

# **Can traffic conflicts be used to estimate road safety? Research effort at UBC**

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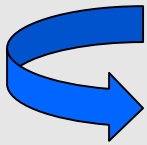
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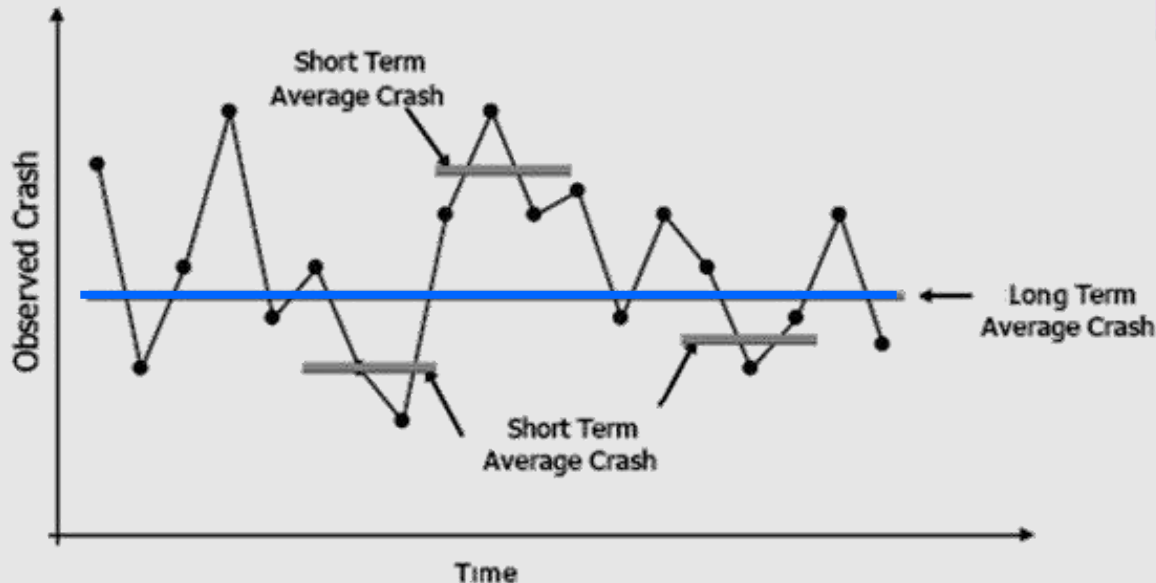


# Introduction

- A collision event is used as a fundamental indicator of safety



It results in injury or property damage



# Introduction

***Estimate of  
Road Safety***



***Estimate of long-term average  
collision frequency of a road  
site, under a given set of  
geometric design and traffic  
volumes in a given time period  
(Hauer, 1997)***

## **Example of road safety measures:**

**observed collisions over 3 years**

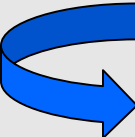
**predicted collision frequency from safety performance functions**

**posterior distribution of collision frequency from Bayesian methods**

**...**

# Collision Data

## Shortcomings

- ✓ **Collision records are in many cases incomplete and lack important details**
- ✓ **Collisions are rare events**
  -  **It is often necessary to observe collisions over a prolonged period to reach statistically valid results**
- ✓ **Confounding factors (regression to the mean, ...)**
- ✓ **Reactive approach**



# Collision-based Safety Evaluations

- ✓ Road safety studies based on collision data rely on solid statistical techniques which were developed over two decades of research
- ✓ Wide literature on Collision prediction models (Safety Performance Functions)



## Mathematical form of CPMs/SPFs:

- e.g., road segments

$$E(Y) = e^{a_0} \cdot L^{a_1} \cdot V_1^{a_2} \cdot e^{\sum_{j=1}^m b_j x_j}$$

traffic exposure

coefficients to be estimated through regression analysis

# **Alternative Safety Measures**

**Also known as “surrogate” measures**

## **Definition**

**1. Must be based on an observable non-crash event that is physically related in a predictable and reliable way to collisions**

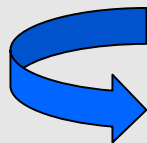
**AND**

**2. There exists a practical method for converting the non-crash events into a corresponding crash frequency**

# Traffic Conflicts

***“An observable situation in which two or more road users approach each other in space and time for such an extent that there is a risk of collision if their movements remain unchanged”***

- Traffic conflict represents the most commonly used surrogate safety measure.
- Time-to-collision (TTC) indicator



# TTC Estimation



Figure 22. Conflict observer in chair.

