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Platform for Collaborative Multimodal Plan-Based Multiparty Dialogue Systems

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Heralding the Intelligent Enterprise

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Purpose

Show that the collaborative multimodal, multi-party plan-based approach can improve upon the state-of-the-art in virtual assistant dialogue systems.

Current “Frame-based” or “Task-Oriented” Dialogue Systems

(see Bobrow et al., 1977; Grosz, 1974; 1977)

Very limited goal --

- System is to perform actions – e.g., Book movie tickets, Reserve restaurants
- System determines which action, and prompts user to provide values for the “slots” (parameters) in the desired action “frame”;

NLP is (mostly) “action/intent’ classification” + “slot-filling”

Prompt user for missing “slot” values;

- E.g, the date & time you want to eat, number of people, etc.
- User supplies atomic value(s) — such as “Monday” “at 7 pm”

Learn “optimal” slot-filling dialogue policies from data

- Mostly academic research competitions, conducted on artificial data, often generated from a user simulator, often operating at the speech act level

Some Limitations of Slot-filling systems

U: *“Please book me a table at Guillaume*

S: *“what time do you want to eat?”*

U: *“before 7 pm”, or*

“not before 7 pm”

“between 7 and 8 pm”

“the earliest time available”

No slot value is provided, only a constraint.

U: *“”Friday before 10”*

Over-answering, with constraints

U: *“what’s the earliest time available?”*

S: *“6 pm”*

U: *“too early”*

S: *“how about 7 pm?”*

U: *“OK”*

Collaborative mixed initiative – User doesn’t fill slot -- parties collaboratively fill the slot.

Collaborative Plan-Based Approach to Dialogue

Origin Story (1975-1980)

- Shopper: Where are the chuck steaks you advertised for 88 cents per pound?
- Butcher: How many do you want?

People infer each others' plans and attempt to help them overcome obstacles to their success.

Communicative actions are part of those plans

Collaborative Plan-Based Approach to Dialogue

- Plan-based

- People can plan their (task-related) communications in the same way as planning domain actions

See Allen/Perrault, 1980; Allen et al., 1995; Appelt, 1982; Brown, 1980; Bruce, 1975; Carberry, 1985; Cohen, 1978; Cohen/Perrault, 1979; Cohen/Levesque, 1990; Ferguson/Allen, 2007; Koller/Stone, 1997; Lesh et al., 1999; Litman/Allen, 1987; Perrault/Allen, 1980; Pollack, 1990; Poesio/Traum, 1997/8; Sadek, 1992; Sidner, 1983/5; Traum/Allen, 1994, and *many many others*. See also Allen et al's TRIPS, TRAINS, COGENT systems.

- Plan communicative actions to affect participants' mental states ("Theory of Mind")

- See Cohen & Perrault, "Elements of a plan-based theory of speech acts", *Cog Sci*, 1979
Perrault & Allen, "A plan-based analysis of indirect speech acts", *Comp Ling*, 1980
Cohen & Levesque, "Intention is Choice plus Commitment", *AIJ*, 1990

