Natural Language Engineering: The Study of Word Sense Disambiguation in Punjabi

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Abstract: Word Sense Disambiguation (WSD) is an important part of Machine Readable Dictionary (MRD) which is extensively used in Expert System/Intelligent Systems. All languages have multiple meanings of words or phrases depending on the context of their usage. WSD draws the correct (intended) meaning using a database called Machine Readable Dictionary (MRD). Some rudimentary designs of MRD have been made for some European Languages. In this paper a preliminary attempt has been made towards the formulation and design of MRD in Punjabi Language using modified Lesk Algorithm which uses a simple method for relating the appropriate word sense relative to set of dictionary meanings of the word or phrase.

1. Introduction

One of the most challenging and active area of Artificial Intelligence is Natural Language Engineering (NLE) also referred as Natural Language Processing (NLP). The main goal of NLE is to design and develop software that will analyze, understand and generate languages that humans use naturally. It also determines a system of symbols, relations and conceptual information that can be used by computer logic to communicate with human. This implementation requires the system to have the capacity to translate, analyze and synthesize language. Natural language is any human "spoken or written language governed by set of rules and conventions sufficiently complex and subtle enough for there to be frequent ambiguity in syntax and meaning." The main tasks of artificial NLE are to replace the human processor with a machine processor and to get a machine to understand the natural language input and then transform it appropriately. The term natural language is used to distinguish human languages (like Punjabi, Hindi, English etc.) from formal or computer languages (like C, C++, Java,



Perl). Natural language processing requires two types of resources: linguistic data in an electronic format and computational engines. Typical electronic linguistic resources are electronic dictionaries and corpora. Computational engines process the linguistic resources and produce an output [1], [2], [3], [4], [5].

2. Word Sense Disambiguation (WSD)

One of the most important issues in the field of natural language engineering is WSD. A lot of research has been carried out on the topic at MIT, Rutgers University, Stanford University in USA and some Indian Universities like IITs etc. In WSD, a system attempts to determine the sense of a word from contextual features. Many words in written languages have more than one meaning and we have to select the meaning which makes the most sense in context. In computational linguistics, WSD is the process of identifying which sense of a word is used in any given sentence, when the word has a number of distinct senses. For example, consider the following phrases in English:

- (a) Do not tell a lie
- (b) Lie down here

There are four conventional approaches to WSD:

- Dictionary and knowledge-based methods
- Supervised methods
- Semi Supervised methods
- Unsupervised methods

[13], [14], [23] explore the subject in detail. The available research primarily focuses on European languages only. No sufficient work has been done on the topic in the context of Indian languages. The present research will deal with WSD in the



context of Indian languages, in general, and with Punjabi language, in particular. An attempt for WSD in case of nouns in Hindi has been made at IIT Bombay [27].

3. Machine Readable Dictionary (MRD)-Structure and Role

Machine Readable Dictionary (MRD) is a dictionary which is stored in a computer as database instead of being printed on paper. MRD contains a rich set of relationships between their senses, and indicate them in variety of ways. It may be single language explanatory dictionary to support translation between two or more languages or a combination of both. WSD approaches utilize external knowledge sources such as MRD, Thesauri, Tagged or Untagged corpora. LDOCE (Longman Dictionary of Contemporary English) has been the most widely used MRD in context of WSD. Such dictionaries are readily available for European languages including English. But no such MRD is define the work available for Punjabi. This work will include design and construction of such a dictionary. It work will lead to better understanding of the relationships between word meanings, which in turn will help in implementation of Lesk Algorithm by retrieving all sense definitions of the words to be disambiguated from MRD.

The following Table gives the basic idea or description of MRD in Punjabi. In this table we are using some words in Punjabi which has more than one meaning for WSD:

Word Synset

sr qlf, srwr, qfs dy Kyzyjf cwky pwqy srkVf, kfnf, sk mfn jl, AppfDl, hvf dy qy clx dlavfj, iksyjfnvr dy clx dlavfj, ij wq, Azdy pqlog dlzor nWiliZwlf krn df Bfv

bld bnlk dlikirafjf Bfv, srlr dyafllyf dfjlV, iesqrlaf dyhll dfgihxf, kivqfivnc docfrjf pljl qlkf dliekfel, BliVaf hieaf,Zikaf hieaf,KVf

puql Bfievfll, abfdl df ihusf, cfh dypuqy gunydysukypuqr, Pul dl puKVl, luhydl pqll kfqr

sfvDfn hisafr, cklhf,Pirqll f, j fgdf, cks, sicg, cgnqf

kwcf ijs ivc ajyimTfs nhl bxl (axpk), axirwiJaf, jOriVHaf nf jf cMgl qrf isikaf nf hNvy axAmbil af, anfVl, kmj r, nrm, nf Br6sjXNg, \$g\frsrkfrl, atkl pwcll iCmfhl-iqmfhl(iemiqhfn), iPkf pYjfn vfl f, jOBpq nf rWK sky jOsQfel nhl, afrjl, ijsdl sjf df \$P\frac{1}{2}\$If nhl hNeaf, k\frac{1}{2}\$If mfl, grB ivcl f QNVIJmihinaf df b\frac{1}{2}\$If, pgzMzl, Cutyrl Amr df mMzf, jOpkkl vhl ivwc drj nf hNvy srmsfr, qfjf

