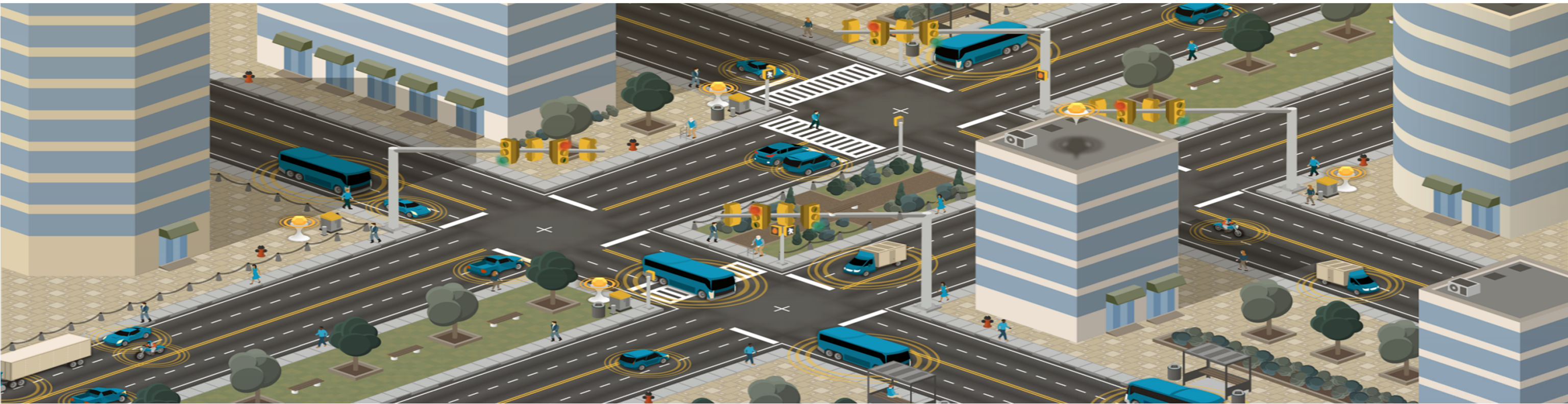




# CONNECTED VEHICLE PILOT Deployment Program



## New York City CV Pilot Deployment Results and Transition Plan



**Jonathan Walker, USDOT; Keir Opie, Cambridge Systematics; Jingqin Gao, NYU; Mohamad Talas, NYCDOT**

ITS Joint Program Office

# Webinar Agenda



## ■ Purpose of this Webinar

- Share the deployment performance results from NYC Connected Vehicle Pilot
- Outline the plan after completion of Phase 3 Operations

## ■ Webinar Content

- Connected Vehicle Pilot Deployment Program Overview
- NYC CV Pilot Deployment Performance Evaluation Results and Transition Plan
- Stakeholder Q&A

## ■ Webinar Protocol

- Please mute your phone during the entire webinar.
- You are welcome to ask questions via chatbox at the Q&A Section.
- The webinar recording and the presentation material will be posted on the CV Pilots website.

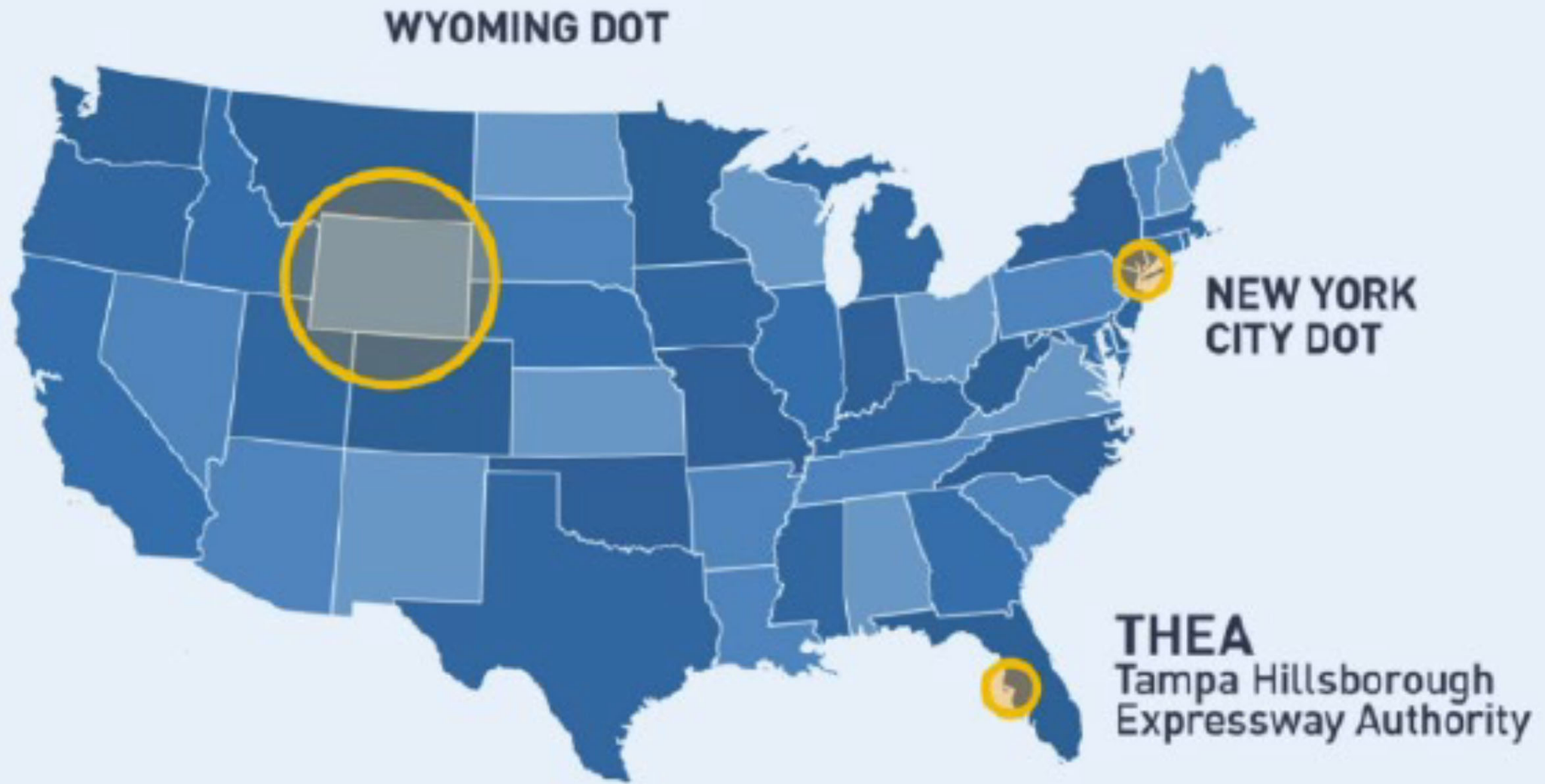
**Program documentation:**

[https://www.its.dot.gov/pilots/cv\\_pubs.htm](https://www.its.dot.gov/pilots/cv_pubs.htm)

# CV Pilot Deployment Program Goals



# The Three Pilot Sites



# Agenda



- Project Overview
- Data Collection and Processing
- Experimental Design & Analysis Plan
- Performance Results & Findings
- Transition Plan
- Perspective on CV Pilot Deployment
- Questions

## *Today's Presenters:*

*Keir Opie*

*NYC CVPD Performance Evaluation Lead  
Cambridge Systematics*

*Jingqin Gao*

*NYC CVPD Safety Analysis Team  
NYU*

*Mohamad Talas*

*NYC CVPD Project Manager  
NYC DOT*



# Project Overview

# NYC CV Pilot Deployment Goals



- Primary Goal:
  - **Improving safety** through the reduction of vehicle and pedestrian crashes, injuries, and fatalities
  
- Secondary Goal:
  - **Improving mobility** and reliability through crash prevention and lower crash severity
  
- Measure System Performance in meeting these goals
  - Data collection system was designed around project performance measures addressing privacy concerns and data collection costs

# NYC CV Infrastructure



- Infrastructure:
  - 470 Roadside Units (RSU)
  - 3000 Vehicles
- Safety applications: 13
- Operations applications: 8
- This is a **large scale** deployment with challenges:
  - Location accuracy – urban canyons
  - RSU density
  - Application arbitration/interference
  - DSRC media only – channel management
  - First full-scale security deployment
  - Security boundary expanded to include all ITS communications
- Utilize edge computing concepts to minimize bandwidth





# NYC CV Safety Applications



## Vehicle-to-Infrastructure (V2I) Pilot Area

- Red Light Violation Warning
- Speed Compliance
- Curve Speed Compliance
- Speed Compliance/Work Zone
- Oversize Vehicle Compliance
  - Over Height
- Emergency Communications and Evacuation Information (Traveler Information)

## Vehicle-to-Vehicle (V2V) Citywide

- Forward Collision Warning
- Emergency Electronic Brake Light
- Blind Spot Warning
- Lane Change Warning/Assist
- Intersection Movement Assist
- Vehicle Turning Right in Front of Bus Warning

## Pedestrian Applications

- Pedestrian in Signalized Intersection Warning – *to vehicles*
- *Mobile Ped Signal System – Vision Disability Pedestrian Navigation Assistance*

# NYC Agencies CVPD Fleets



Agency	Passenger Cars	Pickups and Trucks	Vans	Buses	Vehicle Installations
NYC Dept. of Transportation (DOT)	●	●	●		1238
NYC Dept. of Parks and Recreation (PARKS)	●	●	●		511
NYC Dept. of Correction (DOC)	●	●	●	●	293
NYC Dept. of Environmental Protection (DEP)	●	●	●		159
NYC Dept. of Homeless Services (DHS)	●		●		100
NYC Taxi and Limousine Commission (TLC-DCAS)	●	●	●		98
NYC Human Resources Administration (HRA)	●		●		86
NYC Dept. of Citywide Administrative Services Fleet (DCAS)	●				78
NYC Dept. of Education (DOE)	●	●	●		78
NYC Dept. of Buildings (DOB)	●				69
NYC Administration for Children's Services (ACS)	●	●	●		65
NYC Dept. of Housing, Preservation and Development (HPD)	●				48
NYC Dept. of Health and Mental Hygiene (DHMH)	●	●	●		45
NYC Dept. of Design and Construction (DDC)	●				38
NYC Office of Chief Medical Examiner (OCME)	●	●	●		29
MTA Bus & NYCT				●	14
NYC Emergency Management (OEM)	●				12
NYC Dept. of Consumer Affairs (DCA)	●	●			12
Anheuser-Busch InBev (ABI)			●		10
NYC Dept. of Information Technology and Telecommunications (DOITT)	●				9
NYC Dept. of Probation (DOP)	●				6
NYC CVPD Team Vehicle		●			1
Taxi Limousine Commission (Yellow Cabs)	●				1
<b>Total</b>	<b>1,662</b>	<b>967</b>	<b>269</b>	<b>102</b>	<b>3,000</b>

# Fleet Agency Vehicles



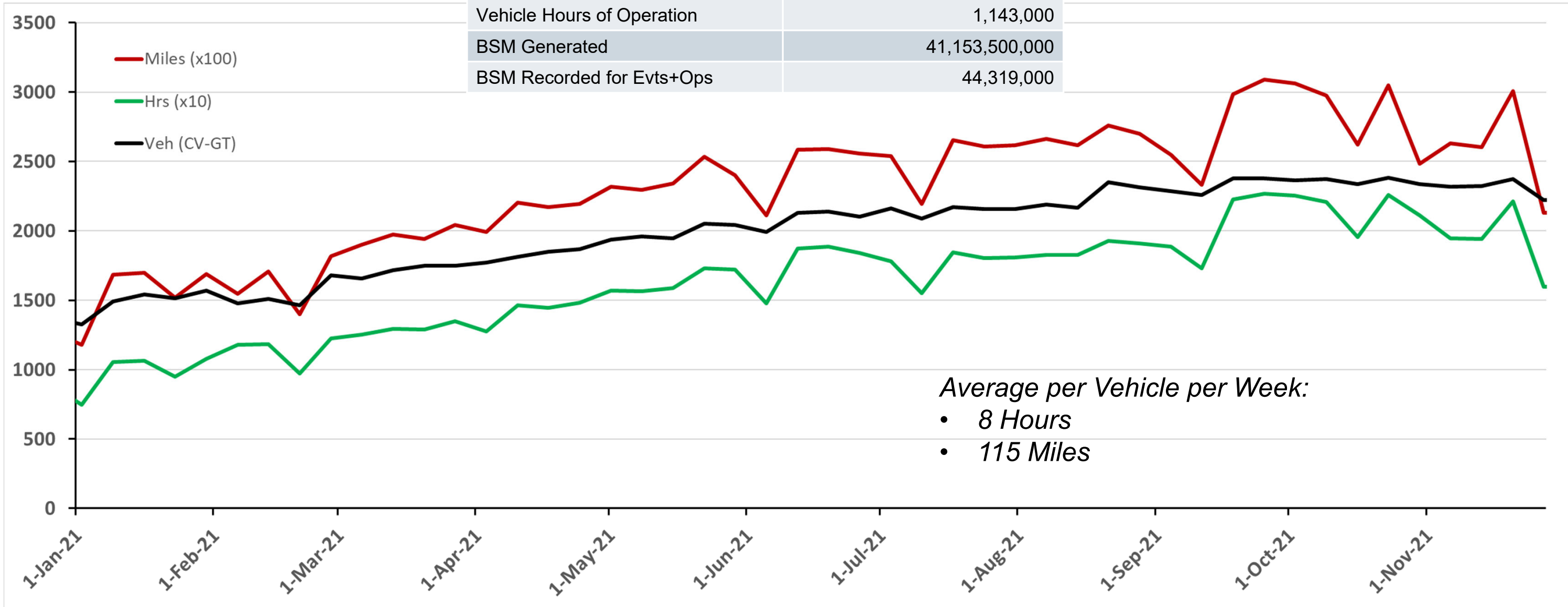
Vehicle CAN bus interface provides speed data for Dead Reckoning to improve location accuracy

