ABDUCTIVE REASONING

Abductive diagnosis

Abduction in natural language processing

Weighted abduction

Abduction in natural language generation

ABDUCTIVE DIAGNOSIS

Information

- Implication (H) effects of errors
- Atoms (S) symptoms observed
- Negated literals (N) symptoms not observed

Explanation for S, i.e., minimal errors F, such that $F \cup H \cup N$ consistent and $F \cup H \models S$

Example

H: measels → fever & rash
 migrane → headache & nausea
 influenza → headache & sorelimbs & fever

N:	empty	S:	fever $\rightarrow 2$ explanations: measles, influenza
			fever, rash \rightarrow 1 explanation: measles
			headache \rightarrow 2 explanations: migrane, influenza
			headache, fever \rightarrow 1 explanation: influenza
N:	¬nausea	S:	headache \rightarrow 1 explanation: influenza
			headache, rash \rightarrow 1 explanation: measles & influenza

INTERPRETATION AS ABDUCTION

Sentence interpretation involves

- Prove the logical form of a sentence together with the constraints that predicates impose on their arguments, allowing for coercion
- Merging redundancies where possible
- Making assumptions where necessary

Example

The Boston office called.

- Resolving reference of "The Boston office"
- Expanding the metonymy
- Determining the implicit relation between Boston and the office

Prove abductively the expression

 $(\exists x,y,z,e) \text{ call'}(e,x) \land \text{ person}(x) \land \text{ rel}(x,y) \land \text{ office}(y) \land \text{ Boston}(z) \land \text{ nn}(z,y)$

INTERPRETATION OF "The Boston office called"



Language Technology

EXAMPLE LOCAL PRAGMATICS

"Disengaged compressor after lube-oil alarm."

To solve local pragmatics, derive the expression

```
(\existse,x,c,k<sub>1</sub>,k<sub>2</sub>,y,a,o) Past(e) ^ disengage'(e,x,c)
```

```
^ compressor(c) ^ after(k_1,k_2) ^ event(k_1)
```

```
^ rel(\mathbf{k}_1, \mathbf{y}) ^ y \in {c,e} ^ event(\mathbf{k}_2) ^ rel(\mathbf{k}_2, \mathbf{a})
```

```
^ alarm(a) ^ nn(o,a) ^ lube-oil(o)
```

Interpretation further requires

- If information for derivation is insufficient, making assumptions is required
- Assumptions are new information, with varying likelyhood, according to linguistic form
- Costs are assigned to assumptions

EXAMPLE LOCAL PRAGMATICS (2)

Costs for interpretation

- Definite NPs generally used referentially Their assumptions should be expensive (10\$) The same holds for selectional constraints
- Indefinite NPs rarely used referentially Their assumptions should be inexpensive (1\$)
- Indefinite NPs should have intermediate costs (5\$)
- Non-nominal propositions are usually new info They should have low costs (3\$)
- Coercion relations and compound nominal relations require interpretation, hence very high costs (20\$)

$$(\exists e, x, c, k_1, k_2, y, a, o) Past(e)^{\$ 3} \land disengage'(e, x, c)^{\$ 3} \land compressor(c)^{\$ 5} \land after(k_1, k_2)^{\$ 3} \land event(k_1)^{\$ 10} \land rel(k_1, y)^{\$ 20} \land y \in \{c, e\} \land event(k_2)^{\$ 10} \land rel(k_2, a)^{\$ 20} \land alarm(a)^{\$ 3} \land nn(o, a)^{\$ 20} \land lube-oil(o)^{\$ 5}$$

WEIGHTED ABDUCTION

Axioms with costs

$$\mathbf{P}_1^{w_1} \wedge \mathbf{P}_2^{w_2} \supset \mathbf{Q}$$

If Q is associated with costs c, the costs of assuming P_1 is cw1, and the costs of assuming P_2 is cw2

If w1 + w2 < 1, most-specific abduction is favored

If w1 + w2 > 1, least-specific abduction is favored

Factorization:

(∃...x,y,...) ... ^ q(x) ^ ... ^ q(y) ^ ... becomes (∃...x,...) ... ^ q(x) ^ if q(x) is associated with lower costs than q(y)

Factorization may override least-specific abduction

$$P_1^{0.6} \land P_2^{0.6} \supset Q_1$$
$$P_2^{0.6} \land P_3^{0.6} \supset Q_2$$
if we want to derive $Q_1 \land Q_2$

ISSUES IN LOCAL PRAGMATICS

Adequate semantic and interpretation requires

• Reference resolution

"The police prohibited the women from demonstrating. They feared violence"

