

# Theoretical foundations of natural language programming and publishing for intelligent agents and robots

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**Abstract**—This paper is an application of ontology theory, conceptual graph and programming languages theories to develop the foundations of natural language programming (NLP) that can be used to produce natural language documents for intelligent agents and human readers. The analysis given reveals three benefits of NLP. First, it is a “conceptualized” programming that enables developers to write less bug prone programs due to clarity of code presentation and enforced structuring of data. Secondly, NLP can aid programming of the all important abstractions for robots: event, action and world model abstractions can be created by sentences. Thirdly, NLP can be used to publish natural language documents by researchers, e.g. English language documents, on the Internet or also in print that is actually code. The paper introduces a large class of intelligent agents that can read such documents. This enables human users and agents a shared understanding of how application systems work.

**Index Terms**—Computer programming, intelligent physical agents, robots, conceptual graphs, machine knowledge.

## I. INTRODUCTION

The paper presents the theoretical background to an information processing system [1] for autonomously operating robotic agents where engineers can use NLP (natural language programming) documents to define their behaviour and later the agents can read new published documents to acquire knowledge such as various physical skills, facts about the environment and also behaviour policies, goal priorities. The NLP documents concerned are similar to engineering booklets with contents, sections, subsections in English, that can appear in HTML, LaTeX(pdf) formats (see numerous examples at [www.system-english.com](http://www.system-english.com)).

Use of NLP documents shortens the route to robotic software update. There is no need for an application engineer to read a journal publication and implement it by programming. The research engineers methodological work is directly utilized by the autonomous vehicle or robot reading NLP documents. Engineers whom are different from the author of the publication can also read the papers and learn the details of how the vehicle operates, how decisions are reached and how skills are performed. Not only will users of the autonomous vehicle clearly understand how it operates but will also know its limitations to avoid misuse or misunderstanding. Users can modify English sentences in the publication to influence the vehicles behaviour or how its skills are performed. Also the legal responsibility for misuse can be shifted to the user.

The process of knowledge distribution among scientists and engineers is by way of publishing at conferences, in journals and books. They describe in technical papers how a control method works, how a navigation method proposed by an author works. Also information processing and decision making procedures are published by research engineers (REs) [1]. Other engineers, lets call them application development engineers (ADEs), read these technical papers, conceptualize the meanings and may opt to implement the methods in practice. There are thousands of control and signal processing engineering papers written every year worldwide. Their impact on industry is patchy as many of the papers are not read or chosen to be ignored due to financial necessities of companies. Knowledge transfer to industry (KTI) is slow as is well known. Cynics may even ask: why do we need to speed up KTI?... we are fine as we are. Others may say: this is a topic for our knowledge engineers ...but artificial intelligence is far from being able to help yet. Reference [2] describes principles and reviews systems developed to support software developers, that also act as mediating tools to provide a platform facilitating knowledge sharing. The need of supporting knowledge collaboration in software development environments, is highlighted in [2], based on the conceptualization of software development as knowledge-intensive and distributed cognitive activity. This principle is extended in [3] for the context of shared knowledge and collaboration between human users, i.e. scientists, engineers, and machines or intelligent agents. It is proposed that documents written in English are distributed to networked devices that human users can also read and hence develop shared understanding with the devices. Hence a system is created that allows humans and machines to communicate intuitively (i.e., by using human language) through intelligent cooperation. This is achieved by creating a coherent meaning-system understandable by both the devices and people. Is it however possible to make a machine to read a technical book written in English sentences to formulate methodologies at the conceptual level? Some would say: natural language interfaces have been developed and you can use various controlled English languages. These need a priori knowledge and definition of meanings .. and you need to know the grammar and what phrases to use .... the problem is a complexity nightmare! And yet this paper is describing an existing system whereby engineers can write documents in English that can also contain equations, figures and images, quotes, numbers, physical quantities, etc. that suitable intelligent agents can read and apply it in their daily work [1,3]. This paper describes a system

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with the following practical features: (1) The authors and readers of these machine readable documents (called system English papers, for short called sEnglish papers) do not need to learn grammar, just to apply common sense. The most important thing for an author is to have conceptual clarity in their area and express that in a suitable style: writing skills are needed. It only takes a few hours to be trained as an author. (2) The autonomous systems can read these special engineering publications in sEnglish. sEnglish publications are of special format with contents, sections and subsections. They may contain images, mathematical formulae to support understanding both to humans and to autonomous systems who read them. (3) The autonomous systems who read these documents do not need access to a central dictionary of meanings. There is no need for building up an infrastructure so that autonomous systems can read sEnglish books. The system can be used immediately, without any investment. Autonomous vehicles (AUVs, UAVs, AGVs and spacecraft), robots, toys can be built that understand natural language documents without a supporting infrastructure to be set up first. We hold the view that software downloads for intelligent autonomous systems are old-fashioned and outdated now. Autonomous systems and robots should read books on mental and physical skills and facts about the world from documents that their user can also read with ease and hence they can have a shared understanding of a selected part of the world.

