Interpretation of Conditionals

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Conditionals

If you pour sugar in your coffee, it tastes great. But if you pour sugar and gasoline in your coffee, it tastes awful.

Material implication

Interpretation of natural language conditionals match predictions by material implication?

$A \rightarrow C$	A/C dependent?	$\neg A \rightarrow \neg C$?
If you mow the lawn, I'll give you \$5.(Geis & Zwicky, 1971)	yes	yes
	Conditional P	erfection

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If the train is on time, we'll join you at the bar.	yes	rather yes
If you scratch on the eight-ball, you will loose the game.(von Fintel, 2001)	yes	no

$A \rightarrow C$	A/C dependent?	$\neg A \rightarrow \neg C$?
If you mow the lawn, I'll give you \$5.(Geis & Zwicky, 1971)	yes	yes
,	Conditional P	Perfection
If the train is on time, we'll join you at the bar.	yes	rather yes
If you scratch on the eight-ball, you will loose the game.(von Fintel, 2001)	yes	no
There are cookies on the sideboard, if you want some.(Austin, 1961)	no	no
	Biscuit Cond	ditionals 🕟 📳

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Simple Indicative Conditionals

If Mary comes to the party, \mathcal{M} Joe will come as well. \mathcal{J}

Expectations

- Joe attending the party depends on whether Mary attends the party
- $P(J \mid M)$ is high
- Speaker doesn't know whether M holds, nor whether J holds

Uncertainty: "If A, C" describes different situations.

Situation 1

	Α	$\neg A$
С	0.9	0.025
¬ <i>C</i>	0.05	0.025

$$P(C \mid A) \approx 0.947$$

 $P(C) = 0.925$

Situation 2

	Α	$\neg A$
С	0.65	0.05
¬ <i>C</i>	0.07	0.23

$$P(C \mid A) \approx 0.903$$

 $P(C) = 0.7$

Conditionals

In which situations would a speaker choose to say "If A, C"?

 \rightarrow When she is not in the position to utter "C"

RSA-model

- takes alternative utterances into account
- reflects uncertainty by making predictions in terms of probability distributions
- gives the possibility to include prior knowledge about context of utterance

Rational-Speech-Act model

$$P_{PL}(s \mid u) \propto P_{speaker}(u \mid s) \cdot P_{bn}(s)$$
 Bayes' rule

$$P_{S}(u \mid s) \propto exp(\alpha \cdot U(u; s))$$
 Softmax

$$U(u; s) = \log P_{LL}(s \mid u) - \cos(u)$$
 Utility function

$$P_{LL}(s \mid u) \propto \delta_{\llbracket u \rrbracket(s)} \cdot P_{bn}(s)$$
 Literal interpretation

$$P_{bn}(s = \langle cn, t \rangle) = P(t \mid cn) \cdot P_{cn}(cn)$$
 Bayes net prior

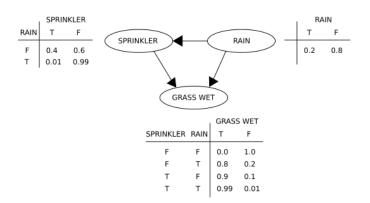
Model components

What do we need to specify?

- 1. Definition of world states
- Prior over world states
- 3. Utterances
- 4. Utterance costs
- 5. Literal Meaning

World States

Causal Bayes nets (Pearl, 1988, 2009)



(https://en.wikipedia.org/wiki/Bayesian_network)

World States

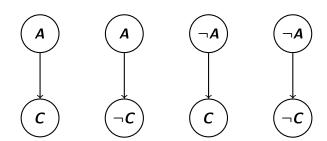
Set of **joint probability distributions** *t*

 \Rightarrow We get a set of **tuples**:

$$\left\{ \begin{array}{l} < t_0, cn_1 >, < t_0, cn_2 >, \dots, < t_0, cn_9 > \\ < t_1, cn_1 >, < t_1, cn_2 >, \dots, < t_1, cn_9 > \\ & \dots \end{array} \right\}$$

Causal networks

1. Dependent networks

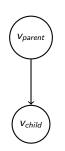


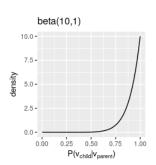
 $(\times 2 \text{ for switched direction})$

