Level-k Reasoning 多層次思考

Joseph Tao-yi Wang (王道一) Lecture 8, EE-BGT

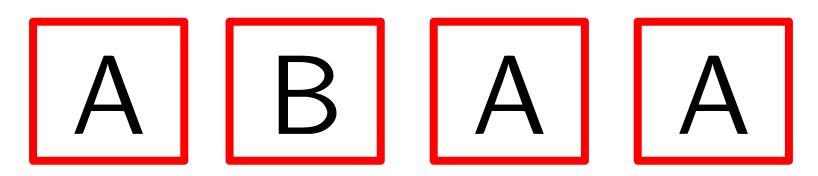
Outline

- Introduction: Initial Deviations from MSE
 - Hide-and-Seek: Crawford & Iriberri (AER07)
 - Initial Joker Effect: Re-asssess O'Neil (1987)
- Simultaneous Dominant Solvable Games
 - Price competition: Capra et al (IER 2002)
 - Traveler's dilemma: Capra et al (AER 1999)
 - -p-BC game: Nagel (AER 95), CHW (AER 98)
- Level-k Theory:
 - Stahl-Wilson (GEB1995), CGCB (ECMA2001)
 - Costa-Gomes & Crawford (AER 2006)

- RTH: Rubinstein & Tversky (1993); Rubinstein,
 Tversky, & Heller (1996); Rubinstein (1998,1999)
- Your opponent has hidden a prize in one of four boxes arranged in a row.
- The boxes are marked as shown below: A, B, A, A.

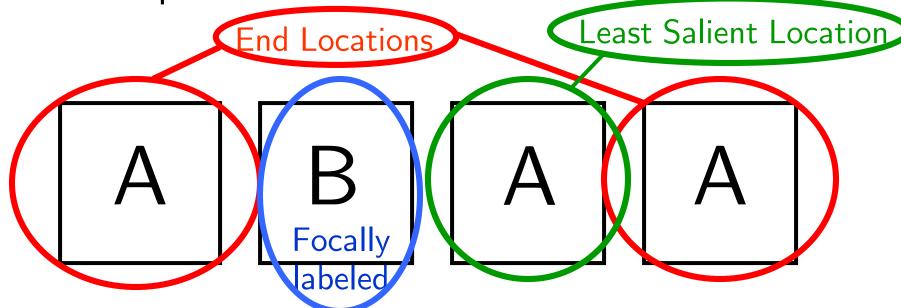


- RTH (Continued):
- Your goal is, of course, to find the prize.
- His goal is that you will not find it.
- You are allowed to open only one box.
- Which box are you going to open?



- Folk Theory: "...in Lake Wobegon, the correct answer is usually 'c'."
 - Garrison Keillor (1997) on multiple-choice tests
- Comment on the poisoning of Ukrainian's presidential candidate (now president):
- "Any government wanting to kill an opponent ...would not try it at a meeting with government officials."
 - Viktor Yushchenko, quoted in Chivers (2004)

- B is distinguished by its label
- The two end A may be inherently salient
- This gives the central A location its own brand of uniqueness as the least salient location

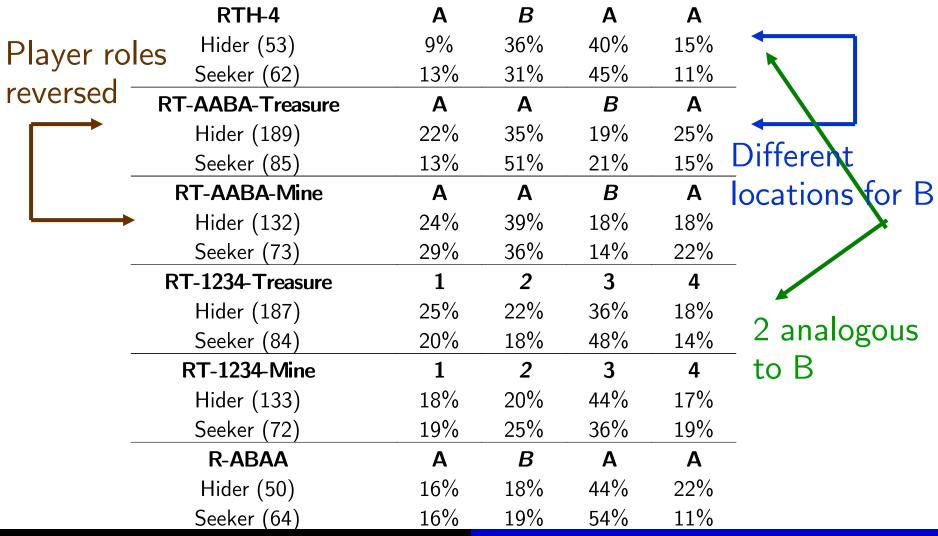


- RTH's game has a unique equilibrium, in which both players randomize uniformly
- Expected payoffs: Hider 3/4, Seeker 1/4

Hider/Seeker	Α	В	Α	Α
Α	0,1	1,0	1,0	1,0
В	1,0	0,1	1,0	1,0
Α	1,0	1,0	0,1	1,0
Α	1,0	1,0	1,0	0,1

- All Treatments in RTH:
- Baseline: ABAA (Treasure Treatment)
- Variants:
 - Left-Right Reverse: AABA
 - Labeling: 1234 (2 is like B, 3 is like central A)
- Mine Treatments
 - Hider hides a mine in 1 location, and Seeker wants to avoid the mine (payoffs reversed)
 - mine hiders = seekers, mine seekers = hiders

Hide-and-Seek Games: Aggregate Results of RTH



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Hide-and-Seek Games: Aggregate Results of RTH

RTH-4	Α	В	Α	Α
Hider (53)	9%	36%	40%	15%
Seeker (62)	13%	31%	45%	11%
RT-AABA-Treasure	Α	Α	В	Α
Hider (189)	22%	35%	19%	25%
Seeker (85)	13%	51%	21%	15%
RT-AABA-Mine	Α	A	В	Α
Hider (132)	24%	39%	18%	18%
Seeker (73)	29%	36%	14%	22%
RT-1234-Treasure	1	2	3	4
RT-1234-Treasure Hider (187)	1 25%	2 22% (3	4 18%
	-			
Hider (187)	25%	22%	36%	18%
Hider (187) Seeker (84)	25% 20%	22% 18%	36% 48%	18% 14%
Hider (187) Seeker (84) RT-1234-Mine	25% 20% 1	22% (18% (36% 48% 3	18% 14% 4
Hider (187) Seeker (84) RT-1234-Mine Hider (133)	25% 20% 1 18%	22% 18% 2 20%	36% 48% 3 44%	18% 14% 4 17%
Hider (187) Seeker (84) RT-1234-Mine Hider (133) Seeker (72)	25% 20% 1 18% 19%	22% 18% 2 20% 25%	36% 48% 3 44% 36%	18% 14% 4 17% 19%

Stylized facts

Hide-and-Seek Games: Aggregate Results of RTH

- Can pool data since no significant differences for Seekers (p=0.48) or Hiders (p=0.16)
 - Chi-square Test across 6 different Treatments

	Α	В	A	Α
Hiders	0.2163	0.2115	0.3654	0.2067
(624)				
Seekers	0.1821	0.2054	0.4589	0.1536
(560)				

Hide-and-Seek Games: Stylized Facts

- Central A (or 3) is most prevalent for both Hiders and Seekers
- Central A is even more prevalent for Seekers (or Hiders in Mine treatments)
 - As a result, Seekers do better than in equilibrium
- Shouldn't Hiders realize that Seekers will be just as tempted to look there?
- RTH: "The finding that both choosers and guessers selected the least salient alternative suggests little or no strategic thinking."

Hide-and-Seek Games: Explaining Stylized Facts

- Can a strategic theory explain this?
- Heterogeneous population with substantial frequencies of L2 and L3 as well as L1 (estimated 19% L1, 32% L2, 24% L3, 25% L4) can reproduce the stylized facts
- More on Level-k later…
 - Let us first see more evidence in DS Games...

