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Embedded implicatures as pragmatic inferences under compositional lexical uncertainty

Christopher Potts

Stanford Linguistics

Paper, code, data: https://github.com/cgpotts/pypragmods





Mike Frank



Dan Lassiter



Roger Levy

Conversational implicature

Definition

Speaker S saying U to listener L conversationally implicates q iff

- **1** S and L mutually, publicly presume that S is cooperative.
- **2** To maintain **1** given U, it must be supposed that S thinks q.
- S thinks that both S and L mutually, publicly presume that L is willing and able to work out that <a>2 holds.

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Example

Ann: What city does Paul live in? Bob: Hmm ... he lives in California.

- (A) Assume Bob is cooperative.
- (B) Bob supplied less information than was required, seemingly contradicting (A).
- (C) Assume Bob does not know which city Paul lives in.
- (D) Then Bob's answer is optimal given his evidence.

Conversational implicature

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Implicature as social, interactional

Implicatures are inferences that listeners make to reconcile the speaker's linguistic behavior with the assumption that the speaker is cooperative.

Implicatures and cognitive complexity

The speaker must believe that the listener will infer that the speaker believes the implicature.

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Two strands of inquiry

Interactional models

- Embrace the social nature of implicatures.
- Derive implicatures from nested belief models with cooperative structure.
- Focus on contextual variability and uncertainty.

Grammar models

- Limit interaction to semantic interpretation.
- Derive implicatures without nested beliefs or cooperativity.
- Place variability and uncertainty outside the theory of implicature.

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My goal

Despite divisive rhetoric, the two sides in this debate are not in opposition, but rather offer complementary insights.

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Plan for today

- Scalar implicature
- 2 Grammar-driven models of implicature
- 3 The compositional lexical uncertainty model
- 4 Experiment: scalars under quantifiers
- 6 Model assessment

Example

A: Sandy's work this term was satisfactory.

- Contextual premise: the speaker A intends to exhaustively answer 'What was the quality of Sandy's work this term?'
- Contextual premise: A has complete knowledge of Sandy's work for the term (say, A assigned all the grades for the class).
- 3 Assume A is cooperative in the Gricean sense.
- The proposition q that Sandy's work was excellent is more informative than p, the content of A's utterance.
- **6** q is as polite and easy to express in this context as p.
- **6** By **1**, q is more relevant than p.
- **7** By **3**-**6**, A must lack sufficient evidence to assert q.
- (3) By (2), A must lack evidence for q because q is false.

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Simplified scalar implicature reasoning

Context: the speaker is a sportscaster who fully observed the outcomes and intends a complete and accurate report:





Simplified scalar implicature reasoning

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Simplified scalar implicature reasoning

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a.	Worlds:	NN	NS	NA	SN	SS	SA AN	AS	AA	
b.	Literal:				SN	SS	SA AN	AS	AA	'at least some'
c.	Implicature:	NN	NS	NA	SN	SS	SA			'not all'
d.	Communicated:				SN	SS	SA			'only some'



Scalar implicatures under universal quantifiers

Every player hit some of his shots.



Scalar implicatures under universal quantifiers

Every player hit some of his shots.





Scalar implicatures under universal quantifiers

Every player hit some of his shots.



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Scalar implicatures under universal quantifiers



a.	Worlds:	NN	NS	NA	SN	SS	SA	AN	AS	AA	
b.	Literal:					SS	SA		AS	AA	'all hit at least some'
c.	Implicature:	NN	NS	NA	SN	SS	SA	AN	AS		'not all hit all'
d.	Result:					SS	SA		AS		'all hit some; not all hit all'
e.	Aux. premise:	NN				SS				AA	'uniform outcomes'
f.	Communicated:					SS					'all hit only some'



Scalar implicatures under non-monotone quantifiers

Exactly one player hit some of his shots.



