Multiword Expressions

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Structure of Course

- a. Introduction
- b. Computational syntax
- c. Extraction/identification
- d. Computational semantics/interpretation
- e. Decomposability/compositionality

INTRODUCTION

Introduction: Structure

- Definitions, properties of MWEs
- Computational challenges
- Component tasks
- MWEs in NLP applications

But First ...

- Research background:
 - Multiword Expression (MWE) Project (CSLI, NTT CS Labs. and Cambridge University)
 - \star Jointly funded by NSF and NTT CS Labs.
 - Primary project aim is to investigate different means for encoding a variety of MWEs in precision grammars
 - Visit us online at: mwe.stanford.edu

What are Multiword Expressions (MWEs)?

- *Definition:* A **multiword expression** (MWE) is:
 - a. decomposable into multiple simplex words
 - b. lexically, syntactically, semantically, pragmatically and/or statistically idiosyncratic

Some Examples

• San Francisco, ad hoc, by and large, Where Eagles Dare, kick the bucket, part of speech, in step, the Oakland Raiders, trip the light fantastic, telephone box, call (someone) up, take a walk, do a number on (someone), take (unfair) advantage (of), pull strings, kindle excitement, fresh air,

MWE or not MWE?

... there is no unified phenomenon to describe but rather a complex of features that interact in various, often untidy, ways and represent a broad continuum between non-compositional (or idiomatic) and compositional groups of words. (Moon 1998)

MWE Markedness

| | Markedness | | | | | |
|-----------------|--------------|--------------|-----|------|------|--|
| | Lex | Syn | Sem | Prag | Stat | |
| ad hominem | | ? | ? | ? | | |
| at first | X | \checkmark | X | X | X | |
| first aid | X | X | | X | ? | |
| salt and pepper | X | X | X | X | | |
| good morning | X | X | X | | | |
| cat's cradle | \checkmark | \checkmark | | X | ? | |

Indicators of MWE-hood

- Institutionalisation/conventionalisation
- Lexicogrammatical fixedness: formal rigidity, preferred lexical realisation, restrictions on aspect, mood, voice, etc.
 - ★ lexicogrammatically fixed MWE: *kick the bucket*
 - ★ lexicogrammatically fixed non-MWE: *look like*
 - * lexicogrammatically non-fixed MWE: keep tabs on

- Semantic/pragmatic non-compositionality: there is a mismatch between the semantics/pragmatics of the parts and the whole
 - * non-compositional MWE: kick the bucket
 * compositional MWE: at first
- Syntactic irregularity:
 - ★ syntactically-irregular MWEs: all of a sudden, the be all and end all of
 - ★ syntactically regular MWEs: kick the bucket, fly off the handle

- Non-identifiability: meaning cannot be predicted from surface form
 - * idiom of decoding (non-identifiable): kick the bucket, fly off the handle
 - * idiom of encoding (identifiable): wide awake, plain
 truth

- Situatedness: the expression is associated with a fixed pragmatic point
 - * situated MWEs: good morning, all aboard
 * non-situated MWEs: first off, to and fro
- Figuration: the expression encodes some metaphor, metonymy, hyperbole, etc
 - ★ figurative expressions: bull market, beat around the bush
 - \star non-figurative expressions: *first off, to and fro*

- Single-word paraphrasability: the expression has a single word paraphrase
 - ★ paraphrasable MWEs: *leave out = omit*
 - ★ non-paraphrasable MWEs: look up
 - \star paraphrasable non-MWEs: *take off clothes = undress*

- Proverbiality: the expression is used "to describe—and implicitly, to explain—a recurrent situation of particular social interest … in virtue of its resemblance or relation to a scenario involving homely, concrete things and relations" (Nunberg *et al.* 1994)
 - informality: the expression is associated with more informal or colloquial registers
 - ★ affect: the expression encodes a certain evaluation of affective stance toward the thing it denotes

- Prosody: the expression has a distinctive stress pattern which diverges from the norm
 - ***** prosodically-marked MWE: *soft spot*
 - * prosodically-unmarked MWE: *first aid, red herring*
 - ***** prosodically-marked non-MWE: *dental operation*

- Substitutability: MWEs characteristically stand in opposition to anti-collocations, i.e. expressions derived through synonym/word order substitution which occur with markedly lower frequency than the base MWE (or not at all):
 - * non-substitutable MWEs: many thanks (cf. *several thanks, *many gratitudes)
 - ★ substitutable MWEs: salt and pepper vs. pepper and salt
 - * non-substitutable non-MWEs: common platypus

Lexicographic Concept of "Multiword"

- *Rough and ready definition:* a lexeme that crosses word boundaries
- Complications with non-segmenting languages (Japanese, Thai, ...) and languages without a pre-existing writing system (Walpiri, Mohawk, ...)
- Also, in English: houseboat vs. house boat, trade off vs. trade-off vs. tradeoff

Exercise: Spot the MWE

| Expression | MWE? | Markedness | | | | |
|---------------------------|------|------------|-----|----------------|------|------|
| | | Lex | Syn | \mathbf{Sem} | Prag | Stat |
| library card | | | | | | |
| at arm's length | | | | | | |
| old tree | | | | | | |
| foreign direct investment | | | | | | |
| the sun | | | | | | |
| at [nine] o'clock | | | | | | |
| to go bush | | | | | | |
| give a demo | | | | | | |
| kick the bucket | | | | | | |
| once upon a time | | | | | | |
| at home | | | | | | |
| in the meantime | | | | | | |
| to read Shakespeare | | | | | | |

MWEs vs. Collocations

- A collocation is an arbitrary and recurrent word combination
- Collocations can be semantically-marked (e.g. *dark horse*) but tend to be compositional (e.g. *strong coffee*)
- Collocations are generally contiguous (binary) word sequences (more often than not $\bar{N}s)$
- Word order variation/flexibility effects generally ignored in collocation research

MWEs vs. Terms

- (Technical) term = a lexical unit consisting of one or more words which represents a concept inside a domain
- Terminology research primarily interested in the synchronic dynamics of terminology, term formation and terminological variation

Challenges Posed by MWEs

- Syntactic idiomaticity
- Semantic idiomaticity
- Pragmatic idiomaticity
- Statistical idiomaticity
- Flexibility
- Productivity

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Syntactic Idiomaticity



hoc

ad

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Semantic Idiomaticity



Decomposability and Syntactic Flexibility

• Decomposability = degree to which the semantics of an MWE can be ascribed to those of its parts

• Consider:

*the bucket was kicked by Kim Strings were pulled to get Sandy the job. The FBI kept closer tabs on Kim than they kept on Sandy. ... the considerable advantage that was taken of the situation

• The syntactic flexibility of an idiom can generally be explained in terms of its decomposability

Challenges Posed by MWEs

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Pragmatic idiomaticity

- The Wheel of Fortune factor how to represent the jumble of phrases stored in the mental lexicon?
- The Monty Python factor mish-mash of language fragments which evoke particular events/individuals/memories

Challenges Posed by MWEs

- Syntactic idiomaticity
- Semantic idiomaticity
- Pragmatic idiomaticity
- Statistical idiomaticity
- Flexibility
- Productivity

Statistical Idiomaticity

| | unblem ished | spotless | flawless | immaculate | impeccable | | |
|------------|--------------|----------|---------------------------|------------|------------|--|--|
| eye | - | — | — | — | + | | |
| gentleman | _ | _ | ? | — | + | | |
| home | ? | + | _ | + | ? | | |
| lawn | _ | _ | ? | + | _ | | |
| memory | _ | _ | + | _ | ? | | |
| quality | _ | _ | _ | _ | + | | |
| record | + | + | + | + | + | | |
| reputation | + | _ | _ | + | + | | |
| taste | _ | _ | _ | _ | + | | |
| | | | Adapted from Cruse (1986) | | | | |

Statistical Idiomaticity and Dialect

- The arbitrariness of some MWEs is brought out well in dialect differences (e.g. OzE vs. AmE):
 - \star phone box vs. phone booth
 - ★ mail man vs. post man
 - \star no through road vs. not a through street

Challenges Posed by MWEs

- Syntactic idiomaticity
- Semantic idiomaticity
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- Statistical idiomaticity
- Flexibility
- Productivity

Flexibility



Mapping the Boundaries of Flexibility

• Cline between full flexibility and full rigidity, e.g.:

Can/could you tell? Are you able to tell? *They might/ought to tell. How might you tell? *How ought they to tell?

Variation in Flexibility

 There is considerable variation in syntactic flexibility between constructions and also within a given construction type:

a green pepper \approx a pepper which is green a red herring \neq a herring which is red the night is young \neq the young night

I handed in the paper = I handed the paper in Kim ran over the dog \approx ? Kim ran the dog over
Challenges Posed by MWEs

- Syntactic idiomaticity
- Semantic idiomaticity
- Pragmatic idiomaticity
- Statistical idiomaticity
- Flexibility
- Productivity

Productivity

• Varying level of productivity for different MWEs:

ad/post/*pre/*in/*apple/... hoc call/ring/phone/*telephone up Melbourne train driver, human language technology, apple juice seat, ...

Motivation: Why Multiword Expressions?

- Pervasiveness in language
 - MWEs estimated to be equivalent in number to simplex words in mental lexicon
- Volatility (domain tuning, terminology, ...)
 - \star axis of evil, make the pie higher, private equity, ...

- Challenge to NLP systems
 - ★ language understanding
 - ★ fluency
 - \star robustness
- Nice interaction between linguistics, statistics and computational linguistics
- MWEs in language learning environments

Lots of interesting crosslingual commonalities/divergences
 * lexical equivalence: in the red vs. no vermelho
 * structural equivalence: in the black vs. no azul
 * semantic equivalence: in a corner vs. encurralado

Computational Tasks/Issues

- Parsing/identification
- Extraction
- Syntactic classification
- Semantic classification
- Representation
- Crosslingual approaches to MWEs

MWEs in NLP Applications

- IR (N-grams)
 - * phrase-based retrieval: mixed results (Salton and Smith 1990; Lewis and Croft 1990)
 - * query expansion: mixed results (Mandala *et al.* 2000)
 - compound nominals more effective than simplex nominals as index terms (Wacholder and Song 2003)
- Tagging

* virtuous circle between MWE identification and

tagging accuracy (Piao et al. 2003)

• Parsing

- ★ MWEs account for 8% of parsing errors with precision grammar (Baldwin *et al.* 2004)
- perfect knowledge of adverbial MWEs shown to enhance parser accuracy
- Information extraction
 - * collocations used extensively in IE tasks (Lin 1998b)
- Machine translation

★ MWEs integral component of symbolic MT systems (Gerber and Yang 1997; Bond and Shirai 1997)

Summary

- What is an MWE?
- What properties are associated with MWEs?
- Why are MWEs challenging for NLP?
- What NLP applications do MWEs feature in?

