Time Series Analysis of Driver Behavior on Curves

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Introduction

- The number of crashes are disproportionally higher on curves
- Analysis of Driver Behavior on Rural Curves using the SHRP 2 NDS Data (Coming soon!)
- Time Series Analysis (Proof-of-concept)

Description of SHRP 2 NDS Data

- 3000 Drivers
- Six States
- Two Years
- 18 Million Traveled Miles
- 2 Petabytes Data
- Data were collected at 10 HZ

HIGHWAY SAFETY The SHRP 2 Naturalistic Driving Study Addressing Driver Performance and

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Composite image showing a driver's head position in relation to the dathboard and wind-

National Academies,

Washington, D.C.

Behavior in Traffic Safety KENNETH L. CAMPBELL

The central goal of the Naturalistic Driving Research Program (SHR 2) is to address the role of driver performance and behavior in traffic safety! This involves understanding how the driver interacts with and adapts to the vehicle, the traffic environment, roadway characteristics, traffic control devices, and other environmental features. The NDS also provides the means to assess the changes in collision risk associated with each of these factors and their interactions.

Driving behavior is a critical factor in nearly all traffic crashes. Driver impairment—primarily due to alcohol—and driver inattention, distraction, drowsiness, and judgment-related errors are believed to be responsible for significant increases in crash risk. After-the-fact crash investigations, however, cannot determine securately a driverk behavior before the The in-vehicle data acquisition system (DAS) unit gathers and stores data from forward radar, four

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- 60

video cameras, accelerometers, vehicle network information, a Geographic Positioning System, and onboard computer vision algorithms. tors. The larger context for exposure enables risk

estimates for various driver behaviors and for other contributing factors. The information will support the development of new and improved safety countermeasures to prevent traffic collisions and injuries.

http://www.trb.org/StrategicHighwayResearchProgram2SHRP2/Pages/The-SHRP-2-Naturalistic-Driving-Study-472.aspx



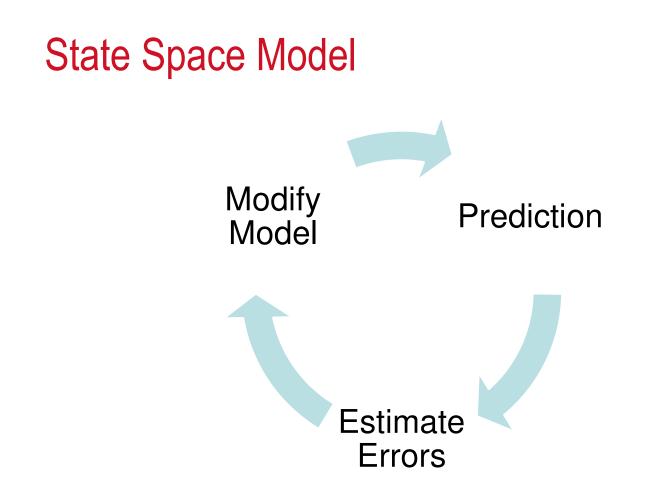


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State Space Model

- Explanatory variables and stochastic time component
- Explanatory variables can evolve over time
- Intervention effects is coded as dummy variable

Observation equation: $Y_t = F_t \theta_t + v_t$, with $v_t \sim N_1(0, V_t)$ State evolution equation: $\theta_t = G_t \theta_{t-1} + \omega_t$, with $\omega_t \sim N_p(0, W_t)$



State Space Model

1. Intervention Analysis

• How the driver interacts with traffic and roadway environment?

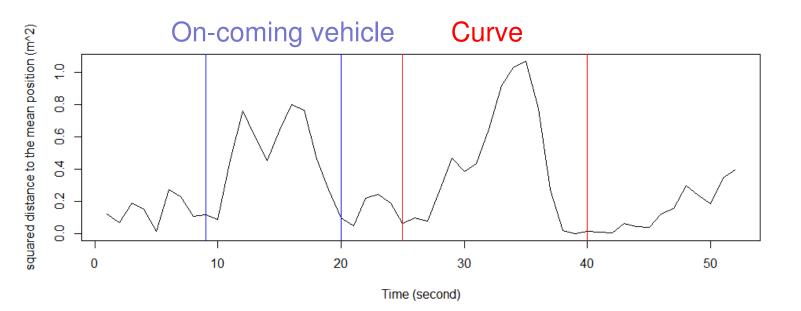
2. Forecasting Analysis

 Can we forecast future position based on past observations?

