

SAV AG SST Nichtlebenmodell

Schätzung der Parameter

Part I

Zusammenfassung

1 Ziel dieses Dokuments

Das vorliegende Dokument verfolgt folgende Ziele:

- a) die Herkunft der derzeitigen Standard-Parameterwerte kurz darzulegen,
- b) die mathematischen Grundlagen betreffend Parameterschätzer unter Einbezug neuerer Entwicklungen in einem Dokument aufzuschreiben und festzuhalten,
- c) die jetzigen Standard-Parameterwerte zu diskutieren,
- d) Änderungsvorschläge und Anregungen für mögliche nächste Schritte zu geben.

2 Zur Historie der Parameter-Schätzer im SST

In 2003/2004 wurde das SST Standard-Modell Nichtleben erarbeitet. Charakteristisch sowohl für die Entwicklung des SST in 2003/2004 wie auch die ersten Jahre danach war der Wille zur Zusammenarbeit sowohl zwischen den einzelnen Gesellschaften wie auch zwischen der Aufsichtsbehörde und den Experten in der Versicherungsindustrie.

Nachdem das Standardmodell entwickelt war, mussten für den ersten Field-Test 2004 noch die zugehörigen Parameter geschätzt werden. Dies musste damals in sehr kurzer Zeit geschehen. Auch dies war eine Gemeinschaftsarbeit, bei der die SAV und die Versicherungsindustrie vertreten durch actuaries aus der Praxis sowie das damalige BPV involviert waren.

Die Korrelationsmatrizen wurden rein "Bayesianisch" ermittelt, d.h. man hat sich überlegt, welche Branchen korreliert sein könnten, und hat dann eine grobe Einteilung (keine Korrelation, Korrelationen von 0.25 und 0.5) vorgenommen.

Die Pareto-Parameter für die Einzel-Grossschäden wurden von mehreren grösseren Gesellschaften mit maximum likelihood geschätzt. Die einzelnen Gesellschaften meldeten ihre Resultate an das BPV, welches dann aufgrund dieser Umfrage die "Standard-Parameter" festlegte. Auch die Standardfaktoren für den Anteil der Einzel-Grossschäden

wurde via Umfrage ermittelt. Die Parameter für Kumulgrossschäden wurde aufgrund spezieller Überlegungen bestimmt.

Analog wurde beim Zufallsrisiko CY vorgegangen. Hier wurden die Variationskoeffizienten der Einzelschadenhöhen von mehreren Gesellschaften ermittelt und an das BPV gemeldet, welches dann die Standard-Werte festlegte.

Für das Parameterrisiko gab es noch keine mathematisch fundierte Theorie. Hier ist man pragmatisch vorgegangen. Beim Parameter-Risiko CY hat man sich bei den Berechnungen wo immer möglich auf Gemeinschaftsstatistiken abgestützt und vereinfachend angenommen, dass angesichts des grossen Volumens das Zufallsrisiko in den Zahlen der Gemeinschaftsstatistik vernachlässigt werden kann. Beim Parameter-Risiko für das Abwicklungsergebnis (Risiko PY) wurden Berechnungen mit der Mack-Formel für das ultimate Reserve-Risiko vorgenommen und daraus die Werte abgeleitet. Für das Zufallsrisiko gibt es keine Standardwerte und die Gesellschaften wurden angehalten, dies aufgrund der gesellschaftseigenen beobachteten Abwicklungsergebnisse in vergangenen Jahren abzuschätzen.

Mittels Umfrage bei den Gesellschaften wurden vom BPV auch die Default-Zahlungsmuster für die einzelnen Branchen ermittelt.

Aus Zeitgründen wurde damals in 2004 auf eine weitergehende Dokumentation bezüglich der Schätzung der Parameter verzichtet.

Die Modellierung von UVG mit dem Replikationsportfolio erfolgte etwas später und war das Resultat von Diskussionen innerhalb einer vom BPV geführten Arbeitsgruppe. Kurz zusammengefasst stecken folgende Überlegungen dahinter: die Finanzierung des Teuerungsteils auf den UVG-Renten ist durch die gesetzliche Grundlage zur Erhebung von Umlagebeiträgen und den Pool-Mechanismus abgesichert, so dass dafür keine zusätzlichen Rückstellungen (ausserhalb des Betrags im UVG-Pool zur Sicherung der indexierten Renten) und kein zusätzliches Risikokapital erforderlich sind. Mit der UVG-Regelung ist das Renten-Deckungskapital (inklusive Teil im UVG-Pool) mit dem 10-jährigen Durchschnitt der 10-jährigen Bundesobligationen zu bedienen. Unter der vereinfachenden Annahme, dass dieses Kapital konstant bleibt über die Zeit, generiert ein Portfolio bestehend aus 10 gleichen Tranchen von 10-jährigen Bundesobligationen automatisch die Verzinsungsverpflichtungen und ist somit ein replizierendes Portfolio für diese Verpflichtungen. Die vereinfachende Annahme eines konstanten Kapitals trifft natürlich nicht voll zu, doch ist man zum Schluss gekommen, dass der Fehler nicht gross ist und dass keine praktikable und bessere Alternativmethode zur Verfügung steht.

Auch die Parameter und die Modellierung von ES wurde aufgrund zwischenzeitlich erfolgter Tarifberechnungen aktualisiert und die bei diesen Berechnungen verwendete Modellierung übernommen.

Ansonsten sind die damals in 2004 ermittelten und festgelegten "Standard-Parameterwerte" unverändert beibehalten worden und entsprechen heute noch den im FINMA-Template erscheinenden Default Standard-Parameterwerten.

3 Zwischenzeitlich erfolgte theoretische Entwicklungen

In der Zwischenzeit gab es verschiedene Weiterentwicklungen auf theoretischem Gebiet. So ist insbesondere eine Formel für das einjährige Abwicklungsrisiko für das chain-ladder Reservierungsmodell entwickelt worden (Merz-Wüthrich [10], aber auch Gisler-Wüthrich [5] sowie Bühlmann & alias [2]). In [5] wird auch gezeigt, wie Credibility zur Schätzung der Abwicklungsfaktoren benutzt werden kann.

In der Masterarbeit von Gradenwitz [6] und in einer nachfolgenden Arbeit von Gisler [4] für das ASTIN-Kolloquium 2009 in Helsinki ("Helsinki-Arbeit") wird das SST-Standard-Modell Nichtleben mit dem Solvency II Modell verglichen und auch auf die Schätzung der Parameter eingegangen. Insbesondere wurden dort Schätzer für die Parameter-Risiken basierend auf einem soliden mathematischen Fundament entwickelt.

4 Zusammenfassung der Ergebnisse

Die diesem Dokument zugrundeliegenden mathematischen Grundlagen wurden in [4] publiziert und sind hier als Teil II beigefügt. In diesem Abschnitt werden die wesentlichen Ergebnisse zusammengefasst.

4.1 CY-Risiken

4.1.1 Zufallsrisiko CY (Normalschäden)

Dazu müssen einerseits die Variationskoeffizienten der Einzelschadenhöhen sowie die erwartete Anzahl Schäden geschätzt werden. Letzteres wird hier nicht kommentiert, da eine Prognose der Schadenanzahl in jeder Gesellschaft im Rahmen des Budgeting Prozesses und des Pricing vorhanden sein sollte.

Die Variationskoeffizienten können ganz einfach mittels der Formel (25) geschätzt werden. Der einzige zu beachtende Punkt ist, dass sich die Verteilung der Einzelschadenhöhen im Laufe der Abwicklung verändert und die Varianz in der Regel grösser wird (Grossschäden wachsen erst im Laufe der Abwicklung in die höheren Layer hinein). Die mit Formel (25) geschätzten Variationskoeffizienten weisen daher bei Long-Tail Branchen für die neueren und noch wenig abgewickelten Schadenjahre einen negativen Bias auf. Dies sollte bei der Schätzung berücksichtigt werden, indem entweder die erhaltenen Variationskoeffizienten für die neueren Jahre extrapoliert werden (z.B. mit chain ladder Prognose wie dargelegt in Abschnitt B.1) oder indem man die Werte für die neuesten zwei bis drei Jahre einfach weglässt und ein Mittel über den Rest der Jahre nimmt.

Die Variationskoeffizienten der einzelnen Branchen sind in der Regel recht stabil über die Zeit. Dieser Parameter lässt sich meist auch von mittleren und kleineren Gesellschaften aufgrund der eigenen Daten ermitteln.

