

Modeling with small data: Credit and Operational Risk

13th congrès des actuaires, Peter Middelkamp, Small Data

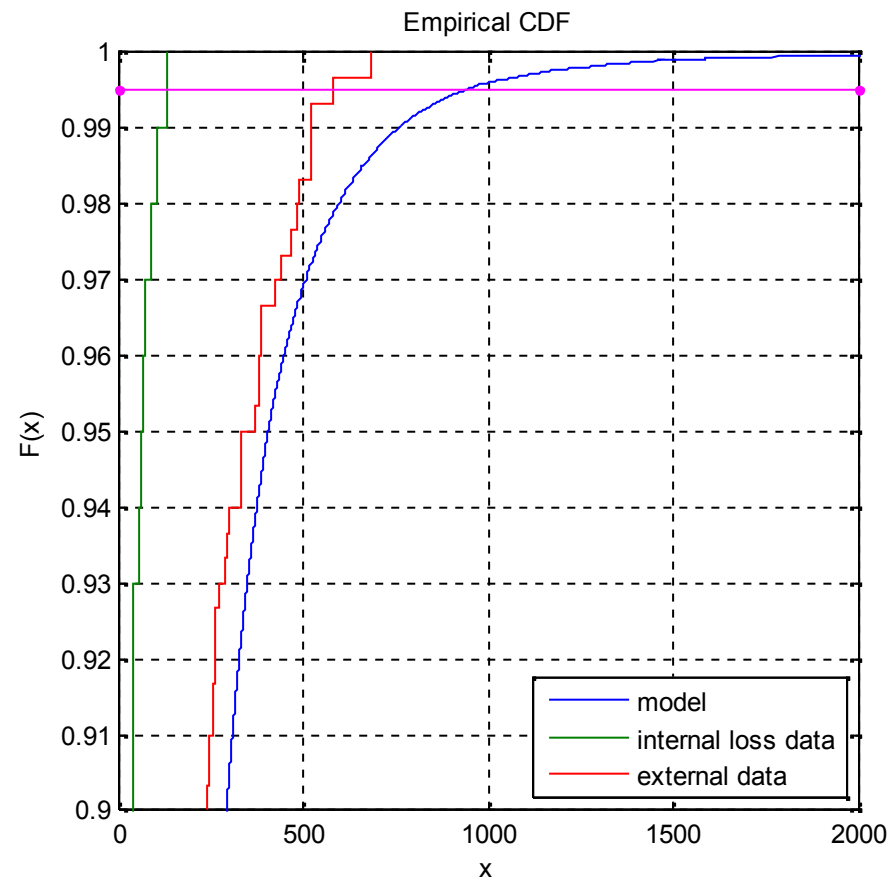


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Example 1: Operational Risk

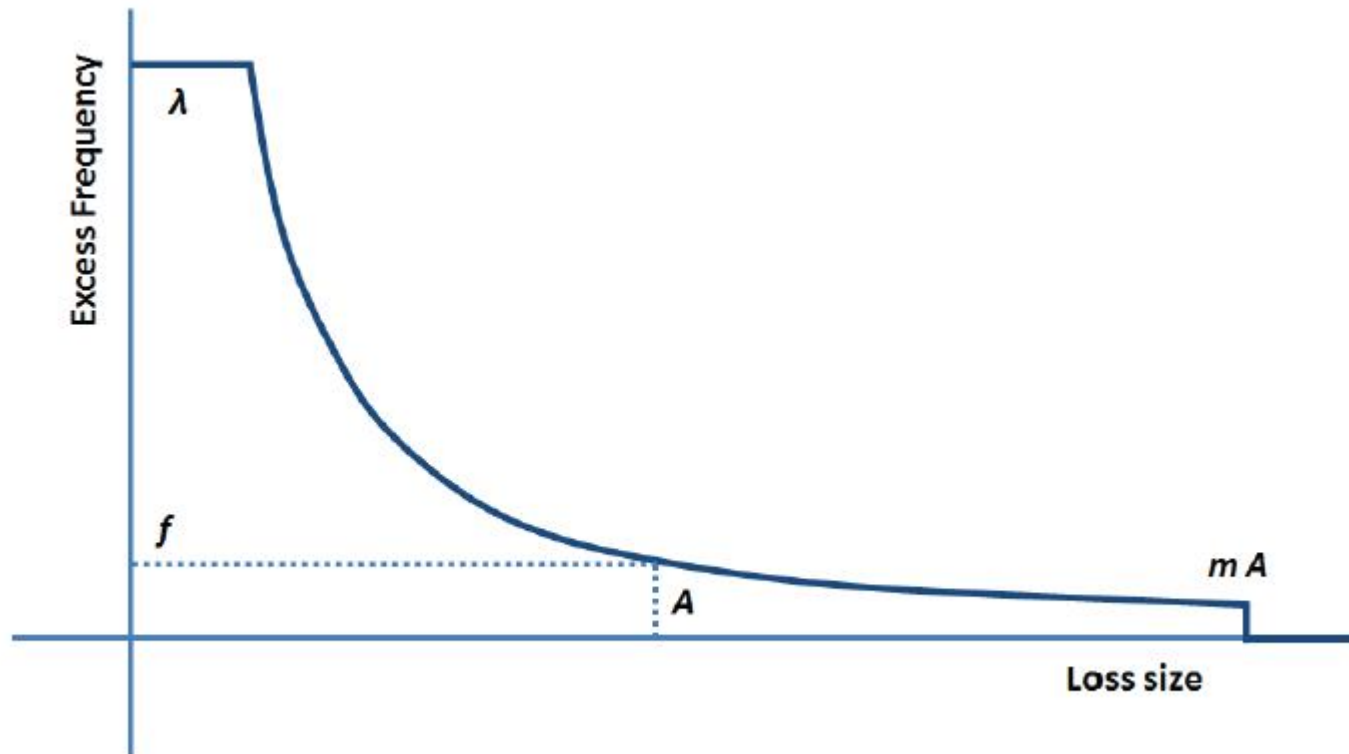
Operational Risk1 : (typically) Available Data

