

Estimating Traffic Crash Counts Using Crowdsourced Data

Volpe Center

Erika Sudderth, PhD

Dan Flynn, PhD

Michelle Gilmore

BTS

Pat Hu

Ed Strocko

OST-P

Paul Teicher

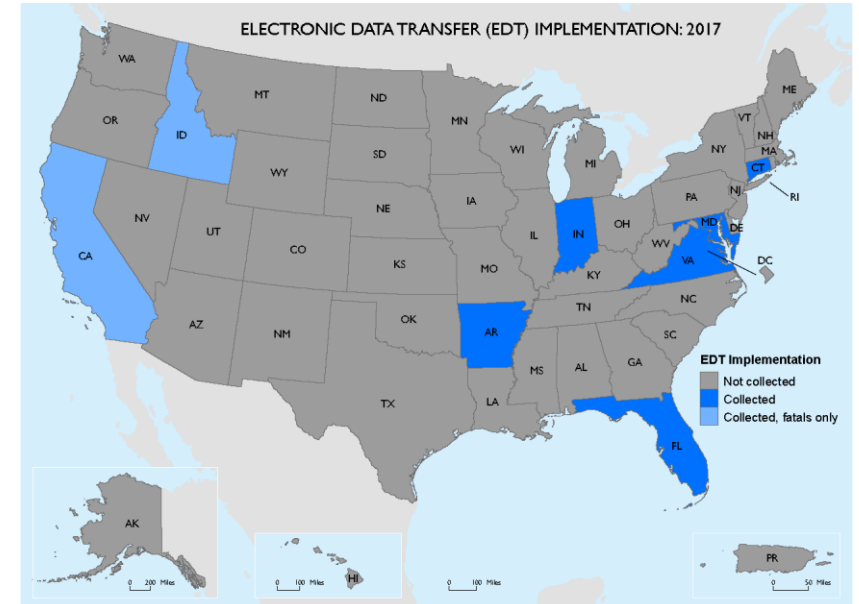
David Winter

2018-10-24



Challenge: Tracking crashes in near real-time

- Crash data are typically available for certain crashes, after several months
- EDT (Electronic Data Transfer) of police accident reports available nightly for nine states
- Waze incident data available where user reported, all 50 states and DC, every 2 minutes
- Waze and EDT could provide near-real time, granular estimates of crashes to inform safety policy and operations



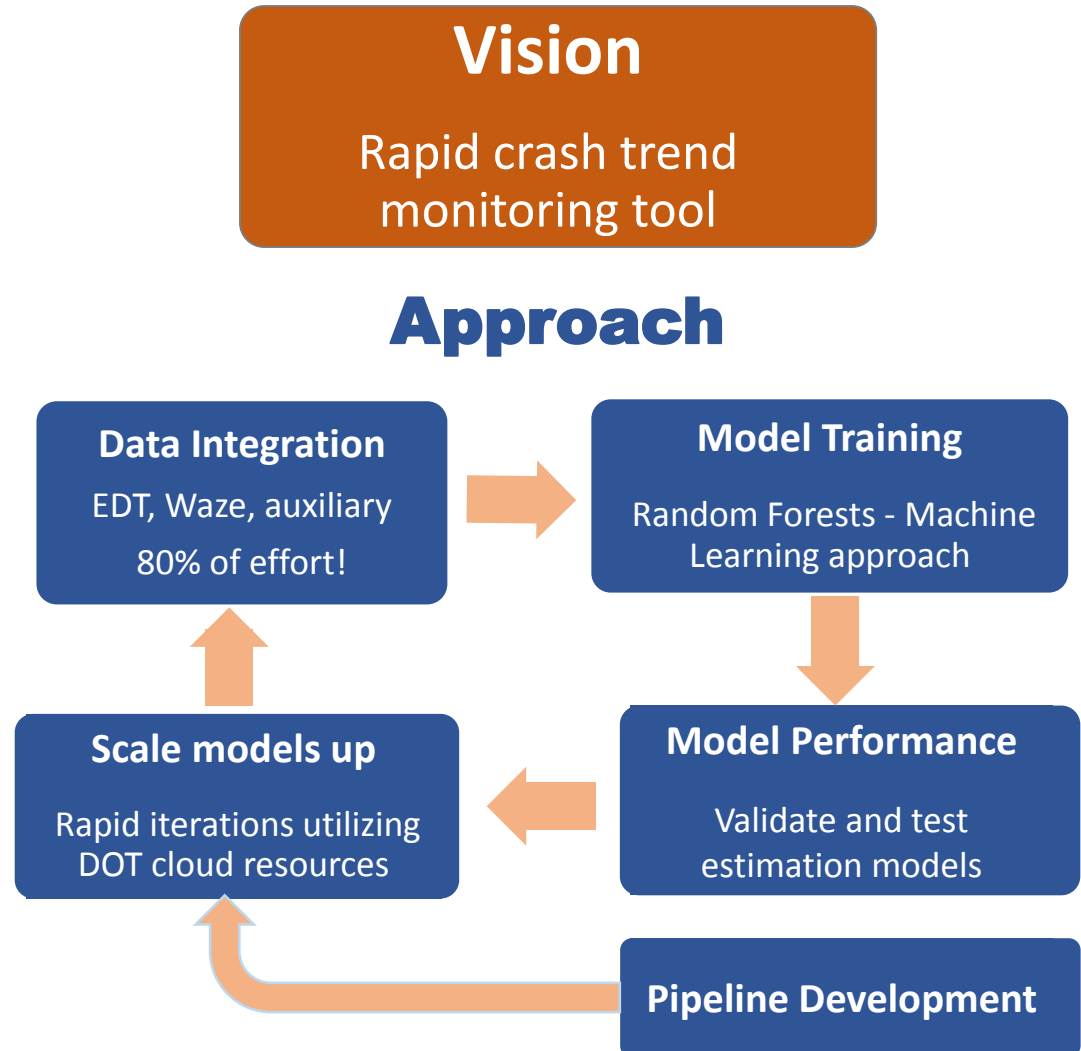
Safety Data Initiative: Waze Pilot Project Overview

