Meaning Representation with Multilayered Extended Semantic Networks

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Abstract

Multilayered Extended Semantic Networks (MultiNet) have been developed along the general line of semantic networks (SN) for the semantic representation of large stocks of natural language information. They allow for a very differentiated meaning representation of natural language expressions and an adequate modelling of cognitive structures. MultiNet has been used for the semantic characterization of lexemes in a large computational lexicon and as a semantic interlingua in natural language interfaces and question-answering systems. Apart from the structural information defined by relations and functions over the nodes of the SN (which is a feature common to all SN), MultiNet is characterized by embedding its conceptual nodes into a multidimensional space of layer attributes and differentiating between an intensional and a preextensional level within the knowledge representation itself. The paper gives an overview of the expressional means of Multi-Net and their use for representing linguistic knowledge and world knowledge.

Introduction

Multilayered Extended Semantic Networks - the so-called MultiNet Paradigm - are both, a formal language to describe the meaning of natural language expressions and a general knowledge representation formalism. They have been developed along the line of semantic networks starting with the work of Quillian (Quillian 1968) and are fully described in (Helbig 2005). One of the design principles of Multi-Net has been the Homogeneity Criterion ((Helbig 2005), p. 4) requiring that a knowledge representation system to be used for natural language processing must be appropriate to describe lexical knowledge as well as world knowledge, linguistic knowledge as well as inferential knowledge. With regard to the computational lexicon HaGenLex, which is one of the largest semantically based lexica (Hartrumpf, Helbig, & Osswald 2003), the expressional means of MultiNet provide the semantic backbone because they are used to specify the meaning structure of lexemes as well as their selectional restrictions. One of the main features of HaGenLex distinguishing it, for instance, from WordNet (Fellbaum 1998) is the fully formal description of the lexemes. Since there is a biunique correspondence between lexicalized concepts and lexemes in our approach, all statements about the semantic representation of concepts in MultiNet have an immediate effect on the characterization of lexemes of HaGenLex. For

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further details, the reader is kindly referred to the cited book (Helbig 2005), Chapter 12. Here, we shall only concentrate on the basic ideas.

MultiNet networks (or for short: Multinets) are hypergraphs the nodes of which represent concepts, and the arcs between the nodes represent relations and functions establishing a semantic connection between these nodes. In contrast to other network formalisms (e.g. KL-ONE (Brachman & Schmolze 1985), SNePS (Shapiro & Rapaport 1992), the arcs are labelled by relations and functions stemming from a strictly fixed set of representational means. (For examples see Table 2.) These relations and functions can be represented as nodes of a second semantic network at a metalevel which are connected with axioms and inference rules describing the logical properties of these metalevel constructs. One characteristic of MultiNet distinguishing it from simple networks and also from Sowa's Conceptual Structures (Sowa 1984) is the rich inner structure of the nodes. Every node of the network belongs to a sort from a given ontology, and nodes are assigned special semantic features. In addition to that, nodes are embedded in a multidimensional space of laverattributes and their values.

The Representational Means of MultiNet

Sorts

MultiNet distinguishes the following classes (sorts) of conceptual entities. The class of all nodes (entities) is [ent]:

(1) Objects [*o*]: There are two types of objects, concrete objects which can be sensually perceived, and such objects for which that is not true, the **abstract objects**.

• Concrete objects [co], which are divided into Substances [s] (Examples: milk, honey, iron) and Discrete objects [d] (Examples: house, apple)

• Abstract objects [ab]:

- Abstract situations [*abs*] comprising Abstract dy-namic situations [*ad*] (Examples: race, robbery, integration, movement) and Abstract static situations [as] (Examples: calmness, equilibrium, awareness, sleep) - Attributes [at]: with measurable attributes like height, weight, length, (subsort [*oa*]), and other attributes like form, trait, charm, (subsort [na]).

