Non-life insurance mathematics

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Non-life insurance from a financial perspective:

for a premium an insurance company commits itself to pay a sum if an event has occured



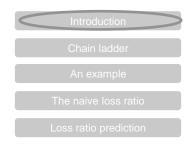
Issues that need to be solved:

- How much premium is earned?
- · How much premium is unearned?
- How do we measure the number and size of unknown claims?
- How do we know if the reserves on known claims are sufficient?

Premium reserves

The premium reserve is split in two parts:

- Provision for unearned premiums
- Provisions for unexpired risks



Earned and unearned premium:

- Written premium is earned evenly/uniformly over the cover period
- The share of the premium that has been earned is the past time's proportion of the total period
- If a larger premium has been received the difference is the unearned premium

Example:

An insurance policy starts on September 1 2012 and is valid until August 31 2013.

The premium for the entire period is 2400.

At 31 December we have received two quarterly premiums or 1200.

We have then earned (4/12)*2400 = 800.

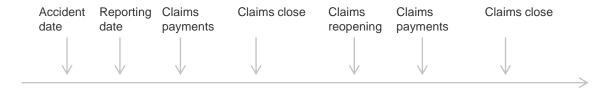
Unearned is 1200-800=400



Unexpired risk reserve

- Regard entire period covered by the insurance
- From a point in time, say 31/12-2012, we look forward to all the claims and expenses that could occur after this point. Call them FC₃₁₁₂
- If FC₃₁₁₂>Future premiums yet not due (FP)+unearned premium reserve (UP) the difference is accounted as unexpired risk reserve
- In example assume FC3112 = 1800>FP+UP=1200+400=1600, so unexpired risk reserve is 200

Claims reserves



Claims reserving issues:

- How do we measure the number and size of unknown claims?
 (IBNR reserve, i.e., Incurred Bot Not Reported)
- How do we know if the reserves on known claims are sufficient? (RBNS reserve, i.e., Reserved But Not Settled)

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Claims occuring:

- A claim event is an event that gives rise to a claim against an insurer by a policy holder
- Gross claim loss: the ultimate cost to the insurer of a claim event, including benefit payments and claims handling expenses
- Net claim loss: deduction of reinsurance recoveries
- Example: settlement delays are considered (RBNS). The process for estimating future reported claim amounts (IBNR) is similar.
- Step 1: Group the claims loss settlement amounts by the year in which the associated claims events occured

Claims	Claims losses
occurence year	settled
2005	3963
2006	4975
2007	5873
2008	6401
2009	6563
2010	6358
2011	6918
2012	3072

Claim payments plus claims handling expenses

The development of claims losses settled

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 Claims losses settled for each claims occurrence year are often not paid on one date but rather over a number of years

Incremental claims loss settlement data presented as a run-off triangle

Increment	al claims loss	Development year										
settlements		0	1	2	3	4	5	6	7			
	2005	1232	946	520	722	316	165	48	14			
e e	2006	1469	1201	708	845	461	235	56				
ence	2007	1652	1416	959	954	605	287					
cur	2008	1831	1634	1124	1087	725						
000	2009	2074	1919	1330	1240							
	2010	2434	2263	1661								
Claims year	2011	2810	4108									
S S	2012	3072										

Comments:

- The development year for a claims settlement amount reflects how long after the claims occurrence year the amount was settled.
 - An amount settled during the claims occurrence year was settled in development year 0
- In the example the largest development year for any claims occurrence years is 7
- The data shown represents the incremental claims losses settled in the development year
- For any cell in the table, the value shown represents the incremental claims loss amount that was settled in calendar year
- Each diagonal set of data represents the amounts settled in a single calendar year
- Green cells represent observed data all red represent time periods in the future for which we wish to estimate the expected claims settlements amounts

The development of claims losses settled

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 Claims losses settled for each claims occurrence year are not paid on one date but rather over a number of years

