

# Current Topics in Finance

## *Week 1: Semiparametric Asset-Pricing Tests*

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# Background

**Oversimplification of** (*some moderately sized subset of*) **the asset-pricing literature:**

1. Write down fully specified model of the world
2. Plug some numbers in
3. See if you can replicate data, *especially* some puzzling aspect(s) of data

**What's the goal here?**

- ▶ Use asset prices as relevant info about the underlying structure of the world
- ▶ If your model explains the data, then maybe we've learned something about beliefs, preferences, stochastic processes we care about, ...

**My approach instead has been:**

- ▶ Make some general(ish) assumptions that allow for statements of the form:

data  $\implies$  underlying structure of the world

*without* fully specified model

- ▶ This is not my invention! Lots of extremely well-known papers take this form

# Background

## Review of one example: Hansen & Jagannathan (1991)

*[simpler version following Shiller (1982), Hansen (1982)]*

- ▶ Consider a single risky asset ( $i$ ) and risk-free asset ( $f$ )
- ▶ Pricing equation and Cauchy-Schwartz inequality give:

$$\mathbb{E}[M(R_i - R_f)] = 0$$

$$\iff \text{Cov}(M, R_i - R_f) + \mathbb{E}[M]\mathbb{E}[R_i - R_f] = 0$$

$$\begin{aligned}\iff \mathbb{E}[R_i - R_f] &= \frac{-\text{Cov}(M, R_i - R_f)}{\mathbb{E}[M]} = \frac{-\text{Corr}(M, R_i - R_f)\sigma(M)\sigma(R_i - R_f)}{\mathbb{E}[M]} \\ &\leq \frac{\sigma(M)\sigma(R_i - R_f)}{\mathbb{E}[M]},\end{aligned}$$

$$\text{so } \text{SR}(R_i - R_f) \equiv \frac{\mathbb{E}[R_i - R_f]}{\sigma(R_i - R_f)} \leq \frac{\sigma(M)}{\mathbb{E}[M]}$$

- ▶ Much more general form of equity premium puzzle: setting  $i$  = market,  $\mathbb{E}[M] = \frac{1}{R_f}$ , get lower bound for  $\sigma(M)$  around 0.4 (and can be made sharper)

# Background

## Question I became interested in:

- ▶ How restrictive is the assumption of rational expectations?
  - ▶ Can we come up with bounds on primitives under general conditions?
  - ▶ Yes, and bounds end up being informative
  - ▶ Either (a) RE doesn't hold, (b) condition under which bound is derived is strongly violated, or (c) risk aversion is extremely high
- ▶ Bound relates asset-price volatility to risk aversion required to rationalize that volatility
- ▶ Why use *volatility* as being informative about expectations?
  1. First became interested after “Taper Tantrum” mid-2013
  2. Long history of volatility bounds, following Shiller (1981)
  3. Encountered a useful starting point (Augenblick & Rabin, 2018) at the right time
- ▶ Done with (most of) throat-clearing — remainder of slides go through JMP

# Introduction

How restrictive is the assumption of **rational expectations** in asset markets?

► **Joint hypothesis problem**  $\implies$  no free answers

► Illustration: Volatility bounds [e.g., Shiller (1981)]:

$$P_t + \text{error} = \text{ex-post fundamental value}$$

$$\implies \text{Var}(P_t) < \text{Var}(\text{ex-post fundamental value}) \quad [\text{Theory}]$$

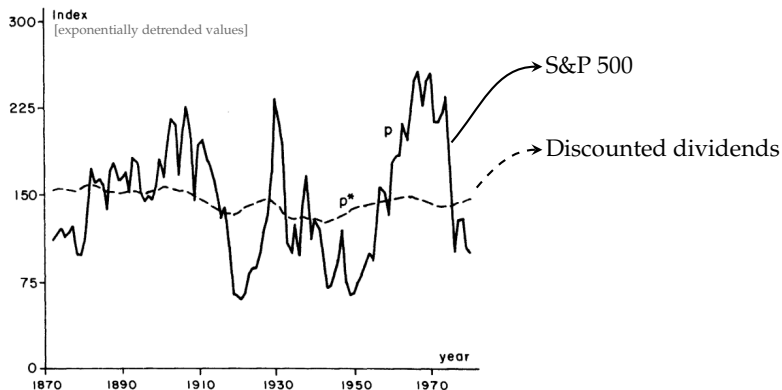
$$\text{Var}(P_t) > \text{Var}\left(\sum_{j=1}^{\infty} \frac{D_{t+j}}{R^j}\right) \quad [\text{Data}]$$

 constant discount rates

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 constant discount rates

► Response [Fama (1991)]:

*“Volatility tests are a useful way to show that expected returns vary, [but] give **no** help on the central issue of whether the variation in expected returns is rational.”*

► Can we say anything with weaker assumptions? **Yes.**

# Our Contributions

1. **Theory:** In general framework, derive bound under RE:

$$\text{Variation in market-implied beliefs} \leq f(\text{risk aversion})$$

[risk-neutral beliefs]                      [SDF slope]

2. **Evidence:**



# Our Contributions

1. **Theory:** In general framework, derive bound under RE:

$$\frac{\text{Variation in market-implied beliefs}}{[\text{second moment}]} \leq \frac{f(\text{risk aversion})}{[\text{SDF slope}]}$$

- ▶ N.B. comparison with Hansen–Jagannathan (1991) bound:

$$\frac{\text{Sharpe ratio for returns}}{[\text{first moment}]} \leq \frac{g(\text{risk aversion})}{[\text{SDF } \textit{volatility}]}$$

2. **Evidence:**

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2. **Evidence:**

- ▶ S&P index options
- ▶ Volatile risk-neutral beliefs  $\implies$  very high required risk aversion

## Possible objections:

- (a) Restriction:  $\mathbb{E}_t[U'_T \mid \text{return state } a] / \mathbb{E}_t[U'_T \mid \text{return state } b]$  constant over  $t$ 
  - ▶ Risk-neutral belief volatility is still an informative moment

# Our Contributions

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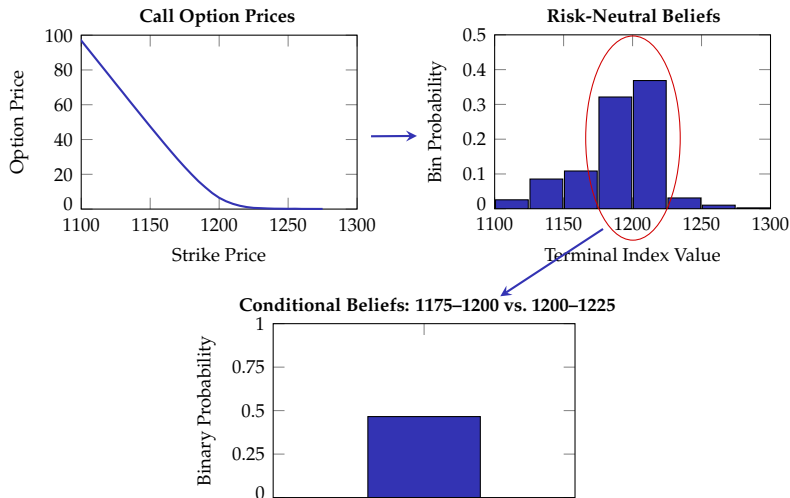
## **Possible objections:**

- (a) Restriction:  $\mathbb{E}_t[U'_T \mid \text{return state } a] / \mathbb{E}_t[U'_T \mid \text{return state } b]$  constant over  $t$
- (b) Measurement noise in option prices
  - ▶ Will address directly

# Preview: Empirical Variation

## S&P 500 Option Prices and Risk-Neutral Beliefs as of **July 1, 2005**

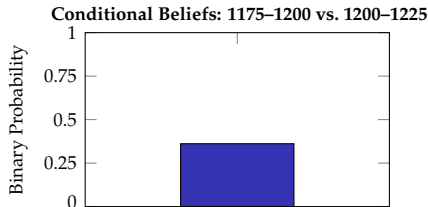
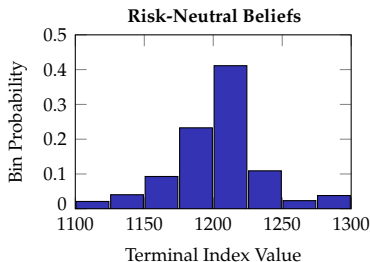
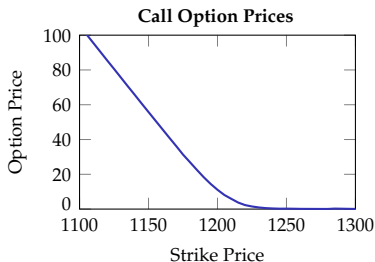
Expiration Date: July 16, 2005