Objectives

- Use crowdsourced data insights to improve transportation safety

Questions

- **Can we integrate DOT data resources at large scales?**
- **Do Waze data support vision of a rapid crash indictor?**



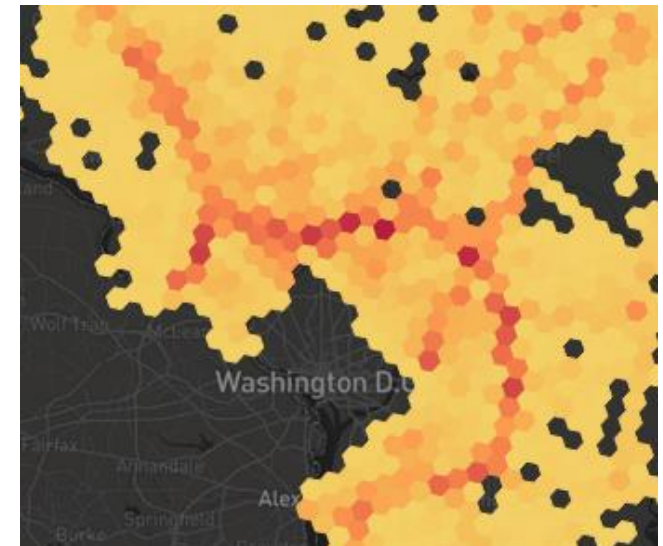
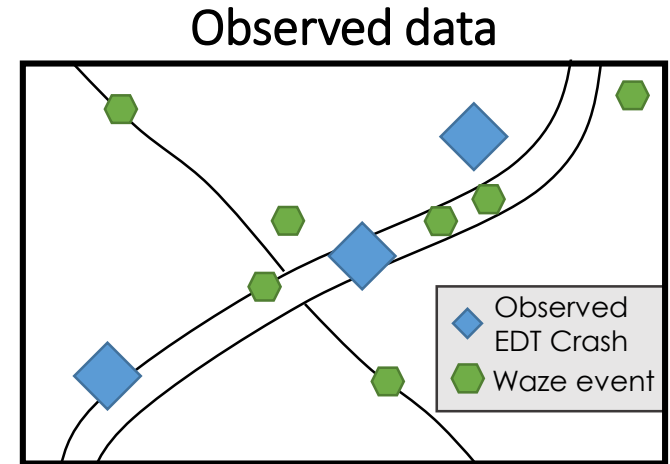
Analysis: Challenges and Solutions

Challenges

- Waze and EDT coordinates do not all align with FHWA road network
- How do we associate Waze events and EDT reports?
- Need to define zeros (time and places with no accidents)

