

Diagnostic Expectations and Credit Cycles

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Introduction

- ▶ Renewal of academic interest on the link between credit expansion and subsequent bust
 - ▶ Schularick, Taylor (2012): credit growth and financial crisis
 - ▶ Mian, Sufi, Verner (2015): household debt and low growth
 - ▶ Baron, Xiong (2014): bank credit and crash risk in stocks
 - ▶ Fahlenbrach et al. (2016): loan growth and bank performance
- ▶ Complementary findings for corporate debt:
 - ▶ Greenwood, Hanson (2013): in credit booms, quality of debt issuers falls. Larger high yield share in bond issuance predicts low (negative) excess returns
 - ▶ Gilchrist, Zakrajsek (2012), Krishnamurthy, Muir (2015): credit tightening anticipates the coming recession
 - ▶ Lopez-Salido, Stein, Zakrajsek (LSZ 2015): low spreads today predicts rise in credit spreads and low growth afterwards

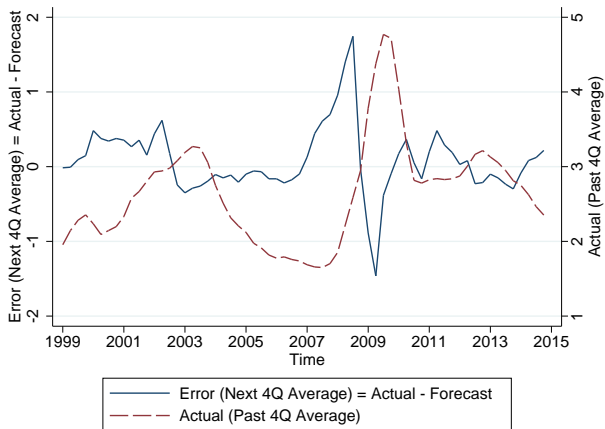
Our Approach

- ▶ Build a behavioral model of credit cycles
- ▶ Micro-founded expectation formation based on Representativeness Heuristic
- ▶ Consistent with available evidence but also with predictable expectations errors
- ▶ Forward looking; immune to the Lucas critique

Related Literature

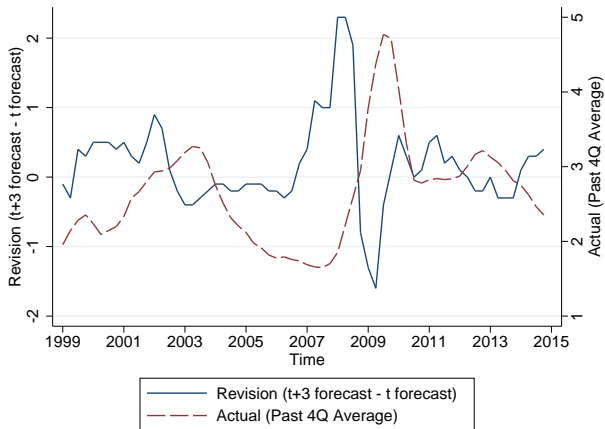
- ▶ Financial frictions
 - ▶ Those models fail to account for predictable returns and errors
 - ▶ Also do not explain where shocks come from
- ▶ Extrapolation
 - ▶ Greenwood and Shleifer 2014, Barberis et al 2015a,b
 - ▶ Our theory micro-founds extrapolation and neglect of risk
- ▶ Limited attention
 - ▶ Sims 2003, Coibion and Gorodnichenko 2012, 2015, Gabaix 2015
 - ▶ These are models of under-reaction, not over-reaction
- ▶ Behavioral models of credit cycles
 - ▶ Gennaioli, Shleifer, and Vishny 2012, 2015, Greenwood, Hanson, and Jin 2016
 - ▶ Our model provides a portable foundation of belief formation

Predictable Expectation Errors of Credit Spreads



Data Source: *Blue Chip Financial Forecasts*

Predictable Reversals in Expectations of Credit Spreads



Data Source: *Blue Chip Financial Forecasts*

This Paper

- ▶ Model of expectation formation based on Gennaioli and Shleifer's (2010) formalization of Kahneman and Tversky's "representativeness" heuristic
- ▶ Inserted into a simple macroeconomic model (no financial frictions), yields many of the previous facts
- ▶ What is representativeness?
 - ▶ How to model it
 - ▶ Implications for Macro-Finance

What is Representativeness?

