

Outlook/Mica

January 18

Evaluation of Adaptive UI

□ For **performance** and **user satisfaction**

- Wizard of Oz Studies
- Simulations using data from a non-adaptive system
- Controlled studies
- Field Studies

LookOut

LookOut

Functions

Support
System
Usage

Support
Info Acquisition/
Decision Making

Support
Learning

Support
Collaboration

Support
Entertainment

Take Over
Routine
Tasks

Adapt
the
Interface

Advice
on
System
Usage

Retrieve Info/
Recommend Objects

Tailor
Info
Presentation

Advice
on
task

Forms of Adaptation

Horvitz Mixed-Initiative principles

1. Significant value-added automation
2. Consider uncertainty about user goals
3. Consider status of user attention in timing services
4. Infer ideal action in light of costs, benefits and **uncertainties**
5. Use dialogue to resolve uncertainty
6. Allow direct invocation and termination
7. Minimize cost of poor guesses
8. Match precision of services with goal uncertainty
9. Mechanisms for user-system collaboration to refine results
10. Socially appropriate behaviors for agent-user interaction
11. Maintaining working memory of recent interactions
12. Continuous learning via observation

Acting Rationally

- AI as study and design of **intelligent agents** that **act rationally** in their environment
 - Their **actions** are **appropriate** for their goals and circumstances
 - They are **flexible** to changing environments and goals
 - They **learn** from experience
 - They make **appropriate choices** given **perceptual limitations** and **limited resources**
- This definition drops the constraint of *cognitive plausibility*
 - Same as building flying machines by understanding general principles of flying (aerodynamic) vs. by reproducing how birds fly