- Recognize how utterances/speech acts fit into a plan

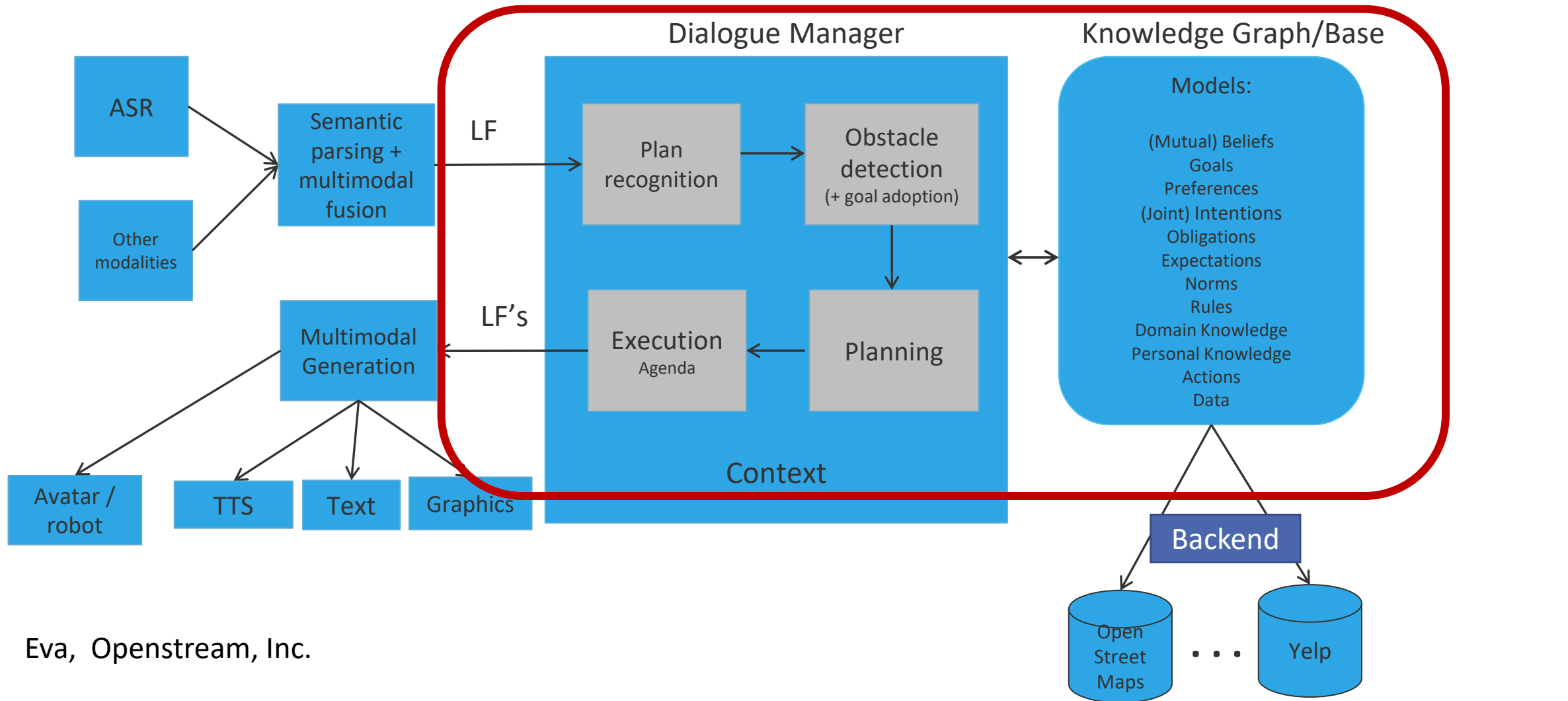
- Represent communicative and domain acts as operators in planning system

- Collaborative

- Joint Commitment/Shared Plans (Cohen/Levesque, 1991; Grosz/Sidner, 1986/90; Levesque et al., 1990)
- Leads to helping where needed, to tell partner when joint goal is achieved or impossible
- Dialogue as Joint Action (Allen et al., 2001; Clark, 1996; C&L, 1991/94; Ferguson/Allen, 2005; Grosz/Sidner, 1986/90; Rich/Sidner, 1997/8; Subramanian et al., 2006), ...
- Cogent Dialogue Shell Galescu et al. (2018)
- Others...

Collaborative Dialogue

Via Plan-based Dialogue Management



Eva, Openstream, Inc.

Plans

Plans as complex mental structures (vs data structures) whose elements are:

Formulas, which can be first-order logic formulas, or logical modality operators applied to Formulas.

- *Modalities*: Belief (*bel*), Persistent Goal (*pgoal*), and *Intend* (<action>).

- *Attitudes towards Actions* --

- *eg, Intend(do (Action)), pgoal(do(Action)), bel(done(Action))*

bel(mary, bel(john, pgoal (mary, do(reserve(john, vittorios, Monday, 7 pm))))).

- *“Mary believes that John thinks that Mary wants to reserve Vittorio’s for Monday at 7 pm”*

- *Relations* between actions and formulas, such as *Achieves, Enables*

Collaborative Plan-based Dialogue Demo

- Constraints
- Mixed initiative
- Multi-agent/person
- Multi-utterance*
- Implicit slot-filling
- Explanation
- Obstacle detection
- Use of context

