] in Davos to assist attendees but to be honest I am a little anxious about it it's a big deal to be invited to Davos is it your first time here, 00:28.7

User, 00:29.1, it is and I'm a little anxious too, 00:31.4

SARA, 00:32.7, I found there's a lot to take in I hope that it isn't overwhelming you, 00:37.2

User, 00:37.9, it is overwhelming a little bit, 00:39.8

Summary of Evaluation Metrics

Interaction

- Total Time
- System Speaking Time
- User Speaking Time
- System's Response Time
- User's Response Time
- System's Interruptions
- User's Interruptions

Verbal

- System's Intentions
- User's Intentions
- System's Conversational Strategies
- User's Conversational Strategies

Task

- Session Rec. Acceptance
- Person Rec. Acceptance

Interpersonal

- Rapport Score
- Mutual Attentiveness
- Coordination
- Positivity

Goal of Socially Aware Systems

Development of a bot that manages *interpersonal rapport* (relationship strength) with users over interactions across time, as well as managing propositional and interactional goals, *in order to improve task performance*.

Automatically recognize rapport-managing conversational strategies from verbal, visual and vocal modalities of speaker and interlocutor, both within the individual and in the dyad.

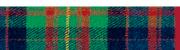
Goal of Building Socially Aware Systems

<u>Theoretical</u>: Understand the nature of rapport in greater detail, by correlating with associated observable verbal (conversational strategies, vocal (voice quality) and visual (non-verbal) cues

<u>Methodologica</u>l: Leverage this understanding to automatically recognize rapport-building strategies by leveraging and developing statistical machine learning techniques

Ineffective Conversation (don't do this with agents)



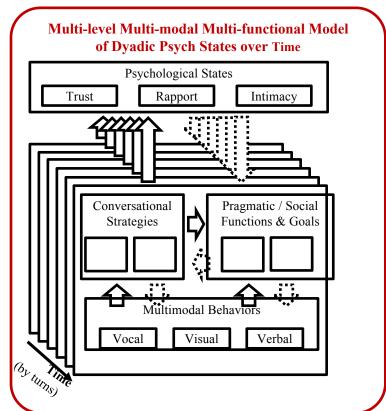


Intimate Conversation (don't do this with agents)



Agent Model of Rapport must be:

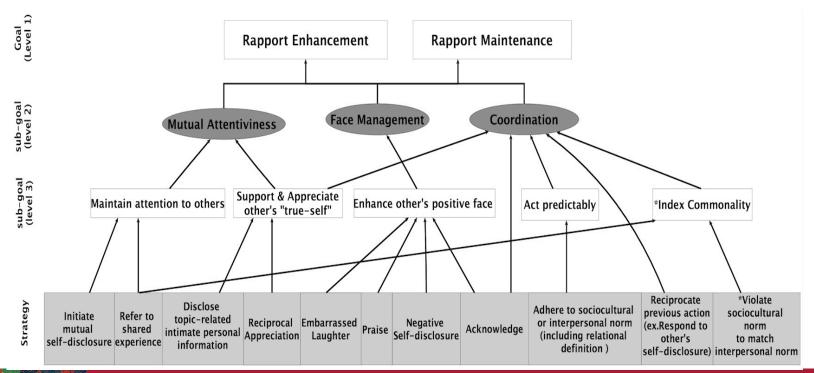
- 1. Dyadic,
- 2. Multi-level:
 differentiate between
 observable signals &
 underlying
 psychological states,
- 3. Sensitive to effect of *time*
- 4. Cross-Modal



with L.P. Morency, 2015

With Ran Zhao

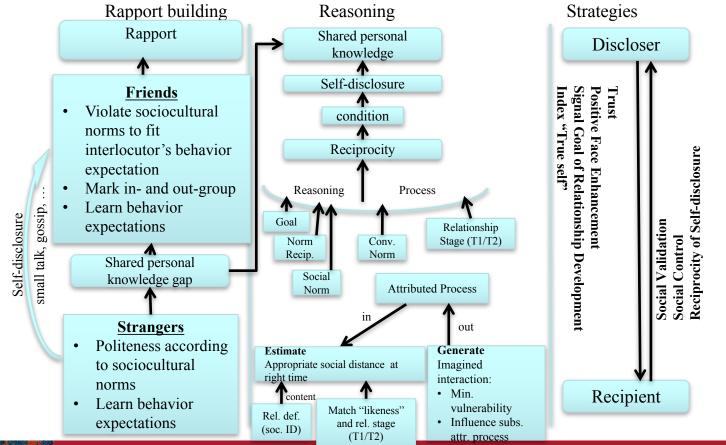
Data- &Theory-Driven Model of Rapport Management



