Modeling Paid and Incurred Losses Together

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Abstract

The modeling skills of actuaries and academicians have developed to the point of their seeking joint models for paid and incurred losses, i.e., models in which paid and incurred losses will inform each other so that their confidence intervals will narrow and the two sets of ultimate losses will be equal. The key to such models is covariance; heteroskedastic models cannot serve the purpose. Properly accounting for covariance in the linear statistical model will provide an exact, sound, and elegant solution to the problem. Moreover, covariance is what distinguishes the same information from like information, and prevents the creation of information out of nothing.

Key concepts: linear statistical model, paid and incurred losses, seemingly unrelated regression (SUR), covariance, variance structure

1. INTRODUCTION

For setting the loss reserves of most casualty lines of insurance, actuaries must turn loss triangles into rectangles. Until the mid-1990s this was largely a deterministic exercise, which involved selecting development factors – perhaps with certain adjustments and sensitivity testing. Since then, business needs have demanded, and advances in theory and computing have allowed, probabilistic modeling of loss triangles. Deterministic methods are waning as actuaries are increasingly asked to estimate statistical properties of loss reserves, especially their probability distributions. However, attempts to answer this need are hampered by the duality (sometimes the multiplicity) of triangles for the same line of business. Triangles usually come in pairs, one of paid losses and another of (case-) incurred losses, which ultimately must reach equality. But when paid and incurred losses are modeled separately, any equality is accidental, and even then devoid of jointly statistical properties. A crasis of models, however artful, is not science, despite appeals to actuarial judgment.¹

According to Gary Venter [2008, 348], "Formal modeling of paid and incurred simultaneously appears to have begun with Halliwell." But the idea received scant attention until a paper by Quarg and Mack [2004 and 2008], which spurred the Venter paper, as well as a yet unpublished paper by Zhang and Clark [2009]. Unlike the Quarg and Mack approach, which "reduces the gap" between the projections, the Halliwell [1997] approach offered an exact

¹ Actuarial judgment functions *within* actuarial science, not to the transcendence or out-guessing of science. "Actuarial judgment is no antidote ..., as if actuaries possessed some expertise or intuition to herd or prod methods into correctness. ... Actuaries must not presume to judge what they cannot scientifically model." [Halliwell, 2007, footnote 5]

solution. Moreover, Quarg and Mack sought the solution in the design matrix of their model, whereas the variance structure was the key for Halliwell. We believe that Halliwell was on the right track, but with some flaws. Our solution to the problem of modeling paid and incurred losses together improves on his variance approach, correcting its flaws. Furthermore, it is easier to understand, simpler to program, as well as theoretically streamlined and elegant.

In the next section we will introduce and comment upon Halliwell's version of the linear statistical model. In Section 3, while discussing basic properties of the model, we will introduce the distinction between statistical sameness and statistical similarity, which arises from differing variance structures and which ensures that information cannot be created *ex nibilo*. Section 4 will apply this theory to a simple example of a joint model of paid and incurred loss triangles, and Section 5 will show a more elaborate application to industry Workers' Compensation losses. Appendix A will deepen insights into the ideas of Section 3 by treating simultaneous equations as a subset of the linear statistical model. Finally, Appendix B will elaborate on permissible variance structures, viz., how with two random variables of known correlation a third may be correlated.

2. GENERAL FORMULATION OF THE LINEAR STATISTICAL MODEL

Many writers present versions of the linear statistical model; however, the version found in Halliwell [1997], despite its initial complexity, is most versatile and general. Moreover, most presentations stop at the estimation of the parameter β . But in his version this is just an intermediate step; the focus is on predictions based on β and their prediction-error variances. The basic form of the linear model is $\mathbf{y}_{t\times 1} = X_{t\times k}\beta_{k\times 1} + \mathbf{e}_{t\times 1}$. It is the error term \mathbf{e} that makes the model statistical (otherwise called probabilistic and stochastic); it is a random vector whose moments are: $E[\mathbf{e}] = \mathbf{0}_{t\times 1}, Var[\mathbf{e}] = \Sigma_{t\times t} = \sigma^2 \Phi_{t\times t}$. For those unfamiliar with multivariate means and variances, especially with the quadratic form $Var[A\mathbf{e}] = AVar[\mathbf{e}]A'$ and with non-negative definite and positive definite matrices, the reader is advised to read Halliwell [1997; Appendix A], Healy [1986; Chapter 7], and Judge [1988, Appendix A].

The matrix X is known as the design, or regressor, matrix; its columns are called regressor variables and independent variables. The vector **y** is called the response variable and the dependent variable, and β is the parameter of the model. But in this formulation, which emphases predictions rather than parameters, the *t* rows of the model are partitioned into t_1 observations and t_2 predictions. In matrix-partitioned form it is expressed:

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$$\begin{bmatrix} \mathbf{y}_{1(t_{1}\times 1)} \\ \mathbf{y}_{2(t_{2}\times 1)} \end{bmatrix} = \begin{bmatrix} X_{1(t_{1}\times k)} \\ \overline{X}_{2(t_{2}\times k)} \end{bmatrix} \beta_{k\times 1} + \begin{bmatrix} \mathbf{e}_{1(t_{1}\times 1)} \\ \mathbf{e}_{2(t_{2}\times 1)} \end{bmatrix},$$

where:

$$E\left[\frac{\mathbf{e}_{1(t_{1}\times 1)}}{\mathbf{e}_{2(t_{2}\times 1)}}\right] = \left[\frac{\mathbf{0}_{t_{1}\times 1}}{\mathbf{0}_{t_{2}\times 1}}\right], \quad Var\left[\frac{\mathbf{e}_{1(t_{1}\times 1)}}{\mathbf{e}_{2(t_{2}\times 1)}}\right] = \left[\frac{\Sigma_{11(t_{1}\times t_{1})}}{\Sigma_{21(t_{2}\times t_{1})}} \left|\frac{\Sigma_{12(t_{1}\times t_{2})}}{\Sigma_{22(t_{2}\times t_{2})}}\right] = \sigma^{2}\left[\frac{\Phi_{11(t_{1}\times t_{1})}}{\Phi_{21(t_{2}\times t_{1})}} \left|\frac{\Phi_{12(t_{1}\times t_{2})}}{\Phi_{22(t_{2}\times t_{2})}}\right]\right]$$

The noun 'observations' is imprecise. What distinguishes observations from predictions is that predictions are missing, or blank, elements of \mathbf{y} (which signals that predictions are desired), whereas "observations" are non-missing, real-valued elements. Usually, they happen to come from observation, such as with loss amounts. However, the key to a joint model of paid and incurred losses is what we will call "tautologous observations," i.e., elements of \mathbf{y} that contain zeroes as the differences by exposure period of incurred ultimate losses from paid ultimate losses. We do not actually need to observe their ultimate equality to know that it will obtain; this is information that we know *a priori* and of which we should make use. The observed and predicted elements can appear in any order; the clearest presentation may not place all the observations in rows above those of the predictions; however, our software will reorder them. Of course, since the variance structure is symmetric, the columns of $Var[\mathbf{e}]$ must likewise be reordered.

The known, or specified, elements of the linear model are the entire² design matrix *X*, the entire variance structure, whether in absolute form Σ or in relative form Φ , and the observed elements of **y**. The modeler desires an estimate of \mathbf{y}_2 , viz., $\hat{\mathbf{y}}_2$. However, that estimate will turn out to be in error by the vector $\mathbf{y}_2 - \hat{\mathbf{y}}_2$, which we will call the prediction error. The formulæ for an estimator of \mathbf{y}_2 and the variance of its prediction error, viz., $\hat{\mathbf{y}}_2$ and $Var[\mathbf{y}_2 - \hat{\mathbf{y}}_2]$, are:

$$\hat{\mathbf{y}}_2 = X_2 \hat{\boldsymbol{\beta}} + \Phi_{21} \Phi_{11}^{-1} \left(\mathbf{y}_1 - X_1 \hat{\boldsymbol{\beta}} \right)$$

$$Var[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}] = (X_{2} - \Phi_{21}\Phi_{11}^{-1}X_{1})Var[\hat{\boldsymbol{\beta}}](X_{2} - \Phi_{21}\Phi_{11}^{-1}X_{1})' + \sigma^{2}(\Phi_{22} - \Phi_{21}\Phi_{11}^{-1}\Phi_{12}),$$

where $\hat{\boldsymbol{\beta}} = (X_1' \Phi_{11}^{-1} X_1)^{-1} X_1' \Phi_{11}^{-1} \mathbf{y}_1$ and $Var[\hat{\boldsymbol{\beta}}] = \sigma^2 (X_1' \Phi_{11}^{-1} X_1)^{-1}$. Or one may use the absolute-variance form, in which ' σ^2 ' is omitted and ' Σ ' replaces ' Φ '. This estimator algebraically

² This conflicts with many linear models of loss triangles, in particular with the Quarg/Mack [2004 and 2008] model, whose predictions two or more periods into the future depend on predictions of the previous periods. The feedback loop 'predictions \rightarrow regressors \rightarrow predictions', which Judge [1988; Chapter 13] calls "Stochastic Regressors," is undesirable for theoretical, numerical-analytic, and æsthetic reasons, some of which Halliwell [2007; Section 5] discusses.

reduces to a linear function of \mathbf{y}_1 , and Halliwell [1997; Appendix C] gives a version of the Gauss-Markov theorem in proof that $\hat{\mathbf{y}}_2$ is the best linear unbiased estimator (BLUE) of \mathbf{y}_2 .³

A few minor conditions need to be made explicit. First, the variance structure (Σ or Φ) must be non-negative definite. Otherwise some random variable consisting of a linear combination of the elements of **e** would have a negative variance. At the very least this implies that that none of the diagonal elements of the variance structure is negative. Second, Σ_{11} or Φ_{11} must be positive definite. Being a block-diagonal part of the variance structure, it must be non-negative definite. However, a positive definite Σ_{11} or Φ_{11} has no variance degeneracy, which guarantees the existence of its inverse. Third, X_1 must be of full column rank, i.e., rank(X_1) = k. The second and third conditions together guarantee that $(X'_1 \Phi_{11}^{-1} X_1)^{-1}$ exists.

Usually the variance structure is known only to within a scale factor; most models posit relative, not absolute, variances. In that case one must estimate σ^2 as $\hat{\sigma}^2 = (\mathbf{y}_1 - X_1\hat{\beta}) \Phi_{11}^{-1} (\mathbf{y}_1 - X_1\hat{\beta}) / (t_1 - k)$, a matrix-weighted "sum of squared residuals" divided by the degrees of freedom. Our software will display a 3×3 matrix titled "SSCP," which stands for "(matrix-weighted) sums of squares and cross products." Define the $t_1 \times 3$ partitioned matrix $V = [\mathbf{y}_1 \mid X_1\hat{\beta} \mid \mathbf{y}_1 - X_1\hat{\beta}]$. Then SSCP = $V \Phi_{11}^{-1} V$, and $\hat{\sigma}^2 = \text{SSCP}_{33} / (t_1 - k)$. Theorems of the linear model state that $\text{SSCP}_{11} = \text{SSCP}_{22} + \text{SSCP}_{33}$ and $\text{SSCP}_{32} = \text{SSCP}_{23} = 0$. SSCP22/SSCP11 is the portion of the observations that the model "explains."⁴

3. BASIC PROPERTIES OF THE LINEAR STATISTICAL MODEL

Before introducing the joint model of paid and incurred losses we must discuss some basic properties of the model and its predictions. First consider the model:

$$\begin{bmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{bmatrix} = \begin{bmatrix} X_1 \\ I_{k \times k} \end{bmatrix} \beta + \begin{bmatrix} \mathbf{e}_1 \\ 0 \end{bmatrix}, \quad Var \begin{bmatrix} \mathbf{e}_1 \\ 0 \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & 0 \\ 0 & 0 \end{bmatrix}$$

One here wants just to predict the parameter ($y_2 = \beta$), so this is a check of the parameter of

³ This estimator is unbiased in that $E[\hat{\mathbf{y}}_2] = \mathbf{y}_2$. There are infinitely many linear-in- \mathbf{y}_1 , unbiased estimators (LUE) of

 $[\]mathbf{y}_2$. But according to the Gauss-Markov theorem, none of them is as good as $\hat{\mathbf{y}}_2$; $\hat{\mathbf{y}}_2$ is the best (B). This means that its prediction-error variance is less than theirs, in the sense that the difference of its prediction-error variance from theirs is positive definite. To the philosophically inclined it is amazing for an estimator to exist that positive-definitely dominates in the linear unbiased universe. It is no less amazing that this BLUE estimator is identical to the maximum-likelihood estimator under the assumption that the error terms are multivariate-normally distributed – an assumption not necessary to the linear statistical model. Such feelings of amazement incline most mathematicians toward a Platonic belief that mathematics is discovered, rather than toward a formalist belief that it is invented. Then again, in this abstruse realm what might be the difference between discovery and invention?

⁴ Though we will call this a rho-square statistic, it differs from the commonly defined statistic that was devised for regression models containing an intercept and that strips away the explanatory power of the intercept. Our definition is appropriate to our intercept-free models.

the model. Applying the formulæ, we confirm:

$$\begin{aligned} \hat{\mathbf{y}}_{2} &= X_{2}\hat{\boldsymbol{\beta}} + \Sigma_{21}\Sigma_{11}^{-1} \left(\mathbf{y}_{1} - X_{1}\hat{\boldsymbol{\beta}} \right) \\ &= I\,\hat{\boldsymbol{\beta}} + 0\Sigma_{11}^{-1} \left(\mathbf{y}_{1} - X_{1}\hat{\boldsymbol{\beta}} \right) \\ &= \hat{\boldsymbol{\beta}} \end{aligned}$$

$$\begin{aligned} Var[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}] &= \left(X_{2} - \Sigma_{21}\Sigma_{11}^{-1}X_{1} \right) Var[\hat{\boldsymbol{\beta}}] \left(X_{2} - \Sigma_{21}\Sigma_{11}^{-1}X_{1} \right)' + \left(\Sigma_{22} - \Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12} \right) \\ &= \left(I - 0\Sigma_{11}^{-1}X_{1} \right) Var[\hat{\boldsymbol{\beta}}] \left(I - 0\Sigma_{11}^{-1}X_{1} \right)' + \left(0 - 0\Sigma_{11}^{-1}0 \right) \\ &= Var[\hat{\boldsymbol{\beta}}] \end{aligned}$$

In fact, Halliwell [1997; 331] began with a "wordier" version of the solution and derived in this manner the streamlined version of the solution that employs $\hat{\beta}$.

The second model is:

$$\begin{bmatrix} A\mathbf{y}_1 \\ -\mathbf{y}_2 \end{bmatrix} = \begin{bmatrix} AX_1 \\ -\mathbf{x}_2 \end{bmatrix} \beta + \begin{bmatrix} A\mathbf{e}_1 \\ -\mathbf{e}_2 \end{bmatrix}, \quad Var\begin{bmatrix} A\mathbf{e}_1 \\ -\mathbf{e}_2 \end{bmatrix} = \begin{bmatrix} A\Sigma_{11}A' & A\Sigma_{12} \\ -\Sigma_{21}A' & \Sigma_{22} \end{bmatrix}$$

The observed part of this model has been transformed by matrix A. The transformation affects even the error term, $e_1 \rightarrow Ae_1$, and one purpose of this exercise is to sensitize the reader to the variance structure. For in general, $Cov[Ae_1, Be_2] = ACov[e_1, e_2]B'$. If A is a nonsingular, or invertible, matrix:

$$\hat{\boldsymbol{\beta}} = \left(\left(AX_{1} \right)' \left(A\Sigma_{11}A' \right)^{-1} \left(AX_{1} \right) \right)^{-1} \left(AX_{1} \right)' \left(A\Sigma_{11}A' \right)^{-1} \left(A\mathbf{y}_{1} \right) \\ = \left(X_{1}'A'A'^{-1}\Sigma_{11}^{-1}A^{-1}AX_{1} \right)^{-1} X_{1}'A'A'^{-1}\Sigma_{11}^{-1}A^{-1}A\mathbf{y}_{1} \\ = \left(X_{1}'\Sigma_{11}^{-1}X_{1} \right)^{-1} X_{1}'\Sigma_{11}^{-1}\mathbf{y}_{1} \\ = Var [\hat{\boldsymbol{\beta}}] X_{1}'\Sigma_{11}^{-1}\mathbf{y}_{1}$$

At this point we've demonstrated that the parameter estimate is unaffected. A similar cancellation of *A* with its inverse continues into the rest of the prediction, as the reader can verify. Hence, a one-to-one transformation of the observations has no effect on the predictions. A corollary to this is that incremental or cumulative models of loss triangles will yield the same predictions, as long as the variance structure is correctly handled.

Now consider a transformation of the predictions:

$$\begin{bmatrix} \mathbf{y}_1 \\ \overline{B}\mathbf{y}_2 \end{bmatrix} = \begin{bmatrix} X_1 \\ \overline{B}X_2 \end{bmatrix} \beta + \begin{bmatrix} \mathbf{e}_1 \\ \overline{B}\mathbf{e}_2 \end{bmatrix}, \quad Var\begin{bmatrix} \mathbf{e}_1 \\ \overline{B}\mathbf{e}_2 \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & \Sigma_{12}B' \\ \overline{B}\Sigma_{21} & \overline{B}\Sigma_{22}B' \end{bmatrix}$$

In this case:

$$\hat{By}_{2} = BX_{2}\hat{\beta} + B\Sigma_{21}\Sigma_{11}^{-1}(\mathbf{y}_{1} - X_{1}\hat{\beta})$$

$$= B\{X_{2}\hat{\beta} + \Sigma_{21}\Sigma_{11}^{-1}(\mathbf{y}_{1} - X_{1}\hat{\beta})\}$$

$$= B\hat{\mathbf{y}}_{2}$$

$$Var\left[B\mathbf{y}_{2} - \hat{B\mathbf{y}}_{2}\right] = \left(BX_{2} - B\Sigma_{21}\Sigma_{11}^{-1}X_{1}\right)Var\left[\hat{\beta}\right]\left(BX_{2} - B\Sigma_{21}\Sigma_{11}^{-1}X_{1}\right)' + \left(B\Sigma_{22}B' - B\Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12}B'\right)$$

$$= B\left\{\left(X_{2} - \Sigma_{21}\Sigma_{11}^{-1}X_{1}\right)Var\left[\hat{\beta}\right]\left(X_{2} - \Sigma_{21}\Sigma_{11}^{-1}X_{1}\right)' + \left(\Sigma_{22} - \Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12}\right)\right\}B'$$

$$= BVar\left[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}\right]B'$$

So the BLUE of a linear combination of predictions is the linear combination of the BLUE of the predictions. Note that here B, unlike the previous A, does not have to be invertible. The most common linear combinations of loss-triangle cells are exposure-period subtotals, i.e., unpaid, IBNR, or even ultimate losses. When we care only for these subtotals, as in the Workers' Compensation model of Section 5, this theorem allows us to bypass the extra time and space of cell-by-cell prediction and to predict them directly.

