

A power law extrapolation – interpolation method for IBNR claims reserving

Werner Hürlimann

Swiss Mathematical Society, Fribourg, Switzerland

Email address:

whurlimann@bluewin.ch

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Abstract: To calculate claims reserves more frequently than the usual yearly periods for which ultimate loss development factors are available, it is necessary to perform an extrapolation prior to the time marking the end of the first development year and an interpolation for each successive development year. A simple power law extrapolation – interpolation method is developed and illustrated for monthly and quarterly sub-periods.

Keywords: Claims Reserving, IBNR Reserve, Loss Development Factors, Interpolation, Extrapolation, Power Law

1. Introduction

Claims reserves are usually the largest single item on an insurance company's balance sheet. Very often reserve fluctuations significantly affect the company's solvency requirements and overall financial position. Any mismatch of reserves has a direct impact on net asset values. Moreover, capital adequacy and reserving adequacy are essentially two sides of the same coin. An insurer whose claims reserves are more than adequate does not need to maintain as much capital as an insurer whose reserves are less than adequate. Setting claims reserves accurately is a gigantic task, especially for a complex multi-line insurer. Reserving has a great impact on virtually everything an insurance company does, from setting prices to establishing solvency margins. Therefore, with the introduction of Solvency II and the new accounting standards for insurance IFRS 4, reserving best practices are more and more important.

By nature, claims reserves are uncertain. Essentially, they are estimates of how much the company will have to pay out in the future on incurred claims, whether or not they have been reported. In simple terms, *claims reserves* consist of three key elements:

- *Case estimates* or *case reserves* are amounts for claims that have been notified to the company but have not yet been fully settled.
- *Incurred but not enough reported* (IBNER) are allowances for any inadequacies in case reserves.
- *Incurred but not reported* (IBNR) are estimated amounts

for claims that have not yet been notified to the company.

Companies seldom distinguish between IBNR and IBNER, instead combining them into a single item, called here simply *IBNR reserve*. In the following, *reported claims* means the sum of the actual *paid claims* and the *case reserves*.

The present note is organized as follows. Section 2 recalls how IBNR reserves are calculated using the standard Chain Ladder method. To report IBNR reserves more frequently than the usual yearly periods, it is necessary to perform an extrapolation prior to the end of the first year and an interpolation for each successive development year. A simple power law method is developed and illustrated for monthly and quarterly sub-periods in Section 3.

2. Calculation of IBNR Reserves

In practice, the calculation of IBNR reserves involves an actuary, either at the initial stage or as part of the audit process. IBNR claims reserving can be described as “squaring the triangle”, that is making use of historic information on the development of paid or reported claims to make estimates about their future development (e.g. Boulter and Grubbs [1], Subotzky and Mazur [23]). For example, at the end of 2014, a company that has been writing a certain class of business since 2005 has 10 annual development points for claims on its 2005 book of business, nine development points for 2006 and one for 2014. A loss triangle can be created with either the reported claims or the paid claims in form of a partially completed table. The rows represent the accident years in which claims incurred and the columns represent the development periods.

Table 1 below is an example of loss triangle. This triangle will form the upper left part of a square (hence the expression, squaring the triangle) and the information in the triangle can be used to fill in the lower right part of the square, which, together with assumptions about the length of the development tail (accident years going beyond 2003), will give an estimate of the ultimate incurred claims. The difference between the estimated ultimate claims and the claims paid to date is the claims reserve, and the difference between the claims reserve and the outstanding case reserves for reported claims is the IBNR reserve.

The topic of claims reserving is well established within actuarial mathematics. Among recent work, one finds a handbook by Radtke and Schmidt [14], an extensive bibliography by Schmidt [21], and Ph.D. theses by Salzmann [17] and Happ [5].

The most commonly used IBNR reserving techniques are the *Chain Ladder* and the *Bornhuetter-Ferguson* methods or an optimal combination of them called *Credible IBNR* method (e.g. Mack [11], Hürlimann [9], Gigante et al. [4]). The methods are deterministic in that they give a point estimate of ultimate claims rather than a range of estimates. Other reserving methods, such as *Bootstrapping* or the *Gamma IBNR* method in Hürlimann [8], are stochastic in that they use runoff triangles to arrive at a distribution of the ultimate claims (see also Wüthrich and Merz [24], Huang and Wu [6]). The statistical estimation of loss development factors in Table

2 is based on the data of Table 1 and uses for simplicity the Chain Ladder method.

