Grounding spatial named entities for information extraction and question answering

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Abstract

The task of named entity annotation of unseen text has recently been successfully automated with near-human performance. But the full task involves more than annotation, i.e. identifying the scope of each (continuous) text span and its class (such as place name). It also involves grounding the named entity (i.e. establishing its denotation with respect to the world or a model). The latter aspect has so far been neglected.

In this paper, we show how geo-spatial named entities can be grounded using geographic coordinates, and how the results can be visualized using off-the-shelf software. We use this to compare a “textual surrogate” of a newspaper story, with a “visual surrogate” based on geographic coordinates.

1 Introduction

The task of named entity annotation of unseen text has recently been successfully automated, achieving near-human performance using machine learning (Zheng and Su, 2002). But many applications also require grounding – i.e., associating each classified text span with a referent in the world or some model thereof. The current paper discusses spatial grounding of named entities that may be referentially ambiguous, using a minimality heuristic that is informed by external geographic knowledge sources. We then apply these ideas to the creation of “visual surrogates” for news articles.

This paper is structured as follows: Section 2 discusses how spatial named entities can be grounded and how this interacts with their extraction and applications. Section 3 describes a geo-spatial resolution algorithm. Section 4 shows how maps can be automatically constructed from named-entity tagged newswire text using resolved place names, hence introducing a new, graphical document surrogate. Section 5 deals with the usefulness of grounded named entities for question answering. Section 6 presents some related work, and Section 7 concludes this paper.

2 Spatial Grounding

Gazetteers are large lists of names of geographic entities, usually enriched with further information, such as their class (e.g., town, river, dam, etc.), their size, and their location (i.e. with respect to some relative or absolute coordinate system such as longitude and latitude).

Appendix A identifies some publicly available sources. UN-LOCODE is the official gazetteer by the United Nations; it is also freely available from the UNECE Web site and contains more than 36,000 locations in 234 countries (UNECE, 1998). The Alexandria Gazetteer (Smith et al., 1996; Frew et al., 1998) is another database of geographical entities, including both their coordinates and relationships such as: in-state-of, in-province-of, in-county-of, in-country-of, in-region-of, part-of and formerly-known-as.

To date, Named Entity Recognition (NER) has only used gazetteers as evidence that a text span could be some kind of place name (LOCATION), even though their finite nature makes lists of names of limited use for classification (Mikheev et al., 1999). Here we use them for spatial grounding – relating linguistic entities of subtype LOCATION to their real-world counterparts.

“World Atlases” and the gazetteers that index them are not the only resources than can be used for grounding spatial terms. In biomedicine, there are are several brain atlases of different species, using various different techniques, and focussing on both normal and disease state; as well as a digital atlas of the human body

1 http://www.unece.org/cefact/locode/service/main.htm
based on data from the Visible Human project. Such atlases and the nomenclatures that label their parts, provide an important resource for biomedical research and clinical diagnosis. For example, the Mouse Atlas (Ringwald et al., 1994) comprises a sequence of 3D (volumetric) reconstructions of the mouse embryo in each of its 26 Theiler States of development. Indexing it is an part-of hierarchy of anatomical terms (such as embryo.organsystem.cardiovascularsystem.heart.atrium), called the Mouse Anatomical Nomenclature (MAN). Each term is mapped to one or more sets of adjacent voxel\(^2\) that constitute the term’s denotation in the embryo. Figure 1 illustrate this linkage (using 2D cross-sections) in the EMAGE database\(^3\).

Just as one might find it useful for information extraction or question answering to ground geographic terms found in previously unseen text, one may also find it useful to ground anatomical terms in previously unseen text. One example of this would be in providing support for the curation of the Gene Expression Database (GXD)\(^4\). This support could come in the form of a named entity recognizer for anatomical parts in text, with grounding against the Mouse Atlas, using the gazetteer-like information in the MAN.