A special pair of models that will illustrate the effect of covariance is the following:

$$\begin{bmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \end{bmatrix} = \begin{bmatrix} X_1 \\ \overline{X}_1 \end{bmatrix} \boldsymbol{\beta} + \begin{bmatrix} \mathbf{e}_1 \\ \mathbf{e}_2 \end{bmatrix}, \quad Var\begin{bmatrix} \mathbf{e}_1 \\ \mathbf{e}_2 \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & | & \mathbf{0} \\ -\mathbf{0} & | & \Sigma_{11} \end{bmatrix}$$

Here the prediction is *like* the observation, the solution being:

$$\begin{aligned} \hat{\mathbf{y}}_{2} &= X_{1}\hat{\boldsymbol{\beta}} + \Sigma_{21}\Sigma_{11}^{-1} \left(\mathbf{y}_{1} - X_{1}\hat{\boldsymbol{\beta}} \right) \\ &= X_{1}\hat{\boldsymbol{\beta}} + 0\Sigma_{11}^{-1} \left(\mathbf{y}_{1} - X_{1}\hat{\boldsymbol{\beta}} \right) \\ &= X_{1}\hat{\boldsymbol{\beta}} \end{aligned}$$

$$\begin{aligned} Var[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}] &= \left(X_{1} - \Sigma_{21}\Sigma_{11}^{-1}X_{1} \right) Var[\hat{\boldsymbol{\beta}}] \left(X_{1} - \Sigma_{21}\Sigma_{11}^{-1}X_{1} \right)' + \left(\Sigma_{22} - \Sigma_{21}\Sigma_{11}^{-1}\Sigma_{12} \right) \\ &= \left(X_{1} - 0\Sigma_{11}^{-1}X_{1} \right) Var[\hat{\boldsymbol{\beta}}] \left(X_{1} - 0\Sigma_{11}^{-1}X_{1} \right)' + \left(\Sigma_{11} - 0\Sigma_{11}^{-1}0 \right) \\ &= X_{1} Var[\hat{\boldsymbol{\beta}}] X_{1}' + \Sigma_{11} \end{aligned}$$

The prediction is *like*, because the variance of \mathbf{e}_2 is like that of \mathbf{e}_1 , 'like' in the sense of identically distributed, but nonetheless uncovaried. But changing the variance structure to $Var\left[\frac{\mathbf{e}_1}{\mathbf{e}_2}\right] = \left[\frac{\sum_{11} \frac{1}{2} \sum_{11} \sum_{11}$

$$\hat{\mathbf{y}}_{2} = X_{1}\hat{\mathbf{\beta}} + \Sigma_{21}\Sigma_{11}^{-1}(\mathbf{y}_{1} - X_{1}\hat{\mathbf{\beta}})$$

$$= X_{1}\hat{\mathbf{\beta}} + \Sigma_{11}\Sigma_{11}^{-1}(\mathbf{y}_{1} - X_{1}\hat{\mathbf{\beta}})$$

$$= X_{1}\hat{\mathbf{\beta}} + (\mathbf{y}_{1} - X_{1}\hat{\mathbf{\beta}})$$

$$= \mathbf{y}_{1}$$

$$Var[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}] = (X_{1} - \Sigma_{11}\Sigma_{11}^{-1}X_{1})Var[\hat{\mathbf{\beta}}](X_{1} - \Sigma_{11}\Sigma_{11}^{-1}X_{1})' + (\Sigma_{11} - \Sigma_{11}\Sigma_{11}^{-1}\Sigma_{11})$$

$$= (X_{1} - X_{1})Var[\hat{\mathbf{\beta}}](X_{1} - X_{1})' + (\Sigma_{11} - \Sigma_{11})$$

$$= 0$$

This version predicts not something *like* \mathbf{y}_1 , but rather something *identical* to \mathbf{y}_1 . Covariance is the key to distinguishing between likeness and sameness.⁵

And finally, we consider pooling two models. For this purpose and for here only, the subscripts '1' and '2' identify the models (e.g., paid and incurred), rather than observations and predictions (here dismissed with a dot '•'). We might be tempted to solve the "super" model:

$$\begin{bmatrix} \mathbf{y}_1 \\ \mathbf{y}_2 \\ \mathbf{\bullet} \end{bmatrix} = \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \\ \mathbf{\bullet} & \mathbf{\bullet} \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} \mathbf{e}_1 \\ \mathbf{e}_2 \\ \mathbf{\bullet} \end{bmatrix}, \quad Var \begin{bmatrix} \mathbf{e}_1 \\ \mathbf{e}_2 \\ \mathbf{\bullet} \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & 0 & | \mathbf{\bullet} \\ 0 & \Sigma_{22} & \mathbf{\bullet} \\ \mathbf{\bullet} & \mathbf{\bullet} & | \mathbf{\bullet} \end{bmatrix}$$

However, the solution for the parameter is:

transformation. In matrix algebra, $Var\begin{bmatrix} \mathbf{e}\\ A\mathbf{e} \end{bmatrix} = \begin{bmatrix} \Sigma & \Sigma A'\\ A\Sigma & A\Sigma A' \end{bmatrix} = \begin{bmatrix} I\\ A \end{bmatrix} \Sigma \begin{bmatrix} I & A' \end{bmatrix}$, and the rank of this larger matrix is

⁵ Appendix A shows covariance to be the solution to the "paradox" of why writing data twice does not increase information. It would increase, if it were *like*, or similar, information; just as repeated sampling of independent, identically-distributed random variables increases information. But repetitions of the *same* information are perfectly correlated with the original, and provide nothing new, not even when the repetition is disguised by a linear $\begin{bmatrix} e \end{bmatrix} \begin{bmatrix} \Sigma & \Sigma A' \end{bmatrix} \begin{bmatrix} I \end{bmatrix} r$

still equal to the rank of Σ . If the off-block-diagonal elements were zero the rank would increase, depending on A, to as much as twice the rank of Σ .

$$\begin{aligned} \operatorname{Var}\begin{bmatrix} \hat{\boldsymbol{\beta}}_{1} \\ \boldsymbol{\beta}_{2} \end{bmatrix} &= \left(\begin{bmatrix} X_{1} & 0 \\ 0 & X_{2} \end{bmatrix}^{\prime} \begin{bmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22} \end{bmatrix}^{-1} \begin{bmatrix} X_{1} & 0 \\ 0 & X_{2} \end{bmatrix} \right)^{-1} \\ &= \left(\begin{bmatrix} X_{1}^{\prime} & 0 \\ 0 & X_{2}^{\prime} \end{bmatrix}^{} \begin{bmatrix} \Sigma_{11}^{-1} & 0 \\ 0 & \Sigma_{22}^{-1} \end{bmatrix} \begin{bmatrix} X_{1} & 0 \\ 0 & X_{2} \end{bmatrix} \right)^{-1} \\ &= \left[\begin{pmatrix} X_{1}^{\prime} \Sigma_{11}^{-1} X_{1} & 0 \\ 0 & X_{2}^{\prime} \Sigma_{22}^{-1} X_{2} \end{bmatrix} \right)^{-1} \\ &= \begin{bmatrix} (X_{1}^{\prime} \Sigma_{11}^{-1} X_{1})^{-1} & 0 \\ 0 & (X_{2}^{\prime} \Sigma_{22}^{-1} X_{2})^{-1} \end{bmatrix} \\ &= \begin{bmatrix} \operatorname{Var}[\hat{\boldsymbol{\beta}}_{1}] & 0 \\ 0 & \operatorname{Var}[\hat{\boldsymbol{\beta}}_{2}] \end{bmatrix} \\ &= \begin{bmatrix} \operatorname{Var}[\hat{\boldsymbol{\beta}}_{1}] & 0 \\ 0 & \operatorname{Var}[\hat{\boldsymbol{\beta}}_{2}] \end{bmatrix} \\ &= \begin{bmatrix} \operatorname{Var}[\hat{\boldsymbol{\beta}}_{1}] & 0 \\ 0 & \operatorname{Var}[\hat{\boldsymbol{\beta}}_{2}] \end{bmatrix} \left(\begin{bmatrix} X_{1} & 0 \\ X_{2}^{\prime} \Sigma_{22}^{-1} \mathbf{y}_{2} \end{bmatrix} \right) \\ &= \begin{bmatrix} \operatorname{Var}[\hat{\boldsymbol{\beta}}_{1}] & 0 \\ 0 & \operatorname{Var}[\hat{\boldsymbol{\beta}}_{2}] X_{2}^{\prime} \Sigma_{22}^{-1} \mathbf{y}_{2} \end{bmatrix} \\ &= \begin{bmatrix} \operatorname{Var}[\hat{\boldsymbol{\beta}}_{1}] X_{2}^{\prime} \Sigma_{22}^{-1} \mathbf{y}_{2} \end{bmatrix} \\ &= \begin{bmatrix} \hat{\boldsymbol{\beta}}_{1} \\ \hat{\boldsymbol{\beta}}_{2} \end{bmatrix} \end{aligned}$$

Hence, simply juxtaposing two (or more) models provides no additional information.⁶ There is no covariance between the two; they are "frictionless" and "like ships passing in the night." But if the off-block-diagonal variance were not zero, the combined model would not reduce to the separate submodels. Judge [1988; Section 11.2] calls covariance-linked models "seemingly unrelated regression" (SUR) models, for they seem to be unrelated if one considers only the model design and ignores the variance structure. The Zhang/Clark [2009] model is an SUR model, tying paid and incurred losses together with covariance; but it does not guarantee the equality of ultimate paid and incurred. The Halliwell model [1997] also qualifies as SUR. The model that we will present in the next section is not an SUR one; rather, tautologous equations

⁶ However, one caveat: The combined model in relative-variance format would invite the modeler to estimate an overall variance scale $\hat{\sigma}^2$, which would be an average of the scale estimates of the submodels weighted according to their respective degrees of freedom.

will be the glue between paid and incurred, and the only tricky part will be the effect of these additional equations on the variance structure.

4. TAUTOLOGY AND A SIMPLE JOINT PAID-INCURRED MODEL

The following example comes from Halliwell [1997; Exhibit 1]. It consists of incremental paid and incurred losses for three accident years at three years of development, development being complete at the third year:

		Paid		Incurred				
	@1	@2	@3	@1	@2	@3		
AY1	50	30	20	75	15	10		
AY2	60	25		75	25			
AY3	45			50				

The AY exposures are equal, and the cells are homoskedastic. The first accident year is mature, and both paid and incurred losses accumulate to 100. The paid and incurred models have the same design matrix (viz., additive, cf. Halliwell [2007; 228]), and the parameter elements are pure premiums by loss type and by age or development period. All cells have the same variance relativity and zero covariance. In symbols, the model of the *hij*th cell is $\mathbf{y}_{hij} = 1 \cdot \beta_{hj} + \mathbf{e}_{hij}$, $Var[\mathbf{e}_{hij}] = \sigma^2 \cdot 1$, where $h \in \{1 = \text{Paid}, 2 = \text{Incurred}\}$, $i \in \{1 = \text{AY1}, 2 = \text{AY2}, 3 = \text{AY3}\}$, and $j \in \{1 = \text{Age1}, 2 = \text{Age2}, 3 = \text{Age3}\}$. If the accident years were not of equal exposure, the model would be $\mathbf{y}_{hij} = \xi_i \beta_{hj} + \mathbf{e}_{hij}$, $Var[\mathbf{e}_{hij}] = \sigma^2 \xi_i$, where ξ_i is the exposure of the *i*th accident year. In this simple, juxtaposed model one would estimate paid development at ages 2 and 3 as 27.5 and 20, and incurred development likewise as 20 and 10. The paid ultimate losses of AY2 and AY3 would be 95 and 92.5, as compared with incurred ultimate losses of 110 and 80.

Exhibit 1 contains the joint model in the standard form:

$$\begin{bmatrix} \mathbf{y}_{1(14\times 1)} \\ \overline{\mathbf{y}_{2(6\times 1)}} \end{bmatrix} = \begin{bmatrix} X_{1(14\times 6)} \\ \overline{X}_{2(6\times 6)} \end{bmatrix} \boldsymbol{\beta}_{6\times 1} + \begin{bmatrix} \mathbf{e}_{1(14\times 1)} \\ \overline{\mathbf{e}_{2(6\times 1)}} \end{bmatrix}, \quad Var\begin{bmatrix} \mathbf{e}_{1} \\ \overline{\mathbf{e}_{2}} \end{bmatrix} = \sigma^{2} \begin{bmatrix} \Phi_{11(14\times 14)} & | & \Phi_{12(14\times 6)} \\ \overline{\Phi_{21(6\times 14)}} & | & \Phi_{22(6\times 6)} \end{bmatrix}$$

The reader will recognize the **y**, *X*, and Φ matrices within the exhibit, and the dotted lines partition the matrices into observations and predictions. Zeroes are present, but not shown. Except for rows 13 and 14, marked 'Diff', and corresponding columns 13 and 14 of Φ , the model would consist of two, unrelated paid and incurred homoskedastic submodels. The '1's in the design matrix represent the unitary exposure slotted into the hj^{th} column of *X* so as to interact

with the hj^{th} element of the model parameter β (implicit in the exhibit, but not shown).

But it is the rows marked 'Diff' that join, or laminate, the paid and the incurred losses. Although we have not observed six of the eighteen cells, we do know that total paid must equal total incurred by AY. For AY2 the difference is:

$$0 = \mathbf{y}_{121} + \mathbf{y}_{122} + \mathbf{y}_{123} - \mathbf{y}_{221} - \mathbf{y}_{222} - \mathbf{y}_{223}$$

= $\beta_{11} + \beta_{12} + \beta_{13} - \beta_{21} - \beta_{22} - \beta_{23} + \{\mathbf{e}_{121} + \mathbf{e}_{122} + \mathbf{e}_{123} - \mathbf{e}_{221} - \mathbf{e}_{222} - \mathbf{e}_{223}\}$
= $\begin{bmatrix} 1 & 1 & 1 & -1 & -1 & -1 \end{bmatrix} \beta + \{\mathbf{e}_{\text{Diff2}}\}$

So this tautology, or difference equation, is equivalent to a zero observation, the design $\begin{bmatrix} 1 & 1 & -1 & -1 & -1 \end{bmatrix}$, and error term that involves six other error terms. Because of heteroskedasticity, the variance of this error term is 6. However, because these are *same* error terms, not *like* ones, $Cov[e_{Diff2}, \pm e_{hij}] = \pm Var[e_{hij}]$. So the reader should now understand row 13 of the model, and similarly, row 14, the tautology for AY3. We could have added a tautology for AY1; however, its variance structure involves no predictions. It would add no new observation, and the resulting 15×15 matrix Φ_{11} would still be of rank 14 (see footnote 5) and thus non-invertible.

In Exhibit 2 we solve for the estimates of β and σ , and derive the SSCP matrix, as explained in Section 2. In the spreadsheet we simplified the notation by dropping subscripts (which are all '1', pertaining to observations) and carets. Both paid and incurred total pure premiums equal 97.916, which is the average of the stand-alone total pure premiums of paid 99.16 and incurred 96.6.

The predictions are estimated in Exhibit 3 according to the formulæ:

$$\hat{\mathbf{y}}_{2} = X_{2}\hat{\boldsymbol{\beta}} + \Phi_{21}\Phi_{11}^{-1}(\mathbf{y}_{1} - X_{1}\hat{\boldsymbol{\beta}})$$

$$Var[\mathbf{y}_{2} - \hat{\mathbf{y}}_{2}] = (X_{2} - \Phi_{21}\Phi_{11}^{-1}X_{1})Var[\hat{\boldsymbol{\beta}}](X_{2} - \Phi_{21}\Phi_{11}^{-1}X_{1})' + \sigma^{2}(\Phi_{22} - \Phi_{21}\Phi_{11}^{-1}\Phi_{12})$$

To help the reader who wishes to reproduce the results, in the bottom half of the exhibit are intermediate calculations. The prediction and the variance of the prediction error are:

Туре	AY	Age	$\hat{\mathbf{y}}_2$	$Var[\mathbf{y}_{2}]$	$\left[-\hat{\mathbf{y}}_{2}\right]$				
Paid	2	3	22.50	79.95	0	39.97	79.95	0	39.97
Paid	3	2	23.75	0	89.94	-29.98	0	29.98	29.98
Paid	3	3	17.50	39.97	-29.98	109.93	39.97	29.98	49.97
Incd	2	3	7.50	79.95	0	39.97	79.95	0	39.97
Incd	3	2	23.75	0	29.98	29.98	0	89.94	-29.98
Incd	3	3	12.50	39.97	29.98	49.97	39.97	-29.98	109.93

In the topmost figure of the exhibit AY subtotals are formed and combined with the paid and incurred to date. For example, the calculation of the prediction-error standard deviation of the

AY3 IBNR is $11.83 = \sqrt{89.94 - 29.98 - 29.98 + 109.93}$. Although it may seem a wonder that the means and variances of the ultimate losses are identical whether one builds them from the paid to date or from the incurred, the model was constructed for this purpose. The identity serves only to confirm that the model was solved without mistake.

This technique is more general than the tautology of 0 = Paid - Incd. In addition to the observations \mathbf{y}_1 of the model $\mathbf{y} = X\beta + \mathbf{e}$, one may know by means other than observation that $\mathbf{z}_{t_3 \times 1} = Q_{t_3 \times t} \mathbf{y}_{t \times 1}$. Hence:

$$Var[\mathbf{z}] = Var[Q\mathbf{e}] = Q\Sigma Q'$$

$$Cov[\mathbf{z}, \mathbf{e}_{1}] = Cov[Q\mathbf{e}, \mathbf{e}_{1}] = QCov[\mathbf{e}, \mathbf{e}_{1}] = QCov\left[\begin{bmatrix}\mathbf{e}_{1}\\\mathbf{e}_{2}\end{bmatrix}, \mathbf{e}_{1}\right] = Q\begin{bmatrix}Cov[\mathbf{e}_{1}, \mathbf{e}_{1}]\\Cov[\mathbf{e}_{2}, \mathbf{e}_{1}\end{bmatrix} = Q\begin{bmatrix}\Sigma_{11}\\\Sigma_{21}\end{bmatrix}$$

$$Cov[\mathbf{z}, \mathbf{e}_{2}] = Cov[Q\mathbf{e}, \mathbf{e}_{2}] = \dots = Q\begin{bmatrix}\Sigma_{12}\\\Sigma_{22}\end{bmatrix}$$

Augmenting the model for the t_3 new observations, we arrive at the form:

 $\mathbf{z} = O(X\beta + \mathbf{e}) = OX\beta + O\mathbf{e}$

$$\begin{bmatrix} \mathbf{y}_{1} \\ \mathbf{z}_{1} \\ \mathbf{z}_{2} \end{bmatrix} = \begin{bmatrix} X_{1} \\ QX \\ -\overline{X}_{2} \end{bmatrix} \beta + \begin{bmatrix} \mathbf{e}_{1} \\ Q\mathbf{e} \\ -\mathbf{e}_{2} \end{bmatrix}, \quad Var \begin{bmatrix} \mathbf{e}_{1} \\ Q\mathbf{e} \\ -\mathbf{e}_{2} \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & [\Sigma_{11} & \Sigma_{12}]Q' & \Sigma_{12} \\ Q\begin{bmatrix} \Sigma_{11} \\ \Sigma_{21} \end{bmatrix} & Q\SigmaQ' & Q\begin{bmatrix} \Sigma_{12} \\ \Sigma_{22} \end{bmatrix} \\ -\overline{\Sigma}_{21} & \overline{\Sigma}_{21} & \overline{\Sigma}_{22} Q' & \overline{\Sigma}_{22} \end{bmatrix}$$

From the juxtaposed model the simple joint model will arise according to this form, if z is zero and:

It is a powerful extension of the linear statistical model; nevertheless, the same two questions inform every statistical model: "What is the equation for each row?" and "How does each row covary with itself and the other rows?"⁷ Next we will apply the joint model to ten accident years

⁷ Halliwell's solution [1997; Exhibit 14] differs from ours only in the prediction error variance, and there only because of a disagreement over the estimate of 2 : his 106.597 versus our 79.948. And this is due to a difference of degrees of freedom, his six versus our eight (106.597/79.948 = 8/6). Halliwell [1997; 247-249] both constrained the variance

of industry Workers' Compensation losses.

5. TAUTOLOGY AND A JOINT WORKERS' COMPENSATION MODEL

Exhibit 4 contains net paid and case-incurred (incurred less bulk and IBNR) triangles for U.S. Worker's Compensation, along with net earned premium and ultimate loss. Ignoring the prior-to-1998 line, we have ten accident years at ten evaluations. But we will project beyond the tenth evaluation (@120 months) to ultimate, our models assuming that at 120 months paid losses are 85.5% of ultimate and incurred are 95.0%. So our model will work with two 10×11 rectangles, each with fifty-five observations. However, to save space we will not predict each future cell, but only the unpaid and IBNR totals by AY. Thus, each of the paid and incurred submodels will have fifty-five observations and ten predictions.

It is not necessary to do so, but we will use the same design matrix for both submodels, an additive design with pure premiums by age (cf. Halliwell [2007; Section 7 and Exhibits 7A and 7B]). Because we have no exposure data, net earned premium will have to suffice. However, due to the underwriting cycle, we must adjust it to a constant loss ratio. Without rate-change information, we needed to develop the triangles to ultimate with standard deterministic methods. It's not desirable, perhaps it even smacks of cheating; but frequently it's a necessary evil, and its circularity does not seem to be vicious. The adjusted, or on-level, premium summary appears in Exhibit 5. The "Selected" losses are the simple average of the booked ultimates and four development methods. The overall loss ratio is 71%, and "Adj Prem" is simply the selected losses divided by this loss ratio, which conserves total premium. Since premium is the exposure base, our pure-premium betas are actually loss ratios.

Recognizing that the volatility, or unit-variance, of the incremental losses varies by age, we must also derive variances, or at least variance relativities, for a heteroskedastic model. We use the additive method on adjusted premium in Exhibits 6. First we derive pure premiums by age in Exhibit 6.1. The pure premium from 120 months to ultimate is calculated as:

$$\beta_{\text{ult}} = \left(\frac{1}{m} - 1\right) \left(\beta_{12} + \ldots + \beta_{120}\right)$$

where m, the maturity at 120 months, is 0.855 for paid and 0.950 for incurred. We assume that variance is proportional to exposure, and the "Selected" rows of Exhibit 6.2 derive the paid and incurred unit variances with some judgmental smoothing and extrapolation. The selected unit variances are then multiplied in Exhibit 6.3 by the adjusted premiums. We will treat these

structure and imposed constraints on β . This seems to have double-counted some observations and reduced the degrees of freedom. Our approach is more easily understood, stays closer to the empirical data, and does not require eigendecompostion of variance matrices. Moreover, it will not disturb the variance relativities of non-homoskedastic models.

as absolute variances, and the "Unpaid" and "IBNR" columns contain the sums of the variances of AY predictions.

Exhibits 7.1 and 7.2 present the separate paid and incurred models. As mentioned, they share the same design matrix. The column marked ' Σ ' contains the homoskedastic variances; the column header ' Σ ', rather than ' Φ ', signals our modeling software to take these as absolute variances. Each model has sixty-six rows, fifty-five observations, ten predictions of AY totals, and one row marked 'Constraint'. Since there is no observation beyond 120 months, the constraint (note its zero variance) allows for the estimation of β_{ult} . In keeping with the two maturities, each constraint is $0 = (1-m)(\beta_{12} + ... + \beta_{120}) - m\beta_{ult}$.

Part of the joint paid-incurred model appears in Exhibit 8.1. A new column 'Type' identifies the paid and incurred submodels, which one can recognize in the block diagonal form. To these one hundred thirty-two rows were added the tautologous observations, 'Ult =' for each accident year. The negative exposure in the incurred half of these ten rows indicates that the zero \mathbf{v} values are the difference of incurred from paid (paid minus incurred). The variance ' Σ ' of each new row is the sum of all the paid and incurred variances of its AY. However, the variance structure of the model at this point is the ' Σ ' column distributed down the main diagonal of a 142×142 matrix. Variance matrices, though large (sometimes exceeding the limit of 256 columns of a Excel spreadsheet) are often sparse. Our software allows us to specify only the columns of non-zero covariance, and to treat the rest of the covariance as zero. In Exhibit 8.2 we have specified how the last ten rows, the tautologous observations covary with all 142 rows. Each column (e.g., the 1998 'Covariance' column) records the ' Σ ' values of its AY, except that 'Type' = 'Incd' rows must be negated, since incurred rows are subtracted in the tautology. Let Cbe the 142×10 covariance matrix of this exhibit. The software knows to insert C into the last ten columns and to insert C' into the last ten rows of the 142×142 variance structure. This completes the joining of the two types of losses. The two modeling questions are answered: "What is the expected value of each row?" and "How does each row vary with itself and covary with the others?"