4.1.2 Parameterrisiko CY (Normalschäden)

Das Parameter-Risiko kann mit der Formel (28) geschätzt werden. Dieser Schätzer ist mathematisch sauber abgestützt und basiert auf dem SST zugrundeliegenden Credibility

Modell.

Gemäss dem Credibility Modell hat das Parameter-Risiko eine spezielle Struktur und kann auch geschrieben werden als

$$\sigma_{param}^2 \simeq \text{Var}(\Theta_1) + \text{Var}(\Theta_2), \quad (1)$$

wobei der erste Term die Varianz des Erwartungswertes der Schadenanzahl und der zweite derjenige des Erwartungswertes der Schadenhöhe widerspiegelt. Diese spezielle Struktur wird im oben genannten Schätzer (28) nicht berücksichtigt.

An Stelle oder in Ergänzung von (28) kann auch (1) verwendet und die dort erscheinenden Komponenten $\text{Var}(\Theta_1)$ und $\text{Var}(\Theta_2)$ direkt geschätzt werden. Dieses Vorgehen ist in Branchen, wo die Jahresrisiken ein "natürliches Volumenmass" bezüglich Frequenz darstellen (z.B. MFH, Hausrat, jedoch nicht KKV, UVG, AH) angebracht. Der erste Term, $\text{Var}(\Theta_1)$ kann dann mit Formel (32) geschätzt werden. Die Schätzung von $\text{Var}(\Theta_2)$ ist etwas komplizierter (siehe Literaturhinweis in Abschnitt B.2). Oft wird der Schätzer einen Wert in der Nähe von Null oder Null ergeben, da die Variabilität in den Schadenhöhen auch bei bekanntem Risikoparameter Θ_2 recht hoch ist und diese eine allfällige zugrundeliegende Variabilität in den "wahren Erwartungswerten" dominiert. Der bedeutendere Term wird in der Regel $\text{Var}(\Theta_1)$ sein, so dass die Vernachlässigung von $\text{Var}(\Theta_2)$ oder eine Bayesianische Schätzung von $\text{Var}(\Theta_2)$ aufgrund von a priori Überlegungen (z.B. wie genau ist die Schätzung der Schadenteuerung (Variationskoeffizient)) in vielen Fällen angebracht und sinnvoll ist.

Als weitere Bemerkung sei angefügt, dass bei kleineren Gesellschaften die Schätzung des Parameter-Risikos aufgrund der eigenen Daten zu wenig aussagekräftig ist und die Verwendung der Standardwerte angezeigt ist. Allerdings wäre eine Überprüfung und Neuschätzung dieser Standardwerte durch die FINMA mehr als angebracht, insbesondere, da diese Werte damals sehr schnell und rudimentär festgelegt wurden.

4.1.3 Schätzung des Pareto-Parameters für die Schadenhöhe der Grossschäden

Der Pareto-Parameter für Pareto-verteilte Grossschäden wird klassisch mit Maximum-Likelihood geschätzt (siehe Formel (33)). In vielen Branchen ist jedoch die Anzahl der beobachteten Einzel-Grossschäden auch bei grösseren Gesellschaften relativ klein, so dass der ML-Schätzer eine erheblich Schätzungenauigkeit aufweist.

Doch gibt es eine gute mathematische Theorie, wie dieser Parameter mittels Credibility geschätzt werden kann (siehe Formel (35)). Die Verwendung dieses Schätzers ist nicht nur für kleine, sondern auch mittlere oder sogar grössere Gesellschaften sinnvoll.

Auch hier ist anzufügen, dass ein update der Default-Werte durch die FINMA mittels Umfrage bei den Gesellschaften angezeigt wäre. Mit den Daten dieses updates könnte auch der in (35) erscheinende Parameter κ ermittelt werden.

4.2 PY-Risiken

Für das einjährige Abwicklungsrisiko gibt es die Formel von Merz-Wüthrich für das chain-ladder Modell (Formeln (38) und (39)), die hier verwendet werden kann. Weitere Forschungen sind im Gange, und Resultate für das einjährige Abwicklungsrisiko für

weitere Reservierungsmodelle sind in den nächsten Jahren zu erwarten. Es lohnt sich, die Literatur hier etwas zu verfolgen.

Es ist absolut zwingend und wichtig zu sehen, dass es beim SST um das Risiko eines "Jahrhundert-Abwicklungsverlustes" geht. Solche "Jahrhundert-Ereignisse" fehlen in der Regel in den beobachteten Abwicklungsdreiecken. Das Reserve-Risiko wird daher unterschätzt, wenn man es einzig mit einem analytischen Modell und der Formel von Merz-Wüthrich ermittelt. Es ist zwingend zu ergänzen durch Szenario-Betrachtungen (siehe dazu auch die Ausführungen in Abschnitt B.4).

Für kleinere Gesellschaften besteht zudem die Schwierigkeit, dass die beobachteten chain-ladder Faktoren sehr volatil sein können. Auch hier könnte man mit Credibility-Theorie aus den beobachteten individuellen Daten und Industrie-weiten Daten eine bessere Schätzung für das Abwicklungspattern und für das Reserve-Risiko erhalten. Die Theorie dazu existiert (siehe [5]). Aufgrund der Beobachtungen von mehreren Gesellschaften könnten nicht nur die a priori zu erwartenden Abwicklungsmuster, sondern auch die für Credibility benötigten Parameterwerte aufgrund geschätzt werden. Dies würde jedoch eine Zusammenarbeit bzw. eine Erfassung von Abwicklungsdreiecken von mehreren Gesellschaften erfordern.

4.3 Korrelationen

Wie in Abschnitt A.3 dargelegt, sind konzeptionell drei Korrelationsmatrizen zu bestimmen:

\mathbf{R}_{CY} , \mathbf{R}_{PY} , $\mathbf{R}_{CY,PY}$ (für CY, PY, sowie Korrelationen zwischen CY und PY). Das ist auch mit den derzeitigen Standard-Parameter für die Korrelationsmatrizen so, doch ist mit den heutigen Werten \mathbf{R}_{PY} die Identitätsmatrix, während $\mathbf{R}_{CY,PY}$ aus lauter Nullen besteht.

Bei der Festsetzung dieser Werte in 2004 war man sich der Wirkung von Kalenderjahr-Effekten zu wenig bewusst. Solche Kalender-Jahr Effekte (z.B. Inflation) induzieren sowohl eine Korrelation des Abwicklungsergebnisses zwischen verschiedenen Branchen wie auch eine Korrelation zwischen Schadenaufwand GJ und Abwicklungsergebnissen, sprich CY- und PY Risiken. Die zur Zeit gültigen Standardwerte für die Korrelationsmatrizen geben daher ein zu optimistisches Bild ab. Das versicherungstechnische Risiko wird damit stark unterschätzt. Zu beachten ist auch, dass in Solvency II bedeutend höhere Korrelationen verwendet werden.

Die Bestimmung der Korrelationsmatrizen mit quantitativen Methoden ist allerdings nicht ganz einfach. Ein mögliches Vorgehen mittels eines "Risk-Driver" Ansatzes ist in Abschnitt B.5 dargelegt.

Fest steht jedoch, dass die derzeitigen Standard-Parameter für die Korrelationsmatrix zu optimistisch sind. Eine Überarbeitung ist nicht nur angezeigt, sondern zwingend nötig. Hier besteht nach Meinung der AG Handlungsbedarf von Seiten der FINMA. Die AG ist auch der Meinung, dass ein VA im Wissen um diese Mängel ab sofort diese zu "optimistischen" Default-Korrelationsannahmen nicht mehr verwenden sollte und etwas vorsichtigere Annahmen (z.B. etwas in Richtung Solvency II) verwenden sollte.

Part II

Parameter estimation in the SST: an extract from the "Helsinki Paper"

Abstract:

This part is mainly an extract of [4] (paper presented at the ASTIN Colloquium 2009 in Helsinki). It concentrates on the part of that paper with regard to the parameter estimation and consists of two parts:

part A: modelling,

part B: the estimation of the corresponding parameters.

