Outline

1. Market Reaction to Biases: Behavioral Finance
2. Market Reaction to Biases: Corporate Decisions
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4. Market Reaction to Biases: Political Economy
5. Welfare Response to Biases
6. Methodology: Markets and Non-Standard Behavior
7. Summary of Evidence
8. Concluding Remarks
1 Market Reaction to Biases: Behavioral Finance

• Who do ‘smart’ investors respond to investors with biases?

• First, brief overview of anomalies in Asset Pricing (from Barberis and Thaler, 2004)

  1. **Underdiversification.**
     
     (a) Too few companies.
     
     – Investors hold an average of 4-6 stocks in portfolio.
     
     – Improvement with mutual funds

     (b) Too few countries.

     – Investors heavily invested in own country.

     – Own country equity: 94% (US), 98% (Japan), 82% (UK)
– Own area: own local Bells (Huberman, 2001)

(c) Own company
– In companies offering own stock in 401(k) plan, substantial investment in employer stock

2. **Naive diversification.**
– Investors tend to distribute wealth ‘equally’ among alternatives in 401(k) plan (Benartzi and Thaler, 2001; Huberman and Jiang, 2005)

3. **Excessive Trading.**
– Trade too much given transaction costs (Odean, 2001)
4. **Disposition Effect in selling**
   - Investors more likely to sell winners than losers

5. **Attention Effects in buying**
   - Stocks with extreme price or volume movements attract attention
     (Odean, 2003)

- Should market forces and arbitrage eliminate these phenomena?
• **Arbitrage:**
  – Individuals attempt to maximize individual wealth
  – They take advantage of opportunities for free lunches

• Implications of arbitrage: ‘Strange’ preferences do not affect pricing

• Implication: For prices of assets, no need to worry about behavioral stories

• Is it true?
• Fictitious example:
  – Asset A returns $1 tomorrow with \( p = .5 \)
  – Asset B returns $1 tomorrow with \( p = .5 \)

  – Arbitrage \( \rightarrow \) Price of A has to equal price of B
  – If \( p_A > p_B \),
    * sell A and buy B
    * keep selling and buying until \( p_A = p_B \)
  – Viceversa if \( p_A < p_B \)
• Problem: Arbitrage is limited (de Long et al., 1991; Shleifer, 2001)

• In Example: can buy/sell A or B and tomorrow get fundamental value

• In Real world: prices can diverge from fundamental value

• Real world example. Royal Dutch and Shell
  – Companies merged financially in 1907
  – Royal Dutch shares: claim to 60% of total cash flow
  – Shell shares: claim to 40% of total cash flow
  – Shares are nothing but claims to cash flow
- Price of Royal Dutch should be $60/40 = 3/2$ price of Shell

- $p_{RD}/p_S$ differs substantially from 1.5 (Fig. 1)

Fig. 1. Log deviations from Royal Dutch/Shell parity. Source: Froot and Dabora (1999).
• Plenty of other examples (Palm/3Com)

• What is the problem?
  
  – Noise trader risk, investors with correlated valuations that diverge from fundamental value

  – (Example: Naive Investors keep persistently bidding down price of Shell)

  – In the long run, convergence to cash-flow value

  – In the short-run, divergence can even increase

  – (Example: Price of Shell may be bid down even more)
• **Noise Traders**

• DeLong, Shleifer, Summers, Waldman (*JPE* 1990)

• Shleifer, *Inefficient Markets*, 2000

• Fundamental question: What happens to prices if:
  – (Limited) arbitrage
  – Some irrational investors with correlated (wrong) beliefs

• First paper on Market Reaction to Biases

• *The* key paper in Behavioral Finance
The model assumptions

A1: arbitrageurs risk averse and short horizon

→ Justification?

* Short-selling constraints
  (per-period fee if borrowing cash/securities)

* Evaluation of Fund managers.

* Principal-Agent problem for fund managers.
A2: noise traders (Kyle 1985; Black 1986)

misperceive future expected price at $t$ by

$$\rho_t \sim \mathcal{N}(\rho^*, \sigma^2_\rho)$$

misperception correlated across noise traders ($\rho^* \neq 0$)

$\rightarrow$ Justification?

* fads and bubbles (Internet stocks, biotechs)

* pseudo-signals (advice broker, financial guru)

* behavioral biases / misperception riskiness
What else?

- \( \mu \) noise traders, \((1 - \mu)\) arbitrageurs

- OLG model
  - Period 1: initial endowment, trade
  - Period 2: consumption

- Two assets with identical dividend \( r \)
  - safe asset: perfectly elastic supply
    \[ \implies \text{price}=1 \text{ (numeraire)} \]
  - unsafe asset: inelastic supply (1 unit)
    \[ \implies \text{price}? \]

- Demand for unsafe asset: \( \lambda^a \) and \( \lambda^n \), with \( \lambda^a + \lambda^n = 1 \).

- CARA: \( U(w) = -e^{-2\gamma w} \) (\( w \) wealth when old)
\[ E[U(w)] = \int_{-\infty}^{\infty} -e^{-2\gamma w} \cdot \frac{1}{\sqrt{2\pi\sigma_w^2}} \cdot e^{-\frac{1}{2\sigma_w^2}(w-w)^2} \, dw \]

\[ = -\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma_w^2}} \cdot e^{-\frac{4\gamma w\sigma_w^2 + w^2 - 2\sigma_w^2 - 2\gamma\sigma_w^2 w}{2\sigma_w^2}} \, dw \]

\[ = -e^{-4\gamma^2\sigma_w^4 + 2\gamma\sigma_w^2 w} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma_w^2}} \cdot e^{-\frac{(w-[2\gamma\sigma_w^2 + \bar{w}])^2}{2\sigma_w^2}} \, dw \]

\[ = -e^{-4\gamma^2\sigma_w^4 + 2\gamma\sigma_w^2 \bar{w}} = e^{-2\gamma(\bar{w} - \gamma\sigma_w^2)} \]

\[ \uparrow \]

\[ \text{max } E[U(w)] \quad \overset{\text{pos. mon. transf.}}{\sim} \quad \text{max } \bar{w} - \gamma\sigma_w^2 \]
Arbitrageurs:

$$\max (w_t - \lambda^a_t p_t)(1 + r)$$

$$+ \lambda^a_t (E_t[p_{t+1}] + r)$$

$$- \gamma (\lambda^a_t)^2 Var_t(p_{t+1})$$

Noise traders:

$$\max (w_t - \lambda^n_t p_t)(1 + r)$$

$$+ \lambda^n_t (E_t[p_{t+1}] + \rho_t + r)$$

$$- \gamma (\lambda^n_t)^2 Var_t(p_{t+1})$$

(Note: Noise traders know how to factor the effect of future price volatility into their calculations of values.)
f.o.c.