Exhibit 9.1 presents some diagnostics. The model had 122 = 2(55+1)+10 observations; however, two of these were constraints. After transformation (see footnote 9), a model with 122 observations in 22 parameters becomes one with 120 observations and 20 parameters. Readers wishing to solve the model in Excel can do so by reformulating '@Ult' predictions in terms of paid and incurred β_{12} , ..., β_{120} (if Excel will invert a 120×120 matrix). Two paid observations, AY 1998@12 and AY 2001@60, are more than two standard units away from expected. This is apparent even from the paid residuals of Exhibit 6.2. However, the diagnostics are not unusual, and we will focus on the estimates of β and \mathbf{y}_2 , as found in Exhibit 9.2. As for 'Betahat', the first eleven elements are paid, and the second eleven incurred. They both sum to 0.696. This purepremium equality would not obtain, if the exposures were adjusted for inflation and changes in claim processing; nonetheless, the ultimate AY equality would still be preserved. The reader may verify that $\hat{\beta}_{12} + ... \hat{\beta}_{120}$ equals 0.855×0.696 for paid and 0.950×0.696 for incurred, as required by the constraints.

The unpaid and IBNR predictions are transferred from Exhibit 9.2 to summary Exhibit 10. The 'Joint Paid-Incurred' box shows agreement of ultimate loss and the standard deviation of its prediction error throughout the ten accident years and in total. The total 'Std Dev' of $\pm 1,196,054$ equals the square root of the sum of the diagonal 11×11 blocks of the 'VarPrdErr' matrix of Exhibit 9.2. We also ran separate paid and incurred models from Exhibits 7, and summarized them in the right side of Exhibit 10. Not surprisingly, the joint model mediates between the separate models: $218,374,758 \in (217,725,562, 219,198,836)$. This holds true by accident year except for AY 2000. But the joint modeling produces second-moment estimates that dominate those of the submodels, both in total and by accident year.

6. CONCLUSION

In a footnote of the introduction we quoted, "Actuaries must not presume to judge what they cannot scientifically model." For a time science may endure competing theories, but eventually one will prevail. Likewise, the *ad hoc* blending of models, especially those arising from paid and incurred data sources, is a stopgap. For at some point it puts knowledge at the mercy of intuition at best, and of whim at worst. Hence actuaries have begun to search for joint models. Here we have shown that the linear statistical model is versatile enough to satisfy the search. The key, as always, is to ask first what all the equations are and second how they covary with each other. We may dub these the "first and second moment" questions. Actuaries have made rapid progress on the first-moment question; we hope that this paper will spur progress on the second.⁸

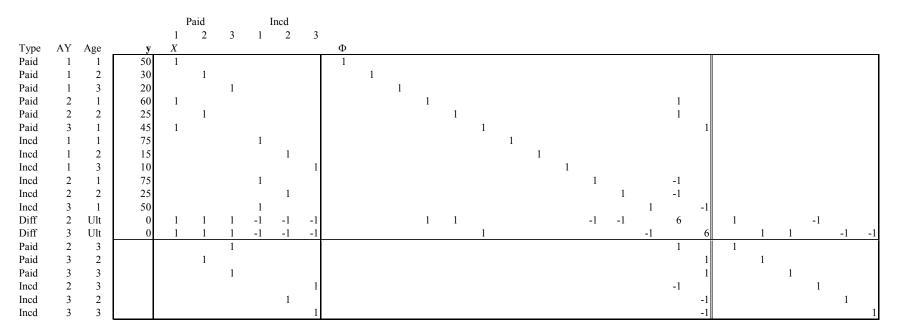
⁸ Since 2000 the topic of generalized linear models (GLM) has received much attention from actuaries and academicians. One is easily lulled into thinking that generalizing is only in one direction. Consider a linear statistical model (LSM) $\mathbf{y} = Xb + \mathbf{e}, Var[\mathbf{e}] = diag(\sigma)$, i.e., a heteroskedastic model. GLM generalizes it with a link function and with a distributional form of \mathbf{e} , for example as: $\mathbf{y} = g^{-1}(Xb) + \mathbf{e} - E[\mathbf{e}], \mathbf{e} \sim distribution(\Theta)$ [Anderson, 2004; 13-14]. The LSM is a GLM whose link is the identity function and whose distribution is multivariate normal. Now the link function can be accommodated with a *non-linear* statistical model (Halliwell [1997; 325-326] and Judge [1988; Chapter 12]). Hence, many consider the advantage of GLM over LSM to reside in non-normal error terms. However, the density of \mathbf{e} is invariably assumed to be the product of the densities of the elements of \mathbf{e} , which implies zero covariance. GLM does not generalize the variance structure beyond heteroskedasticity. Our linear model generalizes the LSM in a different direction from the one in which GLM generalizes it. We do not wish to gainsay GLM; certainly, no one has a panacea. However, to him whose only tool is a hammer everything looks like a nail. Enthusiasm over GLM may distract actuaries from asking the second-moment question. If covariance is key to the joint paid-incurred model, GLM will not provide an acceptable solution.

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REFERENCES

- [1.] Anderson, Duncan, Feldblum, Sholom, et al., "A Practitioner's Guide to Generalized Linear Models," 2004 Discussion Paper Program: Applying and Evaluating Generalized Linear Models, Casualty Actuarial Society, 2004, 1-116, www.casact.org/pubs/dpp/dpp04/04dpp1.pdf.
- [2.] Halliwell, Leigh J., "Conjoint Prediction of Paid and Incurred Losses," 1997 Loss Reserving Discussion Papers, Casualty Actuarial Society, 1997, 241-379, www.casact.org/pubs/forum/97sforum/97sf1241.pdf.
- [3.] "Chain-Ladder Bias: Its Reason and Meaning," Variance, 1:2, 2007, 214-247, www.variancejournal.org/issues/01-02/214.pdf.
- [4.] Healy, M. J. R., Matrices for Statistics, Oxford, Clarendon Press, 1986.
- [5.] Judge, George G., Hill, R. C., et al., Introduction to the Theory and Practice of Econometrics (Second Edition), New York, John Wiley & Sons, 1988.
- [6.] Quarg, Gerhard, and Mack, Thomas, "Munich Chain Ladder: A Reserving Method that Reduces the Gap between IBNR Projections Based on Paid Losses and IBNR Projections Based on Incurred Losses," Variance, 2:2, 2008, 266-299, www.variancejournal.org/issues/02-02/266.pdf. Reprinted from Blätter der Deutschen Gesellschaft für Versicherungs- und Finanzmathematik, 26:4, 2004, 597-630.
- [7.] Venter, Gary G., "Distribution and Value of Reserves Using Paid and Incurred Triangles," Casualty Actuarial Society E-Forum, Fall 2008, 348-375, www.casact.org/pubs/forum/08fforum/15Venter.pdf.
- [8.] Zhang, Yanwei, and Clark, David R., "A Bivariate Approach to Setting Reserves," unpublished as of March 2009.



Joint Paid-Incurred Model

	$X'\Phi^{-1}\mathbf{y}$	$X'\Phi^{-1}X$					
	155	3	0	0	0	0	0
	56.25	0	2.25	0.25	0	-0.25	-0.25
	28.75	0	0.25	1.75	0	-0.25	-0.75
	200	0	0	0	3	0	0
	38.75	0	-0.25	-0.25	0	2.25	0.25
	1.25	0	-0.25	-0.75	0	0.25	1.75
	β	$(X'\Phi^{-1}X)^{-1}$					
Paid1	51.666667	0.3333333	0	0	0	0	0
Paid2	26.25	0	0.4583333	-0.041667	0	0.0416667	0.0416667
Paid3	20	0	-0.041667	0.7083333	0	0.0416667	0.2916667
Incd1	66.666667	0	0	0	0.3333333	0	0
Incd2	21.25	0	0.0416667	0.0416667	0	0.4583333	-0.041667
Incd3	10	0	0.0416667	0.2916667	0	-0.041667	0.7083333
		Var[β]					

Solution of the Joint Paid-Incurred Model

Туре	AY	Age	У	Xβ	e
Paid	1	1	50	51.666667	-1.666667
Paid	1	2	30	26.25	3.75
Paid	1	3	20	20	0
Paid	2	1	60	51.666667	8.3333333
Paid	2	2	25	26.25	-1.25
Paid	3	1	45	51.666667	-6.666667
Incd	1	1	75	66.666667	8.3333333
Incd	1	2	15	21.25	-6.25
Incd	1	3	10	10	0
Incd	2	1	75	66.666667	8.3333333
Incd	2	2	25	21.25	3.75
Incd	3	1	50	66.666667	-16.66667
Diff	2	Ult	0	0	0
Diff	3	Ult	0	0	0

	SSCP	
24868.75	5 24229.167	639.58333
24229.167	24229.167	0
639.58333	3 0	639.58333
100.0%	97.4%	2.6%
t		14
k		6
df		8
σ^2		79.947917

Var[β]					
26.649306	0	0	0	0	0
0	36.642795	-3.331163	0	3.3311632	3.3311632
0	-3.331163	56.629774	0	3.3311632	23.318142
0	0	0	26.649306	0	0
0	3.3311632	3.3311632	0	36.642795	-3.331163
0	3.3311632	23.318142	0	-3.331163	56.629774

Prediction of the Joint Paid-Incurred Model

	AY		Paid	Incd	Unp	aid	IBN	٨R	Ultir	nate			
	1		100	100	0	± 0	0	± 0	100	± 0			
	2		85	100	22.50	± 8.94	7.50	± 8.94	107.50	± 8.94			
	3		45	50	41.25	± 11.83	36.25	± 11.83	86.25	± 11.83			
	Total		230	250	63.75	± 17.31	43.75	± 17.31	293.75	±17.31			
Туре	AY	Age		$\hat{\mathbf{y}}_2$	Std [PE]	Var [Predi	iction Er	ror]					
Paid	2	3	Г	22.50	± 8.94	79.95	0	39.97	79.95	0	39.97		
Paid	3	2		23.75	± 9.48	0	89.94	-29.98	0	29.98	29.98		
Paid	3	3		17.50	± 10.48	39.97	-29.98	109.93	39.97	29.98	49.97		
Incd	2	3		7.50	± 8.94	79.95	0	39.97	79.95	0	39.97		
Incd	3	2		23.75	± 9.48	0	29.98	29.98	0	89.94	-29.98		
Incd	3	3	L	12.50	± 10.48	39.97	29.98	49.97	39.97	-29.98	109.93		
$X_2 - \Phi_{21} \Phi$	$p_{11}^{-1} X_1$						_	$\Phi_{22} - \Phi_{21}$					
0	0	0.5	0	0	0.5			0.5	0	0	0.5	0	0
0	0.75	-0.25	0	0.25	0.25			0	0.75	-0.25	0	0.25	0.25
0	-0.25	0.75	0	0.25	0.25			0	-0.25	0.75	0	0.25	0.25
0	0	0.5	0	0	0.5			0.5	0	0	0.5	0	0
0	0.25	0.25	0	0.75	-0.25			0	0.25	0.25	0	0.75	-0.25
0	0.25	0.25	0	-0.25	0.75			0	0.25	0.25	0	-0.25	0.75
,													
$\Phi_{21}\Phi_{11}^{-1}$													
0	0	0	-0.5	-0.5	0	0	0	0	0.5	0.5	0	0.5	0
0	0	0	0	0	-0.25	0	0	0	0	0	0.25	0	0.25
0	0	0	0	0	-0.25	0	0	0	0	0	0.25	0	0.25
0	0	0	0.5	0.5	0	0	0	0	-0.5	-0.5	0	-0.5	0
0	0	0	0	0	0.25	0	0	0	0	0	-0.25	0	-0.25
0	0	0	0	0	0.25	0	0	0	0	0	-0.25	0	-0.25

Industry Workers' Compensation Net Losses (000)

			Cumulative Paid									
AY	EarnPrem	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	
1998	23,278,084	4,651,588	9,585,142	12,606,256	14,094,760	15,268,042	16,074,584	16,668,642	17,106,835	17,422,941	17,738,999	
1999	21,555,421	4,211,880	9,632,480	12,750,495	14,618,989	15,637,068	16,221,974	16,753,957	17,200,779	17,557,257		
2000	23,495,444	4,553,584	10,366,172	13,709,157	15,579,342	16,724,292	17,365,134	17,961,104	18,432,885			
2001	25,864,065	4,556,995	10,343,323	13,761,573	15,619,782	16,358,074	16,800,979	17,289,118				
2002	29,134,414	4,262,115	9,525,796	12,527,871	14,177,862	15,284,598	15,899,281					
2003	32,391,860	4,274,440	9,451,725	12,390,213	14,138,206	15,283,538						
2004	36,533,278	4,624,395	9,798,635	12,473,626	14,134,508							
2005	39,208,849	4,865,363	9,946,876	12,789,801								
2006	42,065,555	5,130,174	10,724,002									
2007	40,220,014	5,211,936										

AY	Ultimate	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120
1998	20,815,720	10,440,449	14,526,669	16,215,164	17,259,403	18,111,150	18,727,822	19,147,843	19,469,090	19,540,774	19,765,070
1999	21,107,246	10,104,076	14,366,317	16,374,957	17,641,331	18,407,246	18,857,403	19,336,715	19,548,265	19,812,936	
2000	22,339,113	10,614,330	15,701,665	17,701,687	18,844,583	19,668,677	20,094,561	20,402,031	20,739,446		
2001	21,958,321	11,104,926	15,846,924	17,963,819	18,932,871	19,179,055	19,484,216	19,811,447			
2002	21,039,160	10,379,583	15,108,660	16,994,756	17,685,953	18,233,994	18,589,679				
2003	21,658,869	10,932,703	15,324,420	16,898,562	17,695,675	18,314,350					
2004	22,204,956	11,239,343	15,320,398	16,843,250	17,632,172						
2005	23,445,324	11,978,411	15,632,319	17,221,257							
2006	26,885,991	12,468,437	16,822,179								
2007	27,906,944	12,931,177									

Comparison of Ultimates and On-Level Premium

AY	EarnPrem	Ultimate	CL Paid	CL Incd	Add Paid	Add Incd	Selected
1998	23,278,084	20,815,720	20,747,367	20,805,337	20,226,368	20,881,833	20,695,325
1999	21,555,421	21,107,246	20,907,312	21,095,112	20,153,220	21,074,491	20,867,476
2000	23,495,444	22,339,113	22,380,335	22,271,937	21,614,963	22,286,578	22,178,585
2001	25,864,065	21,958,321	21,545,915	21,589,768	21,305,566	21,762,406	21,632,395
2002	29,134,414	21,039,160	20,472,774	20,661,071	21,107,195	21,042,136	20,864,467
2003	32,391,860	21,658,869	20,446,996	20,823,394	21,885,290	21,389,016	21,240,713
2004	36,533,278	22,204,956	20,265,744	20,762,892	23,064,618	21,607,790	21,581,200
2005	39,208,849	23,445,324	20,806,146	21,420,815	24,850,676	22,488,817	22,602,355
2006	42,065,555	26,885,991	22,848,638	23,395,295	28,090,567	24,939,425	25,231,983
2007	40,220,014	27,906,944	24,129,598	25,119,943	28,910,427	26,647,699	26,542,922
Total	313,746,984	229,361,644	214,550,827	217,945,563	231,208,890	224,120,191	223,437,423
AY	Adi Prem	Ultimate	CL Paid	CL Incd	Add Paid	Add Incd	Selected
AY 1998	Adj Prem 29.060.019	Ultimate 89%	CL Paid 89%	CL Incd 89%	Add Paid 87%	Add Incd 90%	Selected 89%
1998	29,060,019	89%	89%	CL Incd 89% 98%	87%	Add Incd 90% 98%	89%
_	29,060,019 29,301,751			89%	87% 93%	90%	
1998 1999 2000	29,060,019 29,301,751 31,142,788	89% 98% 95%	89% 97% 95%	89% 98% 95%	87% 93% 92%	90% 98% 95%	89% 97% 94%
1998 1999	29,060,019 29,301,751	89% 98%	89% 97%	89% 98%	87% 93%	90% 98%	89% 97%
1998 1999 2000 2001	29,060,019 29,301,751 31,142,788 30,375,837 29,297,526	89% 98% 95% 85%	89% 97% 95% 83%	89% 98% 95% 83%	87% 93% 92% 82%	90% 98% 95% 84%	89% 97% 94% 84%
1998 1999 2000 2001 2002	29,060,019 29,301,751 31,142,788 30,375,837	89% 98% 95% 85% 72%	89% 97% 95% 83% 70%	89% 98% 95% 83% 71%	87% 93% 92% 82% 72%	90% 98% 95% 84% 72%	89% 97% 94% 84% 72%
1998 1999 2000 2001 2002 2003	29,060,019 29,301,751 31,142,788 30,375,837 29,297,526 29,825,844	89% 98% 95% 85% 72% 67%	89% 97% 95% 83% 70% 63%	89% 98% 95% 83% 71% 64%	87% 93% 92% 82% 72% 68%	90% 98% 95% 84% 72% 66%	89% 97% 94% 84% 72% 66%
1998 1999 2000 2001 2002 2003 2004	29,060,019 29,301,751 31,142,788 30,375,837 29,297,526 29,825,844 30,303,950	89% 98% 95% 85% 72% 67% 61%	89% 97% 95% 83% 70% 63% 55%	89% 98% 95% 83% 71% 64% 57%	87% 93% 92% 82% 72% 68% 63%	90% 98% 95% 84% 72% 66% 59%	89% 97% 94% 84% 72% 66% 59%
1998 1999 2000 2001 2002 2003 2004 2005	29,060,019 29,301,751 31,142,788 30,375,837 29,297,526 29,825,844 30,303,950 31,737,838	89% 98% 95% 85% 72% 67% 61% 60%	89% 97% 95% 83% 70% 63% 55% 53%	89% 98% 95% 83% 71% 64% 57% 55%	87% 93% 92% 82% 72% 68% 63% 63%	90% 98% 95% 84% 72% 66% 59% 57%	89% 97% 94% 84% 72% 66% 59% 58%

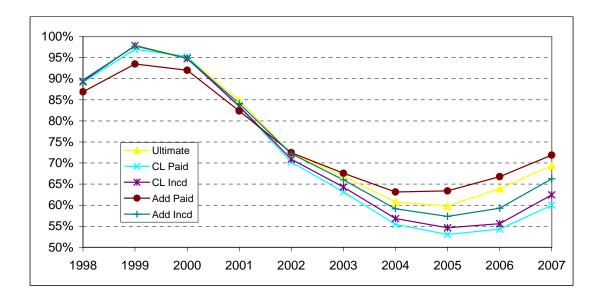


Exhibit 6.1

Additive Projections with On-Level Premium

						Incr	emental Paid							
AY	Adj Prem	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult	Ultimate	LR
1998	29,060,019	4,651,588	4,933,554	3,021,114	1,488,504	1,173,282	806,542	594,058	438,193	316,106	316,058	2,924,110	20,663,109	71%
1999	29,301,751	4,211,880	5,420,600	3,118,015	1,868,494	1,018,079	584,906	531,983	446,822	356,478	318,687	2,948,434	20,824,378	71%
2000	31,142,788	4,553,584	5,812,588	3,342,985	1,870,185	1,144,950	640,842	595,970	471,781	358,902	338,710	3,133,685	22,264,182	71%
2001	30,375,837	4,556,995	5,786,328	3,418,250	1,858,209	738,292	442,905	488,139	460,466	350,063	330,369	3,056,512	21,486,528	71%
2002	29,297,526	4,262,115	5,263,681	3,002,075	1,649,991	1,106,736	614,683	540,138	444,120	337,636	318,641	2,948,009	20,487,825	70%
2003	29,825,844	4,274,440	5,177,285	2,938,488	1,747,993	1,145,332	617,774	549,878	452,129	343,725	324,387	3,001,170	20,572,601	69%
2004	30,303,950	4,624,395	5,174,240	2,674,991	1,660,882	1,071,056	627,677	558,692	459,376	349,235	329,587	3,049,279	20,579,410	68%
2005	31,737,838	4,865,363	5,081,513	2,842,925	1,841,463	1,121,735	657,376	585,128	481,113	365,759	345,182	3,193,561	21,381,119	67%
2006	35,430,317	5,130,174	5,593,828	3,580,408	2,055,705	1,252,242	733,858	653,204	537,087	408,313	385,342	3,565,110	23,895,269	67%
2007	37,271,114	5,211,936	6,503,618	3,766,430	2,162,510	1,317,302	771,986	687,141	564,991	429,527	405,362	3,750,337	25,571,140	69%
	Pure Prem	0.148	0.174	0.101	0.058	0.035	0.021	0.018	0.015	0.012	0.011	0.101	217,725,562	69%
						Increme	ntal Case-Incu	irred						
AY	Adj Prem	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult	Ultimate	LR
~ 1		@12	@24	@30	@40	@00	@12	w04	@ 90	₩100	₩120	@01	Unimate	

AY	Adj Prem	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult	Ultimate	LR
1998	29,060,019	10,440,449	4,086,220	1,688,495	1,044,239	851,747	616,672	420,021	321,247	71,684	224,296	1,015,137	20,805,337	72%
1999	29,301,751	10,104,076	4,262,241	2,008,640	1,266,374	765,915	450,157	479,312	211,550	264,671	226,162	1,023,581	21,093,787	72%
2000	31,142,788	10,614,330	5,087,335	2,000,022	1,142,896	824,094	425,884	307,470	337,415	179,484	240,372	1,087,893	22,272,950	72%
2001	30,375,837	11,104,926	4,741,998	2,116,895	969,052	246,184	305,161	327,231	295,330	175,064	234,452	1,061,102	21,596,099	71%
2002	29,297,526	10,379,583	4,729,077	1,886,096	691,197	548,041	355,685	374,902	284,846	168,850	226,129	1,023,434	20,678,323	71%
2003	29,825,844	10,932,703	4,391,717	1,574,142	797,113	618,675	430,571	381,663	289,983	171,895	230,207	1,041,889	20,861,756	70%
2004	30,303,950	11,239,343	4,081,055	1,522,852	788,922	652,563	437,473	387,781	294,631	174,650	233,897	1,058,590	20,855,966	69%
2005	31,737,838	11,978,411	3,653,908	1,588,938	1,015,906	683,441	458,173	406,129	308,573	182,914	244,964	1,108,680	21,601,428	68%
2006	35,430,317	12,468,437	4,353,742	2,114,552	1,134,100	762,954	511,478	453,379	344,473	204,195	273,464	1,237,667	23,811,342	67%
2007	37,271,114	12,931,177	5,309,716	2,224,415	1,193,022	802,594	538,052	476,935	362,370	214,804	287,672	1,301,970	25,621,849	69%
	Pure Prem	0.358	0.142	0.060	0.032	0.022	0.014	0.013	0.010	0.006	0.008	0.035	219,198,836	70%