Table 1.1: MRD structure in Punjabi [16]



4. Word Sense Disambiguation in Punjabi

The Punjabi language is morphologically rich. As already mentioned, WSD is defined as the task of finding the correct sense of the word in the context. The task needs large amount of word and word knowledge. Words may have different meaning in different contexts. Let us consider the word **v**** (Vatt). This word is used normally in eight different senses in Punjabi.

- v\nt p\nt I df b\nt, Kfl aid df Anc\nt k\nt Example "pfxl r\nt h\nt q\nt vnt Anc\nt kr I v\nt"
- ii. vvit gusf, nrfjgl: Ex. "mY Aus dy mMh qy sucl gl kih iduql! Ausnul vVf vvit ciVaf"
- iii. vlit j rnfl : Ex. "aiDafpk ny mllty dy vlit ky cpl/ mfr idlull"
- iv. vit Jitvi, iving, vi, Bing; Ex. "blZfpyivit cmvi qyvit pik ing jfdyhn"
- v. vlit iqAlVI: Ex. "Gr afeymihmfan vI dK kymlQyvlit nhl pfAnf cfhldf"
- vi. vWt iZwz ivwc pYh vfl f mrWv: Ex. "mYkwcyamrWd Kfigaf hf! hwn iZwz ivwc vWt pYirhf hYv
- vii. vWt psInf, hWh, hWhV, BVfs, BVdfa: Ex. "grml bhq h\! m\n\v\\t q\b kr irhf h\!"
- viii. vlit mliCf ntillvlit cfVny Ex. "srdfr jl svyryr) plug bnx q0bfad mliCf ntillvlit cfVdy hn"

In the above sentences, the word vit has different meanings [8], [15]. There are hundreds of examples of such words in Punjabi. In English language Yarowsky proposed a solution to WSD using the thesaurus and supervised learning approach [14]. The main idea is to compare the context of the word in a sentence with the context constructed from the database and assign correct sense to the word. In this research we will use Punjabi Lexical Resources, Gurmukhi OCR, Harkirat Singh's Shabad Kosh, Punjabi Thesaursus and Online Technical English Punjabi glosses. The work involves the design & construction of a Machine Readable Dictionary (MRD) and its use in implementation of Lesk Algorithm or its modified form for WSD in Punjabi. The study will be a pioneering work on the chosen topic in the context of Indian languages, in general, and Punjabi language, in particular.

Figure 1.1. described the pictorial representation of WSD form Punjabi MRD and given context [28].



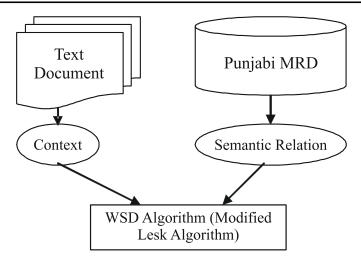


Figure 1.1: To Extraction of synonymy of given context from Punjabi MRD for WSD

5. Lesk Algorithm and its implementation

The original Lesk algorithm [23] proposed a simple algorithm for selecting the appropriate word sense relative to a set of dictionary senses. The **Lesk** algorithm disambiguates a target word by comparing its gloss with those of its surrounding words. The target word is assigned the sense whose gloss has the most overlapping or shared words with the glosses of its neighboring words. **Lesk** demonstrates this algorithm on the words *pine cone*. Using the **Oxford Advanced Learner's Dictionary**, it finds that the word *pine* has two senses:

Sense 1: kind of evergreen tree with needle-shaped leaves

Sense 2: waste away through sorrow or illness.

The word cone has three senses:

Sense 1: solid body which narrows to a point

Sense 2: something of this shape whether solid or hollow

Sense 3: fruit of certain evergreen tree

Each of the two senses of the word *pine* is compared with each of the three senses of the word *cone* and it is found that the words *evergreen tree* occurs in



one sense each of the two words. These two senses are then declared to be the most appropriate senses when the words *pine* and *cone* are used together.

There are two hypothesis that underly this approach. The first is that words appears together in a sentence can be disambiguated by assigning to them the senses that are most closely related to their neighboring words. The second hypothesis is that related senses can be identified by finding overlapping words in their definitions. In this research we will implement Lesk algorithm or modified version of this, keeping in view the linguistic feature of Punjabi to disambiguate the word in Punjabi language.

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