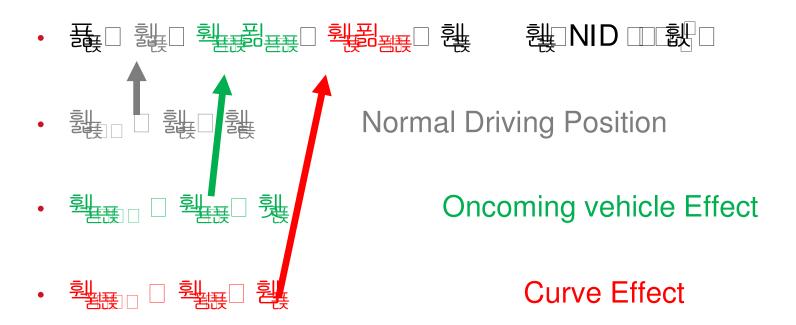
INTERVENTION ANALYSIS WITH STATE SPACE MODEL



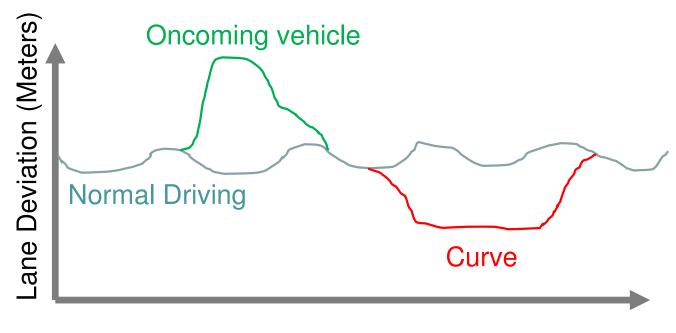
Deviation to the Normal Driving Position



State Space Model – Intervention Analysis

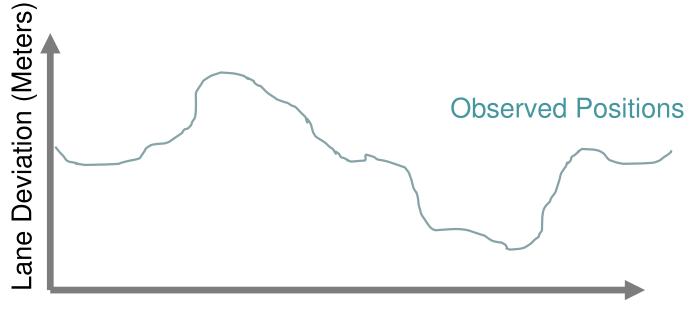


State Space Model – Intervention Analysis



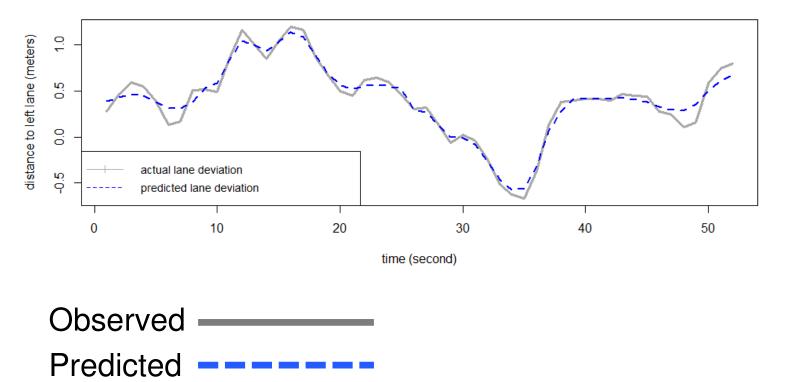
Time (seconds)