Exhibit 6.2

Variances from Additive Method

						Paid Res						
AY	1/AdjPrem	@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult
1998	3.44E-08	359,234	-137,270	84,456	-197,589	146,191	204,630	58,299	-2,327	-18,793	0	
1999	3.41E-08	-116,179	307,595	156,928	168,375	-17,556	-22,012	-8,233	2,638	18,793		
2000	3.21E-08	-46,408	378,332	195,853	63,247	44,246	-4,209	21,812	-311			
2001	3.29E-08	70,286	485,901	348,622	95,770	-335,305	-186,261	-71,879				
2002	3.41E-08	-65,320	151,413	41,415	-49,883	71,250	7,852					
2003	3.35E-08	-131,031	-27,171	-75,561	17,466	91,174						
2004	3.30E-08	148,305	-113,643	-387,373	-97,386							
2005	3.15E-08	177,478	-456,577	-364,340								
2006	2.82E-08	-103,114	-588,580									
2007	2.68E-08	-293,250										
	Zero check	0	0	0	0	0	0	0	0	0	0	
	WSSR	10,180	33,825	15,703	3,150	4,962	2,602	305	0	24	0	
	df	9	8	7	6	5	4	3	2	1	0	
	Unit Var	1,131	4,228	2,243	525	992	651	102	0	24		
	Selected	1,131	4,228	2,243	992	992	651	102	102	55	27	1,928
A. \ /	1/AdjPrem	_				Incurred Re	esiduais					
AY	1/AdiProm			A A A	@ 10	A A A		AAA	@ 0 0	@ 100	@ 400	e lui
		@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult
1998	3.44E-08	48,816	-53,727	-45,865	114,048	225,970	@72 197,156	48,158	38,710	-95,797	@120 0	@Ult
1999	3.44E-08 3.41E-08	48,816 -373,998	-53,727 87,856	-45,865 259,853	114,048 328,446	225,970 134,933	@72 197,156 27,152	48,158 104,356	38,710 -73,338			@Ult
1999 2000	3.44E-08 3.41E-08 3.21E-08	48,816 -373,998 -522,084	-53,727 87,856 650,673	-45,865 259,853 141,358	114,048 328,446 146,037	225,970 134,933 153,467	@72 197,156 27,152 -23,699	48,158 104,356 -91,045	38,710	-95,797		@Ult
1999 2000 2001	3.44E-08 3.41E-08 3.21E-08 3.29E-08	48,816 -373,998 -522,084 242,767	-53,727 87,856 650,673 414,597	-45,865 259,853 141,358 304,004	114,048 328,446 146,037 -3,257	225,970 134,933 153,467 -407,927	@72 197,156 27,152 -23,699 -133,350	48,158 104,356	38,710 -73,338	-95,797		@Ult
1999 2000 2001 2002	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.41E-08	48,816 -373,998 -522,084 242,767 -96,980	-53,727 87,856 650,673 414,597 555,294	-45,865 259,853 141,358 304,004 137,561	114,048 328,446 146,037 -3,257 -246,596	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699	48,158 104,356 -91,045	38,710 -73,338	-95,797		@Ult
1999 2000 2001 2002 2003	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.41E-08 3.35E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218	-53,727 87,856 650,673 414,597 555,294 142,669	-45,865 259,853 141,358 304,004 137,561 -205,924	114,048 328,446 146,037 -3,257 -246,596 -157,591	225,970 134,933 153,467 -407,927	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338	-95,797		@Ult
1999 2000 2001 2002 2003 2004	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.41E-08 3.35E-08 3.30E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748	114,048 328,446 146,037 -3,257 -246,596	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338	-95,797		@ Ult
1999 2000 2001 2002 2003 2004 2005	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.41E-08 3.35E-08 3.30E-08 3.15E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527	-45,865 259,853 141,358 304,004 137,561 -205,924	114,048 328,446 146,037 -3,257 -246,596 -157,591	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338	-95,797		@ Ult
1999 2000 2001 2002 2003 2004 2005 2006	3.44E-08 3.41E-08 3.29E-08 3.41E-08 3.35E-08 3.30E-08 3.15E-08 2.82E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211 -201,163	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748	114,048 328,446 146,037 -3,257 -246,596 -157,591	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338	-95,797		@ Ult
1999 2000 2001 2002 2003 2004 2005	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.41E-08 3.35E-08 3.30E-08 3.15E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748	114,048 328,446 146,037 -3,257 -246,596 -157,591	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338	-95,797		@ Ult
1999 2000 2001 2002 2003 2004 2005 2006	3.44E-08 3.41E-08 3.29E-08 3.29E-08 3.35E-08 3.30E-08 3.30E-08 2.82E-08 2.68E-08 Zero check	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211 -201,163	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748	114,048 328,446 146,037 -3,257 -246,596 -157,591	225,970 134,933 153,467 -407,927 -82,850	@72 197,156 27,152 -23,699 -133,350	48,158 104,356 -91,045	38,710 -73,338 34,628 0	-95,797		@ Ult
1999 2000 2001 2002 2003 2004 2005 2006	3.44E-08 3.41E-08 3.29E-08 3.41E-08 3.35E-08 3.30E-08 3.15E-08 2.82E-08 2.68E-08	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211 -201,163 -396,678	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527 -693,730	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748 -305,240	114,048 328,446 146,037 -3,257 -246,596 -157,591 -181,086	225,970 134,933 153,467 -407,927 -82,850 -23,593	@72 197,156 27,152 -23,699 -133,350 -67,259	48,158 104,356 -91,045 -61,469	38,710 -73,338 34,628	-95,797 95,797	0	@ Ult
1999 2000 2001 2002 2003 2004 2005 2006	3.44E-08 3.41E-08 3.29E-08 3.29E-08 3.35E-08 3.30E-08 3.30E-08 2.82E-08 2.68E-08 Zero check	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211 -201,163 -396,678 0 41,458 9	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527 -693,730	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748 -305,240 0 13,759 7	114,048 328,446 146,037 -3,257 -246,596 -157,591 -181,086 0 8,805 6	225,970 134,933 153,467 -407,927 -82,850 -23,593 0 8,866 5	@72 197,156 27,152 -23,699 -133,350 -67,259	48,158 104,356 -91,045 -61,469	38,710 -73,338 34,628 0	-95,797 95,797 0	0	@ Ult
1999 2000 2001 2002 2003 2004 2005 2006	3.44E-08 3.41E-08 3.21E-08 3.29E-08 3.35E-08 3.35E-08 3.30E-08 3.35E-08 2.68E-08 2.68E-08 Zero check WSSR	48,816 -373,998 -522,084 242,767 -96,980 267,218 402,891 629,211 -201,163 -396,678 0 41,458	-53,727 87,856 650,673 414,597 555,294 142,669 -236,105 -867,527 -693,730 0 69,959	-45,865 259,853 141,358 304,004 137,561 -205,924 -285,748 -305,240 0 13,759	114,048 328,446 146,037 -3,257 -246,596 -157,591 -181,086 0 8,805	225,970 134,933 153,467 -407,927 -82,850 -23,593 0 8,866	@72 197,156 27,152 -23,699 -133,350 -67,259 0 2,121	48,158 104,356 -91,045 -61,469 0 842	38,710 -73,338 34,628 0 274	-95,797 95,797 0 629	0	@Ult

Exhibit 6.3

Absolute Variances from Additive Method

						Р	aid Variance						
AY	Adj Prem	12	24	36	48	60	72	84	96	108	120	@Ult	Unpaid
1998	29,060,019	3.29E+10	1.23E+11	6.52E+10	2.88E+10	2.88E+10	1.89E+10	2.95E+09	2.95E+09	1.59E+09	7.97E+08	5.60E+10	5.60E+10
1999	29,301,751	3.31E+10	1.24E+11	6.57E+10	2.91E+10	2.91E+10	1.91E+10	2.98E+09	2.98E+09	1.61E+09	8.04E+08	5.65E+10	5.73E+10
2000	31,142,788	3.52E+10	1.32E+11	6.99E+10	3.09E+10	3.09E+10	2.03E+10	3.16E+09	3.16E+09	1.71E+09	8.55E+08	6.01E+10	6.26E+10
2001	30,375,837	3.44E+10	1.28E+11	6.81E+10	3.01E+10	3.01E+10	1.98E+10	3.08E+09	3.08E+09	1.67E+09	8.33E+08	5.86E+10	6.42E+10
2002	29,297,526	3.31E+10	1.24E+11	6.57E+10	2.91E+10	2.91E+10	1.91E+10	2.98E+09	2.98E+09	1.61E+09	8.04E+08	5.65E+10	6.49E+10
2003	29,825,844	3.37E+10	1.26E+11	6.69E+10	2.96E+10	2.96E+10	1.94E+10	3.03E+09	3.03E+09	1.64E+09	8.18E+08	5.75E+10	8.54E+10
2004	30,303,950	3.43E+10	1.28E+11	6.80E+10	3.01E+10	3.01E+10	1.97E+10	3.08E+09	3.08E+09	1.66E+09	8.32E+08	5.84E+10	1.17E+11
2005	31,737,838	3.59E+10	1.34E+11	7.12E+10	3.15E+10	3.15E+10	2.06E+10	3.22E+09	3.22E+09	1.74E+09	8.71E+08	6.12E+10	1.54E+11
2006	35,430,317	4.01E+10	1.50E+11	7.95E+10	3.52E+10	3.52E+10	2.30E+10	3.60E+09	3.60E+09	1.94E+09	9.72E+08	6.83E+10	2.51E+11
2007	37,271,114	4.22E+10	1.58E+11	8.36E+10	3.70E+10	3.70E+10	2.42E+10	3.78E+09	3.78E+09	2.05E+09	1.02E+09	7.19E+10	4.22E+11
	_												
						Inc	urred Varianc	e					
		10		~~~	40		70			100	100	0.114	

AY	Adj Prem	12	24	36	48	60	72	84	96	108	120	@Ult	IBNR
1998	29,060,019	1.34E+11	2.54E+11	5.71E+10	5.15E+10	5.15E+10	1.54E+10	8.45E+09	8.45E+09	8.45E+09	5.81E+09	3.12E+10	3.12E+10
1999	29,301,751	1.35E+11	2.56E+11	5.76E+10	5.20E+10	5.20E+10	1.55E+10	8.52E+09	8.52E+09	8.52E+09	5.86E+09	3.14E+10	3.73E+10
2000	31,142,788	1.43E+11	2.72E+11	6.12E+10	5.52E+10	5.52E+10	1.65E+10	9.06E+09	9.06E+09	9.06E+09	6.23E+09	3.34E+10	4.87E+10
2001	30,375,837	1.40E+11	2.66E+11	5.97E+10	5.39E+10	5.39E+10	1.61E+10	8.83E+09	8.83E+09	8.83E+09	6.08E+09	3.26E+10	5.63E+10
2002	29,297,526	1.35E+11	2.56E+11	5.76E+10	5.19E+10	5.19E+10	1.55E+10	8.52E+09	8.52E+09	8.52E+09	5.86E+09	3.14E+10	6.28E+10
2003	29,825,844	1.37E+11	2.61E+11	5.86E+10	5.29E+10	5.29E+10	1.58E+10	8.67E+09	8.67E+09	8.67E+09	5.97E+09	3.20E+10	7.98E+10
2004	30,303,950	1.40E+11	2.65E+11	5.96E+10	5.37E+10	5.37E+10	1.61E+10	8.81E+09	8.81E+09	8.81E+09	6.06E+09	3.25E+10	1.35E+11
2005	31,737,838	1.46E+11	2.78E+11	6.24E+10	5.63E+10	5.63E+10	1.68E+10	9.23E+09	9.23E+09	9.23E+09	6.35E+09	3.40E+10	1.97E+11
2006	35,430,317	1.63E+11	3.10E+11	6.96E+10	6.28E+10	6.28E+10	1.88E+10	1.03E+10	1.03E+10	1.03E+10	7.09E+09	3.80E+10	2.90E+11
2007	37,271,114	1.72E+11	3.26E+11	7.33E+10	6.61E+10	6.61E+10	1.98E+10	1.08E+10	1.08E+10	1.08E+10	7.45E+09	4.00E+10	6.31E+11

Exhibit 7.1

							Paid L	inear Mode	el						
				@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult	
AY 1998	Age	AdjPrem 29,060,019	y 4,651,588	X 29,060,019											3.29E+10
1998		29,060,019	4,933,554	23,000,013	29,060,019										1.23E+10
1998		29,060,019	3,021,114			29,060,019									6.52E+10
1998		29,060,019	1,488,504				29,060,019								2.88E+10
1998		29,060,019	1,173,282					29,060,019							2.88E+10
1998 1998		29,060,019 29,060,019	806,542 594,058						29,060,019	29,060,019					1.89E+10
1998		29,060,019	438,193							29,060,019	29,060,019				2.95E+09 2.95E+09
1998		29,060,019	316,106								20,000,010	29,060,019			1.59E+09
1998		29,060,019	316,058										29,060,019		7.97E+08
1999		29,301,751	4,211,880	29,301,751											3.31E+10
1999		29,301,751	5,420,600		29,301,751	~~~~~~~									1.24E+11
1999 1999		29,301,751 29,301,751	3,118,015 1,868,494			29,301,751	29,301,751								6.57E+10 2.91E+10
1999		29,301,751	1,008,494				29,301,751	29,301,751							2.91E+10 2.91E+10
1999		29,301,751	584,906					20,001,101	29,301,751						1.91E+10
1999		29,301,751	531,983							29,301,751					2.98E+09
1999		29,301,751	446,822								29,301,751				2.98E+09
1999 2000		29,301,751 31,142,788	356,478	04 4 40 700								29,301,751			1.61E+09
2000		31,142,788	4,553,584 5,812,588	31,142,788	31,142,788										3.52E+10 1.32E+11
2000		31,142,788	3,342,985		31,142,700	31,142,788									6.99E+10
2000		31,142,788	1,870,185				31,142,788								3.09E+10
2000		31,142,788	1,144,950					31,142,788							3.09E+10
2000		31,142,788	640,842						31,142,788						2.03E+10
2000 2000		31,142,788	595,970							31,142,788	04 4 40 700				3.16E+09 3.16E+09
2000		31,142,788 30,375,837	471,781 4,556,995	30,375,837							31,142,788				3.16E+09 3.44E+10
2001		30,375,837	5,786,328	50,575,057	30,375,837										1.28E+11
2001		30,375,837	3,418,250			30,375,837									6.81E+10
2001		30,375,837	1,858,209				30,375,837								3.01E+10
2001		30,375,837	738,292					30,375,837							3.01E+10
2001 2001		30,375,837 30,375,837	442,905 488,139						30,375,837	30,375,837					1.98E+10 3.08E+09
2001		29,297,526	400,139	29.297.526						30,375,637					3.31E+10
2002		29,297,526	5,263,681	20,201,020	29,297,526										1.24E+11
2002		29,297,526	3,002,075			29,297,526									6.57E+10
2002		29,297,526	1,649,991				29,297,526								2.91E+10
2002 2002		29,297,526 29,297,526	1,106,736 614,683					29,297,526	29,297,526						2.91E+10 1.91E+10
2002		29,297,526	4.274.440	29.825.844					29,297,520						3.37E+10
2003		29,825,844	5,177,285	20,020,044	29.825.844										1.26E+11
2003	36	29,825,844	2,938,488			29,825,844									6.69E+10
2003		29,825,844	1,747,993				29,825,844								2.96E+10
2003		29,825,844	1,145,332	20 202 050				29,825,844							2.96E+10
2004 2004		30,303,950 30,303,950	4,624,395 5,174,240	30,303,950	30,303,950										3.43E+10 1.28E+11
2004		30,303,950	2,674,991		50,000,000	30,303,950									6.80E+10
2004	48	30,303,950	1,660,882			.,	30,303,950								3.01E+10
2005		31,737,838	4,865,363	31,737,838											3.59E+10
2005		31,737,838	5,081,513		31,737,838										1.34E+11
2005 2006		31,737,838 35,430,317	2,842,925 5,130,174	35,430,317		31,737,838									7.12E+10 4.01E+10
2006		35,430,317	5,130,174	33,430,317	35,430,317										4.01E+10 1.50E+11
2000		37,271,114	5,211,936	37,271,114	20, 100,017										4.22E+10
		Constraint	0	0.145	0.145	0.145	0.145	0.145	0.145	0.145	0.145	0.145	0.145	-0.855	
		29,060,019												29,060,019	5.60E+10
		29,301,751										04 4 40 70 -	29,301,751		5.73E+10
		31,142,788 30,375,837									20 275 927	31,142,788 30,375,837	31,142,788		6.26E+10 6.42E+10
		29,297,526								29,297,526		29,297,526			6.42E+10 6.49E+10
		29,825,844							29,825,844	29,825,844		29,825,844			8.54E+10
2004	Unpd	30,303,950							30,303,950	30,303,950	30,303,950	30,303,950	30,303,950	30,303,950	1.17E+11
		31,737,838					31,737,838		31,737,838	31,737,838	31,737,838				1.54E+11
		35,430,317			37 374 44 4		35,430,317								2.51E+11 4.22E+11
2007	onpa	37,271,114			31,211,114	31,211,114	37,271,114	31,211,114	31,211,114	31,211,114	31,211,114	31,211,114	31,211,114	31,211,114	4.22E+11

Case_Incurred Linear Model

Exhibit 7.2

							case_incui	Tod Emour	Model						
				@12	@24	@36	@48	@60	@72	@84	@96	@108	@120	@Ult	
AY	Age	AdjPrem	У	X	-				-						Σ
1998	12	29,060,019	10,440,449	29,060,019											1.34E+11
1998	24	29,060,019	4,086,220		29,060,019										2.54E+11
1998	36	29,060,019	1,688,495			29,060,019									5.71E+10
1998		29,060,019	1,044,239				29,060,019								5.15E+10
1998 1998	00	29,060,019 29,060,019	851,747 616,672					29,060,019	29,060,019						5.15E+10 1.54E+10
1998		29,060,019	420,021						29,000,019	29,060,019					8.45E+09
1998		29,060,019	321,247							23,000,013	29,060,019				8.45E+09
1998		29,060,019	71,684									29,060,019			8.45E+09
1998	120	29,060,019	224,296										29,060,019		5.81E+09
1999		29,301,751	10,104,076	29,301,751											1.35E+11
1999		29,301,751	4,262,241		29,301,751										2.56E+11
1999		29,301,751	2,008,640			29,301,751									5.76E+10
1999 1999		29,301,751 29,301,751	1,266,374 765,915				29,301,751	29,301,751							5.20E+10 5.20E+10
1999		29,301,751	450,157					29,301,751	29,301,751						1.55E+10
1999		29,301,751	479,312						20,001,701	29,301,751					8.52E+09
1999		29,301,751	211,550								29,301,751				8.52E+09
1999	108	29,301,751	264,671									29,301,751			8.52E+09
2000		31,142,788	10,614,330	31,142,788											1.43E+11
2000		31,142,788	5,087,335		31,142,788										2.72E+11
2000		31,142,788	2,000,022			31,142,788									6.12E+10
2000 2000		31,142,788 31,142,788	1,142,896 824,094				31,142,788	31.142.788							5.52E+10 5.52E+10
2000		31,142,788	425,884					31,142,700	31.142.788						1.65E+10
2000		31,142,788	307,470						51,142,700	31.142.788					9.06E+09
2000		31,142,788	337,415								31,142,788				9.06E+09
2001	12	30,375,837	11,104,926	30,375,837											1.40E+11
2001	24	30,375,837	4,741,998		30,375,837										2.66E+11
2001		30,375,837	2,116,895			30,375,837									5.97E+10
2001		30,375,837	969,052				30,375,837	00.075.007							5.39E+10
2001 2001		30,375,837 30,375,837	246,184 305,161					30,375,837	30,375,837						5.39E+10 1.61E+10
2001		30,375,837	327,231						30,375,637	30,375,837					8.83E+09
2001		29,297,526	10,379,583	29,297,526						30,073,007					1.35E+11
2002		29,297,526	4,729,077		29,297,526										2.56E+11
2002		29,297,526	1,886,096			29,297,526									5.76E+10
2002		29,297,526	691,197				29,297,526								5.19E+10
2002		29,297,526	548,041					29,297,526							5.19E+10
2002		29,297,526	355,685						29,297,526						1.55E+10
2003 2003		29,825,844 29,825,844	10,932,703 4,391,717	29,825,844	29,825,844										1.37E+11 2.61E+11
2003		29,825,844	1,574,142		29,023,044	29,825,844									5.86E+10
2003		29,825,844	797,113			20,020,044	29,825,844								5.29E+10
2003		29,825,844	618,675					29,825,844							5.29E+10
2004	12	30,303,950	11,239,343	30,303,950											1.40E+11
2004		30,303,950	4,081,055		30,303,950										2.65E+11
2004		30,303,950	1,522,852			30,303,950									5.96E+10
2004 2005		30,303,950 31,737,838	788,922	31.737.838			30,303,950								5.37E+10 1.46E+11
2005		31,737,838	11,978,411 3,653,908	51,/3/,838	31,737,838										1.46E+11 2.78E+11
2005		31,737,838	1,588,938		51,151,030	31,737,838									6.24E+10
2005		35,430,317	12,468,437	35,430,317		21,101,000									1.63E+11
2006		35,430,317	4,353,742	,	35,430,317										3.10E+11
2007		37,271,114	12,931,177	37,271,114											1.72E+11
		Constraint	0	0.05	0.05	0.05	0.05	0.05	0.05	0.05	0.05	0.05	0.05		
		29,060,019												29,060,019	3.12E+10
		29,301,751										04 4 40 700	29,301,751		3.73E+10
		31,142,788 30,375,837									20 275 827	31,142,788 30,375,837	31,142,788		4.87E+10 5.63E+10
2001		29,297,526								29 297 526	29,297,526				5.63E+10 6.28E+10
		29,297,526							29,825,844	29,297,526		29,297,526			0.28E+10 7.98E+10
		30,303,950						30,303,950	30,303,950						1.35E+11
		31,737,838						31,737,838	31,737,838	31,737,838	31,737,838	31,737,838	31,737,838	31,737,838	1.97E+11
		35,430,317							35,430,317						2.90E+11
2007	IBNR	37,271,114			37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	37,271,114	6.31E+11

