

Structural GARCH: The Volatility-Leverage Connection

Robert Engle¹ Emil Siriwardane^{1,2}

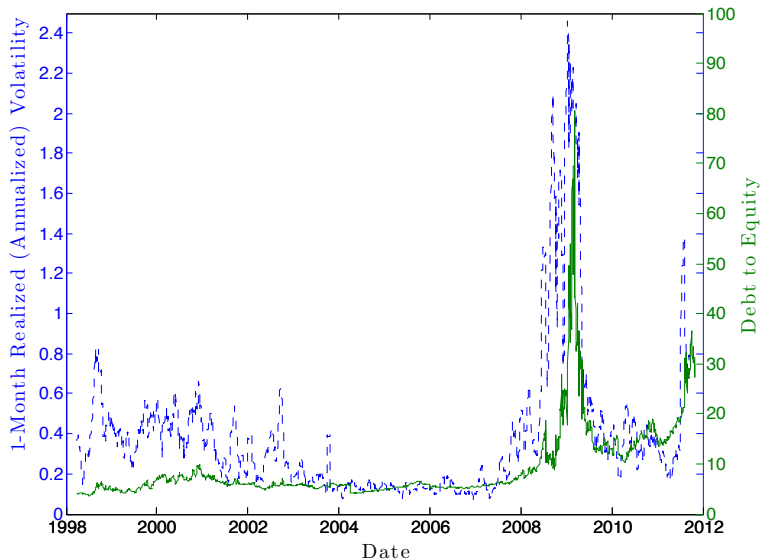
¹NYU Stern School of Business

²U.S. Treasury, Office of Financial Research (OFR)

WFA Annual Meeting: 6/16/2014

Introduction

BAC Leverage and Realized Volatility



Leverage and Equity Volatility

- ▶ Crisis highlighted how leverage and equity volatility are tightly linked
- ▶ “Leverage Effect” has been around - e.g. Black (1976), Christie (1982) - but...
- ▶ A dynamic volatility model that incorporates leverage directly has remained elusive

This Paper

- ▶ GARCH-type model where equity volatility is amplified (non-linearly) by leverage as in structural models of credit
- ▶ **Asset** return series from observed equity series
- ▶ **Assets** have time-varying volatility at high frequencies
- ▶ Statistical test of how leverage affects volatility
- ▶ Two applications:
 1. Systemic Risk: SRISK and Precautionary Capital (today)
 2. Leverage Effect (in the paper)

Theoretical Foundation

Structural Models of Credit

- ▶ Under relatively weak assumptions on the vol process, structural models say $E_t = f(A_t, D_t, \sigma_{A,t}, \tau, r_t)$
 - ▶ A_t = market value of assets
 - ▶ D_t = book value of debt
 - ▶ $\sigma_{A,t}$ = stochastic asset volatility
- ▶ Generic dynamics for assets and asset variance (allow for jumps later):

$$\frac{dA_t}{A_t} = \mu_A(t)dt + \sigma_{A,t}dB_A(t)$$
$$d\sigma_{A,t}^2 = \mu_v(t, \sigma_{A,t})dt + \sigma_v(t, \sigma_{A,t})dB_v(t)$$

- ▶ $B_A(t)$ and $B_v(t)$ potentially correlated

Equity Returns and Equity Volatility

Introducing the Leverage Multiplier

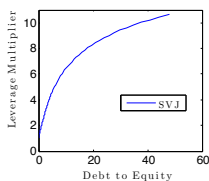
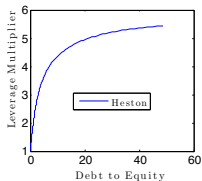
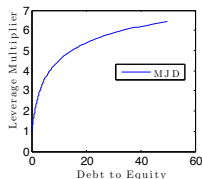
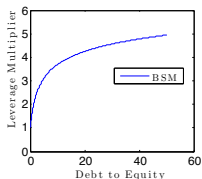
- ▶ Apply Itô Lemma and ignore drift (our model is daily, and daily equity returns ≈ 0):

$$\begin{aligned}\frac{dE_t}{E_t} &= LM_t \sigma_{A,t} dB_A(t) + \frac{v_t \sigma_v(t, \sigma_{A,t})}{E_t 2\sigma_{A,t}} dB_v(t) \\ &\approx LM_t \times \sigma_{A,t} \times dB_A(t) \\ \text{vol}_t \left(\frac{dE_t}{E_t} \right) &\approx LM_t \times \sigma_{A,t}\end{aligned}$$

