



Mack Model

Mack Model



Mack Model

- Mack developed a distribution-free method for computing the Variances of chain ladder age-to-age and age-to-ultimate factors
- The model is applied to weighted average development factors
- While no distribution assumptions are required to estimate factor Variances, they are needed to compute percentiles of ultimate loss and unpaid distributions




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Chain Ladder Assumptions

[1] $E[c(w,d+1)|c(w,1), \dots, c(w,d)] = c(w,d)F(d)$

[2] $c(i,1), \dots, c(i,n)$ & $\{c(j,1), \dots, c(j,n)\}$ are independent for $i \neq j$

[3] $\text{Var}[c(w,d+1)|c(w,1), \dots, c(w,d)] = \alpha_d^2 \sigma(w,d)^2$
– OR –
 $\text{Var}[c(w,d+1)/c(w,d)|\text{Data}] = \alpha_d^2/c(w,d)$
for proportionality constants α



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Mack Mean

- Under these assumptions, the best estimate of the age-to-age factor is a weighted average

$$E[F(d)] = \frac{\sum_w c(w, d)}{\sum_w c(w, d)} \times \frac{\sum_w c(w, d+1)}{\sum_w c(w, d)}$$

- The Ultimate estimate is:
 $E[c(w, n)|D] = c(w, d) \times F(d) \times F(d+1) \times \dots \times F(n-1)$

where D is known data






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Mack Variance

- Since the mean is weighted, the variance is also weighted.
- Variance associated with one age-to-age factor or column of losses, σ_d^2 :



$$\sigma_d^2 = \frac{1}{N-d-1} \sum_{j=1}^{N-d} c(j, d) \left(\frac{c(j, d+1)}{c(j, d)} - F(d) \right)^2$$

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Exercises using Mack Data

- Compute the weighted average age-to-average factors for each column
- Compute the weighted variances for age 1 for the factors in the exercise triangle
- **Bonus:** Calculate weighted averages and variances for age 1 in the complete Mack data triangle

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Answer
(Variance of Column 1)

| Computation of σ_1^2 for Development Age 1 | | | | |
|---|--------|--------------------------------------|-------------------|--------------------|
| Accident Year (w) | c(w,1) | F(w,1) | F(w,1) - F(1) | Weighted Deviation |
| (1) | (2) | (3) | (4) | (5) |
| | | | $[(3) - 2.334]^2$ | (2) x (4) |
| 1981 | 5,012 | 1.650 | 0.47 | 2,345.0 |
| 1982 | 106 | 40.425 | 1,450.90 | 153,795.4 |
| 1983 | 3,410 | 2.637 | 0.09 | 313.3 |
| 1984 | 5,655 | 2.043 | 0.08 | 477.3 |
| Weighted Average [F(I)] | | 2.334 | $\Sigma(5)$ | 156,930.9 |
| | | $\sigma_1^2 = \Sigma(5) / (N-d-1) =$ | | 52,310.3 |

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Variance of Ultimates

- We want variance of future payments or future incurred loss changes
- $MSE[c(w,n)] = E\{[c(w,n) - E\{c(w,n)\}]^2 | D\}$ where D is data
- Iterative rule of expectations
- $MSE[c(w,n)] = Var[c(w,n)|D] + \{E[c(w,n)|D] - E[c(w,n)]\}^2$
- Mean squared error = process variance of Ultimate + Parameter variance of estimate of ultimate
- Does not take into account changes in underlying model in the future.

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Variance of Ultimates

- Iterative computation to get variance of ultimate

$$Var[c(w,n)] = E[c(w,n-1)]\sigma_{n-1}^2 + E[c(w,n-1)]^2 F(n-1)^2 =$$

$$c(w,n-k+1)F(n-k+1)\dots F(n-2)\sigma_{n-1}^2 +$$

$$\{E[c(w,n-2)]^2 F(n-2)^2 F(n-1)^2 + E[c(w,n-2)]F(n-1)^2 \sigma_{n-2}^2\}$$

- Variance of unpaid = variance of ultimate

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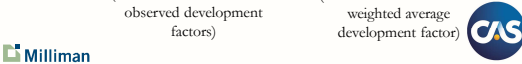
Mack Variance of a Row (AY)

- The Mack formula for the variance of the reserve estimate for accident year w is:

$$Var[R(w, d)] = U(w)^2 \sum_{d=w+1-w}^{w-1} \frac{\sigma_d^2}{F(d)^2} \left(\frac{1}{E[c(w, d)]} + \frac{1}{\sum_{j=1}^{N-d} c(j, d)} \right)$$

Process Variance
 (variance of the column of observed development factors)


Parameter Variance
 (variance of the calculated weighted average development factor)