Taking over routine tasks: Microsoft Lookout

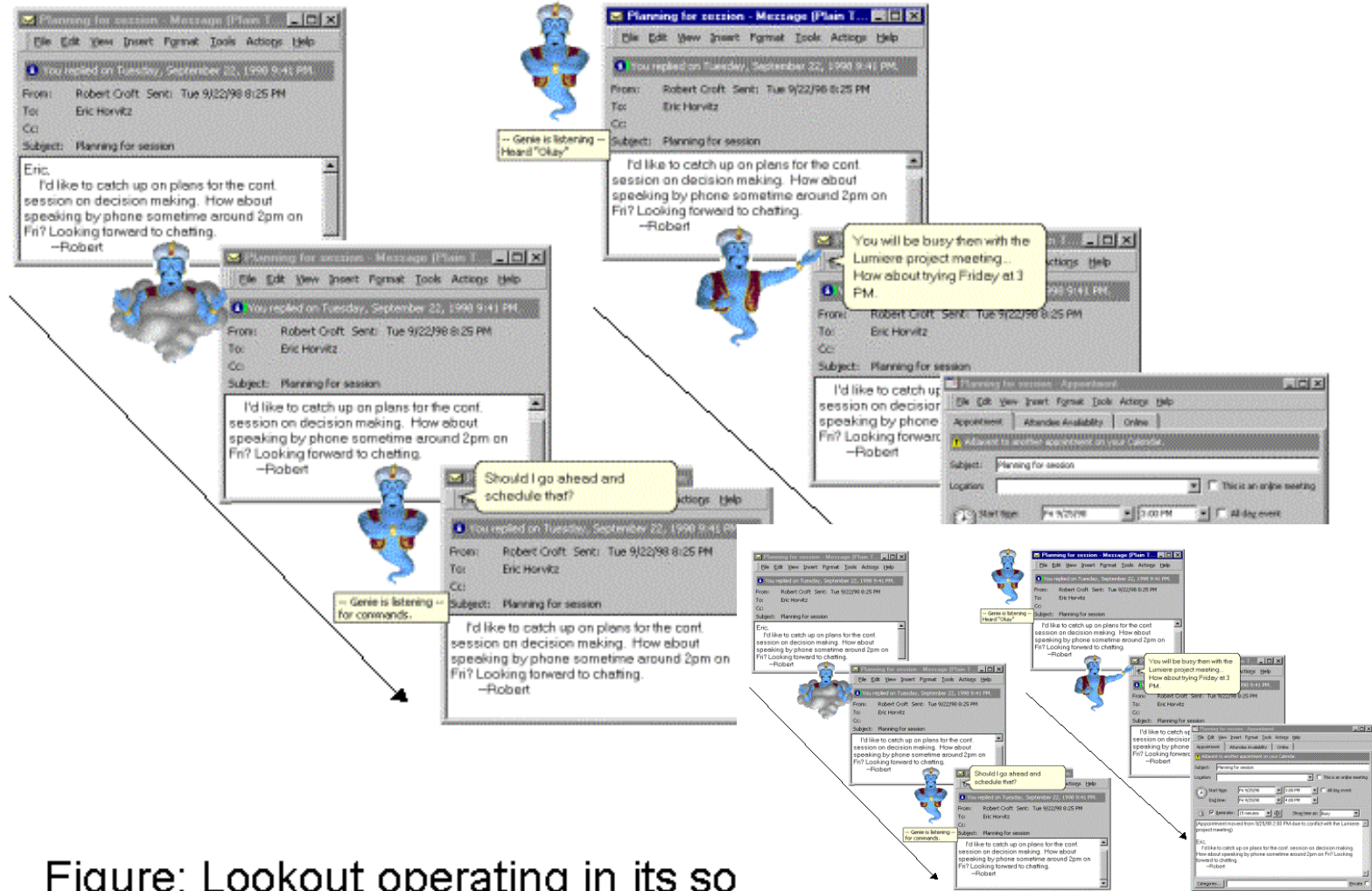


Figure: Lookout operating in its so

Figure: Lookout operating in its social user-interface modality.

Inference/Learning

