# Drive Video Analysis for the Detection of Traffic Near-Miss Incidents



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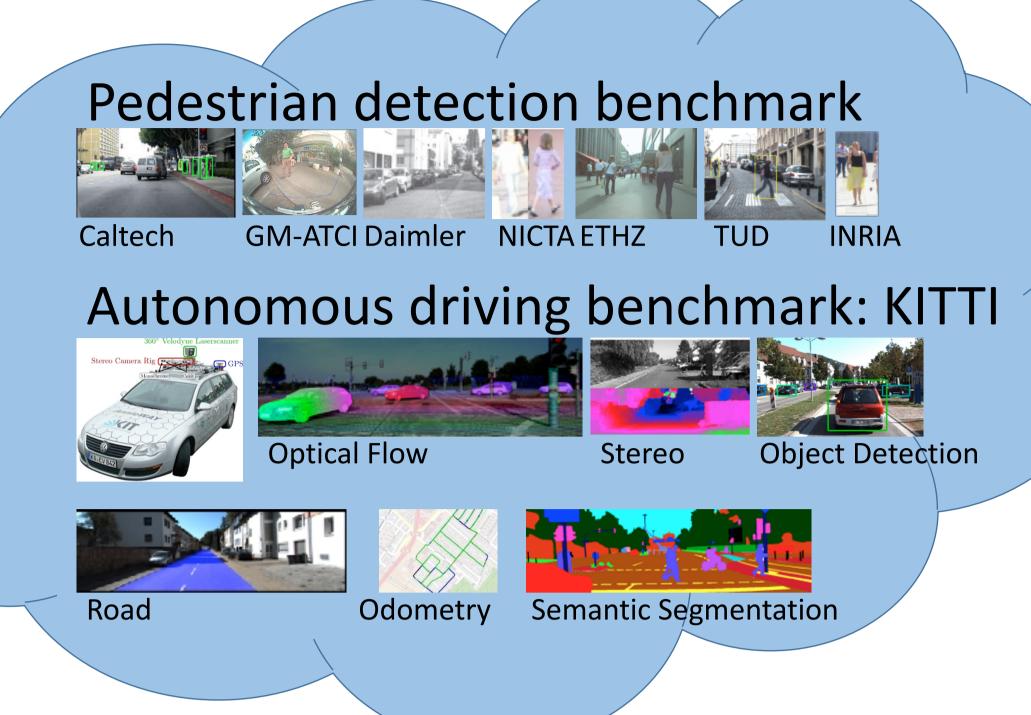
#### **Contribution 1**

We have collected and annotated a novel DB that

contains TRAFFIC NEAR-MISS INCIDENT scenes



On the related works...



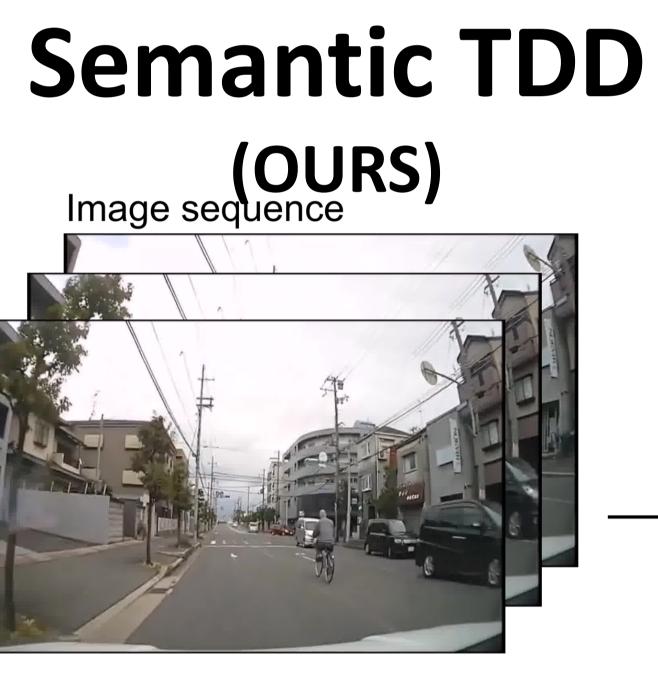
## Our Philosophy

"making sure that analysis of traffic near-miss incidents helps prevent collisions"

