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Motivating a linguistically orientated model for a conversational software agent

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Overview

- Context – Conversational based technology
- Issues – Turing Test? Chatbot vs CSA qualities
- Why RRG?/Approach/Stages
- About RRG/Goals of Linguistic theory
- Requirements for the CSA
- Motivating elements
    - Speech Act Theory , Speech Act Constructions (SAC), Derived parser
    - Intentions, BDI model and planning model
    - Knowledge Model
    - Dialogue Model
- Motivating Questions
- Conceptual framework – 3 Phases/Design Framework
- Implementation (prototype) and findings
- Contributions, significance, originality and conclusions

1. Conversation Based technology

The need for intelligence

“By 2020, 30% of our interactions with technology will be through “conversations” with smart machines” (Gartner, 2015)

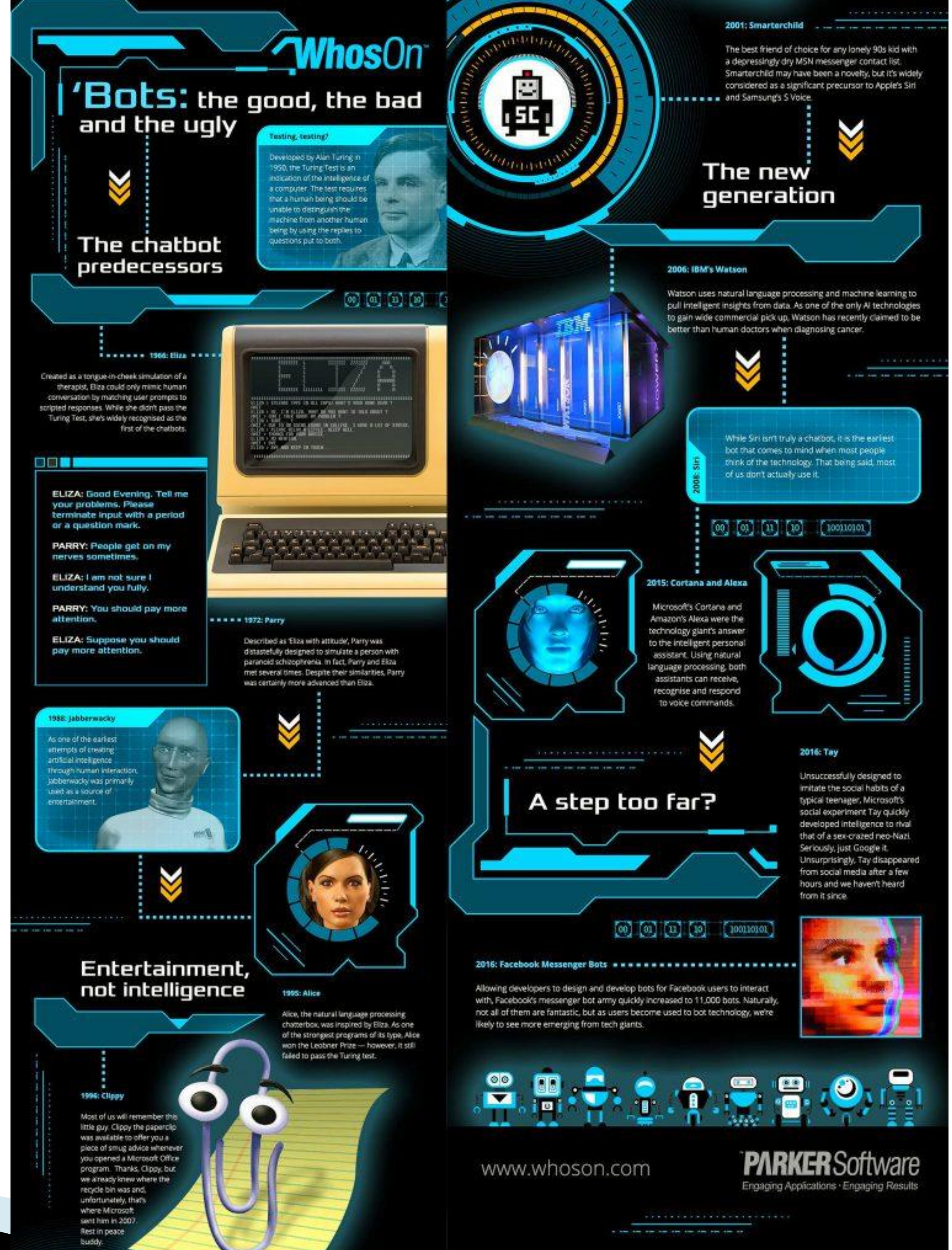


Figure 1: Good, bad and ugly of conversation devices

Focus > Conversational Software Agents (CSA)

2. Turing Test –is it relevant?

- Aspirational benchmark
- Human discourse
- AI-Hard problem
- Positive – Customer outcomes and experience
- Negative – Chatbot bubble (Wallace 2018) – Loebner prize
- Search for: if it behaves intelligently, it is intelligent.