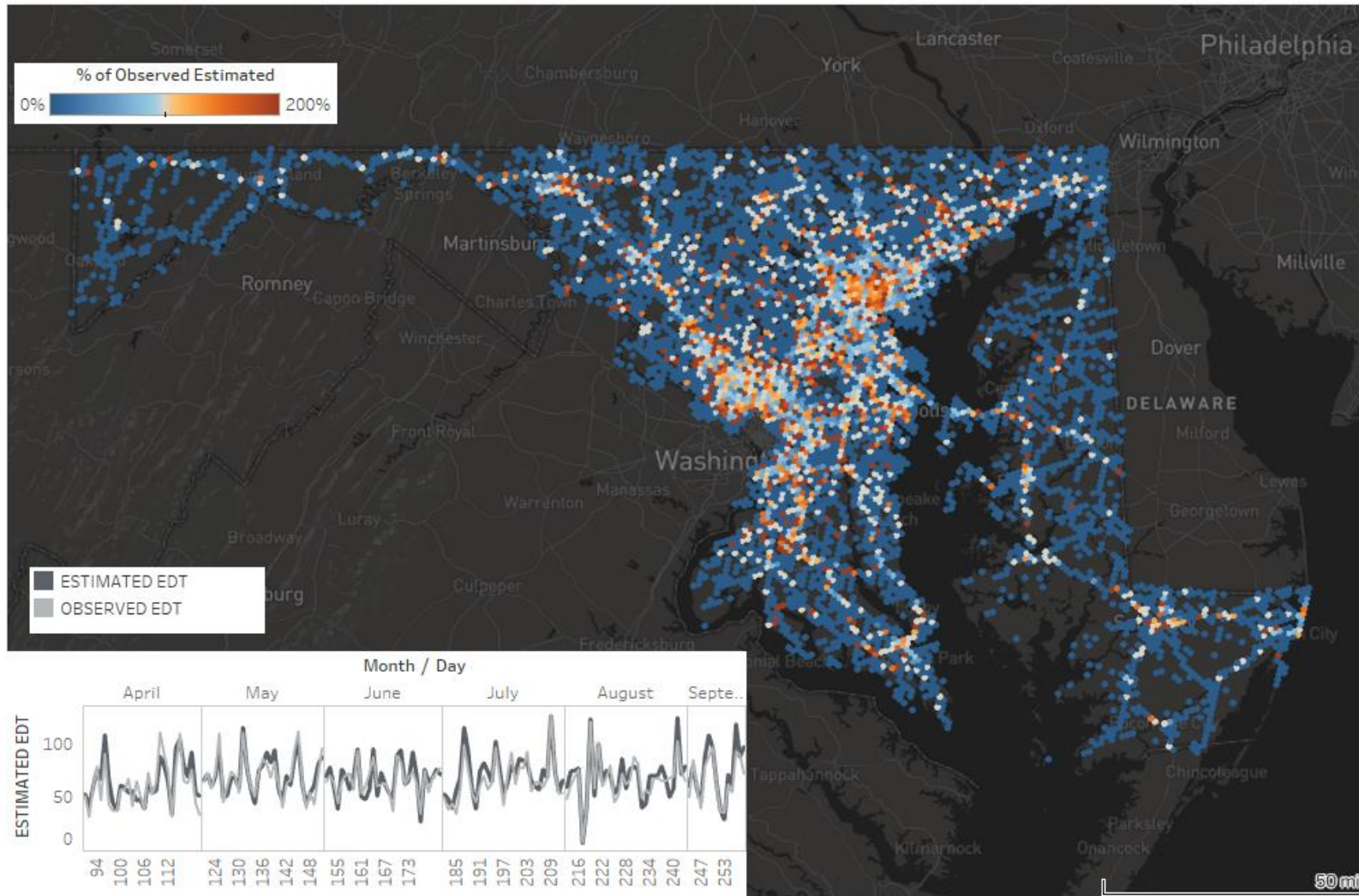
Solutions

- Spatial aggregation of data to hexagonal grids (1-mile area)
- Match Waze to EDT on user-selected buffers in space and time
- Define zeros as grid cells and time periods with 1 or more non-accident Waze events but no EDT reports



Model Performance (April-Sept 2017 in MD)

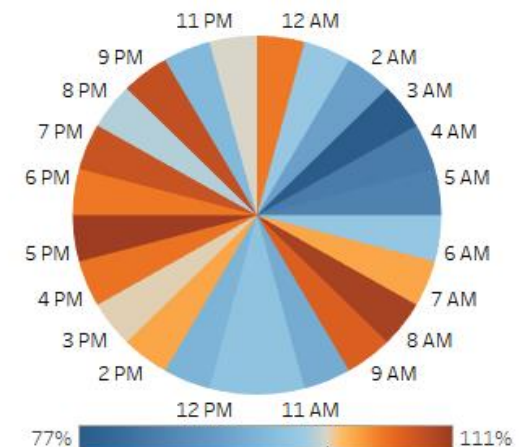
Model estimates highly accurate overall; miss some precise patterns



