

Pinocchio's Pupil:

Using Eyetracking and Pupil Dilation
To Understand Truth-telling and
Deception in Sender-Receiver Games

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Security Analyst Stock Ratings

- 1=Buy
 - 2=Accumulate
 - 3=Neutral
 - 4=Reduce
 - 5=Sell
- 2-1 means:
Short run: Accumulate,
Long run: Buy
 - Short run: 0-12 months
 - Long run: beyond 1 year

Henry Blodgett covers LFMN

\$22.69

Blodgett: 2-1

From: Blodgett, Henry (RSCH)
Sent: Monday, December 04, 2000 1:08 PM
To: Gary, Eve (RSCH)
Subject: RE: Internet ad spend-media conference



LFMN at \$4. I can't believe what a
POS [piece-of-sh-t] that thing is. Shame
on me/us for giving them any benefit
of the doubt.

Blodgett: 2-1

\$4

Some Examples of Strategic Information Transmission

- Managers Inflating Earning Prospectives
 - “My personal belief is that Enron stock is an incredible bargain.”
 - “We will hit our numbers.” ~Kenneth Lay
- Grade Inflation
- Teacher Cheating Student Tests
- Government-Expert
- Doctor-Patient
- Congress Floor-Committee

Strategic Information Transmission (Cheap Talk, Biased Transmission Game)

- Sender sees “secret number” $S=1,2,3,4,5$
- Sender sends message M
- Receiver gets message M , but not S
- Receiver chooses action A
- Payoffs depend on S and A
 - Sender earns most if receiver picks $A=S+b$
 - Receiver earns most if receiver picks $A=S$
 - b is “bias” ($b=0$: truth-telling, $b=2$: babbling)
 - ($b=1$: partition)

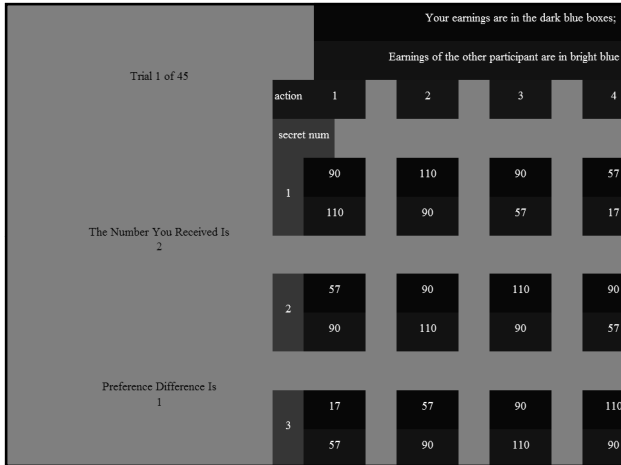
Trail 2 of 45

Your earnings are in the dark blue boxes.
Earnings of the other participant are in light blue boxes.

secret num	action	1	2	3	4	5	secret num
1	1	110	90	57	17	-29	1
1	2	110	90	57	17	-29	1
2	1	90	110	90	57	17	2
2	2	90	110	90	57	17	2
3	1	57	90	110	90	57	3
3	2	57	90	110	90	57	3
4	1	17	57	90	110	90	4
4	2	17	57	90	110	90	4
5	1	-29	17	57	90	110	5
5	2	-29	17	57	90	110	5

The Number You Received is 4

Performance Difference is 0

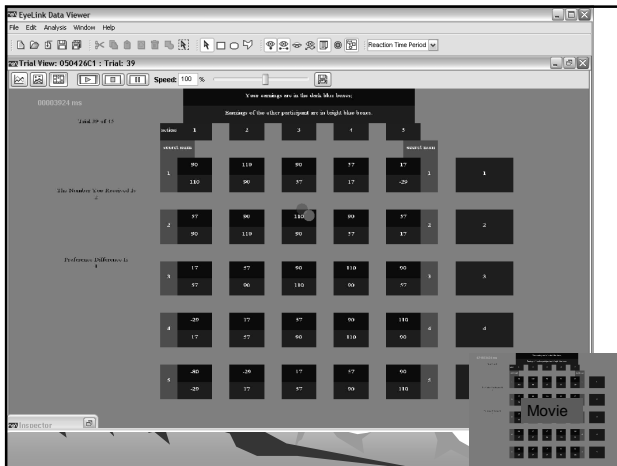


Eyetracking and Pupil Dilation

- Mount Subjects with Eyelink II Eyetracker



- Observe Eye Movements (Info. Search)
- Observe Pupil Dilation (Arousal, Stress, Cognitive Difficulty, ...)



