



# Conditional Probability of Release (CPR) Estimates for Railroad Tank Cars in Accidents



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## 12. ABSTRACT

This report presents estimates and formulas for conditional probabilities of release (CPR) for all common tank car configurations and components in service today, given that the tank car is involved in an FRA-reportable accident, as well as distributions of the quantity of lading lost in the event of a release. The study was funded by the industry partners of the Advanced Tank Car Collaborative Research Program, using data provided by the RSI-AAR Railroad Tank Car Safety Research and Test Project. The study updates and expands upon an earlier Tank Car Safety Project report: RA-05-02, "Safety Performance of Tank Cars in Accidents: Probabilities of Lading Loss", published in 2006.

The purposes of revising the RA-05-02 CPR estimates were to take advantage of data from more recent accidents, and to incorporate factors in the modeling of tank car performance that were not addressed by the earlier report. The new estimates are more reflective of current and future tank cars and accident environments.

CPR estimates for the more recent time period are generally lower than the RA-05-02 estimates for any given car. Some of the newly analyzed variables were revealed to be significant.

## 13. SUBJECT TERMS

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## Conditional Probability of Release (CPR) Estimates for Railroad Tank Cars in Accidents

### Contents

	Page
Executive Summary.....	vi
1. Objectives.....	1
2. Approach.....	2
a) Definition of “conditional probability of release”	2
b) Analytical team	2
c) Data used in the study	2
d) Identification of variables for regression	4
1) Accident-related independent variables	4
2) Car-related independent variables	4
3) Variables considered but not included	6
4) Definition of estimated effective impact speed	7
5) Definition of impact opportunity factor	8
6) Advanced variable selection procedure	9
e) Regression approach	9
f) CPR for one-component losses	11
g) FRA multiplier	12
h) CPRs adjusted for minimum quantity lost	12
i) Quantity lost	13
j) Expected Quantity of Release (EQR)	13
3. Results.....	15
a) CPR formulas	15
1) Shell CPR formula	15
2) Head CPR formula	15
3) Top fittings CPR formula	16
4) Bottom fittings CPR formula	16
b) CPRs for common tank cars and other cars of interest	18
c) Tank thickness effects	19
d) Tank steel type effects	20
e) Top fittings protection effects	21
f) Jacket effects	22
g) Head protection effects	23
h) Speed effects	24
i) Yard/industrial CPRs	27
j) Quantity lost results	27
k) Expected Quantity Released (EQR) Example	33

4. References..... 34

Appendix A: Tank Car Safety Project Collection of Data on Tank Cars Damaged in Accidents

Appendix B: gMCP (Group Minimax Concave Penalty approach)

Appendix C: Final Model Output, Model Fit and Diagnostics

Appendix D: Uncertainty Surrounding CPR Estimates

Appendix E: Calculation of the FRA Accident Data Compatibility Multiplier

Appendix F: Comparison of TWP-17 and RA-05-02 CPR Estimates

Tables

	Page
1. Number of Tank Car Records Used in Each Component Regression, and the Number of Accidents in which Those Cars Were Damaged	3
2. CPRs and CPR <sub>&gt;100s</sub> for Selected Tank Cars Under Average Mainline Derailment Conditions	19
3. CPRs and CPR <sub>&gt;100s</sub> for Selected Groups of Comparable Tank Cars with Different Tank Thicknesses, Under Average Mainline Derailment Conditions	20
4. CPRs and CPR <sub>&gt;100s</sub> for Selected Pairs of Comparable Tank Cars with Different Tank Steel Types, Under Average Mainline Derailment Conditions	21
5. CPRs and CPR <sub>&gt;100s</sub> for Comparable Tank Cars with Different Top Fittings Protection Levels, Under Average Mainline Derailment Conditions	22
6. CPRs and CPR <sub>&gt;100s</sub> for Selected Pairs of Comparable Tank Cars with and without Standard 11-Gauge Jackets, Under Average Mainline Derailment Conditions	23
7. CPRs and CPR <sub>&gt;100s</sub> for Selected Groups of Comparable Tank Cars with Different Levels of Head Protection, Under Average Mainline Derailment Conditions	23
8. CPRs and CPR <sub>&gt;100s</sub> for Selected Cars Derailed at Different Speeds, Under Otherwise Average Mainline Derailment Conditions	24
9. CPRs and CPR <sub>&gt;100s</sub> for Selected Cars Derailed on Different Track Types, Under Average Mainline Derailment Conditions	27
10. Number of Tank Cars that Lost Lading, by Percent of Tank Capacity	32

12. EQRs, CPRs and CPR <sub>&gt;100s</sub> for 30,000-Gallon DOT-111 Tank Cars with Different Risk Reduction Options (RROs), (7/16" Tanks, 119" Shell Inside Diameter), Under Average Mainline Derailment Conditions	33
13. CPR Estimates and Associated Confidence Intervals for Five Tank Car Configurations	D-2
14. TWP-17 CPRs and RA-05-02 CPRs for Selected Tank Cars, Under Average Mainline Derailment Conditions	F-1

Figures

	Page
1. Effective Impact Speed (EIS) Curve for a 10-Car Derailment at 40 mph	8
2. Impact Opportunity Factor Curve for a 15-Car Derailment	9
3. Four Components of a Tank Car That Can Lose Lading	10
4. Relationship Between Tank Thickness and CPR, Illustrated for Jacketed and Non-Jacketed 111A100W1 Tank Cars, Under Average Mainline Derailment Conditions	20
5. Non-Jacketed 111A100W1 Mainline CPR vs. Train Speed as Predicted by TWP-17 Formulas, and Empirical Counts	25
6. Jacketed 111A100W1 Mainline CPR vs. Train Speed as Predicted by TWP-17 Formulas, and Empirical Counts	26
7. Average Quantity of Lading Lost for Each Car Component, Expressed as a Percentage of Car Capacity	28
8. Distribution of Percentage of Car Capacity Lost for Each Car Component	30
9. CPRs for Five Tank Cars, with 95% Confidence Intervals	D-2

## 1. Executive Summary

This report presents estimates and formulas for conditional probabilities of release (CPR) for most common tank car configurations and components in service today, given that the tank car is involved in an FRA-reportable accident, as well as distributions of the quantity of lading lost in the event of a release. The study was funded by the industry partners of the Advanced Tank Car Collaborative Research Program (The American Chemistry Council, The Association of American Railroads, The Chlorine Institute, The Fertilizer Institute, and The Railway Supply Institute), under the ATCCRP designation “TWP-17”. The data used in the analysis were provided by the RSI-AAR Railroad Tank Car Safety Research and Test Project (the Tank Car Safety Project). The study updates and expands upon an earlier Tank Car Safety Project report: RA-05-02, “Safety Performance of Tank Cars in Accidents: Probabilities of Lading Loss”, published in 2006.

The purposes of revising the RA-05-02 CPR estimates were to take advantage of data from more recent accidents, and to incorporate factors in the modeling of tank car performance that were not addressed by the analysis in RA-05-02. The new estimates are more reflective of current and future tank cars and accident environments. Some representative CPR estimates for average derailment conditions are shown in Table E1 (mainline/siding accidents).



Table E1  
CPRs and  $CPR_{>100}$ s for Selected Tank Cars  
Under Average Mainline Derailment Conditions<sup>1</sup>

Car Specification	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	$CPR_{>100}$ <sup>2</sup>
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200 <sup>3</sup>
111A100W1/3	0.4375	A516	Yes	No	No	Yes	119	0.134	0.089
111A100W1/2	0.4375	A516	No	No	No	No	119	0.265	0.194
111A100W1/2/3	0.4375	A516	Yes	No	No	No	119	0.130	0.086
111A100W2	0.5625	A516	No	No	Yes	No	100.625	0.201	0.167
111A100W1	0.4375	TC128B	Yes	Full	Yes	Yes	119	0.064	0.046
111A/S100W1	0.500	TC128B	No	Half	Yes	Yes	119	0.132	0.103
117R100W	0.4375	A516	Yes	Full	No <sup>4</sup>	Yes	119	0.126	0.081
117R100W	0.4375	TC128B	Yes	Full	Yes	Yes	119	0.064	0.046
117R100W	0.500	A516	Yes	Full	No <sup>4</sup>	Yes	119	0.106	0.067
117R100W	0.500	TC128B	Yes	Full	Yes	Yes	119	0.052	0.037
117J100W	0.5625	TC128B	Yes	Full	Yes	Yes	119	0.042	0.029
112J340W	0.625	TC128B	Yes	Full	Yes	No	118.75	0.033	0.023
105J300W	0.562	TC128B	Yes	Full	Yes	No	118.75	0.040	0.028
105A500W	0.779	TC128B	Yes	No	Yes	No	102	0.040	0.030
105J500W	0.797	TC128B	Yes	Full	Yes	No	102	0.031	0.022
112J500I	H1.03 / S0.89	TC128B	Yes	Full	Yes	No	115.34	0.017	0.011
105J600I	H1.136/S0.98	TC128B	Yes	Full	Yes	No	106	0.016	0.011

Criteria for the inclusion of a tank car in the study's dataset include:

- Accident occurred 1980 through 2011, the latest year for which complete data were available at the time of the analysis
- Car was built 1970 or later
- Loaded cars only
- Stub-sill cars with shelf couplers only

<sup>1</sup> Assuming average conditions for a freight car in an FRA-reportable mainline accident: 29 mph train speed at the moment of derailment, 11 cars derailed, and the tank car of interest is the 6<sup>th</sup> derailed car. These averages are derived from FRA mainline and siding freight train accidents for the period 2003-2012. Note that  $CPR_{>100}$  is defined in section 2(h) of the report.

<sup>2</sup> CPR adjusted to exclude releases of 100 gallons or less from an individual car. See Section 2(h).

<sup>3</sup> Preliminary analyses used  $CPR = 0.266$  and  $CPR_{>100} = 0.196$  for non-jacketed DOT/TC-111 cars. Subsequently an adjustment was made to the tank steel assumption that resulted in a slight increase in the CPRs.

<sup>4</sup> Specifications for 117R100W require top fittings protection. However, while there are many different types and levels of top fittings protection today, the data used for this study did not allow those distinctions to be made quantitatively. Top fittings protection was either an impact-resistant protective housing, or none. Some current systems probably perform somewhere in between. Therefore, some examples using "No" for this field are presented, to suggest the range of performance that the real systems probably fall within.

- Tank car classes DOT/TC-111, 211, 105, 112, 114, 120 only
- Tank steel specifications TC128B, A515, A516 only
- Truck capacity 100 tons or more and 4 axles only
- Damaged by impact (as opposed to strictly by fire exposure).

This study focuses strictly on the effects of immediate impacts during a derailment. Subsequent effects of fire exposure, for example, are not addressed.

## Conditional Probability of Release (CPR) Estimates For Railroad Tank Cars in Accidents

### 1. Objectives

This report presents estimates and formulas for conditional probabilities of release (CPR) for most common tank car configurations and components in service today, given that (i.e., under the condition that) the tank car is involved in an FRA-reportable accident, as well as distributions of the quantity of lading lost in the event of a release. The study was funded by the industry partners of the Advanced Tank Car Collaborative Research Program (ATCCRP) (i.e., The American Chemistry Council, The Association of American Railroads, The Chlorine Institute, The Fertilizer Institute, and The Railway Supply Institute), under the ATCCRP designation “TWP-17” (Technical White Paper 17). The information on car features, accident details, tank car damage and lading loss that were used in the analysis was drawn from data that have been compiled since 1970 by the RSI-AAR Railroad Tank Car Safety Research and Test Project (the “Tank Car Safety Project”). This study updates and expands upon an earlier Tank Car Safety Project report: RA-05-02, “Safety Performance of Tank Cars in Accidents: Probabilities of Lading Loss” [Ref. 4], published in 2006.

The purposes of revising the RA-05-02 CPR estimates were to take advantage of data from more recent accidents, and to incorporate factors in the modeling of tank car performance that were not addressed by the analysis in RA-05-02.

The new estimates are more reflective of current and future tank cars and accident environments. The RA-05-02 dataset included accidents that occurred in the period 1965-1997. The TWP-17 analysis uses a more up-to-date dataset that extends through 2011<sup>5</sup> and excludes older data that are considered less relevant to current assessments of tank car safety performance.

Also, during the discussions of an Association of American Railroads (AAR) Tank Car Committee task force<sup>6</sup> and other activities, certain variables were identified as possibly affecting tank car accident performance that had not been explored in RA-05-02, such as tank inside diameter, jacket standoff distance, the presence of heater coils, new variables that measure the severity of the accident environment, and others. Additionally, some variables that were evaluated in RA-05-02 had not been incorporated directly into the regression analyses that produced the CPR estimates then, but instead were addressed by *ad hoc* side analyses. In TWP-17, they were included directly as independent variables in the regressions, yielding a more robust evaluation of the effects of those factors. The most prominent example of this is train speed at the time of derailment.

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<sup>5</sup> Although much of the crude oil fleet was built after this date, the probability estimates for these cars are valid, because all of the features on these cars had been present in the fleet for many years and are well-represented in the dataset. More discussion of this issue can be found in section 2(e).

<sup>6</sup> AAR Tank Car Committee Docket T87.5, “Consider Head and Shell Requirements for Non-Pressure Cars Transporting Packing Group I and II Materials”.

Finally, the identification of significant variables and the determination of coefficients for those variables were accomplished in TWP-17 using more sophisticated statistical approaches than had been possible in RA-05-02. These techniques allow more robust coefficient estimation in cases where the dataset is complex.

Characterization or prediction of a single accident is not an objective of CPR analysis. CPR estimates predict the average behavior of a given car configuration over a group of accidents with similar conditions, and therefore can be used to compare predicted performance among car options. There are numerous factors in any individual accident that can cause a car's performance to diverge from the average performance over all circumstances. Appropriate uses of CPR estimates include incorporating quantitative performance data into risk assessments and supporting the benefits calculations in cost-benefit analyses aimed at selecting among a variety of packaging options.

This study presents the best understanding of statistical tank car accident performance, with respect to impact resistance, that could be developed when it was conducted. CPR estimates, as with any statistically derived estimates, are potentially subject to change over time, as accident environments evolve along with equipment and operating practices, as more and newer data come in, and as research continues. The intent is to present CPR estimates that will be useful tools now and moving forward, until updates or refinements are desired.

## **2. Approach**

### **a) Definition of “conditional probability of release”**

A conditional probability is the probability of a specific event occurring, given that a specific condition is true. For the purpose of this report, the conditional probability of release (CPR) for a tank car is the probability that that single tank car releases any quantity<sup>7</sup> of lading, given that it is derailed in an FRA-reportable<sup>8</sup> accident.

### **b) Analytic team**

The TWP-17 analysis was conducted by a team consisting of experts on tank car safety data, statistics, and structural reliability at the University of Illinois at Urbana-Champaign, led by Dr. Christopher P.L. Barkan, Dr. M. Rapik Saat, and Laura Ghosh, along with Dr. Steve Kirkpatrick of Applied Research Associates and Todd Treichel of the Tank Car Safety Project.

### **c) Data used in the study**

The source of the data used in this study was the Tank Car Safety Project's database of tank cars damaged in accidents. The database contains records on tank cars that were damaged in

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<sup>7</sup> A modification wherein only releases above a certain quantity are considered will also be discussed.

<sup>8</sup> The Federal Railroad Administration requires that railroads report any accident that causes damage to track, equipment and/or structures exceeding a certain dollar threshold that is adjusted each year.

derailments and collisions, but only those cars for which the damage was incurred on tank-car-specific features, including the tank, fittings, jacket, head shields, and so on. For more information on this database, see Appendix A. Some additional data sources that were drawn upon for specific purposes outside of the main analytical calculations are described in Appendix E.

Cars in the Tank Car Safety Project data were included in the study dataset if they met the following criteria:

- Accident occurred 1980 through 2011, the latest year for which complete data were available at the time of the analysis
- Car was built 1970 or later
- Loaded cars only
- Cars with shelf couplers only
- Stub-sill cars only
- Tank car classes DOT/TC-111, -105, -112, -114, -120, AAR-211 only
- Tank steel specifications TC128B, A515, A516 only
- Truck capacity 100 tons or more, and 4 axles only
- All key fields had known values

Cars that were damaged only by fire and not by impact, or that released lading only because of fire exposure, were excluded. This study focuses strictly on the effects of immediate impacts during a derailment.

These criteria are essentially the same as those used in the RA-05-02 study, apart from the time frame.

As explained below, four analyses were conducted, one for each component of the car that can lose lading in an accident. The number of records used in each of the analyses, based on the criteria given above, were as shown in Table 1:

Table 1  
The Number of Tank Car Records Used in Each Component Regression  
And the Number of Accidents in which Those Cars Were Damaged

Component	Tank Cars	Accidents
Shell	7,165	4,993
Head	4,467	2,464
Bottom Fittings	5,484	3,905
Top Fittings	4,467	2,175

The differences in sizes among these datasets are due to unknowns in data fields that eliminated the record from one analysis but not the others. There are fewer accidents than cars because some accidents provided more than one car to the analyses.

d) Identification of variables for regression

The initial set of independent variables (i.e., potential factors that could affect CPR), as reflected in the four full models, was as follows:

1) Accident-related independent variables

- Track type (main & siding, or yard & industrial)
- Accident type (derailment vs. collision)
- Train speed (mph)
- Train speed squared (mph<sup>2</sup>)
- Estimated effective impact speed (mph)
- Estimated effective impact speed squared (mph<sup>2</sup>)
- Impact opportunity factor (unit-less score)
- Number of freight cars derailed
- Date of accident
- Various interactions among these variables

Notes on accident-related variables:

- Mainline and siding accidents were grouped because track and operational conditions are generally similar between the two track types. Similarly, yard accidents and industrial track accidents were grouped.
- Effective impact speed and impact opportunity factor are variables specifically developed for this study, and defined below. These and the number of freight cars derailed are intended as proxies for the amount of impact energy available for damaging a specific tank car in the accident. Although these proxies are not precise models for the behaviors they represent, they help the regression calculation to attribute different outcomes for otherwise apparently similar cars to varying amounts of available impact energy, rather than forcing these differences to be attributed to variables that are not actually driving them, i.e., they allow the regression to do a better job of evaluating the car characteristics of the most interest.
- Date of accident was included to capture any trends or changes in the accident environment over time that were not captured by other variables.
- The specific interactions included can be seen in the output from the statistical software package, provided in Appendix C.

2) Car-related independent variables

- Shell thickness (inches)
- Shell material type (steel specification TC128B, A516, or A515)
- Shell inside diameter (inches)
- Head thickness (inches)

- Head material type (steel specification TC128B, A516, or A515)
- Head shield presence/type (none, half-height, full-height)
- Jacket (yes/no)
- Insulation thickness (jacket standoff; inches)
- Tank capacity (gallons)
- External heater coils (yes/no)
- Top fittings type (pressure-car-style protective housing or not)
- Bottom fittings (equipped yes/no)
- Lading group (as a proxy for internal pressure)
- Year car built
- Car age at time of derailment
- Interactions among these variables

Notes on car-related variables:

- There were 51 records for cars that were known to have had trapezoidal head shields, which was not enough to segregate their performance from that of half-height head shields that are attached to the tank or integral to the jacket.
- Insulation thickness includes a thermal protection blanket if present.
- The only two top fittings configurations with sufficient data for the analysis were non-pressure car fittings with no derailment damage prevention system, and fittings enclosed in a pressure-car-style protective housing per 49 CFR 179.100-12. Although there are now numerous other definitions of top fittings protection, the results for the CPR performance of top fittings protection in this study are the best available for applying to all of these systems at this time.
- Lading groups were created by grouping ladings with similarity in
  - a) internal pressure in the tank, because of previous impact simulations that indicated that this factor could be significant [Ref. 3], and in
  - b) temperature in transportation, in case this might affect the performance of the steel.

However, because the results were internally inconsistent and destabilized the CPR estimates, the lading groups were dropped from the final analysis.

- Bottom fittings protection is only addressed in terms of setting the CPR for bottom fittings to zero when they are absent from a car. The vast majority of cars with bottom fittings have AAR Level A protection, so there is little need to evaluate the effects of different levels, and little data to support an analysis of the uncommon B and C levels. New bottom outlet handle securement approaches have not been in the fleet long enough for there to be sufficient data to analyze those, either.

- The specific interactions included can be seen in the output from the statistical software package, provided in Appendix C.
- Uncommon or hypothetical tank thicknesses can be analyzed with the head and shell formulas derived in this study. The regression process fits a curve, and it is not necessary to have empirical data at every value that might undergo analysis. Good overall fit and predictive capabilities of the final model will preserve the accuracy of estimates for interim thicknesses. Thicknesses that extrapolate beyond the limits of the study's dataset require more care. Typically, values that are beyond but not far from the dataset range can be analyzed with good confidence, but the farther out the value lies from the dataset's range, the more caution must be taken.

### 3) Variables Considered but Not Included

The following variables were considered for inclusion, but omitted from the analysis for the reasons noted.

- Additional tank materials beyond TC128B, A515, A516 – Data exploration prior to analysis found that there were too few records for other materials, and many (other than stainless steel) are of little current interest.
- Normalization of the tank – The status of a significant fraction of tanks in the study was uncertain. Also, previous studies have found that normalization has a minimal effect on the likelihood of a failure (see Ref. 2).
- Ambient temperature – It was decided that commodity was a better proxy for internal pressure, and that ambient temperature was not a reliable proxy for tank material temperature because of the range of loading temperatures. Likewise, geographic location and season were rejected as not being reliable enough indicators of the temperatures that the tank steel would be exposed to.
- Train weight, train length – The available energy in the accident environment for the derailed tank cars generally comes from a more localized portion of the train, not the entire consist.
- Outage – The information was not available in the Tank Car Safety Project database, and there was no reliable source for this information for individual cars, especially given that outage changes during transportation.
- Loaded vs. Residue – Loaded is the more important case for safety analysis, due to the higher quantities of lading available to spill. Residue cars were excluded for homogeneity.
- Tank Class – A single class can encompass a variety of features; specifying actual features was regarded as more directly related to performance.



- Additional/specialty car specifications (e.g., DOT/TC-103, -113-, -115, AAR-204) – These cars are too different in design and failure modes from the major classes, and too few in number to support robust regression analysis.
- Full sills vs stub sills – There are too few recent full-sill records, which are also strongly correlated with certain other features. Full-sill cars were excluded for homogeneity.
- Continuous weld pads on shells – The information was not available in the Tank Car Safety Project database, and the effort to identify these was deemed greater than the value added to the study.
- Pressure car top fittings segregated by commodity – This was found to be significant in the RA-05-02 analysis, but exploratory data analysis performed early in this study did not suggest that this effect would need to be evaluated.

#### 4) Definition of Effective Impact Speed

Although any given accident is assigned a speed at which the train derailed in its TCAD record and FRA report, the cars derailed in that accident do not all derail at the same speed, nor do they impact other objects at the same speed. Their lading loss outcomes are affected by this difference. Therefore it was hypothesized that inclusion of a variable to account for the difference could improve the modeling of variables of interest, including car features.

A new variable was created to measure this effect for any given accident, called Effective Impact Speed, or EIS.

$$EIS = TSP \times (CDR - TCL + 2) / (CDR + 1) \text{ where} \quad (1)$$

TSP = reported train speed

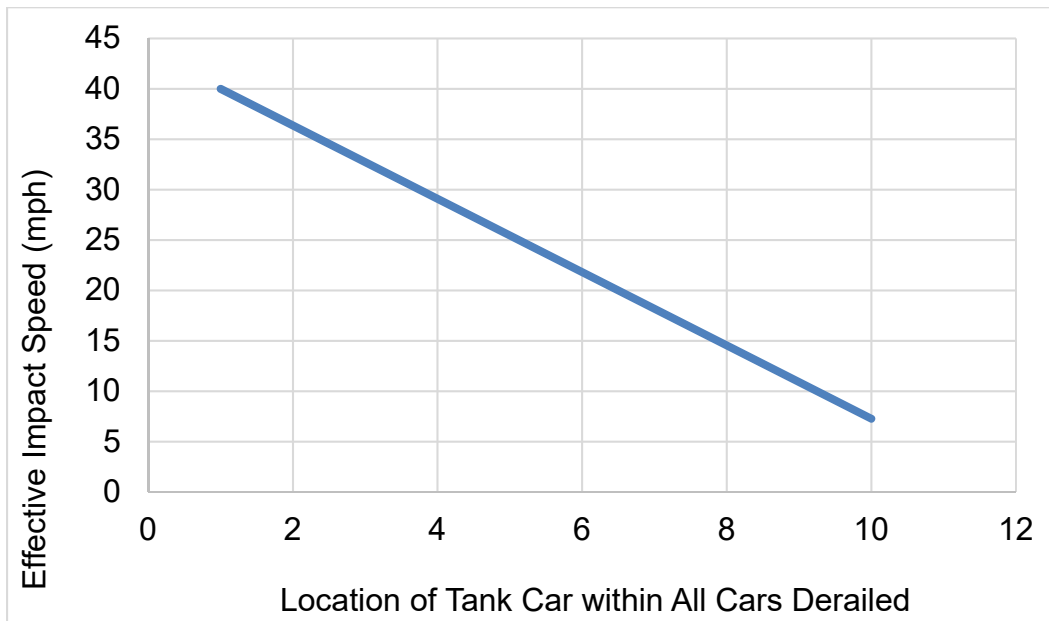
TCL = tank car's location in derailed string (1 = 1<sup>st</sup> car derailed, etc.)

CDR = total number of freight cars derailed

For cars with unknown position in the derailment, or unknown derailment size, EIS was set to TSP/2.

This formula produces a linear reduction in effective impact speed from the front to the back of the derailment. Figure 1 shows an illustrative example:

Figure 1  
Effective Impact Speed (EIS) Curve for a 10-Car Derailment at 40 mph



The formula does not precisely reflect the differences in real impact speeds, and is not intended to. Precision is not necessary because the purpose of this variable is not to analyze its effect on outcomes per se, but to give the regression process an additional data input to use in explaining some of the variation in outcomes that arises due to such an effect, and to minimize the effect of that variation on the characterization of other variables.

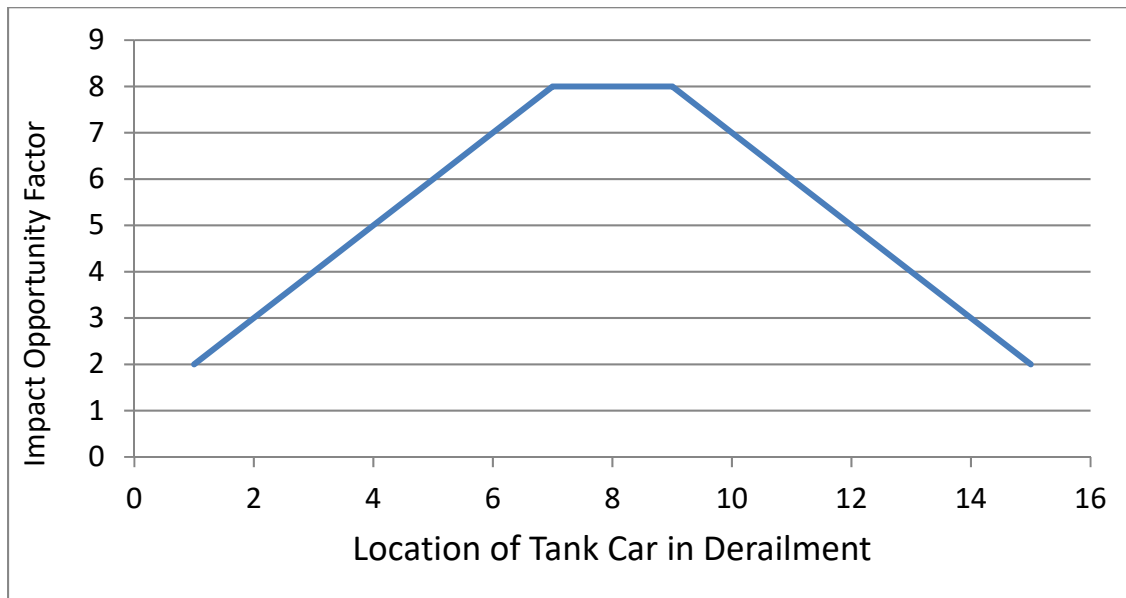
5) Definition of impact opportunity factor

Not all cars in a given accident have the same opportunities to strike or be struck by other freight cars. The cars at the front and back of the derailed string have fewer opportunities to impact other cars than do those towards the middle. The chance that a car will experience an impact capable of leak-causing damage is therefore different within the same accident. In order to account for this, a new variable was created, Impact Opportunity Factor, or IOF.

- IOF = 1 for single-car derailments
- = 2 for a car at either end of a multiple-car derailment
- = 3 for a car one spot inwards from the end of a derailed string of at least 3
- ....and so on up to maximum of 8

This formula produces a curve that peaks in the middle of the derailed string. Figure 2 shows an illustrative example:

Figure 2  
Impact Opportunity Factor Curve for a 15-Car Derailment



As with EIS, the curve does not precisely reflect the behavior of impact opportunities, but it does not need to in order to improve the regression.

6) Advanced variable selection approach

The power of regression is undermined when the independent variables are closely correlated. This is because the process is unable to determine which of two correlated variables is more responsible for a given car's outcome. These datasets do have correlated variables. For example, in the time period under consideration, head shields were far more likely to be found on cars with higher tank thickness than with lower thickness.

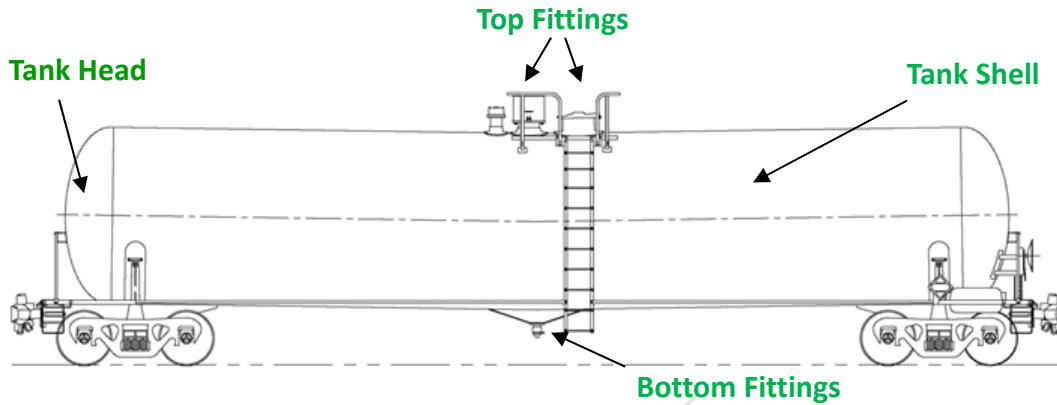
To minimize this problem, an advanced variable selection procedure, the group minimax concave penalty approach (gMCP), was conducted to identify groups of related variables that were more likely to be significant, before beginning each regression. This process helps to separate the effects of correlated variables. The technique resulted in more robust and accurate CPR estimates than would have been possible without such a step. More on gMCP can be found in Appendix B.

e) Regression approach

As in RA-05-02, a regression analysis was applied to four logistic regression formulas, or models. Each of the formulas selected as most representative of the data generates estimated CPRs that are specific to one of four components of the tank car: shell (S), head

(H), top fittings (T) and bottom fittings (B). Accident-caused lading losses come from one (or more) of these four components. A graphic representation of this scheme is shown in Figure 3.

Figure 3  
Four Components of a Tank Car That Can Lose Lading



These component CPRs are then combined using a standard probabilistic formula<sup>9</sup> into a CPR for the car as whole:

$$CPR_{CAR} = 1 - (1 - CPR_H)(1 - CPR_S)(1 - CPR_T)(1 - CPR_B) \quad (2)$$

The regression approach begins with an equation relating the outcomes, in this case the probability of lading loss, with the independent variables, i.e., explanatory factors that are assumed to have a reasonable chance to affect those outcomes. The initial equation is the full model, which includes all of the independent variables. Logistic regression is a standard technique that is used when the dependent variable being estimated is a proportion, and therefore ranges from zero to one, such as a probability estimate.

The observations in the dataset have values for each explanatory term in the model (tank thickness, presence of a jacket, and the others), and binary outcomes of zero or one, e.g., released lading from that component, or did not. The logistic regression procedure estimates the fraction of further observations, i.e., future derailed cars of a similar nature, that will have a value of one, that is, that will release lading. The curve (i.e., subset of significant independent variables and associated coefficients) that best explains the observed data and best predicts future observations is selected as the final regression equation. The independent variables in that final equation, and the coefficients assigned to each such term by the regression computation, provide the formula for estimating CPR.

<sup>9</sup> Use of this formula assumes that the component CPRs are independent, given a set of accident circumstances. This is not strictly true. However, the authors believe that the dependence is not large enough to significantly bias the results, and is additionally mitigated by the design of the full models, for example, including shell parameters in the top fittings model.

The component CPR formulas take the form:

$$\text{CPR}_i = e^{L(X_i)} / [1 + e^{L(X_i)}] \times \text{FRA multiplier}, \text{ where} \quad (3)$$

- $i$  = index to the car component, i.e., head, shell, top fittings or bottom fittings
- $\text{CPR}_i$  = the estimated probability of loss from source  $i$
- $L(X_i)$  = the calculated logistic regression equation for component  $i$
- FRA multiplier as defined below

The CPR for the car as a whole then is:

$$\text{CPR}_{\text{car}} = 1 - \prod_i (1 - \text{CPR}_i) \quad (4)$$

The features and properties of each car in the dataset are incorporated separately into the regression. That is, the input to the regression is not a car specification like “111A100W1”, but a set of values that precisely describe each car’s actual physical characteristics. This approach enables the wide range of characteristics in the fleet to enhance the statistical power and robustness of the analysis. Configurations that were not common in the dataset, but whose *features* are common, e.g., DOT-111 with top fittings protective housing, or DOT-117, can be analyzed accurately because of the component-by-component approach.

f) CPR for one-component losses

Cars that release lading might do so from one of the four components (shell, head, top fittings, bottom fittings), or more than one. Equation (3) above shows the formula for the CPR for one component, including the cases in which the car releases lading from more than one source. If it is desirable to consider releases strictly from one component alone, the component CPRs can be converted into mutually-exclusive ones using Equations (5a) and (5b):

$$\text{CPR}_{j^*} = \text{CPR}_j \times \prod_{i \neq j} (1 - \text{CPR}_i), \text{ and} \quad (5a)$$

$$\text{CPR}_{\text{mult}} = \text{CPR}_{\text{car}} - \sum_i \text{CPR}_{j^*}, \text{ where} \quad (5b)$$

- $\text{CPR}_{j^*}$  = the exclusive CPR for component  $j$ , i.e., releases from  $j$  and no other component,
- $j$  = index to the car component for which an exclusive CPR is desired, i.e., head, shell, top fittings or bottom fittings only,
- $\text{CPR}_{\text{mult}}$  = CPR for releases from more than one source (any combination of two or more).

g) FRA multiplier

The FRA multiplier adjusts the condition on the probability from “given that the car is damaged” (i.e., recorded in the Safety Project accident data) to “given that the car is derailed in an FRA-reportable accident”<sup>10</sup>. This is important when CPR estimates are used as part of a larger quantitative risk assessment, and combined with FRA accident rates and other data based on that metric. The FRA-compatible probability is the form of CPR that is most often used by the industry.