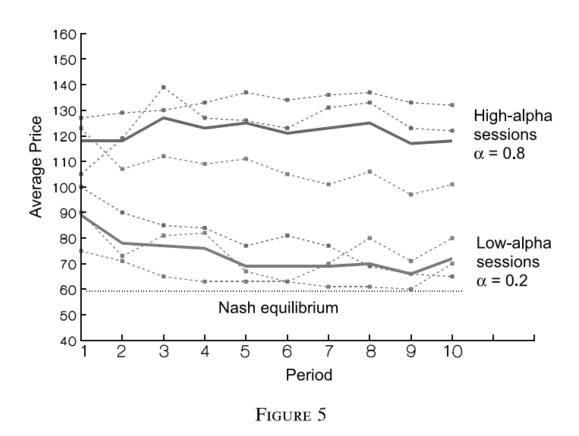
Simultaneous Dominant Solvable Games

- Initial Response vs. Equilibration
- Price Competition
 - Capra, Goeree, Gomez and Holt (IER 2002)
- Traveler's Dilemma
 - Capra, Goeree, Gomez and Holt (AER 1999)
- p -Beauty Contest
 - Nagel (AER 1995)
 - Camerer, Ho, Weigelt (AER 1998)

Price Competition

- Capra, Goeree, Gomez & Holt (IER 2002)
 - Two firms pick prices $p_1 \& p_2$ from \$0.60-\$1.60
 - Both get $(1+\alpha)^* p_1 / 2$ if tied
- But if $p_1 < p_2$:
 - Low-price firm gets $(1 * p_1)$
 - Other firm gets ($\alpha * p_1$)
- α = responsiveness to *best price* (=0.2/0.8)
 - $-\alpha \rightarrow 1$: *Meet-or-release* (low price guarantees)
 - $-\alpha$ <1: Bertrand competition predicts lowest price

Price Competition: Data



AVERAGE PRICES BY SESSION (DASHED LINES) AND TREATMENT (DARK LINE)

Price Competition: Simulation

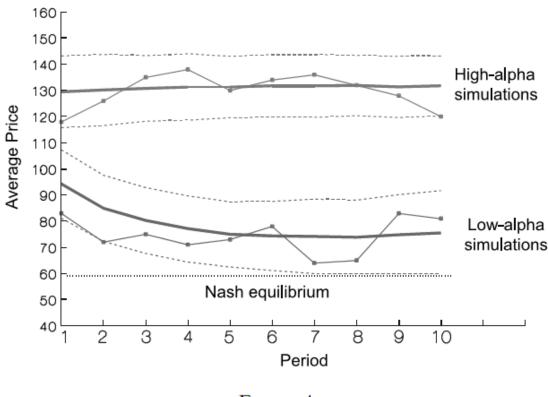


Figure 4

Simulated average prices obtained from 1000 simulations (dark lines) ± 2 standard deviations (dotted lines) and a typical run (lines connecting squares)

Traveler's Dilemma

- Capra, Goeree, Gomez & Holt (AER 1999)
 - Two travelers state claim p_1 and p_2 : 80-200
 - Airline awards both the minimum claim, but
 - reward R to the one who stated the lower claim
 - penalize the other by R
- Unique NE: race to the bottom
 - → lowest claim
 - Like price competition game or p-beauty contest

Traveler's Dilemma: Data

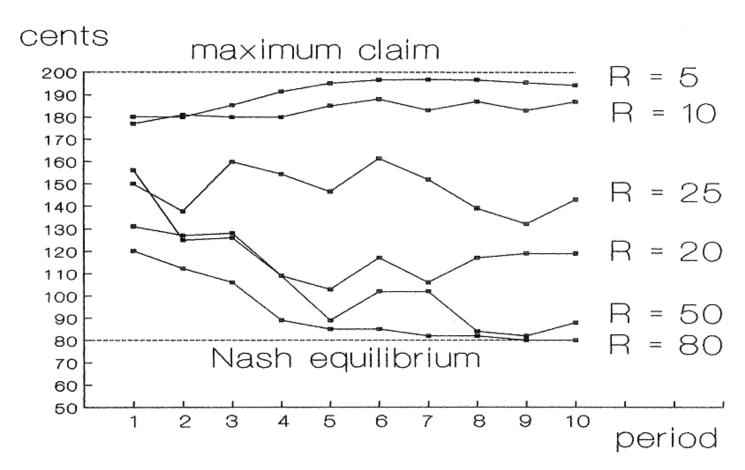
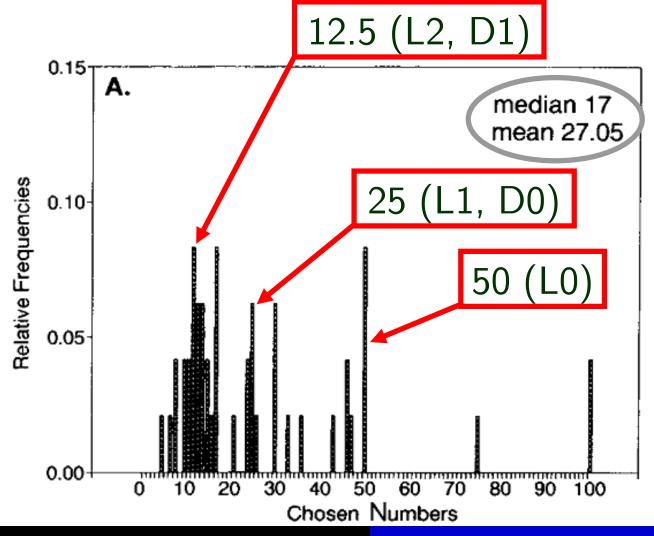


FIGURE 1. DATA FOR PART A FOR VARIOUS VALUES OF THE REWARD/PENALTY PARAMETER

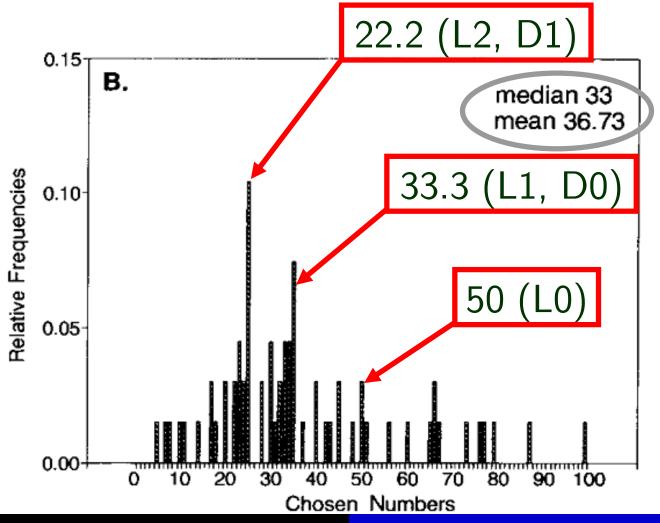
- Each of N players choose x_i from [0,100]
 - 每人選擇0到100之間的數字,希望最接近「所有數字平均乘以p倍」
- Target is p^* (average of x_i)
- Closest x_i wins fixed prize
- (67,100) violates 1st order dominance
 - 選擇67-100的人是選擇(一階的)劣勢策略
- (45, 67] obeys 1 step (not 2) of dominance
 - 選擇45-67的人是選擇除去一階劣勢策略後剩下的(二階)劣勢策略
- 1st Experiment (最早的實驗): Nagel (AER 1995)

Figure 1A of Nagel (AER 1995): p = 1/2



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Figure 1B of Nagel (AER 1995): p = 2/3



Joseph Tao-yi Wang Level-k Reasoning

- Named after Keynes, General Theory (1936)
- "…professional investment may be likened to those newspaper competitions in which the competitors have to pick out the six prettiest faces from a hundred photographs, (專業投資好比報紙上的選美比賽,要從上百張照片挑出最漂亮的六張)
- the prize being awarded to the competitor whose choice most nearly corresponds to the average preferences of the competitors as a whole..." (目標是選擇最接近「平均參賽者會選到的照片」)

- It is not a case of choosing those [faces] that, to the best of one's judgment, are really the prettiest,
 - 「這不是要挑每個人各自認為最漂亮的[臉蛋],
- nor even those that average opinion genuinely thinks the prettiest.
 - 更不是要挑大家公認最漂亮的。
- We have reached the third degree where we devote our intelligences to...
 - 我們已經想到第三層去,

- Anticipating what average opinion expects the average opinion to be.
 - 努力預測一般人心目中認為大家公認最漂亮的會是誰。
- And there are some, I believe, who practice the fourth, fifth and higher degrees."
 - 而且我相信有些人還可以想到第四層、第五層或更高。」
 - Keynes (凱因斯, 1936, p.156)
- Follow-up Studies (後續研究)
 - Camerer, Ho and Weigelt (AER 1998)

Camerer, Ho & Weigelt (AER 1998): Design

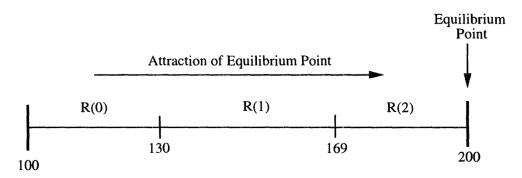


FIGURE 1A. A FINITE-THRESHOLD GAME, FT(n) = ([100, 200], 1.3, n)

