Special Study Large Truck and Bus Analysis for the State of Alabama Data: 2011-2015; Updated 2016



Alabama Department of Transportation and The University of Alabama Center for Advanced Public Safety (CAPS)

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1 Introduction; Crash Overview and Trends

The purpose of this document is to produce information that will be useful to Alabama decision-makers in reducing the number of fatalities and other suffering and loss due to crashes involving large trucks (LTs) and buses in the state. The following definitions of these vehicle types is important to understanding the scope and application of the results given:

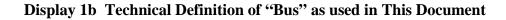
Large truck. The following is the technical logical definition for a Large Truck (LT) obtained from the CARE filter.

Filter Logi	c: Heavy Truck Causal Vehicle C101 🛛 🚽 🗖	×
Logic Tree Logic Text		
 2012-2016 Alabama Integrated Crash I 	Data: Causal Unit (CU) Type is equal to E Single-Unit Truck (3 Axles or Less) Data: Causal Unit (CU) Type is equal to E Truck (6 or 7) with Trailer Data: Causal Unit (CU) Type is equal to E Truck Tractor Only (Bobtail) Data: Causal Unit (CU) Type is equal to E Tractor/Semi-Trailer Data: Causal Unit (CU) Type is equal to E Tractor/Doubles Data: Causal Unit (CU) Type is equal to E Tractor/Triples Data: Causal Unit (CU) Type is equal to E Other Heavy Truck (Cannot Classify) Data: Causal Unit (CU) Type is equal to E Mobile Home Transport Data: Causal Unit (CU) Type is equal to E Mobile Home Transport Data: Causal Unit (CU) Type is equal to E Maintenance/Construction Vehicle Data: Causal Unit (CU) Type is equal to P Truck Tractor* Data: Causal Unit (CU) Type is equal to P Other Truck*	
18271 records selected by this filter.		

Display 1a Technical Definition of a Large Truck (LT)

While the filter definition above is for the causal vehicle (C101), the attribute numbers to which this was applied were both C101 (Causal Unit Type) and C501 (Vehicle 2 Type). These two were logically ORed together to form the subset of Large Trucks (LTs) used in this report. The objective of this specification was to obtain crashes involving all large trucks independent of their CMV specification. It is important in interpreting the results to only apply them to this subset. While it is highly representative of all large cargo vehicle types on the roadway, some might be excluded. The term Large Truck (LT) used in this document will apply only to those defined above.

Bus. Display 1b is the technical CARE filter logical definition for the word "bus" that will be used throughout this report.



8	Filter L	ogic: Heavy Bus Causal Unit (C101) 🛛 🗕 🗖 🗙	
Logic Tree	Logic Text		
2012-2016	-	(OR) Crash Data: Causal Unit (CU) Type is equal to E Other Bus (Seats More than 15) Crash Data: Causal Unit (CU) Type is equal to P Commercial Bus*	
865 records selected	by this filter.		

While the filter specification given is for C101, these attribute specification were applied to both C101 (Causal Unit Type) and C501 (Vehicle 2 Type), and they were logically ORed together to create the subset that was used to produce the results reported in this document. The objective of this specification was to obtain crashes involving all relevant buses independent of their CMV specification. The concern here is with the commercial buses, such as Greyhound, Trailways, etc., and similar interstate buses, not school or local transit buses so they were excluded, as were several other types of buses. So the word "bus" used in this study does not apply to all buses in general, only to those defined above. While the results presented are quite precise for the subset of buses as defined above, the reader must determine the extent to which this subset might infer results for buses in general for any given attribute.

Large truck involved. We define "involvement" to mean that a large truck (as defined above) was either the first (causal unit) or the second unit specified by the reporting officer in the crash report. There is no assumption of fault in the word "involved." A decision was made to exclude from the crash comparisons crashes in which the large truck was the third or higher vehicle; however, they are included in the vehicle counts.

Bus involved. Similarly, a "bus involved" crash is defined as a crash in which either the causal or the second unit was a bus (as defined above), and the other qualifiers for vehicle counts also apply.

This introduction will continue in Section 1.1 by giving some overall crash and victim counts by severity so that an appreciation for the magnitude of LT and bus crashes in the state can be better understood. It will also consider ten-year trends for LTs in Section 1.2.

Time period for the data. The basic study was done using five years of data 2011-2015. Most of these results are comparative and they will remain stable and will not change into 2017 or 2018. However, where it was seen that there were significant changes in 2016, results from this year were added to further qualify and update the results.

1.1 Overall Crash Summaries by Severity

1.1.1 Large Truck (LT) and Bus Involvement Compared to Other Crash Types

This section contains a crash comparisons of LT crashes against bus crashes and all crashes so that their relative importance by the general traffic safety community can be assessed. Display 1.1.1 presents this comparison in a nutshell. In both the LT and the bus results the percent of crashes and the percent of vehicles involved are very close to each other indicating that the number of vehicles in these types of crashes is only slightly greater than that of crashes in general. Clearly the LT crashes greatly exceed those of buses (as defined above), with the LTs having over 20 times the number of crashes, and about the same for number of vehicles involved. However, the number of fatal crashes and persons killed is in the order of magnitude of 50 times higher for LTs than for buses, indicating that their crashes are of significantly higher severity.

Statistic	Number/Frequency	Number for All Crashes	Percent of All			
Large Truck (LT) Crashe	Large Truck (LT) Crashes					
Number of Crashes	27,643	665,380	4.15%			
Total Vehicles Involved	51,626	1,215,712	4.24%			
Total Large Trucks Involved	29,250	-	-			
Number of Fatal Crashes	377	3,917	9.62%			
Number Killed	416	4,392	9.47%			
Number Injured	7,299	120,420	6.06%			
Bus Crashes						
Number of Crashes	(1,370*) 2,345	665,380	0.35%			
Total Vehicles Involved	(2,666*) 4,563	1,215,712	0.37%			
Total Buses Involved	(1,393*) 2,384	-	-			
Number of Fatal Crashes	8	3,917	0.20%			
Number Killed	8	4,392	0.18%			
Number Injured	(615*) 1,053	120,420	0.87%			

Display 1.1.1 Summary of LT and Bus Crashes Relative to All Crashes (2011-2015)

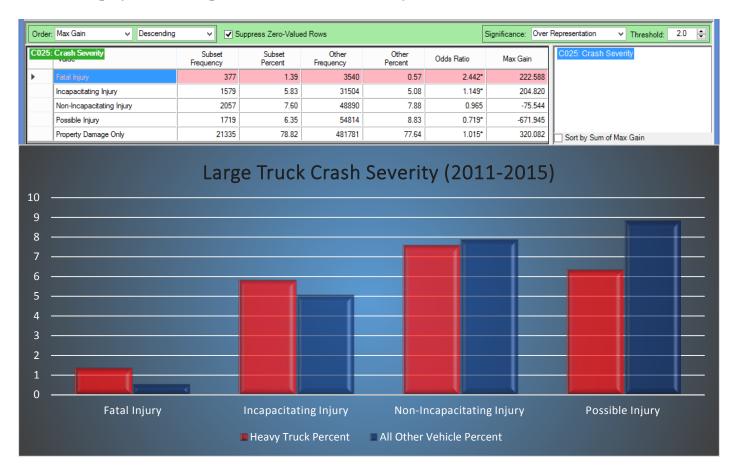
* These values have been increased by proration since an analysis of the data itself indicated that only 58.4% of the crashes were included in the five year subset; see Section 1.3 for details.

The following two sections will compare the severity of LT and bus crashes against vehicle crashes in general. The tool used in these analyses is the CARE Information Mining Prioritization, Analysis Control Technique (IMPACT), which will be describe in detail in the next section.

1.1.2 Severity IMPACT Comparing LT with non-LT Crashes

Display 1.1.2 gives a comparison of the severities of LT crashes with all other crashes. As expected, LT crashes are much more severe to the victims, with fatal crashes being over-represented by a factor of 2.442. In the table above the chart the Odds Ratio indicates this over-representation, and it has a background that has been set to turn red for all values that are twice or more times their expectation. An asterisk (*) on any of the odds ratios indicates that they are statistically significant at a very high alpha level (0.001 or less chance of stating that a difference that exists is merely due to chance). Considering these Odds Ratios, the two highest injury severities are over-represented, and the two lower injury categories are under-represented, while Property Damage Only (PDO) crashes are slightly over-represented (albeit significantly).

The comparision in Display 1.1.2 is not between LTs and buses as given above, it is between crashes that involved LTs and all crashes that did not involve an LT (LT as defined by the filter specification above). In all of the CARE IMPACT tables given in this document, the "Subset Frequency" will either be LTs or buses (in this case it is LTs), and the "Other Frequency" will be all other crashes (the complement of the subset). Similarly with the "Subset Percent" and the "Other Percent," the corresponding definitions apply.



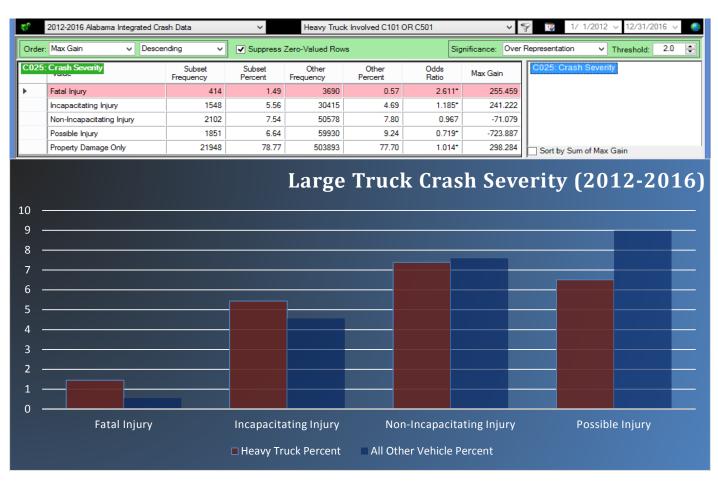
Display 1.1.2a Comparison of LT Crash Severity With All Other Crashes (2011-2015)

The double-bar chart under the table shows the comparisons for the injury categories – the PDO category was removed since it dwarfed the other categories and made the chart unreadable. So the chart results are strictly for crashes in which injuries were recorded. The trimming of the data to improve chart readability will be noted in several display in this document.

In this and all other IMPACT comparison charts, the red bars indicate LTs (or buses if it is a bus comparison); the blue bars indicate the proportion of the crashes for the complement of the subset – all other crashes. In essence the blue bars are an excellent proxy for crashes in general, and they can be considered as an experimental control subset, while the red bars represent various test subsets.

In addition to the disparity between the vehicles which collide, there are three major factors that are highly correlated with crash severity: (1) impact speed, (2) EMS response time, and (3) rural or urban crash location, which has a major effect on the first two factors. We will address the first two of these factors for LTs in the following two sections, but will defer consideration of the rural-urban factor to Section 4.1.1, where it will be considered in the context of other geographical factors.

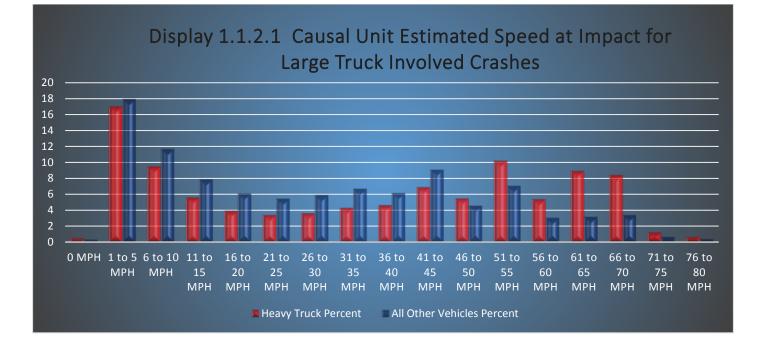
Severity comparison update for 2012-2016. Display 1.1.2b below gives the output comparable to that above but for the 2012-2016 time frame as opposed to 2011-2015. The 2016 calendar year was particularly bad for fatalities with a general 24% increase in all fatal crashes (not just trucks). This effect shows in the fatal crash numbers over the five years. However, the bar charts appear virtually the same as do the odds ratios. This demonstrates that even a dramatic change in one year does not cause the overall conclusions of the various analyses to change. In cases where they do, additional analyses will be performed. I should be noted that there has been a 10% reduction in all fatal crashes for the first 9 months of 2017 as compared to the first 9 months of 2016.



Display 1.1.2b Comparison of LT Crash Severity With All Other Crashes (2016 Updated)

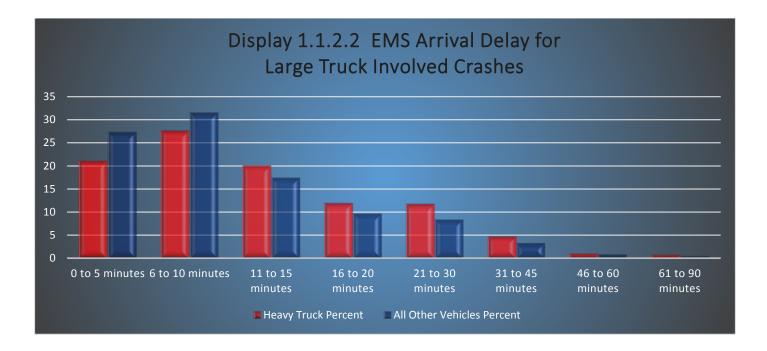
1.1.2.1 Estimated Impact Speed Comparison LT with non-LT

Display 1.1.2.1 presents a picture of the impact speeds for LT crashes. The over-represented cells are indicative of the roadways upon which these vehicles are most often driven. Higher impact speed would clearly be the cause of a higher fatality rate for LT-involved crashes. The general rule of thumb established by crash data within Alabama is that the <u>fatality probability doubles</u> for each increase in impact speed above 40 MPH.



1.1.2.2 EMS Arrival Delay Speed IMPACT comp LT with non-LT

Display 1.1.2.2 indicates that the shortest two arrival intervals are under-represeted, while all of the longer arrival intervals are over-represented. This is fairly typical of the rural-urban mix in which these trucks typically operate, and it is no doubt a contributor to the higher death rate for LTs.

