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# REVISED RATIO OF CRASH SEVERITIES REPORTABLE TO THE MCMIS CRASH FILE

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# Revised Ratio of Crash Severities Reportable to the MCMIS Crash File

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The Motor Carrier Management Information System (MCMIS) Crash file has been developed by the Federal Motor Carrier Safety Administration (FMCSA) to serve as a census file of trucks and buses involved in traffic crashes meeting a specific crash severity threshold. Each state is responsible for identifying cases that meet the MCMIS Crash file criteria and reporting the required data through the SafetyNet system. UMTRI has completed a set of evaluations of state reporting and found that reporting rates range from over 80 percent to less than 10. The present report provides a method of predicting, given a known number of fatal involvements, the number of crash involvements a state should be reporting.

In each state, the number of fatal involvements is well-known, so all states will start with a known quantity, the number of fatal truck and bus crash involvements. It is then hypothesized that the ratio of reportable crash severities, that is, the ratio of fatal involvements to nonfatal involvements, will apply across all the states.

Data from eight states that provide all the information necessary to identify MCMIS-reportable cases were used. A weighted log-linear model is fit to MCMIS data for the eight states that have information recorded for both fatal and nonfatal crashes. The model is then used to estimate the number of nonfatal crashes for a new state in which only the number of fatal crashes is known. Prediction intervals are presented.

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APPROXIMATE CONVERSIONS TO SI UNITS  Symbol When You Know Multiply By To Find Symbol						
	***************************************	LENGTH		<u> </u>		
in	inches	25.4	millimeters	mm		
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mi	miles	1.61	kilometers	km		
2		AREA		2		
in <sup>2</sup>	square inches	645.2	square millimeters	mm²		
ft <sup>2</sup> yd <sup>2</sup>	square feet	0.093	square meters	m² m²		
ac	square yard acres	0.836 0.405	square meters hectares	ha		
mi <sup>2</sup>	square miles	2.59	square kilometers	km²		
		VOLUME				
fl oz	fluid ounces	29.57	milliliters	mL		
gal	gallons	3.785	liters	L		
ft <sup>3</sup>	cubic feet	0.028	cubic meters	m <sup>3</sup>		
yd <sup>3</sup>	cubic yards	0.765	cubic meters	m <sup>3</sup>		
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		MASS				
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		ILLUMINATION				
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fl	foot-Lamberts	3.426	candela/m <sup>2</sup>	cd/m <sup>2</sup>		
		RCE and PRESSURE or S				
lbf	poundforce	4.45	newtons	N		
lbf/in <sup>2</sup>	poundforce per square inch	6.89	kilopascals	kPa		
	APPROXII	MATE CONVERSIONS F	FROM SI UNITS			
Symbol	When You Know	Multiply By	To Find	Symbol		
		LENGTH				
mm	millimeters	0.039	inches	in		
m	meters	3.28	feet	ft		
m	meters	1.09	yards	yd		
km	kilometers	0.621	miles	mi		
		AREA				
mm²	square millimeters	0.0016	square inches	in <sup>2</sup>		
$m_2^2$	square meters	10.764	square feet	ft <sup>2</sup>		
m <sup>2</sup>	square meters	1.195	square yards	yd <sup>2</sup>		
ha km²	hectares square kilometers	2.47 0.386	acres square miles	ac mi²		
KIII	square knometers	VOLUME	Square filles	1111		
mL	milliliters	0.034	fluid ounces	fl oz		
L	liters	0.264	gallons			
$m^3$	cubic meters	35.314	cubic feet	gal ft³		
$m^3$	cubic meters	1.307	cubic yards	yd <sup>3</sup>		
	MASS					
		IVIAGO				
	grams	0.035	ounces	OZ		
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<sup>\*</sup>SI is the symbol for the International System of Units. Appropriate rounding should be made to comply with Section 4 of ASTM E380. (Revised March 2003)

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## 1 Introduction

The Motor Carrier Management Information System (MCMIS) Crash file has been developed by the Federal Motor Carrier Safety Administration (FMCSA) to serve as a census file of trucks and buses involved in traffic crashes meeting a specific crash severity threshold. FMCSA maintains the MCMIS file to support its mission to reduce crashes, injuries, and fatalities involving large trucks and buses. Designing effective safety measures requires accurate and complete crash data to understand the dimensions of the crash problem. The usefulness of the MCMIS Crash file depends upon individual states transmitting a standard set of data items on all trucks and buses involved in traffic crashes that meet the file's crash severity threshold.