Why Eyetracking?

- Observe “Unobservable Variables”
 - Information acquisition (“by-products”)
 - Cognitive difficulty (“energy consumption”)
- Make Better Predictions
 - Lookups and pupil dilation may explain behavior (beyond financial incentives)
- Much Cheaper than fMRI
 - Costs: One eyetracker = one fMRI study
- Pupil dilation already used in lie-detection

Past Use of Eye/mouse-tracking

- Look at critical parameters?
 - Johnson, Camerer et al., J Econ Theory 02
 - Alternating-offer bargaining (“shrinking pie”)
 - Not looking ahead (future stage pie size)
- Look at other’s payoff?
 - Costa-Gomes et al., Econometrica 01
 - Distinguish L1 and D1 by lookup at other’s payoffs
- Direct search strategy measures expertise
 - Hunton and McEwen, Accounting Review 97

Experimental Design

- Subjects were either senders or receivers throughout the experiments
- 3 practice rounds, 45 paid rounds
- Bias $b=0, 1, 2$ [1/3 each or (0.2, 0.4, 0.4)]
- Caltech students recruited via Caltech’s Social Sciences Experimental Lab (SSEL)
- Two designs:
 - Display Bias vs. Hidden Bias

Display Bias vs. Hidden Bias

Display Bias Design

1. Two subjects faced each other 45 rounds
 - Repeated game effect
2. Bias is sufficient
 - No need to look at the payoff table
3. Payoffs are the same
 - Memory effect?

Hidden Bias Design

1. Three pairs randomly matched in 45 rounds
 - Minimize repeated game effect
2. Sender can't see Bias
 - Force to look at payoffs
3. Payoffs perturbed with noise $\sim\{-4, \dots, +4\}$
 - Vary across rounds

Research Question

- What is the behavior (choices)?
 - How does a Level-k model of heterogeneous beliefs explain this data?
- How does subject behavior (choices) match with eyetracking lookup patterns and pupil dilation?
 - Do they support level-k model assumptions?
- Can we predict lies before they happen?

Main Results

1. Choices reproduce theoretical comparative statics; Overcommunication (when bias = 2)
 - A Level-k model explains the choices
2. Lookup results justify level-k assumptions
 - Attention to structure
 - Self-centeredness
 - Incorrect beliefs
 - Strategizing from a truth-telling anchor
3. PDR: The more you lie, the bigger your pupil
4. Predict true state with lookups and message

Choices: Theoretical Predictions

- Game theory:
 - b=0 truth-telling; if S, announce $M=S$
 - b=1 S=1 should announce $M=\{1\}$
S=2,3,4,5, should announce $M=\{2,3,4,5\}$
 - b=2 "babbling", should announce $M=\{1,2,3,4,5\}$
- Comparative Statics with respect to bias b:
 - Information transmission decreases with b

Choices: Theoretical Predictions

- Level-k Model:
 - Start with anchor type L0; higher types best respond (BR) to lower types
- L0:
 - L0 sender: Tell the truth
 - L0 receiver: Follow message (BR to L0 sender)
- L1:
 - L1 sender: Inflate message (BR to L0 receiver)
 - L1 receiver: Discount message (BR to L1 sender)

Choices: Theoretical Predictions

- L2:
 - L2 sender: BR to L1 receiver
 - L2 receiver: BR to L2 sender
- Eq: (=L3 in this game)
 - Eq sender: BR to L2 receiver
 - Eq receiver: BR to Eq sender
- SOPH:
 - BR to the empirical distribution of opponent behavior (know distribution of types...)