In this publication system the manufacturer of a vehicle platform or robot can invite publications from research engineers to enrich the knowledge base of agents controlling these platforms in a format that also merits current standards of scientific and engineering publications. This could for instance be the case for the use of autonomous vehicles, robotic pets and intelligent toys, gardening robots or agricultural unmanned autonomous aerial vehicles. Publications for autonomous systems is a logical next step of information systems in a historical perspective [3]: (a) There was first verbal communication between people. (b) Writing was developed to record events and knowledge. (c) Later book was invented, i.e. multiple copies of written material were printed that masses of people could read. (d) Journal and newspaper publishing and the WWW exists to distribute knowledge. In the computer science community the idea of natural language programming has been largely considered impossible and impractical due to ambiguity. This paper reports about a complete system for autonomous systems unambiguously understanding human readable books to enhance their skills and knowledge about the world. There is a complete software system available today for NLP in sEnglish ('system English') An complete authoring tool (sEAT) [2] and sEnglish reader agent installation system (sERA) [3] is available. When installed on a PC, the reader agent sERA can read web pages that are sEnglish documents and execute sentences and complex actions sequences as instructed by a human instructor of the reader agent. sERA is simple reactive agent (sERA) that understands these papers and can execute commands in sentences.

An important part of NLP documents is their definition of an ontology that is the bases of statements in sentences. As contributive factors to the development and study of ontologies for knowledge sharing and reuse are reviewed in [4]. The following definition is generally accepted by the research community: "Ontologies are explicit formal specification of a shared conceptualization" [5], [6] where: "explicit" means that the type of concepts used, and the constraints on their use are explicitly defined [7]; "formal" means that the ontology should be machine readable; "shared" reflects the notion that an ontology captures consensual knowledge, that is, it is not private to some individual, but accepted by a group. Conceptualization emphasizes the abstract model of some phenomenon in the world by having identified the relevant concept of that phenomenon" [7]. Conceptualization refers to the objects, concepts, and other entities that are assumed to exist in some area of interest and the relationships that hold among them [8]; i.e. addresses the abstract, simplified view of the world that is required for representation [9].

Reference [4] describes the hierarchical abstraction layers of programming and points out that block diagram based abstractions of programs are ill placed, they are too low resolution. Eventually the author advocates the development of "Developmental Mathematics" which can be a complex process. In this paper we neatly fit NLP into the abstractions layers of programming. As shown above NLP provides simultaneous human (through the use of natural language) as well as machine interpretation (translated code) of an NLP sentence. Note that code unambiguity (of NLP sentences) preserves the cornerstone of digital computing: the usefulness of digital computation is its determinism, i.e. its reliability in banking, accounting and computer control of aircraft, machines and robots, etc.

This paper is organized as follows. The next section provides definitions of ontologies. This is followed by a model of human thinking in terms of conceptual graphs. Section IV introduces the NLP document and Section V provides examples. Section VI briefly discusses levels of abstractions for sentences. Section VII is concerned with procedural knowledge sharing among a set of agents. Shared knowledge description is concluded with the involvement of the human operator of the agents that is followed by conclusions.

## II. FUNDAMENTALS

A natural language program document (NLPD) consists of three components: an ontology, a set of sentence templates and an underlying instructional (imperative) programming language. The process of creating and making NLPDs is described in a flow diagram in Figure 1. An expert author of an algorithmic field (for instance in subfields of signal processing, control, science and logic inference) produces an ontology and provides hierarchical definitions of sentence meanings using a programming language for low abstraction level semantics. The sentence meanings, ontology definitions and meta sentences to define the subject area composed into an HMTL document by a document creator algorithm.

Some of the web pages of this document are illustrated in Figures 2,3. The NLPD is utilized either by engineers who can read the document and borrow from it algorithm for their own work, or by an intelligent agent, typically controlling a vehicle, machine, appliance or a robot in general. Machines reading publications directly can shorten the route how results are made useful in intelligent machines - there are no upgrades needed by engineers. The agents seek out their knowledge themselves. This section will first provide formal definitions for the ontologies used and also for the relationship with the underlying programming language. Next formal definitions of human interpretations are given in terms of conceptual graphs. Finally NLP documents are defined and illustrated.

First a formal definition of a traditional ontology is provided here that is short of using the description logic (DL,OWL, see W3C.org) framework as that will not strictly be required for natural language programming.

**Definition 2.1:** An *agent's modelling ontology* is a description logic  $M_{DL} = \{\Gamma, R, =, I, \cup, \cap, \neg, \exists, \forall\}$  with atomic classes  $\Gamma$ , roles  $R$  so that satisfy the following axioms:

- (a) Subclass relationship  $A \supset B$  is defined by  $A \cup B = B$ .
- (b) There are atomic classes *perceived object*, *imagined object*, *mathematical object* that are disjoint and satisfy *modelling object* = ( *perceived object*  $\cup$  *imagined object*  $\cup$  *mathematical object* )
- (c) There is an atomic role *has attribute*.

In NLP the individuals created from classes of  $O$  will become examples of human concepts as represented on a digital computer. To achieve this the  $O$  will be linked to some (popular) imperative programming language: the ontology  $O$  is required to have a subset of classes that are classes in the programming language and any class in  $O$  are subclasses of some basic types (classes) in the programming language. This will be formally described as follows. First we restrict the definition of an ontology that NLP will use.

**Definition 2.2:** A *restricted ontology*  $O = \langle \Gamma | \prec | @ | \Lambda \rangle$  consists of a lattice  $\langle \Gamma | \prec \rangle$  over the class set  $\Gamma$ , an attribute label set  $\Lambda$  and an attribute mapping  $@ : \Gamma \rightarrow 2^{\Gamma(\Lambda)}$  where  $\Gamma(\Lambda) : \Lambda \rightarrow \Gamma$  is a mapping from the attribute label set to the class set  $\Gamma$ . The class set has a universal sup class  $\xi_0$  that is the *modelling object* so that  $\forall \xi \in \Gamma : \xi \prec \xi_0$  and a *universal sub model class*  $\xi_\infty$  such that  $\forall \xi \in \Gamma : \xi_\infty \prec \xi$ . The  $@$  is also required to satisfy the inheritance condition  $\forall \xi, \zeta \in \Gamma : \xi \prec \zeta \Rightarrow @(\xi) \supseteq @(\zeta)$ .

