

Risk Management Group Conference
on
Leading Edge Issues in Operational Risk Measurement

Integrating External Data into the
OR Measurement Approach

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Operational Loss data

- Concern

- Insufficient internal historical loss data, especially for tail events.



- Issues

1. Fail to capture potential risks (i.e. tail events)
2. Unprecedented large loss amount has huge impact on marginal capital (i.e. lack of robustness)

Neither parametric nor non-parametric approach can deal with the above issues

well.	Description	Strength	Weakness
Parametric	Choosing a distribution and estimating its parameters	Generates a potentially fat tail.	-Lacks robustness -Less powerful when parametric assumptions are not met.
Non-parametric	Sampling from the empirical distribution	No parametric assumption about the distributions and parameters.	Does not generate a potentially fat tail.

Smoothed Bootstrap Methodology

- An intermediate solution between parametric distribution and non-parametric bootstrapping.

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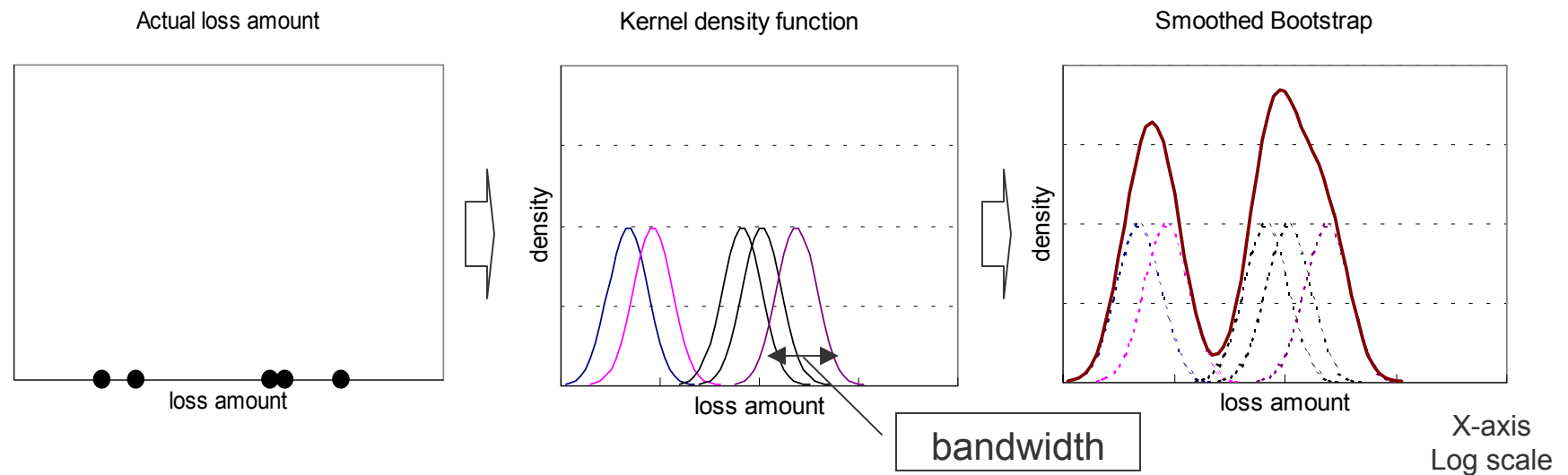


<i>Smoothed bootstrap</i>	<i>Generates a potentially fat tail without any assumptions about distributions.</i>
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- Easy mixture of internal and external data (or scenarios).