3 rounds of IEDS

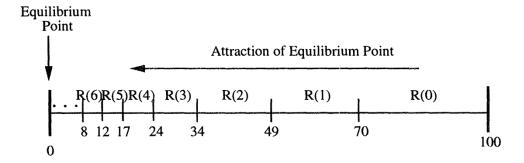


Figure 1B. An Infinite-Threshold Game, IT(n) = ([0, 100], 0.7, n)



Camerer, Ho & Weigelt (AER 1998): Design

TABLE 1—THE EXPERIMENTAL DESIGN 實驗設計 Group size 每組人數: 3 vs. 7 7 Finite → Infinite 先做有限次 $FT(1.3,3) \rightarrow IT(0.7,3)$ $FT(1.3,7) \rightarrow IT(0.7,7)$ (7 groups) 1.3 \rightarrow 0.7 (7 groups) $FT(1.1, 3) \rightarrow IT(0.9, 3)$ $FT(1.1, 7) \rightarrow IT(0.9, 7)$ 再做無限次 (刪劣勢策略) $(7 \text{ groups}) \qquad 1.1 \rightarrow 0.9 \qquad (7 \text{ groups})$ Infinite → Finite 先做無限次 $IT(0.7, 3) \rightarrow FT(1.3, 3)$ $IT(0.7, 7) \rightarrow FT(1.3, 7)$ $(7 \text{ groups}) \qquad 0.7 \rightarrow 1.3 \qquad (7 \text{ groups})$ 再做有限次 $IT(0.9, 3) \rightarrow FT(1.1, 3)$ $IT(0.9, 7) \rightarrow FT(1.1, 7)$ (6 groups) $0.9 \rightarrow 1.1$ (7 groups)

Camerer, Ho and Weigelt (AER 1998)

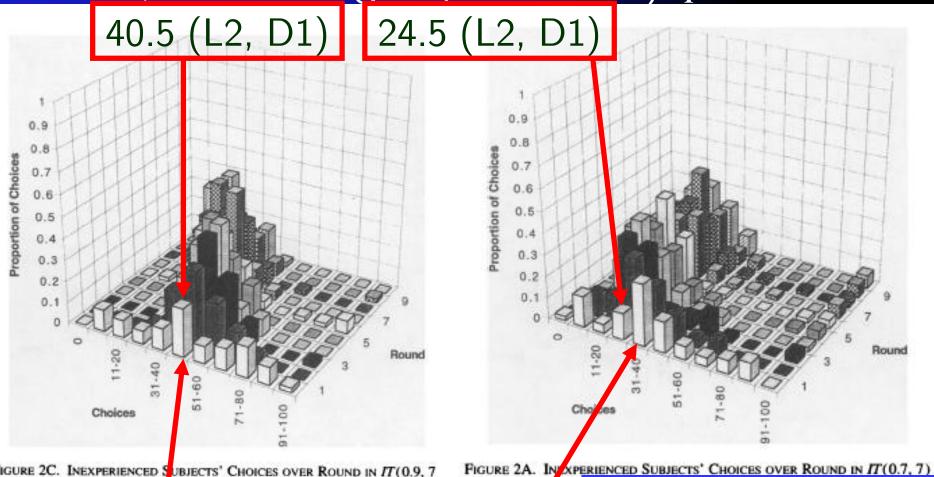
• RESULT 1:

First-period choices are far from equilibrium, and centered near the interval midpoint.

Choices converge toward the equilibrium point over time.

Baseline: IT(0.9,7) and IT(0.7, 7)

Camerer, Ho & Weigelt (AER 1998): p=0.9 vs. 0.7



45 (L1, D0)

35 (L1, D0)

"p=0.7" closer to 0

Camerer, Ho and Weigelt (AER 1998)

- IT(0.9,7) vs. IT(0.7, 7)
- RESULT 2:

On average, choices are closer to the equilibrium point

for games with finite thresholds, and for games with p further from 1.

• Infinite vs. Finite...

Camerer, Ho & Weigelt (1998): Finite Thresholds

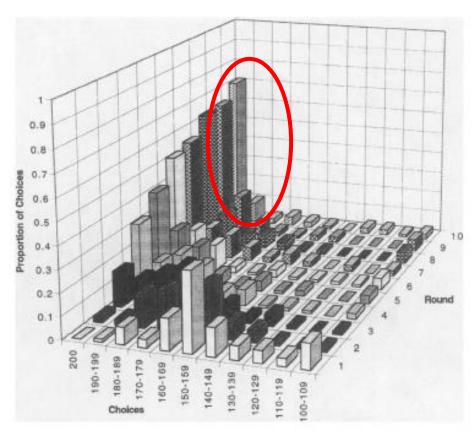


FIGURE 3A. CHOICES OVER ROUND IN FT GAMES PLAYED BY 3-PERSON GROUPS

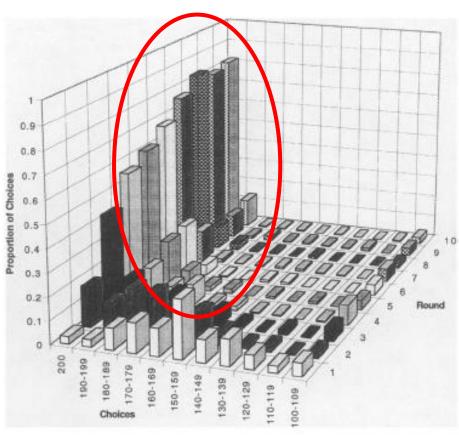


FIGURE 3B. CHOICES OVER ROUND IN FT GAMES PLAYED BY 7-PERSON GROUPS

FT closer to Equilibrium

7-group closer than 3-group

Camerer, Ho and Weigelt (AER 1998)

RESULT 3:

Choices are closer to equilibrium for large (7-person) groups than for small (3-person) groups.

• More on 7-group vs. 3-group...

Camerer, Ho & Weigelt (1998):7-grp vs. 3-grp

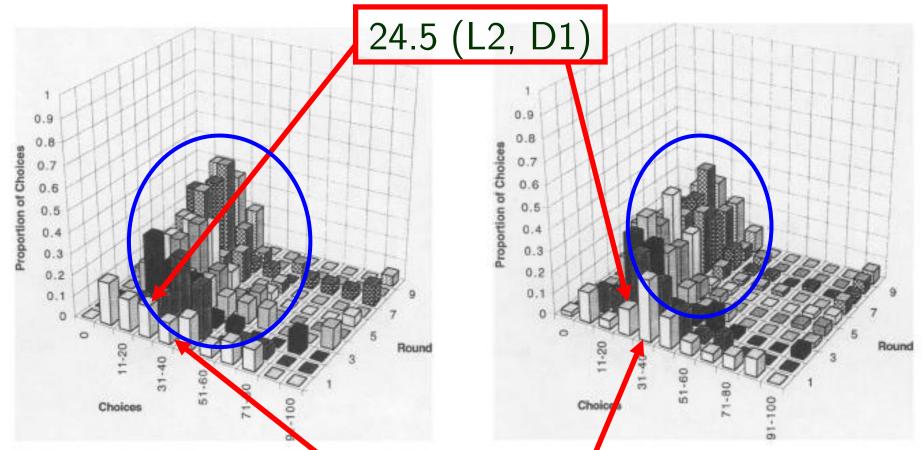
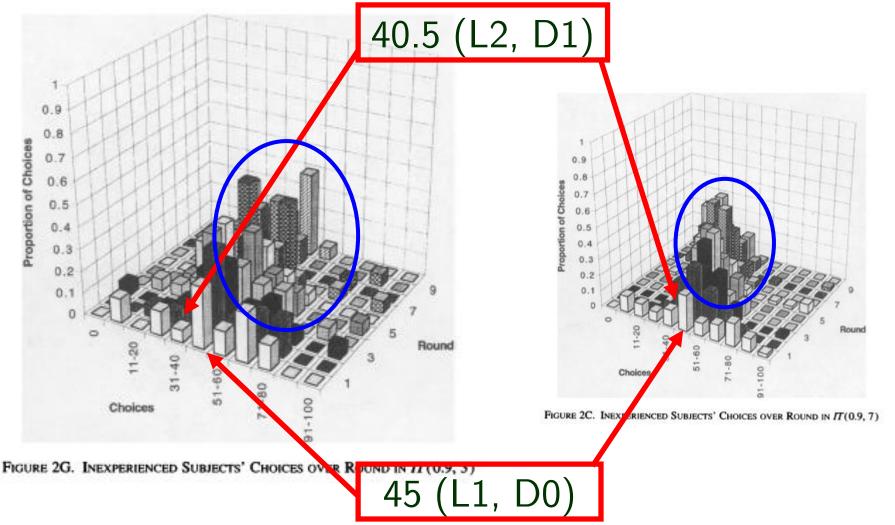


FIGURE 2E. INEXPERIENCED SUBJECTS' CHOICES OVER ROUND IN IT (0.7, 3

FIGURE 2A. INEXPERIENCED SUBJECTS' CHOICES OVER ROUND IN II(0.7, 7)

35 (L1, D0)

Camerer, Ho & Weigelt (1998):7-grp vs. 3-grp



Camerer, Ho and Weigelt (AER 1998)

RESULT 4:

Choices by [cross-game] experienced subjects are no different than choices by inexperienced subjects in the first round, but converge faster to equilibrium.