*Separate demo

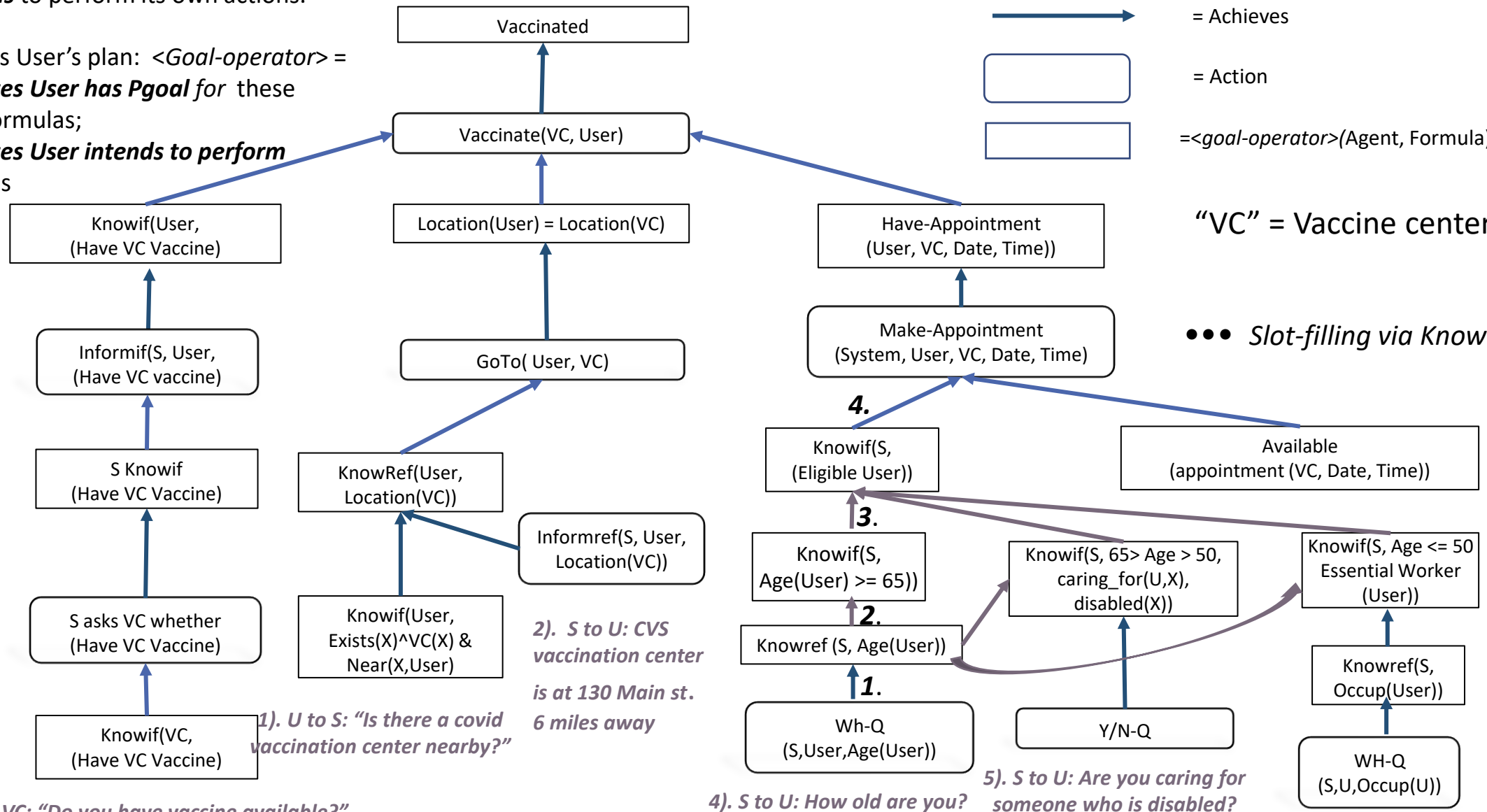
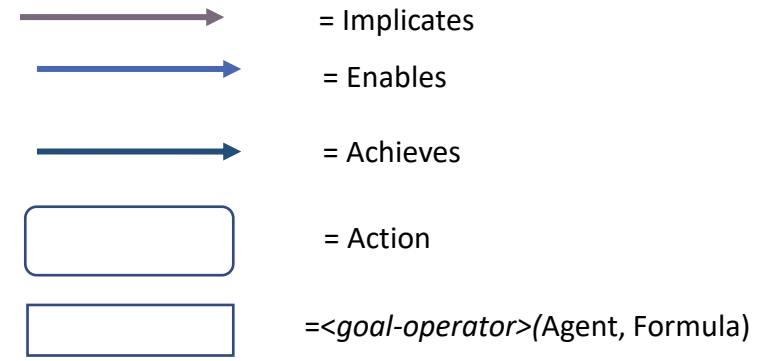
Partial Plan

When S plans: $\langle \text{Goal-operator} \rangle =$

- System has **Pgoal** for these actions and formulas;
- System **intends** to perform its own actions.

When S recognizes User's plan: $\langle \text{Goal-operator} \rangle =$

- **System believes User has Pgoal** for these actions and formulas;
- **System believes User intends to perform** his/her actions



“VC” = Vaccine center

••• Slot-filling via Knowrefs

1). U to S: “Is there a covid vaccination center nearby?”

2). S to U: CVS vaccination center is at 130 Main st. 6 miles away

3). S to VC: “Do you have vaccine available?”

4). S to U: How old are you?

5). S to U: Are you caring for someone who is disabled?

6). S to U: What is your occupation?