State Space Model – Intervention Analysis



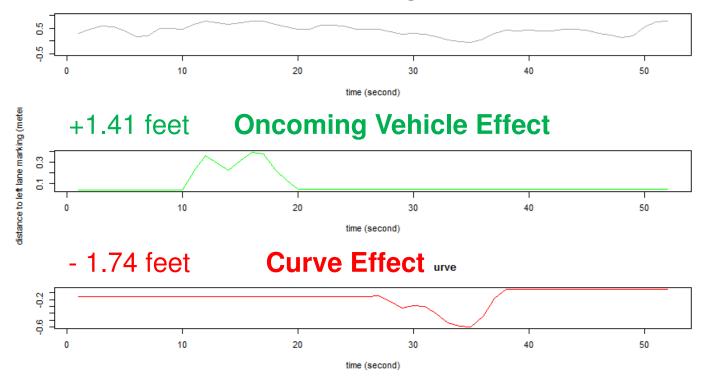
Time (seconds)

State Space Model—Modeling Results



Decomposition for Different Effects

1.44 feet Normal Driving Variation

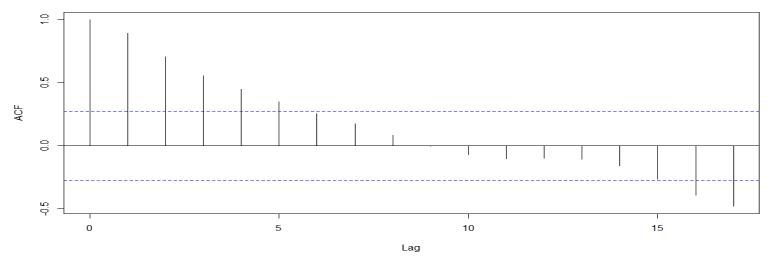




FORECASTING WITH STATE SPACE MODEL

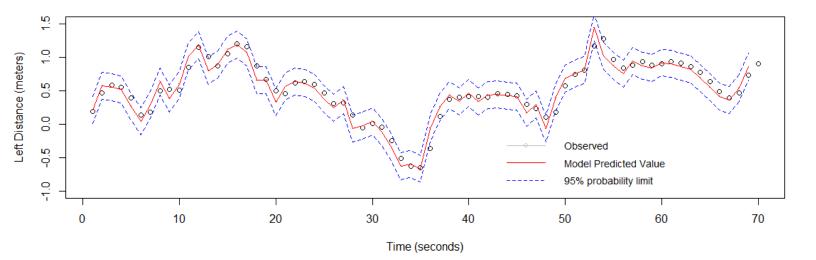
State Space Model - Forecasting

Predict vehicle's future position based on the past observations.

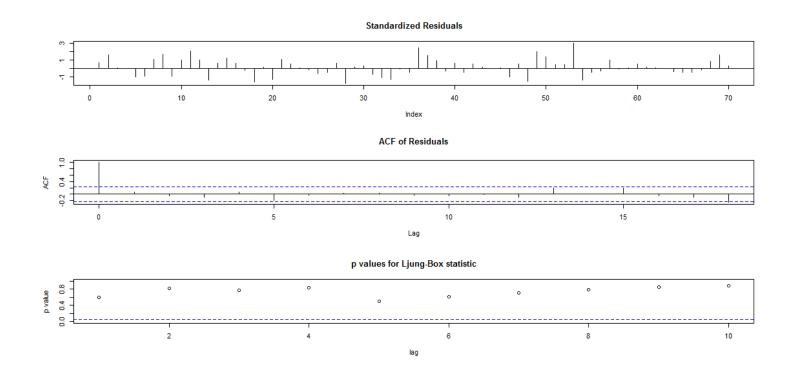


Autocorrelation Function

State Space Model - Forecasting



Model Diagnostics



Summary

Intervention Analysis evaluated the influence of the oncoming vehicle and the curve on driver behavior

Forecasting Analysis successfully predict the future positions

Limitation is difficulty to draw safety implications with statistical significance

Thanks for your attention!

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