Casualty Actuarial Society E-Forum, Spring 2009

Exhibit 8.1

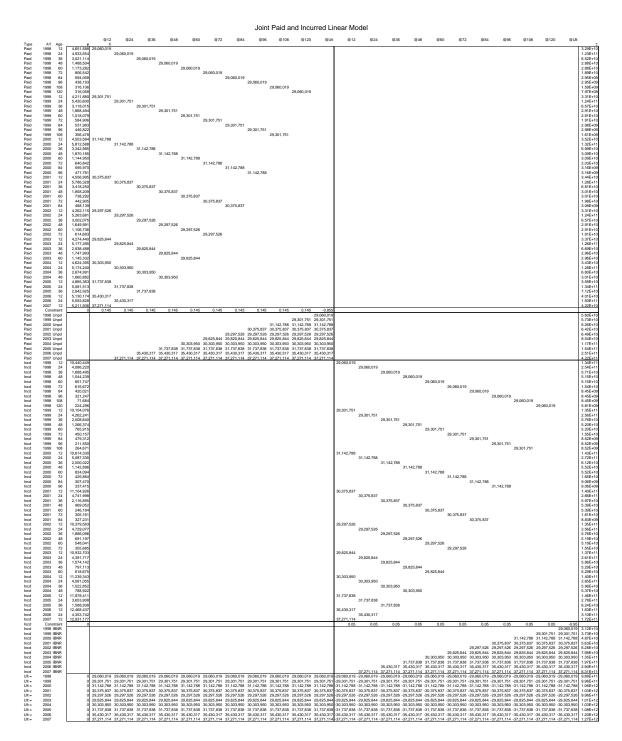
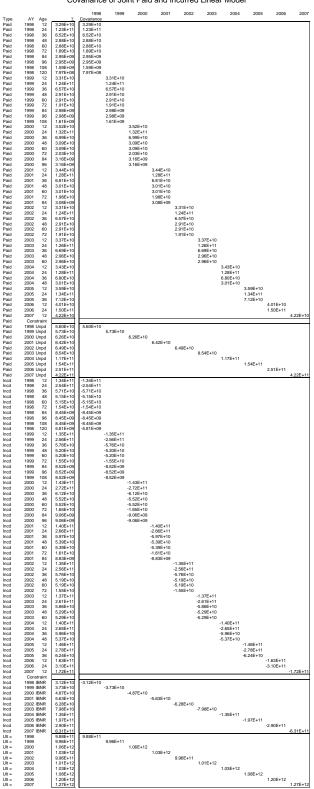


Exhibit 8.2



Covariance of Joint Paid and Incurred Linear Model

Exhibit 9.1

Joint Paid and Incurred Linear Model: Diagnostics

TYPE AY AGE y1 Fitted Resid Student Paid 1988 12 4,651,588 4,2551,411 366,447 2,06 Paid 1998 24 4,933,554 5,081,187 -147,633 -0,44 Paid 1998 36 3,021,114 2,939,870 81,244 0,34 Paid 1998 48 1,488,504 1,688,312 -200,808 1,27 Paid 1998 60 1,173,282 1,034,360 138,922 0,89 Paid 1998 72 80,6542 612,0652 612,065 194,487 1,57	Cnstrnd 0
Paid 1998 36 3,021,114 2,939,870 81,244 0.34 Paid 1998 48 1,488,604 1,689,312 -200,808 -1.27 Paid 1998 60 1,173,282 1,034,360 138,922 0.89 Paid 1998 72 806,542 612,055 194,487 1.57	0
Paid 1998 60 1,173,282 1,034,360 138,922 0.89 Paid 1998 72 806,542 612,055 194,487 1.57	0
Paid 1998 72 806,542 612,055 194,487 1.57	0
	0
Paid 1998 84 594,058 538,316 55,742 1.18 Paid 1998 96 438,193 444,117 -5,924 -0.13	0
Paid 1998 108 316,106 337,302 -21,196 -0.74	0
Paid 1999 12 4,211,880 4,330,869 -118,989 -0.69	0
Paid 1999 24 5,420,600 5,123,455 297,145 0.89 Paid 1999 36 3,118,015 2,964,325 153,690 0.64	0
Paid 1999 48 1,868,494 1,703,364 165,130 1.04	0
Paid 1999 60 1,018,079 1,042,964 -24,885 -0.16 Paid 1999 72 584,906 617,146 -32,240 -0.26	0
Paid 1999 84 531,983 542,793 -10,810 -0.23	0
Paid 1999 108 356,478 340,107 16,371 0.57	0
Paid 2000 12 4,553,584 4,602,979 -49,395 -0.28 Paid 2000 24 5,812,588 5,445,363 367,225 1.07	0
Paid 2000 36 3,342,985 3,150,575 192,410 0.78	0
Paid 2000 48 1,870,185 1,810,387 59,798 0.37 Paid 2000 60 1,144,950 1,108,494 36,456 0.23	0
Paid 2000 72 640,842 655,922 -15,080 -0.12 Paid 2000 84 595,970 576,897 19,073 0.39	0
Paid 2000 96 471,781 475,948 -4,167 -0.09	0
Paid 2001 12 4,556,995 4,489,622 67,373 0.38 Paid 2001 24 5,786,328 5,311,260 475,068 1.40	0
Paid 2001 36 3,418,250 3,072,986 345,264 1.41	0
Paid 2001 48 1,858,209 1,765,803 92,406 0.57 Paid 2001 60 738,292 1,081,195 -342,903 -2.16	0
Paid 2001 72 442,905 639,769 -196,864 -1.56	0
Paid 2001 84 488,139 562,690 -74,551 -1.55 Paid 2002 12 4,262,115 4,330,245 -68,130 -0.39	0
Paid 2002 24 5,263,681 5,122,716 140,965 0.42 Paid 2002 36 3,002,075 2,963,898 38,177 0.16	0
Paid 2002 48 1,649,991 1,703,119 -53,128 -0.34	0
Paid 2002 60 1,106,736 1,042,814 63,922 0.41 Paid 2002 72 614,683 617,057 -2,374 -0.02	0
Paid 2003 12 4,274,440 4,408,332 -133,892 -0.77	0
Paid 2003 36 2,938,488 3,017,345 -78,857 -0.32	0
Paid 2003 48 1,747,993 1,733,831 14,162 0.09 Paid 2003 60 1,145,332 1,061,619 83,713 0.53	0
Paid 2004 12 4,624,395 4,478,997 145,398 0.83	0
Paid 2004 24 5,174,240 5,298,691 -124,451 -0.37 Paid 2004 36 2,674,991 3,065,713 -390,722 -1.60	0
Paid 2004 48 1,660,882 1,761,624 -100,742 -0.63	0
Paid 2005 24 5,081,513 5,549,408 -467,895 -1.35	0
Paid 2005 36 2,842,925 3,210,773 -367,848 -1.48 Paid 2006 12 5,130,174 5,236,686 -106,512 -0.56	0
Paid 2006 24 5,593,828 6,195,043 -601,215 -1.66	0
Paid 2007 12 5,211,936 5,508,761 -296,825 -1.54 Paid Constraint 0 0 0 0.00	0
Incd 1998 12 10,440,449 10,388,110 52,339 0.15 Incd 1998 24 4,086,220 4,135,376 -49,156 -0.10	0
Incd 1998 36 1.688.495 1.735.893 -47.398 -0.21	0
Incd 1998 48 1,044,239 928,957 115,282 0.55 Incd 1998 60 851,747 618,070 233,677 1.12	0
Incd 1998 72 616,672 413,144 203,528 1.82	0
Incd 1998 84 420,021 365,835 54,186 0.67 Incd 1998 96 321,247 273,968 47,279 0.62	0
Incd 1998 108 71,684 157,407 -85,723 -1.25	0
Incd 1999 12 10,104,076 10,474,522 -370,446 -1.06	0
Incd 1999 24 4,262,241 4,169,776 92,465 0.19 Incd 1999 36 2,008,640 1,750,333 258,307 1.15	0
Incd 1999 48 1,266,374 936,684 329,690 1.55	0
Incd 1999 60 765,915 623,211 142,704 0.68 Incd 1999 72 450,157 416,581 33,576 0.30	0
Incd 1999 84 479,312 368,878 110,434 1.37	0
Incd 1999 96 211,550 276,247 -64,697 -0.84 Incd 1999 108 264,671 158,716 105,955 1.54	0
Incd 2000 12 10,614,330 11,132,639 -518,309 -1.44 Incd 2000 24 5,087,335 4,431,764 655,571 1.33	0
Incd 2000 36 2,000,022 1,860,307 139,715 0.60	0
Incd 2000 48 1,142,896 995,536 147,360 0.68 Incd 2000 60 824,094 662,368 161,726 0.75	0
Incd 2000 72 425,884 442,755 -16,871 -0.15	0
Incd 2000 84 307,470 392,055 -84,585 -1.03 Incd 2000 96 337,415 293,604 43,811 0.56	0
Incd 2001 12 11,104,926 10,858,477 246,449 0.69 Incd 2001 24 4,741,998 4,322,623 419,375 0.86	0
Incd 2001 36 2,116,895 1,814,493 302,402 1.32	0
Incd 2001 48 969,052 971,019 -1,967 -0.01 Incd 2001 60 246,184 646,056 -399,872 -1.88	0
Incd 2001 72 305,161 431,851 -126,690 -1.11	0
Incd 2002 12 10,379,583 10,473,012 -93,429 -0.27	0
Incd 2002 24 4,729,077 4,169,174 559,903 1.17 Incd 2002 36 1,886,096 1,750,080 136,016 0.60	0
Incd 2002 48 691,197 936,549 -245,352 -1.16	0
Incd 2002 60 548,041 623,121 -75,080 -0.36 Incd 2002 72 355,685 416,521 -60,836 -0.54	0
Incd 2003 12 10,932,703 10,661,870 270,833 0.77	0
Incd 2003 36 1,574,142 1,781,639 -207,497 -0.91	0
Incd 2003 48 797,113 953,438 -156,325 -0.73 Incd 2003 60 618,675 634,358 -15,683 -0.07	0
Incd 2004 12 11,239,343 10,832,779 406,564 1.14	0
Incd 2004 24 4,081,055 4,312,393 -231,338 -0.48 Incd 2004 36 1,522,852 1,810,199 -287,347 -1.26	0
Incd 2004 48 788,922 968,721 -179,799 -0.84	0
Incd 2005 12 11,978,411 11,345,353 633,058 1.75 Incd 2005 24 3,653,908 4,516,442 -862,534 -1.74	0
Incd 2005 36 1,588,938 1,895,852 -306,914 -1.32 Incd 2006 12 12,468,437 12,665,306 -196,869 -0.52	0
Incd 2006 24 4,353,742 5,041,899 -688,157 -1.32	0
Incd 2007 12 12,931,177 13,323,337 -392,160 -1.01 Incd Constraint 0 0 0 0.00	0
Ult = 1998 0 -2,019 2,019 0.00	0
Ult = 2000 0 -2,163 2,163 0.00	0
Ult = 2001 0 -2,110 2,110 0.00 Ult = 2002 0 -2,035 2,035 0.00	0
Ult = 2003 0 -2,072 2,072 0.00	0
Ult = 2004 0 -2,105 2,105 0.00 Ult = 2005 0 -2,205 2,205 0.00	0
Ult = 2006 0 -2,461 2,461 0.00	0
Ult = 2007 0 -2,589 2,589 0.00	0

SSCP	TTy1	Fitted	Resid
TTy1	21797.28	21712.24	85.03434
Fitted	21712.24	21712.24	0
Resid	85.03434	0	85.03434
rhosq	100.0%	99.6%	0.4%
df	120	20	100
s2hat	181.644	1085.612	0.850343
s2sel			1