In the Chain Ladder method, historical data is examined to estimate loss development factors (LDF) or ratios for each development period. The factors are cumulated and applied to the latest observed numbers (here paid claims) to estimate the ultimate incurred claims. The underlying assumption is that for each year of exposure, a certain percentage of the ultimate claims will have emerged at the end of each development year, and these percentages are consistent across years. So, for example, in Table 1, we can estimate the likely development of 2003 after five years by reference to the actual development of 1994 at 1999, 1995 at 2000, and so on.

For the mathematical specification, consider now a given accident year of a line of business over a development period $(0, T]$ in units of years. The ultimate LDF of the yearly exposure period $(t-1, t]$, $t=1, 2, \dots, T$, of the considered accident year is denoted by F_t (blue line in Table 2 with $T=10$). The further notations are as follows:

S_t : aggregate paid claims for the period $(0, t]$

OS_t : outstanding case reserves for the period $(t, T]$

$IBNR_t$: IBNR reserve for the period $(t, T]$

By definition one has the identity:

$$IBNR_t = (F_t - 1) \cdot S_t - OS_t. \quad (1)$$

Table 1. Loss Triangle of Paid Claims ("A.M. Best" 2004 table for Private Passenger Auto Liability).

Accident Year	Development Period in Months									
	12	24	36	48	60	72	84	96	108	120
1994	16'883'850	31'182'837	37'401'111	40'812'822	42'565'347	43'422'022	43'832'148	44'029'002	44'120'908	44'172'759
1995	17'518'883	31'787'614	38'274'471	41'833'477	43'692'705	44'581'536	45'021'089	45'233'182	45'338'083	
1996	18'137'677	32'509'210	39'097'072	42'817'313	44'826'451	45'792'256	46'264'482	46'470'822		
1997	18'449'658	32'776'770	39'487'465	43'255'912	45'264'843	46'189'365	46'511'626			
1998	18'710'148	33'568'205	40'461'509	44'316'727	46'334'427	47'208'966				
1999	20'553'769	36'347'062	43'531'162	47'472'983	49'515'412					
2000	22'247'399	39'116'657	46'564'786	50'712'030						
2001	23'082'370	40'371'884	48'011'274							
2002	24'245'392	42'085'537								
2003	24'146'487									

Table 2. Loss Development Factors according to the Chain Ladder method.

Period in Months	12-24	24-36	36-48	48-60	60-72	72-84	84-96	96-108	108-120	120-ult.
Chain-ladder factors	1.77805	1.19869	1.09270	1.04487	1.02025	1.00914	1.00455	1.00220	1.00118	1.00000
Ultimate LDF	2.52532	1.42027	1.18485	1.08433	1.03776	1.01716	1.00795	1.00338	1.00118	1.00000
Percent Unpaid Claims	60.40%	29.59%	15.60%	7.78%	3.64%	1.69%	0.79%	0.34%	0.12%	0.00%
Percent Paid Claims	39.60%	30.81%	13.99%	7.82%	4.14%	1.95%	0.90%	0.45%	0.22%	0.12%

3. Extrapolation – Interpolation of Ultimate LDF Patterns

To report IBNR reserves more frequently than the usual yearly periods for which ultimate LDF patterns are available, it is necessary to perform an extrapolation prior to time $t=1$ marking the end of the first year and an interpolation for each successive development year between time $t-1$ and time t . Different and more complex methods of extrapolation –

interpolation have been developed earlier in Sherman [22] and Robbin and Homer [16].

For simplicity, let us focus on monthly and quarterly sub-periods, but the method is valid for sub-periods of arbitrary lengths. We assume that the revealed paid claims in each sub-period of a development year behave proportionally to a power law depending on the elapsed number of sub-periods as follows. Let the amounts of claims paid in the k -th sub-period of the development year $(t-1, t]$ equal $k^\alpha \cdot c_{t-1}^{(\alpha, m)}$ respectively $k^\alpha \cdot c_{t-1}^{(\alpha, q)}$, where $c_{t-1}^{(\alpha, m)}$ and $c_{t-1}^{(\alpha, q)}$

denote appropriate increment constants for the ultimate monthly respectively quarterly LDF patterns and $\alpha \in [0, 1]$. The extreme case $\alpha = 0$ refers to constant revealed paid claims in each sub-period and the extreme case $\alpha = 1$ to a linear increase in the elapsed number of sub-periods. The Figures 1 and 2 yield a picture of this power law method. On the horizontal axis one finds the elapsed time and on the vertical axis the percentage of paid claims in a given development year.