**Contribution 2** 

Risk: {High, Low} Attribute: {Ped., Veh., Cycl.}, and bg

# Benchmarking on the NIDB

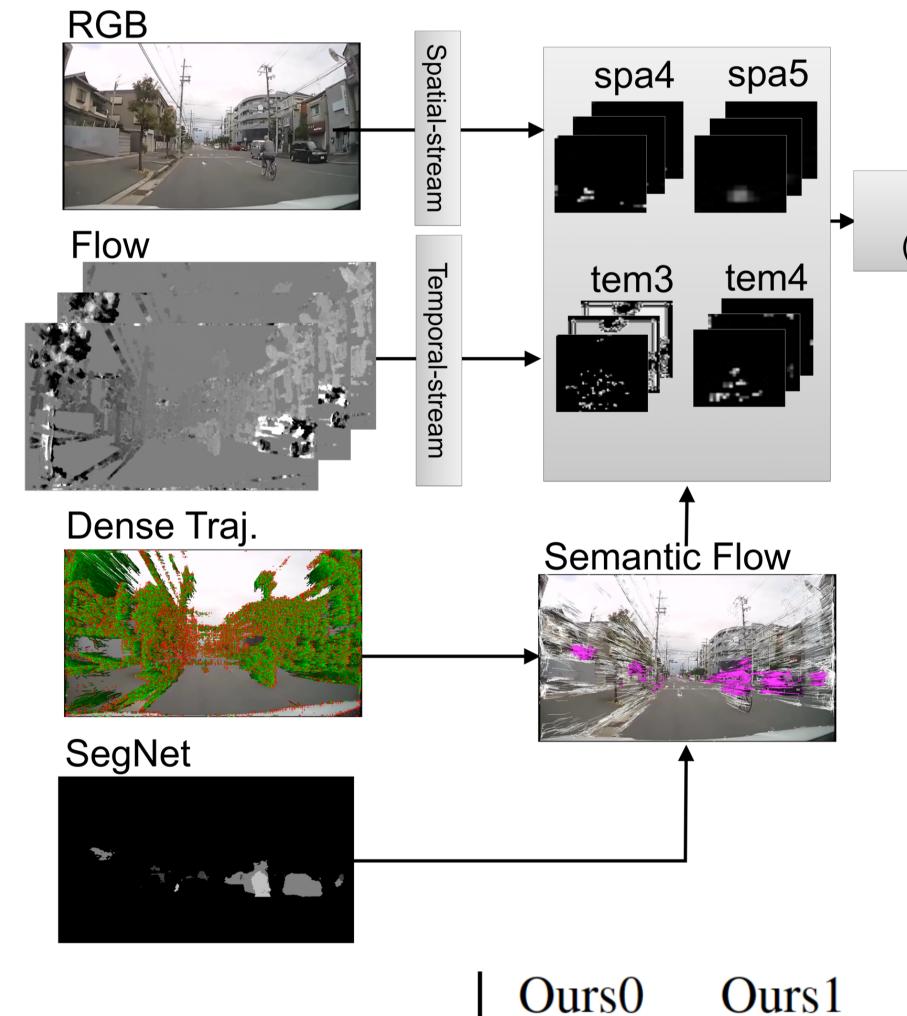


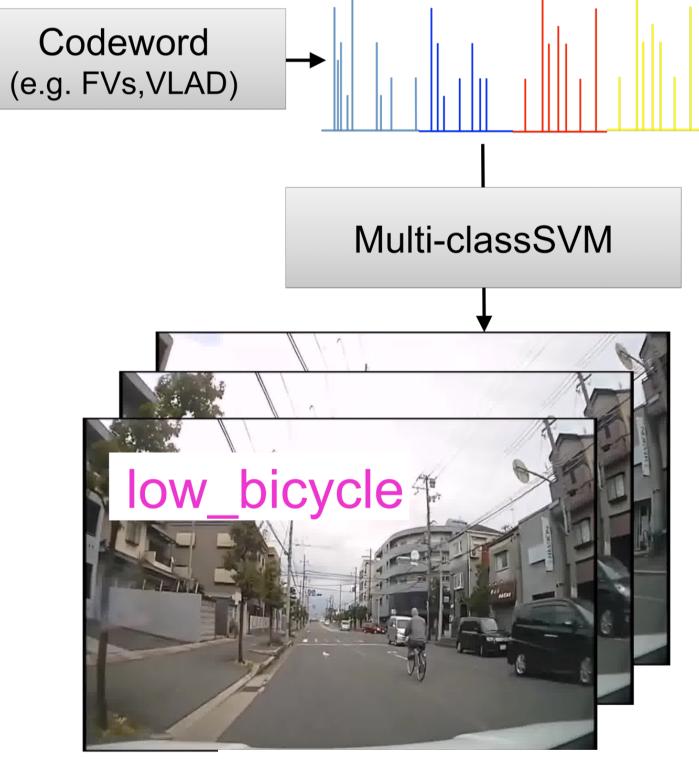
Difference from original TDD

NIDB trained two-stream CNN

Spatial TDD

Semantic flow into TDD





Ours2

62.1

61.3

Motion vector

spa3 spa4 tem3 tem4

IDT, HOG, HOF, MBH VS. (2) 2D Conv SMD, PoT, TS (3) 3D Conv

(1) Hand-crafted feat.

	C3D
Recognition Detection	
	Best on Re (TDD+SemFlo
	Best on De

	Recognition Detection	60-	'
Best on Rec.: Ours1 (TDD+SemFlow+Finetune)		55 50 %	
		Rate (%)	
Best on Det.: Ours2 (+IDT feature)		35 - 30 -	ı
		25	ı

#### Temporal TDD Background fine-tuning Near-miss fine-tuning Foreground & background Extra IDT features 58.6 Recognition task 56.3 60.0 63.2 64.5 48.0 47.4 Temporal Detection task 46.1 49.9 55.1

### **Conclusion & Future Work**

- The purpose of the DB is to directly understand near-miss scenes for selfdriving and ADAS-equipped vehicles
- We are focusing on "traffic accident anticipation" in our on-going work which is annotating additional labels and adaptive loss