COMPUTATIONAL SYNTAX OF MWEs

Case Study: English Resource Grammar

- HPSG-based linguistically-precise open-source grammar
- Compositional semantics based on MRS
- Reversible (parsing and generation)
- Medium coverage
- 8,218 types and 10,625 lexical entries (v06-jun-03)

English Resource Grammar (ERG) in Action

• Leave the report on the desk

```
<h1,e2:PRESENT*:NO_ASPECT*:MOOD:BOOL,
 {h1:imp_m_rel(h3),
 h4:pronoun_q_rel(x6:2PER:REAL_GENDER:ZERO_PRON:-*, h5, h7),
 h8:pron_rel(x6),
 h9:\_leave\_rel(e2, x6, x11:-:REAL\_GENDER:3SG*, v10),
 h12:_def_q_rel(x11, h14, h13),
 h15:_report_rel(x11, v16:BOOL),
 h15:_on_rel(e17:BOOL:NO_TENSE:ASPECT:MOOD, x11, x18:REAL_GENDER:3SG*:-),
 h19:_def_q_rel(x18, h21, h20),
 h22:_desk_rel(x18, v23:BOOL)},
 {h3 QEQ h9,
 h5 QEQ h8,
 h14 QEQ h15.
 h21 QEQ h22}>
```

Basic Syntactic Approach

 Classify different MWE types according to their syntactic flexibility and productivity, and determine the appropriate analysis accordingly





Fixed Expressions

- by and large, in short, kingdom come, every which way, ad hoc (cf. ad nauseum, ad libitum, ad hominem,...), Palo Alto (cf. Los Altos, Alta Vista,...), etc.
- Fixed string which undergoes neither morphosyntactic variation (**in shorter*) nor internal modification (**in very short*)
- Simple words-with-spaces representation is sufficient

Fixed Expressions: Analysis

• Lexical entry for *ad hoc*:

ad_hoc_1 := intr_adj_le &
 [STEM < "ad", "hoc" >,
 SEMANTICS [KEY ad_hoc_rel]].

• Allows very ad hoc, but not *ad very hoc.



Semi-fixed Expressions

- kick the bucket, prostrate oneself, part of speech, San Francisco 49ers
- Adhere to strict constraints on word order and composition
- BUT undergo some lexical variation, e.g.:
 - * inflectional: kick/kicks/kicking/kicked the bucket
 * reflexive pronominal: prostrate him/.../herself
 - **\star determiner selection**: the/those 49ers

Semi-fixed Expressions: Analysis

 Treat as word complex which is lexically variable at particular positions:

```
part_of_speech_1 := n_intr_le &
  [ STEM < "part", "of", "speech" >,
    INFL-POS "1",
    SEMANTICS [KEY part_of_speech_rel ]] & / part_n1.
```

```
kick_the_bucket := v_unacc_le &
  [ STEM < "kick", "the", "bucket" >, INFL-POS "1",
    SEMANTICS [KEY kick_the_bucket_i_rel ]] &
    / kick_v_np*_trans_le.
```

U.S. Sports Team Names

- the (Oakland) Raiders
- an/the [[(Oakland)Raiders] player]
- the [Raiders and 49ers].
- the league-leading (Oakland) Raiders.
- an [[(Oakland) Raider] spokesman]
- *the Oakland 49ers

U.S. Sports Team Names: Analysis

• Name: $\begin{bmatrix} SPR / \langle \rangle \end{bmatrix}$

USTeamName:

$$\begin{array}{c} \text{SPR} & \left\langle \begin{array}{c} \text{Det} \\ \left[\text{definite} \right] \right\rangle \\ \text{NUM} & / \text{plural} \end{array} \end{array}$$

• oakland_raiders_1 := USTeamName & $\begin{bmatrix} LEX-SIGNS & / \langle oakland_1, raiders_1 \rangle \\ SEMANTICS & \langle oakland_raiders_rel \rangle \end{bmatrix}$.

Compound Nouns

- Fully productive = any sequence of nouns can combine to form a MWE (within pragmatic bounds)
- Underspecified semantic relation between the noun modifier and head:

newspaper selection school bus orange juice seat

Compound Nouns: Analysis

- Constructional analysis: ${\rm \bar{N}}~\rightarrow~{\rm N}~{\rm N}$
- Underspecified compound_rel relation between nominal elements, e.g. *cardboard box*:

 $\texttt{cardboard_rel}(x) \land \texttt{box_rel}(y) \land \texttt{compound_rel}(x,y)$

Syntactically-flexible Expressions



Syntactically-flexible Expressions

- write up, let the cat out of the bag, have a shower, ...
- Variable level of flexibility for different expressions
- Basic mechanism of lexical selection

Verb-Particle Constructions

• Verb-Preposition Combinations:

It was like falling off a log/*falling a log off.
They fell quietly off the log.
[Off how many logs] did the drunk fall?

• Verb-Particle Combinations:

- ***** They wrote up the memo/wrote the memo up
- * * Up how many memos did they write?
- \star *They wrote quietly up the memos.

Verb-Particle Constructions

- Compositional: write up, eat/gobble up (activity → accomplishment)
- Noncompositional: look up, throw up
- write up the memo/write the memo up look up the answer/look the answer up

65

Verb-Particle Constructions: Analysis

• Verb selects for particle:

• Assume "joined" word order is canonical (Dehé 2002) and derive "split" word order by way of lexical rule

Verb-Particle Constructions: Productivity

- For fully/semi-productive verb-particles, avoid enumeration through use of lexical rules
- E.g., movement verb + directional particle:

run/walk/... up/down/around/in/...

Light Verbs

• Idiosyncrasy:

make a mistake, *do a mistake, *give a mistake give a demo, do a demo, *make a demo

• Flexibility:

How many demos did Kim give? ...give a revealing demo A demo was given.

Light Verbs



mistake_rel argument_rel ...

- make a mistake/error/boo-boo...
- make an argument/point/statement...

Decomposable Idioms

• take advantage (of), pull strings, keep tabs on

• Flexibility:

They regretted the considerable advantage that had been taken of the unfortunate situation.

Strings had been pulled to get Sandy the job.

The FBI kept closer tabs on Kim than they kept on Sandy.

Flexibility is highly variable.

Decomposable Idioms: Analysis

 $\left[\mathsf{SEMANTICS} \left\langle \left[cat_i_rel \right] \right\rangle \right] \& \ / \ \mathsf{cat_n1}.$
• bag_i :=
$$\left[\text{SEMANTICS } \left\langle \left[bag_i rel \right] \right\} \right] \& / bag_n1.$$

 Use root conditions to ensure that all elements are present:

$$\Big[\mathtt{cat_i}(x) \land \mathtt{bag_i}(y) \land \mathtt{out}(x,y) \Big]$$





Representing institutionalized phrases

- Store matrix of dependency pairs, with the (smoothed) corpus-based frequency of each
- Statistically disprefer rather than symbolically rule out certain word combinations
- Principal use in generation

Simple! ... or then Again?

- To date, we have proposed 4 basic analyses and categorised constructions according to the best fit with those 4 analysis types
- Not all constructions are this compliant!

Test Case: Determinerless PPs

Definition

 Determinerless PPs (PP-Ds) are MWEs comprising a preposition (P) and a singular noun (N_{Sing}) without a determiner:

| Institutional | Media | Metaphor | Temporal | Means/Manner |
|---------------|-------------------|------------|--------------|-----------------|
| at school | on film | on ice | at breakfast | by car |
| $in \ church$ | on TV | at large | on holiday | by train |
| in gaol | to video | at hand | on break | by hammer |
| on campus | $o\!f\!f\ screen$ | at leave | by night | $by \ computer$ |
| $at \ temple$ | in radio | at liberty | by day | via radio |

Crosslinguistic Occurrence of PP–Ds

- Most languages with articles have PP-Ds
- Same semantic types attested in English, Albanian, Tagalog, German, et al. (Himmelman 1998):
 - ★ Institution/Location: *at school*
 - ★ Metaphor/Abstract: *at large*
 - ***** Temporal: *in winter*
 - ★ Means/Manner: *by car*
- Focus of our research on English and Dutch

Corpus- and Lexicon-Occurrence of PP-Ds

- PP-Ds described statically in COMLEX and WordNet, but account for only around 30% and 15%, respectively, of the token occurrences of PP-Ds occurring ≥20 times in the BNC
- $\bullet \approx \! 0.3\%$ of words in BNC are PP–Ds
- \approx 0.2% of parse errors over a sample of the BNC caused by syntactically-marked PP-Ds

The Syntax of PP–Ds

- Variability in syntactic markedness, productivity and nominal modifiability for different PP-D constructions
- Non-productive, non-modifiable PP-Ds: *ex cathedra*, *ad hominem*, *ad nauseum*
- Fully-productive, highly-modifiable PP-Ds: per recruited student that finishes the project
- Most PP–Ds lie between these two extremes

Syntactic Markedness

• Syntactically-unmarked PP-Ds: N_{Sing} is uncountable

E.g. Institutions: in school, in gaol, but *in office (cf. school is over vs. *office is over)

Syntactically-marked PP-Ds: N_{Sing} is strictly countable
E.g. PPs headed by per: per person, but *per information

Nominal Modifiability

- No modification: in *mental/*small hospital
- Idiosyncratic modification: at long/*lengthy/*short last
- Relatively free modification: *at great/considerable/tedious length*
- Modification seldom unrestricted, except in productive constructions