Let  $N = \langle B, L, D, H, P \rangle$  be an instructional (imperative) programming language (can be either object oriented or not) that has a set of basic types  $B$ , a set  $L$  of syntactically correct sets of instructions, declarations  $D$ , subroutines  $H$  and programs  $P$ . The ontology  $O$  is required to be such that any class of  $O$  is a subclass of a type (class) in  $B$ , i.e. of a basic type in  $N$ , as given in the following definition.

**Definition 2.3:** Let  $N = \langle B_N, L_N, D_N, H_N, P_N \rangle$  be an instructional computer language that has a set of variable types  $B_N$ . An ontology  $O = \langle \Gamma | \prec | @ | \Lambda \rangle$  is said to be *supported by*  $N$  if  $B_N \subset \Gamma$  and  $\forall \xi \in \Gamma \exists \zeta \in B_N : \xi \prec \zeta$ .

An ontology  $O$  supported by a programming language  $N$

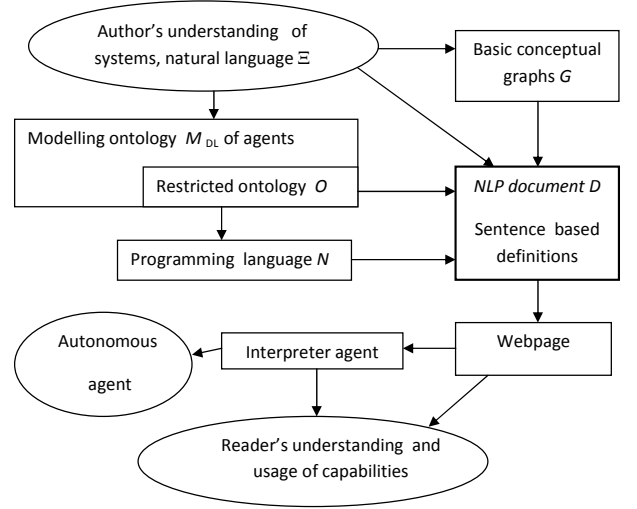


Fig. 1. Flow diagram of NLP document creation and usage.

will often be indexed as  $O_N$  to express this relationship. For instance if the underlying programming language is MATLAB, we will use the notation  $O_M$ , if it is C or C++, then we can use  $O_C$  or  $O_{cpp}$ , respectively. Next a possible relationship of an ontology  $O_N$  to conceptual graphs (CG) will be described.

### III. CONCEPTUAL GRAPHS OF HUMAN THINKING

Formalizing human thinking will not be attempted here, CGs will be used instead to represent human thought as it is generally accepted [10], [5] that CGs are powerful representations of human thought. For completeness a simple and general enough definition of conceptual graphs will be provided here.

**Definition 3.1:** Let  $C$  and  $R$  be sets of classes from two disjoint ontologies  $O_c$  and  $O_r$  respectively. A *basic conceptual graph*  $G = \langle G_v, G_e, L_v \rangle$  is a vertex-labelled directed-bigraph that satisfies the following conditions:

- (a) the vertices  $G_v$  are decomposed into two sets of *concepts*  $C$  and *conceptual relations*  $R$  so that  $G_v = C \cup R$ ;
- (b) the directed edges of  $G$  occur in pairs so that  $G_e \subset C \times R \times C$ .
- (c) Vertices  $G_v$  are labelled by a *referent function*  $L_v : G_v \rightarrow R$  where  $R$  contains of existential, individual, set, universal and question labels as listed in Table 1.

The set of basic conceptual graphs is denoted by  $B_{cg}(O_c, O_r)$ .

Some basic CGs are based on perception of the surrounding physical world and are direct abstractions from perception processes of an agent. Other basic CGs can be obtained through communication by other agents or by operations on conceptual graphs the agent has in memory. Classical work [10] on CG defined four basic operations on conceptual graphs to obtain new ones: (1) *copying*, i.e. creating an identical copy; (2) *restricting* a CG where concepts are replaced by conceptual subclasses or by individual

Referent	Formats	Examples
Existential unique ex. anaphoric	[con: *] [con: @1] [con: #]	DOG: * DOG: @1 DOG: #
Individual	[con: # num] [con: name]	DOG: 453 DOG: Vak
Set generic counted set anaph. set	[con: { $i_1, \dots, i_n$ }] [con: { * }] [con: { * } @ num] [con: { * } #]	DOG: { Vak, Spot } DOG: { * } DOG: { * } @ 4 DOG: { * } #
Universal negative plural set fuzzy set	[con: $\forall$ ] [con: $\forall \neg$ ] [con: { * } $\forall$ ] [con: { * } @ many]	DOG: $\forall$ DOG: $\forall \neg$ DOG: { * } $\forall$ DOG: { * } @ many
Question plural	[con: ?] [con: { * } ?]	DOG: ? DOG: { * }

TABLE I

DEFINITIONS OF REFERENT TYPES IN  $R$ .  $con$  STANDS FOR *concept* THAT IS AN ENTRY FROM  $O_v$  AND  $num$  STANDS FOR *number* THAT IS A NONNEGATIVE INTEGER. THE  $L_v$  MAPS A *concept* IN  $O_v$  TO ONE OF THE LABELS IN COLUMN 2 OF THIS TABLE.

markers; (3) *joining* where identical concepts of individuals are used in two graphs to create a third one that shares the individuals and keeps all other relations from the two graphs. (4) *simplifying*, that means removing duplicate conceptual relations from a graph.

