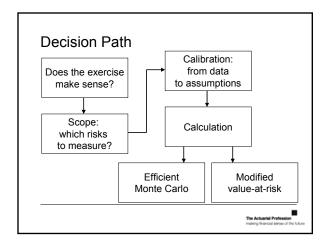


# This is the framework we're discussing

### Assessing Capital based on:

- ■Projected Assets > Liabilities
- ■In one year
- ■With at least 99.5% probability
- ■Applies to life and non-life
- ■Because they're all banks really ●

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### Value at Risk – Market Level Assumptions

	standard deviation	correlati	ons										
Driver													
Equity	20%	100%											
Property	15%	50%	100%										
Yield Curve	0.80%	0%	-30%	100%									
Credit Spread	0.30%	-50%	-40%	10%	100%								
Property loss ratio	10.00%	0%	30%	-10%	25%	100%							
Liability loss ratio	15.00%	20%	10%	-15%	25%	50%	100%						
Inflation	1.00%	-50%	-40%	40%	20%	0%	0%	100%					
Mortality	35%	0%	0%	0%	0%	0%	0%	0%	100%				
Lapses	35%	-30%	-30%	30%	0%	-30%	-30%	20%	0%	100%			
Operational	100%	30%	30%	-30%	-20%	-30%	-30%	20%	0%	40%	100%		
Liquidity	100%	25%	25%	-25%	-50%	-20%	-20%	10%	10%	25%	10%	100%	
Group	100%	20%	20%	-20%	-20%	-20%	-20%	20%	20%	20%	20%	20%	100%

■ Bank VaR typically 200 × 200 correlation matrix

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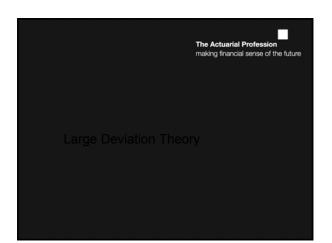
### Fixing the Correlation Matrix

100%								No	ot suf	ficier	nt to have correlations
50%	100%							ho	twoo	n ± 1	100%.
0%	-30%	100%									
-50%	-40% 30%	-10%	100%					Or	nlv po	sitiv	e definite matrices can be
0% 20%	10%	-10%	25% 25%	100%	100%						ation matrices
-50%	-40%	40%	20%	0%	0%	100%		va	lia cc	rreia	ition matrices
-00%	-4U% 0%	40% 0%	20% 0%	0%	0%	0%	100%				
-30%	-30%	30%	0%	-30%	-30%	20%	0%	100%			
30%	30%	-30%	-20%	-30%	-30%	20%	0%	40%	100%		
25%	25%	-25%	-50%	-20%	-20%	10%	10%	25%	10%	100%	
20%	20%	-20%	-20%	-20%	-20%	20%	20%	20%	20%	20%	100%
	ositive d	lefinite co	orrelation	ns							
100%		lefinite co	orrelation	ns				The	larne	r tha	matrix the more likely
	100%	lefinite co	orrelatio	ns							matrix, the more likely
100% 51%	100%		orrelation	ns							matrix, the more likely ve definiteness is a proble
100% 51% -4%	100%	100%		ns 100%							
100% 51% -4% -49%	100% -31% -40%	100%	100%		100%						
100% 51% -4% -49% -1%	100% -31% -40% 29%	100% 9% -9%	100% 25%	100%	100%	100%					
100% 51% -4% -49% -1% 18%	100% -31% -40% 29% 9%	100% 9% -9% -13%	100% 25% 25%	100%		100%					
100% 51% -4% -49% -1% 18% -44%	100% -31% -40% 29% 9% -37%	100% 9% -9% -13% 35%	100% 25% 25% 20%	100% 50% -1%	-2%						
100% 51% -4% -49% -1% 18% -44% 1% -26% 25%	100% -31% -40% 29% 9% -37% 0% -28% 27%	100% 9% -9% -13% 35% -1% 26% -25%	100% 25% 25% 20% 0% 0% -20%	100% 50% -1% 0% -30% -28%	-2% 0% -31% -27%	1% 23% 15%	100% 0% -1%	100% 35%	hat p	ositi	
100% 51% -4% -49% -1% 18% -44% 1% -26%	100% -31% -40% 29% 9% -37% 0% -28%	100% 9% -9% -13% 35% -1% 26%	100% 25% 25% 20% 0% 0%	100% 50% -1% 0% -30%	-2% 0% -31%	1% 23%	100%	it is t	hat p		