A Modelling in the SST

In the following we briefly summarise some of the model assumptions used in the SST

A.1 Modelling of the normal claim amount current year

Model Assumptions A.1 *For each line of business i it is assumed that the current year (=next coming year) is characterised by its risk characteristics $\Theta_i^T = (\Theta_{1i}, \Theta_{2i})$, where Θ_{1i} and Θ_{2i} are independent with $E[\Theta_{1i}] = E[\Theta_{2i}] = 1$, and that conditional on $\Theta_i^T = (\Theta_{1i}, \Theta_{2i})$*

- *the normal claim amount $C_i^{CY,n}$ is compound Poisson distributed*
- *with Poisson parameter $\lambda(w_i, \Theta_{1i}) = w_i \lambda_i \Theta_{1i}$, where w_i is a known weight and where λ_i is the a priori expected claim frequency,*
- *and with claim severities $Y_i^{(\nu)}$ having the same distribution as $\Theta_{2i} Y_i$, where Y_i has a distribution $F_i(y)$ with $E[Y_i] = \mu_i$.*

Remarks:

- The underlying risk characteristics $\Theta_i^T = (\Theta_{1i}, \Theta_{2i})$ reflect the "state of the nature" in the next coming year for lob i . Indeed things like weather conditions, economic situation, change in legislation, etc. might have an impact on the claim number and the claim severity. These "conditions" may vary from year to year and the imposed changes increase the risk and affect big companies as much as small companies, i.e. this risk cannot be diversified. The "true claim frequency" in the coming year will be $\lambda_i(\Theta_{1i}) = \lambda_i \cdot \Theta_{1i}$, hence Θ_{1i} is the random factor by which next year's "true underlying claim frequency" will deviate from the a priori expected claim frequency λ_i . The interpretation for Θ_{2i} is quite analogous, but with respect to the "true underlying expected value of the claim severity".

- Note, that no distributional assumption about the claim severity is made in model assumptions A.1. But it will later be assumed that

$$C_{\bullet} = C_{\bullet}^{CY,n} + C_{\bullet}^{PY}$$

can be approximated by a lognormal distribution with the corresponding first and second moments (see Section A.3). Another idea would be to assume a lognormal distribution on the level of lines of business. This would be a slightly different model, since the sum of lognormal distributions is not lognormal any more. But the difference would presumably be rather small without a significant impact on the result, but it would have the disadvantage to be much more complicated and the distribution of C_{\bullet} could not be expressed by a simple analytical formula any more.

We introduce

$$\tilde{P}_i = E \left[C_i^{CY,n} \right],$$

which is the pure risk premium for normal claims, and

$$X_i = \frac{C_i^{CY,n}}{\tilde{P}_i} \quad (2)$$

the corresponding loss ratio.

One can show that with model assumptions A.1

$$\sigma_i^2 := \text{Var} (X_i) = \sigma_{i,param}^2 + \frac{\sigma_{i,fluct}^2}{\nu_i}, \quad (3)$$

where

$$\sigma_{i,param}^2 = \text{Var} (\Theta_{1i}) + \text{Var} (\Theta_{2i}) + \text{Var} (\Theta_{1i}) \cdot \text{Var} (\Theta_{2i}) \quad (4)$$

$$\simeq \text{Var} (\Theta_{1i}) + \text{Var} (\Theta_{2i}), \quad (5)$$

$$\sigma_{i,fluct}^2 = \text{CoV} a^2 \left(Y_i^{(v)} \right) + 1. \quad (6)$$

and where

$\text{CoV} a \left(Y_i^{(v)} \right)$ = the coefficient of variation of the claim severity,

$\nu_i = w_i \lambda^{(i)}$ = a priori expected number of claims.

Since $\tilde{P}_i = \nu_i \mu_i$ we can also write (3) as

$$\sigma_i^2 := \text{Var} (X_i) = \sigma_{i,param}^2 + \frac{\tilde{\sigma}_{i,fluct}^2}{\tilde{P}_i}, \quad (7)$$

where

$$\tilde{\sigma}_{i,fluct}^2 = \mu_i \sigma_{i,fluct}^2.$$

Note that $\sigma_i^2 = \text{CoV} a^2 \left(C_i^{CY,n} \right)$ is composed of two components, the *parameter risk* and the *random fluctuation risk*. The parameter risk is independent of the size of the company whereas the fluctuation risk decreases with the size of the company respectively with the number of a priori expected claims.

Remark:

- There are *standard values* provided by the supervision authority for $\sigma_{i,param}^2$ and $\sigma_{i,fluct}^2$. Companies might deviate from them based on estimates from their own data. Usually there is sufficient data in most of the companies to estimate $\sigma_{i,fluct}^2$, but there has not been so far a sound methodology to estimate $\sigma_{i,param}^2$ from the own data. Therefore most companies use the default values for $\sigma_{i,param}^2$. In Section B we fill the gap and derive new estimators for estimating $\sigma_{i,param}^2$ from the company-own data.

For calculating the variance of the total normal claim amount (summed up over all lines of business) we have to make assumptions on the correlation of the normal claim amount of different lines of business. Let

$$\begin{aligned}\rho_{ij}^{CY} &: = \text{Corr} \left(C_i^{CY,n}, C_j^{CY,n} \right), \\ X_{\bullet} &: = \frac{C_{\bullet}^{CY,n}}{\widetilde{P}_{\bullet}}.\end{aligned}$$

Then we obtain

$$\text{Var} \left(C_{\bullet}^{CY,n} \right) = \sum_{i,j=1}^I \widetilde{P}_i \sigma_i \widetilde{P}_j \sigma_j \rho_{ij}^{CY} \quad (8)$$

and

$$\sigma^2 := \text{Var} \left(X_{\bullet} \right) = \frac{1}{\widetilde{P}_{\bullet}^2} \sum_{i,j=1}^I \widetilde{P}_i \sigma_i \widetilde{P}_j \sigma_j \rho_{ij}^{CY}. \quad (9)$$

It is convenient to write (8) in matrix notation. Let

$$\begin{aligned}\mathbf{X} &= (X_1, X_2, \dots, X_I)^T, \\ \mathbf{W}_{CY} &= \left(\widetilde{P}_1 \sigma_1, \widetilde{P}_2 \sigma_2, \dots, \widetilde{P}_I \sigma_I \right)^T, \\ \mathbf{R}_{CY} &= \text{Corr} \left(\mathbf{X}, \mathbf{X}^T \right).\end{aligned}$$

Note that \mathbf{R}_{CY} denotes the correlation matrix of \mathbf{X} with the entries

$$\mathbf{R}_{CY} (i, j) = \rho_{ij}^{CY}.$$

(8) written in matrix notation becomes

$$\text{Var} \left(C_{\bullet}^{CY,n} \right) = \mathbf{W}_{CY}^T \cdot \mathbf{R}_{CY} \cdot \mathbf{W}_{CY}. \quad (10)$$

A.2 Modelling of the Reserve Risk

Let

$$\begin{aligned}L_i &= \text{outstanding claims liabilities at 1.1. for lob } i, \\ R_i &= \text{best estimate of } L_i \text{ per 1.1.} = \text{best estimate reserve}, \\ \widetilde{R}_i &= PA_i^{PY} + R_i^{31.12.,PY} = \text{best estimate of } L_i \text{ per 31.12.,}\end{aligned}$$

where PA_i^{PY} are the claim payments for previous years' claims and $R_i^{31.12.,PY}$ are best estimate reserves per 31.12. of previous years' claims

Note that R_i is known at the beginning of the year, whereas \widetilde{R}_i is still a random variable (PA_i^{PY} and $R_i^{31.12.,PY}$ will only be known at the end of the year).

The claim amount PY becomes

$$C_i^{PY} = \widetilde{R}_i - R_i.$$

As for the current year risk, we introduce something analogous to the loss ratio, but with the ingoing reserves R_i (instead of the premiums) as "weight", that is we consider

$$Y_i = \frac{\widetilde{R}_i}{R_i}.$$

Note that $E[Y_i] = 0$ and that $C_i^{PY} = (Y_i - 1)R_i$.

Model Assumptions A.2 (reserve risk SST) *It is assumed that*

$$\tau_i^2 := \text{Var}(Y_i) = \tau_{i,param}^2 + \frac{\tau_{i,fluct}^2}{R_i}. \quad (11)$$

Remarks:

- The SST distinguishes between parameter risk and random fluctuation risk, but the model assumptions A.2 are not written down anywhere in an official document of the SST. As pointed out in [8] the question of how to quantify the reserve is risk is not yet fully answered. Thus the variance structure (11) is my own interpretation and is not yet an integral part of the current SST. I think that from a modelling point of view there are good reasons to assume the above structure, even if this structure is not reflected in the current standard parameters, where the coefficients of variation of the random fluctuation risk do not depend on the size of the reserves. Note that (11) has the same structure as (7).
- As for the current year risk it is again assumed that the variance consists of two components, the parameter risk independent of the weight (=volume of reserves) and the random fluctuation risk which is inversely proportional to this weight. The parameter risk reflects the estimation error which is a risk measure of the deviation of the reserve R_i from the true expected value $E[L_i]$ of the outstanding liabilities L_i , whereas the random fluctuation risk encompasses the pure random fluctuation of the ultimate around $E[L_i]$, which is also called *process error* in claims reserving.