Conversational Strategies

Conversational Strategies	Examples
VSN (Violation of Social Norm)	"man you take forever to write"
SD (Self Disclosure)	"I hate math"
PR (Praise)	"well done"
SE (Reference to Shared Experience)	"I shared m&m's with you last time"
BC (Back Channel)	"yup"
QE (Question Eliciting SD)	"are you an atheist"

Data- & Theory-Driven process model

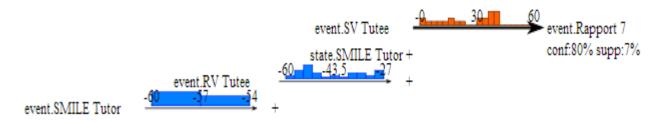


Data-Driven: Temporal association rules

Form of temporal rules

"If event A happens at time t, there is 50% chance of event B happening between time t+3 to t+5"

Temporal association rules: Friends



Example: Friend in high rapport

Tutor: Sweeney you can't do that, that's the whole point {smile} [Violation of Social Norm]

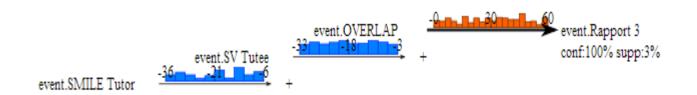
Tutee: I hate you. I'll probably never never do that [Reciprocate Social Norm Violation]

Tutor: Sweeney that's why I'm tutoring you {smile}

Tutee: You're so oh my gosh {smile}. We never did that ever [Violation of Social Norm]

Tutor: {*smile*} What'd you say?

Temporal association rules: Strangers



Example: Stranger in low rapport

Tutee: divide oh this is so hard let me guess: 11[Negative Self-Disclosure]

Tutor: you know

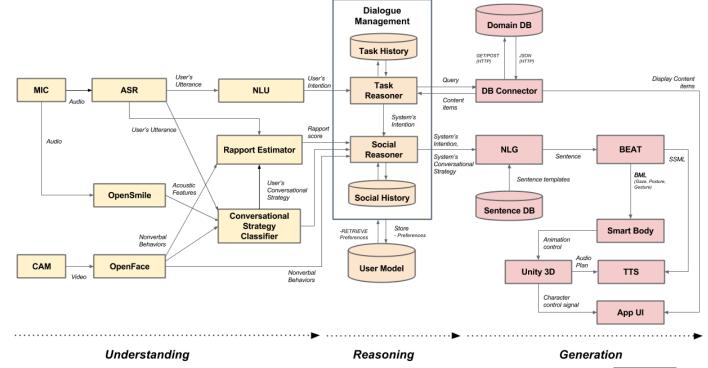
Tutee: 6

Tutor: next problem is is exactly the same {smile}: over 11 equals, 11 x over 11

Tutee: I don't need your help; [Violation of Social Norm]

Tutor: {Overlap} That is seriously like exactly the same.

