

The different flavors of the Iterated Best Response Model of game theoretic pragmatics

Gerhard Jäger

gerhard.jaeger@uni-tuebingen.de

April 28, 2014

ZAS Berlin

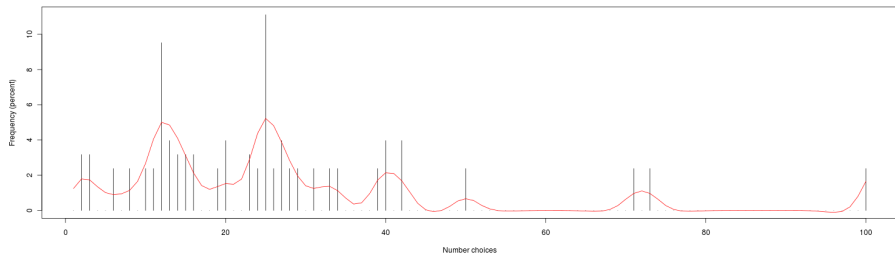
EBERHARD KARLS
UNIVERSITÄT
TÜBINGEN



The Beauty Contest

- each participant has to write down a number between 0 and 100
- all numbers are collected
- the person whose guess is closest to $2/3$ of the arithmetic mean of all numbers submitted is the winner

The Beauty Contest



(data from Camerer 2003, *Behavioral Game Theory*)

Signaling games

- sequential game:
 - 1 **nature** chooses a world w
 - out of a pool of possible worlds W
 - according to a certain probability distribution p^*
 - 2 nature shows w to sender **S**
 - 3 S chooses a message m out of a set of possible signals M
 - 4 S transmits m to the receiver **R**
 - 5 R chooses an action a , based on the sent message.
- Both S and R have preferences regarding R's action, depending on w .
- S might also have preferences regarding the choice of m (to minimize signaling costs).

Tea or coffee?

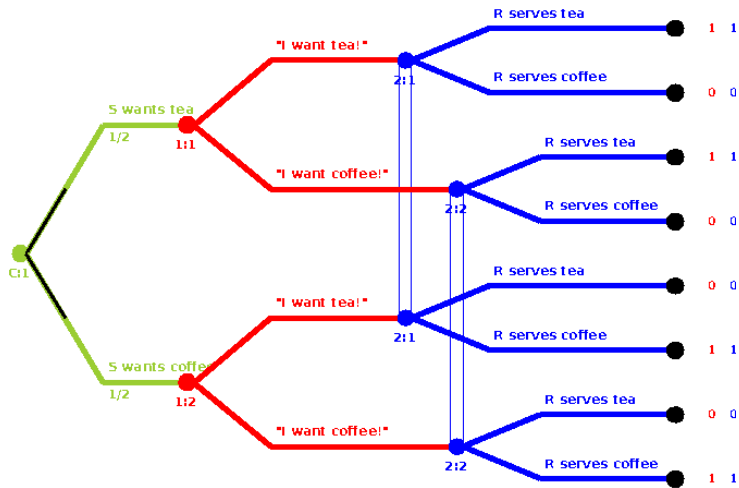
An example

- Sally either prefers tea (w_1) or coffee (w_2), with $p^*(w_1) = p^*(w_2) = 1/2$.
- Robin either serves tea (a_1) or coffee (a_2).
- Sally can send either of two messages:
 - m_1 : *I prefer tea.*
 - m_2 : *I prefer coffee.*
- Both messages are costless.