- ▶ KT (1974): we judge the frequency of an attribute by its similarity to, or representativeness for, the parent population
- ▶ KT (1983): “an attribute is representative of a class if it is very diagnostic; that is, the relative frequency of this attribute is much higher in that class than in a relevant reference class.”
- ▶ KT argue that representativeness lies behind systematic, extensively documented biases in probability judgments:
 - ▶ Base rate neglect, Conjunction Fallacy, Disjunction Fallacy
 - ▶ Example (Linda): an intelligent, single woman in her 30's who was an activist in college is deemed more likely to be a feminist bank teller than a bank teller

How to Model Representativeness?

- ▶ Assess distribution of attribute T in class G

$$h(T = t|G)$$

- ▶ Following KT, define representativeness of $T = t$ for G as:

$$\frac{h(T = t|G)}{h(T = t| - G)}$$

- ▶ Distort $h(T = t|G)$ by inflating the probability of values t that score high, neglect / under-weight values that score low
- ▶ This model yields the KT biases (GS 2010) and accounts for “kernel of truth” in social stereotypes (Bordalo Coffman Gennaioli Shleifer 2015)

Example: Stereotypes

- ▶ Hair color distribution among the Irish

$$h(\text{hair colour}|\text{Irish})$$

- ▶ $T \equiv \{\text{red}, \text{light}, \text{dark}\}$, $G = \text{Irish}$, $-G = \text{World}$

<i>hair colour</i>	<i>red</i>	<i>light</i>	<i>dark</i>
Irish	10%	40%	50%
World	1%	14%	85%

- ▶ The stereotype of Irish overweights red hair:

$$\frac{h(\text{red hair}|\text{Irish})}{h(\text{red hair}|\text{World})} = 10$$

- ▶ Kernel of Truth (Judd and Park 1993, BCGS 2015)
Confirmed in data on political, gender, ethnic groups.

Implications for Macro-Finance?

- ▶ Given data (Irish), inflate prevalence of hair color (red) whose objective probability goes up the most relative to others
- ▶ In a dynamic environment:
 - ▶ given news, agents inflate future states of the world whose *objective* probability goes up the most
 - ▶ the context is lagged information
- ▶ This yields:
 - ▶ extrapolation + neglect of tail risk in a single setup
 - ▶ reversals in the absence of news
 - ▶ excess volatility
 - ▶ immunity to Lucas critique, RE as a special case
 - ▶ no learning, rather beliefs distort true process
 - ▶ model is portable: unify explanation of lab experiments, social stereotypes, macroeconomic predictions

Model Ingredients

- ▶ State of the economy Ω_t at t follows $AR(1)$

$$\omega_t = b \cdot \omega_{t-1} + \epsilon_t$$

- ▶ Diagnostic Expectations about Ω_{t+s}
- ▶ Measure 1 of firms of varying risk (different exposure to Ω_t)
- ▶ Long lived, risk neutral, representative household that supplies capital to these firms (buys risky debt from them)

Diagnostic Expectations

- ▶ After seeing the state ω_t , the agent must represent:

$$h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \omega_t)$$

- ▶ Here $G \equiv \{\Omega_t = \omega_t\}$
- ▶ News assessed relative to $-G$ containing past information
- ▶ Main case: reference is information available at $t - 1$

$$-G \equiv \{\Omega_t = b \cdot \omega_{t-1}\}$$

Then representativeness is:
$$\frac{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \omega_t)}{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = b \cdot \omega_{t-1})}$$