World States

 $Likelihood < table, cn_{dep} >$

What is the likelihood for a given probability distribution to be associated with a particular **dependent** causal network?





$$P(t \mid cn_{dep}) \propto \operatorname{beta}(P(v_{child} \mid v_{parent}) \mid 10, 1) \cdot \tag{1}$$

$$\operatorname{beta}(P(v_{child} \mid v_{parent}) \mid 1, 1) \cdot \tag{2}$$

$$\operatorname{beta}(P(v_{child} \mid v_{parent}) \mid 1, 1) \cdot \tag{3}$$

World States

 $Likelihood < table, cn_{AllC} >$

What is the likelihood for a given probability distribution to be associated with the **independent** causal network?



$$\bigcirc$$

$$A||C \Rightarrow P(A, C) = P(A) \cdot P(C)$$

$$p_1: |P(C \mid A) - P(C \mid \neg A)| \leqslant \epsilon$$

$$p_2: |P(A \mid C) - P(A \mid \neg C)| \leqslant \epsilon$$

$$P(t \mid cn_{ind}) \propto \begin{cases} \text{beta}(0.99 \mid 10, 1) & \text{if } p_1 \text{ and } p2 \text{ hold} \\ \text{beta}(0.01 \mid 10, 1) & \text{else} \end{cases}$$
 (2)

Literal Meaning

1. Set of states to which utterance is applicable

utterance	[u]
If A, C	$ \{ \langle cn, t \rangle P(C \mid A) \geqslant \theta \} $
A and C	$\{\langle cn, t \rangle P(A, C) \geqslant \theta\}$
Α	$\{\langle cn, t \rangle P(A) \geqslant \theta\}$
likely A	$\left \left. \left\{ < cn, t > \right P(A) \geqslant \theta_{likely} \right\} \right $

$$\theta = 0.9, \theta_{likely} = 0.5$$

2. Utterance costs

$$U \rightarrow_{0} L \mid N \mid P \mid C \mid I$$

$$N \rightarrow_{0.125} \neg L$$

$$T \rightarrow_{0} L \mid N$$

$$P \rightarrow_{0.1} \text{ likely } T$$

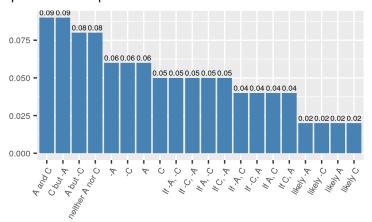
$$C \rightarrow_{0.25} T \text{ and } T$$

$$I \rightarrow_{0.55} \text{ if } T, T$$

$$L \rightarrow_{0} \text{ "A"} \mid \text{"C"}$$

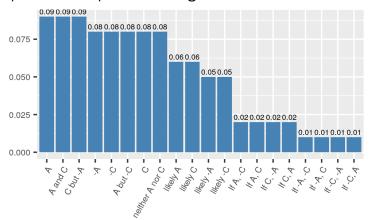
Utterance Costs

Average speaker distribution computed from 1000 Bayes nets sampled from the prior. **Without costs**.



Utterance Costs

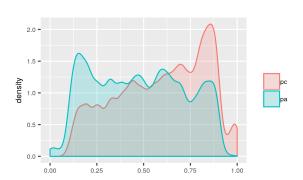
Average speaker distribution computed from 1000 Bayes nets sampled from the prior. **Including costs**.



Speaker uncertainty

Utterance: Any simple indicative conditional "**If A, C**", e.g. *If the sun shines, I'll go climbing.*

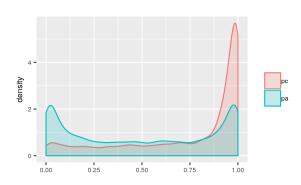
Pragmatic Listener



- P(A) and P(C) mainly in (0.1, 0.9)
- P(C) increases roughly linearly in interval
- *P*(*A*) roughly uniform in interval