Arbitrageurs: \( \frac{\partial E[U]}{\partial \lambda_t^a} = \frac{1}{0} \)

\[ \lambda_t^a = \frac{r + E_t[p_{t+1}] - (1 + r)p_t}{2\gamma \cdot \text{Var}_t(p_{t+1})} \]

Noise traders: \( \frac{\partial E[U]}{\partial \lambda_t^n} = \frac{1}{0} \)

\[ \lambda_t^n = \frac{r + E_t[p_{t+1}] - (1 + r)p_t}{2\gamma \cdot \text{Var}_t(p_{t+1})} + \frac{\rho_t}{2\gamma \cdot \text{Var}_t(p_{t+1})} \]
Interpretation

- Demand for unsafe asset function of:
  - $(+)$ expected return $(r + E_t[p_{t+1}] - (1 + r)p_t)$
  - $(-)$ risk aversion ($\gamma$)
  - $(-)$ variance of return ($Var_t(p_{t+1})$)
  - $(+)$ overestimation of return $\rho_t$ (noise traders)

- Notice: noise traders hold more risky asset than arb. if $\rho > 0$ (and viceversa)

- Notice: Variance of prices come from noise trader risk. “Price when old” depends on uncertain belief of next periods’ noise traders.
• Impose general equilibrium: \( \lambda^n \mu + \lambda^a (1 - \mu) = 1 \) to obtain

\[
1 = \frac{r + E_t[p_{t+1}] - (1 + r)p_t}{2\gamma \cdot Var_t(p_{t+1})} + \frac{\mu \rho_t}{2\gamma \cdot Var_t(p_{t+1})} \quad \text{or}
\]

\[
p_t = \frac{1}{1 + r} \left[ r + E_t[p_{t+1}] - 2\gamma \cdot Var_t(p_{t+1}) + \mu \rho_t \right]
\]

• To solve for \( p_t \), we need to solve for \( E_t[p_{t+1}] = E[p] \) and \( Var_t(p_{t+1}) \)

\[
E[p] = \frac{1}{1 + r} \left[ r + E_t[p] - 2\gamma \cdot Var_t(p_{t+1}) + \mu E[\rho_t] \right]
\]

\[
E[p] = 1 + \frac{-2\gamma \cdot Var_t(p_{t+1}) + \mu \rho^*}{r}
\]
Rewrite $p_t$ plugging in

$$p_t = 1 - 2\gamma \cdot \frac{\text{Var}_t(p_{t+1})}{r} + \frac{\mu \rho^*}{r(1+r)} + \frac{\mu \rho_t}{1+r}$$

$$\text{Var}[p_t] = \text{Var}\left[\frac{\mu \rho_t}{1+r}\right] = \frac{\mu^2}{(1+r)^2} \text{Var}(\rho_t) = \frac{\mu^2}{(1+r)^2} \sigma^2_{\rho}$$

Rewrite $p_t$

$$p_t = 1 - 2\gamma \frac{\mu^2 \sigma^2_{\rho}}{r (1+r)^2} + \frac{\mu \rho^*}{r} + \frac{\mu (\rho_t - \rho^*)}{1+r}$$

Noise traders affect prices!

Term 1: Variation in noise trader (mis-)perception

Term 2: Average misperception of noise traders

Term 3: Compensation for noise trader risk
• Relative returns of noise traders

– Compare returns to noise traders $R^n$ to returns for arbitrageurs $R^a$:

$$\Delta R = R^n - R^a = (\lambda^n_t - \lambda^a_t) [r + p_{t+1} - p_t (1 + r)]$$

$$E(\Delta R | \rho_t) = \rho_t - \frac{(1 + r)^2 \rho_t^2}{2\gamma\mu\sigma^2_\rho}$$

$$E(\Delta R) = \rho^* - \frac{(1 + r)^2 (\rho^*)^2 + (1 + r)^2 \sigma^2_\rho}{2\gamma\mu\sigma^2_\rho}$$

– Noise traders hold more risky asset if $\rho^* > 0$

– Return of noise traders can be higher if $\rho^* > 0$ (and not too positive)

– Noise traders therefore may outperform arbitrageurs if optimistic!

– (Reason is that they are taking more risk)
Welfare

- Sophisticated investors have higher utility
- Noise traders have lower utility than they expect
- Noise traders may have higher returns (if $\rho^* > 0$)
- Noise traders do not necessarily disappear over time
• Three fundamental assumptions

1. OLG: no last period; short horizon

2. Fixed supply unsafe asset (\(a\) cannot convert safe into unsafe)

3. Noise trader risk systematic

• Noise trader models imply that biases affect asset prices:
  - Reference Dependence
  - Attention
  - Persuasion
• Here:
  – Biased investors
  – Non-biased investors

• Behavioral corporate finance:
  – Investors (biased)
  – CEOs (smart)

• Behavioral Industrial Organization:
  – Consumers (biased)
  – Firms (smart)
2 Market Reaction to Biases: Corporate Decisions

- Baker, Ruback, and Wurgler (2005)

- Behavioral corporate finance:
  - biased investors (overvalue or undervalue company)
  - smart managers
  - (Converse: biased (overconfident) managers and rational investors)

- Firm has to decide how to finance investment project:
  1. internal funds (cash flow/retained earnings)
  2. bonds
  3. stocks
• Fluctuation of equity prices due to noise traders

• Managers believe that the market is inefficient
  – Issue equity when stock price exceeds perceived fundamental value
  – Delay equity issue when stock price below perceived fundamental value

• Consistent with
  – Survey Evidence of 392 CFO's (Graham and Harvey 2001): 67% say under/overvaluation is a factor in issuance decision
  – Insider trading

• Go over quickly two examples
• **Long-run performance of equity issuers**
  
  – Market Timing prediction: Companies issuing equity underperform later
  
  – Loughran-Ritter (1995): Compare matching samples of
    * companies doing IPOs
    * companies not doing IPOs but have similar market cap.

