

16 November 2009

Capital Allocation

CAS Annual Meeting

GARY G. VENTER
Senior Advisor to Oliver Wyman
gary.venter@gmail.com

Document number

Some Basics of Capital Allocation

- How to depends on why
- Will assume purpose is to measure risk-adjusted return
 - Purpose of that is:
 - Decide on which business units to grow or shrink
 - Set target profit levels – i.e., pricing risk
- Capital will be allocated in proportion to a risk measure
 - Starts with $\text{Capital} = K * \text{Risk Measure}$
 - E.g., $\text{Capital} = 3 * \text{TVaR}_{98}$ or $\text{Capital} = 7 * \text{TVaR}_{60}$ or $\text{Capital} = 6 * \text{std}$
 - Used to be $K = 1$, but not necessary

Proposed Answer

- Use distortion measures
 - Basically risk measure is mean under transformed probabilities
- Capital is multiple of risk measure
- Allocation by such a risk measure and setting target returns equal across business units ends up pricing by mean of transformed probabilities
- That has some good properties as a pricing rule
- Pricing by other allocations is a back-door into risk pricing that often ignores basic pricing principles
 - Like pricing for all the risk, not just the tail

Pricing Theory 1 – CAPM

- Price for risk is proportional to covariance of risk with market
- Many problems with that including:
 - If risk is not normally distributed, basic investor utility calls for using higher moments as well
 - Empirical studies (Fama-French especially) find that other factors influence prices
 - FF factors may be proxies for risk measures
 - Two higher co-moments do as well as FF
 - Using even more co-moments eliminates FF factor effect
 - Jump risk probably influences prices in addition to moments
 - Jumps make market incomplete and are systematic and not hedgeable
- But even super-CAPM taking all that into account would be wrong for insurance because ...

In Insurance, Specific Risk Matters

- Opacity of accounting makes security analysts and investors sensitive to earnings fluctuations
 - Impact on market price can be much greater than capital loss
- Cost of raising new capital can be high especially when it is needed – distressed firms – so losses from specific risk can be more expensive than market capital costs
- Policyholders tend to be not diversified in their insurance purchases so are adverse to all risk of the insurer
 - Recent market turmoil illustrates that weakening financial position of insurer will force it to lower prices
- Empirical studies support these ideas

Pricing Theory 2 – Arbitrage-Free Pricing

- Arbitrage = positive expected return with no risk of loss
- May exist for a short time but is quickly competed away
- Some so-called arbitrage actually has hidden risks
- In incomplete market same principle applies
 - In complete market no-arbitrage uniquely determines prices
 - In incomplete market no-arbitrage restricts but does not determine prices
- Either way, basic rule to be arbitrage-free is prices have to be means under equivalent transformed probabilities
- Equivalent = no positive probabilities transformed to zero or vice-versa
- Also transforms have to be on event probabilities, not on outcomes of deals – all deal prices have to use same probabilities for the underlying events

Arbitrage-Free Pricing in Insurance

- Main implication is that prices have to be additive
- In effect, policyholders get the benefit of diversification
 - Insurers still get fair price for risk they take on
 - If pricing does not give diversification benefit to policyholders, insurer will eventually lose that business to competitors
 - Advantage of diversification is to improve competitive position
- Suppose not, e.g., suppose insurer can charge pre-diversification prices, pool risk, and cede 100% to a reinsurer at a lower price
 - That would be an arbitrage profit
 - Not likely in competitive insurance market
- Pricing then through equivalent probability transforms
 - Transform probabilities of firm simulations and lines inherit
 - Transform does not change losses, just probabilities of losses
 - Unfortunately still have to decide among transforms
 - Gives target price based on firm risk, to judge market prices

Some Transforms That Have Worked in Market

- Esscher transform:
 - Parameter w , with $c = G^{-1}(1 - 1/w)$
 - $g^*(y) = Kg(y)\exp(y/c)$, $K = 1/E[\exp(Y/c)]$
 - Doesn't always exist but will if there is a maximum loss
- Wang transform, original normal
 - $G^*(y) = \Phi[\Phi^{-1}(G(y)) + \lambda]$
 - Probability transform from shifting normal percentile by λ
 - λ may be positive or negative depending on losses vs. profit, etc.
 - In simulation get g^* by differencing G^*
- Wang transform with T distribution q dof
 - $G^*(y) = T_q[\Phi^{-1}(G(y)) + \lambda]$
 - Puts more weight in tail
 - Unfortunately in both tails
- Empirical studies on bonds, cat bonds, cat reinsurance give support to Esscher and Wang T_5

Example Company

Simulated Company – 1000 Simulations

	LOB 1	LOB 2	LOB 3	TOTAL
<i>Log N Mu</i>	13.796	13.691	13.316	
<i>Log N Sigma (~CV)</i>	20.0%	50.0%	100.0%	
<i>Actual CV</i>	20.2%	53.3%	131.1%	
<i>Skew</i>	61.4%	175.0%	618.5%	
<i>Expected Loss</i>	1,000,000	1,000,000	1,000,000	3,000,000
<i>Simulated</i>	1,009,960	992,037	988,584	2,990,581
<i>Correlation with Total</i>	17%	34%	91%	100%

Capital = 18,000,000 \approx 2.5 TVaR₉₅ \approx 3 TVaR₉₀

Target underwriting profit = 10% of capital = 1,800,000

After tax this is maybe 6.5% and investment income taxed down to be in insurance company, so maybe reasonable