Vehicle Make	Passenger Cars	Pickups and Trucks	Vans	Buses	Vehicle Installations
Chevrolet	165	162	168		495
Chrysler			2		2
Dodge			16		16
Ford	331	714	83		1,128
Freightliner		1			1
IC Corporation				85	85
International				3	3
New Flyer				3	3
Nissan	130				130
Nova				7	7
Orion				4	4
Ram		90			90
Toyota	1,036				1,036
<b>Total</b>	<b>1,662</b>	<b>967</b>	<b>269</b>	<b>102</b>	<b>3,000</b>

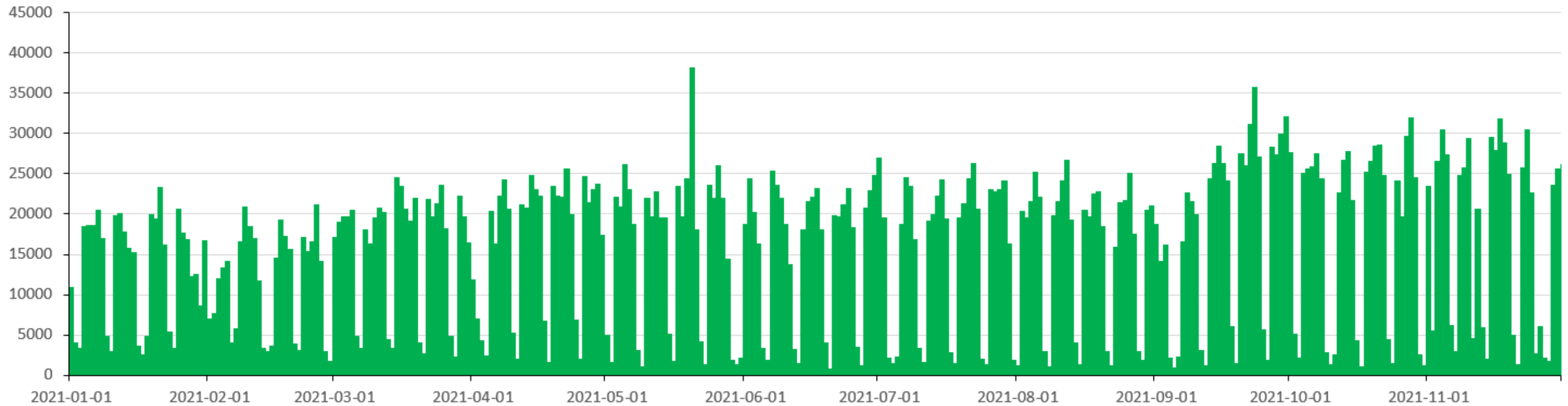
# Fleet Weekly Operations



Estimates for our 3,000 Veh. Fleet	Jan 3 – Nov 27, 2021 (47 wks)
Vehicle Miles Traveled	16,294,000
Vehicle Hours of Operation	1,143,000
BSM Generated	41,153,500,000
BSM Recorded for Evt+Ops	44,319,000



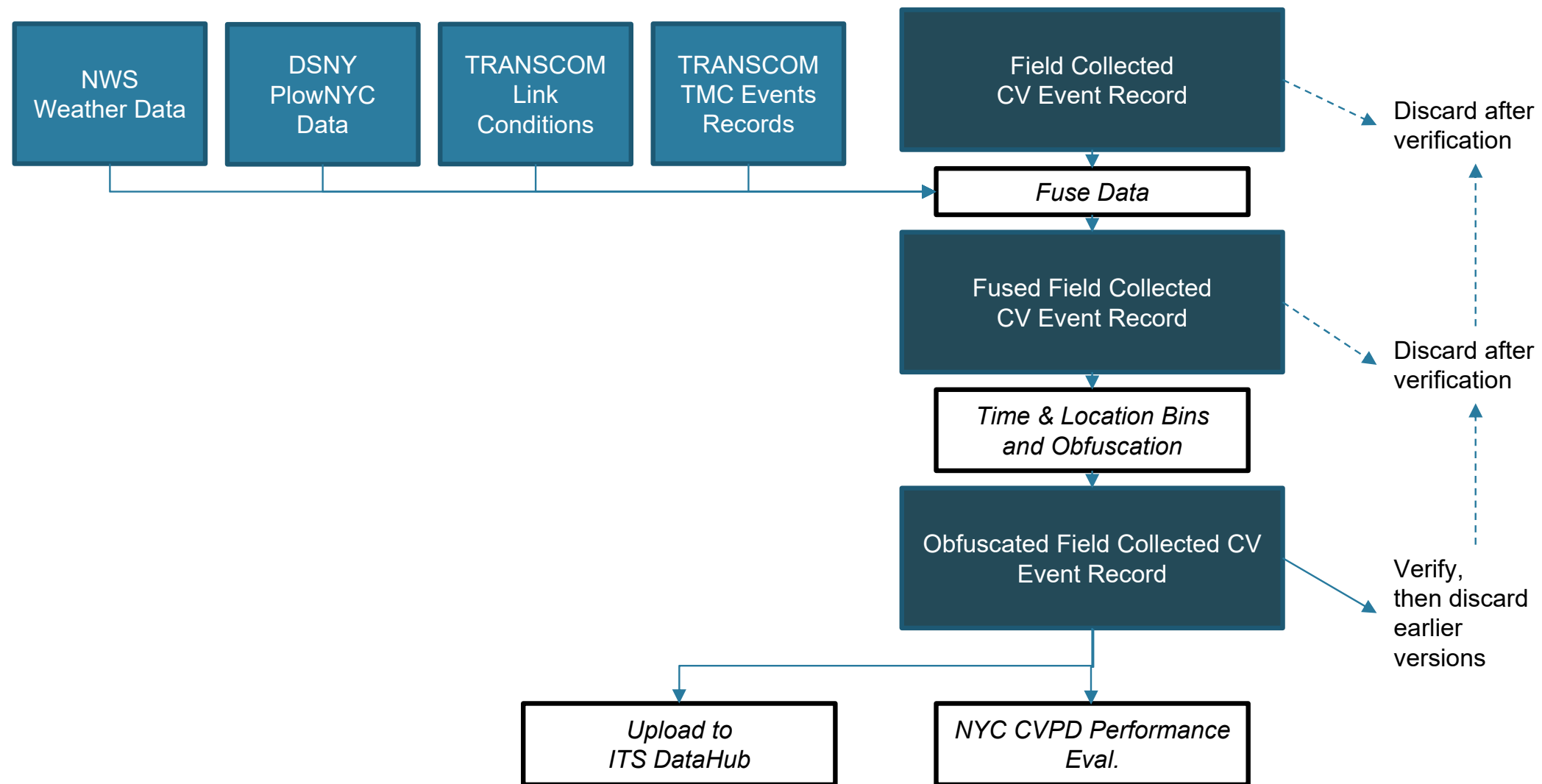
# OBU-OBU V2V Contacts (Daily)





# Data Collection & Processing

# CV Event Record Obfuscation Process



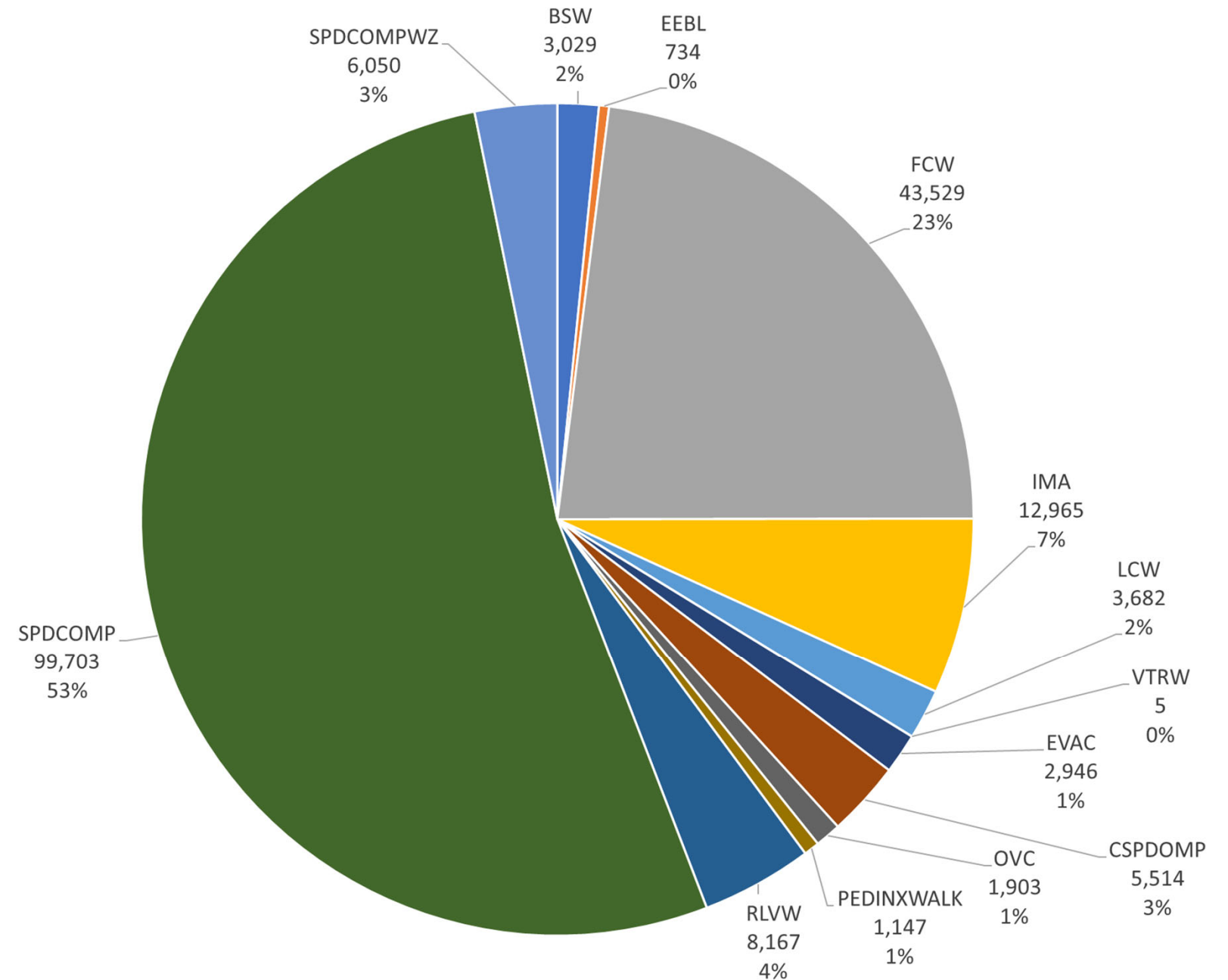


# CV Events by Type (Ingested)

## Jan - Nov 2021: Ingested

- 189,374 Total Events
- 34% V2V, 66% V2I

All events uploaded to the TMC, no scrubbing, checks, or cleaning





# Error and Quality Checks



Month	Received	% Errors	% Old App	% Test	% Released	# Released
<b>Jan-Nov</b>	<b>189,374</b>	<b>12%</b>	<b>5%</b>	<b>3%</b>	<b>80%</b>	<b>151,146</b>
21-01	19,323	29%	19%	4%	47%	9,085
21-02	9,843	10%	10%	5%	75%	7,410
21-03	16,294	6%	7%	8%	79%	12,862
21-04	16,213	7%	4%	5%	84%	13,589
21-05	17,549	10%	4%	3%	83%	14,514
21-06	15,870	10%	3%	2%	84%	13,391
21-07	16,479	11%	3%	2%	84%	13,844
21-08	18,000	11%	3%	2%	85%	15,210
21-09	18,924	11%	3%	1%	85%	16,116
21-10	21,698	10%	3%	1%	86%	18,553
21-11**	19,181	9%	3%	1%	86%	16,572

## Errors:

- Timestamp errors
- Location Errors
- Data Logging Errors
- Early Firmware Versions