The percentage of paid claims within a development year is highest (smallest) for $\alpha = 0$ ($\alpha = 1$). Other choices of the power law exponent $\alpha \in [0, 1]$ lie between these extremes.

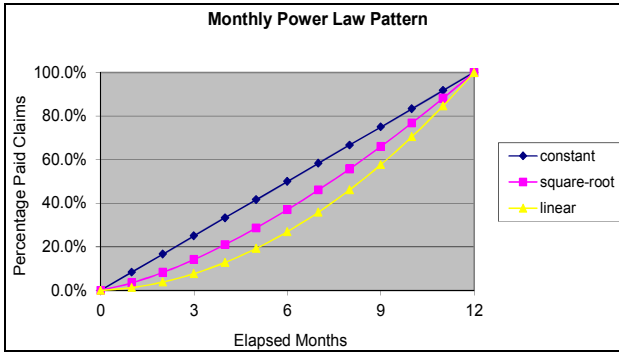


Figure 1. Monthly power-law extrapolation - interpolation pattern.

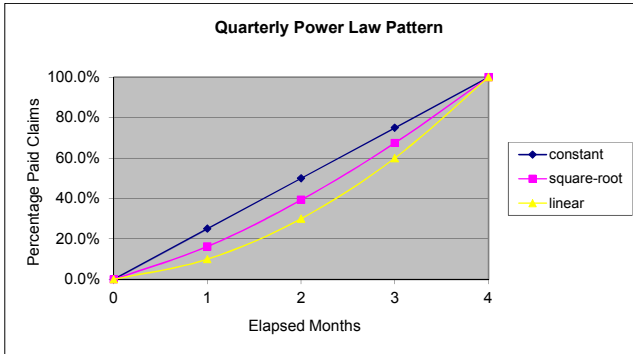


Figure 2. Quarterly power-law extrapolation - interpolation pattern.

The obtained ultimate monthly and quarterly LDF patterns after extrapolation and interpolation are elements of matrices denoted by

$F_{t-1,k}^{(\alpha,m)}$: ultimate monthly LDF pattern for the k -th month of the development period $(t-1, t]$, $t = 1, 2, \dots, T$, $k = 1, 2, \dots, 12$

$F_{t-1,k}^{(\alpha,q)}$: ultimate quarterly LDF pattern for the k -th quarter of the development period $(t-1, t]$, $t = 1, 2, \dots, T$, $k = 1, 2, 3, 4$

To describe the obtained LDF patterns we will need the following quantities:

$U(t-1)$: proportion of unpaid claims at time $t-1$ for the development period $(t-1, t]$, $t = 1, 2, \dots, T$

By definition of the ultimate yearly LDF pattern one has

$$U(0) = 1, \quad U(t-1) = 1 - \frac{1}{F_t}, \quad t = 2, \dots, T. \quad (2)$$

A mathematical analysis yields the following formulas

$$c_{t-1}^{(\alpha,m)} = \frac{U(t-1) - U(t)}{\sum_{k=1}^{12} k^\alpha}, \quad t = 1, 2, \dots, T \quad (3)$$

$$F_{t-1,1}^{(\alpha,m)} = \frac{1}{c_{t-1}^{(\alpha,m)} + 1 - U(t-1)}, \quad (4)$$

$$F_{t-1,k}^{(\alpha,m)} = \frac{1}{k^\alpha \cdot c_{t-1}^{(\alpha,m)} + 1/F_{t-1,k-1}^{(\alpha,m)}}, \quad k = 2, \dots, 12$$

$$c_{t-1}^{(\alpha,q)} = \frac{U(t-1) - U(t)}{\sum_{k=1}^4 k^\alpha}, \quad t = 1, 2, \dots, T \quad (5)$$

$$F_{t-1,1}^{(\alpha,q)} = \frac{1}{c_{t-1}^{(\alpha,q)} + 1 - U(t-1)}, \quad (6)$$

$$F_{t-1,k}^{(\alpha,q)} = \frac{1}{k^\alpha \cdot c_{t-1}^{(\alpha,q)} + 1/F_{t-1,k-1}^{(\alpha,q)}}, \quad k = 2, 3, 4$$