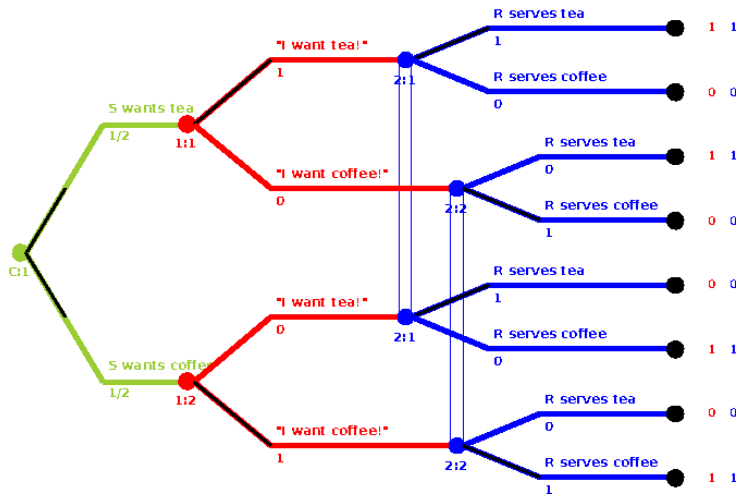
	a_1	a_2
w_1	1, 1	0, 0
w_2	0, 0	1, 1

Table: utility matrix

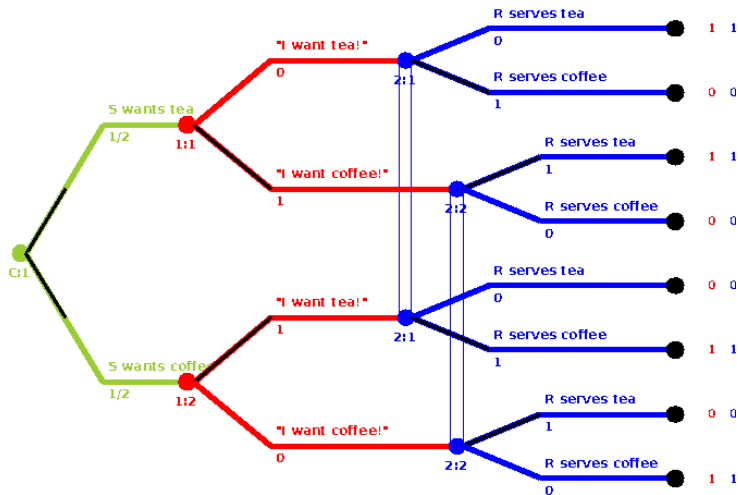
Extensive form



Extensive form



Extensive form



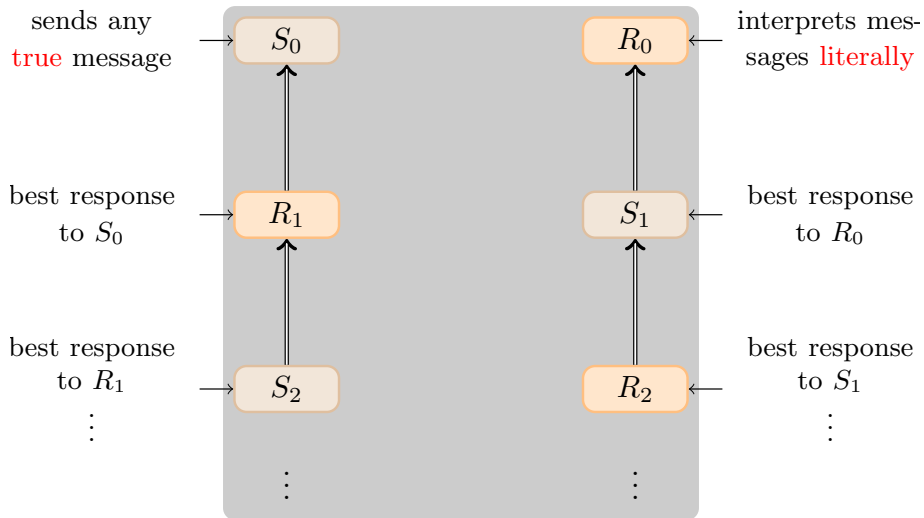
A coordination problem

- two strict Nash equilibria
 - S always says the truth and R always believes her.
 - S always says the opposite of the truth and R interprets everything ironically.
- Both equilibria are equally rational.
- Still, first equilibrium is more reasonable because it employs exogenous meanings of messages for equilibrium selection.
- Criterion for equilibrium selection:

As a default, S and R use/interpret signals according to their literal meaning. They only deviate from this if there self-interest dictates them to do so.

- What exactly does this mean?

