


《STOCHASTIC LOSS RESERVING USING BAYESIAN MCMC MODELS》

模型及效果总结



一、数据说明



二、模型总览



三、模型效果

一、数据说明

- 【数据来源】

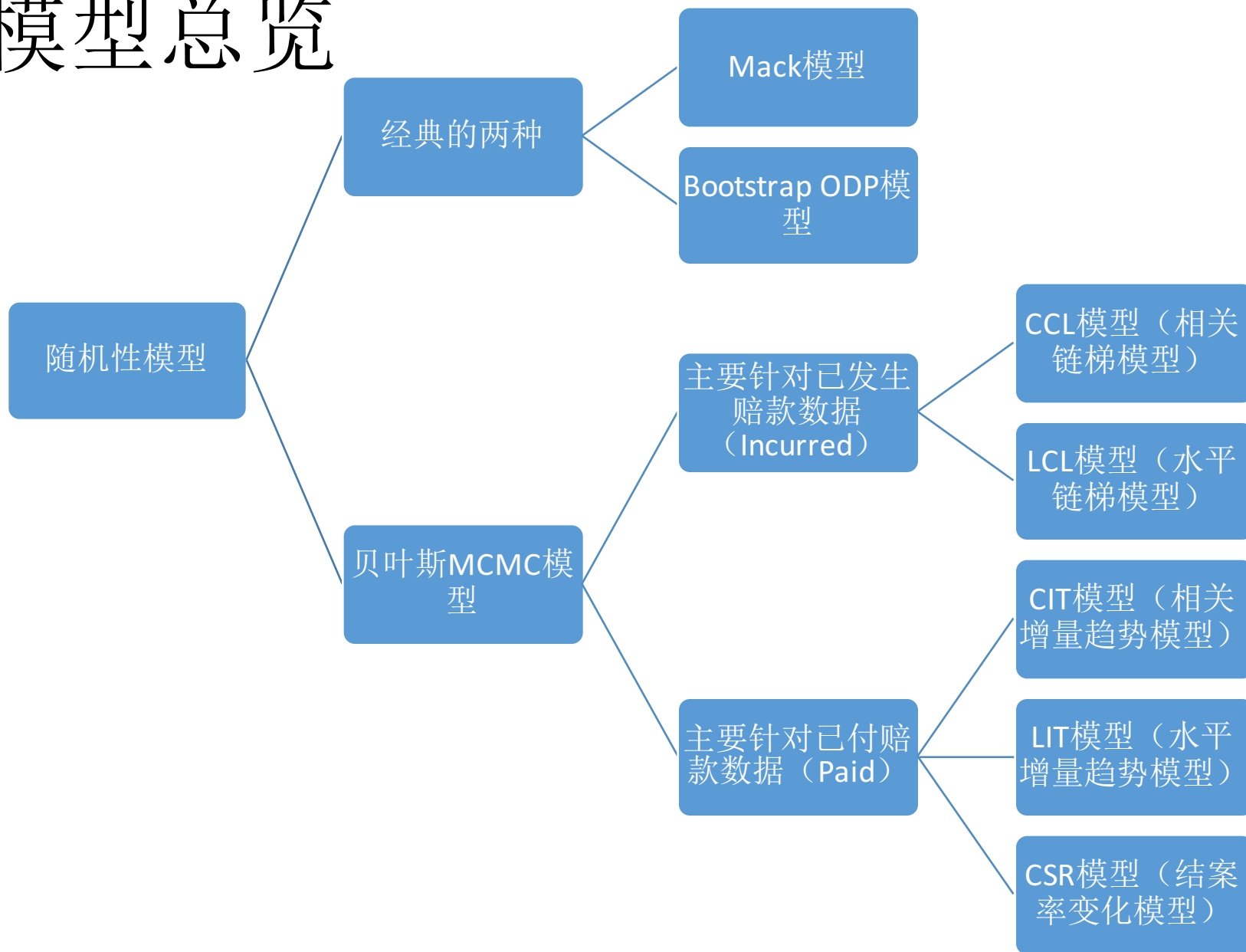
http://www.casact.org/research/index.cfm?fa=loss_reserves_data

- 【数据选取】

4个险种（Commercial Auto, Personal Auto, Workers Compensation, Other Liability）中，每个险种选择50家保险公司的数据。

（数据选取依据和过程详见文章Appendix A。）

二、模型总览



Mack模型 - Mack (1993, 1994)

- 【模型假设】

假设 $\tilde{C}_{w,d}$ 是随机变量:

1. $E[\tilde{C}_{w,d+1} | C_{w,1}, \dots, C_{w,d}] = C_{w,d} \cdot f_d$
2. For any given d , the random variables $\tilde{C}_{v,d}$ and $\tilde{C}_{w,d}$ are independent for $v \neq w$.
3. $\text{Var}[\tilde{C}_{w,d+1} | C_{w,1}, \dots, C_{w,d}] = C_{w,d} \cdot \alpha_d^2$

- 【参数估计】 The Mack estimate for $E[\tilde{C}_{w,K}]$ for $w = 2, \dots, K$ is given by

$$\hat{C}_{w,K} = C_{w,K+1-w} \cdot \hat{f}_{K+1-w} \cdot \dots \cdot \hat{f}_{K-1}$$

where

$$\hat{f}_d = \frac{\sum_{w=1}^{K-d} C_{w,d+1}}{\sum_{w=1}^{K-d} C_{w,d}}$$

Given his assumptions above, Mack then derives expressions for the standard deviations $\text{SD}[\tilde{C}_{w,K}]$ and $\text{SD}[\sum_{w=2}^K \tilde{C}_{w,K}]$

Bootstrap ODP模型 - England and Verrall (2002)

- 【模型假设】

A key assumption made by this model is that the incremental losses are described by the overdispersed Poisson distribution with

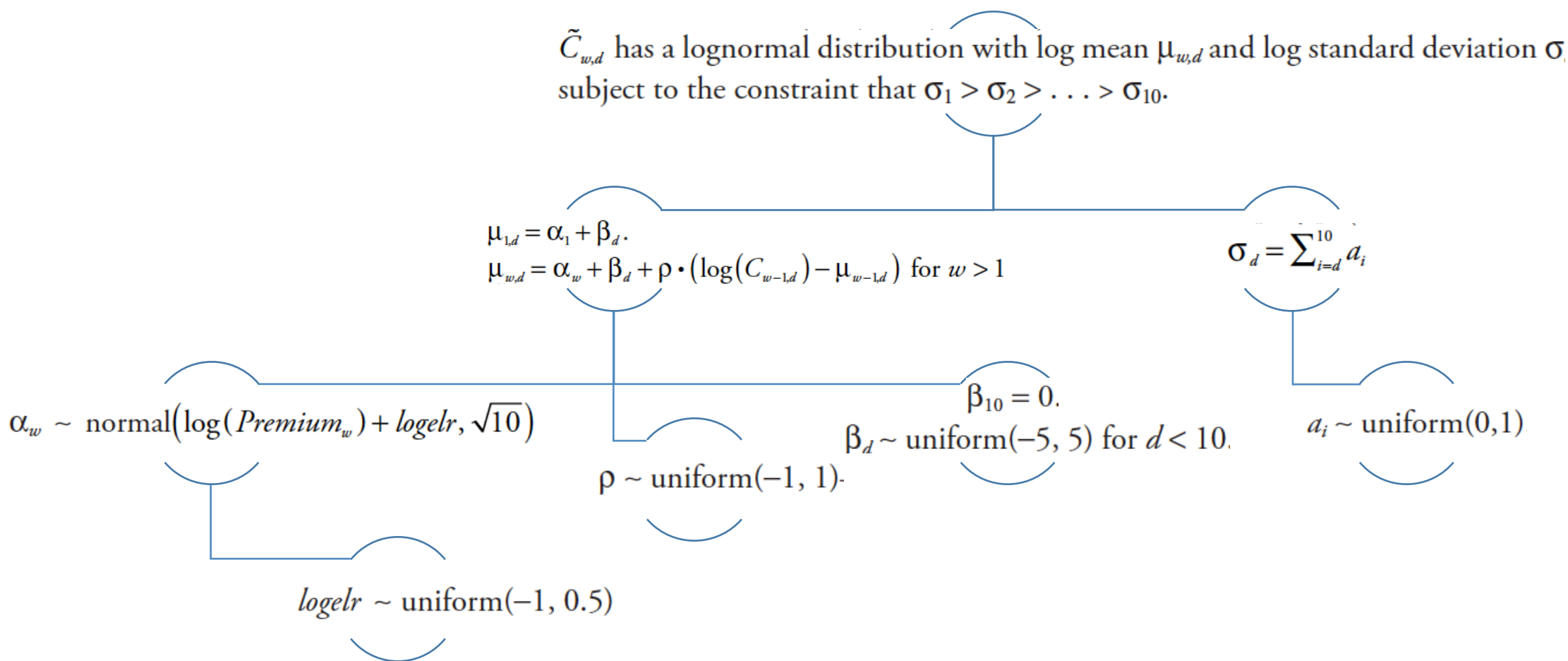
$$E[\tilde{I}_{w,d}] = \alpha_w \cdot \beta_d \quad \text{and} \quad Var[\tilde{I}_{w,d}] = \phi \cdot \alpha_w \cdot \beta_d$$

- 【参数估计】

The parameters of the model can be estimated by a standard generalized linear model (GLM) package.⁹ They then use a bootstrap resampling procedure to quantify the volatility of the estimate.