Results Speaker Uncertainty

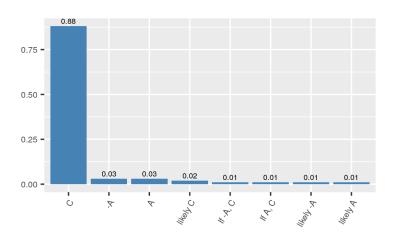
Literal Listener



 is not able to draw inference that C is uncertain

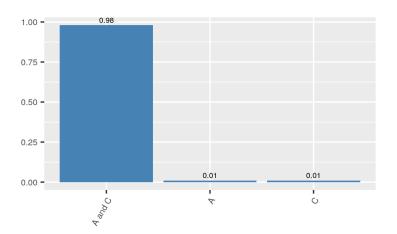
Results

Speaker: C is true (but not $C \wedge A$ nor $C \wedge \neg A$)

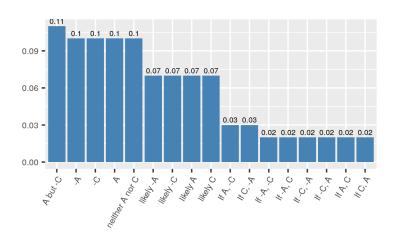


Results

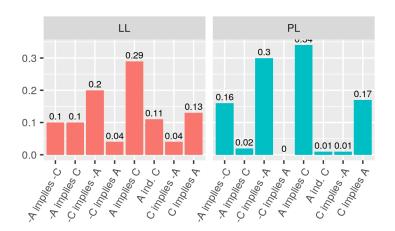
Speaker: C and A are true



Speaker: C is not true



Causal Networks

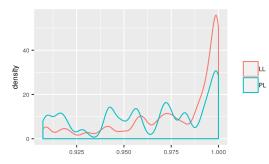


Biscuit conditionals

Prior context

Knowledge about the world: A and C are independent for any Biscuit conditional. P(cn=A||C)=1

P(C)



- Literal Listener is already able to make inference that C is likely true!
- Interpretation arises from world knowledge
- I-implicature

Theory of Implicatures

(Atlas & Levinson, 1981; Levinson, 1987; Horn, 1984)

Q-implicatures



- alternative utterances taken into account, i.e. S could have said "I have 5, 6, ... grandchildren."
- Grice's Maxim of Quantity: Say as much as necessary, but not more.

I-implicatures



- Interpretation arises "from an inference to the best interpretation"
- world knowledge taken into account
- L does not reason about what S could have said



Biscuit Conditionals

The model does not consider syntactical differences, but they may influence the interpretation:

Simple Indicative Conditional

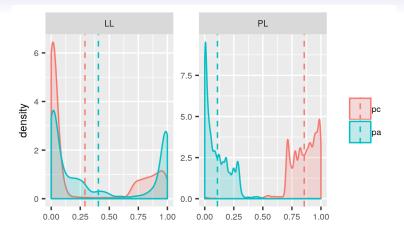
Wenn du mich brauchst, (dann) arbeite ich heute zu Hause. (If you me need, (then) work I at home)

Biscuit Conditional

Wenn du mich brauchst, ich arbeite heute zu Hause. (If you me need, *I work* today at home)

"Sarah and Marian have arranged to go for sundowners at the Westcliff hotel tomorrow. Sarah feels there is some chance that it will rain, but thinks they can always enjoy the view from inside. To make sure, Marian consults the staff at the Westcliff hotel and finds out that in the event of rain, the inside area will be occupied by a wedding party." (Douven & Romeijn, 2011, p.7, lines 1-6)

"If it rains tomorrow, we cannot have sundowners at the Westcliff."

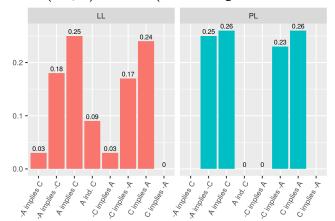


Utterance: "If A, not C" Prior: P(C) > 0.7 favored and probability for A and C being independent very high (P(A||C) = 0.98)

Conditional Perfection

Utterance: "If you mow the lawn, I'll give you \$5."

Prior: $P(\neg A, C)$ is small. (not mowing lawn and listener gets \$5).



Conditional Perfection Strength

