

AN INTRODUCTION TO NATURAL LANGUAGE GENERATION

64-419 Integriertes Seminar Sprachverarbeitung slides by Konstantin S. M. Möllers

How can we produce well-formed (i.e. readable, not just from simple text blocks composed) texts automatically?

-Prof. Michael Hess, University of Zürich

WHY NOT USE BLOCKS?

const tmpl = `The temperature will be between \${x.low}
and \${x.high} degrees.`;

const tmpl = `The temperature will be between \${x.low}
and \${x.high} degrees.`;

```
{
  "date": "2016-06-08",
  "high": 18,
  "low": 13,
  "feelsLike": 14
}
```

```
{
  "date": "2016-06-09",
  "high": 15,
  "low": 15,
  "feelsLike": 15
}
```

The temperature will be between 18 and 13 degrees.

The temperature will be between 15 and 15 degrees.



{
 "date": "2016-06-08",
 "high": 18,
 "low": 13,
 "feelsLike": 14
}

{
 "date": "2016-06-09",
 "high": 15,
 "low": 15,
 "feelsLike": 15
}

You: "Will it be warmer tomorrow?"

```
if (today.high < tomorrow.high)
  return "It's going to be warmer.";</pre>
```

What problems do you expect with this technique? In groups, 3 minutes.

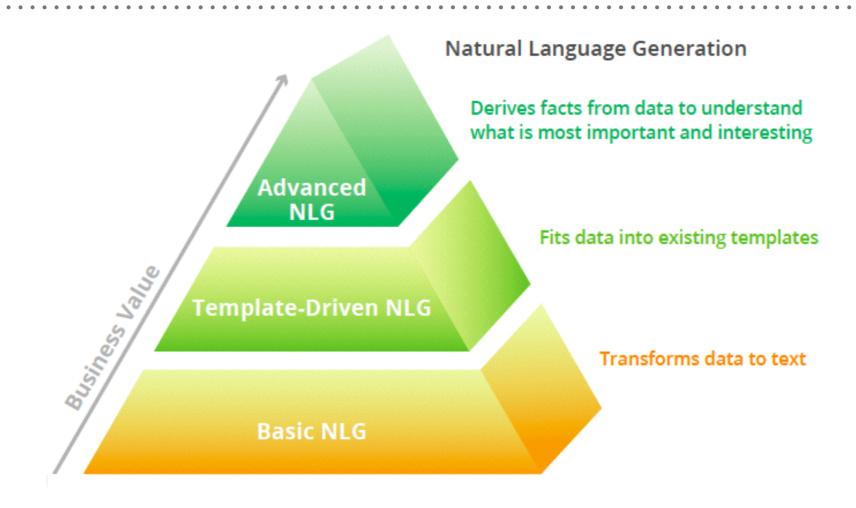
{
 "date": "2016-06-08",
 "high": 18,
 "low": 13,
 "feelsLike": 14
}

{
 "date": "2016-06-09",
 "high": 15,
 "low": 15,
 "feelsLike": 15
}

You: "Will it be warmer tomorrow?"