- Relationships [re]: Examples: analogy, synonymy
- Ideal objects [*io*]: Examples: religion, mercy, justice Abstract temporal objects [*ta*]: Examples: Renais-
- sance, Middle Âges, Easter, holiday, Paleozoic era
- Modalities [mo]: Examples: probability, necessity, intention, permission, desirability

(2) Situations/States of affairs [*si*]: Situations or states of affairs mirror the constellation of objects in space and time. We distinguish;

• Static situations (states) [st]: This sort comprises physical states as well as psychic states. Examples: hunger, illness, drought, tiredness

• **Dynamic situations (events)** [*dy*]: These situations are further classified into

- actions [da] (Examples: work, write, sing, go, sell)

- happenings [dn] (Examples: rain, decay, shine)

(3) Situational descriptors [sd]:

They are seldom lexicalized and often represented by (deictic) adverbial constructs

• Times [t], Examples: yesterday, Monday, tomorrow

• Locations [*l*], Examples: here, there

• **Modalities** [*md*]: They express the position of the speaker with regard to the validity of states of affairs or situations.

Examples: probably, impossible, necessary

(4) Qualities [*ql*]: Qualities or specifications of properties can be classified best by an opposing comparison (see Table 1).

Qualities (in general) $[ql]$				
total qualities [<i>tq</i>]	associative qualities [aq]			
• can be predicatively	• not predicatively usable			
used in natural lan-	in natural language;			
guage;				
Examples: dead,	Ex.: chemical, Newto-			
empty, green, square	nian, philosophical			
gradable qualities [gq]	operational qualities [<i>oq</i>]			
• obtain their full mean-	• they describe the posi-			
ing only in connection	tion in a sequence or are			
with other conceptual	operationally defined;			
objects;				
• predicative use allowed	• they are only defined			
for the corresponding	over generic concepts or			
natural language terms;	pluralities;			
Examples: big, deep,	Examples: fourth, last,			
good, expensive	next, middle			
Relational qualities [rq]				
• they have to be interpreted as relations;				
• they are only usable in connection with pluralities				
and collective concepts				

Examples: inverse, equivalent, similar, estranged

Table 1: Classification of qualities

(5) Quantities [qn]: They express the quantitative aspect of concepts.

- Quantificators [*qf*]: The quantificators are divided into Numerical quantificators [*nu*], Ex.: one, two, ... five
- Nonnumerical quantificators [*nn*], Ex.: all, many, few
- Units of measurement [me] Ex.: kg, meter, mile
- Measurements [m] Ex.: 3 kg, many miles
- (6) Graduators [gr]: Graduators are used for a more detailed specification of properties and quantities. There are
 Qualitative graduators [lg]: Ex.: very, rather
 - Quantitative graduators [ng]:, Examples: almost, nearly, approximately
- (7) Formal entities [*fe*]: In the context of the lexicon, only strings as elements of sort [*fe*] play a certain role for describing proper names.

Sorts together with semantic features are used in the lexicon to describe the semantic properties of lexemes and their selectional restrictions. The sorts are also important for the definition of the formal properties (especially the signatures) of relations and functions. (See Table 2.)

Layer attributes.

Apart from sorts and features, the concepts of Multinet (and therefore also the lexemes of HaGenLex) are classified according to their assignment to certain conceptual layers which are defined by specific layer attributes and their values. MultiNet distinguishes the following layer attributes (where only the first six are used in the lexicon):¹

• FACT: The **facticity** of an entity, i.e. whether it is really existing (value *real*, Example: IBM), not existing (value *nonreal*, Example: UFO), or only hypothetically assumed (value *hypo*, Example: string.1.3 - concept from astrophysics). This attribute is still more important to distinguish real facts from hypothetical assumed situations. (The latter typically occur in the meaning representations of conditional sentences.)

• GENER: The **degree of generality** indicates whether a conceptual entity is generic (value *ge*, Example: admiral) or specific (value *sp*, Example: Nelsson).

• QUANT: The intensional aspect of **quantification** specifies whether the concept is a singleton (value *one*) or a multitude (value *mult*) with the subtypes *fquant* and *nfquant* for fuzzy respective non-fuzzy quantifiers. Examples: **a**(**n**) [QUANT *one*], **several** [QUANT *mult*], **many** [QUANT *fquant*], **all** [QUANT *nfquant*],

• REFER: The **determination of reference**, i. e. whether the concept determines the reference (value *det*) or not (value *indet*). This type of characteristic plays an important part in text processing, especially for reference resolution.

