

Aggregate Volatility Risk

Explaining the Small Growth Anomaly and the New Issues Puzzle

Alexander Barinov

Terry College of Business
University of Georgia

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Playing Field

- **Small growth anomaly** - the smallest growth portfolio has the FF alpha of -0.3% per month (Fama and French, 1993)
- **New issues puzzle** - IPOs and SEOs have the FF alpha of -0.4% per month (Loughran and Ritter, 1995)
- **Cumulative issuance puzzle** - long high net issuance, short low net issuance has CAPM alpha of -0.5% per month (Daniel and Titman, 2006)
- Brav, Geczy, and Gompers (2000): Returns to small growth firms and IPOs/SEOs are driven by a common factor

Contribution

- The common explanation to all three anomalies is aggregate volatility risk
- Small growth firms earn low returns, because they hedge against aggregate volatility risk
- Firm-type story: recent issuers and routine heavy issuers seem to underperform, because they are small growth
- The liquidity explanation of the new issues puzzle (Eckbo and Norli, 2005) in fact picks up aggregate volatility risk

Aggregate Volatility Risk

- Volatility increase means worse future investment opportunities (Campbell, 1993)
- Volatility increase means the need to increase precautionary savings (Chen, 2002)
- Firms with most positive return sensitivity to aggregate volatility changes have lower expected returns (Ang et al, 2006)

Model Setup - Barinov (2007)

- A firm consists of assets in place, B , and growth options (call option on S)

$$dB_t = (r + \pi_B)B_t dt + \sigma_B B_t dW_B$$

$$dS_t = (r + \pi_S)S_t dt + \sigma_S S_t dW_S + \sigma_I S_t dW_I$$

- Growth options volatility consists of the systematic part and the idiosyncratic part
- dW_S and dW_B can be correlated

Main Mechanism: Cross-Section

$$\beta_P = E(P, S) \cdot \beta_S, \quad \frac{\partial E(P, S)}{\partial \sigma_I} < 0$$

- As idiosyncratic volatility goes up
 - The beta of the asset behind the growth option stays constant
 - The growth option elasticity wrt the underlying asset value declines
- Therefore, the growth options beta declines in idiosyncratic volatility

Main Mechanism: Time-Series

- Both IVol and aggregate volatility are high in recessions
- All else constant, higher IVol has two effects, both stronger for high volatility growth firms
 - Risk exposure of growth options decreases
 - Value of growth options increases
- Therefore, high volatility firms are hedges against aggregate volatility risk
- The same is true about small growth firms and new issues

Data 1: Aggregate Volatility

- Aggregate volatility is measured by VIX index (old definition) from CBOE
- VIX index is defined as the implied volatility of S&P100 one-month near-the-money options
- Return sensitivity to VIX changes is from daily regressions of stock excess return on the market excess return and the VIX change, run each firm-month

Data 2: BVIX Factor

- VIX sensitivity portfolios use previous month sensitivity and are held for one month
- BVIX factor is the value-weighted return differential between the most negative and the most positive VIX sensitivity quintiles
- Sample: February 1986 - December 2006 (BVIX availability)

Data 3: Other

- Returns, listing, market value from CRSP
- New issues, their dates, after-issue book and market values from SDC
- IPO/SEO portfolios are held for 3 years starting 1 month after the issue
- Factors, 25 size-B/M portfolios from Kenneth French's website
- Sample: January 1986 to December 2006

Table 1: Is BVIX Priced?

- BVIX earns a bit less than 1% per month of abnormal return
- BVIX betas are significant for 25 Size-B/M portfolios, 25 IVol-M/B portfolios, 48 industry portfolios
- BVIX factor significantly improves the GRS statistic for the alphas
- CAPM+BVIX performs better than FF for 25 IVol-M/B and 48 industry portfolios

Table 1A&1B: Three Puzzles and Business Cycle

- I regress the smallest growth portfolios on MKT and business cycle variables
- In Table 1B I subtract return to the largest value portfolio from LHS
- TARCH is forecast of MKT volatility from TARCH model, TARCH and VIX are logs
- BC is 1 if NBER says recession, HVIX is 1 if VIX in top 25%, HTARCH is 1 if TARCH in top 25%

