

# Language and Computation

week 13, Thursday, April 24

*Tamás Biró*

*Yale University*

tamas.biro@yale.edu

<http://www.birot.hu/courses/2014-LC/>



## Practical matters

- **“Superficial” reading:** JM 25  
<http://birot.hu/courses/2014-LC/readings.txt>
- **Assignment 4:** returned, remarks posted
- **Assignment 5:** posted
- **Python:** if needed, programming section
- **Section/review sections next week:** let Jen know!
- **Final exam:** Fr 05/02, 9 am; Mo 05/05, 7 pm. Room t.b.a.  
In-class, no laptop, open-book—extra copies needed?

# Today

- Machine Translation: traditional approaches
- Machine Translation: statistical approaches
- Summary: humans and computers

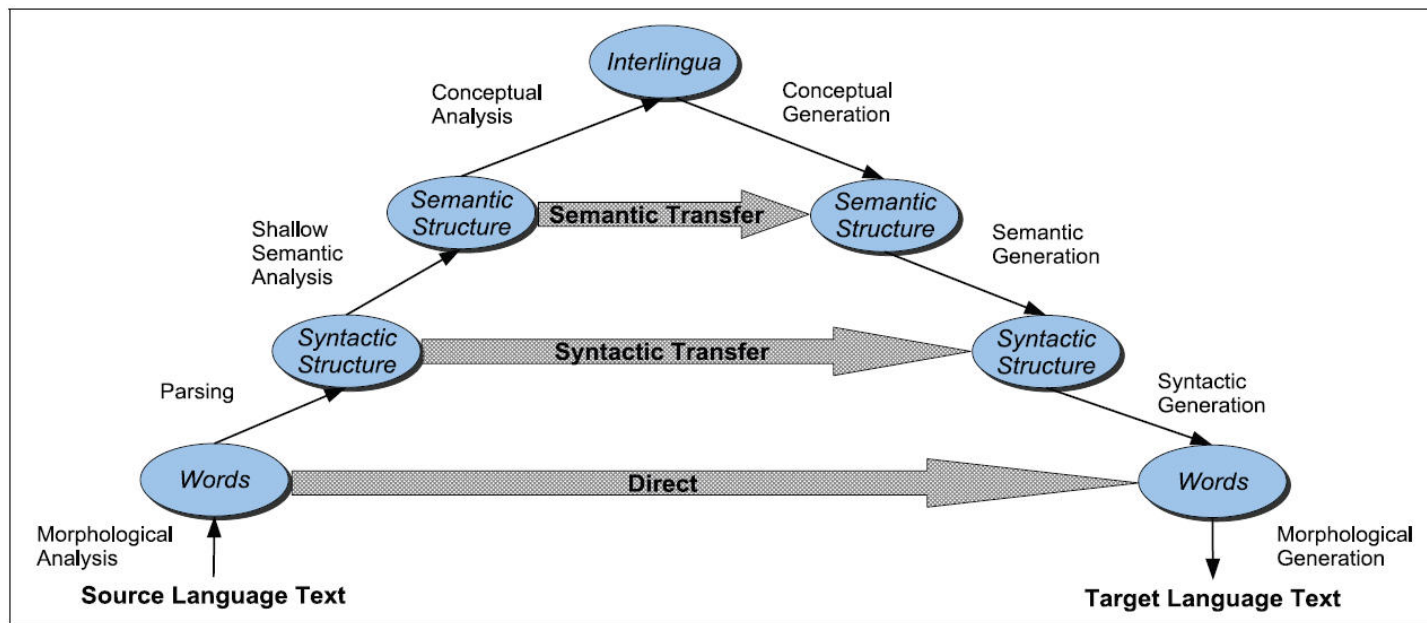


# Classical Machine Translation



# Classical Machine Translation

## The Vauquois Triangle (1968):



# Classical Machine Translation

- Separate MT system for each language pair, or
  - deploying *contrastive knowledge*
  - $n$  languages  $\rightarrow$  there are  $n(n - 1)$  pairs  
(Cf. translation in the European Union: 24 official languages.)
- MT via universal *interlingua*
  - via truly universal semantic representation
  - via English, Esperanto, etc.

# Classical Machine Translation

- Principled ways vs. *ad hoc* solutions:
  - Part-of-speech taggers, parsers, disambiguation, ontologies, etc.
  - Hand-crafted rules
- Balancing three, contradicting goals:
  - High quality
  - No need for human intervention  
(cf. feedback loops in conversational agents)
  - Breadth of domain  
→ portability to new domains, domain adaptation  
“scaling up” to larger domains

# Statistical Machine Translation





# Bayesian MT

(Brown et al. 1993)