So what is the relationship between a place name gazetteer like UN-LOCODE and the Mouse Atlas? The MAN is structured in a similar part-of hierarchy to that of geographical locations:

| USA         | embryo          |
|-------------|-----------------|
| California  | organ system    |
| San Mateo County | cardiovascular system |
| Redwood City | heart            |

Because both gazetteers like UN-LOCODE and biomedical atlases like the Mouse Atlas provide spatial grounding for linguistic terms (Figure 2), both can be used to reason about spatio-temporal settings of a discourse, for instance, to resolve referential ambiguity.

This paper is a rare exception due to its meta-linguistic nature.

3 Place-Name Resolution for Information Extraction

There are many places that share the same (Berlin, Germany ~ Berlin, WI, USA) or similar names (York, UK ~ New York, USA), usually because historically, the founders of a new town had given it a similar or the same name as the place they emigrated from.

When ambiguous place names are used in conversation or in text, it is usually clear to the hearer what specific referent is intended. First, speaker and hearer usually share some extra-linguistic context and implicitly adhere to Grice’s Cooperative Principle and the “maxims” that follow, which require a speaker to provide more identifying information about a location that the recipient is believed to be unfamiliar with. Secondly, linguistic context can provide clues: an accident report on the road between Perth and Dundee promotes an interpretation of Perth in Scotland, while an accident on the road between Perth and Freemantle promotes an interpretation of Perth in Western Australia. Computers, which are bound to select referents algorithmically, can exploit linguistic context more easily than extra-linguistic context, but even the use of linguistic context requires (as always) some subtle heuristic reasoning.

Grounding place names mentioned in a text can support effective visualization – for instance, in a multimedia document surrogate that contains textual, video and map elements (e.g. in a question answering scenario), where we want to ensure that the video shows the region and the map is centered around the places mentioned.

To make use of linguistic context in resolving ambiguous place names, we apply two different minimality heuristics (Gardent and Webber, 2001). The first we borrow (slightly modified) from work in automatic word sense disambiguation (Gale et al., 1992), calling it “one referent per discourse”. It assumes that a place name mentioned in a discourse refers to the same location throughout the discourse, just as a word is assumed to be used in the same one sense throughout the discourse. Neither is logically necessary, and hence both are simply interpretational biases.

The second minimality heuristic assumes that, in cases where there is more than one place name mentioned in some span of text, the smallest region that is able to ground the whole set is the one that gives them their interpretation. This can be used to resolve referential ambiguity by proximity: i.e., not only is the place name Berlin taken to denote the same Berlin throughout a discourse unless mentioned otherwise but so does a Potsdam men-

\(^2\) Pixels are points in the 2D plane \((x, y)\); voxels are 3D generalizations of pixels \((x, y, z)\).

\(^3\) http://genex.hgu.mrc.ac.uk/Emage/database/intro.html

\(^4\) http://genex.hgu.mrc.ac.uk/Resources/GXDQuery1/
tioned together with a Berlin uniquely select the capital of Germany as the likely referent from the set of all candidate Berlins.

To illustrate this “spatial minimality” heuristic, consider Figure 3. Assume that a mention of place A in a text could either refer to A‘ or A“. If the text also contains terms that ground unambiguously to I, J, and K, we assume the referent of A is A‘ rather than A“, because the former leads to a smaller spatial context.

To use this “spatial minimality” heuristic, we start by extracting all place names using a named entity recognizer. We then look up the “confusion set” of potential referents for each place name, e.g. for Berlin: { Berlin, FRG (German capital); Berlin, WI, USA; Berlin, NJ, USA; Berlin, CT, USA; Berlin, NH, USA; Berlin, GA, USA; Berlin, IL, USA; Berlin, NY, USA; Berlin, ND, USA; Berlin, NJ, USA }. Each member of the set of potential referents is associated with its spatial coordinates (longitude/latitude), using a gazetteer. We then compute the cross-product of all the confusion sets. (Each member of the cross-product contains one potential referent for each place name, along with its spatial coordinates.) For each member of the cross-product, we compute the area of the minimal polygon bounding all the potential referents, and select as the intended interpretation, the one with the smallest area. The resulting behaviour is shown in Figure 4, depending on contextually mentioned other places, a different Berlin is selected.

The value of this heuristic needs to be assessed quantitatively against various types of text.