- Internal sources:
 - Own loss history collected over the past 5-10 years
 - Expert judgement of loss potentials
- External sources:
 - Industry loss data from loss data exchanges (eg: ORX)
 - Public information (eg: corporate filings, rating agencies, news, Algo First)



severity distribution per scenario and legal entity

- truncated Pareto with one important input: A = Anticipated Loss Potential for a given confidence level e.g. $f = 99.5\%$.



aggregation with dependence

- Dependence within operational risk:
 - Dependence between legal entities
 - Dependence between event categories
- External dependencies:
 - Dependence to financial markets (more prevailing in banks)
 - Dependence to pandemic events

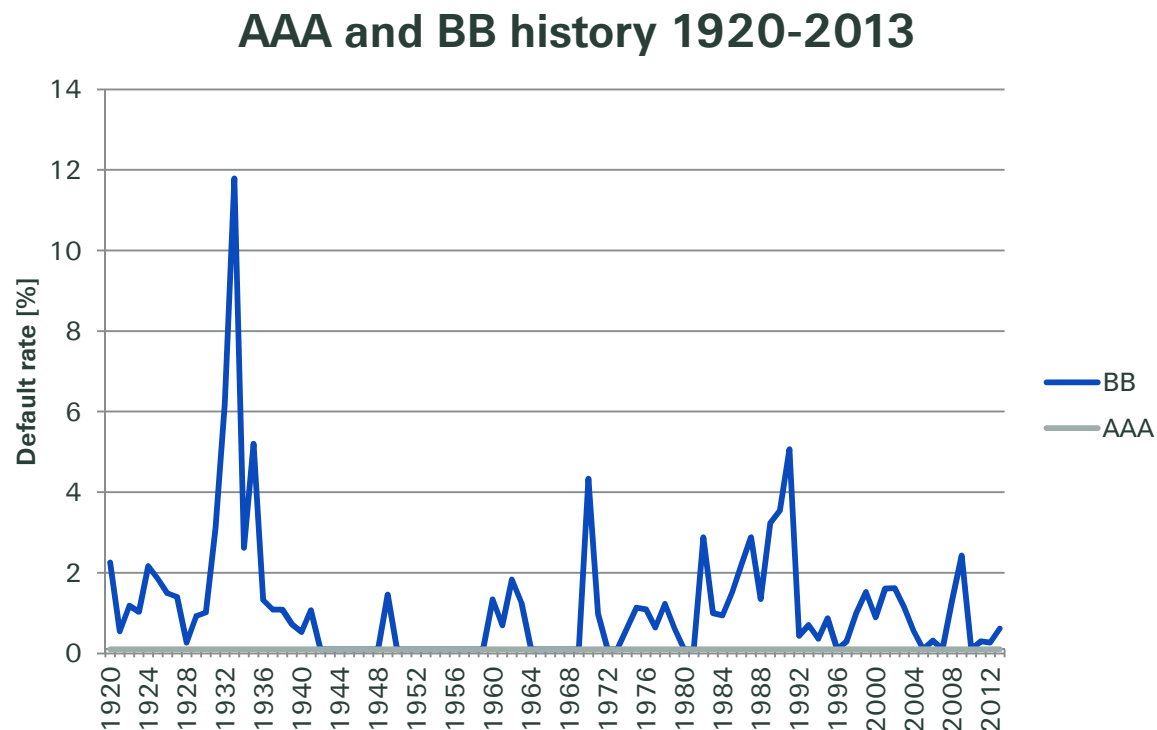
Conclusion Operational risk

- All parameters are mainly based on expert judgement that might be inspired by historic events in the firm and that happened to peers.
- A simple dependence method is a must (e.g. mixture of full dependence, independence)
- Number of parameters should be kept at a minimum to have a stable, easy to understand "well behaved" model.
- Continue to collect data and use it as benchmark for the model.
- Model alternatives (rejected):
 - LDA (not enough data)
 - Standard Formula (based on industry averages)
 - causal models with explicit control modelling (complicated, potentially unstable)
 - combining loss data sources and expert judgement using Bayesian inference (potentially unstable)

