

## *Session 4 Call Paper Program Topics*

### *A Comparative Study of the Performance of Loss Reserving Methods through Simulation*

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# *Agenda*

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- Loss Reserving Methods
  - Loss Development
  - Bühlmann Complementary Loss Ratio
  - Regression Models
- Simulation of Random Loss Triangles
  - Random Reporting Factors
  - Random Backward Development Factors
  - Individual Losses with Changing Severity
  - Pentikainen & Rantala Method
- Comparison Criteria
- Simulation Results
- Conclusions

# *Loss Reserving Methods*

## *Loss Development*

$L_{i,j}$  = losses paid by period j for accident year i

$$i, j = 1, \dots, n.$$

$$f_{i,j} = L_{i,j+1} / L_{i,j}$$

$$\text{Development factor } f_j = \sum_{i=1}^{n-j} f_{i,j} / (n - j)$$

$$\begin{aligned} \text{Age k to ultimate factor } U_k &= \prod_{j=k}^{n-1} f_j \\ U_n &= 1 \end{aligned}$$

$$\text{Ultimate for AYi} = L_{i,n-i+1} * U_{n-i+1}$$

$$\text{Reserve for AYi} = \text{Ultimate}(i) - L_{i,n-i+1}$$

# *Loss Reserving Methods*

## *Buhlman Complementary Loss Ratio*

Incremental loss  $s_{i,j} = L_{i,j} - L_{i,j-1}$

R = inflation rate

Inflated losses  $S_{i,j} = s_{i,j} * (1 + R)^{n-i}$

Average  $M_j = \sum_{i=1}^{n+j-1} S_{i,j} / (n - j + 1)$

$\hat{s}_{i,j} = M_j (1 + R)^{(i-n)}$

Reserve for AYi =  $\sum_{j=n+2-1}^n \hat{s}_{i,j}$

# *Loss Reserving Methods*

## *Regression Models*

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$$Z_{i,j} = \log s_{i,j}$$

Model 1:

$$\begin{aligned} Z_{i,j} &= \mu + \alpha_i + \beta_j + e_{i,j} \\ \alpha_1 &= \beta_1 = 0 \end{aligned}$$

Model 2:

$$\begin{aligned} Z_{i,j} &= \mu + (i-1)\alpha + \beta_j + e_{i,j} \\ \beta_1 &= 0 \end{aligned}$$

Model 3:

$$Z_{i,j} = \mu + (i-1)\alpha + (j-1)\beta + \gamma \log j + e_{i,j}$$

$e_{i,j}$  are random noise assumed  $N(0, \sigma^2)$

# *Loss Reserving Methods*

## *Regression Models*

$$Z = X\theta + e$$

$$Z = [Z_{11}, Z_{12} \dots Z_{1,n}, Z_{2,1} \dots Z_{n,1}]^T$$

$$\text{Model 1: } \theta = [\mu, \alpha_2 \dots \alpha_n, \beta_2 \dots \beta_n]$$

$$X = \begin{bmatrix} 1 & 0 & 0 & \dots & 0, & 0, & \dots & \dots & 0 \\ 1 & 0 & \dots & \dots & 0 & 1 & \dots & \dots & 0 \\ \vdots & & & & & & & & \\ 1 & 0 & \dots & \dots & 0 & 0 & 0 & \dots & 1 \\ 1 & 1 & 0 & 0 & \dots & 0 & \dots & \dots & 0 \\ 1 & 1 & 0 & \dots & 0 & 1 & 0 & \dots & 0 \\ \vdots & & & & & & & & \\ 1 & 0 & \dots & \dots & 1 & 0 & \dots & \dots & 0 \end{bmatrix}$$

e is a random vector.

$$E(e) = 0, V(e) = \sigma^2 I$$

Use method of least squares to estimate  $\theta, \sigma^2$

$$\hat{\theta} = (X^T X)^{-1} X^T Z$$

$$\hat{\sigma}^2 = (Z^T Z - \hat{\theta}^T X^T Z) / (r - p)$$

$$r = n(n+1)/2, \quad p = 2n - 1$$

# *Loss Reserving Methods*

## *Regression Models*

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Model 2:  $\theta = [\mu, \alpha, \beta_2, \dots, \beta_n]^T$

$$X = \begin{bmatrix} 1 & 0 & 0 & \dots & \dots & 0 \\ 1 & 0 & 1 & \dots & \dots & 0 \\ \vdots & & & & & \\ 1 & 0 & 0 & \dots & \dots & 1 \\ 1 & 1 & 0 & \dots & \dots & 0 \\ 1 & 1 & 1 & 0 & \dots & 0 \\ \vdots & & & & & \\ 1 & n-1 & 0 & \dots & \dots & 0 \end{bmatrix}$$