Forms of adaptation

User Model

Inference/Learning:

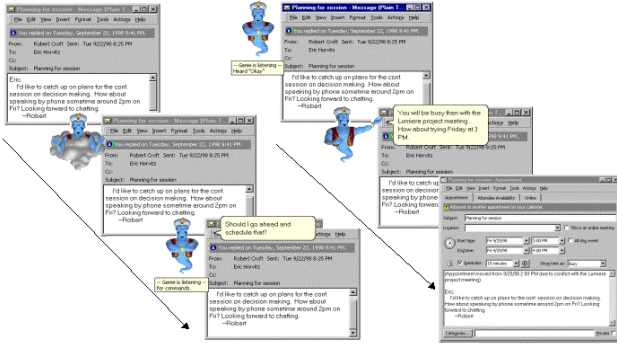
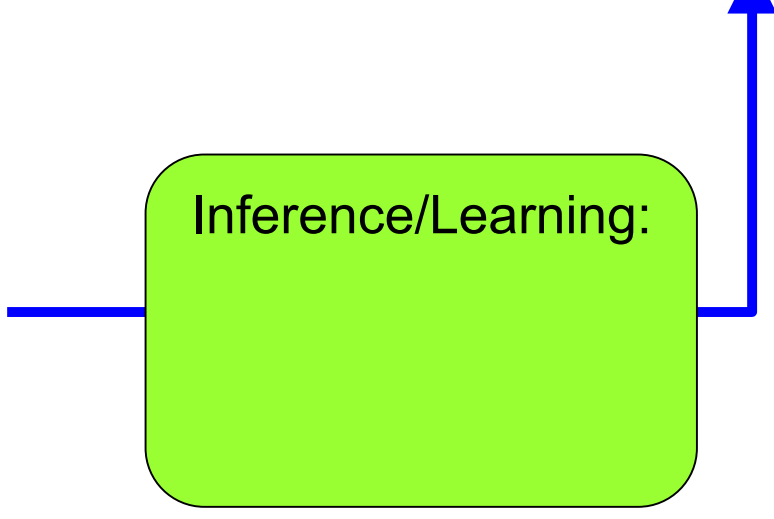
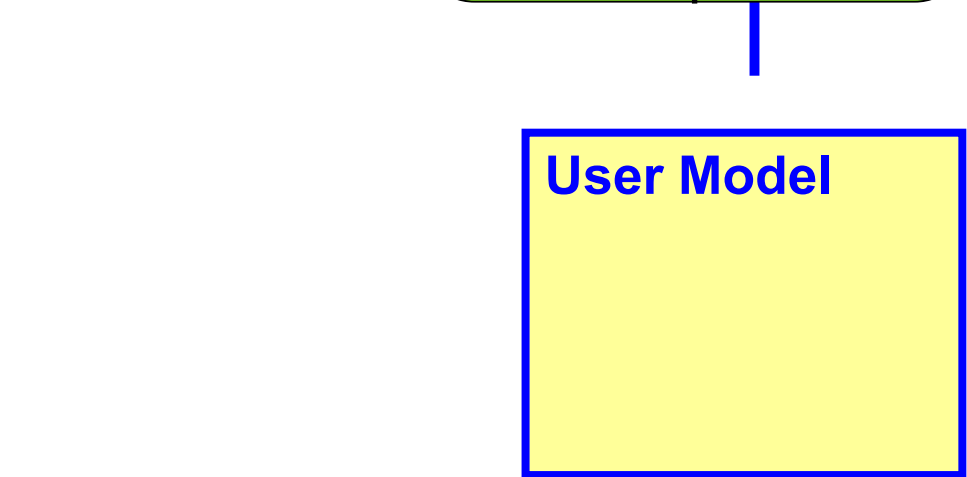
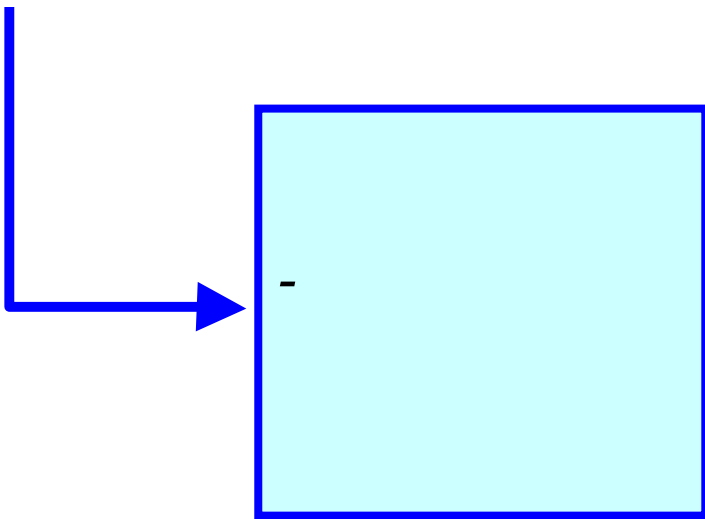