- a. Worlds: NN NS NA SN SS SA AN AS AA
- b. Literal: NS NA SN AN
- c. Local: NS SN SA AS

'exactly one hit at least some' 'exactly one hit only some'



Scalar implicatures under non-monotone quantifiers

Exactly one player hit some of his shots.



- a. Worlds: NN NS NA SN SS SA AN AS AA
- b. Literal: NS NA SN AN
- c. Local: NS SN SA AS

'exactly one hit at least some' 'exactly one hit only some'



Scalar implicatures under non-monotone quantifiers





- a. Worlds: NN NS NA SN SS SA AN AS AA
- b. Literal: NS NA SN AN
- c. Local: NS SN SA AS

'exactly one hit at least some' 'exactly one hit only some'



Scalar implicatures under downward-entailing quantifiers



AN

AA

- a. Worlds: NN NS NA SN SS SA AN AS AA
- b. Literal: NN
- c. Local: NN NA

'none hit some' 'none hit only some'



Scalar implicatures under downward-entailing quantifiers



AN

AA

- a. Worlds: NN NS NA SN SS SA AN AS AA
- b. Literal: NN
- c. Local: NN NA

'none hit some' 'none hit only some'

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Grammar-driven models

Scalar implicature

2 Grammar-driven models of implicature

- 3 The compositional lexical uncertainty model
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Grammar models

Gennaro Chierchia, Danny Fox, and Benjamin Spector (2012), 'The grammatical view of scalar implicatures'

"More specifically, the facts suggest that SIs are not pragmatic in nature but arise, instead, as a consequence of semantic or syntactic mechanisms, which we've characterized with the operator, O. This operator, although inspired by Gricean reasoning, must be incorporated into the theory of syntax or semantics, so that — like the overt operator *only* — it will find its way to embedded positions."

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Exhaustification

Definition (Exhaustification operator)

$$\mathcal{O}_{ALT}(\varphi) = \llbracket \varphi \rrbracket \sqcap \bigsqcup \{ -q : q \in ALT(\varphi) \land \llbracket \varphi \rrbracket \not\sqsubseteq q \}$$

the exhaustified meaning is the literal meaning plus the negation of all stronger alternatives

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Scalar implicatures in logical forms



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Scalar implicatures in logical forms



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Implicit interactionality

Chierchia et al.

"the facts suggest that SIs are not pragmatic in nature but arise, instead, as a consequence of semantic or syntactic mechanisms"

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Implicit interactionality

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"the facts suggest that SIs are not pragmatic in nature but arise, instead, as a consequence of semantic or syntactic mechanisms"

Resolving underspecification pragmatically

The grammatical system specifies a one-to-many mapping from surface forms to logical forms. Only a pragmatic theory can explain how discourse participants coordinate on these LFs.

Implicit interactionality

Chierchia et al.

"the facts suggest that SIs are not pragmatic in nature but arise, instead, as a consequence of semantic or syntactic mechanisms"

Resolving underspecification pragmatically

The grammatical system specifies a one-to-many mapping from surface forms to logical forms. Only a pragmatic theory can explain how discourse participants coordinate on these LFs.

Chierchia et al.

"one can capture the correlation with various contextual considerations, under the standard assumption [...] that such considerations enter into the choice between competing representations (those that contain the operator and those that do not)."

Coordinating on a logical form in context

Example

A: Sandy's work this term was satisfactory. Potential implicature: Sandy's work was not excellent

Available logical forms:

Sandy's work was

- [satisfactory]
- O_{ALT([[satisfactory]])={[[excellent]]}([[satisfactory]])}
- O_{ALT([[satisfactory]])={[[good]],[[excellent]]}([[satisfactory]])}

The compositional lexical uncertainty model

1 Scalar implicature

2 Grammar-driven models of implicature

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Agents
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Definition (Literal listener)

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Definition (Pragmatic speaker)

 $s_1(msg \mid world, Lex) \propto \exp \lambda (\log I_0(world \mid msg, Lex) - C(msg))$