- Compound nominal interpretation "Lube-oil alarm", "Boston office"
- Syntactic ambiguity resolution
 "Disengaged compressor after lube-oil alarm"
- Metonymy resolution

"The Boston office called"

Further issues include definite reference, distinguishing the given and the new, lexical ambiguity, discourse coherence (rhetorical relations)

APPLICATION AREAS

Source: "Interpretation as Abduction" Hobbs et al., (AI Journal)

TACITUS (The Abductive Commonsense Inference Text Understanding System) used for message routing, problem monitoring, database entry and diagnosis Applications so far:

- Equipment failure reports or casuality reports
- Naval operation reports
- Newspaper articles and similar texts on terrorist activities

Example text:

"A cargo train running from Lima to Lorohia was derailed before dawn today after hitting a dynamite charge. Inspector Eugenio Flores died in the explosion. The police reported that the incident took place past midnight in the Carahuaichi-Jaurin area.

TECHNICAL ISSUES FOR FURTHER RESEARCH

Making abduction more efficient

- Exploiting the domain type hierarchy (for filtering of axioms, axioms for incompatibility of assumptions)
- Avoiding transitivity axioms
 (limiting the depth of recursion e.g., for location containment rules)
- Reducing the branch factor of the search

 (for coercion: evaluate class predicates prior to relation predicates)
 (for factoring: avoiding it when this would lead to type violations)

Further pragmatic issues not treated yet:

Resolving quantifier scope ambiguities, metaphor interpretation, recognizing the speaker's plan

A MODERN APPROACH IN THE TACITUS STYLE

Axioms extracted from WordNet and FrameNet, evaluated on textual entailment tasks John(x1):20 & compose(e1,x1,x2):20 & sonata(x2):20 to be interpreted Suppose our knowledge base contains the following axioms:

1) form(e0,x1,x2):90 \rightarrow compose(e0,x1,x2)

2) create art(e0,x1,x2):50 & art piece(x2):40 \rightarrow compose(e0,x1,x2)

3) art piece(x1):90 \rightarrow sonata(x1)

- I1: John(x1):20 & compose(e1,x1,x2):0 & sonata(x2):0 & cost 56 form(e1,x1,x2):18 & art piece(x2):18
- I21 : John(x1):20 & compose(e1,x1,x2):0 & sonata(x2):0 & cost 56 create art(e1,x1,x2):10 & art piece(x2):8 & art piece(x2):18
- I22 : John(x1):20 & compose(e1,x1,x2):0 & sonata(x2):0 & cost 38 create art(e1,x1,x2):10 & art piece(x2):8

The "create art" meaning of compose has been brought forward because of the implicit redundancy in the sentence which facilitated the disambiguation

SYSTEM DESIGN FEATURES

Originally a clean and clear reasoning procedure, introducing optimization measures:

- limitations in time and depth of search
- limiting axioms (eg., axioms must "fit" to the input, enable merging with others)

Exploiting WordNet and FrameNet definitions as axioms

Axiom type	Source	Numb. of axioms
Lexeme-synset mappings	WN 3.0	422,000
Lexeme-synset mappings	WN 2.0	406,000
Synset relations	WN 3.0	141,000
Derivational relations	WN 3.0 (annotated)	35,000
Synset definitions	WN 2.0 (parsed, annotated)	120,500
Lexeme-frame mappings	FN 1.5	50,000
Frame relations	FN 1.5 + corpora	6,000

Interpretation and hypothesis matched

Entailment hypothesized with large cost reductions through facotrization and axiom uses

EXPRESSING CAUSAL AND TEMPORAL RELATIONS

The problem

- Both relations explicit in the knowledge base
- Left implicit on the surface, pragmatically inferable
- Mimic this to generate natural discourse
- Risky, requires adequate model of implicature
- 1) Max entered the office. John greeted him with a smile. He showed Max to the seat in front of his desk and offered him a cup of coffee.
- 2) Max entered the office. Then, John greeted him with a smile. After that, He showed Max to the seat in front of his desk. He then offered Max a cup of coffee.
- **3**) Jon switched off the heating. Judy came in and said the room was too hot.
- 4) Jon switched off the heating. Then, Judy came in and said the room was too hot.
- 1) preferred to 2), but 4) preferred to 3)

DISCOURSE CONSTRAINTS

Connection between eventualities (states or actions)

Relations C: causation, part-whole, temporal relation

• Temporal Coherence

A text is *temporally coherent* if *H* can infer that at least one of the relations *C* holds between the eventualities described in the sentences

• Temporal Reliability

A text is *temporally reliable* if one of the relations in C which H infers hold does in fact hold between the eventualities described in the sentences