Exhibit 9.2

Joint Paid and Incurred Linear Model: Predictions

TYPE	AY AGE	y2hat	StdPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr	VarPrdErr
Paid	1998 Unpd	2,999,915	± 142,339	2.03E+10	3.9E+08	5.28E+08	5.95E+08	6.3E+08	8.06E+08	1.12E+09	1.43E+09	2.04E+09	3.23E+09	2.03E+10	3.9E+08	5.28E+08	5.95E+08	6.3E+08	8.06E+08	1.12E+09	1.43E+09	2.04E+09	3.23E+09
Paid	1999 Unpd	3,398,227	± 158,471	3.9E+08	2.51E+10	2.53E+09	2.43E+09	2.34E+09	2.60E+09	2.88E+09	3.27E+09	4.21E+09	5.59E+09	3.9E+08	2.51E+10	2.53E+09	2.43E+09	2.34E+09	2.60E+09	2.88E+09	3.27E+09	4.21E+09	5.59E+09
Paid	2000 Unpd	3,804,647	± 177,631	5.28E+08	2.53E+09	3.16E+10	3.96E+09	3.77E+09	4.09E+09	4.37E+09	4.85E+09	6.08E+09	7.67E+09	5.28E+08	2.53E+09	3.16E+10	3.96E+09	3.77E+09	4.09E+09	4.37E+09	4.85E+09	6.08E+09	7.67E+09
Paid	2001 Unpd	4,227,840	± 186,625	5.95E+08	2.43E+09	3.96E+09	3.48E+10	4.60E+09	4.95E+09	5.22E+09	5.73E+09	7.08E+09	8.71E+09	5.95E+08	2.43E+09	3.96E+09	3.48E+10	4.60E+09	4.95E+09	5.22E+09	5.73E+09	7.08E+09	8.71E+09
Paid	2002 Unpd	4,662,500	± 192,377	6.3E+08	2.34E+09	3.77E+09	4.60E+09	3.70E+10	5.45E+09	5.71E+09	6.24E+09	7.64E+09	9.26E+09	6.3E+08	2.34E+09	3.77E+09	4.60E+09	3.70E+10	5.45E+09	5.71E+09	6.24E+09	7.64E+09	9.26E+09
Paid	2003 Unpd	5,421,105	± 220,945	8.06E+08	2.60E+09	4.09E+09	4.95E+09	5.45E+09	4.88E+10	7.87E+09	8.50E+09	1.02E+10	1.19E+10	8.06E+08	2.60E+09	4.09E+09	4.95E+09	5.45E+09	4.88E+10	7.87E+09	8.50E+09	1.02E+10	1.19E+10
Paid	2004 Unpd	6,568,927	± 272,535	1.12E+09	2.88E+09	4.37E+09	5.22E+09	5.71E+09	7.87E+09	7.43E+10	1.25E+10	1.46E+10	1.67E+10	1.12E+09	2.88E+09	4.37E+09	5.22E+09	5.71E+09	7.87E+09	7.43E+10	1.25E+10	1.46E+10	1.67E+10
Paid	2005 Unpd	8,692,524	± 320,964	1.43E+09	3.27E+09	4.85E+09	5.73E+09	6.24E+09	8.50E+09	1.25E+10	1.03E+11	1.92E+10	2.15E+10	1.43E+09	3.27E+09	4.85E+09	5.73E+09	6.24E+09	8.50E+09	1.25E+10	1.03E+11	1.92E+10	2.15E+10
Paid	2006 Unpd	13,144,841	± 403,064	2.04E+09	4.21E+09	6.08E+09	7.08E+09	7.64E+09	1.02E+10	1.46E+10	1.92E+10	1.62E+11	3.07E+10	2.04E+09	4.21E+09	6.08E+09	7.08E+09	7.64E+09	1.02E+10	1.46E+10	1.92E+10	1.62E+11	3.07E+10
Paid	2007 Unpd	20,392,908	± 549,291	3.23E+09	5.59E+09	7.67E+09	8.71E+09	9.26E+09	1.19E+10	1.67E+10	2.15E+10	3.07E+10	3.02E+11	3.23E+09	5.59E+09	7.67E+09	8.71E+09	9.26E+09	1.19E+10	1.67E+10	2.15E+10	3.07E+10	3.02E+11
Incd	1998 IBNR	973,844	± 142,339	2.03E+10	3.9E+08	5.28E+08	5.95E+08	6.3E+08	8.06E+08	1.12E+09	1.43E+09	2.04E+09	3.23E+09	2.03E+10	3.9E+08	5.28E+08	5.95E+08	6.3E+08	8.06E+08	1.12E+09	1.43E+09	2.04E+09	3.23E+09
Incd	1999 IBNR	1,142,547	± 158,471	3.9E+08	2.51E+10	2.53E+09	2.43E+09	2.34E+09	2.60E+09	2.88E+09	3.27E+09	4.21E+09	5.59E+09	3.9E+08	2.51E+10	2.53E+09	2.43E+09	2.34E+09	2.60E+09	2.88E+09	3.27E+09	4.21E+09	5.59E+09
Incd	2000 IBNR	1,498,084	± 177.630	5.28E+08	2.53E+09	3.16E+10	3.96E+09	3.77E+09	4.09E+09	4.37E+09	4.85E+09	6.08E+09	7.67E+09	5.28E+08	2.53E+09	3.16E+10	3.96E+09	3.77E+09	4.09E+09	4.37E+09	4.85E+09	6.08E+09	7.67E+09
Incd	2001 IBNR	1.705.511	± 186.625	5.95E+08	2.43E+09	3.96E+09	3.48E+10	4.60E+09	4.95E+09	5.22E+09	5.73E+09	7.08E+09	8.71E+09	5.95E+08	2.43E+09	3.96E+09	3.48E+10	4.60E+09	4.95E+09	5.22E+09	5.73E+09	7.08E+09	8.71E+09
Incd	2002 IBNR	1.972.102	± 192.377	6.3E+08	2.34E+09	3.77E+09	4.60E+09	3.70E+10	5.45E+09	5.71E+09	6.24E+09	7.64E+09	9.26E+09	6.3E+08	2.34E+09	3.77E+09	4.60E+09	3.70E+10	5.45E+09	5.71E+09	6.24E+09	7.64E+09	9.26E+09
Incd	2003 IBNR	2,390,298	± 220,945	8.06E+08	2.60E+09	4.09E+09	4.95E+09	5.45E+09	4.88E+10	7.87E+09	8.50E+09	1.02E+10	1.19E+10	8.06E+08	2.60E+09	4.09E+09	4.95E+09	5.45E+09	4.88E+10	7.87E+09	8.50E+09	1.02E+10	1.19E+10
Incd	2004 IBNR	3.071.266	± 272.536														5.22E+09						
Incd	2005 IBNR	4.261.067	± 320.964														5.73E+09						
Incd	2006 IBNR	7.046.664															7.08E+09						
Incd	2007 IBNR	12.673.667	± 549.291														8.71E+09						
		,,	,																				
Betahat	StdBeta VarBeta	a VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	VarBeta	
0.148	± 0.002 3.59E-06	6 -5.2E-08	-3.5E-08	-2E-08	-2.7E-08	-2.8E-08	-7E-09	-1.2E-08	-1.3E-08	-1.7E-08	5.74E-07	1.5E-08	3.5E-08	0	1.6E-08	2.6E-08	1.4E-08	1.5E-08	2.8E-08	5.8E-08	1.07E-07	1.7E-08	
		6 -5.2E-08	-3.5E-08		-2.7E-08	-2.8E-08							3.5E-08	0				1.5E-08		5.8E-08	1.07E-07		
0.148	± 0.002 3.59E-0	6 -5.2E-08 3 1.48E-05	-3.5E-08 -2.93E-07	-2E-08 -1.49E-07	-2.7E-08	-2.8E-08 -1.61E-07	-7E-09	-1.2E-08	-1.3E-08 -6E-08	-1.7E-08	5.74E-07	1.5E-08	3.5E-08	0	1.6E-08	2.6E-08	1.4E-08	1.5E-08	2.8E-08 1.37E-07	5.8E-08	1.07E-07 4.74E-07	1.7E-08	
0.148 0.175	± 0.002 3.59E-00 ± 0.004 -5.2E-00	5 -5.2E-08 3 1.48E-05 3 -2.93E-07	-3.5E-08 -2.93E-07 8.96E-06	-2E-08 -1.49E-07	-2.7E-08 -1.82E-07 -1.89E-07	-2.8E-08 -1.61E-07	-7E-09 -3.7E-08	-1.2E-08 -5.9E-08	-1.3E-08 -6E-08 -4.3E-08	-1.7E-08 -7.4E-08	5.74E-07 2.33E-06	1.5E-08 6.6E-08	3.5E-08 7.11E-07	0 1.75E-07 2.52E-07	1.6E-08 1.82E-07	2.6E-08 2.25E-07	1.4E-08 9.5E-08	1.5E-08 8.2E-08	2.8E-08 1.37E-07	5.8E-08 2.68E-07	1.07E-07 4.74E-07 3.26E-07	1.7E-08 1.27E-07	
0.148 0.175 0.101	± 0.002 3.59E-00 ± 0.004 -5.2E-00 ± 0.003 -3.5E-00	6 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07	-3.5E-08 -2.93E-07 8.96E-06	-2E-08 -1.49E-07 -1.68E-07 4.6E-06	-2.7E-08 -1.82E-07 -1.89E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07	-7E-09 -3.7E-08 -3.1E-08	-1.2E-08 -5.9E-08 -4.6E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08	-1.7E-08 -7.4E-08 -5.1E-08	5.74E-07 2.33E-06 1.35E-06	1.5E-08 6.6E-08 4.4E-08	3.5E-08 7.11E-07 3.97E-07	0 1.75E-07 2.52E-07 1.16E-07	1.6E-08 1.82E-07 2.44E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07	1.4E-08 9.5E-08 1E-07	1.5E-08 8.2E-08 7.4E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07	1.7E-08 1.27E-07 1.06E-07	
0.148 0.175 0.101 0.058	± 0.002 3.59E-00 ± 0.004 -5.2E-00 ± 0.003 -3.5E-00 ± 0.002 -2E-00	6 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.94E-07	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07	0 1.75E-07 2.52E-07 1.16E-07	1.6E-08 1.82E-07 2.44E-07 2.14E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07	1.4E-08 9.5E-08 1E-07 7.9E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08	
0.148 0.175 0.101 0.058 0.036	± 0.002 3.59E-00 ± 0.004 -5.2E-00 ± 0.003 -3.5E-00 ± 0.002 -2E-00 ± 0.002 -2.7E-00	6 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07 3 -1.61E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.94E-07 4.13E-06	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07	1.6E-08 1.82E-07 2.44E-07 2.14E-07 2.24E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.96E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08	
0.148 0.175 0.101 0.058 0.036 0.021	± 0.002 3.59E-00 ± 0.004 -5.2E-00 ± 0.003 -3.5E-00 ± 0.002 -2E-00 ± 0.002 -2.7E-00 ± 0.002 -2.8E-00	6 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07 3 -1.61E-07 9 -3.7E-08	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07 -3.1E-08	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.94E-07 4.13E-06	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 8.35E-07	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.6E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08	1.6E-08 1.82E-07 2.44E-07 2.14E-07 2.24E-07 1.59E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 2.94E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07 2.08E-07 5.7E-08	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.96E-07 7.8E-08	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08	
0.148 0.175 0.101 0.058 0.036 0.021 0.019	± 0.002 3.59E-0 ± 0.004 -5.2E-0 ± 0.003 -3.5E-0 ± 0.002 -2E-0 ± 0.002 -2.7E-0 ± 0.002 -2.8E-0 ± 0.002 -2.8E-0 ± 0.001 -7E-0	5 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07 3 -1.61E-07 9 -3.7E-08 3 -5.9E-08	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07 -3.1E-08 -4.6E-08	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.94E-07 4.13E-06 -4.3E-08	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 8.35E-07	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08 -1.5E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08 -2.2E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.6E-08 -1.2E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08	1.6E-08 1.82E-07 2.44E-07 2.14E-07 2.24E-07 1.59E-07 2.8E-08	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 2.94E-07 5.2E-08	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07 2.08E-07 5.7E-08	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.96E-07 7.8E-08	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08	
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0.148 0.175 0.101 0.058 0.036 0.021 0.019 0.015 0.012 0.011	$\begin{array}{c} \pm 0.002 & 3.59E-0\\ \pm 0.004 & -5.2E-0\\ \pm 0.003 & -3.5E-0\\ \pm 0.002 & -2E-0\\ \pm 0.002 & -2E-0\\ \pm 0.002 & -2.8E-0\\ \pm 0.001 & -7E-0\\ \pm 0.001 & -7E-0\\ \pm 0.001 & -1.3E-0\\ \pm 0.001 & -1.3E-0\\ \pm 0.001 & -1.7E-0\\ \end{array}$	5 -5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07 3 -1.61E-07 3 -3.7E-08 3 -5.9E-08 3 -6E-08 3 -7.4E-08 7 2.328E-06	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -3.1E-08 -4.6E-08 -4.3E-08 -5.1E-08 1.35E-06	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 6.6E-07	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.8E-08 -4.1E-08 -4.4E-08	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.94E-07 4.13E-06 -4.3E-08 -5.5E-08 -4.4E-08 -4.6E-08	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 -1.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08 -1.5E-08 1.11E-06 -2.2E-08 -2.1E-08 1.35E-07	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.6E-08 -1.2E-08 -2.1E-08 -2.3E-08 9.14E-07	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.35E-07 1.07E-07 1.01E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.8E-08 4.5E-08 5.2E-08	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08 2.1E-08 1.8E-08 1.8E-08	1.6E-08 1.82E-07 2.44E-07 2.14E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 2.94E-07 5.2E-08 6.3E-08 4.8E-08 4.6E-08	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.6E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.8E-08 2.7E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08 7.2E-08 5.4E-08 5.E-08 1.4E-07	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.06E-07 2.42E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.96E-07 7.8E-08 1.38E-07 1.5E-07 1.95E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08 3E-08 2.8E-08 3E-08 3E-08	
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0.148 0.175 0.101 0.058 0.036 0.021 0.019 0.015 0.012 0.011 0.101 0.357 0.142 0.060	$\begin{array}{c} \pm 0.002 & 3.59E{-}0\\ \pm 0.004 & -5.2E{-}0\\ \pm 0.003 & -3.5E{-}0\\ \pm 0.003 & -3.5E{-}0\\ \pm 0.002 & -2E{-}0\\ \pm 0.002 & -2.7E{-}0\\ \pm 0.002 & -2.7E{-}0\\ \pm 0.001 & -7E{-}0\\ \pm 0.001 & -1.2E{-}0\\ \pm 0.001 & -1.2E{-}0\\ \pm 0.001 & -1.7E{-}0\\ \pm 0.001 & -1.7E{-}0\\ \pm 0.001 & -1.7E{-}0\\ \pm 0.001 & -3.5E{-}0\\ \pm 0.004 & -3.5E{-}0\\ \end{array}$	5 -5.2E-08 3 1.448E-05 3 -2.93E-07 3 -1.49E-07 3 -1.49E-07 3 -1.61E-07 9 -3.7E-08 8 -5.9E-08 3 -6E-08 3 -7.4E-08 7 2.328E-06 8 6.10E-07 9 3.71E-07 1.75E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.3E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 -3.1E-08 6.6E-07 2.5E-08 1.89E-07 1.16E-07	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.8E-08 -4.4E-08 -4.4E-08 7.41E-07 3.5E-08 2.1E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.94E-07 -1.94E-07 -4.13E-06 -4.3E-08 -4.3E-08 -5.5E-08 -4.4E-08 5.58E-07 3.5E-08 1.65E-07 8.9E-08	-7E-09 -3.7E-08 -3.1E-08 -3.6E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 1.05E-07 0 3.4E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08 -1.5E-08 1.11E-06 -2.2E-08 -2.1E-08 1.35E-07 1.6E-08 4.8E-08	-1.3E-08 -6E-08 -4.3E-08 -4.3E-08 -4.1E-08 -4.4E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 1.8E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -4.6E-08 -2.1E-08 -2.3E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.07E-07 1.01E-07 1.13E-06 4.8E-08 3.2E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.4E-08 0	3.5E-08 7.11E-07 3.97E-07 1.89E-07 1.65E-07 3.4E-08 4.8E-08 4.5E-08 5.2E-08 3.2E-07 -4.4E-08 3.04E-05 -2.93E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08 2.1E-08 1.8E-08 1.8E-08 1.43E-07 0 -2.93E-07	1.6E-08 1.82E-07 2.44E-07 2.24E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 1.96E-07 -2E-08 -2.81E-07 -1.89E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 2.94E-07 5.2E-08 6.3E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 2.84E-07 -3.3E-08 -3.1E-07 -1.98E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 2.6E-08 2.6E-08 1.22E-07 -1.8E-08 -1.11E-07	1.5E-08 8.2E-08 7.4E-08 9.2E-08 9.2E-08 3.8E-08 2.8E-08 2.8E-08 9.3E-08 9.3E-08 -1.8E-08 -8E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 1.4E-07 -3.5E-08 -1.17E-07	5.8E-08 2.68E-07 1.97E-07 1.26E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.06E-07 2.42E-07 -7.3E-08 -2.04E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.96E-07 7.8E-08 1.38E-07 1.5E-07 1.95E-07 -1.35E-07 -3.31E-07 -1.19E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 3E-08 3E-08 3E-08 3E-08 1.04E-07 7.51E-07 1.51E-06	
0.148 0.175 0.101 0.058 0.036 0.021 0.019 0.015 0.012 0.011 0.357 0.142 0.600 0.032	± 0.002 3.59E-00 ± 0.004 -5.2E-01 ± 0.003 -3.5E-01 ± 0.002 -3.5E-01 ± 0.002 -2.7E-01 ± 0.002 -2.8E-01 ± 0.001 -7.E-00 ± 0.001 -1.2E-01 ± 0.001 -1.7E-00 ± 0.001 -1.7E-00 ± 0.001 -1.7E-00 ± 0.004 1.5E-00 ± 0.006 3.5E-00	5 -5.2E-08 3 1.48E-05 6 -2.93E-07 3 -1.49E-07 3 -1.61E-07 3 -1.61E-07 3 -5.9E-08 3 -5.9E-08 3 -7.4E-08 7 2.328E-06 3 6.6E-08 3 7.11E-07 1 1.75E-07 1 1.82E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.51E-07 -3.1E-08 -4.3E-08 -4.3E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.44E-07	-2E-08 -1.49E-07 -1.68E-07 -1.68E-07 -1.5E-07 -2.2E-08 -3.1E-08 6.6E-07 2.5E-08 1.89E-07 1.16E-07 2.14E-07	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.8E-08 -4.4E-08 7.41E-07 3.5E-08 2.1E-07 1.23E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.14E-07 4.13E-06 -4.3E-08 -5.5E-08 -4.4E-08 -4.4E-08 5.58E-07 3.5E-08 1.65E-07 8.9E-08 1.59E-07	-7E-09 -3.7E-08 -3.1E-08 -3.6E-08 -3.6E-08 -4.3E-08 -4.3E-08 -1.2E-08 -1.2E-08 1.05E-07 0 3.4E-08 1.7E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08 -1.5E-08 1.11E-06 -2.2E-08 1.35E-07 1.6E-08 4.8E-08 2.1E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 1.8E-08 2.7E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.6E-08 -1.2E-08 -2.3E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.07E-07 1.01E-07 1.13E-06 4.8E-08 3.2E-07 1.43E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.4E-08 0	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.5E-08 5.2E-08 3.2E-07 -4.4E-08 3.04E-05 3.04E-05 -2.93E-07 -2.81E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08 2.1E-08 1.8E-08 1.8E-08 1.43E-07 0 -2.93E-07 7.95E-06	1.6E-08 1.82E-07 2.44E-07 2.24E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 1.96E-07 -2E-08 -2.81E-07 -1.89E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 5.2E-08 6.3E-08 4.8E-08 4.8E-08 2.84E-07 -3.3E-08 -3.1E-07 -3.7E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 2.6E-08 2.6E-08 1.22E-07 -1.8E-08 -1.11E-07 -6.5E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.8E-08 9.3E-08 8E-08 -8E-08 -8E-08 -4.2E-08 -7.4E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 1.4E-07 -3.5E-08 -1.17E-07 -5.4E-08 -9.1E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.06E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07	1.07E-07 4.74E-07 3.26E-07 2.82E-07 2.96E-07 7.8E-08 1.38E-07 1.5E-07 1.95E-07 3.8E-07 -1.35E-07 -3.31E-07 -1.19E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08 3E-08 2.8E-08 3E-08 3.28E-08 3.28E-08 3.207 7.51E-07 1.51E-06 3.63E-07	
0.148 0.175 0.101 0.058 0.036 0.021 0.019 0.015 0.012 0.011 0.101 0.357 0.142 0.060	$\begin{array}{c} \pm 0.002 & 3.59E\text{-}0\\ \pm 0.004 & -5.2E\text{-}0\\ \pm 0.003 & -3.5E\text{-}0\\ \pm 0.002 & -3.5E\text{-}0\\ \pm 0.002 & -2E\text{-}0\\ \pm 0.002 & -2.7E\text{-}0\\ \pm 0.001 & \pm 0.001 & -2.8E\text{-}0\\ \pm 0.001 & -1.0E\text{-}0\\ \pm 0.001 & -1.3E\text{-}0\\ \pm 0.001 & -1.3E\text{-}0\\ \pm 0.001 & -1.7E\text{-}0\\ \pm 0.001 & -1.7E\text{-}0\\ \pm 0.001 & 5.5E\text{-}0\\ \pm 0.003 & -1.6E\text{-}0\\ \pm 0.003 & 1.6E\text{-}0\\ \end{array}$	-5.2E-08 1.48E-05 2.93E-07 -1.49E-07 3.1.61E-07 3.1.61E-07 9.3.7E-08 3.5.9E-08 3.66E-08 3.74E-08 7.2328E-06 8.66E-08 7.11E-07 1.75E-07 1.82E-07 1.82E-07 3.182E-07 2.25E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.51E-07 -3.1E-08 -4.3E-08 -4.3E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.44E-07	-2E-08 -1.49E-07 -1.68E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 -3.1E-08 -3.1E-08 -3.1E-08 1.89E-07 1.16E-07 2.14E-07 2.14E-07 2.3E-07	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.8E-08 -4.4E-08 -4.4E-08 -4.4E-08 7.41E-07 3.5E-08 2.1E-07 1.23E-07 2.24E-07 4.18E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.94E-07 -1.94E-07 -1.94E-07 -1.94E-07 -4.3E-08 -5.5E-08 -4.4E-08 5.58E-07 3.5E-08 1.65E-07 8.9E-08 1.59E-07 2.94E-07	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 1.05E-07 0 3.4E-08 1.7E-08 2.8E-08	-1.2E-08 -5.9E-08 -3.1E-08 -3.1E-08 -4.8E-08 -5.5E-08 1.1E-08 1.1E-08 1.35E-07 1.6E-08 4.8E-08 4.8E-08 3.5E-08	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 1.8E-08 2.7E-08 4.8E-08	-1.7E-08 -7.4E-08 -3.1E-08 -3.1E-08 -4.4E-08 -4.6E-08 -1.2E-08 -2.3E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08 2.7E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.01E-07 1.01E-07 1.13E-06 4.8E-08 3.2E-07 1.43E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.47E-05 -4.47E-05 -4.47E-08 -3.3E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.5E-08 5.2E-08 3.2E-07 -4.4E-08 3.04E-05 3.04E-05 -2.93E-07 -2.81E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08 1.7E-08 1.8E-08 1.8E-08 1.43E-07 0 -2.93E-07 -1.98E-07 -1.98E-07	1.6E-08 1.82E-07 2.44E-07 2.24E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 1.96E-07 -2E-08 1.96E-07 -2.81E-07 8.12E-06	2.6E-08 2.25E-07 2.72E-07 2.3E-07 2.34E-07 5.2E-08 6.3E-08 4.8E-08 4.8E-08 4.6E-08 2.84E-07 -3.3E-08 -3.1E-07 -3.7E-07 9.21E-06	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.6E-08 1.22E-07 -1.8E-08 -1.11E-07 -6.5E-08 -1.19E-07 -2.22E-07	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.8E-08 9.3E-08 8E-08 -8E-08 -8E-08 -4.2E-08 -7.4E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 1.4E-07 -3.5E-08 -1.17E-07 -5.4E-08 -9.1E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.89E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.06E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.82E-07 2.82E-07 7.8E-08 1.38E-07 1.5E-07 3.8E-07 -1.35E-07 -1.35E-07 -1.35E-07 -1.35E-07 -1.77E-07 -3-80-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 3E-08 3E-08 3E-08 3E-08 3E-08 3E-08 1.04E-07 7.51E-07 3.51E-07 3.51E-07	
0.148 0.175 0.101 0.058 0.036 0.021 0.015 0.012 0.011 0.357 0.142 0.060 0.032 0.021 0.014	$\begin{array}{c} \pm 0.002 & 3.59E \cdot 0 \\ \pm 0.004 & 5.2E \cdot 0 \\ \pm 0.003 & 3.5E \cdot 0 \\ \pm 0.002 & -3.5E \cdot 0 \\ \pm 0.002 & -2.E \cdot 0 \\ \pm 0.002 & -2.8E \cdot 0 \\ \pm 0.002 & -2.8E \cdot 0 \\ \pm 0.001 & -7.E \cdot 0 \\ \pm 0.001 & -7.E \cdot 0 \\ \pm 0.001 & -1.3E \cdot 0 \\ \pm 0.001 & -1.5E \cdot 0 \\ \pm 0.003 & -1.6E \cdot 0 $	-5.2E-08 3 1.48E-05 3 -2.93E-07 3 -1.49E-07 3 -1.82E-07 3 -1.61E-07 9 -3.7E-08 3 -5.9E-08 3 -7.4E-08 7 2.328E-06 3 6.6E-08 3 7.11E-07 1.75E-07 3 3 2.25E-07 9 9.5E-08	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.5E-08 -4.3E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.52E-07 2.72E-07 1E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 6.6E-07 2.5E-08 1.89E-07 1.16E-07 2.14E-07 2.5E-08 1.89E-07 7.9E-08	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 -1.5E-07 -3.6E-08 -4.8E-08 -4.4E-08 -4.4E-08 -4.4E-08 2.1E-07 1.23E-07 2.24E-07 4.18E-07 1.39E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.14E-07 -1.94E-07 4.13E-06 -4.3E-08 -5.5E-08 -4.4E-08 -4.6E-08 5.58E-07 3.5E-08 1.65E-07 8.9E-08 1.65E-07 2.94E-07 2.94E-07	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 -1.2E-08 1.05E-07 0 3.4E-08 1.7E-08 2.8E-08 5.2E-08 3E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -3.1E-08 -5.5E-08 -1.5E-08 -1.11E-06 -2.2E-08 -3.5E-08 -3	-1.3E-08 -6E-08 -4.3E-08 -2.7E-08 -4.4E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 1.8E-08 2.7E-08 4.8E-08 2.7E-08	-1.7E-08 -7.4E-08 -7.4E-08 -5.1E-08 -3.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -4.4E-08 -2.2E-08 -2.2E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08 2.7E-08 4.6E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.35E-07 1.35E-07 1.35E-07 1.31E-07 4.8E-08 3.2E-07 1.43E-07 2.84E-07 2.84E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.4E-08 0 -2E-08 -3.3E-08 -1.8E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.8E-08 5.2E-08 3.2E-07 -4.4E-08 3.04E-05 -2.93E-07 -2.81E-07 -3.1E-07 -1.11E-07	0 1.75E-07 2.52E-07 1.16E-07 1.23E-07 8.9E-08 1.7E-08 2.1E-08 1.8E-08 1.8E-08 1.8E-08 1.43E-07 0 0 -2.93E-07 7.95E-06 -1.89E-07 -1.98E-07 -6.5E-08	1.6E-08 1.82E-07 2.44E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 -2.81E-07 -1.89E-07 8.12E-00 -3.7E-07 -1.19E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 2.94E-07 5.2E-08 6.3E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 -3.3E-07 -3.3E-07 -3.7E-07 9.21E-06 -2.22E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 2.6E-08 1.22E-07 -1.8E-08 -1.11E-07 -6.5E-08 -1.19E-07 -2.22E-07	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 9.3E-08 9.3E-08 8E-08 8E-08 4.2E-08 7.4E-08 36E-07 8E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-07 3.9E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 1.4E-07 -5.4E-08 -9.1E-08 -1.17E-07 -5.4E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.26E-07 2.08E-07 5.7E-08 1.03E-07 1.03E-07 1.03E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07 -2.29E-07	1.07E-07 4.74E-07 3.26E-07 2.82E-07 2.82E-07 2.82E-07 2.82E-07 1.38E-07 1.5E-07 1.5E-07 1.5E-07 -1.35E-07 -1.35E-07 -1.37E-07 -3.21E-07 -1.68E-07	1.7E-08 1.27E-07 1.0EE-07 6.9E-08 9.6E-08 8.8E-08 3E-08 3E-08 3E-08 3E-08 1.04E-07 7.51E-07 1.51E-06 3.63E-07 3.51E-07 3.51E-07	
0.148 0.175 0.101 0.058 0.021 0.015 0.015 0.012 0.011 0.357 0.142 0.060 0.032 0.021 0.014	$\begin{array}{c} \pm 0.002 & 3.59E\text{-}0\\ \pm 0.004 & -5.2E\text{-}0\\ \pm 0.003 & -3.5E\text{-}0\\ \pm 0.002 & -3.5E\text{-}0\\ \pm 0.002 & -2E\text{-}0\\ \pm 0.002 & -2E\text{-}0\\ \pm 0.001 & -2.8E\text{-}0\\ \pm 0.001 & -1.2E\text{-}0\\ \pm 0.001 & -1.3E\text{-}0\\ \pm 0.001 & -1.3E\text{-}0\\ \pm 0.001 & -1.7E\text{-}0\\ \pm 0.001 & -1.7E\text{-}0\\ \pm 0.001 & 5.5E\text{-}0\\ \pm 0.003 & 1.6E\text{-}0\\ \pm 0.002 & 1.5E\text{-}0\\ \pm 0.002 & 1.5E\text{-}0\\ \end{array}$	-5.2E-08 1.48E-05 2.93E-07 -1.49E-07 -3.7E-08 -5.9E-08 -6.608 -7.4E-08 -6.608 -7.4E-08 -6.608 -7.4E-08	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.68E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.6E-08 -4.3E-08 -5.1E-08 -3.97E-07 2.52E-07 2.52E-07 2.52E-07 7.24E-07 7.4E-08	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 6.6E-07 2.5E-08 1.89E-07 1.16E-07 2.14E-07 2.3E-08 5.5E-08	-2.7E-08 -1.82E-07 -1.82E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.8E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-07 3.6E-08 2.1E-07 1.23E-07 1.23E-07 9.2E-08	-2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.14E-07 -1.94E-07 -1.94E-07 -1.94E-07 -1.94E-07 -3.5E-08 1.65E-07 3.5E-08 1.65E-07 3.5E-08 1.65E-07 2.94E-07 1.72E-07 1.1E-07	-7E-09 -3.7E-08 -3.1E-08 -3.6E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.7E-08 2.8E-08 5.2E-08 3.2E-08 3.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -3.1E-08 -4.8E-08 -5.5E-08 -1.5E-08 1.11E-06 -2.2E-08 1.35E-07 1.6E-08 4.8E-08 3.5E-08 3.6E-08 3.8E-08	-1.3E-08 -6E-08 -4.3E-08 -4.3E-08 -4.1E-08 -2.7E-08 -2.2E-08 9.17E-07 -2.3E-08 9.17E-07 1.7E-08 4.5E-08 1.8E-08 2.7E-08 2.7E-08 2.8E-08 2.8E-08	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -4.4E-08 -2.1E-08 -2.1E-08 -2.2E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08 2.7E-08 4.6E-08 2.7E-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.07E-07 1.07E-07 1.07E-07 1.01E-07 1.32E-07 1.43E-07 1.43E-07 1.43E-07 1.43E-07 1.43E-07 2.84	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.4E-08 0 -2E-08 -3.3E-08 -1.8E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 5.2E-08 3.2E-07 -4.4E-08 3.04E-05 -2.93E-07 -2.93E-07 -3.1E-07 -1.11E-07	0 1.75E-07 2.52E-07 1.6E-07 1.23E-07 1.23E-07 1.7E-08 2.1E-08 1.8E-08 1.8E-08 1.8E-08 1.8E-08 1.43E-07 7.95E-06 -1.89E-07 -6.5E-08 -4.2E-08	1.6E-08 1.82E-07 2.44E-07 2.24E-07 2.24E-07 2.24E-07 2.24E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 2.7E-08 1.96E-07 -2.2E-08 -2.81E-07 -1.89E-07 -1.89E-07 -1.19E-07	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 2.94E-07 5.2E-08 6.3E-08 4.8E-08 4.6E-08 4.6E-08 2.84E-07 -3.3E-07 -3.7E-07 9.21E-06 -2.22E-07 -1.38E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.6E-08 2.6E-08 1.22E-07 -1.8E-08 -1.11E-07 -6.5E-08 -1.19E-07 -2.22E-07 3.42E-06 -8E-06	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.7E-08 9.3E-08 9.3E-08 9.3E-08 -8E-08 -4.2E-08 -7.4E-08 -1.36E-07 -8E-08 2.34E-06	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-08 3.9E-08 5.4E-08 5.4E-08 5.5E-08 1.4E-07 -3.5E-08 -1.17E-07 -5.4E-08 -9.1E-08 -9.9E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.26E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.06E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07 -1.36E-07	1.07E-07 4.74E-07 3.26E-07 1.96E-07 2.82E-07 2.82E-07 2.82E-07 7.8E-08 1.38E-07 1.5E-07 3.8E-07 -1.35E-07 -3.31E-07 -1.19E-07 -1.19E-07 -1.168E-07 -1.74E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08 3E-08 3E-08 3E-08 3E-08 3E-08 1.04E-07 7.51E-06 3.63E-07 3.51E-07 3.51E-07 3.51E-07 7.52E-08	
0.148 0.175 0.101 0.058 0.036 0.021 0.015 0.012 0.011 0.101 0.357 0.142 0.060 0.032 0.021 0.014 0.014 0.014 0.013 0.009	$\begin{array}{c} \pm 0.002 & 3.59E \cdot 0. \\ \pm 0.004 & -5.2E \cdot 0. \\ \pm 0.003 & -3.5E \cdot 0. \\ \pm 0.002 & -3.5E \cdot 0. \\ \pm 0.002 & -3.5E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.001 & -7E \cdot 0. \\ \pm 0.001 & -1.2E \cdot 0. \\ \pm 0.001 & -1.2E \cdot 0. \\ \pm 0.001 & -1.3E \cdot 0. \\ \pm 0.001 & -1.3E \cdot 0. \\ \pm 0.001 & -1.5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.002 & -5E \cdot 0.$	-5.2E-08 3 -1.48E-05 -2.93E-07 -1.49E-07 3 -1.49E-07 3 -1.82E-07 3 -1.82E-07 3 -1.82E-07 3 -1.61E-07 3 -5.9E-08 3 -5.9E-08 3 -5.9E-08 3 -7.4E-08 3 7.11E-07 1 1.75E-07 3 9.5E-08 8 8.2E-08 8 8.2E-08 3 1.37E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.89E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.6E-08 -4.3E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.244E-07 2.72E-07 1E-07 7.4E-08 1.11E-07	-2E-08 -1.49E-07 -1.68E-07 -1.68E-07 -1.68E-07 -2.2E-08 -3.1E-08 -3.5E-07 -3.5E-08 -	-2.7E-08 -1.82E-07 -1.89E-07 -1.5E-07 -5.28E-06 -1.94E-07 -3.6E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -3.5E-08 2.1E-07 1.23E-07 2.24E-07 -1.23E-07 9.2E-08 9.2E-08	2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.14E-07 4.13E-06 4.3E-08 -5.5E-08 -4.4E-08 5.58E-07 3.5E-08 1.65E-07 8.9E-08 1.59E-07 1.294E-07 1.294E-07 1.32E-07	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 -4.3E-07 -1.5E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -3.4E-08 -3.4E-08 -2.2E-08 -3.2E-08 -3.2E-08 -3.2E-08 -3.2E-08 -3.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.6E-08 -5.5E-08 -1.5E-08 -2.2E-08 -2.2E-08 -2.2E-08 -2.2E-08 -3.5E-07 -1.6E-08 -3.5E-08 -3.5E-08 -3.8E-08 -3.8E-08 -3.2E-08	-1.3E-08 -6E-08 -4.3E-08 -4.7E-08 -2.7E-08 -4.1E-08 -4.4E-08 -1.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 1.8E-08 2.7	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -4.6E-08 -1.2E-08 -2.3E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08 2.7E-08 2.7E-08 5.2-08	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.07E-07 1.13E-06 4.8E-08 3.2E-07 1.36E-	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 2.1E-08 4.8E-08 1.47E-05 -4.4E-08 0 0 -2E-08 -3.3E-08 -1.8E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 1.85E-07 3.4E-08 4.8E-08 4.5E-08 5.2E-08 3.04E-05 3.2E-07 -4.4E-08 3.04E-05 -2.93E-07 -2.81E-07 -3.1E-07 -1.11E-07 -8.E-08 -1.17E-07	0 1.75E-07 2.52E-07 2.52E-07 1.23E-07 1.23E-07 1.23E-07 1.7E-08 2.1E-08 1.8E-08 1.8E-08 1.8E-08 1.43E-07 0 -2.93E-07 -1.98E-07 -6.5E-08 -4.2E-08 -5.4E-08	1.6E-08 1.82E-07 2.44E-07 2.24E-07 2.24E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.81E-07 -2E-08 8.12E-06 -3.7E-07 -1.19E-07 -7.4E-08 9.1E-08	2.6E-08 2.25E-07 2.72E-07 2.3E-07 4.18E-07 5.2E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 3.1E-07 -3.3E-07 9.21E-06 9.21E-06 9.21E-06 1.36E-07 -1.36E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.7E-08 4.2E-07 -1.8E-08 -1.11E-07 -6.5E-08 -1.119E-07 3.42E-06 -8E-08 9.5E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 2.8E-08 2.8E-08 2.8E-08 2.8E-08 9.3E-08 -3.8E-08 -4.2E-08 -7.4E-08 -7.4E-08 2.34E-06 2.34E-06 -9.9E-08	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-08 3.9E-08 5.4E-08 5.4E-08 5.4E-08 5.4E-08 5.4E-08 -1.4E-07 -3.5E-08 -1.64E-07 -9.5E-08 -9.9E-08 3.06E-08	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.26E-07 2.08E-07 5.7E-08 1.03E-07 1.06E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07 -1.36E-07 -2.29E-07	1.07E-07 4.74E-07 3.26E-07 2.92E-07 2.92E-07 2.92E-07 7.8E-08 1.38E-07 1.95E-07 3.8E-07 -1.35E-07 -1.35E-07 -1.35E-07 -1.17E-07 -1.77E-07 -1.74E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08 3E-08 2.8E-08 3E-08 2.8E-08 3E-08 3.8E-08 3.8E-08 3.8E-08 3.8E-08 3.654E-07 3.654E-07 3.654E-07 3.654E-07 3.654E-07 9.6E-08	
0.148 0.175 0.101 0.058 0.036 0.021 0.015 0.012 0.011 0.357 0.142 0.060 0.032 0.021 0.014 0.014 0.013 0.009 0.005	$\begin{array}{c} \pm 0.002 & 3.59E \cdot 0. \\ \pm 0.002 & 3.59E \cdot 0. \\ \pm 0.003 & 3.5E \cdot 0. \\ \pm 0.003 & 3.5E \cdot 0. \\ \pm 0.002 & -2.E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.001 & -7.E \cdot 0. \\ \pm 0.001 & -7.2E \cdot 0. \\ \pm 0.001 & -1.3E \cdot 0. \\ \pm 0.002 & -1.3E \cdot 0. \\ \pm 0.002 & -1.6E \cdot 0. \\ \pm 0.002 & 5.8E \cdot 0. \\ \pm 0.002 & 5.8E \cdot 0. \\ \end{array}$	5.2E-08 5.2E-08 1.48E-05 2.93E-07 3 -1.49E-07 3 -1.49E-07 3 -1.49E-07 3 -1.49E-07 3 -1.61E-07 3 -5.7E-08 3 -6E-08 4 -6E-08 5 -7.4E-08 -6E-08 -6E-08 6.6E-08 -6.6E-08 7 7.11E-07 1.52E-07 1.52E-07 2.528E-06 -5.5E-08 8.2E-08 8.2E-08 8.2E-08 1.37E-07 2.68E-07 2.68E-07	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.88E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.6E-08 -4.6E-08 -4.6E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.52E-07 2.52E-07 2.72E-07 7.4E-08 1.11E-07 7.4E-08 1.11E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 -3.1E-08 6.6E-07 2.5E-08 1.89E-07 7.9E-08 5.5E-08 7.6E-08 1.26E-07	-2.7E-08 -1.82E-07 -1.82E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.4E-08 -4.4E-08 -4.4E-08 7.41E-07 3.5E-08 2.1E-07 1.23E-07 2.24E-07 4.18E-07 1.39E-07 9.2E-08 1.21E-07	2.8E-08 -1.61E-07 -1.51E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.94E-07 4.13E-08 -4.3E-08 -4.4E-08 5.58E-07 3.5E-08 1.56E-07 8.9E-08 1.56E-07 2.94E-07 1.172E-07 1.172E-07 1.39E-07	-7E-09 -3.7E-08 -3.1E-08 -3.6E-08 -4.3E-08 8.35E-07 -1.5E-08 -1.2E-08 1.05E-07 0 3.4E-08 5.2E-08 3.2E-08 3.2E-08 3.2E-08 3.2E-08 3.2E-08 3.2E-08 5.2E-08 5.2E-08 5.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.8E-08 -5.5E-08 1.11E-06 -2.2E-08 1.35E-07 1.6E-08 1.35E-07 1.6E-08 4.8E-08 3.5E-08 6.3E-08 3.6E-08 3.6E-08 3.8E-08 7.2E-06 1.03E-07	-1.3E-08 -6E-08 -4.3E-08 -4.3E-08 -2.7E-08 -4.4E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 -2.3E-08 1.07E-07 1.7E-08 4.5E-08 2.7E-08 4.8E-08 2.7E-08 2.8E-08 5.4E-08 5.4E-08 1.13E-07	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -2.1E-08 -2.1E-08 -2.2E-08 9.14E-07 1.01E-07 2.1E-08 5.2E-08 1.8E-08 2.7E-08 4.6E-08 2.7E-08 5.2E-08 1.06E-07	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.01E-07 1.13E-06 4.8E-08 3.2E-07 1.43E-07 1.43E-07 9.3E-08 1.42E-07 2.42E-07	1.5E-08 6.6E-08 4.4E-08 2.5E-08 3.5E-08 3.5E-08 3.5E-08 0 1.6E-08 1.7E-08 4.8E-08 1.47E-05 -4.4E-08 -3.3E-08 -1.8E-08 -1.8E-08 -1.8E-08 -3.5E-08 -7.3E-08	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.8E-08 4.8E-08 4.8E-08 3.2E-08 3.2E-08 3.2E-08 3.2E-07 -4.4E-08 3.04E-05 -2.93E-07 -2.81E-07 -1.11E-07 -2.04E-07	0 1.75E-07 2.52E-07 1.26E-07 1.28E-07 8.9E-08 2.1E-08 1.7E-08 1.8E-08 1.8E-08 1.43E-07 0 -2.93E-07 7.95E-06 -1.98E-07 -1.98E-07 -1.98E-07 -5.5E-08 -4.2E-08 -5.4E-08	1.6E-08 1.82E-07 2.44E-07 2.44E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 2.7E-08 1.96E-07 -2E-08 8.122E-08 -3.7E-07 -1.89E-07 -3.7E-07 -1.19E-07 -7.4E-08 -9.1E-08 -1.3E-07	2.6E-08 2.25E-07 2.7ZE-07 2.3E-07 2.94E-07 5.2E-08 6.3E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-08 4.8E-07 -3.7E-07 -1.98E-07 -1.18E-07 -1.18E-07 -1.18E-07 -1.18E-07 -1.18E-07 -1.18E-07 -1.18E-07 -1.28	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.6E-08 1.22E-07 -1.8E-08 1.22E-07 -6.5E-08 -1.19E-07 -2.22E-07 -8.5E-08 -9.5E-08 -1.3E-07	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.7E-08 -8E-08 -4.2E-08 -7.4E-08 -7.4E-08 -7.4E-08 2.34E-06 -9.9E-08 -9.9E-08 -1.36E-07	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 1.39E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 5.4E-08 5.4E-08 1.4E-07 -3.5E-08 -9.1E-08 -9.5E-08 -9.5E-08 -9.5E-08 -9.5E-08 -2.58E-07	5.8E-08 2.68E-07 1.97E-07 1.89E-07 2.08E-07 5.7E-08 1.03E-07 1.13E-07 1.106E-07 2.42E-07 -7.28E-08 -2.04E-07 -8.2E-08 -1.3E-07 -1.3E-07 -1.3E-07 -2.28E-07 -2.28E-07	1.07E-07 4.74E-07 3.26E-07 2.82E-07 2.96E-07 7.8E-08 1.38E-07 1.5E-07 3.8E-07 -1.35E-07 -3.38E-07 -1.35E-07 -3.38E-07 -1.77E-07 -3.28E-07 -1.68E-07 -1.74E-07 -3.29E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 3E-07 3.51E-07 3.81E-07 7.5E-08 9.6E-08 3E-08	
0.148 0.175 0.101 0.058 0.036 0.021 0.015 0.012 0.011 0.101 0.357 0.142 0.060 0.032 0.021 0.014 0.014 0.014 0.013 0.009	$\begin{array}{c} \pm 0.002 & 3.59E \cdot 0. \\ \pm 0.004 & -5.2E \cdot 0. \\ \pm 0.003 & -3.5E \cdot 0. \\ \pm 0.002 & -3.5E \cdot 0. \\ \pm 0.002 & -3.5E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.002 & -2.8E \cdot 0. \\ \pm 0.001 & -7E \cdot 0. \\ \pm 0.001 & -1.2E \cdot 0. \\ \pm 0.001 & -1.2E \cdot 0. \\ \pm 0.001 & -1.3E \cdot 0. \\ \pm 0.001 & -1.3E \cdot 0. \\ \pm 0.001 & -1.5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.003 & -5E \cdot 0. \\ \pm 0.002 & -5E \cdot 0.$	5 5.2E-08 3 2.52E-08 3 1.48E-05 3 1.48E-07 3 1.49E-07 3 1.49E-07 3 1.49E-07 3 1.49E-07 3 1.42E-07 3 1.52E-07 3 3.7E-08 3 -7.4E-08 3 1.72E-07 1.72E-07 1.22E-07 3 9.5E-08 3 1.37E-07 3 8.2E-08 3 7.47E-08 3 8.2E-08 3 7.47E-08 3 8.26-08 3 7.47E-08 4.32E-07 3 5.32E-07 <	-3.5E-08 -2.93E-07 8.96E-06 -1.68E-07 -1.88E-07 -1.51E-07 -3.1E-08 -4.6E-08 -4.6E-08 -4.6E-08 -4.6E-08 -5.1E-08 1.35E-06 4.4E-08 3.97E-07 2.52E-07 2.52E-07 2.52E-07 2.72E-07 7.4E-08 1.11E-07 7.4E-08 1.11E-07	-2E-08 -1.49E-07 -1.68E-07 4.6E-06 -1.5E-07 -1.14E-07 -2.2E-08 -3.1E-08 -3.1E-08 -3.7E-08 -3.1E-08 -3.7E-08 -3.1E-08 -3.2E-07 1.16E-07 2.3E-07 7.9E-08 5.5E-08 7.6E-08 1.26E-07	-2.7E-08 -1.82E-07 -1.82E-07 -1.5E-07 5.28E-06 -1.94E-07 -3.6E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 -4.4E-08 2.1E-07 1.23E-07 2.24E-07 1.39E-07 9.2E-08 1.24E-07 1.89E-07	-2.8E-08 -1.61E-07 -1.51E-07 -1.154E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.14E-07 -1.34E-08 -4.3E-08 -4.4E-08 -5.58E-07 2.94E-07 1.72E-07 1.1E-07 -1.36E-07 2.96E-07	-7E-09 -3.7E-08 -3.1E-08 -2.2E-08 -3.6E-08 -4.3E-08 -4.3E-07 -1.5E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -1.2E-08 -3.4E-08 -3.4E-08 -2.2E-08 -3.2E-08 -3.2E-08 -3.2E-08 -3.2E-08 -3.2E-08	-1.2E-08 -5.9E-08 -4.6E-08 -3.1E-08 -4.6E-08 -5.5E-08 -1.5E-08 -2.2E-08 -2.2E-08 -2.2E-08 -2.2E-08 -3.5E-07 -1.6E-08 -3.5E-08 -3.5E-08 -3.8E-08 -3.8E-08 -3.2E-08	-1.3E-08 -6E-08 -4.3E-08 -4.3E-08 -2.7E-08 -4.4E-08 -4.4E-08 -1.2E-08 -1.2E-08 -1.2E-08 -2.2E-08 9.17E-07 -2.3E-08 1.07E-07 1.7E-08 4.8E-08 4.8E-08 4.8E-08 5.4E-08 5.4E-08 1.13E-07 1.5E-07	-1.7E-08 -7.4E-08 -5.1E-08 -3.1E-08 -3.1E-08 -4.4E-08 -4.4E-08 -4.4E-08 -2.2E-08 -2.	5.74E-07 2.33E-06 1.35E-06 6.6E-07 7.41E-07 5.58E-07 1.05E-07 1.05E-07 1.07E-07 1.07E-07 1.07E-07 1.07E-07 1.042E-07 1.43E-07 2.84E-07 2.84E-07 2.84E-07	1.5E-08 6.6E-08 4.4E-08 3.5E-08 3.5E-08 3.5E-08 3.5E-08 3.5E-08 3.5E-08 2.1E-08 4.8E-08 1.7E-08 4.8E-08 -3.3E-08 -1.8E-08 3.5E-08 -1.8E-08 -3.5E-08 -7.3E-08 -1.35E-07	3.5E-08 7.11E-07 3.97E-07 1.89E-07 2.1E-07 1.65E-07 3.4E-08 4.8E-08 4.8E-08 4.5E-08 5.2E-07 -3.4E-07 -2.93E-07 -2.93E-07 -3.1E-07 -3.1E-07 -2.04E-07 -2.04E-07	0 1.75E-07 2.52E-07 1.23E-07 8.9E-08 8.9E-08 1.8E-08 1.8E-08 1.8E-08 1.8E-08 1.8E-08 1.8E-08 1.43E-07 0 0 -2.93E-07 -7.95E-06 -1.89E-07 -6.5E-08 -4.2E-08 -8.2E-08 -1.19E-07	1.6E-08 1.82E-07 2.44E-07 2.44E-07 2.24E-07 1.59E-07 2.8E-08 3.5E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.7E-08 2.8E-07 -1.89E-07 -1.42E-07 -1.42E-07 -1.37E-07 -1.77E-07	2.6E-08 2.25E-07 2.3ZE-07 2.3E-07 2.3E-07 5.2E-08 6.3E-08 4.6E-08 2.84E-07 -3.3E-08 -3.3E-08 -3.3E-08 -3.3E-08 -3.3E-07 9.21E-06 -2.22E-07 -1.36E-07 -1.36E-07 -3.2E-07	1.4E-08 9.5E-08 1E-07 7.9E-08 1.39E-07 1.72E-07 3E-08 3.6E-08 2.7E-08 2.7E-08 4.2E-07 -1.8E-08 -1.11E-07 -6.5E-08 -1.119E-07 3.42E-06 -8E-08 9.5E-08	1.5E-08 8.2E-08 7.4E-08 5.5E-08 9.2E-08 1.1E-07 3.2E-08 3.8E-08 2.8E-08 8.2.7E-08 9.3E-08 -1.8E-08 84.2E-08 8.2.34E-06 9.9E-08 2.34E-06 -9.9E-08 1.36E-07 -1.36E-07	2.8E-08 1.37E-07 1.11E-07 7.6E-08 1.21E-07 3.9E-08 7.2E-08 5.4E-08 5.4E-08 5.4E-08 4.1.4E-07 -3.5E-08 -1.7E-07 -5.4E-08 -1.6E-08 -1.6E-08 3.06E-06 -9.9E-08 3.06E-06 -2.58E-07 -3.29E-07	5.8E-08 2.68E-07 1.97E-07 1.26E-07 1.26E-07 1.26E-07 5.7E-08 1.03E-07 5.7E-08 1.03E-07 2.42E-07 -7.3E-08 -2.04E-07 -8.2E-08 -1.3E-07 -1.36E-07 -2.25E-07 4.44E-06 6.94E-06	1.07E-07 4.74E-07 3.26E-07 2.92E-07 2.92E-07 7.8E-08 1.38E-07 1.95E-07 3.8E-07 -1.35E-07 -1.35E-07 -1.35E-07 -1.35E-07 -1.17E-07 -1.77E-07 -1.74E-07	1.7E-08 1.27E-07 1.06E-07 6.9E-08 9.6E-08 8.8E-08 2E-08 3E-08 3E-08 1.04E-07 7.51E-07 1.51E-07 3.63E-07 3.63E-07 3.63E-07 7.9E-08 9.6E-08 1.32E-07	