• Inexperienced vs. Experienced...

Camerer. Ho & Weigelt (1998): Exper. vs. Inexper.

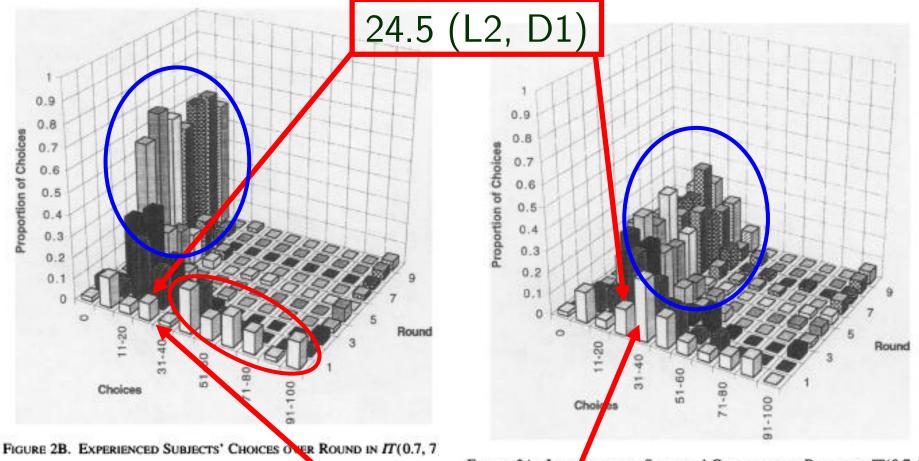
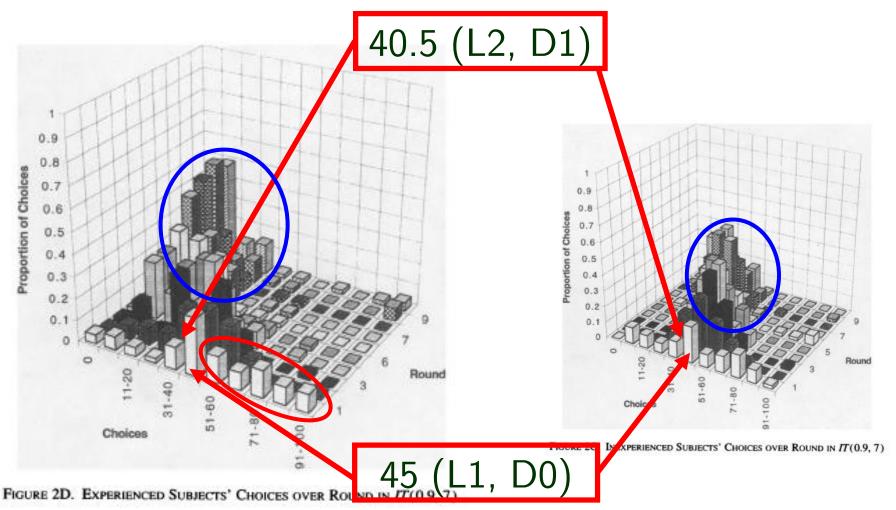


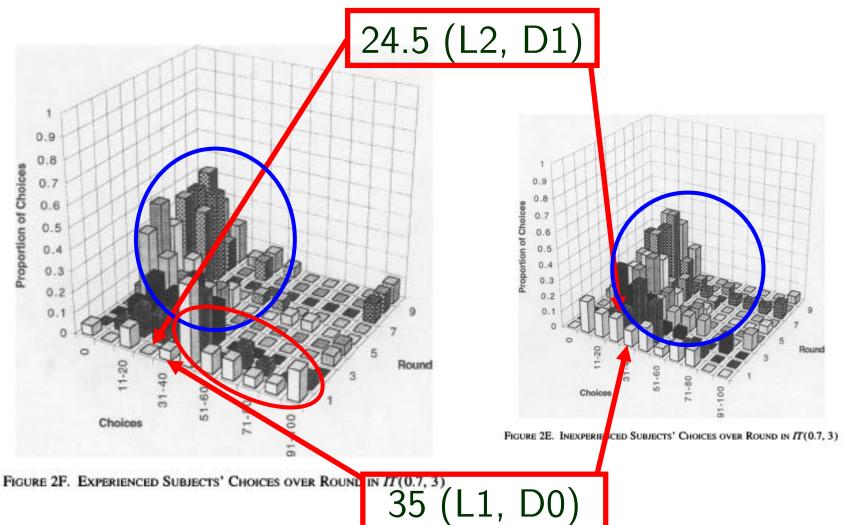
FIGURE 24. INEXTENSED SUBJECTS' CHOICES OVER ROUND IN II(0.7, 7)

35 (L1, D0)

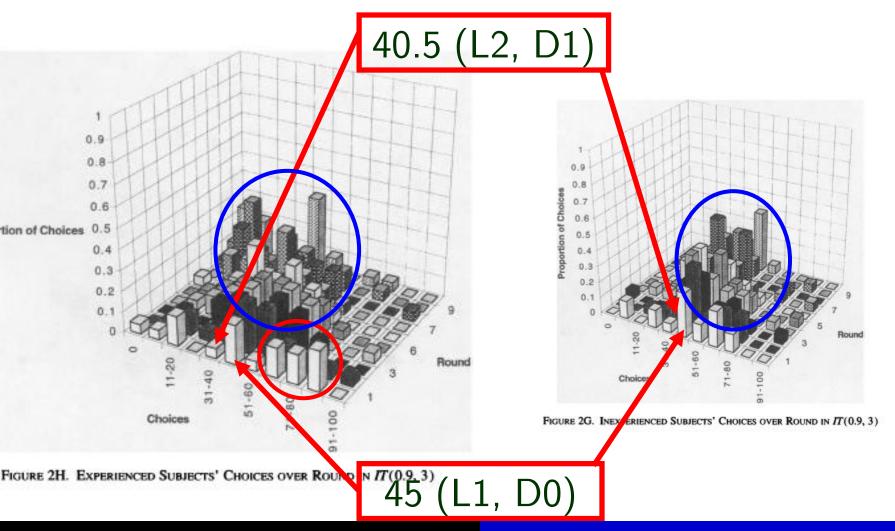
Camerer, Ho & Weigelt (1998): Exper. vs. Inexper.



Camerer, Ho & Weigelt (1998): Exper. vs. Inexper.



Camerer, Ho & Weigelt (1998): Exper. vs. Inexper.



Camerer, Ho and Weigelt (AER 1998)

- Classification of Types
 - Follow Stahl and Wilson (GEB 1995)
- Level-0: pick randomly from $N(\mu, \sigma)$
- Level-1: BR to level-0 with noise
- Level-2: BR to level-1 with noise
- Level-3: BR to level-2 with noise
- Estimate type, error using MLE

Camerer, Ho and Weigelt (AER 1998)

TABLE 3—MAXIMUM-LIKELIHOOD ESTIMATES AND LOG-LIKELIHOODS FOR LEVELS OF ITERATED DOMINANCE (FIRST-ROUND DATA ONLY)

Parameter		data of 3 or 7)	Nagel's data (groups of 16–18)			
estimates	IT(p, n)	FT(p, n)	IT(0.5, n)	IT(2/3, n)		
ω_0	15.93	21.72	45.83 (23.94)	28.36 (13.11)		
ω_1	20.74	31.46	37.50 (29.58)	34.33 (44.26)		
ω_2	13.53	12.73	16.67 (40.84)	37.31 (39.34)		
ω_3	49.50	34.08	0.00 (5.63)	0.00 (3.28)		
μ	70.13	100.50	35.53 (50.00)	52.23 (50.00)		
σ	28.28	26.89	22.70	14.72		
ho	1.00	1.00	0.24	1.00		
-LL	1128.29	1057.28	168.48	243.95		

Type distribution...