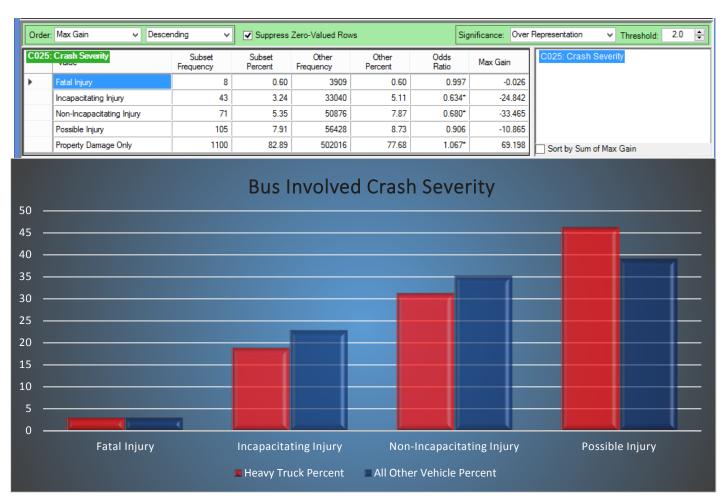


1.1.3 Severity IMPACT: Bus Involved vs non-Bus Involved Crashes

Display 1.1.3 gives a comparison of the severities of bus crashes with all other crashes. Buses are some of the safest vehicles to ride in, and the fatality figure reflects this. *None of the fatalities were recorded to have been bus occupants.* The were: 4 in passenger cars, 1 in an SUV, 2 pedestrians and 1 not recorded.

In the table above the chart the Odds Ratios indicate that the number of fatalities is not significantly different from that of the overall crash population, which has a relative refrequency of about one fatality crash in every 167 crashes (0.06%). All three of the injury classifications are under-represented, with the top most severe being significantly so. The PDO crashes for buses are over-represented, but they have not been shown in the chart in order to be able to see the relative weights of the other severity classifications. Buses had a PDO percentage of about 83%, which is significantly higher than the general PDO rate of about 78%.

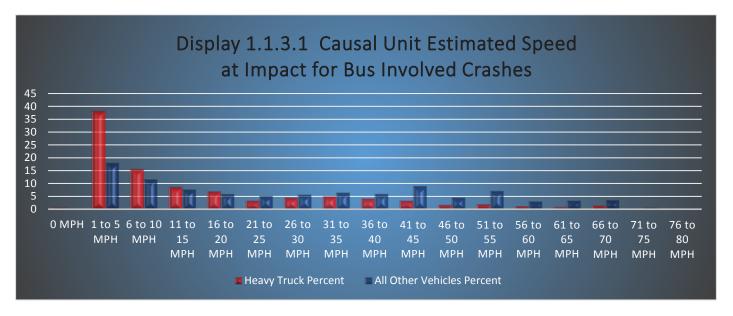
Considerations for factors involved with severity are given for buses (comparable to those for LTs given above) in the next two sections.



Display 1.1.3 Comparison of Bus Crash Severity With All Other Crashes

1.1.3.1 Estimated Impact Speed IMPACT Comparison Bus with non-Bus

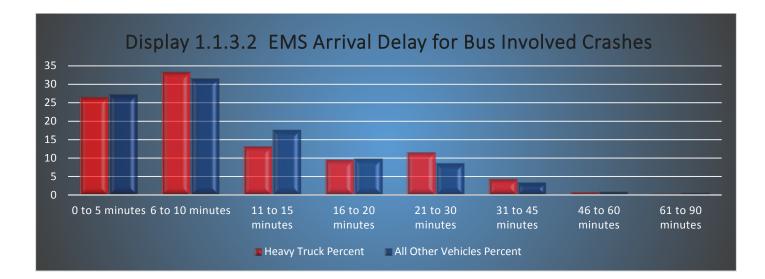
One large reason for the low severity of bus crashes has to do with impact speeds, which from Display 1.1.3.1 show a very large contrast with LTs (Display 1.1.2.1). This is fairly typical for traffic operations in general that are concentrated in urban areas. We would suspect that most buses traval at a slower rate than the traffic in general. The extreme over-representation in the 1-5 MPH category would indicate a large number of backing crashes and collisions with parked vehicles. These will be discussed further in the bus crash analysis in Section 6.



Display 1.1.3.1 Speed of Impact Comparison of Bus and non-Bus Crashes

1.1.3.2 EMS Arrival Delay Speed IMPACT comp Bus with non-Bus

Display 1.1.3.2 shows the comparison between buses and non-buses as far as ambulance arrival time is concerned. Probably because of the low sample sizes involved for bus crashes, none of the values for delay time were significantly different from what would be expected in the general population of crashes. While EMS arrival delay is not an issue in general, it is a major factor in certain remote crashes that involve heavy injury, and further research is warranted on this subject.

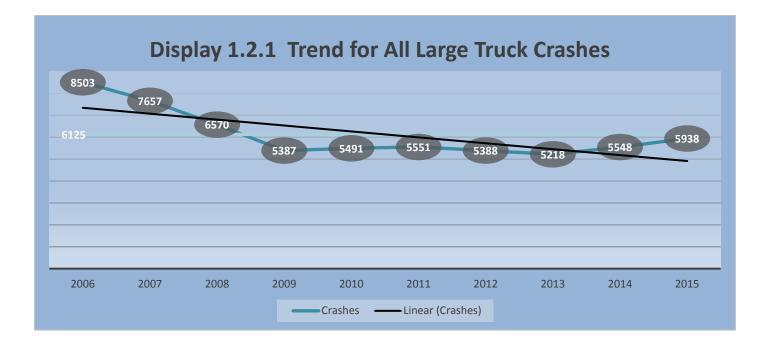


1.2 Trend Analysis for Large Truck Crashes

1.2.1 Total Large Truck (LT) Crash Trend

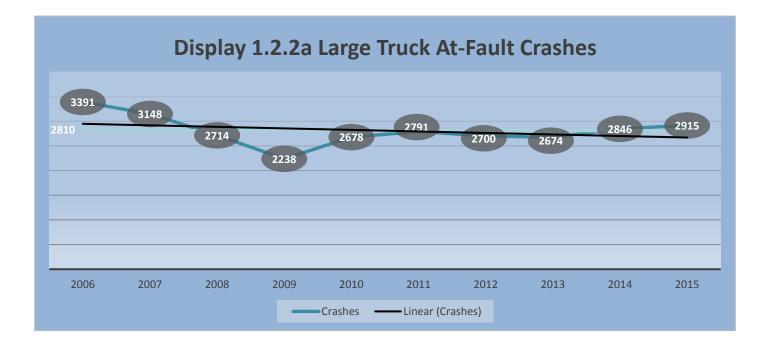
Display 1.2.1 shows the ten year (2006-2015) trend for all large truck (LT) crashes. The display contains (1) the actual data in number of crashes per year; (2) a regression line that indicates the overall trend; and (3) a constant average line for general reference. Clearly the trend is down reflecting the overall trend in traffic crashes over the past ten years, due largely to the reduction in mileage due to the economic downturn. Unfortunately, it is also following the crash pattern of the past three years which is trending back up. The slope of the regression line is -270.00 crashes per year, estimating an average reduction of 2700 crashes over the ten year period, with a correlation coefficient (r-squared) of 0.5364, indicating that 53.64% of the variance in the number of crashes is correlated to the year value.

It is important to recognize that the number of LT crashes is dependent not only on the number of LTs on the road, but also on the other non-LT traffic. In fact, the decline in LT mileage during the relative recession did not diminish nearly to the extent that overall traffic did. LT mileage has a relatively small discretionary proportion since many goods and services continue to be delivered independent of the economic conditions. The same number of trucks in a traffic mix that contains a significantly higher number of passenger vehicles will also result in more crashes. Other factors contributing to the trend included: (1) a leveraged economic effect on the worse drivers – young people and those driving older vehicles; so that a relatively small economic downturn can have a much larger effect on crashes; (2) the cost of fuel, which was higher during the downturn, but is now significantly lower than it was at its lowest level; and (3) the effect of the lack of personal income on the purchase of alcoholic beverages, especially in restaurants and clubs.



1.2.2 Large Truck At-Fault Trend

Display 1.2.2a is limited to those crashes that were caused by large trucks. The overall pattern is about the same but the overall trend is not statistically significant. Display 1.2.2b presents the comparison of total and at-fault crashes for LTs over the ten years. It increased and peaked in the 2011-2013 time frame, but has decreased slightly in the two more recent years. A more detailed at-fault analysis is given in Section 2.1. The results there are slightly different in that only the first two vehicles were considered, whereas in this analysis all LTs were considered, and those with unit numbers 2 or above were considered non-causal.

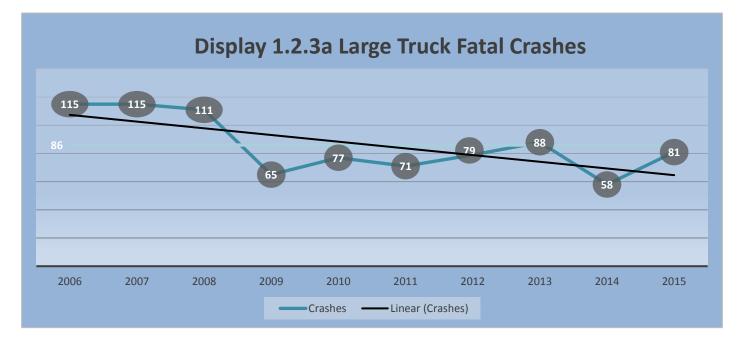


Display	1.2.2b	Large	Truck	Proportion	At-Fault Trend
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Year	All Large Truck Related	Large Truck At-Fault	Large Truck At-Fault Percent
2006	8,503	3,391	39.88%
2007	7,657	3,148	41.11%
2008	6,570	2,714	41.31%
2009	5,387	2,238	41.54%
2010	5,491	2,678	48.77%
2011	5,551	2,791	50.28%
2012	5,388	2,700	50.11%
2013	5,218	2,674	51.25%
2014	5,548	2,846	51.61%
2015	5,938	2,915	49.09%
Total	61,251	28,095	45.87%

1.2.3 Fatality Large Truck Crash Trend

In Section 2.1 it will be demonstrated that large trucks account for only about 20% of fatality crashes in which they are involved, and thus it is expected that the pattern shown in Display 1.2.3a would reflect the overall statewide fatality figures. They average out to be about 9.03% of the total fatalities. Display 1.2.3b shows how this percentage has varied over the ten years of the study. The slope of the regression line is about -4.75 crashes per year, estimating an average reduction of 48 fatality crashes over the ten year period, with a correlation coefficient (r-squared) of 0.4756, indicating that 47.56% of the variance in the number of fatal crashes is correlated to the year value.



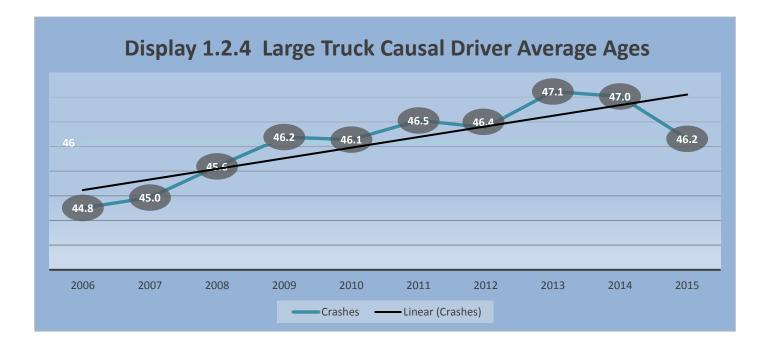
Display 1.2.3b	Comparison	with Statewide Fatalities	
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Year	Large Truck Related	All Fatalities	Large Truck Percent
2006	115	1,207	9.52%
2007	115	1,110	10.36%
2008	111	967	11.48%
2009	65	847	7.67%
2010	77	862	8.93%
2011	71	895	7.93%
2012	79	865	9.13%
2013	88	853	10.32%
2014	58	820	7.07%
2015	81	849	9.54%
Total	845	9275	9.20%

1.2.4 Large Truck Crash Causal Driver Age Trend

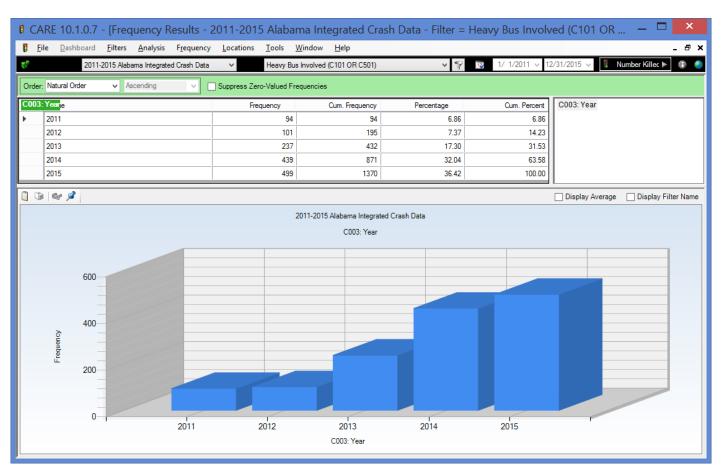
The ten year average causal driver age trend is given in Display 1.2.4. These include both the LT drivers and the non-LT drivers for multi-vehicle LT involved crashes. However, because of the high number of LT drivers in the subset, this will be heavily influenced by their ages. It is expected that this age will generally increase, and so the results in Display 1.2.4 are expected.

The slope of the regression line is 0.2153 average driver age (in years) per year, estimating an average increase of about 2 years over the ten year period, with a correlation coefficient (r-squared) of 0.6961, indicating that 69.61% of the variance in the number of crashes can be attributed to the year value.



