

Econ 219B
Psychology and Economics: Applications
(Lecture 6)

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Outline

1. Social Preferences: Introduction
2. Social Preferences: Gift Exchange
3. Social Preferences: Workplace
4. Social Preferences: Charitable Giving I
5. Methodology: Field Experiments

1 Social Preferences: Introduction

- Laboratory data from ultimatum, dictator, and trust games
—> Clear evidence of social preferences
- **Fehr-Schmidt (QJE, 1999)** and **Charness-Rabin (QJE, 2002)**
- Simplified model of preferences of B when interacting with A :

$$U_B(\pi_A, \pi_B) \equiv \rho\pi_A + (1 - \rho)\pi_B \text{ when } \pi_B \geq \pi_A.$$
$$U_B(\pi_A, \pi_B) \equiv \sigma\pi_A + (1 - \sigma)\pi_B \text{ when } \pi_B \leq \pi_A.$$

- Captures:
 - baseline altruism (if $\rho > 0$ and $\sigma > 0$)
 - differentially so if ahead or behind ($\rho > \sigma$)

- Example: Dictator Game. Have \$10 and have to decide how to share
- **Forsythe et al. (GEB, 1994)**: sixty percent of subjects transfers a positive amount.
- Transfer \$5 if

$$\rho 5 + (1 - \rho)5 = 5 \geq \rho 0 + (1 - \rho)10 \rightarrow \rho \geq 1/2 \text{ and}$$

$$\sigma 5 + (1 - \sigma)5 \geq \sigma 10 + (1 - \sigma)0 \rightarrow \sigma \leq 1/2$$

- Transfer \$5 if $\rho \geq .5 \geq \sigma$

- Taking this to field data? Hard
- **Charitable giving.**
- Qualitative Patterns consistent overall with social preferences:
 - 240.9 billion dollars donated to charities in 2002 (Andreoni, 2006)
 - 2 percent of GDP
- Quantitative patterns, however: Hard to fit with models of social preferences from the lab

- Issue 1:

- Person B with disposable income M_B meets needy person A with income $M_A < M_B$
- Person B decides on donation D
- Assume parameters $\rho \geq .5 \geq \sigma$
- This implies $\pi_A^* = \pi_B^* \rightarrow M_B - D^* = M_A + D^* \rightarrow D^* = (M_B - M_A) / 2$
- Wealthy person transfers half of wealth difference!
- Clearly counterfactual

- Issue 2.

- Lab: Person *A* and *B*.

- Field: Millions of needy people. Public good problem

- Issue 3.

- Lab: Forced interaction.

- Field: Sorting – can get around, or look for, occasions to give

- In addition to payoff-based social preferences, intentions likely to matter
- ρ and σ higher when B treated nicely by A
- Positive reciprocity and negative reciprocity
- More evidence of the latter in experiments

- Other field applications we do not analyze

1. Pricing. When are price increases acceptable?

- **Kahneman, Knetsch and Thaler (1986)**

- Survey evidence

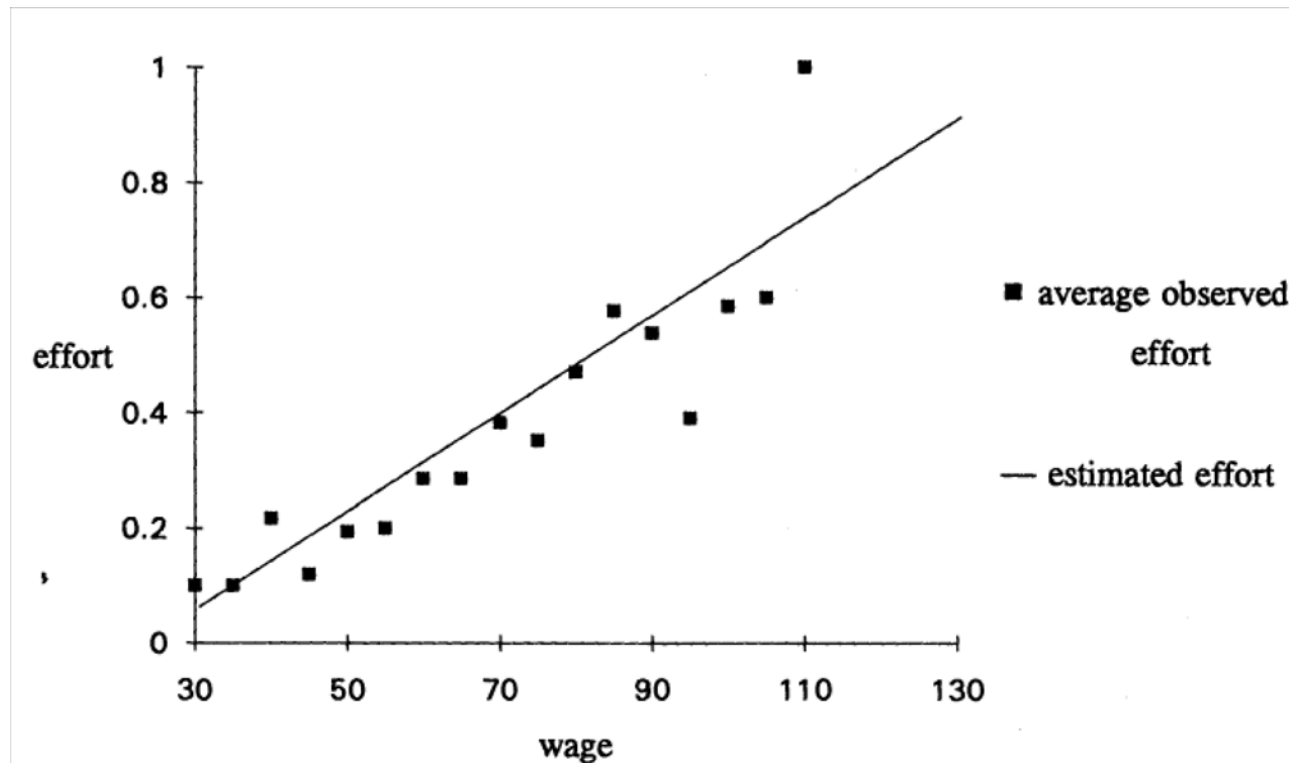
- Effect on price setting

2. Wage setting. Fairness toward other workers → Wage compression

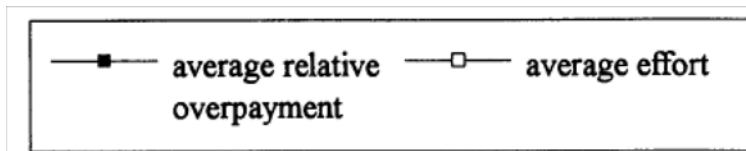
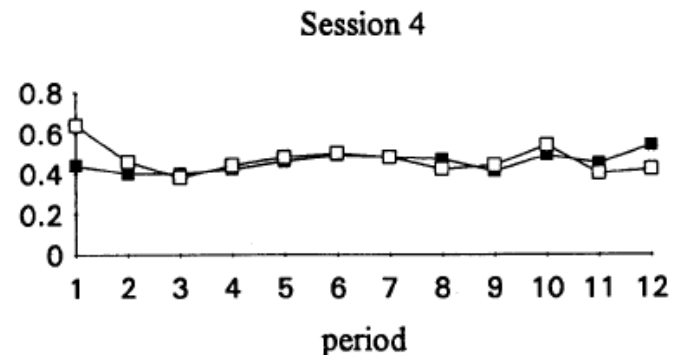
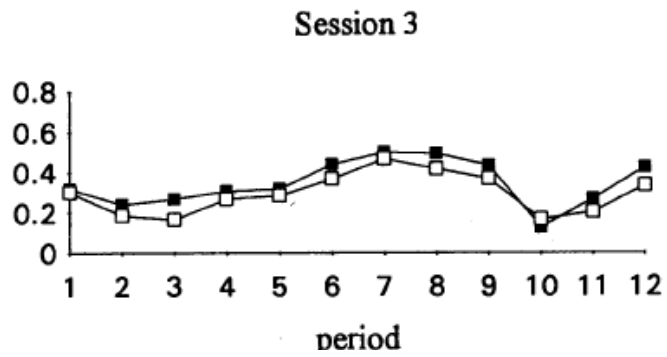
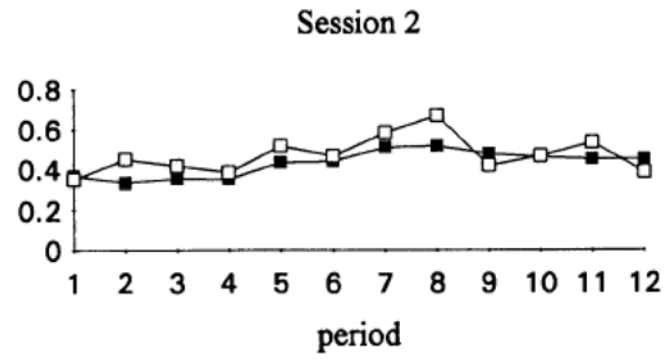
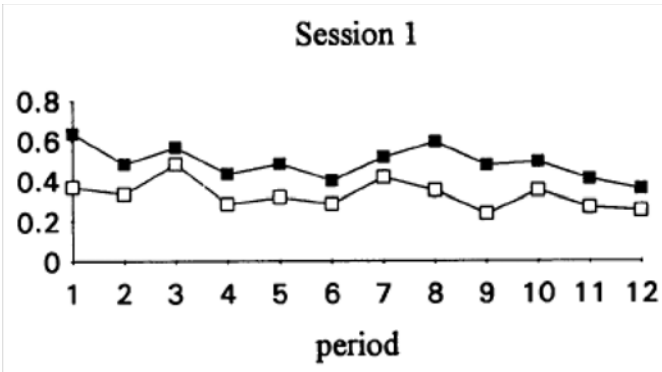
2 Social Preferences: Gift Exchange