Choices: Theoretical Predictions

- In this game, L3 and above are all Eq types
 - May not be true in general
- Comparative Statics with respect to bias b:
 - Information transmission decreases with b
- However, even when standard game theory predicts babbling (b=2), the Level-k model still allows positive information transmission
 - Lower types still send informative messages

Choices: Results

- Aggregate information transmission decreases as bias b increase
 - Correlation between (S, M), (M, A), and (S, A)
 - Receiver Payoffs (“Economic value”, but can be game specific)
- Individual choices are consistent with a Level-k model [Display Bias / Hidden Bias]
 - Hidden Bias: (L0,L1,L2,Eq,SOPH) = (1,4,3,4,4)
 - Display Bias: (L0,L1,L2,Eq,SOPH) = (4,3,4,1,1)

Table 2: Information Transmission [Display Bias / Hidden Bias]

BIAS	r(S, M)	r(M, A)	r(S, A)	Predicted r(S, A)
0	0.99	1.00	0.99	1.00
	0.93	0.92	0.86	
1	0.73	0.74	0.72	0.65
	0.64	0.71	0.49	
2	0.63	0.57	0.59	0.00
	0.34	0.58	0.32	

Table 3: Sender & Receiver's Payoffs [Display Bias / Hidden Bias]

BIAS	u_S (std)	u_R (std)	Predicted u_R (std)
0	109.14 (4.07)	109.14 (4.07)	110.00 (0.00)
	101.30 (17.28)	101.27 (17.69)	
1	93.35 (20.75)	94.01 (19.86)	91.40 (19.39)
	73.28 (37.46)	86.88 (27.59)	
2	41.52 (49.98)	85.52 (25.56)	80.80 (20.76)
	43.31 (52.79)	80.55 (27.57)	

Figure 1: Raw Data Pie Chart (b=0) [Hidden Bias]

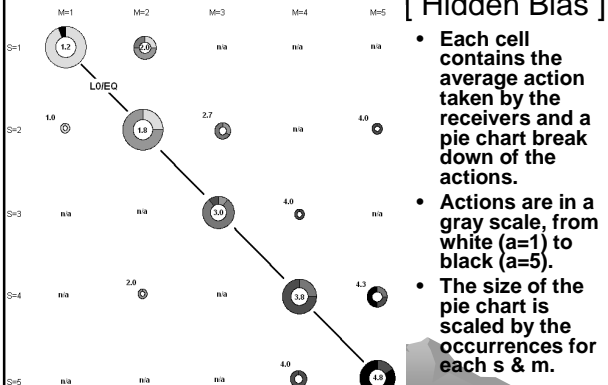
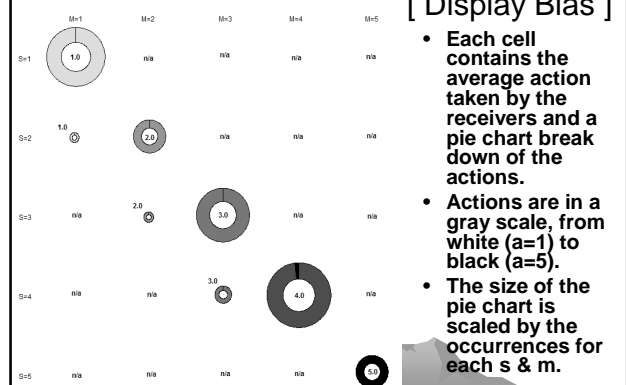


Figure S2: Raw Data Pie Chart (b=0) [Display Bias]