In order to calculate the variance for the reserve risk summed up over all lines of business, we have again to make assumptions about the correlation of the reserve risk of different lob. According to the current SST standard assumption it is assumed that there is no correlation between the reserve risk of different lob.

Discussion and remarks on the correlation assumption for PY risks in the SST

- The current standard assumption of no-correlation between the reserve risks of different lines of business is questionable. Calendar year factors may affect the reserves of several lines of business simultaneously and impose a positive correlation. An obvious calendar year factor is certainly inflation, but there might also be others like change in legislation.
- A change of the correlation assumption can be done well within the SST modelling framework. Hence the no-correlation assumption does not question the SST-standard-model as such, but is rather a matter of the parameter choice within the SST standard model.

Let

$$\begin{aligned} Y_{\bullet} &= \frac{\widetilde{R}_{\bullet}}{R_{\bullet}}, \\ \mathbf{Y} &= (Y_1, Y_2, \dots, Y_I)^T, \\ \mathbf{W}_{PY} &= (R_1\tau_1, R_2\tau_2, \dots, R_I\tau_I)^T, \\ \mathbf{R}_{PY} &= \text{Corr}(\mathbf{Y}, \mathbf{Y}^T). \end{aligned}$$

Then we obtain

$$\text{Var}(C_{\bullet}^{PY}) = \mathbf{W}_{PY}^T \cdot \mathbf{R}_{PY} \cdot \mathbf{W}_{PY}, \quad (12)$$

and

$$\tau^2 := \text{Var}(Y_{\bullet}) = \frac{1}{R_{\bullet}^2} (\mathbf{W}_{PY}^T \cdot \mathbf{R}_{PY} \cdot \mathbf{W}_{PY}). \quad (13)$$

The no-correlation standard assumption means that \mathbf{R}_{PY} is the identity matrix. Then (13) becomes

$$\tau^2 = \frac{1}{R_{\bullet}^2} \sum_{i=1}^I R_i^2 \tau_i^2. \quad (14)$$

A.3 Modelling of the sum of normal claim amount CY and claim amount PY in the SST

Let

$$S_i = C_i^{CY,n} + \widetilde{R}_i, \quad (15)$$

$$Z_i = \frac{C_i^{CY,n} + \widetilde{R}_i}{\widetilde{P}_i + R_i} = \frac{\widetilde{P}_i X_i + R_i Y_i}{\widetilde{P}_i + R_i}. \quad (16)$$

Whereas \widetilde{P}_i and R_i can be considered as the weights attached to X_i and Y_i respectively, the corresponding weight for Z_i is the sum

$$V_i = \widetilde{P}_i + R_i. \quad (17)$$

From (16) we immediately see that

$$\varphi_i^2 := \text{Var}(Z_i) = \frac{\left(\widetilde{P}_i \sigma_i\right)^2 + 2\widetilde{P}_i \sigma_i R_i \tau_i \text{Corr}(C_i^{CY}, C_i^{PY}) + (R_i \tau_i)^2}{V_i^2}. \quad (18)$$

At the end we are interested in the variance of the total claim amount CY and PY summed up over all lines of business. Denote by

$$\mathbf{R}_{CY, PY} = \text{Corr}\left(\mathbf{C}^{CY}, \mathbf{C}^{PYT}\right)$$

the correlation matrix between the claim amounts CY and the claim amounts PY with entries

$$\mathbf{R}_{CY, PY}(i, j) = \text{Corr}(C_i^{CY}, C_j^{PY}).$$

As correlation matrix for the joint random vector of claim amounts CY and claim amounts PY we obtain

$$\begin{aligned} \mathbf{R} &= \text{Corr}\left(\left(\begin{array}{c} \mathbf{C}^{CY} \\ \mathbf{C}^{PY} \end{array}\right), \left(\begin{array}{c} \mathbf{C}^{CY} \\ \mathbf{C}^{PY} \end{array}\right)^T\right) \\ &= \begin{pmatrix} \mathbf{R}_{CY} & \mathbf{R}_{CY, PY} \\ \mathbf{R}_{CY, PY} & \mathbf{R}_{PY} \end{pmatrix}. \end{aligned}$$

Let

$$S_{\bullet} = C_{\bullet}^{CY} + \widetilde{R}_{\bullet}, \quad (19)$$

$$Z_{\bullet} = \frac{C_{\bullet}^{CY} + \widetilde{R}_{\bullet}}{\widetilde{P}_{\bullet} + R_{\bullet}}. \quad (20)$$

Then we obtain

$$\text{Var}(S_{\bullet}) = \begin{pmatrix} \mathbf{W}_{CY} \\ \mathbf{W}_{PY} \end{pmatrix}^T \cdot \mathbf{R} \cdot \begin{pmatrix} \mathbf{W}_{CY} \\ \mathbf{W}_{PY} \end{pmatrix}, \quad (21)$$

$$\text{Var}(Z_{\bullet}) = \frac{1}{V_{\bullet}^2} \left\{ \begin{pmatrix} \mathbf{W}_{CY} \\ \mathbf{W}_{PY} \end{pmatrix}^T \cdot \mathbf{R} \cdot \begin{pmatrix} \mathbf{W}_{CY} \\ \mathbf{W}_{PY} \end{pmatrix} \right\}, \quad (22)$$

where

$$V_{\bullet} = \widetilde{P}_{\bullet} + R_{\bullet}.$$

Since

$$Z_{\bullet} = \sum_{i=1}^I \frac{V_i}{V_{\bullet}} Z_i,$$

we can also express the variance of Z_{\bullet} in terms of the Z-variables, i.e.

$$\begin{aligned} \text{Var}(Z_{\bullet}) &= \sum_{i,j=1}^I \frac{V_i V_j \varphi_i \varphi_j}{V_{\bullet}^2} \text{Corr}(Z_i, Z_j) \\ &= \frac{1}{V_{\bullet}^2} \mathbf{V}^T (\mathbf{W}_{CY}^T \cdot \mathbf{R}_{CY} \cdot \mathbf{W}_{CY} + 2 \cdot (\mathbf{W}_{CY}^T \cdot \mathbf{R}_{CY, PY} \cdot \mathbf{W}_{CY}) + \mathbf{W}_{PY}^T \cdot \mathbf{R}_{PY} \cdot \mathbf{W}_{PY}) \mathbf{V}, \end{aligned} \quad (23)$$

where $\mathbf{V} = (V_1, V_2, \dots, V_I)^T$.

(22) is valid for any correlation matrix \mathbf{R} . The standard assumption in the current SST is that there is no correlation between current year risks and previous year risks, which means that $\mathbf{R}_{CY, PY}$ is a zero-matrix (matrix with only zeros). Together with the assumption that the reserve risks of different lob are not correlated we obtain

$$\text{Var}(Z_\bullet) = \frac{\widetilde{P}_\bullet^2 \sigma^2 + R_\bullet^2 \tau^2}{(\widetilde{P}_\bullet + R_\bullet)^2}. \quad (24)$$

Discussion and remarks on the correlation assumption between CY and PY risks:

- The current standard assumption in the SST that CY risks and PY risks are not correlated, is questionable. There might well be *calendar year factors* affecting both, the CY and the PY results. The most obvious of these factors is claims inflation, which has an impact on both, CY claims and PY claims.

Model Assumptions A.3 (normal claim amount CY plus claim amount PY) *It is assumed that S_\bullet has a lognormal distribution with*

$$\begin{aligned} E[S_\bullet] &= \widetilde{P}_\bullet + R_\bullet, \\ \text{Var}(S_\bullet) &= (\mathbf{W}_{CY}, \mathbf{W}_{PY})^T \cdot \mathbf{R} \cdot (\mathbf{W}_{CY}, \mathbf{W}_{PY}). \end{aligned}$$

Remark:

- This model assumption has to be interpreted in the way that a lognormal distribution is a sufficiently accurate approximation to the distribution of S_\bullet for solvency purposes.

