#### Natural Language and Spatial Reasoning

Stefanie Tellex

#### What is he doing?



#### Where should the robot go?

With your back to the windows, walk straight through the door near the elevators. Continue to walk straight, going through one door until you come to an intersection just past a whiteboard. Turn left, turn right, and enter the second door on your right (sign says "Administrative Assistant").





# What semantic structures can enable a system to understand and use spatial language in realistic situations?

## Methodology

- Pick a subset of language.
- Collect a corpus focusing on that subset.
- Make a program that understands language in the corpus.
- See how well the program works.
- Analyze why the program works.

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- Pick a subset of language.
- Collect a corpus focusing on that subset.
- Make a program that understands language in the corpus.
- See how well the program works.
- Analyze why the program works.
- Do this in more than one domain.

# Outline

- Problem Statement
- Spatial Prepositions
- Spatial Language Video Retrieval
- Direction Understanding
- Next Steps
- Contributions







#### Spatial Language is Pervasive

"Where is the bus stop?"

"Take a right at the elevators."

"Through the years you've never let me down."

"This book jumps across many different topics."

#### Grounding

#### He's going across the kitchen.



#### Grounding

#### He's going across the kitchen.



#### Grounding He's going across the kitchen.



#### Grounding He's going across the kitchen.



#### Grounding He's going across the kitchen.



#### **Spatial Prepositions**

#### **Spatial Prepositions in English**

about, above, across, after, against, along, alongside, amid(st), among(st), around, at, atop, behind, below, beneath, beside, between, betwixt, beyond, by, down, from, in, inside, into, near, nearby, off, on, onto, opposite, out, outside, over, past, through, throughout, to, toward, under, underneath, up, upon, via, with, within, without, far from, in back of, in between, in front of, in line with, on top of, to the left of, to the right of, to the side of

(from Landau and Jackendoff, 1993)

the, of, and, to, a, in, that, is, was, he, for, it, with, as, his, on, be, at, by, i, this, had, not, are, but, from, or, have, an, they, which, one, were, you, all, her, she, there, would, their, we, him, been, has, when, who, will, no, more, if, out, its, so, up, said, what, about, than, into, them, can, only, other, time, new, some, could, these, two, may, first, then, do, any, like, my, now, over, such, our, man, me, even, most, made, after, also, well, did, many, before, must, years, back, through, much, where, your, way, down, should, because, long, each, just, state, people, those, too, how, Mr., little, good, world, make, very, year, still, see, own, work, men, day, get, here, old, between, both, life, being, under, three, never, know, same, last, another, while, us, off, might, great, states, go, come, since, against, right, came, take, used, himself, few, house, American, use, place, during, high, without, again, home, around, small, however, found, mrs, part, school, thought, went, say, general, once, upon, every, left, war, don't, does, got, united, number, hand, course, water, until, always, away, public, something, fact, less, through, far, put, head, think, called, set, almost, enough, end, took, government, night, yet, system, better, four, nothing, told, eyes, city, going, president, why, days, present, point, didn't, look, find, asked, second, group, later, next, room, social, business, knew, program, give, half, side, face, toward, white, five, let, young, form, given, per, order, large, several, national, important, possible, rather, big, among, case, often, early, john, things, looked, ever, become, best, need, within, felt, along, children, saw, church, light, power, least, family, development, interest, others, open, thing, seemed, want, area, god, members, mind, help, country, service, turned, door, done, law, although, whole, line, problem, sense, certain, different, kind, began, thus, means, matter, perhaps, name, times, york, itself, action, human, above, week, company, free, example, hands, local, show, history, whether, act, either, gave, death, feet, today, across, body, past, guite, taken, anything, field, having, seen, word, car, experience, I'm, money, really, class, words, already, college, information, tell, making, sure, themselves, together, full, air, shall, held, known, period, keep, political, real, miss, probably, century, question, seems, behind, cannot, major, office, brought, special, whose, boy, cost, federal, economic, self, south, problems, heard, six, study, ago, became, moment, run, available, job, street, result, short, west, age, change, position, board, individual, reason, 17

