

Impact of agent's answers variability on its believability and human-likeness and consequent chatbot improvements

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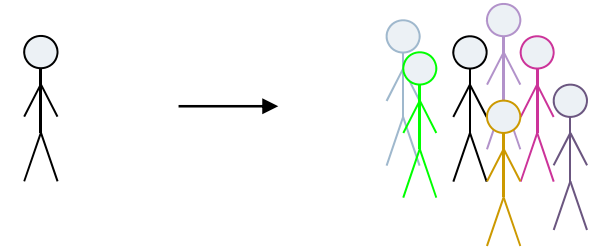
Outline

- Context: assisting novice users with ECA
 - The increasing need for assistance
 - Assisting novice users with ECA
 - Help systems comparison
 - Dialogue system or chatbots?
 - Key issues
- Methodology
- Results
- Conclusion

The increasing need for assistance

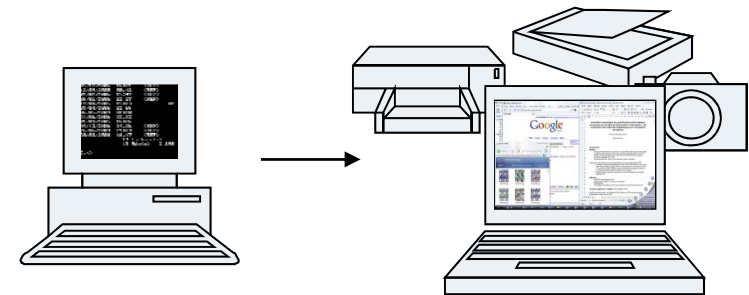
- **Users** evolution:

- In number:
600 millions (2002) →
2 billions (2015 – projection)
- In variety:
from computer scientists to everyone



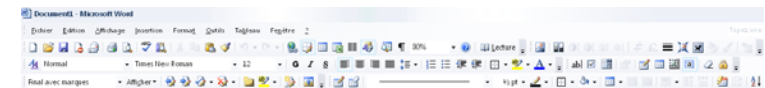
- **Hardware** evolution (Moore's law):

- Application fields
- Interaction fields



- **Software** evolution:

- More numerous
- More complex: in public applications
150 « basic » actions (in menus);
60 dialogue boxes ;
80 tools (through icons).
(Beaudoin-Lafon, 1997)



Assisting novice users with ECA

- **Assisting:** « An Assisting Agent is a software tool with the capacity to resolve help requests, issuing from novice users, about the static structure and the dynamic functioning of software components or services » (Maes, 1994)
- **Conversational:** interaction in unconstrained natural language (NL)
Why?
Frustrated (novice) users spontaneously express use NL
(\Leftrightarrow « thinking aloud effect » (Ummelen & Neutelings, 2000))
- **Embodied:** given a graphical more or less realistic appearance
Why?
Increased agreeability and believability – « Persona Effect »
(Lester, 1997)

