Resolving Shell Nouns

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Abstract

Shell nouns, such as fact and problem, occur frequently in all kinds of texts. These nouns themselves are unspecific, and can only be interpreted together with the shell content. We propose a general approach to automatically identify shell content of shell nouns. Our approach exploits lexico-syntactic knowledge derived from the linguistics literature. We evaluate the approach on a variety of shell nouns with a variety of syntactic expectations, achieving accuracies in the range of 62% (baseline = 33%) to 83% (baseline = 74%) on crowd-annotated data.

1 Introduction

Shell nouns are abstract nouns, such as fact, issue, idea, and problem, which facilitate efficiency by avoiding repetition of long stretches of text. The shell metaphor comes from Schmid (2000), and it captures the various functions of these nouns in a discourse: containment, signalling, pointing, and encapsulating. Shell nouns themselves are unspecific, and can only be interpreted together with their shell content, i.e., the propositional content they encapsulate in the given context. The process of identifying this content in the given context is referred to as shell noun resolution or interpretation. Examples (1), (2), and (3) show usages of the shell nouns fact and issue. The shell noun phrases are resolved to the postnominal that clause, the complement wh clause, and the immediately preceding clause, respectively.1,2

(1) The fact that a major label hadn’t been at liberty to exploit and repackaging the material on CD meant that prices on the vintage LP market were soaring.

(2) The issue that this country and Congress must address is how to provide optimal care for all without limiting access for the many.

(3) Living expenses are much lower in rural India than in New York, but this fact is not fully captured if prices are converted with currency exchange rates.

Observe that the relation between shell noun phrases and their shell content is similar to the relation of abstract anaphora (or cataphora) (Asher, 1993) with backward- or forward-looking abstract-object antecedents. For anaphoric shell noun examples, the shell content precedes the shell noun phrase, and for cataphoric shell noun examples the shell content follows the shell noun phrase.3

Shell nouns as a group occur frequently in argumentative texts (Schmid, 2000; Flowerdew, 2003; Botley, 2006). They play an important role in organizing a discourse and maintaining its coherence (Schmid, 2000; Flowerdew, 2003), and resolving them is an important component of various computational linguistics tasks that rely on

1Note that the postnominal that-clause in (1) is not a relative clause: the fact in question is not an argument of exploit and repackaging.
2All examples in this paper are from the New York Times corpus (https://catalog.ldc.upenn.edu/LDC2008T19)
3We use the terms cataphoric shell noun and anaphoric shell noun for lack of better alternatives.
discourse structure. Accordingly, identifying shell content can be helpful in summarization, information retrieval, and ESL learning (Flowerdew, 2003; Hinkel, 2004).

Despite their importance in discourse, understanding of shell nouns from a computational linguistics perspective is only in the preliminary stage. Recently, we proposed an approach to annotate and resolve anaphoric cases of six typical shell nouns: fact, reason, issue, decision, question, and possibility (Kolhatkar et al., 2013b). This work drew on the observation that shell nouns following cataphoric constructions are easy to resolve. We manually developed rules to identify shell content for such cases. Later, we used these cataphoric examples and their shell content as training data to resolve harder anaphoric examples.

In this paper, we propose a general algorithm to resolve cataphoric shell noun examples. Our long-term goal is to build an end-to-end shell-noun resolution system. If we want to go beyond the six shell nouns from our previous work, and generalize our approach to other shell nouns, first we need to develop an approach to resolve cataphoric shell noun examples. A number of challenges are associated with this seemingly easy task. The primary challenges is that this resolution is in many crucial respects a semantic phenomenon. To obtain the required semantic knowledge, we exploit the properties of shell nouns and their categorization described in the linguistics literature. We evaluate our method using crowdsourcing, and demonstrate how far one can get with simple, deterministic shell content extraction.

2 Related work

Shell-nounhood is a well-established concept in linguistics (Vendler, 1968; Ivanic, 1991; Asher, 1993; Francis, 1994; Schmid, 2000, inter alia). However, understanding of shell nouns from a computational linguistics perspective is only in the preliminary stage.

