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Discussion of

Explaining House Price Dynamics:

Isolating the Role of Nonfundamentals

Kevin J. Lansing, FRB San Francisco

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Overview	U.S. Data	Fundamental vs. Bubble	Comments	Summary

Discussion of "Explaining House Price Dynamics: Isolating the Role of Nonfundamentals" by D. Ling, J. Ooi, and T. Le

> Kevin J. Lansing¹ FRB San Francisco

FRB Dallas/JMCB Conference Housing, Stability, and the Macroeconomy: International Perspectives November 14, 2013

¹Any opinions expressed here do not necessarily reflect the views of the management of the Federal Reserve Bank of San Francisco or of the Board of Governors of the Federal Reserve System

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- Non-fundamental components of sentiment help to predict future house price changes <u>and</u> future sentiment levels.

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- Past house price changes help to predict future sentiment.
 - \Rightarrow Evidence of self-reinforcing feedback.

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Some re	lated finding	in the literature.		

• Investors' expected future returns from surveys are strongly correlated with past 12-month returns. (Case, Shiller & Thompson 2012).

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- Investors' expected returns from surveys are highest after sustained price run-ups, i.e., when price-dividend ratios (or price-rent ratios) are high. (Greenwood & Shleifer 2013).

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Self-reinforcing feedback:

 House prices rose faster in areas where lending standards were weakest, as measured by the prevalence of subprime/exotic mortgages or LTV of first-time home buyers. (Tal 2006, Wheaton and

Nechayov 2008, Mian & Sufi 2009, Pavlov & Wachter 2011, Duca, Muellbauer & Murphy 2012).

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- Past house price appreciation in a given area had a significant positive impact on subsequent loan approval rates in area.

(Dell'Ariccia, Igan, & Laeven 2011, Goetzmann, Peng, & Yen 2012).







Overview 00	U.S. Data 00	Fundamental vs. Bubble ●○○○○	Comments 000	Summary O
Bubbles ve	rsus rational	ly low risk premiu	ms.	
Are the two situa	ations observationa	ally equivalent?		

John Cochrane (2009): "... Crying bubble is empty unless you have an operational procedure for distinguishing them from rationally low risk premiums..."

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$$p_t = d_t + E_t \left(\frac{1}{1+r}\right) p_{t+1}, \quad r = \underbrace{r^{\mathsf{f}} + \mathsf{risk premium}}_{\mathsf{Discount rate} = \mathsf{Expected Return}}$$
$$= d_t + E_t \left[\frac{d_{t+1}}{1+r} + \frac{d_{t+2}}{(1+r)^2} + \frac{d_{t+3}}{(1+r)^3} + \dots, \quad \frac{d_{t+1}}{d_t} = 1 + g + \varepsilon_{t+1}\right]$$
$$\frac{p_t}{d_t} = \frac{1}{r-g}, \quad \text{provided } r > g.$$

A high p-d ratio can be justified by fundamentals if expected return (r) is low because risk premium is low.

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Problem with this story: Survey evidence reveals that expected returns are high when p-d ratios (or price-rent ratios) are high.

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Overview	U.S. D	F	undamental vs. I	Bubble	Comme 000	nts	Summary O

Did housing investors expect low future returns in 2005? Rational model predicts low expected returns at market peaks.





June 13, 2005

June 6, 2005

Overview	U.S. Data	Fundamental vs. Bubble	Comments	Summary
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House price	es and invest	tor expectations in	four U.S. c	ities.

"12-month expectations are fairly well described as attenuated versions of lagged actual 12-month price changes."



Overview	U.S. Data	Fundamental vs. Bubble	Comments	Summary
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Expected I	returns fo	r the U.S. stock ma	arket.	

"Our evidence rules out rational expectations models in which changes in market valuations are driven by the required returns of a representative investor."



Overview	U.S. Data	Fundamental vs. Bubble	Comments	Summary
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A typica Source: IMF	l empirical r WEO 2004, Char	nodel of house pric	C ES. rice Boom"	

Table DO 4. Million Datamatica Harris Data

Are these fundamental explanatory variables?