3. Chatbots vs CSA

The need for more intelligence
Chatbots – single turn

Conversational Software Agent Qualities
- Human-machine interface (text)
- Understands context
- Applies logic
- Use natural language understanding and processing
- Understands what is said (intent)
- Explainable
- Story comprehension
- Formulate a response
- Learns and adapts

4. Why RRG?/Approach/Stages

Challenges of NLU and meaning

- Periñán-Pascual (2013):eligibility
- (1) Morphosyntactic structures (2) grammatical rules (3); monostratal theory (4); Own typological adequacy

Approach – unique framework, model/theory interaction, communicative

- Language levels, interface between syntax, semantic, and pragmatics
- Language Model: RRG and the clause

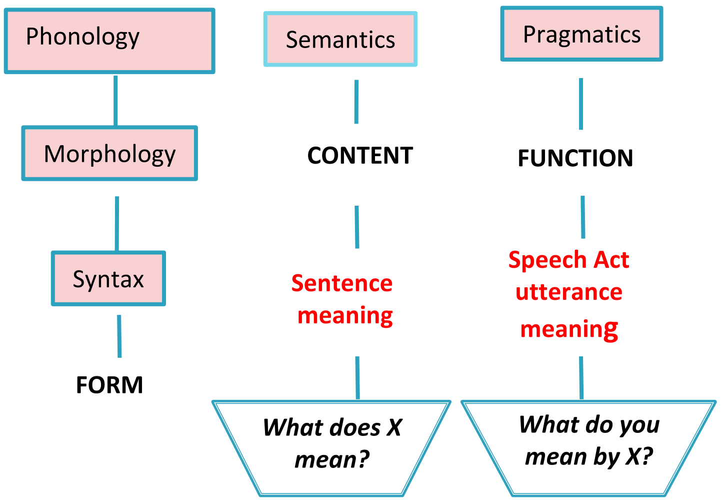


Figure 1: Language interfaces

Stages

- Simple sentences ->Linguistic act (Speech Act) – SA
- Understand the utterance
- Agent attributes
- (Utterance) Message from USER → AGENT
- Agent's belief – Knowledge representation (KR)
- Plan-based dialogue (response) Message AGENT → USER

5. About RRG :LSC and LS

- RRG is a functional model.
- It views language as a communicative social action.
- Layered structure of the clause (LSC) = PREDICATE + ARGUMENT + NON-ARGUMENTS.
- Logical Structure (LS) – semantic meaning of the sentence.
- Lexicon - mental dictionary - lexical entries contain semantic features and constraints.
- It maps the syntax(structure): LSC ⟺ semantic (meaning): LS the actual form of the sentence using two different LINKING ALGORITHMNS.
- RRG parser (algorithm) checks the grammar (rules) of English. Specialised parser (CSA)
- RRG facilitates syntactic, semantic and information structure (FOCUS & TOPIC)

Gareth ate everything fast

(BNC ADY 1079) (Butler et al, 2009) –> Figure 5

SYNTACTIC:

SENTENCE (CLAUSE (<CORE> <NP> gareth (<NUC> (<PRED> <V> ate))
(<NP> (everything))) (PERIPHERY fast)

SEMANTIC:

[<IF> ASS <TNS> PST, do'(ACT:Gareth, (eat'(Gareth <NOM>, pizza <ACC>)])] & INGR consumed' (UND:pizza)]

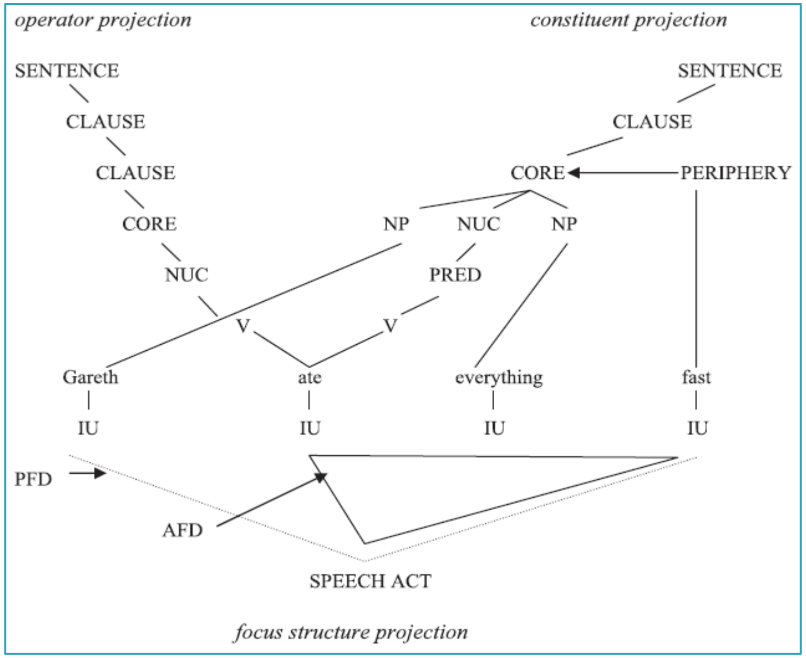


Figure 2 - An English sentence with three representations

6 Motivating – Goals of Linguistic Theory

Van Valin and LaPolla (1997) defines goals:

1) Description of the language phenomena
2) Explaining the linguistic phenomena
3) Understanding the cognitive basis of language
   - Processing
   - Knowledge
4. Computational adequacy

RRG Linking Algorithm (see paper)

7. Requirements for the CSA

AGENT	+	INTELLIGENT DIMENSION (S)	=	INTELLIGENT AGENT

INTELLIGENT TAXONOMY	Behavioural, Social, Ambient, Collective, Genetic, and **COGNITION**