A verification shows that at the extrapolating respectively interpolating times the formulas are consistent with the given ultimate yearly LDF pattern such that

$$F_{t-1,12}^{(\alpha,m)} = F_{t-1,12}^{(\alpha,q)} = F_t, \quad t = 1, 2, \dots, T. \quad (7)$$

In practice one is also interested in the following quantities, where the symbol \bullet stands for monthly (m) or quarterly (q):

$U_{t-1,k}^{(\alpha,\bullet)}$: proportion of unpaid claims at the end of the k -th sub-period of the development year $(t-1, t]$

$P_{t-1,k}^{(\alpha,\bullet)}$: proportion of paid claims during the k -th sub-period of the development year $(t-1, t]$

$AP_{t-1,k}^{(\alpha,\bullet)}$: proportion of aggregate paid claims at the end of the k -th sub-period of the development year $(t-1, t]$

These quantities are obtained using the following formulas:

$$U_{t-1,k}^{(\alpha,\bullet)} = 1 - \frac{1}{F_{t-1,k}^{(\alpha,\bullet)}}, \quad (8)$$

$$U_{0,0}^{(\alpha,\bullet)} = U(0) = 1, \quad U_{t,0}^{(\alpha,\bullet)} = U(t-1)$$

$$P_{t-1,k}^{(\alpha,\bullet)} = U_{t-1,k-1}^{(\alpha,\bullet)} - U_{t-1,k}^{(\alpha,\bullet)} \quad (9)$$

$$AP_{t-1,k}^{(\alpha,\bullet)} = AP_{t-1,k-1}^{(\alpha,\bullet)} + P_{t-1,k}^{(\alpha,\bullet)}, \quad (10)$$

$$AP_{0,0}^{(\alpha,\bullet)} = 0, \quad AP_{t,0}^{(\alpha,\bullet)} = AP_{t-1,12}^{(\alpha,\bullet)}$$

To illustrate, we have calculated the ultimate monthly and quarterly LDF patterns for the given ultimate yearly LDF pattern of Table 2 according to the above power law method

The percentages of unpaid claims (Tables 9 to 14) are less

Table 3. Ultimate LDF Matrix by Year and Month (linear case).

Year	Month												Yearly Pattern	Increment Constants
	1	2	3	4	5	6	7	8	9	10	11	12		
0	196.975	65.658	32.829	19.697	13.132	9.380	7.035	5.472	4.377	3.581	2.984	2.525	2.525	0.508%
1	2.500	2.452	2.383	2.296	2.197	2.088	1.974	1.858	1.743	1.631	1.523	1.420	1.420	0.395%
2	1.417	1.409	1.399	1.385	1.368	1.348	1.326	1.301	1.274	1.246	1.216	1.185	1.185	0.179%
3	1.183	1.181	1.176	1.171	1.164	1.156	1.147	1.136	1.125	1.112	1.099	1.084	1.084	0.100%
4	1.084	1.082	1.081	1.078	1.075	1.071	1.067	1.062	1.057	1.051	1.045	1.038	1.038	0.053%
5	1.037	1.037	1.036	1.035	1.034	1.032	1.030	1.028	1.026	1.023	1.020	1.017	1.017	0.025%
6	1.017	1.017	1.016	1.016	1.015	1.015	1.014	1.013	1.012	1.011	1.009	1.008	1.008	0.012%
7	1.008	1.008	1.008	1.007	1.007	1.007	1.006	1.006	1.005	1.005	1.004	1.003	1.003	0.006%
8	1.003	1.003	1.003	1.003	1.003	1.003	1.003	1.002	1.002	1.002	1.002	1.001	1.001	0.003%
9	1.001	1.001	1.001	1.001	1.001	1.001	1.001	1.001	1.000	1.000	1.000	1.000	1.000	0.002%