Overweighing

- ▶ We assume the distorted distribution $h_t^\theta(\omega_{t+1})$ to be:

$$h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \omega_t) \cdot \left[\frac{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \omega_t)}{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = b \cdot \omega_{t-1})} \right]^\theta \frac{1}{Z_t}$$

- ▶ $\theta \geq 0$ measures the importance of representativeness
- ▶ Rational expectations: special case for $\theta = 0$ or no news $\epsilon_t = 0$
- ▶ Inflate density of future states that have become more likely
- ▶ Denote Diagnostic Expectations by

$$\mathbb{E}_t^\theta(\omega_{t+1}) = \int_{\mathbb{R}} \omega \cdot h_t^\theta(\omega) d\omega$$

Specifying $-G$

- ▶ Alternative: reference is recent diagnostic expectation

$$\frac{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \omega_t)}{h(\Omega_{t+1} = \omega_{t+1} | \Omega_t = \mathbb{E}_{t-1}^\theta(\omega_t))}$$

- ▶ $-G$ influences reaction to news
 - ▶ different specifications imply different lag structures of expectations
- ▶ We proceed with our main case:

$$-G = \{\Omega_t = b \cdot \omega_{t-1}\}$$

and then consider the other cases

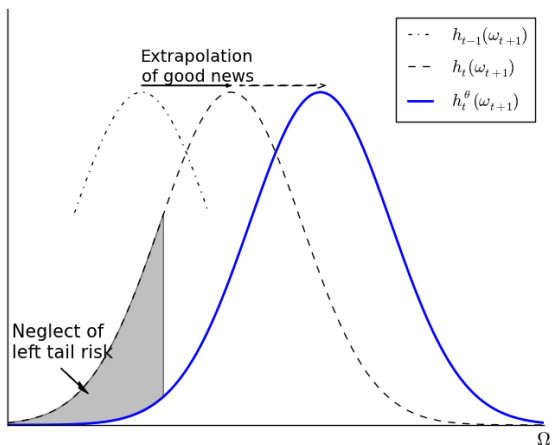
Representation

Proposition 1. When the process for ω_t is AR(1) with normal $(0, \sigma^2)$ shocks, the distribution $h^\theta(\omega_{t+1})$ is also normal, with variance σ^2 and mean:

$$\mathbb{E}_t^\theta(\omega_{t+1}) = \mathbb{E}_t(\omega_{t+1}) + \theta [\mathbb{E}_t(\omega_{t+1}) - \mathbb{E}_{t-1}(\omega_{t+1})]$$

- ▶ \mathbb{E}_t^θ is function of (lagged) rational expectations
- ▶ Kernel of truth: overweight incoming news
- ▶ Context dependence: the path is important

Neglect of Tail Risk (GSV 2012)



Extrapolation

- ▶ Plugging AR(1) in $\mathbb{E}_t^\theta(\omega_{t+1})$ we obtain

$$\mathbb{E}_t^\theta(\omega_{t+1}) - \omega_t = [\mathbb{E}_t(\omega_{t+1}) - \omega_t] + b\theta [\omega_t - \mathbb{E}_{t-1}(\omega_t)]$$

- ▶ Slant toward current objective news $\omega_t - \mathbb{E}_{t-1}(\omega_t)$
- ▶ Neglect of risk and extrapolation follow from the same psychology of context effects

Diagnostic vs. Adaptive Expectations

- ▶ Adaptive Expectations

$$\mathbb{E}_t^a(\omega_{t+1}) = \lambda\omega_t + (1 - \lambda)\mathbb{E}_{t-1}^a(\omega_t), \quad 0 < \lambda < 1$$