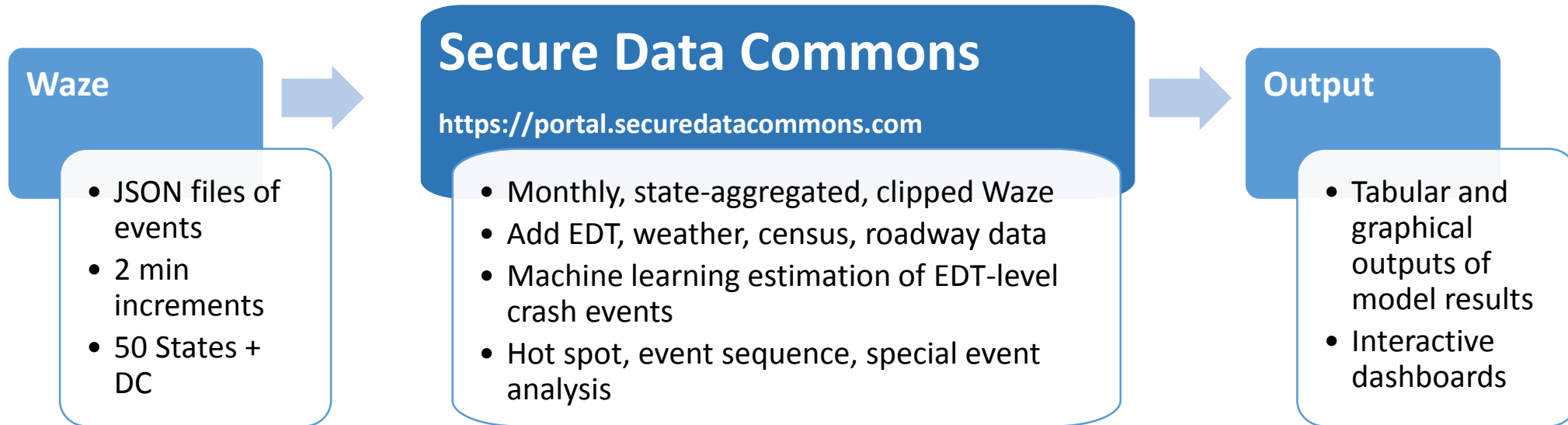
Crashes by Day

Day Of Week	ESTIMATED EDT	OBSERVED EDT	PRCT OBSERVED
Monday	1,089	1,099	99.09%
Tuesday	1,623	1,602	101.31%
Wednesday	1,788	1,709	104.62%
Thursday	1,768	1,694	104.37%
Friday	1,922	1,840	104.46%
Saturday	1,945	1,869	104.07%
Sunday	1,390	1,413	98.37%

% Observed Estimated by Hour



SDI Waze Data Pipeline Development



Technology platform

- AWS S3 buckets for curated data and team working folders
- AWS Redshift database for derived data
- RStudio + Jupyter on virtual computer
- GitHub integration for collaboration (private)



