

Automatic Construction of Linguistic Resources

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Outline of Talk

- > **Introduction:** Linguistic Resources in Human-Computer Interaction
- > **Experiment 1:** Word Categories (Riloff, Shepherd)
- > **Experiment 2:** Refining Taxonomies (Ide, Veronis)
- > **Experiment 3:** Extraction and Structuring of Information (Embley, Capbell, Smith)
- > **Document search** with the help of semantic information
- > **Conclusion/Discussion**

totally: 29 slides

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Typen von semantischen Wissen

- > **Wissen über die Zuordnung zwischen Repräsentation und Fakten**
- > **Wissen um mögliche Inferenzen auf der Repräsentation.**

Experiment 1

A Corpus-Based Approach for Building Semantic Lexicons

Riloff, E. and Shepherd, J (1997). Saltlake City.

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Seedwords

- | | |
|-------------------|---|
| Energy: | <i>fuel gas gasoline oil power</i> |
| Financial: | <i>bank banking currency dollar money</i> |
| Military: | <i>army commander infantry soldier</i> |
| | <i>troop</i> |
| Vehicle: | <i>airplane car jeep plane truck</i> |
| Weapon: | <i>bomb dynamite explosives gun rifle</i> |

Bootstrapping Algorithmus

```
wähle eine Kategorie C  
init seedwords  
categorywords={}  
wiederhole ca. 8 Mal  
    finde alle Kontexte der Seedwords  
    für alle Wörter W, die in den Kontexten  
    auftreten:  
    Score(W,C) = freq(W im Kontext von C) / freq(W)  
    categorywords = categorywords + (W,Score(W,C))  
    kopiere die 5 besten W in seedwords
```

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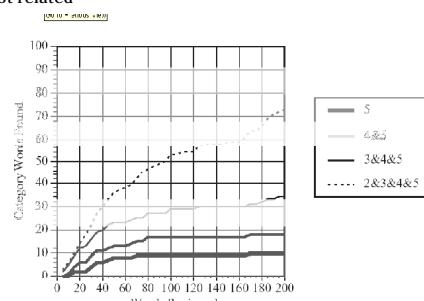
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Words in the Category

Energy: Limon-Covenas^a oligarchs spill staples poles Limon Barrancabermeja Covenas 200,000 barrels oil Bucaramanga pipeline prices electric pipelines towers Cano substation transmission rates pylons pole infrastructure transfer gas fuel sale lines companies power tower price gasoline industries insurance Arauca stretch inc industry forum nationalization supply electricity controls

Measuring Success

- 5: core member of category
1: not related



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Experiment 2

Referencing Taxonomies Extracted from Machine Readable Dictionaries. Ide, N. and Veronis, J. (1994). New York

Typen semantischer Beziehungen

- > zwischen Wortformen (lexical)
 - Synonym:** zwei Worte haben die gleiche Bedeutung.
- > zwischen Wortbedeutungen (semantic)
 - Hyperonym, Hyponym:** Oberbegriff/Spezialisierung
 - Antonym:** Gegenteil
 - Meronym/Holonym:** Bestandteil/ entsteht aus

Mögliche Eigenschaften dieser Beziehungen:

- » transitiv
- » symmetrisch oder reziprok

Die Beziehungen eignen sich zum Aufbau einer **Taxonomie** in Graphnotation.

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Beispiel nyme

Xnyme des Wortes **Fußballspieler**:

Hyperonym - Sportler

Hyponym - Torwart

Meronym - Trikot

Holonym - Mannschaft

Antonym - *Angler

Synonym - ...

Symmetrie: antonym(heiß,kalt)<->antonym(kalt,heiß)

Reziprozität: hyper(Sportler,Fußballspieler)<-> hypo(Fußballspieler,Sportler)

Transitivität: hyper(Fußballspieler,Torwart) && hyper(Sportler,Fußballspieler) -> hyper(Sportler,Torwart)

Extracting Semantic Relations

- > The hypernym is head of the defining noun phrase

dipper	a <i>tadie</i> used for dipping... [CED]
ladle	a long-handled <i>spoon</i> ... [CED]
spoon	a metal, wooden, or plastic <i>utensil</i> ... [CED]

- > or alternatively head of a phrase starting with 'of'.