1.3 Issues with the Trend Analysis for Bus Crashes

Display 1.3 presents the overall summary of bus crashes by year for the five years of the bus dataset.





The data prior to CY2014 is both inconsistent and incomplete, and thus Display 1.3 should be used only for information about the data – it is erroneous to infer that this indicates an increasing number of bus crashes.

There were two reason for the data deficiencies: (1) the shift from paper to eCrash in July 2009; and (2) a recent edict from FMCSA that forced the officers to report buses differently starting sometime mid-2010. The eCrash data elements were structured differently from those within the paper data collection form, and the mapping from one to the other was quite problematic. On top of that, FMCSA mandated another change that was implemented during mid-CY2013. This change is observable in Display 1.3 in that the pre-2013 years are at one level, and the post 2013 years are at a different level. A further complication is that all officers were not trained at reporting bus crashes in the revised way simultaneously. So, it was possible that

while most of the ALEA officers were probably aware of the required change, many local law enforcement might still be reporting in the traditional way.

A number of attempts were made to create filters that would normalize the reported bus crashes over a ten year period. Based on this experimentation, it was determined that the data elements as they currently exist could not create any true picture of reality over this entire period. The recommendation was to allow the data to stabilize and begin measuring any trends starting from CY2014.

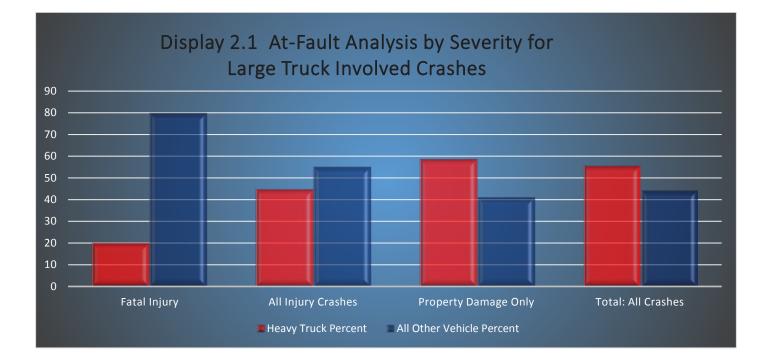
The problem in not getting exactly the same vehicles over the years is NOT as much a problem with IM-PACTs since the filter defines the subset and whatever qualifies goes in and creates that subset which is a sample of all such records (crashes) that qualify. The inferences that we make on the whole based on this sample are valid. To illustrate, if the last two years are considered to be a 100% sample of those two years, then the average of these two is 469 bus crashes per year. This will give an average total of 2,345 over the five year period. The actual total is 1,370. This is an estimated 54.8% sample of the crashes over the five year period. Assuming this sample is internally consistent according to the filter applied, this is an excellent sample size for making the inferences that have been made in the IMPACT analyses. A random sample of this size if far more reliable than a larger sample for which the randomness is of question.

2 At-Fault Analyses by Severity

This At-Fault analyses required that there be two vehicles in the crash, so all single vehicle crashes were excluded from consideration. For LTs, the objective is to determine for any crash involving a large truck and another type of vehicle, what is the probability that the LT was at fault. Similarly for bus crashes. These are discussed in the next two sections, respectively.

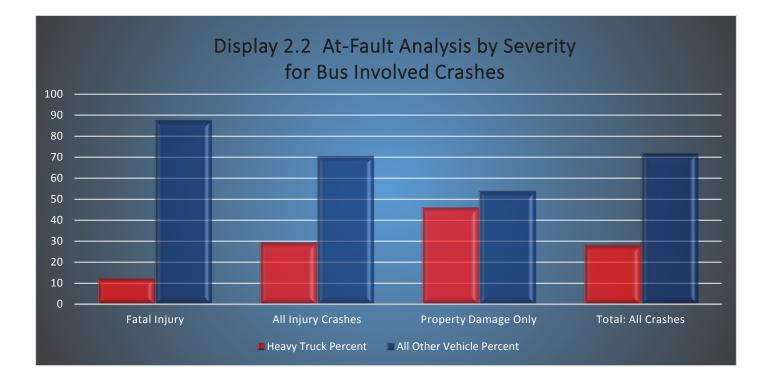
2.1 At Fault Analysis for Large Trucks

Display 2.1 presents the findings from the LT at-fault analysis. All other things being equal, we would expect that all of these bars would be at 50%. For these at-fault charts, any two bar set will sum to 100%. The greater the deviation above 50% for either vehicle type, the greater their chances of causing the crash. As shown in the display, the probability of a LT being at-fault in any given two-vehicle crash is highly dependent on the severity of the crash. It might be expected for crashes in general that LTs would be more often at fault just from the skill that it takes to handle such a large vehicle. The over-representation in PDO and Total (includes PDOs) is somewhat expected. Their dramatic under-representation in fatal crashes is notable since these crashes involve much more detailed investigation, and thus it is not just the opinion of one reporting officer.



2.2 At Fault Analysis for Buses

Display 2.2 presents the findings from the bus at-fault analysis. There were eight fatal crashes in the bus dataset used for analysis (2001-2005), and only one of them was caused by the bus unit. While this is a very small sample size, the relative low probabilities for all crashes, and especially for all injury crashes, tends to validate this finding. It is clear that buses have an excellent record for all severity classifications, but especially for those involving injury.

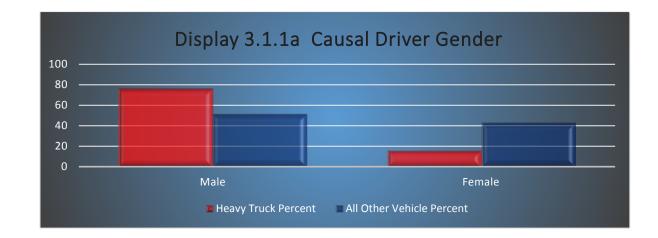


3 Driver Analysis for Large Trucks

The goal of the LT driver analysis is to determine differences between the causal drivers in LT-involved crashes and those in non-LT crashes. All causal drivers are being considered together in these analyses, and there should be no inference that the causal driver is the LT driver. The rationale for including all causal drivers is that countermeasures for LT crashes are as important for the non-LT vehicle and driver as they are for the LT.

3.1 Driver Demographics

3.1.1 Causal Driver Gender IMPACT comparison



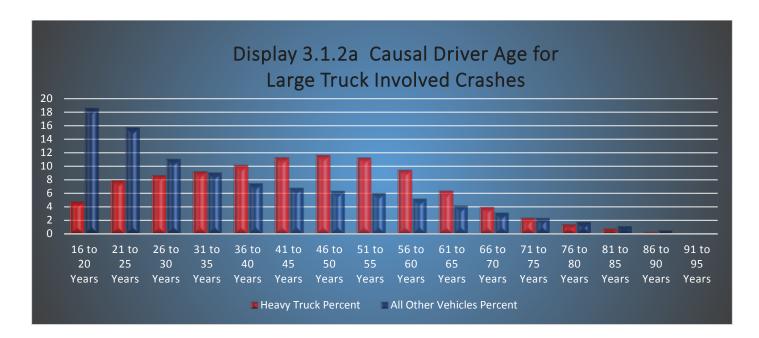
Display 3.1.1a presents the findings from the LT causal driver gender analysis.

Display 3.1.1b Crash Severity by Gender

🔋 CARE 10.1.0.7 - [Crosstab Results - 2011-2015 Alabama Integrated Crash Data - Filter = He 🗕 🗖 💌											
🚦 <u>F</u> ile <u>D</u> ashb	ooard <u>F</u> ilters <u>/</u>	<u>A</u> nalysis <u>C</u> rossta	b <u>L</u> ocations <u>T</u>	ools <u>W</u> indow	<u>H</u> elp			_ 8 ×			
🥵 2011-2015 Alabama Integrated Crash Data 🗸 Heavy Truck Involved C101 OR C501 🗸 🖓 😨 1/ 1/2011 v 12/3											
Suppress Zero Va	Suppress Zero Values: None 🗸 Select Cells: 🖬 🗸 🦃 Column: Crash Severity ; Row: CU Driver Gender 🛺										
	Fatal Injury	Incapacitating Injury	Non- Incapacitating Inju	Possible Injury	Property Damage Only	Unknown	TOTAL	^			
Male	284	1210	1591	1325	16181	423	21014	1 -			
Male	75.33%	76.63%	77.35%	77.08%	75.84%	73.44%	76.02%				
Female	86	344	410	341	3138	101	4420				
remaie	22.81%	21.79%	19.93%	19.84%	14.71%	17.53%	15.99%	×			

It is expected by the general proportion of male truck drivers that males will be predominate in being the causal driver. Even for those crashes caused by the non-LTs, males are over-represented. Thus, the findings of Display 3.1.1a are not surprising. A further analysis by severity, however, shows the female causal drivers to have far more than their share of fatal and injury crashes (all injury severities). The red in the cross-tabulation indicates that the percentage of the cell is more than 10% higher than its expectation (e.g., for fatal injury, 22.81% is greater than 10% higher than 15.99%). These results would infer that when women cause the crash, it results in a much more severe crash. Additional study needs to be done on this surprising finding.

3.1.2 Causal Driver Age



Display 3.1.2 presents the findings from the LT causal driver age analysis.

There is a great disparity between causal driver age of LT-involved crashes and crashes in general. The is seen especially in the younger driver ages (16-30). Obviously at the extreme youngest ages, these individuals are not driving LTs, and the predominance of their crash locations (e.g., near school zones) does not bring them into close proximity to LTs. Instead we observe almost a perfectly normal distribution centered arnd 46-50 years, which tends to match the mean of the professional driver age.

Display 3.1.2b provides a further subdivision of these ages by crash severity. The cells that tend to be overrepresented are at the youngest and oldest extremes.

🕴 CARE 10.1.0.7 - [Crosstab Results - 2011-2015 Alabama Integrated Crash Data - Filt 🗕 🗖 💌										
🚦 <u>F</u> ile <u>D</u> ashl	board <u>F</u> ilters	<u>A</u> nalysis <u>C</u> rossta	b <u>L</u> ocations <u>T</u>	ools <u>W</u> indow	<u>H</u> elp		_ & ×			
¢?	2011-2015 Alabam	a Integrated Crash Da	ata 🗸	Heavy Truck Involv	ed C101 OR C501	~	1/ 1/20			
Suppress Zero Va	lues: None	✓ Select	Cells: 🔳		Column: Crash Seve	erity ; Row: CU Drive	er Age Range 2 👔			
	Fatal Injury	Incapacitating Injury	Non- Incapacitating Inju	Possible Injury	Property Damage Only	Unknown	TOTAL			
0 to 5 Years	0	1	0	0	3	0	4			
	0.00%	0.06%	0.00%	0.00%	0.01%	0.00%	0.01%			
6 to 10 Years	1 0.27%	0.00%	0.00%	0.00%	0.00%	0.00%	1 0.00%			
11 to 15 Years	2	4	1	2	16	1	26			
	0.53%	0.25%	0.05%	0.12%	0.07%	0.17%	0.09%			
16 to 20 Years	29	85	127	89	852	28	1210			
	7.69%	5.38%	6.17%	5.18%	3.99%	4.86%	4.38%			
21 to 25 Years	33	129	190	133	1468	43	1996			
	8.75%	8.17%	9.24%	7.74%	6.88%	7.47%	7.22%			
26 to 30 Years	42	145	155	153	1639	46	2180			
	11.14%	9.18%	7.54%	8.90%	7.68%	7.99%	7.89%			
31 to 35 Years	34	150	202	157	1733	48	2324			
	9.02%	9.50%	9.82%	9.13%	8.12%	8.33%	8.41%			
36 to 40 Years	31	171	210	167	1939	44	2562			
	8.22%	10.83%	10.21%	9.71%	9.09%	7.64%	9.27%			
41 to 45 Years	28	195	209	202	2173	35	2842			
	7.43%	12.35%	10.16%	11.75%	10.19%	6.08%	10.28%			
46 to 50 Years	32	165	218	174	2295	48	2932			
	8.49%	10.45%	10.60%	10.12%	10.76%	8.33%	10.61%			
51 to 55 Years	29	160	211	178	2189	67	2834			
	7.69%	10.13%	10.26%	10.35%	10.26%	11.63%	10.25%			
56 to 60 Years	33	134	195	130	1846	47	2385			
	8.75%	8.49%	9.48%	7.56%	8.65%	8.16%	8.63%			
61 to 65 Years	16	79	111	113	1242	45	1606			
	4.24%	5.00%	5.40%	6.57%	5.82%	7.81%	5.81%			
66 to 70 Years	14	60	76	71	776	20	1017			
	3.71%	3.80%	3.69%	4.13%	3.64%	3.47%	3.68%			
71 to 75 Years	18	28	43	40	462	23	614			
	4.77%	1.77%	2.09%	2.33%	2.17%	3.99%	2.22%			
76 to 80 Years	14	22	29	27	265	10	367			
	3.71%	1.39%	1.41%	1.57%	1.24%	1.74%	1.33%			
81 to 85 Years	7	16	12	12	157	5	209			
	1.86%	1.01%	0.58%	0.70%	0.74%	0.87%	0.76%			
86 to 90 Years	6	6	6	2	60	5	85			
	1.59%	0.38%	0.29%	0.12%	0.28%	0.87%	0.31%			
91 to 95 Years	1	1	0	1	11	0	14			
	0.27%	0.06%	0.00%	0.06%	0.05%	0.00%	0.05%			
More than 95	0	9	2	11	251	0	273			
Years	0.00%	0.57%	0.10%	0.64%	1.18%	0.00%	0.99%			
Not Applicable	5	6	9	4	119	2	145			
	1.33%	0.38%	0.44%	0.23%	0.56%	0.35%	0.52%			
Unknown	0	13	51	53	1839	59	2015			
	0.00%	0.82%	2.48%	3.08%	8.62%	10.24%	7.29%			
CU is Not a	2	0	0	0	0	0	2			
Vehicle	0.53%	0.00%	0.00%		0.00%	0.00%	0.01%			
CU is Unknown	0 0.00%	0	0 0.00%	0	0	0 0.00%	0.00%			
TOTAL	377	1579	2057	1719	21335	576	27643			
	1.36%	5.71%	7.44%	6.22%	77.18%	2.08%	100.00%			