Definition (Literal listener)

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Definition (Pragmatic listener)

 $I_1(world \mid msg, Lex) \propto s_1(msg \mid world, Lex)P(world)$

Definition (Pragmatic speaker)

 $s_1(msg \mid world, Lex) \propto \exp \lambda (\log I_0(world \mid msg, Lex) - C(msg))$

Definition (Literal listener)

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Definition (Lexical uncertainty listener)

$$L(world \mid msg) \propto \sum_{Lex \in L} P_L(Lex)s_1(msg \mid world, Lex)P(world)$$

Definition (Pragmatic speaker)

 $s_1(msg \mid world, Lex) \propto \exp \lambda (\log I_0(world \mid msg, Lex) - C(msg))$

Definition (Literal listener)

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The Rational Speech Acts (RSA) model



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The Rational Speech Acts (RSA) model

N	bosket	Å										
S	4	Å	*				NSA					
		2				A scored	011		N.33	sc	core	d 0
А						A aced	001		S .33		ace	d 0
						0	111		A .33		(0 5
(a) F	ossi	ble v	world	ds		(b)	М	-	(c) Prior	(d) Co	sts
	N	S	A	-		A scored A	aced	0		N	S	A
A scored	0	.5	.5	-	N	0	0	1	A scored	0	.67	.33
A aced	0	0	1		S	.99	0	.01	A aced	0	0	1
0	.33	.33	.33		A	.5	.5	0	0	.99	.01	0
(;	a) <i>l</i> ₀			-		(b) <i>s</i> ₁			(0	:) L		

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1 It's a sofa, not a couch.

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- 1 It's a sofa, not a couch.
- 2 synagogues and other churches

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- 1 It's a sofa, not a couch.
- 2 synagogues and other churches
- 3 superb but not outstanding

- 1 It's a sofa, not a couch.
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- 3 superb but not outstanding
- 4 some . . .

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Definition (Refinement)

- Let φ be a set-denoting expression. X is a refinement of φ iff X ≠ Ø and X ⊆ [[φ]].
- 2 $\mathcal{R}_c(\varphi)$, the set of refinements for φ in context c, is constrained so that $\llbracket \varphi \rrbracket \in \mathcal{R}_c(\varphi)$ and $\mathcal{R}_c(\varphi) \subseteq \wp(\llbracket \varphi \rrbracket) - \emptyset$

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Definition (Refinement)

- Let φ be a set-denoting expression. X is a refinement of φ iff X ≠ Ø and X ⊆ [[φ]].
- 2 R_c(φ), the set of refinements for φ in context c, is constrained so that [[φ]] ∈ R_c(φ) and R_c(φ) ⊆ ℘([[φ]])−Ø

Example

1
$$D = \{a, b\}$$

2 $[[Player A]] = \{Y \subseteq D : a \in Y\}$
 $= \{\{a, b\}, \{a\}\}$
3 $\mathcal{R}_c(Player A) = \begin{cases} \{\{a, b\}, \{a\}\} \\ \{\{a, b\}\} \\ \{\{a\}\} \end{cases}$