In resolving anatomical designators in text, we may employ a variation of the spatial minimality heuristic, based on the fact that no listing will ever be complete with respect to all the existing or new-minted synonyms for anatomical terms.

When grounding the anatomical terms in the text

*In subsequent stages until birth, cytokeratin 8 continues to be expressed in embryonic taste buds distributed in punctate patterns at regular intervals along rows that are symmetrically located on both sides of the median sulcus in the dorsal anterior developing tongue.*

we find no median sulcus within the MAN, only alveolar sulcus, optic sulcus, pre-otic sulcus, sulcus limitans and sulcus terminalis. We just assume that all anatomical terms refer to previously recognized anatomical entities, just as we assume that all geographic terms refer to existing geographic entities and not, for example, some new town called “Berlin” or “London” that is not yet in the gazetteer. Hence median sulcus is assumed to be a synonym for one of the five sulci given in the MAN. At this point, we can invoke the spatial minimality heuristic, looking for the minimal bounding space that includes tongue and one of the five sulci, here yielding “sulcus terminalis”. Thus the spatial minimality heuristic is here pairwise point-point distances, or symbolically, using a hierarchical gazetteer’s relations, such as in-region-of.

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**Table 1:** Comparison between Spatial Grounding in UN-LOCODE and the Mouse Atlas.

| Gazetteer   | Named Entity | Grounding           | Real World               |
|------------|--------------|---------------------|--------------------------|
| UN-LOCODE  | Novosibirsk  | (55°02' N; 82°55' E)(longitude/latitude) | (Novosibirsk)           |
| Mouse Atlas| atrium       | \{(345, 2, 345)\} (set of voxels)       | (part of the heart)     |

Figure 2: Comparison between Spatial Grounding in UN-LOCODE and the Mouse Atlas.

Figure 3: A Place-Name Resolution Method.

Figure 4: Spatial Reference Resolution Using Spatial Minimality.
used with other assumptions to resolve missing or previously unseen terms.

4 Visualization of Geo-Spatial Aspects in Narrative

The usefulness of visual representations to convey information is widely recognized (cf. (Larkin and Simon, 1987)). Here, we use the grounding of named entities in news stories to create a visual surrogate that represents their “spatial aboutness”.

Two news stories were selected from online newspapers on the same day (2003-02-21): one story (Appendix B) reports the tragic death of a baby from London in a Glasgow hospital despite flying it to a Glasgow specialist hospital in the Royal aircraft (BBC News, 2003), and the other report (Appendix C) describes the search of the Californian police for a pregnant women from Modesto, CA, who has disappeared (The Mercury News, 2003).

We use the term “surrogate” to refer to a partial view of a text (e.g. (Leidner, 2002)). Figure 5 shows a textual surrogate in the form of all place names found in a text: an analyst who wants to get a quick overview about the locations involved in some item of news reportage, to decide its local interest or relevance, might find such a surrogate helpful, although the source would still have to be skim-read.

Story A

... Scotland ... Tooting ... London ... Glasgow ... London ... Glasgow ... Northolt ... Glasgow ... Britain ... Prestwick ... Tooting ... Glasgow ... UK ... (Glasgow) ...

Story B

Modesto ... (Southern California) ... (Modesto) ... Los Angeles ... Sacramento ... Berkeley (Marina) ... Fresno ... Oakland ... Modesto ... Los Angeles ... Southern California ... Modesto ... Southern California ... New York ... Long Island ...