Test	Stress Fre	e Assets	Beta Ca	oital required
Base Case		(200)		
Equity	-40%	170	75	39
Property	-25%	183	68	26
Yield Curve	1%	185	-1500	31
Credit Spread	1%	185	-1500	12
Property loss ratio	20%	175	-125	32
Liability loss ratio	20%	170	-150	58
Inflation	1%	180	-2000	52
Mortality	40%	190	-25	23
Lapses	40%	195	-12.5	11
Operational	-1	190	10	26
Liquidity	-1	195	5	13
Group	-1	195	5	13
Total				(334)
Diversification credit				184
Net required				(150)

### Room for Improvement?

- VaR runs instantly and parameters / assumptions are transparent
- Non-zero mean
  - easy to fix
  - take credit for one year's equity risk premium or one year's profit margin in premiums
- Path dependency, overlapping cohorts
  - Add more variables, which can result in huge matrices to estimate
- Company depends linearly on drivers
- Company depends linearly on drivers
   mitigate by careful choice of stress tests
   worst for GI because of reinsurance
   may need mini DFA model to calibrate a VaR model
   Multivariate Normality
   Strong assumption was often supposed lethal
   Before we understood large deviation theory



# Large Deviation Expansions

- In many important examples, we can estimate the moment generating function of net assets
- Large deviation expansions are an efficient way to generate approximate percentiles given moment generating functions
- Exact formulas do exist but they involve numerical integration of complex numbers

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### LD Expansion: The Formula

To estimate  $\operatorname{Prob}\{X \leq c\}$ Where  $\operatorname{Eexp}(pX) = \exp[\kappa(p)]$ Find p where  $\kappa'(p) = c$ 

$$\eta_0 = p \times \sqrt{\frac{2\kappa'(p)}{p} - \frac{2\kappa(p)}{p^2}}$$

$$\eta_1 = \frac{1}{\eta_0} \ln \left[ \frac{p\sqrt{\kappa''(p)}}{\eta_0} \right]$$

$$\eta_2 = \dots$$

**Prob** ~  $\Phi(\eta_0 + \eta_1 + \eta_2 + ...)$ Φ = cumulative normal function Try  $X \sim \text{normal}(\mu, \sigma^2)$   $\kappa(p) = \mu p + \frac{1}{2} \sigma^2 p^2$   $\kappa'(p) = \mu + \sigma^2 p$   $p = \sigma^{-2}(c - p)$   $\eta_0 = \sigma^{-1}(c - p)$   $\eta_1 = 0$ LD expansion exact

Try  $X \sim$  exponential (mean 1) Eexp( $\rho X$ ) = (1- $\rho$ )-1  $\kappa(\rho)$  = -ln(1- $\rho$ )  $\kappa'(\rho)$  = (1- $\rho$ )-1  $\rho$  = 1- $\rho$ -1  $\kappa''(\rho)$  = (1- $\rho$ )-2

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### Comparison $\eta_0 + \eta_1$ with Monte Carlo LD expansion 5.3221 exact 99.5%-ile 5.2983 # sims for same error 350 000 normal 99.5%-ile 3.5758 LD expansion 0.0048 exact 0.5%-ile 0.0050 -0.2 0.2 0.6 0.8 1.2 # sims for same error 180 000 -2] normal 0.5%-ile -1.5758 blue = normal(1,1) red = exponential

### LD Expansion Needs Analytical MGF

### Easy

### **Tricky**

- Normal
- Gamma
- Inverse Gaussian
- Reciprocal Inverse Gaussian
- Generalised hyperbolic
- Poisson / Neg Binomial compounds of the above
- Mixtures of the above
- Pareto
- Lognormal
- Weibull
- Copula approaches

Key question: Is there sufficient data to demonstrate we have a tricky problem?

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# Importance Sampling – How it Works

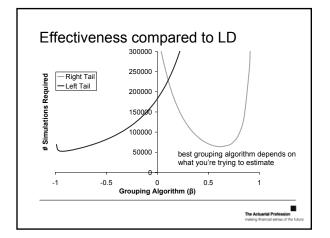
- 25
- Generate 1 000 000 simulations
- Group into 1 000 model points
- Outliers: treat individually
- Near the centre: groups of 5000 observations or more for each model point
- Result: 1 000 model points with as much information as 20 000 independent simulations

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# Importance Sampling: Another View

- We wish to simulate from an exp(1) distribution
  - density  $f(x) = \exp(-x)$
- Instead simulate for an exp(1-β) distribution
  - density  $g(x) = (1-\beta)exp[-(1-\beta)x]$
  - weight  $w(X) = (1-\beta)^{-1} \exp(-\beta X)$
- Use weighted average to calculate statistics
  - equivalent to grouping (yes it does work!)
- Product rule for multiple drivers