Shell nouns take a number of semantic arguments. In this respect, they are similar to the general class of argument-taking nominals as given in the NomBank (Meyers et al., 2004). Similarly, there is a small body of literature that addresses nominal semantic role labelling (Gerber et al., 2009) and nominal subcategorization frames (Preiss et al., 2007). That said, the distinguishing property of shell nouns is that one of their semantic arguments is the shell content, but the literature in computational linguistics does not provide any method that is able to identify the shell content. The focus of our work is to rectify this.

Shell content represents complex and abstract objects. So traditional linguistic and psycholinguistic principles used in pronominal anaphora resolution (see the survey by Poesio et al. (2011)), such as gender and number agreement, are not applicable in resolving shell nouns. That said, there is a line of literature on annotating and resolving personal and demonstrative pronouns, which typically refer to similar kinds of non-nominal abstract entities (Passonneau, 1989; Eckert and Strube, 2000; Byron, 2003; Müller, 2008; Hedberg et al., 2007; Poesio and Artstein, 2008; Navarretta, 2011, inter alia). Also, there have been attempts at annotating the shell content of anaphoric occurrences of shell nouns (e.g., Botley (2006), Kolhatkar et al. (2013a)). However, none of these approaches attempt to annotate and resolve cataphoric examples such (1) and (2).

3 Challenges

A number of challenges are associated with the task of resolving cataphoric shell noun examples, especially when it comes to developing a holistic approach for a variety of shell nouns.

First, each shell noun has idiosyncrasies. Different shell nouns have different semantic and syntactic expectations, and hence they take different types of one or more semantic arguments: one introducing the shell content, and others expressing circumstantial information about the shell noun. For instance, fact typically takes a single factual clause as an argument, which is its shell content, as we saw in example (1), whereas reason expects two arguments: the cause and the effect, with the content introduced in the cause, as shown in example (4). Similarly, decision takes an agent making the decision and the shell content is represented as an action or a proposition, as shown in (5).

(4) **One reason** [that 60 percent of New York City public-school children read below grade level] **effect** is [that many elementary schools don’t have libraries] **cause**.

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4Observe that the postnominal that clause in (4) is not a relative clause, and still it is not the shell content because it is not the cause argument of the shell noun reason.

5Observe that this aspect of shell nouns of taking different numbers and kinds of complement clauses is similar to verbs having different subcategorization frames.
(5) I applaud loudly the decision of\[Greenburgh\]agent to ban animal performances.

Second, the relation between a shell noun and its content is in many crucial respects a semantic phenomenon. For instance, resolving the shell noun reason to its shell content involves identifying a) that reason generally expects two semantic arguments: cause and effect, b) that the cause argument (and not the effect argument) represents the shell content, and c) that a particular constituent in the given context represents the cause argument.

Third, at the conceptual level, once we know which semantic argument represents shell content, resolving examples such as (4) seems straightforward using syntactic structure, i.e., by extracting the complement clause. But at the implementation level, this is a non-trivial problem for two reasons. The first reason is that examples containing shell nouns often follow syntactically complex constructions, including embedded clauses, coordination, and sentential complements. An automatic parser is not always accurate for such examples. So the challenge is whether the available tools in computational linguistics such as syntactic parsers and discourse parsers are able to provide us with the information that is necessary to resolve these difficult cases. The second reason is that the shell content can occur in many different constructions, such as apposition (e.g., parental ownership of children, a concept that allows . . . ), postnominal and complement clause constructions, as we saw in examples (1) and (2), and modifier constructions (e.g., the liberal trade policy that . . . ). Moreover, in some constructions, the content is indefinite (e.g., A bad idea does not harm until someone acts upon it.) or None because the example is a non-shell noun usage (e.g., this week’s issue of Sports Illustrated), and the challenge is to identify such cases.

Finally, whether the postnominal clause introduces the shell content or not is dependent on the context of the shell noun phrase. The resolution can be complicated by complex syntactic constructions. For instance, when the shell noun follows verbs such as expect, it becomes difficult for an automatic system to identify whether the postnominal or the complement clause is of the verb or of the shell noun (e.g., they did not expect the decision to reignite tension in Crown Heights vs. no one expected the decision to call an election). Similarly, shell noun phrases can be objects of prepositions, and whether the postnominal clause introduces the shell content or not is dependent on this preposition. For instance, for the pattern reason that, the postnominal that clause does not generally introduce the shell content, as we saw in (4); however, this does not hold when the shell noun phrase containing reason follows the preposition for, as shown in (6).