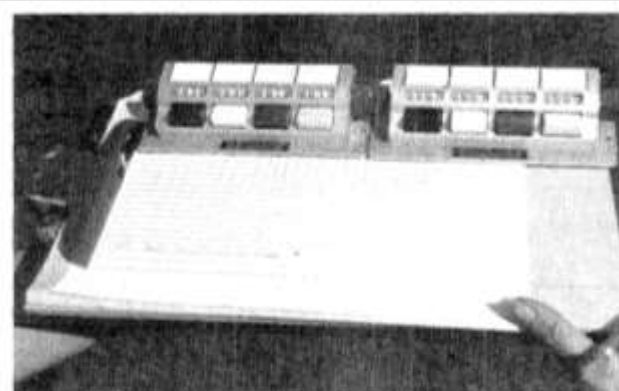
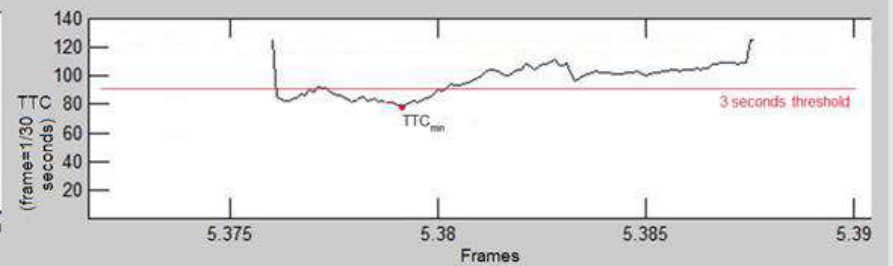
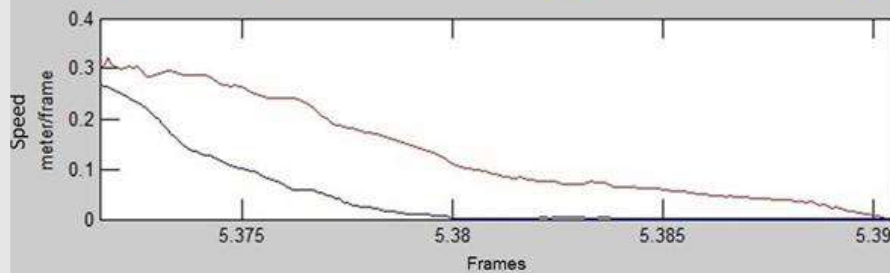
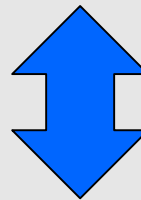


Figure 24. Traffic conflict count board.



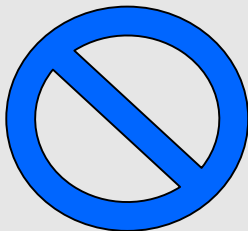


# Can Traffic Conflicts Be Used To Estimate Road Safety?

**2. There exists a practical method for converting the non-crash events into a corresponding crash frequency**



- **The link between conflict reduction and potential collision reduction still needs to be clearly established**

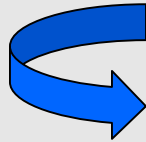


**The correlation between conflicts and collisions is weak, or can vary from site to site**

**Some very recent research has shown models which convert conflicts into collision frequency but results are not uniform**

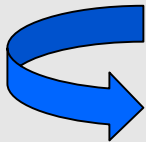
# Research Effort At UBC

1. Comparison of collision-based evaluation with the results of a traffic conflict-based evaluation based on the same set of treatment intersections



*Countermeasure: Smart channel*  
*Case study: Penticton (BC)*

2. Transferring the statistical techniques developed for collision-based safety evaluations to surrogates such as traffic conflicts