Smoothed Bootstrap Methodology

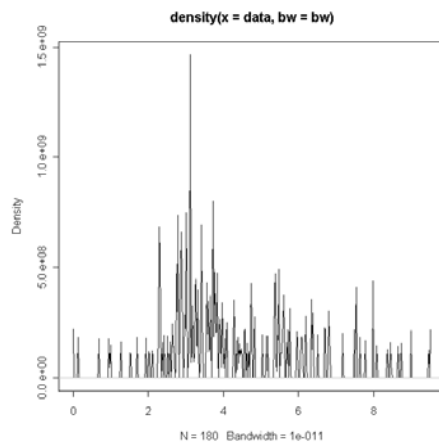
Methodology of smoothing and sampling



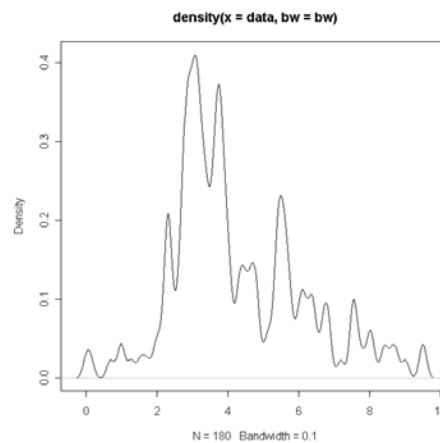
- Instead of re-sampling directly from the empirical distribution, smooth it first then the smoothed distribution is used to generate new samples. (Monte Carlo method)

Smoothed Bootstrap Methodology

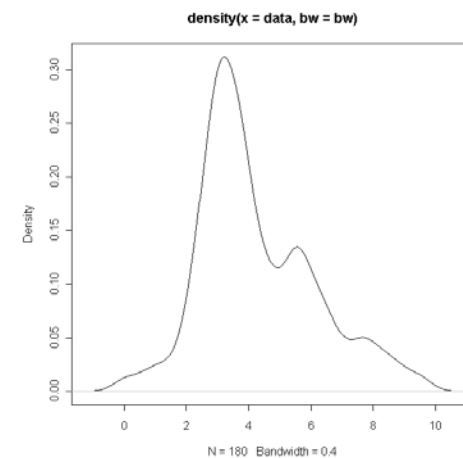
- Tail behavior of the smoothed distribution
 - Sample observations $N=180$



Bandwidth=0



Bandwidth=0.1



Bandwidth=0.4

X-axis
Log scale

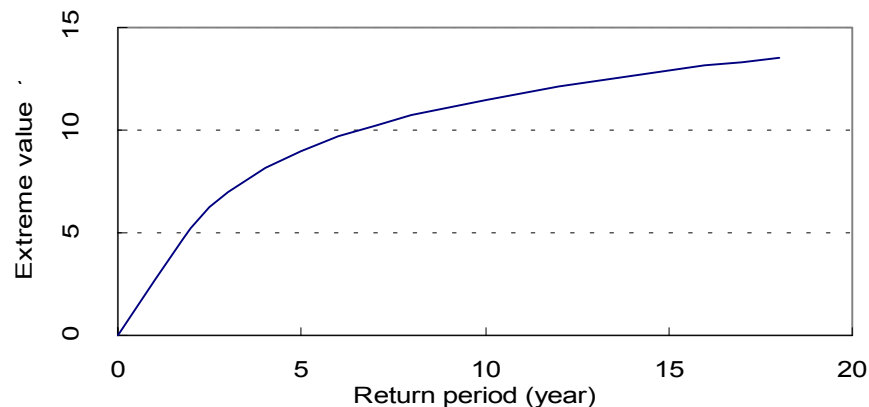


(Smoother
fatter tail)

- The larger the bandwidth, the fatter the tail of the distribution.
- Once the frequency and severity of the tail events are given by internal data, external data or scenarios, the bandwidth can be determined.

Extreme value during a return period

- The frequency and severity of the tail events are described as “Extreme value X during the N -year return period”
 - X is a threshold that is exceeded once per N -year return period on average.
- We get X and N by applying Gumbel distribution.