Model 3:  $\theta = [\mu, \alpha, \beta, \gamma]$

$$X = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & \log 2 \\ \vdots & & & \\ 1 & 0 & n-1 & \log n \\ 1 & 1 & 0 & 0 \\ \vdots & & & \\ 1 & (n-1) & 0 & 0 \end{bmatrix}$$

# *Loss Reserving Methods*

## *Regression Models*

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Define:

$$g(m, t) = \sum_{k=0}^{\infty} \frac{m^k (m+2k)}{m(m+2)\cdots(m+2k)} \frac{t^k}{k!}$$

Unbiased estimate of  $s_{i,j}$

$$\hat{s}_{i,j} = \exp(A_{i,j} \hat{\theta}) \quad g\left(t - p, \frac{1}{2} \left(1 - A_{i,j} (X^T X)^{-1} A_{i,j}^T\right) \hat{\sigma}^2\right)$$

Where  $E Z_{i,j} = [A_{i,j}] \quad \theta$ .

Look for details in Verrall (1994).

Once  $s_{i,j}$  are estimated, accident year reserves can be computed.

# *Simulation of Random Loss Triangles*

## *Random Reporting Factors*

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- Step 1 Generate random number of claims.
  - Step 2 Random claim size for each claim.
  - Step 3 Add all the claims together to get  $S_1$ .
  - Step 4 Uniform random numbers  $X_2, X_3 \dots X_{10}$
  - Step 5  $T_j = 0.1 + 0.5 X_j + 0.5 \log j$
  - Step 6  $U_j = T_1 + T_2 + \dots + T_j$
  - Step 7  $L_{1,j} = S_1 * (1 - \exp(-U_j)), j = 1, 2, \dots, 10. \quad L_{1,11} = S_1$
  - Step 8 Repeat steps 1 through 7 and multiply accident year  $i$  losses by  $(1.06)^{(i-1)}$ .

Note: Claim frequency Poisson with mean 100.

Claim severity is log normal with  $\mu = 7.36, \sigma^2 = 1.510$ .

# *Simulation of Random Loss Triangles*

## *Backward Development Factors*

Steps 1 - 3 As in Method 1

Step 4      Simulate  $X_j \sim N(\mu_j, \sigma^2_j)$ ,  $j=1,\dots,10$

$$\mu_j = (11 - j + (10 - j)^2) / 100$$

$$\sigma_j^2 = (11 - j + (10 - j)^2) / 500$$

Step 5       $f_j = \exp(X_j)$

$$F_j = \prod_{k=j}^{10} f_k$$

$$L_{i,j} = S_1 / F_j$$

Step 6      Repeat Steps 1 through 5 for each accident year and inflate the losses.

# *Simulation of Random Loss Triangles*

## *Individual Losses with Changing Severity*

Step 1 Generate frequency of claims  $N$ , Poisson with mean 100.

Step 2 For each claim, simulate:

- A. Occurrence delay  $x_1$ ; uniform (0,1).
- B. Reporting delay  $x_2$ ; exponential with mean 2.
- C. Claim closing delay  $x_3$ ; exponential with mean 5.
- D. Cap  $x_2$  at  $11 - x_1$ , and  $x_3$  at  $11 - x_1 - x_2$ .

Step 3 Simulate percentile level  $x$  for each claim; uniform (0,1).

Step 4 Claim size  $\ell_k$  for a claim is

$$\begin{aligned}\ell_k &= 0 && k < x_1 + x_2 \\ &= \lambda(k)((1-x)^{1/\theta(k)} - 1), && x_1 + x_2 \leq k \leq x_1 + x_2 + x_3 \\ &= \lambda(r)((1-x)^{1/\theta(r)} - 1), && k > x_1 + x_2 + x_3\end{aligned}$$

Where  $r$  is the smallest integer  $\geq x_1 + x_2 + x_3$

$$\lambda(k) = (1000 + (k-1)50) 1.06^{k-1}$$

$$\theta(k) = 2.5 - 0.05(k-1)$$

Step 5 Add the individual claims  $\ell_k$ .

Step 6 Inflate losses 6% for accident years 2,..11.