The University of Michigan Transportation Research Institute (UMTRI) has prepared a series of reports evaluating the completeness of reporting from selected states. As of this report, UMTRI has completed fifteen evaluations, covering thirteen states. Reporting rates have ranged from less than ten percent of reportable cases to over eighty percent.

Each state is responsible for identifying cases that meet the MCMIS Crash file criteria and reporting the required data through the SafetyNet system. The MCMIS selection criteria are clearly stated and, in theory, easily applied. To comply with the reporting requirements, states have adapted existing systems, developed for different purposes, to identify and capture the correct cases. Each state has its own threshold for reportable crashes, its own system for classifying vehicles, and its own set of information that it collects on the crashes. Given the multiple purposes for which crash data are collected by states, it is often not clear which crashes should be reported and which should not. This is especially true for crashes that do not involve a fatality. Many states do not regularly collect the specific information needed to determine if a nonfatal case is reportable—e.g., whether an injured person was transported for medical attention or whether a vehicle was towed due to disabling damage.

As a consequence of the mismatch between MCMIS requirements and the methodology of states' established systems, there is often no easy way for the states, or FMCSA, to know if the right number of cases is being reported. Some states have thought that they were in compliance and fully reporting to the MCMIS Crash file, but the UMTRI evaluation found significant underreporting. Until all states can be directly evaluated, a method of predicting, within a reasonable range, the number of cases that each state should report could serve as a guidepost or benchmark to the states on where they stand. Such a benchmark could alert the state that a problem exists, and initiate a process to identify a solution.

This paper proposes a method of estimating the number of involvements each state should report to the MCMIS Crash file, even if the state's data system cannot readily identify all the crashes that meet the Crash file criteria. It is based on developing a ratio of reportable crash severities in state data with complete information (though not necessarily reporting). In each state, the number of fatal involvements is well-known, so all states will start with a known quantity, the number of fatal truck and bus crash involvements. It is then hypothesized that the ratio of reportable crash severities, that is, the ratio of nonfatal involvements to fatal involvements, will apply across all the states. If this is true, it will be possible, knowing the number of fatal

involvements that occurred in a state, to predict the number of involvements of lesser severity, and thus predict the total number of cases that the state should report to the MCMIS Crash file.

During the process of evaluating state reporting, UMTRI identified a number of states whose data systems provide the data necessary to apply the MCMIS reporting criteria completely. Virtually all states can identify trucks and buses reasonably well, and all states can identify fatal crashes cleanly. However, most states do not regularly collect the information needed to identify reportable nonfatal crashes: crashes in which an injured person was transported for immediate medical attention or crashes in which at least one vehicle was towed due to disabling damage. States identify persons injured in a crash, and even nominally use the same system to classify injury severity, but not all have taken the next step to capture if the person was transported for medical attention. Similarly, many states record if a vehicle was towed, but not whether the reason for the tow was disabling damage.

UMTRI identified six<sup>1</sup> states whose existing crash data systems can identify all levels of the reporting criteria. Two of the states were evaluated for two different data years, providing a total of eight data files. These data that can be used to estimate the ratio of reportable crash severities, that is, the ratio of nonfatal involvements to fatal involvements.

Why not use the NHTSA's General Estimates System (GES) files? GES is a nationally-representative sample of police-reported crashes. GES cases are sampled from primary sampling units (PSUs) around the country and a standard set of data are coded from the sampled police-reports. The GES data can be used to identify trucks and buses, and crash detail includes whether an injured person was transported for immediate medical attention or a vehicle towed due to disabling damage. Thus, GES nominally can be used to cleanly estimate the number of reportable involvements to the MCMIS Crash file.