Table 1A: Three Puzzles and Business Cycle

	BC	VIX	HVIX	TARCH	HTARCH
S1G1	0.735	0.615	0.588	1.285	1.293
t-stat	<i>0.52</i>	<i>0.56</i>	<i>0.41</i>	<i>1.90</i>	<i>1.22</i>
S2G1	1.454	0.695	0.604	1.048	1.399
t-stat	<i>1.59</i>	<i>1.01</i>	<i>0.78</i>	<i>2.29</i>	<i>2.29</i>
IPO	1.264	0.291	0.001	0.957	1.447
t-stat	<i>0.90</i>	<i>0.34</i>	<i>0.00</i>	<i>1.65</i>	<i>1.86</i>
SEO	0.716	0.224	0.456	0.774	1.017
t-stat	<i>0.87</i>	<i>0.38</i>	<i>0.78</i>	<i>1.94</i>	<i>1.93</i>
Cumlss	0.287	0.423	0.751	0.431	0.198
t-stat	<i>0.38</i>	<i>0.65</i>	<i>0.87</i>	<i>1.14</i>	<i>0.35</i>
MKT	-0.449	-3.539	-2.537	0.114	-0.074
t-stat	<i>-0.27</i>	<i>-3.37</i>	<i>-3.08</i>	<i>0.22</i>	<i>-0.10</i>

Table 1B: Three Puzzles vs. Large Value across Business Cycle

	BC	VIX	HVIX	TARCH	HTARCH
S1G1	1.512	1.620	0.970	1.662	2.069
t-stat	<i>1.03</i>	<i>1.13</i>	<i>0.49</i>	<i>2.26</i>	<i>1.95</i>
S2G1	2.231	1.700	0.987	1.425	2.175
t-stat	<i>2.33</i>	<i>1.65</i>	<i>0.75</i>	<i>2.50</i>	<i>3.18</i>
IPO	2.042	1.296	0.383	1.334	2.223
t-stat	<i>1.39</i>	<i>1.08</i>	<i>0.25</i>	<i>1.96</i>	<i>2.65</i>
SEO	1.493	1.229	0.838	1.151	1.792
t-stat	<i>1.46</i>	<i>1.43</i>	<i>0.80</i>	<i>2.51</i>	<i>3.23</i>
Cumlss	1.064	1.428	1.134	0.808	0.974
t-stat	<i>1.02</i>	<i>1.33</i>	<i>0.78</i>	<i>1.49</i>	<i>1.37</i>

Table 2A: Aggregate Volatility Risk and the Small Growth Anomaly

VW Returns, January 1986 - December 2006

	Small	Size2	Size3	Size4	Big	S-B
α_{CAPM}	-0.926	-0.525	-0.369	-0.058	0.004	-0.930
t-stat	-2.74	-2.35	-1.89	-0.32	0.03	-2.42
α_{FF}	-0.645	-0.280	-0.025	0.225	0.233	-0.879
t-stat	-3.55	-2.57	-0.28	1.77	2.96	-4.37
α_{ICAPM}	-0.440	-0.158	-0.046	0.207	-0.066	-0.374
t-stat	-0.95	-0.53	-0.18	0.86	-0.51	-0.74
β_{BVIX}	-0.495	-0.374	-0.330	-0.270	0.072	-0.567
t-stat	-1.96	-2.22	-2.40	-2.58	2.54	-2.11

Table 2A: Aggregate Volatility Risk and the Negative Size Effect

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BVIX Factor and Small Growth

- The CAPM and the Fama-French model produce large and significant alphas in the two smallest growth quintiles and negative size effect for growth firms
- The ICAPM with BVIX halves the alphas and makes them insignificant
- The results are even stronger if I drop the January 2001 outlier, when the smallest growth firms make 55%

Table 3: Aggregate Volatility Risk and the New Issues Puzzle

	Panel A. IPOs			Panel B. SEOs		
	CAPM	ICAPM	FF	CAPM	ICAPM	FF
α	-0.578	-0.326	-0.406	-0.436	-0.245	-0.415
t-stat	<i>-2.01</i>	<i>-1.08</i>	<i>-2.11</i>	<i>-2.25</i>	<i>-1.22</i>	<i>-3.16</i>
β_{MKT}	1.466	1.423	1.228	1.318	1.286	1.203
t-stat	<i>16.4</i>	<i>15.7</i>	<i>16.7</i>	<i>23.2</i>	<i>23.2</i>	<i>21.6</i>
β_{SMB}			1.048			0.775
t-stat			7.46			7.54
β_{HML}			-0.211			0.019
t-stat			<i>-1.30</i>			<i>0.20</i>
β_{BVIX}		-0.281			-0.203	
t-stat		<i>-1.99</i>			<i>-2.65</i>	
$\Delta\alpha/\alpha$	44%		20%	44%		41%