Incremental claims loss settlement data presented as a run-off triangle

Cumulativ	Cumulative claims loss				Developn	nent year			
settlements		0	1	2	3	4	5	6	7
	2005	1232	2178	2698	3420	3736	3901	3949	3963
e	2006	1469	2670	3378	4223	4684	4919	4975	
enc	2007	1652	3068	4027	4981	5586	5873		
ünc	2008	1831	3465	4589	5676	6401			
၁၁၀	2009	2074	3993	5323	6563				
SI (2010	2434	4697	6358					
Clain year	2011	2810	6918						
ye C	2012	3072							

Comments:

- The data can be presented as cumulative losses settled
- For each claims occurrence year the incremental claims loss settled for a particular development year is the amount settled in that development year
- The cumulative claims losses settled is the total amount settled up to that development year, i.e., the sum of the incremental claims losses settled up to that date.

Assumptions underlying the CLM

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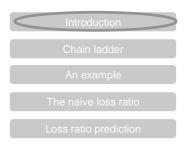
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- Patterns of claims loss settlement observed in the past will continue in the future
- The development of claims loss settlement over the development years follows an identical pattern for every claims occurence year
- But the observed claims loss settlement patterns may change over time:
 - Changes in product design and conditions
 - Changes in the claims reporting, assessment and settlement processes (example: different owners)
 - Change in the legal environment
 - Abnormally large or small claim settlement amounts
 - Changes in portfolio so that the history is not representative for predicting the future (example: strong growth)

Development patterns and development factors

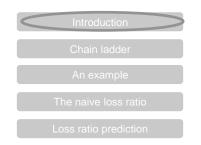


- The insurer may tend towards using any of the following patterns for estimation purposes:
- The proportion of the ultimate cumulative claims losses that is settled in a particular development year (development pattern for incremental claims losses settled)
- The proportion of the ultimate cumulative claims losses that is settled by a particular development year (development pattern for cumulative claims losses settled)
- The ratio of the cumulative losses settled by a particular year to the cumulative claims losses settled by the previous development year (cumulative loss factor)
- The three patterns are equivalent
- CLM relies on the last pattern above holds for all claims occurrence years. For any development year the quotient
 - Expected cumulative claims losses settled up to and including the development year/Expected cumulative claims losses settled up to and including the previous development year

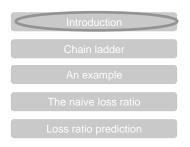
is called the *cumulative claims loss settlement factor* for that development year

• Example: the cumulative claims settlement factor for development year 4 is derived from the cumulative settlement amounts for development years 3 and 4.

Estimating future claims settlement amounts



- Underlying assumption (CLM):
 - the cumulative claims loss settlement factor for a specific development year is assumed to be the same for all claims occurence years
- The CLM estimator for each of the factors is based on the cumulative settlement data for as many claims occurence years as possible



Determining the CLM estimator for the cumulative claims loss settlement factor

Cumulative	Cumulative claims loss settlements				Developm	nent year			
settler			1	2	3	4	5	6	7
	2005	1232	2178	2698	3420	3736	3901	3949	3963
e C	2006	1469	2670	3378	4223	4684	4919	4975	
urence	2007	1652	3068	4027	4981	5586	5072	3736+4684	+5586+6401
ünc	2008	1831	3465	4589	567	6401		=20407	
0001	2009	2074	3993	5323	6563			3736±4684	+5586+6401
	2010	2434	4697	6358				=20407	+3300+0401
Claims year	2011	2810	6918						
CI	2012	3072						20407/1830	0=1,1151
CLM estimate	or for claims								
loss settlen	ment factor		1,9989	1,3140	1,2422	1,1151	1,0491	1,0118	1,0035

Comments:

- These CLM estimators for the cumulative claims loss settlement factors are used to estimate the cumulative claims loss settlement amount in the future
- For each claims occurrence year the last historical observation is used together with the appropriate CLM estimator for the development factor to estimate the cumulative settlement amount in the next development year
- This value is, in turn, multiplied by the estimator for the development factor for the next development year and so on.