## Old App:

- Older SW Firmware with known operational issues for only some CV Apps

## Test:

- Events from our CVPD test vehicles

\*\* November data is not yet finalized, December collection underway

# Obfuscated Event Details



- All Detailed Lat/Long coordinates shifted to artificial cartesian coordinate system (meters)
- All unique IDs for vehicles, intersections (MAP or SPaT), or TIM messages removed;
  - Intersection IDs replaced with new letter codes; unique to one event only
- Time Bins:
  - Date becomes Month and Day of Week
  - Time of day:
    - Overnight period (12:00 am – 6:00 am): NT
    - Morning Peak (6:00 am – 10:00 am): AM
    - Midday Period (10:00 am – 3:00 pm): MD
    - Afternoon Peak (3:00 pm – 8:00 pm): PM
    - Evening Period (8:00 pm – 12:00 am): EV
- Location Bins:
  - NYC Borough (5: MN, QN, BK, BX, SI)
  - Roadway Type
    - RSU Equipped: Avenue or Street
    - Non-Equipped: Freeway, Avenue, or Other

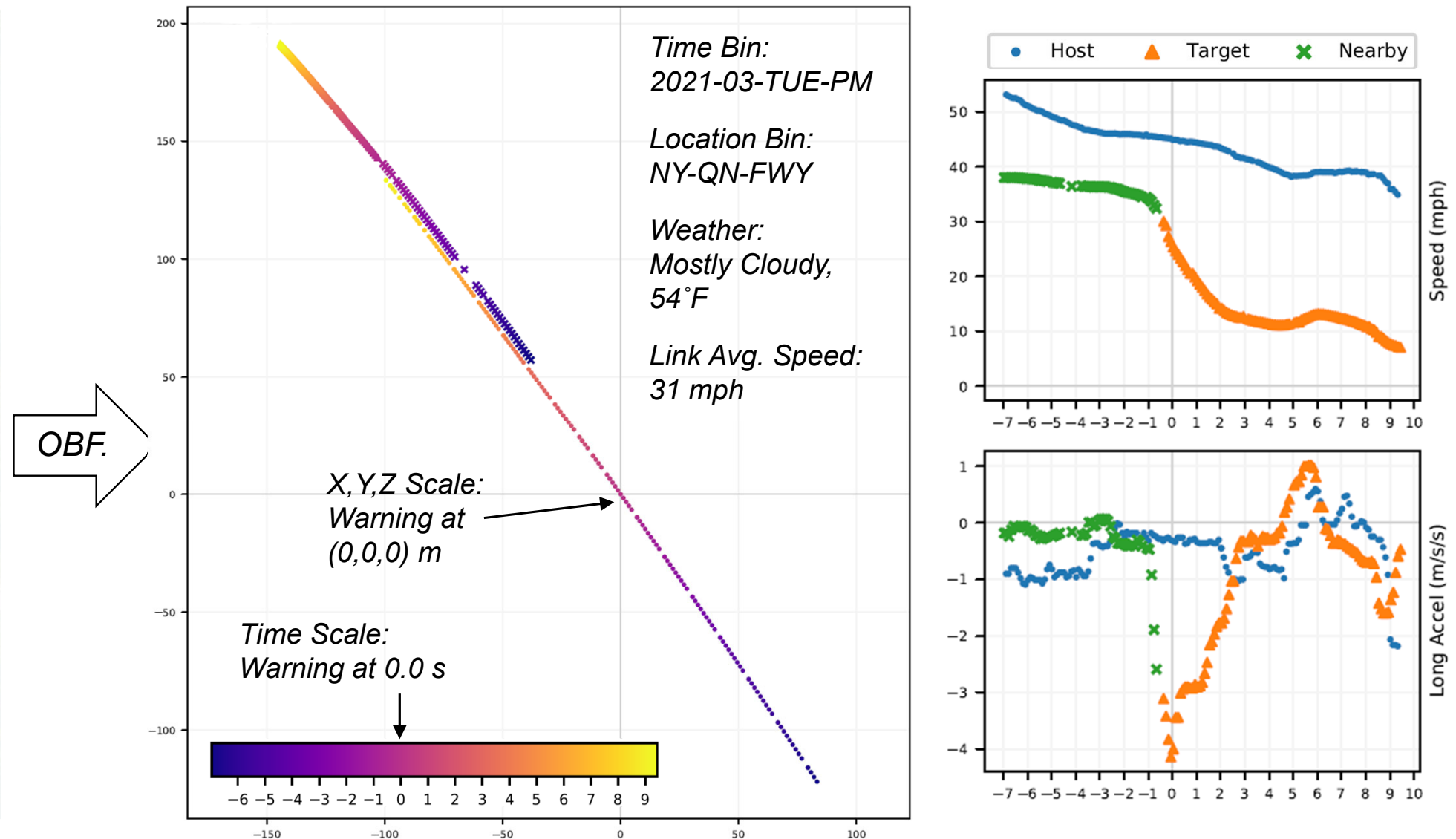


# ASD Event Log Obfuscation: Example

Recorded Event Data:



Obfuscated Event BSM Data:



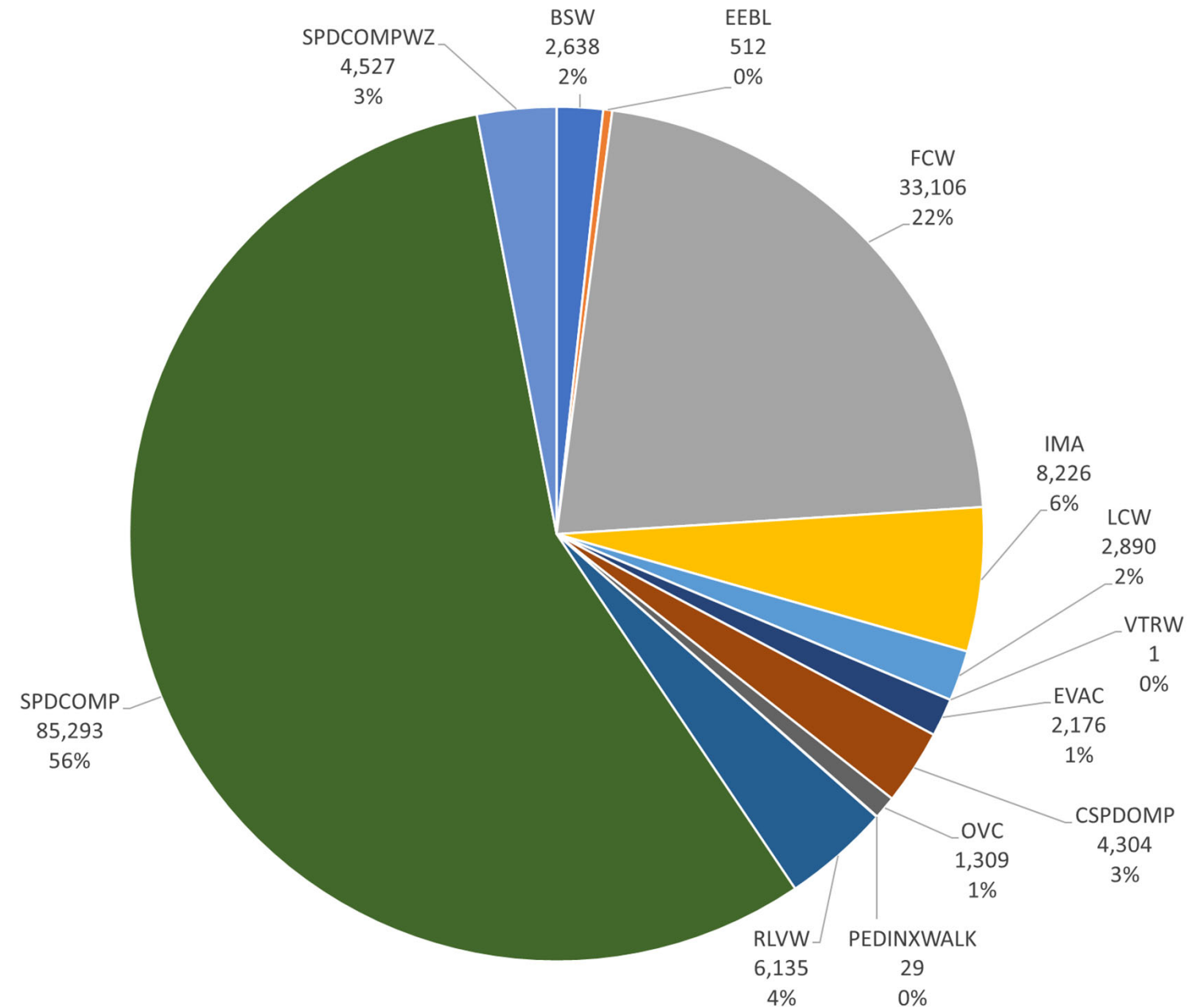


# CV Events by Type (Released)

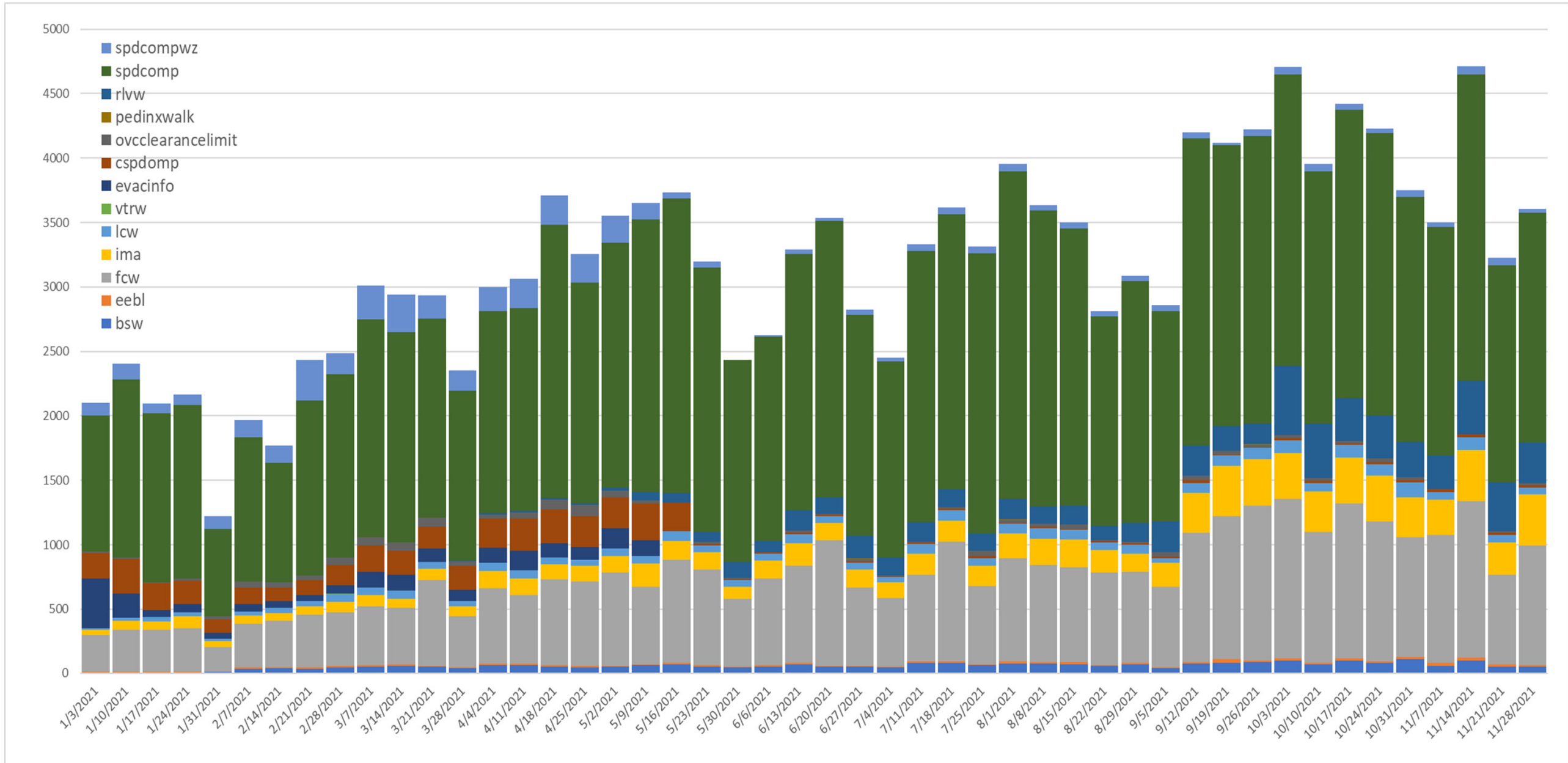
## Jan - Nov 2021: Released

- 153,354 Total Events
- 31% V2V, 69% V2I

Obfuscated event files released to the ITS DataHub



# Weekly Events Released





# Forward Collision Warning (FCW) Events

Jan 1 – Nov 30, 2021:

33,106 events

V2V citywide



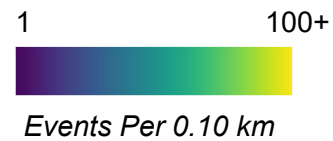


# Speed Compliance (SPDCOMP) Events

Jan 1 – Nov 30, 2021:

85,293 events

V2I areas only





# Experimental Design & Analysis Plan



# Performance Measurement Program



Safety is Job #1.

- Once alerts are activated in a vehicle, they won't be silenced.