The Iterated Best Response sequence



Interpretation games

- How does this relate to linguistic examples?
- There is a quasi-algorithmic procedure (due to Franke 2009) how to construct a game from an example sentence.

What is given?

- example sentence
- set of expression alternatives
- jointly form set of messages
- question under discussion QUD
- set of complete answers to QUD is the set of possible worlds

What do we need?

- interpretation function $\| \cdot \|$
- prior probability distribution p^*
- set of actions
- utility functions

Interpretation games

QUD

- often QUD is not given explicitly
- procedure to construct QUD from expression m and its alternatives $ALT(m)$:
 - Let ct be the context of utterances, i.e. the maximal set of statements that is common knowledge between Sally and Robin.
 - any subset w of $ALT(m) \cup \{\neg m' \mid m' \in ALT(m)\}$ is a possible world iff
 - w and ct are consistent, i.e. $w \cup ct \not\vdash \perp$
 - for any set $X : w \subset X \subseteq ALT(m) \cup \{\neg m' \mid m' \in ALT(m)\}$, $ct \cup X$ is inconsistent

Interpretation games

Game construction

- interpretation function:

$$\|m'\| = \{w \mid w \vdash m\}$$

- p^* is uniform distribution over W
- justified by principle of insufficient reason
- set of actions is W
- intuitive idea: Robin's task is to figure out which world Sally is in
- utility functions:

$$u_{s/r}(w, a) = \begin{cases} 1 & \text{iff } w = a \\ 0 & \text{else} \end{cases}$$

- both players want Robin to succeed

Quantity implicatures

- (1) a. Who came to the party?
b. SOME: Some boys came to the party.
c. ALL: All boys came to the party.

Game construction

- $ct = \emptyset$
- $W = \{w_{\exists-\forall}, w_{\forall}\}$
- $w_{\exists-\forall} = \{\text{SOME}\}, w_{\forall} = \{\text{SOME}, \text{ALL}\}$
- $p^* = (1/2, 1/2)$

- interpretation function:

$$\|\text{SOME}\| = \{w_{\exists-\forall}, w_{\forall}\}$$

$$\|\text{ALL}\| = \{w_{\forall}\}$$

- utilities:

	$a_{\exists-\forall}$	a_{\forall}
$w_{\exists-\forall}$	1, 1	0, 0
w_{\forall}	0, 0	1, 1

The recursion base

- IBR style reasing requires a default assumption that
 - reflects the literal meanings of the messages involved
 - is not the result of higher-order rational reasoning

Honest sender S_0

- in each type, Sally utters a true message
- does not involve a model of how Robin interprets messages

Credulous receiver R_0

- whatever messages Robin observes, he takes it to be true
- does not involve a model of how Sally uses messages

- both S_0 and R_0 are *sets* of (pure) strategies that need not be singleton (due to synonymy/homonymy)

Recursion bases for some-all game

- literal meanings:

	SOME	ALL
$w_{\exists-\forall}$	1	0
w_{\forall}	1	1

- default strategies

S_0	SOME	ALL	R_0	$w_{\exists-\forall}$	w_{\forall}
$w_{\exists-\forall}$	1	0	SOME	β	$1 - \beta$
w_{\forall}	α	$1 - \alpha$	ALL	0	1

$$\alpha, \beta \in \{0, 1\}$$

What does “best response” mean?

- suppose a player knows that their opponent some strategy out of a set A of strategies
- no *a priori* information about how the opponent selects elements of A
- two approaches to model rational behavior:
 - Franke (2009, 2011): a rational response to A is any strategy that maximizes utility against the **uniform** distribution over A
 - Jger (2013): a rational response to A is any strategy that maximizes utility against **some proper** distribution over A
- Terminology:
 - Franke-style rationality is called **best response** (as a technical term)
 - Jäger-style rationality (borrowed from Pearce 1984) is called **cautious response**

Some-all game again

S_0	SOME	ALL
$w_{\exists-\forall}$	1	0
w_{\forall}	α	$1 - \alpha$

R_0	$w_{\exists-\forall}$	w_{\forall}
SOME	β	$1 - \beta$
ALL	0	1

- **IBR** (Iterated Best Response *sensu strictu*; Franke-style rationality):
 $\alpha = \beta = 1/2$
- **ICR** (Iterated Cautious Response; Jäger-style rationality):
 $\alpha, \beta \in (0, 1)$
- in the running example, predictions coincide:

S_1	SOME	ALL
$w_{\exists-\forall}$	1	0
w_{\forall}	0	1

R_1	$w_{\exists-\forall}$	w_{\forall}
SOME	1	0
ALL	0	1

- all four theory variants (S-based vs. R-based, IBR vs. ICR) correctly predict quantity implicature

Biased some-all game

- Rothschild (2013): IBR makes wrong predictions when prior probabilities are non-uniform
- for instance: $p^*(w_{\exists-\forall}) = 1/10, p^*(w_{\forall}) = 9/10$

S_0	p^*	SOME	ALL
$w_{\exists-\forall}$	1/10	1	0
w_{\forall}	9/10	α	$1 - \alpha$

R_0	$w_{\exists-\forall}$	w_{\forall}
SOME	β	$1 - \beta$
ALL	0	1

Biased some-all game; IBR

S_0	p^*	SOME	ALL
$w_{\exists \rightarrow \forall}$	1/10	1	0
w_{\forall}	9/10	1/2	1/2

R_1	$w_{\exists \rightarrow \forall}$	w_{\forall}
SOME	0	1
ALL	0	1

S_2	p^*	SOME	ALL
$w_{\exists \rightarrow \forall}$	1/10	1	0
w_{\forall}	9/10	1/2	1/2

$$R_3 = R_1$$

R_0	$w_{\exists \rightarrow \forall}$	w_{\forall}
SOME	1/2	1/2
ALL	0	1

S_1	p^*	SOME	ALL
$w_{\exists \rightarrow \forall}$	1/10	1	0
w_{\forall}	9/10	0	1

R_2	$w_{\exists \rightarrow \forall}$	w_{\forall}
SOME	1	0
ALL	0	1

$$S_3 = S_1$$

Biased some-all game; ICR

S_0	p^*	SOME	ALL
$w_{\exists-\forall}$	$1/10$	1	0
w_{\forall}	$9/10$	α	$1 - \alpha$

R_1	$w_{\exists-\forall}$	w_{\forall}
SOME	β	$1 - \beta$
ALL	0	1

S_2	p^*	SOME	ALL
$w_{\exists-\forall}$	$1/10$	1	0
w_{\forall}	$9/10$	0	1

R_3	$w_{\exists-\forall}$	w_{\forall}
SOME	1	0
ALL	0	1

$$S_4 = S_2$$

R_0	$w_{\exists-\forall}$	w_{\forall}
SOME	γ	$1 - \gamma$
ALL	0	1

S_1	p^*	SOME	ALL
$w_{\exists-\forall}$	$1/10$	1	0
w_{\forall}	$9/10$	0	1

R_2	$w_{\exists-\forall}$	w_{\forall}
SOME	1	0
ALL	0	1

$$S_3 = S_1$$

Biased some-all game; summary

- prior preference for w_{\forall} is a problem for S-based IBR
- the other three theory variants get the right result

Some but not all

- extension of the previous game with additional message
(1) d. SBNA: Some but not all boys came to the party
- SBNA is only true in $w_{\exists \rightarrow \forall}$ and incurs a nominal cost ($1/100$, for concreteness sake)

	p^*	SOME	ALL	SBNA
$w_{\exists \rightarrow \forall}$	$1/2$	1	0	1
w_{\forall}	$1/2$	1	1	0
c		0	0	$1/100$