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Determining the estimated cumulative claims loss settlements in future periods

Cumulative	claims loss				Developme	nt year			
settler	nents	0	1	2	3	4	5	6	7
	2005	1232	2178	2698	3420	3736	3901	3949	3963
	2006	1469	2670	3378	4223	4684	4919	4975	4993
year	2007	1652	3068	4027	4981	5586	5873	5942	5963
urence	2008	1831	3465	4589	5676	6401	6401 *1,0491 = 6715	6715 *1,0118 = 6794	7 ⁶⁷⁹⁴ *1,0035 = 6818
0001	2009	2074	3993	5323	6563	7319	7678	7768	7796
	2010	2434	4697	6358	7898	8807	9239	9348	9381
Claims	2011	2810	4918	6462	8027	8952	9391	9502	9535
Ö	2012	3072	5686	7471	9280	10349	10857	10985	11024
CLM estimate	or for claims		1,8508	1,3140	1,2422	1,1151	1,0491	1,0118	1,0035

Comments:

- The values shown in the red cells are the estimators for future cumulative claims settled
- These estimates are always based on the latest available cumulative claims settlement amounts for the relevant claims occurrence year, i.e., the estimated future cumulative claims settlements are always based on the last green diagonal of data
- It is now simple to derive the estimated incremental claims settlement amounts for the future periods
- An incremental settlement amount is the difference between tow consecutive cumulative settlement amounts

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Determining the estimated incremental settlement amounts from the estimated cumulative amounts

Incremental	claims loss				Developme	ent year			
settlei	ments	0	1	2	3	4	5	6	7
	2005	1232	946	520	722	316	165	48	14
	2006	1469	1201	708	845	461	235	56	18
year	2007	1652	1416	959	954	605	287	69	21
urence	2008	1831	1634	1124	1087	725	6715- 6401 = 314	6794- 6715 = 79	6818- 6794 = 24
- - - - - - - - - - - - - - - - - - -	2009	2074	1919	1330	1240	756	359	91	28
	2010	2434	2263	1661	1540	909	432	109	33
Claims	2011	2810	4108	1544	1565	924	439	111	34
5	2012	3072	2614	1785	1810	1069	508	128	39

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Determining the estimated incremental settlement amounts from the estimated cumulative amounts

mental	claims loss				Developme	nt year						Estimated claims loss
settler	ments	0	1	2	3	4	5	6	7		Calendar year	settlement amounts
Oottioi		V			Ů		Ů	,	,	^	2013	6855
	2005									7	2014	4718
	2006								18		2015	2181
	2007							69	21		2016	1069+439+109+28=1
	2008						314	79	24	1	2017	652
	2009					756	350		N		2018	162
					1510	000	000	01	20		2019	39
	2010				15/	909	432	111		7_		•
5	2011			15/4	1565	924	430	111	34	1		
0 >	2012		2614	1785	1810	1060	The state of the s	120	30			

Comments:

- Group the estimated incremental claims loss settlement amounts by the year in which they will be settled
- These cash flows can then be discounted to determine the technical provisions
- Norwegian State Treasury Bonds (Statsobligasjoner in Norwegian) may be used as discount factor
 - Example: a cash flow due in 2017 is discounted with a 4 year old Norwegian State
 Treasury Bond etc. Why do we hope that the development year does not exceed 10 ??

Claims reserves

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- Claims reserving is not only about statistical models:
- What is the purpose of the reserving?
- Know your data:
 - Is the history consistant? (relevant when a company has had several owners)
 - Is the history representative when future predictions are to be made? (relevant during strong growth)
 - Has there been significant events that affect settlement practice? (relevant for agencies)
- Should the reserves be calculated gross or net?
- Adjust for inflation
- Know your claims department:
 - How are provisions set?
 - When do they set the provision?
 - How will the lead time be affected by the size of the claims?
 - Are there any backlogs?
 - When do they change the reserves?