CCL模型（相关链梯模型）

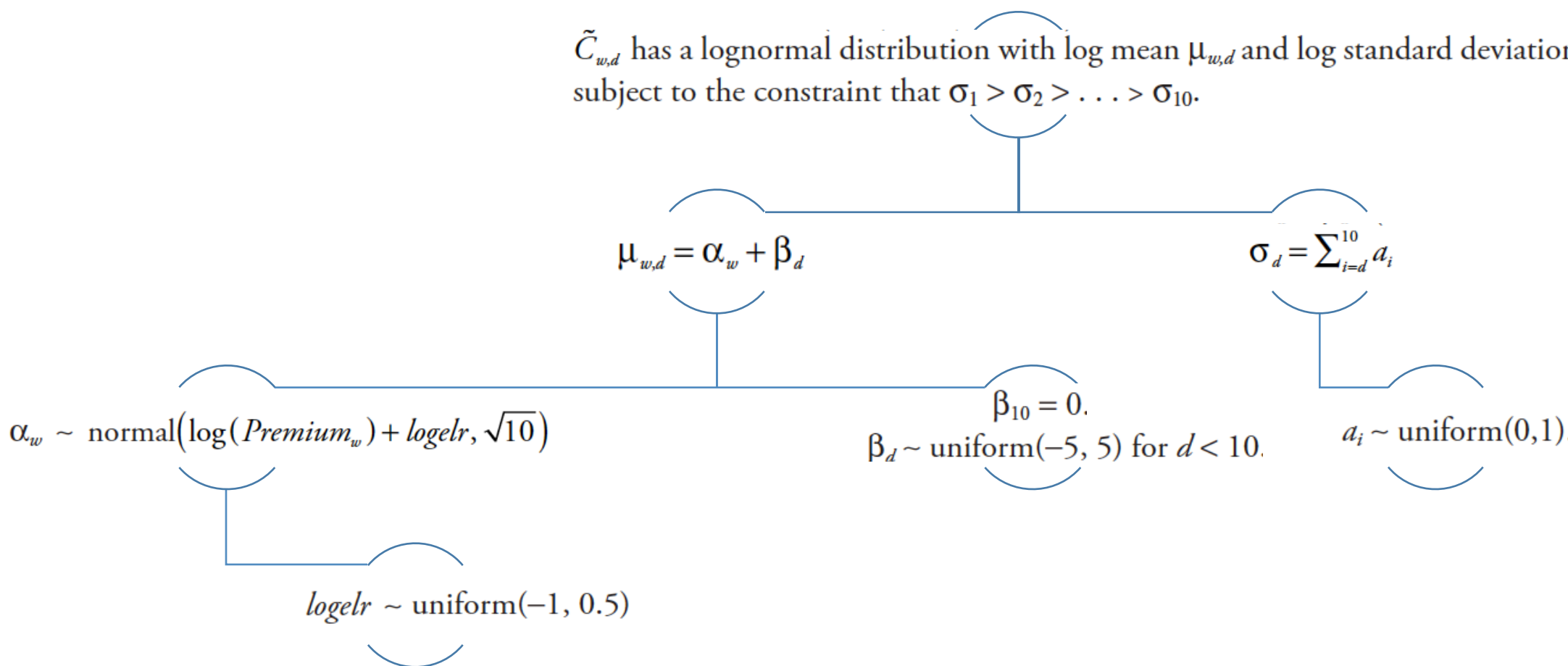
- 【模型假设】假设事故年之间的累计赔款是相关的。



LCL模型（水平链梯模型）

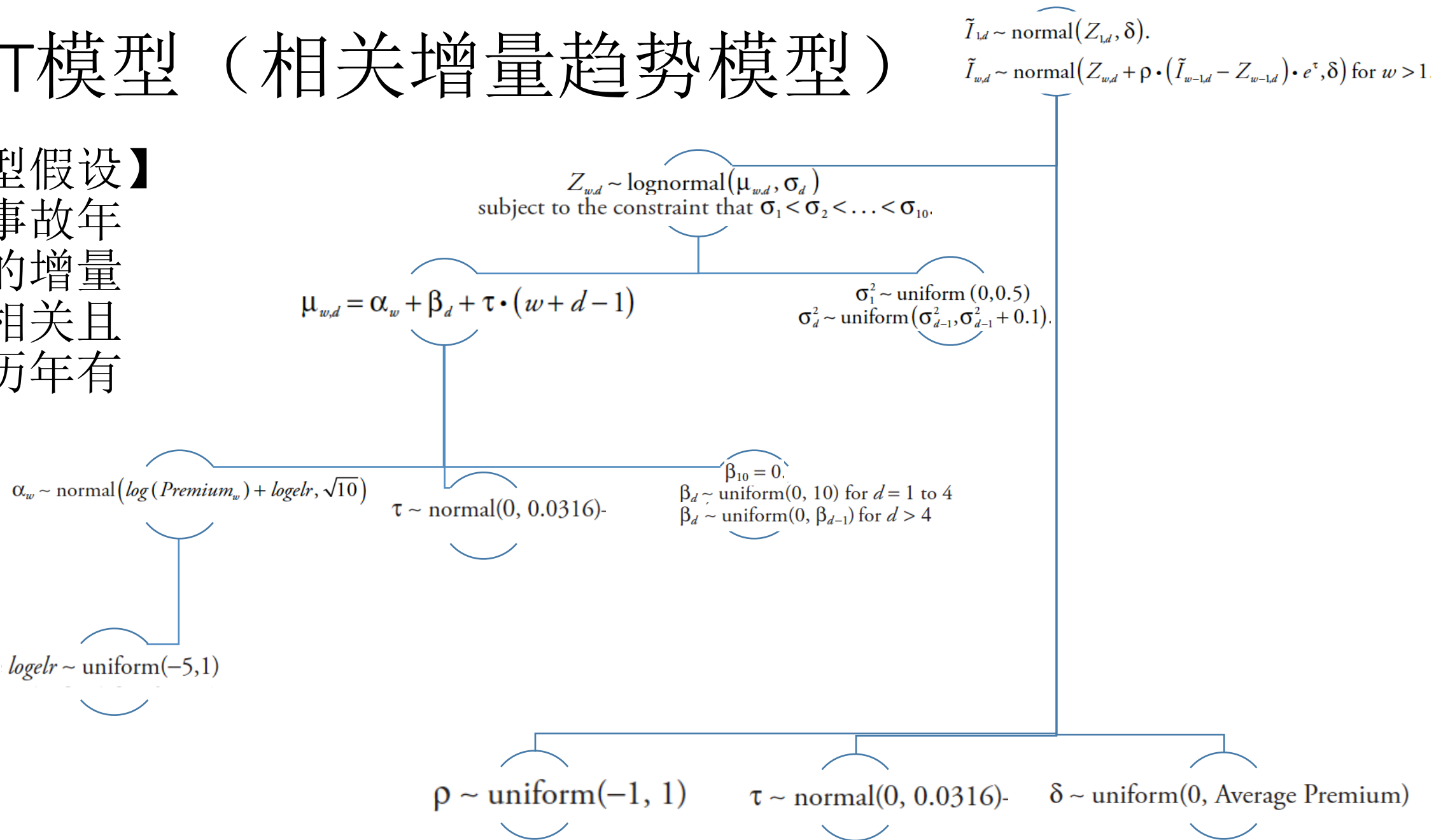
- 【模型假设】CCL模型的特例： $\rho = 0$

$\tilde{C}_{w,d}$ has a lognormal distribution with log mean $\mu_{w,d}$ and log standard deviation σ_d subject to the constraint that $\sigma_1 > \sigma_2 > \dots > \sigma_{10}$.