Temporally *incoherent*, when no relations inferable Temporally *unreliable*, when inferable relation false Temporally *adequate*, when *reliable* and *reliable*

Implicatures calculated via default rules

Manifestations of Gricean-style pragmatic maxims and world knowledge

DISCOURSE RELATIONS (1)

- $e_1 \{ e_2 \text{ eventuality } e_1 \text{ preceeds } e_2 \}$
- $\phi > \psi$ if ϕ then normally ψ
- *Narration* {α,β} > Narration(α,β)
 If clauses α and β are discourse-related, then normally *Narration*(α,β) holds
- Axiom on Narration Narration(α,β) → e_α { e_β
 If Narration(α,β) holds, and α and β describe events e₁ and e₂ then e₁ occurs before e₁
- Explanation {α,β} ^ cause(e_β,e_α) > Explanation(α,β)
 If clauses α and β are discourse-related, and the event described in β caused that described in α, then normally Explanation(α,β) holds

DISCOURSE RELATIONS (2)

- Axiom on Explanation Explanation(α,β) → ¬e_α { e_β
 If Explanation(α,β) holds, then event e₁ described in α does not occur before event e₂
 described in β
- *Push Causal Law* $\{\alpha,\beta\}$ ^ "fall&push" > cause(e_{β},e_{α})

If clauses α and β are discourse-related, and α describes the event e_1 of x falling and β the event e_2 of y pushing x, then normally e_2 causes e_1

Causes precede Effects cause(e₂,e₁) → ¬e₁ { e₂
 If event e₂ causes e₁, then e₁ doesn't occur before e₂

INTERPRETATION (1)

Assuming that text is coherent

Attempt to connect clauses via discourse relations

 $\phi > \psi$ if ϕ then normally ψ

- Max stood up. John greeted him. *Narration* inferred
- John greeted Max. Max stood up. *Narration* inferred too
- Max fell. John pushed him. Narration and push causal law cannot hold both Inferring Narration is *defeasible* here

INTERPRETATION (2)

Causality modeling: Keyness as relation on eventualities $key(e_1,e_2)$ means e_1 is a key event relative to e_2

 $Narration(\alpha,\beta) \rightarrow (\exists e)(key(e,e_{\alpha}) \land (key(e,e_{\beta}) \land \neg(key(e_{\alpha},e_{\beta}) \land \neg(key(e_{\beta},e_{\alpha}) \land \neg(key(e_{\beta},e_{\alpha$

Explanation(α, β) \rightarrow *key*($\mathbf{e}_{\alpha}, \mathbf{e}_{\beta}$)

- ?Max's car broke down. Mary died her hair black. No discourse relation inferable
- ?Everyone laughed. Fred told a joke. Strength of keyness unclear

GENERATION

Knowledge sources

- **Δ** Background knowledge and text purpose
- **EC** Causal and part/whole relations (eventualities)
- **ET** Temporal relations between eventualities
- **EK Keyness of eventualities**

Sketch of the algorithm (three abductive steps)

- From EC, ET, Δ abduce EK fitting speaker's purpose
- From EC, ET, EK, Δ abduce discourse structure D (depth-first left-to-right on EC)
- Build *Conc*, concrete assumptions (depth-first left-to-right on *D*)
 - a) Add concrete assumptions according to current pair of clauses in D
 - b) Nonmonotonic deductive check on Δ and *Conc*

If discourse and event relations included, goto a)

If not, abduce on any rule in Δ and add further concrete assumptions about current clause-pair. Goto b)

GENERATION EXAMPLE 1

Two eventualities e₁ and e₂ and e₂ causes e₁

Falling is the key event

- EC {{ $fall(m,e_1), push(j,m,e_2)$ }, { $cause(e_2,e_1)$ }}
- **ET** $\{\mathbf{e}_2 \mid \mathbf{e}_1\}$

EK $key(e_1,e_2)$

- D $Explanation(\alpha,\beta)$, where $fall(m,e_{\alpha})$ and $push(j,m,e_{\beta})$ D is obtained by abduction of the rules Explanation and Key Event of Explanation
- Conc $\{\langle \gamma, \alpha, \beta \rangle, fall(m, e_{\alpha}), push(j, m, e_{\beta})\}$

NMDC cause(e_{β} , e_{α}) and *Explanation*(α , β)

- Either send *Conc* to the surface grammar: *Max fell. John pushed him.*
 - Or abduce some furtehr concrete assumptions:

 $<\alpha,\beta>$ ^ because(β) > cause(e_{β},e_{α})

Max fell because John pushed him.