Camerer, Ho and Weigelt (AER 1998)

- Robustness checks:
 - High stakes (Fig.1.3 small effect lowering numbers)
 - Median vs. Mean (Nagel 1999 same): BGT Figure 5.1
 - p * (Median+18): Equilibrium is inside
- Subject Pool Variation:
 - Portfolio managers, Econ PhD, Caltech undergrads
 - Caltech Board of Trustees (CEOs)
 - Readers of Financial Times and Expansion
- Experience vs. Inexperience (for the same game)
 - Slonim (EE 2005) Experience good only for 1st round

Level-k Reasoning

- Theory for Initial Response (BGT, Ch. 5)
 vs. Theory for Equilibration (BGT, Ch. 6)
- First: Stahl and Wilson (GEB 1995)
- Better: Costa-Gomes, Crawford & Broseta (Econometrica 2001)
- Best 1: Camerer, Ho and Chong (QJE 2004)
 - Poisson Cognitive Hierarchy
- Best 2: Costa-Gomes & Crawford (AER 2006)

Level-k Theory: Stahl & Wilson (GEB 1995)

- Stahl and Wilson (GEB 1995)
- Level-0: Random play
- Level-1: BR to Random play
- Level-2: BR to Level-1
- Nash: Play Nash Equilibrium
- Worldly: BR to distribution of Level-0, Level-1 and Nash types

Level-k Theory: Stahl & Wilson (GEB 1995)

TABLE IV

PARAMETER ESTIMATES AND CONFIDENCE INTERVALS FOR MIXTURE MODEL WITHOUT RE TYPES

	Estimate	Std. Dev.	95 percent	conf. int.
γ_1	0.2177	0.0425	0.1621	0.3055
μ_2	0.4611	0.0616	0.2014	0.8567
			[0.2360]	0.8567]
γ2	3.0785	0.5743	1.9029	4.9672
			[2.5631	5.0000]
γ3	4.9933	0.9357	1.9964	5.0000
4	0.0624	0.0063	0.0527	0.0774
4	0.4411	0.0773	0.2983	0.5882
74	0.3326	0.0549	0.2433	0.4591
x _O	0.1749	0.0587	0.0675	0.3047
¥ ₁	0.2072	0.0 <u>57</u> 5	0.1041	0.3298
X 2	0.0207	0.0202VD	e distri	ibution
¥ 3	0.1666	1 DEUG	0.0600	0.2057
2 4	0.4306	0.0782	0.2810	0.5723
С	-442.727			

Level-k Theory: CGCB (ECMA 2001)

- Costa-Gomes, Crawford & Broseta (2001)
- 18 2-player NF games designed to separate:
- Naive (L1), Altruistic (max sum)
- Optimistic (maximax), Pessimistic (maximin)
- L2 (BR to L1)
- D1/D2 (1/2 round of DS deletion)
- Sophisticated (BR to empirical)
- Equilibrium (play Nash)

Level-k Theory: CGCB (ECMA 2001)

- Three treatments (all no feedback):
- Baseline (B)
 - Mouse click to open payoff boxes
- Open Box (OB)
 - Payoff boxes always open
- Training (TS)
 - Rewarded to choose equilibrium strategies

Level-k Theory: CGCB (Econometrica 2001)

- Results 1: Consistency of Strategies with Iterated Dominance
- B, OB: 90%, 65%, 15% equilibrium play
 - For Equilibria requiring 1, 2, 3 levels of ID
- TS: 90-100% equilibrium play
 - For all levels
- Game-theoretic reasoning is not computationally difficult, but unnatural.

Result 2: E	- - - - - - - -	e Subject	Decision Rule
Rule	E(u)	Choice (%)	Choice+Lookup (%)
Altruistic	17.11	8.9	2.2
Pessimistic	20.93	0	4.5
Na ï ve	21.38	22.7	44.8
Optimistic	21.38	0	2.2
L2	24.87	44.2	44.1
D1	24.13	19.5	0
D2	23.95	0	0
Equilibrium	24.19	5.2	0
Sophisticated	24.93	0	2.2

Result 3: Information Search Patterns

Subject /	‡ own ¡	payoff	→ other payoff		
Rule	Predicted	Actual	Predicted	Actual	
TS (Equil.)	>31	63.3	>31	69.3	
Equilibrium	>31	21.5	>31	79.0	
Naive/Opt.	<31	21.1	-	48.3	
Altruistic	<31	21.1	-	60.0	
L2	>31	39.4	=31	30.3	
D1	>31	28.3	>31	61.7	

Level-k Theory: CGCB (ECMA 2001)

- Result 3: Information Search Patterns
- Occurrence (weak requirement)
 - All necessary lookups exist somewhere
- Adjacency (strong requirement)
 - Payoffs compared by rule occur next to each other
- H-M-L: % of Adjacency | 100% occurrence

Result 3: Information Search Patterns

TABLE V

AGGREGATE RATES OF COMPLIANCE WITH TYPES' OCCURRENCE AND ADJACENCY FOR TS AND BASELINE SUBJECTS, AND FOR BASELINE SUBJECTS BY

MOST LIKELY TYPE ESTIMATED FROM DECISIONS ALONE, IN PERCENTAGES (— VACUOUS)

Treatment (# subjects)	Altruistic $J = H, M, L, 0$	Pessimistic $j = H, M, L, 0$	Naïve $j = H, M, L, 0$	Optimistic $j = A,0$	L2 $j = H, M, L, 0$	D1 = H, M, L, 0	D2 $j = H, M, L, 0$	Equilibrium $j = H, M, L, 0$	Sophisticated $j = H, M, L, 0$
TS (12) Baseline (45)	3,10,50,27 14,11,51,24	44,7,36,13 74,2,11,14	83,2,0,15 78,4,4,14	86,14 85,15	76,2,0,22 67,14,5,14	92,3,1,5 52,19,15,14	92,3,1,5 50,19,15,14	96,1,1,3 42,23,19,16	75,1,1,24 39,21,20,21
Altruistic (2)	78,6,11,6	56,8,33,3	53,3,42,3	97,3	47,8,39,6	36,6,56,3	33,8,56,3	31,11,56,3	28,14,56,3
Pessimistic (0)	-,-,-,-	-,-,-,-	-,-,-,-	-,-	-,-,-,-	-,-,-,-	-,-,-,-	-,-,-,-	-,-,-,-
Naïve/Optim. (11)	9,5,53,33	85,1,9,5	89,5,3,4	96,4	42,24,3,31	45,22,20,13	43,18,23,16	26,24,28,23	23,23,27,27
L2 (23)	8,12,58,22	72,2,9,17	78,3,0,18	80,20	85,6,3,6	57,20,9,15	54,21,10,15	49,24,12,15	46,22,12,20
D1 (7)	23,21,26,29	59,3,16,23	63,7,6,23	77,23	53,21,6,21	48,17,14,20	45,19,15,21	42,20,17,21	38,14,21,27
D2 (0)	-,-,-,-	-,-,-,-	-,-,-,-	-,-	-,-,-,-	-,-,-,-	-,-,-,-	-,-,-,-	-,-,-,-
Equilibrium (2)	6,8,86,0	100,0,0,0	97,3,0,0	100,0	64,36,0,0	69,17,14,0	67,19,14,0	56,25,19,0	53,19,28,0
Sophisticated (0)	_,_,_,_	_,_,_,_	,,		_,_,_,_	_,_,_,_	_,_,_,_	_,_,_,_	_,_,_,_

Level-k Theory: Cognitive Hierarchy

- Camerer, Ho and Chong (QJE 2004)
- Poisson distribution of level-k thinkers $f(k|\tau)$
 - $-\tau$ = mean number of thinking steps
- Level-0: choose randomly or use heuristics
- Level-k thinkers use k steps of thinking BR to a mixture of lower-step thinkers
 - Belief about others is Truncated Poisson
- Easy to compute; Explains many data

- Costa-Gomes & Crawford (2006)
- 2-Person Guessing Games (p-beauty contest)
 - Player 1's guesses 300-500, target = 0.7
 - Player 2's guesses 100-900, target = 1.5
 - $-0.7 \times 1.5 = 1.05 > 1...$
- Unique Equilibrium @ upper bound (500, 750)
- In general:
- Target1 x Target2 > 1: Nash @ upper bounds
- Target1 x Target2 < 1: Nash @ lower bounds

- 16 Different Games
- Limits:
- $\alpha = [100, 500], \beta = [100, 900],$
- $\gamma = [300, 500], \delta = [300, 900]$
- Target: 1 = 0.5, 2 = 0.7, 3 = 1.3, 4 = 1.5