However, there are a number of reasons why the GES file may not be completely satisfactory for this application. Though the GES variable recording whether a vehicle was towed has a level for towed due to damage, we know through doing the state reporting evaluations that many state police reports do not include that information, so this variable likely underestimates the number of vehicles towed due to disabling damage. In addition, the GES file is a sample drawn through a complex stratified, hierarchical sampling system. Truck and bus crashes are a small sample, relative to automobile crashes. The standard errors for small subsets of the file, such as trucks, are relatively large. Moreover, it is known that GES underestimates the number of fatal truck and bus involvements. For the years 2000 through 2005, GES estimated 2,903 to 5,819 trucks involved in a fatal crash, with an average of about 4,100. The average from UMTRI's Trucks Involved in Fatal Accidents file is over 5,100. Moreover, the 95 percent confidence interval for an estimate in GES of 4,100 fatal truck involvements is ±1,800, or about 40 percent. Since the basis of the ratio is the number of fatal involvements, which it is assumed can be identified precisely in state crash data, the GES estimate is too uncertain to be reliable.

<sup>&</sup>lt;sup>1</sup> One more state, South Dakota, was evaluated after the original, preliminary report on developing a crash severity ratio..

#### 2 Problem statement

The purpose of this report is to develop a method that can be used to assist states in determining if they are in compliance with FMCSA's requirement to report all crash involvements reportable to the MCMIS Crash file. This method is not intended to identify a precise number of reportable cases for each state, but to give guidance as to whether a state's reporting is within an expected range. It is assumed that states can identify fatal involvements quite precisely and with high confidence. Virtually all the states can identify trucks and buses readily. Crashes in which a fatality occurred are equally clearly identifiable. All states identify fatal injuries, and the definition of a fatal traffic accident—death within 30 days of the crash—is standard. Accordingly, one level of the hypothesized ratio, i.e., fatal involvements, should be well established in all states.

It is assumed that the relationship of fatal to nonfatal reportable involvements exists independent of any particular state system. That is, the ratio we are seeking does not depend on a state's definitions or system of collecting data, so the ratio established in one set of states should hold true for other states. As a counter-example, consider the common system for classifying injury severity. Most states use the KABC0 system, which classifies injuries as fatal, incapacitating, non-incapacitating but evident, complaint of pain, and no injury. Fatal injuries are clear and not subject to much interpretation. But the other injury levels are more difficult to classify consistently and can be subject to more interpretation. As a consequence, states vary widely in the relative proportions of A-, B-, and C-injuries.<sup>2</sup> The differing proportions are related not to some underlying difference in the severity of crashes in different states but to variations in the interpretation and application of standard definitions.

In contrast, the MCMIS Crash file criteria do not depend on crash severity standards that are known to be applied unevenly, but instead provide a relatively simple definition that should apply in roughly the same way everywhere. Reportable nonfatal involvements include either an injury transported for immediate medical attention or a vehicle towed due to disabling damage. An injury serious enough to be transported for treatment in Maine likely would also be transported if it occurred in California. There may be some variations from state to state, but they are not expected to be large. Similarly, it is not expected that whether a vehicle is disabled enough to be towed will vary much by region. There may be areas where towing following a crash is more common, but less variation, by state, in judging whether a vehicle has suffered disabling damage. In this way, the choice of criteria for the MCMIS Crash file is astute, specifically because the criteria do not depend on how a state may choose to define an injury severity level and train their officers to identify it.

If it is true that there is a fundamental relationship between fatal and nonfatal (injury/transported and towed/disabled) involvements, then the ratio can be discovered by examining data files with enough detail to identify each crash type. In this process, each state and year of data is one

<sup>&</sup>lt;sup>2</sup> See O'Day, J., *Accident Data Quality*. National Cooperative Highway Research Program Synthesis of Highway Practice, No. 192. Federal Highway Administration, Washington, DC, 1993. O'Day found that the proportion of A injuries varied from 4.9% to 23.8% in a sample of about 20 states. The findings were for 1990-1991 data, but illustrate the point.

observation, one estimate of the underlying ratio. By assembling multiple such observations and fitting a statistical model, it is possible to estimate the true ratio of crash severities that applies across states. This statistical model will allow states to estimate the number of cases that they should be reporting to the MCMIS Crash file, with some margin of error. It is assumed that each state, and FMCSA, will know the correct number of fatal involvements. The model will then predict the number of nonfatal (injury/transported and towed/disabled) involvements that the number of fatal involvements implies.