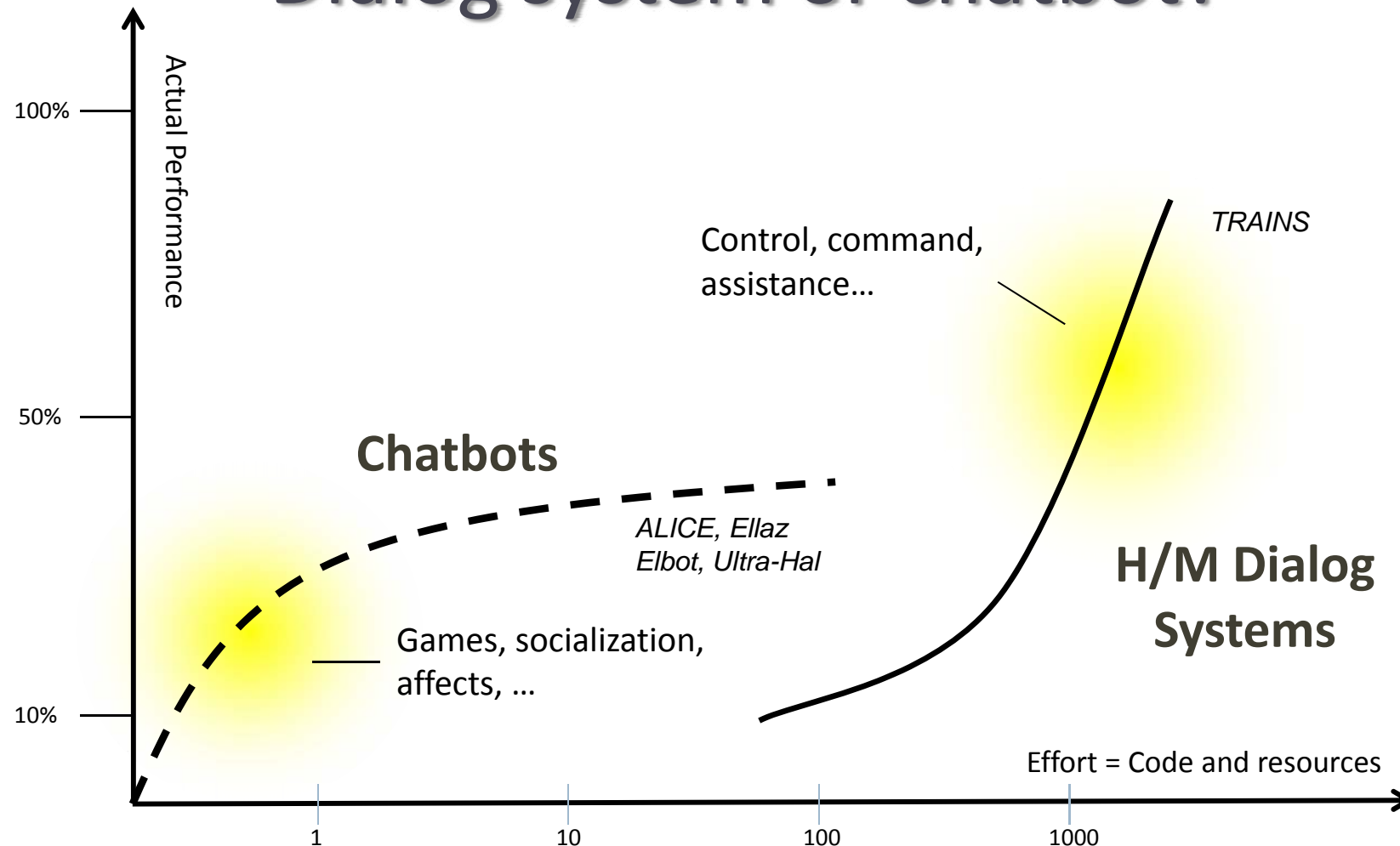
Help systems comparison

Help system	Reactivity	Vocabulary	Task-oriented	Dynamic	Personalized	Proactive
Paper documentation	-	-	-	-	-	-
Electronic documentation	+	-	-	-	-	-
FAQ, How-to, Tutorial	+	=	+	-	-	-
Contextual Help Systems	+	=	=	+	-	-
Assisting Conversational Agent	+	+	+	+	+	=

- **Reactivity**: how fast is it for the user to open the help system when it needs it?
- **Vocabulary**: are there strong constraints or limitations on the words the user has to know to efficiently use the help system? (ex: specific keywords/grammar constructions for NL)
- **Task-oriented**: does the help system explain procedures and not only define concepts?
- **Dynamic**: does the help system change according to the application state?
- **Personalized**: does the help system change according to the user?
- **Proactive**: does the help system appear only when asked for or can it anticipate the user needs (without being intrusive)?

Conclusion: Assisting conversational agents *potentially* seem to be the most efficient way to help novice users.

Dialog system or chatbot?



Chatbots are limited in terms of **genericity** (need to rebuild everytime) (Allen, 1995) and **linguistically** (Wollermann, 2006) – but how far can we push the approach?

Dialog system or chatbot?

- **Advantages: easy, light, precise**
 - They are **easy to develop**: no large semantic analyzer, no complex reasoning tools;
 - They are **light to deploy** in a web-based environment → client architectures can be envisioned;
 - They provide **robust** natural language reactions (Evasive list effect – ELIZA effect);
 - They are tailored and well-suited for the field of **ludo-social** chat;
 - When associated with a given application, they **can** be customized to **be extremely precise**.
- **Drawbacks: lack of genericity, linguistic limits**
 - Minimalistic/**ultra-customized** model of the application;
 - **Minimalistic model** of the dialogue session and of the users;
 - **No semantic analyzer** → lack of precision in the requests (grammar, speech acts, ...);
 - **No formal requests** → class reactions are directly linked to specific linguistics patterns;
 - **No generic reasoning tools**, especially when the function of assistance is concerned.