# Preview: Empirical Variation

## S&P 500 Option Prices and Risk-Neutral Beliefs as of **July 5, 2005**

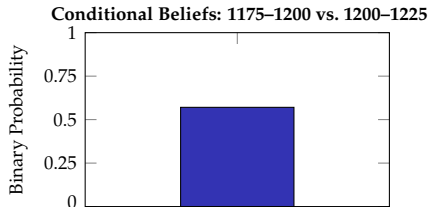
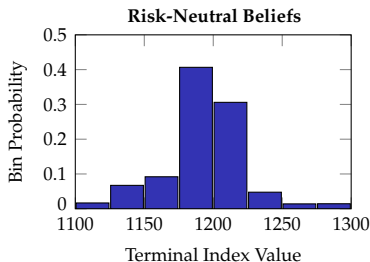
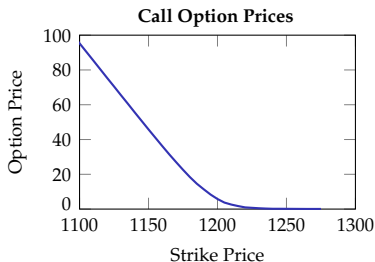
Expiration Date: July 16, 2005



# Preview: Empirical Variation

**S&P 500 Option Prices and Risk-Neutral Beliefs as of July 6, 2005**

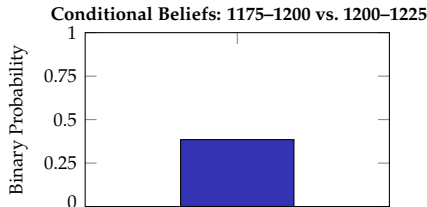
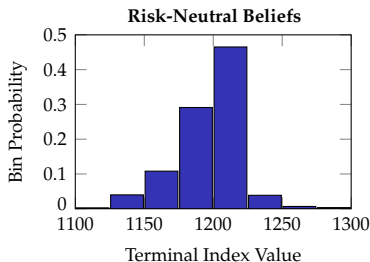
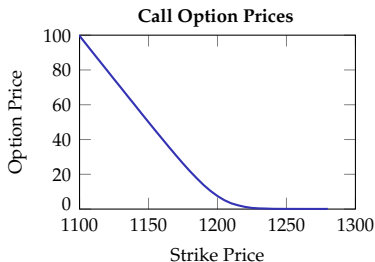
**Expiration Date: July 16, 2005**



# Preview: Empirical Variation

**S&P 500 Option Prices and Risk-Neutral Beliefs as of July 7, 2005**

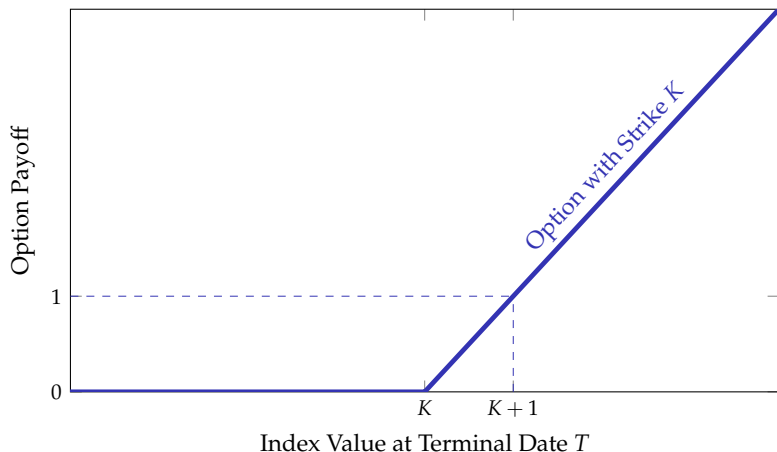
**Expiration Date: July 16, 2005**



# Graphical Intuition: Risk-Neutral Beliefs

Background: Index option prices  $\implies$  risk-neutral beliefs over future index price

- On date  $t$ , can buy or sell **call option** expiring at  $T$  with **strike  $K$** . Payoff:

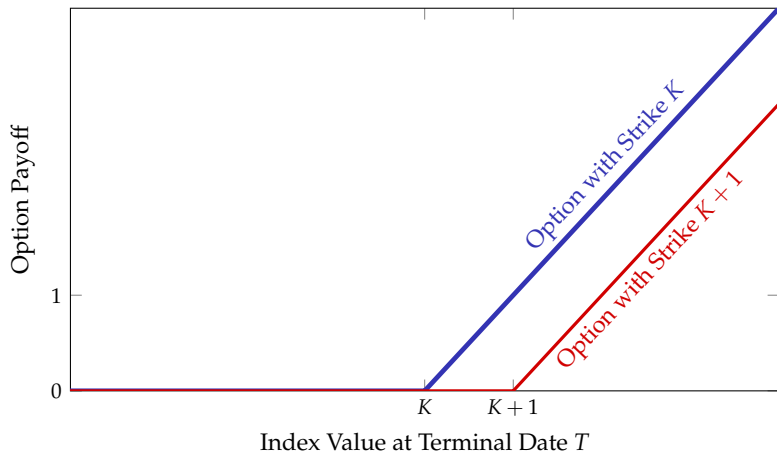




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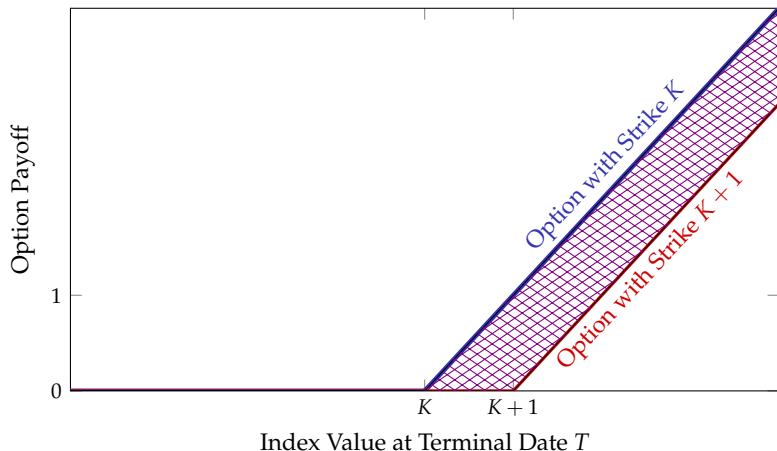
- ▶ On date  $t$ , can buy or sell **call option** expiring at  $T$  with **strike  $K$**
- ▶ With **strike  $K + 1$** ?



# Graphical Intuition: Risk-Neutral Beliefs

Background: Index option prices  $\implies$  risk-neutral beliefs over future index price

- **Payoff** to buying option with **strike  $K$**  + selling **strike  $K + 1$**   $\approx \mathbb{1}\{\text{Index}_T \geq K\}$   
 $\implies$  like an Arrow-Debreu security for the “state”  $\{\text{Index}_T \geq K\}$

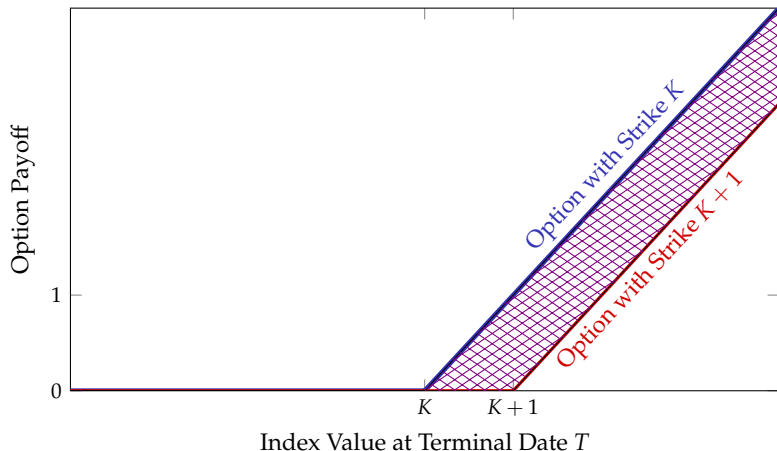


# Graphical Intuition: Risk-Neutral Beliefs

Background: Index option prices  $\implies$  risk-neutral beliefs over future index price