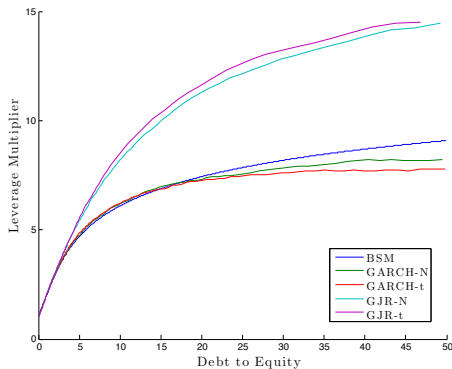
- ▶ $LM_t = LM(E_t/D_t, 1, \sigma_{A,t}, \tau, r_t)$ is the “leverage multiplier”
- ▶ LM_t amplifies asset shocks and volatility
- ▶ Two questions:
 1. How much does the **higher order** term contribute? **Not Much**
 2. What does LM_t look like? **Robust shape across models**

The Leverage Multiplier: Three Basic Properties

Popular Continuous Time Option Pricing Models



Discrete Time: GARCH Option Pricing



1. $LM(0) = 1$. Mechanical, since assets = equity
2. **Monotonically increasing**. More leverage means more risk
3. **Concave**. Reducing leverage more powerful than increasing leverage

Structural GARCH

Our Specification

- ▶ The challenge is choosing the right functional form for LM_t
- ▶ We use simple **transformations** of Black-Scholes-Merton (BSM) functions:

$$LM_t(D_t/E_t, \sigma_{A,t}^f, \tau) = \left[\Delta_t^{BSM} \times g^{BSM} \left(E_t/D_t, 1, \sigma_{A,t}^f, \tau \right) \times \frac{D_t}{E_t} \right]^\phi$$

$g^{BSM}(\cdot)$ is inverse BSM call function. Δ_t^{BSM} is BSM delta

- ▶ $\phi \neq$ specific option pricing model
- ▶ Our parametrization preserves necessary properties of LM , but still allows us to change its scale

The Full Recursive Model

Structural GARCH

$$\begin{aligned}r_{E,t} &= LM_{t-1} \times r_{A,t} \\ &= LM_{t-1} \times \sqrt{h_{A,t}} \times \varepsilon_{A,t}\end{aligned}$$

$$h_{A,t} \sim GJR(\omega, \alpha, \gamma, \beta)$$

$$LM_{t-1} = \left[\Delta_{t-1}^{BSM} \times g^{BSM} \left(E_{t-1}/D_{t-1}, 1, \sigma_{A,t-1}^f, \tau \right) \times \frac{D_{t-1}}{E_{t-1}} \right]^\phi$$

$$\sigma_{A,t-1}^f = \sqrt{\mathbb{E}_{t-1} \left[\sum_{s=t}^{t+\tau} h_{A,s} \right]}$$

So parameter set is $\Theta = (\omega, \alpha, \gamma, \beta, \phi)$

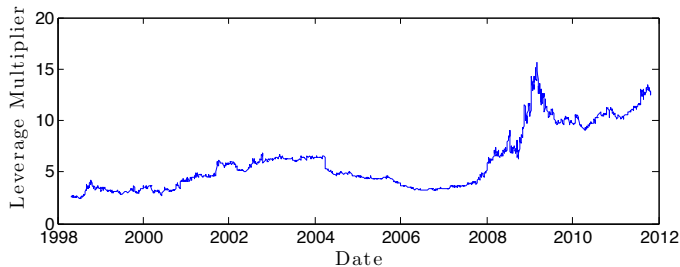
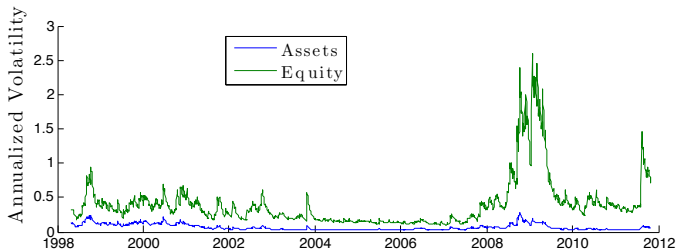
Estimation Results