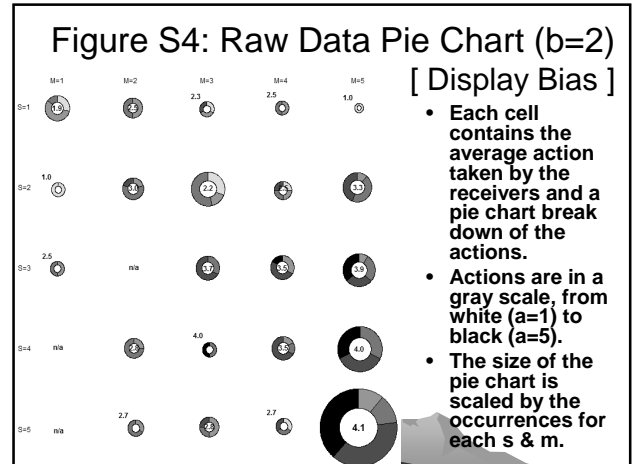
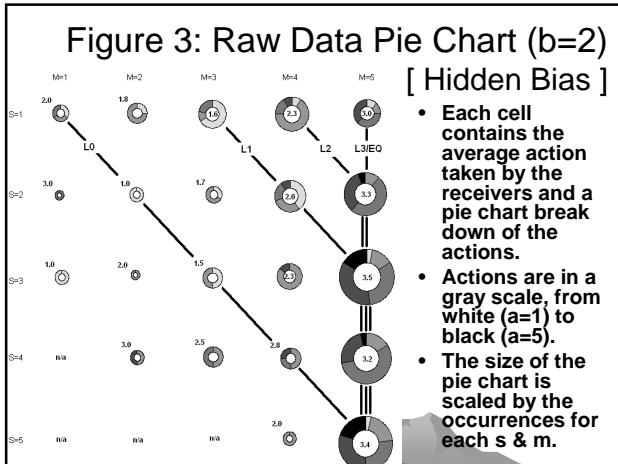
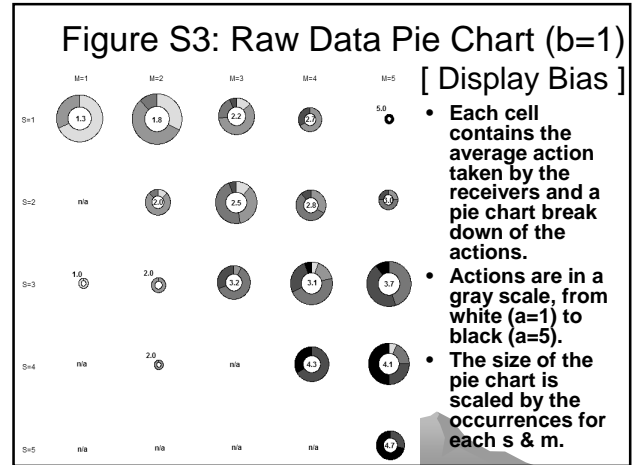
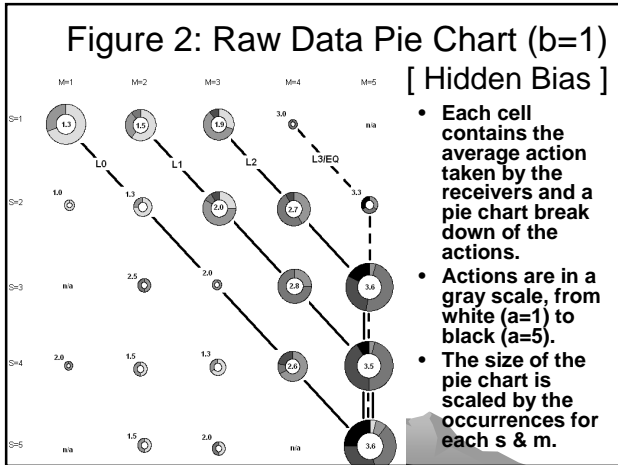


Table 4: Level-k Classification Results [Display Bias / Hidden Bias]

Level	Display Bias	Hidden Bias
L0	#1, #2, #3, #7	#6-3
L1	#4, #9, #10	#1-2, #2-1, #4-1, #5-1
L2	#8, #11, #12	#1-3, #5-2, #6-1
Eq	#6	#2-2, #3-3, #4-3, #5-3
SOPH	#5	#1-1, #2-3, #3-2, #6-2

Note: #3-1 is unclassified; #4-2 dropped due to technical difficulty.

Table 5: Sender Lookup Time (sec.) [Display Bias / Hidden Bias]

BIAS	Response Time		State	Bias	Sender Payoffs	Receiver Payoffs	Sender-to-Receiver Ratio
	1-15	31-45					
0	5.42	2.39	0.65	0.41	0.73	0.27	2.70
	9.78	7.24	0.83	-	2.93	1.71	1.71
1	7.92	5.44	1.47	0.99	2.29	1.05	2.18
	11.77	8.76	0.81	-	3.80	2.66	1.43
2	9.73	8.12	1.72	1.52	3.03	1.50	2.02
	16.84	8.99	0.91	-	4.67	3.26	1.43
all	8.07	5.25	1.34	1.02	2.14	1.00	2.14
	13.47	8.52	0.86	-	3.99	2.72	1.47

Table 6: Lookup Time per Row (sec.)
[Display Bias / Hidden Bias]

BIAS	True State Rows	Other Rows	True-to-Other Ratio
0	0.54	0.11	4.91
	2.76	0.47	5.87
1	2.06	0.32	6.44
	3.88	0.64	6.06
2	2.24	0.57	4.28
	4.29	0.91	4.71
all	1.71	0.36	4.75
	3.83	0.72	5.32