Display 3.1.2b Causal Driver Age by Severity

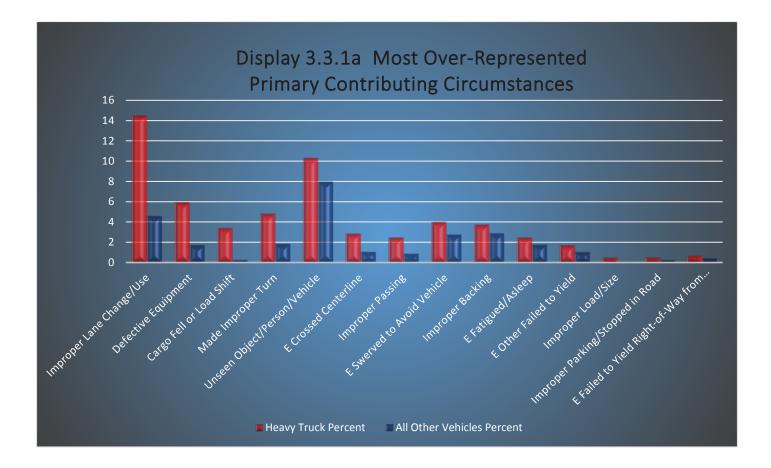
3.3 Driver Related Crash Causes

These are given in their respective sections:

- 3.3.1a Primary Contributing Circumstance (PCC) Over-Representations
- 3.3.1b Primary Contributing Circumstance (PCC) Under-Representations
- 3.3.2 First Harmful Event
- 3.3.3 Manner of Crash
- 3.3.3 Vehicle Maneuver

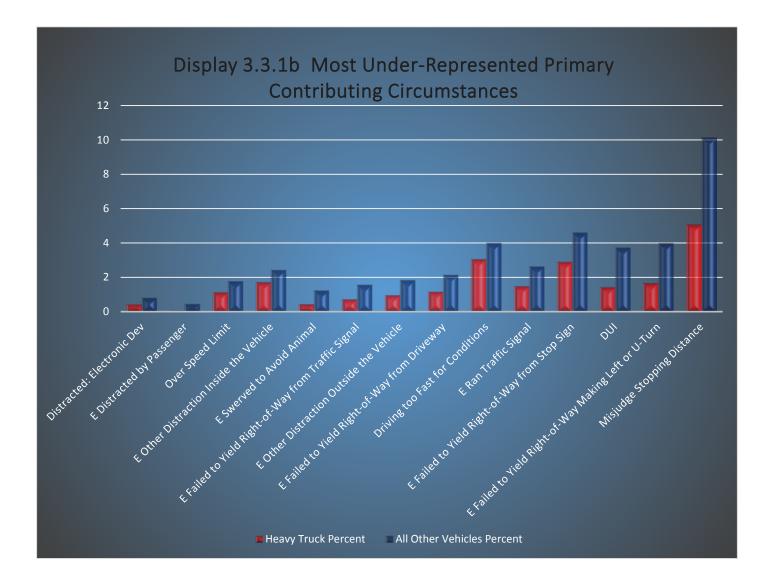
3.3.1a PCC IMPACT comparing LT with non-LT (Over-Representations)

Display 3.3.1a presents the findings from the LT Primary Contributing Circumstance (PCC) analysis. This is the PCC for the entire crash and thus is not linked to any vehicle, causal or otherwise. None of these are particularly surprising in that they are all heavily associated with large trucks.



3.3.1b PCC IMPACT Comparing LT with non-LT (Under-Representations)

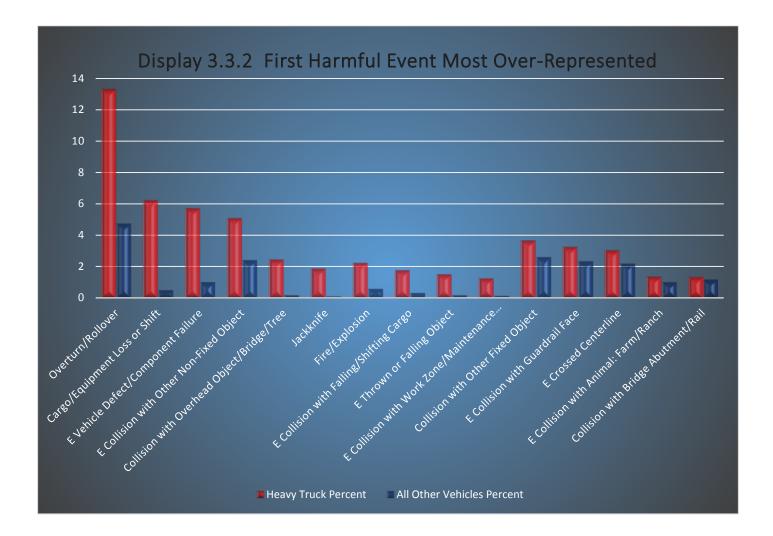
Display 3.3.1b presents the findings from the LT PCC analysis from the other end of the spectrum – those PCCs that are least likely to be associated with LT involved crashes.



Several of these are of note since they tend to reflect the professionalism of the LT drivers, e.g., under-representation in (1) several distracted driving categories, (2) several failure to yield categories, (3) two high speed categories, and (4) misjudge stopping distance, which is highly correlated with lack of drive experience. Also, truckers typically have relatively few problems with DUI.

3.3.2 First Harmful Event IMPACT Comparing LT with non-LT

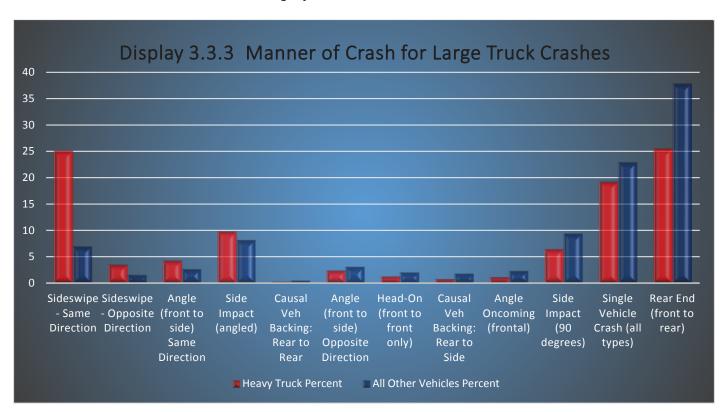
Display 3.3.2 presents the findings from the First Harmful Event (FHE) analysis. This is the FHE for the entire crash and thus is not linked to any vehicle, causal or otherwise.



The highest percentage First Harmful Event (FHE), Collision with Vehicle in Traffic, was excluded from this display in order to keep from dwarfing the other categories. The percentage of Collision with Vehicle in Traffic was 72.48% for non-LTs but was only 70.12% for LTs (significantly under-represented), although still the predominant FHE. As with the over-represented PCC categories, most of the over-represented FHEs are closely associated with large truck issues.

3.3.3 Manner of Crash IMPACT comparing LT with non-LT

Display 3.3.3 presents the findings from the Manner of Crash analysis. This is the Manner of Crash reported for the entire crash and thus is not linked to any vehicle, causal or otherwise.

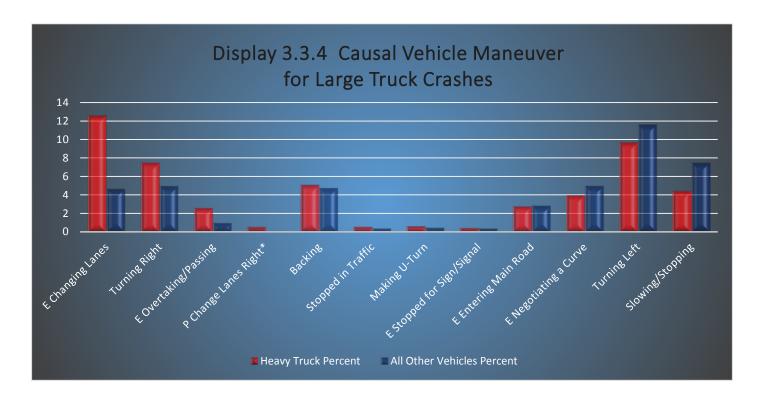


Display 3.3.3 Manner of Crash

The display contains both the over- and under-represented Manner of Crash values. Sideswipes in the same direction are by far the largest over-represented value, which is clearly an issue of the truck blind spots, often resulting from passenger cars staying in these blind spots for too long. Typically this would not be a problem in heavy traffic where cars are essentially forced to stay in the blind spots, since it is unlikely that trucks would change lanes in this situation. This is an issue that can and should be dealt with by educating the general driving public. Cruise control is often at fault since if it is depended upon it can keep cars in blind spots indefinitely. Future innovations need to detect and warn drivers when they are in blind spots since they are often too busy with their electronic devices to notice.

3.3.4 Causal Unit Vehicle Maneuver IMPACT Comparison

Display 3.3.4 presents the findings from the Vehicle Maneuver analysis. This is the Vehicle Maneuver reported for the causal vehicle only.



Changing Lanes in this analysis is closely related to Same Direction Sideswipes in the Manner of Crash analysis in the previous section. It is interesting to contrast Turning Right category with Turning Left, which would seem to be under-represented in that the driver has better visibility on the left side. Slowing/Stopping in this attribute corresponds heavily to Rear Ends in the Manner of Crash and to Misjudging Stopping Distance in the PCC attribute.

4 Large Truck Crash Analysis

Crash analyses are subdivided into two broad classifications (1) geographical and roadway aspects, and (2) time factors.

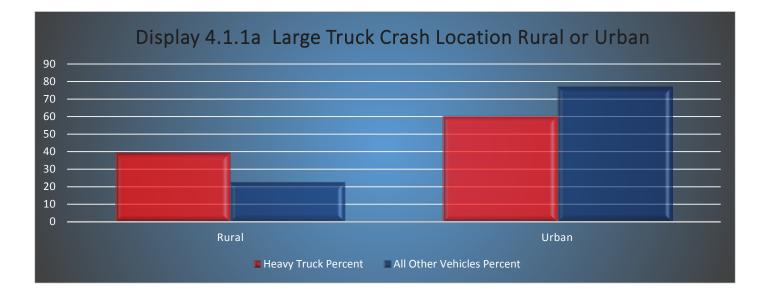
4.1 Geographical and Roadway Aspects

Geographical and roadway aspects are further subdivided as follows:

- 4.1.1 Rural-Urban
- 4.1.2 Highway classification
- 4.1.3 Weather
- 4.1.4 Roadway curvature and grade
- 4.1.5 Workzones

4.1.1 Rural-Urban IMPACT Comparing LT Involved with non-LT

Display 4.1.1a presents the findings from the Rural-Urban analysis. This is and attribute of the crash that is independent of causal vehicle.



While urban crashes outnumber the rural crashes for LTs, they are over-represented in rural areas almost by a factor of 2. This has to do with the types of roads that they run on, which is considered next. The more lethal aspects of rural roadways cannot be underestimated, and it is largely due to the increased speeds in the rural road system. Display 4.4.1 shows the crash severity broken down by the rural-urban classification. It

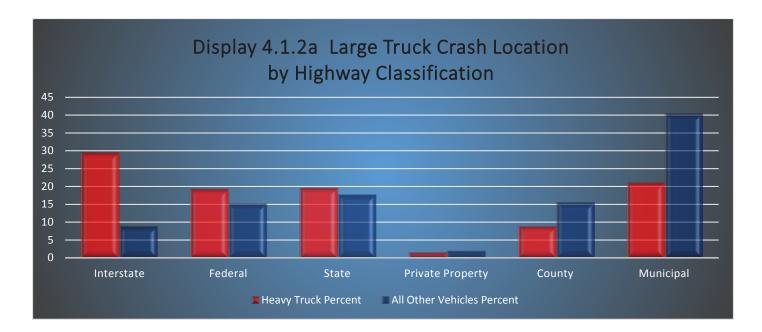
is clear that fatal crashes and those of the higher two severity classifications are over-represented in the rural areas.

CARE 10	.1.0.7 - [Cross	tab Results -	2011-2015 A	Alabama Inte	grated Crash	Data - Filter	= Heavy Tru	ck In 🗕 🗖 🔀	
🖡 <u>F</u> ile <u>D</u> asi	hboard <u>F</u> ilters	<u>A</u> nalysis <u>C</u> rossta	b <u>L</u> ocations <u>T</u>	ools <u>W</u> indow	<u>H</u> elp			_ 8 >	
¢°	😵 2011-2015 Alabama Integrated Crash Data 🗸 Heavy Truck Involved C101 OR C501 v 💡 🔞 1/ 1/2011 v 12/31/2015 v 🕽 🌖								
Suppress Zero V	Suppress Zero Values: None 🗸 Select Cells: 🔹 🌱								
	Fatal Injury	Incapacitating Injury	Non- Incapacitating Inju	Possible Injury	Property Damage Only	Unknown	TOTAL		
Rural	231 61.27%	912 57.76%	1150 55.91%	457 26.59%	8185 38.36%	12 2.08%	10947 39.60%		
Urban	146 38.73%	667 42.24%	907 44.09%	1262 73.41%	13150 61.64%	564 97.92%	16696 60.40%	-	
TOTAL	377	1579 5.71%	2057 7.44%	1719 6.22%	21335 77.18%	576 2.08%	27643 100.00%		
								,	

Display 4.1.1b Crash Severity by Rural-Urban Classification – Large Trucks

4.1.2 Highway Classification IMPACT Comparing LT Involved with non-LT

Display 4.1.2a presents the findings from the Highway Classification analysis. This is an attribute of the crash that is independent of causal vehicle.



