An assistive handwashing system with emotional intelligence

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Agenda

- Problem Statement
 - Motivation
 - Objectives
- 2 Basic Concepts
 - Affect Control Theory (ACT)
 - The BayesACT Framework
- System Design
- Experimental Results
- Discussion

Problem Statement - Motivation

The COACH system



- An assistive system helping people with dementia (e.g. Alzheimer's Disease) completing daily activities
- Works well for some persons, but not as well for others
- Only considers functional states of users

Problem Statement - Objectives

To augment the COACH system with an emotional reasoning engine so that the augmented system:

- is designed in a portable and extensible way
- runs in real-time from the perspective of the user group
- provides a level of functional assistance
- produces the prompts according to the emotional state of a user

Using Emotional Intelligence in Assitive Systems is the future direction.

Concepts - ACT

Affect Control Theory (ACT)

- represents emotions as EPA vectors, where E stands for "evaluation",
 P stands for "potency", and A stands for "activity"
- describes social events by an Actor-Behaviour-Object grammar
- "fundamentals" of identities and behaviours; shared between people within a same culture
- "transient impressions": emotional feelings caused by a specific event

Concepts - ACT

Affect Control Theory (ACT)

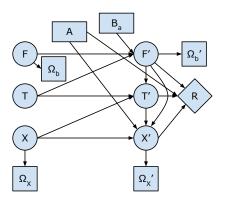
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The ACT Principle

Actors work to experience transient impressions that are consistent with their fundamental sentiments.

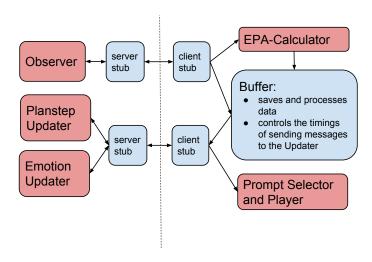
Concepts - BayesACT

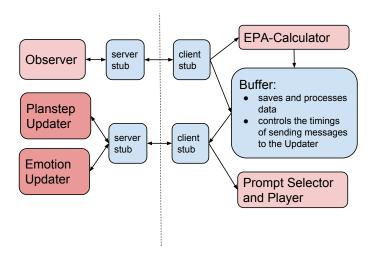
- A Bayesian version of the ACT theory
- Extends the ACT with POMDP model
- Uses a "turn-taking" model and represents state variables for Agent, Behaviour and Client (ABC)



- States $S = \{X, F, T\}$, where $F = \{F_{ij}\}, T = \{T_{ij}\}, i \in \{a, b, c\}, j \in \{e, p, a\}$
- Note: F_c denotes the agent's belief of the client's identity
- Observations $\Omega = \{\Omega_X, \Omega_b\}$
- Actions $\{A, B_a\}$
- Calculate $\{A, B_a\}$ based on $\{X, F, T\}$

Design - Overview





Use the BayesACT framework in the handwashing scenario

• Recall: BayesACT includes states $S = \{X, F, T\}$, observations $\Omega = \{\Omega_x, \Omega_b\}$, and agent actions $\{A, B_a\}$

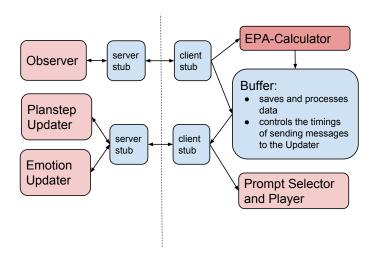
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Design - the EPA-Calculator



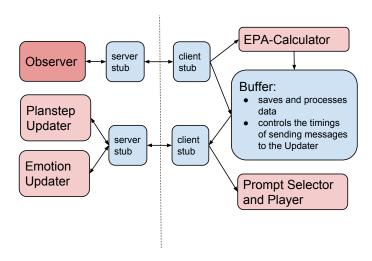
Design - the EPA-Calculator cont.

- Calculates affective meanings of user behaviours
- Feature Selection
 - analysis on facial expressions and speeches?
 - related research for this special application scenario hasn't been done yet

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- Our approach
 - ullet E stays neutral (value = 0); its value is ignored in the reasoning engine
 - P scaled from the expansiveness of the user's two hands
 - A scaled from the moving speeds of the user's hands
 - Used piecewise linear interpolation method
- "Confidence" of Ω_b can be specified in the reasoning engine

Design - the Observer



Design - the Observer cont.