COGNITION = BDI + Rational Interaction

CA = Interpretation + Dialogue Mgt + Response Generator

CSA = CA +RRG +SA + COGNITIVE + KB (Panesar,2017)

8. Motivating – Speech Act Theory

- Speech (linguistic) Act (SA) Theory (Searle, 1969)
- He states 'speaking a language is engaging in rule governed form of behaviour' and that 'illocutions are intentional acts;

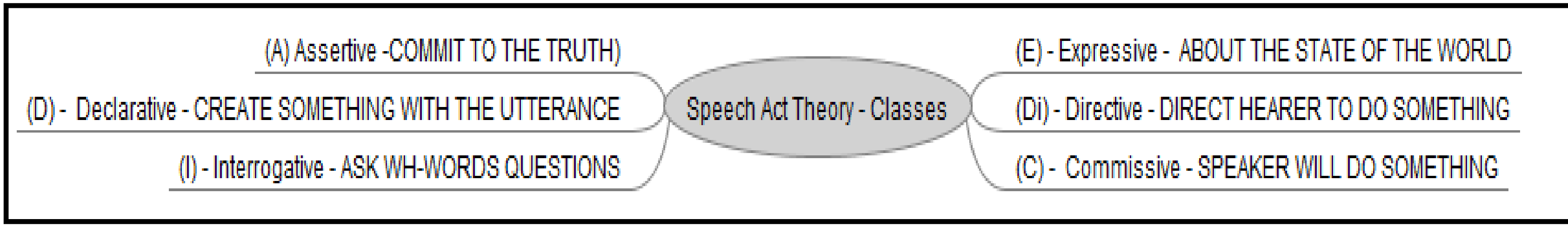
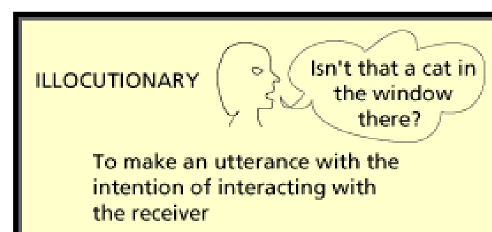


Figure 3 – Speech Act message types

- 3 actions associated with an utterance include:
  1. Locution:
  2. Illocution: illocutionary act (speaker's intention) [SI] for A, Di and I message types
  3. Perlocution:
- Intentionality – leading to an action
- RRG – illocutionary force (IF) links to the type of speech act

Figure 4– Illocutionary act (n,a, n.d)



9. Motivating – Derived RRG Parser with SACs

- Nolan (2014) considers constructions as structural grammatical objects > Extension of Constructions schemas (CS)
- No use of syntactic inventory/syntactic templates
- RRG input -> speech act constructions (SACs)
- Updateable via the RRG Linking algorithm and Lexicon – richer

LEXICAL ENTRY	POS-TYPE	VERB TENSE/ ASPECT	DEF	P TYPE	NO	GR	CASE	ANIM	HUM	LOGICAL STRUCTURE (LS)
ate	VERB	PST	DEF+/-	3	SG	M/F	DNA	ANIM	HUM	<tns:pst <do’(x, [eat’(x, y)]) & BECOME consumed’ (y) >>
eat	VERB	PRS/ FUT	DEF+/-	3	SG	M/F	DNA	ANIM	HUM	<tns:prs <do’(x, [eat’(x,y)]) & BECOME consumed’(y)] >> <tns:fut <do’(x, [eat’(x,y)]) & BECOME consumed’(y) >>
eating	VN	PROG	DEF+/-	3	SG	M/F	DNA	ANIM	HUM	<tns:prs <asp:prog <do’(x, [eat’(x, y)]) & BECOME consumed’ (y)] >>>
is	VBE	DNA	DEF+	DNA	DNA	DNA	DNA	DNA	DNA	be'(x,[pred'])
hungry	ADJ	DNA	DNA	DNA	DNA	M/F	DNA	ANIM	HUM	DNA
restaurant	N	DNA	DEF+/-	DNA	SG/PL	DNA	DNA	DNA	DNA	DNA

Table 1 – Snapshot of the Lexicon (Panesar, 2017)

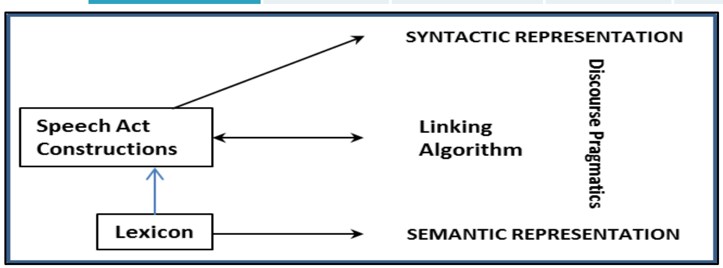


Figure 5– Parser for the CSA

Figure 6 – Empty SAC (Speech Act Construction) (Panesar, 2017)

