
Probabilistic Representations of Linguistic Meaning (“PReLiM”)



Jason Eisner



Jelinek Memorial Workshop, July 2014

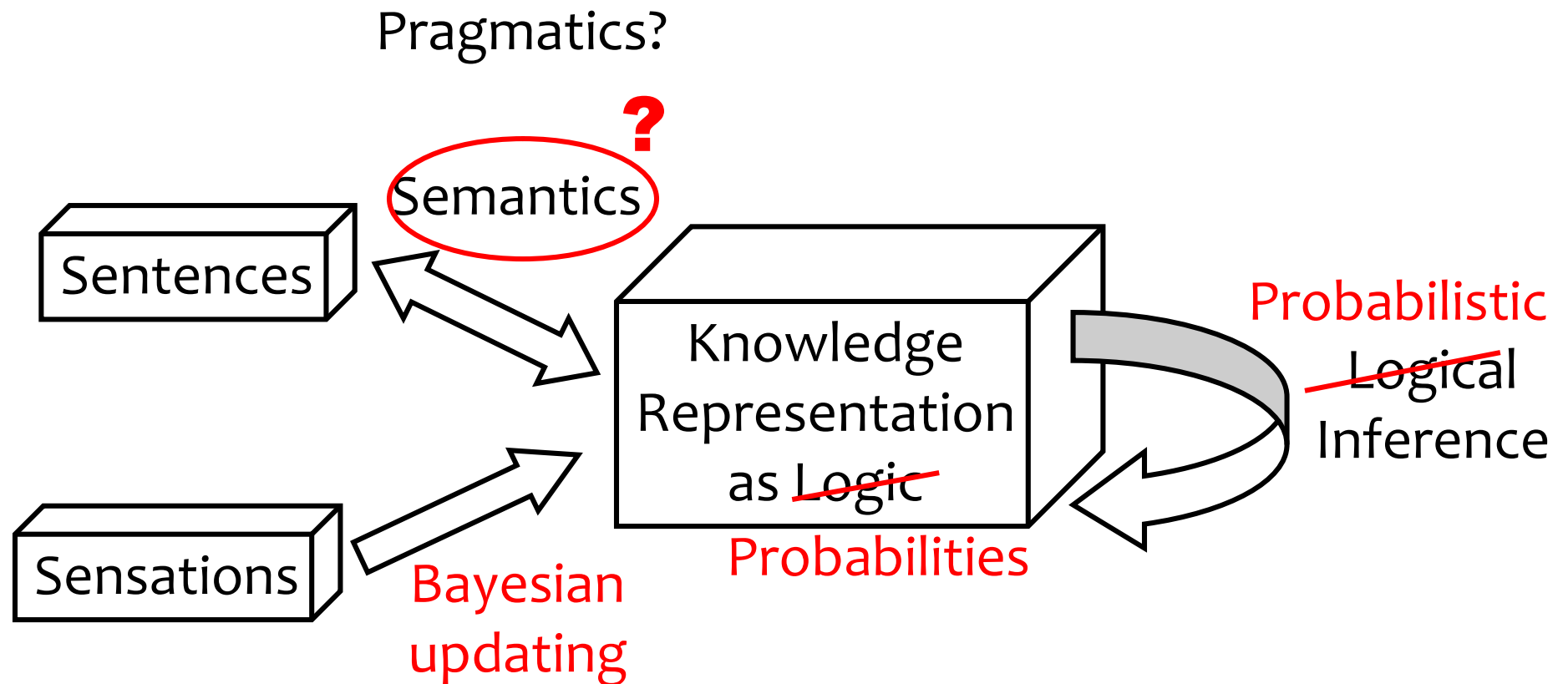
Week 1: A blue-sky workshop

- What should computational semantics look like in 10 years?
- More of what semanticists think about
- Integrated with reasoning and background knowledge (“language of thought”)
- Integrated with pragmatics
- Fuller probabilistic treatment (generative)



Why a “probabilistic treatment”?