► **Payoff** to buying option with **strike  $K$**  + selling **strike  $K + 1$**   $\approx \mathbb{1}\{\text{Index}_T \geq K\}$

$$\implies \text{Price}_t \approx \mathbb{E}_t^*[\mathbb{1}\{\text{Index}_T \geq K\}] = \pi_t^*(\text{Index}_T \geq K)$$



# Outline

1. Background & Intro
2. Theory: Restrictions Under Rational Expectations
  - Two-State Example
  - General Bounds
3. Empirics: Evidence from Index Options
4. Discussion and Conclusions

# Two-State Example: Directly Observed Beliefs

Assumptions (*dropped in general framework*):

- ▶ Representative agent:
  1. Risk-neutral, no discounting
  2. Random terminal consumption:  $C_T = C_{\text{low}}$  or  $C_{\text{high}}$
- ▶ Beliefs:  $\pi_t \equiv$  subjective probability of state  $C_{\text{low}}$
- ▶ Rational expectations:  $\pi_t = \text{Prob}_t(C_T = C_{\text{low}})$
- ▶ Arrow-Debreu security with payoff  $\mathbb{1}\{C_T = C_{\text{low}}\}$ 
  - ▶ Risk neutrality  $\implies$  price  $= \pi_t$

# Two-State Example: Directly Observed Beliefs

Objects we'll keep track of:

1. **Belief movement:** 
$$m_{t_1, t_2} \equiv \sum_{t=t_1+1}^{t_2} (\pi_t - \pi_{t-1})^2$$

► “Volatility”  $\iff$  sum of squared changes in beliefs

2. **Uncertainty resolution:**

$$r_{t_1, t_2} \equiv \underbrace{(1 - \pi_{t_1})\pi_{t_1}}_{\text{uncertainty}_{t_1}} - \underbrace{(1 - \pi_{t_2})\pi_{t_2}}_{\text{uncertainty}_{t_2}}$$

► “Uncertainty”  $\iff$  variance of binomial RV, maximized at 0.5

**Restriction (Augenblick & Rabin, 2018):**

$$\text{Under RE, } \mathbb{E}[m_{t_1, t_2}] = \mathbb{E}[r_{t_1, t_2}].$$

Derivation

- Formalizes “correct” amount of volatility of subjective beliefs
- Derivation uses only martingale property of beliefs:  $\pi_t = \mathbb{E}_t[\pi_{t+1}]$

# Two-State Example: Directly Observed Beliefs

Objects we'll keep track of **for full path**:

1. **Belief movement:** 
$$m \equiv \sum_{t=1}^T (\pi_t - \pi_{t-1})^2$$

► “Volatility”  $\iff$  sum of squared changes in beliefs

2. **Uncertainty resolution:**

$$r \equiv \underbrace{(1 - \pi_0)\pi_0}_{\text{initial uncertainty}} - (1 - \pi_T)\pi_T \xrightarrow{\quad} 0$$

► “Uncertainty”  $\iff$  variance of binomial RV, maximized at 0.5

**Restriction (Augenblick & Rabin, 2018):**

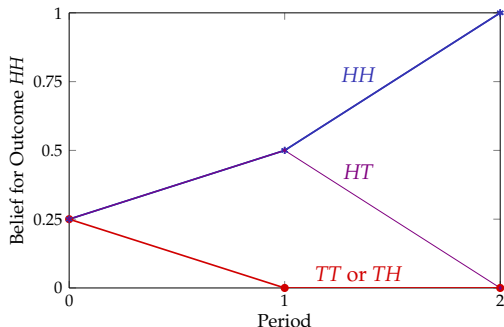
Under RE,  $\mathbb{E}[m] = r$ .

Derivation

- **Intuition:** Changing beliefs  $\iff$  must be learning something (on average)
- Violations can arise from too large (or small) belief revisions

# Two-State Example: Directly Observed Beliefs

$T = 2$ , sequential fair coin flips at  $t = 1$  and  $t = 2$ ,  $C_T = \begin{cases} C_{\text{low}} & \text{if } HH \\ C_{\text{high}} & \text{else} \end{cases}$



**Statistics:** Movement

$$m \equiv \sum_{t=1}^2 (\pi_t - \pi_{t-1})^2$$

Uncertainty Resolution

$$r \equiv (1 - \pi_0) \pi_0$$

**Restriction:**  $\mathbb{E}[m] = r$

Path	Movement	Frequency
HH	5/16	1/4
HT	5/16	1/4
T*	1/16	1/2

$$\left. \begin{array}{l} \text{HH} \\ \text{HT} \\ \text{T*} \end{array} \right\} \Rightarrow \mathbb{E}[m] = 3/16$$

$$= 3/4 \times 1/4 = r \quad \checkmark$$



# Two-State Example with Risk Aversion

**Problem:** Asset prices don't give us subjective beliefs (joint hypothesis problem).

**Assume now:**

- ▶ Utility:  $\mathbb{E}_0 \sum_{t=0}^T U(C_t)$ ,  $U'' < 0$
- ▶ Equilibrium: A-D security with payoff  $\mathbb{1}\{C_T = C_{\text{low}}\}$  now has price

$$q_t(C_{\text{low}}) = \frac{U'(C_{\text{low}})}{U'(C_t)} \pi_t$$

$\implies$  subjective beliefs no longer observable

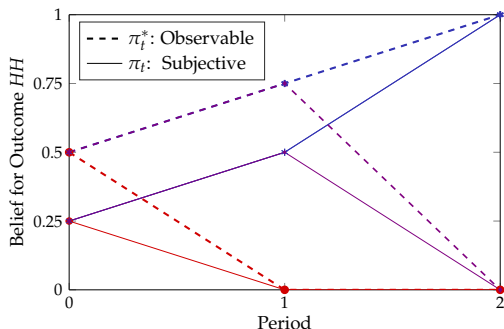
- ▶ Define *risk-neutral belief*:

$$\pi_t^* \equiv \frac{q_t(C_{\text{low}})}{q_t(C_{\text{low}}) + q_t(C_{\text{high}})} = \frac{U'(C_{\text{low}})}{\mathbb{E}_t[U'(C_T)]} \pi_t \geq \pi_t$$

- ▶ Need not follow a martingale under RE  $\implies$  can have  $\mathbb{E}[m^*] > \mathbb{E}[r^*]$

# Two-State Example with Risk Aversion

Back to coin-flip example, with  $U'(C_{\text{low}}) = 3 \times U'(C_{\text{high}})$ :



**Observables:** Movement

$$m^* \equiv \sum_{t=1}^2 (\pi_t^* - \pi_{t-1}^*)^2$$

► Simple variance swap payoff

Uncertainty Resolution

$$r^* \equiv (1 - \pi_0^*)\pi_0^*$$

► Risk-neutral variance

**Restriction:**  $\mathbb{E}[m^* - r^*] \leq ??$

Path	$m^*$	Frequency
HH	$\frac{5}{16} \quad \frac{1}{8}$	$\frac{1}{4}$
HT	$\frac{5}{16} \quad \frac{5}{8}$	$\frac{1}{4}$
$T^*$	$\frac{1}{16} \quad \frac{1}{4}$	$\frac{1}{2}$

$$\left. \begin{array}{l} \text{HH} \\ \text{HT} \\ T^* \end{array} \right\} \Rightarrow \begin{aligned} \mathbb{E}[m^*] &= 5/16 \\ &> (1 - 1/2) \times 1/2 \\ &= r^* \quad \times \end{aligned}$$

# Two-State Example: Result

How much excess movement can there be?

## Proposition

Under RE,

$$\mathbb{E}_0[m^* - r^*] \leq \underbrace{\pi_0^*}_{\text{room for downward movement}} (\underbrace{\pi_0^* - \pi_0}_{\text{indexes excess movement given downward revisions to beliefs}}) = \pi_0^* \left( \pi_0^* - \frac{\pi_0^*}{\pi_0^* + \underbrace{\phi}_{U'(C_{\text{low}})/U'(C_{\text{high}})}} (1 - \pi_0^*) \right)$$

indexes excess movement given  
downward revisions to beliefs

room for downward movement

# General Framework: Setup (I)

Now have to take into account that state space isn't binary.