the, of, and, to, a, in, that, is, was, he, for, it, with, as, his, on, be, at, by, i, this, had, not, are, but, from, or, have, an, they, which, one, were, you, all, her, she, there, would, their, we, him, been, has, when, who, will, no, more, if, out, its, so, up, said, what, about, than, into, them, can, only, other, time, new, some, could, these, two, may, first, then, do, any, like, my, now, over, such, our, man, me, even, most, made, after, also, well, did, many, before, must, years, back, through, much, where, your, way, down, should, because, long, each, just, state, people, those, too, how, Mr., little, good, world, make, very, year, still, see, own, work, men, day, get, here, old, between, both, life, being, under, three, never, know, same, last, another, while, us, off, might, great, states, go, come, since, against, right, came, take, used, himself, few, house, American, use, place, during, high, without, again, home, around, small, however, found, mrs, part, school, thought, went, say, general, once, upon, every, left, war, don't, does, got, united, number, hand, course, water, until, always, away, public, something, fact, less, through, far, put, head, think, called, set, almost, enough, end, took, government, night, yet, system, better, four, nothing, told, eyes, city, going, president, why, days, present, point, didn't, look, find, asked, second, group, later, next, room, social, business, knew, program, give, half, side, face, toward, white, five, let, young, form, given, per, order, large, several, national, important, possible, rather, big, among, case, often, early, john, things, looked, ever, become, best, need, within, felt, along, children, saw, church, light, power, least, family, development, interest, others, open, thing, seemed, want, area, god, members, mind, help, country, service, turned, door, done, law, although, whole, line, problem, sense, certain, different, kind, began, thus, means, matter, perhaps, name, times, york, itself, action, human, above, week, company, free, example, hands, local, show, history, whether, act, either, gave, death, feet, today, across, body, past, guite, taken, anything, field, having, seen, word, car, experience, I'm, money, really, class, words, already, college, information, tell, making, sure, themselves, together, full, air, shall, held, known, period, keep, political, real, miss, probably, century, question, seems, behind, cannot, major, office, brought, special, whose, boy, cost, federal, economic, self, south, problems, heard, six, study, ago, became, moment, run, available, job, street, result, short, west, age, change, position, board, individual, reason, 18

With your back to the windows, walk straight through the door near the elevators. Continue to walk straight, going through one door until you come to an intersection just past a whiteboard. Turn left, turn right, and enter the second door on your right (sign says "Administrative Assistant"). the windows, walk the elevators. Continue to walk you come an intersection turn right, and enter the second door "Administrative Assistant").

the door , going one door a whiteboard. Turn left, (sign says

# Modeling Spatial Prepositions

- Functions that take geometric arguments.
- Classifiers for "to," "across," "towards," "through," "around," etc.
- Library of features.
- Train and test on labeled examples.

#### Across Video

#### peakDistanceToAxes = max(dist(figure, axes))



#### endPointsInGroundBoundingBox



#### displacementFromGround=d2 - d1



#### **Two Different Domains**

Video Retrieval



Direction
Understanding



### The person is going...







#### Corpus

![](_page_29_Picture_1.jpeg)

to the counter. along the east side of the island from the refrigerator. to the cabinet. across the kitchen.

#### Training Classifiers for Spatial Prepositions

Binary classifiers.

Positive examples of "across" are negative examples of "to."

Some exceptions (e.g., "to" and "towards.")

#### Corpus

![](_page_31_Picture_1.jpeg)

Positive example of "to the counter" Negative example of "across the counter"

![](_page_32_Figure_0.jpeg)

![](_page_33_Figure_0.jpeg)

![](_page_34_Figure_0.jpeg)

![](_page_35_Figure_0.jpeg)






#### endPointsInGroundBoundingBox



#### "to the sink"



#### Full model: 0.99 Bounding boxes: 0.76

#### "to the sink"



Full model: 0.10 Bounding boxes: 0.63

#### Results

- Modeled six spatial prepositions.
- Evaluated on a corpus of natural language descriptions of video clips.
- Analyzed which features work.

### Limitations

- Phrases, not sentences or paragraphs.
- All in one room.
- Only six spatial prepositions.

#### **Direction Understanding**

Joint work with Thomas Kollar

### Where should the robot go?





### Structure of Natural Language Directions

Sequential

















# Structure of Natural Language Directions

- Sequential.
- Directive verbs.

#### **Directive Verbs**

# Structure of Natural Language Directions

- Sequential.
- Directive verbs.
- Landmarks.

#### Landmarks

# Structure of Natural Language Directions

- Sequential.
- Directive verbs.
- Landmarks.
- Spatial relations.

### **Spatial Relations**

# Structure of Natural Language Directions

- Sequential.
- Directive verbs.
- Landmarks.
- Spatial relations.

Go through the door.

- Figure: <you>
- Verb: "Go"
- Spatial relation: "through"
- Landmark: "the door."

Continue to walk straight, going through one door until you come to an intersection just past a white board.

V: Continue to walk, SR: straight

Continue to walk straight, going through one door until you come to an intersection just past a white board.

V: going, SR: through, L: one door

Continue to walk straight, going through one door until you come to an intersection just past a white board.