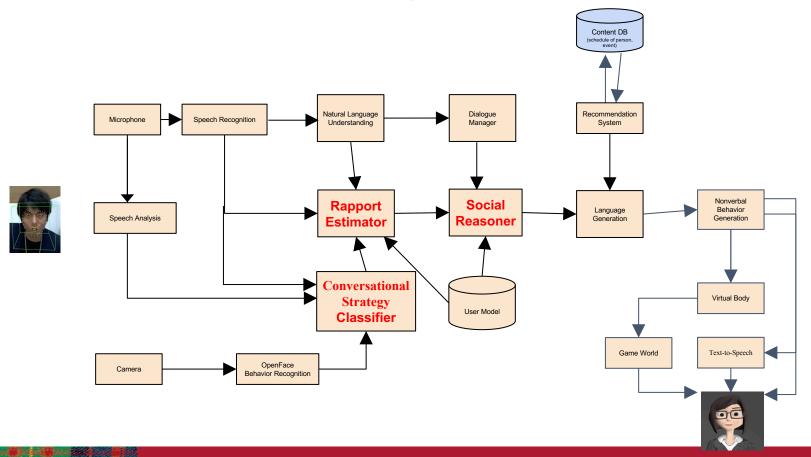
Socially-Aware Agent Architecture



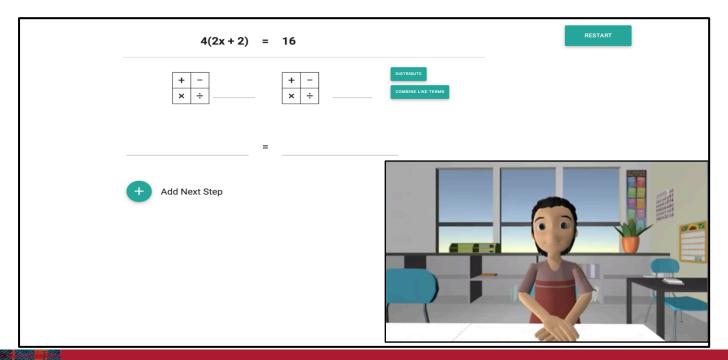


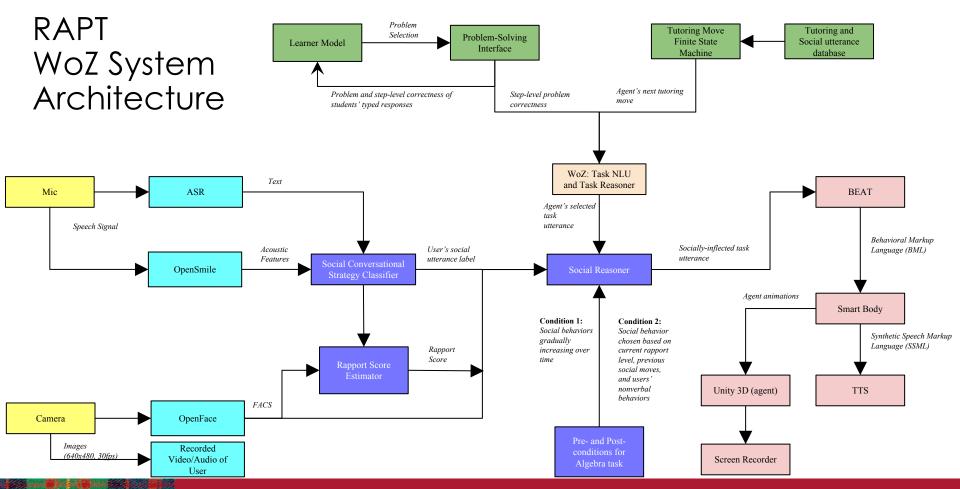


Socially-Aware Agent Architecture



Some Applications: Rapport-Aware Peer Tutor (RAPT)





Evaluation: rule-based vs. adaptive

Control condition: (fixed heuristics for social dialogue usage)

Praise

Decreasing in frequency [Kumar et al., 2010]

Self-Disclosure

Gradually increasing in frequency and intimacy [Ogan, 2011; Bickmore and Schulman, 2010]

Questions eliciting self-disclosure

Gradually increasing in breadth and depth of topics [Altman and Taylor, 1973]

References to shared experiences

Gradually increasing frequency [Kumar et al., 2010; Cassell and Bickmore, 2003]

Violation of social norms

Use only after a given threshold in number of turns or elapsed time [Ogan et al., 2012]

Indirectness

Decreasing in frequency [Madaio et al., 2017]

Evaluation: rule-based vs. adaptive

Experimental condition: (adaptive usage of social dialogue)

Based on:

Current rapport state

Changes in rapport state (increasing, decreasing, maintaining)

User's social behavior (self-disclosure, violation of social norms, etc)