**Probability space:**

- ▶ Discrete probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , filtration  $\mathbb{F} = \{\mathcal{F}_t\}_{t \in \mathbb{N}}$
- ▶ Ex-dividend value of market index:  $V_t^m: \Omega \rightarrow \mathbb{R}_+$
- ▶ Interested in its value on some option expiration date  $T$ 
  - ▶ Call option with strike  $K$  has payoff  $\max\{V_T^m - K, 0\}$
- ▶ Say that **return state**  $s \in \mathcal{S}$  is realized if

$$R_T^m \equiv \frac{V_T^m}{V_0^m} = s$$

- ▶ **Objective** probabilities of return states governed by  $\mathbb{P}: \mathcal{F} \rightarrow [0, 1]$

# General Framework: Setup (II)

## Stochastic discount factor:

- ▶ Absence of arbitrage  $\implies$  existence of strictly positive **stochastic discount factor (SDF)** process  $\{M_t\}$  s.t. price  $S_t$  of claim to random payoff  $X_T$  is

$$S_t(X_T) = \mathbb{E}_t \left[ \frac{M_T}{M_t} X_T \right]$$

- ▶ Standard representative-agent economy:  $\frac{M_{t+1}}{M_t} = \beta \frac{U'(C_{t+1})}{U'(C_t)}$
- ▶ **Risk-neutral measure** for  $T$ -dated payoffs:  $\left. \frac{d\mathbb{P}^*}{d\mathbb{P}} \right|_{\mathcal{F}_t} = \frac{M_T/M_t}{\mathbb{E}_t[M_T/M_t]}$

## Beliefs:

- ▶ Assume prices generated by some marginal trader observing public signals
- ▶ Agent observes signal vector  $\theta_t \in \Theta$ , with  $\mathcal{F}_t = \sigma(\theta_\tau, 0 \leq \tau \leq t)$
- ▶ **Signal-generating process**  $\mathbb{P}(\theta_t | \mathcal{F}_{t-1}, R_T^m)$  governs info on return states
- ▶ **Belief distribution** over return states:  $\{\pi_t(R_T^m = s)\}_{s \in \mathcal{S}}$

# General Framework: Setup (III)

## Definition

An agent has **rational expectations** over return states at  $T$  if and only if both:

- (i) Her date-0 priors coincide with the objective probabilities:

$$\pi_0(R_T^m = s) = \mathbb{P}_0(R_T^m = s) \quad \forall s \in \mathcal{S}.$$

- (ii) She updates according to **Bayes' rule** using the objective likelihood function:

$$\pi_t(R_T^m = s) = \frac{\pi_{t-1}(R_T^m = s) \mathbb{P}(\theta_t | \mathcal{F}_{t-1}, R_T^m = s)}{\mathbb{P}(\theta_t | \mathcal{F}_{t-1})}.$$

- Under RE, risk-neutral belief distribution:

$$\pi_t^*(R_T^m = s) = \frac{\mathbb{E}_t[M_T/M_t | R_T^m = s]}{\mathbb{E}_t[M_T/M_t]} \pi_t(R_T^m = s)$$

# General Framework: Restriction for Identification

- ▶ Work with **conditional** risk-neutral beliefs for state  $s_j$  vs.  $s_{j+1}$ :

$$\begin{aligned}\tilde{\pi}_{t,j}^* &\equiv \pi_t^*(R_T^m = s_j \mid R_T^m \in \{s_j, s_{j+1}\}) = \frac{\pi_t^*(R_T^m = s_j)}{\pi_t^*(R_T^m = s_j) + \pi_t^*(R_T^m = s_{j+1})} \\ \Rightarrow \frac{\tilde{\pi}_{t,j}^*}{1 - \tilde{\pi}_{t,j}^*} &= \underbrace{\frac{\mathbb{E}_t[M_T/M_t \mid R_T^m = s_j]}{\mathbb{E}_t[M_T/M_t \mid R_T^m = s_{j+1}]}}_{:= \phi_{t,j}} \frac{\tilde{\pi}_{t,j}}{1 - \tilde{\pi}_{t,j}}\end{aligned}$$

## Definition

The SDF satisfies **conditional transition independence (CTI)** for the return-state pair  $(s_j, s_{j+1})$  and option expiration date  $T$  if  $\phi_{t,j}$  is constant for all  $0 \leq t < T$  almost surely, and we denote this constant by  $\phi_j$ .

- ▶  $\tilde{\pi}_{t,j}^*$  changes must arise from subjective beliefs, not relative severity of states
- ▶ Requires that  $M_T/M_t$  in  $s_j$  depend in expectation only on  $s_j$  and not on the path of unobservable state variables between  $t$  and  $T$
- ▶  $\phi_j \iff$  utility curvature across states (roughly), and label states so that  $\phi_j \geq 1$

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## What models work?

- ▶ Variable rare disasters
- ▶ Long-run risks
- ▶ Habit formation



# General Framework: Restriction for Identification

- Work with **conditional** risk-neutral beliefs for state  $s_j$  vs.  $s_{j+1}$ :

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What models work?

- **Habit formation**

But permanent shocks to the SDF are admissible in all models.

# Bound in General Framework & Interpretation

## Proposition

$$\tilde{\mathbb{E}}_0[m_j^* - r_j^*] \leq \tilde{\pi}_{0,j}^2 \left( 1 - \frac{1}{\tilde{\pi}_{0,j}^* + \underbrace{\phi_j}_{\mathcal{U}'(C_{\text{low}})/\mathcal{U}'(C_{\text{high}})}} (1 - \tilde{\pi}_{0,j}^*) \right)$$

$\mathbb{E}_t[M_T \mid R_T^m = s_j] / \mathbb{E}_t[M_T \mid R_T^m = s_{j+1}]$

### General framework:

- ▶ **Setting:** Uncertainty over terminal value of market index,  $V_T^m$
- ▶ **State space:** Many return states  $\{s_j\}$  defined by  $R_T^m \equiv \frac{V_T^m}{V_0^m} = s_j$
- ▶ **Physical beliefs:**  $\tilde{\pi}_{t,j} \equiv \pi_t(R_T^m = s_j \mid R_T^m \in \{s_j, s_{j+1}\})$
- ▶ **Risk-neutral beliefs:** SDF  $\{M_t\} \implies \tilde{\pi}_{t,j}^* = \frac{\mathbb{E}_t[M_T \mid R_T^m = s_j]}{\mathbb{E}_t[M_T \mid R_T^m \in \{s_j, s_{j+1}\}]} \tilde{\pi}_{t,j}$
- ▶ **Identification restriction:**  $\phi_j$  is constant

Robustness & simulations

# Bound in General Framework & Interpretation

## Proposition

$$\tilde{\mathbb{E}}_0[m_j^* - r_j^*] \leq \tilde{\pi}_{0,j}^2 \left( 1 - \frac{1}{\tilde{\pi}_{0,j}^* + \underbrace{\phi_j}_{\mathcal{U}'(C_{\text{low}})/\mathcal{U}'(C_{\text{high}})}} (1 - \tilde{\pi}_{0,j}^*) \right)$$

$\mathbb{E}_t[M_T \mid R_T^m = s_j] / \mathbb{E}_t[M_T \mid R_T^m = s_{j+1}]$

### Features of bound and interpretation:

1. Relates **observable values** to **unobserved structural parameter** Discussion
2. Under risk neutrality ( $\phi_j = 1$ ): Bound becomes 0
3. Movement in risk-neutral beliefs still must correspond (on average) to agent learning about return, but now have inequality bound where  $\frac{\partial \text{bound}}{\partial \phi_j} > 0$
4. Identifies min. required  $\phi_j$ , which can be interpreted in terms of relative risk aversion if  $\exists$  rep. agent with utility over index level:

$$\gamma_j = \frac{\phi_j - 1}{(s_{j+1} - s_j)/s_j}$$

# Bound in General Framework & Interpretation

## Proposition

$$\mathbb{E}_0[m_j^* - r_j^*] \leq \tilde{\pi}_{0,j}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0,j}^* + \underbrace{\phi_j}_{\mathcal{U}'(C_{\text{low}})/\mathcal{U}'(C_{\text{high}})}} (1 - \tilde{\pi}_{0,j}^*) \right)$$

$\mathbb{E}_t[M_T \mid R_T^m = s_j] / \mathbb{E}_t[M_T \mid R_T^m = s_{j+1}]$

Taking  $\phi_j \rightarrow \infty$ :