- Laboratory evidence: **Fehr-Kirchsteiger-Riedl (QJE, 1993)**.
 - 5 firms bidding for 9 workers
 - Workers are first paid $w \in \{0, 5, 10, \dots\}$ and then exert effort $e \in [.1, 1]$
 - Firm payoff is $(126 - w) e$
 - Worker payoff is $w - 26 - c(e)$, with $c(e)$ convex (but small)
- Standard model: $w^* = 30$ (to satisfy IR), $e^*(w) = .1$ for all w

- Findings: effort e increasing in w and $Ew = 72$



- These findings are stable over time



- Where evidence of gift exchange in the field?
- **Falk (EMA, 2008)** — field experiment in fund-raising
 - 9,846 solicitation letters in Zurich (Switzerland) for Christmas
 - Target: Schools for street children in Dhaka (Bangladesh)
 - 1/3 no gift, 1/3 small gift 1/3 large gift
 - Gift consists in postcards drawn by kids

Appendix: An example of the included postcards



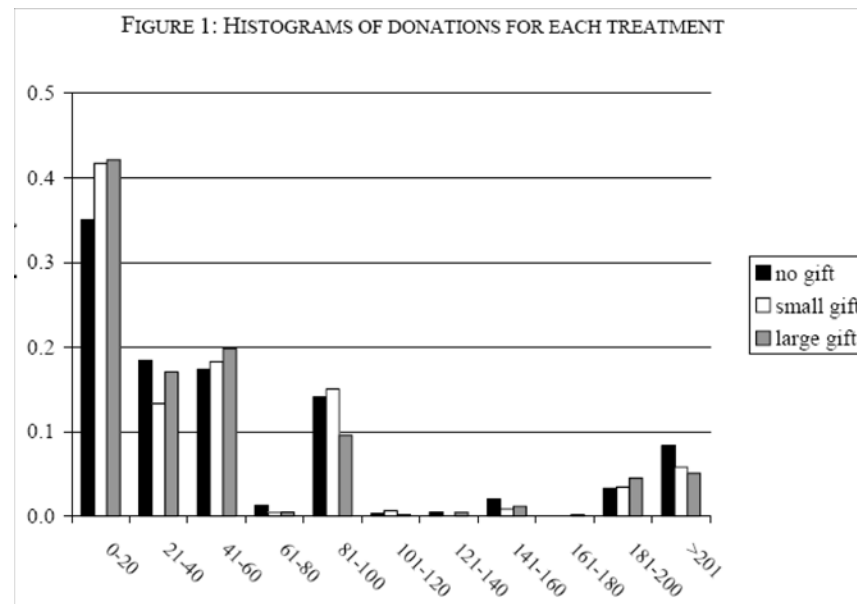
- Short-Run effect: Donations within 3 months

TABLE 1: DONATION PATTERNS IN ALL TREATMENT CONDITIONS

	No gift	Small gift	Large gift
Number of solicitation letters	3,262	3,237	3,347
Number of donations	397	465	691
Relative frequency of donations	0.12	0.14	0.21

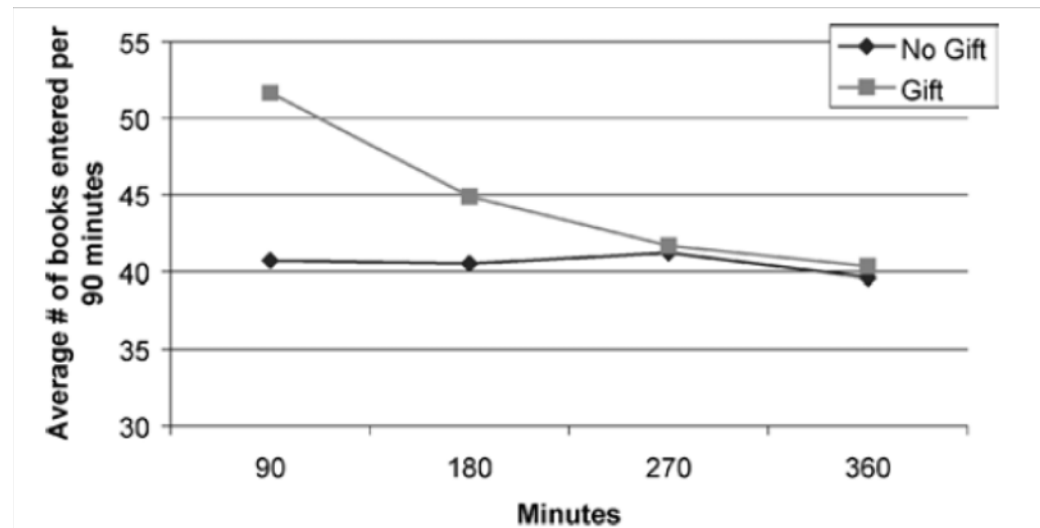
- Large gift leads to doubling of donation probability
- Effect does not depend on previous donation pattern (donation in previous mailing)
- Note: High donation levels, not typical for US

- Small decrease in average donation, conditional on donation (Marginal donors adversely selected, as in 401(k) Active choice paper)

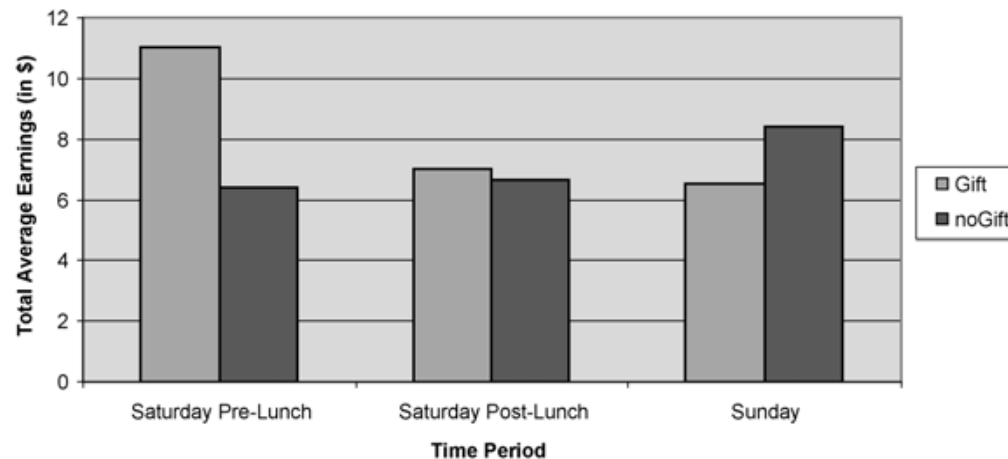


- Limited intertemporal substitution. February 2002 mailing with no gift. Percent donation is 9.6 (control), 8.9 (small gift), and 8.6 (large gift) (differences not significant)

- **Gneezy-List (EMA, 2006)** → Evidence from labor markets
- *Field experiment 1.* Students hired for one-time six-hour (typing) library job for \$12/hour
 - No Gift group paid \$12 ($N = 10$)
 - Gift group paid \$20 ($N = 9$)



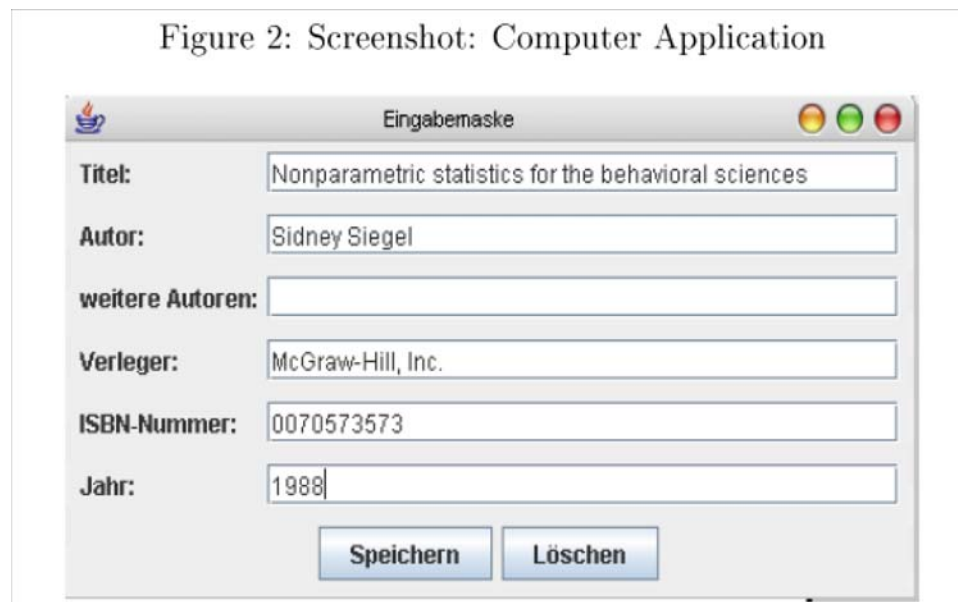
- *Field experiment 2.* Door-to-Door fund-raising in NC for one-time weekend for \$10/hour
 - Control group paid \$10 ($N = 10$)
 - Treatment group paid \$20 ($N = 13$)



- Note: Group coming back on Sunday is subset only (4+9)