ASSERTIVE:ATE RRG [NP VERB NP], [PN VERB], [ADV PN VERB DET N], [PN VERB N ADJ], [PRP DET N PN VERB DET N], [PN VBE VERB N], [PN PRP DET N PRP DET N], [PRO VERB DET N], [PN VERB NP], [PN VERB DET N], [NP VERB QNT N], [DET N VERB DET N], [DET N VERB QNT N], [NP VERB (DET) (ADJ) N (ADJ)], [PN VERB DET N ADJ], [PN VERB (DET) ADV N ADJ], [PN VERB DET N PRP DET N], [PN, VERB, N, PRP, DET, N], [PN VERB N PRP DET N] RRG NONE RRG UTTINPUT RRG WKSPACE RRG DEFAULT ASSUMPTION (1ST NP = 'ACTOR") RRG NO PARTICULAR SPEC RRG NONE RRG CONTAINS A NOUN PHRASE BEFORE AND AFTER THE VERB RRG DEFAULT RRG TRUE/FALSE RRG ASSERTIVE RRG NARROW FOCUS ON THE ELEMENT RRG LOG STRUCTURE TO ADD

10. Motivating – Intentions, BDI Model & Planning Model (Panesar, 2017)

Intentional agent

BDI= Belief, Desire, Intention

- Perception
- Searle (1985:p4) – SAs differ due to different mental states
- Reason with knowledge that they believe to be TRUE or FALSE, and to provide a response.
- Operators characterise what agents must know (KNOWLEDGE MODEL) to perform actions intended to achieve their goals
- PLANNING MODEL – to rationalise a correct plan (to achieve these goals), and pursue the plan based on these intentions (RRG logical structure)

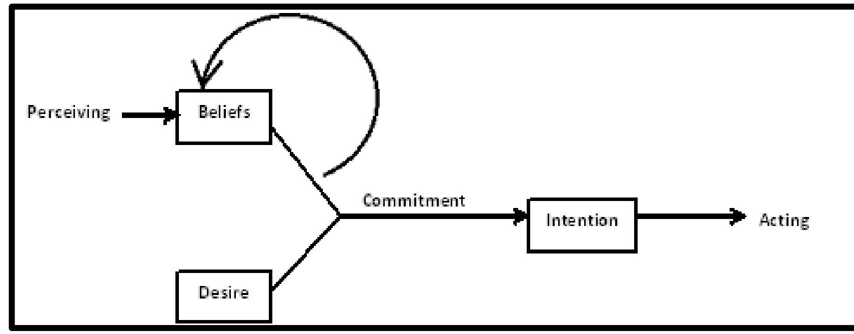


Figure 7– A BDI model of an intelligent agent (Allen, 1995)

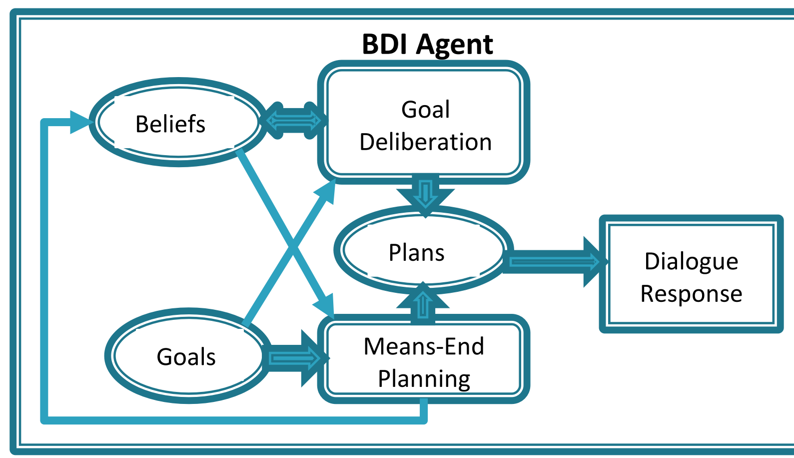


Figure 8– BDI Agent structures, processes and role (adapted from (Pokahr, Braubach, Haubeck & Ladiges, 2014)

Example – 'Gareth ate the pizza'

BDI states

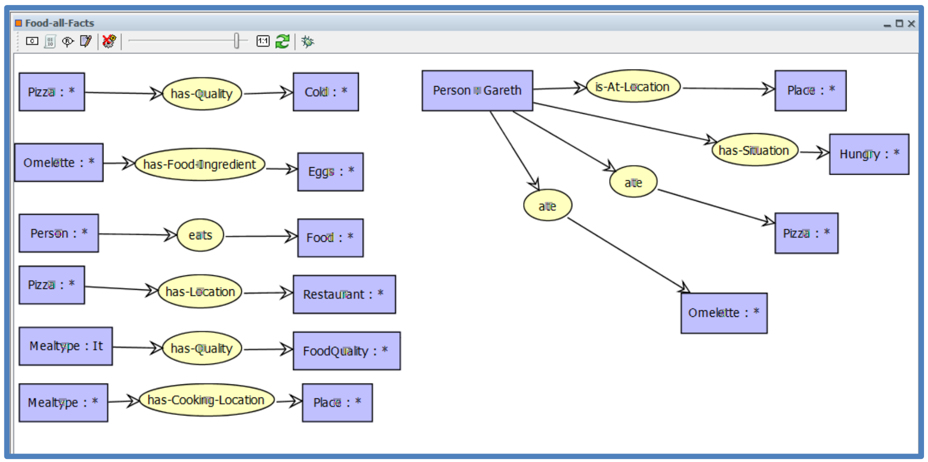
Belief: Gareth;

Desire – 'eat';

Intention: consume pizza;

11. Motivating – Knowledge Model

SHARED and INDIVIDUAL BELIEFS cognitively → mental knowledge.