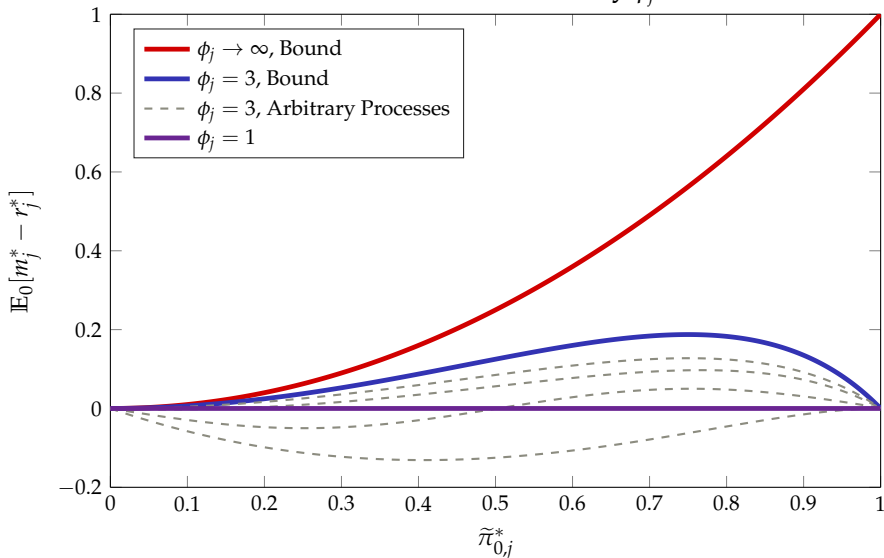
## Corollary

If  $\mathbb{E}_0[m_j^* - r_j^*] > \tilde{\pi}_{0,j}^{*2}$ , then **no** SDF process under which  $\phi_j$  is constant can rationalize the variation in risk-neutral beliefs for the given return-state pair.

- ▶ Can have so much excess vol. that no amount of risk aversion works
- ▶ Contrast with Hansen–Jagannathan bound

# Graphical Intuition: Bound in Proposition 1

Excess Belief Movement vs. Prior by  $\phi_j$  Under RE



# Outline

1. Background & Intro
2. Theory: Restrictions Under Rational Expectations
3. Empirics: Evidence from Index Options
  - Data
  - Baseline Results
  - Channels and Robustness
4. Discussion and Conclusions

# Raw Data and Risk-Neutral Beliefs

## Raw data:

- ▶ Want beliefs about return on market portfolio

⇒ Data on S&P 500 index option prices from OptionMetrics, 1996–2015

Details and cleaning

## Measuring risk-neutral beliefs from options:

- ▶ Breeden and Litzenberger (1978): Index price  $V_T^m$  has risk-neutral CDF

$$\mathbb{P}_t^*(V_T^m \leq v) = 1 + R_{t,T}^f \underbrace{\frac{\partial}{\partial v} q_t^m(v)}_{\text{option price at strike } v}$$

- ▶ Calculate  $\frac{\partial}{\partial v} q_t^m(v)$  numerically following Malz (2014)

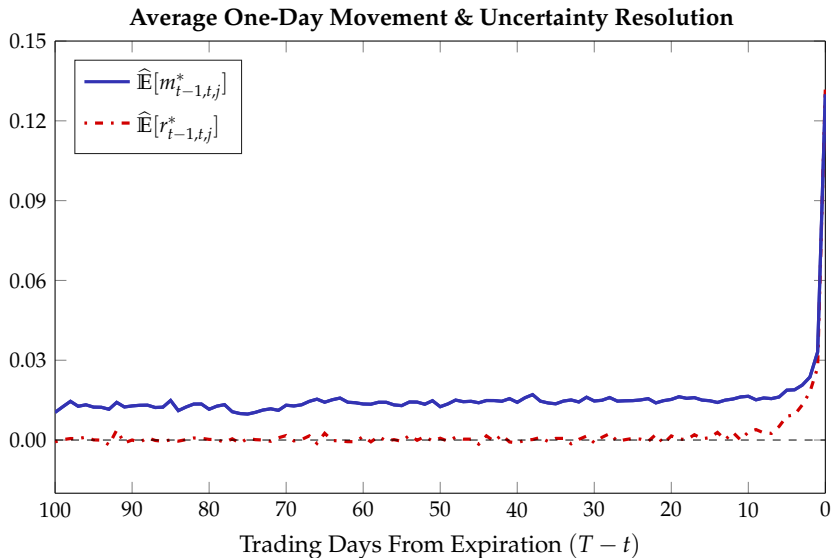
Details and alternative method

- ▶ **Excess-return space:**

$$\mathcal{S}_{\text{baseline}} = \exp\{(-\infty, -11\%), -9\%, -7\%, \dots, 7\%, 9\%, (11\%, \infty)\}$$

- ▶ Use beliefs over  $[-0.10, -0.08]$  for  $-9\%$  state, ...
- ▶ Generally use all states excluding  $(-\infty, -11\%), (11\%, \infty)$

# Summary: Risk-Neutral Belief Variation



*Note:* Empirical averages  $\hat{\mathbb{E}}[\cdot]$  calculated across all expiration dates and state pairs.



# Accounting for Market Microstructure Noise

## Proposition

Assume that observed  $\hat{\pi}_{t,j}^*$  is measured with error:

$$\hat{\pi}_{t,j}^* = \tilde{\pi}_{t,j}^* + \epsilon_{t,j},$$

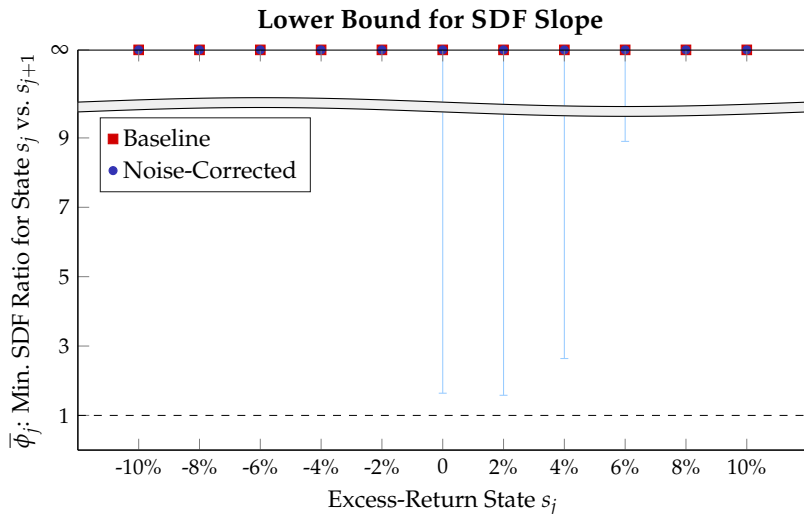
where  $\tilde{\mathbb{E}}[\epsilon_{t,j}] = 0$ ,  $\tilde{\mathbb{E}}[\epsilon_{t,j} \epsilon_{t+1,j}] = 0$ , and  $\tilde{\mathbb{E}}[\epsilon_{t,j} \tilde{\pi}_{t,j}^*] = 0$ . Denoting observed one-period expected excess movement by  $\tilde{\mathbb{E}}_t[\hat{m}_{t,t+1,j}^* - \hat{r}_{t,t+1,j}^*]$ , we have

$$\tilde{\mathbb{E}}_t[\hat{m}_{t,t+1,j}^* - \hat{r}_{t,t+1,j}^*] = \tilde{\mathbb{E}}_t[m_{t,t+1,j}^* - r_{t,t+1,j}^*] + 2\text{Var}(\epsilon_{t,j}).$$

To estimate  $\text{Var}(\epsilon_{t,j})$ , use auxiliary restriction that must hold under RE:

- ▶ Subjective forecasts must be **unbiased**:  $\tilde{\pi}_{t,j} = \mathbb{E}_t[\tilde{\pi}_{T,j}]$
- ▶ Since  $\tilde{\pi}_{T,j} = 0$  or 1, can solve for expected forecast-error variance
- ▶ **Steps**: Conjecture  $\phi_j \rightarrow$  translate from  $\tilde{\pi}_{t,j}^*$  to  $\tilde{\pi}_{t,j} \rightarrow$  calculate expected vs. realized forecast-error variance  $\rightarrow$  assume gap is measurement-error variance  $\rightarrow$  estimate bound  $\rightarrow$  continue until conjectured  $\phi_j =$  estimated  $\phi_j$