Summary of Linear Models

						Joint Paid-Ind	curred			Pa	id	Case-In	curred
AY	EarnPrem	Paid	CaseIncd	Unpaid	Std Dev	IBNR	Std Dev	Ultimate	=	Ultimate	Std Dev	Ultimate	Std Dev
1998	23,278,084	17,738,999	19,765,070	2,999,915	± 142,339	973,844	± 142,339	20,738,914	TRUE	20,663,109	± 239,110	20,780,207	± 177,147
1999	21,555,421	17,557,257	19,812,936	3,398,227	± 158,471	1,142,547	± 158,471	20,955,484	TRUE	20,824,378	± 244,026	21,062,679	± 209,846
2000	23,495,444	18,432,885	20,739,446	3,804,647	± 177,631	1,498,084	± 177,630	22,237,532	TRUE	22,264,182	± 257,624	22,247,195	± 248,293
2001	25,864,065	17,289,118	19,811,447	4,227,840	± 186,625	1,705,511	± 186,625	21,516,958	TRUE	21,486,528	± 262,929	21,577,395	± 268,263
2002	29,134,414	15,899,281	18,589,679	4,662,500	± 192,377	1,972,102	± 192,377	20,561,781	TRUE	20,487,825	± 265,436	20,667,840	± 282,302
2003	32,391,860	15,283,538	18,314,350	5,421,105	± 220,945	2,390,298	± 220,945	20,704,643	TRUE	20,572,601	± 310,529	20,860,557	± 317,408
2004	36,533,278	14,134,508	17,632,172	6,568,927	± 272,535	3,071,266	± 272,536	20,703,435	TRUE	20,579,410	± 367,486	20,871,758	± 408,036
2005	39,208,849	12,789,801	17,221,257	8,692,524	± 320,964	4,261,067	± 320,964	21,482,325	TRUE	21,381,119	± 424,534	21,630,036	± 491,580
2006	42,065,555	10,724,002	16,822,179	13,144,841	± 403,064	7,046,664	± 403,064	23,868,843	TRUE	23,895,269	± 547,487	23,858,441	± 597,042
2007	40,220,014	5,211,936	12,931,177	20,392,908	± 549,291	12,673,667	± 549,291	25,604,844	TRUE	25,571,140	± 709,931	25,642,727	± 867,816
Total	313,746,984	145,061,325	181,639,713	73,313,433	± 1,196,054	36,735,050	± 1,196,055	218,374,758	TRUE	217,725,562	± 1,572,984	219,198,836	± 1,866,883

APPENDIX A

The Effect of Covariance in a Simple Model

This appendix will present variations of an elementary model so as to illustrate the effect of covariance. The top part of Exhibit A.1 shows the basic form (Model 1). Eight quantities are observed (Obs 1–8), and two predictions are desired (Pred 1–2). The model for each y, whether observed or predicted, is $y_i = x_i\beta + e_i$. Except for Pred 2, all x and ϕ values are one; Pred 2 is like a doubling of Pred 1, which makes its variance relativity four. Zeroes in the variance structure are not shown. This is a heteroskedastic model (homoskedastic in the observed part). The formulæ of the linear statistical model were explained in Section 2.