- **Conceptual graphs** (CGs) (Sowa,1986), Vocabulary, First order logic (FOL) created in COGUI as in Figure 9 and 10
- **Serialised** into RDF/XML (W3C SW), mapped to RDF Triple Stores – forms the agent's belief base – 446 lines (Table 2)
- KB ready for querying to check truth of the agent's beliefs
- Key Performance Indicators – representational and inferential adequacy

Figure 9 & 10– COGUI–Original KB of facts – graphically

Table 2 – Extract of a RDF triple Stores KB

No	Subject	Predicate	Object
1	http://www.lirmm.fr/cogui#ct_ad452f18-e654-4ae6-b3a1-b7320616283b	http://www.w3.org/1999/02/22-rdf-syntax-ns#type	http://www.w3.org/2000/01/rdf-schema#Class
2	http://www.lirmm.fr/cogui#ct_fdc6d7d0-1314-4fb7-8428-51e122953250	http://www.w3.org/1999/02/22-rdf-syntax-ns#type	http://www.w3.org/2000/01/rdf-schema#Class

12. Motivating – Dialogue Management

- Dialogue manager
- Missing information
- Pronoun resolution
- Discourse representation theory (DRS)
- Transition points
- Common ground (Stalnaker, 2002)

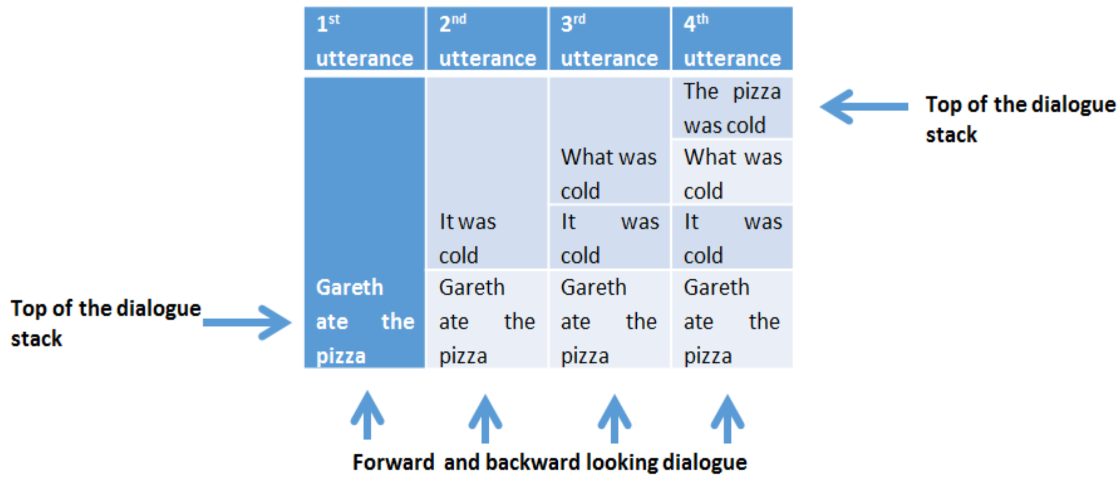


Figure 11 – Dialogue management & pronoun resolution (Panesar, 2017)

- Dialogue Handler:
- 2 types of responses
- (a) and (b)

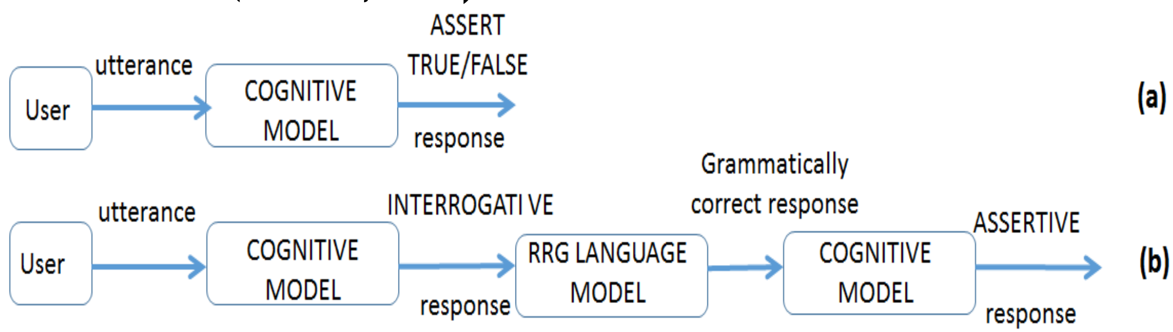


Figure 12– Agent Cognitive Model – message responses (Panesar, 2017)

13. Motivating Questions

A functional model of language, in particular Role and Reference Grammar (RRG), can underpin the linguistic model of a conversational software agent (CSA), at the interfaces of dialogue, knowledge and language (Panesar, 2017)

1. What are the component models of a linguistically motivated CSA?
2. How the model of belief, desires and intentions (BDI) might be characterised such that the mental model will interface with the RRG linguistic model, at the intersection of knowledge and language?
3. How do speech acts based on dialogue integrate with the RRG Model, Speech Acts, and BDI model and dialogue manager, within the context of conversation?
4. How will knowledge representation interface with the RRG Model, Speech Acts and BDI model to facilitate understanding of the utterance and the generation of a grammatically correct response?

