

The Actuarial Profession  
making financial sense of the future

# Operational Risk

Bending the tail of the dragon

Jim Gustafsson  
RSA Scandinavia & University of Copenhagen

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## AGENDA

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## AGENDA



Operational Risk

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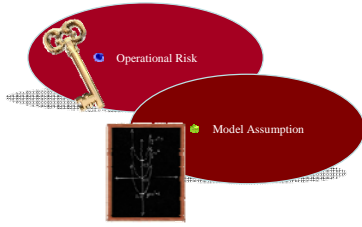
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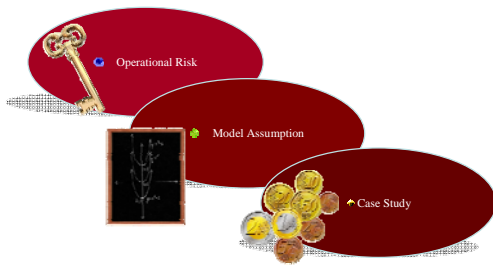
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# AGENDA



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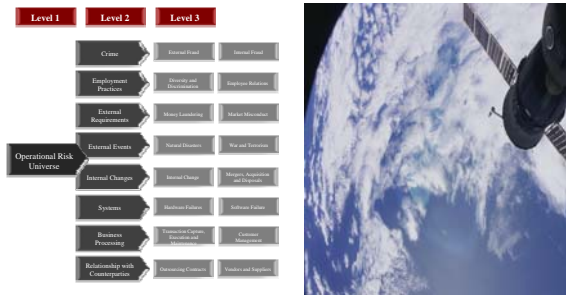
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# Operational Risk Universe



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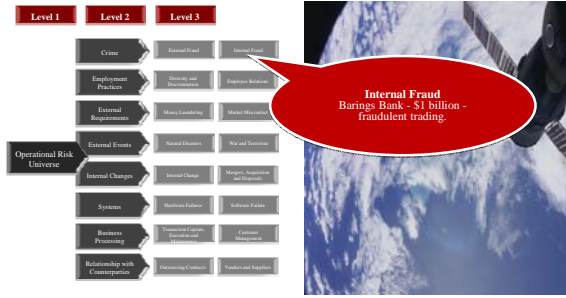
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# Operational Risk Universe



**Internal Fraud**  
Barings Bank - \$1 billion - fraudulent trading.

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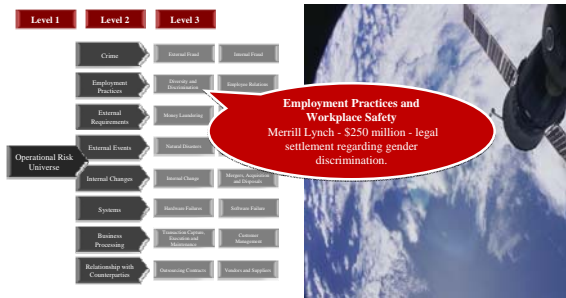
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# Operational Risk Universe



**Employment Practices and Workplace Safety**  
Merrill Lynch - \$250 million - legal settlement regarding gender discrimination.

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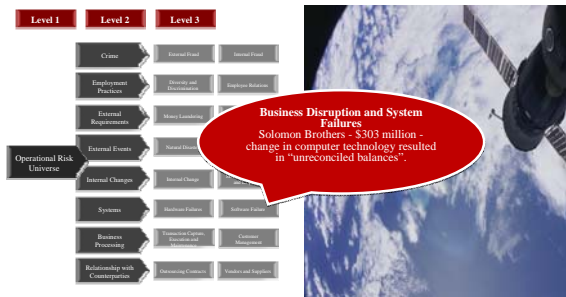
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# Operational Risk Universe



**Business Disruption and System Failures**  
Solomon Brothers - \$303 million - change in computer technology resulted in "unreconciled balances".

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## Operational Risk Universe

Level 1    Level 2    Level 3

Operational Universe

More than 100 losses exceeding \$100 Million over the last decade in the financial industry

Business Processes    Financial Capital    Customer Relationships  
 Relationship with Counterparties    Operating Controls    Vendor and Supplier

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## Operational Risk Loss Characteristic



High frequency low Impact operational risk events



Low frequency high impact operational risk events

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## Operational Risk Quantification

Loss Distribution Approach (LDA)

Monte Carlo Simulation

Frequency Distribution    Severity Distribution

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## Questions we hope to answer...