Process of smoothed bootstrap Application

- How to collect enough data to characterize the maximum value distribution

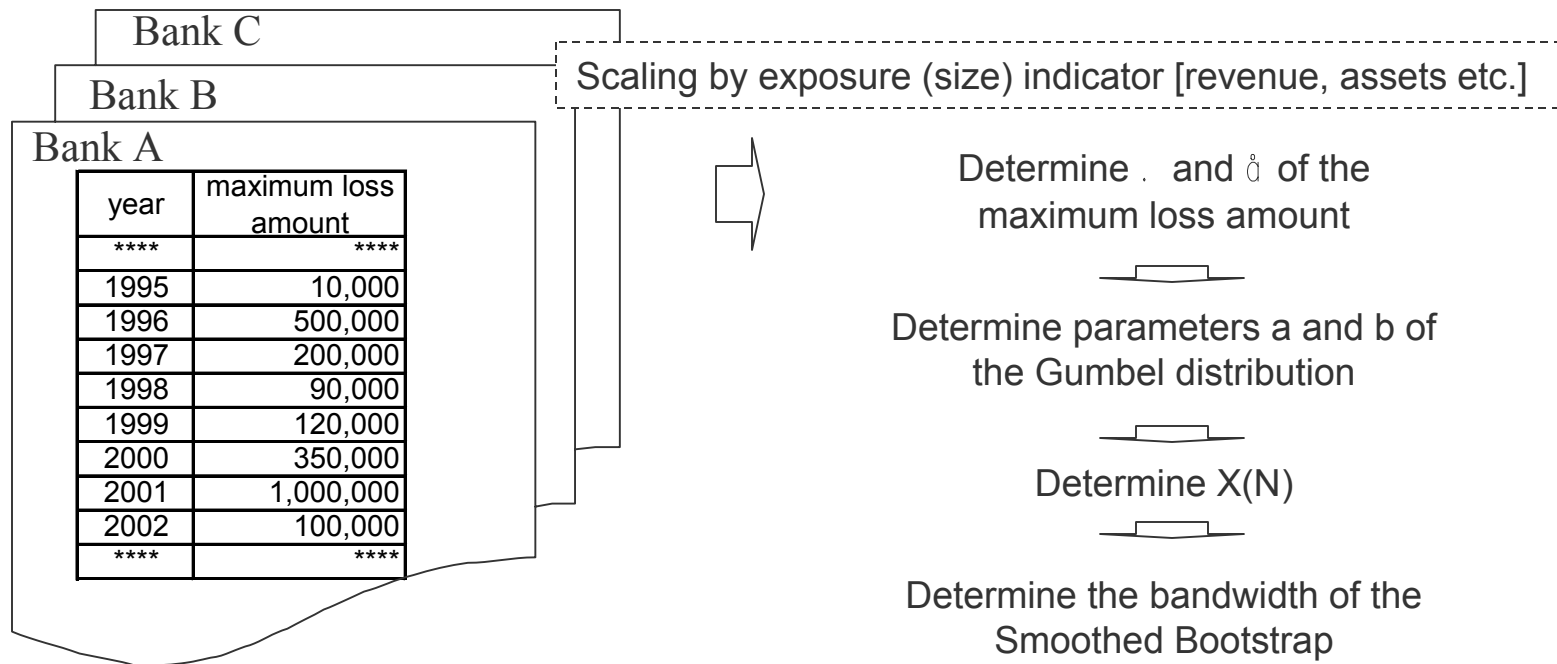
Case . : Use External Data

Case . : Use Scenario Analysis

Application [Case 1: External Data]

■ Case 1: External Data

Select maximum loss amount for each year.



Application [Case 2: Scenario Analysis]

■ Case 2: Scenario Analysis

Scenario data can be collected by using self-assessment including interviews, workshops and questionnaires.

<<Scenario output>>

Scenario type: *****

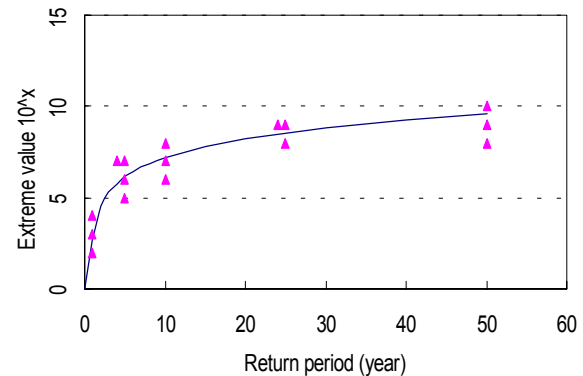
Respondent	Frequency	Severity
a	2	3
b	3	2
c	1	4
d	3	3
e	2	4
	4	2
	1	5

1 : once every 50 years
 2 : once every 25 years
 3 : once every 10 years
 4 : once every 5 years
 5 : once a year

1 : <100K
 2 : >100K and <1mil
 3 : >1mil and <10mil
 4 : >10mil and <100mil
 5 : >100mil



Regression analysis and curve fitting



Plot :
Each assessment

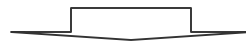
Determine $X(N)$

Determine the bandwidth of the Smoothed Bootstrap

Integration of internal and external data (scenarios)

- X(N) determined by *only* internal data.
 - Does not capture severe tail loss events.
 - Too sensitive to large loss amounts.