Figure 5: A Textual Geo-Spatial Document Surrogate for the Stories in Appendices B and C

We now compare this “baseline” textual surrogate to a graphical representation that draws on the algorithm introduced before. Our simple visualisation method comprises the following components (Figure 6): an open-domain news item is fed into locatagger, our simple named entity tagger for place names based on UNLOCODE. It recognises location names, resolves multi-referential place names and looks up the coordinates:

\[
\begin{align*}
\text{UK} & : 1 \\
\text{Scotland} & : 1 \\
\text{Tooting} & : 2 \\
\text{London} & : 2 \\
\text{Glasgow} & : 5 \\
\text{Northolt} & : 2 \\
\text{Prestwick} & : 1 \\
\text{Britain} & : 1 \\
\text{Modesto} & : 3 \\
\text{SouthernCalifornia} & : 2 \\
\text{LosAngeles} & : 2 \\
\text{Sacramento} & : 1 \\
\text{Berkeley} & : 1 \\
\text{Fresno} & : 1 \\
\text{Oakland} & : 1 \\
\text{NewYork} & : 1 \\
\text{LongIsland} & : 1 \\
\end{align*}
\]

For simplicity, we drop those that correspond to regions (which are represented by sets of points) and feed the remaining list of point coordinates (corresponding to villages and cities) into a map generator to generate a Mercator projection of the geographical area that includes all the points plus 10% of the surrounding area. For this, The Generic Map Tools (GMT) were used, in this case via HTTP.

Figure 7 shows the resulting map for the story in Appendix B. Figure 8 shows the map for the story in Appendix C. Clearly, such a visual surrogate is superior with respect to comprehension time than the textual surrogate presented before. It is interesting so see what happens if we leave out the final paragraph for the map creation (Figure 9): we obtain a “zoomed-in” version of the map. This turns out to be the case for many stories and is due to the convention of news reportage to close a report with linking the narrative to similar events in order to present the event in a wider context.

From the text we obtain a vector of types of all spatial named entities with their frequency of occurrence in the text:

\[\text{http://gmt.soest.hawaii.edu} \]
\[\text{http://stellwagen.er.usgs.gov/mapit/} \]
\[\text{Place names that are too small to be contained in the gazetteer (Tooting/Northold) are ignored.} \]
Figure 7: Automatic Visualization of Story A: A Baby Flown from London to Glasgow for Medical Treatment Dies there.

Figure 8: Automatic Visualization of Story B: A Pregnant Woman is Missing in Modeno, CA (Local View; Final Paragraph Excluded).

Figure 9: Story C: The Final Paragraph Places the Event in Context (Global View; Complete Story).
5 Question Answering

Using grounding knowledge in gazetteers also enables us to answer questions in natural language more effectively:

1. **What is X?**
   - Q: What is Kangigualujjuaq?
   - A: Kangigualujjuaq is a place approximately 1500 kilometers north of Montreal, Canada. (For some place names, many humans cannot tell they refer to places.)

2. **Where is X?**
   - Q: Where is Cannes located?
   - A: [should yield a surrogate based on textual descriptions generated from the gazetteer relations: X is-type Y, X part-of Z and the coordinates, plus a map as generated above, with additional images, e.g. from satellites or picture search engines as available.]

3. **What X is Y part of?**
   - Q: What is the Bad Bergzabern part of?
   - A: Bad Bergzabern is part of the Federal Republic of Germany.
   - Q: Is Andorra la Vella part of Spain?
   - A: No, Andorra la Vella belongs to Andorra.

4. **How far is X from Y?**
   - Q: How far is Cambridge from London?
   - A: The distance between London, England, United Kingdom and Cambridge, England, United Kingdom is 79 km (49 miles or 43 nautical miles). (Note here that the spatial minimality heuristic resolves the question being adressed, so the user can follow up with a different question if this was not what he or she intended.)
   - A: [should give the distance between the two places, e.g. 200 km, as generated from the gazetteer relations: X and Y are both places and the coordinates, plus a map as generated above, with additional images, e.g. from satellites or picture search engines as available.]

Since sophisticated gazetteers are available, answering such questions should not be based on textual extraction from Internet sources, but be based on the gazetteers directly, which reduces noise.

6 Related Work

Whereas a lot of work has been done in the area of spatial databases (Shekhar et al., 1999; Freytag et al., 2000; Paredaens, 1995; Paredaens and Kuipers, 1998), little attention seems to have been given so far to the problem of generating maps from text or annotating maps with information elicited from unstructured documents.