(6) Low tax rates give people an incentive to work, for the simple reason that they get to keep more of what they earn.

4 Linguistic framework

Linguists have studied a variety of shell nouns, their classification, different patterns they follow, and their semantic and syntactic properties in detail (Vendler, 1968; Ivanic, 1991; Asher, 1993; Francis, 1994; Schmid, 2000, inter alia). Schmid points out that being a shell noun is a property of a specific usage of the noun rather than an inherent property of the word. He provides a list of 670 English nouns that tend to occur as shell nouns. A few frequently occurring ones are: problem, notion, concept, issue, fact, belief, decision, point, idea, event, possibility, reason, trouble, question, plan, theory, aim, and principle.

4.1 Lexico-syntactic patterns

Precisely defining the notion of shell-nounhood is tricky. A necessary property of shell nouns is that they are capable of taking clausal arguments, primarily with two lexico-syntactic constructions: Noun + postnominal clause and Noun + be + complement clause (Vendler, 1968; Biber et al., 1999; Schmid, 2000; Huddleston and Pullum, 2002). Schmid exploits these lexico-syntactic constructions to identify shell noun usages. In particular, he provides a number of typical lexico-syntactic patterns that are indicative of either anaphoric or cataphoric shell noun occurrences. Table 1 shows these patterns with examples.

Cataphoric These patterns primarily follow two constructions.

N-be-clause In this construction, the shell noun phrase occurs as the subject in a subject-verb-clause construction, with the linking verb be, and the shell content embedded as a wh clause, that clause, or to-infinitive clause. The linking
Cataphoric

1. **N-be-to**  
   *Our plan is to hire and retain the best managers we can.*

2. **N-be-that**  
   *The major reason is that doctors are uncomfortable with uncertainty.*

3. **N-be-wh**  
   *Of course, the central, and probably insoluble, issue is whether animal testing is cruel.*

4. **N-to**  
   *The decision to disconnect the ventilator came after doctors found no brain activity.*

5. **N-that**  
   *Mr. Shoval left open the possibility that Israel would move into other West Bank cities.*

6. **N-wh**  
   *If there ever is any doubt whether a plant is a poppy or not, break off a stem and squeeze it.*

7. **N-of**  
   *The concept of having an outsider as Prime Minister is outdated.*

Anaphoric

8. **th-N**  
   *Living expenses are much lower in rural India than in New York, but this fact is not fully captured if prices are converted with currency exchange rates.*

9. **th-be-N**  
   *People change. This is a fact.*

10. **Sub-be-N**  
    *If the money is available, however, cutting the sales tax is a good idea.*

Table 1: Lexico-grammatical patterns of shell nouns (Schmid, 2000). Shell noun phrases are underlined, the pattern is marked in boldface, and the shell content is marked in italics.

<table>
<thead>
<tr>
<th>Noun</th>
<th>N-be-to</th>
<th>N-be-that</th>
<th>N-be-wh</th>
<th>N-to</th>
<th>N-that</th>
<th>N-wh</th>
<th>N-of</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>idea</td>
<td>7</td>
<td>2</td>
<td>-</td>
<td>5</td>
<td>23</td>
<td>10</td>
<td>53</td>
<td>91,277</td>
</tr>
<tr>
<td>issue</td>
<td>-</td>
<td>1</td>
<td>5</td>
<td>7</td>
<td>14</td>
<td>2</td>
<td>71</td>
<td>55,088</td>
</tr>
<tr>
<td>concept</td>
<td>1</td>
<td>-</td>
<td>-</td>
<td>6</td>
<td>12</td>
<td>-</td>
<td>79</td>
<td>14,301</td>
</tr>
<tr>
<td>decision</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>80</td>
<td>12</td>
<td>1</td>
<td>5</td>
<td>55,088</td>
</tr>
<tr>
<td>plan</td>
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<td>4</td>
<td>1</td>
<td>-</td>
<td>16</td>
<td>25</td>
<td>2</td>
<td>51</td>
<td>24,025</td>
</tr>
</tbody>
</table>

Table 2: Distribution of cataphoric patterns for six shell nouns in the New York Times corpus. Each column shows the percentage of instances following that pattern. The last column shows the total number of cataphoric instances of each noun in the corpus.