SBNA-game; IBR

S_0	SOME	ALL	SBNA
$w_{\exists-\forall}$	1	0	0
w_{\forall}	1/2	1/2	0

R_0	$w_{\exists-\forall}$	w_{\forall}	c
SOME	1/2	1/2	0
ALL	0	1	0
SBNA	1	0	1/100

R_1	$w_{\exists-\forall}$	w_{\forall}	c
SOME	1	0	0
ALL	0	1	0
SBNA	1	0	1/100

S_2	SOME	ALL	SBNA
$w_{\exists-\forall}$	0	0	1
w_{\forall}	0	1	0

S_3	SOME	ALL	SBNA
$w_{\exists-\forall}$	1	0	0
w_{\forall}	0	1	0

R_3	$w_{\exists-\forall}$	w_{\forall}	c
SOME	1/2	1/2	0
ALL	0	1	0
SBNA	1	0	1/100

SBNA-game; ICR

S_0	SOME	ALL	SBNA
$w_{\exists-\forall}$	1	0	0
w_{\forall}	α	$1 - \alpha$	0

R_1	$w_{\exists-\forall}$	w_{\forall}	c
SOME	1	0	0
ALL	0	1	0
SBNA	1	0	$1/100$

S_2	SOME	ALL	SBNA
$w_{\exists-\forall}$	1	0	0
w_{\forall}	0	1	0

R_0	$w_{\exists-\forall}$	w_{\forall}	c
SOME	β_1	$1 - \beta_1$	0
ALL	0	1	0
SBNA	1	0	$1/100$

S_1	SOME	ALL	SBNA
$w_{\exists-\forall}$	β_2	0	$1 - \beta_2$
w_{\forall}	0	1	0

$$R_2 = R_1$$

SBNA-game; summary

- IBR/R: no scalar implicature for SOME
- IBR/S; ICR: same scalar implicature as in simple some-all game
- simple-minded responses to surprise messages
- in line with general spirit of IBR vs. ICR:
 - IBR: surprise message induces belief in uniform distribution over the set of worlds where the message is true
 - ICR: surprise message induces belief in some proper probability distribution over the set of worlds where the message is true
- Franke (2009, 2011) uses more sophisticated form of forward induction; correctly predicts scalar implicature for both IBR sequences

Lifted SBNA-game with competence assumption

- Pavan (2013) notices a problem with Jäger's (2011) version of ICR
- main difference to Jäger (2013): improper probability distributions are admitted
- Pavan's example:
 - SBNA-game
 - types are *information states*, i.e. sets of possible worlds:

$$t_{\exists \neg \forall} = \{w_{\exists \neg \forall}\}$$

$$t_{\forall} = \{w_{\forall}\}$$

$$t_{\exists} = \{w_{\exists \neg \forall}, w_{\forall}\}$$

- *competence assumption*: $p^*(t_{\exists}) = 0$

Lifted SBNA-game with competence assumption

	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	$1/2$	1	0	1
t_{\forall}	$1/2$	1	1	0
t_{\exists}	0	1	0	0
c		0	0	$1/100$

Lifted SBNA-game with competence assumption; IBR

S_0	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	1/2	1	0	0
t_{\forall}	1/2	1/2	1/2	0
t_{\exists}	0	1	0	0

R_0	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	1/2	1/2	0	0
ALL	0	1	0	0
SBNA	1	0	0	1/100

R_1	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	1	0	0	0
ALL	0	1	0	0
SBNA	1	0	0	1/100

S_1	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	1/2	0	0	1
t_{\forall}	1/2	0	1	0
t_{\exists}	0	1	0	0

S_2	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	1/2	1	0	0
t_{\forall}	1/2	0	1	0
t_{\exists}	0	1	0	0

R_2	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	1/3	1/3	1/3	0
ALL	0	1	0	0
SBNA	1	0	0	1/100

$$R_3 = R_1$$

$$S_3 = S_1$$

Lifted SBNA-game with competence assumption; ICR

S_0	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	$1/2$	1	0	0
t_{\forall}	$1/2$	α_1	$1 - \alpha_1$	0
t_{\exists}	0	1	0	0

R_0	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	β_1	$1 - \beta_1$	0	0
ALL	0	1	0	0
SBNA	1	0	0	$1/100$

R_1	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	1	0	0	0
ALL	0	1	0	0
SBNA	1	0	0	$1/100$

S_1	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	$1/2$	α_2	0	$1 - \alpha_2$
t_{\forall}	$1/2$	0	1	0
t_{\exists}	0	1	0	0