Year	Quarter				Yearly Pattern	Increment Constants
	1	2	3	4		
0	25.253	8.418	4.209	2.525	2.525	3.960%
1	2.343	2.047	1.722	1.420	1.420	3.081%
2	1.393	1.340	1.269	1.185	1.185	1.399%
3	1.174	1.153	1.122	1.084	1.084	0.782%
4	1.079	1.070	1.056	1.038	1.038	0.414%
5	1.036	1.031	1.025	1.017	1.017	0.195%
6	1.016	1.014	1.012	1.008	1.008	0.090%
7	1.007	1.007	1.005	1.003	1.003	0.045%
8	1.003	1.003	1.002	1.001	1.001	0.022%
9	1.001	1.001	1.000	1.000	1.000	0.012%

[illegible]

Table 6. Ultimate LDF Matrix by Year and Quarter (constant case).

Year	Quarter				Yearly Pattern	Increment Constants
	1	2	3	4		
0	10.101	5.051	3.367	2.525	2.525	9.900%
1	2.114	1.818	1.595	1.420	1.420	7.703%
2	1.353	1.292	1.236	1.185	1.185	3.497%
3	1.158	1.132	1.108	1.084	1.084	1.956%
4	1.072	1.061	1.049	1.038	1.038	1.035%
5	1.033	1.027	1.022	1.017	1.017	0.488%
6	1.015	1.013	1.010	1.008	1.008	0.225%
7	1.007	1.006	1.005	1.003	1.003	0.113%
8	1.003	1.002	1.002	1.001	1.001	0.055%
9	1.001	1.001	1.000	1.000	1.000	0.029%

Table 7. Ultimate LDF Matrix by Year and Month (square root case).

Year	Month												Yearly Pattern	Increment Constants
	1	2	3	4	5	6	7	8	9	10	11	12		
0	73.863	30.595	17.814	12.018	8.812	6.819	5.480	4.530	3.826	3.287	2.865	2.525	2.525	1.354%
1	2.460	2.373	2.274	2.170	2.065	1.960	1.859	1.761	1.668	1.581	1.498	1.420	1.420	1.053%
2	1.411	1.397	1.381	1.363	1.344	1.323	1.301	1.279	1.256	1.232	1.209	1.185	1.185	0.478%
3	1.181	1.176	1.169	1.162	1.154	1.146	1.136	1.127	1.117	1.106	1.095	1.084	1.084	0.267%
4	1.083	1.080	1.077	1.074	1.071	1.067	1.062	1.058	1.053	1.048	1.043	1.038	1.038	0.141%
5	1.037	1.036	1.035	1.033	1.032	1.030	1.028	1.026	1.024	1.022	1.020	1.017	1.017	0.067%
6	1.017	1.016	1.016	1.015	1.015	1.014	1.013	1.012	1.011	1.010	1.009	1.008	1.008	0.031%
7	1.008	1.008	1.007	1.007	1.007	1.006	1.006	1.005	1.005	1.004	1.004	1.003	1.003	0.015%
8	1.003	1.003	1.003	1.003	1.003	1.003	1.002	1.002	1.002	1.002	1.001	1.001	1.001	0.008%
9	1.001	1.001	1.001	1.001	1.001	1.001	1.001	1.001	1.000	1.000	1.000	1.000	1.000	0.004%

Table 8. Ultimate LDF Matrix by Year and Quarter (square root case).

Year	Quarter				Yearly Pattern	Increment Constants
	1	2	3	4		
0	15.521	6.429	3.743	2.525	2.525	6.443%
1	2.242	1.934	1.656	1.420	1.420	5.013%
2	1.376	1.317	1.252	1.185	1.185	2.276%
3	1.167	1.143	1.115	1.084	1.084	1.273%
4	1.076	1.066	1.052	1.038	1.038	0.673%
5	1.034	1.030	1.024	1.017	1.017	0.318%
6	1.016	1.014	1.011	1.008	1.008	0.146%
7	1.007	1.006	1.005	1.003	1.003	0.074%
8	1.003	1.003	1.002	1.001	1.001	0.036%
9	1.001	1.001	1.000	1.000	1.000	0.019%

Table 9. Unpaid Claims Matrix by Year and Month (linear case).