1. How much is the capital calculation affected by the choice of model for operational risk assessment?

2. Is it necessary to make use of more sophisticated models?

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## Data Availability

Internal data not collected.

Insufficient internal data.

Insufficient internal data / external data.

Internal data / external data.

Internal data / consortium data / publicly available data.



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## Scenario Analysis Model

Expert opinions on annual frequency and on percentile values of the severity.

Distribution assumption for the severity and frequency distribution.

Find the inverse function of the distribution.

Solve system of equations.

Weibull distribution

$$\begin{cases} -\beta (\log(1 - q_1))^\alpha = x_{q_1} \\ -\beta (\log(1 - q_2))^\alpha = x_{q_2} \end{cases}$$



$$\begin{aligned} \alpha &= \frac{\log(\log(1 - q_2) / \log(1 - q_1))}{\log(x_{q_2} / x_{q_1})} \\ \beta &= \left( \frac{x_{q_1}^\alpha}{\log(1 - q_1)} \right)^{1/\alpha} \end{aligned}$$

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## Parametric Model

- Parametric fit based on internal data / external data.
- Distribution assumption for the severity and frequency distribution.
- Maximum likelihood estimation.
- For example, the lognormal distribution.

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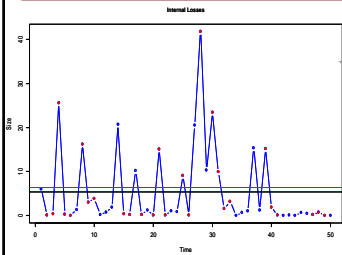
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## Underreporting Model

Underreporting means that not all losses in the company are reported.



- Occurred losses  $(X)_{t=0,1,2,\dots,T}$
- Indicator function  $I(t) = \begin{cases} 1 & \text{if } X_t \text{ is reported} \\ 0 & \text{otherwise} \end{cases}$
- Total number of reported  $N = \sum_{t=0}^{T-1} I(t)$
- Reported losses  $(Y)_{t=0,1,2,\dots,T}$

Ref: [1] & [2]

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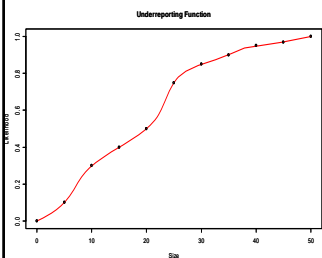
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## Underreporting Model

An underreporting function encodes the likelihood that a loss of particular size is reported.



- Underreporting function  $u(x)$
- Define  $g$  as a function of  $f$  and  $u$   
$$g(x) = \frac{f(x)u(x)^{-1}}{\int f(w)u(w)^{-1} dw}$$
- The probability of observing an operational and loss  $\int g(w)u(w)dw$
- Assume Poisson distribution, then  
 $M \in Po(\lambda)$ , and  
 $N \in Po\left(\lambda \int g(w)u(w)dw\right)$

Ref: [1] & [2]

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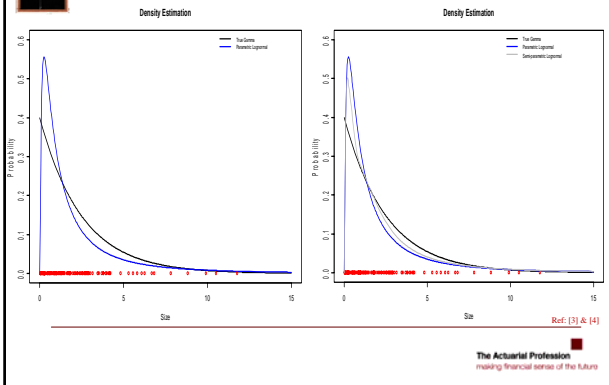
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# Semiparametric Model




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# Semiparametric Model

- When data are limited, the model is close to a parametric model.
- As the number of losses increases, the model becomes more non-parametric.
- Distribution Assumption: Estimate the distribution parameters on the data  $(X_i)_{i=1, \dots, n}$
- Transform losses to the bounded support  $[0, 1]$  with  $Z_i = F_n(X_i)$
- $K(t) = \frac{1}{100} \sum_{i=1}^{100} K_i(Z_i - t)$ ,  $K(t) = \frac{3}{4} (1 - t^2)$ ,  $K_i(t) = \frac{1}{b} K\left(\frac{t}{b}\right)$
- Semi-parametric density:  $\hat{f}(x) = f_n(x)g(x)$