Reason 1: Integration

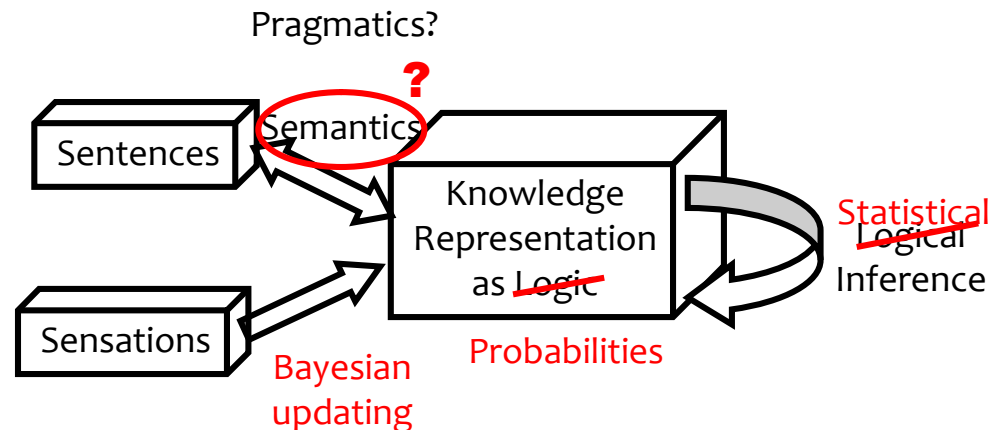


Why a “probabilistic treatment”?

Reason 1: Integration

- This is going to be harder than just Bayes nets.
 - What is the probabilistic version of *modal* logic?
 - How do we deal with richly *structured* beliefs?
 - How about *incomplete, vague, or contradictory* ones?

- We may need some new ideas about probabilistic modeling.



Why a “probabilistic treatment”?

Reason 1: Integration

- Natural language is related to the “language of thought.”
- Fillmore 1982:

“This sentence did not give you this information directly; you had to 'compute' some of it by constructing, in your imagination, a complex context within which each of the lexically signaled framings was motivated.”

(“The decedent while on land and in mufti last weekend ate a typical breakfast and read a novel high in flip strength.”)

- What happens in a car crash?
- Think of a widow. How old is she?
 - How do you know?
 - Compositional semantics: How about “the widow of the firefighter”?
 - “The widow of the village chieftain”?

Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- For some decades, computational linguistics has been trying to reinvent linguistics “so it works”
- From grammars to probability distributions (from “what’s possible?” to “what’s probable?”)
- Starting with phonetics and collocations



Thanks, Fred

Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- **Phonetics:** Gaussian mixtures, n-grams, ...
- **Phonology, morphology:** probabilistic FSTs, stochastic OT, graphical models over many strings, ...
- **History of words:** probabilistic evolutionary models
- **Word collocations:** n-gram models, topic models, embeddings, ...
- **Syntax:** PCFGs, selectional preference models, ...
- **Translation:** synchronous grammars, ...

Model the joint
distribution of elements
of form and meaning



Algorithms for robust
comprehension,
production, and learning

Why a “probabilistic treatment”?

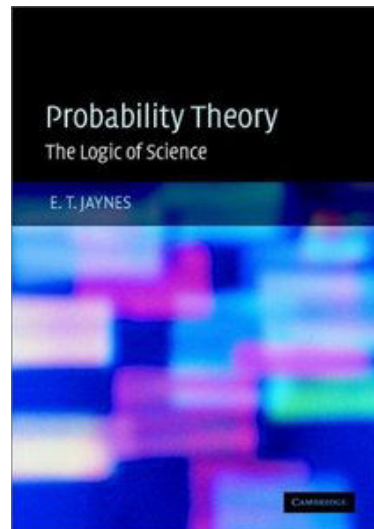
Reason 2: Stochasticization has paid off for computational linguistics

- This is because **knowledge of language** includes **knowledge of probabilities**.
- It’s amazing that we can communicate so effectively at such high bandwidth!
- Seems to require prior probabilities to help reconstruct what is left out.
 - Certainly for syntax ... and also for semantics.

Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- Models get richer in linguistic insight over time.
 - The goal of science is to find the underlying probability distributions that can explain and predict our observations.