Calculated as shown in Appendix E, the FRA multiplier is 1.00. However, this has not been the case in previous studies, and might not be in the future, or additional work might reveal a need for different multipliers for specific contexts, so the term is left in Equation (3) to be thorough and accurate.

h) CPRs adjusted for minimum quantity lost

In some cases, an analysis using CPRs provides better focus on the questions at hand if small releases are excluded from consideration. For example, industry analyses of flammable liquid package options in the past few years have excluded releases of 100 gallons or less from individual tank cars, because the users of the analyses were trying to address the risk of large pool fires resulting from substantial releases of flammable lading. Releases of 100 gallons or less could not produce this scale of incident, and the various package options did not all have the same distribution of loss quantities, so it made sense to adjust the CPRs to account for this.

This adjustment to CPR is calculated by multiplying the unadjusted CPR by the fraction of releases for the specific car configuration in question that are over 100 gallons (or any other quantity<sup>11</sup>), based on empirical quantity-lost distributions.

The approach to handling this is to calculate the CPR according to this formula:

$$CPR_{>X,CAR} = 1 - \prod_{i=H,S,T,B} (1 - CPR_i \times Q(i,X)), \text{ where} \quad (6)$$

- $CPR_{>X,CAR}$  = the quantity-adjusted CPR for the car as a whole,
- $X$  = the gallon threshold; releases of  $X$  gallons or less will be excluded from  $CPR_{>X,CAR}$
- $i$  = index to the car component, , i.e., head, shell, top fittings or bottom fittings,

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<sup>10</sup> It sometimes occurs that cars are damaged in a way that qualifies them for inclusion in the Safety Project accident database, but does not bring the total cost of the accident up to the FRA reportability threshold. The opposite frequently occurs, i.e., a car is derailed as part of an FRA-reportable accident, but not damaged in a way that threatens containment of the lading.

<sup>11</sup> Note that the quantity-lost distribution data necessary to do this for other quantities are not presented here, and are beyond the scope of this report.

- $CPR_i$  = the unadjusted component CPR for component  $i$ , and
- $Q(i,X)$  = the probability that a release from component  $i$  on the car being estimated is greater than  $X$  gallons.

So  $CPR_{>X}$  is the probability that a single tank car releases more than  $X$  gallons, given that it is derailed in an FRA-reportable accident.

i) Quantity lost

The quantity of lading lost, given a release, was also analyzed, separately from CPR. The study characterized the distribution of lading lost from each of the four car components. Initially, regressions were attempted with categories of quantity lost as the dependent variable. However, there are far fewer release observations than damaged car observations, and not enough to support a robust regression. Therefore empirical histograms are provided instead.

j) Expected Quantity Released (EQR)

The term expected quantity of release (EQR) represents the average quantity lost per car derailed. (Note: It is not the average per car releasing. All non-releasing cars contribute zeros to the sum of all losses in the numerator of the calculation, while each releasing car contributes its own specific loss quantity to the numerator.) This metric is sometimes used instead of CPR for car comparisons. EQR is calculated by combining the four component CPRs for a particular car configuration with the quantity-lost distributions for each of those components. In principle the general form of such a calculation would be

$$EQR_i = \sum_{c=H,S,TF,BF} (CPR_{c,i} * Q_{c,i}) \quad (7)$$

where

- $EQR_i$  = Expected Quantity Released for car  $i$
- $CPR_{c,i}$  = CPR for car  $i$  and component  $c = H, S, TF$  or  $BF$
- $Q_{c,i}$  = Average quantity lost from releases from component  $c$  on car  $i$

However, some releases occur from multiple components on the same derailed car. When an accident data record is created, there is generally no way to determine how much of the total loss from the car came from each separate release location, so such data are not available. In equation (7) above, losses from multiple components are double-counted. Therefore, an adjustment must be made. A fifth “component”, or loss source, is created: Multiple, for losses from two or more sources. This source will be associated with its own average loss quantity. (Note that no effort is made to segregate the different pairings and combinations of sources that all contribute to the Multiple category. These differences are often small and are relevant to a small percentage of releases, and therefore this further adjustment is not necessary.)

If  $CPR^*_c$  designates the probability of a loss from component  $c$  alone, then

$$CPR'_c = CPR_c \times \prod_{n \neq c} (1 - CPR_n) \quad (8)$$

where  $c$  and  $n$  are indices to the principal four sources: H, S, TF, and BF.

And

$$CPR'_m = CPR_{car} - \sum_c CPR^*_c \quad (9)$$

where  $CPR_m$  is the probability of a multiple-source release, and  $CPR_{car}$  is the CPR for the whole car.

These single-source CPRs for the four principal sources and Multiple are mutually exclusive, and can therefore be summed to arrive at the overall car CPR. They also allow the proper calculation of EQR for car  $i$ :

$$EQR_i = \sum_{c=H,S,TF,BF,M} (CPR'_c, i * Q_c, i) \quad (10)$$

Safety features that reduce CPR for a component that experiences relatively small releases (e.g., top fittings) will reduce EQR to a smaller degree than they will reduce CPR, and to a smaller degree than alternative features that create the same reduction in CPR, but for a component that typically experiences larger releases (e.g., shell).

In practice, a simplifying assumption needs to be made if an analyst is relying on this report. The data on release quantities for some car/component combinations are not very numerous. Therefore the loss quantity data in the Results section of this report are given for pressure cars and non-pressure cars, rather than a more fine-grained selection of cars and features. The EQR example is calculated this way.

Note that this assumes that pressure car loadings are pressurized, and non-pressure car loadings are not under high pressure. It is assumed that the foremost reason for differences between the two groups of cars in terms of quantities lost given a release has to do with internal pressure and differences in lading behavior. If the cars of interest depart from these assumptions, for example, flammable liquids transported in DOT-120 cars, then the analyst should use the most reasonable assumptions about the distributions of quantity lost.

With this simplified approach, the quantity lost for a given component does not vary with the presence or absence of safety features or tank thickness differences. Only the component CPR portion of the calculation varies. Protection for those components that generally release more lading will still affect the EQR more than protection for components that release less. The EQR still is affected by relevant changes to each component, and can discriminate appropriately among car options in this regard.



### 3. Results

#### a) CPR formulas

The variables that had a significant effect on the four component CPRs in the regressions are shown below. Note that “significant” here means statistically significant, i.e., there is sufficient evidence in the data to conclude that the factor has a measurable effect on CPR. Variables that were not statistically significant have dropped out of the calculated regression equations. It should be noted that "statistically significant" is not the same as "materially significant". The regression coefficient for a given factor may be small enough that the effect of the factor on CPR is quite small, even though we are confident that there is a non-zero effect.

These expressions 1 through 4 below are represented by the term  $L(X_i)$  in the CPR equation (4) in Section 2(e) above.

#### 1) Shell CPR formula

	<u>Coefficient</u>		<u>Variable</u>	
$L_{SHELL} =$	5.6289		(Intercept)	(11)
+	-5.2092	×	Shell thickness (inches)	
+	-1.0986	×	Jacket (yes=1)	
+	-0.0588	×	Shell inside diameter (inches)	
+	-0.0514	×	Train speed (mph)	
+	0.00147	×	Train speed squared (mph <sup>2</sup> )	
+	-0.0254	×	Effective impact speed (mph)	
+	-0.00032	×	Effective impact speed squared (mph <sup>2</sup> )	
+	-0.0953	×	Cars derailed	
+	0.0739	×	Impact opportunity factor (scalar)	
+	0.00729	×	Train speed x cars derailed	
+	-0.00009	×	Train speed squared x cars derailed	
+	0.00736	×	Effective impact speed x Impact opportunity	

#### 2) Head CPR formula

	<u>Coefficient</u>		<u>Variable</u>	
$L_{HEAD} =$	-1.5145		(Intercept)	(12)
+	-3.0699	×	Head shield – Full (yes=1)	
+	-2.8387	×	Head shield – Half/jacket (yes=1)	
+	-0.6773	×	Head shield – Half/no jacket (yes=1)	
+	-1.6871	×	Jacket – no head shield (yes=1)	

+	-0.3447	×	Head thickness (inches)
+	0.5572	×	Heater coils (yes=1)
+	0.000433	×	Train speed (mph)
+	-0.2281	×	Impact opportunity factor (scalar)
+	-0.7613	×	Derailment vs. collision (Der.=1)
+	0.00709	×	Train speed x impact opportunity

### 3) Top fittings CPR formula

	<u>Coefficient</u>		<u>Variable</u>	
$L_{TOPFTG} =$	0.5555		(Intercept)	(13)
+	0.6168	×	Shell material A515 (yes=1)	
+	0.0675	×	Shell material A516 (yes=1)	
+	-2.7467	×	Shell thickness (inches)	
+	-0.5928	×	Jacket (yes=1)	
+	-0.0198	×	Shell inside diameter (inches)	
+	-1.0092	×	Top fittings protective housing (yes=1)	
+	0.8004	×	Derailment vs collision (Der.=1)	
+	0.000483	×	Effective impact speed squared (mph <sup>2</sup> )	
+	0.0195	×	Train speed (mph)	
+	-0.00084	×	Train speed squared (mph <sup>2</sup> )	
+	0.0584	×	Impact opportunity (scalar)	
+	-0.0009	×	Train speed x cars derailed	
+	0.00002	×	Cars derailed x train speed squared	
+	0.00004	×	Train speed squared x impact opportunity	

### 4) Bottom fitting CPR formula

	<u>Coefficient</u>		<u>Variable</u>	
$L_{BOTFTG} =$	-2.3275	×	(Intercept)	(14)
+	-0.873	×	Jacket (yes=1)	
+	-5.2546	×	Shell thickness (inches)	
+	0.00328	×	Train speed (mph)	
+	0.000846	×	Train speed squared (mph <sup>2</sup> )	
+	-0.0275	×	Effective impact speed (mph)	
+	0.0578	×	Cars derailed (integer)	
+	-0.2307	×	Impact opportunity factor (scalar)	
+	-0.00003	×	Train speed squared x cars derailed	
+	0.0111	×	Effective impact speed x impact opportunity	

Note if the car is not equipped with bottom fittings, then  $CPR_{BOTFTG}$  is automatically set to zero.

A few of the factors in these formulas are less intuitive than others at first consideration. For example, the presence of heater coils appears in the Head CPR formula, even though heater coils are not applied to tank heads. Perhaps heater coils affect the way that shells and heads interact to absorb energy during an impact; however, this would require more study to confirm or explain. But in a regression done in conjunction with gMCP, groups of related variables identified as important by that pre-regression process are included together unless there is a strong reason not to do so. These must be interpreted as a group for the following reason. Heater coils appear on jacketed cars, and jacket, head shield type and head thickness form a complex group of correlated factors that are all related to head puncture resistance. In such cases, the gMCP reduces the impact of correlations among the variables, but cannot eliminate it. (Note too that although the effect of coils is statistically significant, it is minimal; for a typical DOT-111, the  $CPR_{>100}$  goes from 0.196 to 0.189 when coils are added.)

Another way that gMCP treats a group of variables together is when it uses an interaction term to temper the effect of a main-effect factor. This can result in a counter-intuitive coefficient sign on some factors. For example, the coefficient on the Impact Opportunity Factor in the Head model is negative, apparently implying that CPR for a given tank car will go down as the number of derailed cars around the tank car (i.e., that it could potentially be impacted by) increases, which is counter-intuitive. However, this factor in this model is tied together with train speed by an interaction term, and the collective effect works in the direction that one would intuitively expect. There are several possible explanations for this analytic result. First, the calculations determined that the overall fit of the resulting formula to the data is best when these are the coefficients. Secondly, there could be complex effects at work that are beyond the capability of these formulations to reflect. For example, some derailments unfold in a linear fashion, while others form the common “accordion” pattern. The opportunities for heads to be damaged are different in these scenarios, and they may occur under different conditions, including train speed. More work would need to be done to settle whether phenomena like that are at work. In the shell model, train speed, cars derailed, and effective train speed receive negative coefficients. Again, interaction terms involving these three factors (and with other variables) are included in the model, and the collective effect works in the expected direction. CPRs move in the expected direction when any of these factors is increased or decreased. Taken collectively, the effects are logical. If this were not the case in this study, the example CPR estimates shown in Table 2 below and elsewhere would produce illogical comparisons, but they do not.

A few other inclusions may need explanation. The formulas consistently find little effect from tank steel specification. This is consistent with previous research using empirical accident data (Ref. 4). However, in the top fittings model, A515 steel is found to significantly differ from A516 or TC128B, and there is a modest difference between A516 and TC128B (Note that TC128B was indicated by setting the A515 and A516 variables to zero). This may be due to the effects on top fittings from the behavior of the tank and fittings nozzle, but more work would be needed to explore the cause. In any event, the result is not counter-intuitive and there is no evident reason to remove the

factor, so it has been retained. (The overall CPR effect of the variable is again modest but not totally negligible; see for example the difference between DOT-111s with TC-128 tanks vs, A516 in Table 4 below, which is attributable to this factor in this model.)

Shell thickness has a significant effect on bottom-fittings CPR, and this also deserves explanation, as it might be thought that all cars with bottom fittings would have the same tank thickness, thereby preventing an effect from that variable being found. Although a substantial majority (about 90%) of the cars in the database that were equipped with bottom fittings had the same shell thickness, 7/16", not all of them did. The remaining 10% (over 500 cars) had thicker shells, which is enough to allow the regression to determine an effect. For example, there are 159 cars with bottom fittings and a shell thickness of 5/8" and only three of these cars had bottom fittings losses (1.9%), compared to 3.0% for 7/16" cars. The data robustly support this finding.

b) CPRs for common tank cars and other cars of interest

Table 2 shows mainline CPR estimates for a variety of common tank cars and other cars of interest, based on the assumed configurations given. The following accident details are also assumed, based on the average mainline freight train accident reported to FRA from 2003 to 2012: train speed of 26 mph, 11 cars derailed, the tank car is the 6<sup>th</sup> car derailed. (Variations in these assumptions would change the CPR estimates in accord with the component CPR formulas.) Also assumed for comparative purposes is that the first car derailed was the first freight car in the train. The latter assumption affects only the EIS calculation, and any value could have been chosen as long as it was held constant from one estimate to the next.

Table 2<sup>12</sup>  
CPRs and CPR<sub>>100S</sub> for Selected Tank Cars  
Under Average Mainline Derailment Conditions

Car Specification	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200 <sup>13</sup>
111A100W1/3	0.4375	A516	Yes	No	No	Yes	119	0.134	0.089
111A100W1/2	0.4375	A516	No	No	No	No	119	0.265	0.194
111A100W1/2/3	0.4375	A516	Yes	No	No	No	119	0.130	0.086
111A100W2	0.5625	A516	No	No	Yes	No	100.625	0.201	0.167
111A100W1	0.4375	TC128B	Yes	Full	Yes	Yes	119	0.064	0.046
111A/S100W1	0.500	TC128B	No	Half	Yes	Yes	119	0.132	0.103
117R100W	0.4375	A516	Yes	Full	No <sup>14</sup>	Yes	119	0.126	0.081
117R100W	0.4375	TC128B	Yes	Full	Yes	Yes	119	0.064	0.046
117R100W	0.500	A516	Yes	Full	No <sup>14</sup>	Yes	119	0.106	0.067
117R100W	0.500	TC128B	Yes	Full	Yes	Yes	119	0.052	0.037
117J100W	0.5625	TC128B	Yes	Full	Yes	Yes	119	0.042	0.029
112J340W	0.625	TC128B	Yes	Full	Yes	No	118.75	0.033	0.023
105J300W	0.562	TC128B	Yes	Full	Yes	No	118.75	0.040	0.028
105A500W	0.779	TC128B	Yes	No	Yes	No	102	0.040	0.030
105J500W	0.797	TC128B	Yes	Full	Yes	No	102	0.031	0.022
112J500I	H1.03 / S0.89	TC128B	Yes	Full	Yes	No	115.34	0.017	0.011
105J600I	H1.136/S0.98	TC128B	Yes	Full	Yes	No	106	0.016	0.011

c) Tank thickness effects

The results confirm that increased tank thickness reduces CPR, all else being equal. Examples can be found in Table 3. Some of these cars are hypothetical, and are only included for illustrative purposes. The comparison would change with changes in other car and accident factors, but increasing tank thickness always reduces CPR.

<sup>12</sup> Table 2 is identical to Table E1 in the Executive Summary.

<sup>13</sup> Preliminary analyses used CPR = 0.266 and CPR<sub>>100</sub> = 0.196 for non-jacketed DOT/TC-111 cars. Subsequently an adjustment was made to the tank steel assumption that resulted in a slight increase in the CPRs.

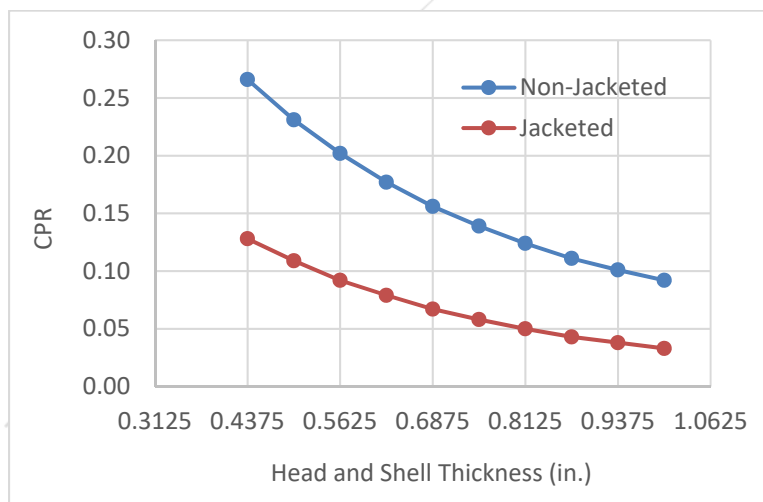
<sup>14</sup> Specifications for 117R100W require top fittings protection. However, while there are many different types and levels of top fittings protection today, the data used for this study did not allow those distinctions to be made quantitatively. Top fittings protection was either an impact-resistant protective housing, or none. Some current systems probably perform somewhere in between. Therefore, some examples using “No” for this field are presented, to suggest the range of performance that the real systems probably fall within.

Table 3  
CPRs and CPR<sub>>100s</sub> for Selected Groups of Comparable Tank Cars  
With Different Tank Thicknesses  
Under Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200
111A100W1	0.5000	A516	No	No	No	Yes	119	0.239	0.173
111A100W1	0.5625	A516	No	No	No	Yes	119	0.208	0.151
111A100W1	0.6250	A516	No	No	No	Yes	119	0.182	0.133
105J300W	0.5625	TC128B	Yes	Full	Yes	No	118.75	0.040	0.028
112J340W	0.625	TC128B	Yes	Full	Yes	No	118.75	0.033	0.023

Figure 4 illustrates the general relationship between tank thickness and CPR. The DOT-111 cars are assumed to have bottom fittings and 119” inside diameter, and no top fittings protection or head shields.

Figure 4  
Relationship Between Tank Thickness and CPR  
Illustrated for Jacketed and Non-Jacketed 111A100W1 Tank Cars  
Under Average Mainline Derailment Conditions



d) Tank steel type effects

The results indicate that there is little difference in empirical CPR performance between TC128B steel and A516, but both provide lower CPRs than A515, all else being equal. Examples can be found in Table 4. The laboratory performance of TC128B and A516 is not equal, but this study finds that the difference in yield strength and other properties does not translate into a significant difference in accident performance. This is possibly

because impacts that deliver forces that fall within the interval between the two strengths might not dominate the accident environments. Similar studies over the years have found the same result. It should be noted that the differences between A516 and TC128B in this table are less than the uncertainty surrounding either estimate, i.e., they are not statistically significant, although they are directionally consistent with lab data.

Table 4  
CPRs and CPR<sub>>100s</sub> for Selected Pairs of Comparable Tank Cars  
With Different Tank Steel Types  
Under Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A515	No	No	No	Yes	119	0.354	0.246
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200
111A100W1	0.4375	TC128B	No	No	No	Yes	119	0.266	0.196
112J340W	0.625	A515	Yes	Full	Yes	No	118.75	0.051	0.033
112J340W	0.625	A516	Yes	Full	Yes	No	118.75	0.034	0.024
112J340W	0.625	TC128B	Yes	Full	Yes	No	118.75	0.033	0.023

e) Top fittings protection effects

The results confirm that there are benefits to protecting the top fittings in a damage-resistant housing. Table 5 shows one comparison. The comparison would change with changes in other car and accident factors, and as noted in the notes on the data in Section 2(d)(2), there are other top fittings protection systems in the fleet now<sup>15</sup> whose

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<sup>15</sup> Another aspect of top fittings can be considered, and further work should be done to determine whether an adjustment should be made to account for this. Top fittings protection as defined in this study is a pressure car protective housing design, and its performance is based on the accident history with that system. Historically, cars with a protective housing were pressure cars, and therefore did not have a separate manway; the housing was mounted on a pressure plate that also sealed the manway. On a DOT/TC-117 car, on the other hand, the manway generally will be separate from the fittings nozzle. Therefore, with respect to top fittings protection overall, the 117 car falls somewhere in between the 111 car and a pressure car. Based on Tank Car Safety Project data for the study period, 29% of top fittings releases from flammable liquid and combustible liquid 111s came from the manway. It may make sense to adjust the benefit of top fittings protection for these cars by that amount. However, there are other factors, including possible protection due the manway's placement directly adjacent to the robust protective housing that might provide some safety benefit in certain types of accident scenarios. Resolution of this question will require analysis beyond this study. CPR studies require revisiting and updating over time as new or more refined data become available and the TWP-17 study is no exception. When new work indicates that CPR estimates need adjustment, supplemental reports will be issued to provide analysts with more current data and statistics.

performance cannot be distinguished in this study. Adding top fittings protection always reduces CPR.

Table 5  
CPRs and CPR<sub>>100s</sub> for Comparable Tank Cars  
With Different Top Fittings Protection Levels  
Under Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200
111A100W1	0.4375	A516	No	No	Yes	Yes	119	0.187	0.151
117R100W	0.4375	A516	Yes	Full	No <sup>16</sup>	Yes	119	0.126	0.081
117R100W	0.4375	A516	Yes	Full	Yes	Yes	119	0.066	0.047

Note that the effectiveness of top fittings protection is based on the design being similar to the traditional pressure-car-style protective housing described in the federal regulations at 49 CFR 179.100-12(c). This is because the CPR estimation is empirically based, and this protection design is the only one represented by meaningful numbers of cars in TCAD. In recent years, many additional types of top fittings protection have been introduced. However, none of these designs have been widespread in the fleet for long enough for their effects to be quantifiable. Therefore, at this time, the best estimate of the performance of any of these systems is the same estimate that has been derived for the traditional pressure-car housing. Future CPR analyses will be better able to distinguish between the various top fittings protection approaches.

The predecessor to this report, RSI-AAR Report RA-05-02 (Ref. 4), found a difference between pressure car housings on different specification cars, for example, chlorine cars vs. anhydrous ammonia cars. However, in the TWP-17 study, no evidence was found for this in the exploratory phase of the project, and the pressure car fittings type variable was not selected for inclusion in the full top fittings model.

f) Jacket effects

The results confirm that there are significant tank damage resistance benefits to a jacket, even though its main purpose is to contain insulation and/or thermal blankets. Table 6 shows some example comparisons. The comparisons would change with changes in other car and accident factors, but the presence of a jacket always confers benefits in this regard. Note that the potential effect of the standoff distance, i.e., the distance between

<sup>16</sup> Specifications for 117R100W require top fittings protection. However, while there are many different types and levels of top fittings protection today, the data used for this study did not allow those distinctions to be made quantitatively. Top fittings protection was either an impact-resistant protective housing, or none. Some current systems probably perform somewhere in between. Therefore, some examples using “No” for this field are presented, to suggest the range of performance that the real systems probably fall within.



the tank and the jacket, was also studied, but this variable was found to not be statistically significant.

Table 6  
CPRs and CPR<sub>>100s</sub> for Selected Pairs of Comparable Tank Cars  
With and Without Standard 11-Gauge Jackets  
Under Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	No	No	No	Yes	119	0.274	0.200
111A100W1	0.4375	Yes	No	No	Yes	119	0.134	0.089
111A100W1	0.5	No	No	No	Yes	119	0.239	0.173
111A100W1	0.5	Yes	No	Yes	No	119	0.114	0.075
112S340W	0.625	No	Half	Yes	No	118.75	0.091	0.071
112J340W	0.625	Yes	Half	Yes	No	118.75	0.034	0.023

g) Head protection effects

The results confirm that there are benefits to head protection systems, and more benefits to full-height protection than for half-height. Table 7 shows some example comparisons. The comparisons in Table 7 would change with changes in other car and accident factors, but the presence of head protection always confers benefits in this regard. The relative benefits are affected by where impacts on the tank head occur (mostly below the centerline) and the protective effects of a jacket head or partial jacket head, which is less resistant to damage than a head shield, but still offers significant resistance.

Table 7  
CPRs and CPR<sub>>100s</sub> for Selected Groups of Comparable Tank Cars  
With Different Levels of Head Protection  
Under Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A516	No	No	No	Yes	119	0.274	0.200
111A100W1	0.4375	A516	No	Half	No	Yes	119	0.252	0.178
111A100W1	0.4375	A516	Yes	No	No	Yes	119	0.134	0.089
111A100W1	0.4375	A516	Yes	Half	No	Yes	119	0.127	0.082
111A100W1	0.4375	A516	Yes	Full	No	Yes	119	0.126	0.081
112J340W	0.625	TC128B	Yes	Half	Yes	No	118.75	0.034	0.023
112J340W	0.625	TC128B	Yes	Full	Yes	No	118.75	0.033	0.023

h) Speed effects (mainline/siding)

Given a specific tank car configuration, CPRs are lower at lower derailment speeds. Note that “derailment speed” means the speed that the train was traveling at the moment of derailment, not the impact speed at which an impactor struck the car or the car struck the ground. The latter speeds are almost always lower than the derailment speed, but they are unknown, so derailment speed must serve as one indicator of how high the impact speeds might have been, relatively. Table 8 shows some example comparisons. The example speeds of 25 and 40 mph are the speed limits for FRA Class 2 and 3 track, respectively. 50 mph is the AAR OT-55 speed limit for key trains [Ref. 1] and the regulatory limit for trains meeting the definition of “high-hazard flammable train”. The comparisons in Table 8 would change with changes in other car and accident factors, and in particular, would change if computed with the average derailment size for a given speed, rather than the overall average.

Table 8  
CPRs and CPR<sub>>100s</sub> for Selected Cars  
Derailed at Different Speeds  
Under Otherwise Average Mainline Derailment Conditions

Car Spec.	Head and Shell Thickness (in.)	Head and Shell Steel Type	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	Mph	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	A516	Yes	No	No	Yes	119	25	0.132	0.087
111A100W1	0.4375	A516	Yes	No	No	Yes	119	40	0.176	0.129
111A100W1	0.4375	A516	Yes	No	No	Yes	119	50	0.241	0.193
117J100W	0.5625	TC128B	Yes	Full	Yes	Yes	119	25	0.041	0.029
117J100W	0.5625	TC128B	Yes	Full	Yes	Yes	119	40	0.064	0.050
117J100W	0.5625	TC128B	Yes	Full	Yes	Yes	119	50	0.102	0.085
105A500W	0.779	TC128B	Yes	No	Yes	No	102	25	0.040	0.030
105A500W	0.779	TC128B	Yes	No	Yes	No	102	40	0.044	0.036
105A500W	0.779	TC128B	Yes	No	Yes	No	102	50	0.066	0.057

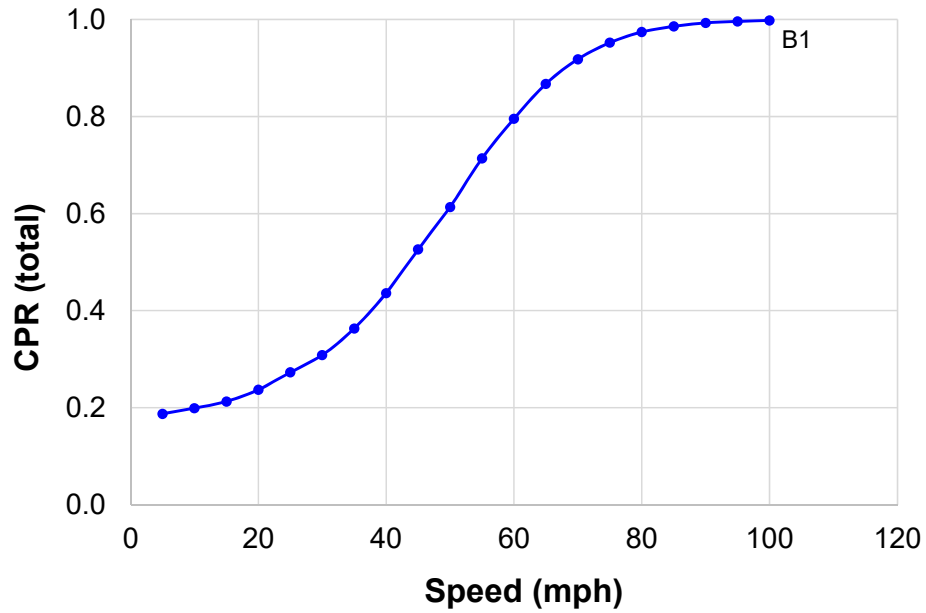
Figure 5 graphically depicts the speed effect on overall car mainline CPR, for non-jacketed 111A100W1s with no additional protective features. Figure 6 shows the same, for jacketed cars. In these figures, at each speed, the average derailment size for that speed is incorporated into the CPR calculation.

Note that the speed effect curve does not go through the origin. On the face of it, CPR at speeds very near zero mph should also be very near zero. When constructing the models for regression, one must choose whether or not to force the intersection to occur at one particular point, such as the origin. This can be done, but often at a cost to the goodness-of-fit elsewhere in the data domain. The curve must in a sense be bent away from a shape that would fit more data points better, in order to go to the origin. A decision was made that the accuracy of analyses at speeds very near zero was less important than the accuracy of analyses at typical mainline speeds. Therefore the speed effect was not forced through [0,0].

Figure 5

Non-Jacketed 111A100W1 Mainline CPR vs. Train Speed  
As Predicted by TWP-17 Formulas, and Empirical Counts

Subfigure 5a  
TWP-17 CPR Predictions



Subfigure 5b  
Empirical Lading-Loss Rates

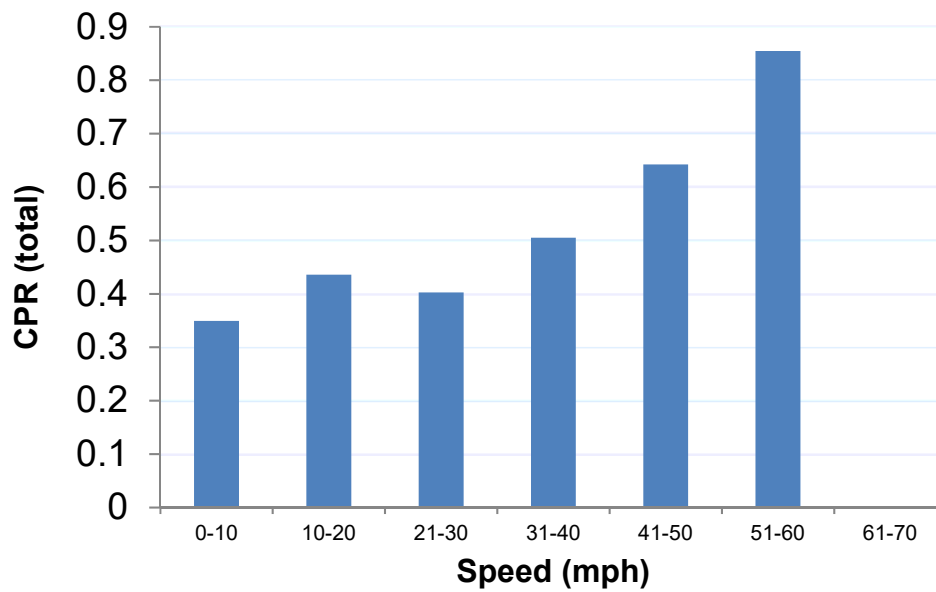
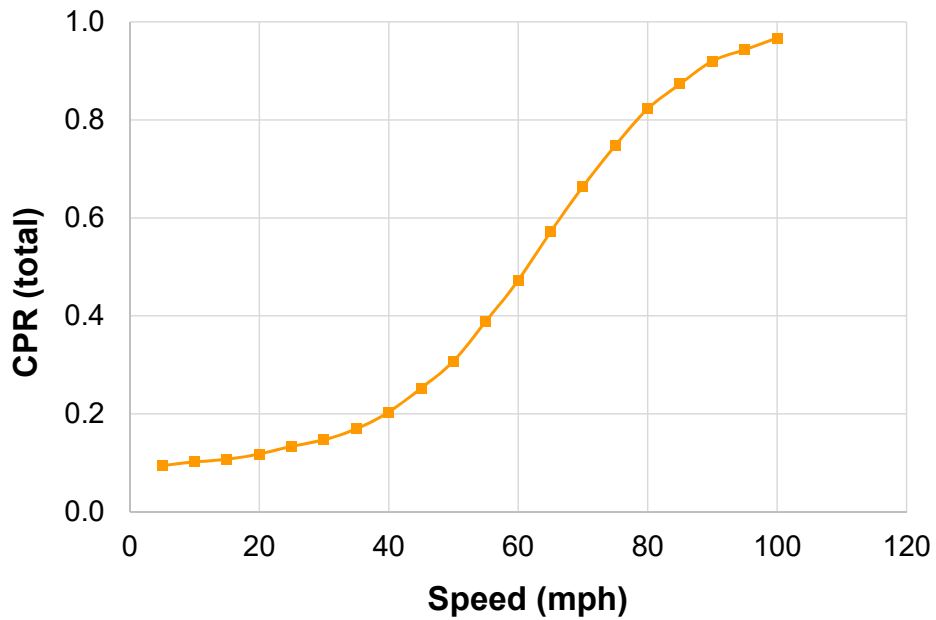


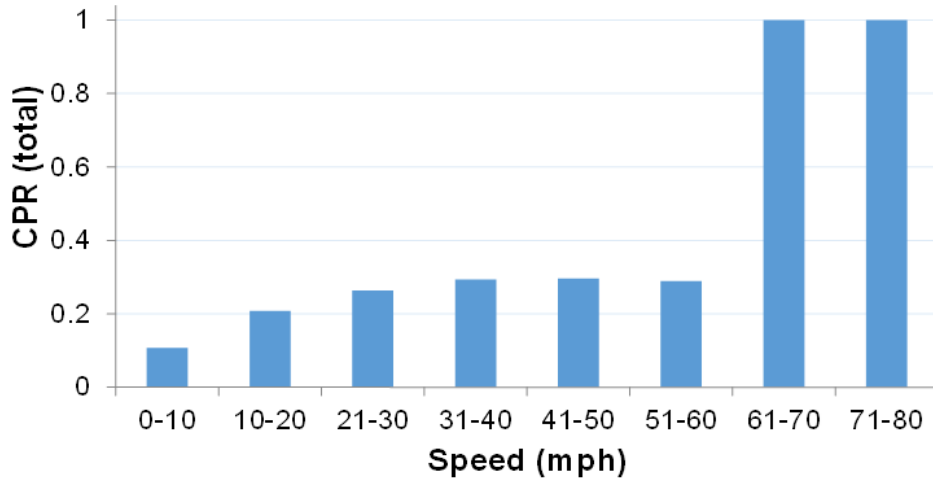
Figure 6

Jacketed 111A100W1 Mainline CPR vs. Train Speed  
As Predicted by TWP-17 Formulas, and Empirical Counts

Subfigure 6a  
TWP-17 CPR Predictions



Subfigure 6b  
Empirical Lading-Loss Rates



i) Yard/Industrial CPRs

CPRs on yard or industrial track are always lower than CPRs for the same cars on mainline or siding track. However, the variable for track type was not statistically significant in any of the component CPR models, so the dominant factors are the lower speeds and smaller derailments. Table 9 shows some example comparisons. The average derailment conditions for mainline track in the FRA accident data are 26 mph, 11 cars derailed, tank car of interest 6<sup>th</sup> in the derailed string. The average conditions for yards are 6 mph, 5 cars derailed and tank car of interest 3<sup>rd</sup> in the derailed string.

