Motivation Experimental Setup Predictions Results Pretending To Analyze Historical Data Conclusion

# THE LUCAS MODEL OF ASSET PRICING: EXPERIMENTS

Elena Asparouhova, Peter Bossaerts, Nilanjan Roy, William Zame

Special Topics in Finance, Melbourne May 2014



Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

#### Motivation

Why experiments on the Lucas asset pricing model?

• underlies most of theoretical macro-finance

- underlies most of theoretical macro-finance
- gives clean predictions
  - cross-sectional
  - intertemporal
  - mutually reinforcing

- underlies most of theoretical macro-finance
- gives clean predictions
  - cross-sectional
  - intertemporal
  - mutually reinforcing
- tests with historical data assume equilibrium

- underlies most of theoretical macro-finance
- gives clean predictions
  - cross-sectional
  - intertemporal
  - mutually reinforcing
- tests with historical data assume equilibrium
  - focus on parametric variations (preferences, consumption, dividends...) of "stochastic Euler equations"

- underlies most of theoretical macro-finance
- gives clean predictions
  - cross-sectional
  - intertemporal
  - mutually reinforcing
- tests with historical data assume equilibrium
  - focus on parametric variations (preferences, consumption, dividends...) of "stochastic Euler equations"
  - weak empirical support

- underlies most of theoretical macro-finance
- gives clean predictions
  - cross-sectional
  - intertemporal
  - mutually reinforcing
- tests with historical data assume equilibrium
  - focus on parametric variations (preferences, consumption, dividends...) of "stochastic Euler equations"
  - weak empirical support
  - experiments can inform us about where the model works and where it potentially fails

# (Stochastic Euler Equations)

$$\beta E \left\{ \frac{\frac{\partial u_i(c_i(t+1))}{\partial c}}{\frac{\partial u_i(c_i(t))}{\partial c}} \left[ p(t+1) + d(t+1) \right] | I(t) \right\} = p(t),$$

# (Stochastic Euler Equations)

$$\beta E \left\{ \frac{\frac{\partial u_i(c_i(t+1))}{\partial c}}{\frac{\partial u_i(c_i(t))}{\partial c}} \left[ p(t+1) + d(t+1) \right] | I(t) \right\} = p(t),$$

• Take historical price and consumption data

# (Stochastic Euler Equations)

$$\beta E \left\{ \frac{\frac{\partial u_i(c_i(t+1))}{\partial c}}{\frac{\partial u_i(c_i(t))}{\partial c}} \left[ p(t+1) + d(t+1) \right] | I(t) \right\} = p(t),$$

- Take historical price and consumption data
- Fit equations for a choice of utility and information

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

## Some Objections

Why lab test test a model that is 'obviously wrong' in the field?

## Some Objections

Why lab test test a model that is 'obviously wrong' in the field?

• Why lab test a model that is 'obviously right' in the field?

## Some Objections

Why lab test test a model that is 'obviously wrong' in the field?

- Why lab test a model that is 'obviously right' in the field?
- Why is the model wrong in the field?

## Some Objections

#### Why lab test test a model that is 'obviously wrong' in the field?

- Why lab test a model that is 'obviously right' in the field?
- Why is the model wrong in the field?
- Models are idealizations; the laboratory is an opportunity to test them in an idealized environment.

 "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."

- "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."
  - People are risk averse in the lab. This is a fact.