No feedback — Elicit Initial Responses

- Define Various Types:
- Equilibrium (EQ): BR to Nash (play Nash)
- Defining L0 as uniformly random
 - Based on evidence from past normal-form games
- Level-k types L1, L2, and L3:
- L1: BR to L0
- L2: BR to L1
- L3: BR to L2

- Dominance types:
 - D1: Does one round of dominance and BR to a uniform prior over partner's remaining decisions
 - D2: Does two rounds and BR to a uniform prior
- Sophisticated (SOPH): BR to empirical distribution of others' decisions
 - Ideal type (if all SOPH, coincide with Equilibrium)
 - See if anyone has a transcended understanding of others' decisions

1 6	Come	11	1.0					CODU
	Game 14. β4γ2	L1 600	L2 525	L3 630	D1 600	D2 611.25	EQ 750	SOPH 630
	6. δ3γ4	520	650	650	617.5	650	650	650
	7. δ3δ3	780	900	900	838.5	900	900	900
	11. δ2β3	350	546	318.5	451.5	423.15	300	420
	16. α4α2	450	315	472.5	337.5	341.25	500	375
	1. α2β1	350	105	122.5	122.5	122.5	100	122
	15. α2α4	210	315	220.5	227.5	227.5	350	262
	13. γ2β4	350	420	367.5	420	420	500	420
	5. γ4δ3	500	500	500	500	500	500	500
	4. γ2β1	350	300	300	300	300	300	300
	10. α4β1	500	225	375	262.5	262.5	150	300
	8. δ3δ3	780	900	900	838.5	900	900	900
	12. β3δ2	780	455	709.8	604.5	604.5	390	695
	3. β1γ2	200	175	150	200	150	150	162
	2. β1α2	150	175	100	150	100	100	132
	9. β1α4	150	250	112.5	162.5	131.25	100	187

- 43 (out of 88) subjects in the baseline made exact guesses (+/- 0.5) in 7 or more games
- Distribution: (L1, L2, L3, EQ) = (20, 12, 3, 8)

TABLE 1—SUMMARY OF BASELINE AND OB SUBJECTS' ESTIMATED TYPE DISTRIBUTIONS

Туре	Apparent from guesses	Econometric from guesses	Econometric from guesses, excluding random	Econometric from guesses, with specification test	Econometric from guesses and search, with specification test
L1	20	43	37	27	29
L2	12	20	20	17	14
L3	3	3	3	1	1
D1	0	5	3	1	0
D2	0	0	0	0	0
Eq.	8	14	13	11	10
Soph.	0	3	2	1	1
Unclassified	45	0	10	30	33

Note: The far-right-hand column includes 17 OB subjects classified by their econometric-from-guesses type estimates.

- No Dk types
- No SOPH types
- No L0 (only in the minds of L1...)
- Deviation from Equilibrium is cognitive
- Cannot distinguish/falsify Cognitive Hierarchy
 - BR against lower types, not just L(k-1)
- But distribution is not Poisson (against CH)
 - Is the Poisson assumption crucial?

- Pseudotypes: Constructed with subjects' guesses in the 16 games (Pseudo-1~Pseudo-88)
- Specification Test: Compare the likelihood of subject's type with likelihoods of pseudotypes
 - Should beat at least 87/8 = 11 pseudotypes
 - Unclassified if failed
- Omitted Type Test: Find clusters that
 - (a) Look like each other, but (b) not like others
 - Pseudotype likelihoods high within, low outside

- 5 small clusters; total = 11 of 88 subjects
- Other clusters?
 - Could find more smaller clusters in a larger sample, but size smaller than 2/88 (approx. 2%)
- Smaller clusters could be treated as errors
 - No point to build one model per subject...
 - A model for only 2% of population is not general enough to make it worth the trouble

- Level-k model explains a large fraction of subjects' deviations from equilibrium
 - (that can be explained by a model)
- Although the model explains only half+ of subjects' deviations from equilibrium,
- it may still be optimal for a modeler to treat the rest of the deviations as errors
 - Since the rest is not worth modeling...

How Level-k Model Explains Hide-and-Seek Games?

- Aggregate RTH Hide-and-Seek Game Results:
- Both Hiders and Seekers over-choose central A
- Seekers choose central A even more than hiders

	A	В	A	Α
Hiders	0.2163	0.2115	0.3654	0.2067
(624)				
Seekers	0.1821	0.2054	0.4589	0.1536
(560)				

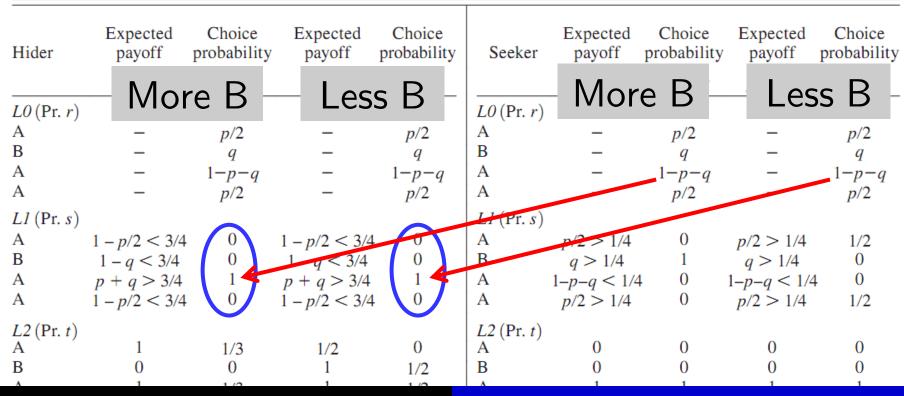
- Can a strategic theory explain this?
- Level-k: Each role is filled by Lk types: L0, L1, L2, L3, or L4 (probabilities to be estimated)
 - Note: In Hide and Seek the types cycle after 24...
- High types anchor beliefs in a naive <u>L0</u> type and adjusts with iterated best responses:
 - 1 best responds to 0 (with uniform errors)
 - L2 best responds to L1 (with uniform errors)
 - Lk best responds to Lk-1 (with uniform errors)

Hide-and-Seek Games: Anchoring Type Level- θ

- L0 Hiders and Seekers are symmetric
 - Favor salient locations equally
- 1. Favor B: choose with probability q > 1/4
- 2. Favor end A: choose with prob. p/2>1/4
 - Choice probabilities: (p/2, q, 1 p q, p/2)
- Note: Specification of Anchoring Type L0 is the key to model's explanatory power
 - See Crawford and Ireberri (AER 2007) for other <u>L0</u>
 - Cannot use uniform <u>L0</u> (coincide with equilibrium)...

- More (or less) attracted to B: p/2 < q (p/2 > q)
- L1 Hiders choose central A

Table 2—Types' Expected Payoffs and Choice Probabilities in RTH's Games when p > 1/2 and q > 1/4



- More (or less) attracted to B: p/2 < q (p/2 > q)
- L1 Seekers avoid central A (pick B or end A)

Table 2—Types' Expected Payoffs and Choice Probabilities in RTH's Games when p > 1/2 and q > 1/4

					I				
Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
	Mor	e R -	_	sB-		Mor	e B	ا ا	ss B
L0 (Pr. r)	14101		L C3	5 0	<i>L0</i> (Pr. r)	17101			
A	_	<i>p</i> /2	_	p/2	A	_	p/2	_	p/2
В	_	\dot{q}	_	\dot{q}	В	_	, q	_	\dot{q}
A	_	1 - p - q	_	1-p-q	A		1-p-q	_	1-p-q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
<i>L1</i> (Pr. s)					LI PLS				
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	0	A	0/2 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4		В	q > 1/4		q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4		A	1 - p - q < 1/q	4 0	1 - p - q < 1/4	4 0
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4		A	p/2 > 1/4	0	p/2 > 1/4	1/2
<i>L2</i> (Pr. <i>t</i>)					L2 (Pr. t)				
A (11.1)	1	1/3	1/2	0	A	0	0	0	0
В	0	0	1	1/2	В	0	0	0	0
٨	1	1./2	1	1./2	Α	1	1	1	1

- More (or less) attracted to B: p/2 < q (p/2 > q)
- L2 Hiders choose central A with prob. in [0.1]

Table 2—Types' Expected Payoffs and Choice Probabilities in RTH's Games when p > 1/2 and q > 1/4