Verb *be* indicates the semantic identity between the shell noun and its content in the given context. The construction follows the patterns in rows 1, 2, and 3 of Table 1.

**N-clause** This construction includes the cataphoric patterns 4–7 in Table 1. For these patterns the link between the shell noun and the content is much less straightforward: whether the postnominal clause expresses the shell content or not is dependent on the shell noun and the syntactic structure under consideration. For instance, for the shell noun *fact*, the shell content is embedded in the postnominal *that* clause, as shown in (1), but this does not hold for the shell noun *reason* in example (4). The N-of pattern is different from other patterns: it follows the construction *N-prepositional phrase* rather than *N-clause*, and since a prepositional phrase can take different kinds of embedded constituents such as a noun phrase, a sentential complement, and a verb phrase, the pattern offers flexibility in the syntactic type of the shell content.

**Anaphoric** For these patterns, the link between the shell noun and the content is created using linguistic elements such as *the, this, that, other, same,* and *such.* For the patterns 8 and 9 the shell content does not typically occur in the sentence containing the shell noun phrase. For the pattern 10, the shell content is the subject in a subject-verb-N construction.

**Pattern preferences** Different shell nouns have different pattern preferences. Table 2 shows the distribution of cataphoric patterns for six shell nouns in the New York Times corpus. The shell nouns *idea, issue,* and *concept* prefer N-of pattern, whereas *plan* and *decision* prefer the pattern N-to. Among all instances of the shell noun *decision* fol-
Idea family
Semantic features: [mental], [conceptual]
Frame: mental; focus on propositional content of IDEA
Nouns: idea, issue, concept, point, notion, theory, ...
Patterns: N-be-that/of, N-that/of

Trouble family
Semantic features: [eventive], [attitudinal], [manner], [deontic]
Frame: general eventive
Nouns: problem, trouble, difficulty, dilemma, snag
Patterns: N-be-to

Reason family
Semantic features: [factual], [causal]
Frame: causal; attentional focus on CAUSE
Nouns: reason, cause, ground, thing
Patterns: N-be-that/why, N-that/why

Table 3: Example families from Schmid (2000). The nouns in boldface are used to evaluate this work.

4.2 Categorization of shell nouns
Schmid classifies shell nouns at three levels. At the most abstract level, he classifies shell nouns into six semantic classes: factual, linguistic, mental, modal, eventive, and circumstantial. Each semantic class indicates the type of experience the shell noun is intended to describe. For instance, the mental class describes ideas and cognitive states, whereas the linguistic class describes utterances, linguistic acts, and products thereof.

The next level of classification includes more-detailed semantic features. Each broad semantic class is sub-categorized into a number of groups. A group of an abstract class tries to capture the semantic features associated with the fine-grained differences between different usages of shell nouns in that class. For instance, groups associated with the mental class are: conceptual, creditive, dubiative, volitional, and emotive.

The third level of classification consists of families. A family groups together shell nouns with similar semantic features. Schmid provides 79 distinct families of 670 shell nouns. Each family is named after the primary noun in that family. Table 3 shows six families: Idea, Plan, Trouble, Problem, Thing, and Reason. A shell noun can be a member of multiple families. The nouns subsumed in a family share semantic features. For instance, all nouns in the Idea family are mental and conceptual. They are mental because ideas are only accessible through thoughts, and conceptual because they represent reflection or an application of a concept. Each family activates a semantic frame. The idea of these semantic frames is similar to that of frames in Frame semantics (Fillmore, 1985) and in semantics of grammar (Talmy, 2000). In particular, Schmid follows Talmy’s conception of frames. A semantic frame describes conceptual structures, its elements, and their interrelationships. For instance, the Reason family invokes the causal frame, which has cause and effect as its elements with the attentional focus on the cause. According to Schmid, the nouns in a family also share a number of lexi-co-syntactic features. The patterns attribute in Table 3 shows prototypical lexi-co-syntactic patterns, which attract the members of the family. Schmid defines attraction as the degree to which a lexicogrammatical pattern attracts a certain noun. For instance, the patterns N-to and N-that attract the shell nouns in the Plan family, whereas the N-that pattern attracts the nouns in the Thing family. The pattern N-of is restricted to a smaller group of nouns such as concept, problem, and issue.7,8

6Table 2 does not include anaphoric patterns, as this paper is focused on cataphoric shell noun examples. Anaphoric patterns are common for all shell nouns: among all instances of a shell noun, approximately 50 to 80% are anaphoric.