![Annual percentage return chart](chart.png)
Similar finding with SEOs

Figure 2. The average annual raw returns for 4,753 initial public offerings (IPOs), and their matching nonissuing firms (top), and the average annual raw returns for 3,702 seasoned equity offerings (SEO), and their matching nonissuing firms (bottom), during the five years after the issue. The equity issues are from 1970 to 1990. Using the first closing postissue market price, the equally weighted average buy-and-hold return for the year after the issue is calculated for the issuing firms and for their matching firms (firms with the same market capitalization that have not issued equity during the prior five years). On each anniversary of the issue date, the equally weighted average buy-and-hold return during the next year for all of the surviving issuers and their matching firms is calculated. For matching firms that get delisted (or issue equity) while the issuer is still trading, the proceeds from the sale on the delisting date are reinvested in a new matching firm for the remainder of that year (or until the issuer is delisted). The numbers graphed above are reported in Table III.
3 Market Reaction to Biases: Employers

- Kahneman, Knetsch and Thaler (1986): Telephone surveys in Canada in 1984 and 1985 —> Ask questions on fairness

Question 4A. A company is making a small profit. It is located in a community experiencing a recession with substantial unemployment but no inflation. There are many workers anxious to work at the company. The company decides to decrease wages and salaries 7% this year.

\[(N = 125) \quad \text{Acceptable 38% Unfair 62\%}\]

Question 4B. …with substantial unemployment and inflation of 12%…The company decides to increase salaries only 5% this year.

\[(N = 129) \quad \text{Acceptable 78\% Unfair 22\%}\]

- A real and nominal wage cut is not fair (Question 4A)
  - A real (but not nominal) wage cut is fair (Question 4B)
• If this is true, expect employers to minimize cases of $w_t - w_{t-1} < 0$

• **Card and Hyslop, 1997**: Examine discontinuity around 0 of nominal wage changes

• Prediction of theory:
• Data sources:
  – 1979-1993 CPS.
    * Rolling 2-year panel
    * Restrict to paid by the hour and to same 2-digit industry in the two years
    * Restrict to non-minimum wage workers
  – PSID 4-year panels 1976-79 and 1985-88

• Use Log Wage changes: \( \log w_t - \log w_{t-1} \)

• Issue with measurement error and heaping at \( \log w_t - \log w_{t-1} = 0 \)

• Construct counterfactual density of LogWage changes
  – Assume symmetry
  – Positive log wage changes would not be affected
• Plots using kernel estimates of density (local smoother)

• Compare the actual distribution and the predicted one

• Evidence from the CPS year-by-year

• Problem more severe in years with lower inflation

• Large effect of nominal rigidities

• Effect on firings?
Figure 4: Smoothed (Kernel) Estimates of Actual and Counterfactual Densities of Real Wage Changes, CPS Samples from 1979-80 to 1982-83
Figure 4 (Continued): Smoothed (Kernel) Estimates of Actual and Counterfactual Densities of Real Wage Changes, CPS Samples from 1983-84 to 1986-87
Figure 4 (Continued): Smoothed (Kernel) Estimates of Actual and Counterfactual Densities of Real Wage Changes, CPS Samples from 1987-88 to 1990-91
4 Market Reaction to Biases: Political Economy

- Interaction between:
  - (Smart) Politicians:
    * Personal beliefs and party affiliation
    * May pursue voters/consumers welfare maximization
    * BUT also: strong incentives to be reelected
  - Voters (with biases):
    * Low (zero) incentives to vote
    * Limited information through media
    * Likely to display biases

- Behavioral political economy
• Examples of voter biases:
  
  – Effect of candidate order (Ho and Imai)
  
  – Imperfect signal extraction (Wolfers, 2004) -> Voters more likely to vote an incumbent if the local economy does well even if... it’s just due to changes in oil prices
  
  – Susceptible to persuasion (DellaVigna and Kaplan, 2007)
  
  – More? Short memory about past performance?
  
• Eisensee and Stromberg (2007): Limited attention of voters
• Setting:
  – Natural Disasters occurring throughout the World
  – US Ambassadors in country can decide to give Aid
  – Decision to give Aid affected by
    * Gravity of disaster
    * Political returns to Aid decision

• Idea: Returns to aid are lower when American public is distracted by a major news event
• Main Measure of Major News: median amount of Minutes in Evening TV News captured by top-3 news items (Vanderbilt Data Set)
- Dates with largest news pressure