Figure 4: Icon Graph (b=1), Hidden Bias

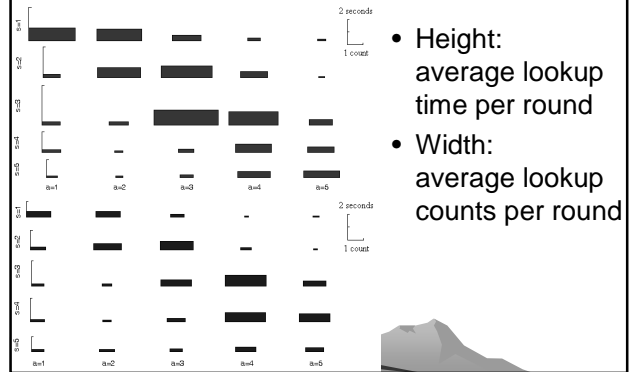


Figure S6: Icon Graph (b=1), Display Bias

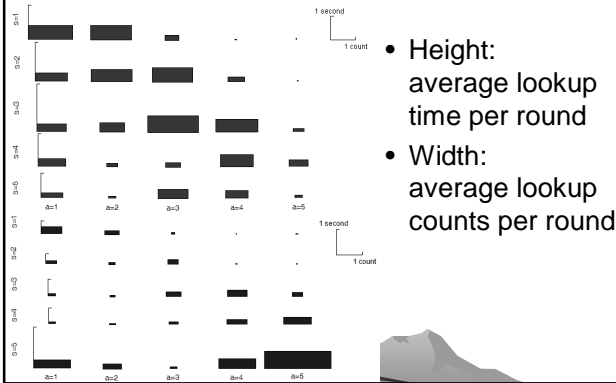


Figure 5: Icon Graph (b=2), Hidden Bias

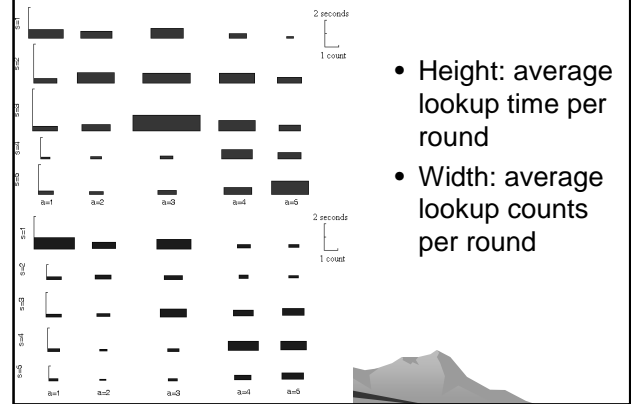
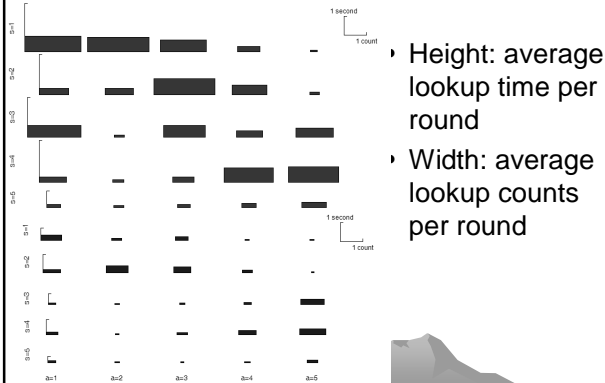


Figure S7: Icon Graph (b=2), Display Bias



$$\begin{aligned}
 PUPIL_i = & \alpha + \sum_{b=1}^2 \beta_{1b} \cdot LIE_SIZE \cdot BIAS_b \\
 & + \sum_{b \neq 2} \beta_{2b} \cdot BIAS_b + \sum_{s \neq 3} \beta_{3s} \cdot STATE_s + \sum_{k \neq 6} \alpha_k \cdot SUBJ_k \\
 & + \sum_{k=1}^K (\gamma_{k,1} ROUND \cdot SUBJ_k + \gamma_{k,2} ROUND^2 \cdot SUBJ_k) + \epsilon
 \end{aligned}$$

- $PUPIL_k$ = Average pupil size at time frame i
- LIE_SIZE = | state – message |
- $BIAS_b$ = Dummy for different biases
- $STATE_s$ = Dummy for different states
- $SUBJ_k$ = Dummy variable for subject k
- $ROUND$ = Number or round

Pupil Dilation

- Pinocchio's Nose?
 - The more you lie, the wider your pupil
- Subjects' pupil sizes before and after their decision are larger if they exaggerate more
 - Random effect with “robust” standard errors (correct serial correlation and heteroscedasticity)
- More so for Display Bias than Hidden Bias
 - Maybe “Hidden Bias” raises “baseline” dilation...