Ref: [3] & [4]

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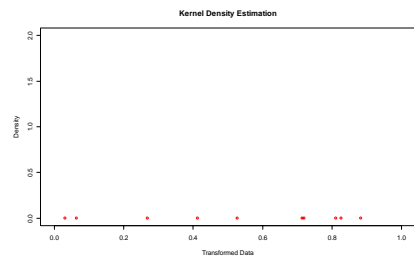
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# Semiparametric Model



Ref: [3] & [4]

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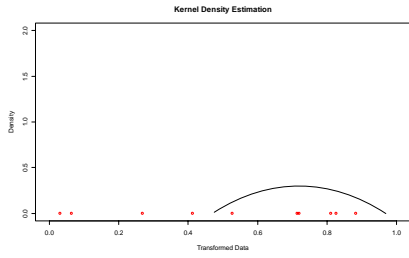
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## Semiparametric Model



Ref: [3] & [4]

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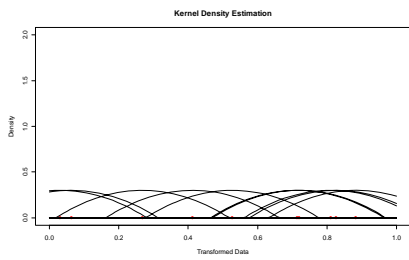
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## Semiparametric Model



Ref: [3] & [4]

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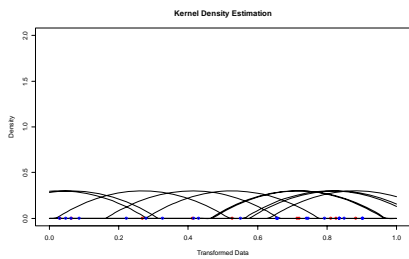
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## Semiparametric Model



Ref: [3] & [4]

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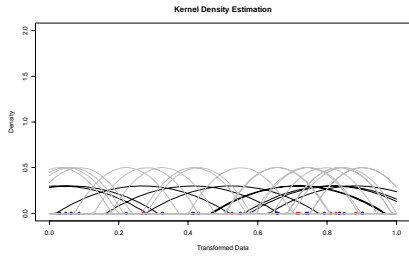
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## Semiparametric Model



Ref: [3] & [4]

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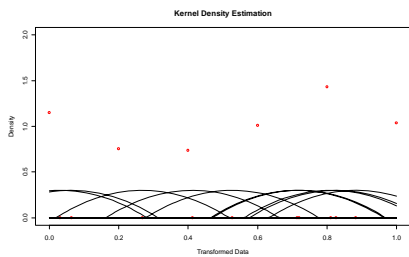
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## Semiparametric Model



Ref: [3] & [4]

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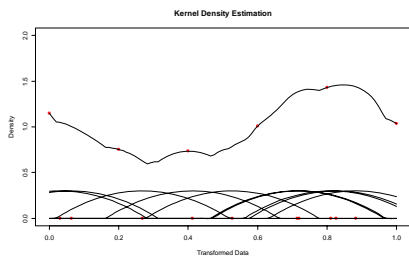
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## Semiparametric Model



Ref: [3] & [4]

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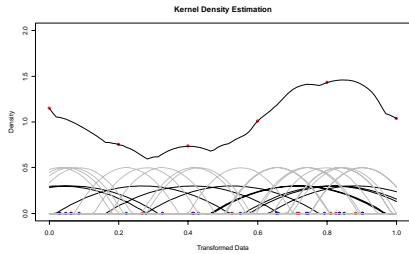
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## Semiparametric Model



Ref: [3] & [4]

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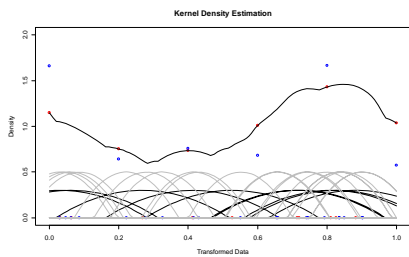
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## Semiparametric Model



Ref: [3] & [4]

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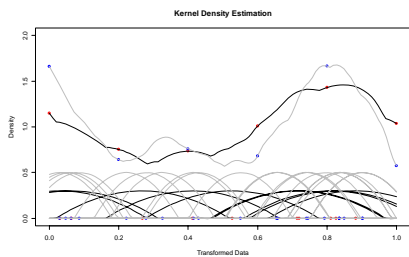
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## Semiparametric Model