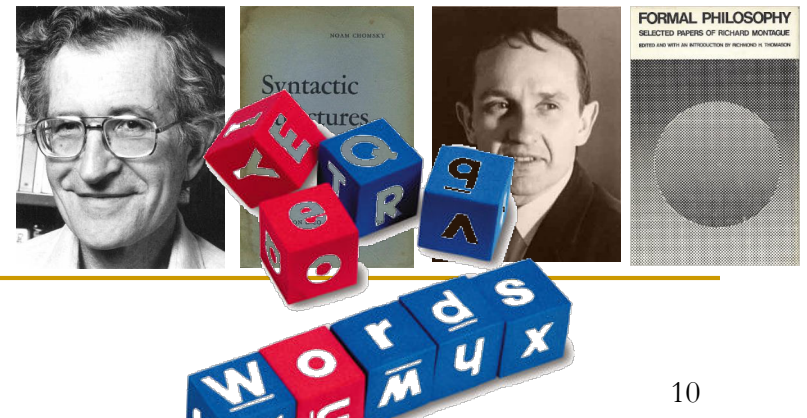
E. T. Jaynes (2003),
Probability Theory:
The Logic of Science



Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- Models get richer in linguistic insight over time.
 - The goal of science is to find the underlying probability distributions that can explain and predict our observations.
 - Linguists shouldn't be alarmed. “Do not think that I have come to abolish the Law or the Prophets; I have not come to abolish them but to fulfill them.”



Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- Models get richer in linguistic insight over time.
 - ❑ The goal of science is to find the underlying probability distributions that can explain and predict our observations (Jaynes 2003).
 - ❑ Linguists shouldn't be alarmed. “Do not think that I have come to abolish the Law or the Prophets; I have not come to abolish them but to fulfill them.”
 - ❑ But we're only human so it takes time. 😊
 - ❑ And in fact, it helps to start simple.

Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- Models get richer in linguistic insight over time.
 - Helps to start simple.

■ Case study: PCFGs

- Formalism just tries to capture the most important things about syntax: phrase types, hierarchical structure
- **Flexible:** no commitment to particular nonterminals or rules
- **Explanatory adequacy?** (*sufficiently strong inductive bias?*)
 - Missing a lot of things that probably belong in UG
 - But you can get more specific theories via priors over grammars
- **Descriptive adequacy?** (*sufficiently weak inductive bias?*)
 - Maybe not, but the formal treatment points the way to more complicated variants (TAG, HPSG, continuous nonterminals ...)

Why a “probabilistic treatment”?

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- Models get richer in linguistic insight over time.
 - Helps to start simple.
- Case study: PCFGs
 - Case study: (stochastic) Optimality Theory
 - Formalism just tries to capture the most important things about phonology: interaction among constraints
 - **Flexible:** no commitment to particular constraints or representations

Why a “probabilistic treatment”?

Reason 2: Stochasticization has paid off for computational linguistics

- Models get richer in linguistic insight over time.
 - Helps to start simple.
- Case study: PCFGs
- Case study: (stochastic) Optimality Theory
 - Case study: probabilistic semantics?
 - What should we capture initially?
 - What are the basic kinds of objects we’re dealing with?
 - How do they fit together?
 - We might not handle verb aspect right away, or physical predicates; but there should be somewhere to fit them in

Why a “probabilistic treatment”?

Reason 3: Lots of specific semantic/pragmatic phenomena ...

- You’re probably going to lose now, since that rook is pinned.
- She may have left. Her car is gone.
- I should cancel the milk.
- Children must be carried. *[sign on the escalator]*
- Most parents are likelier to give in on weekends.
- If you can afford a cheap car, you can afford expensive coffee.
- If you had poured the coffee, it wouldn’t have spilled.