Figure: Lookout operating in its social user-interface modality.



Inference/Learning

Forms of adaptation

User Model

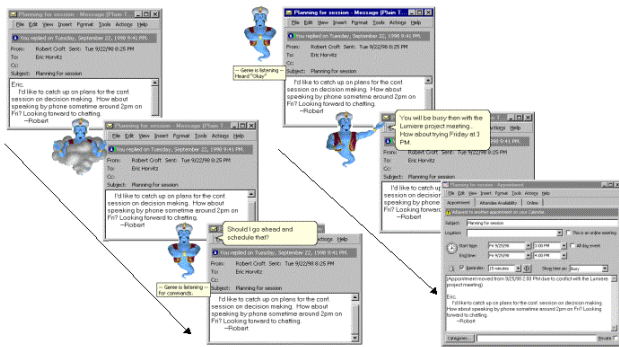


Figure: Lookout operating in its social user-interface modality.

Let's start from Inference/learning from user model to adaptation

Inference from User Model to Adaptation

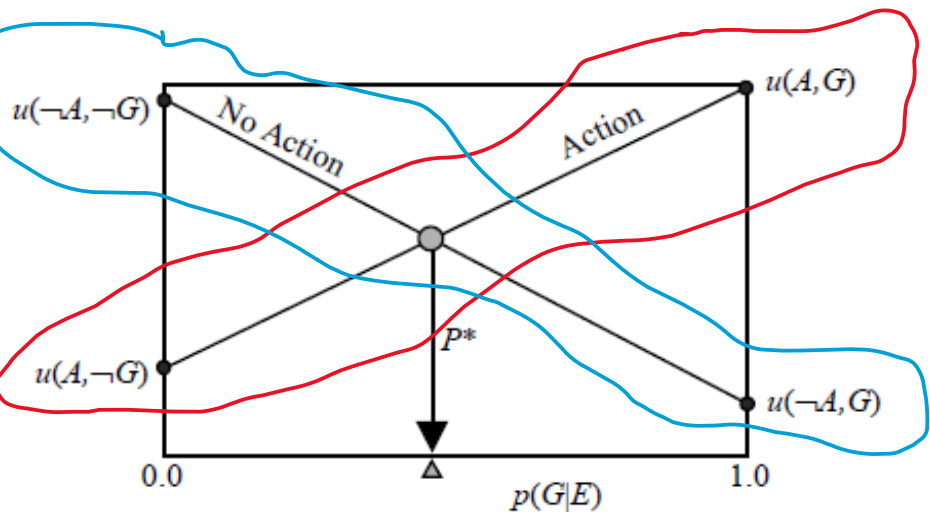
➤ Based on Utility Theory

	Goal	No Goal
Action	$U(A, G)$	$U(A, noG)$
No action	$U(noA, G)$	$U(noA, noG)$

$$eu(A|E) = p(G|E)u(A, G) + p(\neg G|E) u(A, \neg G) =$$

$$p(G|E)u(A, G) + [1-p(G|E)] u(A, \neg G)$$

$$p(\neg G|E) = 1 - p(G|E)$$



Similar equation for **No Action** ($\neg A$)

Chose the behavior with **Max Expected Utility (EU)**

Figure 4. Graphical analysis of the expected utility of action versus inaction, yielding a threshold probability for action.

Inference for Model Application

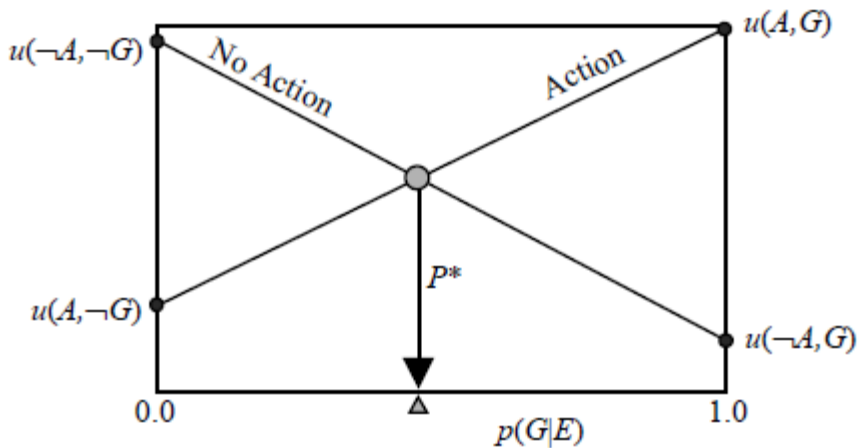


Figure 4. Graphical analysis of the expected utility of action versus inaction, yielding a threshold probability for action.

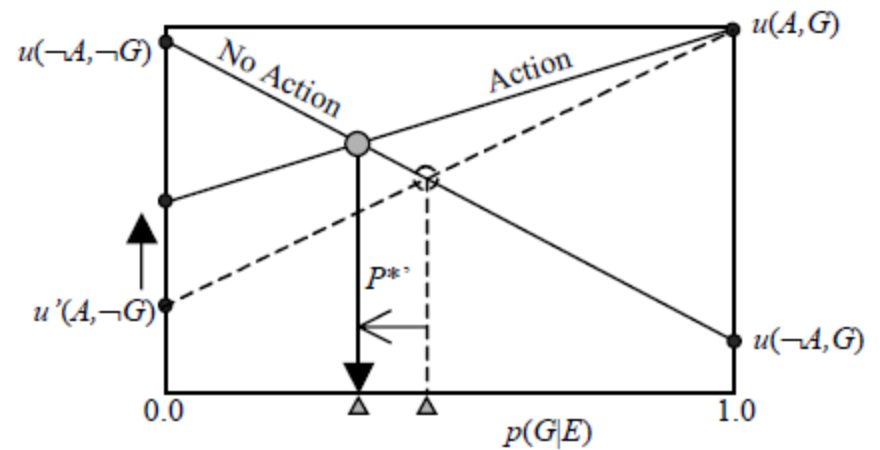


Figure 5. The result of increasing the value of taking erroneous action. Context-dependent shifts in any of the utilities can change the probability threshold for action.