Ref: [3] & [4]

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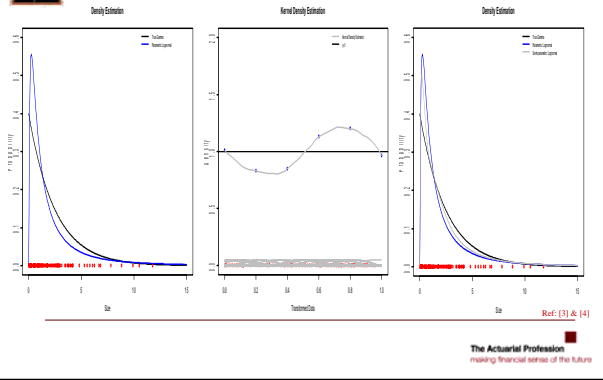
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## Semiparametric Model




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## Mixing Model

- Include prior knowledge from external data.
- Correct the external global start with internal observed data.
- Distribution Assumption: Estimate the distribution parameters on the data  $(Y_{i=1:n})$
- Transform losses to the bounded support  $[0, 1]$  with  $Z_i = F_{\hat{g}}(X_i)$
- Kernel density estimation:  $g(z) = \frac{1}{100} \sum_{i=1}^{100} K_h(Z_i - z)$ ,  $K(z) = \frac{3}{4} (1 - z^2) I_{(0,1)}(z)$
- Semi-parametric density:  $\hat{f}_h(x) = \hat{f}_g(x) g(x)$

Ref: [5]

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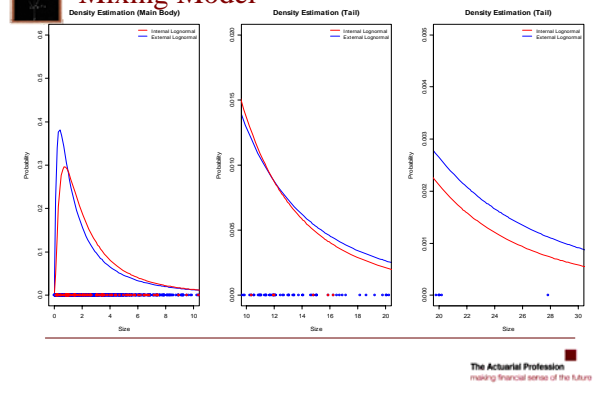
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## Mixing Model




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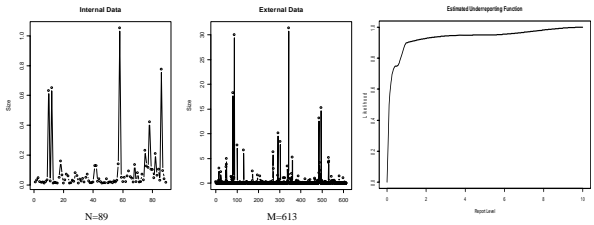
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## Case Study

Scenario Analysis  $\lambda_{0.95} = 0.05$ ,  $\lambda_{0.99} = 1$ ,  $\lambda = 30$




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## Case Study

### Summary Severity Assumption

Model 1	Scenario Analysis	Expert Opinion
Model 2	Parametric Lognormal	Internal Data
Model 3	Semi-Parametric Lognormal	Internal Data
Model 4	Parametric Lognormal Underreporting	Internal Data, Expert Opinion
Model 5	Parametric Lognormal	External Data
Model 6	Mixing	Internal Data, External Data
Model 7	Mixing Credibility	Internal Data, External Data

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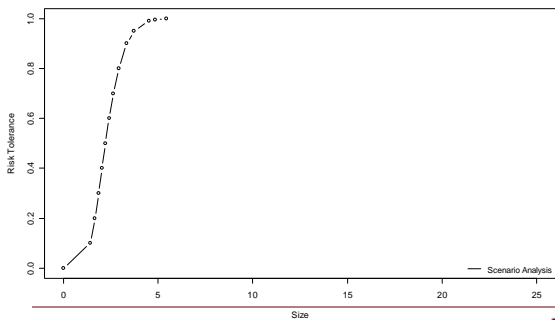
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## Case Study