A.4 Modelling of the Big Claim Amount Current Year

Big claims are defined as single individual claims or claim events (natural catastrophes) with a claim amount exceeding a certain threshold (e.g. 1 m CHF or 5 m CHF). For both types of big claims it is assumed that the following model assumptions hold true.

Model Assumptions A.4 (big claim amount SST)

i) The big claim amount for line of business i has a compound Poisson distribution

$$C_i^{CY, b} = \sum_{\nu=1}^{N_i^b} Y_{i\nu}^b,$$

where the number of big claims N_i^b is Poisson-distributed with Poisson parameter λ_i^b and where the claim severities $Y_{i\nu}^b$, $\nu = 1, 2, \dots, N_i^b$, are independent and independent from N_i^b with distribution function F_i .

ii) $\{C_i^{CY,b} : i = 1, 2, \dots, I\}$ are independent.

Remarks:

- The claim severity distributions F_i are essentially assumed to be Pareto with Pareto parameters α_i . Upper limits can easily be taken into account by truncation of the Pareto distributions. Also xs-reinsurance is no problem to handle by this model.
- Big claim events like hail, wind-storm, flood etc. usually affect several lines of business. However, this is not relevant in this context. Such an event claim might then be associated to the lob with the highest expected claims load. The essential assumption is that all these big claims are independent and compound Poisson.

From these assumptions it follows that

$$C_{\bullet}^{CY,b} = \sum_{i=1}^I C_i^{CY,b}$$

is again compound Poisson with parameter

$$\lambda = \lambda_{\bullet}^b = \sum_{i=1}^I \lambda_i^b$$

and claim severity distribution

$$F = \sum_{i=1}^n \frac{\lambda_i^b}{\lambda_{\bullet}^b} F_i.$$

The compound Poisson-distribution of $C_{\bullet}^{CY,b}$ can for instance be calculated by means of the Panjer-algorithm.

A.5 Aggregation

Let

$$\tilde{T} = C_{\bullet}^{CY,n} + C_{\bullet}^{CY,b} + C_{\bullet}^{PY}$$

be the total claim amount before scenarios. According to model-assumptions A.3, the distribution of $S_{\bullet} = C_{\bullet}^{CY} + C_{\bullet}^{PY}$ is lognormal and according to model-assumptions A.4, the distribution of $C_{\bullet}^{CY,b}$ is compound Poisson. Thus the distribution function $\tilde{F}(x)$ of \tilde{T} is obtained by convoluting the lognormal distribution of S_{\bullet} with the compound Poisson distribution of $C_{\bullet}^{CY,b}$.

A.6 Taking the scenarios into account

Insurance scenarios SC_k^{ins} are situations or events producing an estimated loss c_k (face amount) with probability p_k . In the SST standard model, beside prescribed scenarios, the company should also think about company specific scenarios. Hence the number K of the scenarios considered by a company is not fixed. However, in the standard SST it is

assumed that the scenarios $SC_k^{ins}, k = 1, 2, \dots, K$ exclude each other, i.e. that only one of the scenarios can materialize during the next year. This exclusion property was made because of simplicity reasons and is not such a big restriction as it might seem on the first side. If one wants to incorporate the situation, that e.g. SC_i^{ins} and SC_j^{ins} might occur in the same year, one can add a new scenario SC_{new}^{ins} with face amount $c_{new} = c_i + c_j$ and a corresponding probability p_{new} , for instance $p_{new} = p_i p_j$ if the two scenarios are assumed to be independent.

Let

$$T = C_{\bullet}^{CY,n} + C_{\bullet}^{CY,b} + C_{\bullet}^{PY} + SC_{\bullet}^{ins},$$

be the total claim amount including scenarios, and denote by $F(x)$ the distribution of T . Define

$$\begin{aligned} p_0 &= 1 - \sum_{k=1}^K p_k \text{ and} \\ c_0 &= 0. \end{aligned}$$

Then

$$F(x) = \sum_{k=0}^K p_k \tilde{F}(x - c_k),$$

where $\tilde{F}(\cdot)$ is the distribution before scenarios.

B Parameter Estimation

The standard parameter values in the STT and in solvency II differ substantially. This fact by itself reveals that the estimation of the parameters is a problem on its own.

The SST framework does not yet provide formulas for estimating the parameters from individual data. In the following we present such estimators.

B.1 Estimation of the random fluctuation risk normal claims CY

In Section A.1 we have seen that

$$\sigma_{i,fluct}^2 = CoVa^2 \left(Y_i^{(\nu)} \right) + 1.$$

Hence we have to estimate the coefficient of variation of the claim sizes. For this purpose we calculate first the observed coefficient of variations for different accident years j , that is

$$\widehat{CoVa}_{ij} = \sqrt{\frac{\frac{1}{N_{ij}-1} \sum_{v=1}^{N_{ij}} \left(Y_{ij}^{(\nu)} - \bar{Y}_i \right)^2}{\bar{Y}_i^2}}, \quad (25)$$

where N_{ij} is the number of normal claims of lob i in accident year j , $Y_{ij}^{(\nu)}, \nu = 1, 2, \dots, N_{ij}$, the individual claim amounts and \bar{Y}_i the mean of the $Y_{ij}^{(\nu)}$. However, one should have in

mind that the individual claim amounts $Y_{ij}^{(\nu)}$ in recent accident years j consist to a great part of case estimates. It is a known fact that, especially in long tail lines of business, (25) underestimates in recent accident years the true ultimate coefficient of variation. Therefore, the values resulting from (25) should first be extrapolated to their ultimate value. The following table shows the development triangle of the coefficient of variation in motor liability of a big Swiss insurance company.

Development triangle CoVa claim amounts in motor liability

AY/DY	0	1	2	3	4	5	6	7	8	9
1998	6.1	7.1	8.1	8.5	9.1	9.7	10.1	10.2	10.2	10.9
1999	6.1	7.6	8.1	9.0	10.0	10.3	10.6	10.6	11.2	11.2
2000	5.8	6.9	8.4	8.9	9.4	9.5	9.5	10.3	10.3	10.3
2001	5.8	7.4	8.9	9.5	9.6	9.6	10.4	10.7	10.7	10.7
2002	5.7	7.8	9.0	9.0	9.2	9.9	10.3	10.6	10.6	10.6
2003	6.6	8.5	8.8	9.1	9.8	10.2	10.6	10.9	10.9	10.9
2004	6.6	8.5	9.2	10.1	10.7	11.1	11.5	11.9	11.9	11.9
2005	5.2	7.2	8.7	9.2	9.7	10.1	10.5	10.8	10.8	10.8
2006	5.4	7.5	8.5	9.0	9.5	9.9	10.3	10.6	10.6	10.6
2007	5.7	7.4	8.4	8.8	9.4	9.7	10.1	10.4	10.4	10.4
									mean	10.8

The values above the diagonals are the observed empirical coefficients of variation resulting from (25) of accident years i calculated at different development years. The values below the diagonals are simple chain ladder forecasts. We can see from this table that for instance in the development years 0 or 1 the coefficients of variation are substantially underestimated compared to the ultimate values. We can also see that the observed coefficients of variation are rather stable and that the mean of the extrapolated coefficients of variation (column 9) over the different accident years is an accurate estimator for the coefficient of variation of the claim sizes. Indeed, this parameter can usually be estimated with great accuracy from the own data.

B.2 Estimation of the parameter risk of the normal claim amount CY

For the parameter risk, most companies in Switzerland use the default standard values provided by the supervision authority. One of the reasons is that there has not yet been developed an estimator for the parameter risk for the own data which is based on a sound actuarial basis. In the following we try to fill this gap and we will present such estimators.

We assume that we have observed historical data on the accident years $j = 1, 2, \dots, J$. We consider a specific lob i . To simplify notation we drop in the following the index i and we write for instance X_j for the observation of the specific lob considered in the year j . It is assumed,

- i) that each year is characterised by its characteristics $\Theta_j = (\Theta_{j1}, \Theta_{j2})^T$ and that for each year the model-assumptions A.1 are fulfilled with underlying claim frequency parameter λ_j and claim severity parameter μ_j ,
- ii) that the coefficient of variation for the claim sizes $CoVa(Y_j^{(\nu)})$ is the same for all years,

iii) that random variables belonging to different years are independent and $\Theta_1, \Theta_2, \dots, \Theta_J$ are independent and identically distributed.