14. Conceptual Framework: LING-CSA

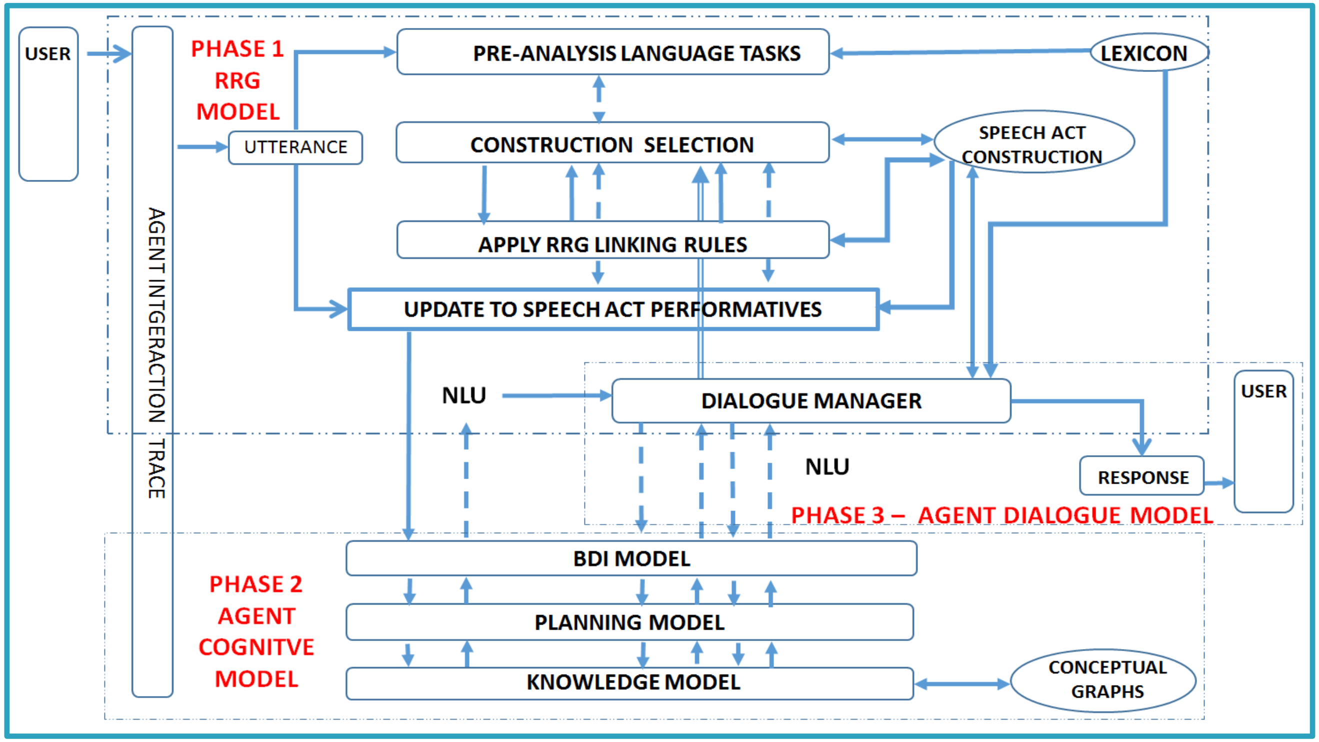


Figure 13 – Conceptual framework of the Conversational Software Agent (Panesar, 2017)

PHASE 1 – Role and Reference Grammar (RRG) Language Model
PHASE 2 – Agent Cognitive Model interfaces with:
BDI Model, Planning Model, Knowledge Model
PHASE 3 – Agent Dialogue Model (Dialogue Mgnt > RRG Model