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### **Extreme Value Theory**

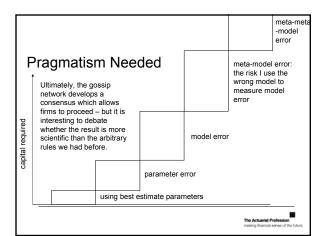
### Central Limit

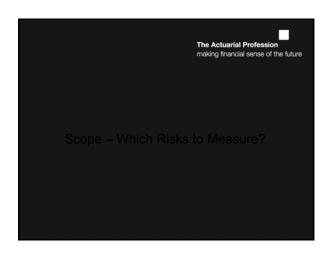
### Extreme Value

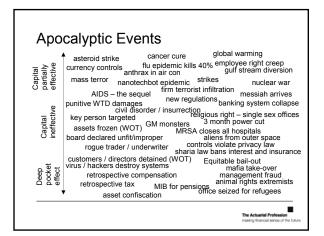
- Finite mean and variance
- Then the average A<sub>n</sub> is asymptotically normal
- Useful theorem because many distributions are covered
- Often need higher terms (eg LD expansion).
- If X₁, X₂, X₃ ... Xn are i.i.d.
   If X has an exponential / Pareto tail
  - Then (X-k|X>k) has an asymptotic exponential / Pareto distribution
  - Many distributions have no limit at all
  - Higher terms in the expansion poorly understood

### **Estimating Extreme Percentiles**

- Suppose "true" distribution is lognormal with parameters  $\mu$ =0,  $\sigma$ <sup>2</sup>=1.
- Simulate for 20 years
- Fit extreme value distribution to worst 10 observations
- Don't need to calculate to see this isn't going to work
- Instability and bias in estimate of 99.5%-ile
- The extreme event: if you have one in the data set its over-represented, otherwise its under-represented.
- Conclusion is invariably a judgment call was 11/09/2001 a 1-in-10 or 1-in-500 event? What's the worst loss I ever had / worst I can imagine call that 1-in-75.
- Problems even worse when trying to estimate correlations / tail correlations / copulas
- Reason to choose a simple model with transparent inputs

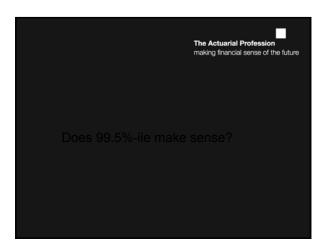








Apocalypse	Insolvent	Insufficient capital to continue	Sufficient cap for next year	ital
3%	0.5%	1.5%	95%	
%		p	robability	100%



# The Consultation Game Statement "Y" Statement "N" Capital Assessment at a 0.5%-ile is a great step forward for the industry. For the first time we have a logical risk-based Statement "N" Capital Assessment at a 0.5%-ile is a daft idea. The models are spurious, yet we have to employ an army of people to fill forms

approach to supervision

business.

which is also useful to the

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with numbers no sane

person has any faith in.

### **Model Consultation Process**

- Every firm must select "Y" or "N" and return this as the response to a consultation process.
- Firms must respond independently of other firms
- Regulator is inclined towards "Y" but can be persuaded to "N" if at least 75% of respondents vote "N".

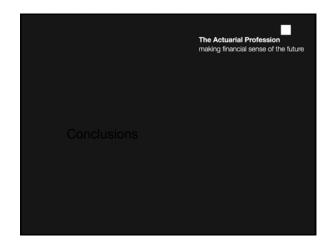
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# Model Consultation Payoff to firm X

	Firm X votes "Y"	Firm X votes "N"
ICA implemented	100	0 Humiliation / Retribution: objections to ICA misconstrued as technical incompetence
ICA scrapped	90 Some wasted effort preparing for ICA	100 Same as top left – so assume adoption of ICA or not is neutral for industry

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# Conclusion: Nash Equilibrium Conclusion: consultation processes tell regulators what they want to hear, irrespective of the underlying merits of what is being consulted. Number of Respondents The Actuarda Profession record from Name Procession and Procession record from Name Pro



### Conclusions

- Existing familiarity of value-at-risk gives it a head start over other approaches.
- Data and scope, but not maths, are the limiting factors for accurate capital calculations.
- If you prefer Monte Carlo, use importance sampling to cut burden by a factor of 5.
- Analytic large deviation theory is as good as 200,000 simulations – but much faster.

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