- "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."
  - People are risk averse in the lab. This is a fact.
    - Whether this is decreasing marginal utility, as in the theory, remains to be seen

- "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."
  - People are risk averse in the lab. This is a fact.
    - Whether this is decreasing marginal utility, as in the theory, remains to be seen
    - But decreasing marginal utility explains phenomena at the market level

- "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."
  - People are risk averse in the lab. This is a fact.
    - Whether this is decreasing marginal utility, as in the theory, remains to be seen
    - But decreasing marginal utility explains phenomena at the market level
- Important message from our work:

individual *↔* market



- "The model relies on risk aversion. Nobody should be risk averse in the lab, so the model cannot possibly be right in the lab."
  - People are risk averse in the lab. This is a fact.
    - Whether this is decreasing marginal utility, as in the theory, remains to be seen
    - But decreasing marginal utility explains phenomena at the market level
- Important message from our work:
  - individual *↔* market
- Contrast economic thinking/social choice thinking



Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

Motivation
Experimental Seture
Predictions
Results
Pretending To Analyze Historical Data

# How one SHOULD think about the experiments (We think)

 We will observe many features that are "defining" for the Lucas model (Pareto efficiency, cross-sectional and intertemporal pricing)

- We will observe many features that are "defining" for the Lucas model (Pareto efficiency, cross-sectional and intertemporal pricing)
- Yet at the same time we observe phenomena that are exactly like in the field:
  - Excess volatility
  - Individuals hardly behave as predicted in the theory

- We will observe many features that are "defining" for the Lucas model (Pareto efficiency, cross-sectional and intertemporal pricing)
- Yet at the same time we observe phenomena that are exactly like in the field:
  - Excess volatility
  - Individuals hardly behave as predicted in the theory
- ... without having to invoke design elements that are claimed to be the reason for these phenomena in the field

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

- We will observe many features that are "defining" for the Lucas model (Pareto efficiency, cross-sectional and intertemporal pricing)
- Yet at the same time we observe phenomena that are exactly like in the field:
  - Excess volatility
  - Individuals hardly behave as predicted in the theory
- ... without having to invoke design elements that are claimed to be the reason for these phenomena in the field
  - Institutions (intermediaries, governments,...), Stochastics (ambiguity, rare events,...), Constraints (incomplete markets, collateral, indivisibilities, ...)

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

#### What we learn...

 Relative prices are correct, and intertemporally prices move with fundamentals

- Relative prices are correct, and intertemporally prices move with fundamentals
- ... but fundamentals explain only 18% of price changes (excess volatility)

- Relative prices are correct, and intertemporally prices move with fundamentals
- ... but fundamentals explain only 18% of price changes (excess volatility)
- Still, substantial Pareto improvements to autarky

- Relative prices are correct, and intertemporally prices move with fundamentals
- ... but fundamentals explain only 18% of price changes (excess volatility)
- Still, substantial Pareto improvements to autarky
- Subject price forecasts are "almost" fulfilled

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

# Take away...

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

# Take away...

• Messages:

### Take away...

- Messages:
  - For theorists: Investigate equilibria where agents make small forecast errors... they look very different from Lucas!

### Take away...

- Messages:
  - For theorists: Investigate equilibria where agents make small forecast errors... they look very different from Lucas!
  - For empiricists: Euler equations might be misguided (because they assume prices are functions of fundamentals only)

### Take away...

#### Messages:

- For theorists: Investigate equilibria where agents make small forecast errors... they look very different from Lucas!
- For empiricists: Euler equations might be misguided (because they assume prices are functions of fundamentals only)
- For policy: excess volatility does not stand in the way of significant Pareto improvements

Standard treatment of the Lucas Model starts with Pareto efficient allocations

 (the equilibrium equations are really first-order conditions of the representative agent that one can construct because of Pareto efficiency!)

- (the equilibrium equations are really first-order conditions of the representative agent that one can construct because of Pareto efficiency!)
- what economic structure could potentially generate Pareto efficiency... while allowing for heterogeneity across agents?