Table 4: New Issues Puzzle in Cross-Section

IPO	MB1	MB2	MB3	3-1	Size1	Size2	Size3	1-3
α_{CAPM}	0.315	-0.463	-0.835	1.150	-0.639	-0.505	0.270	0.909
t-stat	1.14	-1.60	-2.30	3.70	-2.02	-1.69	0.95	2.37
α_{ICAPM}	0.364	-0.338	-0.483	0.846	-0.372	-0.257	0.341	0.713
t-stat	1.41	-1.13	-1.27	2.60	-1.12	-0.84	1.16	1.68
β_{BVIX}	-0.094	-0.135	-0.385	0.291	-0.291	-0.299	-0.100	0.191
t-stat	-1.04	-1.13	-2.11	2.53	-1.86	-2.29	-1.90	1.17
SEO	MB1	MB2	MB3	3-1	Size1	Size2	Size3	1-3
α_{CAPM}	0.022	-0.262	-0.665	0.686	-0.495	-0.376	-0.215	0.280
t-stat	0.10	-1.17	-2.80	2.91	-2.04	-1.94	-1.52	1.14
α_{ICAPM}	0.022	-0.177	-0.340	0.362	-0.269	-0.171	-0.234	0.036
t-stat	0.11	-0.83	-1.25	1.40	-1.01	-0.90	-1.61	0.12
β_{BVIX}	-0.010	-0.092	-0.339	0.329	-0.232	-0.217	0.008	0.240
t-stat	-0.15	-1.61	-3.06	4.04	-2.07	-3.44	0.12	1.55

New Issues Puzzle Explained!

- BVIX explains about 40% of the new issues puzzle and leaves the rest insignificant
- New issues underperformance is driven primarily by small and growth new issues
- BVIX is successful to explain their abysmal returns and why they are different from large and value new issues
- The results are even stronger if I drop January 2001, when IPOs make 39%, and SEOs make 24%

Cumulative Issuance Puzzle

- Cumulative issuance - log growth in market value minus log cumulative returns in the past five years
- It shows the net effect of all equity issuing and retiring activity (but no IPOs!)
- Daniel and Titman (2006) show that firms with high cumulative issuance earn abnormally low future returns
- They say it is managers taking advantage of intangible information mispricing
- I need to show high cumulative issuance means small growth

Table 5B: Size, Market-to-Book, and Cumulative Issuance

	Small	Size2	Size3	Size4	Big
Low	0.104	0.079	0.112	0.069	0.005
t-stat	<i>5.4</i>	<i>3.8</i>	<i>5.3</i>	<i>3.8</i>	<i>0.1</i>
MB2	0.194	0.162	0.122	0.084	0.043
t-stat	<i>7.2</i>	<i>5.9</i>	<i>4.3</i>	<i>3.5</i>	<i>1.0</i>
MB3	0.309	0.262	0.170	0.124	0.031
t-stat	<i>8.0</i>	<i>9.4</i>	<i>9.0</i>	<i>5.8</i>	<i>1.1</i>
MB4	0.450	0.445	0.316	0.210	0.049
t-stat	<i>9.8</i>	<i>9.5</i>	<i>8.3</i>	<i>8.0</i>	<i>1.9</i>
High	0.662	0.721	0.563	0.354	0.074
t-stat	<i>16.8</i>	<i>18.8</i>	<i>11.8</i>	<i>14.7</i>	<i>2.9</i>

Table 6A: Aggregate Volatility Risk and the Cumulative Issuance Puzzle

Panel A. EW Returns

	Lowlss	Medlss	Highlss	H-L
α_{CAPM}	0.574	0.513	-0.065	-0.639
t-stat	3.00	2.37	-0.23	-2.66
α_{ICAPM}	0.503	0.586	0.126	-0.378
t-stat	2.70	2.61	0.41	-1.31
β_{BVIX}	0.070	-0.081	-0.203	-0.273
t-stat	1.41	-1.19	-1.65	-2.09