Inference/Learning

Forms of adaptation

User Model

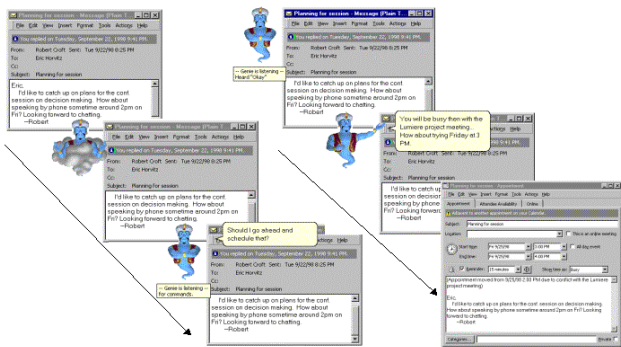


Figure: Outlook operating in its social user-interface modality.

Let's start from this part

Inference/Learning

Find action with
Max EU

Forms of
adaptation

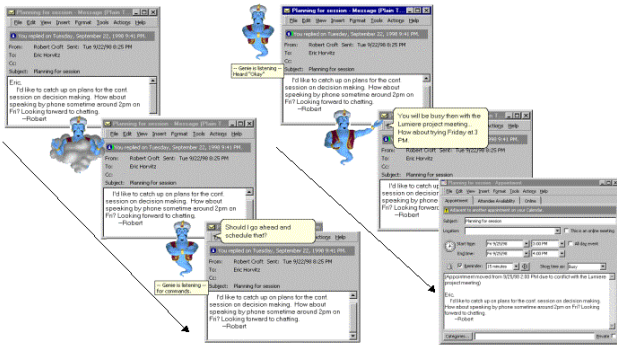


Figure: Lookout operating in its social user-interface modality.

User Model

$U(A, G)$,

$U(A \text{ not} G)$,

$U(\text{not } A, G)$

$U(\text{not } A, \text{not} G)$

$P(G/E)$

Inference/Learning

Find action with Max EU

Forms of adaptation

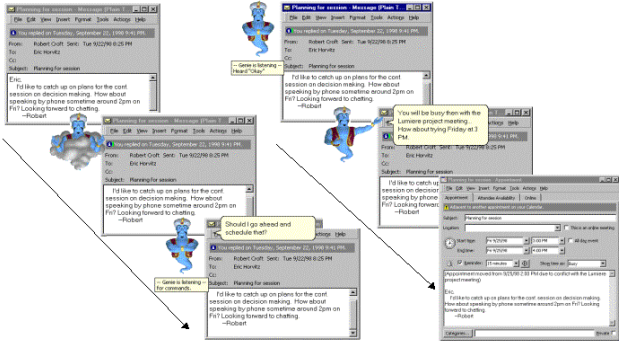
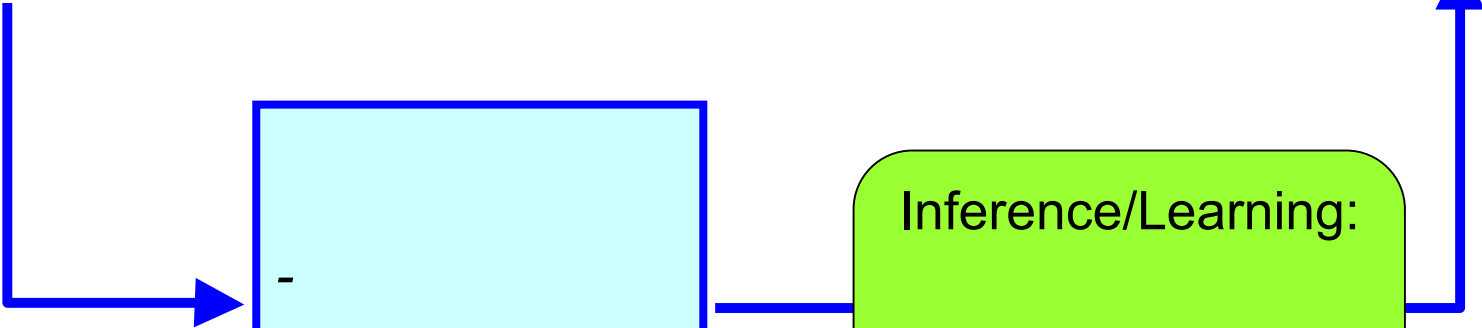
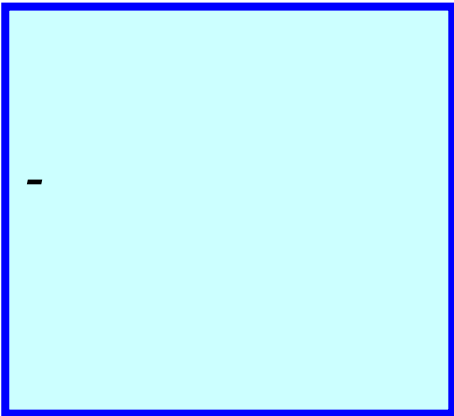