Key issues

Hypothesis: variability improves user's perception of the ECA

1. Technical feasibility: is it **possible** to handle variability with a chatbot architecture?
2. Need: do people **notice** variability?
3. Effect: does it **affect** the perception users have of the agent?
And if yes, how?
4. Can it be **useful** for assistance?

Outline reminder

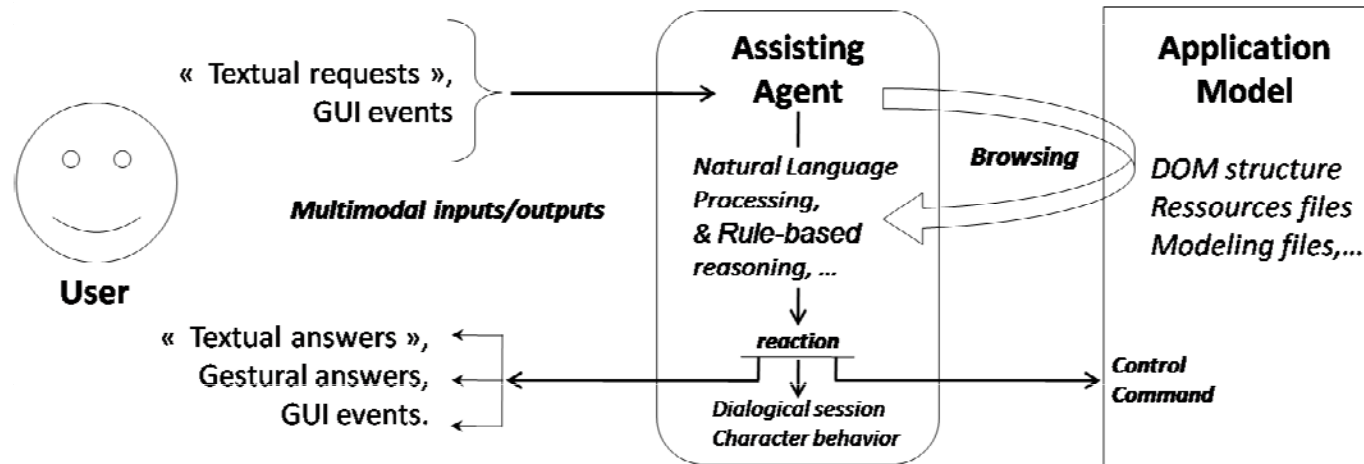
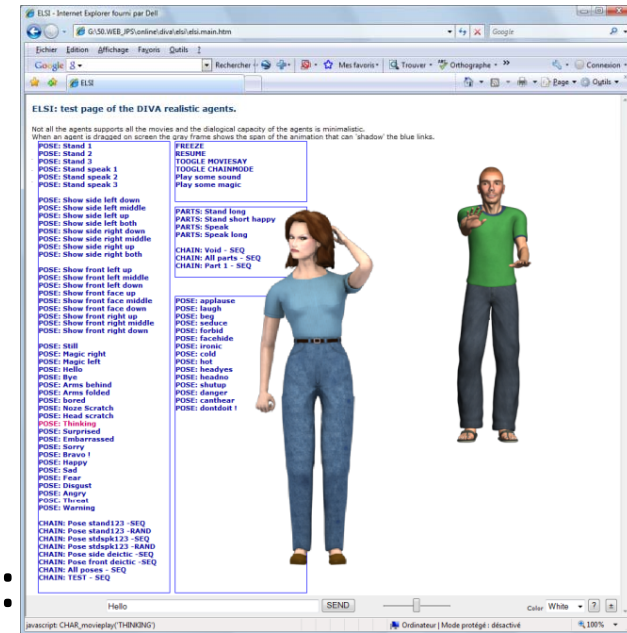
- Context: assisting novice users with ECA
- Methodology
 - Experimental framework: DIVA framework overview
 - Experimental framework: DIVA NLP-chain
 - Experiment principles
 - Experimental protocol
 - Questionnaires
- Results
- Conclusion

DIVA framework overview

- Dom Integrated Virtual Agent:
 - Open programming framework
 - High level of interaction (AJAX)