Empirical Analysis of Modifiability

| | Divergence | | | |
|--------------|------------|-----------|--|--|
| PP | D(PP PP) | D(PP NP) | | |
| on horseback | 0.00 | 0.04 | | |
| before dawn | 0.00 | 0.16 | | |
| to hospital | 0.02 | 0.32 | | |
| up front | 0.03 | 0.26 | | |
| on record | 0.10 | 0.76 | | |
| in diameter | 0.14 | 0.54 | | |
| in school | 0.18 | 0.26 | | |
| on loan | 0.18 | 0.71 | | |
| by decree | 1.62 | 2.07 | | |
| on analysis | 4.29 | 2.81 | | |

Modification Types

| | Obligatory | Optional | Impossible |
|-----------|-------------------------|----------------|------------|
| Noun | at *(eye) | on (summer) | |
| | level | vacation | |
| Adjective | $at \ ^*(long)$ | $in \ (sharp)$ | on~(*very) |
| | range | contrast | top |
| Either | $at \ ^*(company)$ | in (family) | |
| | expense | court | |
| | $at \ ^*(considerable)$ | in (open) | |
| | expense | court | |

Coordination quirks

- Coordinate constructions: from mother to child, room by room.
- Partial selectional mismatches: in brush and ink
- Full selectional mismatches: over mens en wereld "about human being and world", van stadion en hotel
 "of stadium and hotel"

The semantics of PP–Ds

- All PP-Ds show a certain degree of (generally systematic) semantic markedness on the noun:
 - \star institutional: *at school*
 - ***** metaphoric: *on ice*
 - \star generic uses: by car
- Some semantic systematicity to the prepositions

PP–Ds with institutional nouns

- Activity enrichment: in gaol "while being a prisoner" and in school "while attending school", cf. in a/the gaol
- Familiarity enrichment: John is in town "John is in (my/his/this) town", cf. in a/the town
- Overlap between the two: *at work* "while at (my/his) work/working"

Metaphorical PP–Ds

- Examples:
 - ★ English at large, on ice, at last
 - * Dutch op zak "in pocket/possession", aan zet "(lit.) on turn"
- Non-compositional, but some degree of morphological systematicity in English: lastly/at last, edgy/on edge

PP–Ds with generic readings

- Examples: by car, by hand, via email
- Means/manner semantics of noun rare in subject/object position
- Resist referential uses and familiarity enrichments, but allow generic and activity readings:

I travelled to San Francisco by car. They're/It's a great way to travel/#It rattled a lot.

Analysis 1: fixed expression

• Word-with-spaces analysis for fully lexicalised PP-Ds: at large, on track

- Prevents nominal modification, coordination
- Effective at capturing syntactically- and semanticallymarked PP-Ds
- Hard to capture optional preposition selection properties (e.g. on top (of), in front (of)

Analysis 2: simple combination

• Use head-complement rule to combine simplex P and N lexical entries:

at church, in/after/during/... school

- Effective at capturing productive syntactically-unmarked PP—Ds
- Licenses unconstrained nominal modification
- Captures compositional semantics



Analysis 3: N selection

• License the preposition to select for an unsaturated NP (\bar{N}) :

by train/plane/bus/hydrofoil/pogo stick...

• Allows for nominal modification and full productivity



Analysis 4: idiosyncratic modification

- Use idiomatic nominal lexical entries and unary rules to constrain the nominal modifier type (noun, adjective or neither):
 - at *(eye) level, on summer vacation, on (*very) top.
- Specify preposition-noun combinatorics by way of root conditions (a lá decomposable idioms)
- Captures idiosyncratic modification effects

Determining the appropriate analysis

- Consider:
 - * nominal modifiability
 - * productivity
 - ★ semantic markedness
 - * NP saturation (syntactic markedness)

96

An Analytical Challenge

- Some strictly-countable nouns occur prolifically in PP—Ds without marked semantics, e.g.:
 - hand, and Dutch zee "sea" and huis "house"

Exercise: Pick the Right Analysis

| | Analysis | | | | | |
|---------------------|----------|------------|----------|------|-------|--|
| | Fixed | Semi-fixed | Syn-flex | Inst | Other | |
| push up the daisies | | | | | | |
| break the ice | | | | | | |
| in time | | | | | | |
| pecking order | | | | | | |
| from scratch | | | | | | |
| bone of contention | | | | | | |
| short shrift | | | | | | |
| make short work of | | | | | | |
| true candour | | | | | | |
| blow one's top | | | | | | |
| run up | | | | | | |

Summary

• What basic syntactic types of MWE are there?

Test Case:

- What issues do the various analyses address?
- How do we determine the correct analysis for a given MWE?
- Are MWEs really that well behaved?

99

EXTRACTING MWEs

Basic Task Description

- Identify the multiword expression (MWE) types in tagged (or raw) text from observation of the token distribution
- MWE (token) identification vs. (type) extraction

In American romance, almost nothing rates higher than what the movie men have called "meeting cute " - that is, boy-meets-girl seems more adorable if it does n't take place in an atmosphere of correct and acute boredom. Just about the most enthralling real-life example of meeting cute is the Charles MacArthur-Helen Hayes saga : reputedly all he did was give her a handful of peanuts, but he said simultaneously, "I wish they were emeralds . Aside from the comico-romantico content here, a good linguist-anthropologist could readily pick up a few other facts, especially if he had a little more of the conversation to go on. The way MacArthur said his line – if you had the recorded transcript of a professional linguist - would probably have gone like this : A[fj] Primary stresses on emeralds and wish ; ; note pitch 3 (pretty high) on emeralds but with a slight degree of drawl , one degree of oversoftness. Conclusions : The people involved (and subsequent facts bear me out here) knew clearly the relative values of peanuts and emeralds, both monetary and sentimental . And the drawling , oversoft voice of flirtation , though fairly overt , was still well within the prescribed gambit of their culture . In other words , like automation machines designed to work in tandem , they shared the same programming , a mutual understanding not only of English words, but of the four stresses, pitches, and junctures that can change their meaning from black to white . At this point , unfortunately , romance becomes a regrettably small part of the picture ; ; but consider , if you can bear it , what might have happened if MacArthur, for some perverse, undaunted reason, had made the same remark to an Eskimo girl in Eskimo. To her peanuts and emeralds would have been just so much blubber

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Complications in MWE Extraction

 Working out the extent of the collocation (phrase boundary detection)

trip the light \blacksquare trip the light fantastic \blacksquare trip the light fantastic at \blacksquare

• Fine line between collocations and simple default lexical combinations

buy a car/purchase power

Statistical Tests Commonly Used

- Simple frequency: f(x, y)
- Pointwise/specific mutual information: $\log \frac{P(x,y)}{P(x)P(y)}$
- Dice's coefficient: $\frac{2 f(x,y)}{f(x)f(y)}$
- (Student's) *t* score
- (Pearson's) chi-square (χ^2)
- Z score

- Log likelihood
- Selectional association

•

Finding of Evert and Krenn (2001) that simple frequency is as good as a wide range of collocation extraction measures over German Adj-N and P-N-V triple extraction tasks
Bigram Results from the WSJ

| Rank | Frequency | Mutual information | χ^2 | t test |
|------|-------------|---------------------------|--------------------------|-------------|
| 1 | of the | Quadi Doum | Posse Comitatus | of the |
| 2 | $in \ the$ | $Wrong ful \ Discharge$ | LORIMAR TELEPICTURES | $in \ the$ |
| 3 | to NUMB | Seh Jik | Petits Riens | to NUMB |
| 4 | for the | Noo Yawk | Wrongful Discharge | on the |
| 5 | to the | WESTDEUTSCHE LANDESBANK | Tupac Amaru | the company |
| 6 | of NUMB | $Naamloze \ Vennootschap$ | Sary Shagan | about NUMB |
| 7 | on the | $Caisses \ Regionales$ | Outlaw Biker | said it |
| 8 | NUMB to | Centenaire Blanzy | GEMINI SOGETI | for the |
| 9 | that the | Guillen Landrau | Centenaire Blanzy | to be |
| 10 | the company | $Ea\ Matsekha$ | Smith-Corona Typewriters | a share |
| ÷ | | | | |

Why Statistics?

- Pick up on word combinations which occur with "significantly" high relative frequency when compared to the frequencies of the individual words (i.e. f(x, y) as compared to f(x) and f(y))
- Why so many different statistical tests?
 - complications in evaluation (hard to say which is the "best" test, conflicting results from different researchers)

- different corpora have different distributional idiosyncracies
- * different tests have different statistical idiosyncracies

Collocation Extraction: Xtract (Smadja 1993)

Outline

- Automatic method for extracting collocations from raw text based on n-gram statistics
- Basic intuition: collocations are more rigid syntactically and more frequent than other word combinations
- Method: attempt to capture this intuition using the basic statistics of word combinations

Stage 1: Extract Significant Bigrams

- w and w_i co-occur (w_i is a collocate of w) if they are found in a single sentence separated by fewer than 5 words
- a bigram (w, w_i) is **significant** iff:
 - ***** w and w_i co-occur more frequently than chance ***** w and w_i appear in a relatively rigid configuration
- Divide up the set of collocates according to POS

Example Corpus

 \dots multiword₍₋₁₎ expressions \dots \dots multiword₍₋₁₎ expressions \dots \dots dialect₍₋₁₎ expressions \dots \dots dialect₍₋₂₎ and expressions \dots \dots expressions of $interest_{(2)}$ \dots \dots multiword₍₋₁₎ expressions \dots ... collocation extraction ... \dots expressions dialect₍₁₎ \dots \dots multiword₍₋₁₎ expressions \dots \dots multiword₍₋₁₎ expressions \dots

Noun Co-occurrence Table

$$w = expressions, \mathcal{D} = \{-2, -1, 1, 2\}$$

| collocate | multiword | dialect | collocation | interest |
|-----------------------|-----------|---------|-------------|----------|
| conocate | w_1 | w_2 | w_3 | w_4 |
| p_i^{-2} | 0 | 1 | 0 | 0 |
| p_i^{-1} | 5 | 1 | 0 | 0 |
| p_i^1 | 0 | 1 | 0 | 0 |
| p_i^2 | 0 | 0 | 0 | 1 |
| $freq_i$ | 5 | 3 | 0 | 1 |
| $w_i \in \mathcal{C}$ | yes | yes | no | yes |