GENERATION EXAMPLE 2

Two eventualities e₁ and e₂ and e₁ preceeds e₂

No key event

- **EC** $\{\{fall(m,e_1), push(j,m,e_2)\}, \{\}\}$
- **ET** $\{\mathbf{e}_1 \mid \mathbf{e}_2\}$
- **EK** Neither e_1 nor e_2 key
- D *Narration*(α,β), *fall*(m,e_{α}) and *push*(j,m,e_{β}) D is obtained by abduction of the rules Narration and Key Event of Narration
- Conc $\{\langle \gamma, \alpha, \beta \rangle, fall(m, e_{\alpha}), push(j, m, e_{\beta}) \}$
- NMDC cause(e_{β} , e_{α}) and *Explanation*(α , β). The check indicates that the existing set of concrete assumptions will lead to unreliable text. We must therefore add to *Conc* further assumptions about α and β :

Narration will not suffice, since it doesn't add any further assumptions to *Conc*; but if we use this rule about and then, we will add a further concrete Assumption.

 $<\gamma,\alpha,\beta>$ ^ and then(β) > Narration(α,β)

Max fell and then John pushed him.

EXPRESSING CONTENT USING GENERALISATIONS

An Example – Necklace in an exhibition of 20th century jewellery

- 1) This necklace is in the arts-and-crafts style. It is made of silver, amethysts and pearls. It has very elaborate festoons. It has faceted stones.
- 2) This necklace is in the arts-and-crafts style. It is made of silver, amethysts and pearls. Arts-and-crafts jewels tend to be intricately worked; for instance, this piece has very elaborate festoons. However, unusually for arts-and-crafts jewellery, this piece has faceted stones. Most arts-and-crafts jewels (see for example the jewels in case 8) have cabochon stones.

More elaborate text 2) preferable

- More informative and coherent
- Non-obvious relations expressed
- References to related material
- Defeasibility

arts-and-crafts-jewel(x) > intricately-worked(x)

DEFEASIBLE RULES

Vague quantifiers

• Most Xs are Y for X > Y Few Xs are Y for $X > \neg Y$

Coherence relations

• Exemplification

Arts-and-crafts jewels tend to be intricately worked; *for instance*, this piece has very elaborate festoons.

• Amplification

This piece has very elaborate festoons. Indeed, so do most Arts-and-crafts jewels.

• Concession

Arts-and-crafts jewels tend to be intricately worked, *but* this piece has clean, geometric lines.

User model

- Not all arts-and-crafts jewellery is elaborate; *for instance*, this piece has quite plain. *Indeed*, many arts-and-crafts jewels are plain.
- This piece was *also* designed by Jessie King, *but* around 1910.
- If Jessie King designed it, it was probably designed in 1905.

EXPRESSING DEFEASIBLE RULES

Matching defeasible rules with knowledge base

- LHS *and* RHS hold for an object Combination is expected, express regularity
- *Only* LHS holds for an object Exception is worth stating
- LHS holds for an object, but RHS is unclear Expection is expressed by a hedge

Determining content

- **1.** Finding simple facts about an object to be described
- 2. Search through rule base for these facts

General case of a rule must be introduced by a simple fact, otherwise an inexplicable subject change results; the fact is linked to the generalization via the coherence relation *DEFINITION*. The generalization is linked back to another fact by the coherence relation *EXEMPLIFICATION* (for facts in accordance with the rule) or *CONCESSION* (for facts not in accordance with the rule).

Multiple facts may be joined to a generalisation

2. Search mal-rules in user model for misconceptions

EXPRESSING QUANTIFIERS

epresentation	n description
$X \rightarrow Y$	indefeasible rule asserted
$X \twoheadrightarrow \neg Y$	indefeasible rule asserted
$\neg(X \to \neg Y)$	indefeasible rule denied
$\neg(X \to Y)$	indefeasible rule denied
X > Y	defeasible rule asserted
$X > \neg Y$	defeasible rule asserted
$\neg(X > \neg Y)$	defeasible rule denied
$\neg(X > Y)$	defeasible rule denied
	epresentation $X \rightarrow Y$ $X \rightarrow \neg Y$ $\neg (X \rightarrow \neg Y)$ $\neg (X \rightarrow Y)$ X > Y $X > \neg Y$ $\neg (X > \neg Y)$ $\neg (X > \gamma Y)$

Choosing between quantifiers

- Several may be applicable
- For topic introduction, scalar implicature not important (*some*, in case *many* holds)
- For handling misconceptions, some forms preferred

"Some art-deco jewels have cabochon stones. This jewel is a case in point". not focused "Not all art-deco jewels have faceted stones. This jewel has cabochon stones." better