1. Embodied Agents Elsi & Cyril:

2. Natural Language Processing chain:

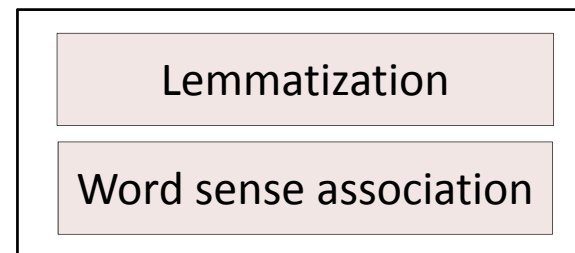


Experimental framework: DIVA NLP-chain

1. Formalization phase

1. Sentences are preprocessed and words are lemmatized;
2. A semantic class (KEY) is associated with each word

« Natural Language request »



« INTERMEDIATE FORMAL REQUEST FORM »

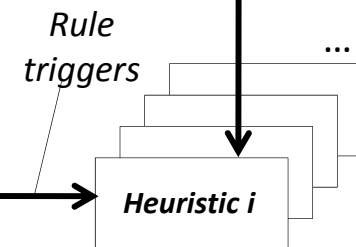
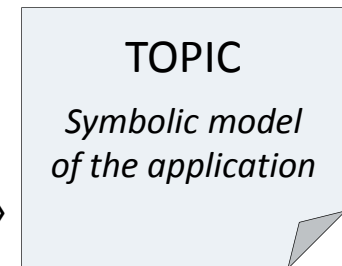


2. Interpretation phase

Interpretation rules are of the form:
Pattern → Reaction

Where reactions are expressed as procedural heuristics achieving reasoning tasks over the description of the application (the topic file).

Customized
Generic



Multimodal response from the assisting agent



Experimental framework: DIVA NLP-chain

« *How old are you?* »

DIVA:

1) **Formalization:**

<QUEST HOW ISOLD TOBE THEAVATAR>

2) **Interpretation:**

<rule id="age" pat="QUEST THEAGE | HOW ISOLD">

<do>

THETOPIC.age.asks++;

If (THETOPIC.age.asks >= 1)

TALK_prepend(['As I said', 'I've told you, ']);

If (THETOPIC.gender = 'female')

TALK.say('It's not polite to ask this.');

</do>

<say>

<p>I'm _THETOPIC.age_. years old</p>

<p>I'm _THETOPIC.age_ ...</p>

<p>My age is _THETOPIC.age_</p>

</say>

</rule>

} variability

} genericity

Classical chatbots (ALICE – AIML):

<category>

<pattern>HOW OLD ARE YOU</pattern>

<template>

<set_it>I</set_it> am 25 years old

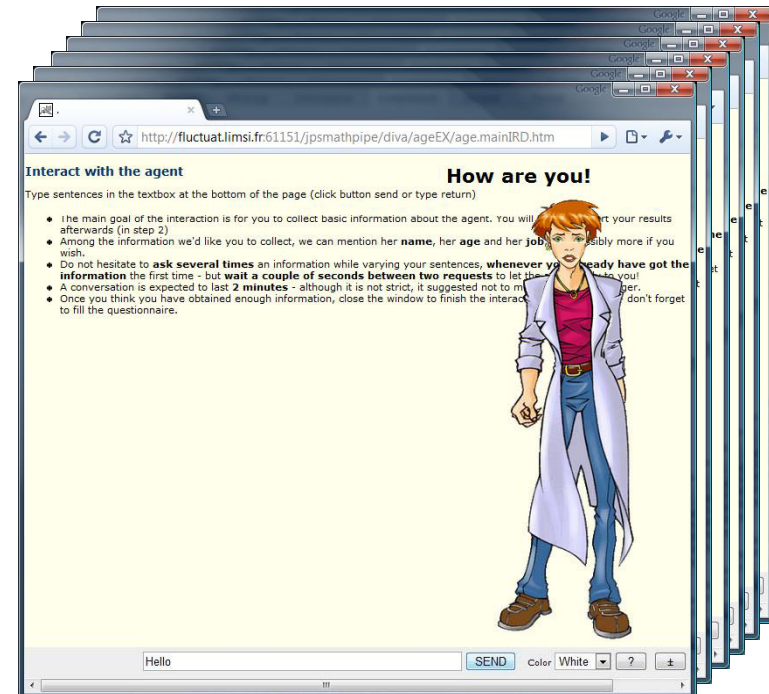
</template>

</category>

1. Matches a user input containing the **exact pattern**
2. Handles a **minimalistic** model of the session (IT)
3. Sends an entirely **predefined** answer