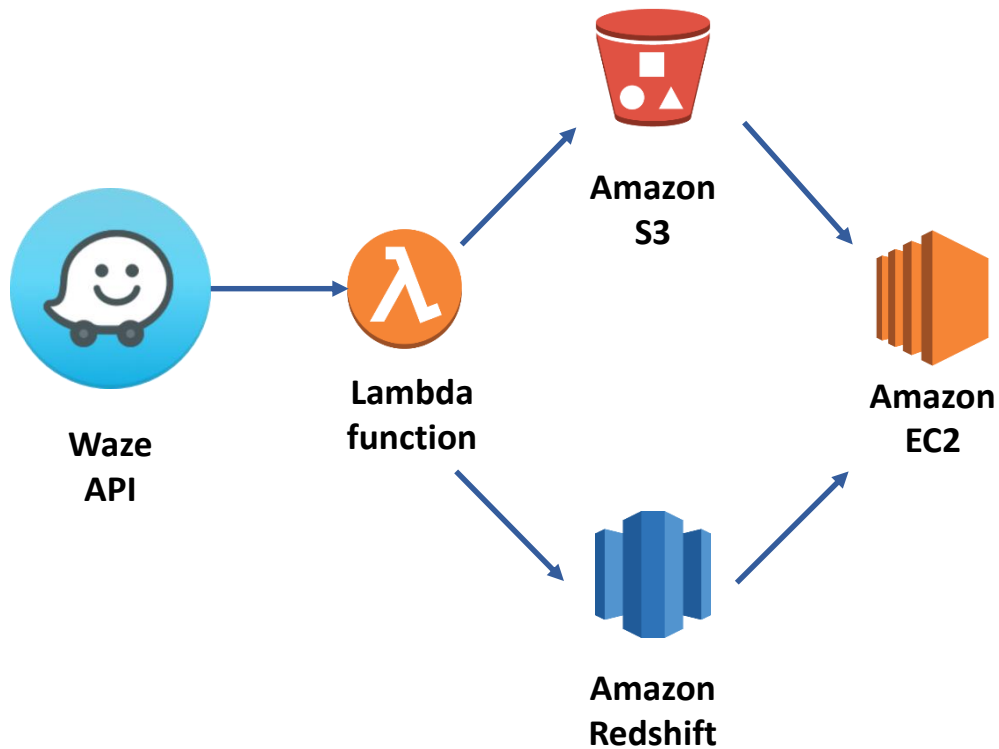
SDI Waze Data Pipeline Development

SDC

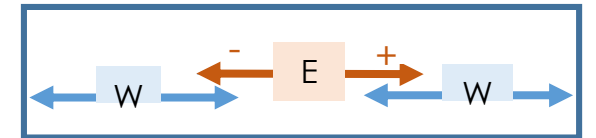
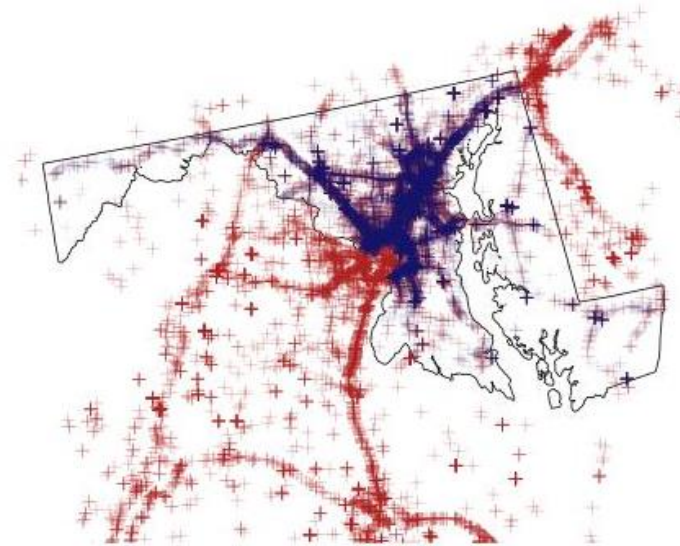
0. Waze Data Ingestion and Curation

1. Query, Clip, Reduce

2. Space-Time Match



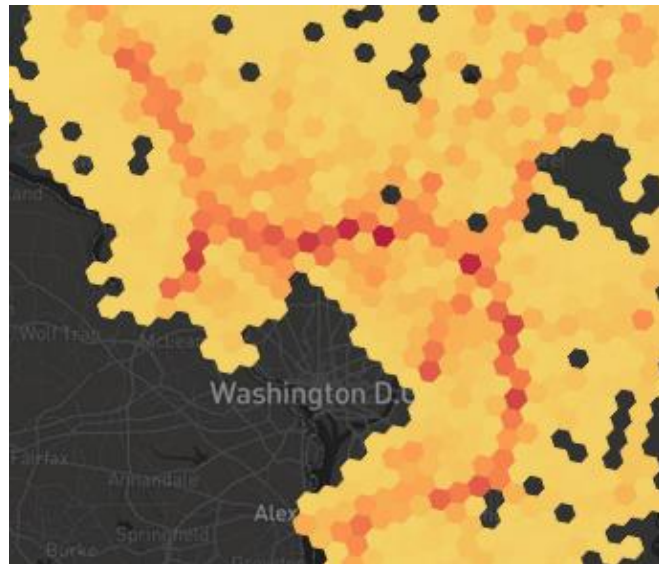
Plotting original and clipped MD



SDI Waze Data Pipeline Development

SDC

3. Grid and Urban Area Overlay

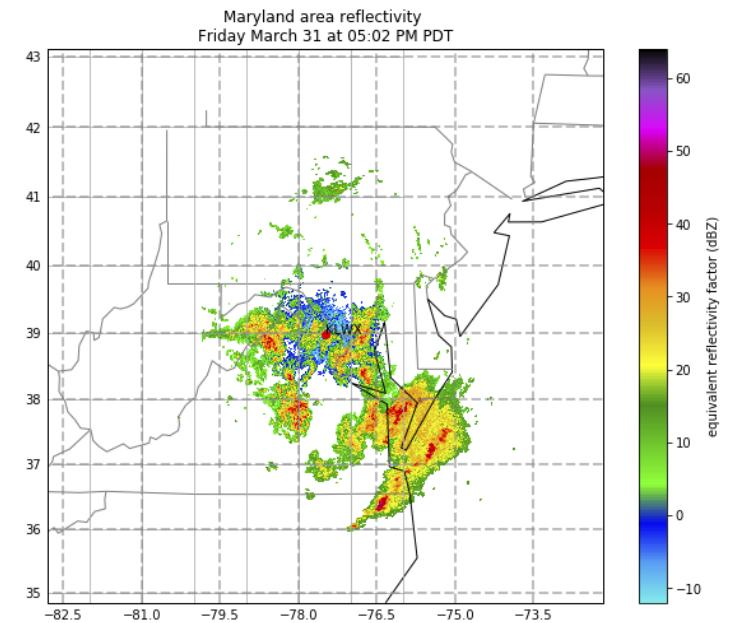


Adding:

- Urban Areas
- Hexagonal grid tessellations

4. Grid Aggregation

5. Weather Overlay



Adding:

