# $\mathbf{G}_{h} \mathbf{0 S t - N N : ~ A ~ R e p r e s e n t a t i v e ~ G o l d ~ S t a n d a r d ~ o f ~}$ German Noun-Noun Compounds 

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## Motivation for a Representative Gold Standard

Interest in systematically exploring factors that have been found to influence the cognitive processing and representation of compounds, such as

- frequency-based factors, i.e., the frequencies of the compounds and their constituents (e.g., van Jaarsveld \& Rattink, 1988; Janssen et al., 2008)
- the productivity (morphological family size), i.e., the number of compounds that share a constituent (de Jong et al., 2002);
- the relationship between compound modifier and head: a teapot is a pot FOR tea, and a snowball is a ball MADE OF snow (Gagné \& Spalding, 2009).
In addition, we were interested in the effect of ambiguity (of both the modifiers and the heads) regarding the compositionality of the compounds.


## Creation of the Gold Standard $\mathbf{G}_{h}$ ost-NN

1. Corpus-based induction of candidate list

- basis: German web corpus DECOW14AX (Schäfer \& Bildhauer, 2012), with 11.7 billion words
- extraction of 365,786 common nouns and their lemmas, according to the Tree Tagger (Schmid, 1994)
- selection of 154,960 two-part noun-noun compounds, according to the morphological analyser SMOR (Faaß et al., 2010)

2. Enrichment of empirical properties

- corpus frequencies of compounds and constituents (i.e., modifiers and heads)
- productivity of the constituents (modifiers and heads), i.e., how many compound types contained a specific modifier or head constituent
- number of senses of the constituents (modifiers and heads) and the compounds, relying on GermaNet (Hamp \& Feldweg, 1997)

3. Random but balanced compound selection

- goal (optimum): random subset of compound candidates balanced across frequency, productivity and ambiguity ranges
- goal (compromise): two main criteria productivity of the modifiers (low/mid/high) and ambiguity of the heads (1, 2 and $>2$ senses)
- random selections of 20/5 compounds across 9 categories

4. Systematic extension of core set

- systematic extension of the $20 \times 9$ and $5 \times 9$ randomly selected compounds by adding all compounds from the original set of compound candidates with either the same modifier or the same head as any of the selected compounds
- extension procedure destroyed the coherent balance of criteria underlying our random extraction, but ensured a variety of compounds with either the same modifiers or the same heads

5. Semantic annotation of gold standard

- size of novel gold standard $\mathrm{G}_{h}$ ost: 868 noun-noun compounds
- semantic relations according to Ó Séaghdha (2007):

BE, HAVE, IN, ABOUT, ACTOR, INST(rument), LEX

- compositionality ratings from experts and via AMT


## $\mathbf{G}_{h}$ ost-NN Properties

Semantic relations in compounds: 300


Compositionality ratings across relation types:


Productivity and compositionality ratings of modifiers and heads:


## Resource

1. The set of 154,960 noun-noun candidate compounds and their constituents, accompanied by corpus frequency, productivity and degree of ambiguity.
2. The final gold standard $G_{h}$ ost-NN of 868 noun-noun compounds and their constituents, accompanied by corpus frequency, productivity, ambiguity, and annotated with semantic relations and compositionality ratings.
3. The carefully balanced $G_{h}$ ost-NN subsets of $20 \times 9$ and $5 \times 9$ compounds and their constituents, categorised according to our 9 categories for modifier productivity and head ambiguity.