# Empirical Implementation of Theoretical Bound



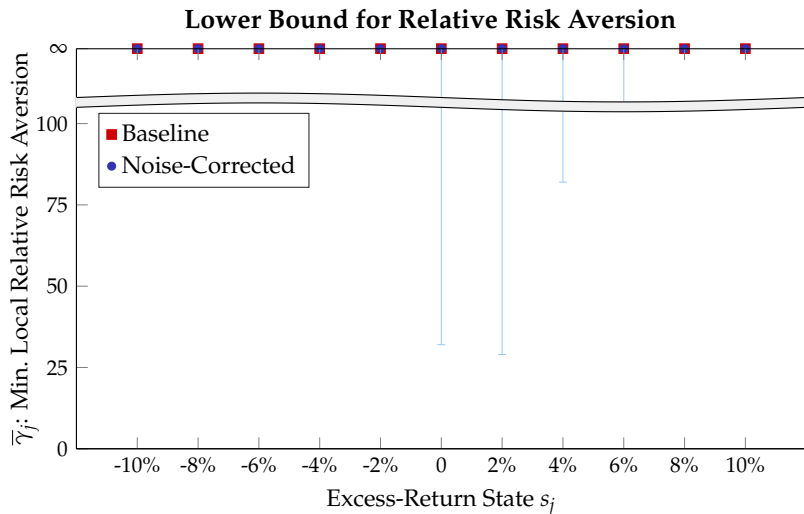
Note: One-sided 95% CIs use block bootstrap with bandwidth of 45 days, 5,000 draws.

[Details](#)

► Aggregate across interior states:  $\hat{\phi} = \infty$ , CI  $[3.0, \infty)$

[Formal justification](#)

# Estimation Results: Risk Aversion



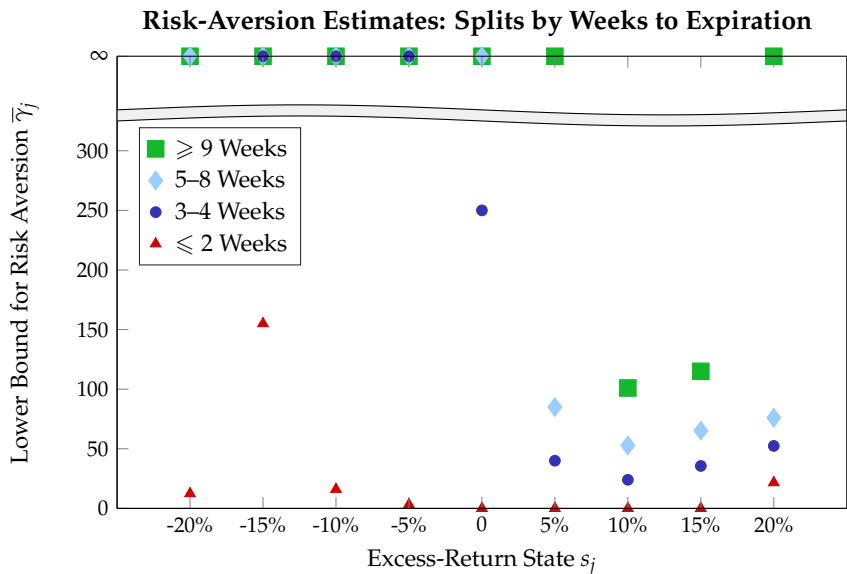
Note: One-sided 95% CIs use block bootstrap with bandwidth of 45 days, 5,000 draws.

[Details](#)

► Aggregate across interior states:  $\hat{\gamma} = \infty$ , CI  $[102, \infty)$

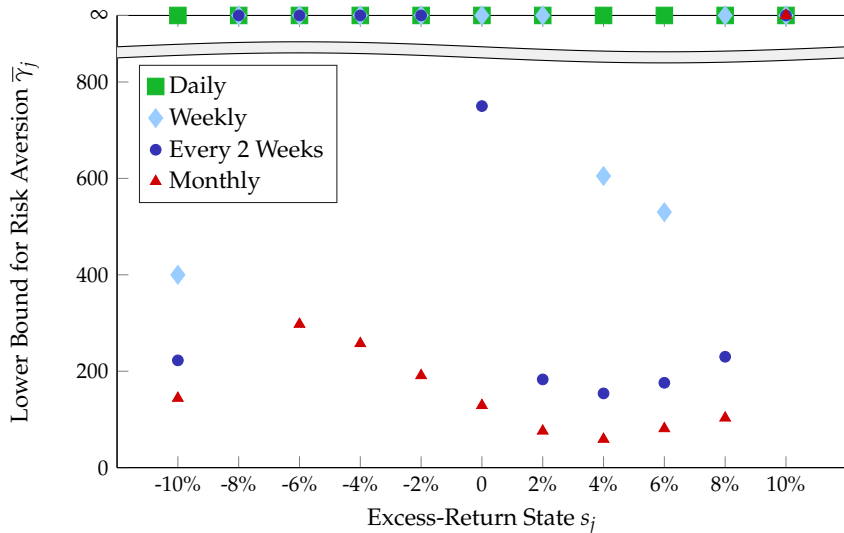
[Formal justification](#)

# Decompositions: Time to Expiration



# Decompositions: Effects of Time Aggregation

Risk-Aversion Estimates: Splits by Sampling Frequency



► Aggregate monthly:  $\hat{\gamma} = 123, \text{CI } [97, \infty)$

# Channels: Reduced-Form Evidence

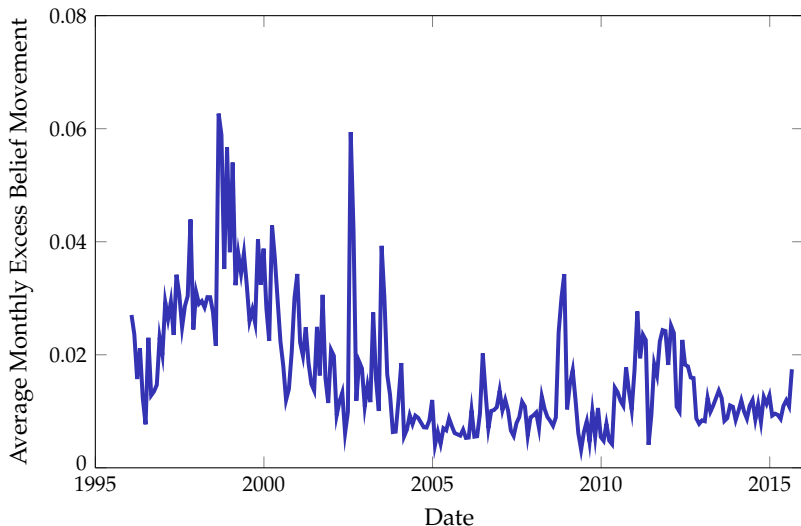
## Regressions for Average Excess Belief Movement by Quarter

	(1)	(2)	(3)	(4)
<b>Liquidity and Limits to Arbitrage</b>				
Bid-Ask Spread	0.2 (1.3)	-0.2 (-0.7)	-0.3* (-0.4)	-0.1 (-1.0)
Broker-Dealer Leverage	-0.1 (-0.4)	0.1 (0.8)	-0.0 (-0.7)	-0.1 (-1.7)
<b>Volatility and Uncertainty</b>				
VIX		0.8** (2.3)	0.9*** (3.4)	0.6* (2.1)
Baker-Bloom-Davis Uncertainty		-0.3 (-1.2)	0.1 (1.5)	0.2* (2.2)
<b>Returns and Valuation</b>				
12-Month S&P Return			0.3** (2.8)	0.3** (2.6)
Price to 10-Year Earnings Ratio			0.6*** (4.1)	0.5*** (4.1)
<b>Time Trend</b>				-0.0* (-2.2)
$R^2$	0.07	0.34	0.72	0.73