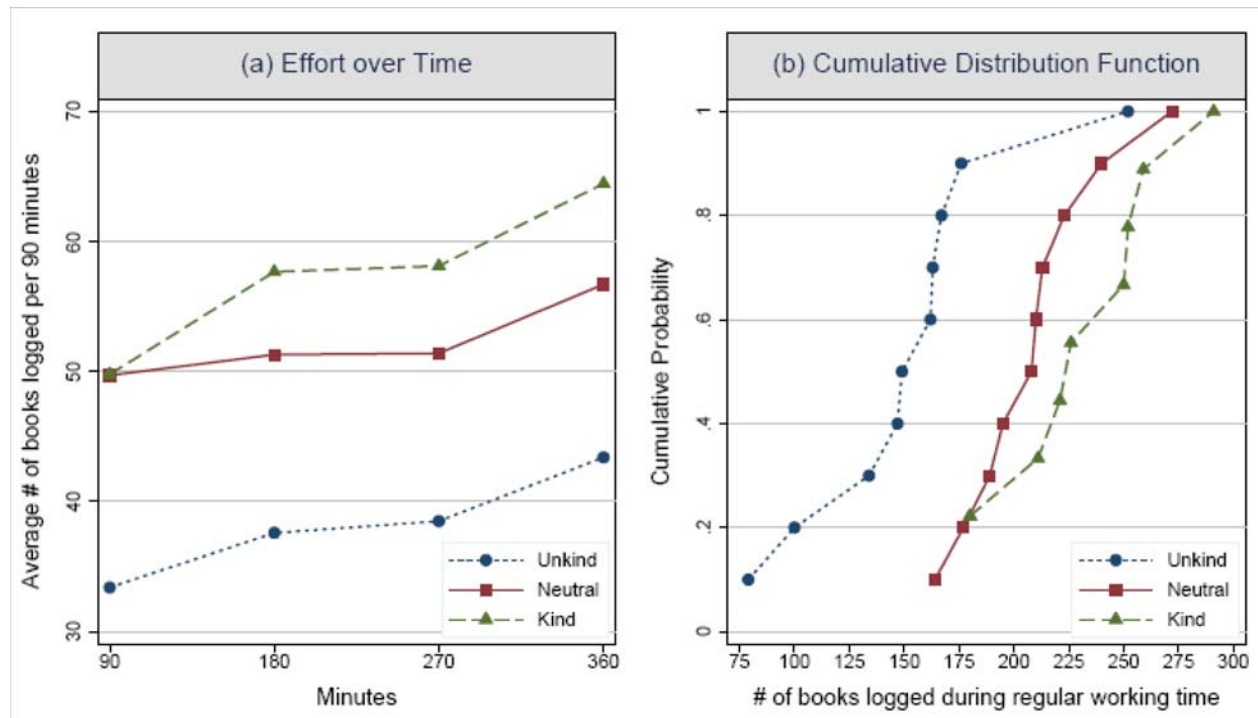
- Evidence of reciprocity, though short-lived
- Issue: These papers test only for positive reciprocity
- Laboratory evidence: negative reciprocity stronger than positive reciprocity
- More difficult to test for negative reciprocity
- Can say that pay is random and see what happens to (randomly) lower paid people

- **Kube-Marechal-Puppe (2007).**
- Field Experiment: Hire job applicants to catalog books for 6 hours



- Announced Wage: '*Presumably*' 15 Euros/hour
 - Control ($n = 10$). 15 Euros/hour
 - Treatment 1 (Negative Reciprocity, $n = 10$). 10 Euros/hour (No one quits)
 - Treatment 2 (Positive Reciprocity, $n = 9$). 20 Euros/hour
- Offer to work one additional hour for 15 Euros/hour

- Result 1: Substantial effect of pay cut
- Result 2: Smaller effect of pay increase
- Result 3: No decrease over time



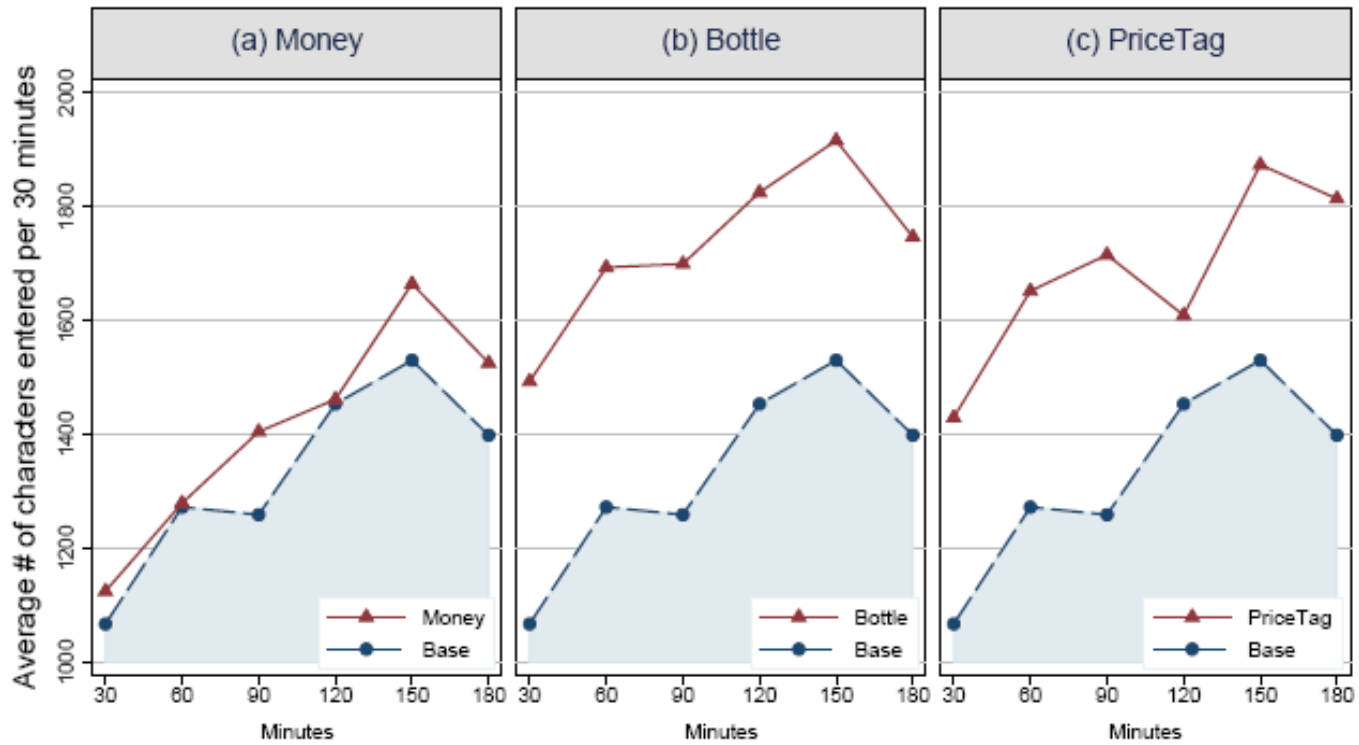
- Notice: No effect on quality of effort (no. of books incorrectly classified)
- Finding consistent with experimental results:
 - Positive reciprocity weaker than negative reciprocity
- Final result: No. of subjects that accept to do one more hour for 15 Euro:
 - 3 in Control, 2 in Pos. Rec., 7 in Neg. Rec.
 - Positive Reciprocity does not extend to volunteering for one more hour

- **Kube-Marechal-Puppe (2008).**
- Field Experiment 2: Hire job applicants to catalog books for 6 hours
- Announced Wage: 12 Euros/hour for 3 hours=36
 - Control ($n = 17$). 36 Euros
 - Treatment 1 (Positive Reciprocity, Cash, $n = 16$). $36 + 7 = 43$ Euros
 - Treatment 2 (Positive Reciprocity, Gift, $n = 15$). 36 Euros plus Gift of Thermos
 - Treatment 3 – Same as Tr. 2, but Price Tag for Thermos

- What is the effect of cash versus in-kind gift?



- Result 1: Small effect of 20% pay increase
- Result 2: Large effect of Thermos \rightarrow High elasticity, can pay for itself
- Result 3: No decrease over time



- Explanation 1. Thermos perceived more valuable
 - –> But Treatment 3 with price tag does not support this
 - Additional Experiment:
 - * At end of (unrelated) lab experiment, ask choice for 7 Euro or Thermos
 - * 159 out of 172 subjects prefer 7 Euro
- Explanation 2. Subjects perceive the thermos gift as more kind, and respond with more effort
- Survey: Ask which is kinder? Thermos rated higher in kindness than 7 Euro

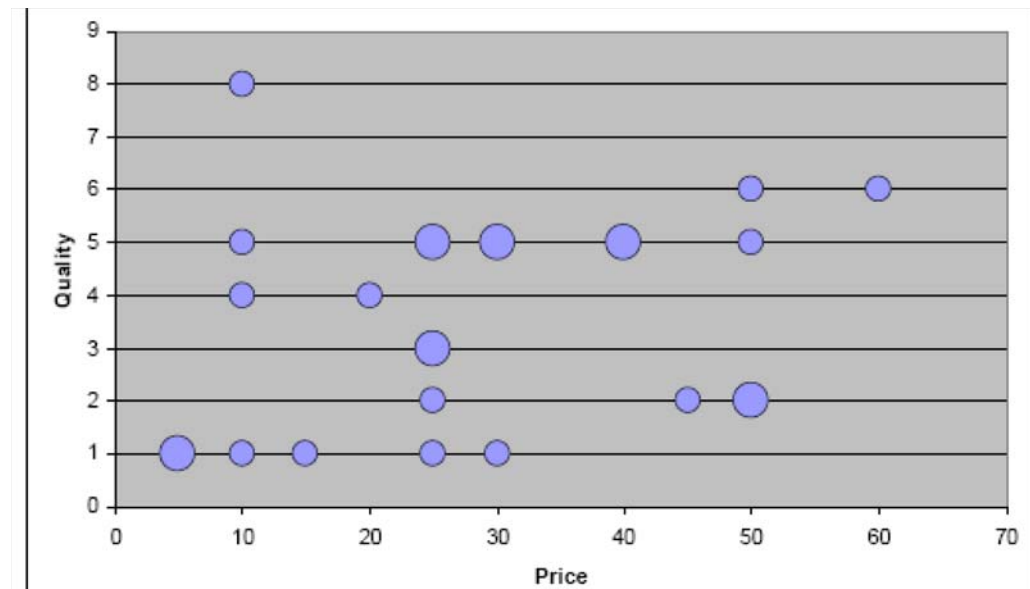
- Wat is missing from these paper? **Model**
- Fehr, Kirchsteiger, and Riedl (QJE, 1993) - Two main model-based explanations:
 - *Reciprocity* (Rabin, 1993; Dufwenberg and Kirchsteiger, 2003): Worker is nice towards firm because firm showed nice intentions
 - *Inequity Aversion* (Fehr and Schmidt, 1999): Worker puts effort because firm had fallen behind in payoffs by putting effort
- Model for Gneezy and List (2006) and follow-up work?
 - Inequity aversion does not predict gift exchange in the field

- Intuition: Firm does not fall behind the worker just because of a pay increase
- Hence, in the field gift exchange, when occurs, is due to reciprocity, not inequity aversion
- Model would also make finer predictions
- Moreover, model would also give a sense of magnitudes
 - How much reciprocity does gift exchange indicate?
 - What are welfare effects of gift?