User Needs related to Performance Measurement

- ***Maintain privacy of users throughout pilot and data collection***
- ***No enforcement***
- ***No driver evaluation***

Performance Measurement Program considers:

- Consider impacts of CV data combined with data from other sources.
- Approach to collecting the performance information.
- Approach to using data collection bins of performance information.
- Control Group vs Treatment Group

**FHWA-JPO-16-302,**

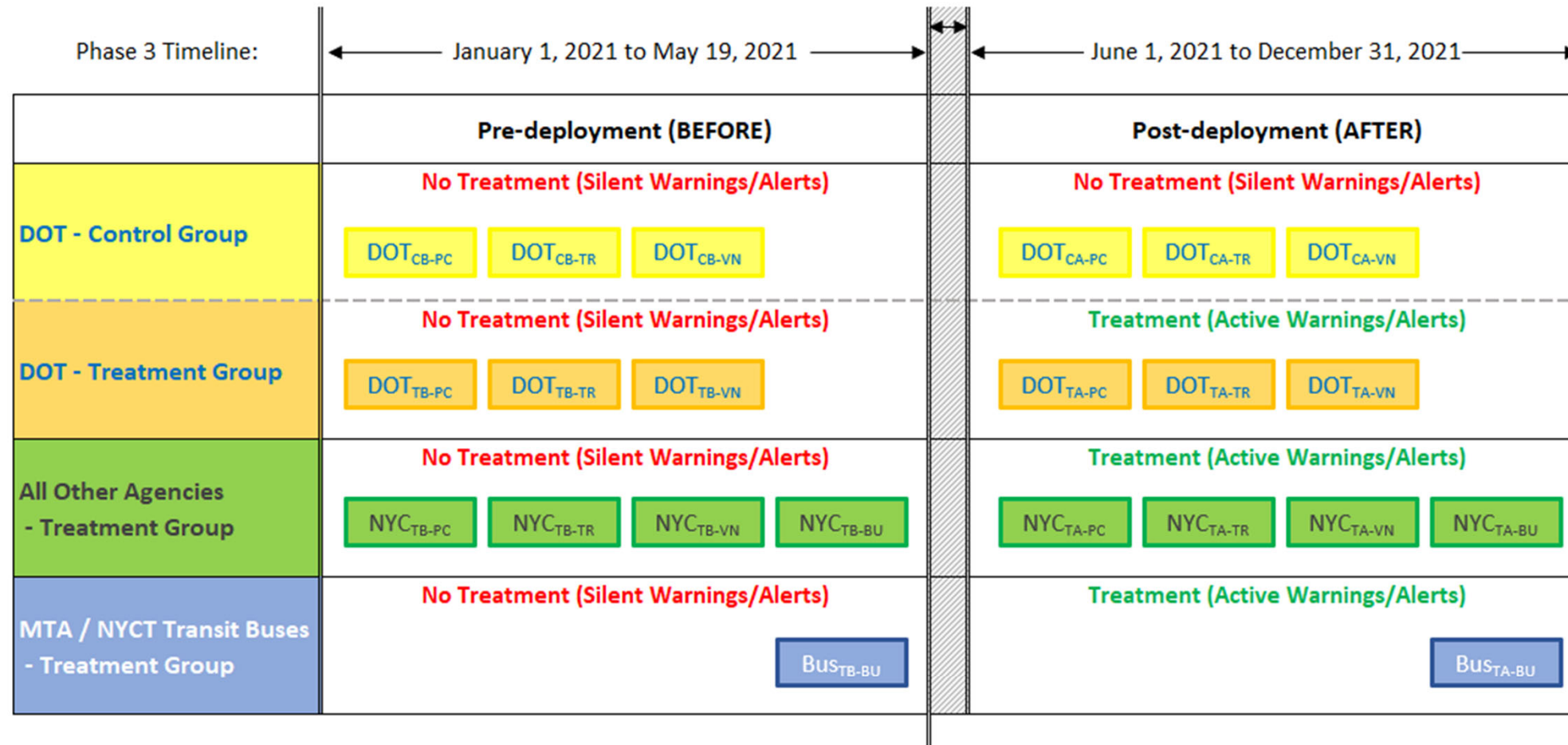
[Performance Measurement and Evaluation Support Plan - NYCDOT](#)

# Vehicle Experimental Design



Fleet Transition: May 20 - May 31, 2021  
Time for the Majority of Events to Transition to Active Alerts

CV fleet vehicles operating “business as normal” citywide



Control Vehicles ~ 7% of Fleet

**LEGEND**

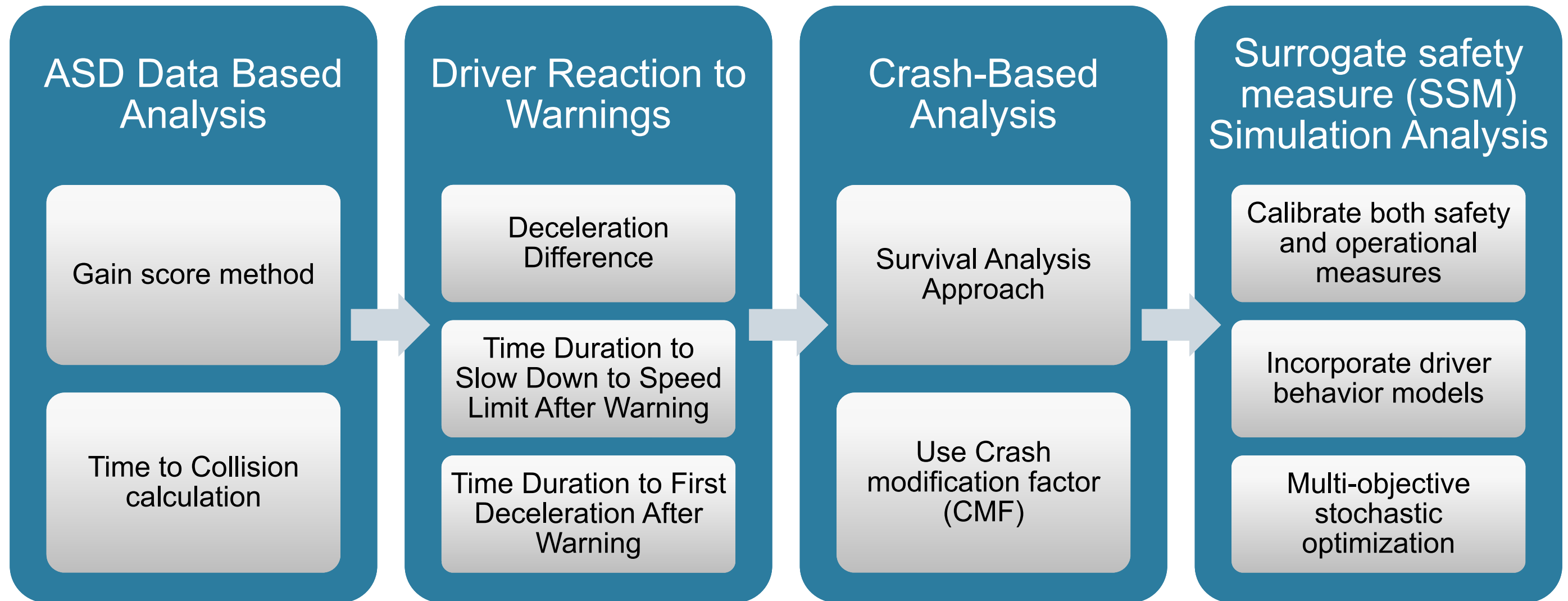
- CB: Control Group Before
- CA: Control Group After
- TB: Treatment Group Before
- TA: Treatment Group After
- PC: Passenger Cars & SUVs
- TR: Pickup and Work Trucks
- VN: Vans
- BU: Buses (Transit or non-Transit)

# Detailed Data Cleaning Prior to Analysis



- Additional steps taken to further clean and filter obfuscated records prior to analysis:
  - Illogical event warning time scales
  - Unreasonably high, zero, or constant speeds
  - Large elevation deltas between host and target vehicles
  - Stationary vehicles
  - Illogical trajectories: large gaps in BSMs, illogical relation to host and target vehicle
  - Detail vehicle trajectory speed cleaning (illogical speeds, speeds inconsistent with trajectory coordinates)
- Clean and repair when feasible
- Remove from consideration where not feasible

# Safety Analysis - Methodology





# Safety Analysis – Methodology (ASD Data)

## Gain Score Method

A commonly used method to analyze before-after control-treatment group design, is adopted (Kim and Steiner 2021).

$$\text{Safety Effect} = \left( \frac{1}{n_{T,A}} \sum_{i=1}^{n_{T,A}} \text{PM}_i - \frac{1}{n_{T,B}} \sum_{j=1}^{n_{T,B}} \text{PM}_j \right) - \left( \frac{1}{n_{C,A}} \sum_{k=1}^{n_{C,A}} \text{PM}_k - \frac{1}{n_{C,B}} \sum_{l=1}^{n_{C,B}} \text{PM}_l \right)$$

Before-after **Treatment** group difference

Before-after **Control** group difference

$n_{T,A}$  and  $n_{T,B}$  represent the total number of events in the treatment group in the after period and before period respectively.

$n_{C,A}$  and  $n_{C,B}$  represent the total number of events in the control group in the after period and before period respectively.

PM is the safety performance measure used in the evaluation.

Kim, Y., Steiner, P.M., 2021. Gain scores revisited: A graphical models perspective. *Sociological Methods & Research* 50 (3), 1353-1375.  
 Hayward, J., 1972. Near miss determination through use of a scale of danger. 51st Annual Meeting of the Highway Research Board 384, 24-34.  
 Gettman, D., Pu, L., Sayed, T., Shelby, S., Siemens, I., 2008. Surrogate safety assessment model and validation. FHWA.

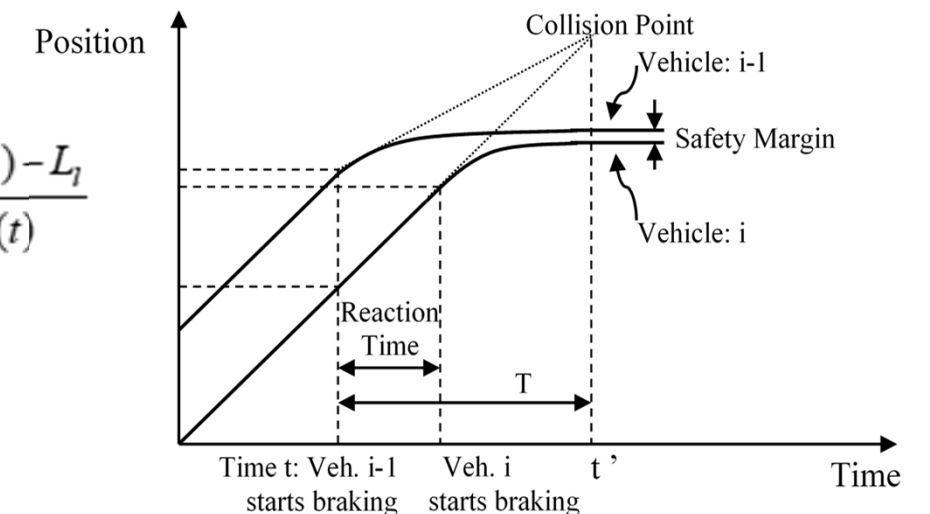
Account for potential unobserved confounding factors.

## Time to Collision (TTC)

The time for two vehicles to collide if they continue at their present speeds and on the same path (Hayward 1972).