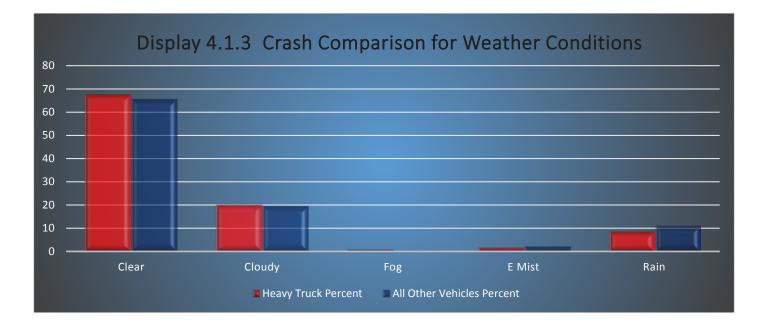
There is no doubt that a primary cause of the over-representations are due mainly to the mileage that LTs put on these various roadway types. It is interesting to do a severity analysis to see the ramification of the various highway classification crashes. This is shown in Display 4.1.2b. It is quite interesting that Interstate highways, which clearly have the highest speeds for LTs are under-represented in fatalities and the highest non-fatal injury classification. We can propose two reasons for this: (1) the forgiveness of the roadway in providing guardrails, impact attenuators and other countermeasures to mitigate the effect of many crashes, (2) clear roadsides and ample shoulders, and (3) the speed with which Emergency Medical Services are able to respond.

CARE 10.1	.0.7 - [Cross	tab Results -	2011-2015	Alabama Inte	grated Crash	Data - Filter	= Heavy Tru	ck Inv — 🗖 🗙		
🚦 <u>F</u> ile <u>D</u> ashb	ooard <u>F</u> ilters <u>/</u>	<u>A</u> nalysis <u>C</u> rosstal	b <u>L</u> ocations <u>T</u>	ools <u>W</u> indow	<u>H</u> elp			_ 8		
¢°	2011-2015 Alabam	a Integrated Crash Da	ita V	Heavy Truck Involve	ed C101 OR C501	×	🌱 🌠 1/ 1	/2011 🗸 12/31/2015 🗸 🕞		
Suppress Zero Values: None 🗸 Select Cells: 🗐 🗸 Column: Crash Severity ; Row: Highway Classifications 🦓										
	Fatal Injury	Incapacitating Injury	Non- Incapacitating Inju	Possible Injury	Property Damage Only	Unknown	TOTAL			
Interstate	103	401	616	427	6532	73	8152]		
Interstate	27.32%	25.40%	29.95%	24.84%	30.62%	12.67%	29.49%]		
Federal	94	423	471	409	3826	110	5333]		
rederal	24.93%	26.79%	22.90%	23.79%	17.93%	19.10%	19.29%]		
State	133	426	478	398	3825	162	5422]		
Sidle	35.28%	26.98%	23.24%	23.15%	17.93%	28.13%	19.61%]		
County	33	175	249	100	1882	28	2467			
County	8.75%	11.08%	12.11%	5.82%	8.82%	4.86%	8.92%			
Municipal	14	152	234	375	4841	197	5813			
Municipal	3.71%	9.63%	11.38%	21.82%	22.69%	34.20%	21.03%			
Private Property	0	2	9	10	425	6	452			
r invate Froperty	0.00%	0.13%	0.44%	0.58%	1.99%	1.04%	1.64%			
P Other*	0	0	0	0	4	0	4]		
r other	0.00%	0.00%	0.00%	0.00%	0.02%	0.00%	0.01%			
TOTAL	377	1579	2057	1719	21335	576	27643]		
TOTAL	1.36%	5.71%	7.44%	6.22%	77.18%	2.08%	100.00%			

Display 4.1.2b Crash Severity by Highway Classification

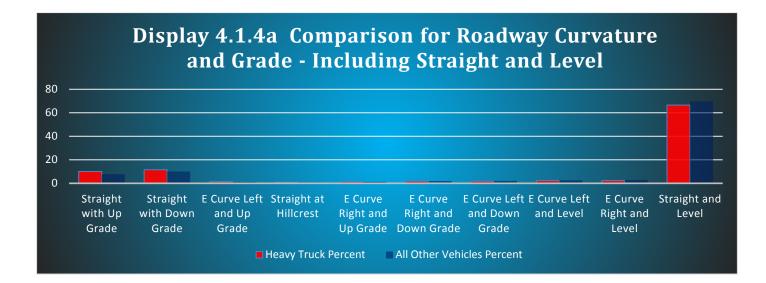
4.1.3 Weather IMPACT comp LT with non-LT

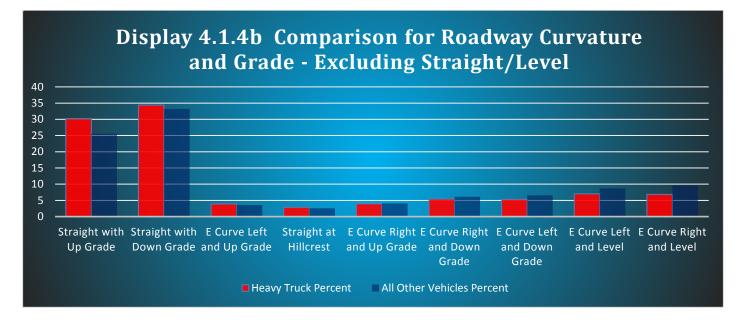
Display 4.1.3 presents the findings from the Weather analysis. LTs were found to be under-represented in rain when compared to other vehicle types. This is probably due to the truck drivers' experience in wet weather.



4.1.4 Roadway Curvature and Grade IMPACT Comparing LT with non-LT

Display 4.1.4a and b present the results for curvature and grade of the causal vehicle. The primary issue seems to be with grades, either up or down. The straight and level category was eliminated from Diplay 4.4.1b in order to make the relative effects of the other categories more visible.





Display 4.1.4b Crash Comparison by Roadway Curvature and Grade – Excluding Straight/Level

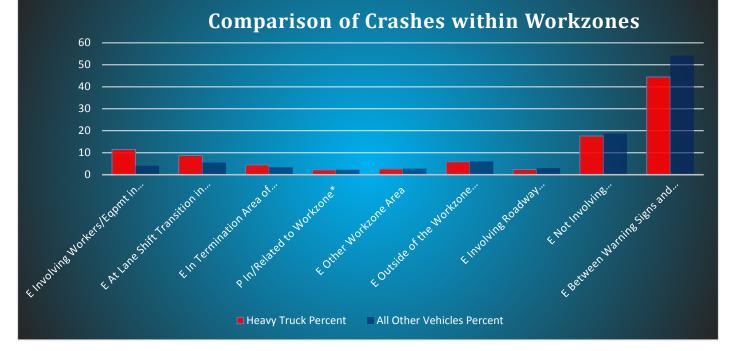
4.1.5 Workzone IMPACT comp LT with non-LT for Crashes in Workzones

During the five year peiord (2011-2015), there were 1,848 LT crashes that involved workzones, which was about 6.7% of all LT involved crashes. Only 1.7% of non-LT vehicles were involved workzones, so the LTs are dramatically over-represented in these crashes. Display 4.1.5 presents the distribution of those LT crashes that occurred within workzones comparing them to the proportions for all other vehicle types. The non-workzone crashes were excluded from Display 4.15, since this amounted to 93.16% of the LT crashes and 98.24% of the nonLT crashes, and thus its inclusion would have made the other attributes unreadable.

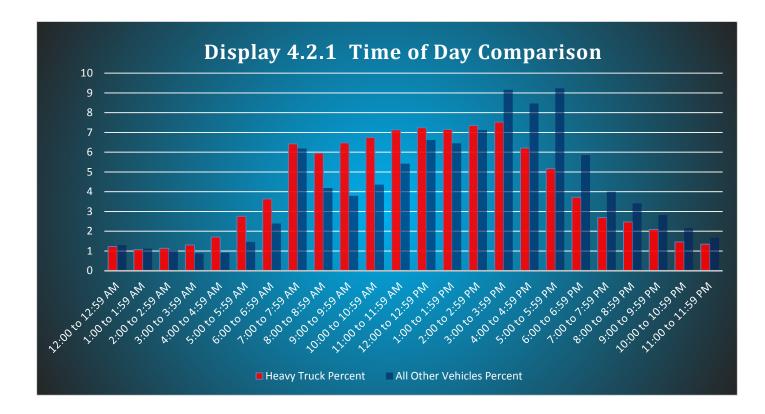
The following factors might be given special consideration: (1) Some of the 213 crashes involving equipment could be LT equipment; (2) the most under-represented should not be neglected because it has the highest frequency – 822; and (3) while the workzone reporting codes are not mutually exclusive, they probably give the best description for any particular crash.

Display 4.1.4 Crash Comparison of Crashes within Workzones

¢?	2011-2015 Alabama Integrated Crash Data	~	Heavy Truck I	nvolved C101 O	R C501	¢	9 😨 1	/ 1/2011 👻 12/31/2015 🗸 🚹 🥞
Orde	r: Max Gain v Descending v	Suppress Zero	-Valued Rows		[Significance:	Over Representat	ion v Threshold: 2.0 🛓
C41	5: CU Workzone Related	Subset Frequency	Subset Percent	Other Frequency	Other Percent	Odds Ratio	Max Gain	C415: CU Workzone Related
•	E Involving Workers/Eqpmt in Activity Area	213	11.53	431	4.07	2.831*	137.753	
	E At Lane Shift Transition in Activity Area	162	8.77	590	5.57	1.573*	58.994	
	E In Termination Area of Workzone	83	4.49	357	3.37	1.332*	20.673	
	P In/Related to Workzone*	41	2.22	239	2.26	0.983	-0.726	
	E Other Workzone Area	49	2.65	292	2.76	0.961	-1.979	
	E Outside of the Workzone Warning Signs	109	5.90	644	6.08	0.969	-3.434	
	E Involving Roadway Conditions in Activity Area	43	2.33	315	2.98	0.782	-11.995	
	E Not Involving Workers/Conditions in Activity Area	326	17.64	1980	18.71	0.943	-19.682	
	E Between Warning Signs and Work Area	822	44.48	5737	54.20	0.821*	-179.604	Sort by Sum of Max Gain

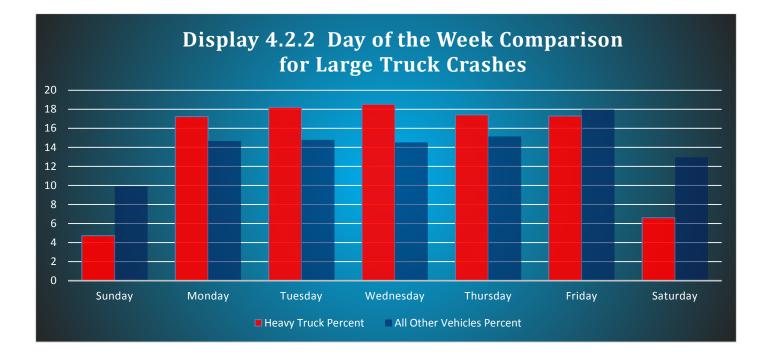


4.2 Time Factors



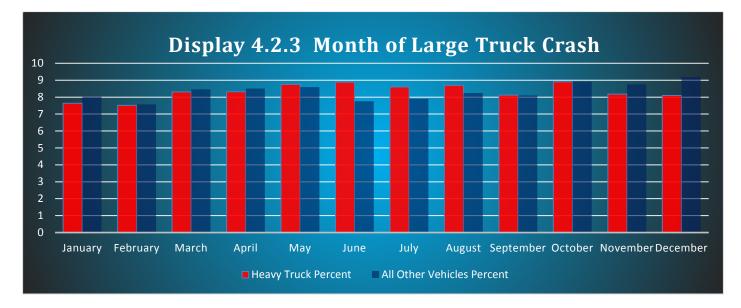
4.2.1 Time of Day IMPACT Comparing LT with non-LT

Time of day is closely related to the times of exposure, and the crash distribution could almost be a proxy of when the LTs are on the road. There is some over-representation early morning from 4 AM through 7 PM, but the frequencies are not high at this time. The morning rush hour is a problem, but LTs are not significantly over-represented in the 7;00 to 7:59 time frame, and the presence of other issues at these times would probably not make either the morning or the afternoon rush hours good times to do enforcement. The most under-represented times are late afternoon on through to before midnight. The optimal times for enforcement would seem to be from 8:00 AM through 2:59 PM.



4.2.2 Day of the Week IMPACT comp LT with non-LT

Similar to time of day, the day of the week distribution, given in Display 4.2.2, reflects the days that the majority of LT vehicles are on the road. Saturday and Sunday are probably way down because of the reluctance of those receiving freight to be working on those days. Friday is the only under-represented week-day, and although the frequency is quite high on Friday, officers are generally working on other issues on Friday, especially in the afternoons. The optimal time for enforcement would seem to be Monday through Thursday.



4.2.4 Month IMPACT comp LT with non-LT

The only pattern of over-representation seems to be in the summer months of June, July and August. This is probably not a significant enough factor to warrant any further study.

5 Large Truck Vehicle and Cargo Analysis

5.1 Analysis by Hazardous Cargo

Display 5.1a presents the number of crashes involving hazardous cargo over the five year period of reporting (2011-2015). The standard LT filter was applied to obtain those in the middle column. It was felt for hazardous materials that all vehicles be included and not just LTs. So an additional run was made that included all vehicles carrying hazardous materials; thus, some part of the entries in Display 5.1a could include hazardous materials carried by pickup trucks and other smaller vehicles. However, it was felt best that all hazardous cargo crashes be considered, due to the potential harm that can come from hazardous materials release. For future reference the following are the hazmat-related variables in Driver-Vehicle dataset: 217, 452, 453, and 456.

Hazardous Cargo	LT Involved in Crashes	All Vehicles Involved in Crashes
Gasoline	272	306
Other Flammable Liquid	202	301
Corrosive Material	76	205
Other	61	91
Oxidizer/Peroxide	24	27
Flammable Solids	22	29
Other Explosives	16	24
Poison	13	16
Radioactive Material	1	1
TOTAL	687	1,000

Display 5.1a Number of Vehicles Involved in Hazardous Cargo

Hazardous materials were recorded to have been release from 107 of the vehicles that were carrying hazardous materials. Display 5.1b gives the types of hazardous materials that were released.