A preliminary report presented a tentative method to estimate reporting ratios for injured/transported and towed/disabled cases.[1]<sup>3</sup> The approach was a simple linear regression model that fit the data well statistically but which produced prediction ranges that were large and did not predict well back to the original data. The model here predicts *nonfatal* reportable involvements, the combination of injured/transported, and towed/disabled cases. Predicting one outcome is more straightforward, produces better results, and meets the basic need to predict the number of cases a state should report to the MCMIS Crash file.

#### 3 Data

We selected states for modeling the distribution of reportable cases that collected the detail necessary to reproduce the MCMIS Crash file reporting criteria. The essential criteria for reporting are displayed in Table 1. Adequate methods of identifying trucks and buses could be developed for all the states evaluated to date (February 2007), with some qualifications. In some states, light vehicles displaying hazardous materials placards were not identified with high confidence, but the number of such vehicles is so small relative to the number of trucks and buses that it should have only an insignificant impact on the analysis.

Truck with GVWR over 10,000 or GCWR over 10,000, or

Vehicle

Bus with seating for at least nine, including the driver, or Vehicle displaying a hazardous materials placard.

Fatality, or Injury transported to a medical facility for immediate medical attention, or Vehicle towed due to disabling damage.

Table 1 Vehicle and Crash Severity Threshold for MCMIS Crash File

Identifying crashes that meet the reporting criteria is the crux of the problem in estimating reportable cases. Table 2 shows the states that were selected for this problem. In all these states, determining the number of reportable fatal involvements can be done fairly cleanly and with minimal ambiguity. Most of the states directly coded the detail needed to identify the different crash severities. An indicator that an injured person was transported for immediate medical attention was critical for selection. The experience of the UMTRI state evaluations showed that

<sup>&</sup>lt;sup>3</sup> See References at the of this paper.

whether an injured person was transported for care does not map cleanly to coded injury severities, so injury severity could not be used as a surrogate. With respect to the towed/disabled criteria, the method some states used to code vehicle damage severity could be used as a substitute for a direct indicator that a vehicle was towed due to disabling damage. The severity scale employed by the states here directly indicate whether a vehicle was disabled. This is not a perfect substitute for towed/disabled, but it is a reasonable surrogate.

State	Data Year	Injured	Transported	Towed	Disabled
Florida	2003	Yes	Yes	No*	Yes
Iowa	2004	Yes	Yes	No*	No*
Louisiana	2005	Yes	Yes	Yes	Yes
Missouri (1)	2001	Yes	Yes	Yes	Yes
Missouri (2)	2005	Yes	Yes	Yes	Yes
Ohio (1)	2000	Yes	Yes	Yes	Yes
Ohio (2)	2005	Yes	Yes	Yes	Yes
South Dakota	2005	Yes	Yes	Yes	Yes

Table 2 States Selected for Modeling the Ratio of Crash Severities

Table 3 shows the data used in modeling the crash severity ratio. Each state and crash year is one observation. It is desirable that the data used in the model covers the range in the number of cases expected to be reported from the fifty states. There was a reasonable range of data available for the modeling effort, including a relatively small state like South Dakota with 19 annual fatal truck and bus crash involvements to Florida, which reported 444 fatal involvements for the data year used. Most of the states available, however, fell into the range

Table 3 Counts of Fatal and Nonfatal Reportable Involvements
Used in Modeling Crash Severity Ratio

	Crash		
State	Fatal Nonfatal		Total
S Dakota	19	434	453
Iowa	68	1,974	2,042
Louisiana	147	4,250	4,397
Missouri 1	155	6,002	6,157
Missouri 2	186	5,946	6,132
Ohio 1	205	8,840	9,045
Ohio 2	211	9,489	9,700
Florida	444	13,353	13,797

between 150 and 200 annual fatalities. The very small states and the very large states were underrepresented in the data available.