CIT模型（相关增量趋势模型）

- 【模型假设】
假设事故年之间的增量赔款相关且随日历年有趋势。

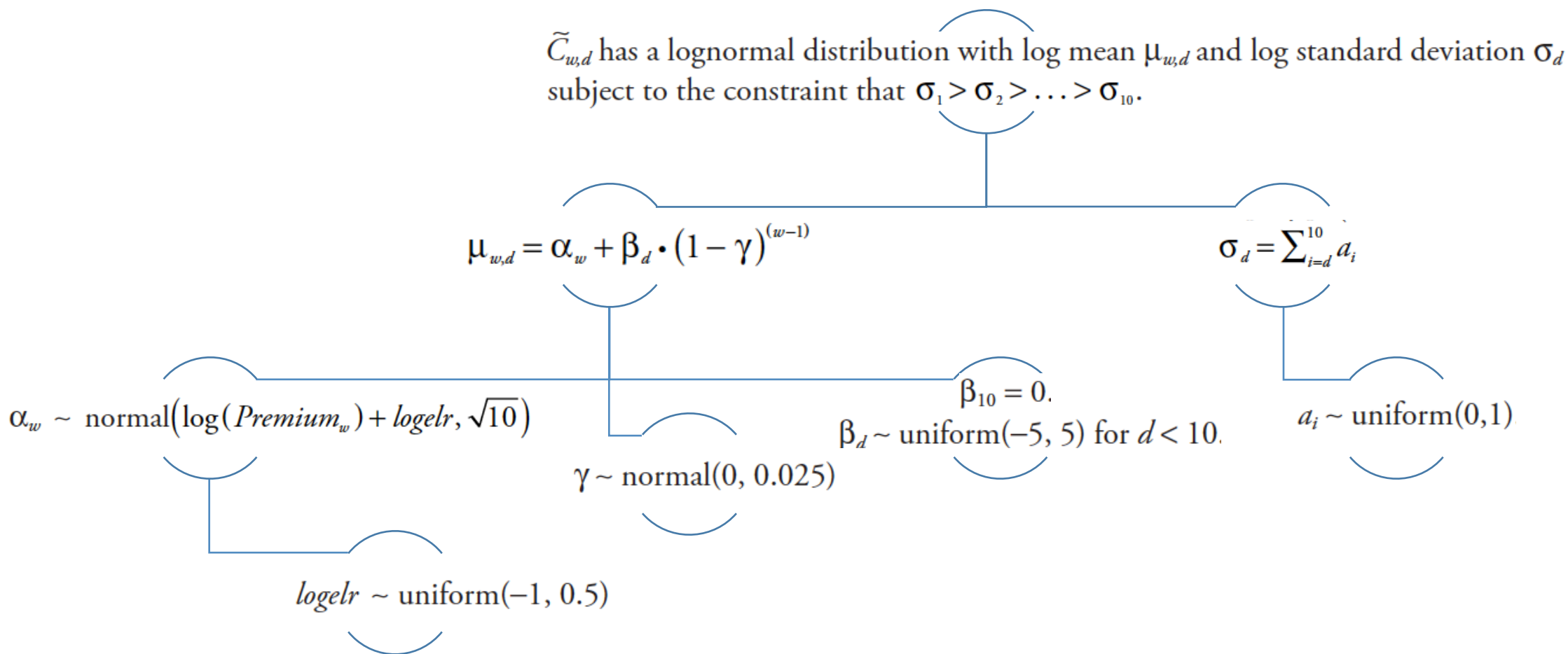


LIT模型（水平增量趋势模型）

- 【模型假设】CIT模型的特例： $\rho = 0$

CSR模型（结案率变化模型）

- 【模型假设】考虑案件处理速度的逐年变化。