- (the equilibrium equations are really first-order conditions of the representative agent that one can construct because of Pareto efficiency!)
- what economic structure could potentially generate Pareto efficiency... while allowing for heterogeneity across agents?
  - impossible to have a complete set of markets

- (the equilibrium equations are really first-order conditions of the representative agent that one can construct because of Pareto efficiency!)
- what economic structure could potentially generate Pareto efficiency... while allowing for heterogeneity across agents?
  - impossible to have a complete set of markets
  - maybe use dynamic completeness and induce a Radner equilibrium? (Duffie-Huang [1985])



### Setting

• stationary (in dividend levels!), infinite horizon

### Setting

- stationary (in dividend levels!), infinite horizon
- two long-lived assets

### Setting

- stationary (in dividend levels!), infinite horizon
- two long-lived assets
  - tree: pays 0 (bad state) 1 (good state) each period probability p = 0.5 (i.i.d.)
  - bond: pays \$0.50 each period

• two (types of) infinitely lived agents

- two (types of) infinitely lived agents
- endowments

- two (types of) infinitely lived agents
- endowments
  - type I
    - 10 trees, 0 bonds
    - income: 15 even periods, 0 odd periods

- two (types of) infinitely lived agents
- endowments
  - type I
    - 10 trees, 0 bonds
    - income: 15 even periods, 0 odd periods
  - type II
    - 0 trees, 10 bonds
    - income: 15 odd periods, 0 even periods

Motivation Experimental Setup Predictions Results Pretending To Analyze Historical Data Conclusion

The Economy
Price Formation
Experimental Timeline
More Design

Why these parameters ???

### Why these parameters ???

• stationary-in-levels (time-invariant) aggregates

### Why these parameters ???

- stationary-in-levels (time-invariant) aggregates
- promote trade (otherwise bubbles Duffy and Crockett [2013])

### Why these parameters ???

- stationary-in-levels (time-invariant) aggregates
- promote trade (otherwise bubbles Duffy and Crockett [2013])
- (may restrict shortsales)

# How Will Prices Be Formed? Trade Through Continuous Electronic Open Book...

# How Will Prices Be Formed? Trade Through Continuous Electronic Open Book...



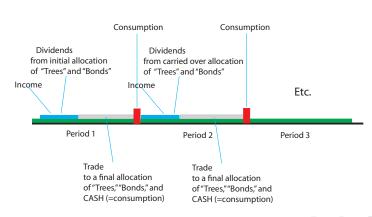
## (Graphical Display Of Book Of Orders)

## (Graphical Display Of Book Of Orders)



### Experimental timeline

### Experimental timeline



### Novel Design Solutions

#### Problems:

discounting in the lab

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab
- stationarity: end of lab session...

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab
- stationarity: end of lab session...

#### Solutions:

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab
- stationarity: end of lab session...

#### Solutions:

random termination

### Novel Design Solutions

#### Problems:

- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab
- stationarity: end of lab session...

#### Solutions:

- random termination
- pay subjects only cash of last period (intermediate payoffs are forfeited)

### Novel Design Solutions

#### Problems:

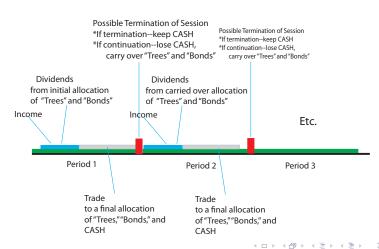
- discounting in the lab
- consumption ("cash") smoothing in the lab
- infinite horizon in the lab
- stationarity: end of lab session...

#### Solutions:

- random termination
- pay subjects only cash of last period (intermediate payoffs are forfeited)
- termination rule: at -10 minutes: reduce to 2-periods ending probabilities = 1/6, 5/6 (exploits separability, iid dividends)

### Back To Experimental Timeline

### Back To Experimental Timeline



90 Q

#### **Equilibrium Notion** Prices Numerical Example – Homogeneity

# (Radner) Equilibrium

### (Radner) Equilibrium

• asset prices p and consumption ("cash") plans  $c^I, c^{II}$ 

## (Radner) Equilibrium

- asset prices p and consumption ("cash") plans  $c^{I}$ ,  $c^{II}$
- such that
  - agents optimize subject to budget constraints
  - markets clear