The calculation of TTC is largely adopted from the Surrogate Safety Assessment Model (SSAM) software (Gettman et al. 2008).

$$\text{TTC}(t) = \frac{X_i(t) - X_f(t) - L_i}{V_f(t) - V_l(t)}$$



$\text{TTC}(t)$ : Time to collision at time  $t$

$v_f(t)$ : Speed of the following vehicle at time  $t$

$v_l(t)$ : Speed of the leading vehicle at time  $t$

$X_i(t) - X_f(t)$ :

Relative distance between the leading and the following vehicle at time  $t$

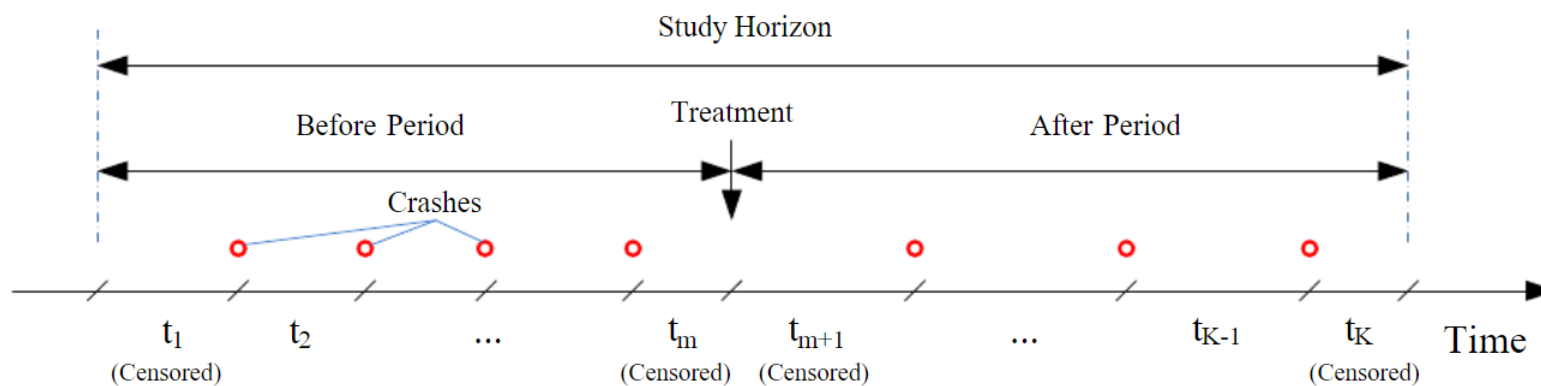
# Safety Analysis – Methodology (Crash Analysis)



## Survival Analysis Approach

- Model time intervals between two consecutive crashes instead of crash frequency
- Relax the assumption of the reference group
  - Allows the evaluation of NYC CVPD since V2V applications can be triggered at any locations

**Crash modification factor (CMF)**, a multiplicative factor used to compute expected number of crashes after implementing a given countermeasure, can be calculate as  $CMF = \exp(\beta_T)$ .



$$f(t_{ij} | \lambda_{ij}) = \lambda_{ij} \exp(-\lambda_{ij} t_{ij})$$

$$\log(\lambda_{ij}) = \beta_0 + \sum_{p=1}^P \beta_p X_{pij} + \beta_T \text{Treatment}_{ij} + \varepsilon_j$$

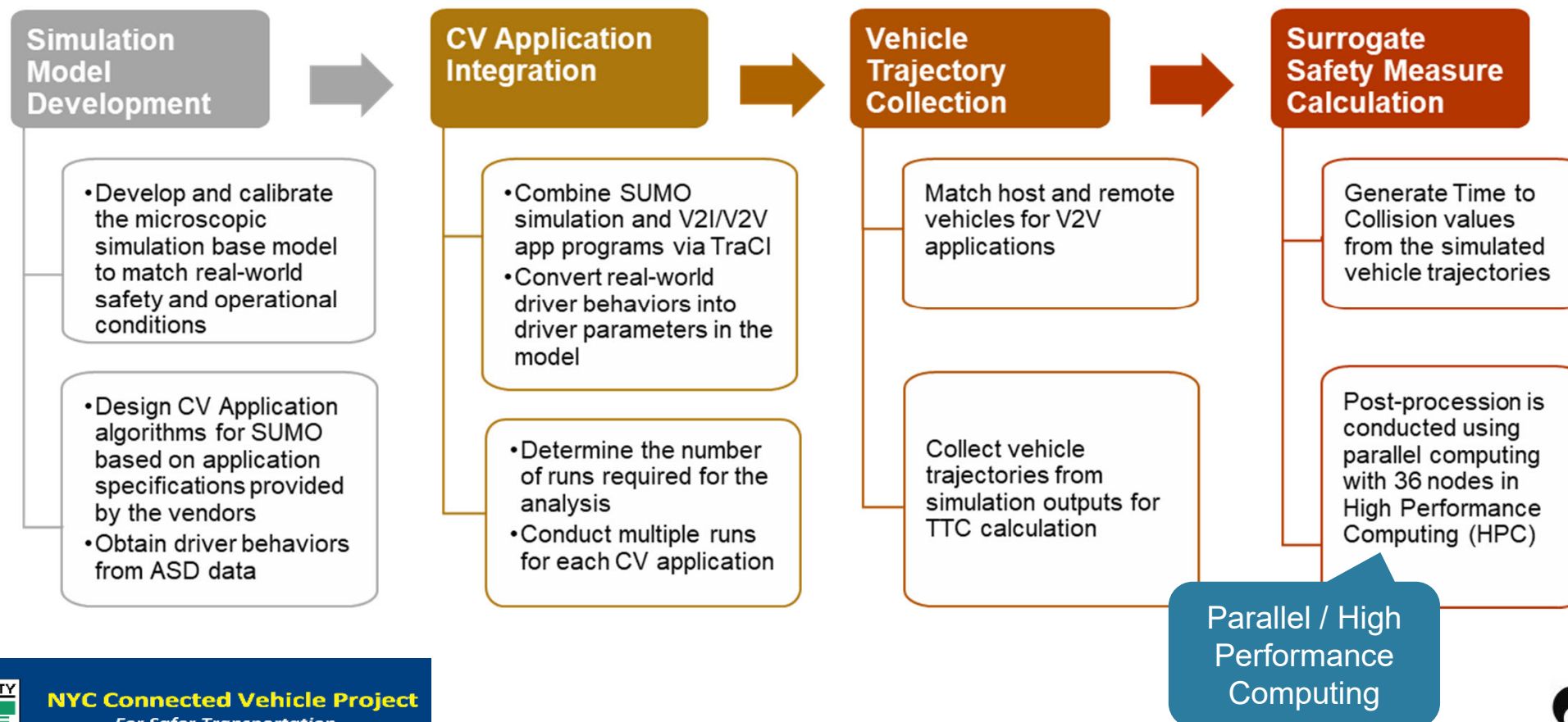
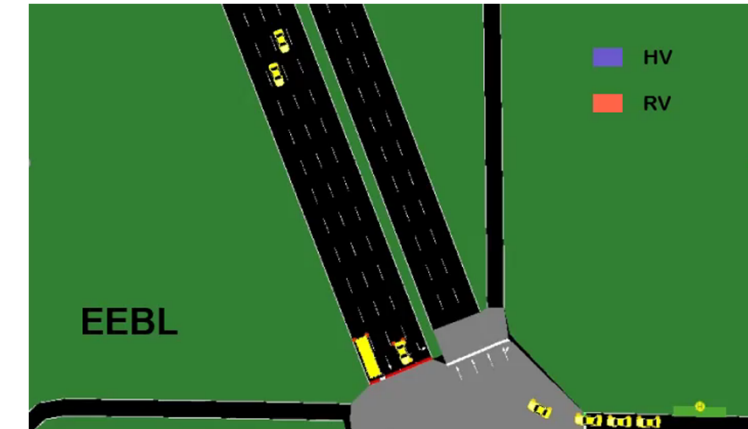
$\lambda_{ij}$  denotes the crash hazard parameter during the  $i$ th time interval at  $j$ th site.  $t_{ij}$  denotes the  $i$ th time interval at  $j$ th site.  $\beta_0$ ,  $\beta_p$ , and  $\beta_T$  are model coefficients to be estimated.

Xie, K., Ozbay, K., Yang, H., Yang, D., 2019. A new methodology for before–after safety assessment using survival analysis and longitudinal data. *Risk analysis* 39 (6), 1342-1357.

# Simulation-Based Surrogate Safety Measures (SSM) Analysis



- Base model is calibrated to both **operational** (volume, travel times, turning movements) and **safety measures** (traffic conflicts severity distributions).
- Open-source micro-simulator SUMO
- Integration of 7 CV App (5 V2V, 2 V2I) and driver behavior model
- Multi-objective stochastic optimization, about 400 simulation hours





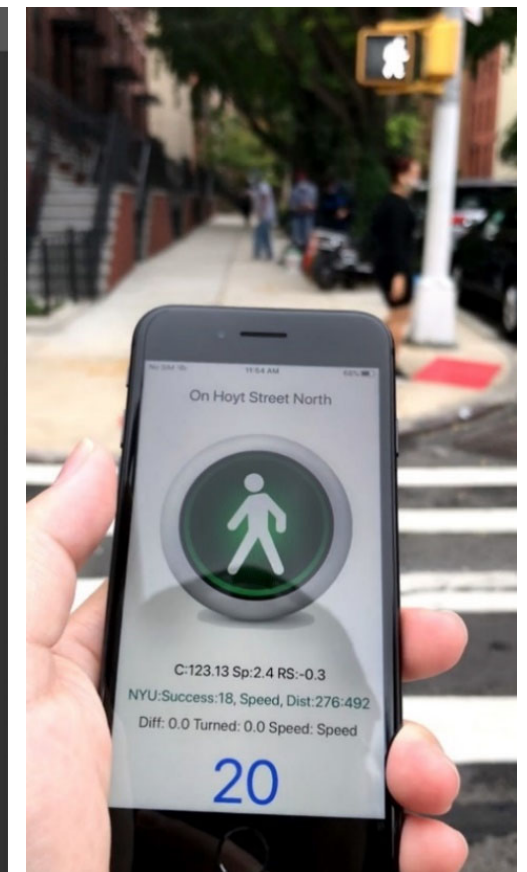
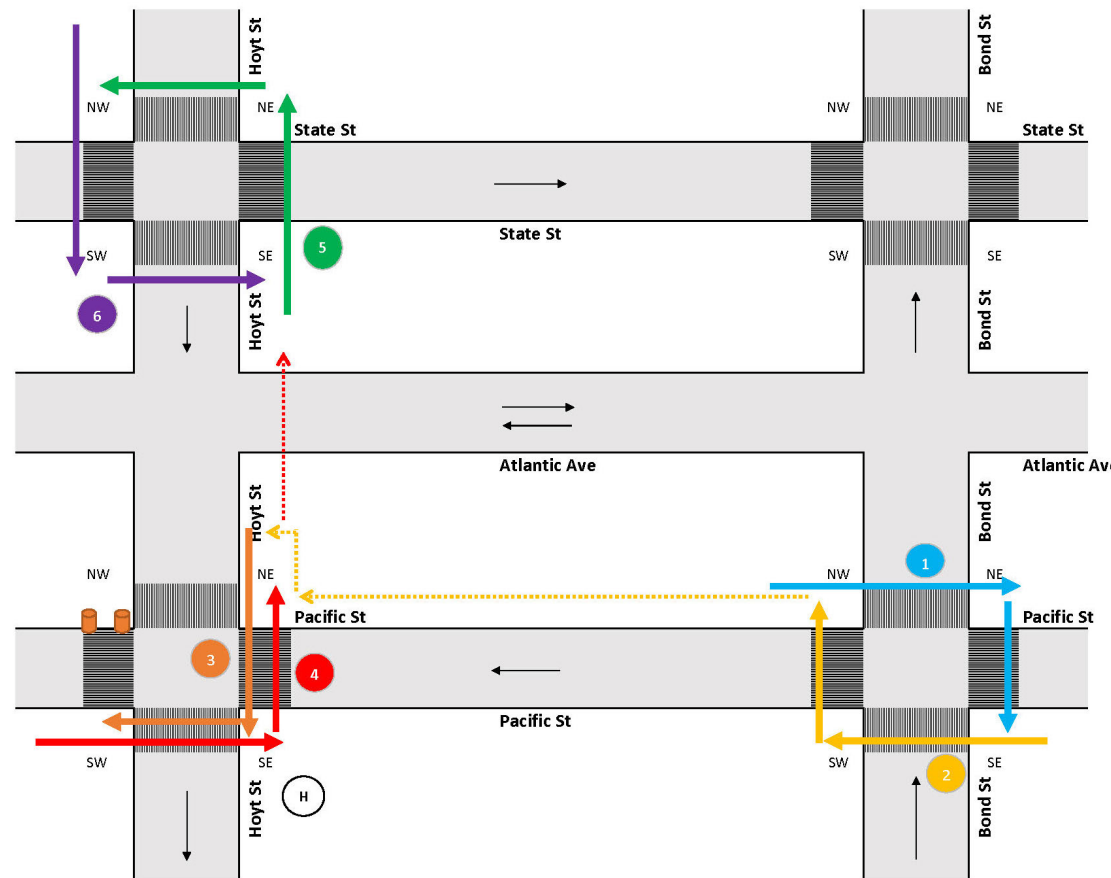
# PED-SIG Experimental Design

- Recruited 24 volunteer participants with vision disabilities via local and national organizations working with blind
- Participants navigated intersections using PID accompanied by NYU Institutional Review Board (IRB)-certified researchers.