# *Simulation of Random Loss Triangles*

## *Pentikainen Rantala Method*

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- Step 1      Simulate  $y \sim N(0.4, 0.05^2)$  for each  $i, j$
  - Step 2      Compute  $q(i, j) = 0.6 * q(i, j-1) + y$   
Where  $q(i, 0) = 1$
  - Step 3      Simulate  $y \sim N(0, 0.015^2)$  and  
Compute  $x(i) = \inf(1) + 0.6(\inf(i) - \inf(1)) + y$   
 $\inf(i+1) = \max[x(i), 0.03]$   
Where  $\inf(1) = 0.06$  and repeat for  $i = 1, \dots, 20$
  - Step 4      Compute  $\text{infe}(i) = 1.01^{(i-1)}$ ,  $i = 1, \dots, 11$
  - Step 5      Compute the incremental losses  
 $s_{i,j} = 500,000 * q(i, j) * \text{infe}(i) * \inf(i+j-1) * x_p(j)$   
Where  $x_p(j)$  is the  $j$ th element of the vector  
[.22, .18, .15, .12, .1, .08, .06, .04, .027, .016, .007]
  - Step 6      Loss triangle can be computed from  $s_{i,j}$

# *Criteria for Comparing Results*

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1. Bias: Average of (actual - estimated reserves)
  2. R.M.S.E.:  $\sqrt{\text{Average}(\text{actual} - \text{estimated reserves})^2}$
  3. Avg. Abs. Dev.: Average  $|\text{actual} - \text{estimated reserves}|$
  4. Average % Error: Average  $\left[ \frac{\text{actual} - \text{estimated reserves}}{\text{actual}} \right]$
  5. Correlation of Actual and Estimated Reserves

Note: Criterion 5 used for total reserve only.

# *Simulation Results*

## *Loss Triangle Simulation\_Random Reporting Factors*

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**TABLE 1**

Actual Total Reserve:      Average =      1,108,298      SD=      244,287

<b>Method of Forecast:</b>	<b>Dev Factor</b>	<b>Buhlmann Loss Ratio</b>	<b>Regression</b>		
			<b>Model 1</b>	<b>Model 2</b>	<b>Model 3</b>
<b>Five Thousand Iterations</b>					
Bias	151,681	5,222	36,486	31,240	51,367
RMSE	466,055	266,874	395,819	328,870	341,537
Avg Abs Dev	364,628	204,674	314,829	254,069	263,444
Avg %	16.84%	4.84%	6.22%	6.75%	8.69%
Corr Actual vs Est	0.25	0.09	0.25	0.15	0.14

# *Simulation Results*

## *Loss Triangle Simulation\_Random Backward Development Factors*

**TABLE 2**

Actual Total Reserve:      Average =      3,665,734      SD=      485,206

<b>Method of Forecast:</b>	<b>Dev Factor</b>	<b>Buhlmann Loss Ratio</b>	<b>Regression</b>		
			<b>Model 1</b>	<b>Model 2</b>	<b>Model 3</b>
<b>Five Thousand Iterations</b>					
Bias	157,684	(8,088)	55,356	15,393	3,125
RMSE	512,092	639,187	481,727	542,257	519,705
Avg Abs Dev	391,022	485,769	373,282	420,438	403,056
Avg %	4.38%	1.23%	1.58%	0.79%	0.47%
Corr Actual vs Est	0.70	0.11	0.70	0.57	0.58

# *Simulation Results*

## *Loss Triangle Simulation\_Individual Losses with Changing Severity*

**TABLE 3**

Actual Total Reserve:      Average =      1,634,559      SD=      252,631

<b>Method of Forecast:</b>	<b>Dev Factor</b>	<b>Buhlmann</b>	<b>Regression</b>		
		<b>Loss Ratio</b>	<b>Model 1</b>	<b>Model 2</b>	<b>Model 3</b>
<b>Five Thousand Iterations</b>					
Bias	30,566	(83,039)	(144,192)	(52,327)	(176,089)
RMSE	413,137	441,109	375,367	299,099	340,506
Avg Abs Dev	356,932	347,340	314,629	259,057	280,243
Avg %	1.39%	-4.36%	-9.49%	-3.31%	-9.52%
Corr Actual vs Est	0.62	0.39	0.68	0.66	0.32

# *Simulation Results*

## *Loss Triangle Simulation\_Pentikainen & Rantala Method*

**TABLE 4**

Actual Total Reserve:      Average =      3,183,654      SD=      330,776

<b>Method of Forecast:</b>	<b>Dev Factor</b>	<b>Buhlmann Loss Ratio</b>	<b>Regression</b>		
			<b>Model 1</b>	<b>Model 2</b>	<b>Model 3</b>
<b>Five Thousand Iterations</b>					
Bias	10,106	(21,441)	5,326	4,789	34,136
RMSE	186,688	186,916	183,351	195,148	201,012
Avg Abs Dev	147,536	147,830	145,029	153,675	157,283
Avg %	0.23%	-0.24%	0.07%	0.06%	0.98%
Corr Actual vs Est	0.89	0.84	0.89	0.88	0.88

# *Conclusion*

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1. Regression models compete well with traditional methods.
  2. Parsimony is important for regression models.
  3. LDF method provides reasonable answers.
  4. Variance is not considered, but RMSE is an indicator of variance if the method is unbiased.

Note:

- i) Regression model results should be better in practice because of user interaction in selecting an appropriate model.
- ii) For individual accident year results, see CAS 1997 Forum.