- X(N) determined by *only* external data (or scenarios).
 - Less sensitive to internal large loss amounts.
 - ✓ Required capital does not increase despite the occurrence of a large internal loss event.
 - ✓ Required capital does not decrease despite the improvement of the internal control.

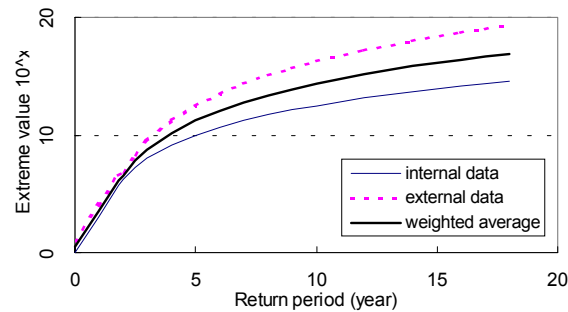


- Determine X(N) by *both* internal and external data (or scenarios).
 - Capture potential risks.
 - Reflect internal control factors.
 - Sensitive, but not too sensitive, to large loss amounts.

Integration of internal and external data (scenarios)

■ Combined distribution of $X_{internal}(N)$ and $X_{external}(N)$

$$Y \equiv \omega X_{internal}(N) + (1 - \omega) X_{external}(N)$$



- ✓ Combined distribution above is again the Gumbel distribution with the parameters $\mu = \omega \mu_{internal} + (1 - \omega) \mu_{external}$ and $\sigma = \omega \sigma_{internal} + (1 - \omega) \sigma_{external}$
- ✓ μ is determined for each event type.
- ✓ Assess weighting of internal and external data using factor analysis or similar method.

Test of Robustness

■ Sample calculation (How much marginal impact does the maximum loss amount have ?)

Sample observation

- Observation period : [n+1] • [n+5] year

	Number of loss events	Total loss amount US\$ mil.	Maximum loss amount US\$ mil.
Event type 1	58	11	5
Event type 2	5,012	39	13
Event type 3	212	110	30
Event type 4
Event type 5



Parametric model

Unexpected loss US\$ mil. (1y, 99.9%)
189
212
1,231
...
...

Smoothed Bootstrap

Unexpected loss US\$ mil. (1y, 99.9%)
150
223
970
...
...

- Internal data
 - Historical data for 5 years
- External data
 - Fixed
- Weighting of internal and external data
 - Internal data : 0.5
 - External data : 0.5

Next year

Sample observation

- Observation period : [n+2] • [n+6] year

	Number of loss events	Total loss amount US\$ mil.	Maximum loss amount US\$ mil.
Event type 1	60	17	6
Event type 2	4,895	55	20
Event type 3	223	322	200
Event type 4
Event type 5



Parametric model

Unexpected loss US\$ mil. (1y, 99.9%)	
354	• 165
312	• 100
3,803	• 2,572
...	
...	

Smoothed Bootstrap

Unexpected loss US\$ mil. (1y, 99.9%)	
174	• 24
343	• 120
1,520	• 550
...	
...	

= Summary =

Smoothed Bootstrap Methodology

- Smoothed Bootstrap Methodology is a desirable solution especially for small data sets.
 - Captures potentially severe tail loss events.
 - Reflects both
 - Internal loss data for calibrating the main body of the severity distribution, and
 - External loss data or scenarios for calibrating the tail of the severity distribution.
 - Robust about outlier data and sensitive to internal control.