Table 7: Cumulative Issuance Puzzle in Cross-Section

	MB1	MB2	MB3	1-3
α_{CAPM}	0.124	-0.319	-1.063	1.187
t-stat	<i>0.55</i>	<i>-1.37</i>	<i>-2.67</i>	<i>4.30</i>
α_{ICAPM}	0.232	-0.118	-0.456	0.688
t-stat	<i>0.97</i>	<i>-0.43</i>	<i>-0.85</i>	<i>1.77</i>
β_{BVIX}	-0.117	-0.211	-0.624	0.507
t-stat	<i>-1.34</i>	<i>-1.84</i>	<i>-2.44</i>	<i>2.58</i>

Explaining the Cumulative Issuance Puzzle

- High cumulative issuance firms are usually small growth
- BVIX factor explains about 45% of the cumulative issuance puzzle and makes the alphas insignificant
- Cumulative issuance puzzle is stronger for growth firms
- BVIX can explain the cross-section of the cumulative issuance puzzle

Liquidity Factor vs. BVIX Factor

- Eckbo and Norli (2005) show that a turnover-based liquidity factor explains IPO underperformance
- Liquidity factor and BVIX have large positive correlation of 0.45
 - Strange, because small firms load negatively on BVIX and should load positively on liquidity risk
 - If turnover picks up uncertainty, the liquidity factor can be a proxy for BVIX
- In two-factor models, BVIX explains returns to the liquidity factor, but not vice versa

Table 9: Horse Race

	S1G1	S2G1	IPO	SEO	Cumlss
α	-0.088	0.090	0.043	-0.019	-0.088
t-stat	<i>-0.23</i>	<i>0.39</i>	<i>0.16</i>	<i>-0.10</i>	<i>-0.46</i>
β_{BVIX}	-0.064	-0.070	0.171	0.073	0.082
t-stat	<i>-0.51</i>	<i>-0.71</i>	<i>1.79</i>	<i>1.07</i>	<i>1.56</i>
β_{LMH}	-1.049	-0.738	-1.099	-0.670	-0.861
t-stat	<i>-7.50</i>	<i>-7.57</i>	<i>-8.94</i>	<i>-9.77</i>	<i>-10.77</i>

Table 9: Horse Race

	S1G1	S2G1	IPO	SEO	Cumlss
α	-0.088	0.090	0.043	-0.019	-0.088
t-stat	<i>-0.23</i>	<i>0.39</i>	<i>0.16</i>	<i>-0.10</i>	<i>-0.46</i>
β_{BVIX}	-0.064	-0.070	0.171	0.073	0.082
t-stat	<i>-0.51</i>	<i>-0.71</i>	<i>1.79</i>	<i>1.07</i>	<i>1.56</i>
β_{LMH}	-1.049	-0.738	-1.099	-0.670	-0.861
t-stat	<i>-7.50</i>	<i>-7.57</i>	<i>-8.94</i>	<i>-9.77</i>	<i>-10.77</i>

Table 10A: Liquidity Factor and Cross-Section of New Issues Puzzle

	Size1	Size2	Size3	3-1
α_{IPO}	0.150	0.199	0.379	0.229
t-stat	0.46	0.75	1.35	0.64
β_{LMH}	-1.069	-0.953	-0.148	0.922
t-stat	-8.64	-6.37	-1.72	5.73
α_{SEO}	0.037	0.079	-0.121	-0.158
t-stat	0.14	0.39	-0.76	-0.60
β_{LMH}	-0.720	-0.616	-0.126	0.594
t-stat	-12.30	-6.19	-1.27	6.05

Liquidity Factor vs. BVIX: Conclusion

- Eckbo and Norli's liquidity factor picks up aggregate volatility risk, not liquidity risk
 - Liquidity factor and BVIX factor are strongly and counterintuitively positively correlated
 - BVIX factor explains returns to the liquidity factor, but not vice versa
 - Smallest growth firms seem to be extraordinary hedges against "liquidity risk"
 - "Liquidity risk" is much lower for the smallest new issues than for the largest ones

Summary

- Aggregate volatility risk measured by the BVIX factor is the common explanation of
 - Small growth anomaly
 - New issues puzzle
 - Cumulative issuance puzzle
- BVIX factor explains the cross-section of the new issues puzzle and cumulative issuance puzzle
- Liquidity factor of Eckbo and Norli (2005) captures aggregate volatility risk, not liquidity risk
- January 2001 is a powerful outlier for small growth