Table 9  
CPRs and CPR<sub>>100s</sub> for Selected Cars  
Derailed on Different Track Types  
Under Average Derailment Conditions for Each Type

Car Spec.	Head and Shell Thickness (in.)	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	M= Main Y= Yard	CPR	CPR <sub>&gt;100</sub>
111A100W1	0.4375	No	No	No	Yes	119	M	0.274	0.200
111A100W1	0.4375	No	No	No	Yes	119	Y	0.075	0.051
117J100W	0.5625	Yes	Full	Yes	Yes	119	M	0.042	0.029
117J100W	0.5625	Yes	Full	Yes	Yes	119	Y	0.010	0.007
105A500W	0.779	Yes	No	Yes	No	102	M	0.040	0.030
105A500W	0.779	Yes	No	Yes	No	102	Y	0.007	0.005

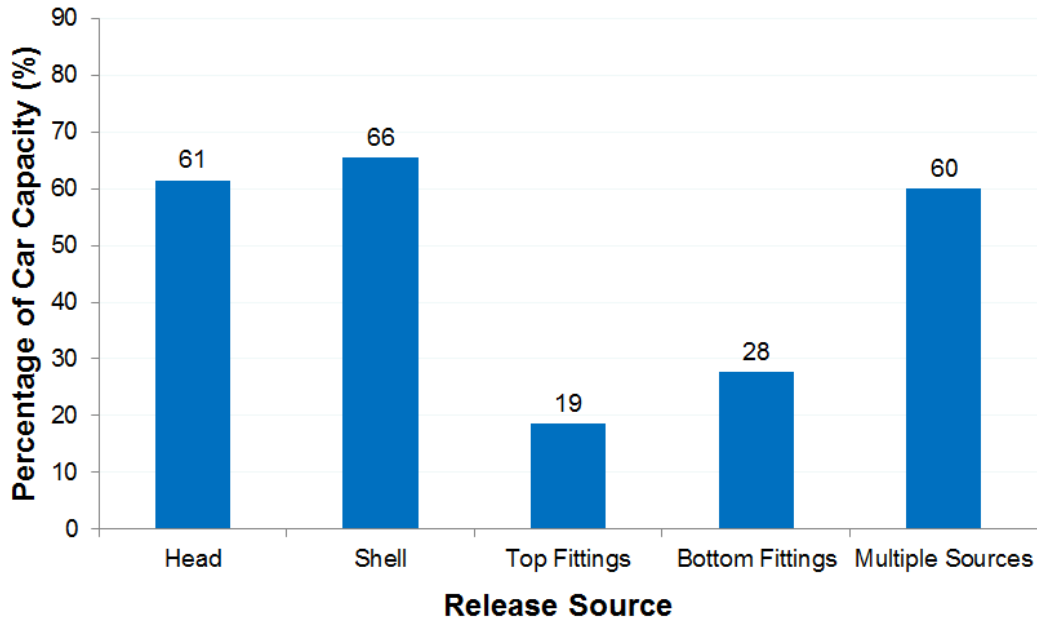
j) Quantity lost results

Because tank cars have a wide range of capacities, a loss of a given gallon volume can represent a different severity of outcome for different cars. For example, a 15,000 gallon loss is 50% of the lading from a 30,000 gallon car, but a total loss from a 15,000 gallon car. Therefore quantity lost data are represented in this study in terms of percentage of car capacity lost. Figure 7 shows the average percentage of car capacity lost in releases from each of the four car components, in four sub-tables based on car type and track type. The average percent-lost figures for each car/component in Figure 7 are shown with only two significant digits in order to avoid presenting a level of precision that may not be supportable, given the relatively small number of release accidents available for the calculation and the number of factors that can affect the quantity lost. Also, the thousandths place on the percentage of a loaded tank represents tens of gallons, and generally a round-off at that level will not bias an analysis.

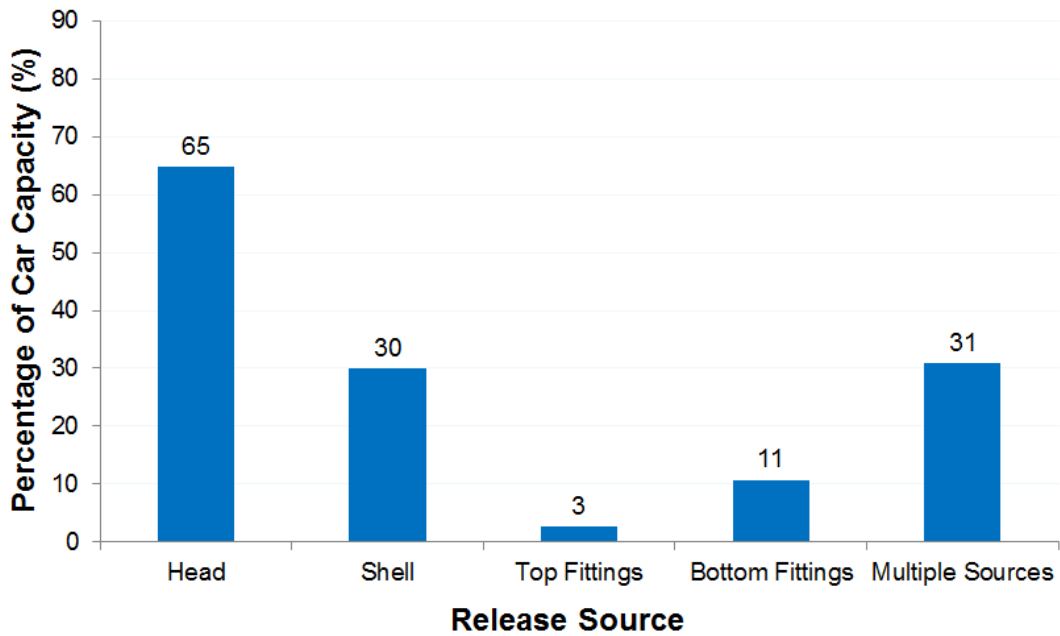
Figure 8 shows the distribution of loss quantities for each component, broken down into five bins of percentages. The bins are not uniform because of how loss quantity contributes to risk. For example, very small releases often present a disproportionately smaller hazard than 20% or 25% losses, so it is not helpful to lump these together. Table 10 shows the data behind Figure 8. Note that there are few releases from pressure cars, and the distributions shown may not be reliable representations of the relative likelihoods of different release sizes.

Figure 7  
Average Quantity of Lading Lost for Each Car Component  
Expressed as a Percentage of Car Capacity

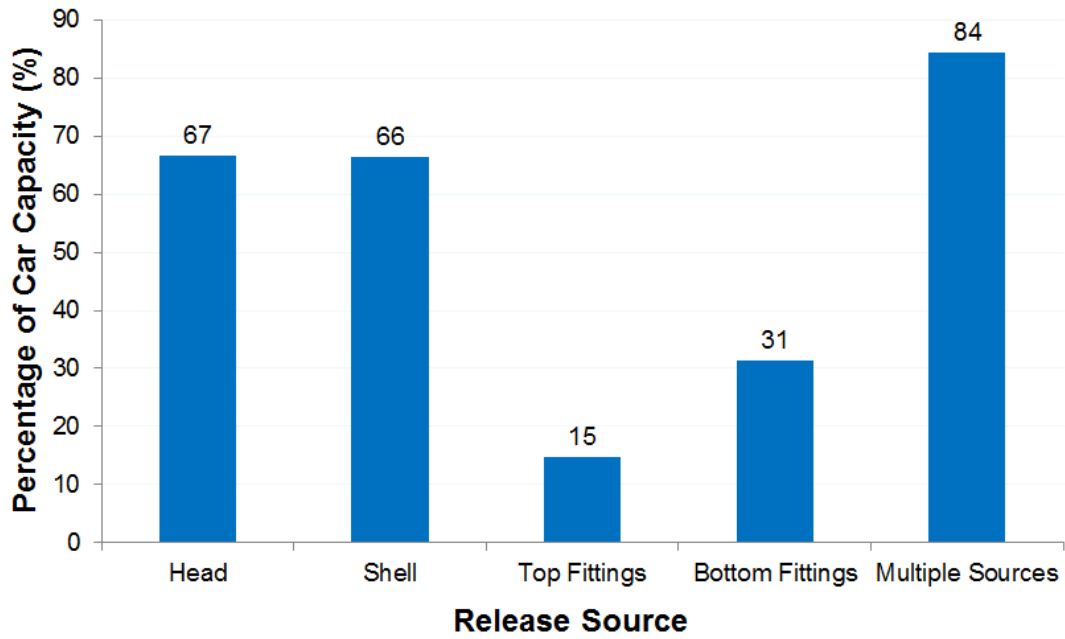
Subfigure 7a  
Non-Pressure Cars, Mainlines and Sidings



Subfigure 7b  
Non-Pressure Cars, Yards and Industrial Track



Subfigure 7c  
Pressure Cars, Mainlines and Sidings



Subfigure 7d  
Pressure Cars, Yards and Industrial Tracks

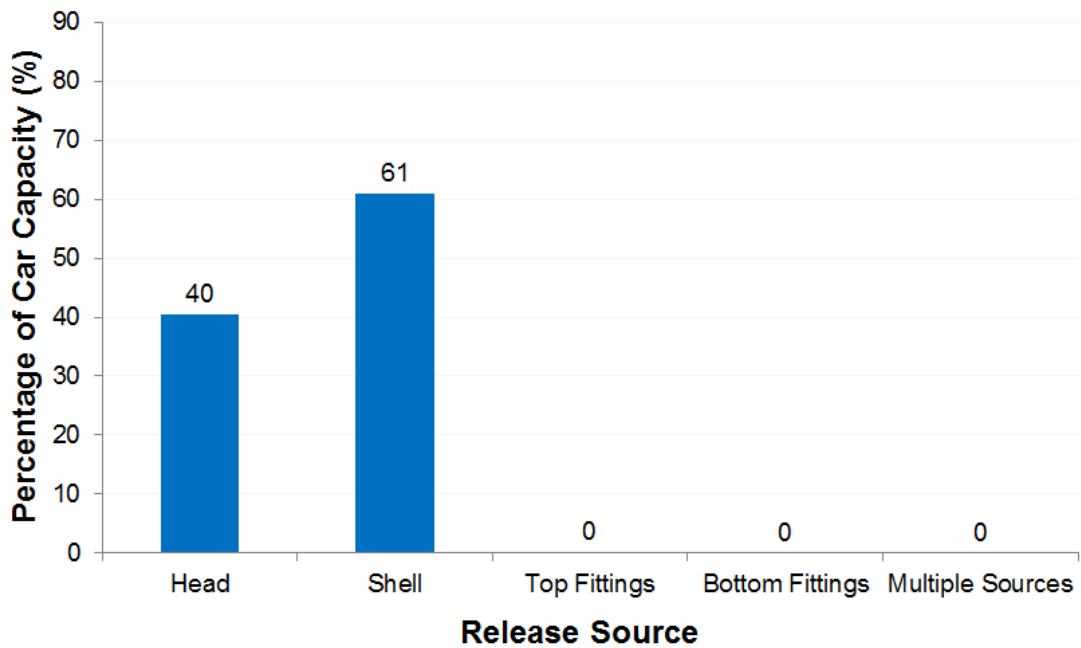
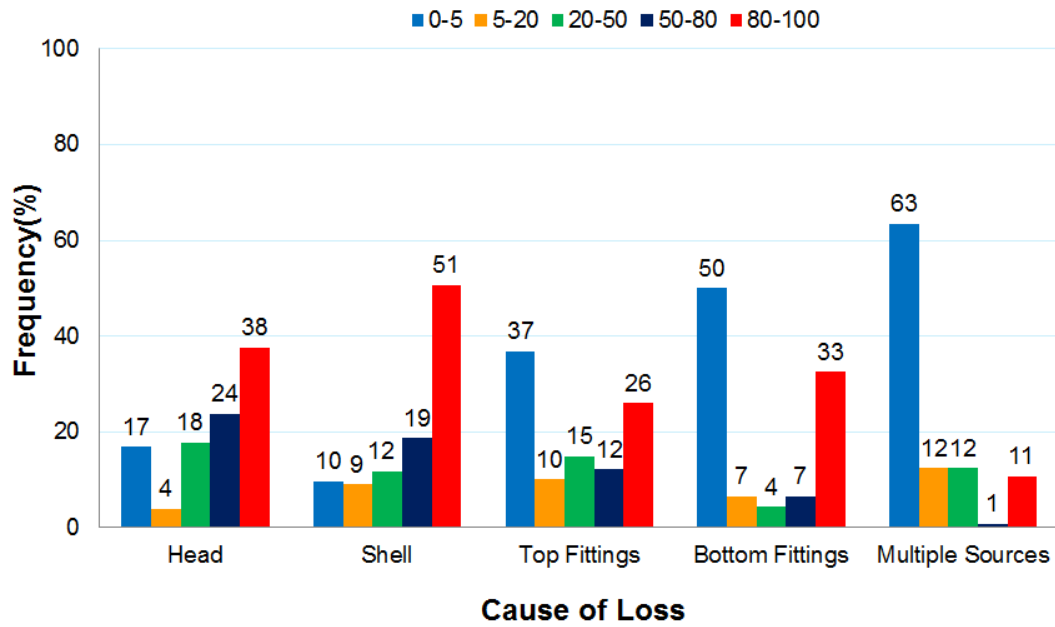
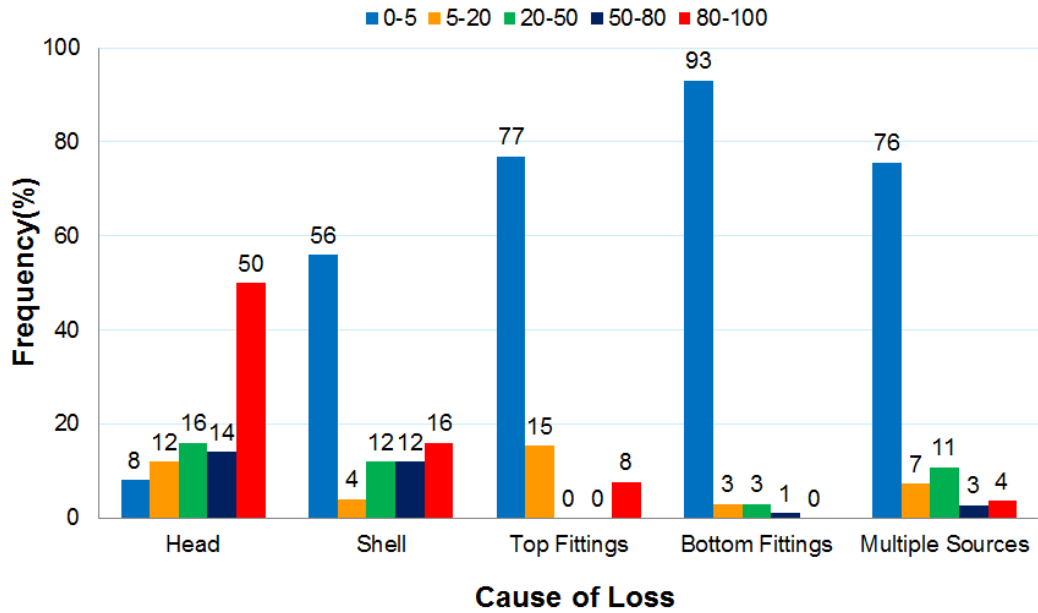


Figure 8  
Distribution of Quantity of Lading Lost, Expressed as a Percentage of Loaded Quantity, for Each Car Component  
(Bar Color Indicates Category of the Quantity Lost, as a Percentage)

Subfigure 8a  
Non-Pressure Cars, Mainline and Sidings

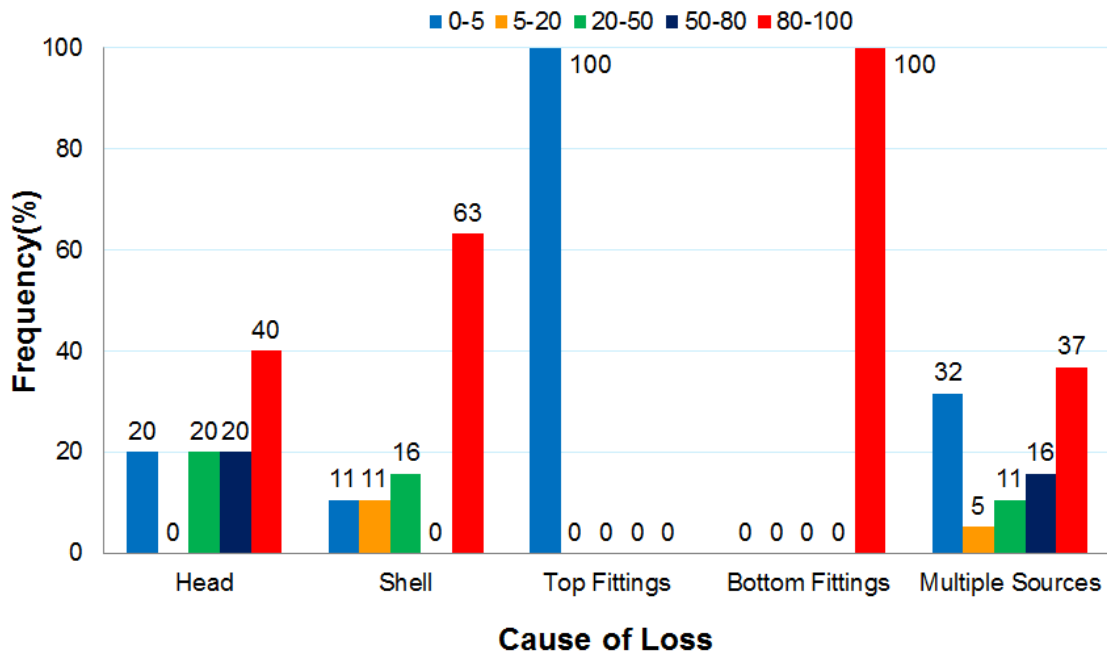


Subfigure 8b  
Non-Pressure Cars, Yards and Industrial Tracks





Subfigure 8c  
Pressure Cars, Mainline and Sidings



Subfigure 8d  
Pressure Cars, Yards and Industrial Tracks

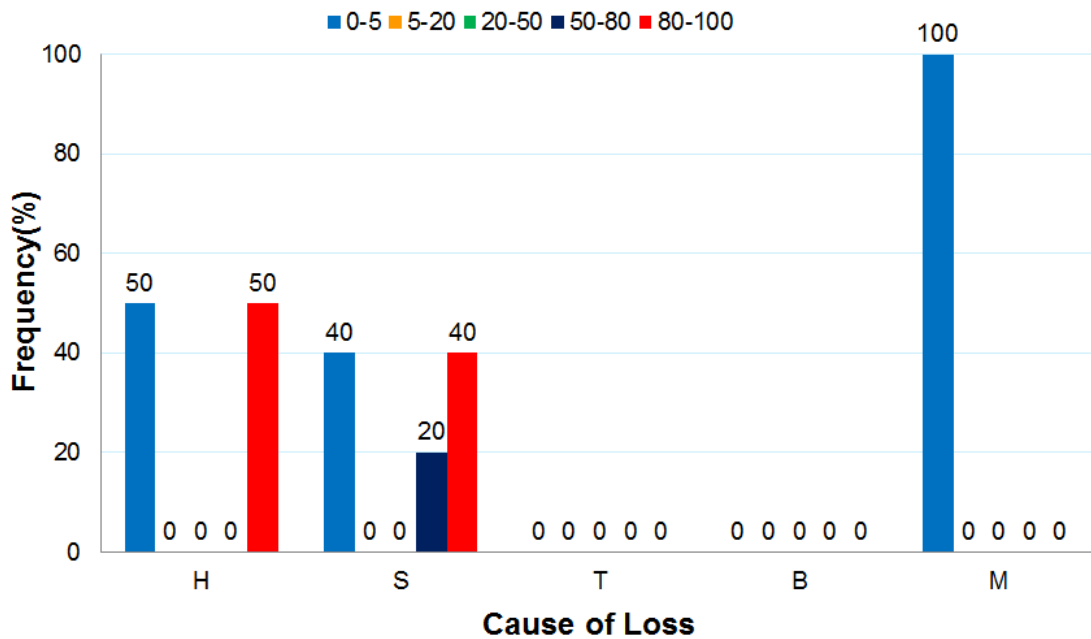


Table 10 provides the count data displayed as percentages in Figure 8.

Table 10  
Number of Tank Cars that Lost Lading, by Percent of Tank Capacity

Subtable 10a  
Non-Pressure Cars on Mainlines and Sidings

Loss Source	Percent of Capacity Lost					Total Cars
	0-5	5-20	20-50	50-80	80-100	
H	17	4	18	24	38	101
S	28	26	34	54	146	288
T	106	29	43	35	75	288
B	23	3	2	3	15	46
M	153	30	30	2	26	241
Total Cars	327	92	127	118	300	964

Subtable 10b  
Non-Pressure Cars, Yards and Industrial Tracks

Loss Source	Percent of Capacity Lost					Total Cars
	0-5	5-20	20-50	50-80	80-100	
H	4	6	8	7	25	50
S	14	1	3	3	4	25
T	10	2	0	0	1	13
B	94	3	3	1	0	101
M	84	8	12	3	4	111
Total Cars	206	20	26	14	34	300

Subtable 10c  
Pressure Cars, Mainlines and Sidings

Loss Source	Percent of Capacity Lost					Total Cars
	0-5	5-20	20-50	50-80	80-100	
H	1	0	1	1	2	5
S	2	2	3	0	12	19
T	6	0	0	0	0	6
B	0	0	0	0	1	1
M	6	1	2	3	7	19
Total Cars	15	3	6	4	22	50

Subtable 10d  
Pressure Cars, Yards and Industrial Tracks

Loss Source	Percent of Capacity Lost					Total Cars
	0-5	5-20	20-50	50-80	80-100	
H	1	0	0	0	1	2
S	2	0	0	1	2	5
T	0	0	0	0	0	0
B	0	0	0	0	0	0
M	9	0	0	0	0	9
Total Cars	12	0	0	1	3	16

k) Expected Quantity Released (EQR) example

To illustrate why EQR calculations might be of interest, here is an example in which a non-jacketed 111A100W1 tank car with a bottom outlet is compared to 1) the same car with the bottom outlet removed, 2) the same car with top fittings protection added, and 3) the same car with a jacket added. Table 11 displays CPR and EQR estimates for these three cars. Note that the percentage of improvement can vary significantly for the same risk reduction measure, depending on the performance metric used.

Table 11  
EQRs, CPRs and CPR<sub>>100S</sub> for 30,000-Gallon DOT-111 Tank Cars  
With Different Risk Reduction Options (RROs)  
(7/16" Tanks, 119" Shell Inside Diameter)  
Under Average Mainline Derailment Conditions

Car	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	CPR	% Reduction	CPR <sub>&gt;100</sub>	% Reduction	EQR (gals.)	% Reduction
Base	No	No	No	Yes	0.274	n/a	0.200	n/a	3,120	n/a
RRO1	No	No	No	No	0.257	3.4 %	0.189	3.6%	3,026	3.0 %
RRO2	No	No	Yes	Yes	0.183	31.2 %	0.149	24.0 %	2,613	16.3%
RRO3	Yes	No	No	Yes	0.128	51.9 %	0.085	56.6 %	1,199	61.6 %

RRO1 = Remove bottom fittings

RRO2 = Add top fittings protection

RRO3 = Add jacket to head and shell

EQRs in Table 11 are based on the shell-full capacity of the tanks, i.e., 30,000 gallons in this case. For purposes of comparisons among car options with the same tank capacity, this is the simplest approach. For other analyses, it might be important to focus on the actual gallons loaded, or the estimated gallons of lading at a typical point in transit, or the tons (which do not vary during the trip). Note that the reported loss quantities from accidents, which the fractions in Figure 7 are based upon, come from cars that have been in transit for some period of time. Vapor lost from the tank is generally not part of the reported loss quantity unless vapor is all that is lost (for example if a fitting is damaged but is above the liquid level so that only vapor is lost). The analyst will need to consider the best approach for a given study.

#### 4. References

1. Association of American Railroads. “Recommended Railroad Operating Practices for Transportation of Hazardous Materials, Circular No. OT-55-Q (CPC-1337)”. 6 September 2018.
2. Hughes, J.P., C.F. Heuer and T.L. Anderson. “Fracture Behavior of Tank Car Steels in Accidents from 1981 through 1994. RSI-AAR Railroad Tank Car Safety Research and Test Project Report RA-03-6-62. December 1998.
3. Kirkpatrick, S. W. “Detailed Puncture Analyses of Tank Cars: Analysis of Different Impactor Threats and Impact Conditions.” DOT/FRA/ORD-13-17. March 2013.
4. Treichel, T.T., J.P. Hughes, C.P.L. Barkan, R.D. Sims, E.A. Phillips, and M.R. Saat. “Safety Performance of Tank Cars in Accidents: Probabilities of Lading Loss”. RSI-AAR Railroad Tank Car Safety Research and Test Project Report RA-05-02. January 2006.

## Appendix A Tank Car Safety Project Collection of Data on Tank Cars Damaged in Accidents

The source of the data used in this study was the Tank Car Safety Project's database of tank cars damaged in accidents, known as the Tank Car Accident Database, or TCAD. The database contains records on tank cars damaged in derailments and collisions, for which the damage was incurred on tank-car-specific features, including the tank, fittings, jacket, head shields, and so on. If there was no damage to these features, the car is not included in TCAD, even if there was damage to running gear, safety appliances, brake equipment, etc.

TCAD includes data on any tank cars damaged in the United States or Canada, regardless of specification, design features, lading (including non-regulated materials), load/residue status, and whether or not lading was released in the accident.

The Tank Car Safety Project began collection and analysis of accident data in 1970. At the time of publication of this report, TCAD contained records on almost 47,000 damaged tank cars and just over 30,000 associated accidents. Not all of these are used in any one analysis. The set of cars used in TWP-17 is described in the Approach section of the main report.

For each car TCAD contains data in three categories: 37 fields describing the accident that damaged the car, 40 fields describing the car's features and lading, and 34 fields describing the damage to and performance of the car.

These data are collected from 28 disparate sources, including public data reported to FRA and PHMSA, proprietary data provided by car owners, car builders and railroads, NTSB, TSB and FRA investigations, Tank Car Safety Project accident investigations, and others. Engineers experienced in assessing tank car damage collate whichever of these sources are available for a particular car, resolve conflicts and gaps, and record an accurate summation of the event.

## Appendix B gMCP

The power of regression is undermined when the independent variables are closely correlated. This is because the process is unable to determine which of two correlated variables is more responsible for the outcome. These datasets do have correlated variables. For example, in the time period under consideration, head shields were far more likely to be found on cars with higher tank thickness than with lower thickness.

To minimize this problem, an advanced variable selection procedure, the group minimax concave penalty approach (gMCP), was conducted to identify groups of related variables that were more likely to be significant, before beginning each regression. This process helps to separate the effects of correlated variables. It also allows both related groups of variables and individual variables to be selected for their importance. The technique resulted in more robust and accurate CPR estimates than would have been possible without such a step.

In brief, gMCP optimizes an objective function with respect to a single group of related variables at a time, cycling through the groups until convergence is reached. The process includes all variables in groups at first, then considers them individually for exclusion. Whether an individual variable enters the regression model is affected by both its own explanatory value and that of its group.

Note that coefficients with confidence limits encompassing zero, and therefore considered not significant in unmodified regression, are often left in the final model with this approach if they were identified as significant in the group selection process. This is because the process shrinks coefficients towards zero, trading some bias for prediction accuracy, and the usual statistical tests must be adjusted. Such variables still explain some of the variation.

Technical details of the approach used are:

- A coordinate descent algorithm was used,
- Continuous variables were standardized to have variance = 1,
- Continuous variables were transformed to center on zero,
- 10-fold cross-validation applied to help counter the possibility that a global minimum is not found, and
- Where interaction terms were included, both main effects were included as well.

The gMCP process was chosen after an attempt with LASSO (least absolute shrinkage and selection operator) had mixed results.

The principal diagnostics used in the gMCP process for selecting the best groupings to include were

- Bayesian Information Criterion (BIC)
- Akaike Information Criterion (AIC)
- Generalized cross-validation (GCV)
- 10-fold cross-validation error (CVE) (six different orders)

Appendix C  
Final Model Output, Model Fit and Diagnostics

Following are the computational outputs for the final selected regression models for the four car components. In all cases the computational procedure was SAS PROC LOGISTIC.

For the regressions, the following diagnostic guidelines were followed:

- The coefficients and associated odds ratios needed to be interpretable.
- The Hosmer-Lemeshow test for goodness-of-fit had to be passed.
- The concordance index and ROC curves had to indicate sufficient predictive accuracy.
- Outliers and high-influence points had to be sparse.
- Strong correlations among variables had to have been addressed by the gMCP process.
- Wald chi-square statistics were examined, even where the gMCP suggested retaining variables that failed a chi-square test

Outliers were retained in the dataset. The binary nature of the outcomes limited the influence of outliers, and the outlier records were not of the sort that could be rejected as errors. There are many variables that affect accident environments, and among thousands of records which account for some but not all of these factors, unusual but valid combinations of these factors will arise, even if rarely. Finally, outliers made up a small percentage of records anyway, and tended to be evenly distributed throughout the domain. Examination of the plots in this chapter supports these conclusions.

## Head

Model Information	
Data Set	WORK.HEADCPR
Response Variable	Head_Loss
Number of Response Levels	2
Model	binary logit
Optimization Technique	Fisher's scoring
Likelihood Penalty	Firth's bias correction

Number of Observations Read	4467
Number of Observations Used	4467

Response Profile		
Ordered Value	Head_Loss	Total Frequency
1	0	4286
2	1	181

*Probability modeled is Head\_Loss='1'.*

Class Level Information		
Class	Value	Design Variables
HSF	0	0
	1	1
HSHJ	0	0
	1	1
HSHNJ	0	0
	1	1
FJKT	0	0
	1	1
HTR	0	0
	1	1
ACCTYP	0	0
	1	1

Intercept-Only Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.