Simple Planning

Backward chaining

- *Effect* \rightarrow *Action*, *Action* \rightarrow *Precond*, (*backward chaining*)
- *Action* \rightarrow *Body* (*hierarchical planning*)

Special Rules:

- *Know-value* – in order to do an action with obligatory arguments, the agent needs to know what those arguments are,
- *Know-if* – if Agent wants to achieve P, Agent needs to know whether P.
- *Reverse a negative state* – If Agent is in a negative state, Agent wants to be in the corresponding positive one

Simple Plan Recognition

Plan Recognition Rules – forward chaining from action, (Allen & Perrault, 1980)

- *Precond* \rightarrow *Action*, *Action* \rightarrow *Effect*, *Body* \rightarrow *Action* (plan parsing)
- *KnowPos*, *KnowNeg* (if you want to know whether P, you may want P to be true/false)

Special Rules

- *Normal Activity* (people want to go places to do the normal activities there),
- *Neg State* \rightarrow *Positive state*.

(If you are in a neg state (e.g., car is broken), you want to be in corresponding positive state (car is fixed))

- *Knowif-exists*: $Knowif(Exists(X)^P(X)) \rightarrow Knowref(X^P(X))$
-

Knowref

- Knowref – agent knows the referent of the description $X^p(X)$ (^ means such that).
 - E.g., John knows Mary’s phone number – ***knowref(john, Ph^phone_number(mary, Ph))***
 - Is a “quantified-in” expression ***$\exists(Ph)$ bel(john, phone_number(mary, Ph))***
 - ***Stronger than John knows Mary has a phone number***
- Can quantify into other mental state operators too:
 - ***$\exists X\#day^p$ goal(john, do(john, reserve(john, Vittorio#restaurant, X#day)))*** --
meaning there is a day that john wants to reserve Vittorio’s restaurant
 - ***knowref(mary, variable:X#day, pred:pgoal(john, do(john, reserve(john, Vittorio#restaurant, X#day))))***
meaning Mary knows the day that john wants to reserve Vittorio’s
 - ***pgoal(mary, knowref(mary, variable:X#day, pred:pgoal(john, do(mary, reserve(john, Vittorio#restaurant, X#day))))***
meaning Mary wants to know the day that John wants Mary to reserve Vittorio’s for John

Speech Acts

- Inform, offer, request, suggest, confirm, informref, informif, whq, ynq,
- Apologize, propose, accept, recommend, greet, ...
- Reason about speech acts just like physical acts

Example: Wh-question

Speaker asks a **Whq** to Listener about the referent of Var such that Pred

“ \wedge ”

Precond: Listener knows the referent of Var \wedge Pred

Mary's phone number

$Ph^{\wedge}phone_number(mary, Ph)$

Effect: Speaker knows the referent of Var \wedge Pred

(Note: Effect doesn't become true by doing the speech act;

Speaker has pgoal that it becomes true and Listener believes that)

Ask someone whom you think knows the answer! \rightarrow Multi-person/agent dialogue

Example of slot-filling question

What time do you want to make an appointment?

System asks a **Whq** to *Usr* about *the time Usr wants to make an appointment*



Slot – quantifying-into a goal

Precond: Listener knows the referent of *the time Usr wants to make appointment*

Effect: Speaker knows the referent of *the time Usr wants to make appointment*

Dialogue occurs when this question is constructed from the Action definitions (e.g., Make-appointment), via Know-val inference during the planning process, and then the plan is executed.

Explanation

- Major issue for AI systems,
 - especially those based on Deep Learning
- “*Why did you say that?*”
- Plan-based systems are *inherently* explainable
- Actions achieve goals, which enable other actions
- Explain an action *relative to the plan* itself (cf Winograd, 1972)
- Example: Why did system ask “4). how old are you?”
- System asked 4) in order to determine if
 - User is eligible for a covid vaccination, which enables
 - System’s making an appointment for User.

Advantages

(see also Cogent system (Galescu et al., 2018))

- ***Principled approach to dialogue***
 - Principled basis for assisting a user by inferring/debugging plans
 - Principled approach to multi-agent/multi-person interaction
- ***Don't "design/script dialogues"***
 - Specify the domain actions and knowledge
 - Speech act interpretation and planning/plan recognition/inference do the rest
 - System reasons about what to say, and why user said what s/he did
- ***Domain-independent reasoning process applied to domain dependent data***
 - Enough reasoning for advances in commercial state-of-the-art
- ***Explanation***
 - System has a plan for everything it says/does; system knows what it is doing
 - Explanation is derived from the plan
- ***Factors the variability***
 - More in the ASR/NLP, less in the Dialogue Manager
 - Of course, need to have better NLU/NLG, for which we use NNs
- ***Basis for choosing emotional expressions for avatars (see Gratch and Marsella, 2004, ...)***

Concluding Remarks

- Principled approach to dialogue
- Virtual Assistant System that actually *assists* its user
- Explainable
- Factors variability
- Basis for multiparty communication
- Basis for multimodal communication