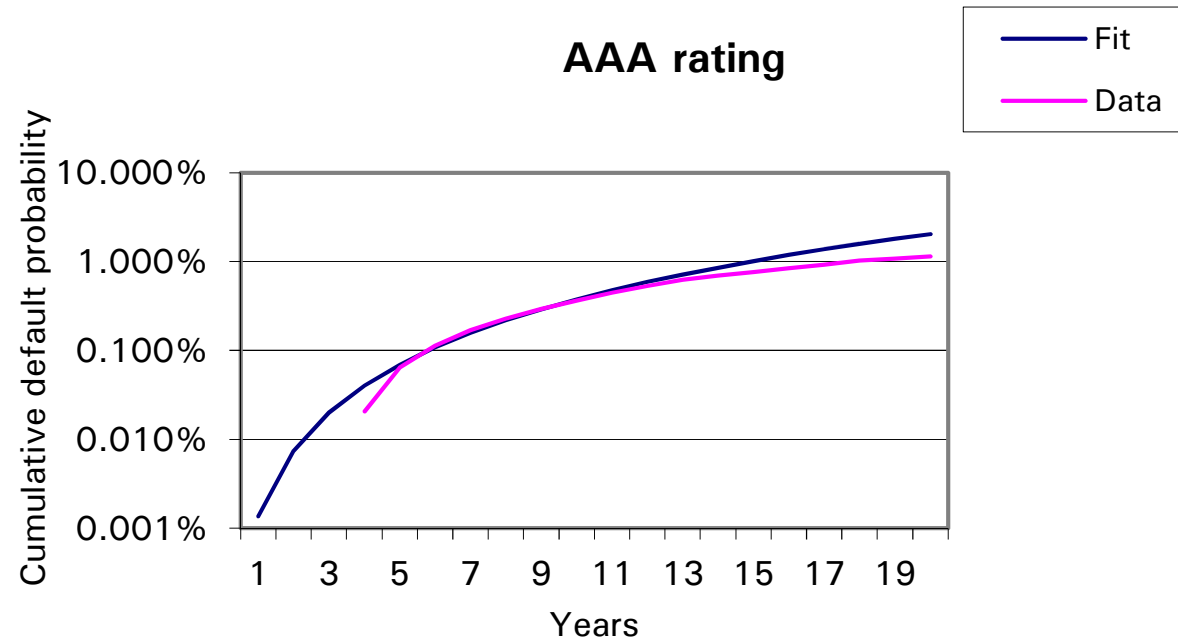
Example 2: Credit Risk

Compared to operational risk, there is plenty of data

- Default history exposes large spikes.
- Some rating categories never experienced a default within one year
- Rating information alone is not enough to determine default risk