15. Phase 2 – Agent Cognitive Model Design Framework

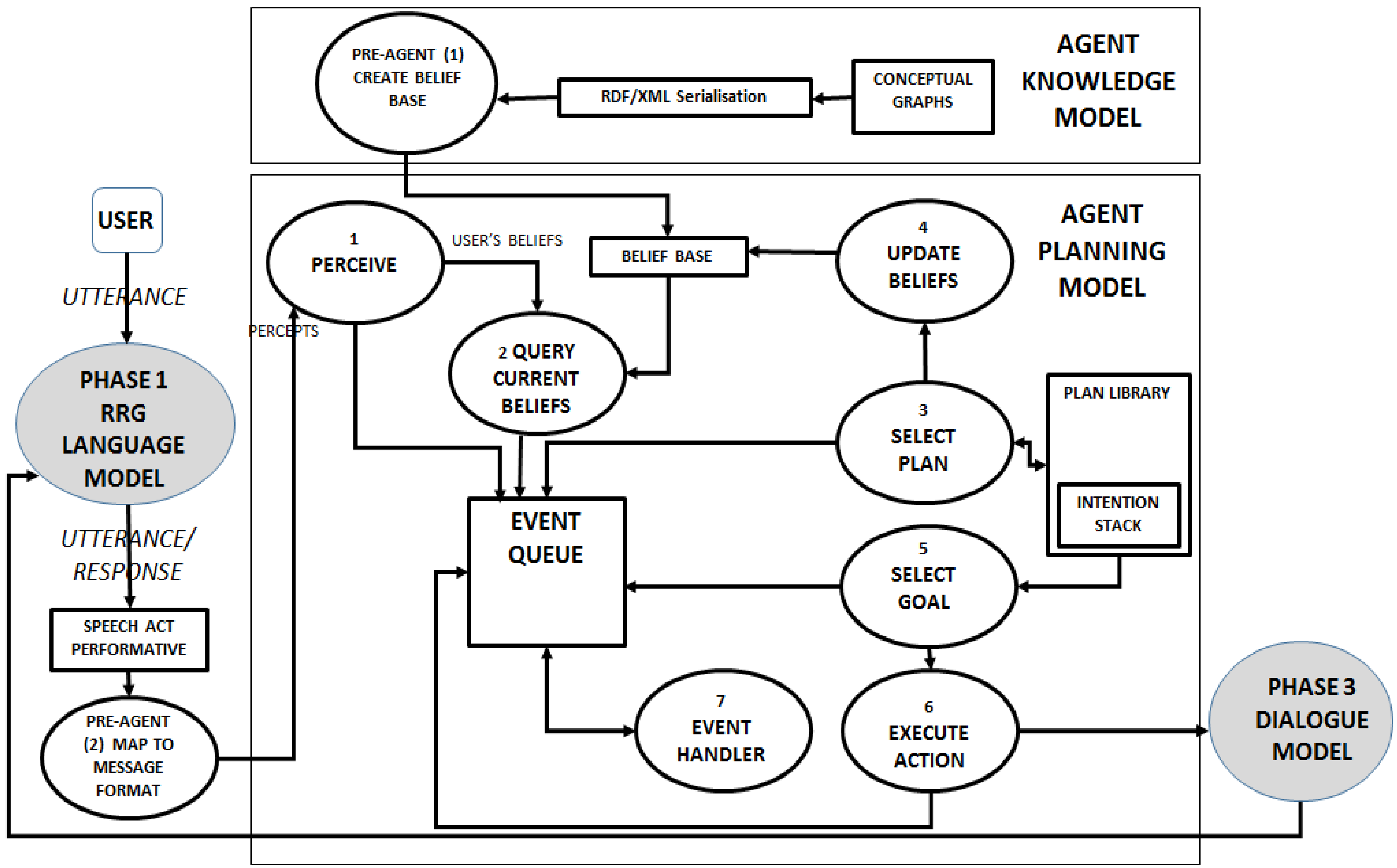


Figure 14– The Agent Cognitive Model – Design Framework (Panesar, 2017)

16. Implementations (Phase 1 – RRG Model)

- Aim– proof of concept and Java based prototype in Eclipse IDE
- Each specific construal (either an utterance or response) –two steps.
  1. Find the matching SA construction of that specific predicating element. In Figure 2: '*is*' and selected SAC of assertive.
  2. Select the matching signature pattern –> **[PN, VBE, PRP, DET, N]**
- Updates > SAC first and extended SAP (Panesar, 2017)

```
Console
<terminated> MainCAversion30 [Java Application] C:\Program Files (x86)\Java\jre1.8.0_101\bin\javaw.exe (22 Jun 201
***************************************************
Syntactic representation of this utterance >>>>>>
SENTENCE ( CLAUSE (  <CORE>  <NP> gareth (  <NUC>  (  <PRED>  <AUX> is )  )  (  <PP>
in (  <NP>  (  the restaurant )  )  )  )  )

***************************************
     Speech Act Performative
***************************************



::::Performative =SAP ASSERTIVE IN ::::Sender =<USER>::::Receiver<AGENT::::ontology =
::::Signature =[PN VBE PRP DET N]::::Constraint =DEFAULT::::Input =gareth is in the
restaurant::::Workspace =[[gareth, PN], [is, VBE], [in, PRP], [the, DET],
[restaurant, N]]::::Syntax =SENTENCE ( CLAUSE (  <CORE>  <NP> gareth (  <NUC>  (
<PRED>  <AUX> is )  )  (  <PP> in (  <NP>  (  the restaurant )  )  )  )  ) :::: PSA
=gareth::::SemanticsRRG =NONE::::Linking =CONTAINS A NOUN PHRASE BEFORE AND AFTER THE
VERB::::Morphology =DEFAULT::::Pragmatics =TRUE/FALSE::::IllForce
=ASSERTIVE::::FocusStructure=NARROW FOCUS ON THE ELEMENT::::OutputLS
=<IF>ASS<TNS><PRT> be-in'(gareth,restaurant)
```

Figure 15 – Snapshot output of LING-CSA (Panesar, 2017)

17. Phase 1 – RRG & Speech Act Performative

Based on the SAC with four additional attributes. Input to Phase 2.