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Example

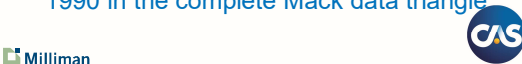
| d | $F(d)$ | σ_d^2 | $\frac{\sigma_d^2}{F(d)^2}$ | $E[c(1,d)]$ | SUM[c(j,d)] | $\frac{1}{E[c(1,d)]} + \frac{1}{\text{SUM}}$ | SUM | $\frac{\sigma_d^2}{F(d)^2}$ |
|-----|--------|--------------|-----------------------------|-------------|-------------|--|--|-----------------------------|
| (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) | (9) |
| 4 | 1.147 | 161.2 | 122.6 | 4.920 | 11.805 | 0.0002880 | 0.04 | |
| 3 | 1.378 | 2,887.3 | 1,519.8 | 3,569 | 16,303 | 0.0003415 | 0.52 | |
| 2 | 1.401 | 161.2 | 82.2 | 2,549 | 21,546 | 0.0004388 | 0.04 | |
| 1 | 2.334 | 52,310.3 | 9,603.8 | 1,092 | 14,183 | 0.0009863 | 9.47 | |
| | | | | $U(w) =$ | 5,642 | $\text{Sum} =$ | | 10.06 |
| | | | | | | | $Var[R(w,d)] = U(w)^2 \times \text{Sum} =$ | 320,349,084 |
| | | | | | | | $SE =$ | 17,898 |



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Exercise

- Compute the standard error for 1984
 - Compute σ_d^2 for ages 2 through 4 for the exercise data (use minimum of 2 & 3 for 4)
 - Assume age 5 is the ultimate valuation.
- Bonus:** Compute the standard error for 1990 in the complete Mack data triangle




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Answer

| Computation of Standard Error of Ultimate for AY 1984 | | | | | | | |
|---|--------|---------------------|------------------------------------|---|----------------------|-------------------------------|------------------------------------|
| d | $F(d)$ | $\text{Sigma } d^2$ | $\frac{\text{Sigma } d^2}{F(d)^2}$ | $E[c(1,d)]$ | $\text{SUM}[c(i,d)]$ | $E[c(1,d)] \times \text{SUM}$ | $\frac{\text{Sigma } d^2}{F(d)^2}$ |
| (1) | (2) | (3) | (3)/(2) ² | (5) | (6) | (7) | (8) |
| 4 | 1.147 | 161.2 | 122.6 | 22.306 | 11.805 | 0.0001295 | 0.02 |
| 3 | 1.378 | 2.887.3 | 1.519.8 | 16.183 | 16.303 | 0.0001251 | 0.19 |
| 2 | 1.401 | 161.2 | 82.2 | 11.555 | 21.546 | 0.0001330 | 0.01 |
| | | | | $U(w) =$ | 25,582 | $\text{Sum} =$ | 0.21 |
| | | | | $\text{Var}[R(w,d)^2] = U(w)^2 \times \text{Sum} =$ | 140,015,175 | $SE =$ | 11,833 |

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Mack Total Variance


- The total unpaid is the sum of the unpaid random variable for each accident year
- $R_{tot} = R_1 + R_2 + \dots + R_N$
- The Variance of a sum equals the sum of the Variances plus twice the co-variances


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Mack Total Variance

- The Mack formula for the Variance of the total unpaid estimate is:

$$SE(R_{tot})^2 = \sum_{w=2}^N \left\{ SE(R_w)^2 + U(w) \left(\sum_{r=w+1}^N c(i,r) \right) \sum_{d=r+1-w}^{r-1} \left(\frac{2\sigma_d^2 / F(d)^2}{\sum_{j=1}^{N-d} c(j,d)} \right) \right\}$$



Covariance term 

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Calculation Pointers



- It is easiest to set up a set of triangles to perform the calculations
 - First create a row of column sums of cumulative losses x the last observation
 - Create a triangle of weighted squared deviations of development factors from their mean
 - Create a projected runoff triangle that computes each estimate of cumulative losses, $E[c(w,d)]$, for all future periods
 - Create a triangle of inverses of projected runoff plus inverse of sum of cumulative losses
 - A spreadsheet showing the calculation for the Mack data is provided



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Using Mack Parameters

- We have a mean and a variance for unpaid (or IBNR) amounts. Now what?
- To get confidence intervals or probability distribution, assumptions must be made
- Assume unpaid (or IBNR) amounts follow a probability distribution, say the Gamma
- Use mean and variance of unpaid (or IBNR) amounts to derive parameters for distribution
- Use this distribution to estimate percentiles and other statistics for unpaid (or IBNR) amounts





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Group Exercise

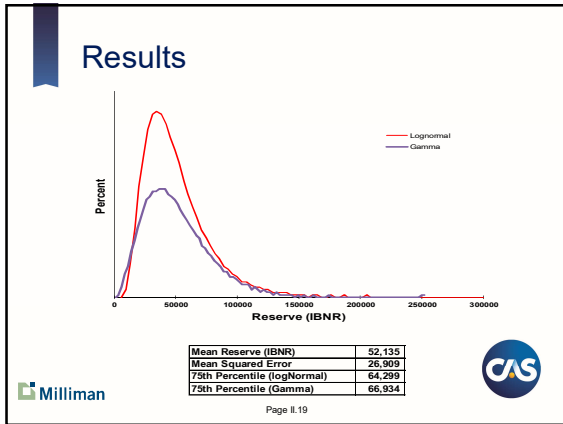
- Compute the variance of the total reserve amount using the Mack data
- Assume total reserve amount follows a lognormal (or Gamma) distribution and compute the parameters μ & σ . Compute the 75th percentile of the reserve (IBNR) amount.