S_2	p^*	SOME	ALL	SBNA
$t_{\exists \rightarrow \forall}$	$1/2$	1	0	0
t_{\forall}	$1/2$	0	1	0
t_{\exists}	0	1	0	0

R_2	$t_{\exists \rightarrow \forall}$	t_{\forall}	t_{\exists}	c
SOME	1	0	0	0
ALL	0	1	0	0
SBNA	1	0	0	$1/100$

$$R_3 = R_1$$

$$S_3 = S_2$$

Lifted SBNA-game with competence assumption; summary

- lifted SBNA-game + competence behaves as simple sbna-game
- scalar implicatures is predicted by IBR/S and ICR, but not by IBR/R
- latter problem can be fixed by means of sophisticated forward induction

Horn-game

	p^*	f_1	f_2
w_1	$2/3$	1	1
w_2	$1/3$	1	1
c		0	$1/100$

Horn-game: IBR

S_0	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	1	0

R_0	w_1	w_2	c
f_1	1	0	0
f_2	1	0	$1/100$

R_1	w_1	w_2	c
f_1	1	0	0
f_2	$1/2$	$1/2$	$1/100$

S_1	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	1	0

S_2	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	0	1

R_2	w_1	w_2	c
f_1	1	0	0
f_2	$1/2$	$1/2$	$1/100$

R_3	w_1	w_2	c
f_1	1	0	0
f_2	0	1	$1/100$

S_3	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	0	1

Horn-game: ICR

S_0	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	1	0

R_0	w_1	w_2	c
f_1	1	0	0
f_2	1	0	$1/100$

R_1	w_1	w_2	c
f_1	1	0	0
f_2	α	$1 - \alpha$	$1/100$

S_1	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	1	0

S_2	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	0	1

R_2	w_1	w_2	c
f_1	1	0	0
f_2	β	$1 - \beta$	$1/100$

R_3	w_1	w_2	c
f_1	1	0	0
f_2	0	1	$1/100$

S_3	p^*	f_1	f_2
w_1	$2/3$	1	0
w_2	$1/3$	0	1

Horn game: summary

- Horn's division of pragmatic labor (M-implicature) is predicted both by IBR and ICR

Generalized Horn game

p^*	f_1	f_2	f_3
w_1 3/6	1	1	1
w_2 2/6	1	1	1
w_2 1/6	1	1	1
c	0	1/100	2/100

Generalized Horn: IBR

S_0	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	1	0	0
w_2	$1/6$	1	0	0

R_0	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	1	0	0	$1/100$
f_2	1	0	0	$2/100$

R_1	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	$1/3$	$1/3$	$1/3$	$1/100$
f_2	$1/3$	$1/3$	$1/3$	$2/100$

S_1	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	1	0	0
w_2	$1/6$	1	0	0

S_2	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	1	0
w_2	$1/6$	0	1	0

R_2	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	$1/3$	$1/3$	$1/3$	$1/100$
f_2	$1/3$	$1/3$	$1/3$	$2/100$

Generalized Horn: IBR (cont.)

S_2	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	1	0
w_2	$1/6$	0	1	0

R_2	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	$1/3$	$1/3$	$1/3$	$1/100$
f_2	$1/3$	$1/3$	$1/3$	$2/100$

R_3	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	0	1	0	$1/100$
f_2	$1/3$	$1/3$	$1/3$	$2/100$

S_3	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	1	0
w_2	$1/6$	0	1	0

S_4	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	1	0
w_2	$1/6$	0	0	1

$$R_4 = R_3$$

Generalized Horn: ICR

S_0	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	1	0	0
w_2	$1/6$	1	0	0

R_1	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	x_1	y_1	$1 - x_1 - y_1$	$1/100$
f_2	x_2	y_2	$1 - x_2 - y_2$	$2/100$

S_2	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	x	$1 - x$
w_2	$1/6$	0	y	$1 - y$

R_0	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	1	0	0	$1/100$
f_2	1	0	0	$2/100$

$$S_1 = S_0$$

$$R_2 = R_1$$

Generalized Horn: ICR (cont.)