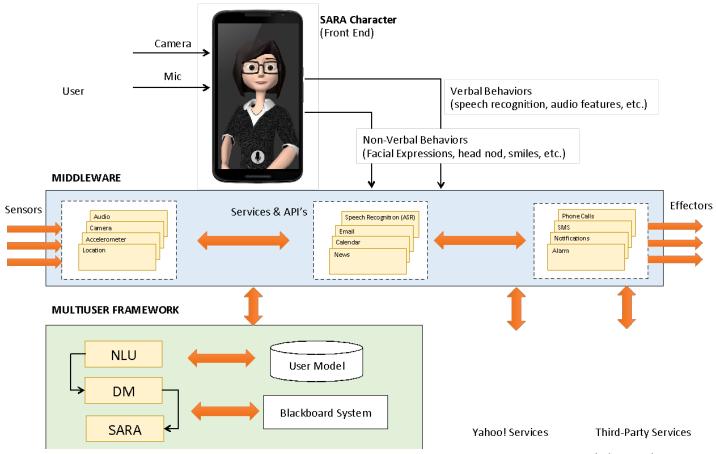
Agent's previous social behavior

User's nonverbal behavior (smiling, nodding, gaze patterns)

Agent's previous tutoring behaviors (feedback, questions, explanations, etc)