Explanatory Variables Real house price (growth) Lagged dependent variable Lagged real house price (growth) 0.521 [0.030]' Reversion Lagged housing affordability ratio -0.144 Fundamentals [0.021]' Real disposable income (per capita, growth) 0.530 [0.199] Short-term interest rate (percent) -0.507 [0.039] Real credit (growth) 0.033 [0.039] Lagged real stock price (growth) 0.033 [0.039] Population growth 1.754 [0.623]'	1se wth) 1 0]* 4 1]*	Real house price (growth 0.521 [0.030]* -0.144	xplanatory Variables agged dependent variable Lagged real house price (growth) Reversion Lagged housing affordability ratio
Lagged dependent variable 0.521 Lagged real house price (growth) 0.521 Reversion -0.144 Lagged housing affordability ratio -0.144 Fundamentals [0.021]* Real disposable income [0.139] Short-term interest rate (percent) -0.507 [0.036]* [0.109] Caged real stock price (growth) 0.033 Lagged real stock price (growth) 0.033 Population growth 1.754	1 0]* 4 1]*	0.521 [0.030]* -0.144	agged dependent variable Lagged real house price (growth) Reversion Lagged housing affordability ratio
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Reversion -0.144 Lagged housing affordability ratio -0.144 Fundamentals [0.021]' Real disposable income (per capita, growth) (per capita, growth) [0.119]' Short-term interest rate (percent) -0.507 Real credit (growth) [0.038]' Lagged real stock price (growth) 0.033 Population growth 1.754	4 1]*	-0.144	Reversion Lagged housing affordability ratio
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[0.009] Population growth 1.754 [0.623]*	3	0.033	Lagged real stock price (growth)
Population growth 1.754 [0.623]*	9]*	[0.009]*	
[0.623]	4	1.754	Population growth
	3]*	[0.623]*	
Bank crisis –2.426	6	-2.426	Bank crisis
[0.952]*	2]*	[0.952]*	

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Comment:	bubbles can	be driven purely	by fundament	tals.
References: Froo	t & Obstfeld (AFR	(Econ	omic Journal 2010)	

$$p_t = d_t + E_t \left(\frac{1}{1+r} \right) p_{t+1}, \qquad \frac{d_{t+1}}{d_t} = 1 + g + \varepsilon_{t+1}$$

 $p_t = p_t^{\mathsf{f}} + p_t^{\mathsf{b}}$

$$\left. \begin{array}{l} p_t^{\rm f} = \frac{d_t}{r-g} \\ \\ p_t^{\rm b} = p_{t-1}^{\rm b} \exp\left(\lambda \, \varepsilon_{t+1}\right) \end{array} \right\} \Rightarrow p_t = {\rm f}\left(p_{t-1}\right)$$

$$\lambda \;=\; \pm \sqrt{rac{2\log(1+r)}{Var(arepsilon_{t+1})}}$$

\Rightarrow Authors' methodology may not detect this type of bubble.

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$$R_{t+1} = \frac{p_{t+1} + d_{t+1}}{p_t} = \frac{(p_{t+1}/d_{t+1} + 1)(1 + g_{t+1})}{p_t/d_t}$$

 $\log(p_t/d_t) \simeq \kappa_0 + \kappa_1 \log(p_{t+1}/d_{t+1}) + g_{t+1} - \log(R_{t+1})$

$$\simeq \kappa_{0} + \sum_{j=1}^{\infty} (\kappa_{1})^{j} [g_{t+j} - \log (R_{t+j})]$$

$$Var [\log (p_{t}/d_{t})] = Cov \left[\log (p_{t}/d_{t}), \sum_{j=1}^{\infty} (\kappa_{1})^{j} g_{t+j} \right]$$

$$- Cov \left[\log (p_{t}/d_{t}), \sum_{j=1}^{\infty} (\kappa_{1})^{j} \log (R_{t+j}) \right]$$

 \Rightarrow Price-rent ratio must predict either future rent-growth or future returns. This motivates the form of forecasting regressions.

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Forecasting	U.S. housin	ng returns with the	price-rent	ratio.

$$\operatorname{Return}_{t \to t+j} = \widehat{b}_0 + \widehat{b}_1 \log \left(\frac{\operatorname{Price}_t}{\operatorname{Rent}_t}\right) + u_{t+1}$$

	1960.Q2 to 2012.Q4	2000.Q1 to 2012.Q4
Forecast Horizon	\widehat{b}_1	\widehat{b}_1
;)	-0.079***	-0.099**
J = Z	(0.013)	(0.043)
i 1	-0.180***	-0.258**
J = 4	(0.024)	(0.043)
:_ 0	-0.405***	-0.736***
J = 0	(0.040)	(0.143)

 \Rightarrow Higher price-rent ratio predicts lower realized returns. But survey evidence shows that investors fail to take this relationship into account when forming their expectations.

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Lessons for policy:

• To guard against costly housing bubbles, regulators should enforce prudent mortgage lending standards.

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Lessons for policy:

- To guard against costly housing bubbles, regulators should enforce prudent mortgage lending standards.
- Debt-to-income limits represent a more prudent lending criteria than loan-to-value limits. (Lim et al 2011, Gelain, et al. 2013)