S_2	p^*	f_1	f_2	f_3
w_1	$3/6$	1	0	0
w_2	$2/6$	0	x	$1 - x$
w_2	$1/6$	0	y	$1 - y$

$$R_2 = R_1$$

R_3	w_1	w_2	w_3	c
f_1	1	0	0	0
f_2	0	x	$1 - x$	$1/100$
f_2	0	y	$1 - y$	$2/100$

$$S_3 = S_2$$

$$S_4 = S_3$$

$$R_4 = R_3$$

Generalized Horn: summary

- IBR predicts isomorphic map between types and signals
- ICR predicts association between maximum-likelihood type and cheapest message; any mapping between lower-likelihood types and higher-cost signals are admitted in fixed point
- IBR result is aesthetically more pleasing, but it is unclear whether there are empirical arguments either way

Free-choice inference

	$m_{\diamond A}$	$m_{\diamond B}$	$m_{\diamond(A \vee B)}$	$m_{\diamond(A \wedge B)}$
$t_{\diamond A}$	1	0	1	0
$t_{\diamond B}$	0	1	1	0
$t_{\diamond AB}$	1	1	1	1
$t_{\diamond A B}$	1	1	1	0
c	0	0	1/100	1/100

Free-choice game: IBR

S_0	$m_{\diamond A}$	$m_{\diamond B}$	$m_{\diamond(A \vee B)}$	$m_{\diamond(A \wedge B)}$
$t_{\diamond A}$	1	0	0	0
$t_{\diamond B}$	0	1	0	0
$t_{\diamond AB}$	1/2	1/2	0	0
$t_{\diamond A B}$	1/2	1/2	0	0

R_1	$t_{\diamond A}$	$t_{\diamond B}$	$t_{\diamond AB}$	$t_{\diamond A B}$	c
$m_{\diamond A}$	1	0	0	0	0
$m_{\diamond B}$	0	1	0	0	0
$m_{\diamond(A \vee B)}$	1/4	1/4	1/4	1/4	1/100
$m_{\diamond(A \wedge B)}$	0	0	1	0	1/100

S_2	$m_{\diamond A}$	$m_{\diamond B}$	$m_{\diamond(A \vee B)}$	$m_{\diamond(A \wedge B)}$
$t_{\diamond A}$	1	0	0	0
$t_{\diamond B}$	0	1	0	0
$t_{\diamond AB}$	0	0	0	1
$t_{\diamond A B}$	0	0	1	0

R_0	$t_{\diamond A}$	$t_{\diamond B}$	$t_{\diamond AB}$	$t_{\diamond A B}$	c
$m_{\diamond A}$	1/3	0	1/3	1/3	0
$m_{\diamond B}$	0	1/3	1/3	1/3	0
$m_{\diamond(A \vee B)}$	1/4	1/4	1/4	1/4	1/100
$m_{\diamond(A \wedge B)}$	0	0	1	0	1/100

S_1	$m_{\diamond A}$	$m_{\diamond B}$	$m_{\diamond(A \vee B)}$	$m_{\diamond(A \wedge B)}$
$t_{\diamond A}$	1	0	0	0
$t_{\diamond B}$	0	1	0	0
$t_{\diamond AB}$	0	0	0	1
$t_{\diamond A B}$	1/2	1/2	0	0

$$R_2 = R_1$$

Free-choice game: ICR

S_0	$m_{\circ A}$	$m_{\circ B}$	$m_{\circ(A \vee B)}$	$m_{\circ(A \wedge B)}$
$t_{\circ A}$	1	0	0	0
$t_{\circ B}$	0	1	0	0
$t_{\circ AB}$	x	$1 - x$	0	0
$t_{\circ A B}$	y	$1 - y$	0	0

R_1	$t_{\circ A}$	$t_{\circ B}$	$t_{\circ AB}$	$t_{\circ A B}$	c
$m_{\circ A}$	1	0	0	0	0
$m_{\circ B}$	0	1	0	0	0
$m_{\circ A \vee B}$	x	y	z	$1 - x - y - z$	$1/100$
$m_{\circ A \wedge B}$	0	0	1	0	$1/100$