Appendix

Smoothed Bootstrap Methodology

■ Define

$$\hat{f}(x) = \frac{1}{b} \sum_{j=1}^n K\left(\frac{x - x_j}{b}\right)$$

where

- » K: Kernel density function
- » x_j : Observations $j=1$ to n
- » b : Bandwidth

■ Suppose

- K : Normal distribution with mean x_j and standard deviation b
- x_j : Log-normal of loss amount ($j=1$ to n)

Smoothed Bootstrap Methodology

- Bandwidth can be determined by the frequency and severity of the tail event, which is described as
“Extreme value X during the N -year return period”

- X is a threshold that is exceeded once per N -year return period on average.
- X and N are related by the following equation :

$$\int_X^{\infty} \frac{1}{b} \sum_{j=1}^n K\left(\frac{x-x_j}{b}\right) dx = \frac{1}{\text{Accumulated loss number for } N \text{ years}} \quad [*1]$$

Probability of loss amount exceeding X

Once every N years

- Once X and N are given by the internal data, external data or scenarios, the bandwidth b can be determined by solving the equation [*1] for b .

Distribution of the maximum value

■ Distribution of the maximum value

- Given independent identically distributed (i.i.d.) random variables X_1, X_2, \dots, X_n with an underlying distribution function F .



- How to obtain the distribution fitting into maximum $M_n = \max\{X_1, X_2, \dots, X_n\}$.

e.g. distribution of the maximum loss amount selected for each year.



■ Extreme Value Theory (EVT)

Distribution of the Maximum Value is of one of the following three types:

- ✓ Type I (Gumbel-type) : The case where F belongs to a class of **medium-tailed** distributions, including normal, exponential, gamma and log-normal distributions.
- ✓ Type II (Frechet-type) : The case where F belongs to a class of **heavy-tailed** distributions, including Pareto, log-gamma, Cauchy etc.
- ✓ Type III (Weibull-type) : The case where F belongs to a class of **short-tailed** distributions, characterized by a finite right endpoint, including uniform and beta distributions.

Gumbel distribution

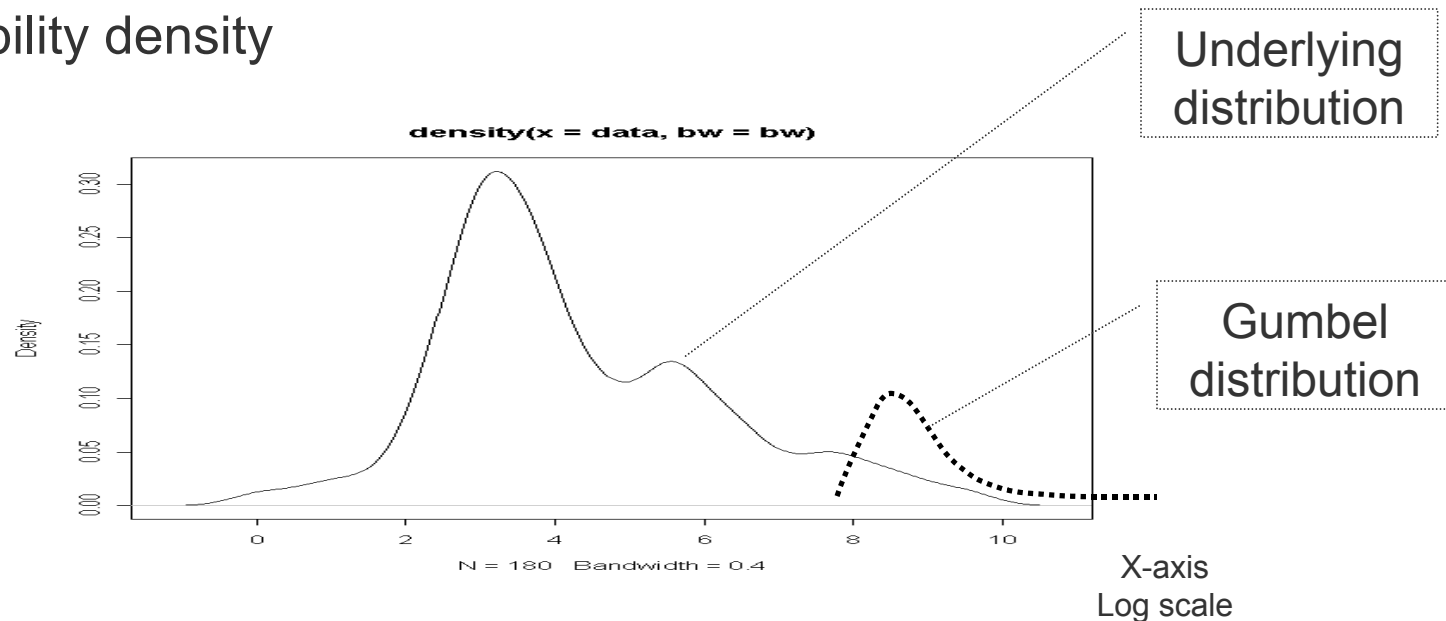
- Extreme value Type I distribution

- The case where underlying distribution F belongs to a class of **medium-tailed** distributions, including normal, exponential, gamma and log-normal distributions.