7Schmid used the British section of COBUILD’S Bank of English for his classification.
8Schmid’s families could help enrich resources such as FrameNet (Baker et al., 1998) with the shell content information.
Resolution algorithm

With this exposition, the problem of shell noun resolution is identifying the appropriate semantic argument of the shell noun representing its shell content. This section describes our algorithm to resolve shell nouns following cataphoric patterns. The algorithm addresses the primary challenge of idiosyncrasies of shell nouns by exploiting Schmid’s semantic families (see Section 4.2). The input of the algorithm is a shell noun instance following a cataphoric pattern, and the output is its shell content or None if the shell content is not present in the given sentence. The algorithm follows three steps. First, we parse the given sentence using the Stanford parser.\(^9\) Second, we look for the noun phrase (NP), where the head of the NP is the shell noun to be resolved.\(^10\) Finally, we extract the appropriate shell content, if it is present in the given sentence.

5.1 Identifying potentially anaphoric shell-noun constructions

Before starting the actual resolution, first we identify whether the shell content occurs in the given sentence or not. According to Schmid, the lexico-syntactic patterns signal the position of the shell content. For instance, if the pattern is of the form N-be-clause, the shell content is more likely to occur in the complement clause in the same sentence. That said, although on the surface level, the shell noun seems to follow a cataphoric pattern, it is possible that the shell content is not given in a postnominal or a complement clause, as shown in (7).

\[ (7) \] Just as weekend hackers flock to the golf ball most used by PGA Tour players, recreational skiers, and a legion of youth league racers, gravitate to the skis worn by Olympic champions. It is the reason that top racers are so quick flash their skis for the cameras in the finish area.

Here, the shell noun and its content are linked via the pronoun it. For such constructions, the shell noun phrase and shell content do not occur in the same sentence. Shell content occurs in the preceding discourse, typically in the preceding sentence.

\[ \text{http://nlp.stanford.edu/software/lex-parser.shtml}\]

\[ \text{We identify such cases, and other cases where the shell content is not likely to occur in the postnominal or complements clauses, by looking for the patterns below in the given order, returning the shell content when it occurs in the given sentence.}\]

Sub-be-N This pattern corresponds to the lexico-grammatical pattern in Figure 1(a). If this pattern is found, there are three main possibilities for the subject. First, if an existential there occurs at the subject position, we move to the next pattern. Second, if the subject is it (example (7)), this or that, we return None, assuming that the content is not present in the given sentence. Finally, if the first two conditions are not satisfied, i.e., if the subject is neither a pronoun not an existential there, we assume that subject contains a valid shell content, and return it. An example is shown in (8).

Note that in such cases, unlike other patterns, the shell content is expressed as a noun phrase.

\[ (8) \] Strict liability is the biggest issue when considering what athletes put in their bodies.

Apposition Another case where shell content does not typically occur in the postnominal or complement clause is the case of apposition. Indefinite shell noun phrases often occur in apposition constructions, as shown in (9).

\[ (9) \] The LH lineup, according to Gale, will feature “cab-forward” design, a concept that particularly pleases him.

In this step, we check for this construction and return the sentential, verbal, or nominal left sibling of the shell noun phrase.

Modifier For shell nouns such as issue, phenomenon, and policy, often the shell content is given in the modifier of the shell noun, as shown in (10).

\[ (10) \] But in the 18th century, Leipzig’s central location in German-speaking Europe and the liberal trade policy of the Saxon court fostered publishing.

We deal with such cases as follows. First, we extract the modifier phrases by concatenating the modifier words having noun, verb, or adjective part-of-speech tags. To exclude unlikely modifier phrases as shell content (e.g., good idea, big
issue), we extract a list of modifiers for a number of shell nouns and create a stoplist of modifiers. If any of the words in the modifier phrases is a pronoun or occurs in the stoplist, we move to the next pattern. If the modifier phrase passes the stoplist test, to distinguish between non-shell content and shell content modifiers, we examine the hypernym paths of the words in the modifier phrase in WordNet (Fellbaum, 1998). If the synset abstraction.n.06 occurs in the path, we consider the modifier phrase to be valid shell content, assuming that the shell content of shell nouns most typically represents an abstract entity.