## (Radner) Equilibrium

- asset prices p and consumption ("cash") plans  $c^{I}$ ,  $c^{II}$
- such that
  - agents optimize subject to budget constraints
  - markets clear
- Agents know structure of uncertainty, true probabilities, asset payoffs, own endowments, utility function, current asset prices

## (Radner) Equilibrium

- asset prices p and consumption ("cash") plans  $c^{I}$ ,  $c^{II}$
- such that
  - agents optimize subject to budget constraints
  - markets clear
- Agents know structure of uncertainty, true probabilities, asset payoffs, own endowments, utility function, current asset prices
- ... and future asset prices

# (Radner) Equilibrium

- asset prices p and consumption ("cash") plans  $c^{I}$ ,  $c^{II}$
- such that
  - agents optimize subject to budget constraints
  - markets clear
- Agents know structure of uncertainty, true probabilities, asset payoffs, own endowments, utility function, current asset prices
- ... and future asset prices
- Equilibrium assumes perfect/correct forecasts!



Equilibrium Notion Prices Numerical Example – Homogeneity

- cross-sectional
  - tree is less expensive than bond tree expected rate of return higher than bond
    - higher consumption beta caution: this is an equilibrium prediction

- cross-sectional
  - tree is less expensive than bond tree expected rate of return higher than bond
    - higher consumption beta caution: this is an equilibrium prediction
- intertemporal
  - price levels correlate perfectly and positively with fundamentals (dividends of tree)

- cross-sectional
  - tree is less expensive than bond tree expected rate of return higher than bond
    - higher consumption beta caution: this is an equilibrium prediction
- intertemporal
  - price levels correlate perfectly and positively with fundamentals (dividends of tree)
- cross-sectional and intertemporal predictions reinforce each other

- cross-sectional
  - tree is less expensive than bond tree expected rate of return higher than bond
    - higher consumption beta caution: this is an equilibrium prediction
- intertemporal
  - price levels correlate perfectly and positively with fundamentals (dividends of tree)
- cross-sectional and intertemporal predictions reinforce each other
- (countercyclical equity premium, or cyclical discount of Tree price relative to Bond price)

Equilibrium Notion Prices Numerical Example – Homogeneity

#### **Allocations**

• dynamic completeness:

- dynamic completeness:
  - Pareto optimality

- dynamic completeness:
  - Pareto optimality
    - smoothing: agents fully insure income fluctuations

- dynamic completeness:
  - Pareto optimality
    - smoothing: agents fully insure income fluctuations
    - diversification: consumption is positively (rank) correlated across states (high, low dividend on tree)

Equilibrium Notion Prices Numerical Example – Homogeneity

- dynamic completeness:
  - Pareto optimality
    - smoothing: agents fully insure income fluctuations
    - diversification: consumption is positively (rank) correlated across states (high, low dividend on tree)
    - (If homothetic utilities: consumption shares are constant across states/periods)

- dynamic completeness:
  - Pareto optimality
    - smoothing: agents fully insure income fluctuations
    - diversification: consumption is positively (rank) correlated across states (high, low dividend on tree)
    - (If homothetic utilities: consumption shares are constant across states/periods)
  - (price risk is hedged)

Equilibrium Notion Prices Numerical Example – Homogeneity

#### Homogeneous Log Utility, $\beta = 5/6$

#### Homogeneous Log Utility, $\beta = 5/6$

 Prices and returns – Tree cheaper; Both assets cheaper in Low state; Countercyclical equity premium and pro-cyclical discount

State	Tree		Bond		Price	Equity
	Price	Return	Price	Return	Discount	Premium
High (H)	\$2.50	3.4%	\$3.12	-0.5%	\$0.62	3.9%
Low $(L)$	\$1.67	55%	\$2.09	49%	\$0.42	6%