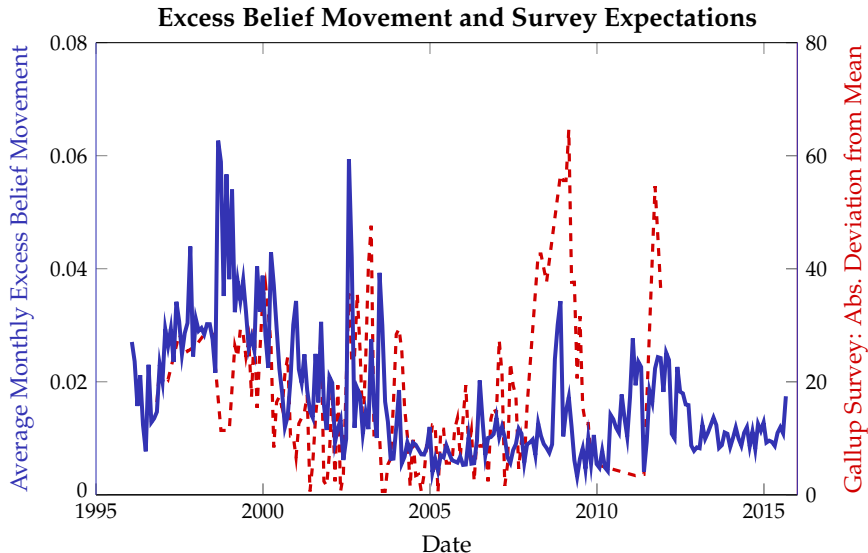
Notes: \*\*\*  $p < 0.01$ ; \*\*  $p < 0.05$ ; \*  $p < 0.1$ . Heteroskedasticity- and autocorrelation-robust  $t$ -statistics in parentheses, using equal-weighted periodogram estimator [Lazarus, Lewis, Stock (2017)]. All variables normalized to unit s.d. except for recession dummy and time trend.  $N = 79$ .

# Channel: Extrapolation $\Rightarrow$ Excess Volatility?

## Excess Belief Movement



# Channel: Extrapolation $\Rightarrow$ Excess Volatility?



*Note:* Gallup data (percent “bullish” less “bearish”) from Greenwood and Shleifer (2014).  
Correlation between two series in chart: 0.30.



# Robustness: Systematic Mean-Reversion vs. Noise

How real is what we're finding?

- Consider a simple statistical model for risk-neutral beliefs:

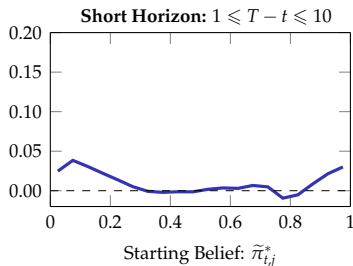
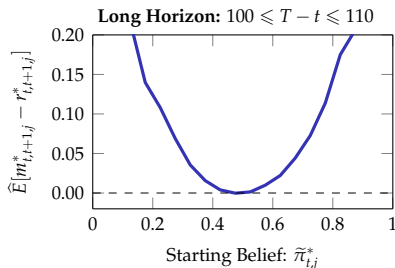
$$\tilde{\pi}_{t+1,j}^* = \mu + \rho(\tilde{\pi}_{t,j}^* - \mu) + v_{t+1}$$

- Setting  $\mu = 1/2$ , this model yields a prediction that:

$$\mathbb{E}[m_{t,t+1,j}^* - r_{t,t+1,j}^*] = 2(1 - \rho)(\tilde{\pi}_{t,j}^* - 1/2)^2$$

⇒ should see parabola for excess movement vs. prior

**Average Daily Excess Risk-Neutral Belief Movement by Starting Belief**



# Robustness: Conditional Transition Independence

- ▶ What if more than 1 state variable determines realization of the SDF (or MU)?
  - ▶ Then shocks to variables unspanned by asset return could violate CTI  
 $\implies \phi_{t,j}$  time-varying within contract

## Proposition

Recall  $\phi_{t,j} \equiv \mathbb{E}_t[M_T/M_t \mid R_T^m = s_j] / \mathbb{E}_t[M_T/M_t \mid R_T^m = s_{j+1}]$ .

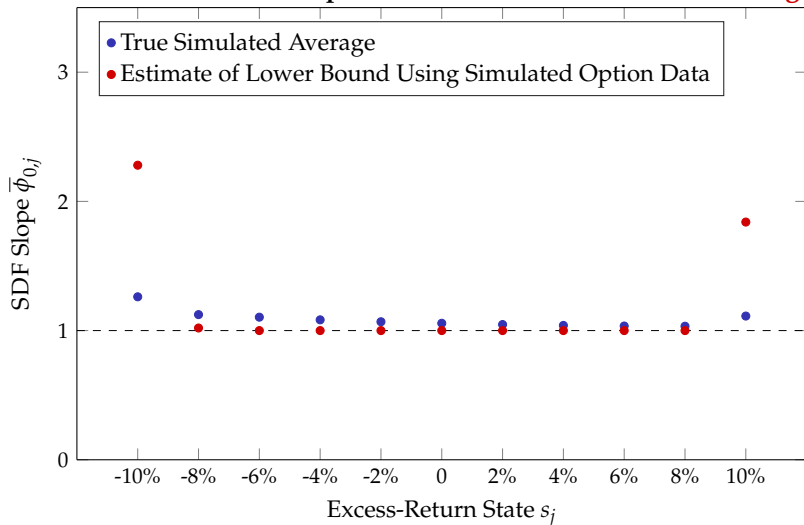
- (a) If  $\phi_{t,j}$  is a martingale, then the bound in Proposition 1 still applies.
- (b) The bound in Proposition 1 applies to an arbitrarily close approximation within a neighborhood of  $\phi_{t,j}$  being a martingale:  $\forall \epsilon > 0, \exists \delta > 0$  s.t. if  $|\tilde{\mathbb{E}}_t[\phi_{t+1,j}] - \phi_{t,j}| < \delta$  a.s., then:

$$\left| \tilde{\mathbb{E}}_0[m_j^* - r_j^*] - \tilde{\pi}_{0,j}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0,j}^* + \phi_{0,j}(1 - \tilde{\pi}_{0,j}^*)} \right) \right| < \epsilon.$$

- ▶ How good an approximation? For now, simulations: habit formation as in Campbell & Cochrane (1999)
- ▶ Not enough variation in  $\phi_{t,j}$  for simulations to match data, and bound still holds

# Robustness: Habit-Formation Simulation Results

## Estimates of SDF Slope: **Estimate vs. True Simulated Average**



Notes: 25,000 simulations of 90-day contracts;  $s_0$  drawn from unconditional distribution.

# Outline

1. Background & Intro
2. Theory: Restrictions Under Rational Expectations
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# Final Notes

## Summary:

- ▶ Examine restrictions on risk-neutral beliefs under RE
- ▶ Derive upper bounds for excess movement in risk-neutral beliefs, corresponding lower bounds on util. curvature needed to rationalize beliefs
- ▶ Associated empirical tests: very high curvature required to rationalize beliefs  
⇒ overreaction to information

## Possible positive models?

- ▶ Extrapolation and/or underweighting of prior relative to news?
  - ▶ Arrow (1982): *"[Evidence from Kahneman and Tversky] typifies very precisely the excessive reaction to current information which seems to characterize all the securities and futures markets."*
- ▶ Heterogeneous beliefs?

**Other:** Evidence from other asset classes (FI, FX), applications to real outcomes, simulations of additional models, ...

# Appendix

# Derivation of Lemma

Consider the conditional expectation of the first term in the movement sum:

$$\begin{aligned}\mathbb{E}_{t_1}[m_{t_1,t_1+1}] &= \mathbb{E}_{t_1}[(\pi_{t_1+1} - \pi_{t_1})^2], \quad \text{where } \mathbb{E}_t[\cdot] \equiv \mathbb{E}[\cdot \mid \mathcal{F}_t] \\ &= \mathbb{E}_{t_1}[\pi_{t_1+1}^2] - 2\mathbb{E}_{t_1}[\pi_{t_1+1}]\pi_{t_1} + \pi_{t_1}^2 \\ &= \mathbb{E}_{t_1}[\pi_{t_1+1}^2] - 2\pi_{t_1}\pi_{t_1} + \pi_{t_1}^2 \quad (\text{martingale prop. of Bayes' rule}) \\ &= \mathbb{E}_{t_1}[\pi_{t_1+1}^2] - \pi_{t_1}^2 + \pi_{t_1} - \mathbb{E}_{t_1}[\pi_{t_1+1}] \quad (\text{same}) \\ &= \mathbb{E}_{t_1}[(1 - \pi_{t_1})\pi_{t_1} - (1 - \pi_{t_1+1})\pi_{t_1+1}] = \mathbb{E}_{t_1}[r_{t_1,t_1+1}].\end{aligned}$$

