Architectures for Spoken Dialogue Systems/ Dialogue Management

## Spoken Dialogue Systems



### Which has the hardest job? Why?

- ASR recognize the words the user spoke
- NLP recognize the meaning of the user's utterance
- DM decide what to answer
- NLG formulate the answer in natural languuage
- TTS speak the answer clearly

### VXML: Strengths

- Simple, straightforward format
- Modelled on HTML, known technology
- Makes design/deployment of simple dialogue systems possible for developers with little/no background in ASR/NL
- Audio server can be hosted remotely, taking away further complication

### VXML: Weaknesses

- State-based dialogue systems inherently limiting
- Underlying technologies typically also limited
  - Simple grammars—isolated words/phrases
  - Grammar-based ASR
  - Limited capabilities for language generation
  - Limited logic/reasoning for dialogue

What else might a spoken dialogue system need to account for?

### Input from the Audio Server

## If barge-in is enabled, how is truncated input interpreted:

User: I'm interested in Thai restaurants in North London. System: *I know of 8 Thai rest*-User: Wait, that's not what I wanted.

User: I'm interested in Thai restaurants in North London. System: I know of 8 Thai restaurants in North London. There's Banh Mi, Thai Palace, Gold-User: Wait, that's the one I wanted.

### Input from ASR

## Can dialogue state constrain recognition choice?

User: I'm going to Dallas on May eighteenth. System: *Okay, where are you leaving from?* User: Dulles.

User: I want to return on May twentieth. System hears:

i want to return on may twelfth i want to return on may twentieth System: So that's returning on May twelfth.

### What information does NLP use?

### Words/phrases are interpreted in context

User: I need to book a flight. System: *Okay, where are you leaving from?* User: Dulles.

### How about NLG?

 Tailor response to fit user model/current history

User: I'm interested in Thai restaurants in North London. System: I know of 8 Thai restaurants in North London. Two of them have very high food quality: Banh Mi and Golden Siam.

### Can TTS use dialogue information?

### Emphasize new/pertinent information

User: I'm interested in Thai restaurants in North London. System: I know of 8 Thai restaurants in North London. Two of them have very high food quality: Banh Mi and Golden Siam.

User: Actually, what about Chinese restaurants. System: Okay, <u>Chinese</u> restaurants in North London.

### What else goes into SDS?

- Meta-level responses
  - Dynamically generated help messages based on current state of dialogue/input/backend data
  - Summary descriptions of backend data
- Fallback mechanisms
  - Descriptive responses when user query results in NULL output from database
- Complex reasoning about domain/database
  - Intelligent ordering of database tuples
  - Incorporation of user preferences
  - Analysis of backend data in light of dialogue context

### What else goes into SDS?

 Global data stores for reprocessing of system output across multiple turns
 Multimodal capabilities (ongoing work)
 Multilingual capabilities

Learning

### Morale: there's a lot to think about

- Dialogue systems involve individually complex components
- Dialogue systems involve complex interactions among these individually complex components
- Dialogue systems are becoming ubiquitous
- The model for most people is humanhuman interaction

## Considerations for dialogue manager

#### Prototyping

- How easy is it to get a 0<sup>th</sup> order iteration up and running?
- What modules are included?
- Are I/O specs standardized/easy to understand?
- How easy is it to expand the system?
- Are modules black boxes?

#### Robustness

- Are fallback mechanisms implemented?
- Is there error catching?

## Considerations for dialogue manager

### Expertise required

- Is there a separate scripting language?
- How is basic functionality (i.e., ASR, response generation) expanded?
- How much computational linguistics, acoustic phonetics, signal processing, UI design is needed?