Estimation Details

- ▶ Estimate for 82 financials via QMLE; iterate over $\tau \in [1, 30]$
- ▶ Equity returns and balance sheet information from Bloomberg
- ▶ D_t is exponentially smoothed book value of debt
 - ▶ smoothing parameter = 0.01, so half-life of weights ≈ 70 days
- ▶ We estimate the model using two approaches for $\sigma_{A,t-1}^f$, then use the highest likelihood:
 1. A dynamic forecast for asset volatility over life of the option
 2. The unconditional volatility of the asset GJR process

Bank of America: Structural GARCH Estimation

$\phi = 1.4$ ($t = 11.4$)



Parameter Values

Cross-Sectional Summary of Estimated Parameters

Parameter	Mean	Mean t-stat	% with $ t > 1.64$
ω	2.7e-06	1.70	47.2
α	0.0458	3.07	86.1
γ	0.0721	2.91	80.6
β	0.9024	80.08	100
ϕ	0.9834	4.00	73.6

- ▶ $(\omega, \alpha, \gamma, \beta)$ are standard GJR parameters - for assets, not equity
- ▶ Average $\tau = 8.34$
- ▶ Leverage matters

Application: SRISK

SRISK

How much would a financial firm need to function normally in another crisis?

- ▶ Acharya et. al (2012) and Brownlees and Engle (2012)
- ▶ Three steps:
 1. GJR-DCC model using firm equity and market index returns
 2. Expected firm equity return if market falls by 40% over 6 months
≡ LRMES
 3. Combine LRMES with book value of debt to determine capital shortfall in a crisis
- ▶ The crisis in this case is a 40% drop in the stock market index over 6 months

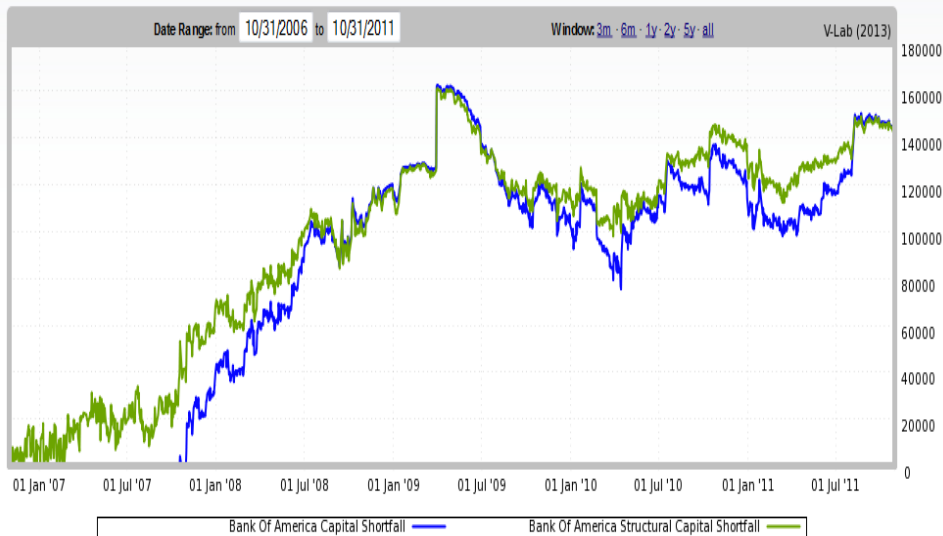
The Role of Leverage?