<b>Model Fit Statistics</b>		
<b>Criterion</b>	<b>Intercept Only</b>	<b>Intercept and Covariates</b>
<b>AIC</b>	1467.509	1308.031
<b>SC</b>	1473.914	1378.480
<b>-2 Log L</b>	1465.509	1286.031

<b>Testing Global Null Hypothesis: BETA=0</b>			
<b>Test</b>	<b>Chi-Square</b>	<b>DF</b>	<b>Pr &gt; ChiSq</b>
<b>Likelihood Ratio</b>	179.4788	10	<.0001
<b>Score</b>	174.8086	10	<.0001
<b>Wald</b>	141.6655	10	<.0001

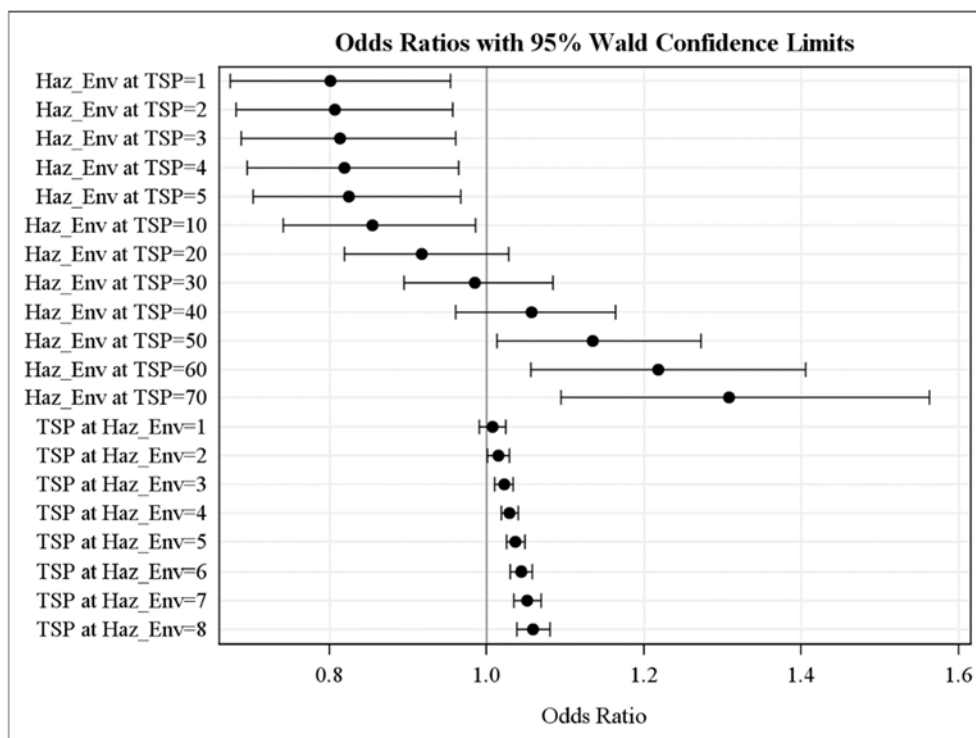
<b>Type 3 Analysis of Effects</b>			
<b>Effect</b>	<b>DF</b>	<b>Wald Chi-Square</b>	<b>Pr &gt; ChiSq</b>
<b>HSF</b>	1	21.0898	<.0001
<b>HSHJ</b>	1	11.2014	0.0008
<b>HSHNJ</b>	1	0.9603	0.3271
<b>FJKT</b>	1	20.4187	<.0001
<b>HMT</b>	1	0.0743	0.7851
<b>HTR</b>	1	2.0189	0.1554
<b>TSP</b>	1	0.0018	0.9664
<b>Haz_Env</b>	1	6.3502	0.0117
<b>ACCTYP</b>	1	8.2247	0.0041
<b>TSP*Haz_Env</b>	1	10.3836	0.0013

Analysis of Maximum Likelihood Estimates						
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept		1	-1.5145	0.6403	5.5954	0.0180
HSF	1	1	-3.0699	0.6685	21.0898	<.0001
HSHJ	1	1	-2.8387	0.8482	11.2014	0.0008
HSHNJ	1	1	-0.6773	0.6911	0.9603	0.3271
FJKT	1	1	-1.6871	0.3734	20.4187	<.0001
HMT		1	-0.3447	1.2641	0.0743	0.7851
HTR	1	1	0.5572	0.3921	2.0189	0.1554
TSP		1	0.000433	0.0103	0.0018	0.9664
Haz_Env		1	-0.2281	0.0905	6.3502	0.0117
ACCTYP	1	1	-0.7613	0.2655	8.2247	0.0041
TSP*Haz_Env		1	0.00709	0.00220	10.3836	0.0013

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	77.5	Somers' D	0.551
Percent Discordant	22.4	Gamma	0.552
Percent Tied	0.1	Tau-a	0.043
Pairs	775766	c	0.776

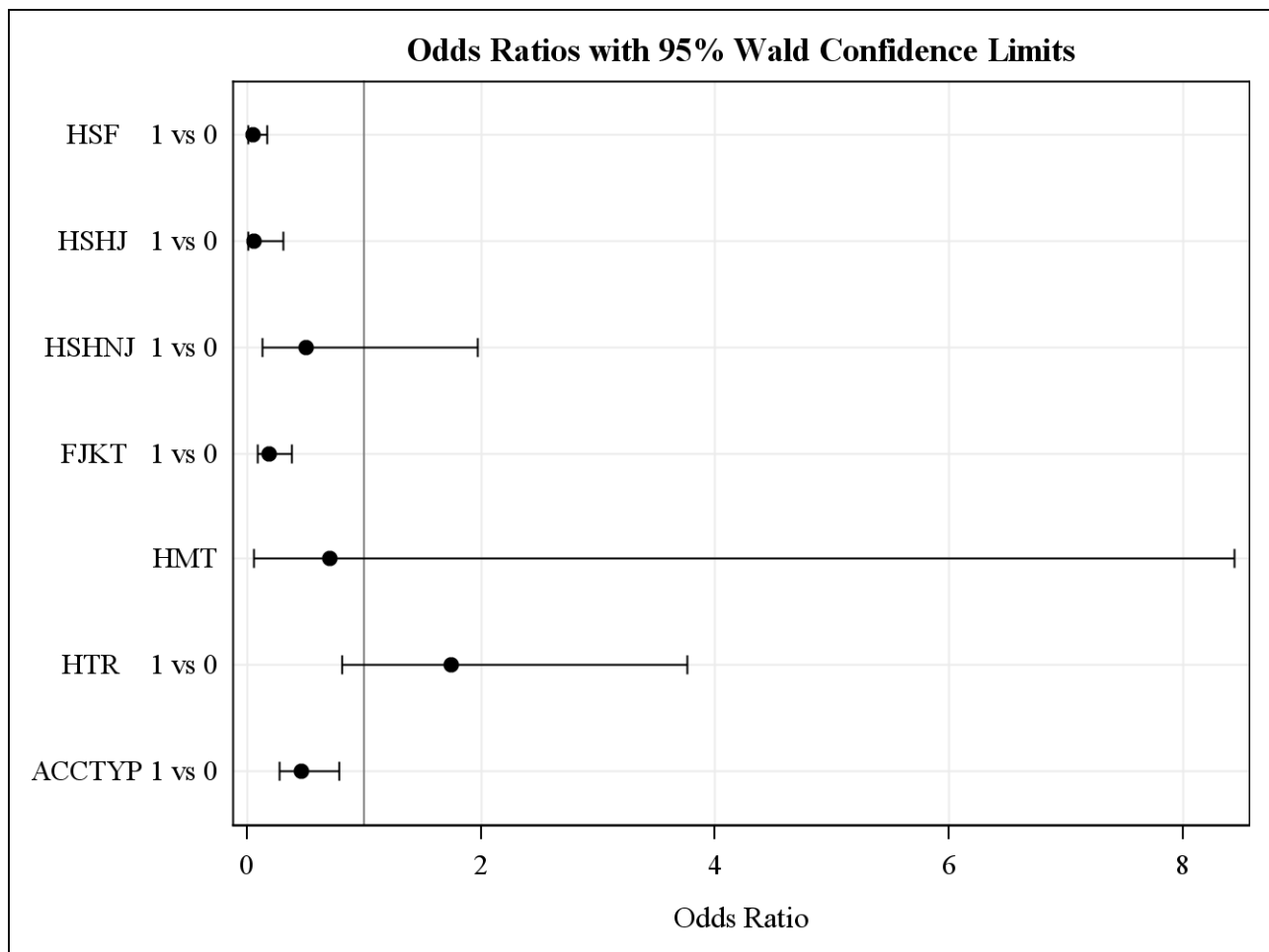
Odds Ratio Estimates and Wald Confidence Intervals			
Label	Estimate	95% Confidence Limits	
Haz_Env at TSP=1	0.802	0.674	0.954
Haz_Env at TSP=2	0.807	0.681	0.957
Haz_Env at TSP=3	0.813	0.688	0.961
Haz_Env at TSP=4	0.819	0.696	0.964
Haz_Env at TSP=5	0.825	0.703	0.967

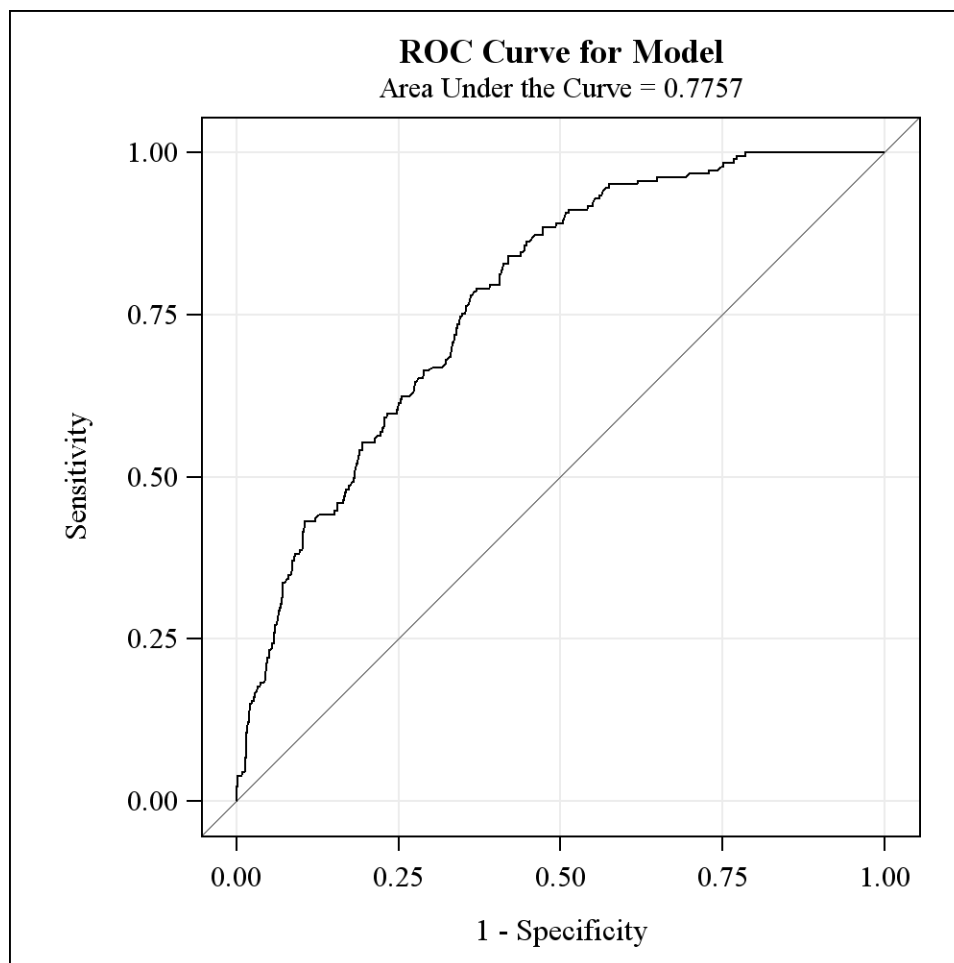
<b>Odds Ratio Estimates and Wald Confidence Intervals</b>			
<b>Label</b>	<b>Estimate</b>	<b>95% Confidence Limits</b>	
Haz_Env at TSP=10	0.855	0.741	0.986
Haz_Env at TSP=20	0.917	0.819	1.028
Haz_Env at TSP=30	0.985	0.895	1.084
Haz_Env at TSP=40	1.057	0.960	1.164
Haz_Env at TSP=50	1.135	1.013	1.272
Haz_Env at TSP=60	1.218	1.056	1.406
Haz_Env at TSP=70	1.308	1.095	1.562
TSP at Haz_Env=1	1.008	0.991	1.024
TSP at Haz_Env=2	1.015	1.001	1.028
TSP at Haz_Env=3	1.022	1.011	1.033
TSP at Haz_Env=4	1.029	1.019	1.040
TSP at Haz_Env=5	1.037	1.025	1.048
TSP at Haz_Env=6	1.044	1.030	1.058
TSP at Haz_Env=7	1.051	1.034	1.069
TSP at Haz_Env=8	1.059	1.038	1.080



Parameter Estimates and Wald Confidence Intervals				
Parameter		Estimate	95% Confidence Limits	
Intercept		-1.5145	-2.7693	-0.2596
H <sub>SF</sub>	1	-3.0699	-4.3801	-1.7597
H <sub>SHJ</sub>	1	-2.8387	-4.5011	-1.1763
H <sub>SHNJ</sub>	1	-0.6773	-2.0318	0.6773
F <sub>JKT</sub>	1	-1.6871	-2.4188	-0.9553
H <sub>MT</sub>		-0.3447	-2.8223	2.1329
H <sub>TR</sub>	1	0.5572	-0.2114	1.3258
TSP		0.00043	-0.0197	0.0206
		3		
Haz_Env		-0.2281	-0.4055	-0.0507
ACCTYP	1	-0.7613	-1.2817	-0.2410
TSP*Haz_Env		0.00709	0.00278	0.0114

Odds Ratio Estimates and Wald Confidence Intervals				
Effect	Unit	Estimate	95% Confidence Limits	
<b>HSF 1 vs 0</b>	1.000	0.046	0.013	0.172
<b>HSHJ 1 vs 0</b>	1.000	0.059	0.011	0.308
<b>HSHNJ 1 vs 0</b>	1.000	0.508	0.131	1.969
<b>FJKT 1 vs 0</b>	1.000	0.185	0.089	0.385
<b>HMT</b>	1.000	0.708	0.059	8.440
<b>HTR 1 vs 0</b>	1.000	1.746	0.809	3.765
<b>ACCTYP 1 vs 0</b>	1.000	0.467	0.278	0.786



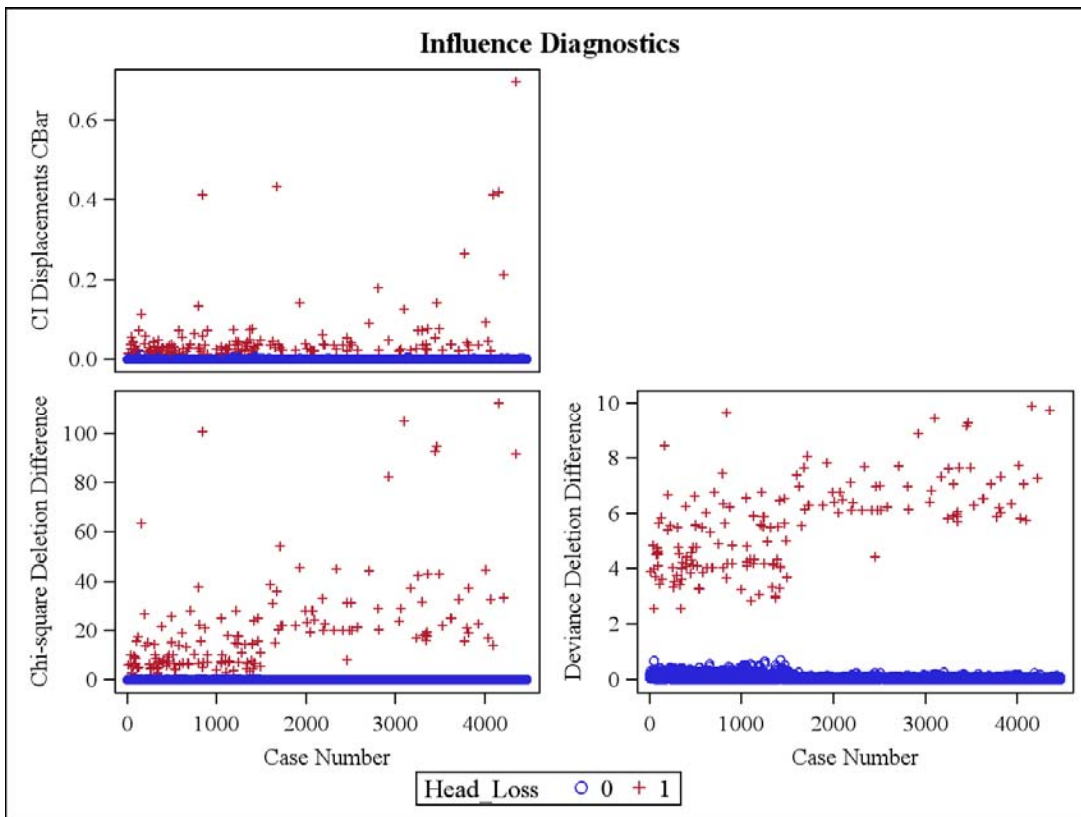
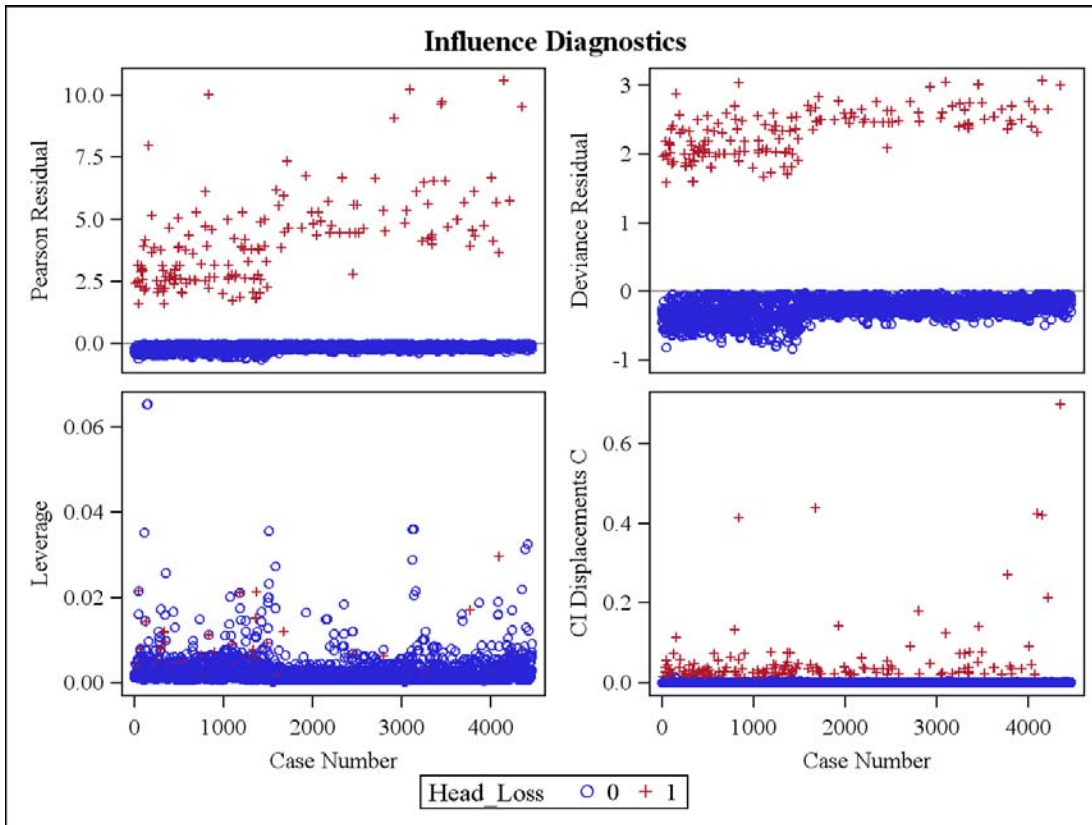


Estimated Correlation Matrix								
Parameter	Intercept	HSF1	HSHJ1	HSHNJ1	FJKT1	HMT	HTR1	TSP
Intercept	1.0000	0.2361	0.2025	0.2119	0.2660	-0.9180	-0.3742	-0.1869
HSF1	0.2361	1.0000	0.0934	0.0996	0.1628	-0.2982	-0.1381	0.0386
HSHJ1	0.2025	0.0934	1.0000	0.0819	0.1223	-0.2496	-0.1029	0.0268
HSHNJ1	0.2119	0.0996	0.0819	1.0000	0.1511	-0.2572	-0.1276	0.0122
FJKT1	0.2660	0.1628	0.1223	0.1511	1.0000	-0.3468	-0.9020	0.0239
HMT	-0.9180	-0.2982	-0.2496	-0.2572	-0.3468	1.0000	0.4014	-0.0604
HTR1	-0.3742	-0.1381	-0.1029	-0.1276	-0.9020	0.4014	1.0000	-0.0179
TSP	-0.1869	0.0386	0.0268	0.0122	0.0239	-0.0604	-0.0179	1.0000
Haz_Env	-0.1858	0.0456	0.0266	0.0144	0.0511	-0.0785	-0.0237	0.6047
ACCTYP1	-0.0162	-0.0111	-0.0042	0.0021	-0.0256	0.0434	0.0244	-0.4457
TSPHaz_Env	0.1810	-0.0437	-0.0281	-0.0169	-0.0342	0.0800	0.0239	-0.8600

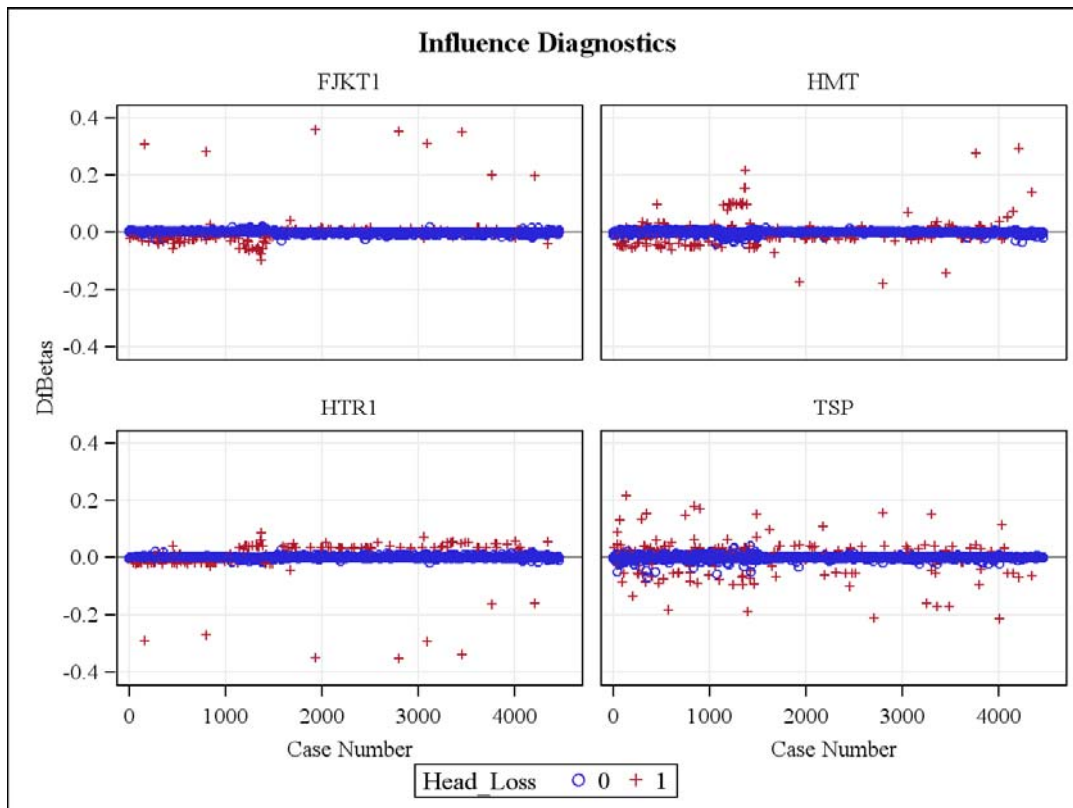
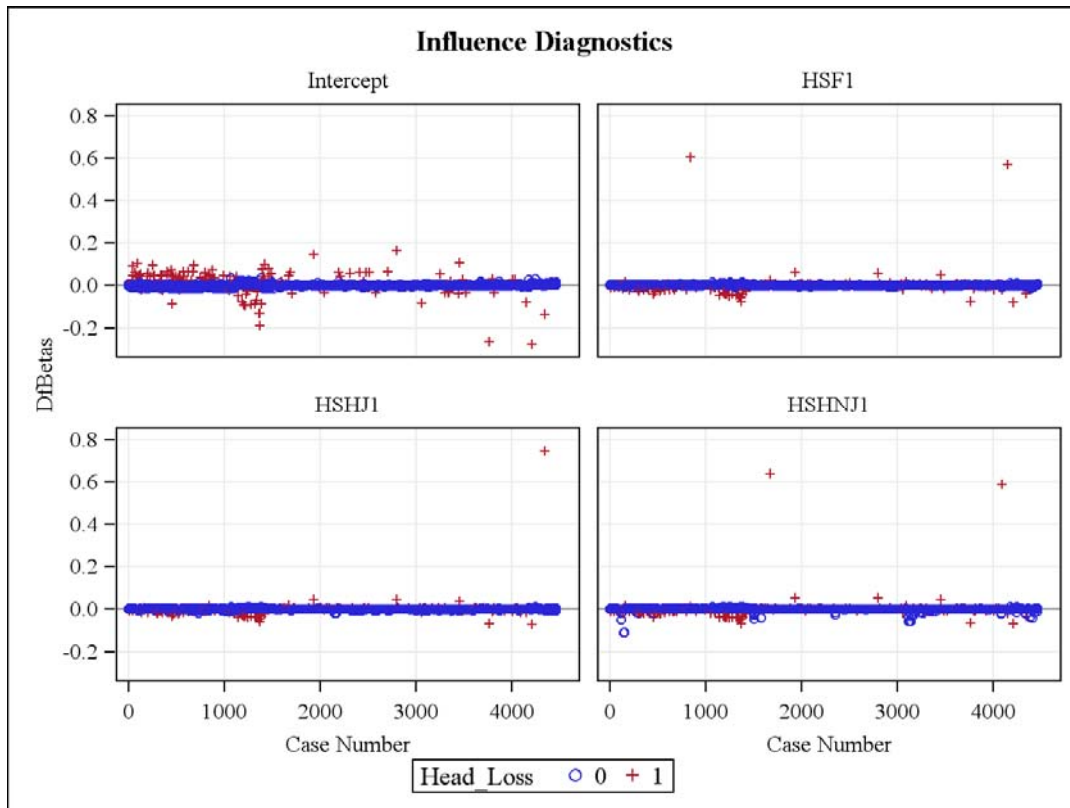
Estimated Correlation Matrix			
Parameter	Haz_Env	ACCTYP1	TSPHaz_Env
Intercept	-0.1858	-0.0162	0.1810
HSF1	0.0456	-0.0111	-0.0437
HSHJ1	0.0266	-0.0042	-0.0281
HSHNJ1	0.0144	0.0021	-0.0169
FJKT1	0.0511	-0.0256	-0.0342
HMT	-0.0785	0.0434	0.0800
HTR1	-0.0237	0.0244	0.0239
TSP	0.6047	-0.4457	-0.8600
Haz_Env	1.0000	-0.5980	-0.8494
ACCTYP1	-0.5980	1.0000	0.4361
TSPHaz_Env	-0.8494	0.4361	1.0000

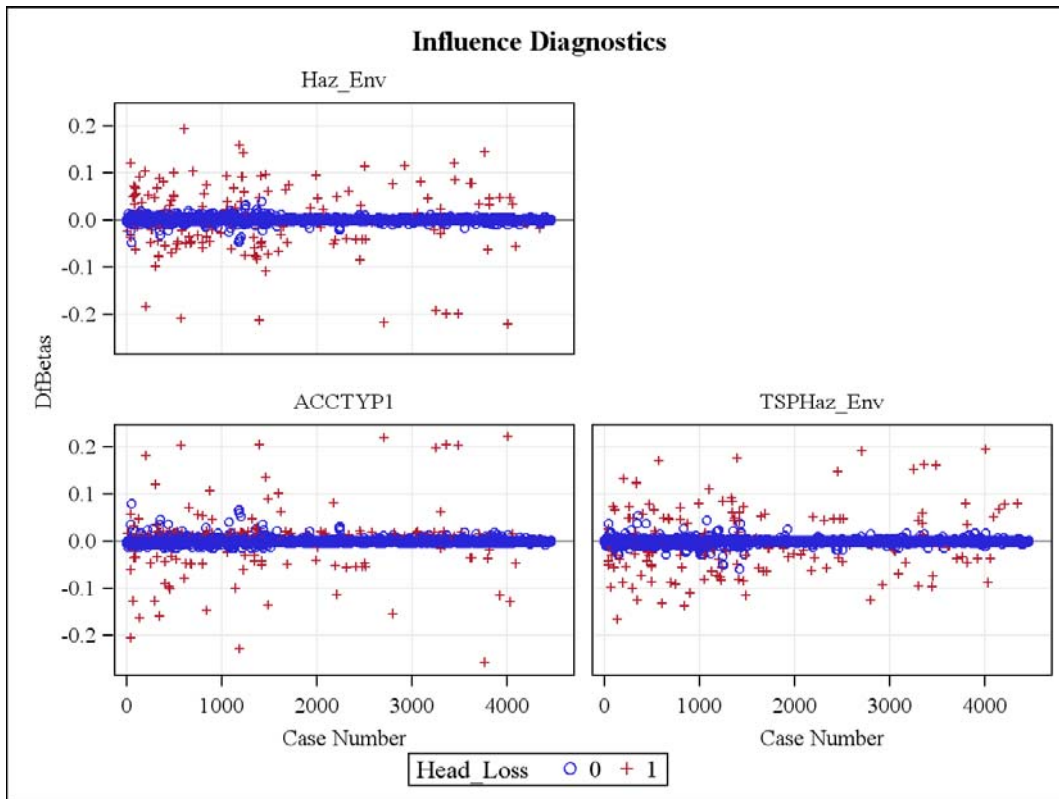
Partition for the Hosmer and Lemeshow Test					
Group	Total	Head_Loss = 1		Head_Loss = 0	
		Observed	Expected	Observed	Expected
1	445	0	1.07	445	443.93
2	447	0	2.76	447	444.24
3	448	7	4.74	441	443.26
4	452	2	7.50	450	444.50
5	449	12	10.86	437	438.14
6	438	17	14.19	421	423.81
7	448	25	19.14	423	428.86
8	447	24	23.67	423	423.33
9	447	27	35.12	420	411.88
10	446	67	66.73	379	379.27

Hosmer and Lemeshow Goodness-of-Fit Test		
Chi-Square	DF	Pr > ChiSq
13.6615	8	0.0910









## Shell

Model Information	
Data Set	WORK.SHELL_CPR
Response Variable	Shell_Loss
Number of Response Levels	2
Model	binary logit
Optimization Technique	Fisher's scoring
Likelihood Penalty	Firth's bias correction

<b>Number of Observations Read</b>	7165
<b>Number of Observations Used</b>	7165

<b>Response Profile</b>		
<b>Ordered Value</b>	<b>Shell_Loss</b>	<b>Total Frequency</b>
1	0	6815
2	1	350

*Probability modeled is Shell\_Loss='1'.*

<b>Class Level Information</b>						
<b>Class</b>	<b>Value</b>	<b>Design Variables</b>				
<b>Shell_Lad_Grp</b>	<b>A</b>	0	0	0	0	0
	<b>D</b>	1	0	0	0	0
	<b>G</b>	0	1	0	0	0
	<b>H</b>	0	0	1	0	0
	<b>L</b>	0	0	0	1	0
	<b>O</b>	0	0	0	0	1

<b>Intercept-Only Model Convergence Status</b>
Convergence criterion (GCONV=1E-8) satisfied.

<b>Model Convergence Status</b>
Convergence criterion (GCONV=1E-8) satisfied.

<b>Model Fit Statistics</b>		
<b>Criterion</b>	<b>Intercept Only</b>	<b>Intercept and Covariates</b>
<b>AIC</b>	2648.441	1999.073
<b>SC</b>	2655.318	2122.858
<b>-2 Log L</b>	2646.441	1963.073

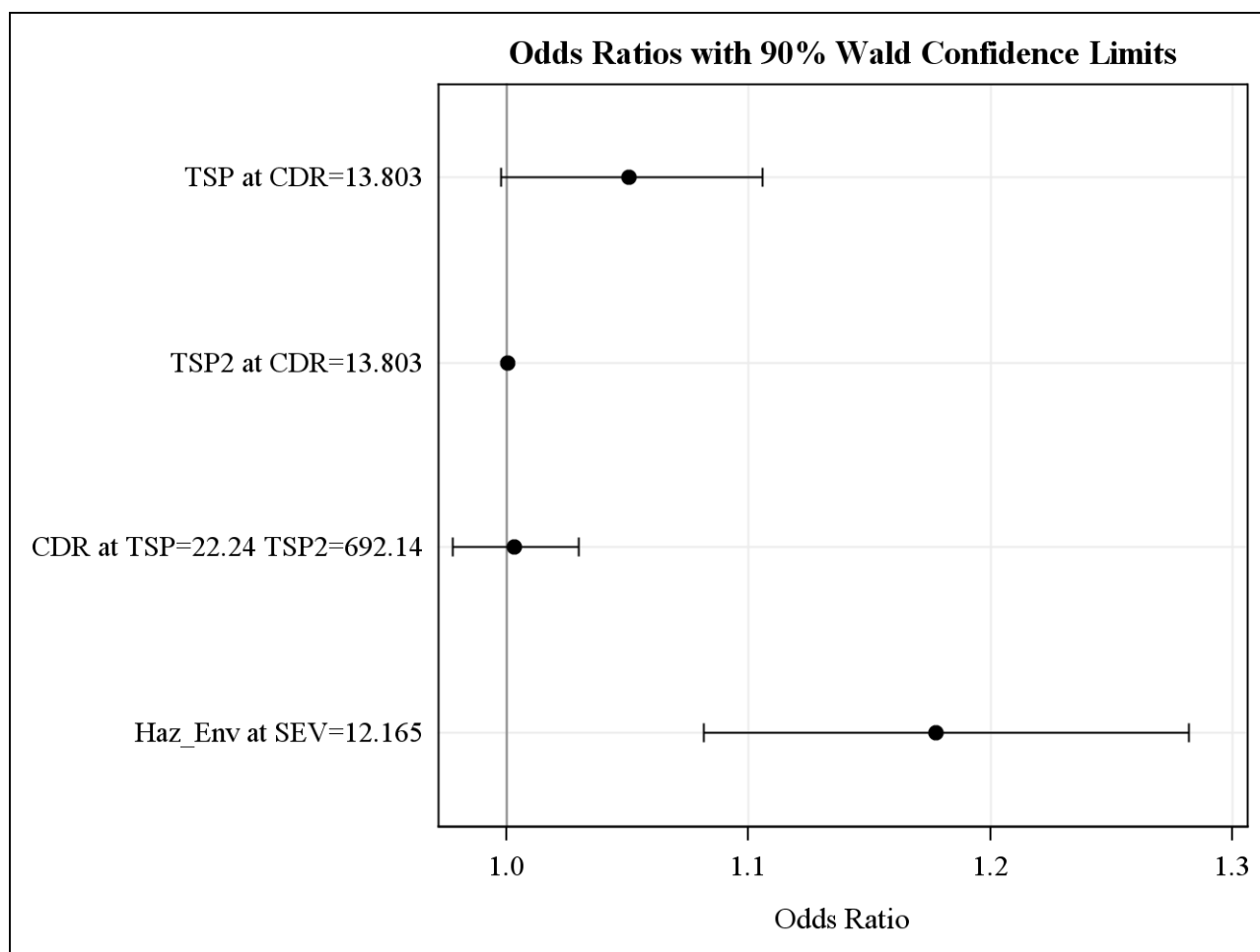
<b>Testing Global Null Hypothesis: BETA=0</b>			
<b>Test</b>	<b>Chi-Square</b>	<b>DF</b>	<b>Pr &gt; ChiSq</b>
<b>Likelihood Ratio</b>	683.3684	17	<.0001
<b>Score</b>	926.4661	17	<.0001
<b>Wald</b>	561.1429	17	<.0001

<b>Type 3 Analysis of Effects</b>			
<b>Effect</b>	<b>DF</b>	<b>Wald Chi-Square</b>	<b>Pr &gt; ChiSq</b>
<b>STS</b>	1	21.0475	<.0001
<b>JKT</b>	1	53.9564	<.0001
<b>SID</b>	1	25.1491	<.0001
<b>TSP</b>	1	2.5472	0.1105
<b>TSP2</b>	1	8.2004	0.0042
<b>SEV</b>	1	0.4567	0.4992
<b>SEV2</b>	1	0.2924	0.5887
<b>CDR</b>	1	6.1236	0.0133
<b>Haz_Env</b>	1	0.8297	0.3624
<b>TSP*CDR</b>	1	19.3758	<.0001
<b>TSP2*CDR</b>	1	20.9714	<.0001
<b>SEV*Haz_Env</b>	1	5.2947	0.0214
<b>Shell_Lad_Grp</b>	5	37.1427	<.0001

Analysis of Maximum Likelihood Estimates						
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept		1	5.6289	1.3984	16.2030	<.0001
STS		1	-5.2092	1.1355	21.0475	<.0001
JKT		1	-1.0986	0.1496	53.9564	<.0001
SID		1	-0.0588	0.0117	25.1491	<.0001
TSP		1	-0.0514	0.0322	2.5472	0.1105
TSP2		1	0.00147	0.000514	8.2004	0.0042
SEV		1	-0.0254	0.0376	0.4567	0.4992
SEV2		1	-0.00032	0.000594	0.2924	0.5887
CDR		1	-0.0953	0.0385	6.1236	0.0133
Haz_Env		1	0.0739	0.0811	0.8297	0.3624
TSP*CDR		1	0.00729	0.00166	19.3758	<.0001
TSP2*CDR		1	-0.00009	0.000020	20.9714	<.0001
SEV*Haz_Env		1	0.00736	0.00320	5.2947	0.0214
Shell_Lad_Grp	D	1	-0.7976	0.3437	5.3845	0.0203
Shell_Lad_Grp	G	1	-0.9089	0.2699	11.3439	0.0008
Shell_Lad_Grp	H	1	-0.4355	0.2406	3.2760	0.0703
Shell_Lad_Grp	L	1	1.4745	0.3781	15.2044	<.0001
Shell_Lad_Grp	O	1	-1.5734	0.6590	5.6993	0.0170

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	86.0	Somers' D	0.721
Percent Discordant	13.9	Gamma	0.721
Percent Tied	0.0	Tau-a	0.067
Pairs	2385250	c	0.860