Statistics of Expectation • $\overline{f} = \frac{\sum_{w_i \in \mathcal{C}} freq_i}{|\mathcal{C}|} = \frac{5+3+1}{3} = 3$ (frequency average) • $\sigma = \sqrt{\frac{\sum_{w_i \in \mathcal{C}} (freq_i - \overline{f})^2}{|\mathcal{C}|}} = \sqrt{\frac{(5-3)^2 + (3-3)^2 + (1-3)^2}{3}} \approx 1.63$ • $k_i = \frac{freq_i - \overline{f}}{\sigma}$ (strength)

(pair count average)



• $\overline{p}_i = \frac{\sum_{j \in \mathcal{D}} p_i^j}{|\mathcal{D}|}$

Collocation Filters

• Strength: $k_i > k_{\alpha}(=1)$

→ select frequent collocates

- Spread: $U_i > U_0(=1)$
 - → select spiky distributions
- Peakiness: $p_i^j \ge \overline{p}_i + (k_\beta \times \sqrt{U_i})$

 $k_{\beta} = 0.5^{1}$

→ identify interesting spikes

¹Value of 10 suggested for k_{β} in Smadja (1993)

Back to our Example: Strength

•
$$w_1$$
 (multiword)
* $k_1 = \frac{5-3}{1.63} = 1.22 > 1 \square$
• w_2 (dialect)
* $k_2 = \frac{3-3}{1.63} < 1 \square$
• w_4 (interest)
* $k_3 = \frac{1-3}{1.63} < 1 \square$

Spread and Peakiness

• w_1 (multiword) $\star \overline{p}_1 = \frac{0+5+0+0}{4} = 1.25$ $\star U_1 = \frac{(0-1.25)^2 + (5-1.25)^2 + (0-1.25)^2 + (0-1.25)^2}{4} \approx 20.31 > 1 \checkmark$ p_1^{-1} 5 > 3.500 < 3.50 🛛 p_1^{\perp} p_{1}^{2} 0 < 3.50 X

Stage 2: Bigrams to N-grams

- Independent filter to detect larger N-grams
- Method: for each fixed-distance collocate (w, w_i^j) , extract out contiguous word sequences where $\max(p(word[i])) > T(=0.75)$

Example

 $w = resistance, w_i^{-3} = path$

Concordances from the BNC:

... trod the $path_{(-3)}$ of least resistance , finding the $path_{(-3)}$ of least <u>resistance</u> will along the $path_{(-3)}$ of least <u>resistance</u> the safest $path_{(-3)}$ of least <u>resistance</u> through took the $path_{(-3)}$ of least <u>resistance</u> and ...

→ the path of least resistance is a rigid noun phrase

Reflections

- (At the time) groundbreaking research on collocation extraction
- Not effective at extracting out low-frequency words
- Difficulties in evaluating the results of collocation extraction (applies to this day)
- Difficulties in extracting non-contiguous (predicative) collocations such as verb particles

Linguistics in Collocation Extraction

- Apply statistical measures to (head) bigrams in a given dependency relation (e.g. subject-verb)
 - ★ filters out stop words, produces "collocations" of predefined type for direct use in parsing, etc
- Look beyond contiguous bigrams, to bigrams occurring within a "collocational window" of fixed size (e.g. within 3-4 words of each other)
- Utilise linguistic qualities of collocations:

- * limited internal modifiability (applicable as a postfilter)
- * limited substitutability (contrast with anticollocations, e.g. (strong/*powerful) coffee)
- * non-compositional semantics

Substitutability

Lexicalisation

Concept



Substitutability

- Most immediate means of testing substitutability via synonyms (Pearce 2001b)
- Synonyms accessible from thesauri, but word sense disambiguation is generally needed to isolate which synset(s) over which to apply substitution test
- Possibilities of getting at synonyms via distributional analysis (possibly based on dependency pairs)

Collocation Extraction and Evaluation

- Difficulties in evaluation collocation extraction techniques due to lack of gold-standard datasets (what is MWE?)
- Precision generally evaluated according to pre-compiled LR or relative to corpus
- How to evaluate recall?
- How much is good enough?

Verb-particle Extraction (Baldwin and Villavicencio 2002; Baldwin (to appear))

Verb-particle Constructions (VPCs)

- VPC = verb + obligatory particle(s)
 - ***** intransitive:

Kim calmed down v_particle_le

***** transitive:

Kim handed in the paperKim handed the paper in

v_particle_np_le

Kim gets Sandy down

v_np_prep_particle_only_le

Linguistic Properties of VPCs

- Transitive VPCs undergo the particle alternation (*hand in* the paper vs. *hand* the paper *in*)
- With transitive VPCs, pronominal objects must be expressed in the split configuration (*hand it in* vs. **hand in it*)
- Manner adverbs cannot occur between the verb and particle (*hand it promptly in)

Extracting VPCs: Task Description

- Extract out full list of VPCs attested in a given corpus (cf. generation of independent list of VPCs)
- Make no assumptions about corpus annotation (use only information from pre-processors)
- Base extraction method on basic linguistic properties of VPCs
- Develop technique to be robust over low-frequency VPCs

The Joys of VPC Extraction

- Limited coverage of linguistic tests
- Variable word order
- Variable window length
- Structural/analytical ambiguity:

★ hand [the paper] [in] [here] vs. hand [the paper] [in here] vs. hand [the paper in here]
★ hand [in] [the paper] vs. hand [in the paper]

Corpus Analysis of VPCs

- Generate gold-standard VPC data by taking intersection of VPCs in Alvey Tools data, COMLEX v3.0 and ERG (total of 3,205 entries)
- Take random sample of 1,000 VPCs (1,577 LEs) and manually check for occurrences of each lexical entry (valence) in the Brown Corpus, WSJ and BNC
- Estimate the frequency of attested VPCs by voting across a range of extraction methods (explained later)

VPC Frequency Distribution



www.csse.unimelb.edu.au/~tim/pubs/altss2004.pdf

Analysis of VPC Results

| Corpus | Attested LEs | Ave. freq | Median freq |
|--------|--------------|-----------|-------------|
| Brown | 21.4% | 2.3 | 1 |
| WSJ | 21.2% | 3.4 | 1 |
| BNC | 69.9% | 89.6 | 7 |

Why do We Need Extraction?

 Rather then extracting VPCs, why not just use a pre-compiled broad-coverage, general-purpose VPC dictionary?

| Resources | VPCs | Verbs | Particles |
|------------------|-------|-------|-----------|
| A+C+E | 3,156 | 1,400 | 45 |
| BNC | 7,070 | 2,542 | 48 |
| $A+C+E \cap BNC$ | 2,014 | 1,149 | 28 |
| A+C+E - BNC | 1,138 | 251 | 17 |
| BNC - A+C+E | 5,056 | 1,393 | 20 |
| A+C+E+BNC | 8,208 | 2,793 | 65 |

Tasks

- a. **Shallow lexical acquisition:** extraction of VPC types without valence information (e.g. *calm down, hand in*)
- b. Deep lexical acquisition: extraction of VPC lexical entries (e.g. calm down = v_particle_le ^ v_particle_np_le, blaze away = v_particle_le)

Evaluation

- Use standard measures of precision, recall and F-score $(\beta=1)$
- Calculate relative to the manually-determined corpus attestations of the 1,000 VPCs (for each corpus), cast in terms of the relative task

Classifier design

- a. Generate feature vectors based on various statistics of VPC occurrence from training and test corpora, and build classifier using TiMBL v4.2 (*k*-NN)
- b. Evaluation according to 10-fold cross validation
 - hold out test VPCs in training corpus data
 - test corpus annotations only used in evaluation (not in training data)

Training data

- a. **Corpus attestation data:** manually-annotated corpus attestation of 1,000 VPCs/1,577 LEs
- b. **Gold-standard dictionary data:** 3,205 valenceannotated VPC types (4,597 LEs)
 - apply closed-world assumption in classifying VPC training data

Method-1: Simple POS-based Extraction

 Identify particles using dedicated POS tag (RP in Penn and CLAWS2 tagsets)

• PROCEDURE:

- a. tag the data using a tagger and lemmatise using morph
 b. for each particle, search back to the left up to 6 words
 to find governing verb
- c. filter data according to set of 73 canonical particles
- d. classify as transitive if split or immediately followed by NP, otherwise intransitive

Method-1: Feature Representation

- Features describing frequency of intransitive and transitive VPC types:
 - INTRANS TRANS

Method-1: Example

country_NN fund_NNS offer_VBP an_DT easy_JJ way_NN to_TO get_VB a_DT taste_NN of_IN foreign_JJ stock_NNS without_IN the_DT hard_JJ research_NN of_IN seek_VBG out_RP individual_JJ company_NNS ._.
Method-1: Example

country_NN fund_NNS offer_VBP an_DT easy_JJ way_NN to_TO get_VB a_DT taste_NN of_IN foreign_JJ stock_NNS without_IN the_DT hard_JJ research_NN of_IN seek_VBG out_RP individual_JJ company_NNS ._.