三、模型效果

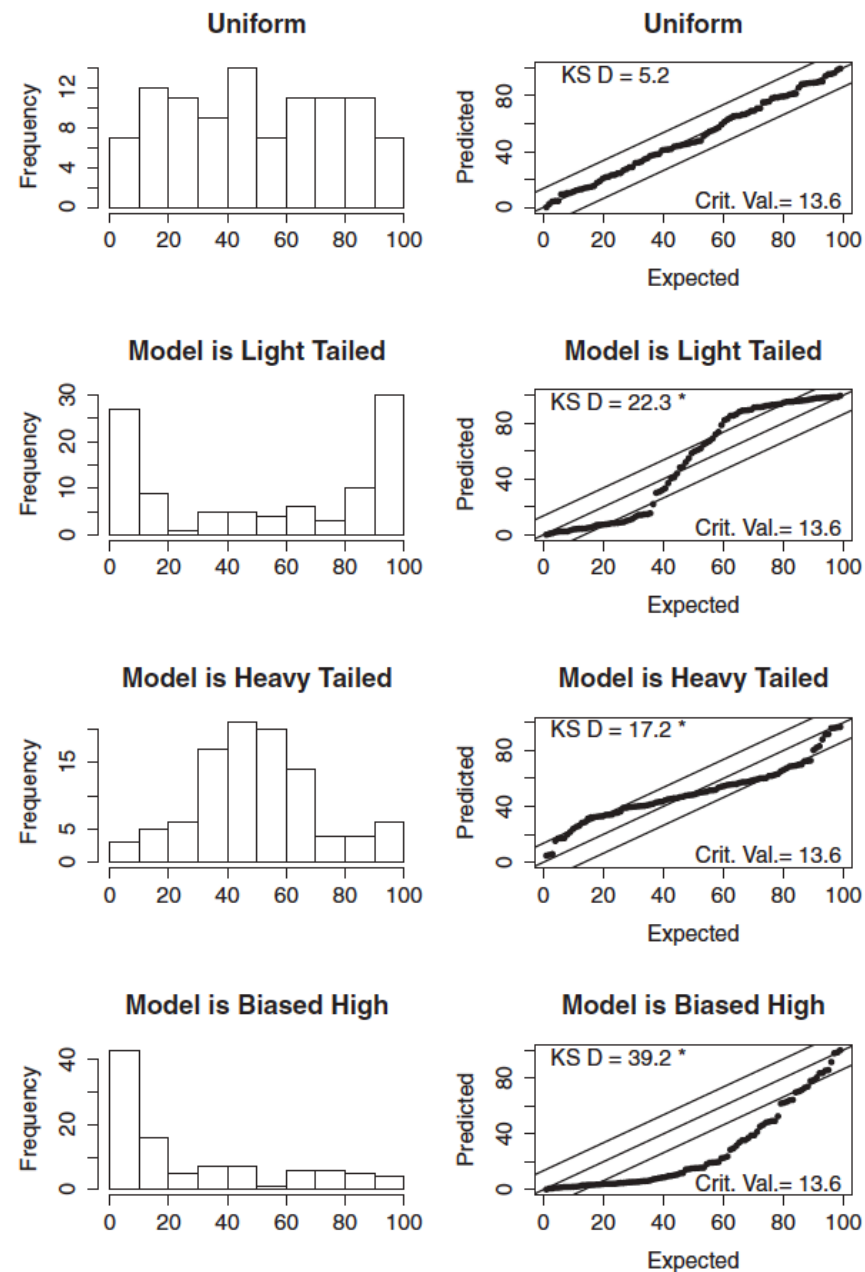
- 【评价标准】

假设 $\sum_{w=1}^{10} C_{w,10}$ 服从对数正态分布，均值和标准差分别为通过模型得到的 $\sum_{w=1}^{10} \tilde{C}_{w,10}$ 的均值和标准差。

