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# Crashes and near-crashes on horizontal curves along rural two-lane highways: Analysis of naturalistic driving data

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## ABSTRACT

Prior research has shown the probability of a crash occurring on horizontal curves to be significantly higher than on similar tangent segments, and a disproportionately higher number of curve-related crashes occurred in rural areas. Challenges arise when analyzing the safety of horizontal curves due to imprecision in integrating information as to the temporal and spatial characteristics of each crash with specific curves. The second Strategic Highway Research Program (SHRP 2) conducted a large-scale naturalistic driving study (NDS), which provides a unique opportunity to better understand the contributing factors leading to crash or near-crash events. This study utilizes high-resolution behavioral data from the NDS to identify factors associated with 108 safety critical events (i.e., crashes or near-crashes) on rural two-lane curves. A case-control approach is utilized wherein these events are compared to 216 normal, baseline-driving events. The variables examined in this study include driver demographic characteristics, details of the traffic environment and roadway geometry, as well as driver behaviors such as in-vehicle distractions. Logistic regression models are estimated to discern those factors affecting the likelihood of a driver being crash-involved. These factors include high-risk behaviors, such as speeding and visual distractions, as well as curve design elements and other roadway characteristics such as pavement surface conditions. This paper also discusses limitations and lessons learned from working with the SHRP 2 NDS data.

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## 1. Introduction

The safety of horizontal curves has drawn significant interest in the research literature (Bonneson et al., 2009; Felipe et al., 2007; Schurr, McCoy, Pesti, & Huff, 2002). Curves are shown to experience disproportionately higher crash rates than similar tangent roadways. The Federal Highway Administration (FHWA) notes that crash rates on horizontal curves are three times higher than crash rates on tangent segments (FHWA, 2014). A total of 7911 fatal crashes occurred on horizontal curves in 2015 (National Highway Traffic Safety Administration (NHTSA), 2015). These types of crashes account for about one quarter of all motor-vehicle fatalities annually in the United States. The majority of curve-related crashes occur in rural areas, especially on rural two-lane highways.

An extensive number of prior studies have sought to understand the relationship between crash risk and horizontal curve characteristics. This includes various studies that have investigated the relationships between crash frequency or severity and curve characteristics, such as radius/degree of the curve and the curve length (Council, 1998; Khan,

Bill, Chitturi, & Noyce, 2013; Schneider, Zimmerman, Savolainen, & Moore, 2010; Suh, Park, Park, & Chon, 2006; Torbic et al., 2004). Collectively, research has shown crashes occur more frequently and more severely on curves. Additional research has examined how these increased crash risks are affected by drivers' curve negotiation behavior from a human factors perspective. The research has shown drivers to have difficulty in effectively recognizing the presence and sharpness of upcoming curves due to visual distortion (Charlton, 2007; Hassan & Easa, 2003; Hong, Iwasaki, Furuichi, & Kadoma, 2006; Kandil, Rotter, & Lappe, 2010; Mars, 2008; Shinar, Mcdowell, & Rockwell, 1977; Suh et al., 2006; Wooldridge, Fitzpatrick, Koppa, & Bauer, 2000; Zakowska, 2000). Additional studies have focused on driver speed selection while navigating curves with various geometric characteristics. The research has shown mean and 85th percentile speeds to be impacted by the curve radius, the curve length, the tangent distance, and the speed limits. (Bella, 2007; Bonneson & Pratt, 2009; Collins, Fitzpatrick, Bauer, & Harwood, 2007; Donnell, Ni, Adolini, & Elefteriadou, 2001; Felipe & Navin, 2007; Fitzpatrick et al., 1999; Montella, Pariota, Galante, Imbriani, & Mauriello, 2014; Schurr et al., 2002). Several studies have also examined vehicle lateral position during curve navigation. These studies have contrasted patterns of curve trajectories, showing drivers to exhibit curve cutting, swinging, drifting, and correcting in contrast to ideal behavior (Abele & Møller, 2011; Bella, 2013; Bertola, Balk, & Shurbutt, 2012; Charlton, 2007; Gunay & Woodward, 2007; Hallmark,

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83 2014; Reymond, Kemeny, Droulez, & Berthoz, 2001; Spacek, 2005; Suh  
 Q28 Q29 et al., 2006; Taylor, Abu-Lebdeh, & Rai, 2005).