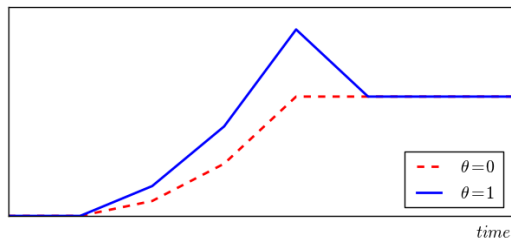
- ▶ Diagnostic Expectations

$$\mathbb{E}_t^\theta(\omega_{t+1}) = b(1 + \theta)\omega_t - b\theta\mathbb{E}_{t-1}(\omega_t)$$

- ▶ Overreaction + reversal (rather than momentum)
- ▶ Forward looking:
 - ▶ Extrapolate only if process is stochastic and persistent, $b > 0$
 - ▶ No mistakes for i.i.d. case, $b = 0$
 - ▶ Immune to Lucas Critique

Sequences of News

- ▶ Random walk, $\omega_t = \omega_{t-1} + \epsilon_t$



- ▶ accelerating good news cause sustained optimism
- ▶ when good news stop, boom is followed by bust

Non-Fundamental Reversals

- ▶ Suppose fundamentals follow a random walk ($b = 1$) . Then:

$$\mathbb{E}_t^\theta(\omega_{t+1}) = \omega_t + \theta (\omega_t - \omega_{t-1})$$

- ▶ As a consequence:

$$\begin{aligned}\mathbb{E}_t[\mathbb{E}_{t+1}^\theta(\omega_{t+2}) - \mathbb{E}_t^\theta(\omega_{t+1})] \\ &= \mathbb{E}_t [(\omega_{t+1} - \omega_t) (1 + \theta) - \theta (\omega_t - \omega_{t-1})] \\ &= -\theta (\omega_t - \omega_{t-1})\end{aligned}$$

- ▶ Excess optimism at t systematically wanes at $t + 1$ even in the absence of news. A boom is followed by a bust.

Model Ingredients

- ▶ State of the economy Ω_t at t follows $AR(1)$

$$\omega_t = b \cdot \omega_{t-1} + \epsilon_t$$

- ▶ Diagnostic Expectations about Ω_{t+s}
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- ▶ Long lived, risk neutral, representative household that supplies capital to these firms (buys risky debt from them)

Firms

- ▶ Each firm is identified by its risk $\rho \in \mathbb{R}$ (which is common knowledge), and produces output:

$$y(k|\omega_t, \rho) = \begin{cases} k^\alpha & \text{if } \omega_t \geq \rho \\ 0 & \text{if } \omega_t < \rho \end{cases}$$

- ▶ At t , firm ρ borrows at the interest rate $r_{t+1}(\rho)$ to install capital $k_{t+1}(\rho)$. Maximize expected profit:

$$\max_{k_{t+1}(\rho)} [k_{t+1}(\rho)^\alpha - k_{t+1}(\rho) \cdot r_{t+1}(\rho)] \cdot \mu_t^\theta(\rho)$$

where “perceived creditworthiness” is:

$$\mu_t^\theta(\rho) = \int_{\rho}^{+\infty} h_t^\theta(\omega) d\omega$$

Households

- ▶ The representative household solves:

$$\max_{D_{s+1}(\rho)} \mathbb{E}_t^\theta \left[\sum_{s=t}^{+\infty} \beta^{s-t} c_s \right]$$

with budget constraint

$$c_s + \int_{\mathbb{R}} D_{s+1}(\rho) f(\rho) d\rho = w + \int_{\mathbb{R}} l(\rho, \omega_s) \cdot [r_s(\rho) D_s(\rho) + \pi_s(\rho)] f(\rho) d\rho$$

- ▶ We assume the endowment is large enough to obtain interior solutions despite risk neutrality, $w \geq (\alpha\beta)^{\frac{1}{1-\alpha}}$

Equilibrium

- ▶ Firms invest until MPK in case of success equals contract interest rate

$$k_{t+1}(\rho) = \left[\frac{\alpha}{r_{t+1}(\rho)} \right]^{\frac{1}{1-\alpha}}$$