Equilibrium Notion Prices Numerical Example – Homogeneity

#### Homogeneous Log Utility, $\beta = 5/6$

Equilibrium Notion Prices Numerical Example – Homogeneity

### Homogeneous Log Utility, $\beta = 5/6$

 Holdings and trading: Type I (receives income in Even periods and buys Trees to hedge price risk)

Period	Tree	Bond	(Total)
Odd	7.57	0.62	(8.19)
Even	2.03	7.78	(9.81)
(Trade in Odd)	(+5.54)	(-7.16)	(-1.62)

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging Individual Choices

#### Sessions/Replications

## Sessions/Replications

Session	Place	Replication Number	Periods (Total, Min, Max)	Subject Count
1	Caltech*	4	(14, 1, 7)	16
2	Caltech	2	(13, 4, 9)	12
3	UCLA*	3	(12, 3, 6)	30
4	UCLA*	2	(14, 6, 8)	24
5	Caltech*	2	(12, 2, 10)	20
6	Utah*	2	(15, 6, 9)	24
(Overall)		15	(80, 1, 10)	

(Starred sessions ended with prematurely halted replication)



Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging Individual Choices

# Tree cheaper; Both assets cheaper in low state; But discount counter-cyclical

# Tree cheaper; Both assets cheaper in low state; But discount counter-cyclical

Data	Tree Price	Bond Price	Discount (Bond - Tree)
Mean	2.75	3.25	0.50
St. Dev.	0.41	0.49	0.40
High (State)	2.91	3.34	0.43
Low (State)	2.66	3.20	0.54
Difference	0.24	0.14	-0.11
across states			

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging Individual Choices

# Discount (of tree price) and price differential across states are positively correlated

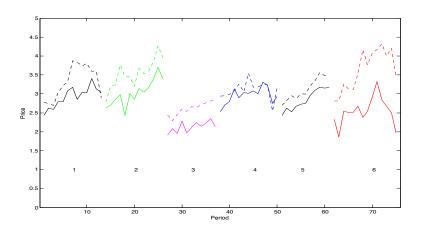
# Discount (of tree price) and price differential across states are positively correlated

Correlation is between the average (per replication)
difference between bond and tree price, and the average
(per replication) difference of prices (of a security)
between high and low states.

	Tree	Bond
Correlation	0.80	0.52
(St. Err.)	(0.40)	(0.40)

#### Prices move with fundamentals – but noisily

## Prices move with fundamentals - but noisily



# Apparent trend is not significant once allowing for influence of state (change)

# Apparent trend is not significant once allowing for influence of state (change)

Table 10: OLS regression of changes in period-average transaction prices. (\* = significant at p=0.05.)

Explanatory	Tree Price Change		Bond Price Change	
Variables	Estim.	(95%  Conf. Int.)	Estim.	(95%  Conf. Int.)
Change in State Dummy				
(None=0; High-to-Low=-1,	0.19*	(0.08, 0.29)	0.10	(-0.03, 0.23)
Low-to-High=+1)				
$R^2$		0.18		0.04
Autocor. (s.e.=0.13)		0.18		-0.19

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

Prices: Cross-Sectional
Prices: Temporal
Consumption Across Types
Price Hedging
Individual Choices

#### Results in Returns

#### Results in Returns

Table 7: Average returns across securities and states (High or Low aggregate dividend).

State	Tree	Bond	Equity
			Premium
High	12.8 (%)	15.9	-3.1
Low	17.8	16.1	1.7
Difference	-5.0	-0.2	-4.8
Average	16.1	16.0	0.1

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging

Significant smoothing and diversification across states – to extent that consumption shares are constant (mixed-effects two-factor ANOVA)

# Significant smoothing and diversification across states – to extent that consumption shares are constant (mixed-effects two-factor ANOVA)

	Sta	ites	Periods		
	High Low		Odd	Even	
Type I	14.93 (19.75)	7.64 (4.69)	7.69 (2.41)	13.91 (20.65)	
Type II	15.07 (10.25)	12.36 (15.31)	14.72 (20)	11.74 (5)	
ANOVA p:					
Factors	0.0	09	C	).27	
Interaction	0.23				

(Autarky cash holdings in parentheses)



Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data

Prices: Cross-Sectional
Prices: Temporal
Consumption Across Types
Price Hedging

Individual Choices

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging

### A closer look at trading

 Subjects did not hedge price risk (much) – they did not expect prices to move with fundamentals?