V: going, SR: until, L: you come to an intersection just past a white board

Continue to walk straight, going through one door until you come to an intersection just past a white board.

F: you, V: come, SR: to, L: an intersection just past a white board

Continue to walk straight, going through one door until you come to an intersection just past a white board.

F: an intersection, SR: just past, L: a white board



#### Model

*argmax*<sub>path</sub> p(path, directions, observations)

### Model

p(path, directions, observations) =

*p*(*directions*|*path*, *observations*)*p*(*path*, *observations*) (definition of conditional probability)

p(directions | path, observations) p(path) p(observations)(path and observations are independent)

 $p(sdc_1...sdc_M|vp_1...vp_{M+1}, o_1...o_K) p(vp_1...vp_{M+1}) p(o_1...o_K)$ (topological map of viewpoints, and a sequence of SDCs)

#### **Topological Map**


## Model

p(path, directions, observations) =

*p*(*directions*|*path*, *observations*)*p*(*path*, *observations*) (definition of conditional probability)

p(directions | path, observations) p(path) p(observations)(path and observations are independent)

 $p(sdc_1...sdc_M|vp_1...vp_{M+1}, o_1...o_K) p(vp_1...vp_{M+1}) p(o_1...o_K)$ (topological map of viewpoints, and a sequence of SDCs)

 $\prod_{i=1}^{M} p(sdc_{i}|vp_{i}, vp_{i+1}, o_{1}...o_{K}) \prod_{i=1}^{M} p(vp_{i+1}|vp_{1}...vp_{i}) p(o_{1}...o_{K})$ (Paths are Markov)

## Model

p(path, directions, observations) =

*p*(*directions*|*path*, *observations*)*p*(*path*, *observations*) (definition of conditional probability)

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## Probability of Directions Given a Possible Movement

 $p(sdc_i|vp_i,vp_{i+1},o_1...o_K)$ 

=  $p(figure_i, verb_i, spatial\_relation_i, landmark_i | vp_i, vp_{i+1}, o_1...o_K)$ (The structure of a spatial description clause.)

## **Spatial Description Clause**

Go through the door.

- Figure: <you>
- Verb: "Go"
- Spatial relation: "through"
- Landmark: "the door."

## Probability of Directions Given a Possible Movement

 $p(sdc_i|vp_i,vp_{i+1},o_1...o_K)$ 

=  $p(figure_i, verb_i, spatial\_relation_i, landmark_i|vp_i, vp_{i+1}, o_1...o_K)$ (The structure of a spatial description clause.)

 $= p(figure_i | vp_i, vp_{i+1}, o_1 \dots o_K) p(verb_i | vp_i, vp_{i+1}) \times p(spatial\_relation_i | landmark_i, v_i, v_{i+1}, o_1 \dots o_K) \times p(landmark_i | v_i, v_{i+1}, o_1 \dots o_K)$ 

(Factor according to certain independence assumptions.)

## Probability of Directions Given a Possible Movement

 $p(sdc_i|vp_i,vp_{i+1},o_1...o_K)$ 

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(Factor according to certain independence assumptions.)

#### Observations



#### **Contextual relationships**



Tags include: desktop, monitor, computer, keyboard, and mouse

## Probability of Directions Given a Possible Movement

 $p(sdc_i|vp_i,vp_{i+1},o_1...o_K)$ 

=  $p(figure_i, verb_i, spatial\_relation_i, landmark_i|vp_i, vp_{i+1}, o_1...o_K)$ (The structure of a spatial description clause.)

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(Factor according to certain independence assumptions.)

## Probability of Directions Given a Possible Movement

 $p(sdc_i|vp_i,vp_{i+1},o_1...o_K)$ 

=  $p(figure_i, verb_i, spatial\_relation_i, landmark_i|vp_i, vp_{i+1}, o_1...o_K)$ (The structure of a spatial description clause.)

 $= p(figure_{i} | vp_{i}, vp_{i+1}, o_{1}...o_{K}) p(verb_{i} | vp_{i}, vp_{i+1}) \times \frac{p(spatial\_relation_{i} | landmark_{i}, v_{i}, v_{i+1}, o_{1}...o_{K})}{p(landmark_{i} | v_{i}, v_{i+1}, o_{1}...o_{K})}$ 

(Factor according to certain independence assumptions.)

# Learning Distributions for Spatial Relations



# Learning Distributions for Spatial Relations

Draw a path going down the hallway.



#### Performance of Classifiers



85

## High Scoring Examples of "past"







## Low Scoring Examples of "past"



#### Corpus of Natural Language Directions

**Instructions:** 

Write, in whatever way is natural to you, directions from the start label to the end label, with the goal of communicating this route to another person who does not have a map.