Basic conceptual graphs are timeless and do not contain references to the context how the “thought” was generated, e.g. communication, belief, desire or whether it occurred in the past or will be valid in the future. These modalities are clearly important for human and also for artificial agents such as the belief-desire-intention agents. Context boxes [10], [5] are used to represent negation, modality and propositional attitudes like “know” or “believe” and indirect speech. Conceptual graphs have been extended by agent’s perspectives and temporal localization [6] where agent’s perspectives and temporal localizations are also described. Using these references we provide a sufficiently general definition of compound conceptual graphs (CCGs) to provide background for the concepts of natural language programming.

**Definition 3.2:** Let  $O_c$  and  $O_r$  be two disjoint ontologies for concepts and conceptual relations with associated basic conceptual graphs  $B_{cg}(O_c, O_r)$ . The set of compound conceptual graphs  $C_{cg}(O_c, O_r)$  is the smallest set that satisfies the following conditions:

- $B_{cg}(O_c, O_r) \subseteq C_{cg}(O_c, O_r)$ .
- Let  $T_l$  be a set of temporal labels. For any  $x \in T_l$ ,  $g \in C_{cg}(O_c, O_r)$  also  $T(x, g) \in C_{cg}(O_c, O_r)$ .
- Let  $T_r$  be a set of temporal relation labels on the real time-line. For any  $x \in T_r$ ,  $G, H \in C_{cg}(O_c, O_r)$  also  $T(G, x, H) \in C_{cg}(O_c, O_r)$ .
- Let  $A_l$  be a set of agent aspect labels. For any  $x \in A_l$ ,  $G \in C_{cg}(O_c, O_r)$  also  $A(x, G) \in C_{cg}(O_c, O_r)$ . The set of A-enveloped aspect graphs will be denoted by  $A(O_c, O_r) \subset C_{cg}(O_c, O_r)$ .
- Let  $G \in C_{cg}(O_c, O_r)$  and  $c$  a concept in  $G$ . The compound graph  $G'$  obtained from  $G$  by replacing all occurrences of  $c$  by a  $A \in A(O_c, O_r)$  is also a compound

graph:  $G' \in C_{cg}(O_c, O_r)$ .

Large parts of human knowledge can be modelled by sets of conceptual graphs (CGs) that can also have modalities of time, perception and communication, etc.

Here we will aim to formalise human thought processes but make the following assumption.

**Assumption 3.1:** Human knowledge base  $H_{KB} = \{G_i \mid i \in J\}$  about the present and the past of the world is represented by a time varying set of conceptual graphs with abstraction modalities of time, intention, desirability, likelihood and communication.

Now we come to the definition of a NLP text that are related to CGs.

#### IV. NLP DOCUMENTS

In the following an NLP *text* is understood as a sequence of ASCII characters. A *word* is a sequence of characters without space. A *proper name* is a word starting with a capital letter but not all words starting with capital letters are proper names, for instance all sentences start with capital letters.

**Definition 4.1:** Let  $\Xi$  be a natural language with word set  $\Xi_w$ . A *word* is a sequence of characters without space. A *proper name* is a single word that starts with a capital letter. An NLP *class name* is a finite sequence of words from  $\Xi_w$ . An *attribute name* is a finite sequence of words from  $\Xi_w$ .

- 1) An ontology  $O$  is called  $\Xi$  compliant if all of its class names and attributes are finite sequences from  $\Xi_w$ .
- 2) An NLP *sentence* is a sequence of words that starts with an initial capital and ends with ., ! or ?. A sentence may contain proper names, class names and attribute names as sub-sequences of the sentence. The set of all feasible, but not necessarily plausible, sentences is named  $\tilde{S}(\Xi, O, S)$ .
- 3) An NLP *text* is a finite sequence of NLP sentences. The set of all feasible, but not necessarily plausible, sentences is named  $\tilde{X}(\Xi, O, S)$ .

This definition with minimalist conditions sets constraints to form words and on sequences of words to form sentences and NLP text. The semantics of NLP meanings is however defined in terms of a broader data set to reflect context sensitivity of meanings. Semantics will be defined in terms of procedures run on a digital computer with input-output devices such as sensors and actuators.

To secure conceptual graph correspondence for each sentence  $s \in S$  defined by  $m$ , we also need a conceptual description function as defined.

**Definition 4.2:** For each  $s \in S$  the *conceptual description*  $d(s) \in \tilde{X}(\Xi, O, S)$  is a formalized NLP text description.

While we are not going into the details of an algorithm that parses NLP sentences to obtain conceptual graphs, we are making the following assumption.

**Assumption 4.1:** There is conceptual resolvent function  $F_{CG} : \tilde{X}(\Xi, O, S) \rightarrow B_{cg}(O_c, O_r)$  that uniquely identifies a  $g(s) = F_{CG}(d(s), s) \in B_{cg}(O_c, O_r)$  in terms of a CG with referenced vertices belonging to conceptual classes in  $O$  and

conceptual relations in  $O_r$  as permitted by  $B_{cg}(O_c, O_r)$  of  $\in H_{KB}$ .

For any set  $S$  the notation  $S^*$  will be used for the set of its sub-sequences.