Year	Month												Yearly Pattern
	1	2	3	4	5	6	7	8	9	10	11	12	
0	99.5%	98.5%	97.0%	94.9%	92.4%	89.3%	85.8%	81.7%	77.2%	72.1%	66.5%	60.4%	60.4%
1	60.0%	59.2%	58.0%	56.5%	54.5%	52.1%	49.3%	46.2%	42.6%	38.7%	34.3%	29.6%	29.6%
2	29.4%	29.1%	28.5%	27.8%	26.9%	25.8%	24.6%	23.1%	21.5%	19.7%	17.8%	15.6%	15.6%
3	15.5%	15.3%	15.0%	14.6%	14.1%	13.5%	12.8%	12.0%	11.1%	10.1%	9.0%	7.8%	7.8%
4	7.7%	7.6%	7.5%	7.2%	7.0%	6.7%	6.3%	5.9%	5.4%	4.9%	4.3%	3.6%	3.6%
5	3.6%	3.6%	3.5%	3.4%	3.3%	3.1%	2.9%	2.7%	2.5%	2.3%	2.0%	1.7%	1.7%
6	1.7%	1.7%	1.6%	1.6%	1.5%	1.4%	1.4%	1.3%	1.2%	1.1%	0.9%	0.8%	0.8%
7	0.8%	0.8%	0.8%	0.7%	0.7%	0.7%	0.6%	0.6%	0.5%	0.5%	0.4%	0.3%	0.3%
8	0.3%	0.3%	0.3%	0.3%	0.3%	0.3%	0.3%	0.2%	0.2%	0.2%	0.2%	0.1%	0.1%
9	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%

Table 10. Unpaid Claims Matrix by Year and Quarter (linear case).

Year	Quarter				Yearly Pattern
	1	2	3	4	
0	96.0%	88.1%	76.2%	60.4%	60.4%
1	57.3%	51.2%	41.9%	29.6%	29.6%
2	28.2%	25.4%	21.2%	15.6%	15.6%
3	14.8%	13.3%	10.9%	7.8%	7.8%
4	7.4%	6.5%	5.3%	3.6%	3.6%
5	3.4%	3.1%	2.5%	1.7%	1.7%
6	1.6%	1.4%	1.1%	0.8%	0.8%
7	0.7%	0.7%	0.5%	0.3%	0.3%
8	0.3%	0.3%	0.2%	0.1%	0.1%
9	0.1%	0.1%	0.0%	0.0%	0.0%

Table 11. Unpaid Claims Matrix by Year and Month (constant case).

Year	Month												Yearly Pattern
	1	2	3	4	5	6	7	8	9	10	11	12	
0	96.7%	93.4%	90.1%	86.8%	83.5%	80.2%	76.9%	73.6%	70.3%	67.0%	63.7%	60.4%	60.4%
1	57.8%	55.3%	52.7%	50.1%	47.6%	45.0%	42.4%	39.9%	37.3%	34.7%	32.2%	29.6%	29.6%
2	28.4%	27.3%	26.1%	24.9%	23.8%	22.6%	21.4%	20.3%	19.1%	17.9%	16.8%	15.6%	15.6%
3	14.9%	14.3%	13.6%	13.0%	12.3%	11.7%	11.0%	10.4%	9.7%	9.1%	8.4%	7.8%	7.8%
4	7.4%	7.1%	6.7%	6.4%	6.1%	5.7%	5.4%	5.0%	4.7%	4.3%	4.0%	3.6%	3.6%
5	3.5%	3.3%	3.2%	3.0%	2.8%	2.7%	2.5%	2.3%	2.2%	2.0%	1.8%	1.7%	1.7%
6	1.6%	1.5%	1.5%	1.4%	1.3%	1.2%	1.2%	1.1%	1.0%	0.9%	0.9%	0.8%	0.8%
7	0.8%	0.7%	0.7%	0.6%	0.6%	0.6%	0.5%	0.5%	0.5%	0.4%	0.4%	0.3%	0.3%
8	0.3%	0.3%	0.3%	0.3%	0.2%	0.2%	0.2%	0.2%	0.2%	0.2%	0.1%	0.1%	0.1%
9	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%

Table 12. Unpaid Claims Matrix by Year and Quarter (constant case).