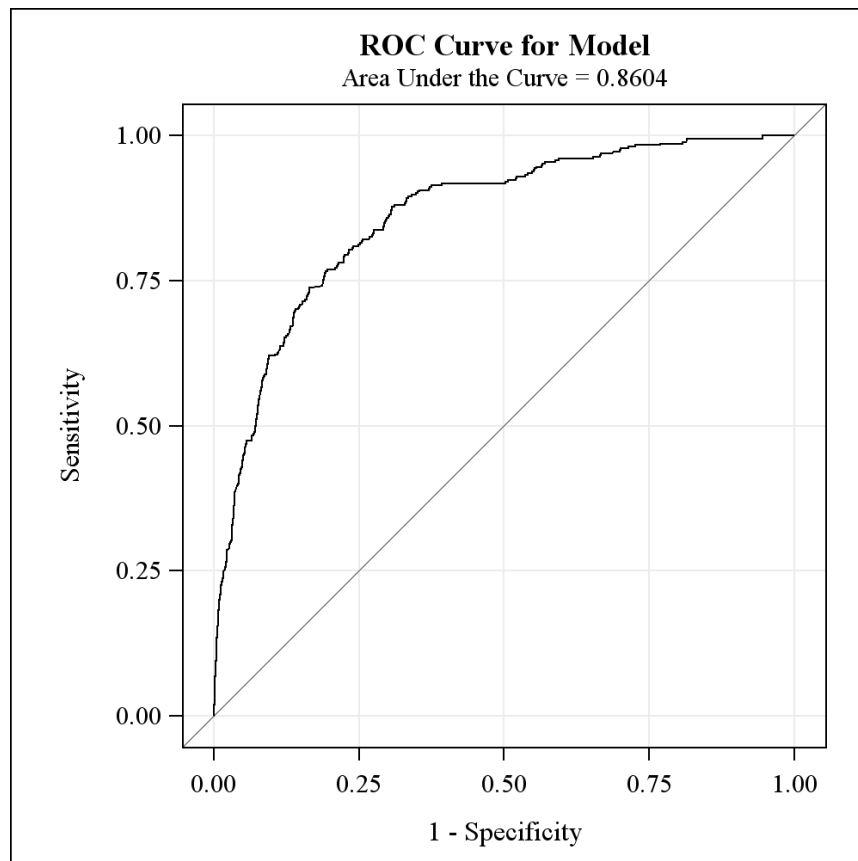
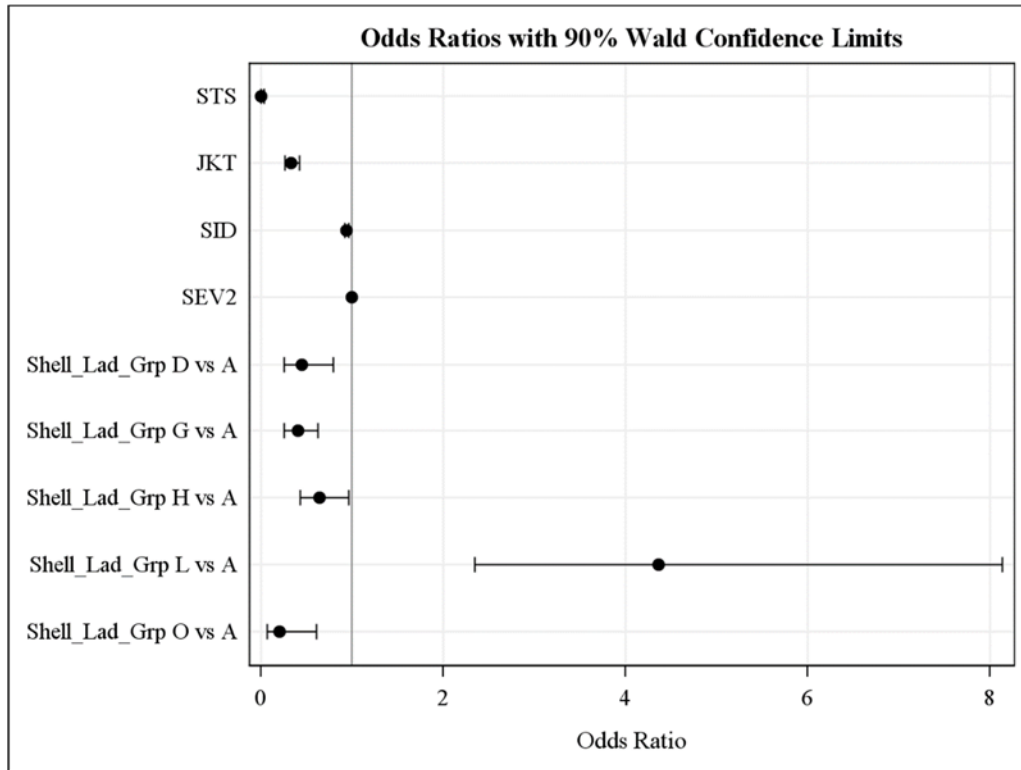
Odds Ratio Estimates and Wald Confidence Intervals			
Label	Estimate	90% Confidence Limits	
TSP at CDR=13.803	1.050	0.998	1.106
TSP2 at CDR=13.803	1.000	0.999	1.001
CDR at TSP=22.24 TSP2=692.14	1.003	0.977	1.030
Haz_Env at SEV=12.165	1.177	1.081	1.282



Parameter Estimates and Wald Confidence Intervals			
Parameter	Estimate	90% Confidence Limits	
Intercept	5.6289	3.3288	7.9290
STS	-5.2092	-7.0769	-3.3416
JKT	-1.0986	-1.3446	-0.8526

<b>Parameter Estimates and Wald Confidence Intervals</b>				
<b>Parameter</b>		<b>Estimate</b>	<b>90% Confidence Limits</b>	
<b>SID</b>		-0.0588	-0.0781	-0.0395
<b>TSP</b>		-0.0514	-0.1045	0.00157
<b>TSP2</b>		0.00147	0.000627	0.00232
<b>SEV</b>		-0.0254	-0.0873	0.0365
<b>SEV2</b>		-0.00032	-0.00130	0.000656
<b>CDR</b>		-0.0953	-0.1587	-0.0320
<b>Haz_Env</b>		0.0739	-0.0595	0.2073
<b>TSP*CDR</b>		0.00729	0.00457	0.0100
<b>TSP2*CDR</b>		-0.00009	-0.00012	-0.00006
<b>SEV*Haz_Env</b>		0.00736	0.00210	0.0126
<b>Shell_Lad_Grp D</b>	<b>D</b>	-0.7976	-1.3629	-0.2322
<b>Shell_Lad_Grp G</b>	<b>G</b>	-0.9089	-1.3528	-0.4650
<b>Shell_Lad_Grp H</b>	<b>H</b>	-0.4355	-0.8312	-0.0397
<b>Shell_Lad_Grp L</b>	<b>L</b>	1.4745	0.8525	2.0965
<b>Shell_Lad_Grp O</b>	<b>O</b>	-1.5734	-2.6574	-0.4893

<b>Odds Ratio Estimates and Wald Confidence Intervals</b>				
<b>Effect</b>	<b>Unit</b>	<b>Estimate</b>	<b>90% Confidence Limits</b>	
<b>STS</b>	1.000	0.005	<0.001	0.035
<b>JKT</b>	1.000	0.333	0.261	0.426
<b>SID</b>	1.000	0.943	0.925	0.961
<b>SEV2</b>	1.000	1.000	0.999	1.001
<b>Shell_Lad_Grp D vs A</b>	1.000	0.450	0.256	0.793
<b>Shell_Lad_Grp G vs A</b>	1.000	0.403	0.259	0.628
<b>Shell_Lad_Grp H vs A</b>	1.000	0.647	0.436	0.961
<b>Shell_Lad_Grp L vs A</b>	1.000	4.369	2.345	8.137
<b>Shell_Lad_Grp O vs A</b>	1.000	0.207	0.070	0.613





Estimated Correlation Matrix										
Parameter	Intercept	STS	JKT	SID	TSP	TSP2	SEV	SEV2	CDR	Haz_Env
Intercept	1.0000	-0.2336	-0.3312	-0.8920	-0.0189	-0.0021	-0.1754	0.1426	-0.1402	-0.0919
STS	-0.2336	1.0000	-0.2434	-0.1353	-0.0069	0.0020	0.0199	-0.0121	-0.0155	0.0183
JKT	-0.3312	-0.2434	1.0000	0.3924	0.0329	-0.0241	-0.0086	0.0274	-0.0218	0.0256
SID	-0.8920	-0.1353	0.3924	1.0000	-0.0593	0.0899	0.0735	-0.1012	0.0823	-0.0527
TSP	-0.0189	-0.0069	0.0329	-0.0593	1.0000	-0.9264	-0.5110	0.3930	0.1271	-0.3800
TSP2	-0.0021	0.0020	-0.0241	0.0899	-0.9264	1.0000	0.2946	-0.2421	0.0560	0.2192
SEV	-0.1754	0.0199	-0.0086	0.0735	-0.5110	0.2946	1.0000	-0.8387	-0.1324	0.4716
SEV2	0.1426	-0.0121	0.0274	-0.1012	0.3930	-0.2421	-0.8387	1.0000	0.0153	-0.0677
CDR	-0.1402	-0.0155	-0.0218	0.0823	0.1271	0.0560	-0.1324	0.0153	1.0000	-0.3333
Haz_Env	-0.0919	0.0183	0.0256	-0.0527	-0.3800	0.2192	0.4716	-0.0677	-0.3333	1.0000
TSPCDR	0.1304	0.0185	0.0024	-0.0426	-0.3973	0.2718	0.1215	0.0007	-0.9110	0.3333
TSP2CDR	-0.0961	-0.0180	0.0012	-0.0064	0.5720	-0.5452	-0.0831	-0.0255	0.6980	-0.2737
SEVHaz_Env	0.1416	-0.0293	-0.0336	-0.0069	0.3323	-0.2047	-0.5512	0.0921	0.2817	-0.8580
Shell_Lad_GrpD	-0.3216	-0.4743	0.4515	0.5160	-0.0127	0.0291	0.0249	-0.0385	-0.0220	-0.0431
Shell_Lad_GrpG	-0.3307	0.0713	-0.0868	0.3307	0.0232	-0.0279	0.0348	-0.0545	-0.0348	-0.0245
Shell_Lad_GrpH	0.0685	0.1451	-0.2567	-0.1235	-0.0338	0.0439	0.0059	-0.0296	0.0206	-0.0179
Shell_Lad_GrpL	0.3594	-0.3226	-0.1659	-0.2488	0.0210	-0.0159	-0.0262	0.0011	0.0465	-0.0801
Shell_Lad_GrpO	-0.1304	0.0310	-0.0468	0.1274	-0.0006	-0.0036	-0.0064	-0.0031	-0.0013	-0.0051

Estimated Correlation Matrix					
Parameter	TSPCDR	TSP2CDR	SEVHaz_Env	Shell_Lad_GrpD	Shell_Lad_GrpG
Intercept	0.1304	-0.0961	0.1416	-0.3216	-0.3307
STS	0.0185	-0.0180	-0.0293	-0.4743	0.0713
JKT	0.0024	0.0012	-0.0336	0.4515	-0.0868
SID	-0.0426	-0.0064	-0.0069	0.5160	0.3307
TSP	-0.3973	0.5720	0.3323	-0.0127	0.0232
TSP2	0.2718	-0.5452	-0.2047	0.0291	-0.0279
SEV	0.1215	-0.0831	-0.5512	0.0249	0.0348
SEV2	0.0007	-0.0255	0.0921	-0.0385	-0.0545

Estimated Correlation Matrix					
Parameter	TSPCDR	TSP2CDR	SEVHaz_Env	Shell_Lad_GrpD	Shell_Lad_GrpG
CDR	-0.9110	0.6980	0.2817	-0.0220	-0.0348
Haz_Env	0.3333	-0.2737	-0.8580	-0.0431	-0.0245
TSPCDR	1.0000	-0.9238	-0.2838	0.0122	-0.0247
TSP2CDR	-0.9238	1.0000	0.2326	-0.0210	0.0576
SEVHaz_Env	-0.2838	0.2326	1.0000	0.0101	0.0125
Shell_Lad_GrpD	0.0122	-0.0210	0.0101	1.0000	0.1536
Shell_Lad_GrpG	-0.0247	0.0576	0.0125	0.1536	1.0000
Shell_Lad_GrpH	-0.0138	-0.0018	0.0267	-0.0926	0.0972
Shell_Lad_GrpL	-0.0629	0.0686	0.0719	0.0306	-0.0349
Shell_Lad_GrpO	0.0071	-0.0050	0.0142	0.0451	0.0949

Estimated Correlation Matrix			
Parameter	Shell_Lad_GrpH	Shell_Lad_GrpL	Shell_Lad_GrpO
Intercept	0.0685	0.3594	-0.1304
STS	0.1451	-0.3226	0.0310
JKT	-0.2567	-0.1659	-0.0468
SID	-0.1235	-0.2488	0.1274
TSP	-0.0338	0.0210	-0.0006
TSP2	0.0439	-0.0159	-0.0036
SEV	0.0059	-0.0262	-0.0064
SEV2	-0.0296	0.0011	-0.0031
CDR	0.0206	0.0465	-0.0013
Haz_Env	-0.0179	-0.0801	-0.0051
TSPCDR	-0.0138	-0.0629	0.0071
TSP2CDR	-0.0018	0.0686	-0.0050
SEVHaz_Env	0.0267	0.0719	0.0142
Shell_Lad_GrpD	-0.0926	0.0306	0.0451
Shell_Lad_GrpG	0.0972	-0.0349	0.0949

**Estimated Correlation Matrix**

Parameter	Shell_Lad_GrpH	Shell_Lad_GrpL	Shell_Lad_GrpO
Shell_Lad_GrpH	1.0000	0.0707	0.0367
Shell_Lad_GrpL	0.0707	1.0000	-0.0187
Shell_Lad_GrpO	0.0367	-0.0187	1.0000

**Partition for the Hosmer and Lemeshow Test**

Group	Total	Shell_Loss = 1		Shell_Loss = 0	
		Observed	Expected	Observed	Expected
1	711	2	2.09	709	708.91
2	717	3	4.37	714	712.63
3	719	6	6.74	713	712.26
4	681	5	8.78	676	672.22
5	704	13	11.51	691	692.49
6	718	1	14.43	717	703.57
7	717	27	18.75	690	698.25
8	668	34	27.96	634	640.04
9	717	53	53.76	664	663.24
10	813	206	207.71	607	605.29

**Hosmer and Lemeshow  
Goodness-of-Fit Test**

Chi-Square	DF	Pr > ChiSq
20.2341	8	0.0095

Classification Table									
Prob Level	Correct		Incorrect		Percentages				
	Event	Non-Event	Event	Non-Event	Correct	Sensitivity	Specificity	False POS	False NEG
0.000	350	0	6815	0	4.9	100.0	0.0	95.1	.
0.020	320	4082	2733	30	61.4	91.4	59.9	89.5	0.7
0.040	275	5211	1604	75	76.6	78.6	76.5	85.4	1.4
0.060	244	5724	1091	106	83.3	69.7	84.0	81.7	1.8
0.080	227	5949	866	123	86.2	64.9	87.3	79.2	2.0
0.100	212	6160	655	138	88.9	60.6	90.4	75.5	2.2
0.120	194	6243	572	156	89.8	55.4	91.6	74.7	2.4
0.140	173	6313	502	177	90.5	49.4	92.6	74.4	2.7
0.160	166	6392	423	184	91.5	47.4	93.8	71.8	2.8
0.180	153	6442	373	197	92.0	43.7	94.5	70.9	3.0
0.200	146	6478	337	204	92.4	41.7	95.1	69.8	3.1
0.220	135	6525	290	215	93.0	38.6	95.7	68.2	3.2
0.240	104	6611	204	246	93.7	29.7	97.0	66.2	3.6
0.260	94	6652	163	256	94.2	26.9	97.6	63.4	3.7
0.280	86	6677	138	264	94.4	24.6	98.0	61.6	3.8
0.300	82	6701	114	268	94.7	23.4	98.3	58.2	3.8
0.320	79	6716	99	271	94.8	22.6	98.5	55.6	3.9
0.340	79	6726	89	271	95.0	22.6	98.7	53.0	3.9
0.360	74	6731	84	276	95.0	21.1	98.8	53.2	3.9
0.380	72	6735	80	278	95.0	20.6	98.8	52.6	4.0
0.400	69	6744	71	281	95.1	19.7	99.0	50.7	4.0
0.420	62	6753	62	288	95.1	17.7	99.1	50.0	4.1
0.440	53	6765	50	297	95.2	15.1	99.3	48.5	4.2
0.460	48	6773	42	302	95.2	13.7	99.4	46.7	4.3
0.480	43	6777	38	307	95.2	12.3	99.4	46.9	4.3
0.500	36	6785	30	314	95.2	10.3	99.6	45.5	4.4

Classification Table									
Prob Level	Correct		Incorrect		Percentages				
	Event	Non-Event	Event	Non-Event	Correct	Sensitivity	Specificity	False POS	False NEG
0.520	31	6787	28	319	95.2	8.9	99.6	47.5	4.5
0.540	26	6791	24	324	95.1	7.4	99.6	48.0	4.6
0.560	20	6797	18	330	95.1	5.7	99.7	47.4	4.6
0.580	12	6800	15	338	95.1	3.4	99.8	55.6	4.7
0.600	9	6803	12	341	95.1	2.6	99.8	57.1	4.8
0.620	4	6808	7	346	95.1	1.1	99.9	63.6	4.8
0.640	3	6808	7	347	95.1	0.9	99.9	70.0	4.8
0.660	2	6810	5	348	95.1	0.6	99.9	71.4	4.9
0.680	1	6812	3	349	95.1	0.3	100.0	75.0	4.9
0.700	1	6812	3	349	95.1	0.3	100.0	75.0	4.9
0.720	1	6812	3	349	95.1	0.3	100.0	75.0	4.9
0.740	0	6813	2	350	95.1	0.0	100.0	100.0	4.9
0.760	0	6814	1	350	95.1	0.0	100.0	100.0	4.9
0.780	0	6814	1	350	95.1	0.0	100.0	100.0	4.9
0.800	0	6814	1	350	95.1	0.0	100.0	100.0	4.9
0.820	0	6814	1	350	95.1	0.0	100.0	100.0	4.9
0.840	0	6814	1	350	95.1	0.0	100.0	100.0	4.9
0.860	0	6815	0	350	95.1	0.0	100.0	.	4.9

## Top Fittings

Model Information	
Data Set	WORK.TFCPR
Response Variable	TF_Loss
Number of Response Levels	2
Model	binary logit
Optimization Technique	Fisher's scoring
Likelihood Penalty	Firth's bias correction

<b>Number of Observations Read</b>	4467
<b>Number of Observations Used</b>	4467

<b>Response Profile</b>		
<b>Ordered Value</b>	<b>TF_Loss</b>	<b>Total Frequency</b>
1	0	3946
2	1	521

*Probability modeled is TF\_Loss='1'.*

<b>Class Level Information</b>												
<b>Class</b>	<b>Value</b>	<b>Design Variables</b>										
<b>SMS</b>	128	0	0									
	515	1	0									
	516	0	1									
<b>JKT</b>	0	0										
	1	1										
<b>TFT</b>	0	0										
	1	1										
<b>ACCTYP</b>	0	0										
	1	1										
<b>TF_Lading_Grp</b>	<b>A</b>	0	0	0	0	0	0	0	0	0	0	0
	<b>B</b>	1	0	0	0	0	0	0	0	0	0	0
	<b>D</b>	0	1	0	0	0	0	0	0	0	0	0
	<b>E</b>	0	0	1	0	0	0	0	0	0	0	0
	<b>F</b>	0	0	0	1	0	0	0	0	0	0	0
	<b>G</b>	0	0	0	0	1	0	0	0	0	0	0
	<b>H</b>	0	0	0	0	0	1	0	0	0	0	0
	<b>K</b>	0	0	0	0	0	0	1	0	0	0	0
	<b>M</b>	0	0	0	0	0	0	0	1	0	0	0

Class Level Information													
Class	Value	Design Variables											
	<b>O</b>	0	0	0	0	0	0	0	0	0	1	0	0
	<b>Q</b>	0	0	0	0	0	0	0	0	0	0	1	0
	<b>T</b>	0	0	0	0	0	0	0	0	0	0	0	1

Intercept-Only Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics		
Criterion	Intercept Only	Intercept and Covariates
AIC	3040.970	2535.692
SC	3047.375	2702.208
-2 Log L	3038.970	2483.692

Testing Global Null Hypothesis: BETA=0			
Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	555.2789	25	<.0001
Score	503.7059	25	<.0001
Wald	353.1420	25	<.0001

<b>Type 3 Analysis of Effects</b>			
<b>Effect</b>	<b>DF</b>	<b>Wald Chi-Square</b>	<b>Pr &gt; ChiSq</b>
SMS	2	22.1531	<.0001
STS	1	1.9989	0.1574
JKT	1	17.9973	<.0001
SID	1	1.7225	0.1894
TFT	1	4.1741	0.0410
ACCTYP	1	13.7555	0.0002
SEV2	1	8.1144	0.0044
TSP	1	1.1729	0.2788
TSP*CDR	1	3.0099	0.0828
TSP2	1	3.9557	0.0467
CDR*TSP2	1	3.6516	0.0560
Haz_Env	1	1.5996	0.2060
TSP2*Haz_Env	1	1.5749	0.2095
TF_Lading_Grp	11	114.8696	<.0001

<b>Analysis of Maximum Likelihood Estimates</b>						
<b>Parameter</b>		<b>DF</b>	<b>Estimate</b>	<b>Standard Error</b>	<b>Wald Chi-Square</b>	<b>Pr &gt; ChiSq</b>
Intercept		1	0.5555	1.9739	0.0792	0.7784
SMS	515	1	0.6168	0.2007	9.4462	0.0021
SMS	516	1	0.0675	0.1958	0.1191	0.7300
STS		1	-2.7467	1.9427	1.9989	0.1574
JKT	1	1	-0.5928	0.1397	17.9973	<.0001
SID		1	-0.0198	0.0151	1.7225	0.1894
TFT	1	1	-1.0092	0.4939	4.1741	0.0410
ACCTYP	1	1	0.8004	0.2158	13.7555	0.0002
SEV2		1	0.000483	0.000170	8.1144	0.0044
TSP		1	0.0195	0.0180	1.1729	0.2788
TSP*CDR		1	-0.00090	0.000516	3.0099	0.0828
TSP2		1	-0.00084	0.000423	3.9557	0.0467

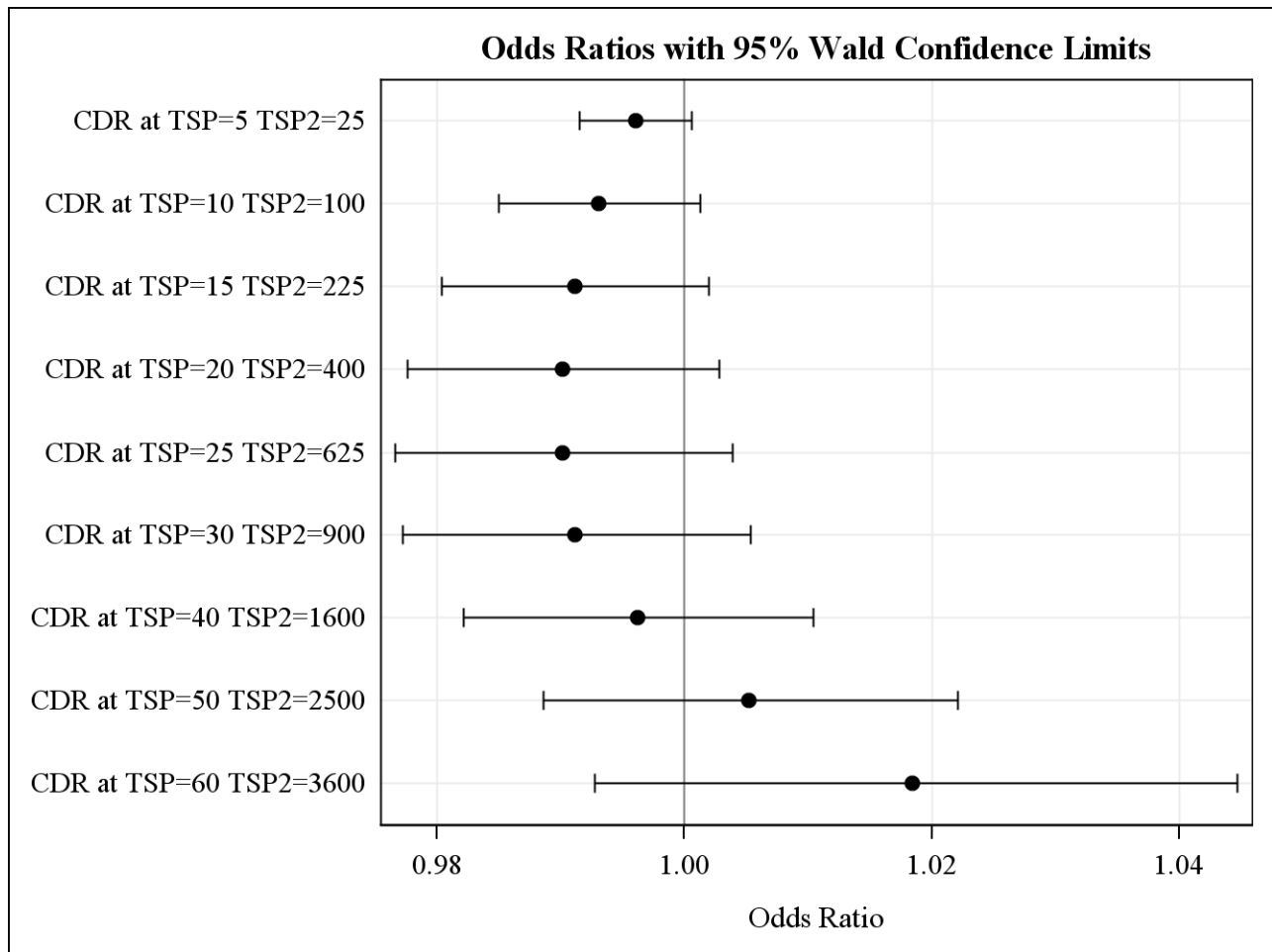


Analysis of Maximum Likelihood Estimates						
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
CDR*TSP2		1	0.000020	0.000010	3.6516	0.0560
Haz_Env		1	0.0584	0.0462	1.5996	0.2060
TSP2*Haz_Env		1	0.000040	0.000032	1.5749	0.2095
TF_Lading_Grp	B	1	0.1975	0.2177	0.8231	0.3643
TF_Lading_Grp	D	1	0.8302	0.3485	5.6741	0.0172
TF_Lading_Grp	E	1	0.6831	0.2061	10.9844	0.0009
TF_Lading_Grp	F	1	0.3576	0.2261	2.5020	0.1137
TF_Lading_Grp	G	1	-2.0358	0.4077	24.9345	<.0001
TF_Lading_Grp	H	1	-1.8599	0.4182	19.7778	<.0001
TF_Lading_Grp	K	1	-0.9286	0.4530	4.2012	0.0404
TF_Lading_Grp	M	1	1.1469	0.5453	4.4242	0.0354
TF_Lading_Grp	O	1	-1.3968	0.5753	5.8938	0.0152
TF_Lading_Grp	Q	1	-1.1367	0.6854	2.7502	0.0972
TF_Lading_Grp	T	1	-2.3648	1.4465	2.6728	0.1021

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	79.9	Somers' D	0.598
Percent Discordant	20.1	Gamma	0.598
Percent Tied	0.0	Tau-a	0.123
Pairs	2055866	c	0.799

Odds Ratio Estimates and Wald Confidence Intervals			
Label	Estimate	95% Confidence Limits	
CDR at TSP=5 TSP2=25	0.996	0.991	1.001
CDR at TSP=10 TSP2=100	0.993	0.985	1.001
CDR at TSP=15 TSP2=225	0.991	0.980	1.002

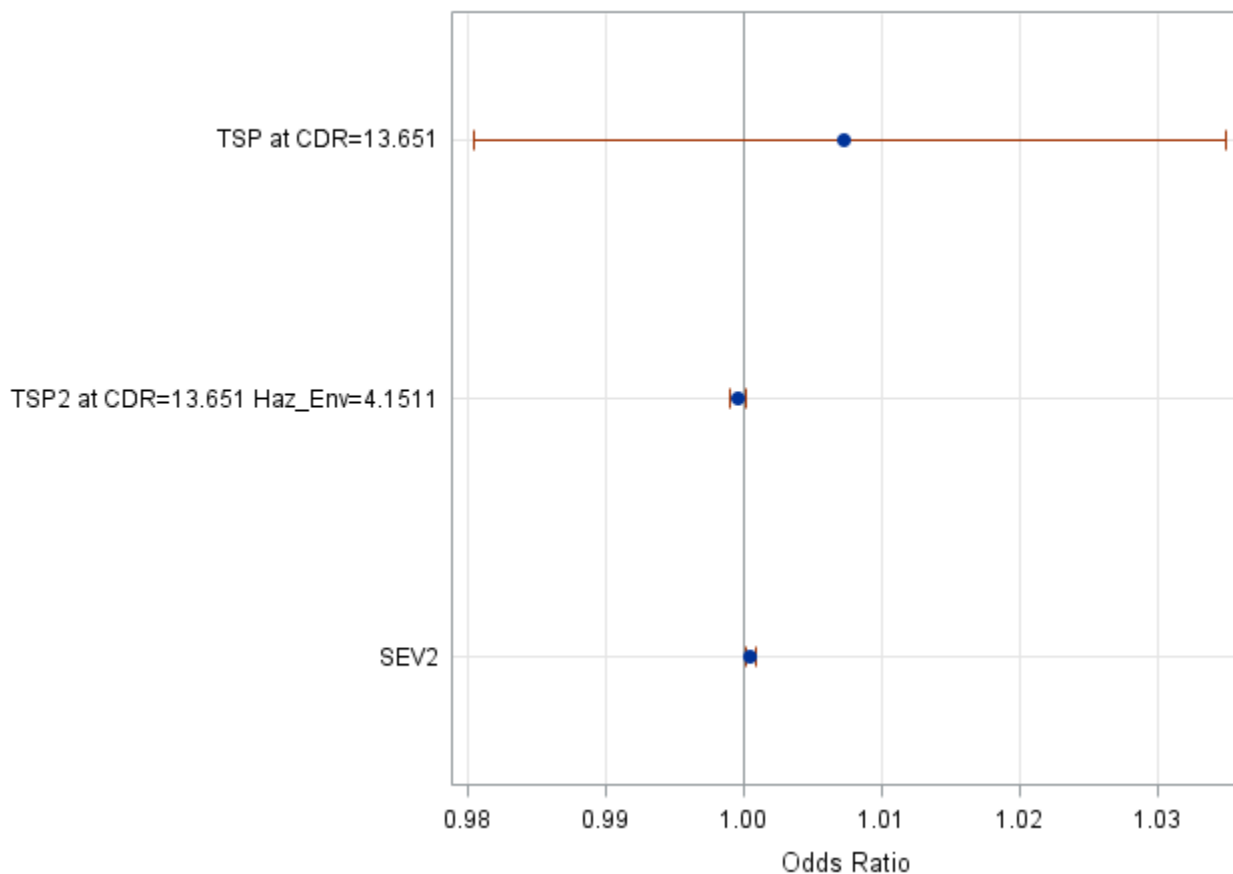
Odds Ratio Estimates and Wald Confidence Intervals			
Label	Estimate	95% Confidence Limits	
CDR at TSP=20 TSP2=400	0.990	0.978	1.003
CDR at TSP=25 TSP2=625	0.990	0.977	1.004
CDR at TSP=30 TSP2=900	0.991	0.977	1.005
CDR at TSP=40 TSP2=1600	0.996	0.982	1.010
CDR at TSP=50 TSP2=2500	1.005	0.989	1.022
CDR at TSP=60 TSP2=3600	1.018	0.993	1.045



**Odds Ratio Estimates and Wald Confidence Intervals**

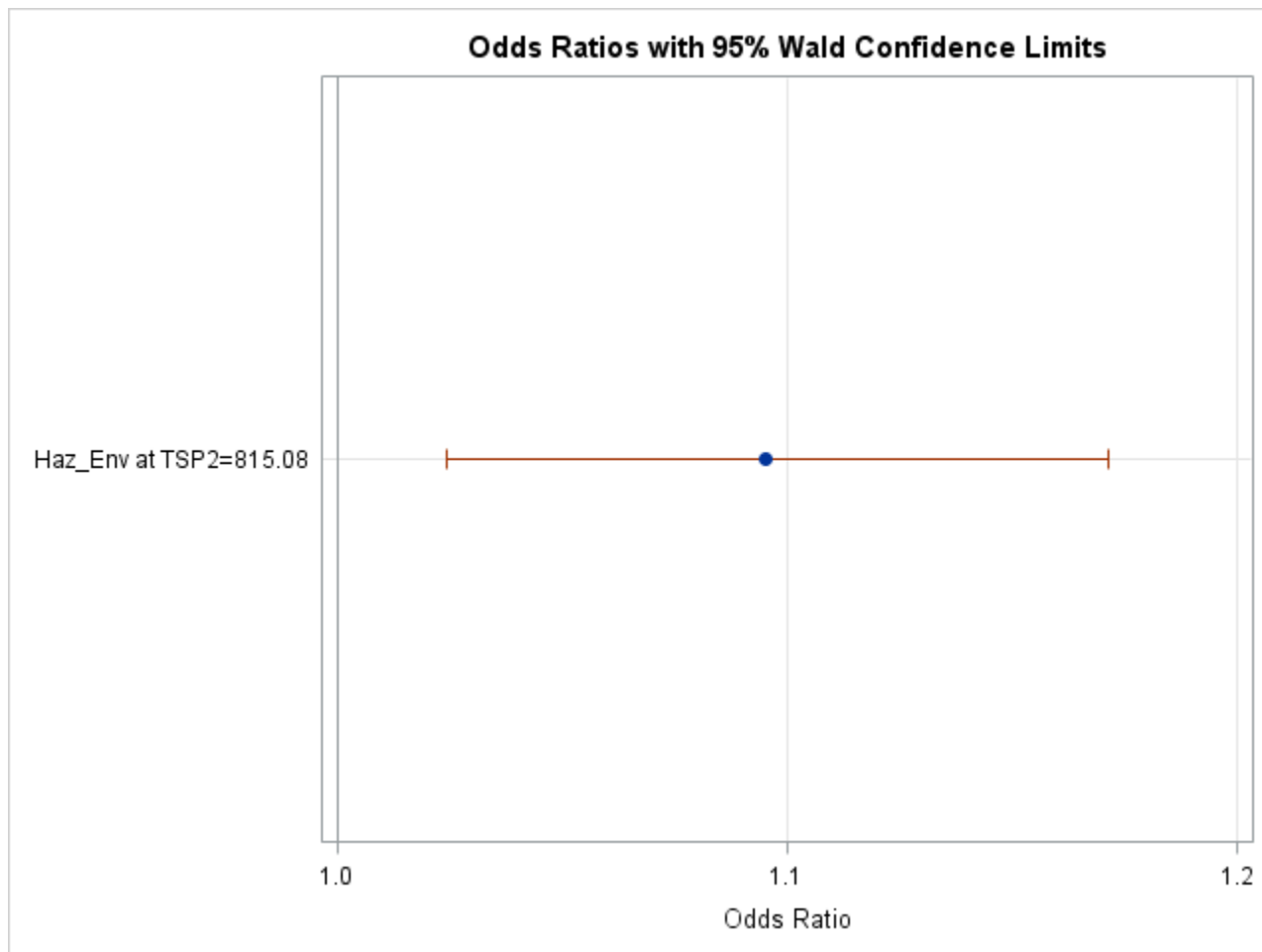
Label	Estimate	95% Confidence Limits	
TSP at CDR=13.651	1.007	0.980	1.035
TSP2 at CDR=13.651 Haz_Env=4.1511	1.000	0.999	1.000
SEV2	1.000	1.000	1.001