Experiment principles (1)

- Three (linked) parameters actually tested:
 - **Responsivity**: the requested information is in the answer
 - **Variability**: twice the same question can lead to different answers
 - **Dependence**: variability with a memory of previous questions
- Differences: one only answer when requested its **age**.
- 6 female agents, visually identical
- Interaction through chatbox at the bottom of the window



Experiment principles (2)

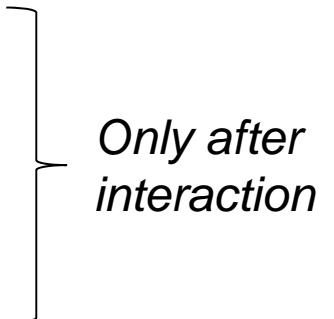
« *How old are you?* »

	Responsive	Variable	Dependent	1 st reply	2 nd reply	3 rd reply
1	✓	✓	✓	I'm 25	I told you I'm 25	I won't answer to that again
2	✓	✓	✗	I'm 25	25 years old	I'm 25 years old
3	✓	✗	-	I'm 25	I'm 25	I'm 25
4	✗	✓	✓	I won't tell you	I said I won't tell you this	Stop insisting!
5	✗	✓	✗	I won't tell you	It's a secret	I will not tell you
6	✗	✗	-	I won't tell you	I won't tell you	I won't tell you

Experimental protocol

- User's objective: retrieving information about an agent
 - Free chat
 - Suggestions:
 - Examples given: name, age, job...
 - Short interaction (< 2 minutes)
- Interaction with **two agents**:
 - Case 1 or Case [2..6]
 - Case [2..6] or Case 1
- Three **questionnaires**:
 - One after each interaction (5-point Likert scales)
 - Final comparative questionnaire

Questionnaires

- 7 parameters evaluated:
 - **Variability**: not always answering the same way
⇔ noticing variability
 - **Cooperation**: if information requested could be obtained ⇔ noticing responsiveness

Only after interaction

- **Precision**: « 25 years old » / « young »
- **Relevance**: the agent remains in the topic of conversation
- **Believability**: the agent being a female is believable
- Human-likeness: same answer could come from a human being
- Global satisfaction: overall feeling about conversation

Outline reminder

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- Results
 - Raw results
 - Comparative questionnaire results
 - Post-interaction questionnaire results
- Conclusion

Raw results

- 21 subjects, over the internet
 - Sex: 14 men / 7 women
 - Age: 20-60 (62% in 26-30)
 - Origin: Chinese/French mainly
 - Studies: university level (85%)
 - Computer science knowledge: disparate (42% below 3/5)
- 38 post-interaction questionnaires
- 19 final questionnaires

Comparative questionnaire results

- Globally: 1 vs all
if a difference is made, **1 is preferred**, for every parameter
- Individually: 1 vs [2-6]
if a difference is made, **1 is preferred**, except:
 - 4 ($\neg RVD$) is perceived as more human-like
 - 6 ($\neg R \neg V$) is perceived as more relevant
- Discussion:
 - Not giving the age of a woman is not problematic: parameters interdependancy
 - Variability is even more crucial in that case (4 vs 5-6): expectation of a high level behavior

Post-interaction questionnaire results

- Sample too small to obtain many statistically significant results
- Many expected results:
 - Satisfaction: $RVD > \neg R \neg V$
 - Cooperation: $RVD > 5$, $RVD > \neg R \neg V$
 - Precision: $RVD > \neg RVD$, $RVD > \neg R \neg V$
- Some unexpected ones:
 - Precision: $RVD < R \neg V$
 - Believability: $RVD < RV \neg D$
 - Human-likeness: $RVD < R \neg V$
- Discussion:
 - Variability can make the agent look more imprecise
 - If the rest of the behavior doesn't follow, it is interpreted as mistakes

Conclusion

- **Possibility** to handle variability with a chatbot architecture
- Users **notice** variability in agents
- Agents with variability are perceived as:
 - more believable,
 - more human-like......but **coherence** is crucial!
- Can it be useful for assistance?
 - Indirectly yes:
 - chat is important (~40%) even for assisting agents only (*Bouchet&Sansonnet, 2007*)
 - improved user's satisfaction
 - reduced « motivational paradox » (*Carroll&Rosson, 1987*)
 - Directly? Upcoming experiment
- Variant: behaviours affecting every parameter
- Study of parameters influence on each other (ex: gender/age)