Compositional semantics under lexical uncertainty

Refinements	Lexica	Semantic composition
$(\{\{a, b\}, \{a\}\})$	$[Player A] = \{\{a, b\}, \{a\}\}$ $[scored] = \{a, b\}$	<pre>[[Player A]]([[scored]]) = 1</pre>
$\mathcal{R}_{c}(Player A) = \begin{cases} \{\{a,b\}\} \\ \{\{a\}\} \end{cases}$	$[Player A] = \{\{a, b\}, \{a\}\}$ $[scored] = \{a\}$	$[\![Player A]\!]([\![scored]\!]) = 1$
({a, b})	$[Player A] = \{\{a, b\}, \{a\}\}$ $[scored] = \{b\}$	[[<i>Player A</i>]]([[scored]]) = 0
$\mathcal{R}_{c}(\text{scored}) = \left\{ \{a\} \\ \{b\} \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \$	$[Player A] = \{\{a, b\}\}$ $[scored] = \{a, b\}$	$\llbracket Player A \rrbracket (\llbracket scored \rrbracket) = 1$
	$[Player A] = \{\{a, b\}\}$ $[scored] = \{a\}$	[[<i>Player A</i>]]([[scored]]) = 0
	[[<i>Player A</i>]] = {{ <i>a, b</i> }} [[<i>scored</i>]] = { <i>b</i> }	$[\![Player A]\!]([\![scored]\!]) = 0$
	<pre>[[Player A]] = {{a}} [[scored]] = {a, b}</pre>	$[\![Player A]\!]([\![scored]\!]) = 0$
	<pre>[[Player A]] = {{a}} [[scored]] = {a}</pre>	$[\![Player A]\!]([\![scored]\!]) = 1$
	[[<i>Player A</i>]] = {{a}} [[<i>scored</i>]] = { <i>b</i> }	$[\![Player A]\!]([\![scored]\!]) = 0$

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Simple scalar implicature

		N S A	
,		A scored 0 .71 .29	
L		A aced 0 0 1	
		0 .75 .25 0	
	1	\downarrow	\searrow
	A scored A aced 0	A scored A aced 0	A scored A aced 0
0.	N 0 0 1	N 0 0 1	N 0 0 1
51	S .99 0 .01	S .99 0 .01	S 0 01
	A .33 .67 0	A 0 .99 .01	A .5 .50
		\downarrow	↓
	N S A	N S A	N S A
,	A scored 0 .5 .5	A scored 0 1 0	A scored 0 0 1
<i>I</i> 0	A aced 0 0 1	A aced 0 0 1	A aced 0 0 1
	0.33.33.33	0.33.33.33	0.33.33.33
		\rightarrow	
	N S A	N S A	N S A
	A scored 0 1 1	A scored 0 1 0	A scored 0 0 1
М	A aced 0 0 1	A aced 0 0 1	A aced 0 0 1
	0 1 1 1	0 1 1 1	0111
	<u>↑</u>	<u>↑</u>	<u>↑</u>
0	$[scored] = \{\langle S, a \rangle, \langle A, a \rangle\}$	$[scored] = \{(S, a)\}$	$[scored] = \{\langle \mathbf{A}, \mathbf{a} \rangle\}$
Ľ	$[aced] = \{\langle A, a \rangle\}$	$\llbracket aced \rrbracket = \{ \langle A, a \rangle \}$	$\llbracket aced \rrbracket = \{\langle A, a \rangle\}$

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored	0.0	0.0	0.0	0.24	0.19	0.16	0.18	0.16	0.07
Player A aced	0.0	0.0	0.0	0.0	0.0	0.0	0.36	0.3	0.34
Player B scored	0.0	0.24	0.18	0.0	0.19	0.16	0.0	0.16	0.07
Player B aced	0.0	0.0	0.36	0.0	0.0	0.3	0.0	0.0	0.34
some player scored	0.0	0.14	0.11	0.14	0.17	0.14	0.11	0.14	0.05
some player aced	0.0	0.0	0.22	0.0	0.0	0.19	0.22	0.19	0.18
every player scored	0.0	0.0	0.0	0.0	0.31	0.27	0.0	0.27	0.14
every player aced	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0
no player scored	0.31	0.14	0.12	0.14	0.06	0.05	0.12	0.05	0.01
no player aced	0.18	0.19	0.08	0.19	0.14	0.06	0.08	0.06	0.0
0	0.01	0.01	0.32	0.01	0.01	0.15	0.32	0.15	0.0