^o slice	any of various <i>utensils</i> ... [CED]
-----------------------	--

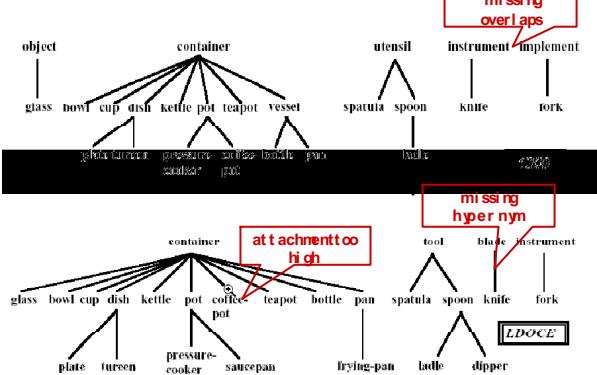
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Problems with Taxonomy Extraction

- > **Attachment too high** - description uses a hypernym of the actual hypernym.
- > **Missing hypernyms** - other way of description used.
- > **Missing overlaps**: - out of several possible hypernyms only one is given.
- > **Or-conjoined heads**: - description yields several heads not all being a universal hypernym
- > **Circularity** - words are described by each other.

Errors in the Graph

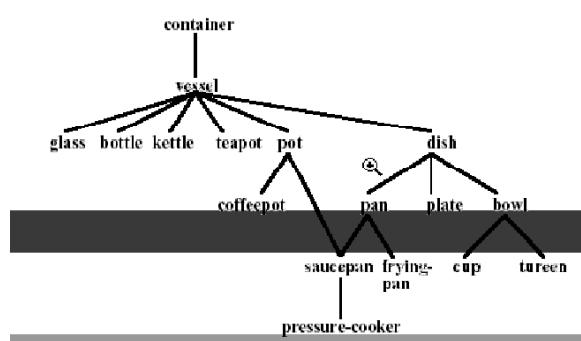


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Merging hierarchies

WORD	CORBUILD	COLLINS	LDOCE	OALD	W9	Combined
ladle	spoon	spoon	spoon	spoon	spoon	spoon
basin	container	container	container	bowl	vessel	bowl
ewer	jug	jug OR pitcher	container	pitcher	pitcher OR jug	pitcher
saucepans	pot	pan	pot	pot	pan	pot AND pan
grill	(absent)	device	(absent)	device	utensil	device AND utensil
fork	tool	implement	instrument	implement	implement	tool, implement AND instrument

Corrected Graph



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Experiment 3

Ontology-Based Extraction and Structuring of Information from Data-Rich Unstructured Documents.

Ebley, D., Campbell, D., Smith, R. (1998)

Raw data

Criteria

- > "data-richness"
- > "narrow ontological breadth"

'96 CHEV Monte Carlo Z34, loaded, bright Red
15,000 actual miles! A great buy at \$14,990,
\$750 to 1000 down. MURDOCK CHEVROLET 298-8090

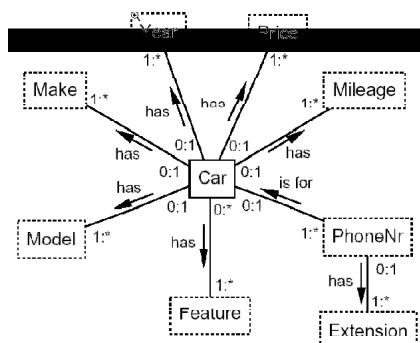
'94 CHEV Corsica, 88,281 miles. Ask for #16. \$4,900.
Government Surplus 523 5885

'93 EGRD 00, red, auto., p/w, p/l, sunroof, loaded, 120K,
new trans., new diff. Runs perfect. Must sell. \$3300 obo.
ccall Nate, 554-4414