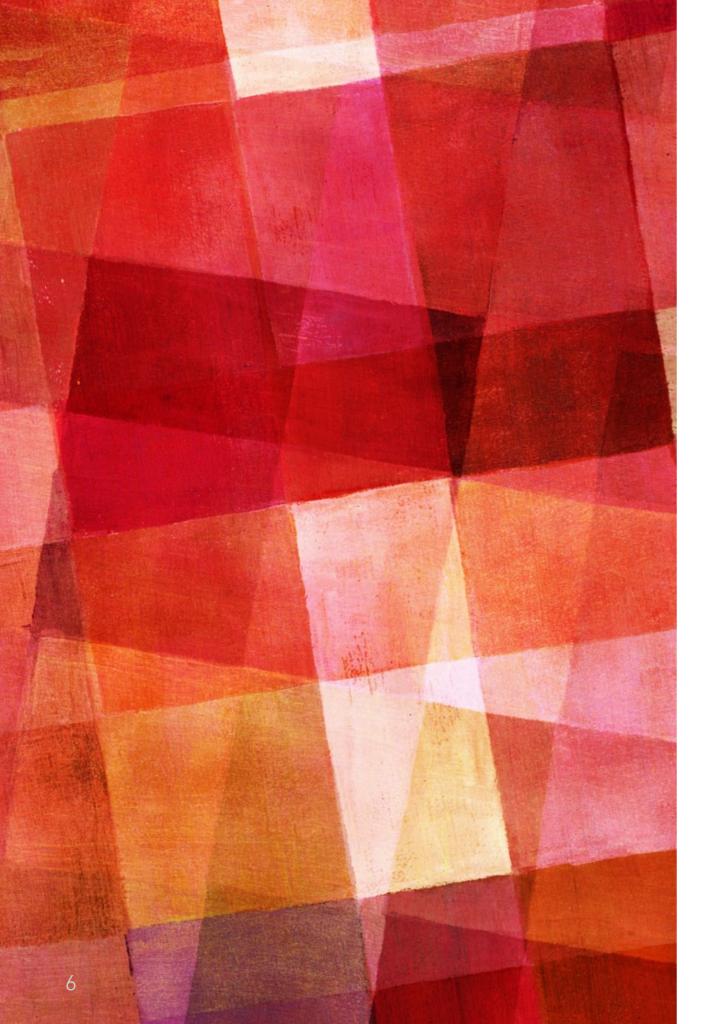
```
if (today.high < tomorrow.high)
return "It's going to be warmer.";</pre>
```

Computer: "For my CPU it's colder, but you will feel warmer!"



- ➤ automated earning reports
- > statement of benefits generation
- ➤ weather forecasts
- personalized advertising messages

[Glascott15]



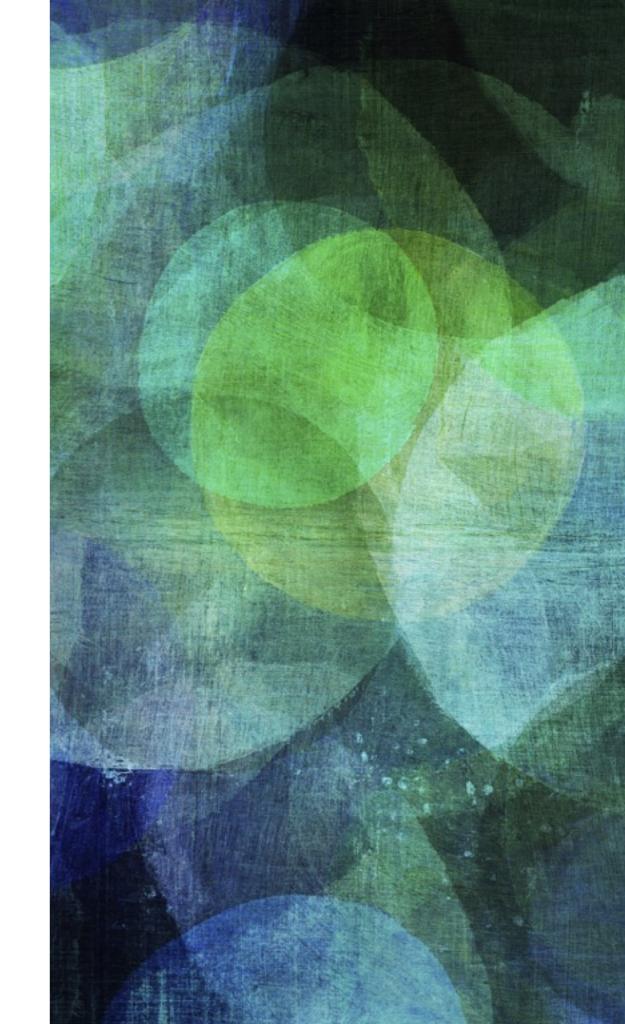
OUTLINE

1. Motivation



- 2. How it works
- 3. Realizations
- 4. Conclusion

HOW IT WORKS







STRATEGIC COMPONENT

What should be said?

TACTICAL COMPONENT

How should it be said?

[Hess05]

PLANNING A DISCOURSE — LOCATION PLAN

Points is strategic or tactical? Imagine some points of your own!