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.24					
Player A aced					1		0.36		
Player B scored		0.24							
Player B aced			0.36						
some player scored					0.17				
some player aced			0.22				0.22		
every player scored					0.31				
every player aced									1.0
no player scored	0.31								
no player aced		0.19		0.19					
0			0.32				0.32		

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.24					
Player A aced							0.36		
Player B scored		0.24							
Player B aced			0.36						
some player scored					0.17				
some player aced			0.22				0.22		
every player scored					0.31				
every player aced									1.0
no player scored	0.31								
no player aced		0.19		0.19					
0			0.32				0.32		

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.24					
Player A aced							0.36		
Player B scored		0.24							
Player B aced			0.36						
some player scored					0.17				
some player aced			0.22				0.22		
every player scored					0.31				
every player aced									1.0
no player scored	0.31								
no player aced		0.19		0.19					
0			0.32				0.32		

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.24					
Player A aced							0.36		
Player B scored		0.24							
Player B aced			0.36						
some player scored					0.17				
some player aced			0.22				0.22		
every player scored					0.31				
every player aced									1.0
no player scored	0.31								
no player aced		0.19		0.19					
0			0.32				0.32		

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.24					
Player A aced							0.36		
Player B scored		0.24							
Player B aced			0.36						
some player scored					0.17				
some player aced			0.22				0.22		
every player scored					0.31				
every player aced									1.0
no player scored	0.31								
no player aced		0.19		0.19					
0			0.32				0.32		

- **1** $\mathcal{R}_c(\text{Player A}) = \{ [\text{Player A}], [only Player A] \}$
- 2 $\mathcal{R}_c(\text{Player B}) = \{ [\![\text{Player B}]\!], [\![only Player B]\!] \}$
- 3 R_c(some) = {[[some]], [[some and not all]]}
- 6 R_c(scored) = {[[scored]], [[scored and didn't ace]]}

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored	0.0	0.0	0.0	0.45	0.11	0.22	0.15	0.05	0.02
Player A aced	0.0	0.0	0.0	0.0	0.0	0.0	0.42	0.36	0.22
Player B scored	0.0	0.45	0.15	0.0	0.11	0.05	0.0	0.22	0.02
Player B aced	0.0	0.0	0.42	0.0	0.0	0.36	0.0	0.0	0.22
some player scored	0.0	0.25	0.09	0.25	0.06	0.12	0.09	0.12	0.01
some player aced	0.0	0.0	0.24	0.0	0.0	0.21	0.24	0.21	0.11
every player scored	0.0	0.0	0.0	0.0	0.61	0.16	0.0	0.16	0.07
every player aced	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0
no player scored	0.61	0.0	0.16	0.0	0.0	0.0	0.16	0.0	0.06
no player aced	0.19	0.17	0.1	0.17	0.13	0.07	0.1	0.07	0.0
0	0.15	0.13	0.13	0.13	0.1	0.09	0.13	0.09	0.05

Overview	Scalar implicature	Grammar-driven models	Our model	Experiment	Model assessment	Conclusion	Appendix
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- **1** $\mathcal{R}_c(\text{Player A}) = \{ [\text{Player A}], [only Player A] \}$
- 2 $\mathcal{R}_c(\text{Player B}) = \{ [\text{Player B}], [\text{only Player B}] \}$
- 3 R_c(some) = {[[some]], [[some and not all]]}
- **5** $\mathcal{R}_c(\text{scored}) = \{ [[\text{scored}]], [[\text{scored and didn't ace}]] \}$

	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.45					
Player A aced							0.42		
Player B scored		0.45							
Player B aced			0.42						
some player scored		0.25		0.25					
some player aced			0.24				0.24		
every player scored					0.61				
every player aced									1.0
no player scored	0.61								
no player aced	0.19								
0	0.15								

