Motivation Experimental Setup Predictions Results Pretending To Analyze Historical Data Conclusion

THE LUCAS MODEL OF ASSET PRICING: EXPERIMENTS

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Motivation

Why experiments on the Lucas asset pricing model?

• underlies most of theoretical macro-finance

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- gives clean predictions
 - cross-sectional
 - intertemporal
 - mutually reinforcing

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 - cross-sectional
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- 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)

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

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

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Some Objections

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- 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.

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- Important message from our work:

individual *↔* market



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 - individual *↔* market
- Contrast economic thinking/social choice thinking



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How one SHOULD think about the experiments (We think)

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 - 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, ...)

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What we learn...

 Relative prices are correct, and intertemporally prices move with fundamentals

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- Still, substantial Pareto improvements to autarky
- Subject price forecasts are "almost" fulfilled

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Take away...

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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

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 - impossible to have a complete set of markets
 - maybe use dynamic completeness and induce a Radner equilibrium? (Duffie-Huang [1985])



Setting

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- 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

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The Economy
Price Formation
Experimental Timeline
More Design

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- (may restrict shortsales)

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

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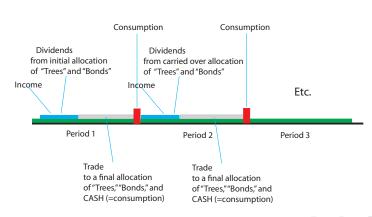
(Graphical Display Of Book Of Orders)

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Experimental timeline

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Novel Design Solutions

Problems:

discounting in the lab

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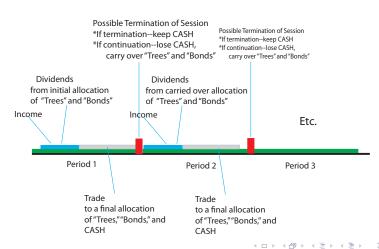
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Solutions:

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- 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

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90 Q

Equilibrium Notion Prices Numerical Example – Homogeneity

(Radner) Equilibrium

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- Equilibrium assumes perfect/correct forecasts!



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- (countercyclical equity premium, or cyclical discount of Tree price relative to Bond price)

Equilibrium Notion Prices Numerical Example – Homogeneity

Allocations

• dynamic completeness:

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 - (price risk is hedged)

Equilibrium Notion Prices Numerical Example – Homogeneity

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

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 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

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 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)

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

Sessions/Replications

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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)



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

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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			

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

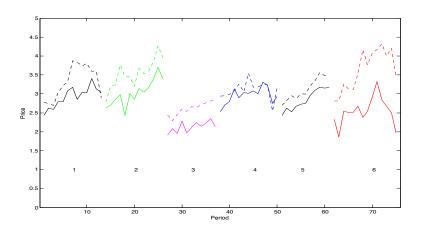
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Correlation is between the average (per replication)
difference between bond and tree price, and the average
(per replication) difference of prices (of a security)
between high and low states.

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

Prices move with fundamentals – but noisily

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

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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

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Prices: Temporal
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Results in Returns

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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)

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	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)



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Prices: Cross-Sectional
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A closer look at trading

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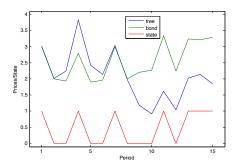
- Subjects did not hedge price risk (much) they did not expect prices to move with fundamentals?
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- If agents do not expect prices to move with fundamentals, the resulting equilibrium is VERY different from Lucas model!

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

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

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- 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...

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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!)

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GMM Tests...

• using returns and aggregate consumption data (only!)

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- using returns and aggregate consumption data (only!)
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We should:

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

Results Sensitive To Instruments!

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

Instruments	β	γ	χ^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			

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- 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.

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The Future

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- 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!)
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