Repeating for all periods and using L.I.E. yields the stated result. □

# Two-State Example: Non-Constant Discount Rates

**What would using the underlying price [Shiller (1981)] give us?**

- ▶ Consider extreme DGP: No info revealed until date  $T$ , so  $\pi_0^* = \dots = \pi_{T-1}^*$
- ▶ Price of claim to  $C_T$  is

$$\mathbb{E}_t \left[ \beta^{T-t} \frac{U'(C_T)}{U'(C_t)} C_T \right]$$

- ▶ Consider deterministic consumption stream  $C_0 \neq C_1 \neq \dots$
- $\Rightarrow$  *arbitrary* price variation as  $C_t$  changes, but *no* variation in expected payoff  $C_T$
- ▶ Paper discusses cases with time-varying risk premia



# Main Results: Implementation

$$\mathbb{E}_{0_i}[m_{T_{i,j}}^* - r_{T_{i,j}}^*] \leq \underbrace{\tilde{\pi}_{0_i, T_{i,j}}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0_i, T_{i,j}}^* + \phi_{i,j}(1 - \tilde{\pi}_{0_i, T_{i,j}}^*)} \right)}_{:= mr_{UB,i,j}^*}$$

## Issue:

- ▶ Only observe one draw  $m_{T_{i,j}}^* - r_{T_{i,j}}^*$  per contract, rather than  $\mathbb{E}_{0_i}[m_{T_{i,j}}^* - r_{T_{i,j}}^*]$
- ▶ But  $\frac{\partial^2 mr_{UB,i,j}^*}{\partial \phi_{i,j}^2} < 0$ , so Jensen's inequality (and L.I.E.) give that:

$$\begin{aligned} \mathbb{E}[m_{T_{i,j}}^* - r_{T_{i,j}}^*] &\leq \mathbb{E} \left[ \tilde{\pi}_{0_i, T_{i,j}}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0_i, T_{i,j}}^* + \phi_{i,j}(1 - \tilde{\pi}_{0_i, T_{i,j}}^*)} \right) \right] \\ &\leq \mathbb{E} \left[ \tilde{\pi}_{0_i, T_{i,j}}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0_i, T_{i,j}}^* + \mathbb{E}[\phi_{i,j}](1 - \tilde{\pi}_{0_i, T_{i,j}}^*)} \right) \right], \end{aligned}$$

where expectation is over all expiration dates  $T_i$

- ▶ So min.  $\mathbb{E}[\phi_{i,j}]$  solving inequality is lower bound of average ratio of SDF across states  $\implies$  info on reasonableness of pricing model required under RE
- ▶ If no such  $\mathbb{E}[\phi_{i,j}]$  exists, then no SDF or risk aversion can explain  $\mathbb{E}[m_{T_{i,j}}^* - r_{T_{i,j}}^*]$

# Raw Data: Details and Cleaning

## Details of data:

- ▶ End-of-day prices for calls and puts, Jan. 1996–Aug. 2015, all traded strikes  $\implies$  685 expiration dates  $T_i$ ; 4,949 trading dates; 7,385,062 option prices
- ▶ Also obtain underlying index price, dividend yield, and risk-free rates from OptionMetrics, and hand-collect option settlement values from CBOE

## Data cleaning:

- ▶ Drop any options with: bids of 0, Black-Scholes implied vol. more than 100%, greater than 6 months to maturity [Constantinides, Jackwerth, Savov (2013)], and any trading date–expiration date combos with fewer than 3 listed prices
- ▶ Calculate end-of-day price as average of listed bid and ask prices
- ▶ Cleaning for conditional risk-neutral probabilities: to avoid noisy measurement, only use date–state pairs meeting  $\pi_{t,T_{i,j}}^* + \pi_{t,T_{i,j}+1}^* \geq 5\%$

# Spline Details

- ▶ Calculate  $\frac{\partial}{\partial v} q_{t,T_i}(v)$  numerically following Malz (2014):
  1. Transform call and put price schedules for each date—expiration date set into Black-Scholes implied volatilities
  2. Fit clamped cubic splines to interpolate implied vols between strike prices for both calls and puts
  3. Average the calculated call and put implied vols at 1,900 strike prices
  4. Invert Black-Scholes implied volatility function to transform resulting implied vols back into call prices
  5. Numerically difference the resulting smoothed call-price schedule
- ▶ We only use Black-Scholes implied vols for smoothing and then transform vols back into prices, so doesn't require Black-Scholes model to be correct
- ▶ “Clamped” cubic spline: Sets slope of implied vol schedule to be zero at boundary strike-price values, and sets all implied vols below minimum observed strike price to value at minimum price (likewise for max.)
- ▶ This guarantees monotonically decreasing and convex call price schedule, which maintains no-arbitrage restrictions
- ▶ This is an *interpolating* spline: passes through all observed data (or *knot*) points

# Alternative Smoothing Method

## Bliss and Panigirtzoglou (2004) spline:

- ▶ Natural spline in implied vol–delta ( $\Delta = \frac{\partial q}{\partial v}$ ) space, weighted by vega ( $v = \frac{\partial q}{\partial \sigma}$ )
- ▶ Smoothing spline: Penalizes squared second derivative of spline, with weight  $\lambda = 0.01$  relative to deviations from observed data
- ▶ Force horizontal implied vol extrapolation by adding pseudo-data three strike intervals above/below observed strikes with implied vols equal to min./max. observed values before calculating spline
- ▶ Use out-of-the-money options (puts below current forward price, calls above)

In both cases, we set any negative probability values to 0 and renormalize the distribution accordingly. We calculate Bliss and Panigirtzoglou (2004) values to check robustness of main results.

# Estimation and Inference

## Estimation:

- ▶ Define pricing-kernel ratio as before:  $\phi_{i,j} \equiv \frac{\mathbb{E}_t[M_{T_i}/M_t \mid R_{T_i}^m = s_j]}{\mathbb{E}_t[M_{T_i}/M_t \mid R_{T_i}^m = s_{j+1}]} \geq 1$
- ▶ Main result:  $\mathbb{E}[m_{T_{i,j}}^* - r_{T_{i,j}}^*] \leq \mathbb{E}\left[\tilde{\pi}_{0_i, T_{i,j}}^{*2} \left(1 - \frac{1}{\tilde{\pi}_{0_i, T_{i,j}}^* + \mathbb{E}[\phi_{i,j}](1 - \tilde{\pi}_{0_i, T_{i,j}}^*)}\right)\right]$
- ▶ Denote  $\bar{\phi}_j \equiv \min \left\{ \mathbb{E}[\phi_{i,j}] \right\}$  solving inequality
- ▶ Sample estimator for each state pair solves moment condition:

$$\frac{1}{N} \sum_{i=1}^N \left[ m_{T_{i,j}}^* - r_{T_{i,j}}^* - \tilde{\pi}_{0_i, T_{i,j}}^{*2} \left( 1 - \frac{1}{\tilde{\pi}_{0_i, T_{i,j}}^* + \hat{\phi}_j (1 - \tilde{\pi}_{0_i, T_{i,j}}^*)} \right) \right] = 0$$

## Inference:

- ▶ CIs using block bootstrap: Partition sample into blocks of length  $B$  days, then resample (with replacement) all paths in entire block & re-estimate
- ▶ Addresses possible temporal dependence and correlation across state pairs

# Interpretation: Priors or Updating?

- ▶ Recall: Testing null of *both* correct prior and correct updating
- ▶ Are we really finding excess volatility of beliefs, or just miscalibrated priors?

## Proposition

Assuming CTI holds for the return-state pair  $(s_j, s_{j+1})$ , the effects of an incorrect physical prior,  $\tilde{\pi}_{0,j} \neq \mathbb{P}_0(R_T^m = s_j \mid R_T^m \in \{s_j, s_{j+1}\})$ , are limited as follows:

- (i) If  $\tilde{\pi}_{0,j}^* < \mathbb{P}_0(R_T^m = s_j \mid R_T^m \in \{s_j, s_{j+1}\})$ , then under Bayesian updating, it must be the case that  $\tilde{\mathbb{E}}_0[m_j^* - r_j^*] \leq \max\{\tilde{\pi}_{0,j}^{*2}, (1 - \tilde{\pi}_{0,j}^*)^2\}$ .
  - (ii) Otherwise, an incorrect prior cannot by itself lead to  $\tilde{\mathbb{E}}_0[m_j^* - r_j^*] > \tilde{\pi}_{0,j}^{*2}$ .
- ▶ Thus  $\hat{\phi}_j = \infty \implies$  likely excess volatility, not just bad priors