Examples: this [REFER det] – a(n) [REFER indet]

• CARD: The **cardinality** characterizes the extensional aspect of a multitude. Such cardinalities are useful, among others, for the disambiguation of coreferences. In the lexicon, they are only used in connection with numerals transferring their cardinality to the representative of a complex noun phrase (when used in such a combination). Examples: **three** [CARD 3], **five boys** [CARD 5]

• ETYPE: The **type of extensionality** of an entity with values: nil – no extension, 0 – individual which is no set (e.g. Napoleon), 1 – entity with a set of elements from type [ETYPE 0] as extension (e.g. crew, family) etc.

• VARIA: The **variability** describes whether an object is conceptually varying (value *var*) or not (value *con*). Example: *"This dog* [VARIA *con*] *bites every cat* [VARIA *var*]."

Relations and Functions

MultiNet provides a collection of about 140 relations and functions to describe the interconnections between concepts (i. e. nodes of the semantic network). An overview of the signatures and strongly abbreviated definitions of relations which are typically used for the semantic representation of situations or states of affairs is given in Table 2.

In principle, all of these formal constructs can be used to describe lexical entries (lexemes) too. By means of

¹It should be noticed that the full potential of the layer attributes and their use can only be recognized in the combination of determiners and quantificators with lexemes designating concepts. This interplay is explained and dealt with in greater detail in (Hartrumpf & Helbig 2002).

Relation	Signature	Short Characteris- tics		
AFF	$[si \cup abs] \times [o \cup si]$	Affected object		
AGT ATTR	$ \begin{bmatrix} si \cup abs \end{bmatrix} \times o \\ [o \cup l \cup t] \times at $	Agent of an action Specification of an at-		
AVRT	$[dy \cup ad] imes o$	tribute Event averting/tur-		
		ning away from an ob- ject		
BENF	$[si \cup abs] \times [o \setminus abs]$	Benefactee of a situa- tion		
CIRC	$si imes [ab \cup si]$	Relation between sit- uation and circum-		
CAUS	si' imes si'	stance Causality		
COND DPND	$\tilde{si} \times \tilde{si}$	Conditional relation		
	$ent_{ext} \times ent_{ext}$	Dependency of two concepts		
EXP	$[si \cup abs] \times o$	Experiencer of an event		
INIT	$[dy \cup ad] \times [o \cup si]$	Relation specifying an initial state		
INSTR	$[si \cup abs] \times co$	Instrument of an ac-		
MCONT	$[si \cup o] \times [o \cup si]$	tion Relation between a mental process		
METH	$[si \cup abs] \times [dy \cup ad \cup io]$	and its content Method used in an ac-		
MEXP	$[st \cup abs] \times d$	tion Mental carrier of a		
	~	state or process		
MODL	$si \times md$	Relation specifying a restricting modality		
OBJ	$si \times [o \cup si]$	Neutral object in a sit- uation		
OPPOS	$[si \cup o] \times [si \cup o]$	Entity being opposed by a situation		
ORNT	$[si \cup abs] \times o$	Orientation towards		
PARS	$ [co \times co] \cup [io \times io] \cup \\ [t \times t] \cup [l \times l] $	something Part-whole relation		
POSS	$\begin{bmatrix} \iota \land \iota \end{bmatrix} \cup \begin{bmatrix} \iota \land \iota \end{bmatrix} \\ o \times o$	Relation between pos-		
PROP	o imes p	sessor and possession Relation between ob-		
PURP	$si \times [o \cup si]$	ject and property Purpose of an action		
RSLT SCAR	$[si \cup abs] \times [o \cup si]$	Result of an event Carrier of a state		
SSPE	$ \begin{matrix} [st \cup as] \times o \\ [st \cup as] \times ent \end{matrix} $	Entity specifying a		
SUB	$[o \setminus abs] imes [\overline{o} \setminus \overline{abs}]$	state (State specifier) Relation of concep- tual subordination		
CLID C		(for objects)		
SUBS	$[si \cup abs] \times [\overline{si} \cup \overline{abs}]$	Relation of concep- tual subordination		
TEMP	$[si \cup t \cup o] \times [t \cup si \cup abs]$	(for situations) Relation specifying the temporal embed-		
VAL	$\dot{at} \times [o \cup qn \cup p \cup fe \cup t]$	ding of a situation Relation between at- tribute and its value		
		L		