- ▶ Households buy debt until expected return equals inverse discount factor

$$r_{t+1}(\rho) \cdot \mu_t^\theta(\rho) = \frac{1}{\beta} \quad \leftrightarrow \quad r_{t+1}(\rho) = \frac{1}{\beta \mu_t^\theta(\rho)}$$

- ▶ Equilibrium spread between risky firm ρ and safe firm $\rho \rightarrow -\infty$

$$S\left(\rho, \mathbb{E}_t^\theta(\omega_{t+1})\right) \equiv r_{t+1}(\rho) - \frac{1}{\beta} = \frac{1}{\beta} \left(\frac{1}{\mu_t^\theta(\rho)} - 1 \right)$$

Equilibrium

- ▶ Debt issuance/installed capital of firm ρ :

$$k_{t+1}(\rho) = \left[\alpha \beta \mu_t^\theta(\rho) \right]^{\frac{1}{1-\alpha}} = \left[\frac{\alpha}{1/\beta + S(\rho, \mathbb{E}_t^\theta(\omega_{t+1}))} \right]^{\frac{1}{1-\alpha}}$$

- ▶ Total debt issued and investment (full depreciation):

$$K_{t+1} = \int_{\mathbb{R}} \left[\alpha \beta \mu_t^\theta(\rho) \right]^{\frac{1}{1-\alpha}} f(\rho) d\rho$$

- ▶ Future output in state ω_{t+1} :

$$Y_{t+1}(\omega_{t+1}) = \int_{-\infty}^{\omega_{t+1}} \left[\alpha \beta \mu_t^\theta(\rho) \right]^{\frac{1}{1-\alpha}} f(\rho) d\rho$$

Spreads and Issuance

- ▶ Define average spread at t (inverse measure of optimism)

$$S_t = \int_{\mathbb{R}} S\left(\rho, \mathbb{E}_t^\theta(\omega_{t+1})\right) f(\rho) d\rho$$

- ▶ **Proposition.** Higher S_t (lower optimism at t) causes:
 - ▶ disproportionate rise in spread of riskier firms: $\frac{\partial^2 S}{\partial S_t \partial \rho} > 0$
 - ▶ disproportionate decline in debt issuance and investment by riskier firms: $\frac{\partial}{\partial S_t} \frac{k_{t+1}(\rho_1)}{k_{t+1}(\rho_2)} < 0$ for $\rho_1 > \rho_2$
 - ▶ holds for $\theta \geq 0$
- ▶ Accordingly, GH (2013) show junk share rises as spreads fall

Dynamics of Credit Spreads

- ▶ Linearise model for $\mathbb{E}_t^\theta(\omega_{t+1})$ near long-term mean $\bar{\omega} = 0$

$$S_t = \sigma_0 - \sigma_1 \mathbb{E}_t^\theta(\omega_{t+1})$$

- ▶ **Proposition.** Average spread S_t follows process:

$$S_t = (1 - b)\sigma_0 + b \cdot S_{t-1} - (1 + \theta)b\sigma_1\epsilon_t + \theta b^2\sigma_1\epsilon_{t-1}$$

- ▶ for $\theta = 0$ (rational expectations), spreads follow AR(1), just like fundamentals
- ▶ for $\theta > 0$, spreads instead follow ARMA(1,1)

Dynamics of Credit Spreads

- ▶ Spreads follow ARMA(1,1):

$$S_t = (1 - b)\sigma_0 + b \cdot S_{t-1} - (1 + \theta)b\sigma_1\epsilon_t + \theta b^2\sigma_1\epsilon_{t-1}$$

- ▶ As S_t is function of expectations at t :
 - ▶ has autoregressive component, $S_t \sim b \cdot S_{t-1}$
 - ▶ but S_{t-1} overreact to news at $t - 1$
 - ▶ $t - 1$ overreaction subsides at t , does not contaminate S_t
 - ▶ add correction term $\theta b^2\sigma_1\epsilon_{t-1}$ (moving average component)