5.2 Resolving remaining instances
At this stage we are assuming that the shell content occurs either in the postnominal clause or the complement clause. So we look for the patterns below, returning the shell content when found.

N-be-clause The lexico-grammatical pattern corresponding to the pattern N-be-clause is shown in Figure 1(b). This is one of the more reliable patterns for shell content extraction, as the be verb suggests the semantic identity between the shell noun and the complement clause. The be-verb does not necessarily have to immediately follow the shell noun. For instance, in example (2), the head of the NP The issue that this country and Congress must address is the shell noun issue, and hence it satisfies the construction in Figure 1(b).

N-clause Finally, we look for this pattern. An example of this pattern is shown in Figure 1(c). This is the most common (see Table 2) and trickiest pattern in terms of resolution, and whether the shell content is given in the postnominal clause or not is dependent on the properties of the shell noun under consideration and the syntactic construction of the example. For instance, for the shell noun decision, the postnominal to-infinitive clause typically represents shell content. But this did not hold for the shell noun reason, as shown in (11).

(11) The reason to resist becoming a participant is obvious.

Here, Schmid’s semantic families come in the picture. We wanted to examine a) the extent to which the previous steps help in resolution, and b) whether knowledge extracted from Schmid’s families add value to the resolution. So we employ two versions of this step.

Include Schmid’s cues (+SC) This version exploits the knowledge encoded in Schmid’s semantic families (Section 4.2), and extracts postnominal clauses only if Schmid’s pattern cues are satisfied. In particular, given a shell noun, we determine the families in which it occurs and list all possible patterns of these families as shell content cues. The postnominal clause is a valid shell content only if it satisfies these cues. For instance, the shell noun reason occurs in only one family: Reason, with the allowed shell content patterns N-that and N-why. Schmid’s patterns suggest that the postnominal to-infinitive clauses are not allowed as shell content for this shell noun, and thus this step will return None. This version helps correctly resolving examples such as (11) to None.

Exclude Schmid’s cues (–SC) This version does not enforce Schmid’s cues in extracting the postnominal clauses. For instance, the Problem family does not include N-that/wh/to/of patterns, but in this condition, we nonetheless allow these patterns in extracting the shell content of the nouns from this family.

6 Evaluation data
We claim that our algorithm is able to resolve a variety of shell nouns. That said, creating evaluation data for all of Schmid’s 670 English shell
nouns is extremely time-consuming, and is therefore not pursued further in the current study. Instead we create a sample of representative evaluation data to examine how well the algorithm works a) on a variety of shell nouns, b) for shell nouns within a family, c) for shell nouns across families with completely different semantic and syntactic expectations, and d) for a variety of shell patterns from Table 1.

6.1 Selection of nouns
Recall that each shell noun has its idiosyncrasies. So in order to evaluate whether our algorithm is able to address these idiosyncrasies, the evaluation data must contain a variety of shell nouns with different semantic and syntactic expectations. To examine a), we consider the six families shown in Table 3. These families span three abstract categories: mental, eventive, and factual, and five distinct groups: conceptual, volitional, factual, causal, and attitudinal. Also, the families have considerably different syntactic expectations. For instance, the nouns in the Idea family can have their content in that or of clauses occurring in N-clause or N-be-clause constructions, whereas the Trouble and Problem families do not allow N-clause pattern. The shell content of the nouns in the Plan family is generally represented with to-infinitive clauses. To examine b) and c), we choose three nouns from each of the first four families from Table 3. To add diversity, we also include two shell nouns from the Thing family and a shell noun from the Reason family. So we selected a total of 12 shell nouns for evaluation: idea, issue, concept, decision, plan, policy, problem, trouble, difficulty, reason, fact, and phenomenon.

6.2 Selection of instances
Recall that the shell content varies based on the shell noun and the pattern it follows. Moreover, shell nouns have pattern preferences, as shown in Table 2. To examine d), we need shell noun examples following different patterns from Table 1. We consider the New York Times corpus as our base corpus, and from this corpus extract all sentences following the lexi-co-grammatical patterns in Table 1 for the twelve selected shell nouns. Then we arbitrarily pick 100 examples for each shell noun, making sure that the selection contains examples of each cataphoric pattern from Table 1. These examples consist of 70% examples of each of the seven cataphoric patterns, and the remaining 30% of the examples are picked randomly from the distribution of patterns for that shell noun.