Table 2: Relations for the description of lexical entries Sort symbols can be marked by the following signs:

 \overline{o} – generic concept with [GENER ge];

 \dot{o} – individual concept with [GENER *sp*];

 \tilde{o} – hypothetical entity with [FACT *hypo*].

the attribute NET shown in Figure 3, all lexemes can be embedded into a larger semantic network to describe their meaning. Thus, the NET entry of the lexeme/concept verarbeiten.1.1 of Figure 3 (English meaning: "to produce something from something") indicates that the affected object x2 (relation AFF) is a raw material and the result (relation RSLT) is a product. Table 2 shows important relations and functions of MultiNet used in this paper.

These expressional means (shortly subsumed under the term "relations") can be divided into different groups:

• Relations describing situations, among them such relations like AGT (Agent), OBJ (neutral object), AFF (affected object which is changed by an event) etc. They are also used to describe the argument structure (the valencies) of content words.

• Relations characterizing the interconnection between situations, like COND (conditional relation) or CAUS (causality).

• Set relations describing the connection between the extensions of different concepts, like ELMT (element relation), *DIFF (set difference) etc.

• Relations characterizing conceptual objects, like SUB (subordination of concepts), PARS (part-whole-relationship) or LOC (describing the location of an object).

• Relations comparing or opposing concepts. Among these we number:

(X SYNO Y) – X and Y are synonyms, (X ANTO Y) – X and Y are antonyms (this relation comprises the next three relations), (X CNVRS Y) – X and Y are converse, (X COMPL Y) – X and Y are complementary, (X CONTR Y) – X and Y are contrary, (X ASSOC Y) – X and Y are semantically associated with each other.

• Relations expressing a change of sorts, where the third and the fourth letter in the name indicate the sorts from which respective to which the change is carried out. The letters have the following meaning: A – sort [*ab*] (abstract concept), E - [dy] (event), P - [p] (property), S - [st] (state). Thus the following relations hold: (long CHPA length), (hot CHPE (to) heat), (contain CHSA content), (ill CHPS illness).

One great advantage of MultiNet is the fact that the same expressional means used in the lexicon are also used to describe whole knowledge bases by conceptual networks. By that there is a direct way from the lexical representation of word senses, via the results of the syntanctic-semantic analysis of natural language expressions into the knowledge base representing a large stock of natural language information and also to the semantics of question-answering (Hartrumpf 2004). The expressional means of MultiNet are also the carriers of logical inferences and axiomatic rules as shown in the description of the INSTR relation below.

Every relation is described by means of the same schema having the following structure:

relational-definition :: = [$\langle name \rangle$, $\langle algebraic signature \rangle$, $\langle verbal definition \rangle$, $\langle mnemonic hint \rangle$, $\langle question patterns asking for that relation \rangle$, $\langle explanation consisting of commentaries and axioms \rangle$]

Table 3 gives a short description of the relation INSTR connecting a specific participant playing the cognitive role of an instrument with an event.

To support an effective work with MultiNets a graphical tool MWR for the knowledge engineer has been developed. It allows for an automatic generation of semantic networks by writing a natural language expression in the input field of MWR (see Figure 1) and calling the syntactic-semantic

C-Role – Instrument

[Deep case relation]

INSTR: $[si \cup abs] \times co$

Definition: The relation (s INSTR o) establishes a connection between an action s and the instrument o which is used to carry out s.

Mnemonics: instrument – (Ge: Instrument) (s INSTR o) – [s is carried out/sustained with o]

Question pattern: What instrument/tool is used for s? By means of which $\langle o \rangle$ [be] s carried out/sustained? Through what medium $\langle o \rangle$ [be] $\langle ent \rangle$ transferred?

Explanation: As it can be seen from the definition, instruments are always concrete objects. To represent the abstract aid by means of which an action is carried out or a state is sustained, the relation METH is also provided in MultiNet. Examples:

• Peter [drives]^{INSTR_{arg1}} with [his car]^{INSTR_{arg2} through the town.}

• $Max [draws]^{INSTR_{arg1}}$ the figure with [the computer]^{INSTR_{arg2}}.