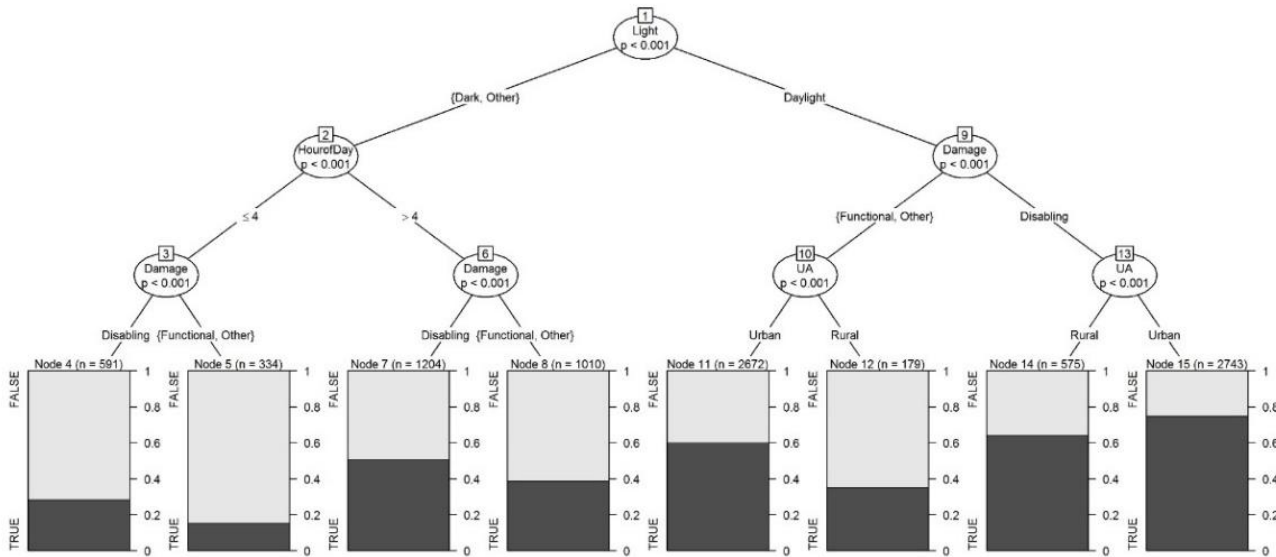
- Raster weather reflectivity

SDI Waze Data Pipeline Development

ATA

ATA + Local

6. Modeling



Adding:

- FARS
- HPMS road class
- AADT
- LEHD

7. Visualization and Reporting



Statistical Approach: Supervised Classification

Random Forests

- Machine learning approach which minimizes overfitting
- Trained models on 70% of data using EDT reports as our labeled “ground-truth”
- Tested model performance using 30% of data to compare estimated EDT crashes with observed EDT crashes
- Rigorously trained and tested data feature combinations (50+ models)
- Best crash estimation models minimize False Positives and False Negatives

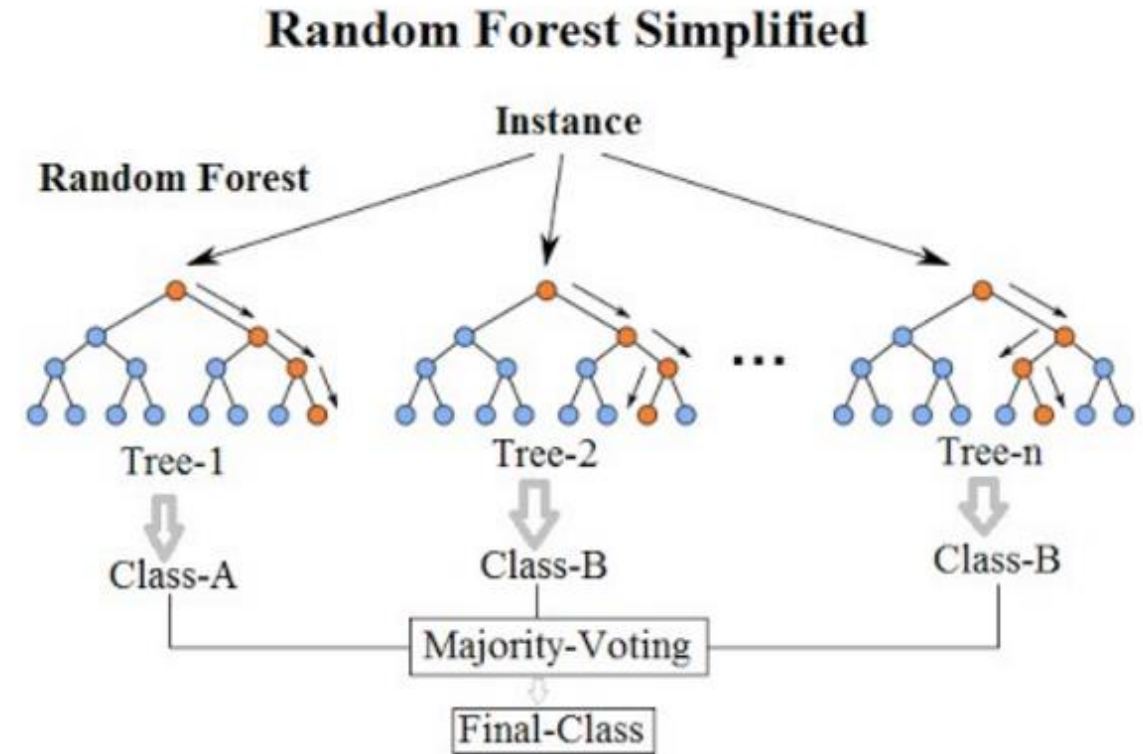


Image credit: <https://medium.com/@williamkoehrsen/random-forest-simple-explanation-377895a60d2d>

Results – what have we learned?