**Definition 4.3:** Let  $N$  be a computer programming language with syntactically correct code set  $L_N$  and let  $O$  be an ontology compliant with natural language  $\Xi$ . A *meaning function*  $m : S \rightarrow S^* \cup L_N \cup \{\emptyset\}$  maps any sentence to a sequence of sentences or to a code in  $L_N$  or to the empty set to express that there is no meaning. An *interpreter* is a tuple  $I = \langle S, m, L_N \rangle$ . An interpreter  $I$  is called *semantically complete* if there is a *translator*  $T : S \rightarrow L_N$  so that

- 1) For all  $s \in S$  the  $m(s) \neq \emptyset$ .
- 2)  $T(m(s)) = \text{concatenation}(T(s_1), T(s_2), \dots, T(s_k))$  if  $m(s) = \{s_1, s_2, \dots, s_k\}$

Practically useful are sets of semantically complete sentences that use some ontology that is compliant with a natural language. As meaning of natural language sentences is context dependent and can be ambiguous, a most desirable property for sentence meaning is to retain the most important characteristics of *von Neumann architectures* that is determinism of code meanings. This is the property that makes today's computers reliable for bank transactions or airplane ticket search and reservation systems, etc. Determinism is however not defined globally, that would be untenable in the global world of knowledge, but locally in the context of a single NLP document.

**Definition 4.4:** NLP document is a tuple  $D = \langle \Xi, O, S, m, d, L_N, N \rangle$  of natural language  $\Xi$ , ontology  $O$ , sentence set  $S$ , meaning function  $m$ , conceptualization descriptions  $d$  and syntactically correct codes  $L_N$  (in terms of programming language  $N$ ), if the following are satisfied:

- $O$  and  $S$  are  $\Xi$  compliant.
- The interpreter  $I = \langle S, m, L_N \rangle$  is semantically complete.
- There is a unique translator  $T$  such that  $T(s) \in L_N, \forall s \in S$  and all  $s \in S$  are acceptable in  $\Xi$ .

The property of unique compilation is to preserve the main benefit of digital computers in that they are deterministic.

## V. EXAMPLES OF NLP DOCUMENTS

Figures 2,3 illustrate the web pages of an NLP document, sentences and document format for modelling of the environment and goals. Figure 4 shows the graphical user interface of the sEnglish authoring tools. The reader can find and download a number of NLP documents from [www.system-english.com](http://www.system-english.com).

## VI. LEVELS OF ABSTRACTIONS

Assume  $\Xi_p = \{O, S\}$  is a conceptual program. Any  $s \in S$  has an associated set of subroutine calls denoted by  $H(s) = \{h_i \in L_N, i = 1, \dots, n_s\}$ . Each  $h_i$  has a set of input object classes corresponding to ontology classes in  $\Gamma$  of  $O$ . An individual object under a class  $\gamma \in \Gamma$  is simply a structure with field names  $\{\lambda \mid \exists \chi \in \Gamma : \langle \lambda, \chi \rangle \in @(\gamma)\}$ .

An  $s \in S$  within a  $\Xi_p = \{O, S\}$  is called to have an *elementary meaning* if  $m(s) \in L_N$  and it is called to have

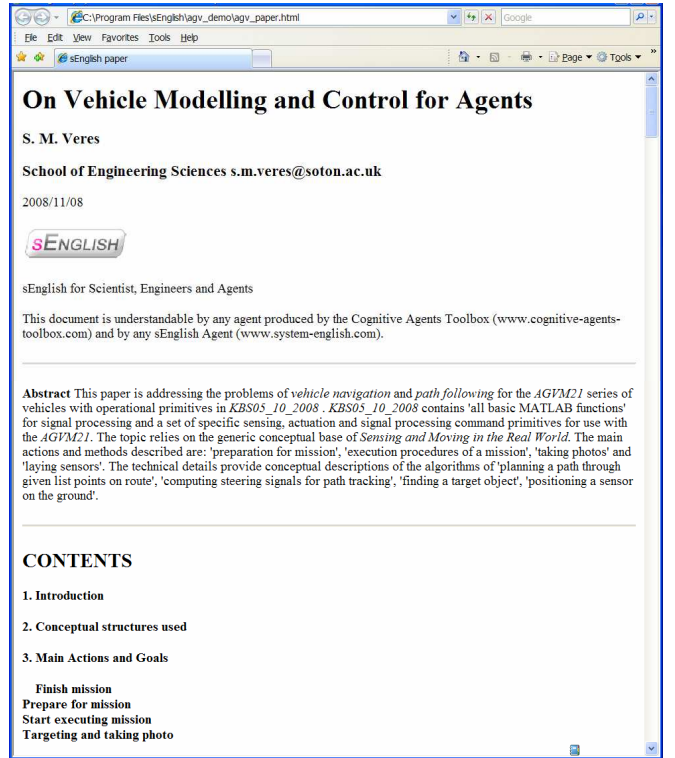


Fig. 2. Title page of an NLP document in “system English” that is short for sEnglish. See numerous examples of machine generated sEnglish documents at [www.system-english.com](http://www.system-english.com).

an *explained meaning* if  $m(s) = \{s_1, s_2, \dots, s_k\}$ ,  $s_i \in S, i = 1, \dots, k$ .

**Definition 6.1:** A sentence  $s \in S$ , that has elementary meaning only, is called *zero-level abstraction D-abstraction*. An  $s \in S$  is called to be of a *level k D-abstraction* if  $k$  is the smallest positive integer such that there is a sequence of  $s_i \in S, i = 0, 1, \dots, k-1$  such that  $m(s_0) \in L_N, s_i \in m(s_{i+1}), i = 1, \dots, k-1, s_k = s$ .