UMTRI's Trucks Involved in Fatal Accidents (TIFA) and Buses Involved in Fatal Accidents (BIFA) can be used to provide an accurate distribution of the annual expected reportable truck

<sup>\*</sup> Vehicle damage severity used as surrogate for towed due to damage

and bus fatal involvements for each state. The two files should include all truck and bus fatal involvements that are reportable to the MCMIS Crash file, since the definitions of a reportable truck or bus are compatible with the MCMIS definitions. Only light vehicles transporting hazardous materials that require a placard are not included, but there are only a small number of such vehicles each year so they would not affect the overall distribution.

Figure 1 shows the distribution of states by the annual average number of truck and bus fatal involvements. There are about nine states each year with 25 or fewer MCMIS-reportable fatal involvements, eight with between 26 and 50, and five with between 76 and 100. Almost 55 percent of the states have 100 or fewer reportable cases annually. The circled numbers on the figure show the number of states used in the analysis within each interval. The available states cover the range, but the smaller states are underrepresented.

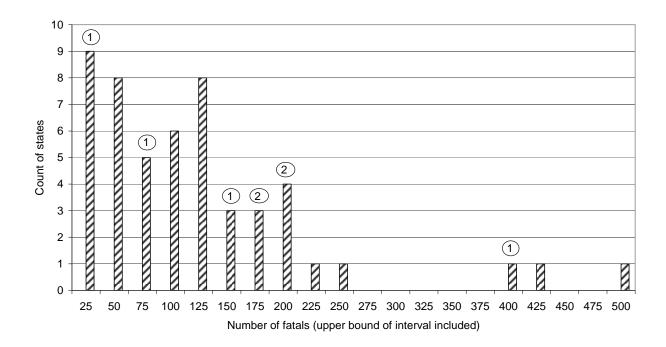


Figure 1 Counts of States by Average Annual Number of Fatal Truck and Bus Involvements TIFA 1999-2003, BIFA 1999-2003

## 4 Model and Methods

The goal is to predict the number of nonfatal crashes from the number of fatal crashes for states that only have data recorded for the number of fatal crashes. Ninety-percent prediction intervals for the estimates are also desired. To accomplish this goal, a weighted log-linear model is fit to MCMIS data for eight states that have information recorded for both fatal and nonfatal crashes. The model is then used to estimate the number of nonfatal crashes for a new state in which only the number of fatal crashes is known. Prediction intervals are presented, instead of confidence intervals, because data from the new state were not used to estimate the regression line in the modeling process. Since there is more uncertainty in predicting nonfatal crashes for a state not

used in the modeling process, prediction intervals are wider than confidence intervals. Table 4 shows data for the eight states that have numbers of crashes recorded for both fatal and nonfatal crashes. The logs of the count of fatals and nonfatals are also shown.

Table 4 Data Used in the Modeling Process					
State	Fatals	Nonfatals	Log Fatals	Log Nonfatals	
S Dakota	19	434	2.9444	6.0730	
Iowa	68	1,974	4.2195	7.5878	
Louisiana	147	4,250	4.9904	8.3547	
Missouri 1	155	6,002	5.0434	8.6998	
Missouri 2	186	5,946	5.2257	8.6905	
Ohio 1	205	8,840	5.3230	9.0870	
Ohio 2	211	9,489	5.3519	9.1579	
Florida	444	13 353	6.0958	9 4995	

**Table 4 Data Used in the Modeling Process** 

Two decisions were made with respect to the modeling procedure: the data should be analyzed on the log scale, and weights should be incorporated to reflect the idea that larger states should receive more weight than smaller states. The decision to analyze the data on the log scale is based on a scatterplot of the log of the nonfatal crashes by the log of the fatal crashes. The scatterplot shows a strong linear association between the two variables, with a correlation of about 0.97. Crash numbers are often in the thousands, and crash data are often analyzed using log-linear models. The decision to incorporate weights into the regression model is based on the realization that smaller states have smaller numbers of fatal crashes, and therefore more uncertainty in their numbers. States with larger numbers of crashes should receive more weight than states with fewer numbers of crashes.