Thought Experiment with Structural GARCH

- ▶ Firm experiences sequence of negative equity (asset) shocks
- ▶ Level of leverage goes up rapidly
- ▶ Leverage multiplier increases, equity vol amplification higher
- ▶ Painfully obvious in the crisis, so build into SRISK

Bank of America

Capital Shortfall: 2006-2011



Precautionary Capital

Defining Precautionary Capital

e.g. How much additional equity would a bank need, today, to be 90% sure they won't need bailout money in a future crisis?

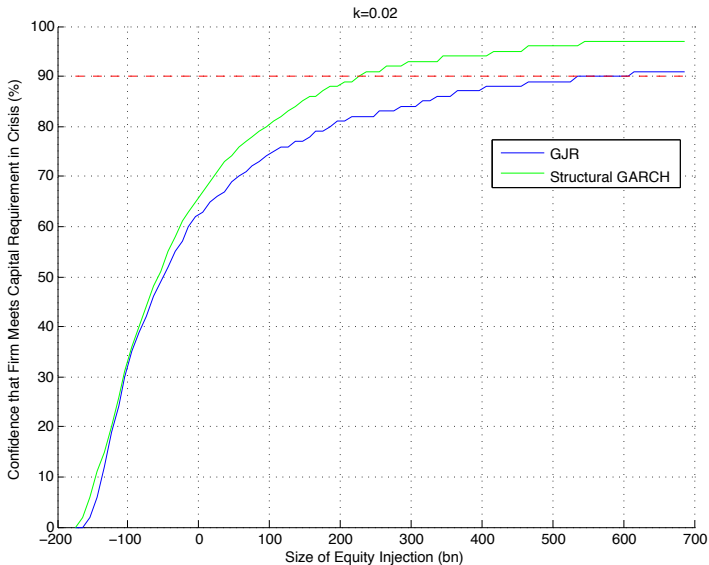
- ▶ SRISK: how much capital would a firm need in a financial crisis to return to a equity/asset ratio of $k\%$?
- ▶ Precautionary Capital: How much capital do we have to add to the firm *today* so that we can have a level of certainty, α , that the firm meets a capital requirement of $k\%$ in a crisis?
- ▶ Uses the quantiles of the future return distribution
- ▶ We set $k = 2\%$ and vary α

Primary Takeaway in a Nutshell

- ▶ Standard volatility models don't have a channel for leverage, so adding equity to the firm today won't reduce future risk
- ▶ Structural GARCH: **reducing leverage today reduces future risk**
 - ▶ The effect is further enhanced by the concavity of the LM
- ▶ Engle and Siriwardane (2014) use this idea to suggest a risk-based total leverage capital requirement

Precautionary Capital: BAC

BAC on 10/1/2008: $E_0 = 173.9$ bn; $D_0 = 1,670.1$ bn



What's Next

Other Applications

- ▶ Endogenous Crisis Probability with Structural GARCH
- ▶ Estimation of Distance to Crisis
- ▶ Endogenous Capital Structure and Leverage Cycles
- ▶ Counter-cyclical Capital Regulation
- ▶ Model of CDS Volatility

Appendix

Ignore Higher Order Terms

$$\frac{dE_t}{E_t} = LM_t \sigma_{A,t} dB_A(t) + \frac{v_t \sigma_v(t, \sigma_{A,t})}{E_t 2\sigma_{A,t}} dB_v(t)$$

How much do the **higher order terms** contribute?

- ▶ Not much. Simple intuition...
- ▶ Volatility mean reversion speed \ll typical debt maturities, so ...
- ▶ Total volatility over option is effectively constant
- ▶ We verify in paper for variety of option pricing models

Dynamic Forecast vs Constant Forecast

- ▶ Estimate two types of models:
 1. Using a dynamic forecast for asset volatility over life of the option
 2. Using unconditional volatility of GJR process
- ▶ Then take the model that delivers the highest likelihood
- ▶ A few outliers where ϕ hits lower bound (exclude from subsequent analysis):
 - ▶ SCHW, JNS, LM, BK, BLK, NTRS, CME, CINF, TMK, UNH