Notation

R	eporting	year		
Accident				
year	1	2	3	
1998	C_{11}	C_{12}	C_{13}	
1999	C_{21}	C_{22}		
2000	C_{31}			

- Cij Cumulative claims from accident year i, reported through the end of period j
- Dij = Cij- Cij-1 Incremental claims from accident year i, reported in period j
- Cim Ultimate claims for accident year i, where
- m is the last development period that is known and
- Ri = E[Cim] Cij is the reserve for accident year i
- fi one period loss development factor. Also called age-to-age factor or link ratio
- Fij Development factor from accident year i, period j, to ultimate
- Li Claims relative to an exposure for accident year i
- Pi A measure of exposure for accident year i
- Ak Experience up to development period k

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Chain ladder builds on that cumulative claims in a period are proportional to the claims in the preceding period. The proportionality factor depends on the number of periods since outset, but is expected to be the same for all periods. It is assumed that:

(CL1)
$$E[C_{ij+1} | C_{i1}, C_{i2}, ..., C_{ij}] = C_{ij} * f_j$$

Observe that fi does not depend on accident year.

(CL2) The vectors
$$\{C_{i1}, C_{i2}, ..., C_{im}\}$$

and $\{C_{k1}, C_{k2}, ..., C_{km}\}$ are independent if $i \neq k$

CL1 just brings us one step ahead, whereas we want to get to the end. To get there we are going to utilize the rule of double expectation:

(Lemma 6.1) If
$$E[Z]$$
 is finite, then $E[E[Z|X]]$

Using this lemma and CL1, we find

Loss ratio prediction

$$E[C_{ij+k} \mid C_{i1}, C_{i2}, ..., C_{ij}] =$$

$$= E[E[C_{ij+k} \mid C_{i1}, C_{i2}, ..., C_{ij+k-1}] \mid C_{i1}, C_{i2}, ..., C_{ij}] =$$

$$= E[C_{ij+k-1} * f_{j+k-1} \mid C_{i1}, C_{i2}, ..., C_{ij}] =$$

$$= E[C_{ij+k-1} \mid C_{i1}, C_{i2}, ..., C_{ij}] * f_{j+k-1} =$$

$$= E[E[C_{ij+k-1} \mid C_{i1}, C_{i2}, ..., C_{ij+k-2}] \mid C_{i1}, C_{i2}, ..., C_{ij}] * f_{j+k-1} =$$

$$= E[C_{ij+k-2} * f_{j+k-2} \mid C_{i1}, C_{i2}, ..., C_{ij}] * f_{j+k-1} =$$

$$= E[C_{ij+k-2} \mid C_{i1}, C_{i2}, ..., C_{ij}] * f_{j+k-1} * f_{j+k-2} =$$

$$= ... = C_{ii} * f_{i} * f_{i+1} * f_{i+2} * * f_{i+k-1}$$

$$(6.2)$$

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We could rewrite CL1 on the following form:

(CL1')
$$E[C_{ij+1}/C_{ij} | C_{i1}, C_{i2}, ..., C_{ij}] = f_{ij}$$

Thus, we could use observed ratios C_{ij+1}/C_{ij} as unbiased estimators of f_j. Before combining estimates of the same f_j we make a further assumption,

(CL3)
$$Var[C_{ij+1} | C_{i1}, C_{i2}, ..., C_{ij}] = C_{ij} * \sigma_j^2$$

Observe that the last factor in the variance is no depending on accident year. We also need the following Lemma

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(Lemma 6.2) Suppose X_i are uncorrelated random variables with the same mean, but with variances σ_i^2 . Then the best linear unbiased estimator of the mean is given by

$$\sum_{i=1}^{n} w_{i} X_{i} \text{ where } w_{i} = \frac{C_{ij}}{\sum_{k=1}^{m-j} C_{kj}} \text{ and } \sum_{i=1}^{n} w_{i} = 1$$

