

CAS Centennial
November 2014

A Deeper Understanding of Experience Rating -- Split and Unsplit

Ira Robbin, PhD



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Credibility of Experience

– Conceptual Drivers

- Volatility of Actual Loss
 - Process Risk = Noise
 - More Noise \Rightarrow Less Credibility
- Belief in Initial Estimate
 - Parameter Risk = Uncertainty in Initial Estimate
 - More Parameter Risk \Rightarrow More Credibility
 - *The Less You Know*(in advance)

Pop-quiz 1

- A simple class plan is replaced with one that is more refined. The z for individual risk experience will:
 - A. Increase B. Stay the same C. Decline
- The expected mean loss for a risk is initially \$1,000 with standard deviation of \$100. The z -weighted estimate is \$1,100. The standard deviation of that estimate is:
 - A. Greater than \$100 B. = \$100 C. Less than \$100

Notation

RV	A
Condl Mean	$\mu(\theta)$
Condl Variance	$\sigma^2(\theta)$
UnCondl Mean	μ or E
Process Variance	$\sigma^2 = E[\sigma^2(\theta)]$
Parameter Variance	$\tau^2 = \text{Var}(\mu(\theta))$
Total Variance	$\lambda^2 = \sigma^2 + \tau^2$

Credibility Estimate and MSE

$$\mu^*(z) = zA + (1 - z)E$$

- *Given arbitrary credibility, z , the mean square error (MSE) is given as:*

$$\begin{aligned}\varepsilon^2 &= E[(\mu^*(z) - \mu(\theta))^2] \\ &= z^2 \cdot \sigma^2 + (1 - z)^2 \tau^2\end{aligned}$$

Optimal Credibility

- *The z which minimizes mean square error is given as z^* where:*

$$z^* = \frac{\tau^2}{\tau^2 + \sigma^2} = \frac{\tau^2}{\lambda^2}$$

What increases optimal credibility?

- *Reducing process risk*
- *Increasing parameter risk*

Error Reduction in Estimate of the Mean

- *It can be proved that Optimal Mean Square Error is given as:*

$$\varepsilon_0^2(NS) = \tau^2 (1 - z^*)$$

- *Initial mean square error is τ^2*
- *z^* = the factor by which parameter risk is reduced using optimal weighting*

Example of Error Reduction

- Let $\tau^2 = 100$ and $\sigma^2 = 300$.

It follows that:

$$\lambda^2 = 400 \text{ and } z^* = 100/400 = 25\%$$

$$\begin{aligned} \text{Optimal MSE} &= E[(\mu - \mu^*)^2] \\ &= \tau^2 \cdot (1 - z^*) = 100 \cdot (1 - .25) = 75 \end{aligned}$$

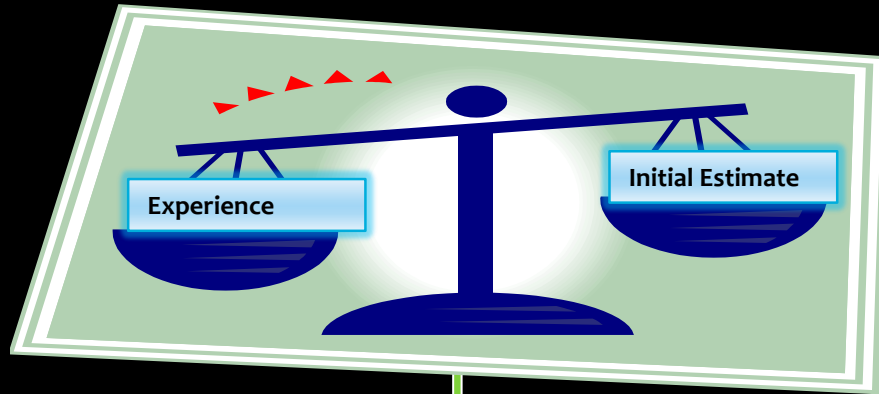
Split Credibility

Split Credibility – US Workers Comp

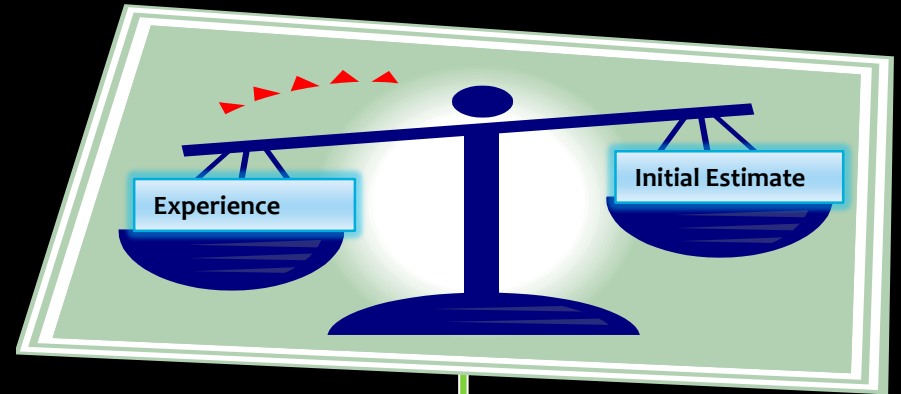
- US Workers Comp Experience Rating
- Split-z procedure
- Primary and Excess Split
- Accidents capped at state per accident limit
- Tables of primary and excess Z by risk size
- Tables of expected primary and excess loss rates by class using d ratios