<table>
<thead>
<tr>
<th>Year</th>
<th>Date</th>
<th>Main News Story</th>
</tr>
</thead>
<tbody>
<tr>
<td>2003</td>
<td>14 Aug</td>
<td>New York City Blackout</td>
</tr>
<tr>
<td></td>
<td>22 Mar</td>
<td>Invasion of Iraq: Day 3</td>
</tr>
<tr>
<td>2002</td>
<td>11 Sep</td>
<td>9/11 Commemoration</td>
</tr>
<tr>
<td></td>
<td>24 Oct</td>
<td>Sniper Shooting in Washington: Arrest of Suspects</td>
</tr>
<tr>
<td>2001</td>
<td>13 Sep</td>
<td>9/11 Attack on America: Day 3</td>
</tr>
<tr>
<td></td>
<td>12 Sep</td>
<td>9/11 Attack on America: Day 2</td>
</tr>
<tr>
<td>2000</td>
<td>26 Nov</td>
<td>Gore vs. Bush: Florida Recount - Certification by Katherine Harris</td>
</tr>
<tr>
<td></td>
<td>8 Dec</td>
<td>Gore vs. Bush: Florida Recount - Supreme Court Ruling</td>
</tr>
<tr>
<td>1999</td>
<td>1 Apr</td>
<td>Kosovo Crisis: U.S. Soldiers Captured</td>
</tr>
<tr>
<td></td>
<td>18 Jul</td>
<td>Crash of Plane Carrying John F. Kennedy, Junior</td>
</tr>
<tr>
<td>1998</td>
<td>16 Dec</td>
<td>U.S. Missile Attack on Iraq</td>
</tr>
<tr>
<td></td>
<td>18 Dec</td>
<td>Clinton Impeachment</td>
</tr>
<tr>
<td>1997</td>
<td>23 Dec</td>
<td>Oklahoma City Bombing: Trial</td>
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<tr>
<td></td>
<td>31 Aug</td>
<td>Princess Diana's Death</td>
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<tr>
<td>1996</td>
<td>18 Jul</td>
<td>TWA Flight 800 Explosion</td>
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<tr>
<td></td>
<td>27 Jul</td>
<td>Olympic Games Bombing in Atlanta</td>
</tr>
<tr>
<td>1995</td>
<td>3 Oct</td>
<td>O.J. Simpson Trial: The Verdict</td>
</tr>
<tr>
<td></td>
<td>22 Apr</td>
<td>Oklahoma City Bombing</td>
</tr>
<tr>
<td>1994</td>
<td>17 Jan</td>
<td>California Earthquake</td>
</tr>
<tr>
<td></td>
<td>18 Jan</td>
<td>O.J. Simpson Arrested</td>
</tr>
<tr>
<td>1993</td>
<td>17 Jan</td>
<td>U.S. Missile Attack on Iraq</td>
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<tr>
<td></td>
<td>20 Apr</td>
<td>Waco, Texas: Cult Standoff Ends in Fire</td>
</tr>
<tr>
<td></td>
<td>1 May</td>
<td>Los Angeles Riots</td>
</tr>
</tbody>
</table>
• 5,000 natural Disasters in 143 countries between 1968 and 2002 (CRED)
  – 20 percent receive USAID from Office of Foreign Disaster Assistance
    (first agency to provide relief)
  – 10 percent covered in major broadcast news
  – OFDA relief given if (and only if) Ambassador (or chief of Mission) in
    country does Disaster Declaration
  – Ambassador can allocate up to $50,000 immediately

• Estimate

\[
\text{Relief} = \alpha \text{News} + \beta X + \varepsilon
\]

• Below: \textit{News} about the Disaster is instrumented with:
  – Average News Pressure over 40 days after disaster
  – Olympics
TABLE IV
EFFECT OF THE PRESSURE FOR NEWS TIME ON DISASTER NEWS AND RELIEF

<table>
<thead>
<tr>
<th></th>
<th>Dependent variable: News</th>
<th></th>
<th>Dependent variable: Relief</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>News Pressure</td>
<td>-0.0162</td>
<td>-0.0163</td>
<td>-0.0177</td>
<td>-0.0142</td>
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<tr>
<td></td>
<td>(0.0041)***</td>
<td>(0.0041)***</td>
<td>(0.0057)***</td>
<td>(0.0057)***</td>
</tr>
<tr>
<td>Olympics</td>
<td>-0.1078</td>
<td>-0.1079</td>
<td>-0.0871</td>
<td>-0.1111</td>
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<tr>
<td></td>
<td>(0.0470)**</td>
<td>(0.0470)**</td>
<td>(0.0628)</td>
<td>(0.0413)***</td>
</tr>
<tr>
<td>World Series</td>
<td>-0.1133</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.1065)</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>log Killed</td>
<td></td>
<td>0.0665</td>
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<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.0040)***</td>
<td></td>
<td></td>
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<tr>
<td>log Affected</td>
<td>0.0123</td>
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<td></td>
<td></td>
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<tr>
<td></td>
<td>(0.0024)***</td>
<td></td>
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<td></td>
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<tr>
<td>imputed log Killed</td>
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<td>0.0481</td>
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<td>0.0442</td>
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<td></td>
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<td>(0.0034)***</td>
<td></td>
<td>(0.0037)***</td>
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<tr>
<td>imputed log Affected</td>
<td></td>
<td>0.0151</td>
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<td>0.0304</td>
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<tr>
<td></td>
<td></td>
<td>(0.0020)***</td>
<td></td>
<td>(0.0020)***</td>
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<tr>
<td>Observations</td>
<td>3212</td>
<td>3212</td>
<td>2926</td>
<td>3212</td>
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<tr>
<td>R-squared</td>
<td>0.1799</td>
<td>0.1797</td>
<td>0.3624</td>
<td>0.2875</td>
</tr>
</tbody>
</table>

Linear probability OLS regressions. All regressions include year, month, country and disaster type fixed effects. Regressions with imputed values ((4) and (5)) also include fixed effects for the interaction of missing values and disaster type. Robust standard errors in parentheses: * significant at 10%; ** significant at 5%; *** significant at 1%.

- 1st Stage: 2 s.d increase in News Pressure (2.4 extra minutes) decrease
  * probability of coverage in news by 4 ptg. points (40 percent)
  * probability of relief by 3 ptg. points (15 percent)
• Is there a spurious correlation between instruments and type of disaster?