Method-1: Taggers

mxpost: Penn-based MaxEnt tagger

fnTBL: Penn-based TBL tagger

RASP: CLAWS2-based HMM tagger used in RASP

| Method-1: | Results | (Corpus-based |
|-----------|---------|---------------|
|-----------|---------|---------------|

| Tagger | Task | Precision | Recall | $F_{\beta=1}$ |
|--------|-------------------------|-----------|--------|---------------|
| | Shallow | .973 | .877 | .922 |
| mxpost | Deep _{intrans} | .570 | .447 | .502 |
| | Deep _{trans} | .886 | .824 | .854 |
| | Shallow | .979 | .825 | .896 |
| fnTBL | Deep _{intrans} | .573 | .447 | .503 |
| | Deep _{trans} | .894 | .776 | .831 |
| | Shallow | .971 | .735 | .837 |
| RASP | Deep _{intrans} | .600 | .525 | .560 |
| | Deep _{trans} | .834 | .707 | .765 |

Method-1: Results (Dictionary-based)

| Tagger | Task | Precision | Recall | $F_{\beta=1}$ |
|--------|-------------------------|-----------|--------|---------------|
| | Shallow | .973 | .876 | .922 |
| mxpost | Deep _{intrans} | .645 | .658 | .651 |
| | Deep _{trans} | .871 | .842 | .857 |
| | Shallow | .979 | .822 | .894 |
| fnTBL | Deep _{intrans} | .663 | .627 | .644 |
| | Deep _{trans} | .856 | .832 | .844 |
| | Shallow | .963 | .537 | .690 |
| RASP | Deep _{intrans} | .652 | .451 | .533 |
| | Deep _{trans} | .829 | .442 | .577 |

Method-1: Results

- Good results for the shallow task and transitive VPCs, lesser so for intransitive VPCs
- mxpost and fnTBL roughly equivalent, RASP significantly worse
- Corpus-based training data generally produces higher precision, dictionary-based training data higher recall

Method-2: Simple Chunk-based Extraction

 Identify particles using dedicated CoNLL-2000 chunk tag (PRT)

• PROCEDURE:

- a. chunk-parse tagged/lemmatised data using fnTBL
- b. for each (canonical) particle, search back to the left up to 6 words to find governing verb
- c. only allow noun, preposition and adverb chunks between verb and particle
- d. valence determination similar to POS-based extraction

Method-2: Feature Representation

 Features describing frequency of intransitive and transitive VPC types:

INTRANS INTRANS_L INTRANS^{χ} TRANS TRANS_L TRANS^{χ} MI where:

$$(IN)TRANS_{L} = freq(linguistic test data)$$
$$(IN)TRANS = freq(other VPC instances)$$
$$(IN)TRANS_{\mathcal{X}} = \frac{freq((IN)TRANS)}{freq((IN)TRANS) + freq(TRANS)}$$
$$MI = MI(V; P)$$

Method-2: Example

 $\begin{bmatrix} O & `` \end{bmatrix} \begin{bmatrix} PP & \text{instead of} \end{bmatrix} \begin{bmatrix} VP & \text{buy} \end{bmatrix} \begin{bmatrix} NP & \text{mask} \end{bmatrix} \begin{bmatrix} PP & \text{for} \end{bmatrix}$ $\begin{bmatrix} NP & \text{your kid} \end{bmatrix} \begin{bmatrix} O & P \end{bmatrix} \begin{bmatrix} ADVP & \text{just} \end{bmatrix} \begin{bmatrix} VP & \text{cut} \end{bmatrix} \begin{bmatrix} PRT & \text{out} \end{bmatrix}$ $\begin{bmatrix} NP & \text{the columnist} \end{bmatrix} \begin{bmatrix} NP & \text{instead of} \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix}$

Method-2: Example

 $\begin{bmatrix} O & `` \end{bmatrix} \begin{bmatrix} PP & \text{instead of} \end{bmatrix} \begin{bmatrix} VP & \text{buy} \end{bmatrix} \begin{bmatrix} NP & \text{mask} \end{bmatrix} \begin{bmatrix} PP & \text{for} \end{bmatrix}$ $\begin{bmatrix} NP & \text{your kid} \end{bmatrix} \begin{bmatrix} O & P \end{bmatrix} \begin{bmatrix} ADVP & \text{just} \end{bmatrix} \begin{bmatrix} VP & \text{cut} \end{bmatrix} \begin{bmatrix} PRT & \text{out} \end{bmatrix}$ $\begin{bmatrix} NP & \text{the columnist} \end{bmatrix} \begin{bmatrix} NP & \text{instead of} \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix} \begin{bmatrix} O & \dots \end{bmatrix}$

Method-2: Results

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| | Shallow | .991 | .736 | .845 |
| Corpus | Deep _{intrans} | .596 | .489 | .537 |
| | Deep _{trans} | .936 | .724 | .817 |
| | Shallow | .987 | .735 | .842 |
| Dict | Deep _{intrans} | .634 | .614 | .624 |
| | Deep _{trans} | .881 | .751 | .811 |

Method-2: Results

- Higher precision than Method-1, but recall goes down considerably
 - cause: low chunk recall over particles
- Slightly disappointing results

Method-3: Chunk Grammar-based

- Improve recall by looking also at canonical particles occurring as non-particle (PP, ADV) chunks
- Use chunk grammar to determine the syntactic relation between verbs and "particle candidates"
- Classify instances as:
 - * unambiguously intransitive/transitive VPC
 * unambiguously intransitive/transitive non-VPC
 * possible intransitive/transitive VPC

Method-3: Identifying VPCs

- Use chunk grammar to:
 - * check that the chunks which occur between the verb and particle are maximally an NP and particle premodifier adverb chunk (*back*, *right*, ...)
 - * check for a clause boundary or NP immediately after the particle/preposition/adverb chunk
 - check the clause context of the verb chunk for possible extraposition of an NP verbal complement
- Check congruity with linguistic properties of VPCs

Method-3: Structural Ambiguity

 $\begin{bmatrix} NP & we \end{bmatrix} \begin{bmatrix} VP & may ask \end{bmatrix} \begin{bmatrix} NP & question \end{bmatrix} \begin{bmatrix} SBAR & as \end{bmatrix} \begin{bmatrix} NP & you \end{bmatrix} \begin{bmatrix} VP & go \end{bmatrix} \begin{bmatrix} ADVP & along \end{bmatrix} \begin{bmatrix} O & J & \cdots & \checkmark \end{bmatrix}$

 $\begin{bmatrix} NP & \text{it} \end{bmatrix} \begin{bmatrix} VP & \text{wo n't do} \end{bmatrix} \begin{bmatrix} NP & \text{any good} \end{bmatrix} \begin{bmatrix} PP & \text{for} \end{bmatrix} \\ \begin{bmatrix} NP & \text{anybody} \end{bmatrix} \begin{bmatrix} SBAR & \text{unless} \end{bmatrix} \begin{bmatrix} NP & \text{employee} \end{bmatrix} \begin{bmatrix} VP & \text{know} \end{bmatrix} \\ \begin{bmatrix} PP & \text{about} \end{bmatrix} \begin{bmatrix} NP & \text{it} \end{bmatrix} \begin{bmatrix} O & . \end{bmatrix} \qquad \checkmark$

 $\begin{bmatrix} VP & nonperform \end{bmatrix} \begin{bmatrix} NP & loan \end{bmatrix} \begin{bmatrix} VP & will & make \end{bmatrix} \begin{bmatrix} PP & up \end{bmatrix}$ $\begin{bmatrix} NP & only about 0.5 \% \end{bmatrix} \begin{bmatrix} PP & of \end{bmatrix} \begin{bmatrix} NP & the combine bank \end{bmatrix}$ $\begin{bmatrix} NP & 's total loan \end{bmatrix} \begin{bmatrix} ADJ & outstanding \end{bmatrix} \dots$???

Method-3: Attachment Disambiguation

- For cases of structural ambiguity, attempt to resolve using log likelihood ratio (verb-particle (VP), verb-NP₁ head (VN₁), NP₁ head-particle (N₁P) and particle-NP₂ head (PN₂):
 - [$_{VP}$ hand] [$_{NP_1}$ the paper] [$_{PP}$ in] [$_{NP_2}$ here] VPC realised iff:

 $VP \times VN_1 > N_1P \times PN_2$ $VP \times VN_1 > VP \times PN_2$

Method-3: Feature Representation

 Features describing frequency of positive/negative diagnostics for each (intrans/trans) VPC type:

 $INTRANS_+ INTRANS_- (INTRANS_{ATT}) TRANS_+ TRANS_- (TRANS_{ATT})$

Method-3: Results

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| | Shallow | .984 | .891 | .935 |
| Corpus | Deep _{intrans} | .670 | .848 | .748 |
| | Deep _{trans} | .889 | .821 | .853 |
| | Shallow | .982 | .790 | .876 |
| Dict | Deep _{intrans} | .753 | .672 | .710 |
| | Deep _{trans} | .877 | .762 | .815 |

Method-3: Results

- Appreciable gain in recall over Method-1 and Method-2 (greater robustness over low-frequency data)
- More credible results over deep processing tasks
- Corpus-based training data markedly better than dictionary-based training data

Method-4: Parser-based

- Use full parser to resolve attachment ambiguity
- **Parser:** RASP (tag sequence-based parser)
- Read VPCs off RASP output directly
- Valence determination directly from RASP output (presence of dobj for head verb)

Method-4: Example (Full Parse)

(|He:1_PPHS1| |be+ed:2_VBDZ| |wound+ed:3_VVN| |,:4_,| |but:5_CCB| |fight+ed:6_VVD| |on:7_RP|) 1 ; (-7.655)

(|ncsubj| |fight+ed:6_VVD| |He:1_PPHS1| _)

(|ncsubj| |wound+ed:3_VVN| |He:1_PPHS1| |obj|)

(|aux| _ |wound+ed:3_VVN| |be+ed:2_VBDZ|)

(|ncmod| _ |fight+ed:6_VVD| |on:7_RP|)

(|conj| _ |wound+ed:3_VVN| |fight+ed:6_VVD|)

Method-4: Example (Partial Parse)

... |to:15_TO| |bring:16_VVO| |out:17_RP|
|the:18_AT| |sheen:19_NN1|) 0 ; ()

(|ncsubj| |bring:16_VV0| |child+s:12_NN2| _) (|dobj| |bring:16_VV0| |sheen:19_NN1| _) (|ncmod| _ |bring:16_VV0| |out:17_RP|) (|detmod| _ |sheen:19_NN1| |the:18_AT|) (|xcomp| |to:15_T0| |hair:14_NN1| |bring:16_VV0|)

Method-4: Feature Representation

Features describing frequency of each (intrans/trans)
 VPC type, for full and partial parses:

INTRANSfull INTRANSpartial TRANSfull TRANSpartial

Method-4: Results

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| | Shallow | .975 | .715 | .825 |
| Corpus | Deep _{intrans} | .632 | .656 | .644 |
| | Deep _{trans} | .861 | .705 | .775 |
| | Shallow | .975 | .715 | .825 |
| Dict | Deep _{intrans} | .643 | .639 | .641 |
| | Deep _{trans} | .865 | .705 | .777 |

Method-4: Results

- Recall down as compared to Method-3
- Results superior to RASP tagger (parser pre-processor) but below those of the other taggers
- Very little difference between corpus- and dictionarybased training data

Method Combination

- Combine methods to consolidate on relative strengths (precision/recall)
- Combination by concatenating feature vectors for individual methods