### Approaches to building more complex SDS

- Architectures:
  - Information State Update model (University of Edinburgh)
  - Galaxy Communicator (MIT)
- Dialogue Management schemes:
  - Information State Update approach (University of Edinburgh)
  - Data-driven (MIT)

### Commonalities among "advanced" architectures

- Unify sets of software servers/agents, each performing different task
- Control flow of information among servers
- Are rule-based at some level
- Have stores for global variables

# Design considerations for research architectures

- Sequential rules vs. blackboardUnification of all HLT servers
- Common IO specs
- Plug-and-play

### **Open-Agent Architecture**

- Allows integration of software agents for prototyping dialogue system
- Agents conform to conventions of framework
- Use common language for communication
- "Facilitator" mediates interaction among agents
- Facilitator maintains ordering constraints implicitly

### **Information-State Update Model**

#### Core: Dialogue Move Engine

- Receives input from other agents (e.g., ASR)
- Updates internal state to reflect new information
- Calls other agents (e.g., TTS)

#### Declarative representation of dialogue modelling

- Specification of contents of dialogue
- Datatypes for information state
- Update rules for dealing with dynamic information
- Control strategy

# DIPPER: an implementation of the ISU model

 Update language independent of any particular programming language

 Incorporates many off-the-shelf OAA agents

## Galaxy Communicator

- Sequential rules
- Configuration specifically aimed at spoken dialogue systems
- Multiple servers interacting with one central hub

### **Basic components**

#### Hub

- Keeps track of global state
- Mediates interaction among servers
- Controls logging, global parameters
- Servers
  - Stateless
  - Connect to hub via control file
- Token
  - Global store for attributes
  - Unless otherwise specified, attributes disappear with new turn



Control Strategy
<ul> <li>A set of ordered rules is a "program"</li> <li>Simple syntax supports boolean and arithmetic tests applied to hub variables</li> </ul>
<ul> <li>All rules that apply are simultaneously executed</li> <li>Relevant input variables are packaged into a frame and sort to target sonyor</li> </ul>
<ul> <li>Frame is queued by hub when target server is busy</li> <li>Each program has a separate name</li> </ul>
<ul> <li>The "main" program controls processing for user queries</li> </ul>
<ul> <li>Other programs control module-to-module sub- dialogues and asynchronous I/O</li> </ul>

## Control Strategy (cont'd)

- Upon start-up, hub sends a "welcome" frame to each server
  - Server-specific initializations
- Hub polls continuously for new inputs or replies
- New inputs generate new tokens
- Tokens are processed according to program rules
- Replies modify existing tokens
- Tokens destroyed when no further rules apply
- Multiple users are managed via distinct sessions
  - Retain state for user's dialogue; e.g., language, domain, discourse context, etc.

### Sample rule

RULE:	:parseFrame & !:requestFrame → contextTracking
<b>RETRIEVE:</b>	:historyFrame
IN:	:parseFrame
OUT:	:requestFrame :historyFrame :domain
STORE: :historyFrame	
24 4 9 3 5	A PROPERTY OF THE PROPERTY OF

- Boolean tests on attributes in global token
- IN and OUT keys specify specific attributes to retrieve from and store in token

RETRIEVE and STORE used for attributes in global store

- Rules can also
  - Specify logging parameters (both into and out of operation)
  - Rename variables

### Turn Management (the heart of Dialogue Management)

- Phases of turn management
- Making turn management domain independent
- Making turn management data-driven
  - Using data to determine what to say
  - Using data to determine concepts

# Roles of dialogue management in information retrieval domains

- Resolve ambiguities
  - Ambiguous input constraint (e.g. Miami, Florida or Miami, Ohio)
  - Pragmatic considerations (e.g., too many flights to speak)
- Inform and guide user
  - Suggest subsequent sub-goals (e.g., what time?)
  - Offer dialogue-context dependent assistance upon request
  - Provide plausible alternatives if requested information unavailable
  - Initiate clarification sub-dialogues for confirmation
- Influence other system components
  - Adjust language model due to dialogue context
  - Adjust discourse history due to pragmatics: "Christmas" = Dec25
  - Set up context for system initiative: "where to?" = destination