• No correlation with severity of disaster

<table>
<thead>
<tr>
<th>TABLE V</th>
</tr>
</thead>
<tbody>
<tr>
<td>CORRELATIONS BETWEEN INSTRUMENTS AND THE SEVERITY OF DISASTERS</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>News Pressure</th>
<th>Olympics</th>
</tr>
</thead>
<tbody>
<tr>
<td>log Killed</td>
<td>-0.0082</td>
<td>0.0003</td>
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<tr>
<td>(0.0113)</td>
<td>(0.0010)</td>
<td></td>
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<tr>
<td>log Affected</td>
<td>0.0005</td>
<td>-0.0006</td>
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<tr>
<td>(0.0068)</td>
<td>(0.0006)</td>
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<tr>
<td>p-value: F-test of joint insignificance</td>
<td>0.75</td>
<td>0.62</td>
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<tr>
<td>Observations</td>
<td>5212</td>
<td>5212</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.3110</td>
<td>0.2035</td>
</tr>
</tbody>
</table>

OLS regressions with the instruments News Pressure and Olympics as dependent variables, and including year, month, country and disaster type fixed effects. Robust standard errors in parentheses.* significant at 10%; ** significant at 5%; *** significant at 1%. The F-test tests the joint significance of log Killed and log Affected in the regression.
• OLS and IV Regressions of Reliefs on presence in the News

• (Instrumented) availability in the news at the margin has huge effect: Almost one-on-one effect of being in the news on aid

<table>
<thead>
<tr>
<th>Table VI</th>
<th>DEPENDENT VARIABLE: Relief</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>OLS</strong></td>
<td><strong>IV</strong></td>
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<tr>
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<tr>
<td>News</td>
<td>0.2886</td>
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<td>(0.0200)***</td>
</tr>
<tr>
<td>News*abs(Pr(news)-0.5)</td>
<td>-0.4922</td>
</tr>
<tr>
<td></td>
<td>(0.1059)***</td>
</tr>
<tr>
<td>abs(Pr(news)-0.5)</td>
<td>0.5374</td>
</tr>
<tr>
<td></td>
<td>(0.0943)***</td>
</tr>
<tr>
<td>log Killed</td>
<td>0.0486</td>
</tr>
<tr>
<td></td>
<td>(0.0046)***</td>
</tr>
<tr>
<td>log Affected</td>
<td>0.0358</td>
</tr>
<tr>
<td></td>
<td>(0.0024)***</td>
</tr>
<tr>
<td>imputed log Killed</td>
<td>0.0378</td>
</tr>
<tr>
<td></td>
<td>(0.0038)***</td>
</tr>
<tr>
<td>imputed log Affected</td>
<td>0.0375</td>
</tr>
<tr>
<td></td>
<td>(0.0020)***</td>
</tr>
</tbody>
</table>

F-stat, instruments, 1st stage
Over-id restrictions, $\chi^2$ (p-value) 11.0 6.1 11.1 0.51 (0.47) 0.64 (0.42) Observations 5212 2926 5212 5212 5027 5212 5212 2926 R-squared 0.2443 0.4225 0.3800 0.3860

All regressions include year, month, country, and disaster type fixed effects. Regressions with imputed values ((3), (4) and (5)) also include fixed effects for the interaction of missing values and disaster type. Robust standard errors in parentheses: * significant at 10%; ** significant at 5%; *** significant at 1%.
• Second example: Theory/History paper, Glaeser (2005) on Political Economy of Hatred

• Idea: Hatred has demand side and supply side
  – Demand side:
    * Voters are susceptible to hatred (experiments: ultimatum game)
    * Media can mediate hatred
  – Supply side:
    * Politicians maximize chances of reelection
    * Set up a hatred media campaign toward a group for electoral gain
    * In particular, may target non-median voter
• Idea:
  – Group hatred can occur, but does not tend to occur naturally
  – Group hatred can be due to political incentives
  – Example 1: *African Americans in South, 1865-1970*
    * No hatred before Civil War
    * Conservative politicians foment it to lower demand for redistribution
    * Diffuse stories of violence by Blacks
  – Example 2: *Hatred of Jews in Europe, 1930s*
    * No hatred before 1920
    * Jews disproportionately left-wing
    * Right-wing Hitler made up Protocol of Elders of Zion
5 Welfare Response to Biases

- Need for government/social planner intervention?
  - No if:
    * Sophistication about biases
    * Markets to correct biases exist
  - Potentially yes if:
    * Naivete’ of agents
    * Missing markets
    * Example: sin taxes on goods

- Government intervention does not need to be heavy-handed:
  - Require active decision
  - Change default
• **Benartzi-Thaler, 2004** (First Behavioral paper in JPE for 15 since 1991!)

• Setting:
  – Midsize manufacturing company
  – 1998 onward
  – Company constrained by anti-discrimination rules —> Interested in increasing savings

• Features of SMT 401(k) plan:
  – No current increase in contribution rate
  – Increase in contribution rate by 3% per future pay increase
  – Can quit plan at any time
• Biases targeted:

1. Self-control
   - Desire to Save more
   - Demand for commitment

2. Partial naivete’
   - Partial Sophistication $\rightarrow$ Demand of commitment
   - Partial Naiveté $\rightarrow$ Procrastination in quitting plan

3. Loss Aversion with respect to nominal wage cuts
   - Hate real wage cuts
   - Accept nominal wage cuts
• Solutions:
  1. Increase savings in the future (not in present)
  2. Set default so that procrastination leads to more (not less) savings
  3. Schedule increase only at time of pay raise

• Implementation:

<table>
<thead>
<tr>
<th>Table 1: Participation Data for the First Implementation of SMarT</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of plan participants prior to the adoption of the SMarT plan</td>
</tr>
<tr>
<td>Number of plan participants who elected to receive a recommendation from the consultant</td>
</tr>
<tr>
<td>Number of plan participants who implemented the consultant’s recommended saving rate</td>
</tr>
<tr>
<td>Number of plan participants who were offered the SMarT plan as an alternative</td>
</tr>
<tr>
<td>Number of plan participants who accepted the SMarT plan</td>
</tr>
<tr>
<td>Number of plan participants who opted out of the SMarT plan between the first and second pay raises</td>
</tr>
<tr>
<td>Number of plan participants who opted out of the SMarT plan between the second and third pay raises</td>
</tr>
<tr>
<td>Number of plan participants who opted out of the SMarT plan between the third and fourth pay raises</td>
</tr>
<tr>
<td>Overall participation rate prior to the advice</td>
</tr>
<tr>
<td>Overall participation rate shortly after the advice</td>
</tr>
</tbody>
</table>
• Result 1: High demand for commitment device