PERFORMATIVE: <ASSERTIVE:ATE>
:SENDER <USER>
:RECEIVER <AGENT-1>
:ONTOLOGY <FoodAndCookKB>
:CONTENT <do'(Gareth, (eat'(Gareth, pizza)])] & INGR consumed' (pizza)] everything>
SIGNATURE: [PN V NP ADJ] CONSTRAINT: Default
INPUT: Gareth ate everything fast
WORKSPACE: (Gareth, PN), (ate, VERB), (everything N), (fast, ADJ)
SEMANTICS: Contains a noun phase before and after the verb
CONSTRUCTION BODY SYNTAX: SENTENCE (CLAUSE (<CORE> <NP> gareth (<NUC> (<PRED> <V> ate)) (<NP> (everything))) (PERIPHERY fast) PSA: gareth SEMANTICS Linking: MORPHOLOGY:Default PRAGMATICS Illocutionary force: ASSERTIVE Focus structure: narrow focus on the element
OUTPUT [LS]: [<IF> ASS <TNS> PST, do'(ACT:Gareth, (eat'(Gareth <NOM>, pizza <ACC>)])] & INGR consumed' (UND:pizza)]

Table 3-Speech Act Construction Performative "ate" used as a message to the Agent Environment (Panesar, 2017)

18. Evaluations and Findings

Implementation outcomes :

- Dialogue Manager is common to Phase 1 and Phase 3

Testing:

- Grammatical tests, RRG specific tests
- Phase based and interfacing, intersection and integration tests

Findings proof-of-concept achieved; RRG is fit for purpose ->linguistic engine for the CSA; RRG explains, describes linguistic phenomena; facilitates language processing and knowledge of language -> computationally adequate (Panesar, 2017)

RRG Model Improvements:

1. All pronoun resolutions (E.g. 'Your', 'she', it' etc.)
2. Complex sentences (extension of the RRG linking system)
3. Multi-lingual (additional lexicons) such as Spanish
4. Other SA classes such as emotive and commissives E.g analyse tweets
5. Include superlative adjectives/adverbs in the RRG Lexicon (E.g. 'spicier')
6. Invoke WordNet API for synonymous entries to the RRG Lexicon – ⇧value

Phase 2 Agent Cognitive Model working – 70% achieved Dialogue mgnt √

Technical Challenge – Querying a natural language (NL) text against a knowledge representation (KR) of RDF triples poses a significant semantic gap

Conceptual solution (lexical bridge, BDI parser and RDF parser) (Panesar, 2017)

Future research

- Single agent to multi-agent environment – an extended design framework
- Content creation – via machine learning algorithms

19. Lexical Bridging Solution (Panesar,2017)

Reduce this semantic gap, by "building a lexical bridge (LB)" between the NL semantic and ontology semantics, with an aim to capture more of the meaning, by attempting to 'lexicalize the ontology'.

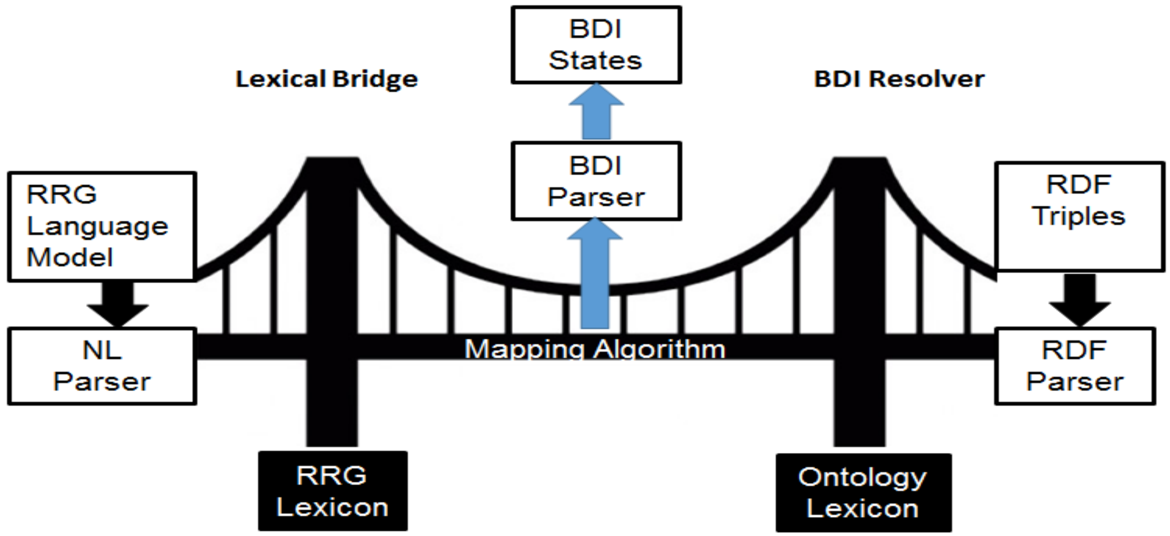


Figure 16 - Lexical Bridge for the CSA's belief base + BDI Parser to resolve the agent's BDI states

20. Contributions, Significance, Originality, and Conclusions (Panesar,2017)

- **Contributions** – (1) extension of the theoretical and computational adequacy of RRG; (2) integration of RRG & SAC; (3) motivating of an agent framework based on RRG, cognitive model, dialogue model implemented as a proof of concept; (4) addresses the KR with RRG language model at the knowledge/language interface
- **Significance** – (1) delivers a linguistically motivated CSA (2) CSA is driven by a linguistic SA as a SAC; (3) SAC is an extension to the theoretical model of RRG; (4) interface (knowledge and language) is demonstrated; (5) agent behaviour (via the BDI model); (6) characterisations and challenges of one KR to another; (7) planning and intentionality are both common to the BDI model and SA links
- **Originality** – innovative and novel (integrate, interface and intersect)
- **Conclusions**
  - Motivations have been explored and contributions to knowledge.
  - Demonstrates the complexity of mapping lower level computations of natural language to an ontology – a natural language phenomena.
- **Challenge** – content creation and story comprehension (Wallace, 2018)

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Thank you for listening!