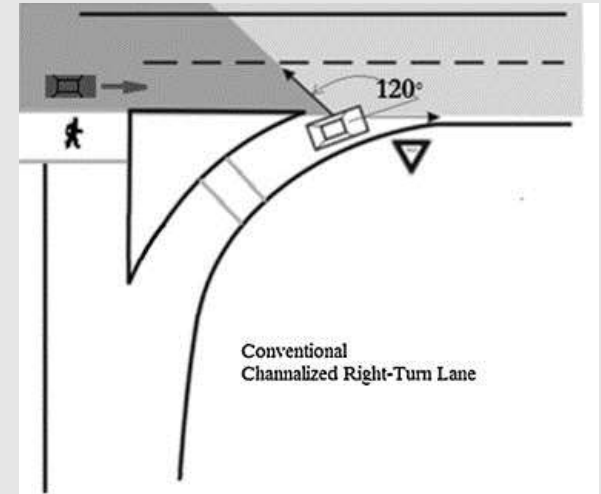


*Case study: Surrey (BC)*

# Smart Channels

## Channelized right turns

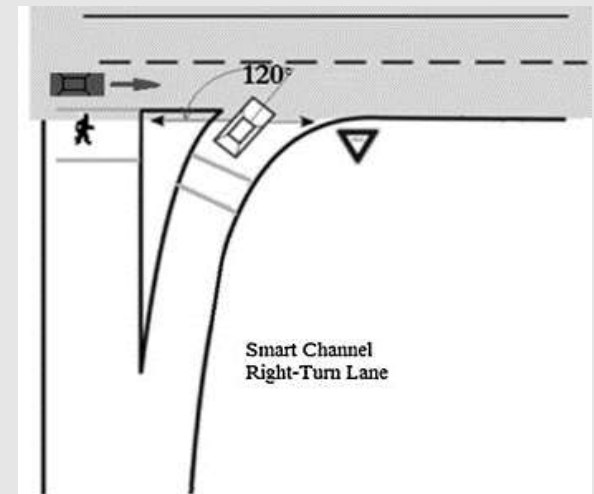
implemented at intersections with high right-turn traffic volumes to reduce vehicle delay



## “Smart” channel

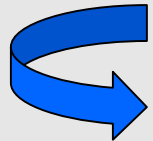
alternative right-turn design more pedestrian-friendly

Drivers are provided with a better view of the traffic stream they are to merge with

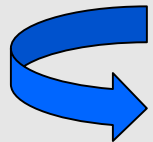


# Project Goals

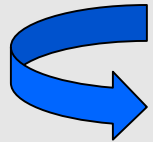
**At first, a before-and-after evaluation using traffic conflicts was carried out for 3 smart channels in Penticton (Autey, Sayed et al., 2012)**



**BA study using collision and traffic volume data for the same treated intersections and state-of-the-art statistical technique**



**Compare the results of the collision-based evaluation and the conflict-based evaluation conducted earlier to validate traffic conflict technique**



**Better understanding of the link between road safety (collisions) and conflicts**

# Review Of Data Sources



- **3 treated, 14 control and 6 comparison sites**
  - a) comparison sites: site right-turn channel not improved
  - b) control sites: nearby intersections comparable for geographic proximity and comparability to a treatment site.
- **Traffic volumes from 2007 to 2011**
- **Collision data (PDO and F+I) gathered in time frames of 1 and 4 months**
- **Months of implementation removed (summer 2010)**

# Full Bayesian Evaluation (I)

- Collision reduction from OR
- OR = 1 no change; OR < 1 safety improvement
- $OR = \frac{A/C}{B/D}$
- Poisson-lognormal intervention model (advanced SPF)

$$Y_{it} | \theta_{it} \sim \text{Poisson}(\theta_{it})$$

$$\ln(\theta_{it}) = \ln(\mu_{it}) + \varepsilon_i$$

$$\ln(\mu_{it}) = \alpha_0 + \alpha_1 T_i + \alpha_2 t + \alpha_3 [t - (t_{B,i} + 1)] I_{it} + \alpha_4 T_i t + \alpha_5 T_i [t - (t_{B,i} + 1)] I_{it} + \alpha_6 T_i I_{it} + \beta_1 \ln(V_{1,it}) + \beta_2 \ln(V_{2,it})$$

with

$$\varepsilon_i \sim \text{Normal}(0, \sigma_\varepsilon^2)$$

coefficients to be estimated  
with MCMC techniques

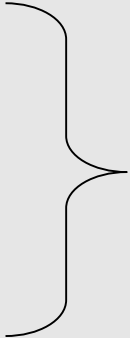
# Full Bayesian Evaluation (II)

**Account for confounding factors:**

- ✓ **Exposure effect**
- ✓ **Regression-to-the-mean phenomenon**
- ✓ **Unrelated effects**
- ✓ **Trend effects**

**Type of models developed:**

- ✓ **Univariate Analysis**
- ✓ **Multivariate Analysis**
- ✓ **Analysis with Matched Pair Sites**



**“Best” models  
selected for  
comparison**

# Results Comparison

- Collisions versus traffic conflicts reduction

Intersection/severity	Traffic conflicts (average hourly conflicts)	Collisions (on 4-month basis)
T1 – Channel/Green	33%	36.4% <sup>a</sup>
T2 – Channel/Warren	57%	65.1% <sup>a</sup>
T3 – Channel/Duncan	55%	55.1% <sup>a</sup>
Overall	51%	55.6% <sup>a</sup>
Total severity	Conflict severity: 41%	PDO: 67.5% <sup>b</sup>
		F + I: 47.6% <sup>b</sup>

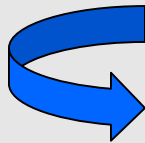
a PLNI model with matched pairs.

b MPLNI model.