Full Combination Results

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| | Shallow | .978 | .957 | .967 |
| Corpus | Deep _{intrans} | .638 | .830 | .721 |
| | Deep _{trans} | .893 | .890 | .891 |
| | Shallow | .979 | .857 | .914 |
| Dict | Deep _{intrans} | .660 | .781 | .715 |
| | Deep _{trans} | .845 | .796 | .820 |

Results of Method Combination

- System combination produces F-score superior to individual methods
- Still disappointing results for intransitive VPCs

 \rightarrow try selective system combination (chunker, chunk grammar, RASP)

Selective Combination Results

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| 6 | Shallow | .978 | .942 | .960 |
| Corpus | Deep _{intrans} | .651 | .893 | .753 |
| | Deep _{trans} | .899 | .874 | .886 |
| | Shallow | .979 | .824 | .895 |
| Dict | Deep _{intrans} | .651 | .717 | .683 |
| | Deep _{trans} | .863 | .754 | .805 |

Best F-score for intransitive VPCs (corpus-based training data)

| Full Combination | | Results (Brown) | | |
|-------------------------|-------------------------|-----------------|--------|---------------|
| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
| | Shallow | .956 | .888 | .921 |
| Corpus | Deep _{intrans} | .783 | .692 | .735 |
| | Deep _{trans} | .840 | .766 | .801 |
| | Shallow | .973 | .555 | .706 |
| Dict | Deep _{intrans} | .675 | .314 | .429 |
| | Deep _{trans} | .765 | .571 | .654 |

Full Combination Results (WSJ)

| Training data | Task | Precision | Recall | $F_{\beta=1}$ |
|---------------|-------------------------|-----------|--------|---------------|
| | Shallow | .938 | .916 | .927 |
| Corpus | Deep _{intrans} | .664 | .669 | .667 |
| | Deep _{trans} | .875 | .812 | .843 |
| | Shallow | .983 | .559 | .713 |
| Dict | Deep _{intrans} | .561 | .274 | .368 |
| | Deep _{trans} | .830 | .612 | .704 |

Reflections

- A range of methods proposed for shallow/deep lexical acquisition of VPCs from unannotated corpora
- Blurring of the token/type distinction used to boost performance
- Method robust over extremely low-frequency data (vital for many MWE types)

Predicting VPC Productivity

- Verb semantics are often a good predictor of verbparticle combinatorics (for compositional VPCs)
- Try using Levin classes to predict productive verbparticle combinations (e.g. aspectual up)

Levin-based Productivity (vs. Dictionaries)



Levin-based Productivity (vs. Dict + BNC)



Summary

- What is the basis of collocation extraction methods?
- How do collocation and MWE extraction methods differ?
- What properties of MWEs make collocation extraction techniques unsuitable?
- In what way can MWE extraction circumvent the issue of lexical coverage?
MWE INTERPRETATION: COMPOUND NOUNS

Compound Nominals and Nominalisations

- **Compound nominal:** N made up of two or more nouns, e.g.:
 - telephone box/booth, river bed, radar footprint, chest X-ray
- Nominalisation: subclass of compound nominals in which the head noun is deverbal, e.g.:

machine performance, museum construction, family worker, student education, satellite observation

Compound Nominals and NLP

- Compound nominals generally processed in three steps:
 - a. **Identification** of compound nominals in some corpus A film interpretation of the book which satirises
 - black assimilation into white society.
 - b. Syntactic analysis of the structure

engine oil filter \rightarrow [[engine oil] filter]

c. Interpretation of the semantics

film interpretation \rightarrow OBJ

• We will focus on interpretation (Step 3)

Identification

• Compound nominals easily detectable from the output of a tagger:

A_DT film_NN interpretation_NN of_IN the_DT book_NN which_WDT satirises_VBZ black_NN assimilation_NN into_IN white_NN society_NN

Identification

• Compound nominals easily detectable from the output of a tagger:

A_DT film_NN interpretation_NN of_IN the_DT book_NN which_WDT satirises_VBZ black_NN assimilation_NN into_IN white_NN society_NN

Identification

- Largely a question of POS tagger post-correction (Lapata and Lascarides 2003)
- Subtle questions about how to detect compound nominals which are part of larger lexical items (e.g. social services committee)

NN Corpus Occurrence

• Estimate of English and Japanese NN corpus occurrence:

| | BNC | Reuters | Mainichi |
|------------------|-------|---------|----------|
| Token coverage | 2.6% | 3.9% | 2.9% |
| Total no. types | 265K | 166K | 889K |
| Ave. token freq. | 4.2 | 12.7 | 11.1 |
| Singletons | 60.3% | 44.9% | 45.9% |

• Highly productive, high frequency of occurrence

Syntactic Analysis

Adjacency (Resnik 1993) vs. dependency (Lauer 1995a) in syntactic analysis, e.g. woman aid worker:
 woman aid > aid worker [Adj] aid worker > woman worker [Dep]
 → [[woman aid] worker]

woman aid < aid worker</th>[Adj]aid worker < woman worker</td>[Dep] \rightarrow [woman [aid worker]]

Interpretation

- Compound nominals are largely unrestricted semantically
 - diesel truck/oil/tanker, phone book, cloud bus, apple juice seat
- Nominalisations tend to occur with subject or object interpretation:
 - *machine performance, museum construction, student education* **BUT ALSO** *soccer competition*

Why Compound Nominal Interpretation?

- Component of language understanding
- (Partial) interpretation required for MT into certain languages:
 - cf. Italian: *coltello da pane* "bread knife", *porta a vetri* "glass door", *succo di limone* "lemon juice"

186

Interface between semantics and stress assignment:

pint jar/mile run/six-figure salary/...
pantry shelf/garage door/bedroom furniture/...

wood box/water bucket/gin bottle/...
daisy chain/cable network/sugar cube/...

BUT rubber boots/steel plate/gold medal/...

Interpreting Compound Nominals

- Possible to interpret compound by way of:
 - * system of "semantic relations" ACTIVITY, CHANGE, PERSON-AFFLICTED, ... steel can = MADE-OF(can,steel) * paraphrasing with prepositional "hidden variable" $P(n2, p, n1) \approx P(n2, p, *)P(*, p, n1)$

 $baby \ chair = chair \ FOR \ babies$

Semantic Theories

- Every linguist has her own theory, but with commonalities
- Import of syntax, semantics, discourse and knowledge representation in different theories
- Claims that finite enumeration of semantic relations are psychologically untenable (Downing 1977)

Example Theory 1: Levi (1978)

- 4 roles for nominalisations:
 - * ACT, PRODUCT, AGENT, PATIENT truck driver = AGENT student discontinuation = ACT
- 9 recoverably deletable predicates for compound nominals:
 - * IN, FOR, FROM, ABOUT
 * CAUSE, MAKE, HAVE, USE, BE power station = MAKE steel box = USE baby crocodile = BE

(prepositional) (relative clauses)

Example Theory 2: Lauer (1995a)

- Interpret compound nominals according to 7 prepositions:
 - **\star of**: state law = law OF state
 - ★ for: baby chair = chair FOR baby
 - *** in**: morning prayer = prayer IN morning
 - *** at**: airport food = food AT airport
 - \star on: Sunday television = television ON Sunday
 - *** from**: reactor waste = waste FROM reactor
 - *** with**: gun man = man WITH gun
 - *** about**: war story = story ABOUT war

Example Theory 3: Copestake (2003)

- Cateogrise compounds as first category that "fits":
 - a. **listed compounds**: home secretary
 - b. hypernmic compounds: tuna fish, oak tree
 - c. deverbal compounds: satellite observation
 - d. relational compounds: jazz fan
 - e. made-of compounds: steel sword, polystyrene box
 - f. prepositional compounds: airshow accident
 - g. non-deverbal verb compounds: *oil town*
 - h. non-paraphrasable compounds: *listeria society*

Exercise: Analyse the NN

| NN | Interpretation | | | |
|---------------------|----------------|-------|-----------|--|
| | Levi | Lauer | Copestake | |
| machine translation | | | | |
| cold virus | | | | |
| cardboard box | | | | |
| state premier | | | | |
| tax module | | | | |
| disk cylinder | | | | |
| relaxation class | | | | |
| darts competition | | | | |
| tennis coach | | | | |
| city protest | | | | |
| telephone number | | | | |

Open Questions

- Is there a definitive categorical system of compound nominal interpretation types? (splitters and lumpers)
- Can any one system work for all domains and compound nominal types?
- What systems of interpretation work in different domains?
- To what degree is interpretation required?

The Disambiguation of Nominalisations (Lapata 2002)

Basic Outline

- Task: binary classification of nominalisations as having a SUBJ or OBJ interpretation (ignore nominalisations such as *soccer competition* — i.e. constrain the space in such a way that interpretation is a well-defined task)
- Assumption: $P(rel|n_1, n_2) \approx P(rel|v_{n_2}, n_1)$
- **Problem:** getting accurate estimates of $P(rel|v_{n_2}, n_1)$

196

Basic Model

$$RA(rel, n_1, n_2) = \log_2 \frac{P(\text{OBJ}|n_1, n_2)}{P(\text{SUBJ}|n_1, n_2)}$$
$$P(rel|n_1, n_2) \approx \frac{f(v_{n_2}, rel, n_1)}{\sum_i f(v_{n_2}, rel_i, n_1)}$$

Resources

- Derive frequency estimates from the BNC
- Estimate $f(v_{n_2}, rel, n_1)$ from output of dependency parser (Cass)
- Determine base verb form of nominalisation based on NOMLEX and CELEX
- Hand-annotate/filter 796 nominalisations extracted from BNC

Observation

- Of 796 items in gold-standard nominalisation set, 47% not attested in BNC in either a verb-object or verbsubject relation
- How to get accurate estimates of $f(v_{n_2}, rel, n_1)$?
- Answer: smoothing based on the frequencies of observed verb-argument pairs

Smoothing

a. **Discounting:** redistribute probability from observed events to unobserved events

- b. **Class-based smoothing:** word-to-class distributional similarity
- c. Distance-weighted averaging: word-to-word distributional similarity