## User Interface for the PED-SIG Mobile Phone Application

### Highlights:

- Four semi-protected intersections
- Six predefined routes, each made up of two crosswalk crossings
- Data collected from operational data logs, field observation and pre- and post-experiment surveys







# Performance Results & Findings

# Qualitative Results – Driver Surveys



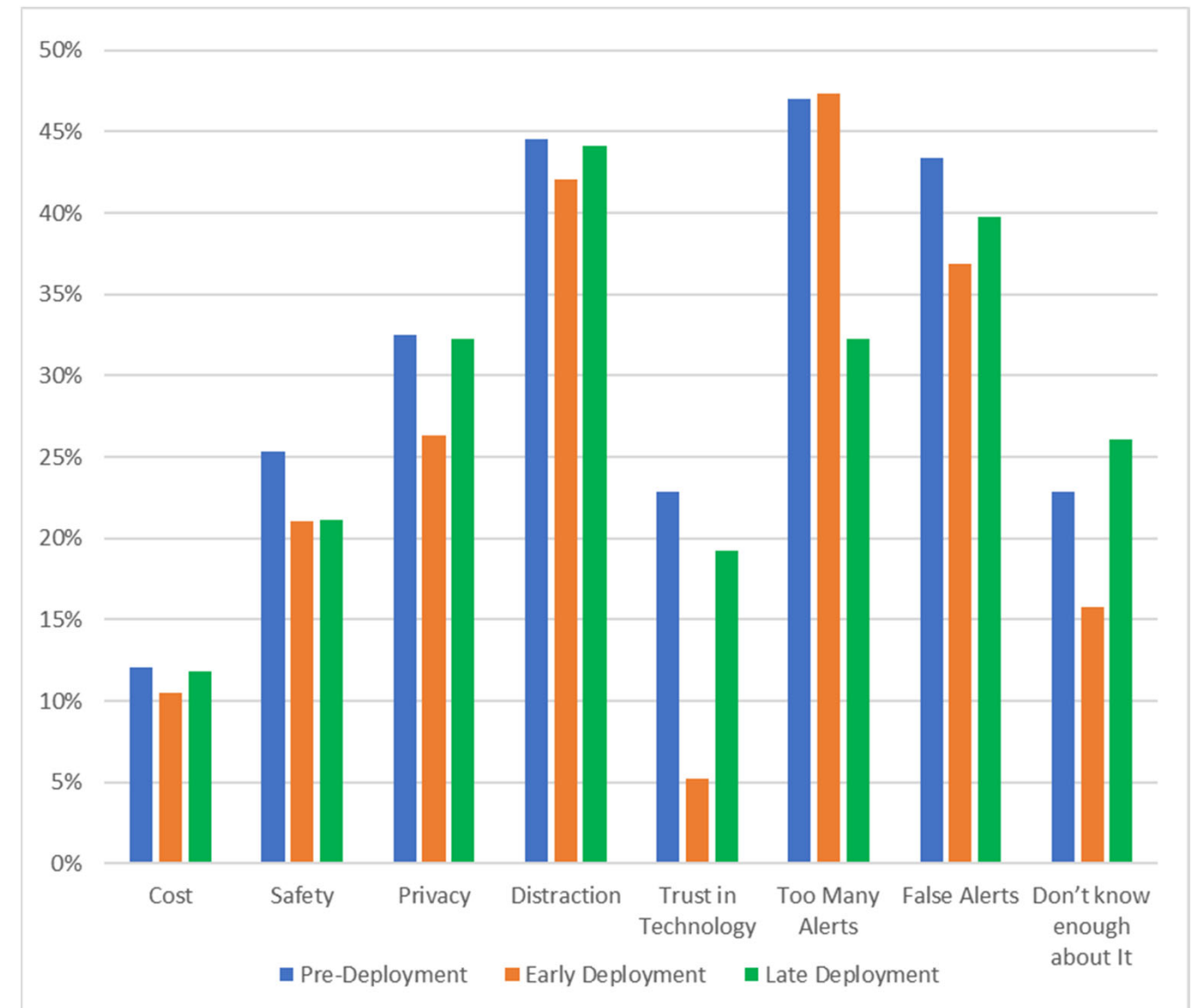
- Three surveys conducted:
  - Pre-deployment: Last month of before period
  - Early-Deployment: 1-2 Months into after period
  - Late-Deployment: 4-5 Months into after period
- Focus areas for questions:
  - Vehicle Usage on a typical day
  - Attitudes and perception of CV technology and Driving Safety in NYC
  - User experience with CV Apps (after period surveys only)
  - Basic demographics

# Driver Survey Findings



- Typical respondent:
  - DOT employee
  - Drives in Manhattan and Queens
  - More than 10 years experience driving in NYC for work
  - Drives 20-50 miles\* per day
    - *\*higher than typical vehicle use*
  - Likelihood of a crash or near crash with:
    - Another Vehicle: Slightly likely
    - A Pedestrian or Bicyclist: Slightly likely
    - Off-road or infrastructure: Not at all likely

## Concerns about CV Technology:



# Driver Survey Findings – User Experience (1)

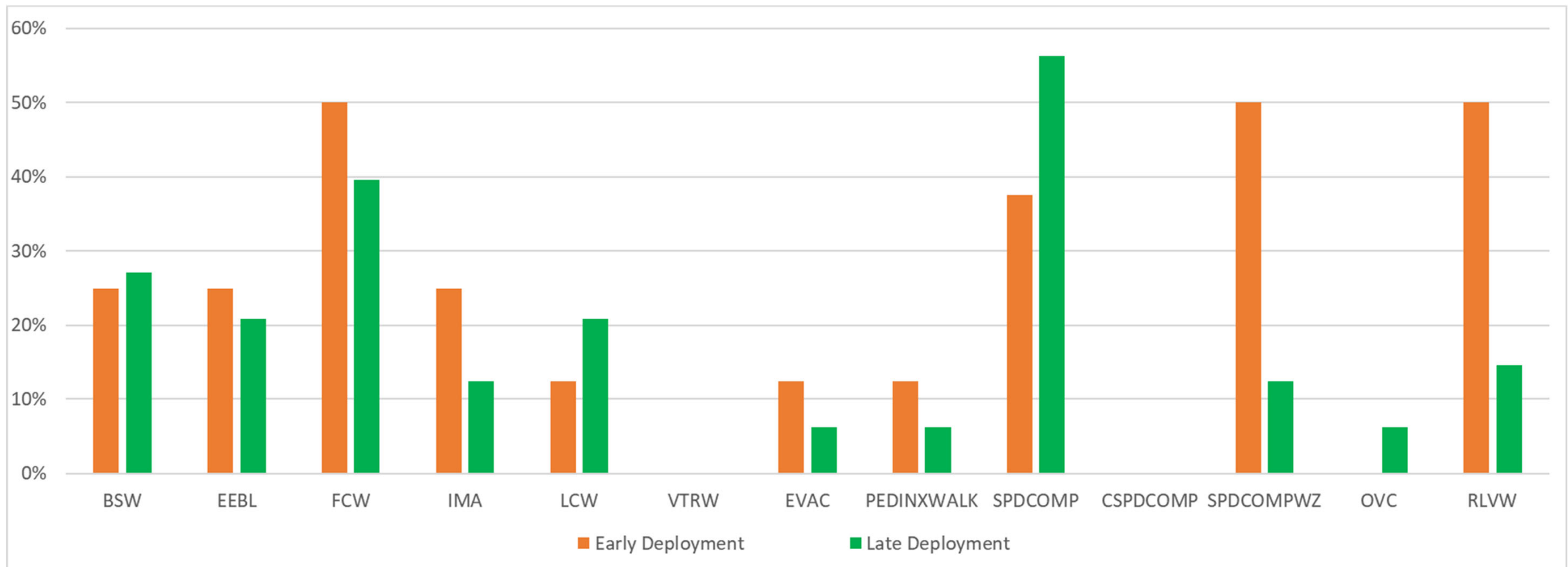


- Frequency of alerts heard varied evenly from never to many times per day
- Majority found alerts too loud
  - Somewhat loud 38%, Much too loud 22%
- *Majority found alerts distracting:*
  - 17% Slightly, 30% Moderately, 19% Very, 23% Extremely
- *Majority found alerts helpful:*
  - 23% Slightly, 27% Moderately, 8% Very, 1% Extremely
- Have the alerts affected how you drive?
  - No: 70%
  - Yes: 30% (split 20% very negative, 23% negative, 48% positive, 10% very positive)

# Driver Survey Findings – User Experience (2)



- 38% reported alerts helped them drive more safely
  - Apps reported as helpful:



# Analytic Results



- App Specific Analysis that follows:
  - ASD-based Analysis:
    - Number of events refers:
      - CV event file warning logs included in the analysis
      - From January - April (before period) and June - September (after period).
      - Events remaining after all error checking, obfuscating, cleaning, and filtering is complete
  - SSM-based Analysis:
    - Simulation of CV apps based on driver behavior response analysis from event logs
    - Assumed a 5% CV market penetration rate

# Safety Analysis Results

## SPDCOMP: Speed Compliance




- Number of events:

- 40,635



Statistically significant at 95%

- ASD-based Analysis:

- Compared to silent warning scenarios, there are additional 47 events per 1,000 SPDCOMP events that driver slowed to the speed limit when treatment was enabled 

- Driver behavior response:

- 0.148 m/s<sup>2</sup> extra deceleration on average after speed compliance warnings were issued 

- 0.619 s reduction in time duration to slow down to speed limit 

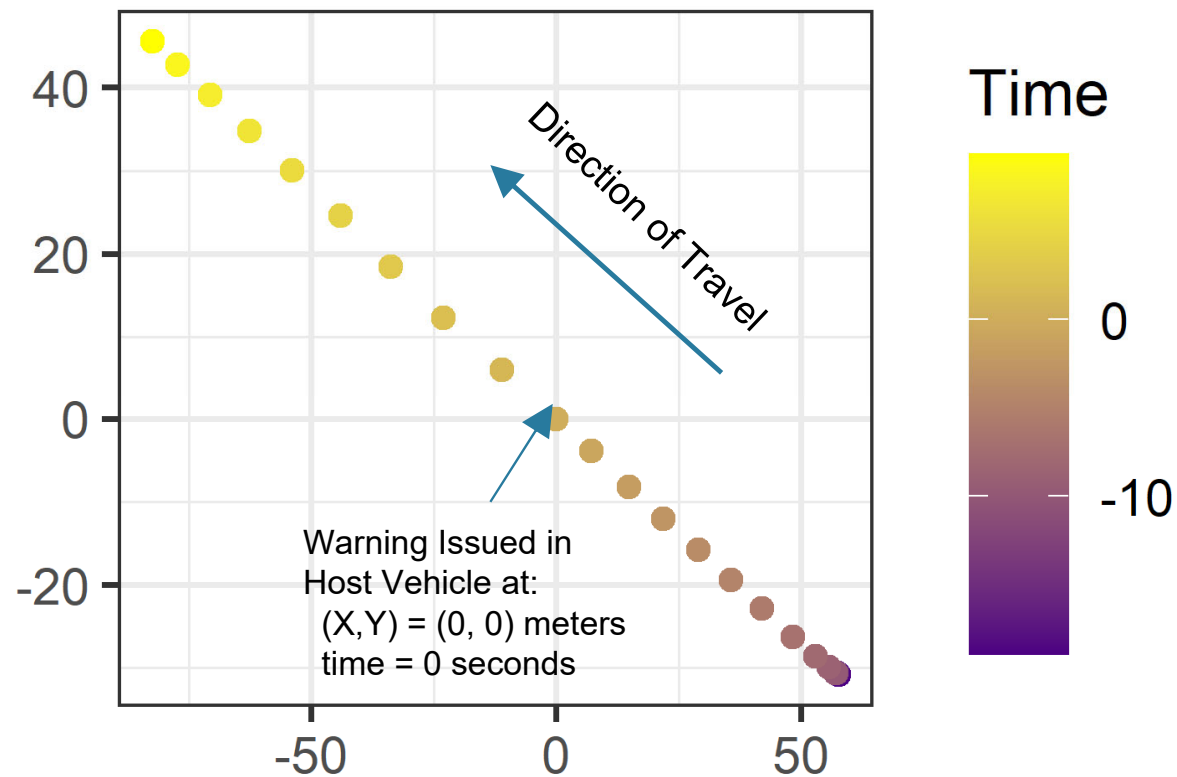
- Minor reduction (<1 s)

# Safety Analysis Results

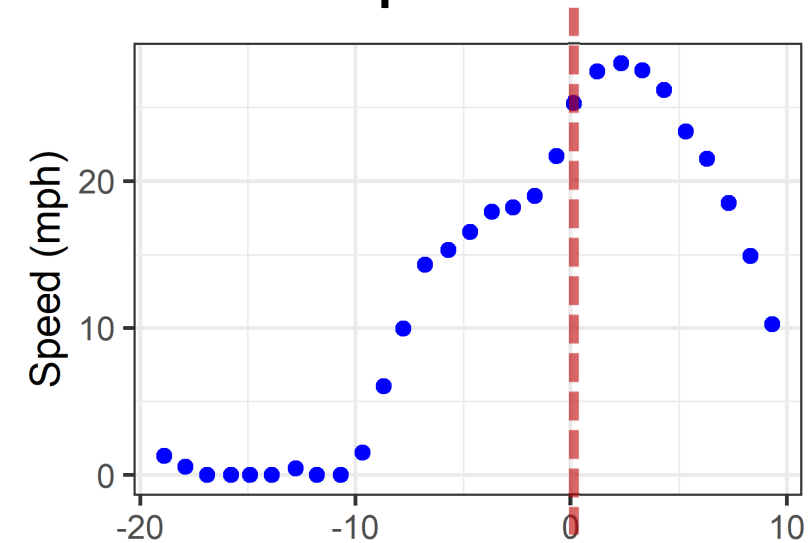
## SPDCOMP: Driver Response Example



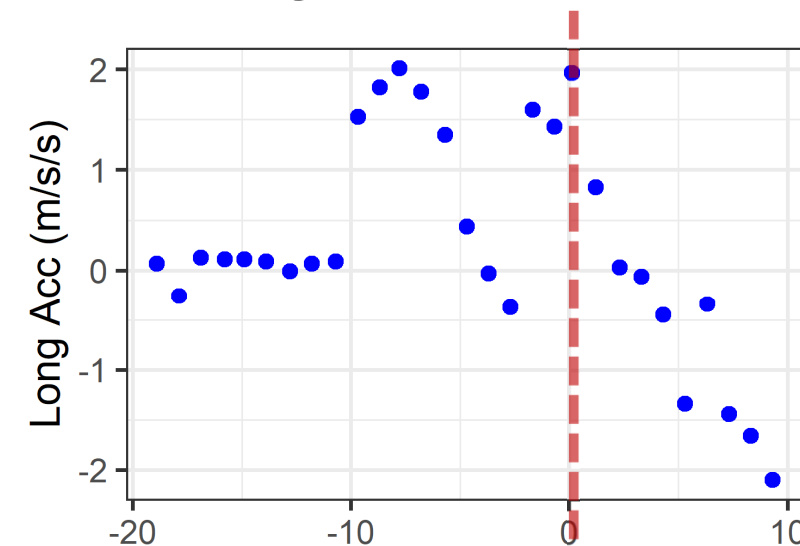
### Vehicle Trajectory



### Speed



### Long Acceleration





# Safety Analysis Results

## CSPDCOMP: Curve Speed Compliance



### ■ Number of events:

- 27



Statistically significant at 95%

### ■ ASD-based Analysis:

- 8.750 mph reduction in vehicle speeds at curve entry



- 0.691 m/s<sup>2</sup> reduction in lateral acceleration in the curve



### ■ Driver behavior response:

- 0.908 m/s<sup>2</sup> decrease in deceleration difference





- In general, drivers did not decelerate after being given the curve speed compliance warning