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
<i>L0</i> (Pr. <i>r</i>)	Mor	e B	Les	sB-	$\frac{1}{L0(\text{Pr. }r)}$	Mor	e B	Les	s B
A	_	<i>p</i> /2	_	<i>p</i> /2	A	_	<i>p</i> /2	_	p/2
В	_	\boldsymbol{q}	_	q	В	_	\boldsymbol{q}	_	q
A	_	1 - p - q	_	1 - p - q	A	_	1 - p - q	_	1 - p - q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
L1 (Pr. s)					<i>L1</i> (Pr. s)				
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4	0	В	q > 1/4	1	q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4	1	A	1-p-q < 1/4	4 0	1 - p - q < 1/4	0
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
<i>L2</i> (Pr. <i>t</i>)					<i>L2</i> (Pr. t)				
A	1	1/3	1/2	0	A	0	0	0	0
В	0	0	1	1/2	В	Û	0	0	0
A	1	1/3	1	1/2	A	1	1	1	1
A	1	1/3	1/2	0	A	0	0	0	0

- More (or less) attracted to B: p/2 < q (p/2 > q)
- L2 Seekers choose central A for sure Table 2—Types' Expected Payoffs and Choice Probabilities in RTH's Games when p > 1/2 and q > 1/4

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
<i>L0</i> (Pr. <i>r</i>)	Mor	e B	Less	B -	$\frac{1}{L0(\text{Pr. }r)}$	Mor	e B	Les	sB-
Α	_	p/2	_	p/2	A	_	p/2	_	p/2
В	_	\boldsymbol{q}	_	q	В	_	q	_	q
A	_	1 - p - q	_	1 - p - q	A	_	1-p-q	_	1-p-q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
<i>L1</i> (Pr. s)					<i>L1</i> (Pr. s)				
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4	0	В	q > 1/4	1	q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4	1	A	1 - p - q < 1/4	0	1-p-q < 1/4	0
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
<i>L2</i> (Pr. <i>t</i>)					<i>L2</i> (Pr. t)				
A	1	1/3	1/2	0	A	0	0	0	0
В	0	0	1	1/2	В	Û	0	0	0
A	1	1/3	1	1/2	A	1	> 1	1	7 1
A	1	1/3	1/2	0	A	0	0	0	0

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
$_{L0\mathrm{(Pr.r)}}$ More B — Less B –					$\frac{1}{L0(\text{Pr. }r)}$	Mor		Les	s B
A B	_	p/2	_	p/2	A B	_	p/2	_	p/2
A	_	<i>q</i>	_	<i>q</i>	A	_	<i>q</i>	_	<i>q</i>
A	_	1-p-q $p/2$	_	1-p-q $p/2$	A	_	1-p-q	_	1-p-q
		p/2		p/2			p/2		p/2
L1 (Pr. s)	4 10 - 214	0		0	L1 (Pr. s)		0	0 - 14	4.10
A	1 - p/2 < 3/4		1 - p/2 < 3/4		A	p/2 > 1/4	0	p/2 > 1/4	1/2
B A	1 - q < 3/4	0 1	1 - q < 3/4	0 1	В	q > 1/4	1	q > 1/4	. 0
A A	p + q > 3/4		p + q > 3/4		A A	1-p-q < 1/4	1 0 0	1-p-q < 1/4	1/2
	1 - p/2 < 3/4	U	1 - p/2 < 3/4	. 0		p/2 > 1/4	U	p/2 > 1/4	1/2
L2 (Pr. t)				0	<i>L2</i> (Pr. <i>t</i>)	0	0		0
A	1	1/3	1/2	0	A	0	0	0	0
В	0	0	1	1/2	В	0	0	0	0
A A	1	1/3	1 /2	1/2 0	A	1	0	1	1
	1	1/3	1/2	U	A	0	0	1	U
<i>L3</i> (Pr. <i>u</i>)					£3 (Pr. u)				
A	1	1/3	1	1/3	A	1/3	1/3	0	0
В	1	1/3	I	1/3	В	0	0	1/2	1/2
A	0	0	0	0	A	1/3	1/3	1/2	1/2
A	1	1/3	1	1/3	A	1/3	1/3	0	0
L4 (Pr. v)					L4 (Pr. v)				
A	2/3	0	1	1/2	A	1/3	1/3	1/3	1/3
В	1	1	1/2	0	В	1/3	1/3	1/3	1/3
A	2/3	0	1/2	0	A	0	0	0	0
Λ	2/2	Λ	1	1 /2	i A	1./2	1 /2	1./2	1./2

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
$\overline{B_{LO(Pr,r)}}$ More B — Less B —						Mor	$rac{1}{2}$ R $-$	$ $ $_{\Box}$	s B $^-$
L0 (Pr. r)	10101		LCS		<i>L0</i> (Pr. r)	10101		LCS	55 D
A	_	p/2	_	p/2	A	_	p/2	_	p/2
В	_	q	_	q	В	_	q	_	q
A	_	1 - p - q	_	1 - p - q	A	_	1-p-q	_	1 - p - q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
<i>L1</i> (Pr. s)					<i>L1</i> (Pr. s)				
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	. 0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4	0	В	q > 1/4	1	q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4		A	1-p-q < 1/4		1-p-q < 1/4	
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4		A	p/2 > 1/4	0	p/2 > 1/4	1/2
10(D)	P		P		10(5)	P / = / -		P =	-, -
L2 (Pr. t)	1	1 /2	1 /2	0	L2 (Pr. t)	0	0	0	0
A B	0	1/3 0	1/2 1		A B	0	0	0	0
A	1	1/3	1	1/2	A	1	1	1	1
A	1	1/3	1/2	1/2 0	A	0	0	0	0
	1	1/3	172	U		U	O	U	U
<i>L3</i> (Pr. <i>u</i>)					<i>L3</i> (Pr. <i>u</i>)				
A	1	1/3	1	1/3	A	1/3	1/3	0	0
В	1	1/3	1	1/3	В	0	0	1/2	1/2
A	0	0	0	0	A	1/3	1/3	1/2	1/2
A	1	1/3	1	1/3	A	1/3	1/3	0	0
L4 (Pr. v)					L4 (Pr. v)				
A A	2/3	0	1	1/2	A	1/3	1/3	1/3	1/3
В	1	1	1/2	0	В	1/3	1/3	1/3	1/3
A	2/3	0	1/2	0	A	0	0	0	0
Α	2/2	0	1	1./2	Λ	1./2	1./2	1./2	1./2

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
<i>L0</i> (Pr. <i>r</i>)	Mor	re B	Les		$\boxed{\frac{1}{L0(\text{Pr. }r)}}$	Mor		Les	s B
A	_	p/2	_	p/2	A	_	p/2	_	p/2
В	_	q	_	q	В	_	q	_	q
A	_	1-p-q	_	1-p-q	A	_	1-p-q	_	1-p-q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
L1 (Pr. s)					<i>L1</i> (Pr. s)				
A	1 - p/2 < 3/4		1 - p/2 < 3/4		A	p/2 > 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4	0	В	q > 1/4	1	q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4		A	1-p-q < 1/4		1 - p - q < 1/4	
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	. 0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
<i>L2</i> (Pr. <i>t</i>)					<i>L2</i> (Pr. <i>t</i>)				
A (11.1)	1	1/3	1/2	0	A	0	0	0	0
В	0	0	1	1/2	В	0	0	0	0
A	1	1/3	1	1/2	A	1	1	1	1
A	1	1/3	1/2	0	A	0	0	0	0
<i>L3</i> (Pr. <i>u</i>)					<i>L3</i> (Pr. <i>u</i>)				
A	1	1/3	1	1/3	A	1/3	1/3	0	0
В	1	1/3	1	1/3	В	0	0	1/2	1/2
A	0	0	0	0	A	1/3	1/3	1/2	1/2
A	1	1/3	1	1/3	A	1/3	1/3	0	0
<i>L4</i> (Pr. v)					<i>L4</i> (Pr. v)	_			
A (11. V)	2/3	0		1/2	A (11. V)	1/3	1/3	1/3	1/3
В	1	1	1/2	0	B	1/3	1/3	1/3	1/3
A	2/3	0	1/2	0	A	0	0	0	0
٨	2/2		1	1/2	Λ	1./2	1./2	1/2	1./2