计算原始数据中 $\sum_{w=1}^{10} C_{w,10}$ 在该对数正态分布中的百分位数（percentile）。

如果所有选取样本得到的百分位数的分布接近均匀分布，则说明模型效果较好。

Figure 1. *p-p* Plots Test for Uniformity



For Incurred Loss Data: $CCL > LCL > Mack$

Figure 2. p - p Plots for the Mack Model on Incurred Loss Triangles

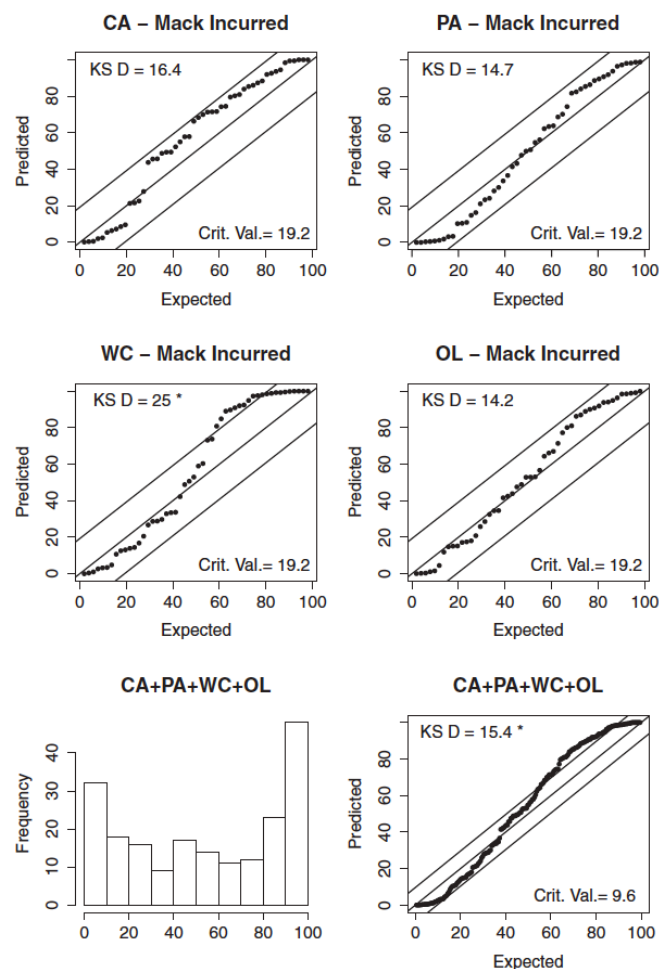


Figure 8. p - p Plots for the LCL Model on the Incurred Loss Triangles

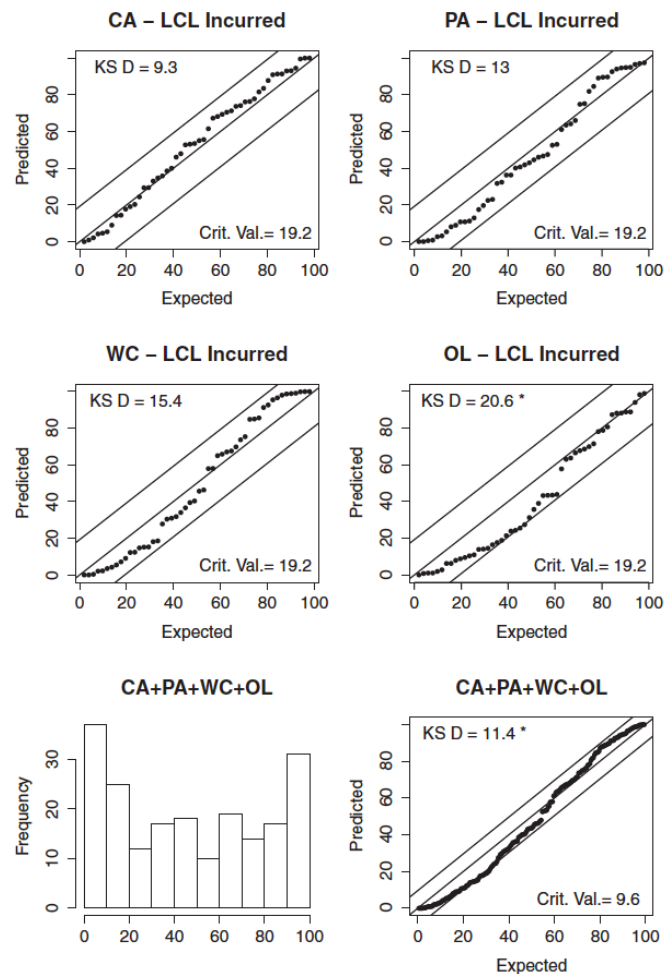
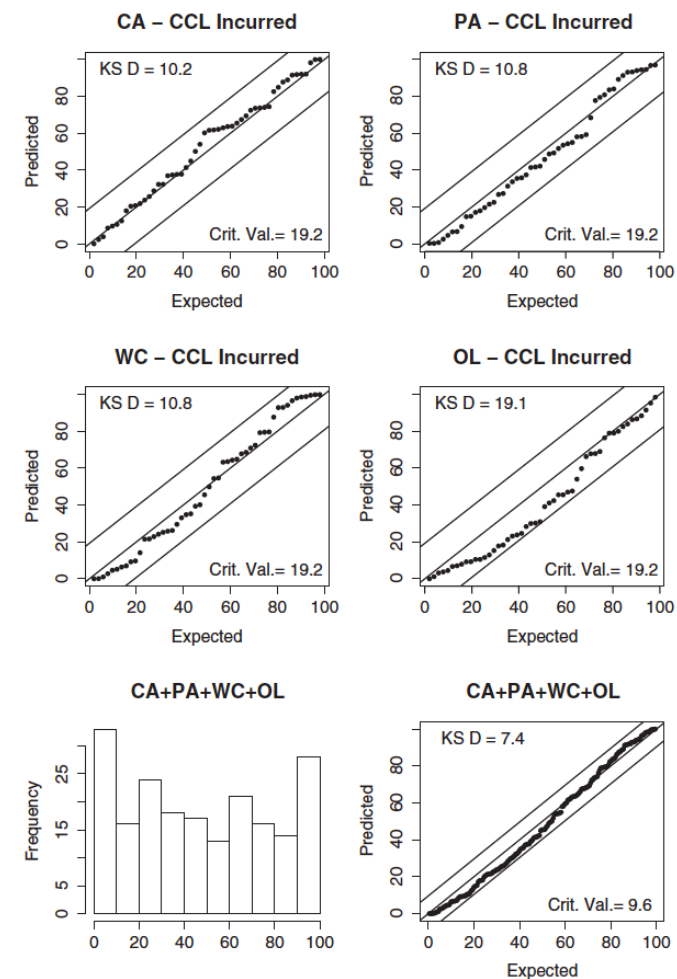