We can integrate DOT data resources at large scales

- Our data integration and analysis pipeline can support rapid crash estimates (when/where Waze signal present)
- Successfully integrated transportation data that are not originally intended to track traffic safety

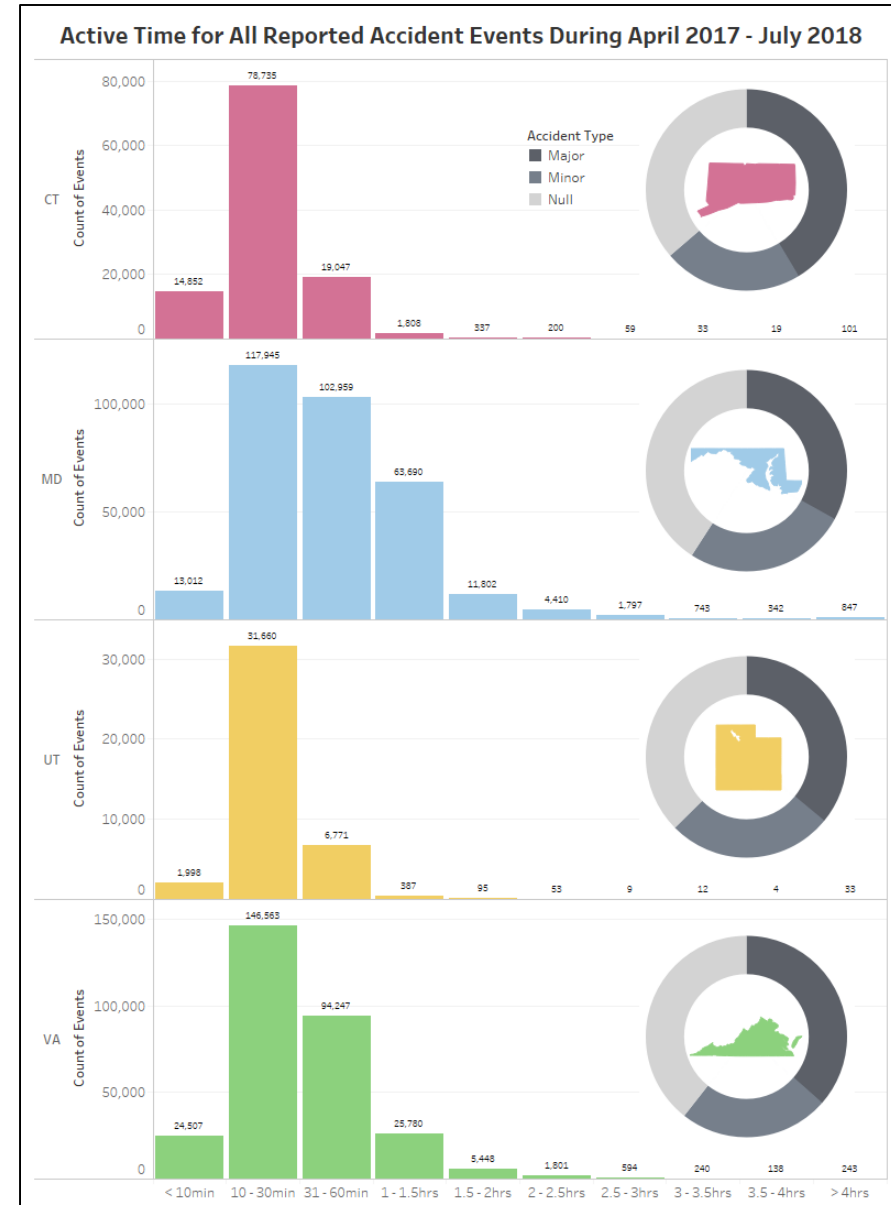
Waze data support rapid crash indicator

- With Waze signal, models produce good overall estimates for multiple states
- Foundation for tool for rapid tracking of traffic safety trajectories



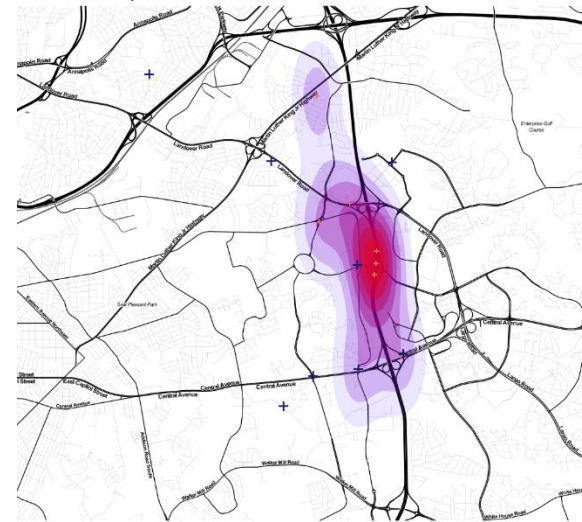
Results – what have we learned?