• Result 2: Phenomenal effects on savings rates

<table>
<thead>
<tr>
<th>Participants who Did Not Contact the Financial Consultant</th>
<th>Participants who Accepted the Consultant's Recommended Saving Rate</th>
<th>Participants who Joined the SMarT Plan</th>
<th>Participants who Declined the SMarT Plan</th>
<th>All</th>
</tr>
</thead>
<tbody>
<tr>
<td>Participants initially choosing each option*</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pre-advice</td>
<td>29</td>
<td>79</td>
<td>162</td>
<td>45</td>
</tr>
<tr>
<td>First pay raise</td>
<td>6.6</td>
<td>4.4</td>
<td>3.5</td>
<td>6.1</td>
</tr>
<tr>
<td>Second pay raise</td>
<td>6.5</td>
<td>9.1</td>
<td>6.5</td>
<td>6.3</td>
</tr>
<tr>
<td>Third pay raise</td>
<td>6.8</td>
<td>8.9</td>
<td>9.4</td>
<td>6.2</td>
</tr>
<tr>
<td>Fourth pay raise</td>
<td>6.6</td>
<td>8.7</td>
<td>11.6</td>
<td>6.1</td>
</tr>
<tr>
<td></td>
<td>6.2</td>
<td>8.8</td>
<td>13.6</td>
<td>5.9</td>
</tr>
</tbody>
</table>

* There is attrition from each group over time. The number of employees who remain by the time of the fourth pay raise is 229.
• Second implementation: Simple letter sent, no seminar / additional information + 2% increase per year

• Lower take-up rate (as expected), equally high increase in savings

<table>
<thead>
<tr>
<th>TABLE 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>AVERAGE SAVING RATES FOR ISPAT INLAND (%)</td>
</tr>
<tr>
<td>Employees Who Were Already Saving on May 31, 2001</td>
</tr>
<tr>
<td>Joined SMarT (N=615)</td>
</tr>
<tr>
<td>Pre-SMarT (May 2001)</td>
</tr>
<tr>
<td>First pay raise (October 2001)</td>
</tr>
</tbody>
</table>

Note.—The sample includes 5,817 employees who are eligible to participate in the 401(k) plan and have remained with the company from May 2001 through October 2001. The sample includes 414 employees who were already saving at the maximum rate of 15 percent, although they were not allowed to join the SMarT program. The reported saving rates represent the equally weighted average of the individual saving rates.
• Third Implementation with Randomization:
  – Division A: Invitation to attend an informational seminar (40% do)
  – Division O: ‘Required’ to attend information seminar (60% do)
  – 2 Control Divisions

• Two differences in design:
  – Increase in Savings take place on April 1 whether pay increase or not (April 1 is usual date for pay increase)
  – Choice of increase in contr. rate (1%, 2%, or 3%) (Default is 2%)
  – Increases capped at 10%

• Results: Sizeable demand for commitment, and large effects on savings + Some spill-over effects
<table>
<thead>
<tr>
<th>Date</th>
<th>Employees Who Were Already Saving in December 2001</th>
<th>Employees Who Were Not Saving in December 2001</th>
<th>All Employees</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Joined SMaRT</td>
<td>Did Not Join SMaRT</td>
<td>Joined SMaRT</td>
</tr>
<tr>
<td>Observations</td>
<td>7,405</td>
<td>7,053</td>
<td>14,458</td>
</tr>
<tr>
<td>Pre-SMaRT (December 2001)</td>
<td>5.65</td>
<td>.00</td>
<td>2.90</td>
</tr>
<tr>
<td>Post-SMaRT (March 2002)</td>
<td>5.76</td>
<td>.70</td>
<td>3.20</td>
</tr>
</tbody>
</table>

A. Control Group

B. Test Group (Divisions A and O Combined)

<table>
<thead>
<tr>
<th>Date</th>
<th>Employees Who Were Already Saving in December 2001</th>
<th>Employees Who Were Not Saving in December 2001</th>
<th>All Employees</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Joined SMaRT</td>
<td>Did Not Join SMaRT</td>
<td>Joined SMaRT</td>
</tr>
<tr>
<td>Observations</td>
<td>180</td>
<td>339</td>
<td>36</td>
</tr>
<tr>
<td>Pre-SMaRT (December 2001)</td>
<td>5.25</td>
<td>5.38</td>
<td>.00</td>
</tr>
<tr>
<td>Post-SMaRT (March 2002)</td>
<td>6.83</td>
<td>5.72</td>
<td>5.03</td>
</tr>
</tbody>
</table>

C. Division A

<table>
<thead>
<tr>
<th>Date</th>
<th>Employees Who Were Already Saving in December 2001</th>
<th>Employees Who Were Not Saving in December 2001</th>
<th>All Employees</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Joined SMaRT</td>
<td>Did Not Join SMaRT</td>
<td>Joined SMaRT</td>
</tr>
<tr>
<td>Observations</td>
<td>66</td>
<td>190</td>
<td>10</td>
</tr>
<tr>
<td>Pre-SMaRT (December 2001)</td>
<td>5.47</td>
<td>5.48</td>
<td>.00</td>
</tr>
<tr>
<td>Post-SMaRT (March 2002)</td>
<td>7.32</td>
<td>5.97</td>
<td>6.80</td>
</tr>
</tbody>
</table>

D. Division O

<table>
<thead>
<tr>
<th>Date</th>
<th>Employees Who Were Already Saving in December 2001</th>
<th>Employees Who Were Not Saving in December 2001</th>
<th>All Employees</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Joined SMaRT</td>
<td>Did Not Join SMaRT</td>
<td>Joined SMaRT</td>
</tr>
<tr>
<td>Observations</td>
<td>114</td>
<td>149</td>
<td>26</td>
</tr>
<tr>
<td>Pre-SMaRT (December 2001)</td>
<td>5.14</td>
<td>5.25</td>
<td>.00</td>
</tr>
<tr>
<td>Post-SMaRT (March 2002)</td>
<td>6.55</td>
<td>5.41</td>
<td>4.35</td>
</tr>
</tbody>
</table>

Note.—The “test” group consists of individuals at Divisions A and O.
• Issues: Saving too much? Ask people if would like to quit plan