Overview 9	Scalar implicature	Grammar-driven models	Our model	Experiment	Model assessment	Conclusion	Appendix
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- **1** $\mathcal{R}_c(\text{Player A}) = \{ [\text{Player A}], [only Player A] \}$
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Player A aced							0.42		
Player B scored		0.45							
Player B aced			0.42						
some player scored		0.25		0.25					
some player aced			0.24				0.24		
every player scored					0.61				
every player aced									1.0
no player scored	0.61								
no player aced	0.19								
0	0.15								

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Player A aced							0.42		
Player B scored		0.45							
Player B aced			0.42						
some player scored		0.25		0.25					
some player aced			0.24				0.24		
every player scored					0.61				
every player aced									1.0
no player scored	0.61								
no player aced	0.19								
0	0.15								

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	NN	NS	NA	SN	SS	SA	AN	AS	AA
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Player A aced							0.42		
Player B scored		0.45							
Player B aced			0.42						
some player scored		0.25		0.25					
some player aced			0.24				0.24		
every player scored					0.61				
every player aced									1.0
no player scored	0.61								
no player aced	0.19								
0	0.15								

Overview	Scalar implicature	Grammar-driven models	Our model	Experiment	Model assessment	Conclusion	Appendix
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	NN	NS	NA	SN	SS	SA	AN	AS	AA
Player A scored				0.45					
Player A aced							0.42		
Player B scored		0.45							
Player B aced			0.42						
some player scored		0.25		0.25					
some player aced			0.24				0.24		
every player scored					0.61				
every player aced									1.0
no player scored	0.61								
no player aced	0.19								
0	0.15								

Experiment: scalars under quantifiers

Scalar implicature

- 2 Grammar-driven models of implicature
- 3 The compositional lexical uncertainty model
- 4 Experiment: scalars under quantifiers
- 6 Model assessment

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Experiment display



Overview	Scalar implicature	Grammar-driven models	Our model	Experiment	Model assessment	Conclusion	Appendix
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Experiment display



Other experiment details

- 800 participants recruited via Mechanical Turk (no participants or responses excluded)
- Between-subjects design
- 3 training items; 23 fillers; 9 target sentences:

Every		all)
Exactly one	player hit	none	of his shots.
No		some)

- Worlds: {NNN, NNS, NNA, NSS, NSA, NAA, SSS, SSA, SAA, AAA}
- Average 80 responses per target-world pair
- Visual display of worlds and jersey colors randomized

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Results



Percentage True responses

Appendix



Percentage True responses





Percentage True responses

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Model assessment

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Set-up

1 $D = \{a, b, c\}$

2 $W = \{NNN, NNS, NNA, NSS, NSA, NAA, SSS, SSA, SAA, AAA\}$

$\mathbf{3} \ \mathbf{M} = \\ \left\{ Q(player)(hit(S(shot))) : \begin{array}{l} Q \in \{exactly \ one, \ every, \ no\} \\ S \in \{every, \ no, \ some\} \end{array} \right\} \cup \{\mathbf{0}\}$

④
$$C(\mathbf{0}) = 5$$
; $C(m) = 0$ for all $m \in M - \{\mathbf{0}\}$

6 Flat state prior

6 Flat lexicon prior

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Models

- Literal semantics: the predicted values are the output of the literal listener l₀
- Pixed-lexicon pragmatics: the predicted values are the output of L considering only one lexicon
- **Our Constrained refinement**: the inferences of the uncertainty listener *L* with the largest space of refinements
- A Neo-Gricean refinement: as in 'Unconstrained refinement', but with just neo-Gricean refinements

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Comparisons with humans



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Comparisons with humans



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Overall assessment

	Pearson	Spearman	MSE	
Literal semantics	.938 (.926947)	.762 (.754770)	.0065 (.00570075)	
Fixed-lexicon pragmatics	.924 (.911932)	.757 (.749–.766)	.0079 (.00720090)	
Unconstrained uncertainty	.945 (.936950)	.794 (.767820)	.0038 (.00350044)	
Neo-Gricean uncertainty	.959 (.950962)	.809 (.808820)	.0034 (.00310040)	

Table: Overall assessment with 95% confidence intervals obtained via non-parametric bootstrap over subjects.