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Inference/Learning:



Inference/Learning
Find action with
Max EU

Forms of
adaptation

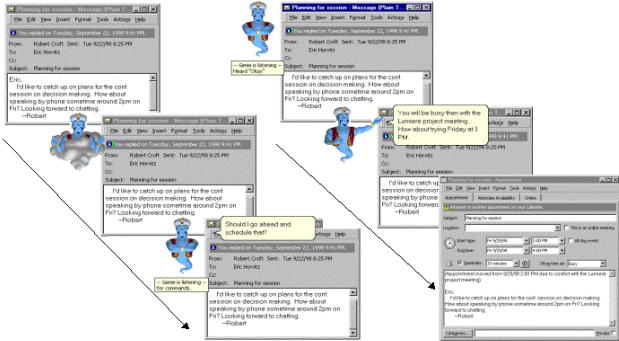
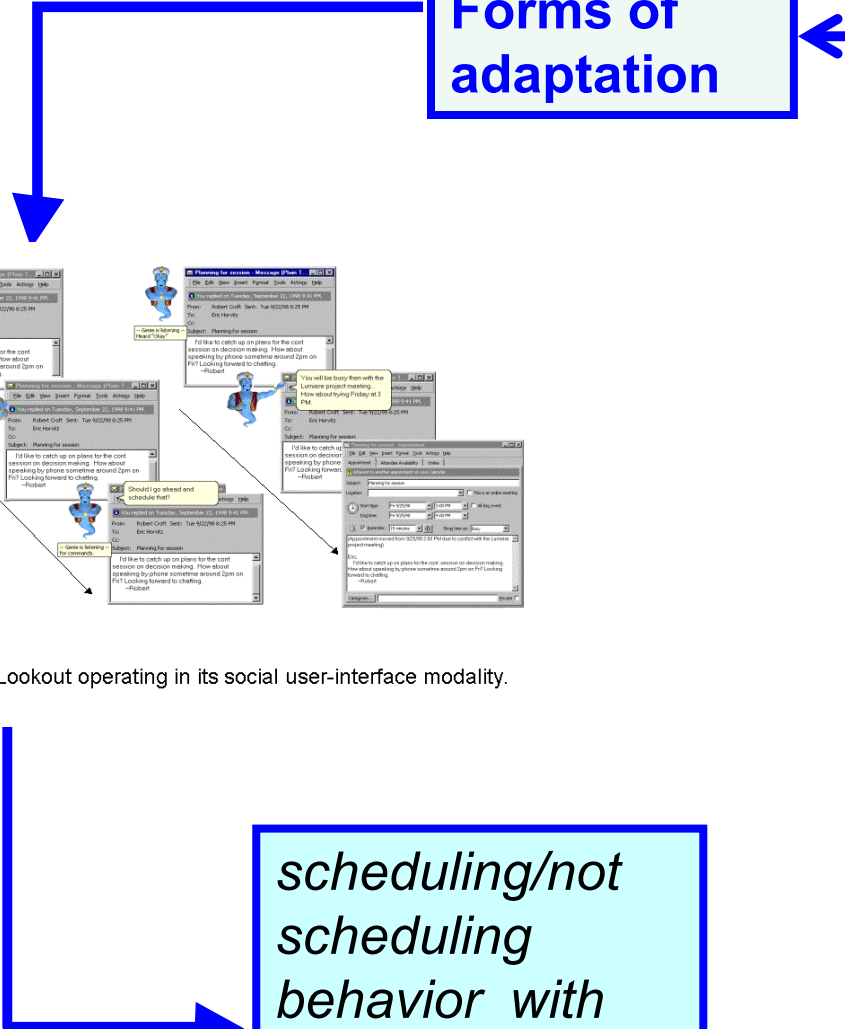