From Lemma 6.2 it turns out that

$$\hat{f}_{j} = rac{\displaystyle\sum_{i=1}^{m-j} C_{ij+1}}{\displaystyle\sum_{i=1}^{m-j} C_{ij}}$$

Proof: Form the Lagrangian

Constraint

 $L(w_1,\ldots,w_n,\lambda) = \sum_{i=1}^n w_i^2 \sigma_i^2 + \lambda (1-\sum_{i=1}^n w_i)$ $\operatorname{Var} \sum_{i=1}^n w_i X_i \text{ , which we want to minimize}$

Solve the system

$$\begin{cases}
\frac{\partial}{\partial w_k} L(w_1, ..., w_n, \lambda) = 0, & k = 1, ..., n \\
\frac{\partial}{\partial \lambda} L(w_1, ..., w_n, \lambda) = 0
\end{cases}$$
(1)

$$\frac{\partial}{\partial \lambda} L(w_1, ..., w_n, \lambda) = 0$$
(2)

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Proof: Starting with (1) yields

$$\frac{\partial}{\partial w_k} L(w_1, ..., w_k, \lambda) = 2w_k \sigma_k^2 - \lambda = 0$$

$$\Leftrightarrow \lambda = 2w_k \sigma_k^2 \Leftrightarrow w_k = \frac{\lambda}{2\sigma_k^2}$$

Continuing with (2) yields

$$\frac{\partial}{\partial \lambda} L(w_1, ..., w_k, \lambda) = 1 - \sum_{i=1}^n w_i = 0$$

$$\Leftrightarrow \sum_{i=1}^{n} w_{i} = 1 \stackrel{(3)}{\Leftrightarrow} \sum_{i=1}^{n} \frac{\lambda}{2\sigma_{i}^{2}} = 1 \Leftrightarrow \lambda = 2 \frac{1}{\sum_{i=1}^{n} \sigma_{i}^{-2}}$$
(4)

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Combining (3) and (4) yields

$$w_{k} = \frac{\sum_{i=1}^{n} \sigma_{i}^{-2}}{2\sigma_{i}^{2}} = \frac{\sigma_{i}^{-2}}{\sum_{i=1}^{n} \sigma_{i}^{-2}}$$
(5)

Rewriting CL3 gives

(CL3')
$$Var[C_{ij+1}/C_{ij} | C_{i1}, C_{i2}, ..., C_{ij}] = \frac{1}{C_{ij}^2} C_{ij} * \sigma_j^2 = \frac{\sigma_j^2}{C_{ij}}$$

and the weights are thus by (5)

$$w_{i} = (\sigma_{j}^{2} / C_{ij})^{-1} / \sum_{k=1}^{m-j} (\sigma_{j}^{2} / C_{kj})^{-1} = C_{ij} / \sum_{k=1}^{m-j} C_{kj}$$

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and

$$\hat{f}_{j} = \sum_{i=1}^{m-j} w_{i} f_{ij}^{\hat{}} = \sum_{i=1}^{m-j} \frac{C_{ij}}{\sum_{k=1}^{m-j} C_{kj}} \frac{C_{ij+1}}{C_{ij}} = \frac{\sum_{i=1}^{m-j} C_{ij+1}}{\sum_{i=1}^{m-j} C_{ij}}$$
(6)