Results on crucial items

	'everysome'			'ex	'exactly onesome'			'no… some'		
	Ρ	S	MSE	Р	S	MSE	Р	S	MSE	
Literal	.99	.86	.0002	.80	.70	.0180	.88	.52	.0346	
Fixed-lexicon	.93	.85	.0027	.80	.70	.0179	.88	.52	.0346	
Unconstrained	.88	.84	.0043	.98	.94	.0007	.76	.57	.0097	
Neo-Gricean	.82	.88	.0087	.94	.87	.0036	.93	.89	.0028	

Table: Assessment of crucial items. 'P' = 'Pearson'; 'S' = 'Spearman'.



Conclusion

- A synthesis of Gricean and grammar-driven approaches in a single formal, quantitative model.
- Key components: lexical uncertainty and recursive modeling of speaker and listener agents.
- Next steps: experiments with different sentences, and with different notions of refinement.
- Code and data available to facilitate such investigations: https://github.com/cgpotts/pypragmods



Conclusion

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- Key components: lexical uncertainty and recursive modeling of speaker and listener agents.
- Next steps: experiments with different sentences, and with different notions of refinement.
- Code and data available to facilitate such investigations: https://github.com/cgpotts/pypragmods

Thanks!



Binary and Likert response experiments

Binary



Likert





Binary and Likert response experiments





Likert





Binary and Likert response experiments





Likert



Overview	Scalar implicature	Grammar-driven models	Our model	Experiment	Model assessment	Conclusion	Appendix
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Model assessment

	Pearson	Spearman	MSE	
Literal semantics	.938 (.926947)	.762 (.754–.770)	.0065 (.00570075)	
Fixed-lexicon pragmatics	.924 (.911932)	.757 (.749–.766)	.0079 (.00720090)	
Unconstrained uncertainty	.945 (.936950)	.794 (.767–.820)	.0038 (.00350044)	
Neo-Gricean uncertainty	.959 (.950962)	.809 (.808820)	.0034 (.00310040)	

Table: Binary

	Pearson	Spearman	MSE
Literal semantics	.935 (.910947)	.756 (.742764)	.0079 (.00650099)
Fixed-lexicon pragmatics	.920 (.894932)	.751 (.736759)	.0094 (.00800114)
Unconstrained uncertainty	.929 (.905938)	.794 (.765815)	.0052 (.00450067)
Neo-Gricean uncertainty	.950 (.927956)	.805 (.795812)	.0046 (.00380062)

Table: Likert

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Parameter exploration

			C(0)	λ	k
Literal semantics	Pearson Spearman MSE	.94 .76 .0065			
Fixed lexicon pragmatics	Pearson	.93	1	.1	1
	Spearman	.76	0	.2	1
	MSE	.0069	1	.1	1
Unconstrained uncertainty	Pearson	.97	1	.1	1
	Spearman	.80	1	.1	1
	MSE	.0022	1	.1	1
Neo-Gricean uncertainty	Pearson	.98	1	.1	1
	Spearman	.81	1	.2	1
	MSE	.0018	1	.1	1

Table: Best models found in hyper-parameter exploration, as assessed against the binary-response experiment. Searched λ : [0.1,5] in increments of .1; L_k for $k \in \{1, 2, 3, 4, 5, 6\}$; $C(\mathbf{0}) \in \{0, 1, 2, 3, 4, 5, 6\}$. The literal listener is not affected by any of the parameters explored.

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Parameter exploration



Figure: L_1 , using parameters in the range that seem to be nearly optimal for all of these models: $\lambda = 0.1$; $C(\mathbf{0}) = 1$.

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Parameter exploration



Figure: L_1 , using the parameters we originally chose: $\lambda = 1$; $C(\mathbf{0}) = 5$.