Split Credibility

Bucket 1



Bucket 2



Revised Estimate Bucket 1

Revised Estimate Bucket 2

Final Revised Estimate

Split Credibility – Basic Formula

- Final Estimate = Sum of Credibility Weighted Estimates

$$\begin{aligned}\mu^* &= \mu_1^* + \mu_2^* \\ &= \{z_1 A_1 + (1 - z_1) E_1\} \\ &\quad + \{z_2 A_2 + (1 - z_2) E_2\}\end{aligned}$$

Credibility Example-Loss Data

Loss Experience			
Claim Number	Total Loss	Primary	
		Loss	Excess Loss
1	1,000	1,000	-
2	1,500	1,500	-
3	2,500	2,500	-
4	4,000	4,000	-
5	15,000	5,000	10,000
6	80,000	5,000	75,000
Total	104,000	19,000	85,000

Based on Split point = 5,000

Credibility Example – Non-Split vs Split

Experience Rating				
No Split				Z-wtd
Plan	Actual Loss	Credibility	Expected Loss	estimate
Total	104,000	50%	100,000	102,000
Split Plan				Z-wtd
Actual Loss	Credibility	Expected Loss	estimate	
Primary	19,000	70%	30,000	22,300
Excess	85,000	20%	70,000	73,000
Combined Split Estimate				95,300

Notation

RV	A_1	A_2	$A = A_1 + A_2$
Condl Mean	$\mu_1(\theta)$	$\mu_2(\theta)$	$\mu(\theta)$
Condl Variance	$\sigma_1^2(\theta)$	$\sigma_2^2(\theta)$	$\sigma^2(\theta)$
UnCondl Mean	μ_1	μ_2	μ
Process Variance	σ_1^2	σ_2^2	σ^2
Parameter Variance	τ_1^2	τ_2^2	τ^2
Total Variance	λ_1^2	λ_2^2	λ^2
Process Cov		ρ	
Parameter Cov		π	
Total Cov		κ	

Split Credibility Notation- Total Variances

➤ *Total Component Variances*

- $\lambda_1^2 = \sigma_1^2 + \tau_1^2$ and $\lambda_2^2 = \sigma_2^2 + \tau_2^2$

➤ *Total Covariance*

- $\kappa = \rho + \pi$

➤ *Total Variances*

- *Total:* $\lambda^2 = \lambda_1^2 + \lambda_2^2 + 2\kappa$

- *Process:* $\sigma^2 = \sigma_1^2 + \sigma_2^2 + 2\rho$

- *Parameter:* $\tau^2 = \tau_1^2 + \tau_2^2 + 2\pi$

Optimal Split Credibility Formulas

$$z_1 = \frac{\lambda_2^2 (\tau_1^2 + \pi) - \kappa (\tau_2^2 + \pi)}{D}$$

$$z_2 = \frac{\lambda_1^2 (\tau_2^2 + \pi) - \kappa (\tau_1^2 + \pi)}{D}$$

where $D = \lambda_1^2 \lambda_2^2 - \kappa^2$

Mean Square Error with Optimal Credibilities

- It can be proved that minimal MSE is given as :

$$\varepsilon_0^2(SP) = (\tau_1^2 + \pi)(1 - z_1^*) + (\tau_2^2 + \pi)(1 - z_2^*)$$

- Error reduction interpretation
 - Each component starts with its own parameter risk plus the parameter covariance
 - The separate “z” are the factors by which the parameter error is reduced for each component

When Is a Split Plan Effective?

- Most extreme improvement is achieved if:
 - One component gets all the parameter risk
 - The other gets all the process risk
 - Covariances are zero
- Intuition: A Split works to the degree that it separates noise from signal!
- Split plan improves on No-split Plan when the Split Induces a Differential Allocation of Process and Parameter Risk.

Example-Split Plan A

Unsplit Plan		Split Plan A	Combined	Primary	Excess	CoVar
<i>Process Var</i>	300.0	<i>Process Var</i>	300.0	200.0	60.0	20.0
<i>Parameter Var</i>	100.0	<i>Parameter Var</i>	100.0	66.7	20.0	6.7
<i>Total Var</i>	400.0	<i>Total Var</i>	400.0	266.7	80.0	26.7
		<i>D</i>	20,622			
<i>Credibility</i>	25%	<i>Credibility</i>		25%	25%	
<i>MSE</i>	75.0	<i>MSE</i>	75.0			

Example-Split Plan B

Unsplit Plan		Split Plan B	Combined	Primary	Excess	CoVar
<i>Process Var</i>	300.0	<i>Process Var</i>	300.0	150.0	130.0	10.0
<i>Parameter Var</i>	100.0	<i>Parameter Var</i>	100.0	80.0	10.0	5.0
<i>Total Var</i>	400.0	<i>Total Var</i>	400.0	230.0	140.0	15.0
		<i>D</i>	31,975			
<i>Credibility</i>	25%	<i>Credibility</i>		37%	7%	
<i>MSE</i>	75.0	<i>MSE</i>	67.9			

Quiz 2

- Which was a more effective split?
 - A
 - B
 - Both equally effective
- Split Plan B does not work as well as a no-split plan because component 2 has a credibility of only 7% versus 25% for the non-split plan.
 - True
 - False

Quiz 3: True or False

Assume credibilities are determined by minimizing MSE:

- Q1. Splitting produces two layers each with higher z than the z of the original unsplit losses.
- Q2. $Z_p \leq Z_e$

Conclusions and Questions

- The Split Credibility approach works in practice
- With arbitrary frequency severity model and arbitrary priors, splitting does not necessarily offer much improvement.
- With the right model and reasonable data and a well chosen split point, split credibility can be superior to unsplit credibility.
- Having caps on large losses, raises credibility of the excess layer and improves overall performance
- Questions??