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User Model
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 $U(\text{not } A, G)$,
 $U(\text{not } A, \text{not}G)$
 $P(G/E)$

Inference/Learning:
SVM text classifier

*scheduling/not
scheduling
behavior with
previous emails*



Inference for Model Application

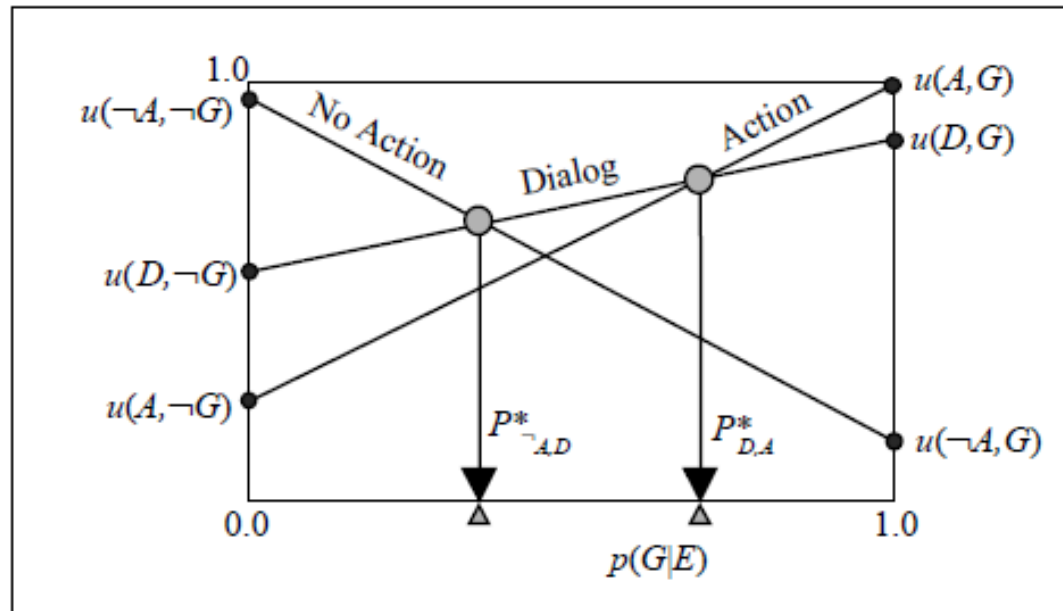


Figure 6. Adding a second action option consisting of dialog with users about their goals. In this case, the graphical analysis highlights the origin of two threshold probabilities for guiding the action of autonomous services.

User's input in LookOut

□ Explicit

- Self-reports on $U(G, A)$

□ Non Explicit

- Previous scheduling behaviors

Acquisition mechanisms in LookOut

□ Knowledge-Based (or Expert-Based)

- Define rules (deterministic or probabilistic) to identify relevant user properties based on existing theories/knowledge

□ Data-Based

- Learn relevant user features from data (e.g. labeled or unlabelled example behaviors)

□ Hybrid

Domain Model in LookOut

Closed World (e.g. domain to be taught in educational application)

- Usually well defined
- Rich representations are possible

Open World (e.g. the Web)

- Ill defined
- Requires to deal with lower levels of representation

Interface Features Important for Mixed Initiative

Interface Features Important for Mixed Initiative

- ❑ Multiple interaction modalities
- ❑ Variable dwell time for a response
- ❑ Don't take final action without user approval