The simultaneous occurrence of INSTR and METH in one and the same situation is shown by the following example:

• The results had been obtained [with a PASCAL program]^{INSTR_{arg2}} following the [method of Cooley-Tucker]^{METH_{arg2}}.

 $\begin{array}{l} \text{The following connection can be established between} \\ \text{INSTR and the relation PURP representing a purpose :} \\ \bullet (v_1 \; \text{SUBS s}) \land (v_1 \; \text{AGT } k_1) \land (v_1 \; \text{INSTR } k_2) \rightarrow \\ \exists \; v_2 \; (v_2 \; \text{SUBS use}) \land (v_2 \; \text{AGT } k_1) \land \\ \quad (v_2 \; \text{OBJ } k_2) \land (v_2 \; \text{PURP } v_1) \end{array}$

Table 3: Abbreviated definition of the instrument relation

analysis. The result of the semantic interpretation is automatically displayed in graphical form on the main window of MWR.

MWR can also be used by a lexicographer out of his workbench LIA (Hartrumpf, Helbig, & Osswald 2003) to specify the NET attribute of lexical entries. In that, the lexicographer is simply calling the MWR tool by clicking on the NET entry of a given lexeme which is shown and edited by means of LIA. MWR represents this entry as a semantic net, and the network can be graphically manipulated with MWR and stored back into the lexicon in an internal format.

For the sake of illustration, Figure 1 shows the semantic representation of the well-known Donkey-sentence "*Every farmer who owns a donkey beats it.*" automatically created by means of MWR. The values of the layer attributes of "donkey-node" c3 express that c3 is a specific concept [GENER = sp] whose representative at the preextensional level is varying [VARIA = var] in dependency of the extension of the farmer node. (The dependency is expressed by the DPND relation).

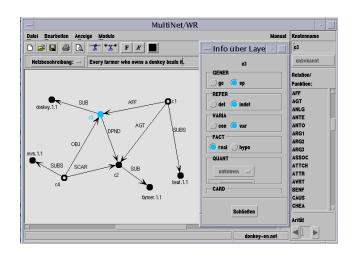


Figure 1: Semantic representation of a Donkey-sentence with layer information of node $c3 = \langle a \text{ donkey which is beaten by a farmer} \rangle$

Lexical Knowledge and Semantic features.

On the one hand, the classification of concepts by sorts is not fine-grained enough to describe all semantic restrictions for constituents which are potential fillers for the slots opened by the valencies of lexemes, e.g. there is no MultiNet sort distinguishing the agents admissible to the German word "essen" – English "eat (human)" from word "fressen" – English "eat (animal)" (\rightarrow [human +] vs [animal +]).

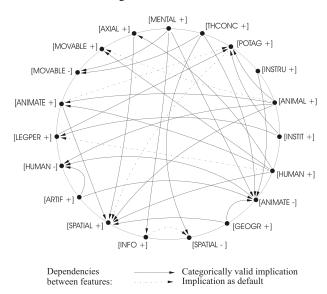
		Example values	
Name	Meaning	+	_
animal animate artif axial geogr human info	living being artifact object having an axis	fox tree house pencil the Alps woman book	person stone tree sphere table ape grass
instit instru legper	institution typical instrument juridical/natural per- son	UNO hammer firm	apple tiger animal
mental method movable potag spatial	mental concept method movable object potential agent object with spatial extension	pleasure recursion car motor table	length book forest poster idea
thconc	theoretical concept	physics	pleasure

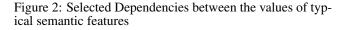
Table 4: Features for the semantic fine-characterization of objects (Hartrumpf, Helbig, & Osswald 2003)

On the other hand, the application of the full repertory of expressional means of MultiNet for describing the selectional restriction of content words – even if principally possible –

would be to unwieldy. For that reason, a compromise has been found in using typical semantic features (or semantic markers) which have over a long time been proven as powerful enough to govern large classes of semantic selections. It must be emphasized, however, that all these features could be omitted and expressed by concepts and relations of Multi-Net.

The features allow a very fine-grained specification of lexemes to characterize the most important selectional restrictions of verbs, nouns and adjectives. They also warrant an effective syntactic-semantic analysis, which should not be overburdened with logical inferences.





As it can easily be seen, the values of the semantic features for a certain lexeme are not independent of each other. Figure 2 shows some of the most important dependencies which can be used in the inheritance based lexicon as well as during syntactic-semantic analysis.