Discounting

• Katz's backing-off:

$$P(rel|n_1, n_2) = \begin{cases} \alpha \frac{f(v_{n_2}, rel, n_1)}{\sum_i f(v_{n_2}, rel, n_1)} & \text{if } f(v_{n_2}, rel, n_1) > 0 \\\\ \beta \frac{f(rel, n_1)}{f(n_1)} & \text{if } f(rel, n_1) > 0 \\\\ (1 - \alpha - \beta) \frac{f(rel)}{\sum_i f(rel_i)} & \text{otherwise} \end{cases}$$

• Estimate α and β by Good-Turing estimation

Class-based Smoothing

- Map observed verb-argument tuples onto the WordNet/Roget classes of the noun, distributing equally across all synsets the noun is categorised as belonging to
- Calculate $f(v_{n_2}, rel, n_1)$ by averaging across the classes that n_1 occurs in
- Closed world assumption for nouns

Distance-weighted Averaging

• Use confusion probability or Jensen-Shannon divergence to estimate the distributional similarity between v_{n_2} and each verb w'_1 , and estimate $f(v_{n_2}, rel, n_1)$ according to:

$$f_s(v_{n_2}, rel, n_1) = \sum_{w'_1} \sin(v_{n_2}, w'_1) f(w'_1, rel, n_1)$$

Confusion Probability

$$P_{C}(w_{1}|w_{1}') = \sum_{rel,w_{2}} P(w_{1}|rel,w_{2})P(rel,w_{2}|w_{1}')$$
$$= \sum_{rel,w_{2}} \frac{f(w_{1},rel,w_{2})f(w_{1}',rel,w_{2})}{f(rel,w_{2})} \frac{f(w_{1}',rel,w_{2})}{f(w_{1}')}$$

Jensen-Shannon Divergence

$$J(w_1, w'_1) = \frac{1}{2} \bigg[D \Big(m(w_1) || n(w_1, w'_1) \Big) + D \Big(m(w_2) || n(w_1, w'_1) \Big) \bigg]$$

$$W_J(w_1, w'_1) = 10^{-\beta J(w_1, w'_1)}$$

where

$$\begin{split} m(w) &= P(rel, w_2 | w) \\ n(w_1, w_1') &= \frac{1}{2} \Big(m(w_1) + m(w_1') \Big) \\ D\Big(m(w_1) || n(w_1, w_1') \Big) &= \\ \sum_{rel, w_2} P(rel, w_2 | w_1) \log \frac{P(rel, w_2 | w_1)}{\frac{1}{2} \Big(P(rel, w_2 | w_1) + P(rel, w_2 | w_1') \Big)} \end{split}$$

Evaluation

- Annotator agreement = 89.7%
- Take 2,000 nearest neighbour verbs w_1' distance-weighted averaging methods, $\beta=5$
- Baseline accuracy of 61.5% (OBJ interpretation)

Results

- Confusion probability and WordNet-based smoothing tend to do the best overall
- Good results for system classification, combined with context modelling in the form of the right word context of the compound nominal (85% test accuracy)

Reflections

- Interesting task-oriented smoothing experiment
- What to do with non-SUBJ/OBJ nominalisations?
- What to do with prepositional verbs, verb particles?
- Influence of pragmatics on interpretation

Classifying the Semantic Relations in Noun Compounds (Rosario and Hearst 2001)

Basic Outline

- **Task:** interpretation of (2-word) compound nominals within the biomedical domain
- Method: use lexical or conceptual knowledge about the component nouns to interpret the whole (context-independent)
- **Resource:** MeSH (biomedical thesaurus)

Semantic Roles

- Compound nominals interpreted via 18 (out of 38) relations:
 - ★ more specific than case roles, and less specific than IE template fillers
 - ★ customised to the biomedical domain (e.g. polio survivors → PERSON-AFFLICTED)
 - * thresholded for frequency
 - ★ overlapping (multiclass classification possible: cell growth → ACTIVITY + CHANGE)

211

Method

- Class-based model: describe NN according to the concatenation of the MeSH representations of N₁ and N₂ (up to level N)
- Lexical model: describe NN by its component words (closed-word assumption)
- Learner: neural network (feed-forward network with one hidden layer)
Results

- Over closed data, the lexical and class-based models perform equivalently (\approx 60%)
- Over open data, the class-based model performs better (unsurprisingly)
- Suggestion that N_2 has a stronger impact on the interpretation than N_1

Reflections

- Question of interpretation system sidestepped to some degree by picking a technical domain
- Multiclassification awkward effect, which raises questions about the appropriateness of the interpretation system
- Possibility for a hybrid approach combining the classbased and lexical models?
- No systematic treatment of lexicalised nominals

Integrating Symbolic and Statistical Representations (Copestake and Lascarides 1997)

Basic Outline

• Basic method:

- a. use the grammar/lexicon to delimit the range of potential interpretations of a given NN
- b. use "productivity" probabilities to rank the individual interpretations
- c. use pragmatics to filter out interpretations which produce discourse incoherence within a given context
- Possible to derive non-standard interpretations for a compound nominal (e.g. *garbage man*)

Semantic Hierarchy



Estimating Productivity

• Estimate productivity based on the number of attested forms of a given schemata:

$$\operatorname{Prod}(cmp_schema) = \frac{M+1}{N}$$

where N is the number of pairs of senses which match cmp_schema and M is the number of attested forms

• Cf. substitution tests for collocations/compositionality

Applying the Productivity Estimates

- Interpretations for *cotton* bag based on analysis of fabric/container NNs in the BNC (based on WordNet):
 - MADE-OFP = 0.84PURPOSE-PATIENTP = 0.14GENERAL-NNP = 0.02
- Prediction that the default interpretation for *cotton* bag is MADE-OF

Interface with Pragmatics

- Model pragmatics with SDRT and world knowledge with DICE
- Use SDRT and DICE to filter out interpretations that produce discourse incoherence:
 - a. Mary sorted her clothes into various bags made from plastic
 - b. She put her skirt into the cotton bag

Reflections

- Rare instance of method which provides direct handling of the lexicon-pragmatics interface
- Implausible interpretations supported explicitly, but dispreferred
- Difficulties in collecting productivity statistics
- Question of real-world applicability of SDRT/pragmatic reasoning

Summary

- What basic types of compound noun are there, and how do they differ?
- What types of theory are there for interpreting compound nouns?
- What are their strengths and weaknesses?
- What computational techniques can be employed to interpret compound nouns?

SEMANTIC COMPOSITIONALITY

Semantic Decomposability and Compositionality

• Decomposability = degree to which the semantics of an MWE can be ascribed to those of its parts

kick the bucket \rightarrow die' spill the beans \rightarrow reveal'(secret')

- Compositionality = degree to which the semantics of the parts of an MWE contribute towards those of the whole
- Domain considerations: *monosodium glutamate* in chemistry vs. health domains

Syntactic Compositionality

- Degree to which the syntactic properties of the parts of an MWE combine to make up the syntax of the whole
 - ★ Fixed expressions: by and large, San Francisco
 ★ Verb particles: eat up vs. chicken out
- Syntactic compositionality binary effect; noncompositional MWEs lexicalised
- Semantic decomposability continuum of regularity with more subtle effects and syntactic corollaries

Decomposability and Syntactic Flexibility

• Consider:

*the bucket was kicked by Kim Strings were pulled to get Sandy the job. The FBI kept closer tabs on Kim than they kept on Sandy. ... the considerable advantage that was taken of the situation

• The syntactic flexibility of an idiom can generally be explained in terms of its decomposability

Ideal Research Objective

- Automatically decompose a given MWE/align component words with semantic primitives
- Classification of MWEs into 3 classes:
 - a. non-decomposable MWEs (e.g. kick the bucket, shoot the breeze, hot dog)
 - b. idiosyncratically decomposable MWEs (e.g. spill the beans, let the cat out of the bag, radar footprint)
 - c. simple decomposable MWEs (e.g. kindle excitement, traffic light)

Realistic Short-term Objective

- Demarcate simple decomposable MWEs from idiosyncratically decomposable and non-decomposable MWEs (roughly equivalent to endocentric vs. exocentric distinction)
- Binary distinction vs. mapping onto continuum of relative decomposability

Approaches to Evaluation

- Dictionary based: binary evaluation, based on prediction that non-compositional MWEs will be lexically listed
- **Similarity based**: relative similarity of the parts to the whole (e.g. relative to WordNet)

 $sim(pig metal, metal) \gg sim(pig metal, pig)$

• Entailment based: binary evaluation, based on whether the whole "entails" the parts or not

Susan finished up her paper \models Susan finished her paper

Ranking based: describe MWE compositionality by way of continuous/discrete scale of compositionality
 comp(put up) ≥ comp(eat up) ≥ comp(gun down)

231

Exercise: Rate the Compositionality

| VPC | Compositionality | | | | |
|---------------------------|------------------|----------------|--------|--------|------|
| | Dic | \mathbf{Sim} | Ent(V) | Ent(P) | Rank |
| $get \ down_{trans}$ | | | | | |
| $piss \ of\!f_{trans}$ | | | | | |
| $pay off_{trans}$ | | | | | |
| $lift \ out_{trans}$ | | | | | |
| $roll \ back_{trans}$ | | | | | |
| $dig up_{trans}$ | | | | | |
| $lie \ down_{intrans}$ | | | | | |
| $wear \ on_{intrans}$ | | | | | |
| $chicken \ out_{intrans}$ | | | | | |
| hand out_{trans} | | | | | |

Automatic Identification of Non-Compositional Phrases (Lin 1999)

Basic Method

• Use substitution as a test of compositionality:

 $\begin{array}{l} red \ tape \ \rightarrow \ \underline{yellow} \ tape, \ red \ \underline{cassette} \\ economic \ impact \ \rightarrow \ \underline{political} \ impact, \ economic \\ \underline{effect} \end{array}$

Evaluate based on a dictionary of idioms

System Resources

POS-conditioned thesaurus (nouns, verbs, adjectives/adverbs)

* derived from dependency data (Minipar):

Collocation data

* dependency tuples (H,R,M) with high log-likelihood ratio (H = head, R = relation, M = modifier)

(Point-wise) Mutual Information

• Measure of the level of association between two events A and B:

$$MI(A,B) = \log_2 \frac{P(A,B)}{P(A)P(B)}$$

- Commonly used in collocation extraction
- Not appropriate for low-frequency events