6.3 Crowdsourcing annotation
We designed a crowdsourcing experiment to obtain the annotated data for evaluation. We parse each sentence using the Stanford parser, and extract all possible candidates, i.e., arguments of the shell noun from the parser’s output. Since our examples include embedding clauses and sentential complements, the parser is often inaccurate. For instance, in example (12), the parser attaches only the first clause of the coordination (that people were misled) to the shell noun fact.

(12) The fact that people were misled and information was denied, that’s the reason that you’d wind up suing.

To deal with such parsing errors, we consider the 30-best parses given by the parser. From these parses, we extract a list of eligible candidates. This list includes the arguments of the shell noun given in the appositional clauses, modifier phrases, postnominal that, wh, or to-infinitive clauses, complement clauses, objects of postnominal prepositions of the shell noun, and subject if the shell noun follows subject-be-N construction. On average, there were three candidates per instance.

After extracting the candidates, we present the annotators with the sentence, with the shell noun highlighted, and the extracted candidates. We ask the annotators to choose the option that provides the correct interpretation of the highlighted shell noun. We also provide them the option None of the above, and ask them to select it if the shell content is not present in the given sentence or the shell content is not listed in the list of candidates.

CrowdFlower We used CrowdFlower\(^{11}\) as our crowdsourcing platform, which in turn uses various worker channels such as Amazon Mechanical Turk\(^{12}\). CrowdFlower offers a number of features. First, it provides a quiz mode which facilitates filtering out spammers by requiring an annotator to pass a certain number of test questions before starting the real annotation. Second, during annotation, it randomly presents test questions with known answers to the annotators to keep them on their toes. Based on annotators’ responses to the test questions, each annotator is assigned a trust

\(^{11}\)http://crowdflower.com/

\(^{12}\)https://www.mturk.com/mturk/welcome
score: an annotator performing well on the test questions gets a high trust score. Finally, CrowdFlower allows the user to select the permitted demographic areas and skills required.

**Settings** We asked for at least 5 annotations per instance by annotators from the English-speaking countries. The evaluation task contained a total of 1200 instances, 100 instances per shell noun. To maintain the annotation quality, we included 105 test questions, distributed among different answers. We paid 2.5 cents per instance and the annotation task was completed in less than 24 hours.

**Results** Table 4 shows the agreement of the crowd. In most cases, at least 3 out of 5 annotators agreed on a single answer. We took this answer as the gold standard in our evaluation, and discard the instances where fewer than three annotators agreed. The option *None of the above* was annotated for about 30% of the cases. We include these cases in the evaluation. In total we had 1,257 instances (1,152 instances where at least 3 annotators agreed + 105 test questions).

<table>
<thead>
<tr>
<th>Nouns</th>
<th>LSC</th>
<th>A–SC</th>
<th>A+SC</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>≥5</td>
<td>≥4</td>
<td>≥3</td>
</tr>
<tr>
<td>idea</td>
<td>53</td>
<td>67</td>
<td>95</td>
</tr>
<tr>
<td>issue</td>
<td>44</td>
<td>65</td>
<td>95</td>
</tr>
<tr>
<td>concept</td>
<td>40</td>
<td>56</td>
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</tr>
<tr>
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<td>plan</td>
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<td>55</td>
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<td>42</td>
<td>61</td>
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<td>69</td>
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<tr>
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<td>68</td>
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<tr>
<td>phenomenon</td>
<td>39</td>
<td>56</td>
<td>95</td>
</tr>
<tr>
<td>all</td>
<td>46</td>
<td>63</td>
<td>96</td>
</tr>
</tbody>
</table>

Table 4: Annotator agreement on shell content. Each column shows the percentage of instances on which at least $n$ or fewer than $n$ annotators agree on a single answer.

Table 5 shows the evaluation results for the LSC baseline, the algorithm without Schmid’s cues (A–SC), and the algorithm with Schmid’s cues (A+SC). The A–SC condition in all cases and the A+SC condition in some cases outperform the LSC baseline, which proves to be rather low, especially for the shell nouns with strict syntactic expectations (e.g., *fact* and *reason*). Thus we see that our algorithm is adding value.