In groups, 5 minutes.

Select Landmarks on the Way

Use Nominalization

Inform the User Where she/he Currently is

Tell that there is a Traffic Light

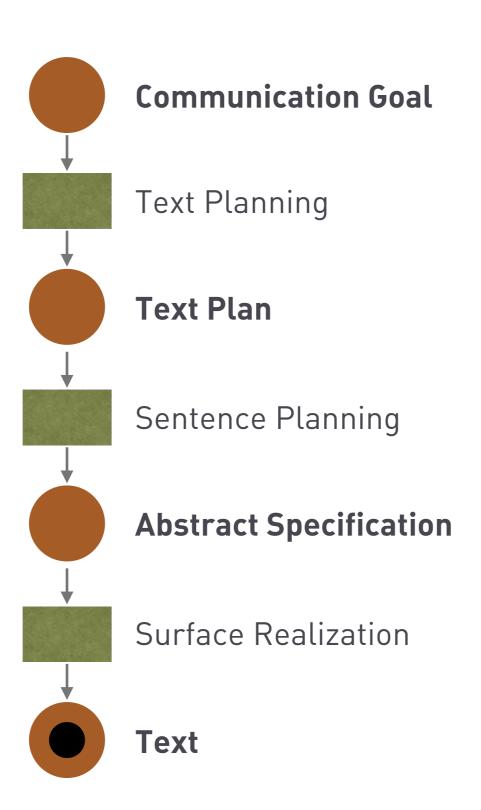
Merge two Similar Sentences

Ensure the User that she/he is Still on the Right Way

STRATEGY

TACTIC

DEFAULT NLG ARCHITECTURE



[Horacek10]

TEXT GENERATION SITUATION

A text generation situation can be described by a quadruple $\langle k, c, u, d \rangle$, where

- \triangleright k is the **knowledge base**,
- \succ c the communication goal,
- ➤ u the user model and
- \blacktriangleright d the discourse history.

TEXT PLANNING

```
const c = Describe(Route, A, B);

const k = [
    At(User, A),
    Path(A, C), Path(C, D), Path(D, B),
    Left(Path(A, C), Path(C, D)),
    Straight(Path(C, D), Path(D, B)),
    Type(C, "church"),
    Type(D, "town-hall"),
    Property(D, "big"),
    ...
];
```

```
const l = [
  be(User, at(A)),

  go(User, from(A), to(C)),
  go(User, from(C), to(D)),
  go(User, from(D), to(B)),

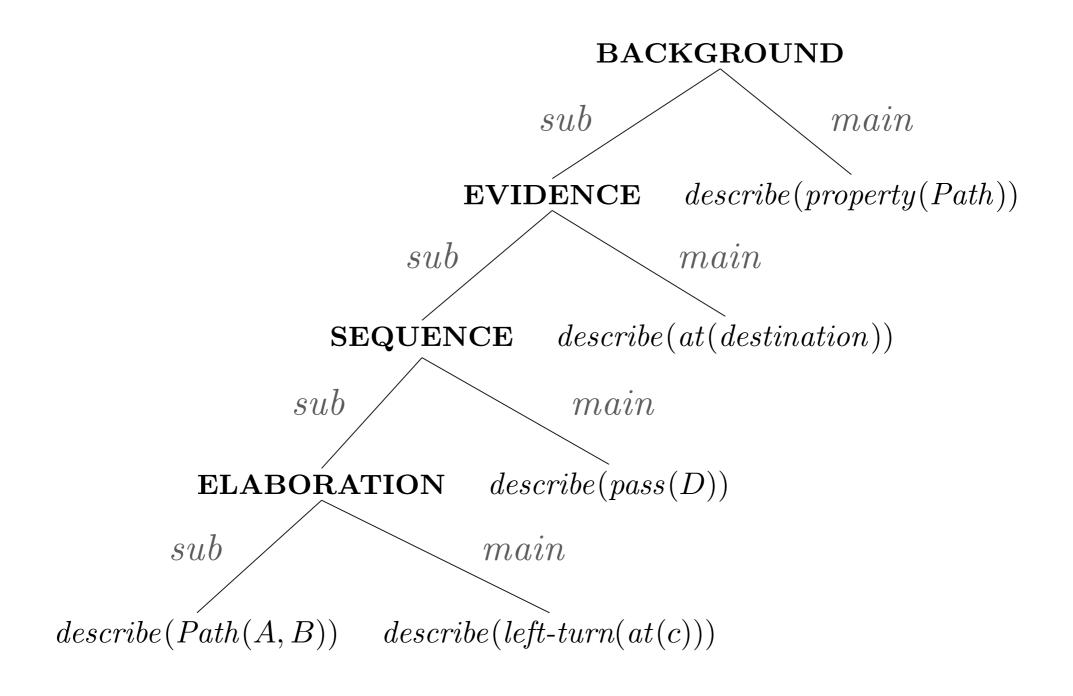
  turn(User, at(C), direction(left)),
  turn(User, at(D), direction(none)),

  church(C), town_hall(D), big(D), ...
];
```