- Cumulative distribution function

$$G(x, a, b) = \exp(-\exp(-a(x - b)))$$

- Probability density



Extreme value during a return period

- We get $X(N)$ (“Extreme value X during the N -year return period”) by applying Gumbel distribution as follows :

- Frequency is the inverse of the return period N .



- X is the value of the upper $1/N^{\text{th}}$ percentile of the Gumbel cumulative distribution.

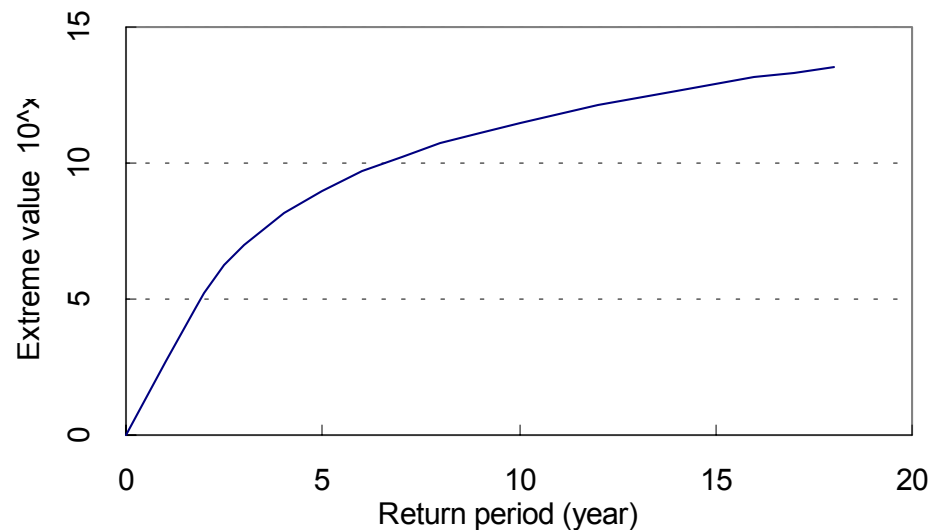
$$G(X, a, b) = \exp(-\exp(-a(X - b))) = 1 - \frac{1}{N} \quad [*2]$$



Extreme value during a return period

- Solving equation [*2] for X , we get

$$X(N) = b + \frac{1}{a} \left[-\ln \left\{ -\ln \left(1 - \frac{1}{N} \right) \right\} \right] \quad [*3]$$



Parameter estimation for the Gumbel distribution

■ Parameter estimation

- Estimators with the moment method for the Gumbel distribution are

$$a = \frac{\pi}{\sigma\sqrt{6}} \quad b = \mu - \frac{\sigma\sqrt{6}}{\pi}\gamma \quad [*4]$$

where

- μ and σ are the mean and standard deviation of the maximum loss amount observed for a given period, respectively.
- γ is Euler's number (constant 0.5772).

Process of smoothed bootstrap

- Observe maximum loss amount for each year.
- Calculate μ and σ of the maximum loss amount.
- Determine parameters a and b of the Gumbel distribution.

$$a = \frac{\pi}{\sigma\sqrt{6}} \quad b = \mu - \frac{\sigma\sqrt{6}}{\pi} \gamma \quad [*4]$$

- Determine $X(N)$.

$$X(N) = b + \frac{1}{a} \left[-\ln \left\{ -\ln \left(1 - \frac{1}{N} \right) \right\} \right] \quad [*3]$$

- Determine the bandwidth of the Smoothed Bootstrap.

$$\int_x^{\infty} \frac{1}{b} \sum_{j=1}^n K\left(\frac{x-x_j}{b}\right) dx = \frac{1}{\text{Accumulated loss number for } N \text{ years}} \quad [*1]$$