File Dashb	oard <u>F</u> ilters	<u>A</u> nalysis <u>C</u> rosstal	o <u>T</u> ools <u>W</u> ind	low <u>H</u> elp			_ 8
\$	2011-2015 Alabam	na Integrated Driver-Ve	hicle D ∨	All records (do not ap	ply a filter)		→ → → → → → → → → → → → →
Suppress Zero Val	ues: None	✓ Select	Cells: 🔳 🔻 🌹				Column: E Hazardous Released ; Row: Hazardous Cargo 💽
	Yes	No	Unknown	Not Applicable/No Hazardous Cargo	Record from Paper System	TOTAL	
None	0	0	0	50061	1323	51384	
	0.00%	0.00%	0.00%	4.50%	1.82%	4.23%	_
Gasoline	32	261	1	12	0	306	
	29.91%	26.77%	0.00%	0.00%	0.00%	0.03%	_
Other Flammable			1	11	24	301	
Liquid	26.17%	24.31%	0.00%	0.00%	0.03%	0.02%	_
Flammable Solids	2	12	0	2	13	29	
	1.87%	1.23%	0.00%	0.00%	0.02%	0.00%	_
Oxidizer/Peroxide	4	21	0	0	2	27	
	3.74%	2.15%	0.00%	0.00%	0.00%	0.00%	_
Other Explosives	2	20	0	2	0	24	
	1.87%	2.05%	0.00%	0.00%	0.00%	0.00%	_
Poison	2	10	0	4	0	16	
	1.87%	1.03%	0.00%	0.00%	0.00%	0.00%	_
Radioactive Material	0	1	0	0	0	1	
Material	0.00%	0.10%	0.00%	0.00%	0.00%	0.00%	_
Corrosive Material	11	67	0	127	0	205	
	10.28%	6.87%	0.00%	0.01%	0.00%	0.02%	_
Other	4	64	2	9	12	91	
	3.74%	6.56%	0.01%	0.00%	0.02%	0.01%	_
Unknown	22 20.56%	282 28.92%	29781	118	22 0.03%	30225 2.49%	
Not Applicable			99.99%	0.01%			-
Not Applicable (Not a CMV)	0.00%	0.00%	0.00%	1061689 95.47%	71430 98.08%	1133119 93.20%	
(-
TOTAL	107 0.01%	975 0.08%	29785 2.45%	1112035 91.47%	72826 5.99%	1215728 100.00%	_
	0.01%	0.00%	Z.40 %	31.47 %	0.33%	100.00%	

Display 5.1b Type of Hazardous Cargo Released (All Vehicles)

5.2 Vehicle Defects

Display 5.2 presents the summary of the Contributing Vehicle Defects variable from the Driver-Vehicle dataset. No IMPACT was run because of the unique nature of a large number of these values; i.e., many of them do not occur in any non-LT vehicle. The defect values are ordered by the frequency to surface those that might be most critical.

	le Dashboard Filters Analysis Fr 2011-2015 Alabama Integrated Dr		Vindow Help Heavy Truck Involve	d C101 OR C501		- 1/ 1/2011 v 12/31/2015 v
					• A	
_	Frequency V Descending	Suppress Ze	ero-Valued Frequencies			
2	Contributing Vehicle Defect	Frequency 💌	Cum. Frequency	Percentage	Cum. Percent	D215: E Placard Required
	None	45898	45898	88.90	88.90	D216: E Placard Status D217: Hazardous Cargo
	Unknown	2787	50631	5.40	98.07	D217: Hazardous Cargo D218: E Hazardous Released
	Not Applicable	974	51605	1.89	99.96	D219: Attachment
	E Tire Blowout/Separation	577	46951	1.12	90.94	D220: Oversized Load Requiring Per
	Other	421	47844	0.82	92.67	D221: Had Oversized Load Permit
	Brakes	416	46314	0.81	89.71	D222: Contributing Vehicle Defect
	Wheels	136	47128	0.26	91.29	D223: Speed Limit
	Power Train	87	47305	0.17	91.63	D224: Estimated Speed at Impact D225: Citation Issued
	Trailer Hitch/Coupling	84	47218	0.16	91.46	D226: Vehicle Damage
	Steering	60	46374	0.12	89.83	D227: Vehicle Towed
	E Improper Tread Depth	41	46992	0.08	91.02	D230: Areas Damaged #1
	Suspension	29	47408	0.06	91.83	D231: E Areas Damaged #2
	E Tail Lights	25	47359	0.05	91.73	D232: E Areas Damaged #3 D233: Point of Initial Impact
	Fuel System	20	47335	0.03	91.67	D321: Driver Seating Position
			47379			D322: Driver Victim/Occ Type
	Tum Signal	20		0.04	91.77	D323: Driver Safety Equipment
	E Body/Doors	15	47423	0.03	91.86	D324: Driver Airbag Status
	P Tires*	11	51616	0.02	99.98	D325: Driver Age
	P Cargo	7	51626	0.01	100.00	D327: Driver Ejection Status
	E Headlights	5	47333	0.01	91.68	D328: Driver Injury Type D329: Driver First Aid By
	Windows/Windshield	3	47131	0.01	91.29	D320: Driver Transport Immediate
	E Mirrors	3	47134	0.01	91.30	D331: E Driver Transport Type
	P Restraint System	3	51619	0.01	99.99	D401: E Involved Road/Bridge
	Exhaust	2	47328	0.00	91.67	D402: E Road Surface Type
	E Wipers	0	47128	0.00	91.29	D403: Roadway Condition D404: E Environmental Contributing
	E Cruise Control	0	47408	0.00	91.83	D404: E Environmental Contributing D405: Contributing Material in Roady
	P Lights*	0	51616	0.00	99.98	<

Display 5.2 Frequency and Relative Frequency of Contributing Defects

5.3 Vehicle Attachment

Display 5.3 presents the summary of the Vehicle Attachments variable from the Driver-Vehicle dataset. No IMPACT was run because most of these values do not occur in any non-LT vehicle. The attachment values are ordered by the frequency to surface those that might be most critical.

	RE 10.1.0.7 - [Frequency Res			tegrated Drive	er-Vehicle Data	
¢.	2011-2015 Alabama Integrated Dr	iver-Vehicle D V	<u>V</u> indow <u>H</u> elp Heavy Truck Involv		¥ 9	_ <i>⊟</i> >
	Frequency Descending Attachment None E Other Semi Trailer E E Log Trailer Unknown E Large Utility (2+ Axles) Tanker P Semi Trailer* Other Not Applicable Double/Triple Trailer E Small Utility (1 Axle) Towed Vehicle Pole Trailer Mobile Home Boat Trailer E	Suppress 2 Frequency 27316 14675 14675 1862 1539 1517 1021 859 753 774 704 478 266 164 145 266 145 266 6 6	ro-Valued Frequencie Cum. Frequency 27316 43975 29300 49983 46236 47546 51546 48444 50687 44453 444719 46525 447691 27438 46309	s Percentage 52.91 28.43 3.61 2.98 2.94 1.98 1.66 1.46 1.36 0.93 0.52 0.32 0.28 0.24 0.13	Cum. Percent 52.91 85.18 56.75 96.82 89.56 92.10 99.85 93.84 98.18 86.11 86.62 90.12 92.38 53.15 89.70	D215: E Placard Required D216: E Placard Status D217: Hazardous Cargo D218: E Hazardous Released D219: Attachment D220: Oversized Load Requiring Permit D221: Had Oversized Load Permit D222: Contributing Vehicle Defect D223: Speed Limit D224: Estimated Speed at Impact D225: Citation Issued D226: Vehicle Damage D227: Vehicle Towed D230: Areas Damaged #1 D231: E Areas Damaged #2 D232: E Areas Damaged #3 D233: Point of Initial Impact D321: Driver Seating Position D322: Driver Victim/Occ Type D323: Driver Safety Equipment
	P Utility Trailer* Camper Trailer P 4 Wheel Trailer* E Steerable Front Axle	66 52 14 7	51612 46361 51626 46243	0.13 0.10 0.03 0.01	99.97 89.80 100.00 89.57	D324: Driver Airbag Status D325: Driver Age D327: Driver Ejection Status D328: Driver Inium Type
]]	, & Ø	· · ·				Display Average Display Filter Name

Display 5.3 Frequency and Relative Frequency of Vehicle Attachments

6 Bus Crash Analyses

The following indicates the sections and subjects of bus analyses that have been given above:

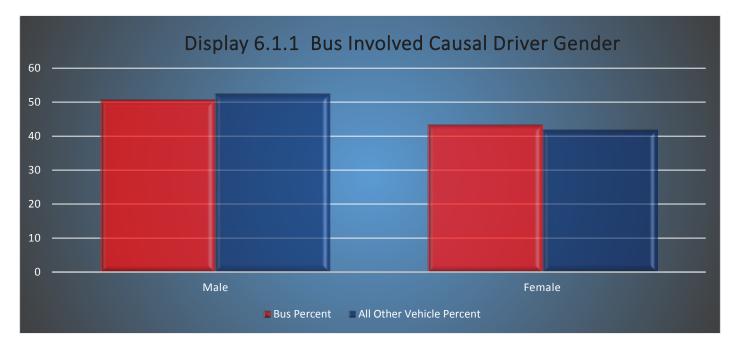
Section	Bus Subject Covered
1	Definition of the buses considered in this report
1.1.1	Bus involvement compared to other crash types
1.1.3	Severity of bus crashes
1.1.3.1	Analysis of bus impact speeds
1.1.3.2	Analysis of bus ambulance delay times
2.2	At fault analysis for buses

This section will continue by presenting bus crash causal driver demographics (6.1), general geographical and roadway bus crash analysis (6.2), and time factors for bus crashes (6.3).

6.1 Driver Demographics for Bus Involved Crashes

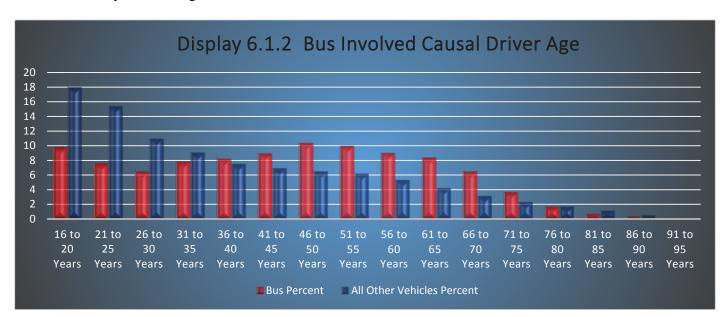
6.1.1 Causal Driver Gender for Bus Involved Crashes

Display 6.1.1 shows that bus causal driver gender is not significantly different from that of crashes in general. This amounts to about 50-52% male, 42-44% female, and 5.5-6.0% unknown.



6.1.2 Causal Driver Age for Bus Involved Crashes

Display 6.1.2 shows a distribution for buses that is much the same as for large trucks (see Section 3.1.2). The only notable difference is in the 16-20 and the 21-25 year olds that are not nearly as under-represented for buses as they are for large trucks.

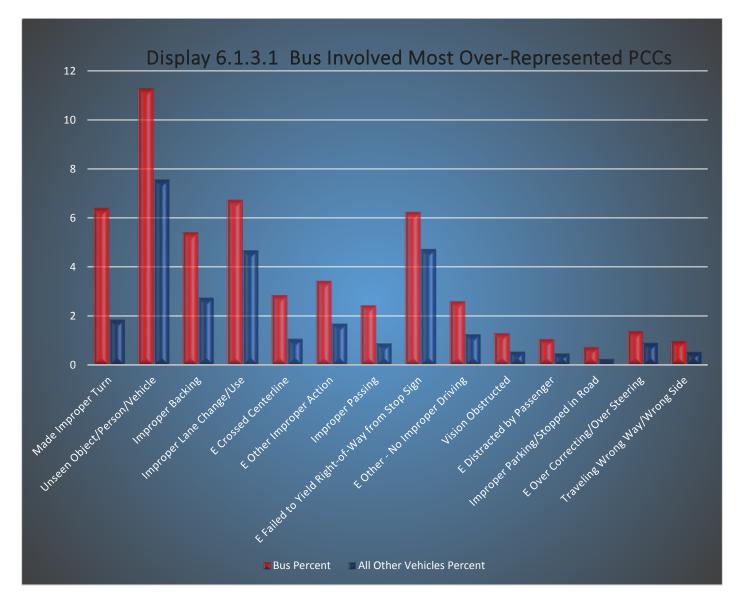


6.1.3 Driver Related Causes for Bus Involved Crashes

In this section we will consider causal driver Primary Contributing Circumstances (6.1.3.1), First Harmful Event (6.1.3.2), Manner of Crash (6.1.3.3), and Causal Unit Vehicle Maneuver (6.1.3.4)

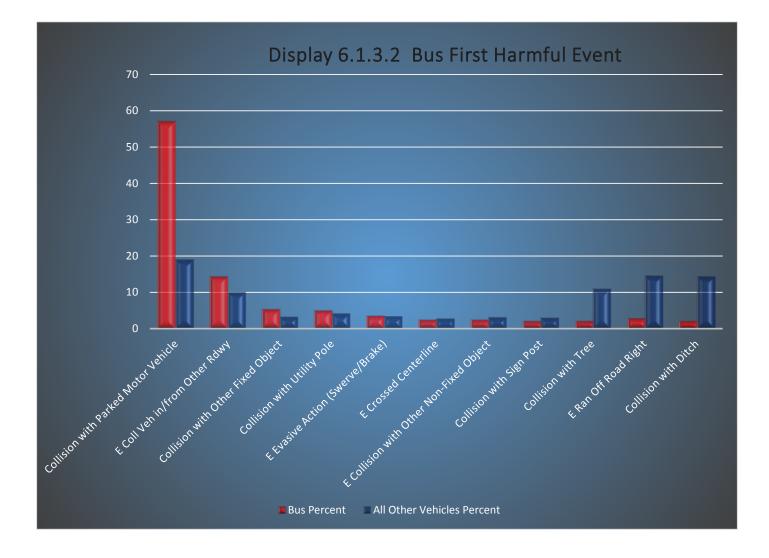
6.1.3.1 Primary Contributing Circumstances (PCCs) for Bus Involved Crashes

Generally the most over-represented categories are those associated with the control of a larger vehicle and the restrictions on the ability of the driver to see and be aware of hazards. This chart includes only the top 14 over-represented categories; it should be noticed that all categories in the chart are all over-represented. This covered all significantly over-represented categories for which there were at least 10 bus crashes.