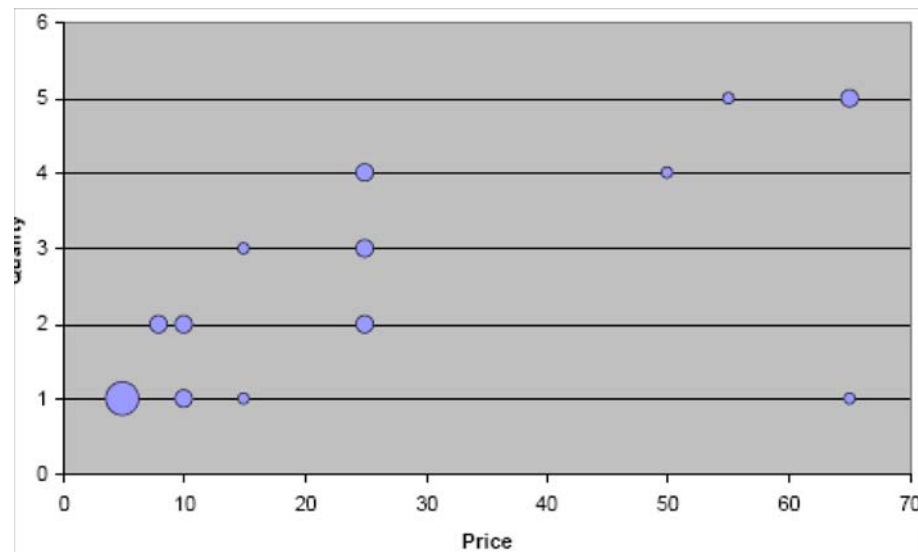
- Key unobservable is *cost of effort*: How costly is it to increase effort at margin?
- If not costly \rightarrow minuscule reciprocity can generate gift exchange
- If costly \rightarrow reciprocity needs to be sizeable
- Additional treatments varying announced pay (but holding sample constant) can identify elasticity
- Frontier in this literature

- **List (JPE, 2006)**. Test of social preferences from sellers to buyers
- Context: sports card fairs \rightarrow Buyers buying a particular (unrated) card from dealers
- Compare effect of laboratory versus field setting
- *Treatment I-R*. Clever dual version to the **Fehr-Kirchsteiger-Riedl (1993)** payoffs
 - Laboratory setting, abstract words
 - Buyer pay $p \in \{5, 10, \dots\}$ and dealer sells card of quality $q \in [.1, 1]$
 - Buyer payoff is $(80 - p)q$
 - Dealer payoff is $p - c(q)$, with $c(q)$ convex (but small)
- Standard model: $p^* = 5$ (to satisfy IR), $q^*(p) = 0.1$ for all p

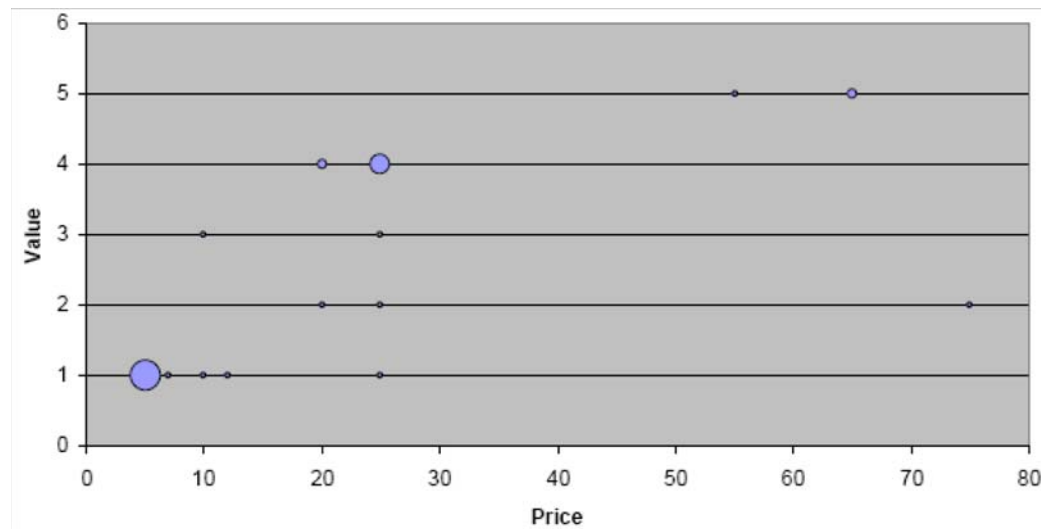
- Effect: Substantial reciprocity
 - Buyers offer prices $p > 0$
 - Dealers respond with increasing quality to higher prices



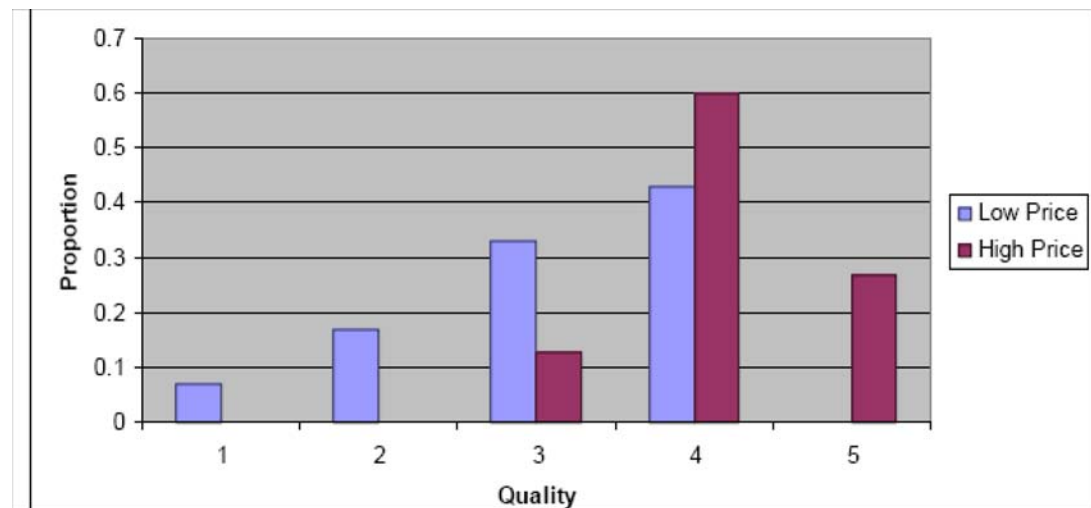
- *Treatment I-RF*. Similar result (with more instances of $p = 5$) when payoffs changed to
 - Buyer payoff is $v(q) - p$
 - Dealer payoff is $p - c(q)$, with $c(q)$ convex (but small)
 - $v(q)$ estimated value of card to buyer, $c(q)$ estimate cost of card to dealer



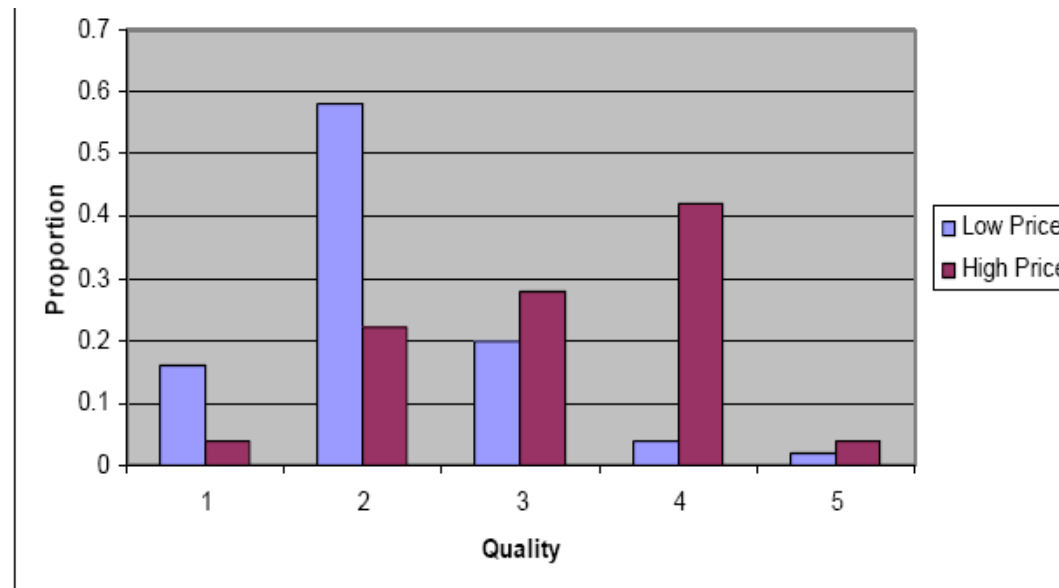
- *Treatment II-C*. Same as Treatment I-RF, except that use context (C) of Sports Card
- Relatively similar results



- *Treatment II-M* → Laboratory, real payoff (for dealer) but...
 - takes place with face-to-face purchasing
 - Group 1: Buyer offers \$20 for card of quality PSA 9
 - Group 2: Buyer offers \$65 for card of quality PSA 10
 - Substantial “gift exchange”



- *Treatment III* → In field setting, for real payoffs (for dealer)
 - Group 1: Buyer offers \$20 for card of quality PSA 9
 - Group 2: Buyer offers \$65 for card of quality PSA 10
 - Lower quality provided, though still “gift exchange”



