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Using Computer Vision to Process Vehicle Dashboard Displays in Transportation Safety Research

Kristin Jiating Chen, Alexander Cates, Rick Huey, Marcelo Simas,
James Jenness, Gonzalo Rivero

Outline

- Introduction of Research Problem
- Nature of data
 - Videos
 - Icons
- Methodology – Machine learning pipeline with OpenCV
- Results
 - Model performance
 - Future research

Introduction – Background of the Overall Study

- Understand driver behavior in the context of driver-assist systems in Toyota Safety Sense system (TSS)

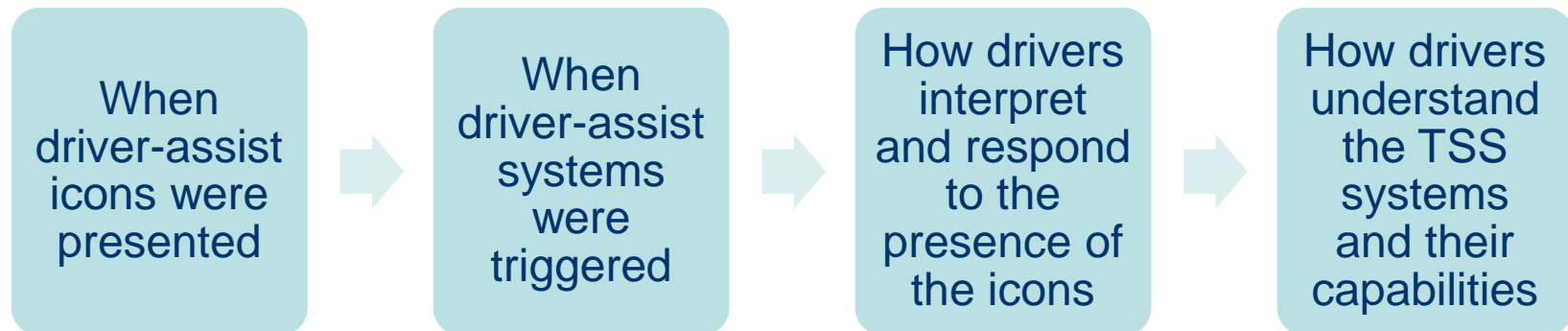
- Driver-assist systems

Adaptive Cruise Control

Lane Departure Alert

Pre-Collision System

- when the systems are triggered, related driver-assist icons appear on the vehicle dashboard displays



Introduction – Research Problem

- Problem
 - Identify the presence of icons on the vehicle dashboard displays
- Pilot study
 - Collect data by recording the central dashboard displays while driving instrumented Toyota vehicles

Nature of Data – Videos

- Video recordings of central dashboard display
 - Pilot study data
 - 200+ 1-min videos
 - Study data (*estimated*)
 - 500+ 1-min videos per vehicle per week:
 - 10-20 instrumented Toyota vehicles
 - 12 weeks participation per vehicle

Nature of Data – Icons

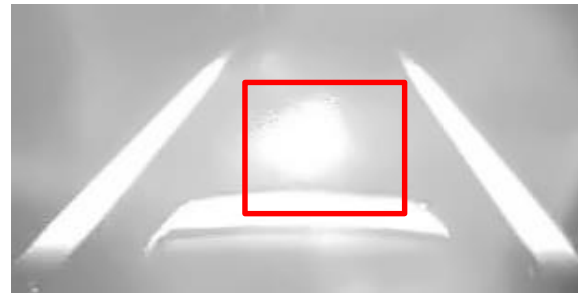
- 10 icons
 - Adaptive Cruise Control
 - Lane Departure Alert
 - Pre-Collision System



Leading Car

Headway Bar Indication

Low resolution!
Glare!