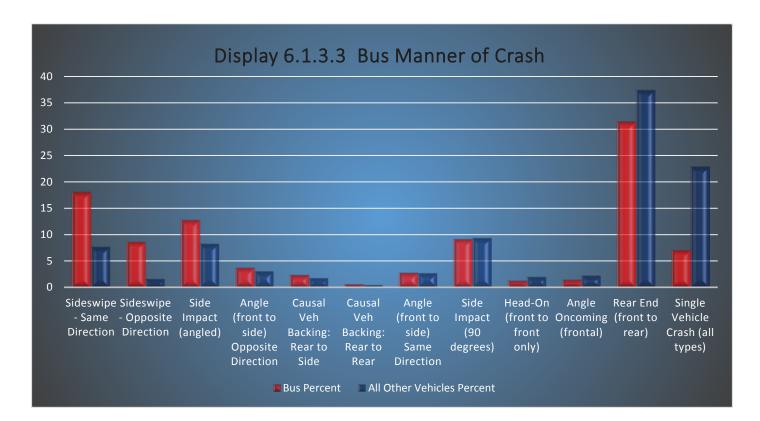
6.1.3.2 First Harmful Event for Bus Involved Crashes

The highest First Harmful Event for buses was Collision with Vehicle in Traffic, which amounted to 1035 (75.66% of the) bus crashes. This is significantly over-represented from the expected value from all other vehicles of 68.34%. Display 6.1.3.2 has this value (Collision with Vehicle in Traffic) removed, so it only represents a little over 24% of the bus crashes. Also, all categories with five or less bus crashes over the five year period were excluded. Of those remaining, the first two shown were the only two that were significantly over-represented for buses: Collision with Parked Motor Vehicle, and Collision with Vehicle in (or from) another Roadway.



6.1.3.3 Manner of Crash for Bus Involved Crashes

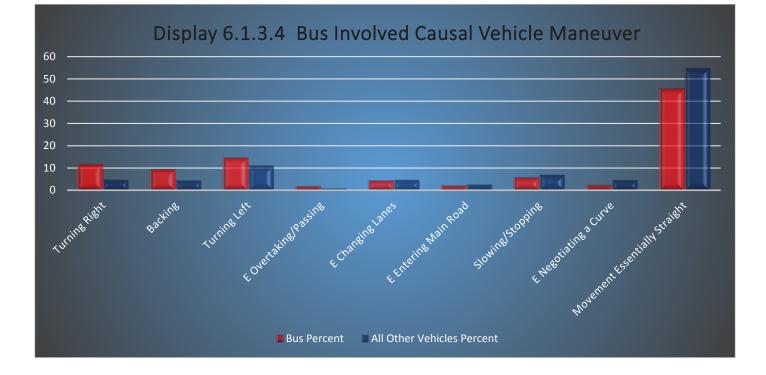
This appears to be one attribute that buses and large trucks have in common. They are both significantly over-represented in both sideswipe categories and in Side Impact (angled).



6.1.3.4 Causal Unit Vehicle Maneuver for Bus Involved Crashes

Display 6.1.3.4 shows four vehicle maneuvers where buses are significantly over-represented (frequency, over-representation factor):

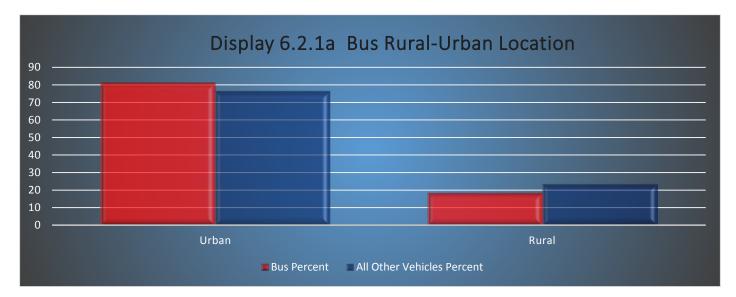
- Turning Right (151, 2.43);
- Backing (122, 2.08);
- Turning Left (188, 1.33); and
- Overtaking/Passing (26, 1.98).



6.2 Bus Crash Analysis

6.2.1 Bus Rural-Urban

The bus rural-urban location comparison given in Display 6.2.1a shows a considerably different pattern than that given for large trucks, which rural crashes were over-represented. In this case urban crashes are significantly over-represented when compared to crashes involving all other vehicle types.



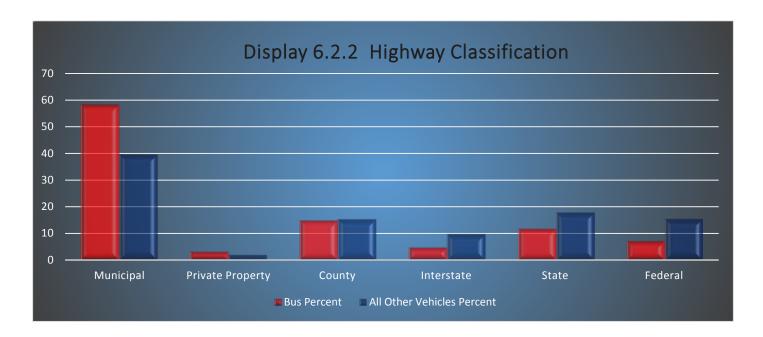
A severity comparison is given in Display 6.2.1b, which shows amazingly similar results as that obtained when considering bus crashes, as given in Display 4.1.1b. This is due to the lower speeds on urban road-ways, and the fact that buses have a much higher proportion of urban crashes than large trucks accounts for their generally lower fatality and severe injury rates.

Display 6.2.1b Crash Severity by Rural-Urban Classification – Buses

CARE 10.	1.0.7 - [Cross	tab Results -	2011-2015 A	Alabama Inte	grated Crash	Data - Filter	= Heavy Bus	s Inv — 🗖 🗙					
🚦 <u>F</u> ile <u>D</u> ash	board <u>Filters</u>	<u>A</u> nalysis <u>C</u> rossta	b <u>L</u> ocations <u>T</u>	ools <u>W</u> indow	<u>H</u> elp			_ & ×					
🎸 2011-2015 Alabama Integrated Crash Data 🗸 Heavy Bus Involved C101) OR C501 🗸 🖓 😨 1/ 1/2011 v 12/31/2015 v 🕽 🌖													
Suppress Zero Values: Rows and Column: V Select Cells: 💷 🖓 Column: Crash Severity ; Row: Rural or Urban 👰													
	Fatal Injury	Incapacitating Injury	Non- Incapacitating Inju	Possible Injury	Property Damage Only	Unknown	TOTAL						
Rural	3 37.50%	20 46.51%	23 32.39%	8 7.62%	202 18.36%	1 2.33%	257 18.76%	-					
Urban	5 62.50%	23 53.49%	48 67.61%	97 92.38%	898 81.64%	42 97.67%	1113 81.24%	-					
TOTAL	8 0.58%	43 3.14%	71 5.18%	105 7.66%	1100 80.29%	43 3.14%	1370 100.00%	-					
	_							-					

6.2.2 Bus Highway Classification

Display 6.2.2 shows a dramatic over-representation of crashes on municipal roadways. All of the differences are significant with the exception of county roads, which were essentially equal in proportion.

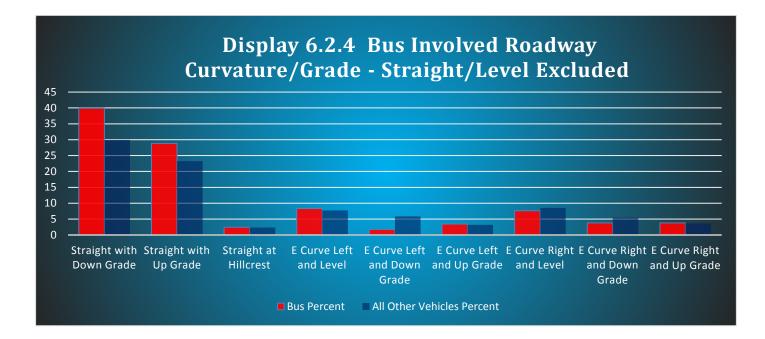


6.2.3 Bus Weather Analysis

As with large trucks, weather does not seem to be a major factor for bus crashes in comparison to non-bus crashes. Over the five year period, buses had only 89 crashes in the rain, which accounted for 6.51% as opposed to 11.11% for all other crashes, a statistically significant difference that indicates greater skill in bus drivers in dealing with inclement weather.

6.2.4 Bus Roadway Curvature and Grade

The vast majority of bus crashes occur on straight and level roadways, 78.69%, which is significantly higher than the 69.08% of all other vehicles. The major reason for this is that they tend to operate on more straight and level roadways than the traffic in general. Since this category would dwarf all of the others in comparison, the Straight and Level category was omitted from Display 6.2.4, which indicates that buses are over-represented in both the down-grade and the up-grade categories, with the Straight with Down Grade being statistically significant.



6.2.5 Bus Workzone Analysis

Only 20 of the 1,370 bus crashes were reported to have occurred in workzones. This is about 1.46% of bus crashes, as compared to 1.81% for all other vehicles. So buses are a slight bit under-represented in workzones, although this cannot be considered significant with such a small sample size. Safety considerations for buses in workzones should be about the same as all other vehicles.

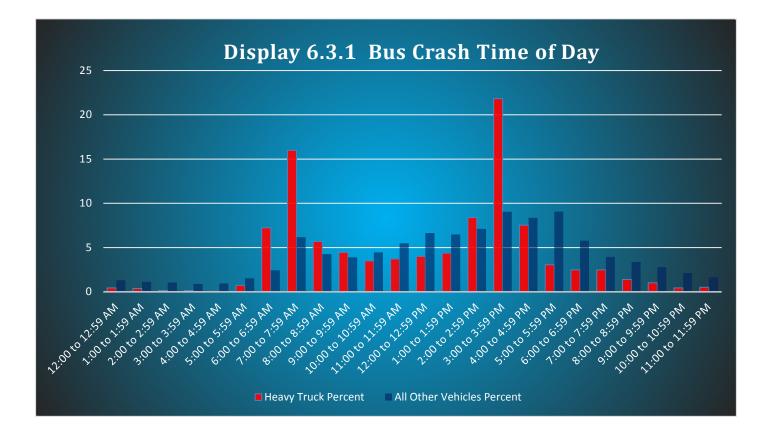
6.3 Bus Time Factors

6.3.1 Bus Time of Day

While the bus time of day distribution tends to follow the times when buses are most in operation, there are three hours when bus crashes are particularly over-represented that should be of note:

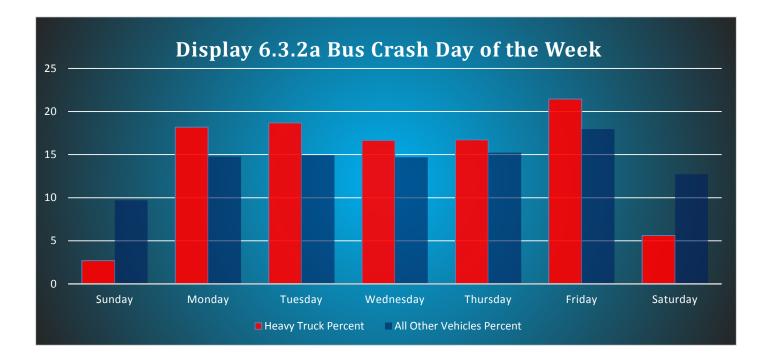
- 6:00-6:59 AM when buses have about three times their expected proportion of crashes;
- 7:00-7:59 AM which is the second highest in frequency (219 crashes over the five years); while this time slot has a lower over-representation ratio (2.59 as opposed to 3), its frequency (219) is over twice that of the previous hour; and
- 3:00-3:59 PM, the early afternoon rush hour, which has the highest proportion, with a frequency of 299 crashes and a over-representation ratio of over double (2.411).

The extremely large combined early morning and early afternoon rush hour over-representations may be a pattern that need to be investigated carefully, and it should be subject to further research. While no conclusive results can be reached without further analyses, it might be hypothesized that fatigue could be an issue with drivers coming to the end of a long period of driving and at the same time encountering heavy traffic. Other causes for this pattern should also be proposed and evaluated.



6.3.2 Bus Day of the Week

The distribution given in Display 6.3.2a probably reflects the days at which buses are in operation. It also reflects the density of other traffic as was certainly true of the time of day analysis, reflecting what seems to be a multiplier effect of traffic density when it comes to bus crash causation. In this case Monday, Tuesday and Friday's over-representations were statistically significant. These days (along with the two other week-days) would have much higher rush hour traffic than would Saturday and Sunday, so there is distinct correlation between the time of day and the day of the week.



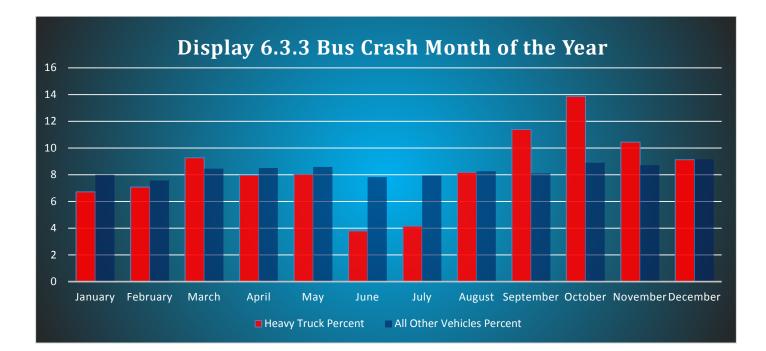
Display 6.3.2b shows the time of day by day of the week distribution for bus crashes. Note particularly the numbers on the 6:00-7:59 AM lines and those on the 3:00-3:59 PM. These show a stark contrast between what is happening during the week as opposed to the weekends.