To be able to use the algorithm suggested by formula (6.2) with the estimators from (6) above we need to prove that estimates are uncorrelated. Define the set of experience up to development period k by

$$A_k = \{C_{ij} \mid j \le k, i \le m\}$$

Then we have

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$$\begin{split} E[f_{j} * f_{k}] &= E[E[f_{j} * f_{k} | A_{k}]] = E f_{j} * [E[f_{k} | A_{k}]] \\ &= E f_{j} * [E[(\sum_{v=1}^{m-k} C_{vk+1}) / (\sum_{v=1}^{m-k} C_{vk}) | A_{k}]] = \\ &= E[f_{j} * [E[(\sum_{v=1}^{m-k} C_{vk+1}) | A_{k}] / (\sum_{v=1}^{m-k} C_{vk})] = \\ &= E[f_{j} * [\sum_{v=1}^{m-k} E[C_{vk+1} | C_{vk}] / (\sum_{v=1}^{m-k} C_{vk})] = \\ &= E[f_{j} * [f_{k} * (\sum_{v=1}^{m-k} C_{vk}) / (\sum_{v=1}^{m-k} C_{vk})] = \\ &= E[f_{j}] * f_{k} = E[f_{j}] * E[f_{k}] \end{split}$$

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Thus we have shown that the estimates of f_j and f_k are uncorrelated. If we combine this with (6.2) it shows that the following ultimate estimator is unbiased,

$$E[C_{im} \mid C_{ij}] = C_{ij} * \hat{f}_{j} * ... * \hat{f}_{m-1}$$

Chain ladder - example

- Example: one fire/combined product (fire/combined in Norway includes home, contents and cabin)
- The triangle shows the payments for the different years
- How do we fill out the blanks?

	1	2	3	4	5
2008	7 008 148	25 877 313	31 723 256	32 718 766	33 019 648
2009	30 105 220	65 758 082	76 744 305	79 560 296	
2010	89 181 138	171 787 015	201 380 709		
2011	109 818 684	198 015 728			
2012	97 250 541				

We start by estimating f₁, f₂, f₃, f₄

$$\hat{f}_{1} = \frac{\sum_{i=1}^{5-1} C_{i1+1}}{\sum_{i=1}^{5-1} C_{i1}} = \frac{C_{12} + C_{22} + C_{32} + C_{42}}{C_{11} + C_{21} + C_{31} + C_{41}} =$$

$$=\frac{25.9+65.8+171.8+198.0}{7.0+30.1+89.2+109.8}=1.954$$

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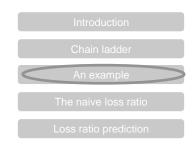
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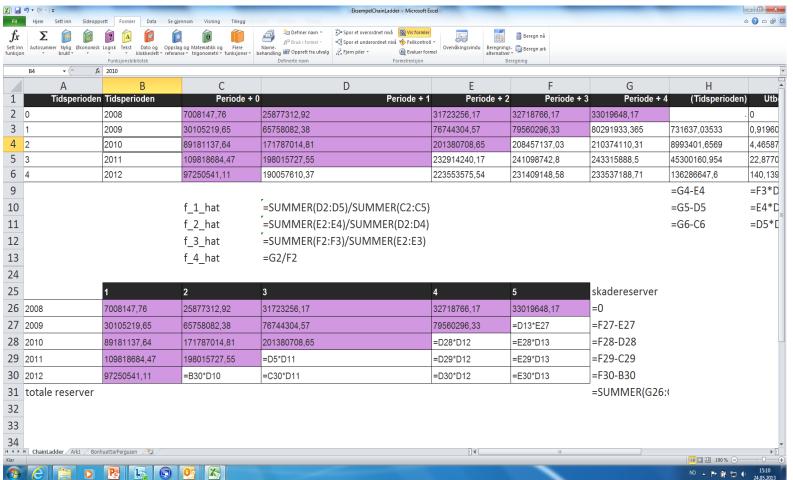
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f_1_hat	1,954
f_2_hat	1,176
f_3_hat	1,035
f_4_hat	1,009

Chain ladder - example



Then we use the formula $E[C_{im} \mid C_{ij}] = C_{ij} * f_j * ... * f_{m-1}$ to fill out the blanks



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for a premium an insurance company commits itself to pay a sum if an event has occured