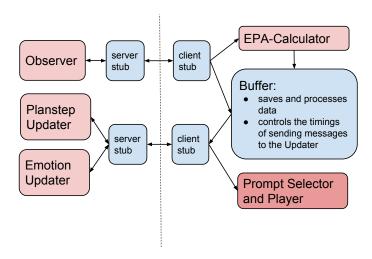
- Step 1: Get the locations of the user's hands
 - bases on a body tracker implemented in previous work
 - obtains locations of body parts from depth images taken from an overhead perspective
 - was trained using partially labeled, unbalanced data, and is configurable and re-trainable



Design - the Observer cont.

- Step 2: Map locations to user behaviours
 - if hands are close to an object, then there's high probability of performing the behaviour corresponded to the object
- Observation noise handled by the observation function in the reasoning engine

Design - the Prompt Selector and Player



Design - the Prompt Selector and Player cont.

- The prompt dataset
 - 30 audio-visual prompts generated in previous study
 - created using the USC Virtual Human Toolkit
 - EPA values of videos evaluated by human raters





- A proper prompt is selected as the final prompt if it:
 - has the same propositional labels as the desired prompt
 - has the closest emotional (EPA) values as the desired prompt

Experiments - Latency of the system

Average latency of the system

- 46.80ms for obtaining user behaviours
- 1.65s for calculating and updating functional and emotional beliefs
- 1.70s in total

The system runs in real-time from the perspective of its user group

Experiments - Two laboratory tests

Two laboratory tests

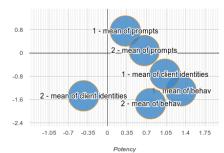
- washed hands while the system observed and assisted in real time
- acted more powerfully and more actively in the first test than the second
- link to video of test #2
- results shown in thesis

Another 15 tests were also run. Results are in the Appendix.

Experiments - Conclusion

- Functionality performance
 - sometimes false positively recognizes an user behaviour
 - in general, is able to produce propositionally useful system prompts
- Emotionality performance

Comparison of P and A values in the two tests



Generally, user behaviours with higher *P* and higher *A* values lead to

- client identities with higher *P* and higher *A* values
- system prompts with lower P and higher A values
- as predicted by the Affect Control Theory

Discussion

Contribution

- Designed and implemented a prototypical hand-washing system that satisfies the objectives
- Tests also indicated a correlation between the EPA values of user behaviours, user identities, and system prompts

Future Work

- Improve the EPA-Calculator
- Improve the prompt generation process
- Improve the Planstep- and Emotion- Updater
- Conduct clinical trials for the system

Acknowledgement

This work is based on previous works of:

- Hoey, J., Schroder, T., & Alhothali, A. (2013, September). Bayesian affect control theory. In Affective Computing and Intelligent Interaction (ACII), 2013 Humaine Association Conference on (pp. 166-172). IEEE.
- Czarnuch, S., & Mihailidis, A. (2014 (in review)). Depth image hand tracking from an overhead perspective using partially labeled, unbalanced data: Development and real-world testing. IEEE Journal of Biomedical and Health Informatics.
- Malhotra, A., Yu, C., Schroder, T., & Hoey, J. (2014 (in review)). An exploratory study into the use of an emotionally aware cognitive assistant.

I'd like to take this opportunity to thank:

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 - Stephen Czarnuch, Aarti Malhotra and Celia Yu
 - Families and friends, especially Chengbo Li, Xiao Yang, Enxun Wei and Luyi Lin

The end

Thank you!

- Questions?
- Comments?

Concepts - BayesACT cont.

Model formulation

• The deflection $\phi(F, T)$ between F and T:

$$\phi(f,t) \propto e^{-(f'-t')\Sigma^{-1}(f-t)} \tag{1}$$

• The probability of a post-action fundamental sentiment f':

$$Pr(f'|f,t,x,b_a,\phi) \propto e^{-\phi(f',t')-\xi(f',f,b_a,x)}$$
 (2)

where t' can be computed from $\{f', t, x\}$ by empirically derived prediction equations of ACT.