Year	Quarter				Yearly Pattern
	1	2	3	4	
0	90.1%	80.2%	70.3%	60.4%	60.4%
1	52.7%	45.0%	37.3%	29.6%	29.6%
2	26.1%	22.6%	19.1%	15.6%	15.6%
3	13.6%	11.7%	9.7%	7.8%	7.8%
4	6.7%	5.7%	4.7%	3.6%	3.6%
5	3.2%	2.7%	2.2%	1.7%	1.7%
6	1.5%	1.2%	1.0%	0.8%	0.8%
7	0.7%	0.6%	0.5%	0.3%	0.3%
8	0.3%	0.2%	0.2%	0.1%	0.1%
9	0.1%	0.1%	0.0%	0.0%	0.0%

Table 13. Unpaid Claims Matrix by Year and Month (square root case).

Year	Month												Yearly Pattern
	1	2	3	4	5	6	7	8	9	10	11	12	
0	98.6%	96.7%	94.4%	91.7%	88.7%	85.3%	81.8%	77.9%	73.9%	69.6%	65.1%	60.4%	60.4%
1	59.3%	57.9%	56.0%	53.9%	51.6%	49.0%	46.2%	43.2%	40.1%	36.7%	33.2%	29.6%	29.6%
2	29.1%	28.4%	27.6%	26.7%	25.6%	24.4%	23.1%	21.8%	20.4%	18.8%	17.3%	15.6%	15.6%
3	15.3%	15.0%	14.5%	14.0%	13.4%	12.7%	12.0%	11.2%	10.4%	9.6%	8.7%	7.8%	7.8%
4	7.6%	7.4%	7.2%	6.9%	6.6%	6.2%	5.9%	5.5%	5.0%	4.6%	4.1%	3.6%	3.6%
5	3.6%	3.5%	3.4%	3.2%	3.1%	2.9%	2.7%	2.6%	2.4%	2.1%	1.9%	1.7%	1.7%
6	1.7%	1.6%	1.6%	1.5%	1.4%	1.4%	1.3%	1.2%	1.1%	1.0%	0.9%	0.8%	0.8%
7	0.8%	0.8%	0.7%	0.7%	0.7%	0.6%	0.6%	0.5%	0.5%	0.4%	0.4%	0.3%	0.3%
8	0.3%	0.3%	0.3%	0.3%	0.3%	0.3%	0.2%	0.2%	0.2%	0.2%	0.1%	0.1%	0.1%
9	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.1%	0.0%	0.0%	0.0%	0.0%	0.0%

Table 14. Unpaid Claims Matrix by Year and Quarter (square root case).

Year	Quarter				Yearly Pattern
	1	2	3	4	
0	93.6%	84.4%	73.3%	60.4%	60.4%
1	55.4%	48.3%	39.6%	29.6%	29.6%
2	27.3%	24.1%	20.2%	15.6%	15.6%
3	14.3%	12.5%	10.3%	7.8%	7.8%
4	7.1%	6.2%	5.0%	3.6%	3.6%
5	3.3%	2.9%	2.3%	1.7%	1.7%
6	1.5%	1.3%	1.1%	0.8%	0.8%
7	0.7%	0.6%	0.5%	0.3%	0.3%
8	0.3%	0.3%	0.2%	0.1%	0.1%
9	0.1%	0.1%	0.0%	0.0%	0.0%

Let us conclude with a brief account of some related claims reserving literature and possible future developments.

Usually, claims reserving models assume independence between different accident years. For this reason, they fail to model claims inflation appropriately, because claims inflation acts on all accident years simultaneously. A model that accounts for accident year dependence in runoff triangles has been proposed by Salzmänn and Wüthrich [18].

Predictions of claims reserves often rely on individual loss triangles, where each triangle corresponds to a different line of business. Since different lines of business are often dependent it is necessary to develop models for loss triangle dependence. Examples that use copulas are Regis [15] and de Jong [2].

To take into account solvency purposes (e.g. the Solvency II project) it is necessary to adapt the classical claims reserving models. Some typical developments include Merz and Wüthrich [12], Hürlimann [7], Savelli and Clemente [19], Pira et al. [13], Eling et al. [3], Salzmänn [17] and Happ [5].

Another direction concerns the development of claims reserving models based on multiple risk factors. Besides [7] and [20] we would like to point out [10], where the use of stochastic LDF's is advocated.

The integration of the presented simple extrapolation – interpolation method in these and other recent claims reserving techniques and the study of its impact might be a topic for future research.

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