- However, “gift exchange” behavior depends on who the dealer is
 - Local dealer (frequent interaction): Strong “gift exchange”
 - Non-Local dealer (frequent interaction): No “gift exchange”
- This appears to be just rational behavior
- *Treatment IV.* → Test a ticket market before (*IV-NG*) and after (*IV-AG* and *IV-G*) introduction of certification
 - No “gift exchange” in absence of certification(*IV-NG*)
 - “gift exchange” only for local dealers

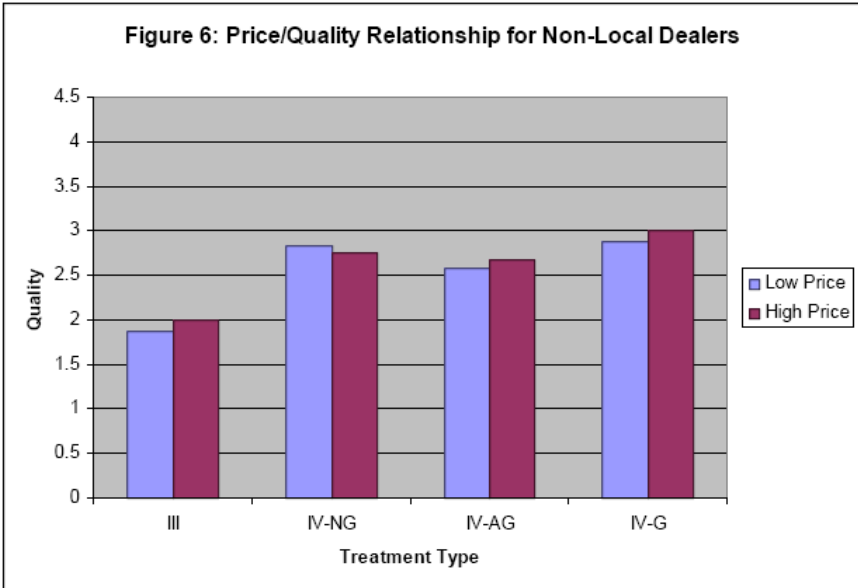
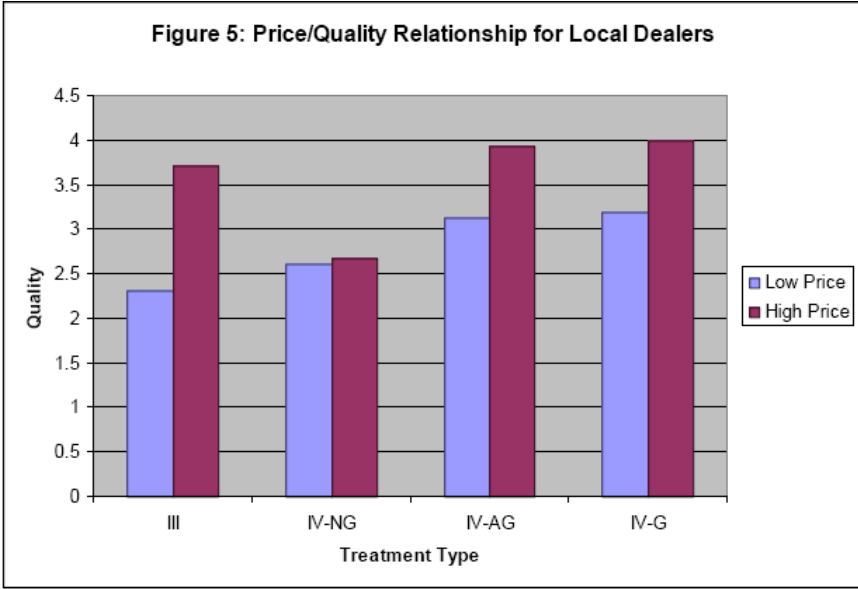


Table 1. Experimental Design

Treatment I	Treatment I-R <i>Replicate lab studies</i> $n = 25$	Treatment I-RF <i>Extend to field values</i> $n = 25$	Treatment I-RF1 <i>Extend to one-shot environment</i> $n = 27$
Treatment II	Treatment II-C <i>Adds market context</i> $n = 32$	Treatment II-MS20 <i>Adds market interaction</i> $n = 30$	Treatment II-MS65 <i>Adds market interaction</i> $n = 30$
Treatment III	Treatment IIIS20 <i>Naturally occurring sportscards</i> $n = 50$	Treatment IIIS65 <i>Naturally occurring sportscards</i> $n = 50$	
Treatment IV	Treatment IV-NG <i>Naturally occurring tickets before grading was available</i> $n = 60$	Treatment IV-AG <i>Naturally occurring tickets post-grading announcement</i> $n = 54$	Treatment IV-G <i>Naturally occurring tickets when grading service is available</i> $n = 36$

Notes: Each cell represents one (or two, in the case of Treatment IV) unique treatment. For example, Treatment I-R in row 1, column 1, denotes that 25 dealer and 25 nondealer observations were gathered to replicate the laboratory gift exchange studies in the literature.

Table 3: Marginal Effects Estimates for the Sellers' Quality^{a,b}

Variable	Treatment Type									
	I-R	I-RF	I-RF1	II-C	II-M	III	IV-NG	IV-AG	IV-G	IV-P
Price	0.05* (1.8)	0.05^ (3.3)	0.10^ (5.0)	0.06^ (4.2)	0.02^ (4.4)	0.02^ (6.6)	-0.001 (0.01)	0.02^ (2.1)	0.02 (1.1)	0.02^ (2.6)
Constant	0.6 (0.7)	-0.4 (0.7)	-0.8 (1.7)	-0.6 (1.7)	1.6^ (6.2)	0.6^ (3.1)	1.7^ (8.0)	1.6^ (5.8)	1.8^ (3.3)	1.7^ (7.3)
θ	---	\$0.72^ (3.6)	\$1.3^ (5.5)	\$0.77^ (4.2)	0.45^ (2.1)	\$0.21^ (5.0)	\$0.01 (0.3)	\$0.17 (1.1)	\$0.23 (1.1)	\$0.21^ (2.3)
Person Random Effects	YES	YES	NO	NO	YES	YES	YES	YES	YES	YES
N	25	25	27	32	60	100	60	54	36	90

^aDependent variable is the sellers' product quality given to the buyer. IV-P pools IV-AG and IV-G data. θ is the monetary gift exchange estimate, computed as $\partial v(q)/\partial P$.

^bt-ratios (in absolute value) are beneath marginal effect estimates.

^ Significant at the .05 level.

* Significant at the .10 level.

Table 4: Marginal Effects Estimates for the Sellers' Quality Split by Dealer Type^{a,b,c}

Variable	Treatment Type									
	III _L	III _N	IV-NG _L	IV-NG _N	IV-AG _L	IV-AG _N	IV-G _L	IVG _N	IV-P _L	
Price	0.03^ (8.6)	0.004 (0.7)	0.002 (0.2)	-0.005 (0.5)	0.04^ (2.1)	0.003 (0.3)	0.04^ (2.7)	0.003 (0.1)	0.04^ (4.8)	
Constant	0.6^ (4.1)	0.6^ (4.6)	1.6^ (5.0)	1.8^ (5.2)	1.7^ (5.2)	1.5^ (4.6)	1.8^ (5.0)	1.8* (1.7)	1.8^ (10.0)	
θ	\$0.31^ (5.2)	\$0.01 (0.5)	\$0.02 (0.4)	-\$0.006 (0.5)	\$0.32 (1.4)	\$0.02 (0.6)	\$0.42 (1.5)	\$0.03 (0.1)	\$0.35^ (2.1)	
Person Random Effects	YES	YES	YES	YES	YES	YES	YES	YES	YES	
N	70	30	36	24	30	24	20	16	50	

- Conclusion on gift exchange and social preferences
 - Reciprocation and gift exchange are present in field-type setting (Falk)
 - They disappear fast (Gneezy-List)...
 - ...Or maybe not (Kube et al.)
 - They are stronger on the negative than on the positive side (Kube et al.)
 - Not all individuals display them – not dealers, for example (List)
 - Laboratory settings may (or may not) matter for the inferences we derive

3 Social Preferences: Workplace

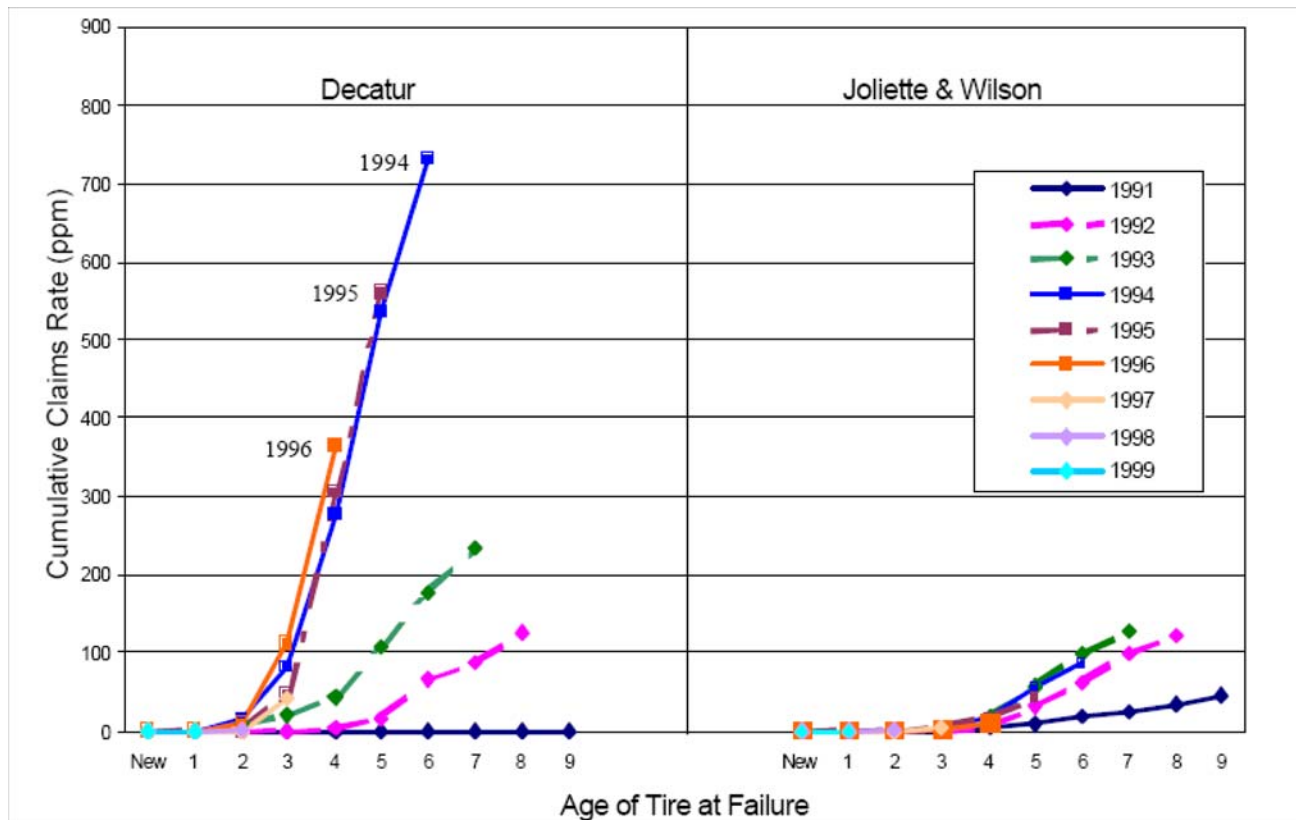
- In the workplace, do workers respond in kind to generous behavior by employers?
- Basis for some efficiency wage models
 - Natural Experiment: Krueger-Mas (2004)
 - Field Experiment on Social Preferences: Bandiera-Barankay-Rasul (2005)
 - Field Experiments on Gift Exchange: Kube-Marechel-Puppe and Gneezy-List