Participants

‘Big, architectural thinkers’

- Jason Eisner, JHU (CS/cogsci)
- Oren Etzioni, UW/Allen Institute (CS)
- Shalom Lappin, KCL (philosophy)
- Dan Lassiter, Stanford (psycholinguistics)
- Percy Liang, Stanford (CS/stats)
- Staffan Larsson, Gothenburg (philosophy/linguistics)
- David McAllester, TTI-Chicago (CS)
- James Pustejovsky, Brandeis (linguistics)
- Benjamin Van Durme, JHU (CS/cogsci)

- JHU students:
 - Nick Andrews (CS)
 - Drew Reisinger (cogsci)
 - Darcey Riley (CS)
 - Rachel Rudinger (CS)

Virtual Participants

Couldn't be here this week but asked to be involved

- [Joshua Tenenbaum](#), MIT (cogsci)
- [Noah Goodman](#), Stanford (psych)
- [Barbara Partee](#), UMass (ling)
- [Gerhard Jaeger](#), Tübingen (ling)
- [Michael Franke](#), Amsterdam (ling)
- [Christopher Potts](#), Stanford (ling)
- [Kyle Rawlins](#), JHU (ling)
- [Robin Cooper](#), Gothenburg (phil)
- [Igor Douven](#), Groningen (phil)
- [Luke Zettlemoyer](#), UW (CS)
- [Ido Dagan](#), Bar-Ilan (CS)
- [Len Schubert](#), Rochester (CS)
- [Dan Klein](#), Berkeley (CS)
- JHU students:
 - [Frank Ferraro](#) (CS)
 - [Pushpendre Rastogi](#) (CS)

(Kyle and Frank are joining next week)

Setting the stage

- **Mon am: Introductions**
 - Our goals and interests
 - Our desiderata and warnings of pitfalls
- **Mon pm: James Pustejovsky (plenary talk)**
 - Why it is important to distinguish "possible" from "probable"
meaning shifts: How distributions impact linguistic theory
- **Tue am: Taking stock**
 - Hard examples
 - What's already understood
 - Planning our time

Towards a probabilistic language of thought

- **Tue am:** Shalom Lappin (plenary talk)
 - A Rich Probabilistic Type Theory for the Semantics of Natural Language
- **Tue pm:** Knowledge representation
 - Belief, theory of mind
 - Metaphor and meaning shift
 - Chalktalk by Darcey/Jason on locally renormalized PCFG?
- **Tue pm:** Oren Etzioni (plenary talk)
 - Semantics, Science, and 10-Year-Olds
- **Wed am:** Worlds and situations
 - Generics, quantifiers
 - Modals, conditionals, counterfactuals
 - Chalktalk by Drew on dialogue scenario?

Pragmatics

- **Wed am:** Dan Lassiter (plenary talk)
 - Bayesian Pragmatics
- **Wed pm:** Pragmatics
 - Meta-reasoning (chalktalk by Dan?)
 - Presuppositions and implicatures
 - Game theory

Linguisticization

- **Wed pm: David McAllester (plenary talk)**
 - The Problem of Reference
- **Thu am: Linguisticization**
 - Lexical semantics, event semantics
 - Framing
 - Linguistic marking
 - definiteness, information structure, modality, evidentials, classifiers, conventional implicatures

Grounding

- **Thu am:** Staffan Larsson (plenary talk)
 - Perceptual Semantics and Coordination in Dialogue
- **Thu pm:** Grounding
 - Perception
 - Vagueness
 - Temporal and spatial reasoning

Making it happen

- **Thu pm:** Percy Liang (plenary talk)
 - The State of the Art in Semantic Parsing
- **Fri am:** Remaining difficult issues
 - E.g., imprecise language, contradictory beliefs, linguistic ambiguity about contrast sets
- **Fri am:** Martha Palmer (plenary talk)
 - Designing Abstract Meaning Representations for Machine Translation
- **Fri pm:** Practical next steps toward “semantic AI”
 - Chalktalks by Rachel, Nick?

Goals

- What are the constraints on a full theory?
 - Listen closely to each other ...
- Can we agree on a baseline theory (like PCFG)?
 - Or at least rule out directions that won't work?
- What conceptual / mathematical work still needs to be done?
- What first steps can we take toward actually building semantic AI?
 - Starting with the remaining 3 weeks
(student work on PReLiM; Martha's CLAMR group)

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- What should computational semantics look like in 10 years? Let's get the foundations right.
- More of what semanticists think about
- Integrated with reasoning and background knowledge (“language of thought”)
- Integrated with pragmatics
- Fuller probabilistic treatment (generative)