Application. Mobile from end to apps

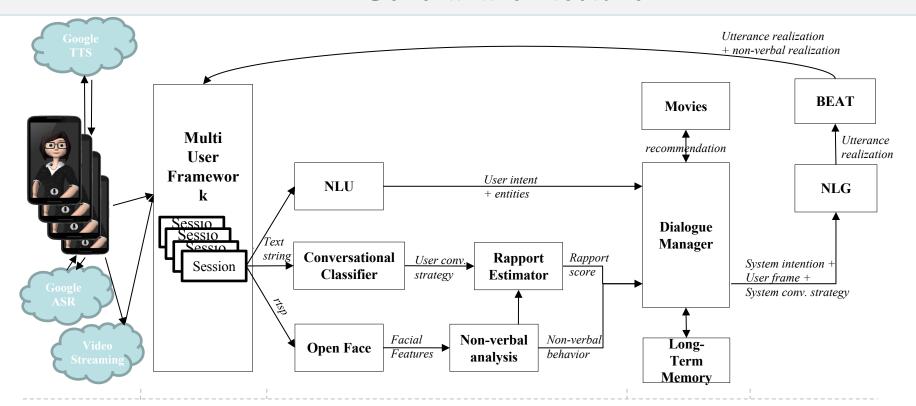




with Yahoo InMind Team

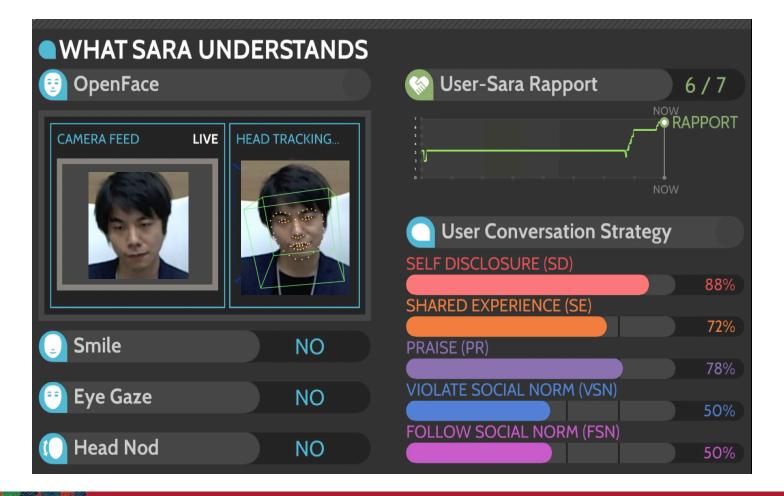
Integrated InMind Dialog Architecture

General architecture

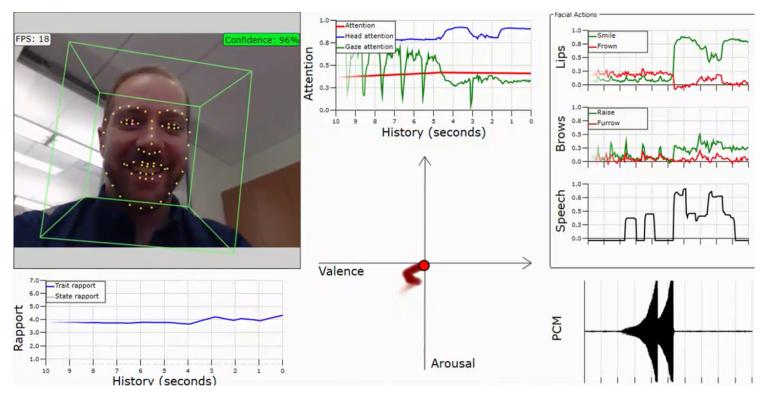


SARA: Socially Aware Robot Assistant

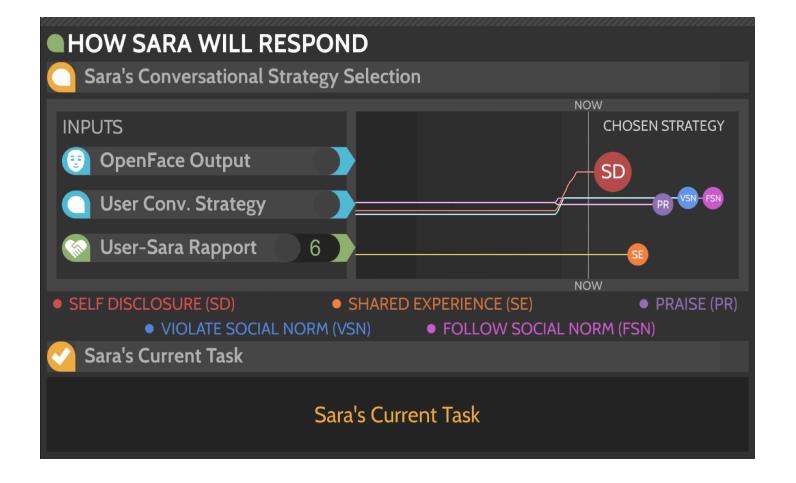


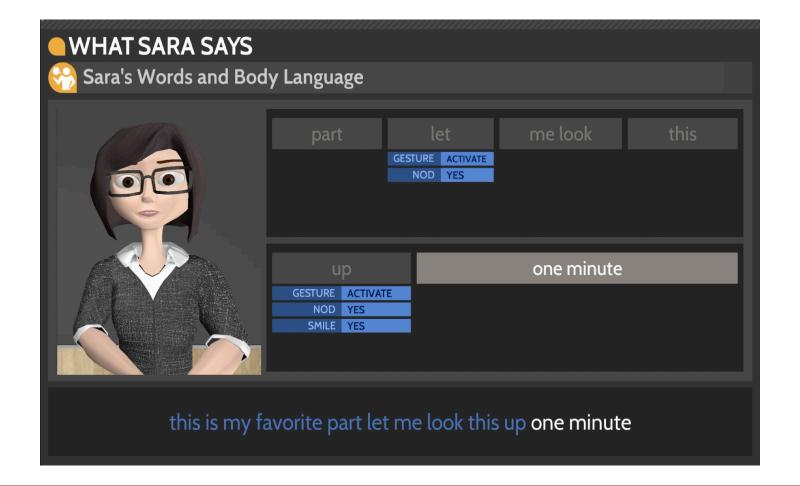


Tracking Facial Movements



OpenFace: L.P. Morency



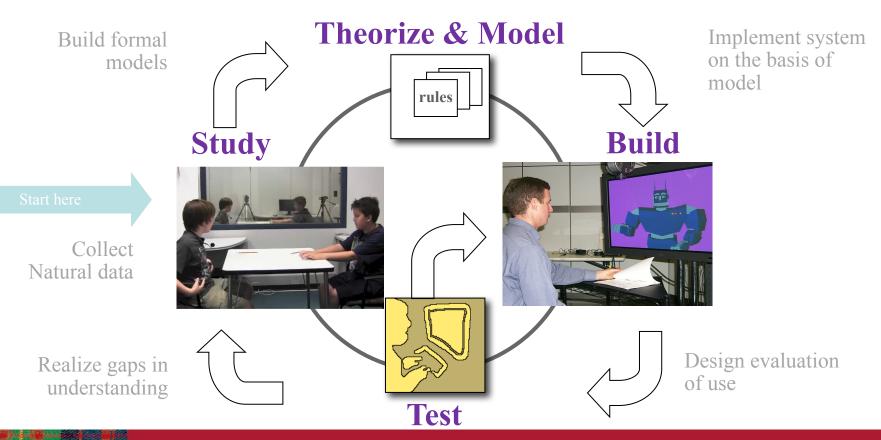


Carnegie Mellon University

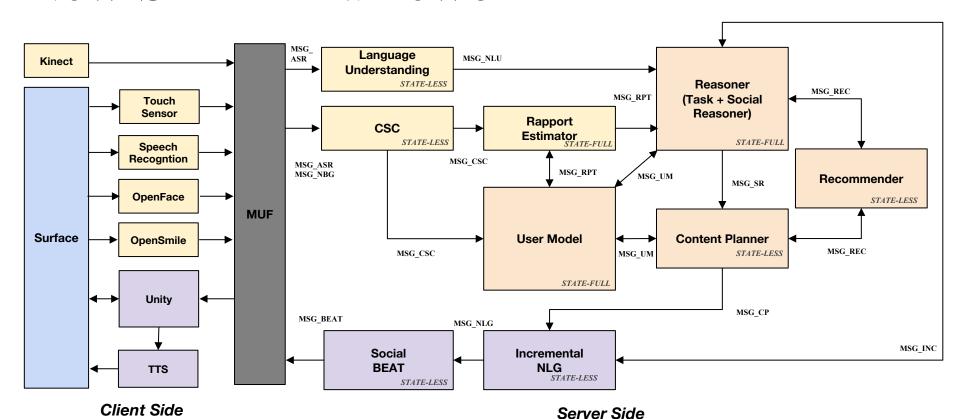
SARA: Socially-Aware Robot Assistant at Davos



Methodology



New SARA Framework



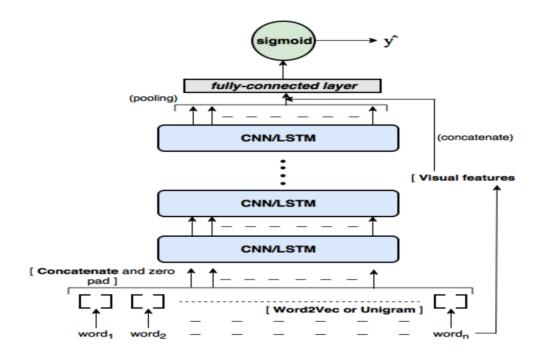
Indirectness Strategy Classifier

- Corpus
 - RAPT 2013: indirectness annotation of peer-tutoring corpus
 - ConLL 2010 shared task on uncertainty detection
 - Wikipedia dataset (Wikipedia articles)
 - BioScope dataset (abstracts and articles from biomedical literature)

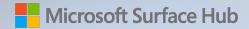
•	Code Definition		Example	Distribution
	Apology	Apologies used to soften direct speech acts	Sorry, its negative 2.	7.7%
	Qualifiers	Qualifying words for reducing intensity or certainty	You just add 5 to both sides.	66.1%
	Extenders	Indicating uncertainty by referring to vague categories	You have to multiply and stuff.	3.6%
	Subjectivizer	Making an utterance seem more subjective to reduce intensity	I think you divide by 3 here.	22.6%

IN (Indirect Delivery) "so I think what I'm gonna do is make that 15 minus 3 a 12"

Architecture: Indirectness Classifier



Pranav Goel, Yoichi Matsuyama, Michael Madaio & Justine Cassell, "I think it might help if we multiply, and not add": Detecting Indirectness in Conversation, International Workshop on Spoken Dialog System Technology (IWSDS 2018). – to appear



SARA Receptionist



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