From the above assumptions it follows that the random variables X_j are independent with

$$\text{Var}(X_j) = \sigma_{param}^2 + \frac{\sigma_{fluct}^2}{\nu_j} \quad (26)$$

$$\simeq \sigma_{param}^2 + \frac{\tilde{\sigma}_{fluct}^2}{\tilde{P}_j}, \quad (27)$$

where

$$\begin{aligned} \nu_j &= \text{number of a priori expected claims,} \\ \tilde{P}_j &= E[C_j^{CY,n}], \\ \sigma_{fluct}^2 &= CoVa^2(Y^{(v)}) + 1, \end{aligned}$$

and with

$$\begin{aligned} E[X|\Theta] &= \Theta_1\Theta_2, \\ \text{Var}(X|\Theta) &= \frac{1}{w\lambda} \cdot \Theta_1\Theta_2^2 (CoVa^2(Y) + 1). \end{aligned}$$

Hence the observations $X_j, j = 1, 2, \dots, J$, fulfil the conditions of the Bühlmann-Straub model (see for instance [1], Chapter 4). However, we do not have the "standard situation" of the Bühlmann Straub model, where we have a collective of risks $i = 1, 2, \dots, I$, and where for each of these risks observations $X_{ij}, j = 1, 2, \dots, n$, over several years are available. Here the risks are the different years $j = 1, 2, \dots, J$ and for each of these "risks", one has only one observation X_j . In the standard situation of the Bühlmann-Straub model, the observations over several years for each risk are used to estimate the within risk variance component. But here the within risk variance is σ_{fluct}^2 , which can be estimated otherwise as described in Section B.1.

Therefore we can use the standard estimators in the Bühlmann-Straub model (see e.g. [1], Chapter 4.8) and obtain

$$\hat{\sigma}_{param}^2 = c \cdot \left\{ \frac{J}{J-1} \sum_{j=1}^J \frac{w_j}{w_{\bullet}} (X_j - \bar{X})^2 - \frac{J \hat{\sigma}_{fluct}^2}{n_{\bullet}} \right\}, \quad (28)$$

where

$$\begin{aligned} X_j &= \frac{C_j^{CY,n}}{\tilde{P}_j}, \\ C_j^{CY,n} &= \text{normal claim amount CY of year } j, \\ w_j &= \tilde{P}_j = \text{pure risk premium in year } j, \end{aligned}$$

$$c = \frac{J-1}{J} \left\{ \sum_{j=1}^J \frac{w_j}{w_{\bullet}} \left(1 - \frac{w_j}{w_{\bullet}} \right) \right\}^{-1},$$

$$\widehat{\sigma}_{fluct}^2 = \widehat{CoVa}^2(Y^{(v)}) \text{ estimator of the coefficient of}$$

variation of the claim sizes,

$$n_{\bullet} = \text{observed number of claims.}$$

Whereas the total claim amounts $C_j^{CY,n}$ are known figures and recorded in the files of a company, this is not the case for the corresponding pure risk premiums $\tilde{P}_j = E \left[C_j^{CY,n} \right]$. Hence, before being able to apply (28) one has to determine \tilde{P}_j , which are used for calculating the X_j .

Denote by

$$LR_j = \frac{C_j^{CY,n}}{P_j}$$

the observed claims ratio for normal claims in year j , where P_j is the earned premium. Under the assumption that

$$E[LR_j] = \mu_{LR} \tag{29}$$

is the same for all years, it is suggested to use

$$\tilde{P}_j = \overline{LR} \cdot P_j$$

where

$$\overline{LR} = \sum_{j=1}^J \frac{P_j}{P_{i_{\bullet}}} LR_j.$$

Often (29) is not fulfilled because of things like business cycles and premium policies. Then the premiums have first to be adjusted by such effects, such that the loss ratios calculated with the adjusted "on level" premiums fulfils (29).

Finally, we have to decide which weights w_j should be used in (28). We suggest to take the \tilde{P}_j as weights, which is justified by the variance property (27).

The estimator (28) is based on the properties and the variance structure of X_j . It does, however not make use of the fact that according to (5)

$$\sigma_{param}^2 \simeq \text{Var}(\Theta_1) + \text{Var}(\Theta_2). \tag{30}$$

An alternative approach to (28) is therefore to estimate $\text{Var}(\Theta_1)$ and $\text{Var}(\Theta_2)$ directly, which then involves the claim frequencies and the claim averages. Hence the following estimators are recommended for lines of business where there is a "natural" volume measure, which is necessary that the claim frequencies are comparable over time. This is the case for personal lines like motor liability, household insurance etc., but not for corporate property, corporate liability or industrial fire insurance.

Let

$$F_j = \frac{N_j}{\nu_j},$$

where $N_j =$ number of claims in year j ,

$\nu_j =$ a priori expected number of claims in year j .

Note that F_j is a "standardised" frequency with $E[F_j] = 1$. If there are no trends in the non-standardised claim frequency, that is if

$$E\left[\frac{N_j}{JR_j}\right] = \lambda \text{ for all } j, \quad (31)$$

then one can put

$$\begin{aligned} \nu_j &= \hat{\lambda} \cdot JR_j, \\ \text{where } JR_j &= \text{number of risks in year } j, \\ \hat{\lambda} &= \frac{N_{\bullet}}{JR_{\bullet}}. \end{aligned}$$

If there are trends, then use $\nu_j = \hat{\lambda}_j \cdot JR_j$, where $\hat{\lambda}_j$ is the trend adjusted estimate of the a priori claim frequency in year j .

Under model-assumptions A.1, where N_j is assumed to be conditionally Poisson, it holds that

$$\text{Var}(F_j) = \text{Var}(\Theta_1) + \frac{\lambda}{\nu_j}$$

and one can show that

$$\widehat{\sigma_{\Theta_1}^2} = \left(c \cdot \frac{\nu_{\bullet}}{J}\right)^{-1} \left(\frac{V_F}{\bar{F}} - 1\right) \quad (32)$$

where

$$\begin{aligned} V_F &= \frac{1}{J-1} \sum_{j=1}^J \nu_j (F_j - \bar{F})^2, \\ \bar{F} &= \sum_{j=1}^J \frac{\nu_j}{\nu_{\bullet}} F_j, \\ c &= \sum_{j=1}^J \frac{\nu_j}{\nu_{\bullet}} \left(1 - \frac{\nu_j}{\nu_{\bullet}}\right). \end{aligned}$$

is an unbiased estimator of $\text{Var}(\Theta_1)$.

For estimating $\text{Var}(\Theta_2)$ one has to look at the observed claims averages

$$\bar{Y}_j = \frac{1}{N_j} \sum_{v=1}^{N_j} Y_j^{(\nu)}$$

in different years. Because of inflation or possible other trends, one has first to adjust the claim sizes in a given year j by these factors to bring them on the same level, which can be done for instance by linear regression. After these adjustments, the claim sizes and claim averages fulfil the Bühlman-Straub credibility model for claim sizes as presented in Chapter 4.11 of [1] and one can use this theory to estimate $\text{Var}(\Theta_2)$.

B.3 Estimation of the Pareto Parameters for Big Claims

We consider again a specific lob i and drop the index i to simplify notation. Assume that Y_1, Y_2, \dots are the big claims above a certain threshold c and we assume that they are Pareto-distributed with Pareto-parameter ϑ .

Assume you have observed n such claims. It is well known that

$$\widehat{\vartheta} = \left(\frac{1}{n-1} \sum_{\nu=1}^n \ln \left(\frac{Y_{\nu}^b}{c} \right) \right)^{-1} \quad (33)$$

is an unbiased estimator of ϑ with

$$\begin{aligned} E \left[\widehat{\vartheta} \right] &= \vartheta, \\ CoVa \left(\widehat{\vartheta} \right) &= \frac{1}{\sqrt{n-2}}. \end{aligned} \quad (34)$$

The number of observed big claims of an individual company in the considered lob might be rather small. From (34) it is seen that the uncertainty of this estimate then becomes fairly big. Hence, it is desirable to take also into account in some way the standard value of the SST, which can be considered as an estimate gained from industry-wide data.