Lane Line Indication

Methodology

- Technique
 - OpenCV: open source computer vision library
- Machine learning Pipeline utilized OpenCV Python API and R Shiny



Methodology

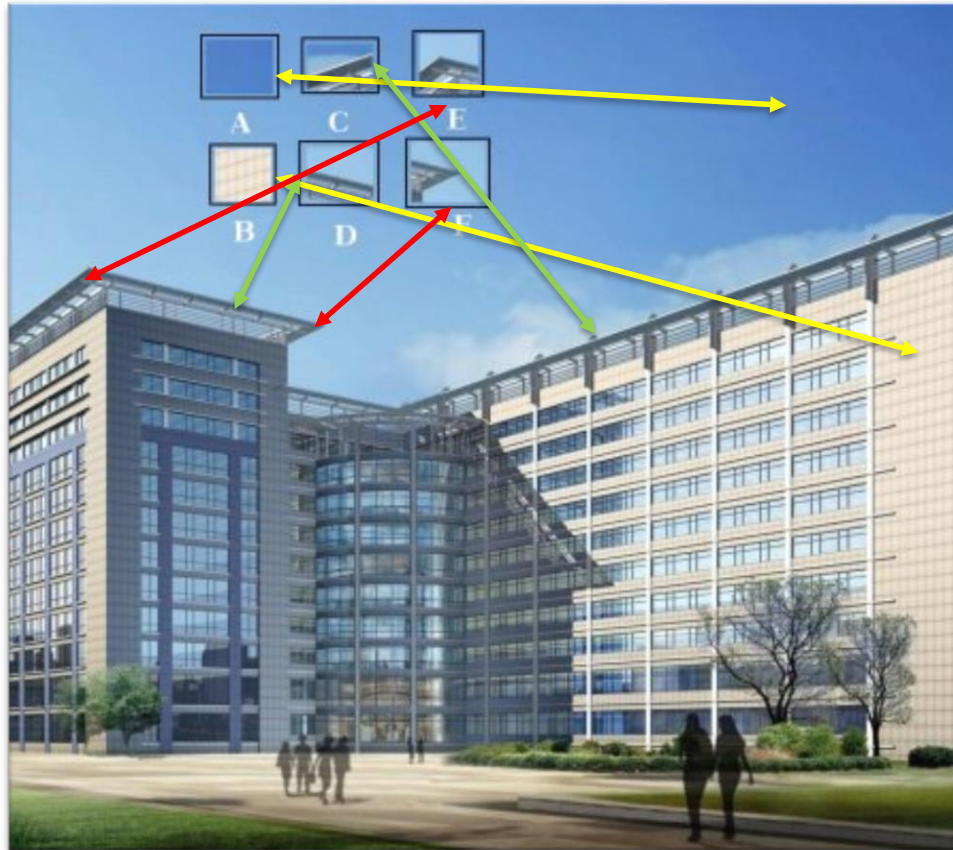
– Frame Extraction

- Extract frames from videos per half-second

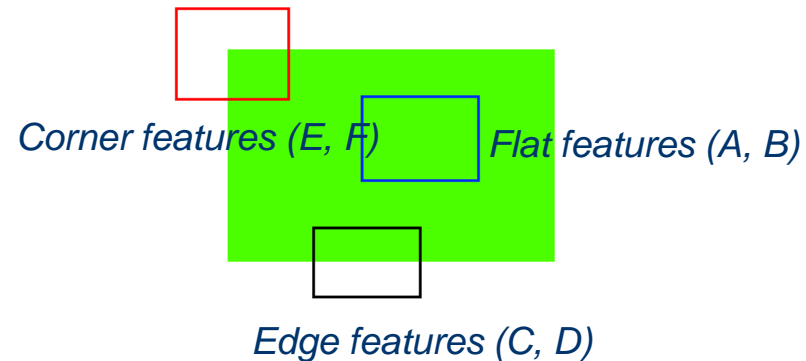


Methodology