That said, we observe a wide range of performance for different shell nouns. On the upside, adding Schmid’s cues helps resolving the shell nouns with strict syntactic expectations. The A+SC results for the shell nouns *idea*, *issue*, *concept*, *decision*, *reason*, and *fact* outperform the baseline and the A–SC results. In particular, the A+SC results for the shell nouns *fact* and *reason* are markedly better than the baseline results. These nouns have strict syntactic expectations for the shell content clauses they take: the families *Thing* and *Certainty* of the shell noun *fact* allow only a *that* clause, and the *Reason* family of the shell noun *reason* allows only *that* and *because* clauses for the shell content. These cues help in correctly resolving examples such as (11) to *None*, where the postnominal *to-infinitive clause*.

Note that we only extract subordinating clauses (e.g., (SBAR (IN that) (clause))) and to-infinitive clauses, and not relative clauses.

13. Note that we only extract subordinating clauses (e.g., (SBAR (IN that) (clause))) and to-infinitive clauses, and not relative clauses.
describes the purpose or the goal for the reason, but not the shell content itself.

On the down side, adding Schmid’s cues hurts the performance of more versatile nouns, which can take a variety of clauses. Although the A–SC results for the shell nouns plan, policy, problem, trouble, difficulty, and phenomenon are well above the baseline, the A+SC results are markedly below it. That is, Schmid’s cues were deleterious. Our error analysis revealed that these nouns are versatile in terms of the clauses they take as shell content, and Schmid’s cues restrict these clauses to be selected as shell content. For instance, the shell noun problem occurs in two semantic families with N-be-that/of and N-be-to as pattern cues (Table 3), and postnominal clauses are not allowed for this noun. Although these cues help in filtering some unwanted cases, we observed a large number of cases where the shell content is given in postnominal clauses, as shown in (13).

\begin{equation}
\text{(13) I was trying to address the problem of unreliable testimony by experts in capital cases.}
\end{equation}

Similarly, the Plan family does not allow the N-of pattern. This cue works well for the shell noun decision from the same family because often the postnominal of clause is the agent for this shell noun and not the shell content. However, it hurts the performance of the shell noun policy, as N-of is a common pattern for this shell noun (e.g., ... officials in Rwanda have established a policy of refusing to protect refugees...). Other failures of the algorithm are due to parsing errors and lack of inclusion of context information.

8 Discussion and conclusion

In this paper, we proposed a general method to resolve shell nouns following cataphoric constructions. This is a first step towards end-to-end shell noun resolution. In particular, this method can be used to create training data for any given shell noun, which can later be used to resolve harder anaphoric cases of that noun using the method that we proposed earlier (Kolhatkar et al., 2013b).

The first goal of this work was to point out the difficulties associated with the resolution of cataphoric cases of shell nouns. The low resolution results of the LSC baseline demonstrate the difficulties of resolving such cases using syntax alone, suggesting the need for incorporating more linguistic knowledge in the resolution.

The second goal of this work was to examine to what extent knowledge derived from the linguistics literature helps in resolving shell nouns. We conclude that Schmid’s pattern and clausal cues are useful for resolving nouns with strict syntactic expectations (e.g., fact, reason); however, these cues are defeasible: they miss a number of cases in our corpus. It is possible to improve on Schmid’s cues using crowdsourcing annotation and by exploiting lexico-syntactic patterns associated with different shell nouns from a variety of corpora.

One limitation of our approach is that in our resolution framework, we do not consider the problem of shell noun usage constructions, but the shell content can be expressed in a variety of other constructions. A robust machine learning approach that incorporates context and deeper semantics of the sentence, along with Schmid’s cues, could mitigate this limitation.

This work opens a number of new research directions. Our next planned task is clustering different shell nouns based on the kind of complements they take in different usages similar to verb clustering (Merlo and Stevenson, 2000; Schulte im Walde and Brew, 2002).

Acknowledgements

We thank the anonymous reviewers for their comments. We also thank Suzanne Stevenson, Gerald Penn, Heike Zinsmeister, Kathleen Fraser, Aida Nematzadeh, and Ryan Kiros for their feedback. This research was financially supported by the Natural Sciences and Engineering Research Council of Canada and by the University of Toronto.

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