- ➤ define a concept
- ➤ compare two objects
- ➤ describe available information

[Horacek10, McKeown85]

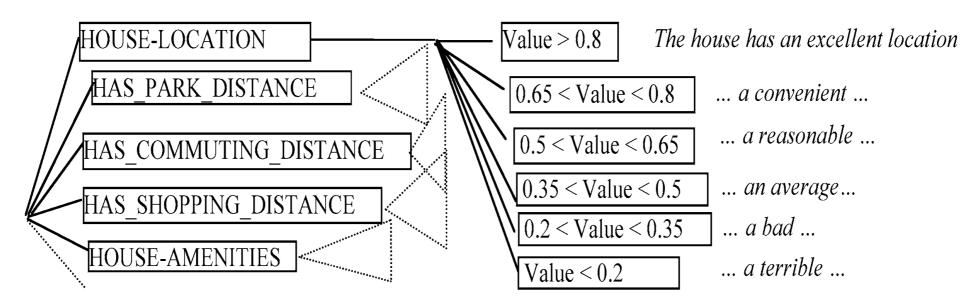
TEXT PLANNING RESULT



[Horacek10, McKeown85]

SENTENCE PLANNING

➤ lexicalization: find lexical elements to represent predicates



➤ aggregation: gaining sentence representations out of clauses

```
const clauses = [
   boil(Hans, Water),
   bake(Hans, Pie),
];
assertEquals(aggregate(clauses), [
   Hans,
   and(boil(Water), bake(Pie)),
]);
```