- Subjects did not hedge price risk (much) they did not expect prices to move with fundamentals?
- (Significant correlation between prices and fundamentals cannot easily be detected in 10-15 rounds)

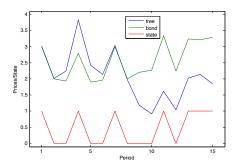
- Subjects did not hedge price risk (much) they did not expect prices to move with fundamentals?
- (Significant correlation between prices and fundamentals cannot easily be detected in 10-15 rounds)
- If agents do not expect prices to move with fundamentals, the resulting equilibrium is VERY different from Lucas model!

- Subjects did not hedge price risk (much) they did not expect prices to move with fundamentals?
- (Significant correlation between prices and fundamentals cannot easily be detected in 10-15 rounds)
- If agents do not expect prices to move with fundamentals, the resulting equilibrium is VERY different from Lucas model!
- ... but very much like in our experiments (stochastic drift, etc.)

Prices: Cross-Sectional Prices: Temporal Consumption Across Types Price Hedging

# Prices when agents do not expect prices to move with fundamentals

# Prices when agents do not expect prices to move with fundamentals



(Consumption share of Type I agent fluctuates between 39 and 44%.)

# Analysis of price expectations

## Analysis of price expectations

 Adam, Marcet and Nicolini (2012) also point out that even with only small mistakes in expectations about prices (assuming everyone knows underlying dividend processes!), equilibrium prices may look very different from the Lucas equilibrium – much more like in "the real world."

## Analysis of price expectations

- Adam, Marcet and Nicolini (2012) also point out that even with only small mistakes in expectations about prices (assuming everyone knows underlying dividend processes!), equilibrium prices may look very different from the Lucas equilibrium – much more like in "the real world."
- But Adam, Marcet and Nicolini (2012) do not point out that equilibrium allocations could still be pretty much the same as in the Lucas equilibrium – and close to optimal!

# Analysis of price expectations

- Adam, Marcet and Nicolini (2012) also point out that even with only small mistakes in expectations about prices (assuming everyone knows underlying dividend processes!), equilibrium prices may look very different from the Lucas equilibrium – much more like in "the real world."
- But Adam, Marcet and Nicolini (2012) do not point out that equilibrium allocations could still be pretty much the same as in the Lucas equilibrium – and close to optimal!
- ... because our agents trade consistent with their expectations, and their expectations are almost self-fulfilling?



### Individual choices are all over...

#### Individual choices are all over...

Table 12: End-Of-Period Asset Holdings Of Three Type I Subjects. Initial allocations: 10 Trees, 0 Bonds. Data from one replication in the first Caltech session.

Subject	1	2	3	4	5	6
Trees:						
3	4	4	3	4	3	4
5	1	1	0	1	1	3
7	7	10	13	15	19	20
Bonds:						
3	3	5	3	5	3	4
5	8	15	14	15	16	17
7	2	3	0	4	0	4

(So, individuals are not "representative" of what happens at the market level!)

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

#### GMM Tests...

• using returns and aggregate consumption data (only!)