| Nouns |  |  |  |  |  | Frequencies |  |  | Productivities |  | Ambiguities |  | Relation | Ratings |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Com | mpound | Modifier |  | Head |  | Compound | Modifier | Head | Modifier | Head | Modifier | Head |  | Modifier | Head |
| Stadthotel | city hotel | Stadt | city | Hotel | hotel | 3,405 | 4,053,206 | 1,199,856 | 543 | 59 | 1 | 1 | IN | 3.35 | 5.35 |
| Stadtrand | suburb | Stadt | city | Rand | border | 25,099 | 4,053,206 | 523,473 | 543 | 98 | 1 | 2 | HAVE | 4.94 | 4.25 |
| Stadtwerk | public services | Stadt | city | Werk | plant | 107,754 | 4,053,206 | 1,354,148 | 543 | 366 | 1 | 6 | ACTOR | 3.81 | 3.69 |
| Sonnenenergie | solar energy | Sonne | sun | Energie | energy | 25,398 | 832,636 | 1,191,333 | 155 | 30 | 3 | 2 | INST | 4.58 | 5.44 |
| Sonnenkönig | Sun King | Sonne | un | König | king | 2,680 | 832,636 | 494,221 | 155 | 109 | 3 | 3 | LEX | 1.94 | 5.50 |
| Sonnenscheibe | solar disc | Sonne | sun | Scheibe | slice | 3,155 | 832,636 | 364,567 | 155 | 96 | 3 | 4 | BE | 4.56 | 3.75 |
| Sonnenseite | sunny side | Sonne | sun | Seite | side | 7,279 | 832,636 | 5,508,445 | 155 | 256 | 3 | 6 | IN | 4.00 | 4.31 |
| Sonnenstrahl | sunbeam | Sonne | sun | Strahl | beam | 44,612 | 832,636 | 32,182 | 155 | 27 | 3 | 3 | HAVE | 5.13 | 4.69 |
| Sonnenuhr | sundial | Sonne | sun | Uhr | clock | 8,407 | 832,636 | 4,507,590 | 155 | 63 | 3 | 2 | INST | 3.75 | 5.31 |
| Jeanshose | jeans | Jeans | jeans | Hose | trousers | 2,971 | 66,789 | 273,665 | 19 | 61 | 1 | 1 | BE | 5.25 | 5.44 |
| Latzhose | overall | Latz | bib | Hose | trousers | 3,296 | 5,324 | 273,665 | 1 | 61 | 2 | 1 | HAVE | 3.54 | 5.23 |
| Strumpfhose | tights | Strumpf | stockings | Hose | trousers | 20,535 | 26,331 | 273,665 | 13 | 61 | 1 | 1 | BE | 4.35 | 4.42 |
| Kirchspiel | parish | Kirche | church | Spiel | game | 6,583 | 1,761,187 | 4,122,168 | 319 | 403 | 3 | 6 | LEX | 4.44 | 3.13 |
| Machtspiel | power game | Macht | power | Spiel | game | 4,408 | 806,162 | 4,122,168 | 169 | 403 | 2 | 6 | ABOUT | 4.63 | 3.44 |
| Ritterspiel | knights' tournament | Ritter | knight | Spiel | game | 2,365 | 115,484 | 4,122,168 | 47 | 403 | 1 | 6 | ACTOR | 3.94 | 4.75 |
| Testspiel | tryout | Test | test | Spiel | game | 37,800 | 660,169 | 4,122,168 | 100 | 403 | 3 | 6 | BE | 4.25 | 5.19 |
| Windspiel | wind chimes | Wind | wind | Spiel | game | 2,284 | 551,317 | 4,122,168 | 88 | 403 | 3 | 6 | INST | 4.31 | 2.94 |
| Winterspiel | winter games | Winter | winter | Spiel | game | 16,067 | 721,552 | 4,122,168 | 207 | 403 | 1 | 6 | IN | 4.43 | 5.14 |
| Würfelspiel | game of dice | Würfel | dice | Spiel | game | 4,408 | 80,371 | 4,122,168 | 14 | 403 | 2 | 6 | INST | 4.94 | 5.56 |
| Bergkette | mountain chain | Berg | mountain | Kette | chain | 8,799 | 564,178 | 207,479 | 205 | 139 | 2 | 4 | BE | 5.13 | 2.56 |
| Halskette | necklace | Hals | neck | Kette | chain | 8,707 | 271,703 | 207,479 | 39 | 139 | 3 | 4 | IN | 3.94 | 5.44 |
| Handelskette | trade chain | Handel | trade | Kette | chain | 6,509 | 428,611 | 207,479 | 240 | 139 | 1 | 4 | INST | 4.75 | 3.38 |
| Hotelkette | hotel chain | Hotel | hotel | Kette | chain | 6,410 | 1,199,856 | 207,479 | 134 | 139 | 1 | 4 | BE | 5.00 | 3.13 |
| Produktionskette | production chain | Produktion | production | Kette | chain | 2,738 | 579,419 | 207,479 | 244 | 139 | 2 | 4 | HAVE | 4.69 | 3.19 |
| Schneekette | snow chains | Schnee | snow | Kette | chain | 5,167 | 324,839 | 207,479 | 95 | 139 | 1 | 4 | INST | 4.19 | 4.21 |