**Odds Ratios with 95% Wald Confidence Limits**



**Odds Ratio Estimates and Wald Confidence Intervals**

Label	Estimate	95% Confidence Limits	
Haz_Env at TSP2=815.08	1.095	1.024	1.171

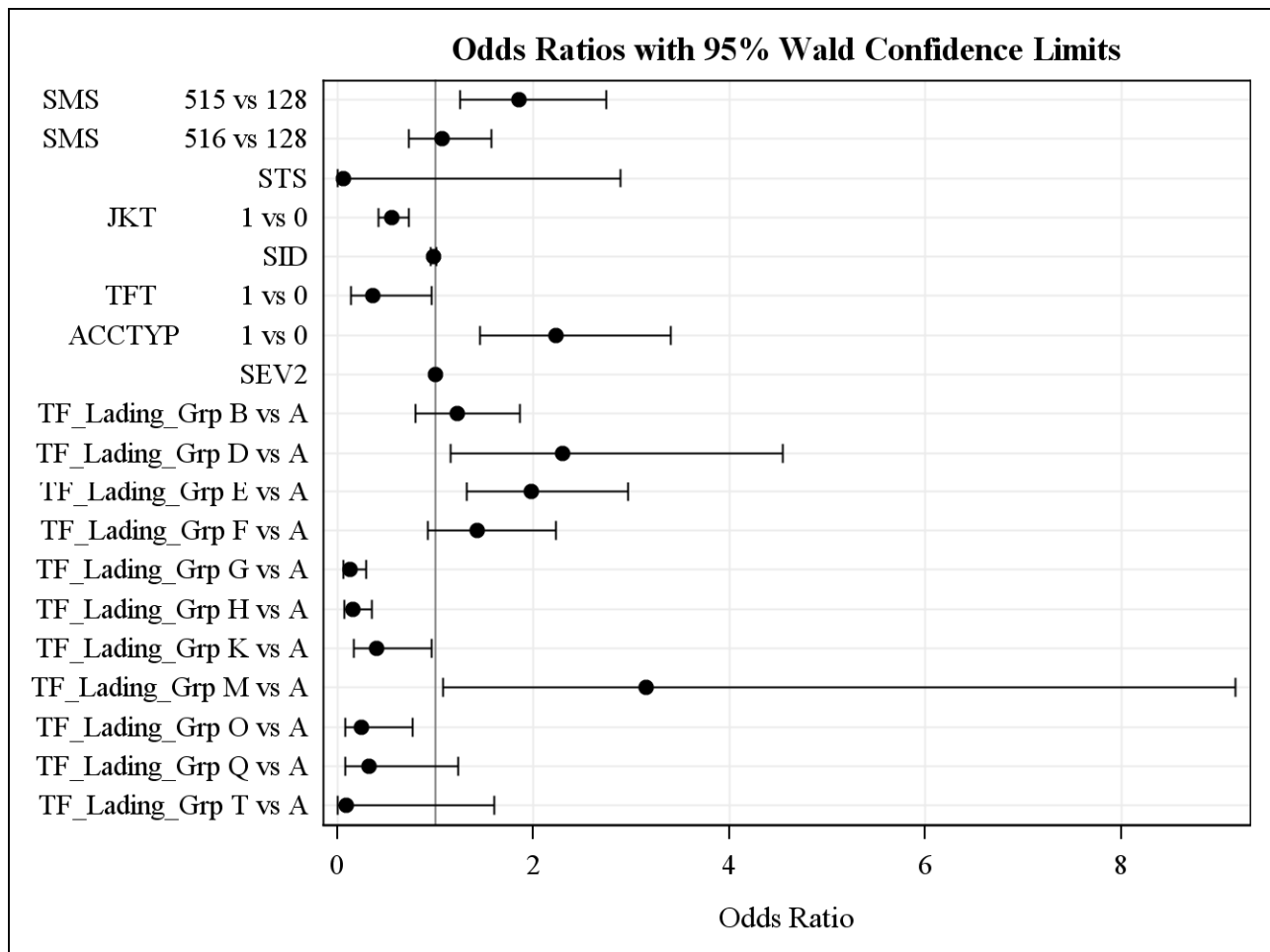


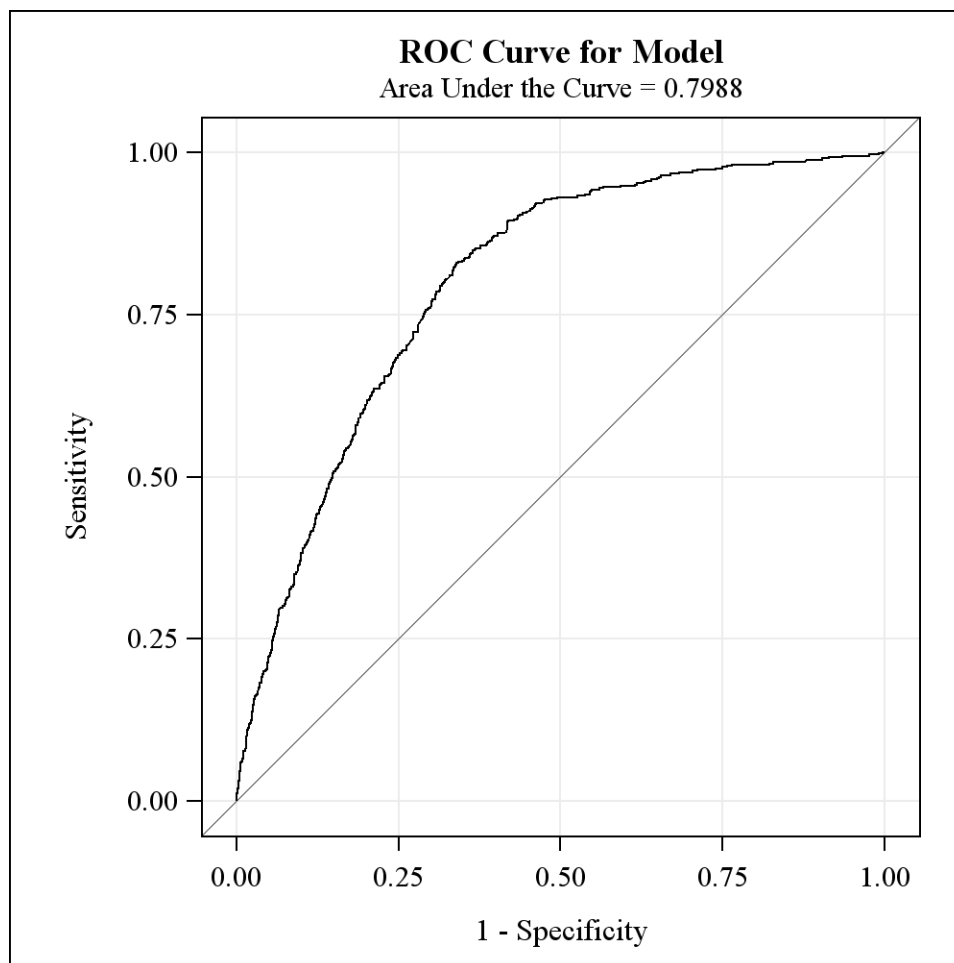
Parameter Estimates and Wald Confidence Intervals				
Parameter		Estimate	95% Confidence Limits	
Intercept		0.5555	-3.3132	4.4243
SMS	515	0.6168	0.2234	1.0101
SMS	516	0.0675	-0.3161	0.4512
STS		-2.7467	-6.5543	1.0610
JKT	1	-0.5928	-0.8667	-0.3189
SID		-0.0198	-0.0494	0.00978
TFT	1	-1.0092	-1.9773	-0.0410
ACCTYP	1	0.8004	0.3774	1.2234
SEV2		0.000483	0.000151	0.000815
TSP		0.0195	-0.0158	0.0547
TSP*CDR		-0.00090	-0.00191	0.000116
TSP2		-0.00084	-0.00167	-0.00001

Parameter Estimates and Wald Confidence Intervals				
Parameter		Estimate	95% Confidence Limits	
CDR*TSP2		0.000020	-5.13E-7	0.000040
Haz_Env		0.0584	-0.0321	0.1489
TSP2*Haz_Env		0.000040	-0.00002	0.000103
TF_Lading_Grp	B	0.1975	-0.2291	0.6241
TF_Lading_Grp	D	0.8302	0.1471	1.5133
TF_Lading_Grp	E	0.6831	0.2791	1.0870
TF_Lading_Grp	F	0.3576	-0.0855	0.8008
TF_Lading_Grp	G	-2.0358	-2.8349	-1.2368
TF_Lading_Grp	H	-1.8599	-2.6796	-1.0402
TF_Lading_Grp	K	-0.9286	-1.8165	-0.0406
TF_Lading_Grp	M	1.1469	0.0782	2.2156
TF_Lading_Grp	O	-1.3968	-2.5244	-0.2691
TF_Lading_Grp	Q	-1.1367	-2.4800	0.2067
TF_Lading_Grp	T	-2.3648	-5.1998	0.4702

Odds Ratio Estimates and Wald Confidence Intervals					
Effect		Unit	Estimate	95% Confidence Limits	
SMS	515 vs 128	1.000	1.853	1.250	2.746
SMS	516 vs 128	1.000	1.070	0.729	1.570
STS		1.000	0.064	0.001	2.889
JKT	1 vs 0	1.000	0.553	0.420	0.727
SID		1.000	0.980	0.952	1.010
TFT	1 vs 0	1.000	0.365	0.138	0.960
ACCTYP	1 vs 0	1.000	2.226	1.459	3.399
SEV2		1.000	1.000	1.000	1.001
TF_Lading_Grp	B vs A	1.000	1.218	0.795	1.867
TF_Lading_Grp	D vs A	1.000	2.294	1.158	4.542
TF_Lading_Grp	E vs A	1.000	1.980	1.322	2.965
TF_Lading_Grp	F vs A	1.000	1.430	0.918	2.227
TF_Lading_Grp	G vs A	1.000	0.131	0.059	0.290
TF_Lading_Grp	H vs A	1.000	0.156	0.069	0.353

Odds Ratio Estimates and Wald Confidence Intervals				
Effect	Unit	Estimate	95% Confidence Limits	
TF_Lading_Grp K vs A	1.000	0.395	0.163	0.960
TF_Lading_Grp M vs A	1.000	3.148	1.081	9.167
TF_Lading_Grp O vs A	1.000	0.247	0.080	0.764
TF_Lading_Grp Q vs A	1.000	0.321	0.084	1.230
TF_Lading_Grp T vs A	1.000	0.094	0.006	1.600





Estimated Correlation Matrix										
Parameter	Intercept	SMS515	SMS516	STS	JKT1	SID	TFT1	ACCTYP1	SEV2	TSP
Intercept	1.0000	-0.5739	-0.4458	-0.4950	-0.2591	-0.8903	0.4340	-0.0469	0.0068	-0.0295
SMS515	-0.5739	1.0000	0.8032	0.0317	-0.1052	0.5937	0.1653	0.0184	0.0032	0.0213
SMS516	-0.4458	0.8032	1.0000	-0.0027	-0.2012	0.4690	0.2078	0.0195	0.0081	0.0040
STS	-0.4950	0.0317	-0.0027	1.0000	0.0825	0.0715	-0.7664	-0.0097	0.0394	0.0085
JKT1	-0.2591	-0.1052	-0.2012	0.0825	1.0000	0.2451	-0.3062	-0.0329	0.0225	0.0462
SID	-0.8903	0.5937	0.4690	0.0715	0.2451	1.0000	-0.1605	0.0044	-0.0283	-0.0041
TFT1	0.4340	0.1653	0.2078	-0.7664	-0.3062	-0.1605	1.0000	0.0180	-0.0540	-0.0438
ACCTYP1	-0.0469	0.0184	0.0195	-0.0097	-0.0329	0.0044	0.0180	1.0000	-0.0508	-0.0830
SEV2	0.0068	0.0032	0.0081	0.0394	0.0225	-0.0283	-0.0540	-0.0508	1.0000	0.0225
TSP	-0.0295	0.0213	0.0040	0.0085	0.0462	-0.0041	-0.0438	-0.0830	0.0225	1.0000
TSPCDR	-0.0038	0.0316	0.0373	-0.0157	-0.0600	0.0345	0.0861	0.0840	-0.0585	-0.7215
TSP2	0.0104	-0.0404	-0.0280	-0.0292	-0.0241	0.0219	0.0493	-0.0322	-0.2147	-0.8850

Estimated Correlation Matrix										
Parameter	Intercept	SMS515	SMS516	STS	JKT1	SID	TFT1	ACCTYP1	SEV2	TSP
CDRTSP2	-0.0049	-0.0153	-0.0118	0.0255	0.0414	-0.0312	-0.0866	-0.0939	0.0445	0.7638
Haz_Env	-0.0429	-0.0277	-0.0324	0.0100	0.0456	0.0108	-0.0389	-0.4782	0.0689	-0.1610
TSP2Haz_Env	0.0183	0.0604	0.0376	0.0230	-0.0010	-0.0109	-0.0057	0.2932	0.1670	0.2044
TF_Lading_GrpB	0.3478	-0.2178	-0.1656	-0.0209	-0.1124	-0.4602	0.2854	-0.0036	-0.0351	-0.0736
TF_Lading_GrpD	0.0691	0.1122	0.0176	-0.6945	0.1974	0.2313	0.5822	0.0258	-0.0423	-0.0457
TF_Lading_GrpE	0.3273	-0.1134	-0.1138	-0.0118	-0.0267	-0.4589	0.3065	-0.0136	-0.0402	-0.0757
TF_Lading_GrpF	0.0993	-0.0510	-0.0127	-0.0266	-0.2977	-0.1598	0.2860	0.0139	-0.0580	-0.0710
TF_Lading_GrpG	-0.1257	0.0877	0.0954	0.0104	-0.1161	0.1144	0.1133	0.0023	-0.0522	-0.0330
TF_Lading_GrpH	0.2062	-0.1160	-0.0690	-0.0154	-0.1830	-0.2635	0.1767	0.0047	-0.0523	-0.0294
TF_Lading_GrpK	0.2008	-0.1231	-0.0773	-0.0199	-0.1931	-0.2529	0.1672	0.0036	-0.0451	0.0063
TF_Lading_GrpM	0.1077	-0.0400	-0.1047	-0.0129	-0.1080	-0.1386	0.1105	-0.0088	0.0090	0.0027
TF_Lading_GrpO	-0.0879	0.0614	0.0708	0.0086	-0.0846	0.0764	0.0828	-0.0107	-0.0205	-0.0184
TF_Lading_GrpQ	0.0879	-0.0898	-0.0704	-0.0008	0.0033	-0.1222	0.0674	0.0025	0.0017	-0.0217
TF_Lading_GrpT	0.0443	-0.0342	-0.0286	-0.0008	-0.0309	-0.0624	0.0405	0.0167	-0.0101	-0.0155

Estimated Correlation Matrix							
Parameter	TSPCDR	TSP2	CDRTSP2	Haz_Env	TSP2Haz_Env	TF_Lading_GrpB	TF_Lading_GrpD
Intercept	-0.0038	0.0104	-0.0049	-0.0429	0.0183	0.3478	0.0691
SMS515	0.0316	-0.0404	-0.0153	-0.0277	0.0604	-0.2178	0.1122
SMS516	0.0373	-0.0280	-0.0118	-0.0324	0.0376	-0.1656	0.0176
STS	-0.0157	-0.0292	0.0255	0.0100	0.0230	-0.0209	-0.6945
JKT1	-0.0600	-0.0241	0.0414	0.0456	-0.0010	-0.1124	0.1974
SID	0.0345	0.0219	-0.0312	0.0108	-0.0109	-0.4602	0.2313
TFT1	0.0861	0.0493	-0.0866	-0.0389	-0.0057	0.2854	0.5822
ACCTYP1	0.0840	-0.0322	-0.0939	-0.4782	0.2932	-0.0036	0.0258
SEV2	-0.0585	-0.2147	0.0445	0.0689	0.1670	-0.0351	-0.0423
TSP	-0.7215	-0.8850	0.7638	-0.1610	0.2044	-0.0736	-0.0457
TSPCDR	1.0000	0.5827	-0.9464	-0.0920	-0.0305	0.0841	0.0427
TSP2	0.5827	1.0000	-0.6733	0.3313	-0.5356	0.0629	0.0686
CDRTSP2	-0.9464	-0.6733	1.0000	0.0854	-0.0005	-0.0800	-0.0701
Haz_Env	-0.0920	0.3313	0.0854	1.0000	-0.6837	-0.0164	-0.0355
TSP2Haz_Env	-0.0305	-0.5356	-0.0005	-0.6837	1.0000	-0.0230	-0.0075
TF_Lading_GrpB	0.0841	0.0629	-0.0800	-0.0164	-0.0230	1.0000	0.2020



Estimated Correlation Matrix							
Parameter	TSPCDR	TSP2	CDRTSP2	Haz_Env	TSP2Haz_Env	TF_Lading_GrpB	TF_Lading_GrpD
TF_Lading_GrpD	0.0427	0.0686	-0.0701	-0.0355	-0.0075	0.2020	1.0000
TF_Lading_GrpE	0.1014	0.0639	-0.0949	-0.0091	-0.0120	0.7669	0.2341
TF_Lading_GrpF	0.1218	0.0900	-0.1149	-0.0033	-0.0643	0.5652	0.2269
TF_Lading_GrpG	0.0161	0.0348	-0.0175	-0.0143	-0.0038	0.2154	0.1635
TF_Lading_GrpH	0.0332	0.0240	-0.0383	-0.0134	0.0089	0.3889	0.0770
TF_Lading_GrpK	0.0161	0.0038	-0.0159	-0.0222	-0.0021	0.3585	0.0661
TF_Lading_GrpM	0.0253	0.0008	-0.0405	-0.0477	0.0265	0.2583	0.0962
TF_Lading_GrpO	0.0452	0.0142	-0.0360	0.0057	-0.0018	0.1560	0.1095
TF_Lading_GrpQ	0.0191	0.0216	-0.0212	-0.0006	-0.0028	0.2158	0.0714
TF_Lading_GrpT	0.0189	0.0112	-0.0143	0.0047	-0.0035	0.1053	0.0269

Estimated Correlation Matrix					
Parameter	TF_Lading_GrpE	TF_Lading_GrpF	TF_Lading_GrpG	TF_Lading_GrpH	TF_Lading_GrpK
Intercept	0.3273	0.0993	-0.1257	0.2062	0.2008
SMS515	-0.1134	-0.0510	0.0877	-0.1160	-0.1231
SMS516	-0.1138	-0.0127	0.0954	-0.0690	-0.0773
STS	-0.0118	-0.0266	0.0104	-0.0154	-0.0199
JKT1	-0.0267	-0.2977	-0.1161	-0.1830	-0.1931
SID	-0.4589	-0.1598	0.1144	-0.2635	-0.2529
TFT1	0.3065	0.2860	0.1133	0.1767	0.1672
ACCTYP1	-0.0136	0.0139	0.0023	0.0047	0.0036
SEV2	-0.0402	-0.0580	-0.0522	-0.0523	-0.0451
TSP	-0.0757	-0.0710	-0.0330	-0.0294	0.0063
TSPCDR	0.1014	0.1218	0.0161	0.0332	0.0161
TSP2	0.0639	0.0900	0.0348	0.0240	0.0038
CDRTSP2	-0.0949	-0.1149	-0.0175	-0.0383	-0.0159
Haz_Env	-0.0091	-0.0033	-0.0143	-0.0134	-0.0222
TSP2Haz_Env	-0.0120	-0.0643	-0.0038	0.0089	-0.0021
TF_Lading_GrpB	0.7669	0.5652	0.2154	0.3889	0.3585
TF_Lading_GrpD	0.2341	0.2269	0.1635	0.0770	0.0661
TF_Lading_GrpE	1.0000	0.5633	0.2109	0.3926	0.3575
TF_Lading_GrpF	0.5633	1.0000	0.2793	0.3305	0.3147

Estimated Correlation Matrix					
Parameter	TF_Lading_GrpE	TF_Lading_GrpF	TF_Lading_GrpG	TF_Lading_GrpH	TF_Lading_GrpK
TF_Lading_GrpG	0.2109	0.2793	1.0000	0.1304	0.1208
TF_Lading_GrpH	0.3926	0.3305	0.1304	1.0000	0.2115
TF_Lading_GrpK	0.3575	0.3147	0.1208	0.2115	1.0000
TF_Lading_GrpM	0.2695	0.2320	0.0950	0.1458	0.1387
TF_Lading_GrpO	0.1548	0.2033	0.1261	0.0935	0.0881
TF_Lading_GrpQ	0.2257	0.1671	0.0662	0.1109	0.1037
TF_Lading_GrpT	0.1082	0.0895	0.0354	0.0584	0.0547

Estimated Correlation Matrix				
Parameter	TF_Lading_GrpM	TF_Lading_GrpO	TF_Lading_GrpQ	TF_Lading_GrpT
Intercept	0.1077	-0.0879	0.0879	0.0443
SMS515	-0.0400	0.0614	-0.0898	-0.0342
SMS516	-0.1047	0.0708	-0.0704	-0.0286
STS	-0.0129	0.0086	-0.0008	-0.0008
JKT1	-0.1080	-0.0846	0.0033	-0.0309
SID	-0.1386	0.0764	-0.1222	-0.0624
TFT1	0.1105	0.0828	0.0674	0.0405
ACCTYP1	-0.0088	-0.0107	0.0025	0.0167
SEV2	0.0090	-0.0205	0.0017	-0.0101
TSP	0.0027	-0.0184	-0.0217	-0.0155
TSPCDR	0.0253	0.0452	0.0191	0.0189
TSP2	0.0008	0.0142	0.0216	0.0112
CDRTSP2	-0.0405	-0.0360	-0.0212	-0.0143
Haz_Env	-0.0477	0.0057	-0.0006	0.0047
TSP2Haz_Env	0.0265	-0.0018	-0.0028	-0.0035
TF_Lading_GrpB	0.2583	0.1560	0.2158	0.1053
TF_Lading_GrpD	0.0962	0.1095	0.0714	0.0269
TF_Lading_GrpE	0.2695	0.1548	0.2257	0.1082
TF_Lading_GrpF	0.2320	0.2033	0.1671	0.0895
TF_Lading_GrpG	0.0950	0.1261	0.0662	0.0354
TF_Lading_GrpH	0.1458	0.0935	0.1109	0.0584

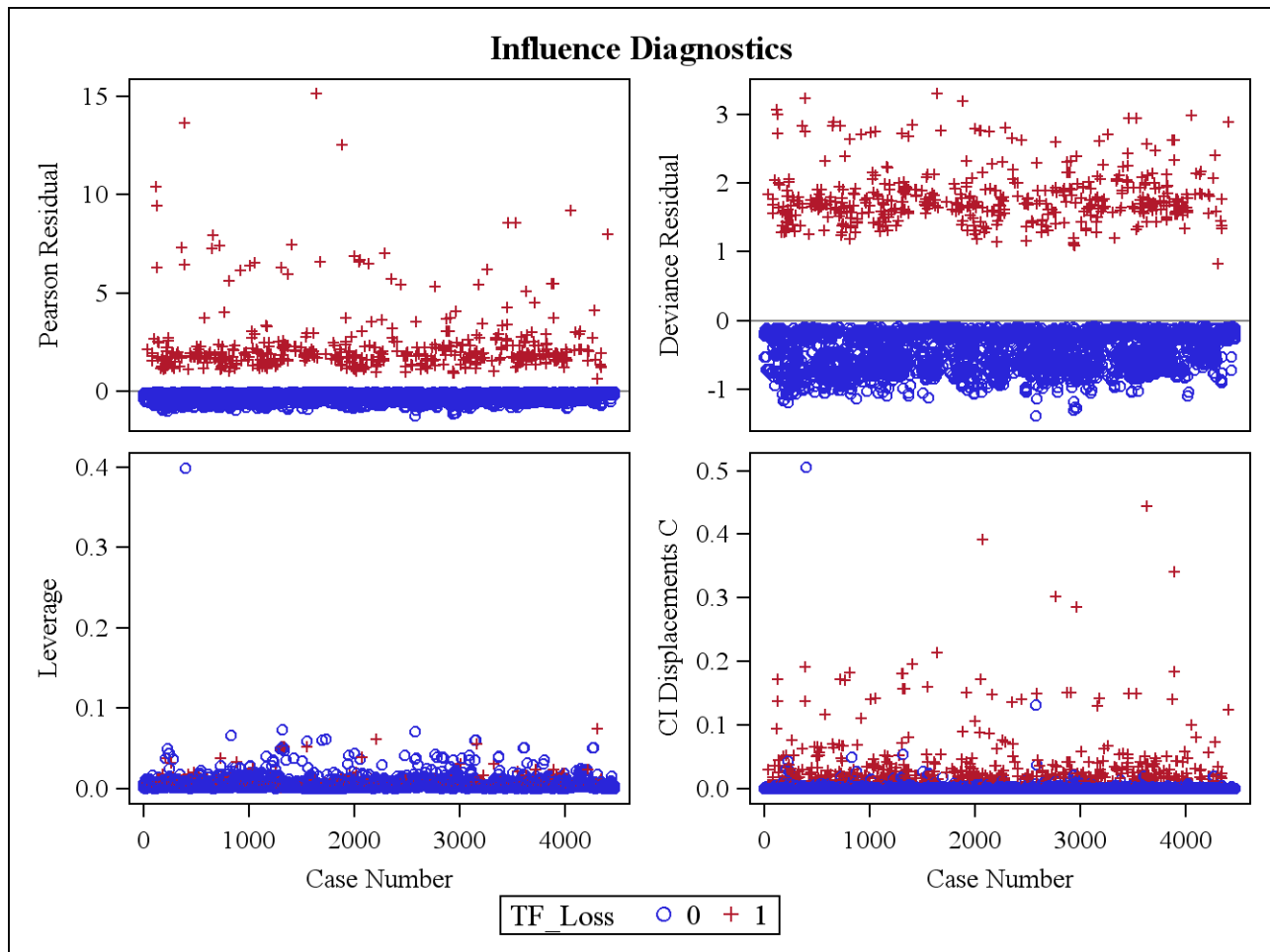
Estimated Correlation Matrix				
Parameter	TF_Lading_GrpM	TF_Lading_GrpO	TF_Lading_GrpQ	TF_Lading_GrpT
TF_Lading_GrpK	0.1387	0.0881	0.1037	0.0547
TF_Lading_GrpM	1.0000	0.0674	0.0743	0.0380
TF_Lading_GrpO	0.0674	1.0000	0.0484	0.0265
TF_Lading_GrpQ	0.0743	0.0484	1.0000	0.0317
TF_Lading_GrpT	0.0380	0.0265	0.0317	1.0000

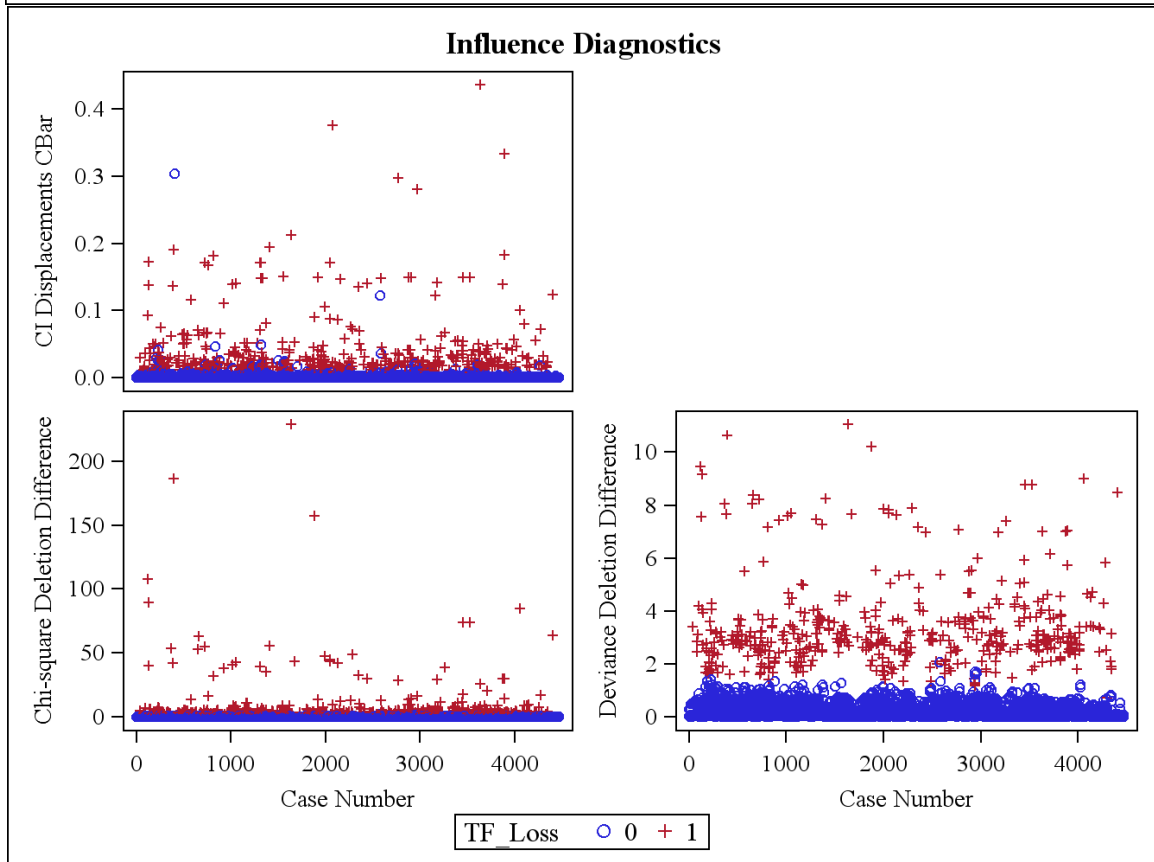
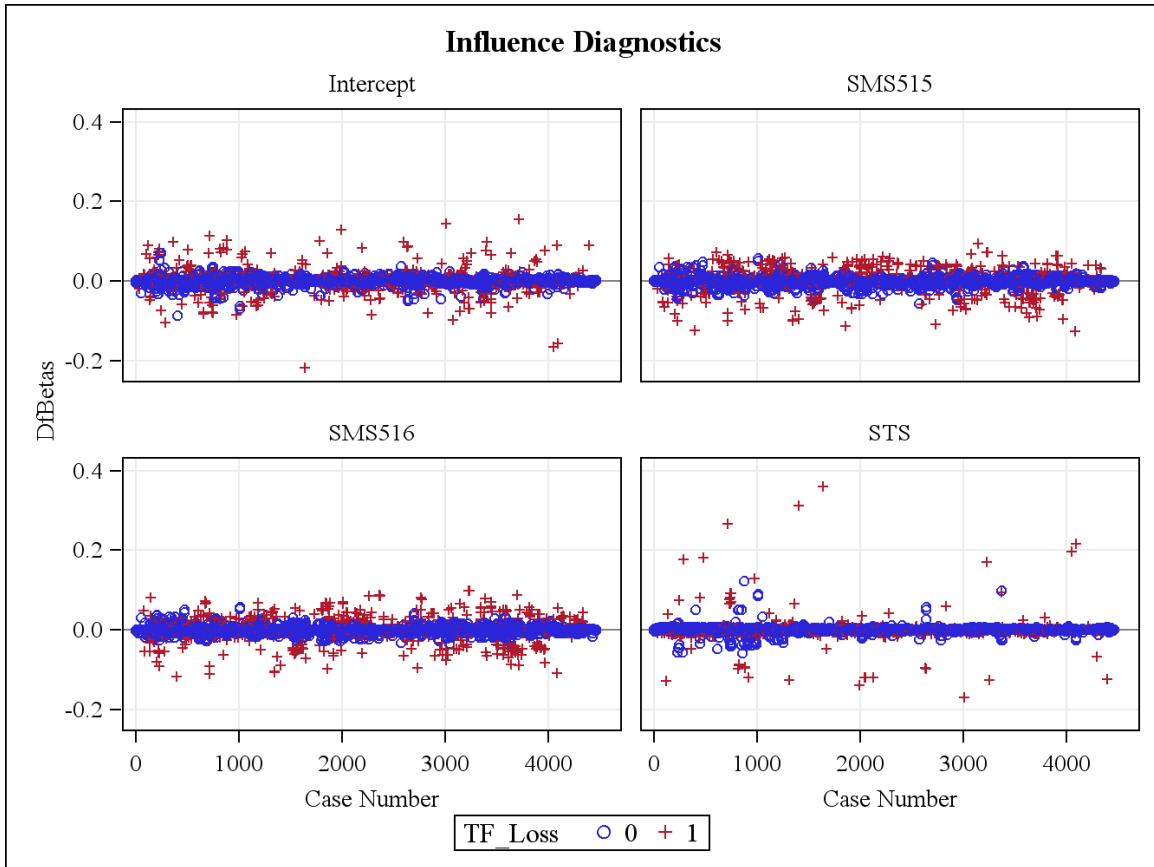
Partition for the Hosmer and Lemeshow Test					
Group	Total	TF_Loss = 1		TF_Loss = 0	
		Observed	Expected	Observed	Expected
1	448	6	3.85	442	444.15
2	447	4	6.81	443	440.19
3	451	9	8.95	442	442.05
4	447	11	11.87	436	435.13
5	447	19	22.06	428	424.94
6	447	40	47.45	407	399.55
7	447	76	70.64	371	376.36
8	447	88	92.26	359	354.74
9	447	111	112.62	336	334.38
10	439	157	153.09	282	285.91

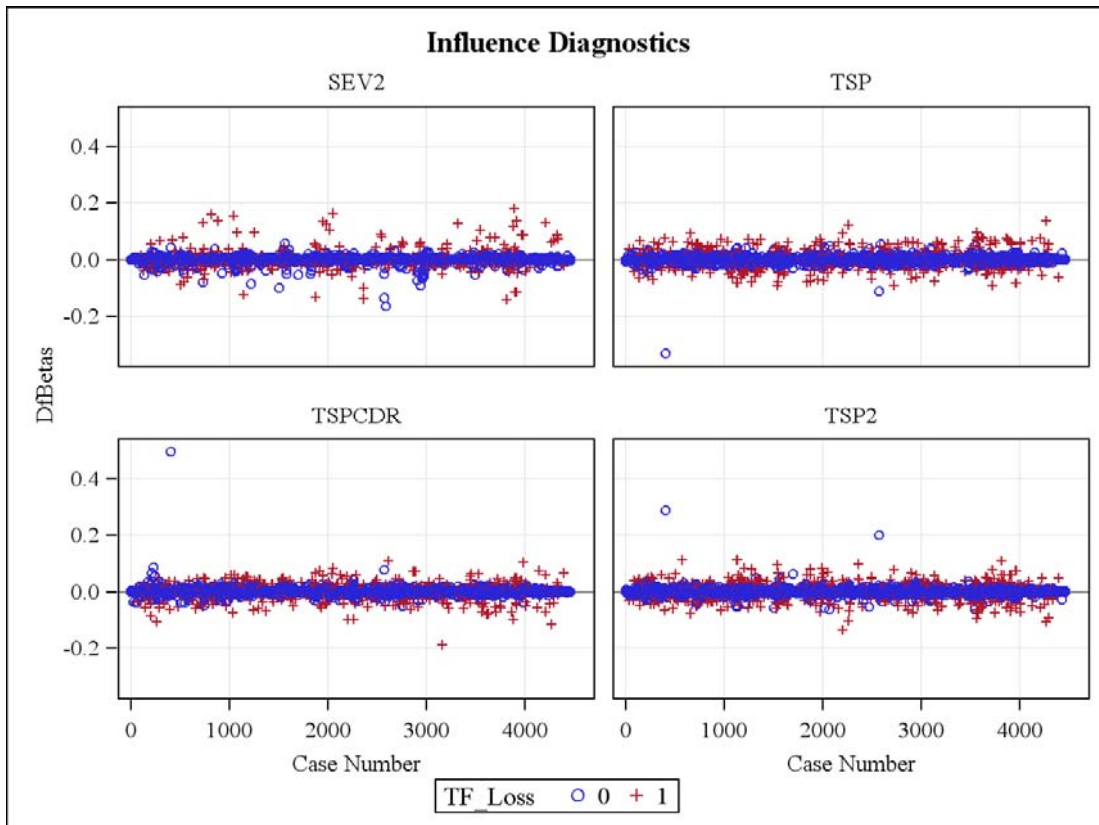
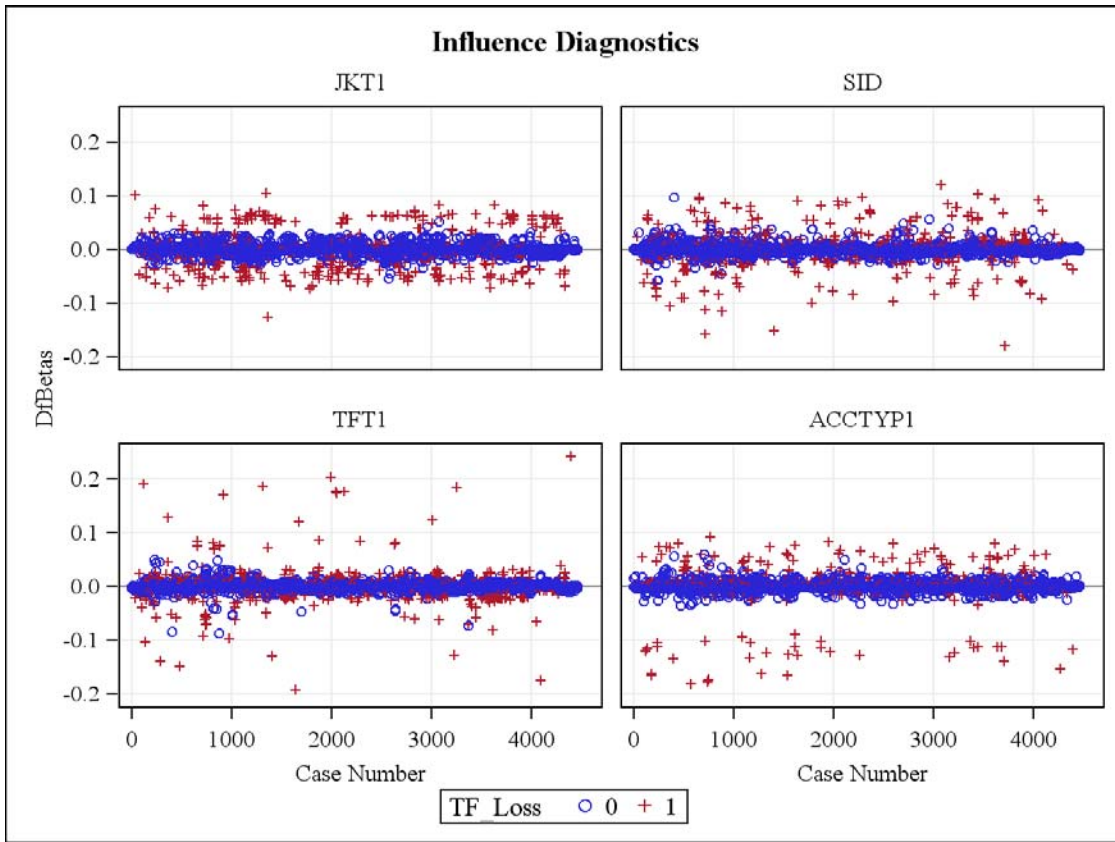
Hosmer and Lemeshow Goodness-of-Fit Test		
Chi-Square	DF	Pr > ChiSq
5.1212	8	0.7445