Indeed, the estimator $\widehat{\vartheta}$ fulfils the Bühlmann-Straub credibility model and we can use credibility techniques to combine the individual experience with the a priori estimate given by the standard value of the SST. The credibility estimator is given by (see [11] or [1], Chapter 4.14)

$$\widehat{\vartheta}^{cred} = \alpha \widehat{\vartheta} + (1 - \alpha) \vartheta_0 \quad (35)$$

where $\widehat{\vartheta}$ is as in (33) and where

$$\begin{aligned} \alpha &= \frac{n-2}{n-1+\kappa}, \\ \vartheta_0 &= \text{standard value from the SST}, \\ \kappa &= CoVa(\Theta)^{-2}. \end{aligned}$$

Here $CoVa(\Theta)$ denotes the coefficient of variation of the Pareto parameter within the different companies. It could be estimated for instance by the supervision authority from the data of the different companies, or one can assess it in a pure Bayesian way, let's say to assume that it is 25%, which would result in a value of 16 for κ . For instance if your individual estimator is based on 10 observed big claims, then you would give a credibility weight of 32% to your own Pareto-estimate and if your estimator is based on 30 observed big claims, then the credibility weight given to your own estimate would be 62%.

B.4 Estimation of the Reserve Risk

We believe that the reserve risk should be determined based on the techniques used for estimating these reserves. The most known formula for estimating the variance or mean squared error of the reserves is the famous formula of Mack for the chain ladder reserving

method (see [9]). However, Mack's formula measures the ultimate reserve risk whereas for solvency purposes we need the one-year reserve risk.

The one-year reserve risk is best understood when we look at the recursive formula for the chain ladder reserving method. This has been done in [2]. The formula derived there coincides with the result found in [10].

In the following, we briefly summarize these results.

Assume that at time I , there is given a development triangle or trapezoid

$$\mathcal{D}_I = \{C_{i,j} : 0 \leq i \leq I, 0 \leq j \leq J, i + j \leq I\},$$

where $C_{i,j}$ denote cumulative claims payments or cumulative incurred claims of accident year i in development year j and where $C_{i,J}$ is the ultimate claim. Let

$$\begin{aligned}\widehat{C}_{i,J} &= C_{i,I-i} \prod_{j=I-i}^{J-1} \widehat{f}_j, \\ \widehat{R}_i &= \widehat{C}_{i,J} - C_{i,I-i}^{paid}, \\ \widehat{\sigma}_j^2 &= \frac{1}{I-j-1} \sum_{i=0}^{I-j-1} C_{i,j} \left(F_{i,j} - \widehat{f}_j\right)^2.\end{aligned}\tag{36}$$

$$\begin{aligned}F_{ij} &= \frac{C_{i,j+1}}{C_{i,j}}, \\ S_j^{[k]} &= \sum_{i=0}^k C_{i,j}, \\ \widehat{f}_j &= \frac{S_{j+1}^{[I-j-1]}}{S_j^{[I-j-1]}}.\end{aligned}\tag{37}$$

Note that the \widehat{f}_j are the well known chain ladder factors, $\widehat{C}_{i,J}$ is the chain ladder forecast for the ultimate claim and \widehat{R}_i is the chain ladder reserve for accident year i .

Then in [5] and [10] the following result has been derived for the one-year reserve risk.

Theorem B.1 (one-year mse) *The one-year mean square error can be estimated as follows:*

i) single accident year i

$$mse\left(\widehat{R}_i\right) = C_{i,I-i}\Gamma_{I-i} + C_{i,I-i}^2\Phi_{I-i}\tag{38}$$

where

$$\begin{aligned}\Gamma_{I-i} &= \widehat{\sigma}_{I-i}^2 \left(1 + \frac{C_{i,I-i}}{S_{I-i}^{[i]}}\right) \prod_{j=I-i+1}^{J-1} \widehat{f}_j, \\ \Phi_{I-i} &= \frac{\widehat{\sigma}_{I-i}^2}{S_{I-i}^{[i]}} \prod_{j=I-i+1}^{J-1} \widehat{f}_j^2,\end{aligned}$$

ii) all accident years

$$mse\left(\widehat{R}_{\bullet}\right) = \sum_{i=I-J+1}^I mse\left(\widehat{R}_i\right) + 2 \sum_{I-J+1 \leq i < k \leq I}^I C_{i,I-i} \widehat{C}_{k,I-i} (\Gamma_{I-i} + \Phi_{I-i}), \quad (39)$$

where $\widehat{C}_{k,I-i}$ is the chain ladder forecast of $C_{k,I-i}$.

Remarks:

- Since the reserves \widehat{R}_i are proportional to $C_{i,I-i}$, model assumptions A.1 (or (11), respectively) are consistent with model assumptions A.2 (or (38) respectively). In (38) the term $C_{i,I-i}\Gamma_{I-i}$ corresponds to the random fluctuation risk and the term $C_{i,I-i}^2\Phi_{I-i}$ to the parameter risk.
- We believe that the parameter risk for the chain ladder reserving method can be treated in a rigorous mathematical way only in a Bayesian set-up (see [2], [?], [5]). With the Bayesian set-up and taking a non-informative prior the resulting estimates are slightly different from Mack's formula for the ultimate risk or the above formula for the one-year risk. Mack's formula and the above formula are then obtained by a first order Taylor approximation from the corresponding results with the Bayesian approach.

When looking at the reserve risk from a solvency point of view, we are interested in the 100- or 200-year adverse event. What are such events that first come into our mind? There are things like a high inflation or a change in legislation like a decrease of the technical interest rate to calculate the lump sums in motor liability insurance, hence situations with a high adverse calendar year effect. But such events are usually not observed in the triangles and not captured by chain-ladder or similar reserving methods. Hence they are also not reflected by the above estimator of the reserve risk. Thus the question arises whether we do the right thing when looking at the reserve risk from a solvency point of view. The answer is that the above method is adequate for the reserve risk except in extraordinary situations with a huge adverse calendar year effect. Hence, it seems absolutely necessary to complement the reserve risk calculation with a scenario of a hyper-inflation or another adverse situation which might happen. Indeed, nobody knows whether the actual finance crisis will not lead to high inflation in some years.

The following example is taken from [6]. First you find below a development triangle in private-liability of a major Swiss insurance company. For confidentiality reasons the figures are multiplied by a constant factor. The figures above the diagonal are the observed cumulative payments, whereas the figures below the diagonal show the chain ladder forecasts.