- Pr(x'|x, f', t', a): how the application progresses
- $Pr(\omega_b|f)$ and $Pr(\omega_x|x)$: observation functions for the client behaviour sentiment and system state

Update X'_{ps} based on Ω_x and $\{X_{ps}, X_{behav}, X_{aw}, F, T\}$

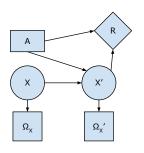
- SampleXVar() and evalSampleXVar()
- Pseudocode of SampleXVar() (on next page)
- $Pr: X_{behav} \rightarrow \Delta(\Omega_X)$ used in evalSampleXVar()

Update X'_{ps} based on Ω_x and $\{X_{ps}, X_{behav}, X_{aw}, F, T\}$ cont.

```
1: if Deflection(F, T) is high then
                                                 19:
                                                               aw = low and not moving forward
       threshold = high
                                                  20:
                                                            else
 3. else
                                                 21.
                                                               aw stays high and moving forward
                                                  22:
 4:
       threshold = low
                                                            end if
 5: end if
                                                 23:
                                                         end if
6: if aw high then
                                                 24 else
                                                  25:
 7:
       if prompted then
                                                         if prompted then
8.
          if random_prob() < threshold then
                                                 26.
                                                            if random_prob() > threshold and
9:
             aw = low and not moving forward
                                                            prompt correct then
10:
          else if prompt wrong then
                                                 27:
                                                               move on and aw high
11.
             aw = low and not moving forward
                                                  28:
                                                            else
12:
          else if likely then
                                                  29:
                                                               unlikely: aw high and not moving
13.
             moving forward
                                                               forward
14.
          else if random_prob() < threshold
                                                            end if
                                                 30:
                                                 31:
          then
                                                         else
15.
             aw = low and not moving forward
                                                  32:
                                                            unlikely: aw high and moving forward
16:
                                                 33:
          end if
                                                         end if
                                                 34: end if
17:
       else
18:
          if random_prob() < threshold then
```

Concepts - POMDP

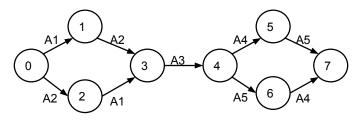
Partially Observable Markov Decision Process (POMDP)



- A timeslice of a POMDP process
- ullet Variables: $\{X, A, \Omega_X\}$
- $Pr: X \to \Delta(\Omega_X)$, $Pr: X \times A \to \Delta(X)$
- Reward Function: R(A, X')

Solution - Representing "Functional States"

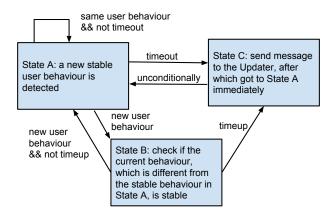
Planstep Definition and Update Diagram



- Eight plansteps: (0) "off/dirty/dry", (1) "on/dirty/dry", (2) "off/soapy/dry", (3) "on/soapy/dry", (4) "on/clean/wet", (5) "off/clean/wet", (6) "on/clean/dry", (7) "off/clean/dry"
- Five behaviours: A1 to A5 are "turn on water", "put on soap", "rinse hands", "turn off water", and "use towel", respectively.

Solution - The Buffer

- Between the Observer, the EPA-Calc, and the Reasoning Engine
- Controls timings of sending messages



Smoothes EPA values calucated by the Calculator

Experiments - Parameter values used in laboratory experiments

Param.	Value	Defined in which component
n	10	EPA-Calc
distance	$\{-\infty, 0, 8, 40, 128, 160, +\infty\}$	EPA-Calc
potency	$\{-4.3, -4.3, 0, 1, 2, 4.3, 4.3\}$	EPA-Calc
difference	$\{-\infty, 0, 3.5, 17.5, 35, 70, +\infty\}$	EPA-Calc
activity	$\{-4.3, -4.3, -2, -1, 0, 4.3, 4.3\}$	EPA-Calc
alpha	0	Buffer
timeout	300	Buffer
timeup	1	Buffer
β_a^0	0.001	Updater
β_c^0	2.0	Updater
γ	(100000, 1.0, 0.5)	Updater
N	2000	Updater
f_a^0	[1.5, 0.51, 0.45]	Updater
f_c^0	Different in each test	Updater