– Feature Matching between Icons and Extracted Frames



- Feature: distinct patterns
- Corner feature – good feature

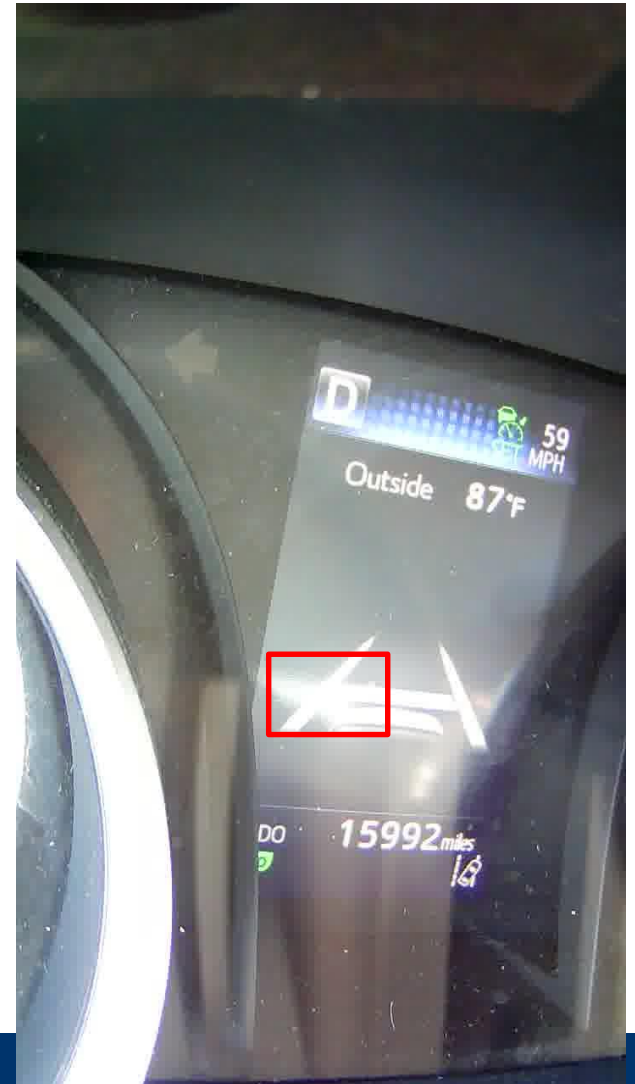


Methodology

– Feature Matching



- Feature detection
 - Corner features



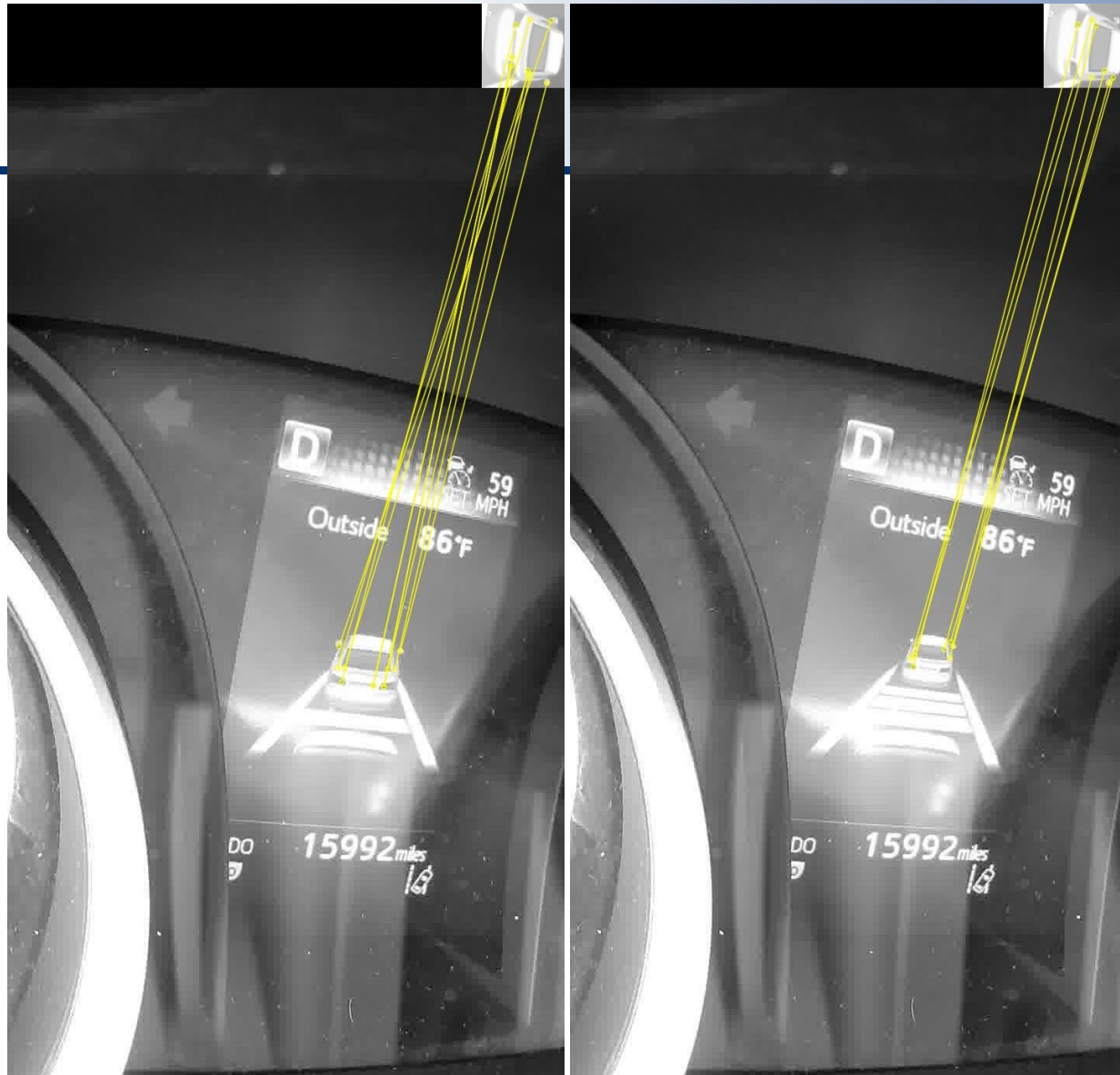
Methodology

— Feature Matching

- Feature description
 - Rotation and Scale invariant



Leading Car



Methodology

– Feature Matching

- Feature matching between icons and extracted frames
 - matching criteria: 3 features matched (tunable)



Methodology

– Image Preprocessing before Feature Matching

- Crop the dashboards from the frames - focus on region-of-interest
- Deskew the dashboards
- Denoise