Zero level  $D$ -abstractions are also called *subconscious* abstractions. The term “ $D$ -abstraction” is short for “deterministic abstractions”. Although  $D$ -abstractions are unlike human abstractions in flexibility, they are abstractions as they abstract away the details into their developed meaning in terms of other sentences and eventually code in  $L_N$ . Note however that even zero-level abstractions abstract away from the individuals the meaning refers to as available objects (inputs): a zero level sentence  $s \in S$  is defined by the set of classes of its outputs and inputs in the corresponding calls  $H(s) = \{h_i \in L_N, i = 1, \dots, n_s\}$ . Hence zero level abstractions are defined by classes and not individuals.

Any  $D$ -abstraction at any level is well defined through a series of meaning definitions and hence deterministic. This property is a bit unlike human concepts where definitions are not as well determined. The connections between abstractions in an NLP  $\Xi_p = \{O, S\}$  and between human concepts, in terms of human ontologies  $\langle O_c, O_r \rangle$ , are best defined via the



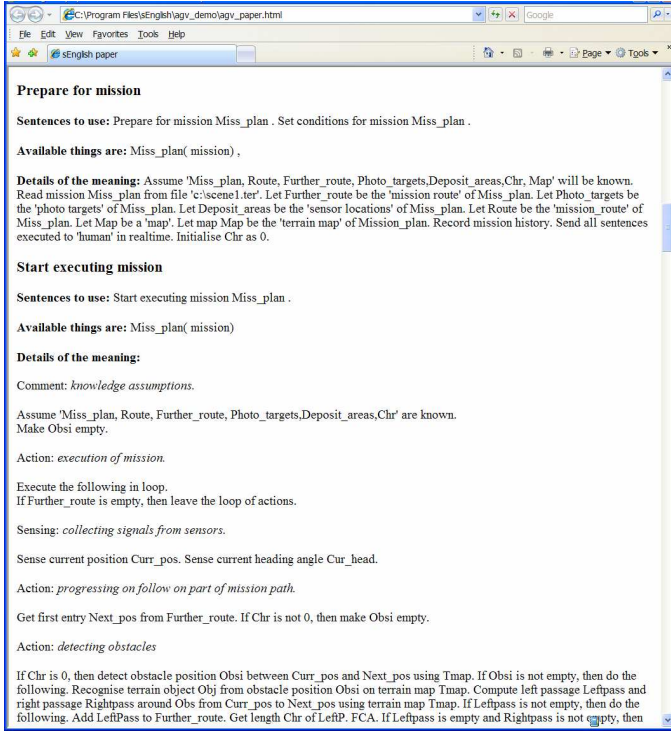


Fig. 3. Example section of an NLP document

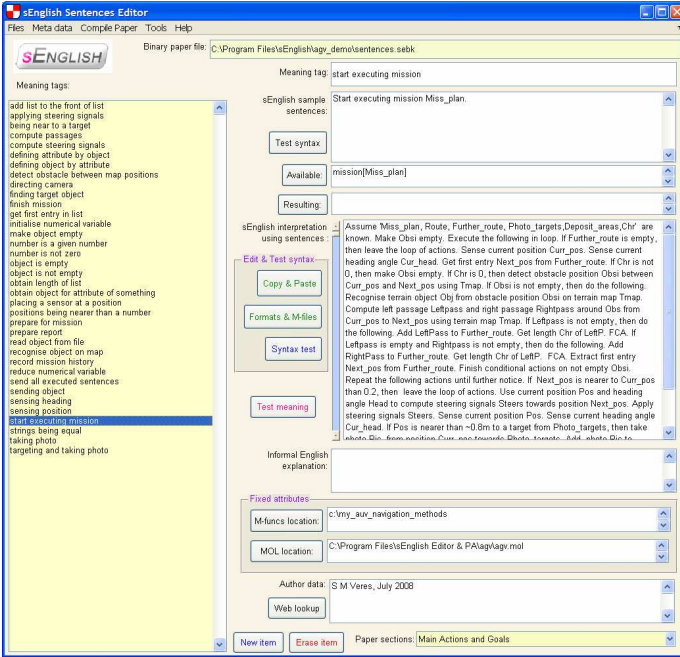


Fig. 4. Sentences editor for an NLP document that is part of the *Authoring Tools* for sEnglish documents.

human interpretation function  $\Psi$ .  $\Psi$ , as defined in Definition 8.1, associates a sentence with a set of conceptual graphs in the human knowledge base  $H_{KB}$ .

## VII. PROCEDURAL KNOWLEDGE SHARING AMONG AGENTS

To be able to define shared skills and knowledge among agents via NLP documents we provide formal definition of a very broad class agents [11], [12]. The term “skill” is used for procedural capabilities such as interaction with the environment and “knowledge” also includes models of relations in the current environment the agent is in.  $2^S$  will denote the set of subsets of a set  $S$  and  $S^*$  will denote the set of all finite sequences by elements of  $S$ .

Let the set of states of the agent’s environment be defined by  $S = \{s_1, s_2, \dots\}$ . The effector capability of an agent is represented by a set  $A = \{a_1, a_2, \dots\}$  of actions. At the most basic level an agent can then be viewed as a function

$$action : S^* \rightarrow A$$

which maps finite sequences of environment states to actions. Intuitively speaking an agent decides what action to perform on the basis of its history i.e. its experience to date. The non-deterministic behaviour of the environment can be modelled as a function

$$env : S \times A \rightarrow 2^S$$

which maps the current state of the environment and the agent’s action to a set of environment states that could result from performing the action. If all the sets in the range are singletons then the environment is called deterministic.