Refer to Mack Model workbook for results



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- ### Open issues
- Covariance term for the oldest year
 - Multiply ultimate for year by
 1. Sum of ultimates for all subsequent years
 2. Times the factor variance (σ_x^2) for last age-to-age factor
 3. Divide by square of last age-to-age factor
 - For Other years, need a sum of the ratio computed in 2 and 3
 - Tail Factors? Recursion formula is useful
 - Assumption testing
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- ### Testing Assumption 1
- $$[1] E[c(w,d+1)|c(w,1), \dots, c(w,d)] = c(w,d)F(d)$$
- Graph two consecutive cumulative development periods.
 - Does it look like a linear relationship?
 - Does it look like the intercept is zero?
 - Fit a regression of the form

$$c(w,d+1) = \alpha + \beta c(w,d)$$
 - Test α for statistical significance
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Testing Assumption 1

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Testing Assumption 2

[2] $\{c(i,1), \dots, c(i,n)\}$ & $\{c(j,1), \dots, c(j,n)\}$ are independent for $i \neq j$

- Mack suggests that the independence assumption can be tested by searching for calendar year effects
- His test looks for the probability you would observe "s" large or small development factors (relative to the median) along a diagonal out of j factors

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Testing Assumption 2

| Development Factors | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
|---------------------|--------|-------|-------|-------|-------|-------|-------|-------|-------|
| 1 | 1.850 | 1.319 | 1.082 | 1.147 | 1.195 | 1.113 | 1.033 | 1.003 | 1.009 |
| 2 | 40.425 | 1.259 | 1.977 | 1.292 | 1.132 | 0.993 | 1.043 | 1.033 | |
| 3 | 2.637 | 1.543 | 1.163 | 1.161 | 1.186 | 1.029 | 1.026 | | |
| 4 | 2.943 | 1.364 | 1.349 | 1.102 | 1.113 | 1.029 | | | |
| 5 | 8.759 | 1.656 | 1.400 | 1.171 | 1.099 | | | | |
| 6 | 4.260 | 1.816 | 1.105 | 1.228 | | | | | |
| 7 | 7.217 | 2.723 | 1.125 | | | | | | |
| 8 | 5.142 | 1.887 | | | | | | | |
| 9 | 1.722 | | | | | | | | |
| Median | 4.260 | 1.599 | 1.163 | 1.166 | 1.132 | 1.033 | 1.033 | 1.018 | 1.009 |

Conclusion?



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Testing Assumption 3

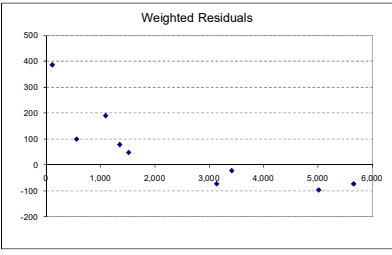
[3] $\text{Var}[c(w,d+1) | c(w,1), \dots, c(w,d)] = \alpha_w^2 \sigma^2$
for unknown proportionality constants α

- Graph weighted residuals v. cumulative losses
 $[c(w,d+1) - c(w,d)F(d)]/c(w,d)^{1/2}$ v. $c(w,d)$





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Testing Assumption 3





Conclusion?



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Some Tail Factor Methods

- Industry tail factor
- Judgmental selections
- Generalized Bondy
- Simple Inverse Power $F(d) = 1 + (b/d)^a$
 - Can use regression to fit





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Mack: Tail Variances

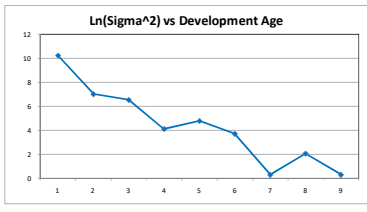
- Look at pattern of standard deviations or variances from triangle
- Judgmentally select SD (or Var) for development ages with no data
- If you use industry factors to get tail, variance should be larger than indicated by triangle, because entity specific tail may be different from industry (an additional uncertainty)





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Tail Variances

- Plot $\ln(\text{Sigma}^2)$ vs. development age
- If relationship looks linear, fit regression to get variances in tail





| Development Age | Ln(Sigma ²) |
|-----------------|-------------------------|
| 1 | 10.5 |
| 2 | 7.0 |
| 3 | 6.5 |
| 4 | 4.5 |
| 5 | 5.0 |
| 6 | 4.0 |
| 7 | 0.5 |
| 8 | 2.5 |
| 9 | 0.5 |



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Exercises



- Use judgment to select tail variances for the Mack data.
- Fit a regression to $\ln(\text{Sigma}^2)$ vs development age for the Mack data and use it to estimate the ages 9 and 10 variances.
- What is the variance of the next (age 10-11) factor?



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Questions on Mack Model?



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