Figure 9. p - p Plots for the CCL Model on the Incurred Loss Triangles



For Paid Loss Data: $CSR > CIT \approx LIT > \text{Bootstrap ODP} > \text{Mack}$

Figure 4. p - p Plots for the Mack Model on Paid Loss Triangles

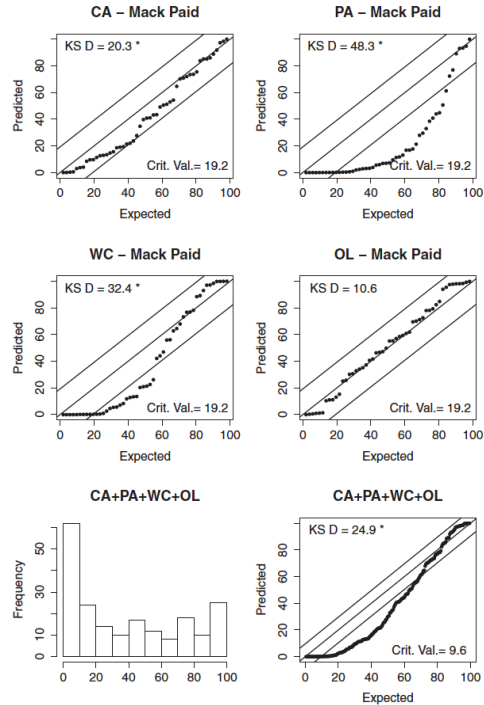


Figure 3. p - p Plots for the Bootstrap ODP Model on Paid Loss Triangles

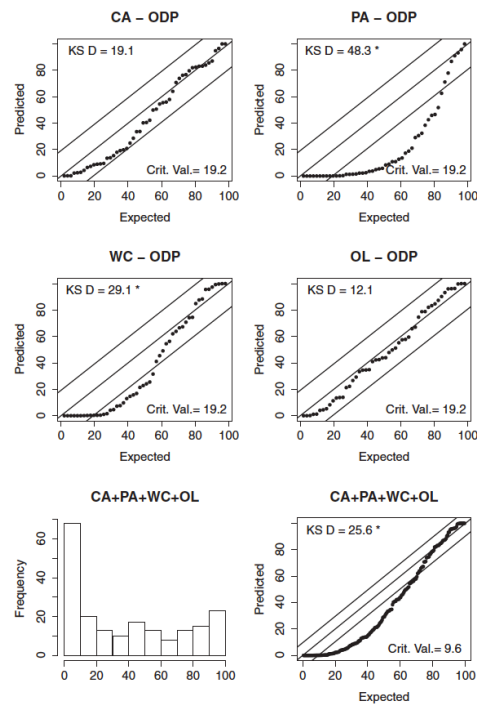


Figure 20. p - p Plots for the LIT Model

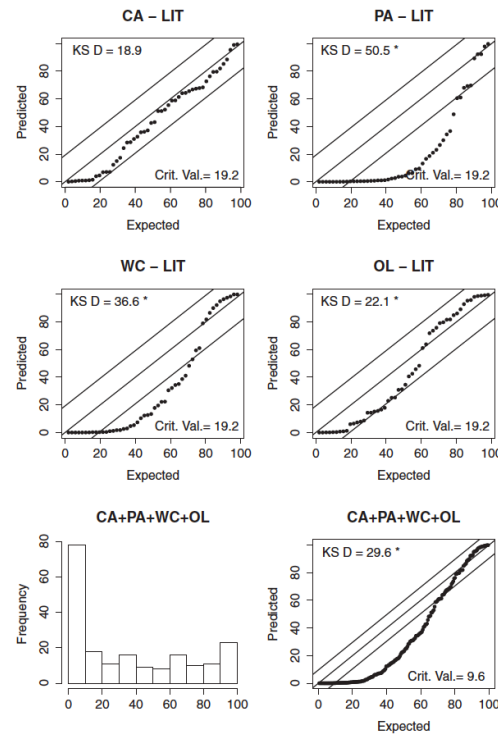


Figure 19. p - p Plots for the CIT Model

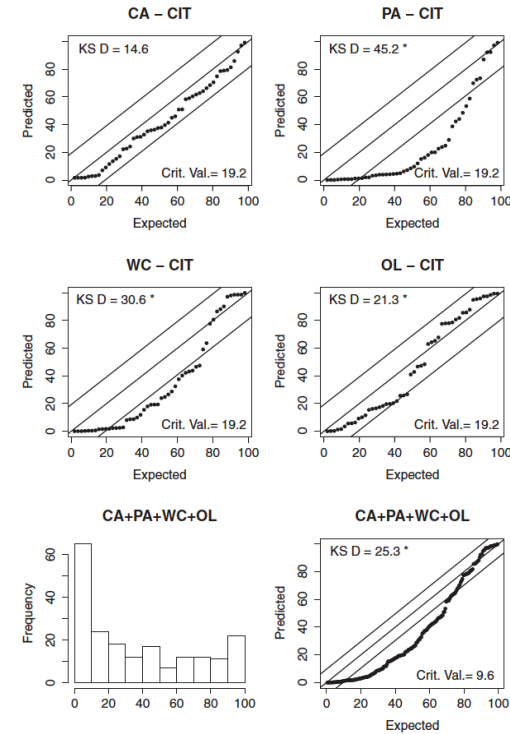


Figure 22. p - p Plots for the CSR Model