- using returns and aggregate consumption data (only!)
- using traditional instruments

- using *returns* and aggregate consumption data (only!)
- using traditional instruments
- assuming power utility (wlog because we only have two states each period)

- using returns and aggregate consumption data (only!)
- using traditional instruments
- assuming power utility (wlog because we only have two states each period)

We should:

- using *returns* and aggregate consumption data (only!)
- using traditional instruments
- assuming power utility (wlog because we only have two states each period)

#### We should:

 reject (prices too volatile; discount on tree is countercyclical)

- using returns and aggregate consumption data (only!)
- using traditional instruments
- assuming power utility (wlog because we only have two states each period)

#### We should:

- reject (prices too volatile; discount on tree is countercyclical)
- find significant risk aversion

#### Results Sensitive To Instruments!

#### Results Sensitive To Instruments!

Table 15: GMM Estimation And Testing Results For Three Different Sets Of Instruments.

Instruments	β	$\gamma$	$\chi^2$ test
	(p value for $\beta = 5/6$ )	$(p \text{ value for } \gamma = 0)$	(p  value)
constant 1,	0.86	-0.01	7.124
lagged consumption growth,	(0.003)	(0.917)	(0.310)
lagged asset returns			
constant 1,	0.86	-0.18	0.731
lagged consumption growth	(0.029)	(0.162)	(0.694)
high state dummy,	0.86	0.16	14.349
low state dummy,	(0.002)	(0.001)	(0.006)
lagged consumption growth			

Motivation Experimental Setup Predictions Results Pretending To Analyze Historical Data Conclusion

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

#### Conclusions

 The cross-sectional pricing implications of the Lucas model are born out in the experimental data

- The cross-sectional pricing implications of the Lucas model are born out in the experimental data
- The intertemporal variation (predictability) in asset prices is far less than predicted (given cross-sectional difference).
   Prices exhibit excessive volatility.

- The cross-sectional pricing implications of the Lucas model are born out in the experimental data
- The intertemporal variation (predictability) in asset prices is far less than predicted (given cross-sectional difference).
   Prices exhibit excessive volatility.
- Subjects seem to have anticipated this and therefore reduce their demands to hedge against price risk; still, these anticipations are inconsistent in equilibrium (prices will – and do – depend on tree dividends even if this is not anticipated...)

- The cross-sectional pricing implications of the Lucas model are born out in the experimental data
- The intertemporal variation (predictability) in asset prices is far less than predicted (given cross-sectional difference).
   Prices exhibit excessive volatility.
- Subjects seem to have anticipated this and therefore reduce their demands to hedge against price risk; still, these anticipations are inconsistent in equilibrium (prices will – and do – depend on tree dividends even if this is not anticipated...)
- Nevertheless, the risk sharing properties of the Lucas equilibrium emerge: allocations are OK even if prices are excessively volatile.

Motivation Experimental Setup Predictions Results Pretending To Analyze Historical Data Conclusion

Motivation
Experimental Setup
Predictions
Results
Pretending To Analyze Historical Data
Conclusion

#### The Future

• What happens in the "long run"? (Incentive problems!)

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)
- The role of emotions

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)
- The role of emotions
  - To what extent are emotions part of the neoclassical math? (Prediction errors.) To what extent do they explain the variance of price changes not captured by the neoclassical model (> 80%)?

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)
- The role of emotions
  - To what extent are emotions part of the neoclassical math? (Prediction errors.) To what extent do they explain the variance of price changes not captured by the neoclassical model (> 80%)?
  - Contagion?

- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)
- The role of emotions
  - To what extent are emotions part of the neoclassical math? (Prediction errors.) To what extent do they explain the variance of price changes not captured by the neoclassical model (> 80%)?
  - Contagion?
- Introducing money...



- What happens in the "long run"? (Incentive problems!)
- Security design (to facilitate learning of future prices)
  - Introducing redundant securities such as options
  - Replacing consol bond with shorter-maturity options (optimal maturity of fixed-income securities?)
- The role of emotions
  - To what extent are emotions part of the neoclassical math? (Prediction errors.) To what extent do they explain the variance of price changes not captured by the neoclassical model (> 80%)?
  - Contagion?
- Introducing money...
- Introduce "rational bubbles"