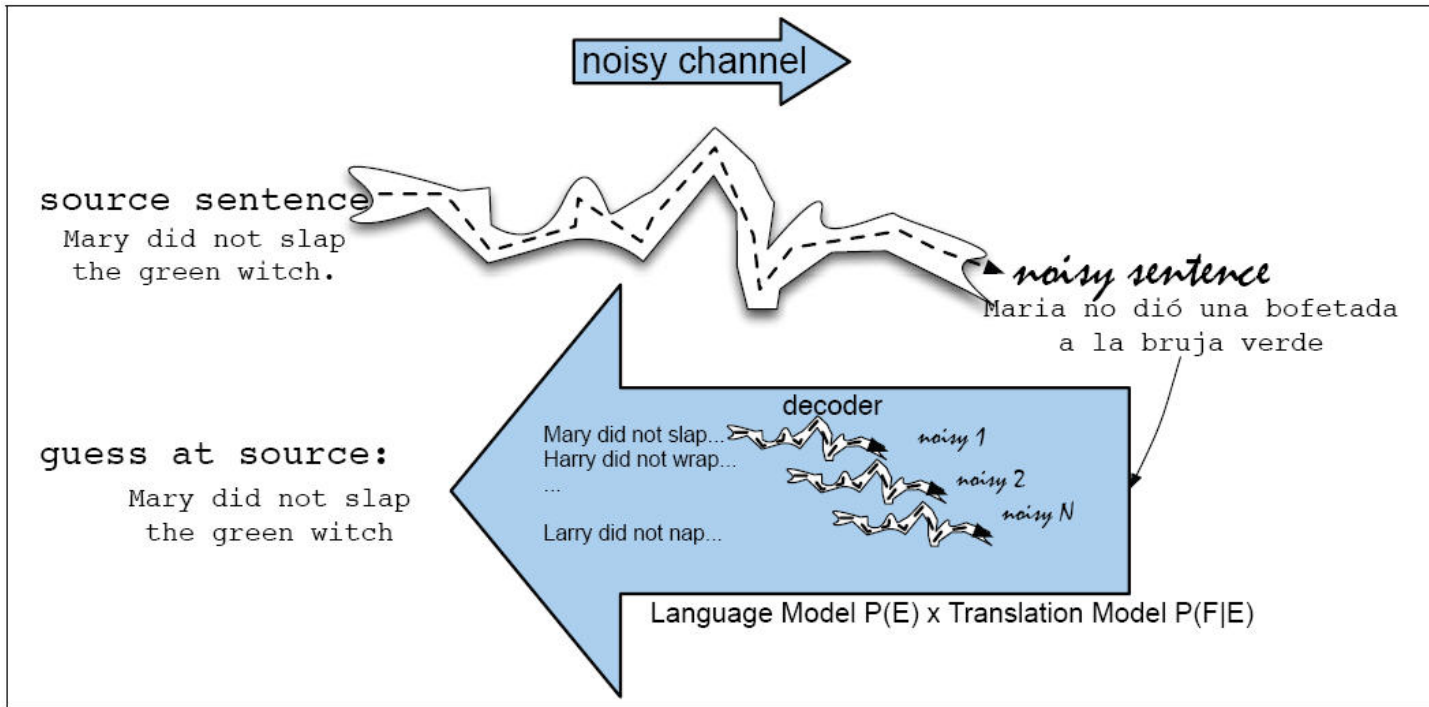
- Translating French sentence  $f$  to English  $e$ :  
Given  $f$ , which  $e$  maximizes  $\text{Prob}(e|f)$ ?
- Bayes' theorem:

$$\text{Prob}(e|f) = \frac{\text{Prob}(e) \cdot \text{Prob}(f|e)}{\text{Prob}(f)}$$

- Hence, find  $e$  that maximizes  $\text{Prob}(e) \cdot \text{Prob}(f|e)$ .



# Bayesian MT



# Bayesian MT

- A classical search problem:  
find  $e$  such that  $e = \arg \max_{e'} \text{Prob}(e') \cdot \text{Prob}(f|e')$ .
- Problem decomposed. Parameters estimated from corpora:
  - **Language model**  $\text{Prob}(e)$ : quality of English translation  
Estimated piecemeal from corpus of English.  
No need to care for correspondence with French.
  - **Translation model**  $\text{Prob}(f|e)$ : E-F correspondences  
Estimated piecemeal from **aligned** bilingual corpora.  
No need to care for quality of the generated English text.
  - **Decoder**: similar idea to spelling, speech, etc.

# Summary: humans and computers



# Goals of doing “Language and Computation”

- Linguistics: computational linguistics
- Computer science: Natural Language Processing
- Electrical engineering: Speech Recognition
- Psychology, cognitive science: Computational Psycholinguistics



# Goals of doing “Language and Computation”

- Supporting (theoretical) linguistics:  
novel approaches to language analysis.
- Computational issues raised by (theoretical) linguistics and cognitive science:  
E.g., parsability, learnability, mental implementation of theoretical constructs.
- Language related tasks when answering the needs of the computer users: processing natural language data.

# Humans and computers

- Human–computer interaction
- Should computers imitate humans?
  - Performance? (e.g., prone to errors)
  - Mechanism?

E.g., parsing algorithms vs. human parsing.

E.g., the Chomsky hierarchy and natural languages.

Thank you for your attention  
during the entire semester!