Figure 2 shows a scatter plot of the coefficient of variation against the mean number of annual fatal involvements for each state. The plot shows that states with smaller mean annual fatal involvements have more variability in the annual number than larger states. The coefficient of variation (CV) is calculated by dividing the standard deviation by the mean number of fatals. States with high CV's have more variation in the count of fatals than states with low CVs. In other words, where there are many fatal involvements annually, the number does not fluctuate from year to year as much relatively as states with few fatal involvements. Because of this it is not desirable to treat each state as an observation with equal weight in fitting the model. On the other hand, most of the states have fewer than 100 fatal involvements annually, and using the count of fatals to weight the states would give the few large states excessive leverage in the model. According, we weighted the states by the square root of the number of fatal involvements. This choice gives more weight to the large states, which have less variability in

the number of fatals, but still gives appropriate weight to the smaller states, which is where most of the data are.

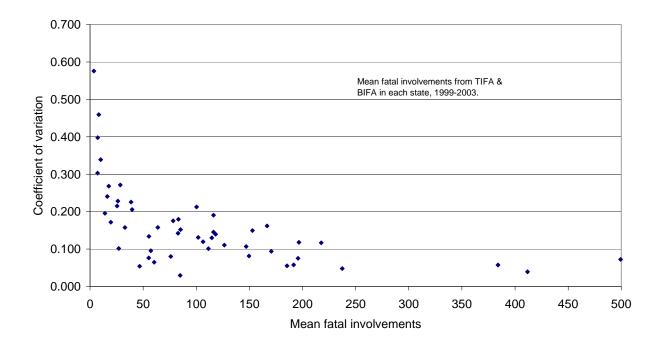


Figure 2 Variation in the Annual Number of Fatal Involvements Reportable From the States

The model for analyzing the data shown in Table 4 takes the form

$$\log y_i = \beta_0 + \beta_1 \log(x_i) + \varepsilon_i, \qquad \varepsilon_i \sim N(0, \sigma^2 / \sqrt{x_i}), \qquad i = 1, ..., 8$$

where  $y_i$  is the number of nonfatal crashes for state i,  $x_i$  is the number of fatal crashes,  $\beta_0$  and  $\beta_1$  are the intercept and slope parameters, respectively, and  $\varepsilon_i$  are the error terms. The random component of the model is contained in the error terms. These terms are modeled as Normal random variables with mean 0 and variance  $\sigma^2/\sqrt{x_i}$ . In standard regression models the variance is constant at  $\sigma^2$ , but since this is a weighted regression, each case receives a different weight. In this regression model each state is being weighted by the square root of the number of fatals. This ensures that states with more fatals get more weight than states with fewer fatals. In other words, the variances of the error terms are smaller for states with more fatals. Weighting states by the square root was arrived at by trial and error. Preliminary results showed that weighting by the number of fatals was too severe and weighting on the log scale was too mild. Weighting by the square root appears to be a reasonable compromise.

After fitting this model, the estimated equation is

$$\log \hat{y}_i = 3.0983 + 1.0835 \log(x_i).$$

Note that the slope parameter, 1.0835, is very close to one. This means that for a unit increase in the log fatals, the estimated log nonfatals increase by a little bit more than one unit. Suppose now that it is desired to estimate the number of nonfatal crashes for a new state with 100 fatal crashes. Applying the fitted equation to the new state gives

$$3.0983 + 1.0835\log(100) = 8.0878$$

and a 90 percent prediction interval is (7.5867, 8.5888). Exponentiating these results back to the original scale gives an estimated  $\exp(8.0878) = 3,254$  nonfatal crashes and the 90 percent prediction interval is (1972, 5371).

An Excel spreadsheet is provided that allows the user to input the number of fatal crashes for a new state. The output produced consists of the estimated number of nonfatal crashes and the 90 percent prediction interval. Using the present model, if the number of fatals for a new state is 200, the estimated number of nonfatals is 6,896 and the 90 percent prediction interval is (4551, 10450). Figure 3 below shows the scatterplot, the fitted regression line, and 90 percent prediction intervals based on the fit to eight states.