- **Krueger-Mas (JPE, 2004).**

- Setting:

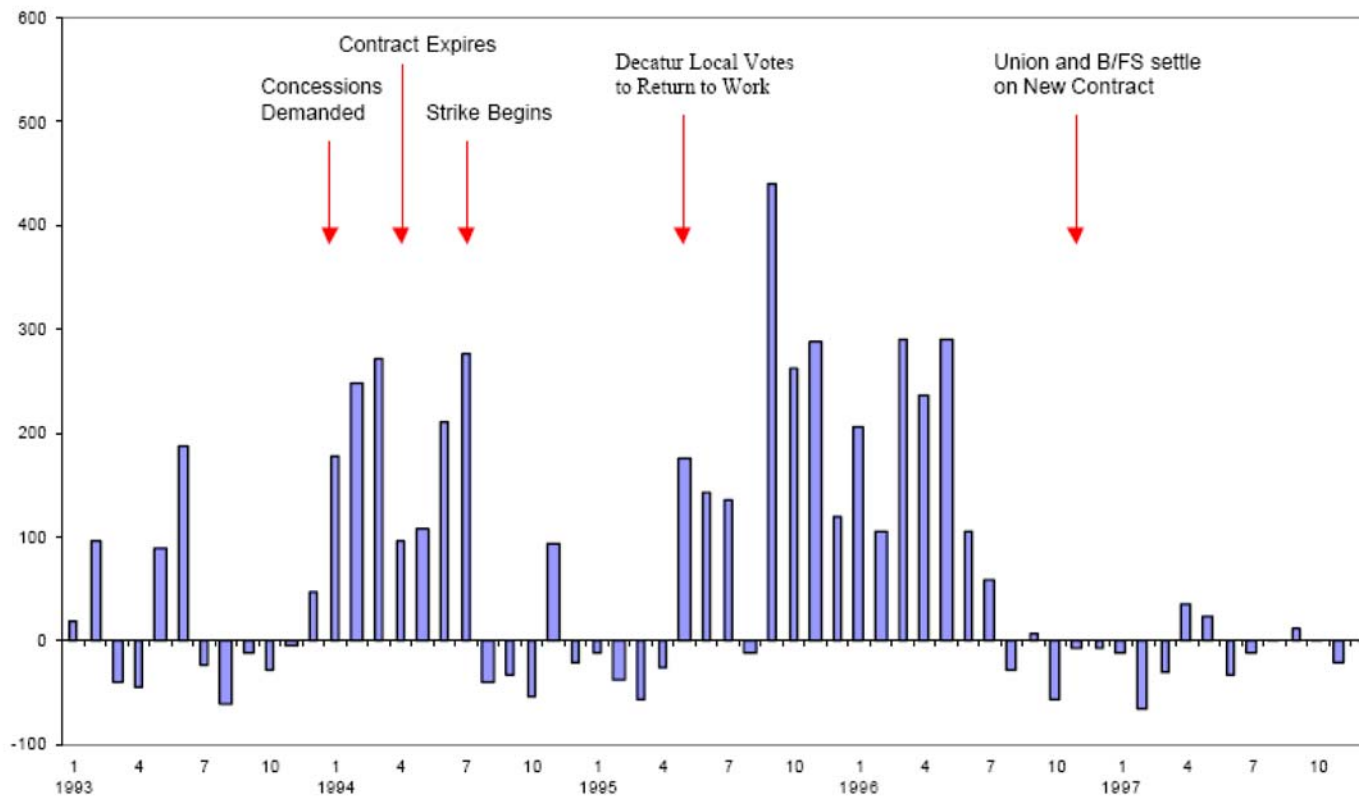
- Unionized Bridgestone-Firestone plant
- Workers went on strike in July 1994
- Replaced by replacement workers
- Union workers gradually reintegrated in the plant in May 1995 after the union, running out of funds, accepted the demands of the company
- Agreement not reached until December 1996

- Do workers sabotage production at firm?
 - Examine claims per million tires produced in plants affected
 - Compare to plant not affected by strike (Joliet & Wilson)



- Ten-fold increase in number of claims
- Similar pattern for accidents with fatalities
- Possible explanations:
 - Lower quality of replacement workers
 - Boycotting / negative reciprocity by unionized workers
- Examine the timing of the claims

Figure 8: Difference in the Number of Complaints per million Tires Produced by Month: Decatur Plant minus Joliette and Wilson Plants.



Source: Authors' calculations based on NHTSA complaints data. Records with missing data are excluded.

- Two time periods with peak of claims:
 - Beginning of Negotiation Period
 - Overlap between Replacement and Union Workers
- Quality not lower during period with replacement workers
- Quality crisis due to Boycotts by union workers
- Claims back to normal after new contract settled
- Suggestive of extreme importance of good employer-worker relations

- **Bandiera-Barankay-Rasul (QJE, 2005).**
- Test for impact of social preferences in the workplace
- Use personnel data from a fruit farm in the UK
- Measure productivity as a function of compensation scheme
- Timeline:
 - First 8 weeks of the 2002 picking season → Fruit-pickers compensated on a relative performance scheme
 - * Per-fruit piece rate is decreasing in the average productivity.
 - * Workers that care about others have incentive to keep the productivity low
 - Next 8 weeks → Compensation switched to flat piece rate per fruit
 - Switch announced on the day change took place

- Dramatic 50 percent increase in productivity

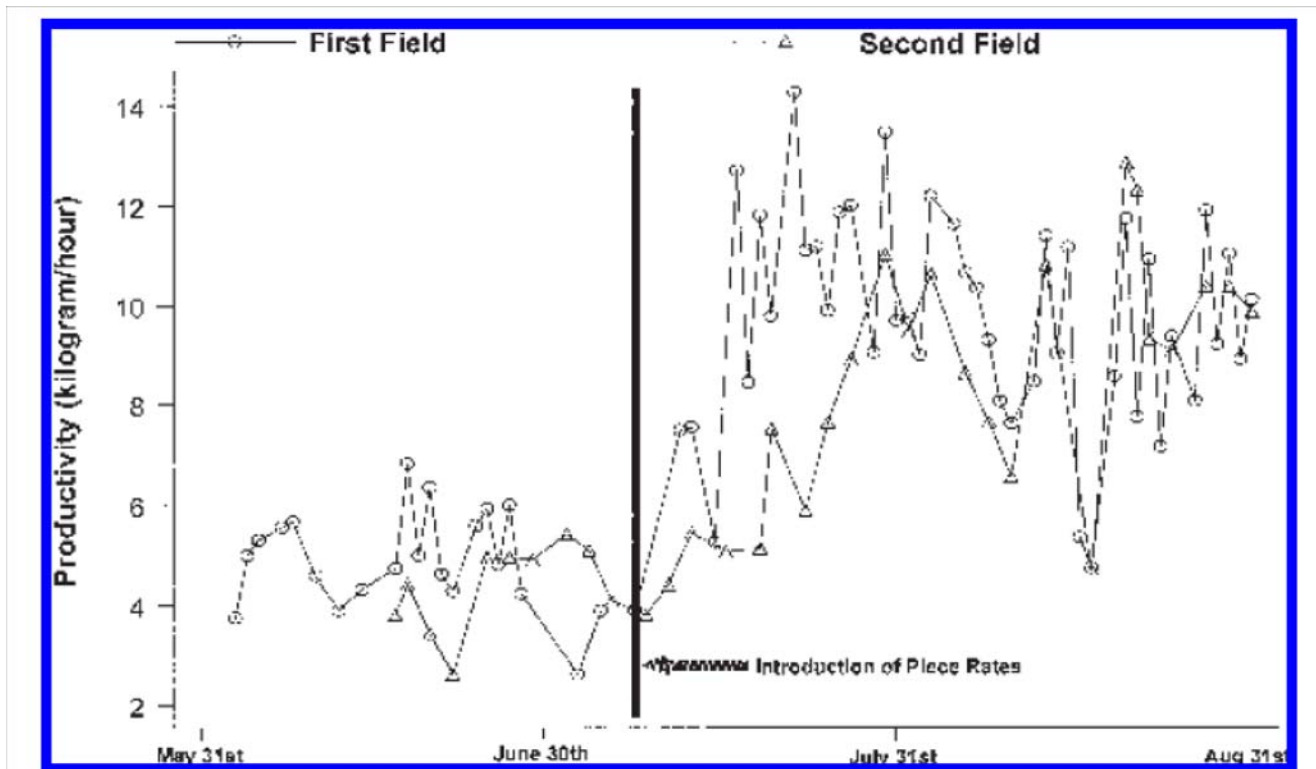


FIGURE I
Productivity (kilogram/hour) over the Season

- No other significant changes

	Relative incentives	Piece rates	Difference
Worker productivity (kg/hr)	5.01 (.243) [4.53, 5.49]	7.98 (.208) [7.57, 8.39]	2.97***
Kilos picked per day	Confidential		23.2***
Hours worked per day	Confidential		-.475
Number of workers in same field	41.1 (2.38)	38.1 (1.29)	-3.11
Daily pay	Confidential		1.80
Unit wage per kilogram picked	Confidential		-.105***

*** denotes significance at 1 percent. Sample sizes are the same as those used for the productivity regressions. Standard errors and confidence intervals take account of the observations being clustered by field-day. Productivity is measured in kilograms per hour. Daily pay refers to pay from picking only. Both daily pay and the unit wage per kilogram picked are measured in UK Pounds Sterling. Some information in the table cannot be shown due to confidentiality requirements.