development triangle cumulative payments

acc. year	dev. year																				Ultimate		
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19			
1979	3670	5817	6462	6671	6775	7002	7034	7096	7204	7326	7395	7438	7468	7541	7580	7677	7794	7919	8047	8051	8152	8152	
1980	4827	7600	8274	8412	9059	9197	9230	9277	9331	9435	9864	9621	9654	9704	9743	9745	9745	9748	9749	9749	9749	9749	
1981	7130	10852	11422	12024	12320	12346	12379	12426	12460	12500	12508	12547	12557	12624	12639	12644	12647	12666	12711	12713	13063	13063	
1982	9244	13340	13758	13853	13894	13942	13980	14001	14012	14070	14085	14098	14111	14118	14120	14133	14141	14145	14166	14167	14167	14167	
1983	10019	14223	15403	15579	15732	15921	16187	16420	16531	16559	16568	16585	16597	16614	16617	16619	16681	16681	16687	16711	16748	16748	
1984	9966	14599	15181	15431	15506	15538	15906	16014	16537	16833	16951	17038	17040	17195	17298	17307	17308	17308	17309	17310	17311	17311	
1985	10441	15043	15577	15784	15926	16054	16087	16107	16311	16366	16396	16414	16419	16426	16480	16480	16480	16480	16482	16482	16482	16482	
1986	10578	15657	16352	16714	17048	17289	17632	17646	17662	17678	17693	17700	17706	17706	17712	17713	17718	17719	17719	17719	17719	17719	
1987	11214	16482	17197	17518	18345	18480	18505	18520	18578	18633	18671	18689	18746	18772	18774	18804	18804	18804	18806	18806	18809	18809	
1988	11442	17621	18465	18693	18882	18965	20464	20522	20558	20607	20611	21250	21257	22505	22509	22516	22518	22522	22523	22527	22611	22611	
1989	11720	17779	18655	18940	19098	19368	19970	20162	20195	20415	20510	20594	20657	20752	20823	20899	20983	21124	21622	21627	21708	21708	
1990	13293	20689	21696	22439	22798	23054	23394	23554	23888	24061	24096	24301	24366	24389	24391	24784	24784	24784	24794	24894	24900	24993	24993
1991	15063	22917	23543	24032	24156	24232	24360	24410	24884	24896	24968	25031	25172	25459	25460	25470	25734	25774	25877	25883	25980	25980	25980
1992	16986	23958	25090	25392	25546	26099	26129	26149	26166	26231	26328	26743	27023	27048	27075	27168	27234	27276	27386	27392	27494	27494	27494
1993	16681	24867	25871	26463	26941	27120	27164	27183	27250	27490	27497	27561	27565	27582	27706	27787	27855	27898	28010	28017	28121	28121	28121
1994	17595	25152	26140	27090	27568	27637	27805	28003	28223	28240	28245	28250	28257	28297	28347	28430	28499	28543	28658	28665	28772	28772	28772
1995	16547	25396	26506	27043	27866	27882	27903	27933	28492	28545	28564	28686	28695	28897	28948	29033	29104	29149	29266	29273	29382	29382	29382
1996	15449	22702	23909	24690	26083	26525	26567	26640	26695	26801	26814	27018	27078	27269	27317	27397	27463	27506	27616	27623	27726	27726	27726
1997	18043	26918	28256	28569	28964	29268	29344	29393	30159	30966	31122	31190	31410	31466	31558	31635	31684	31811	31819	31937	31937	31937	31937
1998	17655	26241	27369	28063	28346	28780	29024	29180	29250	29575	29660	29809	29875	30085	30138	30227	30300	30347	30469	30476	30590	30590	30590
1999	16789	25547	27099	27801	27919	28052	30020	30035	30456	30663	30750	30905	30973	31192	31247	31338	31415	31463	31590	31597	31715	31715	31715
2000	15538	23830	25202	26462	27056	27480	27564	27612	27873	28062	28143	28284	28347	28546	28597	28681	28751	28795	28911	28917	29025	29025	29025
2001	15113	23405	26822	27711	28080	29203	29647	29751	30032	30236	30323	30475	30542	30758	30812	30903	30978	31025	31150	31157	31274	31274	31274
2002	14543	22674	23603	24159	24242	24425	24767	24854	25089	25260	25332	25459	25515	25695	25741	25816	25879	25919	26023	26029	26126	26126	26126
2003	14590	22337	23442	24031	24665	24942	25292	25380	25621	25794	25868	25998	26056	26239	26286	26363	26427	26468	26574	26580	26680	26680	26680
2004	13976	21527	22615	23242	23653	23918	24254	24339	24569	24736	24806	24931	24986	25162	25207	25281	25342	25381	25483	25489	25585	25585	25585
2005	12932	20118	21309	21824	22209	22459	22774	22853	23070	23226	23293	23410	23462	23627	23669	23738	23796	23832	23928	23934	24023	24023	24023
2006	12538	20357	21435	21953	22341	22592	22909	22989	23206	23364	23431	23548	23600	23767	23809	23879	23937	23974	24070	24076	24166	24166	24166
2007	12888	19413	20441	20935	21304	21544	21846	21922	22130	22280	22344	22456	22506	22664	22704	22771	22826	22861	22953	22959	23045	23045	23045

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The next table shows the resulting chain ladder reserves and the corresponding standard-deviation for each accident year and for the total. The ultimate reserve risk calculated with Mack's formula amounts to 13.3%, whereas the one year reserve risk according to formula (39) reduces to 6.6%.

acc. year	Cl- Reserve	ultimate (Mack)		one year	
		CoVa	sqrt mse	Vko_CDRi	sqrt mse
1979-1987	-	-	-	-	-
1988	84	206.4%	174	206.4%	174
1989	86	198.0%	170	30.3%	26
1990	199	129.2%	257	91.0%	181
1991	246	111.7%	275	36.7%	90
1992	326	92.1%	301	32.7%	107
1993	415	78.9%	328	31.0%	129
1994	475	70.8%	336	14.4%	68
1995	687	73.6%	506	54.9%	377
1996	708	70.0%	495	12.9%	91
1997	971	60.6%	589	26.0%	252
1998	1'015	59.7%	605	19.7%	200
1999	1'259	51.4%	647	16.3%	205
2000	1'413	46.8%	662	17.6%	249
2001	1'627	42.8%	696	7.1%	116
2002	1'701	48.4%	823	31.3%	532
2003	2'015	43.0%	867	12.4%	249
2004	2'343	41.2%	964	16.3%	381
2005	2'714	34.4%	933	9.7%	262
2006	3'809	28.6%	1'088	14.7%	558
2007	10'157	13.1%	1'328	7.9%	801
Total	32'250	13.3%	4'294	6.6%	2'131

Finally, we should consider an adverse extraordinary scenario and also take it into account when considering the reserve risk. If we assume that claims inflation increases by

3 %-points (additional inflation to the one already existent in the development triangle) and stays on this level for 10 years, then this would create a reserve loss of 4'401 or 13.6% of the reserves. Inflation scenarios would also affect the reserves of other lob, and its impact should be taken into account as a reserve scenario for the total business of a company.

The observations in the development triangles might vary quite a lot due to random fluctuations, in particular for small and medium sized companies. It would therefore be helpful to know an industry wide payment pattern and a technique how to combine the individual pattern with the industry wide pattern to obtain an optimal estimate. Here we just want to mention that such a technique was presented in [5]. The main idea is to use credibility to obtain credibility estimates of the chain ladder factors, which are then a weighted mean between the industry wide factors and the factors obtained from the company's own triangle.

B.5 On the Estimation of the Correlation Matrices

We believe that we should first think about the reasons why risks of different lob or the premium and reserve-risk of the same lob are correlated. The main reasons for such correlations are calendar year effects affecting the different risks simultaneously.

To understand the impact of diagonal factors let us have a closer look at the reserve risk. However, the same considerations could also be applied to the CY-risk.

Using the same notation as in Section A.2 we find

$$\widetilde{R}_i = Y_i R_i \tag{40}$$

with

$$\begin{aligned} E[Y_i] &= 1, \\ \text{Var}(Y_i) &= \tau_i^2. \end{aligned}$$

Assume that (40) reflects the situation before calendar year effect and that the reserves risk of different lob are independent, that is

$$\text{Cov}(Y_i, Y_j) = 0 \text{ for } i \neq j.$$

Add now a diagonal effect Δ independent of \widetilde{R}_i with $E[\Delta] = 1$ affecting different lob simultaneously (e.g. claims inflation) and denote by

$$\begin{aligned} \widetilde{R}_i^* &= Y_i R_i \Delta, \\ C_i^{PY*} &= \widetilde{R}_i^* - R_i, \end{aligned}$$

the a posteriori best estimate of the outstanding liabilities L_i , respectively the claim amount PY after calendar year effect. Then

$$\begin{aligned} \text{Var}(\widetilde{R}_i^*) &= R_i^2 (E[Y_i^2] E[\Delta^2] - 1) \\ &= R_i^2 \{ (E[Y_i^2] - 1) (1 + E[\Delta^2] - 1) + E[\Delta^2] - 1 \} \\ &= R_i^2 \{ \tau_i^2 + \sigma_\Delta^2 + \tau_i^2 \sigma_\Delta^2 \} \\ &\simeq R_i^2 \{ \tau_i^2 + \sigma_\Delta^2 \}, \end{aligned}$$

where in the last equation we have assumed that $\tau_i^2 \sigma_\Delta^2 \ll \tau_i^2 + \sigma_\Delta^2$ and where

$$\sigma_\Delta^2 = \text{Var}(\Delta).$$

Under the assumption that the calendar year effect is also effective for lob j we obtain

$$\begin{aligned} \text{Cov}(\widetilde{R}_i^*, \widetilde{R}_j^*) &= \text{Cov}(C_i^{PY*}, C_j^{PY*}) \\ &= R_i R_j \text{Var}(\Delta), \\ \text{Corr}(\widetilde{R}_i^*, \widetilde{R}_j^*) &= \text{Corr}(C_i^{PY*}, C_j^{PY*}) \\ &= \frac{\sigma_\Delta^2}{\sqrt{\tau_i^2 + \sigma_\Delta^2} \sqrt{\tau_j^2 + \sigma_\Delta^2}}. \end{aligned} \tag{41}$$

(41) is an intuitive and handy formula, which gives quite a good qualitative insight: correlation induced by a calendar year effect becomes smaller the smaller σ_Δ^2 is compared to τ_i^2 and τ_j^2 . As an example consider claims inflation. Assume that the yearly standard deviation for claims inflation is 1% and that the reserve risks before the calendar year effect "inflation" for lob i and j are 3%. Then we obtain from (41) that

$$\text{Corr}(C_i^{PY*}, C_j^{PY*}) = 11\%.$$

If we increase the calendar year inflation to 3%, the correlation would be 50%. This shows for instance, that the correlation induced by varying inflation is bigger in countries with high inflation than in countries with low inflation. The same considerations can also be applied to other calendar year effects and might help experts to assess the correlation matrices.

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