CARE 10.1	.0.7 - [Cross	stab Results -	2011-2015	Alabama Inte	grated Crash	n Data - Filter	= Heavy Bus	s Involve –	
🖡 <u>F</u> ile <u>D</u> ashb		<u>A</u> nalysis <u>C</u> rosstał		ools <u>W</u> indow	 Help				_ 8 :
¢?	2011-2015 Alabar	ma Integrated Crash Da	ta 🗸	Heavy Bus Involved	I (C101 OR C501)	~	Se 1/ 1	1/2011 y 12/31/201	5 🗸 🚯 🍯
Suppress Zero Val	ues: None	✓ Select	Cells: 🔳 🔻					of the Week ; Row: T	ime of Day 🛛 🖓
Suppress Zero val					1	1		1	ime of Day
	Sunday	Monday	Tuesday	Wednesday	Thursday	Friday	Saturday	TOTAL	
12:00 Midnight to 12:59 AM	2 5.41%	1 0.40%	1 0.39%	0.00%	0.00%	0.00%	2 2.60%	6 0.44%	
1:00 AM to 1:59	2	1	0	0	0	1	1	5	
AM 2:00 AM to 2:59	5.41% 1	0.40%	0.00%	0.00%	0.00%	0.34%	1.30% 0	0.36%	
2.00 AM to 2.59 AM	2.70%	0.00%	0.00%	0.00%	0.00%	0.34%	0.00%	0.15%	
3:00 AM to 3:59 AM	1 2.70%	0	0 0.00%	0	0	0	1 20%	2 0.15%	
4:00 AM to 4:59	0	0.00%	0.00%	0.00%	0.00%	0.00%	1.30%	1	
AM	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	1.30%	0.07%	
5:00 AM to 5:59 AM	1 2.70%	3 1.20%	2 0.78%	2 0.88%	1 0.44%	0.00%	1 1.30%	10 0.73%	
6:00 AM to 6:59	0	20	22	20	18	18	1	99	
AM 7:00 AM to 7:59	0.00%	8.03% 41	8.59% 45	8.77% 45	7.86%	6.12% 47	1.30%	7.23%	
AM 67.55	2.70%	16.47%	17.58%	19.74%	17.03%	15.99%	1.30%	15.99%	
8:00 AM to 8:59 AM	2 5.41%	17	16	14	9	15	5	78	
9:00 AM to 9:59	5.41%	6.83% 11	6.25% 8	6.14% 11	3.93%	5.10% 13	6.49%	5.69% 61	
AM	13.51%	4.42%	3.13%	4.82%	3.93%	4.42%	5.19%	4.45%	
10:00 AM to 10:59 AM	2 5.41%	9 3.61%	10 3.91%	6 2.63%	8 3.49%	10 3.40%	3 3.90%	48 3.50%	
11:00 AM to 11:59	1	6	11	9	9	11	4	51	
AM 12:00 Near to	2.70%	2.41%	4.30% 10	3.95% 5	3.93%	3.74%	5.19% 4	3.72% 55	
12:00 Noon to 12:59 PM	5.41%	6.02%	3.91%	2.19%	2.18%	4.76%	4 5.19%	4.01%	
1:00 PM to 1:59 PM	4	6	12	4	9	20	5	60	
2:00 PM to 2:59	10.81%	2.41% 21	4.69% 27	1.75%	3.93% 27	6.80% 20	6.49% 1	4.38% 115	
PM	2.70%	8.43%	10.55%	7.89%	11.79%	6.80%	1.30%	8.39%	
3:00 PM to 3:59 PM	1 2.70%	60 24.10%	45 17.58%	58 25.44%	46 20.09%	83 28.23%	6 7.79%	299 21.82%	
4:00 PM to 4:59	3	24.10%	21	16	19	18	5	103	
PM	8.11%	8.43%	8.20%	7.02%	8.30%	6.12%	6.49%	7.52%	
5:00 PM to 5:59 PM	0.00%	5 2.01%	9 3.52%	6 2.63%	9 3.93%	7 2.38%	6 7.79%	42 3.07%	
6:00 PM to 6:59 PM	3	2	5	5	10	3	6	34	
7:00 PM to 7:59	8.11% 2	0.80%	1.95% 7	2.19%	4.37% 7	1.02%	7.79%	2.48%	
PM	5.41%	1.20%	2.73%	1.75%	3.06%	1.02%	10.39%	2.48%	
8:00 PM to 8:59 PM	2 5.41%	6 2.41%	4 1.56%	0.00%	1 0.44%	4 1.36%	2 2.60%	19 1.39%	
9:00 PM to 9:59	1	0	1.00%	4	1	4	3	1.35%	
PM	2.70%	0.00%	0.39%	1.75%	0.44%	1.36%	3.90%	1.02%	
10:00 PM to 10:59 PM	0.00%	1 0.40%	0	0.00%	0.44%	2 0.68%	2 2.60%	6 0.44%	
11:00 PM to 11:59	0	0	0	1	1	0	5	7	
PM	0.00%	0.00%	0.00%	0.44%	0.44%	0.00%	6.49% 0	0.51%	
Unknown	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	
TOTAL	37 2.70%	249 18.18%	256 18.69%	228 16.64%	229 16.72%	294 21.46%	77 5.62%	1370 100.00%	

Display 6.3.2b Bus Crash Time of Day by Day of the Week

6.3.3 Bus Month of the Year

Display 6.3.3 shows that the best months over the past five years for bus crashes has been June and July. The two worst months are September and October. Additional analyses would be warranted to compare these two time periods to determine what may be the causes for these disparities.



7 County Summaries for 2014 and 2015

7.1 Large Trucks and Buses by County

Display 7.1 Large	Truck and Bus	Crashes by Severit	y for 2014 and 2015

	Tot	al Large	Truck a	nd Bus	Crashes	;		La	arge Tru	ck Crasl	nes		Bus Crashes						
NUMBER (PER	SONS	PER	SONS	NUM	BER OF	PER	SONS	PER	SONS	NUM	BER OF	PERSONS		PER	SONS	
NONDER	JF CRAS		KIL	LED	INJ	JRED	CRA	SHES	KIL	LED	INJU	JRED	CRA	SHES	KIL	LED	INJU	JRED	
COUNTY	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	
Autauga	55	62	3	1	18	20	50	53	3	1	16	18	5	9	0	0	2	2	
Baldwin	173	148	2	1	37	43	165	137	2	1	35	38	8	11	0	0	2	5	
Barbour	31	36	2	0	10	10	27	36	2	0	10	10	4	0	0	0	0	0	
Bibb	20	22	3	2	9	12	16	20	2	2	8	12	4	2	1	0	1	0	
Blount	35	38	0	0	13	12	34	38	0	0	13	12	1	0	0	0	0	0	
Bullock	12	10	0	0	2	5	12	10	0	0	2	5	0	0	0	0	0	0	
Butler	59	70	2	0	16	18	58	69	2	0	16	18	1	1	0	0	0	0	
Calhoun	137	122	0	0	27	40	129	109	0	0	27	31	8	13	0	0	0	9	
Chambers	35	40	0	0	12	10	34	36	0	0	12	6	1	4	0	0	0	4	
Cherokee	30	35	1	2	11	9	30	33	1	2	11	9	0	2	0	0	0	0	
Chilton	62	61	1	1	13	15	60	60	1	1	10	15	2	1	0	0	3	0	
Choctaw	20	25	0	0	10	9	19	25	0	0	10	9	1	0	0	0	0	0	
Clarke	55	50	1	1	20	19	50	45	1	1	19	16	5	5	0	0	1	3	
Clay	15	14	1	0	8	8	15	13	1	0	8	6	0	1	0	0	0	2	
Cleburne	93	74	0	1	32	20	92	74	0	1	29	20	1	0	0	0	3	0	
Coffee	37	56	0	0	12	16	33	49	0	0	12	15	4	7	0	0	0	1	
Colbert	55	83	0	2	16	25	53	80	0	2	14	25	2	3	0	0	2	0	
Conecuh	55	62	0	2	14	19	53	62	0	2	14	19	2	0	0	0	0	0	
Coosa	22	19	3	0	10	2	20	19	3	0	10	2	2	0	0	0	0	0	
Covington	22	24	0	1	2	15	22	23	0	0	2	6	0	1	0	1	0	9	
Crenshaw	13	13	0	0	4	4	13	12	0	0	4	4	0	1	0	0	0	0	
Cullman	104	141	2	3	15	23	104	139	2	3	15	23	0	2	0	0	0	0	
Dale	39	52	0	1	13	13	36	51	0	1	13	13	3	1	0	0	0	0	
Dallas	61	58	1	3	22	23	57	51	1	3	22	23	4	7	0	0	0	0	
Dekalb	70	62	1	2	20	17	69	60	1	2	20	16	1	2	0	0	0	1	
Elmore	51	50	0	0	13	16	45	40	0	0	12	16	6	10	0	0	1	0	
Escambia	57	64	0	1	31	29	54	60	0	1	17	28	3	4	0	0	14	1	
Etowah	123	125	5	3	37	28	118	119	5	3	35	28	5	6	0	0	2	0	
Fayette	14	16	1	0	5	3	12	16	1	0	5	3	2	0	0	0	0	0	
Franklin	25	28	0	0	13	9	24	26	0	0	4	9	1	2	0	0	9	0	
Geneva	17	15	0	1	9	1	17	15	0	1	9	1	0	0	0	0	0	0	
Greene	41	65	1	0	10	13	38	65	1	0	10	13	3	0	0	0	0	0	
Hale	19	21	0	0	4	11	18	20	0	0	4	10	1	1	0	0	0	1	

	Tot	al Large	Truck a	nd Bus	Crashes	;		La	rge Tru	ck Crasł	nes		Bus Crashes						
NUMBER C	DF CRAS	HES		SONS		SONS		BER OF		SONS		SONS		BER OF		SONS		SONS	
				LED		JRED		SHES		LED		JRED		SHES		LED		JRED	
COUNTY	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	2014	2015	
Henry	19	27	0	1	3	6	18	27	0	1	3	6	1	0	0	0	0	0	
Houston	130	127	3	0	26	22	121	115	3	0	25	22	9	12	0	0	1	0	
Jackson	53	75	1	2	16	29	50	70	1	2	16	29	3	5	0	0	0	0	
Jefferson	1,093	1,260	6	9	217	254	981	1,137	4	9	181	216	112	123	2	0	36	38	
Lamar	15	18	0	0	6	7	15	18	0	0	6	7	0	0	0	0	0	0	
Lauderdale	46	41	0	2	7	15	40	37	0	2	7	14	6	4	0	0	0	1	
Lawrence	40	47	3	2	9	34	39	43	3	2	9	17	1	4	0	0	0	17	
Lee	196	191	0	0	72	36	183	179	0	0	68	35	13	12	0	0	4	1	
Limestone	79	91	0	0	35	37	73	87	0	0	33	29	6	4	0	0	2	8	
Lowndes	27	29	0	0	9	8	27	28	0	0	9	8	0	1	0	0	0	0	
Macon	75	64	4	1	23	23	73	62	4	0	23	23	2	2	0	1	0	0	
Madison	205	225	5	2	81	43	180	190	5	2	55	40	25	35	0	0	26	3	
Marengo	28	38	0	1	16	26	28	36	0	1	16	15	0	2	0	0	0	11	
Marion	34	51	0	1	37	8	33	48	0	1	10	5	1	3	0	0	27	3	
Marshall	69	83	0	0	27	33	64	78	0	0	21	33	5	5	0	0	6	0	
Mobile	471	494	2	4	117	127	420	424	2	4	104	110	51	70	0	0	13	17	
Monroe	25	17	0	2	9	8	24	17	0	2	8	8	1	0	0	0	1	0	
Montgomery	369	363	2	8	102	132	335	328	2	8	86	114	34	35	0	0	16	18	
Morgan	134	124	1	0	23	37	125	122	1	0	22	34	9	2	0	0	1	3	
Perry	4	12	0	1	4	5	4	12	0	1	4	5	0	0	0	0	0	0	
Pickens	32	28	1	2	14	21	31	25	1	1	14	6	1	3	0	1	0	15	
Pike	51	64	1	0	9	18	46	63	1	0	7	18	5	1	0	0	2	0	
Randolph	18	14	0	0	6	4	16	13	0	0	6	4	2	1	0	0	0	0	
Russell	90	119	1	1	51	58	81	108	1	1	24	53	9	11	0	0	27	5	
Shel by	179	228	1	3	37	43	169	212	1	3	34	41	10	16	0	0	3	2	
St Clair	116	124	0	2	27	42	110	121	0	2	18	42	6	3	0	0	9	0	
Sumter	68	57	1	3	41	23	66	57	1	3	41	23	2	0	0	0	0	0	
Talladega	107	104	3	5	34	38	102	100	3	5	32	37	5	4	0	0	2	1	
Tallapoosa	24	7	1	0	4	0	21	7	1	0	4	0	3	0	0	0	0	0	
Tuscaloosa	401	429	3	5	118	119	366	392	3	5	108	90	35	37	0	0	10	29	
Walker	77	61	1	2	29	17	76	56	1	2	26	17	1	5	0	0	3	0	
Washington	19	16	0	1	6	9	19	16	0	1	6	9	0	0	0	0	0	0	
Wilcox	17	17	1	0	7	15	17	16	1	0	7	8	0	1	0	0	0	7	
Winston	19	31	0	2	7	8	18	29	0	2	6	8	1	2	0	0	1	0	

Display 7.1 Large Truck and Bus Crashes by Severity for 2014 and 2015 (continued)