Q30 Q31 While above-mentioned studies have provided important insights  
 85 into the factors affecting safety on horizontal curves, some important  
 86 gaps remain in the literature. While driver behavior has been shown  
 87 to contribute to >90% of all crashes (NHTSA, 2008; Rumar, Evans, &  
 88 Schwing, 1985; Salmon, Regan, & Johnston, 2005; Spainhour, Brill,  
 89 Sobanjo, Wekezer, & Mtenga, 2005; Treat et al., 1979), research has  
 90 been somewhat limited as to how curve geometries affect driver behav-  
 91 ior and crash risk at a fine level of detail. For example, much of the  
 92 extant research in this area focuses on analyses of police-reported  
 93 crash data. However, crash report forms often have limited information  
 94 as to driver behavior during the period immediately preceding a crash.  
 95 This is exacerbated by the fact that crash reports are difficult to integrate  
 96 with associated roadway characteristics due to the lack of temporal and  
 97 spatial accuracy. Driver behavior on curves has also been assessed  
 98 through driving simulator studies, though these studies can be criticized  
 Q32 as to their validity in replicating real-world behaviors (Casali & Frank,  
 100 1986; Fisher, Rizzo, Caird, & Lee, 2011).

101 Naturalistic driving studies (NDS) have emerged as an innovative  
 102 means to observe real-world driving behavior with the potential to pro-  
 103 vide important insights that are not possible through alternative study  
 104 designs. NDS provides an unobtrusive observation method to study  
 105 driver behavior in natural settings without any experiment controls  
 Q33 (Dingus et al., 2014; PROLOGUE, 2009; Schagen & Sagberg, 2012).  
 107 Such studies allow researchers to observe how drivers naturally interact  
 108 with other vehicles, roadway characteristics, and traffic environments.

109 The study described in this paper utilizes data collected as a part of  
 110 the NDS conducted through the second Strategic Highway Research  
 111 Program (SHRP2) to better understand the role of driver behavior in  
 112 crash and near-crash events. The SHRP2 NDS involved continuous mon-  
 113 itoring of more than 3000 drivers, resulting in unparalleled information  
 114 as to how driver behavior varies with respect to the driving environ-  
 115 ment, providing unique opportunities to better understand how driver  
 116 behavior and other factors affect crash risk. The SHRP2 NDS provides  
 117 a variety of pre-crash information that is not available in police-  
 118 reported crash data. For example, details as to in-vehicle distractions  
 119 can be discerned from in-vehicle driver face cameras. External factors,  
 120 such as traffic flow and pavement conditions, can also be determined  
 121 from the accompanying forward road-view video. Detailed vehicle  
 122 information is also available through 10-Hz vehicle Controller Area  
 123 Network (CAN) bus signals, allowing for determination of speed, accel-  
 124 eration, yaw rate, and other kinematic data. These critical variables are  
 125 generally difficult to collect using traditional study designs. Ultimately,  
 126 the study demonstrates how NDS data can be exploited to examine  
 127 safety issues on rural two-lane curves.

## 128 2. Data

### 129 2.1. SHRP2 NDS overview

130 The SHRP2 NDS resulted in the collection of detailed data from  
 131 more than 3000 drivers from October 2010 to November 2013  
 Q34 (Dingus et al., 2014). More than 5 million trips and 1 million h of driving  
 133 data were collected across study sites in six states in the United States,  
 134 including Florida, Indiana, New York, North Carolina, Pennsylvania,  
 135 and Washington.