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Graphical Ontology



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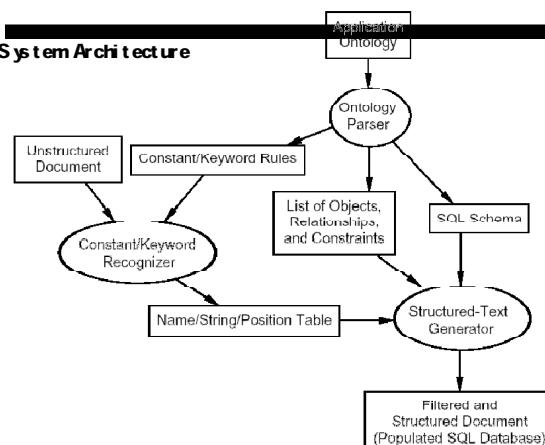
Ontology File

```

Car [0..1] has Year [1..*];
Year {regexp[2]: "\d{2}": ([^\$\d]|")\d{2}\{^\,dkk\";
      "\d{2}": ((^\$\d|^")\d{2},|^"\d",
      "\d{2}": \b'\d{2}\b" );
Car [0..1] has Make [1..*];
Make {regexp[10]: "\baudi\b", "\bchevy\b", ... };
Car [0..1] has Model [1..*];
Model {regexp[10]: "88 : \bolds\$*\$*88\b",
      "80 : \baudi\$*\$*80\b", "\bacclaim\b", ... };
Car [0..1] has Mileage [1..*];
Mileage {regexp[8]: "\b[1-9]\d{2}\d{3}: [\$\d][1-9]\d{2},\d{3}[^\d] "
           {contxt: "\bmile\b", "\bmi\b", "\bmi\b" };
Car [0..*] has Feature [1..*];
Feature {regexp[20]:
  -- Colors
  "\baqua\$-metallic\b", "\bbeige\b", ...
  -- Transmission
  "\b[1-6]\$spd\b", "auto : \bauto(\.\.)",
  -- Accessories
  "\bboot\$-rack\b", "\bspoiler\b", ...
}
  
```

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System Architecture



name/string/position table

Year	96	2	3
Make	CHEV	5	8
Model	Monte Carlo	10	20
Year	94	23	24
Feature	Red	42	44
Mileage	15,000	46	51
KEYWORD(Mileage)	miles	60	64
Price	14,990	84	89
Price	750	93	95
Mileage	1000	100	103
Make	CHEVROLET	120	128
PhoneNr	298 8090	130	137

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Extracted information

Year	Make	Model	Price
94	DODGE		4,995
94	DODGE	Intrepid	10,000
91	DODGE	Neon	3,500
90	FORD	Probe	
88	FORD	Escort	1000

Example of application

- > Using Wordnet in Text retrieval

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Die Wordnet – Datenbank: Browser

Zu Nomen kann der Browser anzeigen:

- > **Synonyme nach Häufigkeit**
- > **Synonyme nach Ähnlichkeit**
- > „Coordinate Terms“ – Alternative Begriffe des gleichen Hyperonyms
- > Hyperonyme und Hyponyme
- > **Bekanntheit (abgeschätzt durch Polysemie)**

Zu den Relationen ist auch eine für Menschen lesbare Definition (gloss) gespeichert.



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Die Wordnet – Datenbank: Datenstruktur

Dies ist die Datenstruktur von Lexicographer Files, die manuell erstellt werden.

```
{ canine, [ dog1, cat! ], pooth, canid,@ }
```

canid ist Hyperonym
Relationeller Wortform canine gloss

```
{ collie, dog1,@ (large multi- colored dog with pointy nose) }
```

Dog (Bedeutung 1) ist Oberbegriff des Sysets

```
{ hound, hunting_dog, pack, #m dog1,@ }
```

Synonyme ein hound ist *Mitglied eines packs

```
{ dog }
```

Begriff ohne Synonyme

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Using WordNet in Text retrieval

> Text-Suche als Vektorraummodell

Mehrdeutigkeit bei der Suche:

- > **Homographen** verringern Precision
- > **Synonyme** verringern Recall

Concept Matching:

- > Queries und Dokumente als semantic concordances
- > Semantic Tagging **mit Hilfe von „Moods“**
- > **automatisches** „Semantic Tagging“ insb. bei Query ein Problem

Query Expansion:

- > Synsets **aus der Umgebung werden der hinzugefügt.**
- beide Verfahren scheitern am „Semantic Tagging“

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Conclusion and discussion

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References

- Riloff, E. and Shepherd, J. (1997). A Corpus- Base Approach for Building Semantic Lexicons. Saltlake City.
Refining Taxonomies Extracted from Machine Readable Dictionaries. Ide, N. and Yatskov, J. (1994). New York.
Ebley, D., Campbell, D., Smith, R. (1998). Ontology-Based Extraction and Structuring of Information from Data-Rich Unstructured Documents. Provo.
<http://www.cogsci.princeton.edu/~wn/>
» Insbesondere **Five** Papers on Wordnet“
Fellbaum, C., (1998). WordNet, an electronical lexical database . Cambridge, MA: MIT Press.
» Insbesondere: Voorhees E. M.. Using WordNet for Text Retrieval.

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