Mutual Information and Compositionality

• Scaling up to 3 events A, B and C, where B and C are conditionally independent given A:

$$MI(A, B, C) = \log_2 \frac{P(A, B, C)}{P(B|A)P(C|A)P(A)}$$
$$MI(H, R, M) = \log_2 \frac{\frac{|H \ R \ M|}{|* \ * \ * |}}{\frac{|H \ R \ *| \ |* \ R \ M| \ |* \ R \ *|}{|* \ R \ *| \ |* \ R \ *| \ |* \ R \ *|}}$$
$$= \log_2 \frac{\frac{|H \ R \ M| \ |* \ R \ *|}{|H \ R \ *| \ |* \ R \ M|}}{|H \ R \ *| \ |* \ R \ M|}$$

Definition of Compositionality

- A phrase α is non-compositional iff there is no β s.t.:
- (a) β can be produced by substitution of the components of α for any of 10 most-similar words, and
 (b) there is an overlap between the 95% confidence interval of the MI values of α and β
- 10 most-similar words tested for each of H and M (R fixed)

Word Similarity: Lin (1998)

$$sim(w_1, w_2) = \frac{\sum_{(r,w)\in T(w_1)\cap T(w_2)} MI(w_1, r, w) + MI(w_2, r, w)}{\sum_{(r,w)\in T(w_1)} MI(w_1, r, w) + \sum_{(r,w)\in T(w_2)} MI(w_2, r, w)}$$

Example 1: spill (one's) guts

- (*spill*,V:comp1:N,*gut*):
 - ★ spill: leak, pour, spew, ..., spray
 - \star gut: intestine, instinct, foresight, ..., charisma
- Check for each of (*leak*,V:comp1:N,*gut*), (*spill*,V:comp1:N,*inte* ... in the collocation database
- None found, so *spill (one's) guts* is non-compositional

Example 2: red tape

• (*tape*,N:adj:N,*red*):

* tape: videotape, cassette, videocassette, ..., audio
* red: yellow, purple, pink, ..., shade

- Find (tape,N:adj:N,yellow), (tape,N:adj:N,orange), (tape,N:adj:N,black) in the collocation database but with very different MI values
- *red tape* is non-compositional

MI Confidence Interval: the Z-test

 Possible to calculate the "true" MI of (H,R,M) according to the Z-test:

$$\overline{p} \pm z_N \sqrt{\frac{\overline{p}(1-\overline{p})}{n}} = \frac{k}{n} \pm z_N \sqrt{\frac{\frac{k}{n}(1-\frac{k}{n})}{n}} \approx \frac{k \pm z_N \sqrt{k}}{n}$$

where \overline{p} is the MLE of p, n is |* * *|, k is |H R M|, and z_N is a constant determined by the confidence level N, e.g. $z_{0.95} = 1.96$

Applying the Z-test

 Determine the "fit" between two MI values by calculating the Z-score interval for the putative noncompositional MWE and determining whether the MI of the second falls into that interval

Evaluation

- Evaluate the method relative to an idiom dictionary
- OK precision, and significant numbers of the extracted MWEs not contained in the dictionary appear to be non-compositional based on manual inspection

Reflections

- Is substitution really a good test for noncompositionality?
 - * institutionalised phrases: frying pan, salt and pepper, many thanks
 - ***** productive MWEs: *call/phone/ring up*
- Look to alternative methods

A Statistical Approach to the Semantics of Verb-particles (Bannard *et al.* 2003)

Basic Method

- Define similarity in terms of distributional similarity, i.e. assume that if an MWE is compositional, it will occur in the same lexical context as its parts
- Divide up compositionality to look at verb and particle similarity independently
- Evaluate against human judgements

Verb-particle constructions (VPCs)

 VPC = A verb plus one or more obligatory (prepositional) particles

Peter put the picture up Susan finished up her paper Philip gunned down the intruder Barbara and Simon made out

 Assumption: VPCs are not always fully compositional or fully non-compositional, but rather populate a continuum between the two extremes
Compositionality by Entailment

- Peter put the picture up
 |= Peter put the picture somewhere
 |= the picture was up
- Susan finished up her paper \models Susan finished her paper
- Philip gunned down the intruder \models the intruder was down
- Barbara and Simon made out

Obtaining Human Judgements

- a. Extract VPCs from British National Corpus (Baldwin and Villavicencio 2002)
- b. Randomly select 5 sentence tokens for each of 40 randomly selected VPC types
- c. Present native English speakers with tokens and asked whether verb and/or particle is implied by the VPC $\,$
- d. Response: Yes, No or Don't Know

Example Sentence Tokens: round up

A dog started to round up sheep.

In three years they had rounded up fifty captive orangs.

Owned by Jo Rutherford, Trigo rounded up the milking herd and brought it back to the milking parlour in Devon.

On 9 August, 349 Arrests were made as the miltary swooped to round up serving and former IRA activists.

Ten days later, when the agents moved in to <u>round up</u> their targets, El-Jorr checked out and returned to Cyprus, charging the hotel bill to his American Express card as instructed.

Human Judgements

- Does round up imply round?
- Does round up imply up?
- Obtain gold-standard analysis by taking majority judgement (ignore **Don't Know** responses)

Sample Judgements

| VPC | Component word | Yes | No | Don't Know |
|-------------|----------------|-----|----|------------|
| dig up | dig | 21 | 5 | 0 |
| | up | 18 | 7 | 1 |
| stay up | stay | 20 | 5 | 1 |
| | up | 21 | 5 | 0 |
| brighten up | brighten | 9 | 16 | 1 |
| | up | 16 | 10 | 0 |
| add up | add | 12 | 14 | 0 |
| | up | 19 | 6 | 1 |

Binary Classification Tasks (×4)

- a. Is the item completely compositional?
- b. Does the item include at least one item that is compositional?
- c. Does the verb contribute its simplex meaning?
- d. Does the particle contribute its simplex meaning?

Classification Methods

- Four different classifiers implemented
- Method 1 based on Lin (1999), Methods 2-4 address theoretical concerns with this model
- All methods based on co-occurrence vector representation of VPC and component words

- Reimplemented Lin (1999) over VPCs.
- Tested over all four tasks assuming that the substitutability of each item will give us some insight into its semantic contribution
- Reconstruct Lin's thesaurus to include all of verbs, nouns, adjectives/adverbs and prepositions.

- Similar to Lin (1999) except for use of knowledge-free approach to obtaining thesaurus
- Very similar to Schütze (1998) "context space" method
- Similarities from pairwise comparison of all verbs, particles and VPCs
- \bullet Obtain thesaurus by taking the N most similar words of a given POS

257

- Use same method of substitution
- A component is said to be contributing simplex meaning if expression formed by substitution occurs among the nearest 100 verb-particle constructions

- Hypothesis is that if a verb or particle is contributing simplex meaning to a VPC then it will be semantically similar to the VPC according to cosine measure
 - ★ a verb is judged to be contributing simplex meaning if it occurs within the 20 most similar items to the VPC.
 - ★ a particle is judged to be contributing simplex meaning if it is in top 10 most similar items to the VPC.

Results

- Mixed at best!
- Methods 3 and 4 tend to perform better than methods
 1 and 2

Detecting a Continuum of Compositionality in Phrasal Verbs (McCarthy *et al.* 2003)

Basic Method

- Compare different methods for modelling compositionality based on distributional similarity and statistical tests traditionally used in collocation extraction
- Map 111 VPCs onto a ranked list, based on human judgements over an 11-point compositionality scale
- Evaluate according to the rank order correlation with the gold-standard ranked list

System Resources

• Build thesaurus a lá Lin (1998), based on dependency tuple output of RASP

Similarity-based Methods

- overlap: relative overlap between the top N neighbours of the VPC and its simplex verb
- same particle: the number of VPCs which select for the same particle as the given VPC amongst the top N neighbours of that VPC
- sameparticle simplex: the value for sameparticle minus the number of top N neighbours of the simplex verb which select for that same particle

- **simplexasneighbour:** does the simplex verb occur in the top 50 neighbours of the VPC?
- rankofsimplex: what is the rank of the simplex verb in the neighbours of the VPC?
- overlapS: the overlap of neighbours in the top N neighbours of the VPC and simplex verb, where VPC neighbours are converted to simplex verbs in the VPC case

Statistical Methods



- Log-likelihood ratio
- (Point-wise) mutual information
- Simple frequency of the VPC
- Simple frequency of the simplex verb

Resource-based Method

- Binary test for the occurrence of the VPC in:
 - ★ WordNet
 - ★ Alvey Tools (ANLT) VPC data
 - * Alvey Tools (ANLT) prepositional verb data

Correlation with the Gold-standard Data

- For binary tests (**simplexasneighbour**, WordNet, ANLT), use the Mann-Whitney U test (rank sum test)
- For other methods, map each output value onto a rank and apply the Spearman Rank Correlation test (rank test)
- In each case, calculate the Z score and the probability of the null hypothesis (i.e. no correlation between the output of the method and the gold-standard data)

Results

- same particle simplex best performer of similaritybased methods
- MI best performer of statistical tests
- Question of how to apply the results to the proposed task of lexical acquisition?

Overall Reflections

- Promising results observed for detecting compositionality/decon but less so for determining the semantic contribution of individual words in an overall MWE
- What about MWEs where the simplex words don't occur with that same POS (e.g. *chicken out*)
- Effects of polysemy (e.g. run down, run over)
- How to move on to actually semantically decompose an MWE?

Summary

- How do decomposability and compositionality differ?
- What methods have been proposed for generating goldstandard compositionality data?

WRAP-UP

Overall Conclusions

- MWEs are frequent, fun and funky in all sorts of ways
- There's much, much more to MWEs than our old friend kick the bucket
- More work needs to be done in developing gold-standard resources to encourage others to enter the fray
- Most of the research problems are far from resolved: lots of room for everyone to play in!

MWE Resources

- MWE project page: mwe.stanford.edu
- On-line MWE data: mwe.stanford.edu/resources
- On-line bibliographies:
 - * mwe.stanford.edu/bib.html
 - * www.ims.uni-stuttgart.de/euralex/bibweb/

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