136 The data collected as a part of the SHRP2 NDS includes extensive  
 137 information detailing the drivers and vehicles involved in the study.  
 138 In addition, summary information is provided for a sample of the  
 139 5 million trips that were conducted over the course of the NDS. This  
 140 includes general information as to roadway and traffic characteristics,  
 141 as well as aggregate-level summaries of operational data (e.g., speed,  
 142 acceleration) and driver behavior (e.g., seatbelt use, cell phone use)  
 143 over the course of each trip. More detailed information is provided for  
 144 specific events, which are vehicle traces of approximately 20-second in

**Table 1**  
 Definitions of event types (SHRP2 NDS website, 2015).

Event type	Description	
Crash	Any contacts that the subject vehicle has with an object, either moving or fixed, at any speed in which kinetic energy is measurably transferred or dissipated. Also includes non-premeditated departures of the roadway where at least one tire leaves the paved or intended travel surface of the road.	t1.1 t1.2 t1.3 t1.4
Near-crash	Any circumstance that requires a rapid evasive maneuver by the subject vehicle or any other vehicle, pedestrian, cyclist, or animal to avoid a crash. Near Crashes must meet the following four criteria: 1) Not a Crash; 2) Not pre-meditated; 3) Evasion required; 4) Rapidity required.	t1.5
Safety critical event	The combination of all crash and near-crash events were referred as safety critical event in this study.	t1.6
Baseline	An epoch of data selected for comparison to any of the conflict types listed above rather than due to the presence of conflict. For SHRP2, these baselines are 21 s long and were randomly selected with a goal of 20,000 baselines, a minimum of 1 baseline per driver.	t1.7

length. Trips may include multiple events, with each event being 145  
 classified as a crash, near-crash, or baseline event. These events are 146  
 defined in Table 1. 147

Given the relatively small number of crash events occurring along 148  
 horizontal curves on two-lane highways within the NDS data, both 149  
 crash and near-crash events were combined and are classified as “safety 150  
 critical events” in the remainder of this paper. Several prior studies have 151  
 shown near-crash events to be a reasonable crash surrogate in natural- 152  
 istic driving studies (Wu, Aguero-Valverde, & Jovanis, 2014; Wu & 153  
 Jovanis, 2012, 2013). In addition to the safety critical events, a sample 154  
 of baseline events were included as a control to represent normal driv- 155  
 ing activities. The baseline events were sampled from normal driving 156  
 activities in proportion to the number of miles driven by each driver 157  
 over the course of the study period. An alternative is to use lane 158  
 encroachment as a surrogate for lane departure crashes, but the lateral 159  
 position measurement was found to contain large number of missing 160  
 values and outliers, so lane encroachment is not used in this study. 161

In addition to the general driver, vehicle, trip, and event information 162  
 described previously, forward roadway-view videos were also utilized 163  
 to examine pre-crash scenarios for better understanding the causes of 164  
 crash and near-crash events. The forward roadway view videos were 165  
 manually reviewed by our own research team to extract roadway and 166  
 traffic-related information. 167

Safety critical and baseline events are available on the SHRP2 InSight 168  
 website (<https://insight.shrp2nds.us/home>), which was developed and 169  
 is hosted by VTTI. Data reductionists at the Virginia Tech Transportation 170  
 Institute (VTTI) manually coded the event detail data. Over 76 variables 171  
 were reduced and are available including crash type, severity, driver 172  
 actions, presence of passengers, environmental factors, event narrative, 173  
 etc. A brief video clip of the forward roadway is included along with 174  
 graphical display of select vehicle kinematics (i.e., speed, acceleration, 175  
 distance into trip, wiper status). High-level roadway and traffic charac- 176  
 teristics are also included such as intersection type, traffic control, 177  
 alignment, and level of service. This information is provided in an 178  
 Event Detail Table, which can be sorted using various filters such as 179  
 the type of roadway. 180

Filters were used to first select roadway departure events. This 181  
 was accomplished using incident type (i.e., road departure left) or 182  
 precipitating event (i.e., subject over left lane line). A total of 1631 183  
 roadway departure crashes and near-crashes were identified from 184  
 the Event Detail Table. Next safety critical events were filtered by 185  
 setting number of lanes (i.e., Lanes = 2). No variable exists which 186  
 specifically identifies an event as rural so the variable for locality 187  
 was set to “open country” or “open residential.” Presence of a curve 188  
 was identified through a variable for alignment which indicated 189  
 curve left or curve right. 190

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