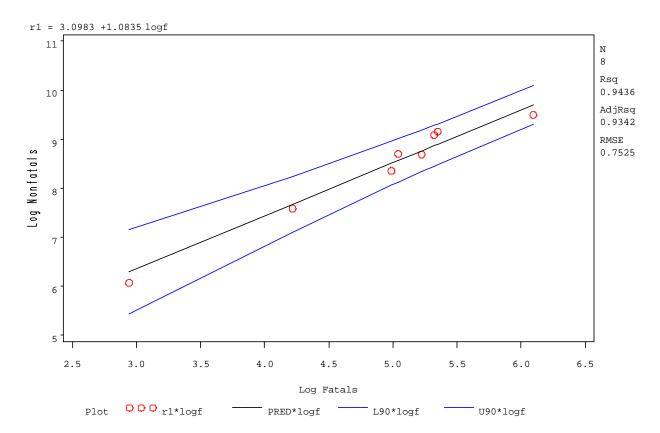


Figure 3 Scatterplot, Fitted Regression Line, and 90 Percent Prediction Intervals for Eight States

# 5 Application

The model can be applied to individual states to estimate reportable cases, although it is not valid to predict outside of the range of values covered in the model. However, the model is based on a range that covers most of the states: There are only eight states<sup>4</sup> that average fewer than 19 truck or bus fatal involvements annually, and only one (Texas) that averages over 444. Table 5 compares the predicted and actual values for nonfatal involvements for the states used in generating the model. Generally, the model estimates are reasonably close to the observed values. Almost all the predictions are within 20 percent of the actual number, and Iowa and Missouri 2 are within 20 percent. Agreement is less good for South Dakota (24 percent) and Florida (23 percent) but all the predicted numbers are within the 90 percent prediction interval.

	Nonfatal Involvements		90% Prediction Interval	
State	Actual Predicted		Lower bound	Upper bound
South Dakota	434	538	228	1,271
Iowa	1,974	2,143	1,213	3,785
Louisiana	4,250	4,940	3,160	7,723
Missouri 1	6,002	5,232	3,368	8,129
Missouri 2	5,946	6,375	4,181	9,721
Ohio 1	8,840	7,083	4,683	10,713
Ohio 2	9,489	7,309	4,843	11,029
Florida	13,353	16,363	11,003	24,336

Table 5 Comparison of Actual and Nonfatal and 90 Percent Prediction Intervals

Figure 4 displays the data in Table 5 graphically. Note that all the observed values for nonfatal involvements are within the 90 percent prediction intervals. The range is relatively large for Florida. Both predictions for Ohio fall substantially below the actual number of nonfatal involvements, but within the range predicted by the model.

<sup>&</sup>lt;sup>4</sup> The District of Columbia, Rhode Island, Alaska, Hawaii, Vermont, New Hampshire, North Dakota, and Delaware.

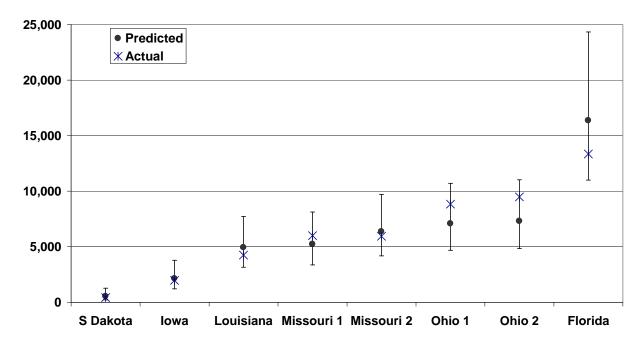


Figure 4 Actual, Predicted and 90 Percent Prediction Interval for Case States

The model presented here is improved from the models in the earlier attempt to model the ratios of cases reported to the MCMIS Crash file. There is clearly a relationship between the number of fatal involvements and nonfatal involvements. The previous work showed that the relationship is linear and the association between the counts of fatal and injury/transported or fatal and tow/disabled involvements rather good, with R<sup>2</sup> statistics of 0.87 and 0.85 respectively. But the models were not entirely satisfactory because they did not predict back to the original data well, and the confidence intervals were too wide to provide useful guidance to the states.