<table>
<thead>
<tr>
<th>TABLE 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>MEDIAN INCOME REPLACEMENT RATIOS (%)</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>INCOME</th>
<th>25</th>
<th>35</th>
<th>45</th>
<th>55</th>
</tr>
</thead>
<tbody>
<tr>
<td>A. Pre-SMarT</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$25,000</td>
<td>57</td>
<td>57</td>
<td>56</td>
<td>55</td>
</tr>
<tr>
<td>$50,000</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>54</td>
</tr>
<tr>
<td>$75,000</td>
<td>48</td>
<td>49</td>
<td>46</td>
<td>43</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>B. Post-SMarT</th>
</tr>
</thead>
<tbody>
<tr>
<td>$25,000</td>
</tr>
<tr>
<td>$50,000</td>
</tr>
<tr>
<td>$75,000</td>
</tr>
</tbody>
</table>

NOTE.—The table displays the median income replacement ratios for different age and income profiles, using investment advice software by Financial Engines. The projections are based on the following assumptions: no defined-benefit pension, statutory social security benefits, employee saving rate of 4 percent before SMarT and 14 percent thereafter, employer match of 50 cents on the dollar up to 6 percent, portfolio mix of 60 percent stocks and 40 percent bonds, and retirement age of 65.

• General equilibrium effect of increase in savings on returns
  - Why didn’t a company offer it? How about teaching people?
• Psychology & Economics & Public Policy:
  
  – Leverage biases to help biased agents
  
  – Do not hurt unbiased agents (cautious paternalism)

• SMartT Plan is great example:
  
  – From Design of an economist...
  
  – ...to Research Implementation with Natural Experiment and Field Experiment
  
  – ...to Policy Implementation into Law passed in Congress: *Automatic Savings and Pension Protection Act*
• Research agenda:
  
  – Identify biases (persuasion? reference dependence? self-control?)
  
  – Design contract/institution
  
  – Field experiment
  
  – Good luck!
6 Methodology: Markets and Non-Standard Behavior

- Why don’t market forces eliminate non-standard behavior?

- Common Chicago-type objection

- **Argument 1.** Experience reduces non-standard behavior.
  - Experience appears to mitigate the endowment effect (List, 2003 and 2004).
  - Experience improves ability to perform backward induction (Palacios-Huerta and Volji, 2007 and 2008)
  - **BUT:** Maybe experience does not really help (Levitt, List, and Reiley, 2008)
What does experience imply in general?

* Feedback is often infrequent (such as in house purchases) or noisy (such as in financial investments) → not enough room for experience

* Experience can exacerbate a bias if individuals are not Bayesian learners (Haigh and List 2004)

* Not all non-standard features should be mitigated by experience. Example: social preferences

* Debiasing by experienced agents can be a substitute for direct experience. However, as Gabaix and Laibson (2006) show, experienced agents such as firms typically have little or no incentive to debias individuals
- *Curse of Debiasing* (Gabaix-Laibson 2006)
  - Credit Card A teaser fees on $1000 balance:
    - $0 for six months
    - $100 fee for next six months
  - Cost of borrowing to company $100 → Firm makes 0 profit in Perfectly Competitive market
  - Naive consumer:
    - Believes no borrowing after 6 months
    - Instead keeps borrowing
    - Expects cost of card to be $0, instead pays $100
• Can Credit Card B debias consumers and profit from it?
  – Advertisement to consumers: ‘You will borrow after 6 months!’
  – Offer rate of
    * $50 for six months
    * $50 for next six months
• What do consumers (now sophisticated) do?
  – Stay with Card A
    * Borrow for 6 months at $0
    * Then switch to another company
• No debiasing in equilibrium
• System of transfers:
  – Firms take advantage of naive consumers
  – Sophisticated consumers benefit from naive consumers

• Related: Suppose Credit Card B can identify naive consumer
  – What should it do?
  – If debias, then lose consumer
  – Rather, take advantage of consumer
• **Argument 2.** Even if experience or debiasing do not eliminate the biases, the biases will not affect aggregate market outcomes
  
  – Arbitrage $\implies$ Rational investors set prices
  
  – However, limits to arbitrage (DeLong et al., 1991) $\implies$ individuals with non-standard features affect stock prices
  
  – In addition, in most settings, there is no arbitrage!
    
    * Example: Procrastination of savings for retirement
    
    * (Keep in mind SMRT plan though)
  
  – Behavioral IO: Non-standard features can have a disproportionate impact on market outcomes
    
    * Firms focus pricing on the biases
    
    * Lee and Malmendier (2007) on overbidding in eBay auctions
Bidders with bias have *disproportionate* impact

Opposite of Chicago intuition

<table>
<thead>
<tr>
<th>Table V. Disproportionate Influence of Overbidders</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
</tr>
<tr>
<td><strong>Auction-level sample</strong></td>
</tr>
<tr>
<td>Does the auction end up overbid?</td>
</tr>
<tr>
<td>No</td>
</tr>
<tr>
<td>78</td>
</tr>
<tr>
<td>(Percent)</td>
</tr>
<tr>
<td>56.52%</td>
</tr>
<tr>
<td>Yes</td>
</tr>
<tr>
<td>60</td>
</tr>
<tr>
<td>Total</td>
</tr>
<tr>
<td>138</td>
</tr>
<tr>
<td>100.00%</td>
</tr>
<tr>
<td><strong>Bidder-level sample</strong></td>
</tr>
<tr>
<td>Does the bidder ever overbid?</td>
</tr>
<tr>
<td>No</td>
</tr>
<tr>
<td>670</td>
</tr>
<tr>
<td>(Percent)</td>
</tr>
<tr>
<td>83.02%</td>
</tr>
<tr>
<td>Yes</td>
</tr>
<tr>
<td>137</td>
</tr>
<tr>
<td>Total</td>
</tr>
<tr>
<td>807</td>
</tr>
<tr>
<td>100.00%</td>
</tr>
<tr>
<td><strong>Bid-level sample</strong></td>
</tr>
<tr>
<td>Is the bid an over-bid?</td>
</tr>
<tr>
<td>No</td>
</tr>
<tr>
<td>2,101</td>
</tr>
<tr>
<td>(Percent)</td>
</tr>
<tr>
<td>89.29%</td>
</tr>
<tr>
<td>Yes</td>
</tr>
<tr>
<td>252</td>
</tr>
<tr>
<td>Total</td>
</tr>
<tr>
<td>2,353</td>
</tr>
<tr>
<td>100.00%</td>
</tr>
</tbody>
</table>