- Is this due to response to change in piece rate?
 - No, piece rate went down → Incentives to work less (susbt. effect)

- Results robust to controls
- Results are stronger the more friends are on the field

	(1a)	(1b)	(2a)	(2b)
	Relative incentives	Relative incentives	Piece rates	Piece rates
Share of workers in the field who are friends	-1.68*** (.647)	-5.52** (2.36)	.072 (.493)	1.17 (1.60)
Share of workers in the field who are friends × number of workers in same field		1.60** (.684)		-.285 (.501)
Number of workers in same field		.182 (.117)		.085 (.069)
Marginal effect of group size (at mean friends' share)		.236** (.110)		.076 (.065)
Worker fixed effects	Yes	Yes	Yes	Yes
Field fixed effects	Yes	Yes	Yes	Yes
Other controls	Yes	Yes	Yes	Yes
Adjusted R^2	.3470	.3620	.3065	.3081
Number of observations (worker-field-day)	2860	2860	4400	4400

- Two Interpretations:
 - Social Preferences:
 - * Work less to help others
 - * Work even less when friends benefit, since care more for them
 - Repeated Game
 - * Enforce low-effort equilibrium
 - * Equilibrium changes when switch to flat pay

- Test: Observe results for tall plant where cannot observe productivity of others (raspberries vs. strawberries)

- Compare Fruit Type 1 (Strawberries) to Fruit Type 2 (Raspberries)
 - No effect for Raspberries

DEPENDENT VARIABLE = LOG OF WORKER'S PRODUCTIVITY
(KILOGRAM PICKED PER HOUR PER FIELD-DAY)
ROBUST STANDARD ERRORS REPORTED IN PARENTHESES, ALLOWING FOR CLUSTERING
AT FIELD-DAY LEVEL

	(1) Fruit type 2	(2) Fruit type 1	(3) Fruit types 1 and 2 combined
Piece rate dummy (P_t)	-.063 (.129)	.483*** (.094)	
Piece rate \times fruit type 2			-.100 (.095)
Piece rate \times fruit type 1			.490*** (.092)

- \rightarrow No Pure Social Preferences. However, can be reciprocity
- Important to control for repeated game effects \rightarrow Next papers

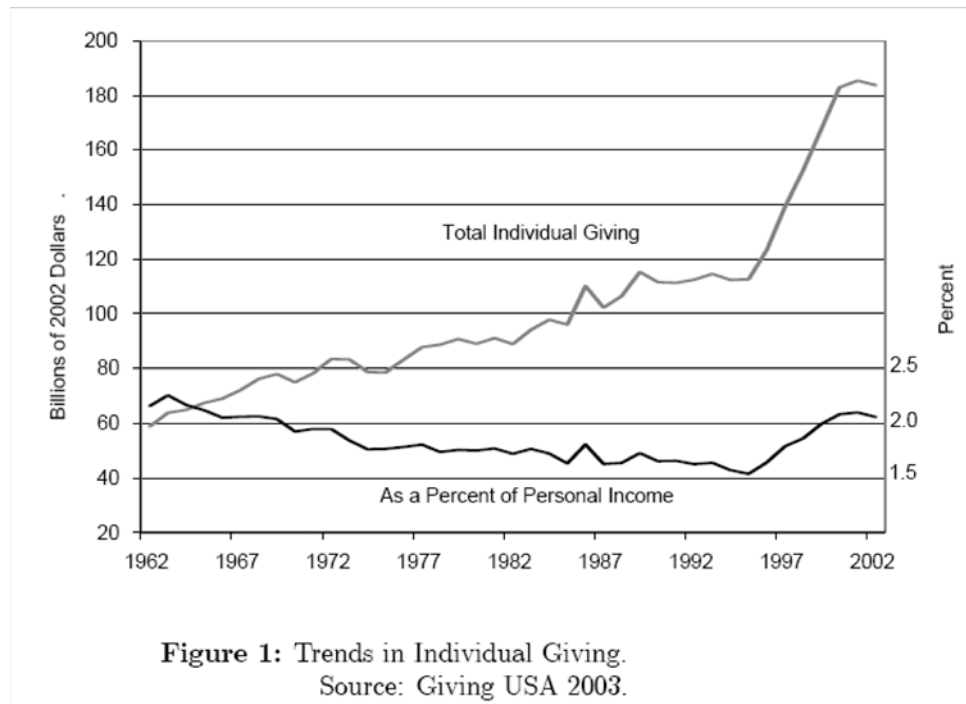
4 Social Preferences: Charitable Giving I

- Andreoni (2004). Excellent survey of the theory and evidence
- Stylized facts:
 - US Giving very large: 1.5 to 2.1 percent GDP!
 - Most giving by individuals (Table 1)

Source of gifts	Billions of dollars	Percent of total
Individuals	183.7	76.3
Foundations	26.9	11.2
Bequests	18.1	7.5
Corporations	12.2	5.1
Total for all Sources	240.9	100

Source: Giving USA, 2003

- – Giving fairly constant over time (Figure 1)



- Giving by income, age, and education (Table 2 – no controls)
 - Giving as percent of income fairly stable
 - Increase for very rich (tax incentives matter here)

Table 2
Private philanthropy by income, age, and education of the giver, 1995

	Percent of households who give	Average amount given by those who give	Percent of household income
All contributing households	68.5	1,081	2.2
<i>Household Income</i>			
under \$10,000	47.3	324	4.8
10,000–19,000	51.1	439	2.9
20,000–29,999	64.9	594	2.3
30,000–39,999	71.8	755	2.2
40,000–49,999	75.3	573	1.3
50,000–59,999	85.5	1,040	1.9
60,000–74,999	78.5	1,360	2.0
75,000–99,999	79.7	1,688	2.0
100,000 or above	88.6	3,558	3.0

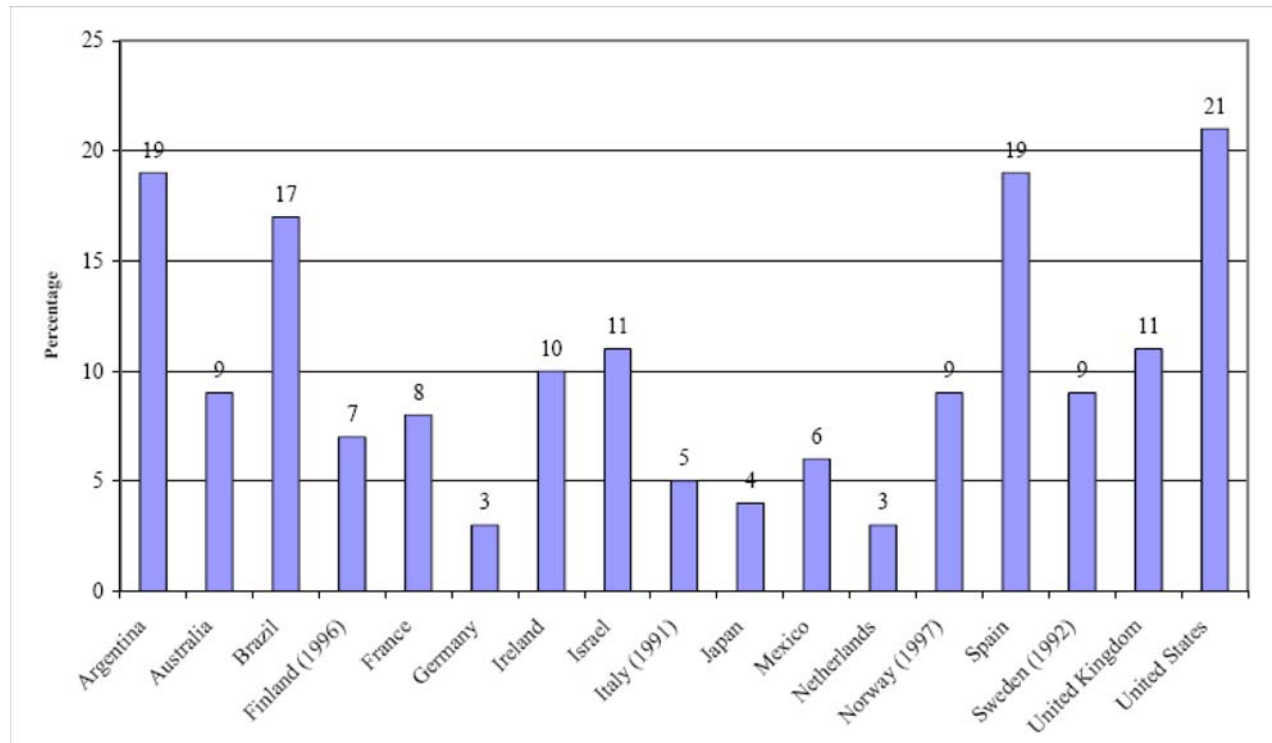
- Giving to whom? (Table 3)
 - Mostly for religion
 - Also: human services, education, health
 - Very little international donations

Table 3
Private Philanthropy by Type of Charitable Organization, 1995.