# Safety Analysis Results

## SPDCOMPWZ: Speed Compliance in Work Zone



- Number of events:
  - 2,665
- Driver behavior response:
  - There is an extra  $0.427 \text{ m/s}^2$  deceleration from the drivers on average after being issued the warnings 
  - 2.260 s reduction in time duration to slow down to speed limit 



Statistically significant at 95%



U.S. Department of Transportation








# Safety Analysis Results

## V2V Applications



 Statistically significant at 95%

 Marginally insignificant at 95%

CV App	# of Events	15 <sup>th</sup> Percentile TTC - ASD		15 <sup>th</sup> Percentile TTC – SIM (5% CV)	
FCW	12,255	0.198 s increase		1.60 s increase	
EEBL	107	Insignificant according to 95%		1.58 s increase	
BSW	738	Insignificant according to 95%		2.43 s increase	
LCW	873	0.265 s increase		2.03 s increase	
IMA	2,666	2.951 s increase		-	

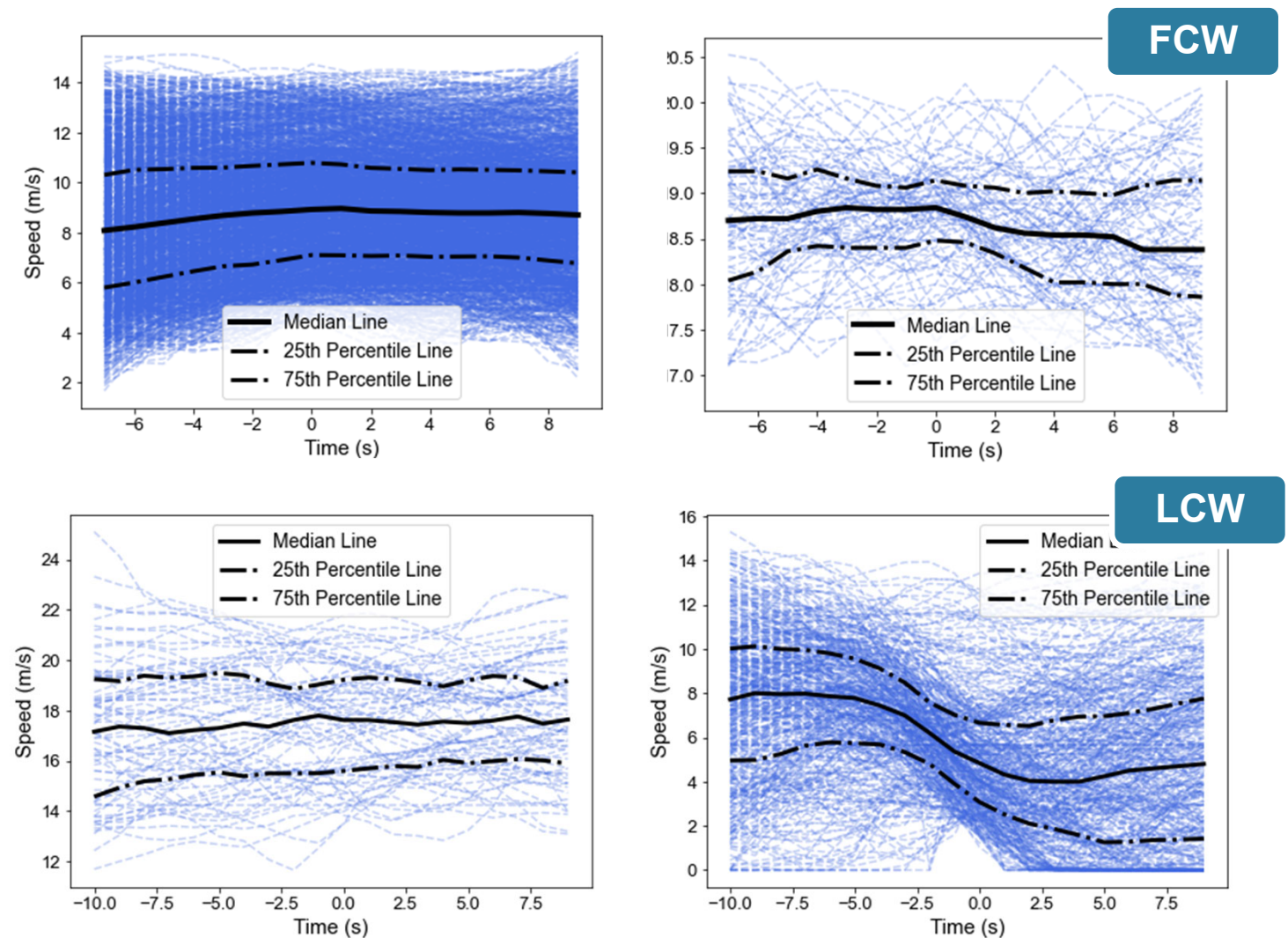
# Safety Analysis Results

## V2V Applications - Driver Responses



	# Events in Treatment Group	Reacted	No Reaction
FCW	2,263	18%	82%
EEBL	68	100%	0%
BSW	473	84%	16%
LCW	472	84%	16%
RLVW	1,114	100%	0%





### Clustering of Driver Responses Based on Speed-Time Relationship




# Safety Analysis Results

## RLVW: Red Light Violation Warning



- Number of events: 2,073
- ASD-based Analysis:
  - Compared to silent warning scenarios, there are approximately 152 fewer likely red-light violations\* per 1,000 RLVW events when treatment was enabled 
- Driver behavior response:
  - Drivers tended to decelerate approximately 0.137 m/s<sup>2</sup> more after RLVWs were issued 
  - 0.083 s reduction in time duration to first deceleration after warning 
- Simulation-based SSM Analysis:
  - 1.20 s increase in 15<sup>th</sup> percentile TTC values 

\*Drivers did not come to a full stop after receiving a red-light violation warning.


 Marginally insignificant at 95%

 Statistically significant at 95%

# Safety Analysis Results

## PEDINXWALK: Pedestrian in Crosswalk



- Number of events: 20
- Simulation-based SSM analysis
  - 1.80 s increase in 15<sup>th</sup> percentile TTC values  for vehicle to pedestrian conflicts



Statistically significant at 95%



Thermal Sensor  
Pedestrian Detection

# Safety Analysis Results

## Crash-Based Analysis





### Challenges:




- **Crash records cannot be linked to instrumented vehicles** due to the privacy/liability concerns.
- Thus, **crashes involving all vehicles** in NYC during the study horizon are used for crash analysis.
- Other safety-related **confounding factors that occurred simultaneously with the NYC CVPD**, including the COVID-19 pandemic, Vision Zero projects, planned special events, and so on.
- The crash analysis results should be interpreted as **a combined treatment effect for all the potential safety-related “treatments”** that occurred simultaneously around NYC during the NYC CVPD implementation period and may not be solely due to the CV applications.

# Evaluation Results - Crash-Based Analysis



## Rear-end Crashes (FCW & EEBL)

- Number of rear-end crashes (NYPD crash database): 4,581
- Crash modification factors (CMFs):
  - Injury crashes: 0.947 
    - Reduction in injury rear-end crashes after account for crash exposure
  - Property damage only (PDO) crashes: 0.906 
    - Reduction in PDO rear-end crashes after account for crash exposure

-  Statistically significant at 95%
-  Marginally insignificant at 95%
-  Statistically insignificant at 95%

## Side-swipe Crashes (BSW & LCW)

- Number of side-swipe crashes (NYPD crash database): 1,471
- CMFs:
  - Injury & Fatal crashes: 0.985 
  - PDO crashes: 0.850 
    - Reduction in PDO side-swipe crashes after account for crash exposure

### Crash modification factor (CMF)

A multiplicative factor to compute expected # crashes after implementing a given countermeasure.

### Crash Exposure

Traffic volume data are used as crash exposure to account for the COVID-19 recovery.

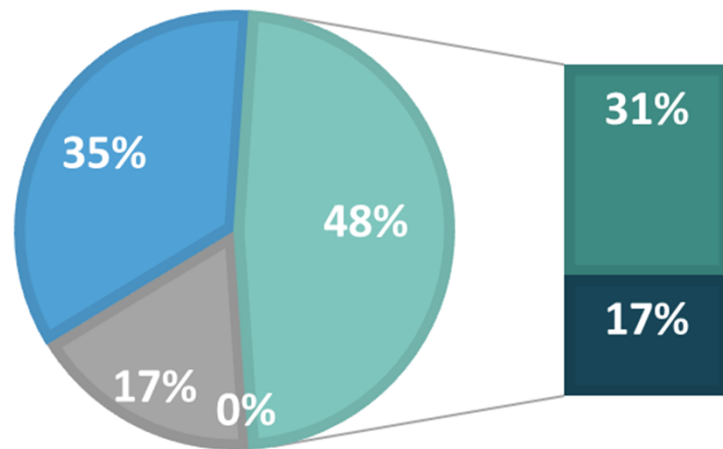


# PED-SIG Evaluation Results



## Overall PED-SIG Application Rating

■ Poor 
 ■ Fair 
 ■ Good 
 ■ Very Good 
 ■ Excellent



**14**  
**100%**  
**96%**

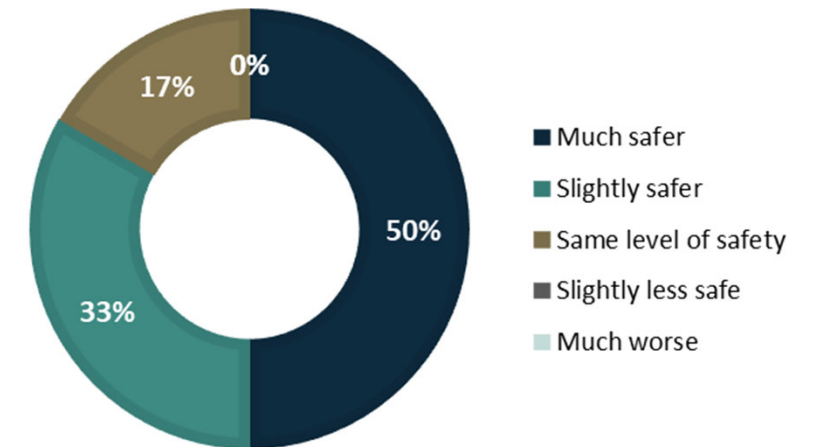
Participants were recruited from 14 local and national organizations working with blind communities.

Anticipated that pedestrians would benefit from the use of PED-SIG technologies.

Felt they were given sufficient time to cross the intersection when using the PED-SIG application.

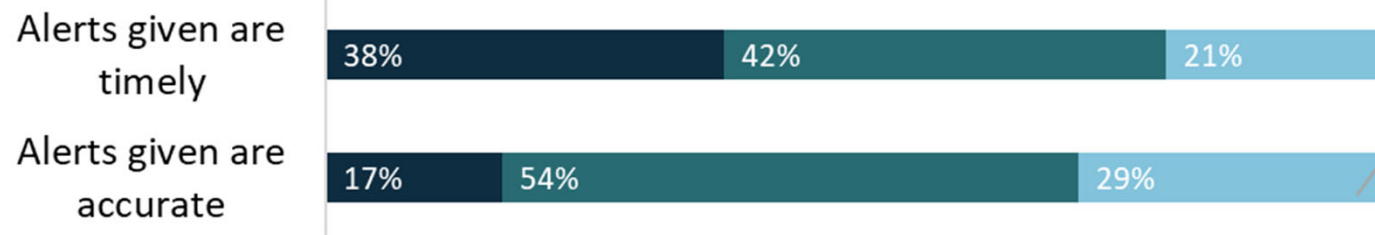
**PED-SIG**

## PED-SIG Application Safety Perception



## PED-SIG Application Alerts Were Accurate/Timely

■ Always 
 ■ Mostly 
 ■ Sometimes 
 ■ Never



## The main problems experienced:

- Location information provided was not always accurate (75%)
- Slow responses (25%)
- Orientation was not accurate (21%).