An agent is purely reactive if its action only depends on the current state of the environment, so its action can be described by a function  $action : S \rightarrow A$ . This definition of reactive agents is a bit simplistic to be useful and the first architectural issue is to split the action mapping into perception and action:

$$see : S \rightarrow P, \quad act : P^* \rightarrow A$$

where  $P$  is a nonempty set of percepts that is composed of relations of objects with classes in a restricted ontology  $\mathcal{O}$ . To be more precise let  $M_{\mathcal{O}}$  be the set of all objects with class in  $\Gamma_{\mathcal{O}}$ . Relations of objects from  $M_{\mathcal{O}}$  are unary, binary, tertiary, etc. partial functions with Boolean or real-vector values. For any positive integer  $i$  the  $R_i(V)$  denotes the set of  $i$ -argument partial functions  $f$  that are defined as  $f : \Pi_{k=1}^i M_{\mathcal{O}} \rightarrow V$ . The set of all relation functions over  $M_{\mathcal{O}}$  is  $R(V) = \cup_{i \geq 0} R_i(V)$ , finally the set of all possible relations is denoted by  $R$ . We now define the perception  $P$  as some subset of  $R$  and  $P^*$  as sequences of subsets of  $R$ .

This means that  $act$  now maps sequences of percepts to actions. Let’s now introduce a set of agent states  $K$ . Then the next state is defined by a function  $next : K \times P \rightarrow K$  and the action is decided by a new function  $action : K \rightarrow A$  that maps the set of states into actions. State-based agent description assumes that an initial state exists.

Sentences defined in NLP documents can now be used in various ways:

- (a) Relational sentences describing the relations in  $P$  are created from sensor signals.
- (b) Meanings of instructional sentences describing how actions are executed in terms of modelling objects in  $M_O$ . These can be “physical” actions such as feedback/feedforward control or actions can also be “mental” actions such as various types of path planning or planning of steps to achieve a more abstract goal. Assessing priorities in a deliberative agent is also type of mental action.

The above described general agent class with the state transition *next* function can cover all kinds of layered and deliberative agent architectures. The details of how state transition is organized in an agent is not the concern of this paper. We are only interested to define a large class of agents that can be fully documented in an NLP document and can make use of new documents read.

*Definition 7.1:* An agent  $A(O, see, next, action)$  is called *NLP documentable* if there is an NLP document  $D$  that describes

- (i) a agent-world modelling ontology;  $O$
- (ii) the execution of its *see* function for perception and communication;
- (iii) the execution of its *next* and *action* functions.

#### A. Agents sharing a conceptual base

We now assume that in a set  $A$  of NLP documented agents each agent  $a$  is based on the same document  $D = \langle \Xi, O, S, m, L_N, N \rangle$ . There will be no specific assumption made about the agent architecture and decision making procedures but the following will be shared among the agents.

- 1) Each agent  $a \in A$  uses ontology  $O$  to model the set of objects and systems  $M$  of the shared physical environment in which they operate.
- 2) Each agent  $a \in A$  uses the same set of sentences  $M_r$  to express relationships among objects and systems of the environment in terms of their perception definitions.

In terms of modelling relations the NLP sentences used by agents can be of the following kind:

- 1) Query sentences that have for meaning the description of a procedure in terms of NLP sentences.
- 2) Instructional sentences that have meanings described by sequences of NLP sentences describing the steps of the procedure.
- 3) Goal sentences that have meanings described by a mixed sequence of instructional sentences and other goal sentences (subgoals).
- 4) Behaviour constraint sentences in terms of relationships among environmental objects (including the agents “body”) that describe what is not allowed while achieving goals.

Note that if something is not behaviourally constrained than it is allowed for any agents during its course of actions to achieve a goal.

#### B. Agents reading new NP documents

It is now clear that if human authors write about new perception or action procedures in an NLP document that is targeted for an agent class than the following holds.

*Theorem 7.1:* A perception or action sentence definition  $S$  by other NLP sentences in a new NLP document  $D$  is useful for agents in an agent class  $A$  if the following conditions are satisfied:

- (I) The ontology used by  $D$  is an extension of the base ontology of the agent class  $A$ .
- (II)  $S$  occurs in the base NLP document of agent class  $A$ .
- (III) Underlying programming language calls used in the meaning definition of  $S$  in  $D$  are contained in the routine library  $H$  of  $N$  used by the agent class  $A$ .

□

This means that researchers can look up the basic perception, action, planning, decision making and other sentences of a robot family and write new procedures in an HTML published NLP document. The difference relative to currently available automatic software updates (through the Internet) is that in our systems these Internet publications are actually readable English (or other) documents that can be the basis of professional knowledge sharing as that is customary today in journal publications. The next section provides a formal description of knowledge sharing with humans.

### VIII. SHARED KNOWLEDGE DEFINITIONS WITH HUMANS

Apart from clear presentation of knowledge sharing among a set of artificial agents, the most useful feature of NLP documents is that defined meanings also map to conceptual structures that represent human thinking. The relationship of a document  $D = \langle \Xi, O, S, m, d, L_N, N \rangle$  to human thinking, and onto knowledge sharing among a set of human users and programmers, is formulated by its human interpretation functions  $\Psi$ .