S_2	$m_{\circ A}$	$m_{\circ B}$	$m_{\circ(A \vee B)}$	$m_{\circ(A \wedge B)}$
$t_{\circ A}$	1	0	0	0
$t_{\circ B}$	0	1	0	0
$t_{\circ AB}$	0	0	0	1
$t_{\circ A B}$	0	0	1	0

R_0	$t_{\circ A}$	$t_{\circ B}$	$t_{\circ AB}$	$t_{\circ A B}$	c
$m_{\circ A}$	x_1	0	y_1	$1 - x_1 - y_1$	0
$m_{\circ B}$	0	x_2	y_2	$1 - x_2 - y_2$	0
$m_{\circ A \vee B}$	x_3	y_3	z_3	$1 - x_3 - y_3 - z_3$	$1/100$
$m_{\circ A \wedge B}$	0	0	1	0	$1/100$

S_1	$m_{\circ A}$	$m_{\circ B}$	$m_{\circ(A \vee B)}$	$m_{\circ(A \wedge B)}$
$t_{\circ A}$	x	0	$1 - x$	0
$t_{\circ B}$	0	y	$1 - y$	0
$t_{\circ AB}$	0	0	0	1
$t_{\circ A B}$	x_2	y_2	$1 - x_2 - y_2$	0

R_2	$t_{\circ A}$	$t_{\circ B}$	$t_{\circ AB}$	$t_{\circ A B}$	c
$m_{\circ A}$	x	0	0	$1 - x$	0
$m_{\circ B}$	0	y	0	$1 - y$	0
$m_{\circ A \vee B}$	x_3	y_3	0	$1 - x_3 - y_3$	$1/100$
$m_{\circ A \wedge B}$	0	0	1	0	$1/100$

Free-choice game: ICR (cont.)

S_2	$m_{\circ A}$	$m_{\circ B}$	$m_{\circ(A \vee B)}$	$m_{\circ(A \wedge B)}$
$t_{\circ A}$	1	0	0	0
$t_{\circ B}$	0	1	0	0
$t_{\circ AB}$	0	0	0	1
$t_{\circ A B}$	0	0	1	0

R_3	$t_{\circ A}$	$t_{\circ B}$	$t_{\circ AB}$	$t_{\circ A B}$	c
$m_{\circ A}$	1	0	0	0	0
$m_{\circ B}$	0	1	0	0	0
$m_{\circ A \vee B}$	0	0	0	1	1/100
$m_{\circ A \wedge B}$	0	0	1	0	1/100

$$S_4 = S_2$$

R_2	$t_{\circ A}$	$t_{\circ B}$	$t_{\circ AB}$	$t_{\circ A B}$	c
$m_{\circ A}$	x	0	0	$1 - x$	0
$m_{\circ B}$	0	y	0	$1 - y$	0
$m_{\circ A \vee B}$	x_3	y_3	0	$1 - x_3 - y_3$	1/100
$m_{\circ A \wedge B}$	0	0	1	0	1/100

$$S_3 = S_1$$

Free-choice game: summary

- IBR and ICR/S predict free-choice inference
- ICR/R fails

Summary IBR vs. ICR

	IBR/S	IBR/R	ICR/S	ICR/R
<i>biased some-all</i>	X	✓	✓	✓
<i>sbna</i>	✓	X	✓	✓
<i>lifted sbna+competence</i>	✓	X	✓	✓
<i>Horn</i>	✓	✓	✓	✓
<i>generalized Horn</i>	✓	✓	X	X
<i>free choice</i>	✓	✓	✓	X

Summary IBR vs. ICR

- problems for vanilla IBR/R are solved in sophisticated version of Franke (2009, 2011)
- problems for ICR not unsurmountable (more sophisticated forward induction protocol; preference ordering over types in R_0)
- IBR: computationally simple and straightforward
- ICR: difficult (often virtually impossible) to compute solution manually for larger games
- automated solution requires Linear Programming
- while LP is generally fast, number of LP constraints grow exponentially with number of types/message
- ICR is computationally unfeasible already for moderately-sized games — at least in the worst case (as IA and equilibrium-based approaches)
- like IA, ICR is less sensitive to priors than IBR