Joseph Tao-yi Wang Level-k Reasoning

Hider	Expected payoff	Choice probability	Expected payoff	Choice probability	Seeker	Expected payoff	Choice probability	Expected payoff	Choice probability
<i>L0</i> (Pr. <i>r</i>)	- Mor	e B	Les		$\boxed{\frac{1}{L0(\text{Pr. }r)}}$	Mor		Les	s B
A	_	p/2	_	p/2	A	_	p/2	_	p/2
В	_	q	_	q	В	_	q	_	q
A	_	1-p-q	_	1-p-q	A	_	1-p-q	_	1-p-q
A	_	p/2	_	p/2	A	_	p/2	_	p/2
L1 (Pr. s)					<i>L1</i> (Pr. s)				
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	. 0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
В	1 - q < 3/4	0	1 - q < 3/4	0	В	q > 1/4	1	q > 1/4	0
A	p + q > 3/4	1	p + q > 3/4	1	A	1-p-q < 1/4	4 0	1-p-q < 1/4	. 0
A	1 - p/2 < 3/4	0	1 - p/2 < 3/4	. 0	A	p/2 > 1/4	0	p/2 > 1/4	1/2
<i>L2</i> (Pr. t)					<i>L2</i> (Pr. <i>t</i>)				
A A	1	1/3	1/2	0	A A	0	0	0	0
В	0	0	1	1/2	В	0	0	0	0
A	1	1/3	1	1/2	A	1	1	1	1
A	1	1/3	1/2	0	A	0	0	0	0
<i>L3</i> (Pr. <i>u</i>)					<i>L3</i> (Pr. <i>u</i>)				
Α	1	1/3	1	1/3	A	1/3	1/3	0	0
В	1	1/3	1	1/3	В	0	0	1/2	1/2
A	0	0	0	0	A	1/3	1/3	1/2	1/2
A	1	1/3	1	1/3	A	1/3	1/3	0	0
(D		2,0			(5		2,0		
<i>L4</i> (Pr. <i>v</i>)	2/2	0	1	1./2	<i>L4</i> (Pr. v)	1/2	1/2	1/2	1/2
A B	2/3 1	1	1 /2	1/2 0	A B	1/3	1/5	1/3	1/3
A	-	0	1/2 1/2	0	A	0	1/3	0	0
Λ .	2/3	0	1/2	1/2	A .	1/2	1/2	1/2	1/2

Joseph Tao-yi Wang Level-k Reasoning

Hide-and-Seek Games: Explain Stylized Facts

- Given L0 playing (p/2, q, 1 p q, p/2),
 - L1 Hiders choose central A (avoid L0 Seekers)
 - L1 Seekers avoid central A (search for L0 Hiders)
- L2 Hiders choose central A with prob. in [0,1]
- L2 Seekers choose central A for sure
- L3 Hiders avoid central A
- L3 Seekers choose central A w/ prob. in [0,1]
- L4 Hiders and Seekers both avoid central A

Hide-and-Seek Games: Explain Stylized Facts

- Heterogeneous Population (L0, L1, L2, L3, L4) = (r, s, t, u, v) with r=0, t, u large and s not too large can reproduce the stylized facts
- Need s < (2t+u)/3 (More B) or s < (t+u)/2 (Less B)
- estimated r = 0, s=19%, t=32%, u=24%, v=25%

Joseph Tao-yi Wang Level-k Reasoning

Hide-and-Seek Games: Out of Sample Prediction

- Estimate on one treatment and predict other five treatments
 - 30 Comparisons: 6 estimations, each predict 5
- This Level-k Model with symmetric L0 beats other models (LQRE, Nash + noise)
 - Mean Squared prediction Error (MSE) 18% lower
 - Better predictions in 20 of 30 comparisons

HS Level-k Model Ported to Joker Game

- Can Level-k Reasoning developed from the Hide-and-Seek Game predict results of other games?
 - Try O'Neil (1987)'s Joker Game
- Stylized Facts:
 - Aggregate Frequencies close MSE
 - Ace Effect (A chosen more often than 2 or 3);
 - Not captured by QRE

Neill (1987) The Joker Game: O'

	Α	2	3	J	MSE	Actual	QRE		
Α	-5	5	5	-5	0.2	0.221	0.213		
2	5	-5	5	-5	0.2	0.215	0.213		
3	5	5	-5	-5	0.2	0.203	0.213		
J	-5	-5	-5	5	0.4	0.362	0.360		
MSE	0.2	0.2	0.2	0.4	 Actual frequite close QRE bette get the Actual frequite 				
Actual	0.226	0.179	0.169	0.426					
QRE	0.191	0.191	0.191	0.427					

 Actual frequencies are quite close to MSE

 QRE better, but cannot get the Ace effect

HS Level-k Model Ported to Joker Game

- Level-k model w/ symmetric L0 (favor A&J)
- $LO: (a, (1-a-j)/2, (1-a-j)/2, j), a, j>\frac{1}{4}$
 - A and J, 'face' cards and end locations, are more salient than 2 and 3...
- Higher Lk type BR to L(k-1) (Table A3-A4)
- Challenge: To get the Ace Effect (without L0), need a population of almost all L4 or L3
 - This is an empirical question, but very unlikely

HS Level-k Model Ported to Joker Game

- Could there be no Ace Effect in the initial rounds of O'Neil's data?
 - The Level-k model predicts a Joker Effect instead!
- Crawford and Ireberri asked for O'Neil's data
 - And they found...
- Initial Choice Frequencies
 - -(A, 2, 3, J) = (8%, 24%, 12%, 56%) for Player 1
 - -(A, 2, 3, J) = (16%, 12%, 8%, 64%) for Player 2

Table 5. Comparison of the Leading Models in O'Neill's Game										
Model	Parameter estimates Observed or predicted choice frequencies						MSE			
		Player	А	2	3	J				
Observed frequencies		1	0.0800	0.2400	0.1200	0.5600	-			
(25 Player 1s, 25 Player 2s)		2	0.1600	0.1200	0.0800	0.6400	-			
Equilibrium without		1	0.2000	0.2000	0.2000	0.4000	0.0120			
perturbations		2	0.2000	0.2000	0.2000	0.4000	0.0200			
Level- k with a role-symmetric	a > 1/4 and $j > 1/4$	1	0.0824	0.1772	0.1772	0.5631	0.0018			
<i>L0</i> that favors salience	3j - a < 1, a + 2j < 1	2	0.1640	0.1640	0.1640	0.5081	0.0066			
Level- k with a role-symmetric	a > 1/4 and $j > 1/4$	1	0.0000	0.2541	0.2541	0.4919	0.0073			
L0 that favors salience	3j - a < 1, a + 2j > 1	2	0.2720	0.0824	0.0824	0.5631	0.0050			
Level- k with a role-symmetric	a < 1/4 and $j < 1/4$	1	0.4245	0.1807	0.1807	0.2142	0.0614			
L0 that avoids salience		2	0.1670	0.1807	0.1807	0.4717	0.0105			
Level- k with a role-asymmetric $L0$	a < 1/4 $i > 1/4$									

for which player is a seeker and avoids it for $3j_1$ - a_1 < 1, a_1 + $2j_1$ <

 $a_1 < 1/4, j_1 > 1/4;$ $a_2 > 1/4, j_2 < 1/4$

that favors salience for locations

Joseph Tao-yi Wang Level-k Reasoning

0.1804

0.2729

0.2729

0.2739

0.0291

Conclusion

- Limit of Strategic Thinking: 2-3 steps
- Theory (for initial responses)
- Level-k Types:
 - Stahl-Wilson (GEB 1995), CGCB (ECMA 2001)
 - Costa-Gomes and Crawford (AER 2006)
 - Chen, Huang and Wang (mimeo 2013)
- Cognitive Hierarchy:
 - CHC (QJE 2004)

Applications

- p -Beauty Contest:
 - Costa-Gomes and Crawford (AER 2006)
 - Chen, Huang and Wang (mimeo 2013)
- MSE:
 - Hide-and-Seek: Crawford and Iriberri (AER 2007)
 - LUPI: Ostling, Wang, Chou and Camerer (AEJ 2011)
- Auctions:
 - Overbidding: Crawford and Iriberri (AER 2007)
 - Repeated eBay Auctions: Wang (2006)

More Applications

- Coordination-Battle of the Sexes (Simple Market Entry Game):
 - Camerer, Ho and Chong (QJE 2004)
 - Crawford (2007)
- Pure Coordination Games:
 - Crawford, Gneezy and Rottenstreich (AER 2008)
- Pre-play Communication:
 - Crawford (AER 2003)
 - Ellingsen and Ostling (AER 2011)

More Applications

- Strategic Information Communication:
 - Crawford (AER 2003)
 - Cai and Wang (GEB 2006)
 - Kawagoe and Takizawa (GEB 2008)
 - Wang, Spezio and Camerer (AER 2010)
 - Brown, Leveno and Camerer (AEJ 2012)
 - Lai, Lim and Wang (GEB 2015)
 - Battaglini, Lai, Lim and Wang (work-in-progress)

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