*Definition 8.1:* A pair of time varying functions  $\Psi' = [\Psi'_r, \Psi'_c]$ , is called the *human interpretation* of sentences in an NLP document  $D = \langle \Xi, O, S, m, L_N, N \rangle$  if the following are satisfied:

- $\Psi'_c : \Gamma_O^I \rightarrow 2^C$  maps current individuals of classes in ontology  $\Gamma$  of  $O$  to a set of feasible conceptual interpretation among the concepts  $O_c$  of human CGs.
- $\Psi'_r : S \rightarrow 2^{H_{KB}}$  is a mapping consistent with  $d : S \rightarrow B_{cg}(O_c, O_r)$ .

Note that the difference between the basic CGs  $B_{cg}(O_c, O_r)$  and  $H_{KB}$  is that the latter represents a stream of compound CGs of human interpretation of the current and past state of the world.

For the human interpreter a sentence  $s \in S$  can have varying meanings, i.e. conceptual graphs depending on the context the sentence is uttered in. In general we cannot aim here to formalize human interpretation due to lack of insufficient knowledge about human thinking. There is however a special context for the human mind where meanings of sentences  $s \in S$  can be well defined.

**Definition 8.2:** The human context where a human agent  $A_h$  have read and remembers the meanings of definitions in  $D = \langle \Xi, O, S, m, d, L_N, N \rangle$  is called a *human-machine context*  $X(A_h, \Xi_p)$ . The human agent  $A_h$  is called *nlp-compliant* if it chooses to interpret all  $s \in S$  with their unique translator  $T(s)$  associated with  $D$ .

Note that for humans the way the  $\Xi_p = \langle O, S, m \rangle$  interprets a sentence can be only one of the possibilities. Nlp-compliant human behaviour means that the human chooses to interpret sentences according to  $\Xi_p = \{O, S\}$ . Given a set of humans sharing the knowledge of a  $\Xi_p = \langle O, S, m \rangle$  and they are nlp-compliant, then they have shared meaning of any  $s \in S$ , irrespectively whether they are agent programmers or users of an agent.

**Theorem 8.1:** Any NLP document  $D = \langle \Xi, O, S, m, d, L_N, N \rangle$  provides shared unambiguous understanding of sentence meanings in  $S$  for nlp-compliant humans and software agents.

**Proof.** Any sentence  $s \in S$  in an NLP  $D = \langle \Xi, O, S, m, d, L_N, N \rangle$  corresponds to a code  $T(s) \in L_N$  that defines its meaning for an agent residing on a computer and using  $D$ . On the other hand any sentence  $s \in S$  also has a unique interpretation  $\Psi_r(s) \in H_{KB}$  for an nlp-compliant human agent. The correspondence  $T(s) \Leftrightarrow \Psi_r(s)$  creates a shared meaning between the software agent and the human as well as a definition of possible meaning for the human.  $\square$

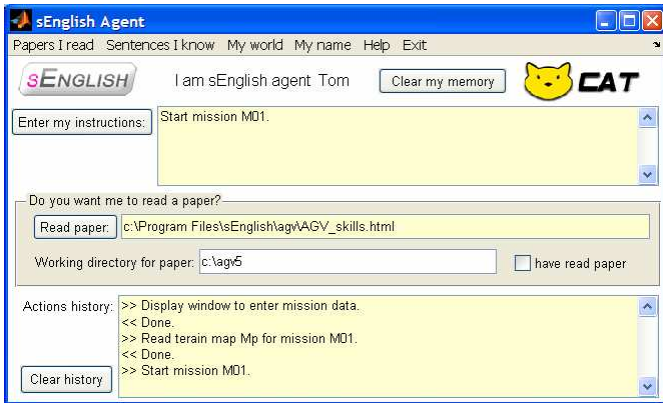


Fig. 5. Graphical user interface of an sEnglish reader agent for web pages that can be installed on any PC.

The usefulness of natural language program definitions stems from the fact that

- The sentence meanings define the precise meaning that may be ambiguous to the human interpreter
- A team of humans can share the understanding of natural language sentences definitions and adopt their high level decision making ability the meaning definitions provided. This is especially so if the NLP sentence definitions and modelling object classes represent generally acceptable examples of human concepts.

## IX. CONCLUSIONS

This paper has described the theory of natural language programming and theoretical foundations NLP and its use by agents. NLP has existed for a couple of years [2], [3] in software that is now well tested and this paper fills in the gap in terms of precise theoretical definitions. Users of sEnglish can find a formal description that reveals its simplicity and explains its practical strength. This theory clarifies the advantages an NLP publishing system over traditional programming of robots in ADA, Stateflow/Simulink or Python. A brief summary is:

- Through the documents human users can share the knowledge of the agents and that reduces misunderstanding when complex intelligent behaviour is needed during autonomous missions.
- The "published" papers can be distributed to agents of a specific control program that define their deliberative or reactive behaviour. Instead of reprogramming, the autonomous systems updates/learns from publications as humans do.
- The sEnglish publications can be distributed within a company or on the Internet as "proper publications" to make colleagues aware of results in sensory signal processing, navigation, environment modelling and adaptive/learning control methods.

The theory also reveals that NLP can be considered as further development of two traditional approaches. On the one hand NLP is an extension of object oriented programming as it uses object classes and its methods are defined by sentences. On the other hand NLP is developing the paradigm of graphical user interfaces (GUIs) further for complex system where GUIs are not sophisticated enough to express internal logic of system operations. In such cases an NLP document can define the conceptual and logical structuring of an application while some NLP sentences can be used to invoke GUIs for data entry and display of data.

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