Total Loss Distribution




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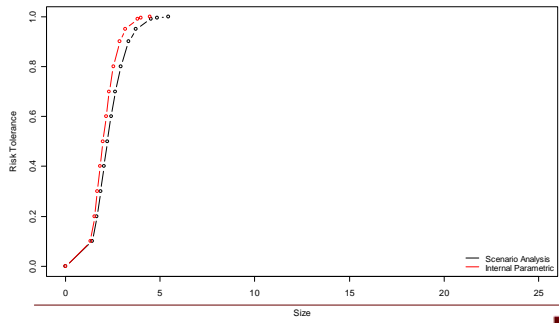
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## Case Study

Total Loss Distribution



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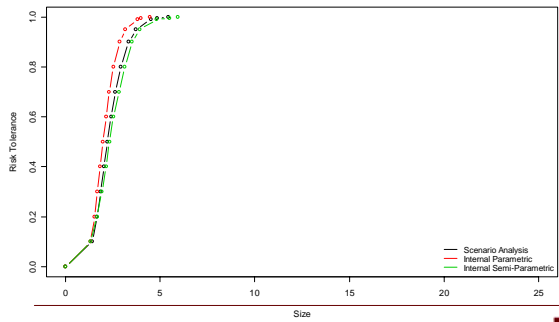
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## Case Study

Total Loss Distribution



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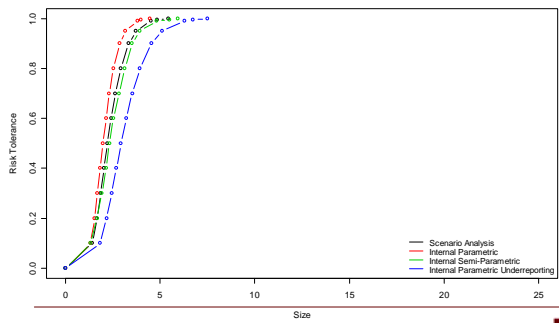
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## Case Study

Total Loss Distribution



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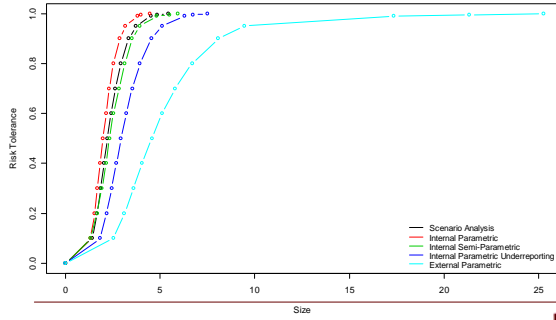
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## Case Study

Total Loss Distribution



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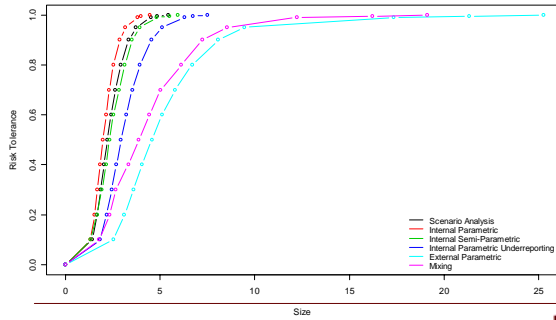
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## Case Study

Total Loss Distribution



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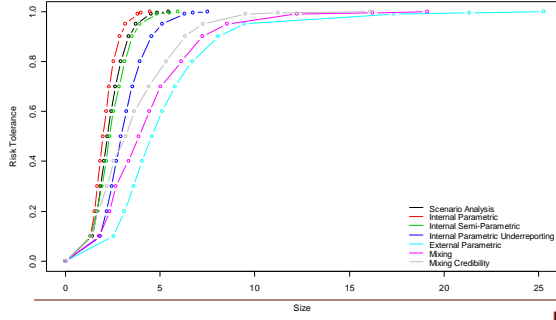
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## Case Study

Total Loss Distribution



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## Questions we hope to answer...

1. How much is the capital calculation affected by the choice of model for operational risk assessment?

Very much! The model choice depends on the information available.

2. Is it necessary to make use of more sophisticated models?

'All models are wrong, but some models are useful'  
- Box (1979).

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Thanks for your attention.

Questions?

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## References

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[7] Gustafsson, J., Nielsen, J.P., Pritchard, P. and Roberts, D. (2006). Quantifying Operational Risk Guided by Kernel Smoothing and Continuous credibility. *The ICFAI Journal of Financial Risk Management*, Vol. III, No. 3, 30-44.

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