Credit Spreads Forecasts

- ▶ To compute forecasts of spreads, note:

$$\mathbb{E}_t^\theta \left(\mathbb{E}_{t+s}^\theta (\omega_{t+T}) \right) = \mathbb{E}_t^\theta (\omega_{t+T})$$

- ▶ diagnostic expectations satisfy law of iterated expectations
- ▶ Revisions of expectations are unpredictable to investors. Forecasts of credit spreads then follow:

$$\mathbb{E}_t^\theta (S_{t+T}) = \sigma_0 \left(1 - b^T \right) + b^T S_t$$

- ▶ Actual spreads follow ARMA(1,1) but forecasts follow AR(1)
 - ▶ introduces systematic errors, that can account for our motivating evidence

Credit Spreads Forecasts

- ▶ **Proposition.** Conditional on information at t :

- ▶ forecast error at $t + 1$ is predictable:

$$\mathbb{E}_t [S_{t+1} - \mathbb{E}_t^\theta (S_{t+1})] = \theta b^2 \sigma_1 \epsilon_t$$

- ▶ revision of forecasts are predictable:

$$\mathbb{E}_t [\mathbb{E}_{t+s}^\theta (S_{t+T}) - \mathbb{E}_t^\theta (S_{t+T})] = \theta b^{T+1} \sigma_1 \epsilon_t$$

- ▶ Good news predict that S_t and $\mathbb{E}_t^\theta (S_{t+1})$ are too low, and that future spreads are revised upwards
- ▶ Consistent with our evidence:
 - ▶ negative correlation between S_t and error $S_{t+1} - \mathbb{E}_t^\theta (S_{t+1})$
 - ▶ also between S_t and revision $\mathbb{E}_{t+s}^\theta (S_{t+T}) - \mathbb{E}_t^\theta (S_{t+T})$

Predictable Returns and Excess Volatility

- ▶ **Corollary.** Let $S_t^r \equiv S_t|_{\theta=0}$ be rational spread. For $\theta > 0$:
 - ▶ avg returns are predictably low (high) on good (bad) news

$$S_t - S_t^r = -\theta b \sigma_1 \epsilon_t$$

- ▶ spreads exhibit excess volatility

$$\text{Var}_{t-1}[S_t] = (1 + \theta)^2 \text{Var}_{t-1}[S_t^r]$$

- ▶ Consistent with evidence
 - ▶ high junk shares predict low (even negative) returns (GH 2013)
 - ▶ fundamentals account for small share of volatility (Collin-Dufresne et al 2001)

Non-fundamental boom-bust cycles

Corollary. Let $\theta > 0$ and S_{t-1} be low due to good news $\epsilon_{t-1} > 0$. Then, controlling for fundamentals at $t - 1$:

- ▶ spreads predictably rise at t
- ▶ aggregate investment at t , and aggregate production at $t + 1$, predictably drop
- ▶ consistent with Lopez-Salido, Stein, Zakrajsek (2015)

Summary

- ▶ We present a psychologically founded, forward looking model of expectation formation.
 - ▶ in a simple macro model, it reproduces several features of credit cycles.
 - ▶ also reproduces facts about spreads forecasts
- ▶ Features in common with RE model:
 - ▶ spread compression in good times
 - ▶ relatively high issuance by high yield firms
- ▶ Features arising from diagnostic expectations ($\theta > 0$)
 - ▶ extrapolative expectations of fundamentals
 - ▶ excess volatility of credit spreads
 - ▶ predictably low returns in good times
 - ▶ systematic non-fundamental reversals (bad credit shocks)

Future Avenues

- ▶ Expectations
 - ▶ consider a variety of time series and their expectations
 - ▶ understand sources of under- and over-reaction
- ▶ Financial Frictions
 - ▶ Special role of debt / risk misallocation (GSV 2012, 2013)
 - ▶ Asymmetric effect of busts