#### Corpus of Natural Language Directions

- Study 1 (Stata 3)
  - 10 people, 16 routes
- Study 2 (Stata 8)
  - 15 people, 10 routes
- Study 3 (Stata 1)
  - 15 people, 10 routes
- Study 4 (outdoors)
  - ongoing



#### Human Performance

	Stata 3	Stata 8	Stata 1
All directions	63%	85%	86%
Best Direction Giver	87%	100%	100%
Worst Direction Giver	43%	30%	20%

## Inference

- Global.
  - Viterbi-style algorithm.
  - Last-SDC only.
  - Landmarks only.
- Local.
- Random.



















## Improving Direction Understanding Results

- Polygon landmarks.
- Add more spatial relations.
- Use more complex environments.
- Changing the model.
  - Alignment.
  - Conditional Random Fields.
- Exploration.

## Next Steps

- Video retrieval.
  - Collect a corpus of paragraph-length descriptions of a person's activity.
- Direction generation.
- Compare spatial prepositions across domains.
- Create a lexicon of spatial prepositions.



## Schedule

- December 15, 2010 JAIR, direction understanding
- January 22, 2010 SIGIR, video retrieval
- March 1, 2010 TiCS on generation.
- April 1, 2010 JAIR, lexicon of spatial prepositions.
- April 14, 2010 thesis outline to Deb.
- May 14, 2010 thesis draft to Deb.
- May 31, 2010 thesis draft to committe.
- July, 2010 thesis defense

#### **Related Work**



## Related Work – Video Retrieval

- Katz et al. (2004) "Show me all the cars leaving the garage."
- Ivanov and Wren (2006) Query by example trajectory
- Fleischman et al. (2006) "Show me people making coffee."
- Naphade et al. (2006) Large-Scale Concept Ontology for Multimedia

## Contributions

- Create a library of features for grounding spatial prepositions.
- Analyze which features perform best for specific prepositions.
- Compare semantics of spatial prepositions in two different domains.
- Model higher level structures.

#### Acknowledgements

Tom Kollar Deb, Yuri, Boris, Cynthia, Cogmac ML support staff Gremio, Dave, Carie, Kai-yuh, Lin Ania & Seth & Moby Grandma Patches and Grandma Nicholas Mom, Dad, Staci & Seth, Shannon & Jon, Scott


### Contributions

- Create a library of features for grounding spatial prepositions.
- Analyze which features perform best for specific prepositions.
- Compare semantics of spatial prepositions in two different domains.
- Model higher level structures.

### Collaboration

- My thesis
  - Library of features for spatial prepositions.
  - Analyze which features perform the best.
  - Comparison of meanings of spatial relations in different domains.
- Tom's thesis
  - Grounding landmarks with flickr.
  - Planning to find objects.
- Both
  - Spatial description clauses.
  - MRF model of spatial language.

### **Directions With Only Landmarks**

the windows,

the elevators.

you

an intersection the second door

"Administrative Assistant").

the door one door a whiteboard. (sign

## Why is this a Media Lab thesis?

- Towards a new (or old) form of human-machine interaction.
- Multi-modal, multi-domain interface.
- Defining a new problem, with a new corpus.

- CRF chunker tags each word.
- Trained from annotated data.
- Features (tri-grams)
  - o Words
  - Part-of-speech tags
  - Label sequence
- Greedy algorithm groups tagged words together.

Continue to walk straight, going through one door until you come to an intersection just past a white board.

Continue to walk straight, going through one door until you come to an intersection just past a white board.

V: Continue to walk straight

Continue to walk straight, going through one door until you come to an intersection just past a white board.

V: going, SR: through, L: one door

Continue to walk straight, going through one door <mark>until</mark> you come to an intersection just past a white board.

SR: until, L: you come to an intersection

Continue to walk straight, going through one door until you come to an intersection just past a white board.

SR: just past, L: a white board



### **Computational Linguistics**

- Maron (1961)
  - First bag-of-words paper, using Naïve Bayes.
- Cleverdon (1967)
  - First corpus-based information retrieval evaluation.

### **Related Work**

- Cognitive Semantics
  - Landau and Jackendoff (1993)
  - Talmy (2005)
- Cognitive Science
  - Regier (1992)
  - Regier and Carlson (2001) attention vector sum for "above"
  - Carlson and Covey (2005) "How far is near? Inferring distance from spatial descriptions."

#### **Related Work**

- Turing (1950)
- Winograd (1971)

- Shrdlu

• Harnad (1990)

- Grounding