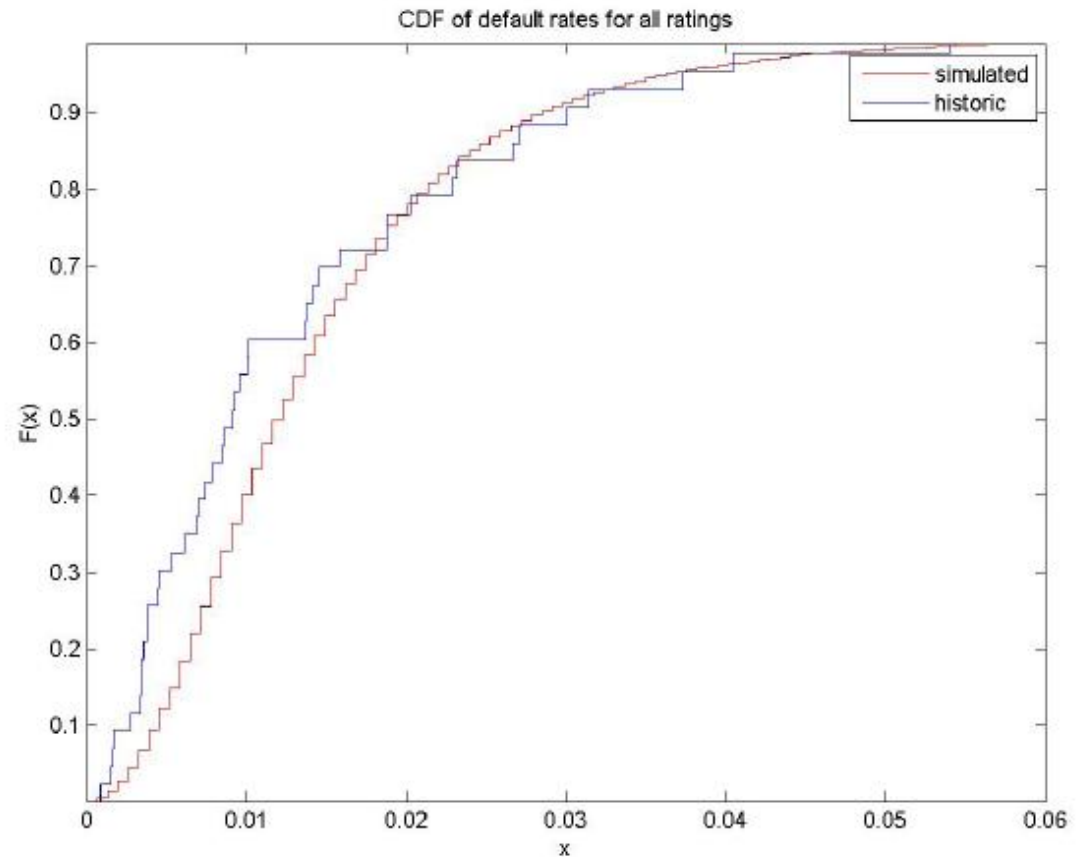
Multi year default probabilities can be used to estimate one year default probabilities



- Note that a minimum default probability of 0.03% is applied to corporates, 0.01% to sovereigns.

Modelled default probabilities should be able to reproduce and go beyond historic realisations

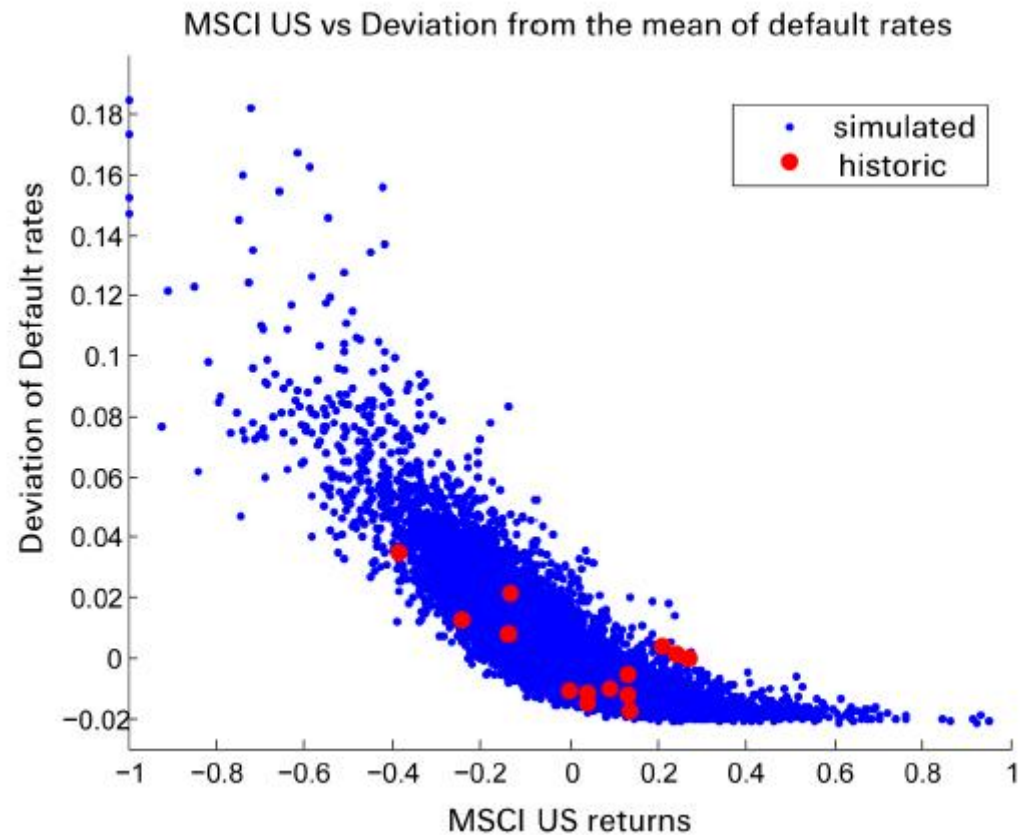
- Credit portfolio risk models capture historic default history (back-test)



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Credit portfolio risk models should also capture historic dependencies to financial markets (back-test)

- Default rates are higher when equity markets are down/ credit spreads widen (not that there is sometimes the default rate lags behind financial markets)





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