The work presented here perhaps most closely resembles that of (Mackinlay, 1986; Casner, 1990; Roth and Hefley, 1993) who describe systems that generate multimedia representations, potentially including maps, but from a formalized knowledge source rather than unstructured text, and (Waldinger et al., 2002), who describe Geo-Logica, a system that can transform requests to see a place (formulated in natural language) into a three-dimensional aerial view.

7 Summary and Conclusion

Rich gazetteers can be helpful in determining unique geospatial referents and have many applications.

Starting with a description of how spatial named entities can be grounded in the physical world using gazetteer knowledge in two different domains, we have described a heuristic method to resolve place names using such gazetteers.

We have then applied our method to the task of drawing maps from text automatically. These (geo-)graphical document surrogates give an overview about where a reported news event takes place. We do not know of previous attempts to create geographical maps automatically from unconstrained newswire text.

Many venues for further research can be conceived, for instance, a finer granularity of analysis could reveal more named entity classes (including e.g. brigades and other artifacts) and their relative orientation. Maps as created here could also be used to produce an animation from multiple documents (with different time-stamps) on the same topic to illustrate how the importance of places shifts over time, which could aid analysts in their timely identification of regions of crisis.

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A Some Online Gazetteer Sites

All URLs cited as of Friday 2003-02-21.
B  Story Royal Mercy Flight Baby Dies

A baby was flown to Scotland for specialist treatment on an aircraft normally used to carry the Royal Family and has died in hospital.

Khola, who was to have been taken to Yorkhill Hospital in Glasgow, died in hospital last month after being given an undiagnosed heart condition.

She was transferred to the Royal Hospital for Sick Children in Glasgow and the team who transported Khola, and the Queen for allowing the use of her plane. We also wish to thank everyone who prayed for our daughter.

Khola was taken to Yorkhill Hospital An RAF Hercules had initially been scheduled to take the baby to Glasgow last month. However, it burst a tyre as it landed at Northolt.

It was then decided to fly the baby and the medical team to Glasgow in an aircraft from the Royal Squadron.

A Royal Navy helicopter at Prestwick was mobilised and flown in bad weather to RAF Northolt where the specialist medical team on board was transferred to St George’s Hospital in Tooting.

The baby was stabilised in an incubator throughout the night before the transfer to Glasgow. Doctors at Yorkhill said Khola had been in need of respiratory support.

Four hospitals in the UK can provide extracorporeal membrane oxygenation (ECMO) for breathing difficulties, but Glasgow was the only hospital able to accommodate Khola.

We wish to thank everyone who prayed for our daughter.

Khola's parents

C  Story News Crews Wait and Watch as Police Search Home of Missing Woman


The media circus was only the latest in a long line that have sprung up around high-profile criminal cases, where the news is scant but the demand for information – even the smallest morsel – soars off the charts. “People have been interested in this type of story since pre-biblical times,” said Joe Saltzman, associate dean at the Annenberg School for Communications at the University of Southern California. “We live in a global village where everyone is our neighbor, and this is an interesting story about a man whose pregnant wife is kidnapped.”

As Modesto police completed their second day combing through the home that missing mother-to-be Laci Peterson shared with her husband, Scott, more than a dozen news crews from Los Angeles to Sacramento camped out front. Many of the reporters huddled under a tent hastily erected by one reporter as it rained and hailed.

Detective Doug Ridenour said Wednesday that investigators had completed their work, removing about 95 bags of evidence. On Tuesday, Amy Rocha, the sister of the missing woman, accompanied detectives into the modest three-bedroom home for about two hours. But on Wednesday, investigators worked alone – completing their task about 5 p.m. They also planned to do a walk-through with Scott Peterson.

“We’ve been able to accomplish a lot,” Ridenour told reporters at the Peterson house. “But here at this point we just don’t have the significant evidence we need to find Laci or to move in another direction.”

Ridenour would not reveal when detectives would analyze what they took away or when they would release a detailed description of what they found.

Kornberg, who is best known for representing a man the public loved to hate—Joey Buttafuoco, who had a torrid affair with Long Island teenager Amy Fisher – said the only way to end such sideshows is to convert the American system of justice “to the English system, where you announce the arrest of an individual and then you announce the verdict when it’s over.”