Agg mean 2,990,581

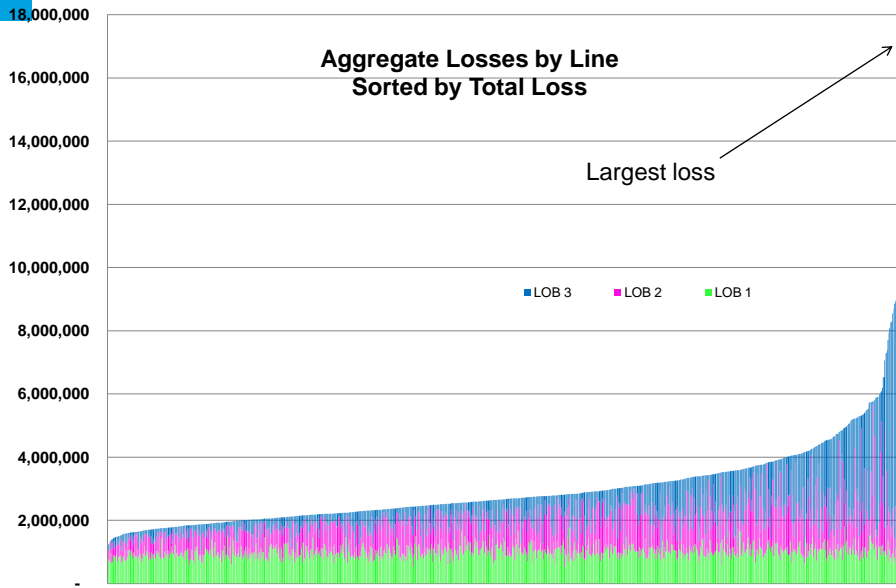
Stdev 1,358,875

CV 45.4%

Skew 318.9%

Skew if lognormal 145.7%

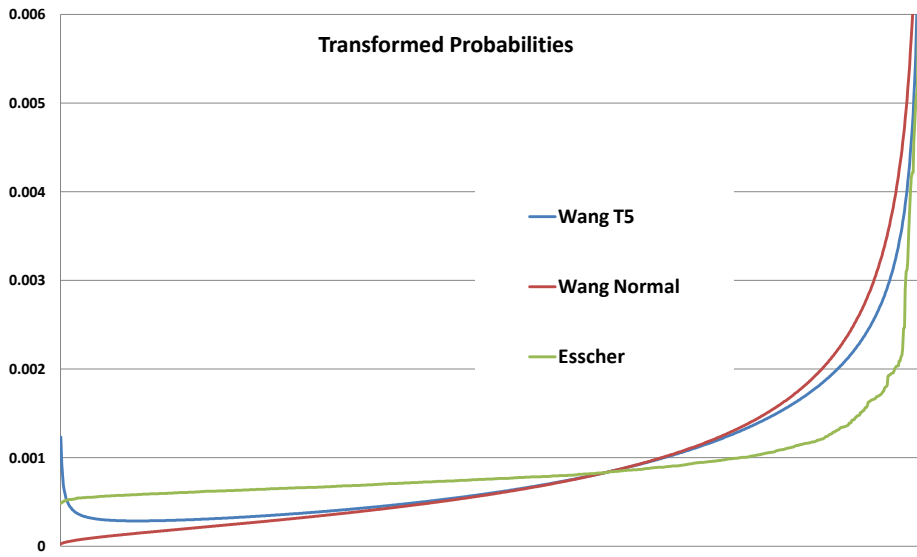
Line Contributions to Total Distribution



Applying Transforms

- Each has a free parameter
- Find parameter value that makes transformed mean = mean + target profit
- Original probability of each scenario = $0.001 = 1/1000$
- Esscher probability:
 - $w = 3.57$, $c = 3,281,072$, $K = 1/E[\exp(Y/c)] = 1/2.968$
 - $g^*(y) = K \exp(y/c) / 1000$
- Normal Wang, set scenario probability = $1/1001$ so cumulative probability < 1
 - $\lambda = -1.03005$, $G^*(y) = \Phi[\Phi^{-1}(G(y)) + \lambda]$
 - $G^*(1000/1001) = 0.9803$, $g^*(\text{largest scenario}) = 1 - 0.9803 = 0.0197$
 - $G^*(999/1001) = 0.9677$, $g^*(2^{\text{nd}} \text{ largest}) = 0.9803 - 0.9677 = 0.0126$

Resulting Probabilities



Allocation and Load

	<u>Total</u>	<u>LOB 1</u>	<u>LOB 2</u>	<u>LOB 3</u>	
Capital	18,000,000	278,954	307,078	17,413,968	Esscher
	18,000,000	304,445	1,299,012	16,396,543	Wang T5
	18,000,000	390,663	1,892,154	15,717,183	Wang Normal
Expected profit	1,800,000	27,895	30,708	1,741,397	Esscher
	1,800,000	30,444	129,901	1,639,654	Wang T5
	1,800,000	39,066	189,215	1,571,718	Wang Normal
Price	4,790,581	1,037,856	1,022,745	2,729,980	Esscher
	4,790,581	1,040,405	1,121,939	2,628,238	Wang T5
	4,790,581	1,049,026	1,181,253	2,560,302	Wang Normal
Load	60.2%	2.8%	3.1%	176.2%	Esscher
	60.2%	3.0%	13.1%	165.9%	Wang T5
	60.2%	3.9%	19.1%	159.0%	Wang Normal