Table 7: Pupil Size Regressions [Display Bias]

PUPIL _t	secs	(-1.2, -0.8)	(-0.8, -0.4)	(-0.4, 0.0)	(0.0, 0.4)	(0.4, 0.8)
Constant	α	99.59 (2.45)	99.78 (2.41)	104.62 (2.19)	111.81 (1.84)	109.95 (2.07)
LIE_SIZE	β_{10}	1.20 (3.21)	6.41 (6.38)	3.92 (3.06)	-3.91 (2.76)	0.58 (7.36)
* BIAS						
interactions	β_{11}	2.79* (1.19)	3.40** (1.17)	3.28** (0.97)	4.55*** (0.86)	4.20*** (0.73)
	β_{12}	3.49*** (0.99)	3.71*** (0.98)	3.04*** (0.84)	2.90** (0.87)	3.28** (0.90)

Note: t-Test p-values lower than *5%, ** 1%, and *** 0.1%.

Table 7: Pupil Size Regressions [Hidden Bias]

PUPIL _t	secs	(-1.2, -0.8)	(-0.8, -0.4)	(-0.4, 0.0)	(0.0, 0.4)	(0.4, 0.8)
Constant	α	107.27 (2.81)	108.03 (2.55)	106.19 (2.57)	109.56 (2.05)	108.67 (2.16)
LIE_SIZE	β_{10}	2.83 (1.85)	2.36 (2.22)	3.07 (2.46)	5.35** (1.16)	5.57* (2.19)
* BIAS						
interactions	β_{11}	-1.02 (1.26)	-0.46 (1.31)	-0.36 (1.28)	2.16^ (1.21)	2.64* (1.15)
	β_{12}	2.06* (0.86)	1.52^ (0.79)	1.47* (0.75)	1.83* (0.75)	2.00** (0.74)

Note: t-Test p-values lower than ^10%, *5%, ** 1%, and *** 0.1%.

Predicting True States

- We can “figure out” the true state using message, lookup patterns and pupil dilation
 - Message
 - Most lookup rows (self and other)
- We can increase actual receiver payoffs using these predictions
 - Paired: Improve by 6-8% (86→92, 93→100)
 - Random Match: Improve by 16–21% (80, 87→98~101)
(As high as “actual payoffs” when b=0!)

Table 8: Predicting True States

	Y	Display Bias	Much Stronger Lookup Effect
MESSAGE*B	Message sent	0.64* (0.22)	0.46** (0.12)
MESSAGE*B	Message sent	0.91** (0.23)	0.42** (0.09)
ROW _{self} * BIAS=1	β_{31}	0.98** (0.21)	1.07** (0.24)
ROW _{self} * BL	Lookups	1.00** (0.27)	1.72* (0.20)
ROW _{other} * BIAS=1	β_{41}	0.25 (0.16)	1.27** (0.22)
ROW _{other} * BIAS=2	β_{42}	0.39* (0.17)	0.44** (0.15)
		Actual	Hold-out
Aver. predicted u_R (b=1)		93.4 (22.3)	100.7* (2.4)
			87.5 (28.8)
Aver. predicted u_R (b=2)		86.2 (23.8)	91.8* (3.4)
			80.9 (26.9)
			101.7** (2.1)
			98.0** (2.2)

16-21% Increase; Near Actual Payoff at b=0 (=100.9)

Conclusion

- More deception as bias increases
- But “not enough” deception (as equilibrium theory)
- Level-k model predicts individual behavior
- Look at “true state rows” in the payoff table
- Pupil dilation correlated with deception
- Prediction gain: receiver payoffs increase by 16-21% (6-8% for Display Bias)
 - Hidden Bias: close to “actual payoffs” at b=0

Conclusion

- Friedrich Nietzsche (1878)
 - “Why do almost all people tell the truth in ordinary everyday life? --Certainly not because a god has forbidden them to lie. The reason is, firstly because it is easier; for lying demands invention, dissimulation and a good memory.” (Human, All Too Human, II.54)
- Mark Twain
 - “If you tell the truth, you don’t have to remember anything.”
- Can lie-detection be done?