# PED-SIG Evaluation Results (1)



**Operation data logs and field observations**  
Approximately **170 runs**, each made up of two crosswalk crossings, were completed.

## Performance Measures:

- Pedestrian crossing speed
- Pedestrian crossing travel time
- Waiting time at intersection for crossing
- Times out of crosswalk



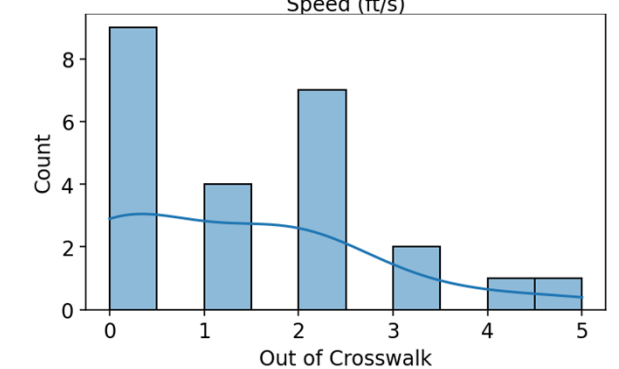
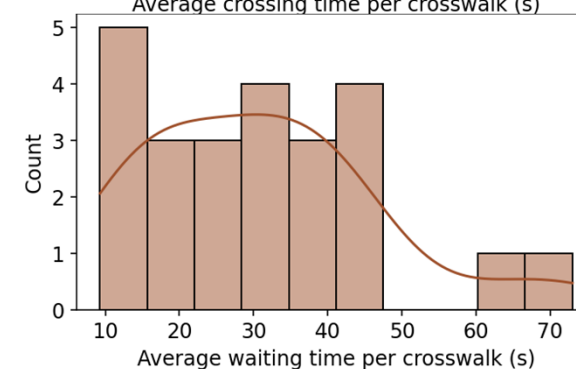
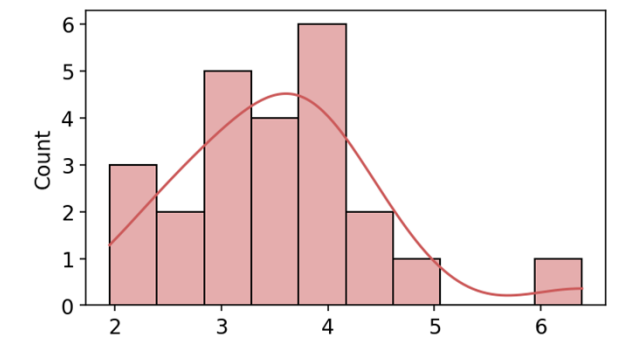
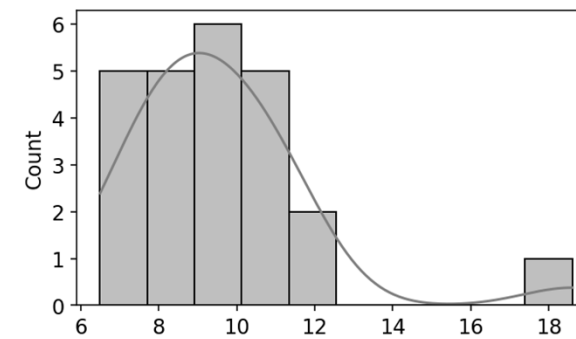


# PED-SIG Evaluation Results (2)

## Main Findings:

- 63% of the participants veered off the crosswalk at least once.
- 54% of them crossed the streets faster than the 3.5 ft/s assumption used for signal timing design.
- The waiting time per crosswalk varies among different participants.

	Crossing speed (ft/s)	Avg crossing time per crosswalk (s)	Avg waiting time per crosswalk (s)	Out of Crosswalk (Avg #times)
<b>Mean</b>	3.6	9.6	31.0	1.4
<b>Std</b>	0.9	2.4	15.9	1.4
<b>15<sup>th</sup> /85<sup>th</sup> Percentile</b>	[2.6, 4.2]	[7.7, 11.0]	[14.9, 43.0]	[0.0, 2.3]





# CV Transition Plan

# Next Steps



- **Original plan** anticipated a sustainable investment for future connected vehicles
- **Driven by FCC First Report and Order** to clear the lower 45 MHz by July 2022
- **Additional complications** create substantial complexity with respect future temporary configurations
- **Suspend DSRC transmissions** after completing the project's Phase 3 Operation
- New York City is not abandoning CV concepts



# NYCDOT Perspective on CV Pilot Deployment



- **Successful project!** Produced large data set for future research
- **Equipping and managing a fleet is challenging** with many “black box” components
- **Many Lessons Learned and Contributions to CV Community**
  - Urban Deployment and Performance Reliability
  - Maturity of CV Deployment, Operations, and Maintenance **at scale**
  - Managing Security Aspects for CV Deployment

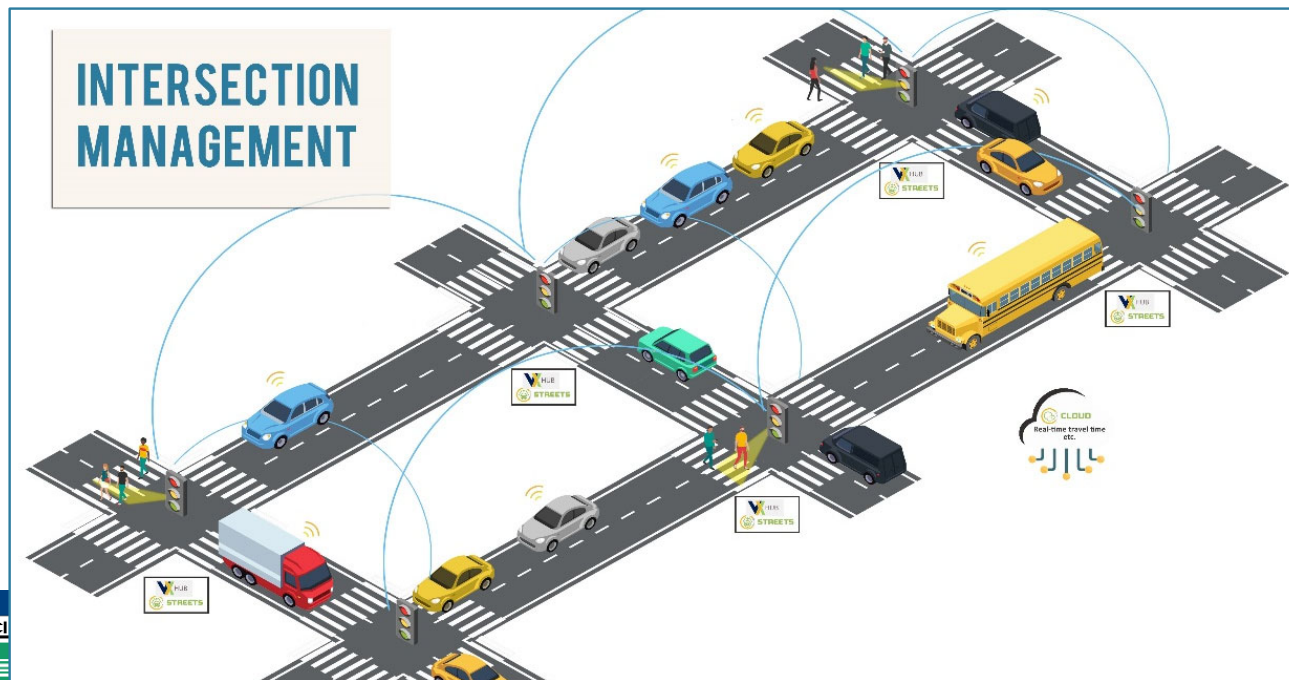
# NYC Cooperative Driving Use Cases



## CD for ACV Use Cases Mapping to NYC:

- 1) Pedestrian and bicyclist safety through cooperation
- 2) Cooperative work zones
- 3) Cooperative intersection management.

See full use cases [here](https://c2smart.engineering.nyu.edu/wp-content/uploads/2021/12/CD_for_ACV_NYU_UseCase_DataAnalysis.pdf) ([https://c2smart.engineering.nyu.edu/wp-content/uploads/2021/12/CD\\_for\\_ACV\\_NYU\\_UseCase\\_DataAnalysis.pdf](https://c2smart.engineering.nyu.edu/wp-content/uploads/2021/12/CD_for_ACV_NYU_UseCase_DataAnalysis.pdf))





# FUTURE OF CVs IN NYC



With deployed CV infrastructure and experience in NYC, we could:

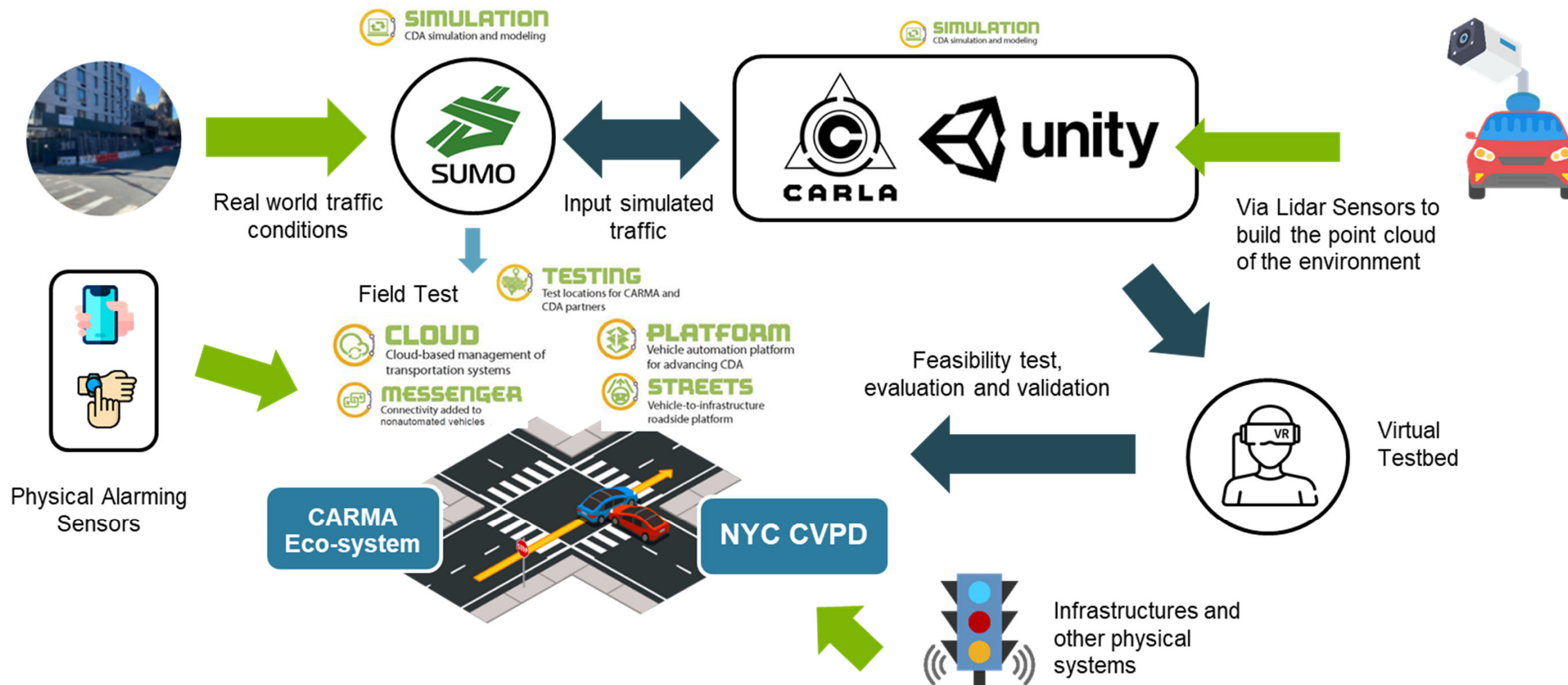
- Enable “Cooperative Perception”
  - **Cooperative Driving for Advanced Connected Vehicles** (CD for ACVs) at urban intersections in NYC
- Develop co-simulation environments for safe and realistic testing of CD for ACV
  - Cyber-physical testbed for prototype testing, learning, evaluation and integration of FHWA's CARMA.





# Potential Cyber-Physical Testbed

NYC could integrate its facilities for an expanded CARMA cyber-physical testbed.



# Questions?



Photo Courtesy: MTA New York City Transit

# Stay Connected



## Contact for CV Pilots Program/Site AORs:

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- Walter During, Evaluation COR, [Walter.During@dot.gov](mailto:Walter.During@dot.gov)

## Visit CV Pilot and Pilot Site Websites for more Information:

- CV Pilots Program: <http://www.its.dot.gov/pilots>
- NYCDOT Pilot: <https://www.cvp.nyc/>
- Tampa (THEA): <https://www.tampacvpilot.com/>
- Wyoming DOT: <https://wydotcwp.wyoroad.info/>