Simplifying the approach by modeling only the nonfatal involvements and using the log transform improved the results. We also provide prediction intervals, rather than confidence intervals. In estimating the regression line for predicting the number of nonfatal crashes from the number of fatal crashes, eight states were used. These eight states represent a sample from a larger population of states. Since there is uncertainty in a sample, a confidence interval is often desired for the estimate of the number of nonfatal crashes for a state, given the number of fatal crashes. Often, 90 or 95 percent confidence intervals are calculated for each state. A 90 percent confidence interval is calculated by a procedure, such that if this procedure were repeated over and over again, 90 percent of the confidence intervals would *trap* the true number of nonfatal crashes in the population. Thus, we are 90 percent confident that our estimate traps the true number of nonfatal crashes in the population for a particular state. By collecting a sample and calculating our estimates, we only perform this procedure once. Confidence intervals apply to states that were used in the estimation process.

Prediction intervals, on the other hand, apply to out-of-sample states. In other words, prediction intervals are used for new states that were not used to estimate the regression line. Intuitively, prediction intervals are wider than confidence intervals. Thus, the prediction problem begins by first fitting a model to a sample of states. Once the model is fitted, an estimate of the number of

nonfatal crashes can be predicted for a new state not used in the estimation process. Since the new state is out-of-sample, a prediction interval should be reported. The interpretation is similar to that of a confidence interval. We are 90 percent certain that the interval traps the population value for the new state.

The prediction intervals provide reasonable guidance to the states in terms of expected number of nonfatal involvements. The range of the intervals is about  $\pm 20$  percent. This may be regarded as relatively wide, but it should be recognized that there is a lot of variability in the underlying data. There are several sources of this variability. One is in the number of fatal involvements. Figure 2 showed how the amount of variability from year to year differs with the mean number of fatal involvements, such that the variability is much higher in states with few fatals. This is because whether a person is killed in a particular crash is highly random. When there are many fatal crashes in a state, the randomness tends to wash out, but when there are only a few, the randomness can have a substantial effect on the absolute number of fatals.

There is also no doubt significant measurement error in the counts of nonfatal involvements determined in the state data. Even though the states selected coded all the information needed to identify crash involvements that meet the MCMIS Crash file reporting criteria, it is important to remember that the source of the data is an individual police officer completing a crash report. The UMTRI evaluations have shown that the accuracy of reported cases vary widely. The reporting officers often work in difficult conditions; protecting life and property, rather than accurate crash data, is their primary mission; and quality control is difficult and expensive. All these factors contribute to variability in the underlying data.

Adding further states to the model may improve the estimates and narrow the prediction intervals. Of particular interest would be to add states in areas not well-covered by the set of states available for the model at this time. States with fewer than 19 fatals, between 70 and 150, and more than 200 would help fill gaps in the range of states covered by the model. However, the prediction intervals available in the current model should provide meaningful guidance to the states.

#### 6 References

- 1 Green, P., and Blower, D., Preliminary Ratio of Crash Severities Reportable to the MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. January 2007. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 2 Blower, D., and Matteson, A., Evaluation of Florida Crash Data Reported to MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. December 2004. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 3 Blower, D., and Matteson, A., Evaluation of Iowa Crash Data Reported to the MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. August 2006. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 4 Blower, D., and Matteson, A., Evaluation of 2005 Louisiana Crash Data Reported to the MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. December 2006. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 5 Blower, D., and Matteson, A., Evaluation of Missouri Crash Data Reported to MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. January 2004. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 6 Blower, D., and Matteson, A., Evaluation of 2005 Missouri Crash Data Reported to MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. September 2006. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 7 Blower, D., and Matteson, A., Patterns of MCMIS Crash File Underreporting in Ohio. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. August 2003. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.
- 8 Green, P.E., and Matteson, A., Evaluation of 2005 Ohio Crash Data Reported to MCMIS Crash File. University of Michigan Transportation Research Institute, Ann Arbor, Michigan. November 2006. Sponsor: Federal Motor Carrier Safety Administration, U.S. D.O.T.