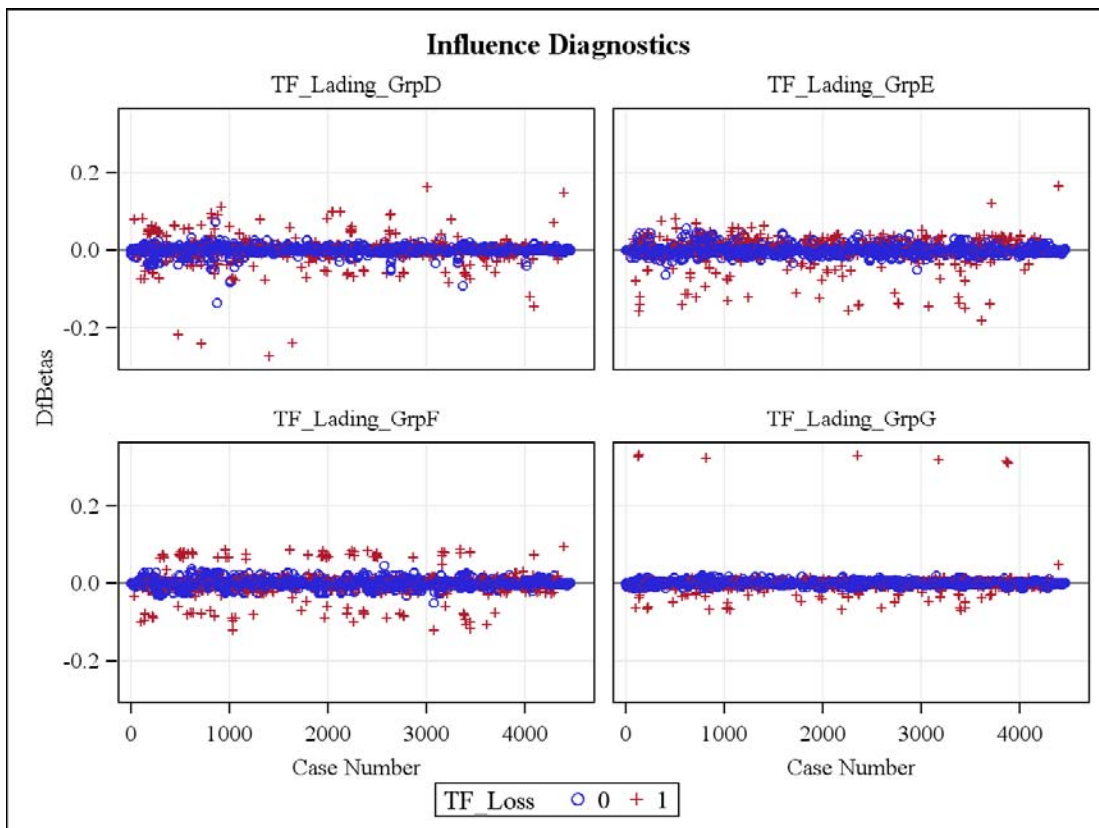
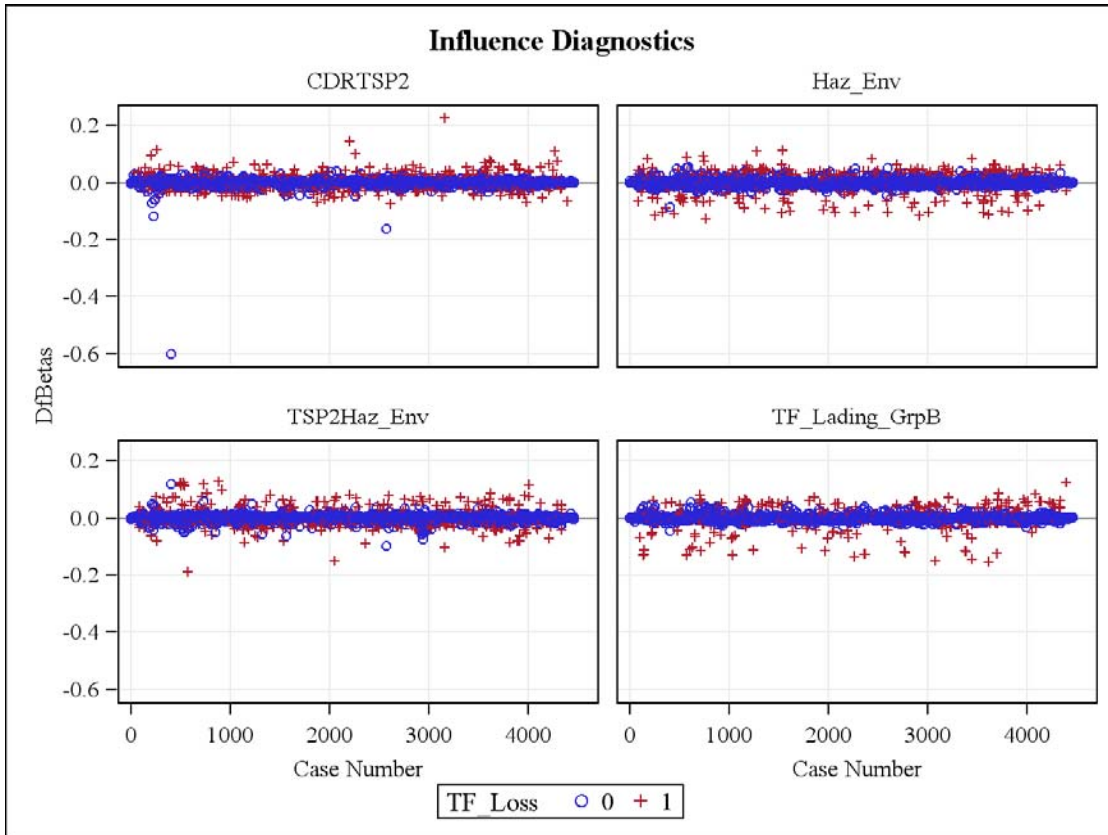
<b>Classification Table</b>									
<b>Prob Level</b>	<b>Correct</b>		<b>Incorrect</b>		<b>Percentages</b>				
	<b>Event</b>	<b>Non-Event</b>	<b>Event</b>	<b>Non-Event</b>	<b>Correct</b>	<b>Sensitivity</b>	<b>Specificity</b>	<b>False POS</b>	<b>False NEG</b>
<b>0.000</b>	521	0	3946	0	11.7	100.0	0.0	88.3	.
<b>0.020</b>	503	1136	2810	18	36.7	96.5	28.8	84.8	1.6
<b>0.040</b>	485	1901	2045	36	53.4	93.1	48.2	80.8	1.9
<b>0.060</b>	476	2085	1861	45	57.3	91.4	52.8	79.6	2.1
<b>0.080</b>	467	2215	1731	54	60.0	89.6	56.1	78.8	2.4
<b>0.100</b>	453	2318	1628	68	62.0	86.9	58.7	78.2	2.8
<b>0.120</b>	436	2501	1445	85	65.7	83.7	63.4	76.8	3.3
<b>0.140</b>	416	2640	1306	105	68.4	79.8	66.9	75.8	3.8
<b>0.160</b>	383	2769	1177	138	70.6	73.5	70.2	75.4	4.7
<b>0.180</b>	351	2954	992	170	74.0	67.4	74.9	73.9	5.4
<b>0.200</b>	315	3103	843	206	76.5	60.5	78.6	72.8	6.2
<b>0.220</b>	278	3236	710	243	78.7	53.4	82.0	71.9	7.0
<b>0.240</b>	231	3387	559	290	81.0	44.3	85.8	70.8	7.9
<b>0.260</b>	175	3550	396	346	83.4	33.6	90.0	69.4	8.9
<b>0.280</b>	145	3660	286	376	85.2	27.8	92.8	66.4	9.3
<b>0.300</b>	102	3741	205	419	86.0	19.6	94.8	66.8	10.1
<b>0.320</b>	87	3793	153	434	86.9	16.7	96.1	63.8	10.3
<b>0.340</b>	82	3823	123	439	87.4	15.7	96.9	60.0	10.3
<b>0.360</b>	64	3843	103	457	87.5	12.3	97.4	61.7	10.6
<b>0.380</b>	52	3871	75	469	87.8	10.0	98.1	59.1	10.8
<b>0.400</b>	39	3891	55	482	88.0	7.5	98.6	58.5	11.0
<b>0.420</b>	29	3912	34	492	88.2	5.6	99.1	54.0	11.2
<b>0.440</b>	12	3924	22	509	88.1	2.3	99.4	64.7	11.5
<b>0.460</b>	10	3932	14	511	88.2	1.9	99.6	58.3	11.5
<b>0.480</b>	9	3933	13	512	88.2	1.7	99.7	59.1	11.5
<b>0.500</b>	5	3937	9	516	88.2	1.0	99.8	64.3	11.6
<b>0.520</b>	4	3941	5	517	88.3	0.8	99.9	55.6	11.6
<b>0.540</b>	2	3941	5	519	88.3	0.4	99.9	71.4	11.6
<b>0.560</b>	1	3943	3	520	88.3	0.2	99.9	75.0	11.7

Classification Table									
Prob Level	Correct		Incorrect		Percentages				
	Event	Non-Event	Event	Non-Event	Correct	Sensitivity	Specificity	False POS	False NEG
0.580	1	3944	2	520	88.3	0.2	99.9	66.7	11.6
0.600	1	3945	1	520	88.3	0.2	100.0	50.0	11.6
0.620	1	3945	1	520	88.3	0.2	100.0	50.0	11.6
0.640	1	3945	1	520	88.3	0.2	100.0	50.0	11.6
0.660	1	3945	1	520	88.3	0.2	100.0	50.0	11.6
0.680	1	3946	0	520	88.4	0.2	100.0	0.0	11.6
0.700	0	3946	0	521	88.3	0.0	100.0	.	11.7

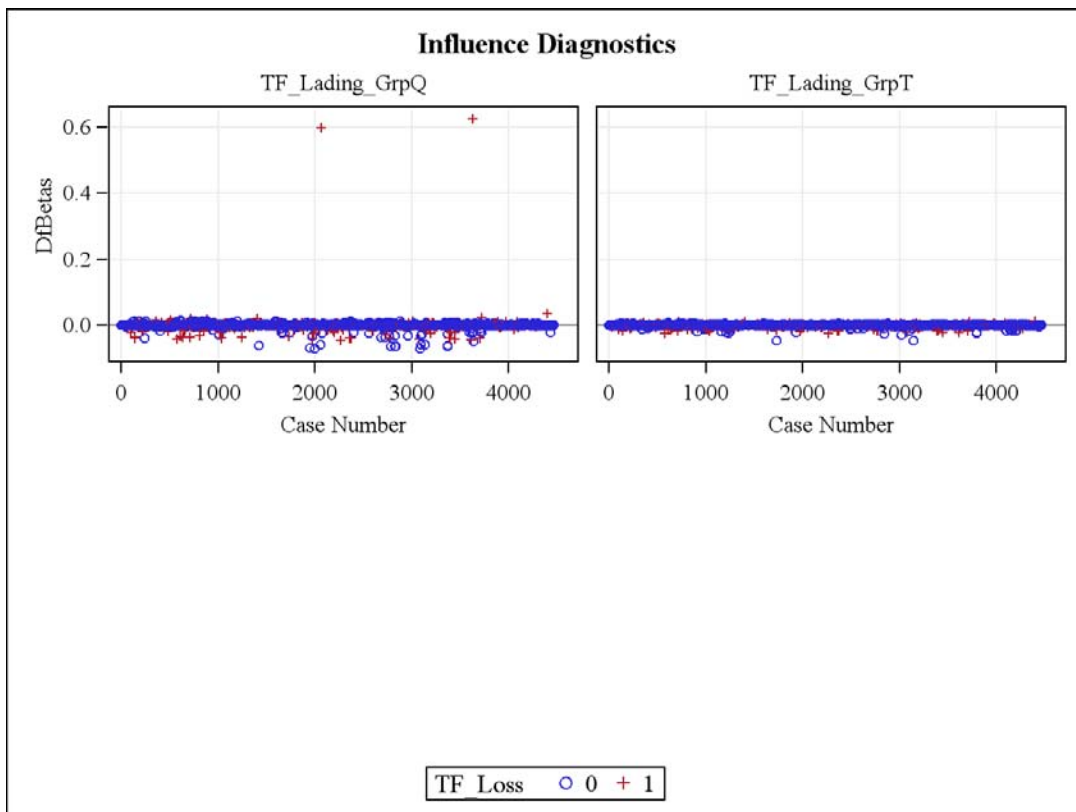
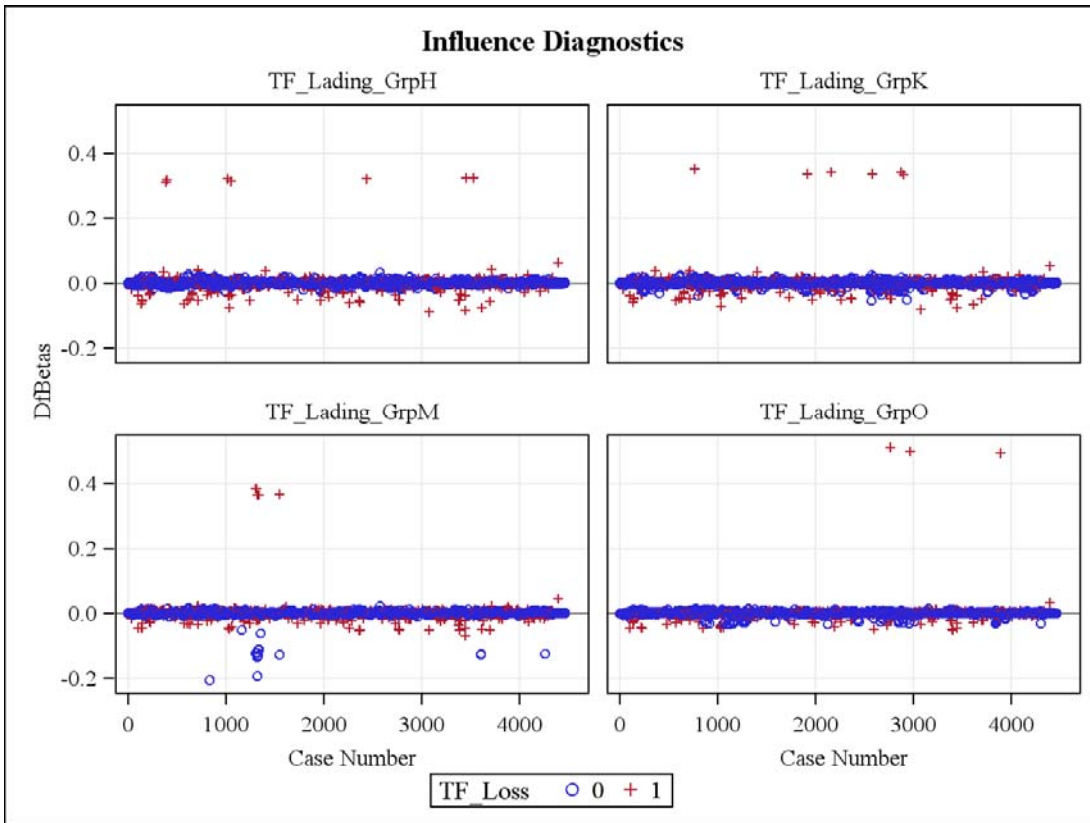












## Bottom Fittings

Model Information	
Data Set	WORK.BFCPR2
Response Variable	BF_Loss
Number of Response Levels	2
Model	binary logit
Optimization Technique	Fisher's scoring
Likelihood Penalty	Firth's bias correction

Number of Observations Read	5484
Number of Observations Used	5484

Response Profile		
Ordered Value	BF_Loss	Total Frequency
1	0	5344
2	1	140

*Probability modeled is BF\_Loss='1'.*

Class Level Information							
Class	Value	Design Variables					
BF_Lading_Grp	A	0	0	0	0	0	0
	D	1	0	0	0	0	0
	E	0	1	0	0	0	0
	F	0	0	1	0	0	0
	J	0	0	0	1	0	0
	K	0	0	0	0	1	0
	N	0	0	0	0	0	1

<b>Intercept-Only Model Convergence Status</b>
Convergence criterion (GCONV=1E-8) satisfied.

<b>Model Convergence Status</b>
Convergence criterion (GCONV=1E-8) satisfied.

<b>Model Fit Statistics</b>		
<b>Criterion</b>	<b>Intercept Only</b>	<b>Intercept and Covariates</b>
<b>AIC</b>	1208.387	1053.618
<b>SC</b>	1214.996	1159.372
<b>-2 Log L</b>	1206.387	1021.618

<b>Testing Global Null Hypothesis: BETA=0</b>			
<b>Test</b>	<b>Chi-Square</b>	<b>DF</b>	<b>Pr &gt; ChiSq</b>
<b>Likelihood Ratio</b>	184.7683	15	<.0001
<b>Score</b>	204.9466	15	<.0001
<b>Wald</b>	164.4706	15	<.0001

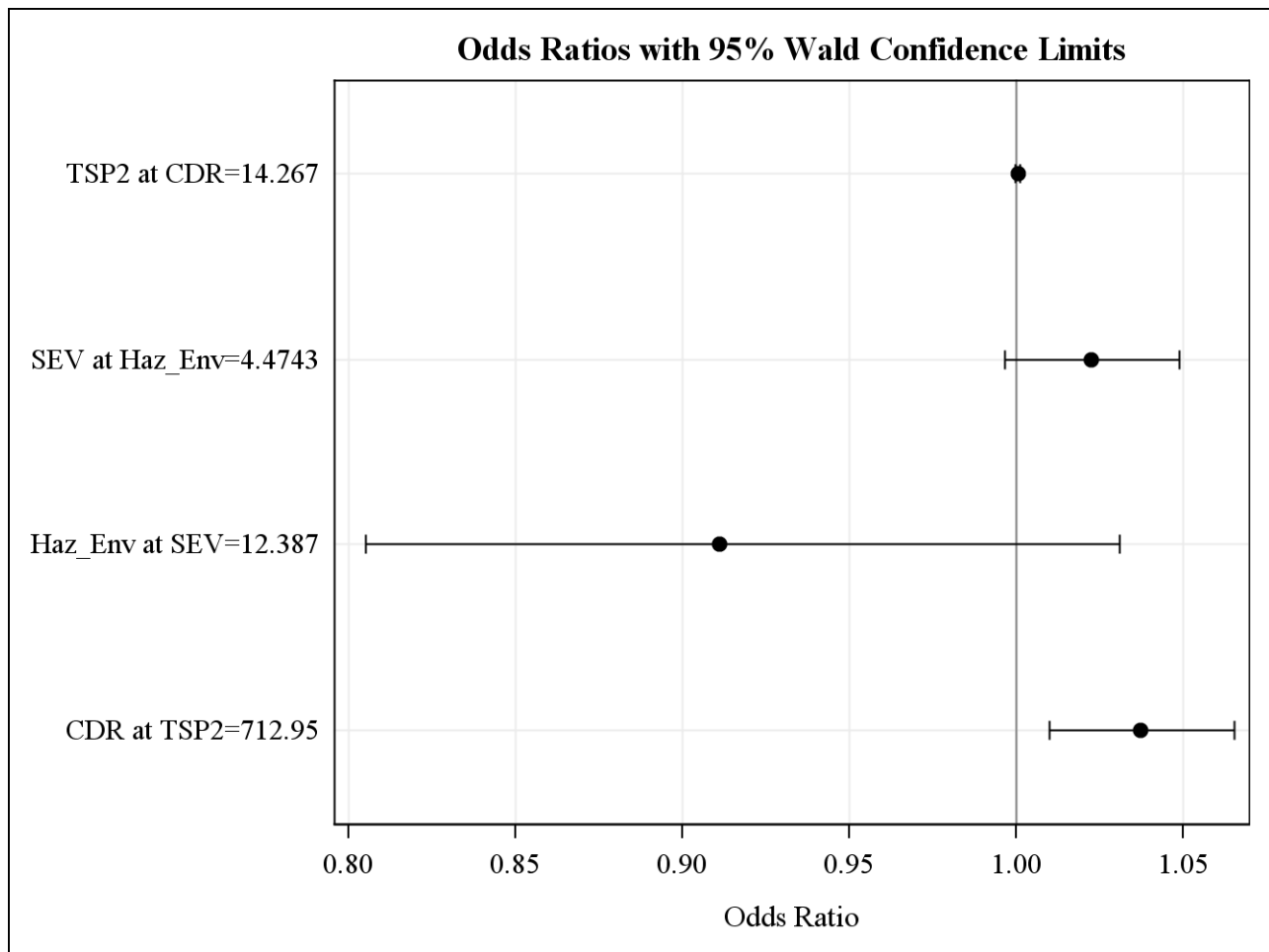
<b>Type 3 Analysis of Effects</b>			
<b>Effect</b>	<b>DF</b>	<b>Wald Chi-Square</b>	<b>Pr &gt; ChiSq</b>
<b>JKT</b>	1	14.3077	0.0002
<b>STS</b>	1	0.5032	0.4781
<b>TSP</b>	1	0.0156	0.9006
<b>TSP2</b>	1	2.5579	0.1097
<b>SEV</b>	1	1.6980	0.1926
<b>CDR</b>	1	9.0908	0.0026
<b>Haz_Env</b>	1	5.7701	0.0163
<b>TSP2*CDR</b>	1	8.1470	0.0043

Type 3 Analysis of Effects			
Effect	DF	Wald Chi-Square	Pr > ChiSq
SEV*Haz_Env	1	8.0248	0.0046
BF_Lading_Grp	6	56.9025	<.0001

Analysis of Maximum Likelihood Estimates						
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept		1	-2.3275	3.2833	0.5025	0.4784
JKT		1	-0.8730	0.2308	14.3077	0.0002
STS		1	-5.2546	7.4073	0.5032	0.4781
TSP		1	0.00328	0.0263	0.0156	0.9006
TSP2		1	0.000846	0.000529	2.5579	0.1097
SEV		1	-0.0275	0.0211	1.6980	0.1926
CDR		1	0.0578	0.0192	9.0908	0.0026
Haz_Env		1	-0.2307	0.0960	5.7701	0.0163
TSP2*CDR		1	-0.00003	0.000010	8.1470	0.0043
SEV*Haz_Env		1	0.0111	0.00392	8.0248	0.0046
BF_Lading_Grp	D	1	0.3864	1.1245	0.1181	0.7311
BF_Lading_Grp	E	1	1.7277	0.2664	42.0498	<.0001
BF_Lading_Grp	F	1	1.7051	0.3310	26.5384	<.0001
BF_Lading_Grp	J	1	2.7727	0.9190	9.1030	0.0026
BF_Lading_Grp	K	1	1.5623	0.4157	14.1269	0.0002
BF_Lading_Grp	N	1	0.9318	0.3887	5.7454	0.0165

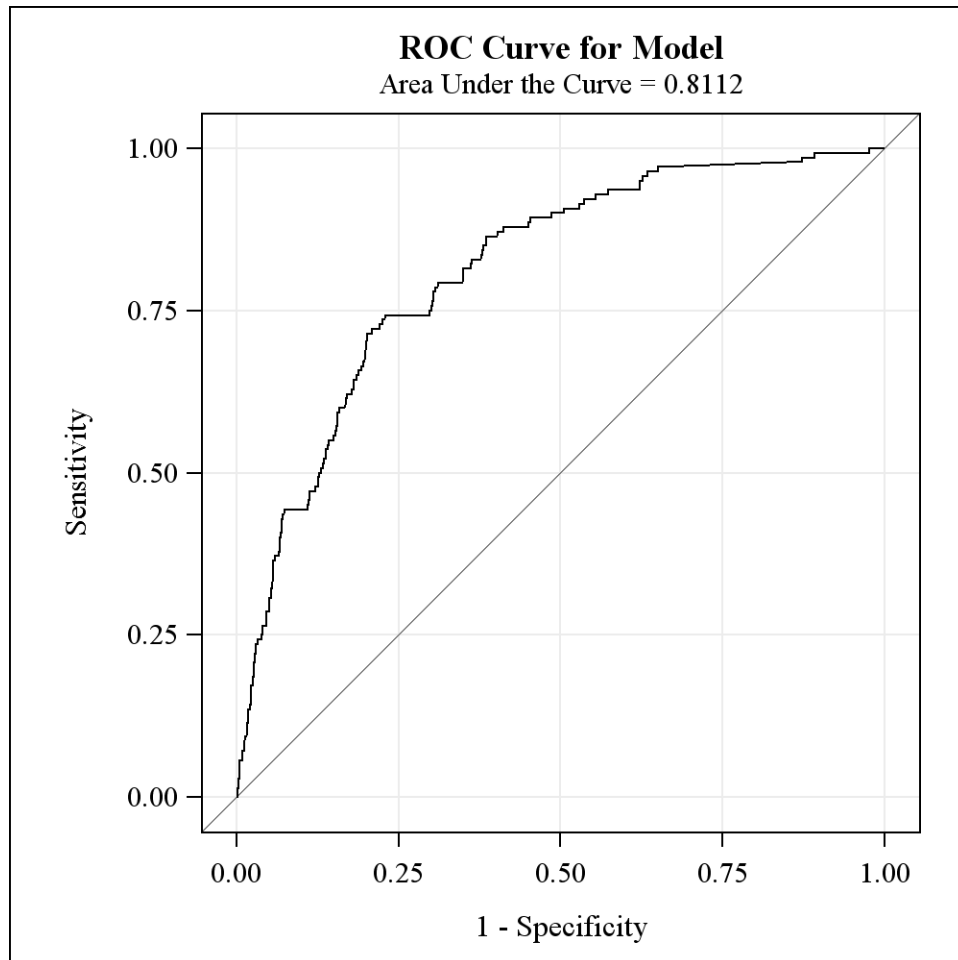
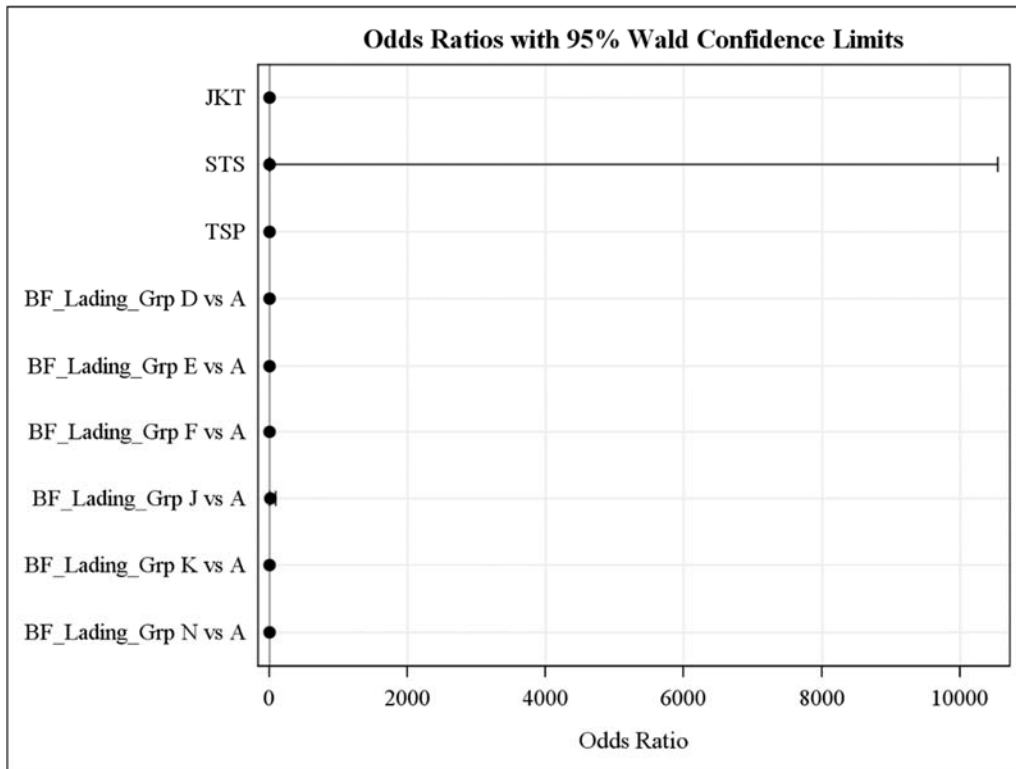
Association of Predicted Probabilities and Observed Responses			
Percent Concordant	81.0	Somers' D	0.622
Percent Discordant	18.8	Gamma	0.624
Percent Tied	0.2	Tau-a	0.031
Pairs	748160	c	0.811

Odds Ratio Estimates and Wald Confidence Intervals			
Label	Estimate	95% Confidence Limits	
TSP2 at CDR=14.267	1.000	1.000	1.001
SEV at Haz_Env=4.4743	1.022	0.997	1.049
Haz_Env at SEV=12.387	0.911	0.805	1.031
CDR at TSP2=712.95	1.037	1.010	1.065



Parameter Estimates and Wald Confidence Intervals				
Parameter		Estimate	95% Confidence Limits	
Intercept		-2.3275	-8.7626	4.1076
JKT		-0.8730	-1.3254	-0.4207
STS		-5.2546	-19.7726	9.2634
TSP		0.00328	-0.0482	0.0548
TSP2		0.000846	-0.00019	0.00188
SEV		-0.0275	-0.0690	0.0139
CDR		0.0578	0.0202	0.0954
Haz_Env		-0.2307	-0.4190	-0.0425
TSP2*CDR		-0.00003	-0.00005	-9.35E-6
SEV*Haz_Env		0.0111	0.00342	0.0188
BF_Lading_Grp	D	0.3864	-1.8175	2.5904
BF_Lading_Grp	E	1.7277	1.2055	2.2499
BF_Lading_Grp	F	1.7051	1.0564	2.3538
BF_Lading_Grp	J	2.7727	0.9715	4.5739
BF_Lading_Grp	K	1.5623	0.7476	2.3770
BF_Lading_Grp	N	0.9318	0.1699	1.6937

Odds Ratio Estimates and Wald Confidence Intervals				
Effect	Unit	Estimate	95% Confidence Limits	
JKT	1.000	0.418	0.266	0.657
STS	1.000	0.005	<0.001	>999.999
TSP	1.000	1.003	0.953	1.056
BF_Lading_Grp D vs A	1.000	1.472	0.162	13.335
BF_Lading_Grp E vs A	1.000	5.628	3.338	9.487
BF_Lading_Grp F vs A	1.000	5.502	2.876	10.525
BF_Lading_Grp J vs A	1.000	16.001	2.642	96.917
BF_Lading_Grp K vs A	1.000	4.770	2.112	10.772
BF_Lading_Grp N vs A	1.000	2.539	1.185	5.440



Estimated Correlation Matrix									
Parameter	Intercept	JKT	STS	TSP	TSP2	SEV	CDR	Haz_Env	TSP2CDR
Intercept	1.0000	-0.1025	-0.9896	-0.0331	0.0547	-0.0856	0.0454	-0.0952	-0.0613
JKT	-0.1025	1.0000	0.0563	0.0054	-0.0365	0.0267	-0.0924	0.0289	0.0968
STS	-0.9896	0.0563	1.0000	-0.0009	-0.0233	0.0203	-0.0383	0.0157	0.0475
TSP	-0.0331	0.0054	-0.0009	1.0000	-0.8785	-0.1803	-0.4684	-0.2672	0.4433
TSP2	0.0547	-0.0365	-0.0233	-0.8785	1.0000	-0.0886	0.6452	0.0872	-0.7409
SEV	-0.0856	0.0267	0.0203	-0.1803	-0.0886	1.0000	-0.2408	0.6827	0.2978
CDR	0.0454	-0.0924	-0.0383	-0.4684	0.6452	-0.2408	1.0000	-0.2158	-0.8308
Haz_Env	-0.0952	0.0289	0.0157	-0.2672	0.0872	0.6827	-0.2158	1.0000	0.2194
TSP2CDR	-0.0613	0.0968	0.0475	0.4433	-0.7409	0.2978	-0.8308	0.2194	1.0000
SEVHaz_Env	0.0786	-0.0172	-0.0103	0.2090	-0.0833	-0.7890	0.1171	-0.8150	-0.2098
BF_Lading_GrpD	0.8309	0.0822	-0.8565	0.0034	0.0148	-0.0057	-0.0012	-0.0120	-0.0239
BF_Lading_GrpE	-0.0604	0.4235	-0.0143	0.0156	-0.0283	0.0138	-0.0053	0.0081	0.0371
BF_Lading_GrpF	-0.0016	-0.2183	-0.0363	0.0370	-0.0080	0.0200	0.0561	-0.0171	-0.0485
BF_Lading_GrpJ	-0.0223	-0.0382	0.0011	0.0457	-0.0503	-0.0108	-0.0070	0.0242	0.0290
BF_Lading_GrpK	-0.0238	-0.1632	-0.0062	0.0931	-0.0745	-0.0103	-0.0704	-0.0036	0.0561
BF_Lading_GrpN	-0.0657	0.2186	0.0191	0.0204	-0.0626	0.0541	-0.0841	0.0286	0.0827