Encapsulation of Partial Networks

It is important in a knowledge base to distinguish that part of knowledge which is defining the meaning of a concept (*definitional knowledge*) from that part where the concept in question is only mentioned (*assertional knowledge*). This distinction can also be met in other formalisms as for instance in SB-ONE (Allgayer & Reddig-Siekmann 1990). In MultiNet the definitional knowledge characterizing a concept C is comprised in a concept capsule attached to the semantic representative of C. In Figure 4 the capsule (which is drawn as a box) in the upper part characterizes the generic concept van while the capsule in the lower part describes the individual concept (a van built by Ford).

This distinction is important during inferential answer finding to select that knowledge which provides the basis for answering questions like "*What is a van*?". Orthogonally to this classification there is another partition of knowledge in MultiNet discerning between *categorical knowledge* with regard to a concept (marked by K in the graphical representation of Figure 4), *prototypical or default knowledge* (marked by D) and *situational knowledge* (marked by S).

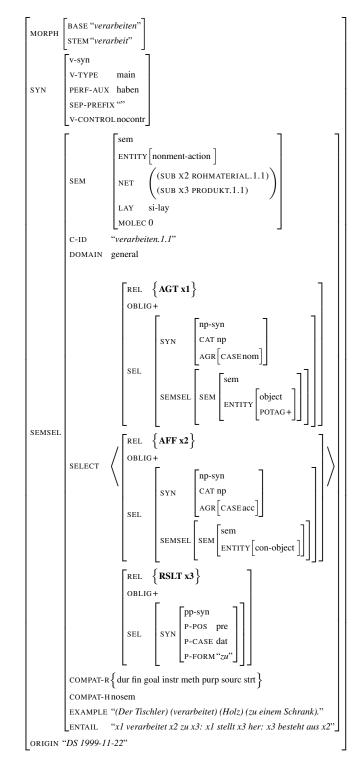


Figure 3: Formal characterization of the German lexeme verarbeiten.1.1 which corresponds to the reading "*to process something*"

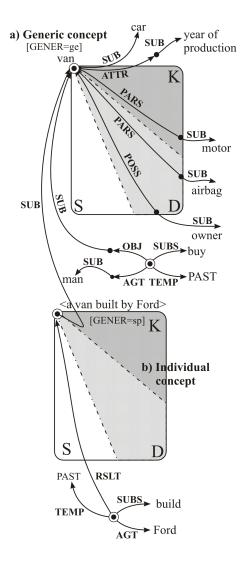


Figure 4: Partial networks (concept capsules) defining the extent of meaning of two different concepts

Taking the concept van from Figure 4 we have the following distinction: That a van is a car having a motor and a year of production is categorical knowledge. Having an airbag as a part is assumed as default knowledge for a van. Default knowledge and categorical knowledge together constitute that knowledge which is immanently connected with a concept. Categorical knowledge is connected with monotonic reasoning, while default knowledge has to be treated with methods of non-monotonic reasoning. As mentioned already, there is a third kind of knowledge, the so-called situational knowledge. In Figure 4 the information that "Ford built a van" is classified as situational knowledge with regard to this special van. Often (but not always) situational knowledge and assertional knowledge coincide. However, the lower capsule shows that the situational knowledge that a certain van has been produced by Ford must be classified as definitional knowledge with regard to this special concept node. The role of this cross-classification into different knowledge types for the answering of special question classes (especially for different types of so-called "Essay questions") is explained more detailed in (Glöckner & Helbig 2005).

Conclusion

MultiNet is one of the most comprehensively described knowledge representation paradigms which have been used in practically relevant NLP applications.² It is also the basic formalism for the description of lexemes/concepts in one of the largest semantically oriented and formally described computational lexica. MultiNet is distinguished from other semantic network formalisms with regard to its expressional means by a multilayered structure, a rich inner structure of nodes and encapsulation of partial networks to build complex concept descriptions. The MultiNet formalism provides also the target language for a natural language interpreter which has been used (and still is in use) for the semantic annotation of large NL corpora with millions of sentences and for question-answering over these corpora.

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²http://pi7.fernuni-hagen.de/research/research.html#Projects