Type of Charity	Percent of Households who give	Average amount given by those who give	Percent of total household contributions
Arts, culture and humanities	9.4	221	2.6
Education	20.3	335	9.0
Environment	11.5	110	1.6
Health	27.3	218	8.1
Human Services	25.1	285	9.5
International	3.1	293	1.1
Private and community foundations	6.1	196	1.4
Public or Societal benefit	10.3	127	1.7
Recreation	7.0	161	1.4
Religious	48.0	946	59.4
Youth Development	20.9	140	3.8
Other	2.1	160	0.3

Source: Author's calculations, data from Independent Sector, Giving and Volunteering, 1995.

- Compare to giving in other countries (Figure 2)
 - In US non-profits depend more on Charitable contributions



- What else do we know?
- Until 1990s, very limited research on charitable giving
- Then:
 1. Evidence by Jim Andreoni and others on fund-raising, and especially on crowding out prediction (see below)
 2. Field experiments by John List and others

- Focus on Field Experiments. First paper: **List and Lucking-Reiley (2002)** focuses on seed money
 - Capital campaign to raise money for computer center at Univ. Central Florida
 - 3,000 letters assign to 6 treatments
 - Randomization of seed money, i.e., how much money was already raised
 - Randomization of whether refund promised if threshold not matched

TABLE 1
RESULTS OF THE FIELD EXPERIMENT

	10	10R	33	33R	67	67R
A. Experimental Design						
Number of solicitations mailed	500	500	500	500	500	500
Seed money (%)	10%	10%	33%	33%	67%	67%
Seed money (\$)	\$300	\$300	\$1,000	\$1,000	\$2,000	\$2,000
Refund offered?	no	yes	no	yes	no	yes
B. Results						
Number of contributions	17	20	33	31	42	40
Participation rate	3.4%	4.0%	6.6%	6.2%	8.4%	8.0%
Total contributions	\$202	\$379	\$805	\$863	\$1,485	\$1,775
Mean amount given	\$11.88	\$18.95	\$24.39	\$27.84	\$35.36	\$44.38
Standard error of mean amount	\$2.27	\$3.13	\$2.50	\$4.59	\$2.26	\$6.19

- Huge effect of the seed money, less so of refund
- Interpretation: Presumably signalling of quality

- More recent work: **Landry et al. (QJE, 2006)**
 - Door-to-door fund-raising as opposed to mailer
 - Test different form of solicitation
 - * Seed Money or not
 - * Lottery or not
 - Examines also features of solicitor

- Main finding: Female attractiveness matters, male attractiveness does not

TABLE IV
DICHOTOMOUS CONTRIBUTION DECISION AND SOLICITOR CHARA

	Model A	Model B	Model C	Model D	Model E
Overall constant—	0.27**	0.28**	0.25**	0.27**	0.26**
VCM is baseline	(0.03)	(0.08)	(0.07)	(0.08)	(0.07)
VCM with seed	-0.11**	-0.08	-0.07	-0.06	-0.07
money	(0.04)	(0.06)	(0.05)	(0.05)	(0.05)
Single-prize lottery	0.20**	0.19**	0.20**	0.21**	0.19**
	(0.06)	(0.06)	(0.06)	(0.06)	(0.06)
Multiple-prize lottery	0.15**	0.18**	0.20**	0.21**	0.20**
	(0.05)	(0.05)	(0.05)	(0.05)	(0.05)
Solicitor beauty		0.07**			
rating		(0.03)			
Beauty—male			-0.02	-0.03	-0.04
solicitor			(0.04)	(0.04)	(0.04)
Beauty—female			0.12**	0.13**	0.12**
solicitor			(0.04)	(0.04)	(0.04)

- What does this teach us about charitable giving in general? That more affects giving than just pure altruism

- Charitable giving important phenomenon – How do we understand it?
- **Model 1.** Social preferences: Giving because caring for welfare of others
- Problem (i): Amounts given off relative to lab experiments
- Problem (ii): Model predicts crowding out of giving:
 - If government spends on income of needy group, corresponding one-on-one decrease in giving
 - Evidence of crowding out: Limited crowd-out
- Problem (iii): Model predicts giving to one highest-value charity—Instead we observe dispersion across charities
- Problem (iv): In-person or phone requests for giving raise much more than impersonal requests (mail)

- **Model 2.** Andreoni (1994): Warm-Glow or Impure altruism.
 - Agent gets utility $v(g)$ directly from giving
 - Utility $v(g)$ sharply concave
- Can explain (i), (ii), and (iii) – See Problem Set 3
- Does not directly explain (iv) – Can assume though that warm-glow is triggered more by in-person giving

- **Model 3.** Giving is due to social pressure
 - Pay a disutility cost S if do not give when asked
 - No disutility cost if can avoid to meet the solicitor
- Can explain (i), (ii), and (iii): Give small amounts to charities, mostly because asked
- Can also explain (iv): Give more in higher social pressure environments
- Key prediction differentiating Models 2 and 3:
 - Model 2: Agent seeks giving occasions to get warm glow
 - Model 3: Agents avoids giving occasions to avoid social pressure
- **DellaVigna, List, and Malmendier (2009):** Next time

5 Methodology: Field Experiments

- Field Experiments combine advantages of field studies and natural experiments:
 - Field setting (External Validity)
 - Randomization (Internal Validity)
- Common in Development, Public, Psychology and Economics, (Labor)
- Uncommon in IO (except for Demand estimation), Corporate Finance, Asset Pricing, Macro
- Difficulties: large sample (costly) and getting approval for implementation

- Definition 1. Card, DellaVigna, and Malmendier (*JEP* 2011) Randomized allocation to treatment and control groups for study purposes in a field setting
 - Excludes studies with no randomization (Bandiera et al., 2005 and on)
 - Includes social experiments run by the government
 - Includes experiments run by firms (Ausubel, 1999)
 - Excludes incidental randomization (i.e., lottery winnings, or Vietnam draft number)

- Definition 2. Harrison and List (*JEL* 2004): Broader definition, does not emphasize randomized allocation
 - But then how to separate from natural experiments?
 - Emphasis on laboratory versus field: 4 groups
 1. *(Conventional) Laboratory Experiment*
 2. *Artefactual Laboratory Experiment*. This is laboratory experiment in the field (i.e., on non-students)
 3. *Framed Field Experiment*. Experiment in the field with natural setting, but people aware of experimental treatments
 4. *Natural Field Experiment*. Experiment in the field, subjects unaware of manipulations

- What to do if planning a field experiment?

- **Advice 1.** Read how-to manuals and previous field experiments: **Duflo-Glennerster-Kremer (NBER, 2006)**
 - * Great discussion of practical issues: Compliance, Sample Size,...
 - * Discussion of statistical issue, such as power tests
 - * Targeted toward development

- **Advice 2.** Choose what type of Experiment
 - *Large-Scale Experiment.* Example: Bandiera et al. (2005)
 - * More common in Development
 - * Convince company or organization (World Bank, Government)
 - * Need substantial funding
 - * Example among students:
 - Damon Jones: field experiment on tax preparers
 - However (also Damon): H&R Block experiment fell through after 1-year plans
 - Safeway (research center at Stanford, Kristin Kiesel in charge)

– *Small-Scale Experiment*. Example: Falk (2008)

- * More common in Psychology and Economics

- * Need to convince non-profit or small company

- * Limited funds needed – often company will pay

- * Example among students:

- Dan Acland: projection bias and gym attendance

- Vinci Chow: commitment devices for on-line computer game play

- Pete Fishman: small video store randomized advertising

- **Advice 3.** Need two components:
 1. Interesting economic setting:
 - Charity, Gym, Village in Kenya
 - Does Video Games matter? Yes, increasingly so
 2. Economic model to test
 - Examples: Self-control, reciprocity, incentives
 - Avoid pure data-finding experiments
 - Insurance. If you can, pick a case where ‘either’ result is interesting
 - Best scenario: Do a field experiment tied to a model to infer parameters

- **Advice 4.** Two key issues: Power calculations and Pilots
 - *Power calculations.* Will your sample size be enough?
 - * Crucial to do ex ante to avoid wasting time and money
 - * Simple case:
 - Assume outcome binary variable, dep.variable is share p doing 1 (Ex: giving to charity, taking up comm. device)
 - Standard error will be $\sqrt{p(1-p)/n}$
 - Example: $p = .5$, s.e. is .05 with $n = 100$, .025 with $n = 400$
 - *Pilots.* So many things can go wrong – try to do small pilot
 - * Use to spot problems in implementation
 - * Do not use pilot as data analysis (sample too small)

- **Advice 5.** Other practical issues:
 - Mostly refer to **Duflo-Glennerster-Kremer (NBER, 2006)**
 - Approval from Humans Subjects!
 - * At Berkeley, takes about 2 months
 - * More about this later
 - Keep in mind implementation of randomization
 - * Example: Cross Designs hard to implement correctly
 - * Example: **Green-Gerber (APSR, 2001)** on voter turnout:
 - cross-randomize phone calls, mailings, in-person visits
 - Hard to implement → Lead to loss of randomization

- * OK to do if requires just computerized implementation (ex: loan offers)
- Monitor what happens in the field *continuously*
- Build in data redundancy to catch measurement error or implementation problems
- * Example: ‘Did you see a flyer on the door?’ in DellaVigna-List-Malmendier (2009)

- **Advice 6.** Start looking soon for funding

- Funding harder to obtain for graduate students

- Good options:

- * IBER: \$1,000 administered quickly (one week or so)

- * Russel Sage Small Grant Program: \$5,000 (\$2,500 for paying subjects) (two to three months)

- * NSF dissertation improvement grant website (<http://www.nsf.gov/funding/pg>)

- * Look at CVs of assistant professors in your field or job market students (Jonas' advice)

- * Ask your advisor → May know of some funding sources