Estimated Correlation Matrix				
Parameter	SEVHaz_Env	BF_Lading_GrpD	BF_Lading_GrpE	BF_Lading_GrpF
Intercept	0.0786	0.8309	-0.0604	-0.0016
JKT	-0.0172	0.0822	0.4235	-0.2183
STS	-0.0103	-0.8565	-0.0143	-0.0363
TSP	0.2090	0.0034	0.0156	0.0370
TSP2	-0.0833	0.0148	-0.0283	-0.0080
SEV	-0.7890	-0.0057	0.0138	0.0200
CDR	0.1171	-0.0012	-0.0053	0.0561
Haz_Env	-0.8150	-0.0120	0.0081	-0.0171
TSP2CDR	-0.2098	-0.0239	0.0371	-0.0485
SEVHaz_Env	1.0000	0.0049	-0.0030	-0.0164
BF_Lading_GrpD	0.0049	1.0000	0.2121	0.1202
BF_Lading_GrpE	-0.0030	0.2121	1.0000	0.4110
BF_Lading_GrpF	-0.0164	0.1202	0.4110	1.0000



Estimated Correlation Matrix				
Parameter	SEVHaz_Env	BF_Lading_GrpD	BF_Lading_GrpE	BF_Lading_GrpF
BF_Lading_GrpJ	0.0233	0.0337	0.1694	0.1542
BF_Lading_GrpK	0.0096	0.0792	0.3282	0.3572
BF_Lading_GrpN	0.0027	0.1135	0.5169	0.2916

Estimated Correlation Matrix			
Parameter	BF_Lading_GrpJ	BF_Lading_GrpK	BF_Lading_GrpN
Intercept	-0.0223	-0.0238	-0.0657
JKT	-0.0382	-0.1632	0.2186
STS	0.0011	-0.0062	0.0191
TSP	0.0457	0.0931	0.0204
TSP2	-0.0503	-0.0745	-0.0626
SEV	-0.0108	-0.0103	0.0541
CDR	-0.0070	-0.0704	-0.0841
Haz_Env	0.0242	-0.0036	0.0286
TSP2CDR	0.0290	0.0561	0.0827
SEVHaz_Env	0.0233	0.0096	0.0027
BF_Lading_GrpD	0.0337	0.0792	0.1135
BF_Lading_GrpE	0.1694	0.3282	0.5169
BF_Lading_GrpF	0.1542	0.3572	0.2916
BF_Lading_GrpJ	1.0000	0.1286	0.1201
BF_Lading_GrpK	0.1286	1.0000	0.2403
BF_Lading_GrpN	0.1201	0.2403	1.0000

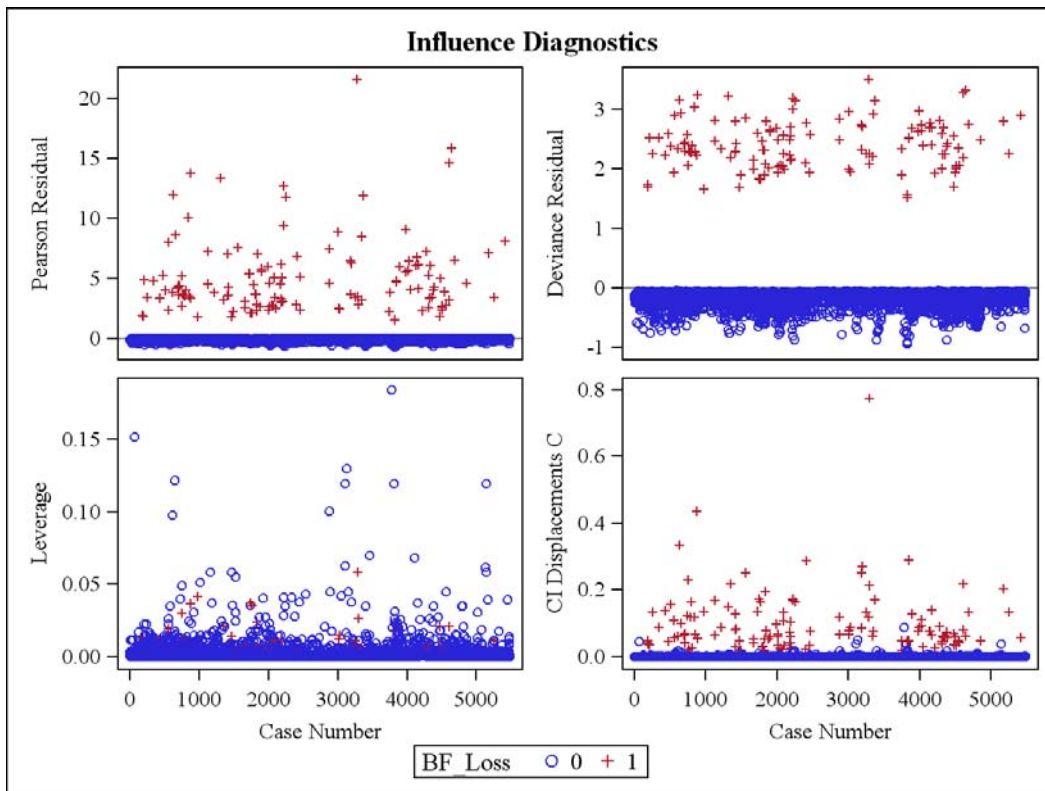
Partition for the Hosmer and Lemeshow Test					
Group	Total	BF_Loss = 1		BF_Loss = 0	
		Observed	Expected	Observed	Expected
1	548	1	1.53	547	546.47
2	212	2	0.94	210	211.06
3	1056	1	5.58	1055	1050.42
4	548	6	4.32	542	543.68
5	548	5	7.12	543	540.88
6	549	11	9.82	538	539.18

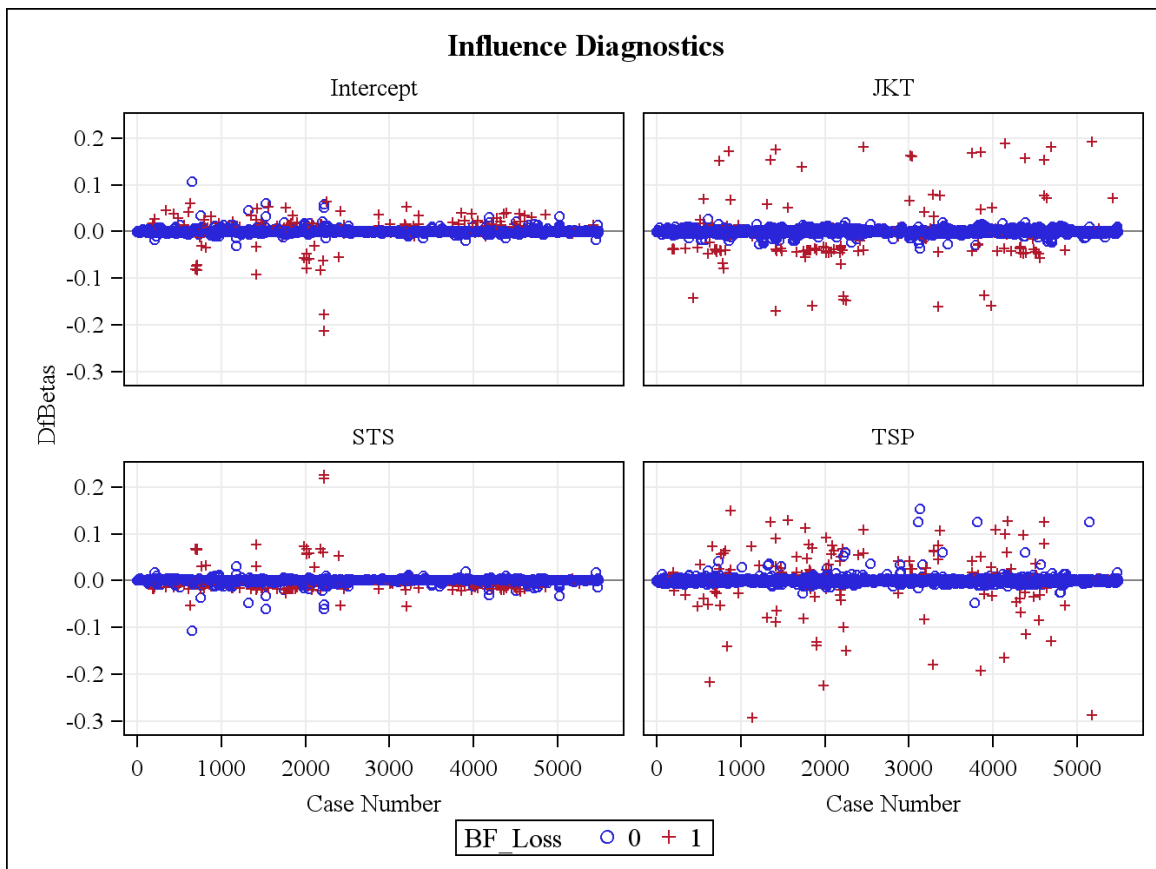
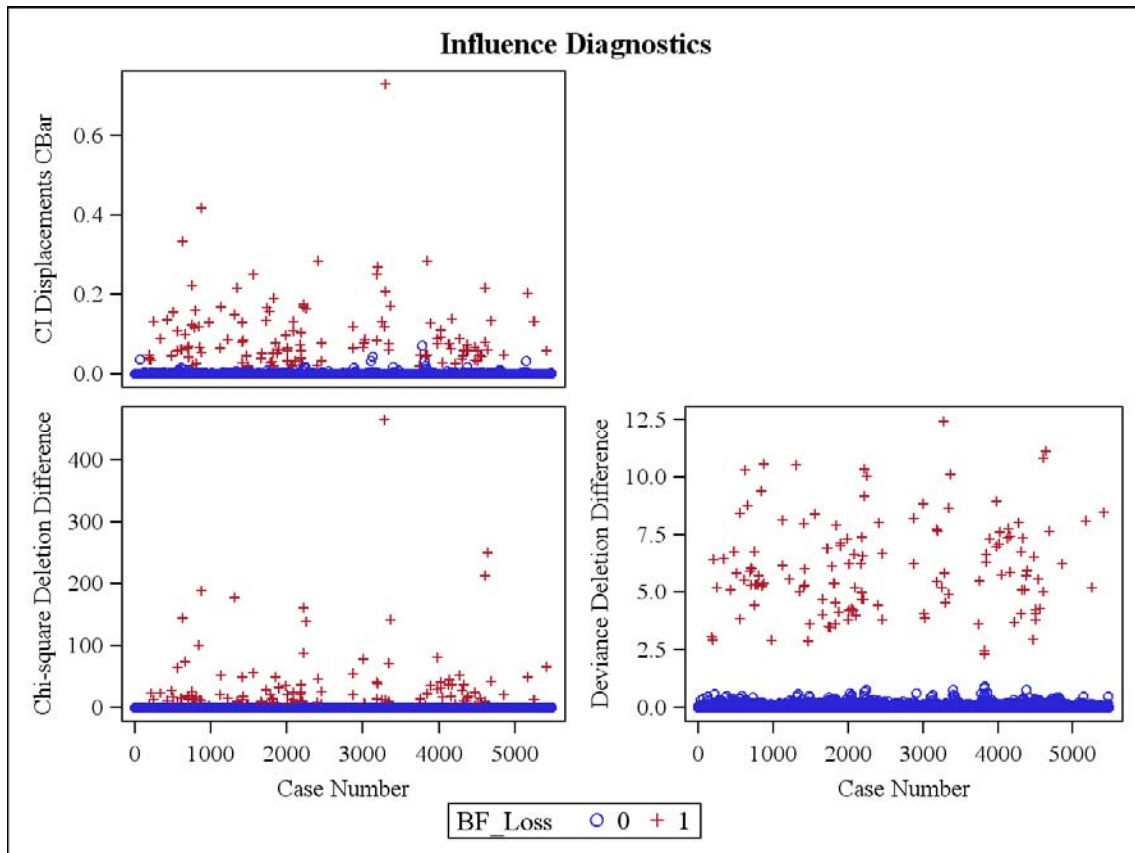
Partition for the Hosmer and Lemeshow Test					
Group	Total	BF_Loss = 1		BF_Loss = 0	
		Observed	Expected	Observed	Expected
7	548	10	14.40	538	533.60
8	548	21	18.63	527	529.37
9	548	31	33.81	517	514.19
10	379	52	50.66	327	328.34

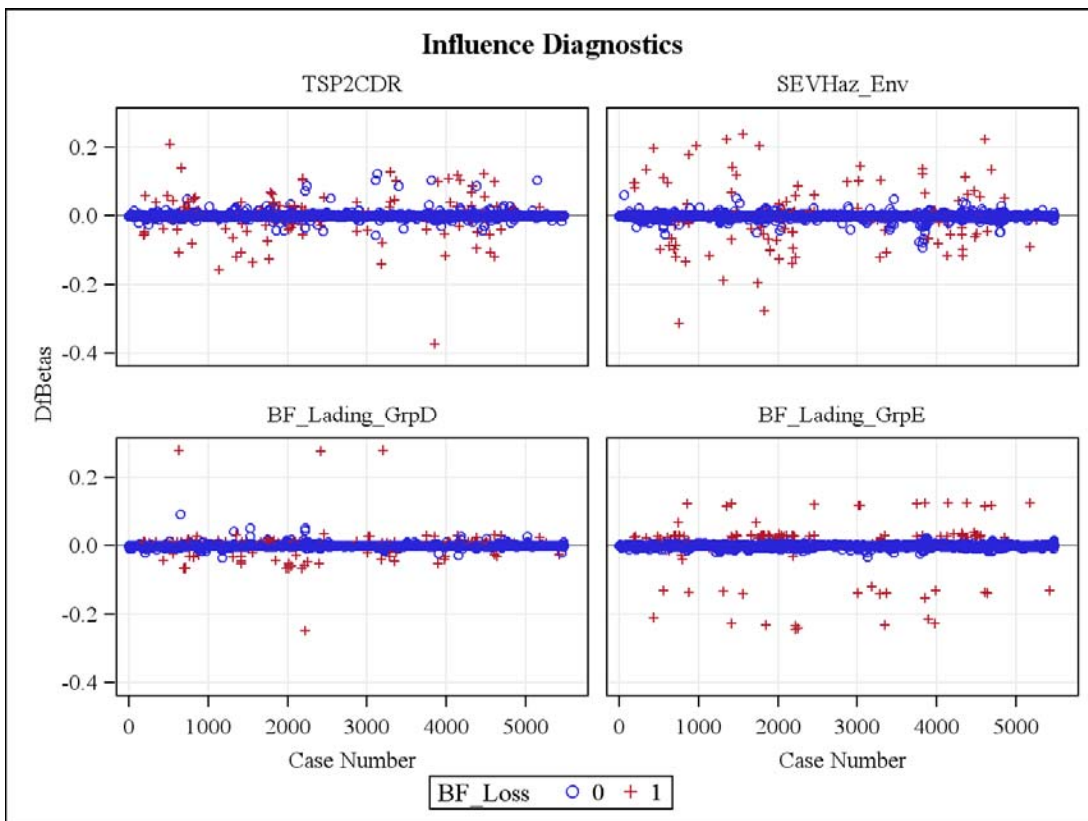
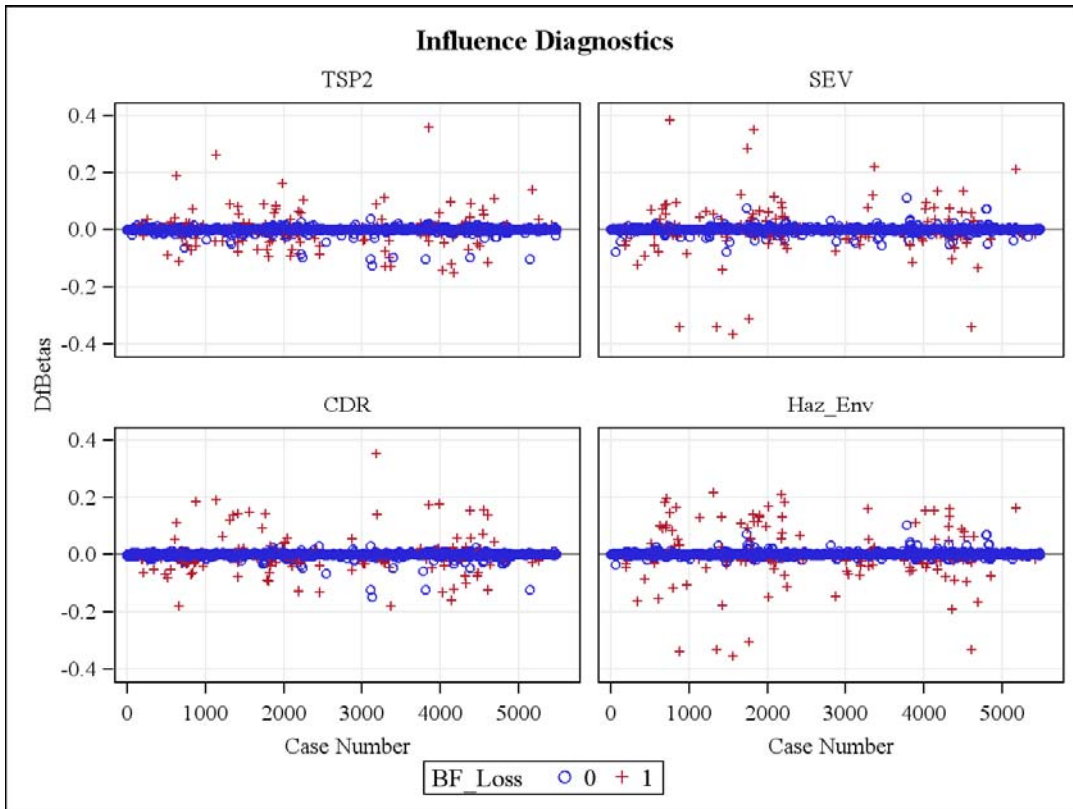
Hosmer and Lemeshow Goodness-of-Fit Test		
Chi-Square	DF	Pr > ChiSq
8.5866	8	0.3784

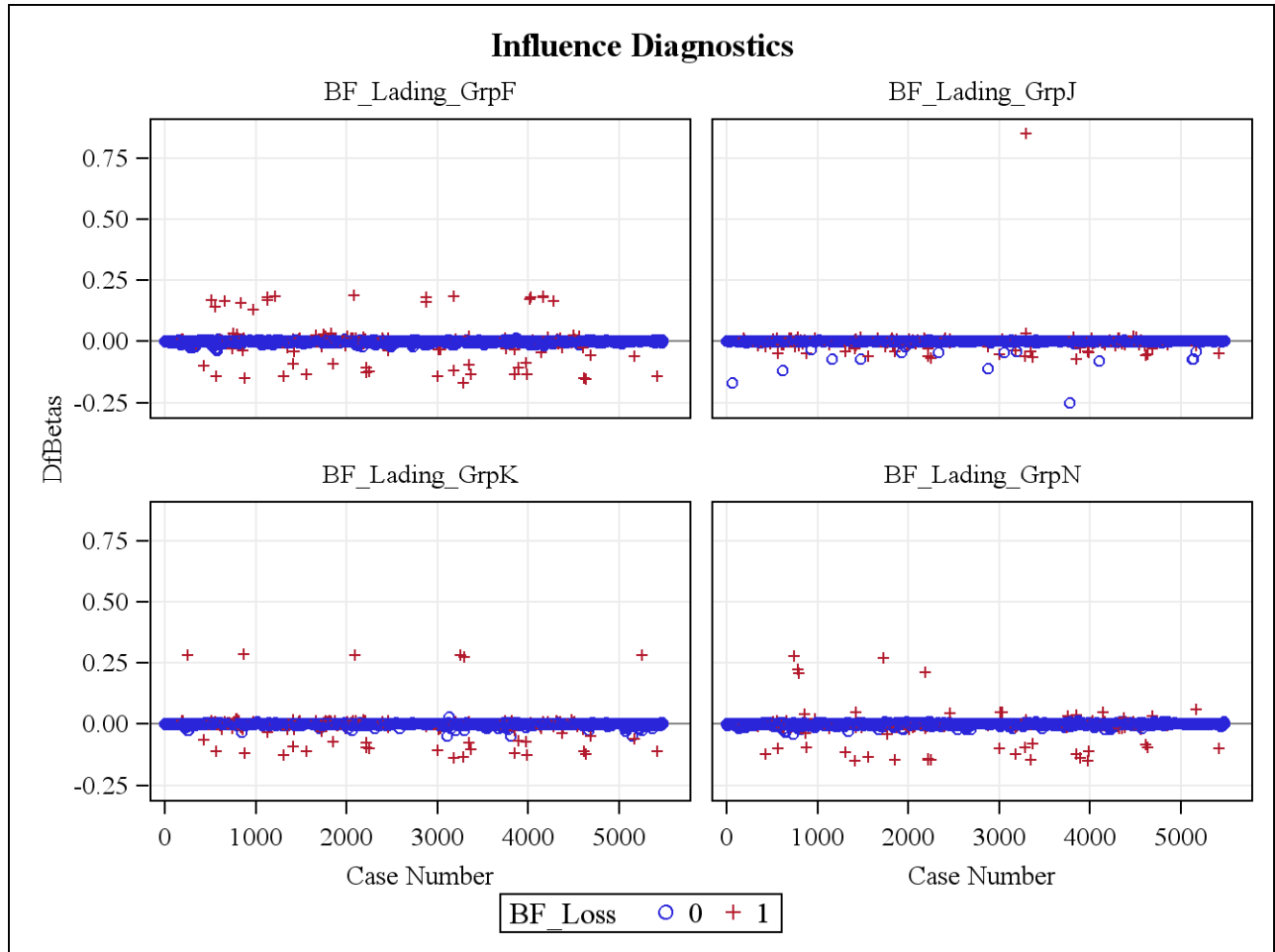
Classification Table									
Prob Level	Correct		Incorrect		Percentages				
	Event	Non-Event	Event	Non-Event	Correct	Sensi-tivity	Speci-ficity	False POS	False NEG
0.000	140	0	5344	0	2.6	100.0	0.0	97.4	.
0.020	112	3321	2023	28	62.6	80.0	62.1	94.8	0.8
0.040	83	4399	945	57	81.7	59.3	82.3	91.9	1.3
0.060	63	4672	672	77	86.3	45.0	87.4	91.4	1.6
0.080	39	5034	310	101	92.5	27.9	94.2	88.8	2.0
0.100	31	5135	209	109	94.2	22.1	96.1	87.1	2.1
0.120	26	5189	155	114	95.1	18.6	97.1	85.6	2.1
0.140	17	5224	120	123	95.6	12.1	97.8	87.6	2.3
0.160	12	5260	84	128	96.1	8.6	98.4	87.5	2.4
0.180	8	5287	57	132	96.6	5.7	98.9	87.7	2.4
0.200	8	5303	41	132	96.8	5.7	99.2	83.7	2.4
0.220	6	5311	33	134	97.0	4.3	99.4	84.6	2.5
0.240	2	5321	23	138	97.1	1.4	99.6	92.0	2.5
0.260	2	5332	12	138	97.3	1.4	99.8	85.7	2.5
0.280	2	5334	10	138	97.3	1.4	99.8	83.3	2.5
0.300	1	5337	7	139	97.3	0.7	99.9	87.5	2.5
0.320	0	5337	7	140	97.3	0.0	99.9	100	2.6

Classification Table									
Prob Level	Correct		Incorrect		Percentages				
	Event	Non-Event	Event	Non-Event	Correct	Sensi- tivity	Speci- ficity	False POS	False NEG
0.340	0	5341	3	140	97.4	0.0	99.9	100	2.6
0.360	0	5343	1	140	97.4	0.0	100.0	100	2.6
0.380	0	5344	0	140	97.4	0.0	100.0	.	2.6









## Appendix D Uncertainty Surrounding CPR

All estimates/predictions in the study are accompanied by uncertainty. The methods and the final models were chosen to limit and, to the extent possible, minimize this uncertainty, but it cannot be eliminated.

To assist in placing the appropriate weight on the estimates for decision-making purposes, it can be worthwhile to examine a confidence interval around key CPRs.

A large-sample 95% confidence interval for s prediction is given by

$$(L^{lo} = \hat{L} - 1.96\hat{\sigma}_L, L^{hi} = \hat{L} + 1.96\hat{\sigma}_L) \quad (15)$$

This transforms to a confidence interval for P of

$$\left( \frac{e^{L^{lo}}}{1 + e^{L^{lo}}}, \frac{e^{L^{hi}}}{1 + e^{L^{hi}}} \right) \quad (16)$$

When comparing two CPRs, if one falls within the confidence interval around the other, then the two estimates are not statistically distinct, i.e., there is no evidence that the true CPRs for those two cars are actually different.

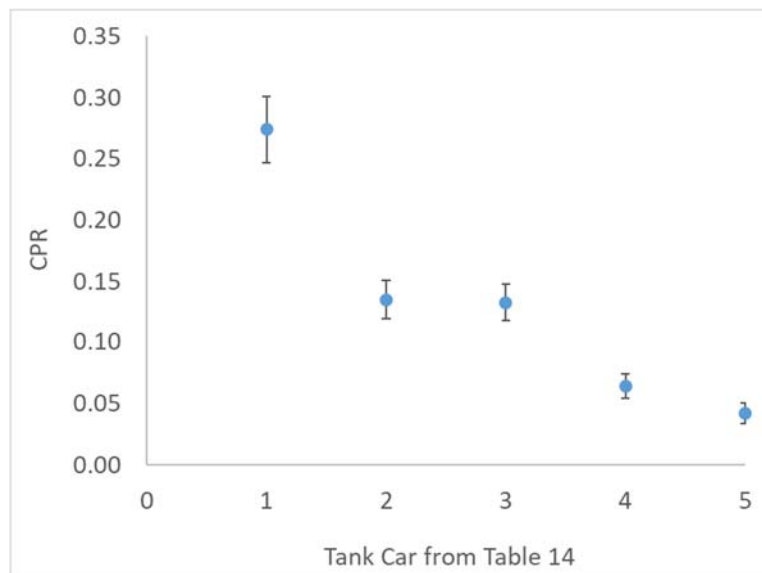
Table 12 provides examples.

Table 12  
CPR Estimates and Associated Confidence Intervals for Five Tank Car Configurations

	Car	Estimated Mainline CPR	Lower 95% Confidence Bound	Higher 95% Confidence Bound
1	111A100W1, 7/16" tank, no jacket, bottom fittings, 119" ID	0.274	0.239	0.293
2	111A100W1, 7/16" tank, jacket, bottom fittings, 119" ID	0.134	0.119	0.151
3	CPC-1232, 1/2" tank, no jacket, half-height head shield, top fittings protection, bottom fittings, 119" ID	0.132	0.117	0.148
4	CPC-1232, 7/16" tank, jacket, full head shield, top fittings protection, bottom fittings, 119" ID	0.064	0.054	0.074
5	DOT-117, 9/16" tank, jacket, full head shield, top fittings protection, bottom fittings, 119" ID	0.042	0.034	0.050

Figure 9 shows the same data, and makes clear that cars 2 and 3, for which the CPRs fall within each other's confidence intervals, are not statistically significantly different in CPR, but the CPR differences between all other pairings are statistically significant.

Figure 9  
CPRs for Five Tank Cars, with 95% Confidence Intervals





Appendix E  
Calculation of the FRA Accident Data Compatibility Multiplier

Tank cars are entered in TCAD when they experience damage to tank-car-specific features in an accident. Therefore, if the fraction (cars releasing lading)/(cars in the database) is calculated, this results in an estimate of the probability of lading loss, conditional on being in the database, i.e., being damaged.

However, such an estimate is not particularly useful in a broader risk assessment or cost-benefit analysis. It cannot be connected to accident rate data, for example. If the denominator can be converted into the number of tank cars derailed in FRA-reportable accidents, on the other hand, then the resulting probability estimate can be combined with published FRA accident rates. Such a conversion would account for tank cars that are derailed in reportable accidents, but not damaged on the tank, jacket, fittings, or head shield. Also, tank cars that are in TCAD but derailed in accidents that did not meet the FRA cost threshold for reporting would also be accounted for.

Two approaches to estimating the adjustment, which takes the form of a multiplicative factor referred to as the FRA multiplier, are explored below.

The first approach uses a sample of data from Burlington Northern Santa Fe Railway (BNSF) covering the period 2000-2010. In this dataset, it was known that each record corresponded to an FRA-reportable accident, and that these were all of the tank cars derailed in FRA-reportable accidents on the BNSF system in that period. All data were for mainline accidents. These data were compared to data in TCAD that were associated with accidents on BNSF.

The maximum likelihood estimator (MLE) of the ratio between BNSF/FRA and TCAD loss probability estimates is calculated using equation (17):

$$\hat{M} = \frac{CPR_{FRA}}{CPR_{SP}} = \frac{L_{FRA}/N_{FRA}}{L_{SP}/N_{SP}} \quad (17)$$

where:

- $\hat{M}$  = estimated FRA multiplier
- $L_{FRA}$  = number of tank cars with lading loss in FRA-reportable accidents,
- $N_{FRA}$  = number of tank cars derailed in the FRA-reportable accidents.
- $L_{SP}$  = number of tank cars with lading loss in the TCAD data, and
- $N_{SP}$  = number of tank cars in the TCAD data

The  $(100-\alpha)\%$  large sample asymptotic confidence interval of estimated multiplier can be expressed as:

$$\exp(\ln \hat{M} \pm z_{1-\alpha/2} SE) \tag{18}$$

where

$$SE = \sqrt{\frac{1}{L_{FRA}} - \frac{1}{N_{FRA}} + \frac{1}{L_{SP}} - \frac{1}{N_{SP}}} \tag{19}$$

Data for the BNSF mainline accidents in 2000-2010 were as follows:

	BNSF Records Known to Be FRA-Reportable	All TCAD BNSF Records
Cars with loss, L	115	66
Cars in dataset, N	470	355
CPR, L/N	0.245	0.186

The estimated FRA multiplier for mainlines arising from this procedure is

$$\hat{M} = \frac{CPR_{FRA}}{CPR_{SP}} = \frac{L_{FRA}/N_{FRA}}{L_{SP}/N_{SP}} = 1.32 \tag{20}$$

with a corresponding 95% confidence interval is (1.01, 1.72). The 99% confidence interval is (0.92, 1.88). Note that the value of one falls within the 99% interval, but not the 95% interval. The P-value for a test of difference from one is 0.046. Whether there is statistically significant evidence that the mainline multiplier is different from one is dependent on the analyst's selection of significance level. At 5% significance level, the multiplier appears to be greater than one. At 1% significance level, no difference is found.

It was decided to consider this analysis as insufficient evidence of a multiplier different from one, for the following reasons:

- 1) The p-value is borderline,
- 2) The form of the confidence interval is defensible for sufficiently large sample sizes, and incurs some uncertainty at moderate sample sizes,
- 3) It is unclear how representative BNSF data for the whole rail industry, and
- 4) The observations are not, strictly speaking, independent, since some come from the same accident.

A second approach was to attempt to assign a FRA-reportable status to a larger group of cars in TCAD for 2000-2010. Some railroads indirectly inform the Tank Car Safety Project when an accident was reported to FRA by using the same report submission number for their FRA report and for information provided to the Project. Tank cars identified this way, damaged in mainline accidents, broke down as follows:

	TCAD Records Known to Be FRA-Reportable	All TCAD Mainline Records for the Period
Cars with loss, L	303	376
Cars in dataset, N	1,310	1,699
CPR, L/N	0.231	0.221

For this dataset,

$$\hat{M} = \frac{CPR_{FRA}}{CPR_{SP}} = \frac{L_{FRA}/N_{FRA}}{L_{SP}/N_{SP}} = 1.05. \quad (21)$$

The corresponding 95% confidence Interval is (0.91, 1.19). The confidence interval contains the value one. Therefore, even at 95% confidence, there is no evidence of a difference between the loss ratios from FRA-reportable TCAD accidents and all of TCAD for mainlines. This result reinforces the earlier conclusion that the mainline FRA multiplier should be set at 1.

The previous study, RA-05-02 [Ref. 4], provided separate FRA multipliers for mainline/siding and yard/industrial track types. In this study the track type variable was included in the regressions directly, and found to be insignificant on its own. Therefore, it was deemed adequate to use appropriate train speeds and derailment sizes for the two different track types, and let those differences drive any difference in CPRs.

It is still possible that the degree of correlation between an accident being FRA-reportable and that accident causing TCAD-recordable damage could differ between track types. Further work could be done to resolve this. Also, the magnitude of multipliers and the need for multipliers could change over time so the term has been kept in the model in case such a need arises.

Note that the FRA multiplier does not have any bearing on comparisons among car configurations, as the percent difference between CPRs for different cars remains the same no matter what multiplier is used.

Appendix F  
Comparison of TWP-17 and RA-05-02 CPR Estimates

Table 13 compares the CPR estimates derived in this study to those for the same cars using the RA-05-02 formulas. Note that TWP-17 CPRs are consistently lower. This could be due to a number of factors that cannot be verified. The distribution of accident environments has probably changed over time. The robustness of materials and/or designs that are lumped under one term in this study, such as “TC-128” or “head shield”, may have generally improved over time. In any event, the lower rates of release are validated by the broader data in the Tank Car Safety project database; note that the mainline FRA multiplier is 1.0.

Table 13  
TWP-17 CPRs and RA-05-02 CPRs for Selected Tank Cars  
Under Average Mainline Derailment Conditions

Car Specification	Head and Shell Thickness (in.)	Jacket	Head Shield	Top Fittings Protection	Bottom Fittings	Shell Inside Diameter (in.)	TWP-17 CPR	RA-05-02 CPR	Difference
111A100W1	0.4375	No	No	No	Yes	119	0.274	0.353	-0.079
111A100W1/3	0.4375	Yes	No	No	Yes	119	0.134	0.207	-0.073
111A100W1/2	0.4375	No	No	No	No	119	0.265	0.310	-0.045
111A100W1/2/3	0.4375	Yes	No	No	No	119	0.130	0.170	-0.040
111A100W2	0.5625	No	No	Yes	No	100.625	0.201	0.202	-0.001
111A100W1	0.4375	Yes	Full	Yes	Yes	119	0.064	0.150	-0.086
111A/S100W1	0.500	No	Half	Yes	Yes	119	0.132	0.241	-0.109
117R100W	0.4375	Yes	Full	No	Yes	119	0.126	0.188	-0.062
117R100W	0.4375	Yes	Full	Yes	Yes	119	0.064	0.150	-0.086
117R100W	0.500	Yes	Full	No	Yes	119	0.106	0.176	-0.070
117R100W	0.500	Yes	Full	Yes	Yes	119	0.052	0.137	-0.085
117J100W	0.5625	Yes	Full	Yes	Yes	119	0.042	0.126	-0.084
112J340W	0.625	Yes	Full	Yes	No	118.75	0.033	0.076	-0.043
105J300W	0.5625	Yes	Full	Yes	No	118.75	0.040	0.086	-0.046
105A500W	0.779	Yes	No	Yes	No	102	0.040	0.073	-0.033
105J500W	0.797	Yes	Full	Yes	No	102	0.031	0.060	-0.029
112J500I	H1.03 / S0.89	Yes	Full	Yes	No	115.34	0.017	0.049	-0.032
105J600I	H1.136/S0.98	Yes	Full	Yes	No	106	0.016	0.047	-0.031