Overbidding is defined using the final price.
7 Summary of Evidence

- Update type of evidence encountered so far

- Empirical evidence of type 1 (DellaVigna and Malmendier, 2006; Odean, 1999; Sydnor, 2009):

- **Menu choice.** Need to observe:
  - menu of options
  - later utilization
  - Use revealed preferences to make inferences from contract choice in (a)
  - Compare to actual utilization in (b)
  - Worries: hard to distinguish unusual preferences (self-control) and wrong beliefs (naïveté, overconfidence)
• Simple example.
  – Agent can choose action $X_1$ or $X_2$
  – Upon choice of $X_i$, agent chooses $x_i$

• Prediction of standard theory:

  If Choose $X_1$, then $Eg(x_1) \geq \bar{g}$

• Consider consumers choosing $X_1$

• Choice of $x_1$ conditional on $X_1$ $\rightarrow$ Estimate $Eg(x_1)$

• Then, reject standard theory if

  \[ Eg(x_1) < \bar{g} \] among those choosing $X_1$
• DellaVigna and Malmendier (2006) on health clubs
  – Choice of
    * Monthly contract ($X_M$), lump-sum fee $L = $80
    * Pay-per-visit ($X_P$) at $p = $10
  – Observe number of visits $v_i$, upon choice of $X_i$.
  – Prediction of standard theory:
    
    \[
    \text{If Choose } X_M, \text{ then } E_M[v] \geq L/p
    \]
    
    (This is “if” statement, “only if” part does not hold)
  – Use data to estimate $E_M[v]$ and conclude
    
    \[
    E_M[v] < L/p
    \]
    
    \(\rightarrow\) Rejection of standard theory
• Empirical evidence of types 2 and 3 share same idea, with different identification strategies

• Observe two situations, treatment situation $T$ and control situation $C$

• Observe outcome $x_i$ ($i = T, C$)

• Comparative statics prediction of different models:
  – Standard model:
    \[
    Ex_T \leq Ex_C
    \]
  – Alternative model:
    \[
    Ex_T > Ex_C
    \]

• Compare empirically $Ex_T$ and $Ex_C$ to test standard vs. alternative model
Empirical evidence of type 2 (Benartzi and Thaler, 2004; Choi et al., 2001; Huberman and Regev, 2001; Madrian and Shea, 1999; Wolfers and Zitzewitz, 2003):

**Natural Experiments**

- At time $t$, change in regime
  - Simple difference: Look at $(\text{After } t - \text{Before } t)$
  - Double Difference: Look at $(\text{After } t - \text{Before } t)_{Treatment} - (\text{After } t - \text{Before } t)_{Control}$
- Worries:
  - Endogeneity of change
  - Other changes occurring at same time
  - How many observations? Maybe $n = 1$?
Empirical evidence of type 3 (Ariely and Wertenbroch, 2002; Ausubel, 2004; Duflo and Saez, 2003; Falk and Ichino, 2004; Fehr and Goette, 2004; Hossain and Morgan, 2003; List's work):

Field experiments

- Naturalistic setting

- Explicitly Randomize treatment
  * Plus: Randomization ensures clean identification
  * Plus: Inference takes place in the field
  * Minus: Costly to run → Sample usually small
• Empirical evidence of type 4 (Barber and Odean, 2004; Camerer et al., 2001; DeGeorge et al., 1999; Farber, 2004; Genesove and Mayer, 2003; Malmendier and Tate, 2004; Odean, 1998):

• Correlational studies
  – Variables $x$ and $y$. Standard theory predicts
    $$ Cov(x, y) \geq 0 $$
  – Behavioral theory predicts
    $$ Cov(x, y) < 0. $$
  – Most commonly available evidence
  – Minus: Hard to infer causality
  – Minus: Hard unless theory makes sign prediction on correlation
• Empirical evidence of type 5 (Laibson, Repetto, and Tobacman, 2006; Paserman, 2004; Fang and Silverman, 2006; Conlin, O’Donoghue, and Vogelsang, 2007; DellaVigna, List, and Malmendier, 2009):

• **Structural Identification**
  
  – Write down model
  
  – Test prediction based on theory
    * Minus: Often hard to know what is driving results
    * Minus: Very time-consuming
    * Plus: Can estimate underlying parameters \((\beta, \hat{\beta})\)
    * Plus: Can do welfare and policy evaluations
  
  – Compromise: Do calibrations
8 Concluding Remarks

- How to complete a dissertation and be (approximately) happy

1. Know yourself, and put yourself to work
   - Do you procrastinate?
   - Are you afraid of undirected research?
   - Not enough intuition?
   - Not enough technicality?
   - Work in team with a classmate!
2. Economics is about techniques, and about ideas

- *Rule 1. Study the techniques*

- Everyone needs a reasonable knowledge of:
  * Modelling skills (decisions, game theory, contracts)
  * Econometrics (asymptotics, applied metrics)
  * (At least) One Field (methodology, questions, previous research)
- Rule 2. Think of interesting ideas

- Start from new idea, not from previous papers. Ex.: Mas-Moretti on Safeway data

- Think of an idea that can fix a broken literature (Levitt). Ex.: Fehr-Goette on cab drivers

- Rule 3. Explore technique you need for idea

  * Idea come first

  * It will be much easier to learn technique once you have an interesting problem at hand
3. What are good ideas?

- 1% of GDP (Glaeser)
- New questions (better) or unknown answers
- Questions you care about (comparative advantage: List)
- Socially important topics, if you can
4. Look for occasions to learn:

- Attend seminars
- Attend job market talks
- Do not read too much literature
- Discuss ideas with peers, over lunch, with yourself
- Get started on some data set
- Be curious
5. Above all, do not get discouraged...

   - Unproductive periods are a fact of life

   - Ideas keep getting better (and economics becomes more fun) with exercise

   - Work hard

   - Keep up the exercise!