# Discussion

- ✓ **The similarity of overall and location-specific reductions in conflicts and collisions was remarkable**



**This provides strong support for using traffic conflicts in BA studies**

- ✓ **Proportionality constant  $\approx 1$**
- ✓ **Limits of the evaluation (TTC < 3 s, other conflict indicators, other safety countermeasures ...)**

# Second Project: Goals

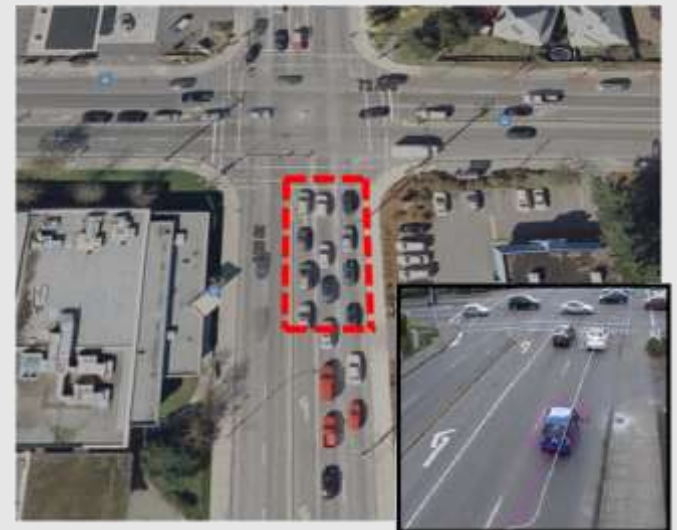
**Transferring the statistical techniques developed for collision-based safety evaluations to surrogates such as traffic conflicts**



**We introduced traffic conflicts in place of collision data in the statistical models used for road safety evaluations**

# Data (rear-end conflicts)

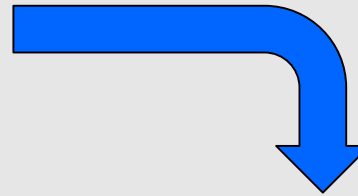
- Automatic extraction of conflicts from 8 video cameras for 2 signalized intersections in Surrey
- Data was gathered only during daytime hours (approximately 7 AM to 8 PM) in two different days
- Traffic conflicts were defined by using the TTC conflict indicator



# Conflicts-based SPFs

$$Y_{it} | \theta_{it} \sim \text{Poisson}(\theta_{it})$$

$$\ln(\theta_{it}) = \ln(\mu_{it}) + \varepsilon_i$$



Hourly traffic volume

Length of count of conflicts

$$\mu_t = \exp(\beta_0) \cdot V_t^{\beta_1} \cdot \exp(\beta_2 L_t)$$

coefficients to be  
estimated through  
regression analysis

# Estimated Model Parameters

Distribution	Variable	Parameter	Mean	Standard Deviation
Poisson-Gamma	Intercept	$\beta_0$	-6.756	1.17
	Traffic volume	$\beta_1$	1.25	0.171
	Length	$\beta_2$	0.006	0.002
	Overdispersion	$\kappa$	1.718	0.196
	Dispersion	$1/\kappa$	0.589	0.066
	Deviance information criteria	DIC	Value= 1096.98	
Poisson-Lognormal	Intercept	$\beta_0$	-5.59	1.115
	Traffic volume	$\beta_1$	1.198	0.169
	Length	$\beta_2$	0.005	0.003
	extra-Poisson variation	$\sigma_\varepsilon$	0.803	0.051
	Deviance information criteria	DIC	Value=1105.19	

# Model validation

- ✓ Global goodness-of-fit tests based on Bayesian p-values were performed

$$\text{Bayes p-value} = \Pr [T(Y_{t,\text{rep}}, \beta) \geq T(Y_t, \beta) | Y_t]$$

probability that the simulated (replicated) data set,  $Y_{t,\text{rep}}$ , could be more extreme than the observed one,  $Y_t$

**$T_1$ : global measure of lack of fit (i.e., sums-of-squares residuals)**

**$T_2$ : fit in the upper tail of the distribution (i.e., max value)**

Poisson-Gamma Model	Bayesian p-value	
	$T_1$	$T_2$
Surrey - data set I	0.392	0.760

# Conclusions

- ✓ **Conflict-based SPFs were successfully developed using Bayesian statistical techniques**
- ✓ **The results demonstrated that the Poisson-Gamma model always outperformed the Log-Normal**
- ✓ **The number of conflicts involving lane changes or braking at intersections increases more rapidly than traffic volume**
- ✓ **A first framework toward the development of a promising field of conflict-based evaluations alternative to collision-based ones was provided**

# Questions