In the middle of the exhibit is shown the estimate of the parameter: $\hat{\beta} = 97.875 \pm 2.074$. Those unfamiliar with this formulation of the linear model should at least recognize that Model 1 is equivalent to the simple average. At the bottom of the exhibit are various diagnostics (left side) and weighted sums of squares and crossproducts ("SSCP", right side). The diagonal elements of the 3×3 SSCP matrix must satisfy the equation $d_1 = d_2 + d_3$, and the fit accounts for 99.7% of d_1 . Eight (unrelated) observations and one parameter make for seven degrees of freedom, and the estimate of the variance scale is $\hat{\sigma}^2 = 240.875/7 = 34.411$.

The estimator of β linearly depends on the error terms of the observations. Because the error terms of the predictions do not covary with those of the observations, nether do they covary with $\hat{\beta}$. Hence:

$$Var[Pred] = \begin{bmatrix} 1 \\ 2 \end{bmatrix} Var[\hat{\beta}] \begin{bmatrix} 1 & 2 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix} \hat{\sigma}^{2}$$
$$= \begin{bmatrix} 4.301 & 8.603 \\ 8.603 & 17.205 \end{bmatrix} + \begin{bmatrix} 34.411 & 0 \\ 0 & 137.643 \end{bmatrix}$$
$$= \begin{bmatrix} 38.712 & 8.603 \\ 8.603 & 154.848 \end{bmatrix}$$

(We explain the model in this manner for the benefit of those unfamiliar with matrix algebra and multivariate statistics; nevertheless, we encourage them to study these fields until they become natural. See references.) We conclude Model 1 by saying that Pred 2 is *like* Pred 1, but on twice the scale. The covariance between the two predictions is solely due to their reliance on the estimate of β ; it has to do with parameter variance or

uncertainty.

Model 2 of Exhibit A.2 is identical to Model 1 except that the variance relativity of Obs 4 is zero, rather than one. The column 'Constrnd' signals a variance degeneracy with a '1' for this observation. Whether this be unrealistic, it is at least instructive as a limiting case. The model says that in one observation we were able to see β without the obfuscation of an error term. So $\hat{\beta} = 96 \pm 0$, plain and simple. The error terms readjust, and according to the formulation of our software⁹ there are seven (stochastic) observations and zero estimated parameters, which again makes for seven degrees of freedom. In this model it is easier to see that Pred 2 is like Pred 1, but on twice the scale. If, in addition, the variance of any other observation were set to zero, the resulting two equations would be inconsistent.

Things become very interesting with Model 3 (Exhibit A.3), in which Obs 2 and Obs 6 covary, as well as Obs 7 and Pred 2. Both covariances imply perfect positive correlations. In the latter case, we show non-negative definiteness by:

$$Var\left[\begin{bmatrix} 2 & -1 \begin{bmatrix} Obs 7 \\ Pred 2 \end{bmatrix}\right] = \begin{bmatrix} 2 & -1 \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} = 0$$

So the error term of Pred 2 equals (with probability one) the error term of Obs 7. It is not *like* twice the error term of Obs 7; rather, it is the *same* as twice the error term. And because its dependent variable is twice that of Obs 7, Pred 2 is not *like* twice Obs 7; it *is* twice Obs 7. Without even estimating β we know Pred 2 to be $2 \cdot 90 = 180 \pm 0$. Covariance makes the difference between like and same.

But within the observations themselves is a variance degeneracy. Though all the observation error terms are *alike*, the error term of Obs 7 *is* that of Obs 3. The 'Cnstrnd' column indicates with '0.5' the dependency of the two observations. In effect, it says that the same observation is written twice, and should be counted once. Thus instead of eight observations, there are really seven, which with one parameter estimated makes for six degrees of freedom. When covariance is properly considered, one cannot create information *ex nihilo*, i.e., by repeating the same observation. One cannot fool the model

⁹ It would take us too far afield to detail how our software solves the linear statistical model when the variance of the observations Σ_{11} is not positive definite. But briefly, it eigen-decomposes the observations, and treats the once-transformed rows whose eigenvalues are zero as constraints on β . Then it transforms constrained β space into a lower-dimensional, unconstrained γ -space. The twice-transformed model (cf. "TTy1" in SSCP, where "TT" stands for "twice-transformed") is solved, and transformed back. It is a theorem that the solution of a linear model is invariant to any invertible, or one-to-one, transformation of the observations, i.e., $A\mathbf{y}_1 = AX_1\beta + A\mathbf{e}_1$, for any nonsingular A. But one must not forget to transform the covariances: $\Sigma_{12} \rightarrow A \Sigma_{12}$ and $\Sigma_{21} \rightarrow \Sigma_{21}A'$.

even by repeating a linear combination of observations, for $Var\begin{bmatrix} \mathbf{y}_1\\ A\mathbf{y}_1 \end{bmatrix} = \begin{bmatrix} \Sigma_{11} & \Sigma_{11}A'\\ A\Sigma_{11} & A\Sigma_{11}A' \end{bmatrix}$

contains no more information than Σ_{11} contains. Note that although Obs 2 and Obs 6 are redundant, they are consistent. If they were not both 93, they would be inconsistent equations.

Lastly, Model 4 in Exhibit A.4 is a mixture of Models 2 and 3. The reason for changing Obs 6 from 93 to 94 will soon appear. Similarly to Model 3:

$$Var\left[\begin{bmatrix} 2 & -1 \begin{bmatrix} Obs & 2 \\ Obs & 6 \end{bmatrix}\right] = \begin{bmatrix} 2 & -1 \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} = 0$$

i.e., $\operatorname{Prob}[2e_2 - 1e_6 = 0] = 1$, or simply, $e_6 = 2e_2$. But here Obs 2 and Obs 6 are not redundant. Nevertheless, they are equivalent to three equations in three variables:

$$93 = \beta + \boldsymbol{e}_2$$

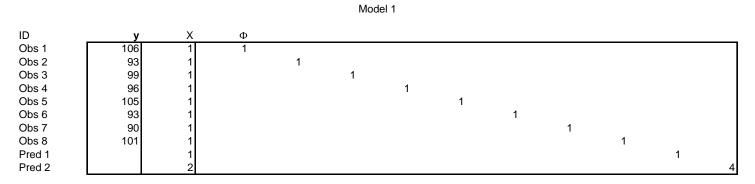
$$94 = \beta + \boldsymbol{e}_6$$

$$\boldsymbol{e}_6 = 2\boldsymbol{e}_2$$

whose solution is $\beta = 92$, $e_2 = 1$, $e_6 = 2$. In Model 2 β was like a gem lying on the surface; here we had to pan a little for it. So $\hat{\beta} = 92 \pm 0$. If we had not changed Obs 6, the equations still would have been consistent, but the error terms would have both been zero – a much less interesting result. The 'Cnstrnd' column indicates the variance degeneracy between the two observations; but their counting as one observation is apportioned inversely according to their 1:4 variance relativities.

Granted, the behavior of Models 2–4 depends on perfect correlation, and is akin to imagining relativistic effects at the speed of light. As long as $\Sigma 11$ has no variance degeneracy, i.e., is positive definite, the observations consist of t_1 consistent equations in $t_1 + k$ variables. They comprise a system of equations that can be solved only probabilistically. However, these limiting cases confirm the conservation of information. Just as there is no magic, just illusion, so too only by trickery can someone produce information out of nothing.

Consequently, to include in a joint paid-incurred model tautologous observations for completely observed exposure periods, correctly accounting for covariance, will furnish no additional information. For it's simply a linear combination of old information. Moreover, for numerical-analytic reasons it's dangerous, since the software must decide when small eigenvalues should be treated as zeroes. Without these redundant equations the joint model will be of full rank.

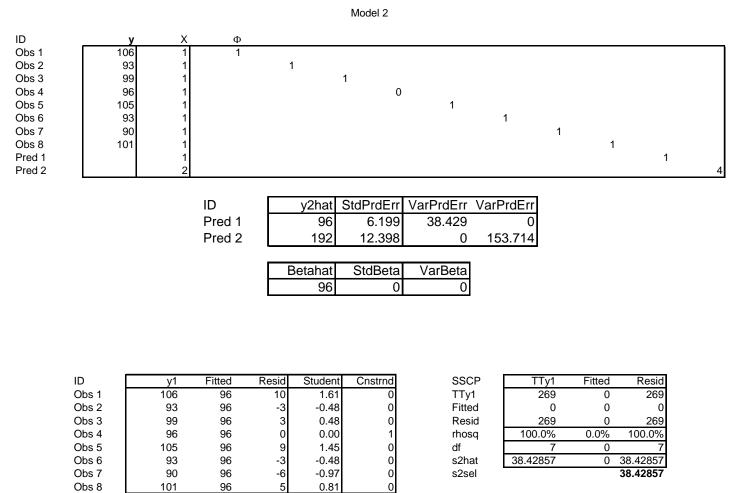


ID	y2hat	StdPrdErr	VarPrdErr	VarPrdErr
Pred 1	97.875	6.222	38.712	8.603
Pred 2	195.750	12.444	8.603	154.848

Betahat	StdBeta	VarBeta
97.875	2.074	4.301

ID	y1	Fitted	Resid	Student	Cnstrnd	SSCP
Obs 1	106	97.875	8.125	1.48	0	TTy1
Obs 2	93	97.875	-4.875	-0.89	0	Fitted
Obs 3	99	97.875	1.125	0.21	0	Resid
Obs 4	96	97.875	-1.875	-0.34	0	rhosq
Obs 5	105	97.875	7.125	1.30	0	df
Obs 6	93	97.875	-4.875	-0.89	0	s2hat
Obs 7	90	97.875	-7.875	-1.44	0	s2sel
Obs 8	101	97.875	3.125	0.57	0	

Ρ	TTy1	Fitted	Resid
1	76877	76636.13	240.875
d	76636.13	76636.13	0
d	240.875	0	240.875
q	100.0%	99.7%	0.3%
	8	1	7
at	9609.625	76636.13	34.41071
el			34.41071

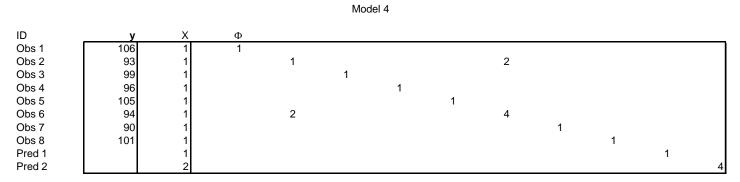


					Mod	lel 3						
ID	у	х	Φ									
Obs 1	106	1	1									
Obs 2	93	1		1				1				
Obs 3	99	1			1							
Obs 4	96	1				1						
Obs 5	105	1					1					
Obs 6	93	1		1				1				
Obs 7	90	1							1			2
Obs 8	101	1								1		
Pred 1		1									1	
Pred 2		2							2			4

ID	y2hat	StdPrdErr	VarPrdErr	VarPrdErr
Pred 1	98.571	6.380	40.707	0
Pred 2	180	0	0	0

Betahat	StdBeta	VarBeta
98.571	2.256	5.088

ID	y1	Fitted	Resid	Student	Cnstrnd	SSCP	TTy1	Fitted	Resid
Obs 1	106	98.571	7.429	1.34	0	TTy1	68228	68014.29	213.7143
Obs 2	93	98.571	-5.571	-1.01	0.5	Fitted	68014.29	68014.29	0
Obs 3	99	98.571	0.429	0.08	0	Resid	213.7143	0	213.7143
Obs 4	96	98.571	-2.571	-0.47	0	rhosq	100.0%	99.7%	0.3%
Obs 5	105	98.571	6.429	1.16	0	df	7	1	6
Obs 6	93	98.571	-5.571	-1.01	0.5	s2hat	9746.857	68014.29	35.61905
Obs 7	90	98.571	-8.571	-1.55	0	s2sel			35.61905
Obs 8	101	98.571	2.429	0.44	0				



ID	y2hat	StdPrdErr	VarPrdErr	VarPrdErr
Pred 1	92	8.586	73.714	0
Pred 2	184	17.171	0	294.857

Betahat	StdBeta	VarBeta
92	0	0

ID	y1	Fitted	Resid	Student	Cnstrnd	SSCP	TTy1
Obs 1	106	92	14	1.63	0	TTy1	516
Obs 2	93	92	1	0.12	0.8	Fitted	0
Obs 3	99	92	7	0.82	0	Resid	516
Obs 4	96	92	4	0.47	0	rhosq	100.0%
Obs 5	105	92	13	1.51	0	df	7
Obs 6	94	92	2	0.12	0.2	s2hat	73.71429
Obs 7	90	92	-2	-0.23	0	s2sel	
Obs 8	101	92	9	1.05	0		

)	TTy1	Fitted	Resid
	516	0	516
	0	0	0
	516	0	516
	100.0%	0.0%	100.0%
	7	0	7
	73.71429	0	73.71429
			73.71429

APPENDIX B

Correlation Constraints among Three Random Variables

Our solution to the joint model involved the addition of tautologous observations which covary with certain of the paid and incurred observations. Often the loss observations are not inter-correlated. According to statistical and econometric terminology (e.g., Judge [1988], Chapter 9), the variance structure of such observations is homo- or heteroskedastic, as opposed to autocorrelated. In our simple example they were homoskedastic; in the Workers' Compensation example they were heteroskedastic. If z be a tautologous observation that involves loss observation x (so that $Cov[z, x] \neq 0$; in our models, $Cov[z, x] = \pm Var[x]$), and x does not covary with any other loss observation, then we may assume that z does not secondarily covary with any other loss observation. But in general, for z to covary with x and for x to covary with y places a transitive tendency for z to covary with y. Ignoring secondary covariance may lead one to create models whose variance structure is not non-negative definite. Such models would be defective, because a variance structure is legitimate if and only if it is non-negative definite.

So our task here is to solve an interesting problem: If the correlation between two random variables is ρ , what are legitimate values of x and y, the correlations of a third random variable with the first two? Mathematically expressed, for what values of x and y is $\rho \quad 1 \quad y$ non-negative definite? To express the problem as correlation is simpler and no less general than to express it as covariance.

First, because the correlation coefficient is bounded, $-1 \le \rho, x, y \le 1$. Hence, regardless of ρ , allowable pairs (x, y) must be on or within the square whose four corners are $(\pm 1, \pm 1)$. And second, it is a theorem of matrix algebra that a matrix Σ is non-negative definite if and only if it has a "square root," i.e., a real-valued matrix *W* such that $\Sigma = WW'$. The Cholesky decomposition yields a suitable square *W* that is lower-triangular (i.e., zero above the main diagonal).¹⁰

The Cholesky decomposition of the correlation matrix is:

$$\begin{bmatrix} 1 & \rho & x \\ \rho & 1 & y \\ x & y & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ \rho & \sqrt{1 - \rho^2} & 0 \\ x & a & b \end{bmatrix} \begin{bmatrix} 1 & \rho & x \\ 0 & \sqrt{1 - \rho^2} & a \\ 0 & 0 & b \end{bmatrix},$$

¹⁰ Cf. Halliwell [1997; Appendix A], Healy [1986; 54f], and Judge [1988; 961].

where:

$$y = \rho x + a\sqrt{1-\rho^2}$$
$$1 = x^2 + a^2 + b^2$$

Hence, the correlation matrix is non-negative definite if and only if real values of *a* and *b* exist that solve these two last equations (in which ρ , *x*, and *y* are given).

Of course, $\sqrt{1-\rho^2} \ge 0$. If $\sqrt{1-\rho^2} = 0$, y must equal ρx , and a may be any real number. Setting a to zero in this case gives the most leeway for b to be real. Hence, a Cholesky decomposition exists if and only if $b^2 \ge 0$. With this information we derive the inequality:

$$x^{2} - 2\rho xy + y^{2} = x^{2} - \rho^{2} x^{2} + y^{2} - 2\rho xy + \rho^{2} x^{2}$$

$$= x^{2} (1 - \rho^{2}) + (y - \rho x)^{2}$$

$$= x^{2} (1 - \rho^{2}) + (a\sqrt{1 - \rho^{2}})^{2}$$

$$= (x^{2} + a^{2})(1 - \rho^{2})$$

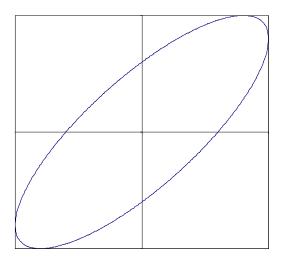
$$= (x^{2} + a^{2} + 0)(1 - \rho^{2})$$

$$\leq (x^{2} + a^{2} + b^{2})(1 - \rho^{2})$$

$$= 1 \cdot (1 - \rho^{2})$$

Therefore, the correlation matrix is non-negative definite if and only if $x^2 - 2\rho xy + y^2 \le (1 - \rho^2)$.

Legitimate (x, y) points satisfy the equation $x^2 - 2\rho xy + y^2 \le (1 - \rho^2)$. The region of these legitimate points is symmetric about the two lines $y = \pm x$. In fact, it is an ellipse whose axes are on those lines, its half lengths along the lines being $\sqrt{1 \pm \rho}$ respectively. Here is a graph is of the ellipse when $\rho = 0.8$:



Interior points of the ellipse, i.e., $x^2 - 2\rho xy + y^2 < (1 - \rho^2)$, produce positive definite matrices; boundary points indicate a linear dependence among the three random variables. If $\rho = 0$, the ellipse becomes the unit circle. In the case that $\rho = \pm 1$, the ellipse degenerates into the respective lines $y = \pm x$, as expected. The origin is always a legitimate candidate for (x, y), since for any two random variables there exists a third uncorrelated with either one of them.

The area of the ellipse is $\pi\sqrt{1-\rho^2}$. That this is maximized for $\rho = 0$ means that one may accommodate new random variables most freely into a universe of uncorrelated random variables. We can integrate the area of ellipse(ρ) over $\rho: \int \pi\sqrt{1-\rho^2} d\rho = \pi \frac{\pi}{2} = \frac{\pi^2}{2}$. From this we conclude that the probability of constituting a legitimate correlation structure by randomly sampling ρ , x, and y from a Uniform [-1, 1] distribution is $\frac{\pi^2}{2}/2^3 = \left(\frac{\pi}{4}\right)^2 \approx 61.7\%$.¹¹

¹¹ The legitimacy equation has the three-way symmetric form $x^2 + y^2 + z^2 - 2xyz \le 1$, as well as the determinant form $\begin{vmatrix} 1 & x & y \\ x & 1 & z \\ y & z & 1 \end{vmatrix} \ge 0$. However, non-negative (positive) definiteness means more than a non-

negative (positive) determinant. A symmetric $n \times n$ matrix is non-negative (positive) definite if and only if all its subdeterminants, of which there are 2^n -1, are ≥ 0 (> 0). However, as a test of definiteness this is much less efficient than the Cholesky decomposition.