reference expressions: identify an and inform about objects

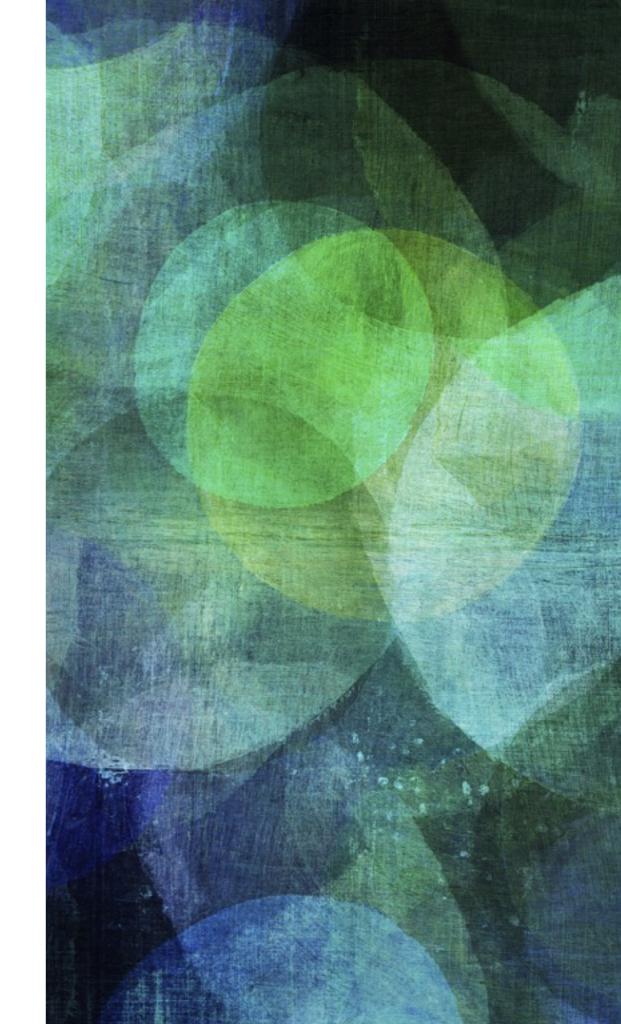


SURFACE REALIZATION

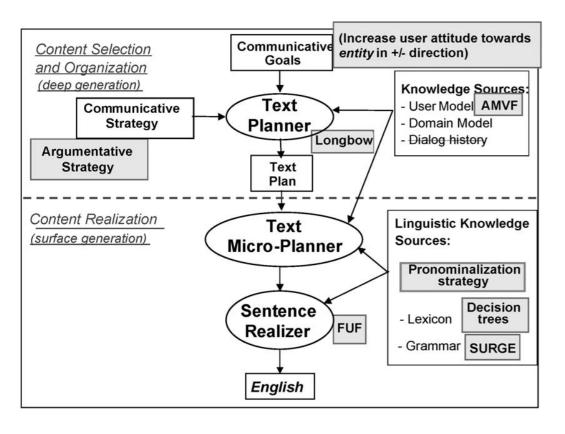
- > structure-driven: using a grammar
 - ➤ Direction free: like Prolog, interpretation agnostic
 - Top-down: left-recursive resolution of the input
- ➤ lexeme-driven: using a lexicon
 - ➤ Shake-and-bake: try every combination of lexicon entries and check grammar
 - ➤ Chart generation: subsumes lexicon entries under input

[Horacek10]

REALIZATIONS



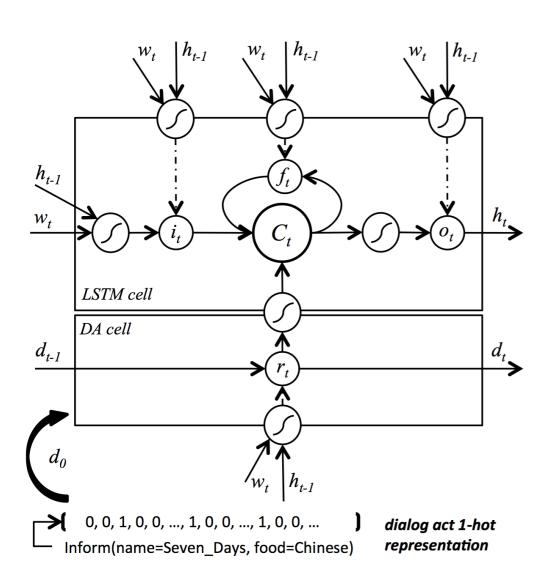
USING ARGUMENTATION THEORY



- ➤ **GEA**: evaluative argument generator
 - ➤ for doctors to advise patients
 - ➤ for salesman to advise customers
- ➤ using rhetoric means
- ➤ domain specific model extracted
- evaluation using empirical means

[Walker+04, Carenini+06]

USING NEURAL NETWORKS



- ➤ LSTM: long short-term-memory
- ➤ trained from unaligned data
- > jointly optimizes