Methodology

– Image Preprocessing before Feature Matching

- Crop into 3 parts
 - Narrow down the target area



Adaptive Cruise Control

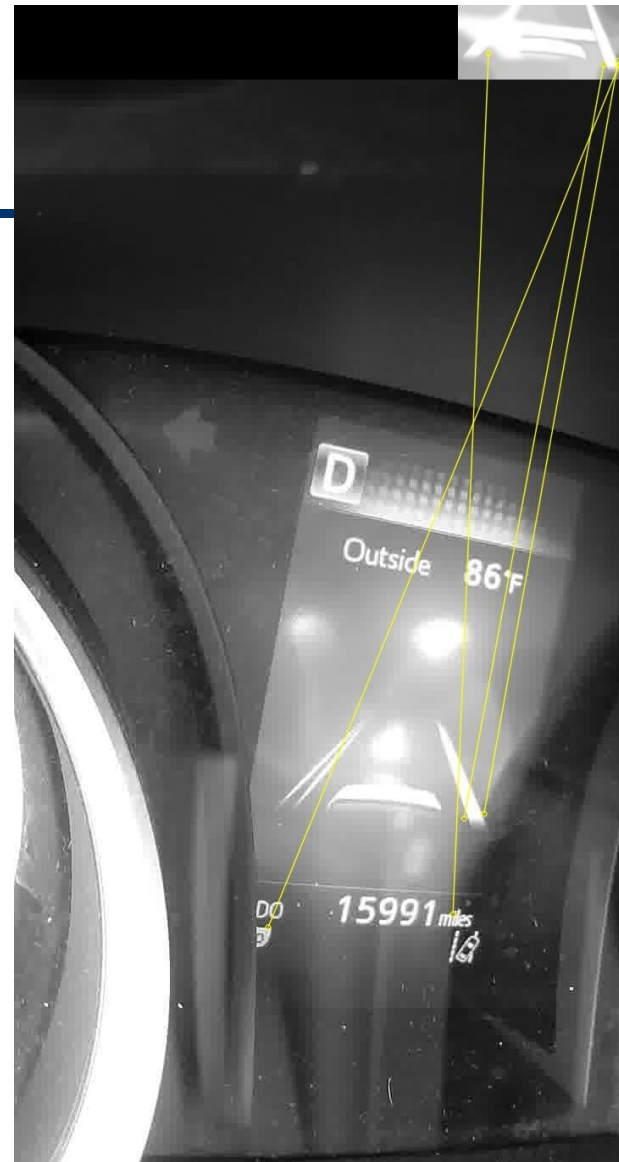


Lane Departure Alert



Methodology

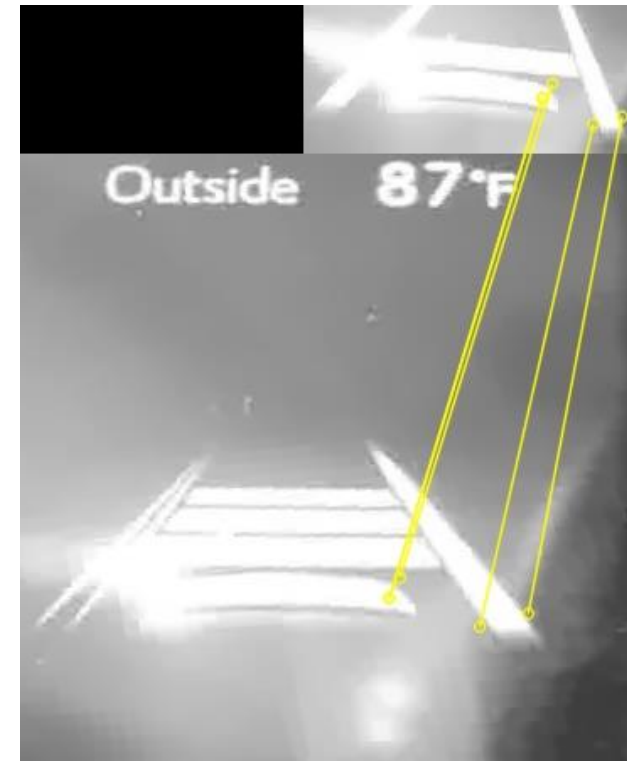
– Feature Matching (after Image Preprocessing)



False Positive



< 3 matched features
True Negative



> 3 matched features
True Positive

Methodology

– Visualization

- Interactive tool to visualize matching results and other information about drivers and road conditions



The screenshot shows a video player interface with three video thumbnails. The top thumbnail, labeled 'Video 1', shows a person's face with a yellow star sticker. The middle thumbnail, labeled 'Video 2', shows a dark scene with a bright light source. The bottom thumbnail, labeled 'Video 3', shows a car's dashboard with a navigation screen. A red box highlights a small white square on the navigation screen in the bottom thumbnail. The interface includes tabs for 'Video', 'Audio', and 'High-G', a 'Play All' button, and a timestamp '8/24/2018, 4:48:10 AM'.

Results

– Model Performance

- Classification Accuracy: 0.9

- Sample data: 10 icons, 29 *selected* frames, 290 combinations

n	Predicted Negative	Predicted Positive	
Actual Negative	213	4	217
Actual Positive	25	48	73
	238	52	290

- True Positive Rate: 0.66 True Negative Rate: 0.98
- False Positive Rate: 0.018 False Negative Rate: 0.342

- Runtime: ~ 4 hours with *4 threads multiprocessing*

- Week 1 vehicle #1 Data: 10 icons, 57,000 frames, 570,000 combinations from more than 500 1-min videos
- Frame extraction: 10 minutes
- Image preprocessing: ~2 hours
- Feature matching: ~2 hours

Results

– Conclusion & Future Research

- Current research
 - Tune 3-feature matching criteria to balance FN/FP error
- Future research
 - Train customized model
- Conclusion: Works well!
 - 90% accuracy
 - Leads to 66% true positive rate and 98% true negative rate
 - Reasonable computation time
 - Meets computation requirement when data scales up

Thank you for listening!

- Contact information

Kristin Chen KristinChen@Westat.com

Alexander Cates AlexanderCates@Westat.com

- Citation

ANN ARBOR. Mich, (2017, November 15). *Toyota's Collaborative Safety Research Center to Study Societal Acceptance of Connected and Automated Vehicle Technologies*.

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https://docs.opencv.org/3.4/db/d27/tutorial_py_table_of_contents_feature2d.html