Sentence Planning

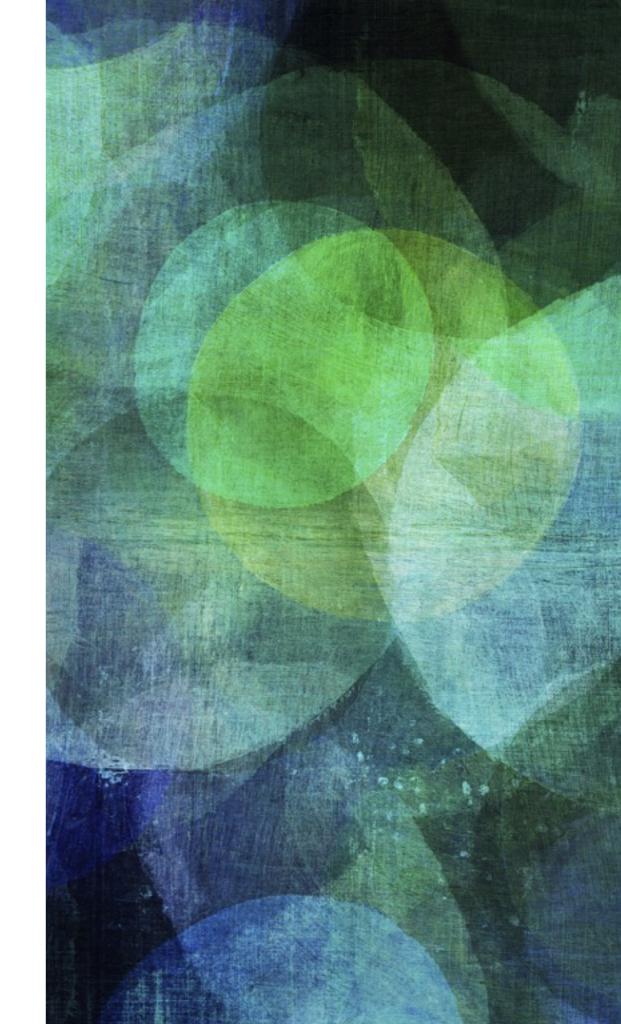


Surface Realization

USING NEURAL NETWORKS

#	Example Dialogue Acts and Realizations from SF Restaurant Domain
1	inform(name="red door cafe", goodformeal="breakfast", area="cathedral hill", kidsallowed="no")
	red door cafe is a good restaurant for breakfast in the area of cathedral hill and does not allow children .
	red door cafe is a good restaurant for breakfast in the cathedral hill area and does not allow children .
	red door cafe is a good restaurant for breakfast in the cathedral hill area and does not allow kids .
	red door cafe is good for breakfast and is in the area of cathedral hill and does not allow children .
	red door cafe does not allow kids and is in the cathedral hill area and is good for breakfast .
2	informonly(name="dosa on fillmore and kiss seafood", pricerange="expensive", near="lower pacific heights")
	there is no place other than dosa on fillmore and kiss seafood that are expensive near to lower pacific heights .
	dosa on fillmore and kiss seafood is the only expensive restaurant near lower pacific heights .
	the only listed restaurant near lower pacific heights in the expensive price range is dosa on fillmore and kiss seafood.
	i apologize , dosa on fillmore and kiss seafood is the only expensive restaurant near lower pacific heights .
	i apologize , dosa on fillmore and kiss seafood are the only expensive restaurants near lower pacific heights .
#	Example Dialogue Acts and Realizations from SF Hotel Domain
3	inform(type="hotel",count="182",dogsallowed="dontcare")
	there are 182 hotels if you do not care whether dogs are allowed .
	there are 182 hotels if you do not care whether they allow dogs .
	182 hotels are available if dogs allowed or not is not an issue .
	there are 182 hotels if allowing dogs or not is not an issue.
	there are 182 hotels if whether dogs are allowed does not matter .
4	informonly(name="red victorian bed breakfast",acceptscreditcards="yes",near="haight",hasinternet="yes")
	red victorian bed breakfast is the only hotel near haight and accepts credit cards and has internet .
	red victorian bed breakfast is the only hotel near haight and has internet and accepts credit cards .
	red victorian bed breakfast is the only hotel near haight that accept credit cards and offers internet .
	the red victorian bed breakfast has internet and near haight, it does accept credit cards.
	the red victorian bed breakfast is the only hotel near haight that accepts credit cards , and offers internet .

CONCLUSION



SUMMARY

- ➤ well-formed text generation has high interest in economy
- ➤ flexibility for heterogenous domain, like in dialog systems
- > strategic (,,what") vs. tactical (,,how") component
- ➤ default NLG architecture:
 - 1. text planning
 - 2. sentence planning
 - 3. surface realization
- ➤ Carenini and Moore use rhetoric to generate evaluative arguments
- ➤ Wen et al. use recurrent neural networks to optimize the generated natural language w.r.t. its variation

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IMAGE SOURCES

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