- Potential for Waze data to support analysis of roadway incident clearance times
- Sequence of event analysis shows potential for crash precursor early warning
- Waze data can evaluate impact of special events using heat maps
- Beginning partnerships with state agencies to deliver usable tool

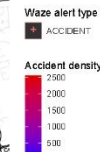
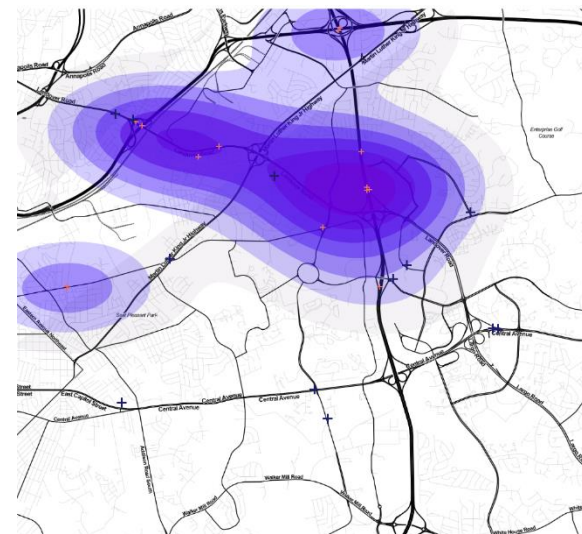


Results – what have we learned?

- Potential for Waze data to support analysis of roadway incident clearance times
- Sequence of event analysis shows potential for crash precursor early warning
- Waze data can evaluate impact of special events using heat maps
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Special event



No special event

Next Steps

- Full year modeling on multiple states
- Partnerships with state or local DOTs to identify use cases
- Cross-state Waze data assessment & dashboard
- Applications of segment-based models

Potential Applications

Rapid crash trend monitoring tool

- Flag anomalies
- Short-term intervention assessment
- Cross-state comparisons
- Effectiveness models

- Incident Duration
- Clearance Times
- Secondary Crashes

Additional Slides

Evaluating Model Performance

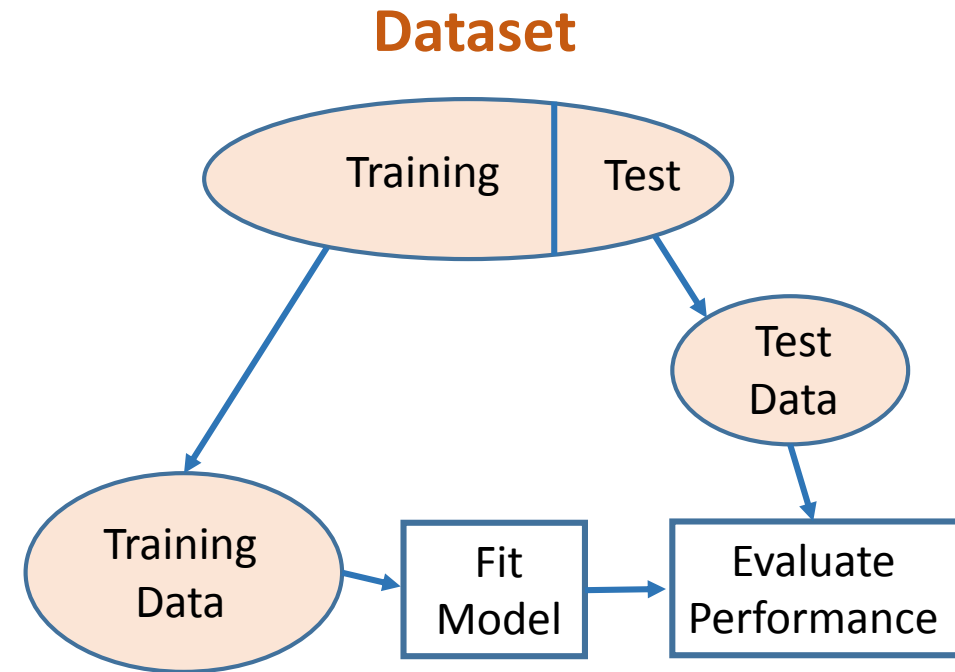
Divide data into training and testing subsets

- Training data: Select **70%** of observations (random by rows, whole days, or whole weeks)
- Test data: Remaining **30%** of observations

Training: fit model parameters with a large set of known EDT crashes, associated Waze events and other predictors

Testing: apply fitted model parameters to a new set of Waze events and other predictors to generate estimated EDT crashes

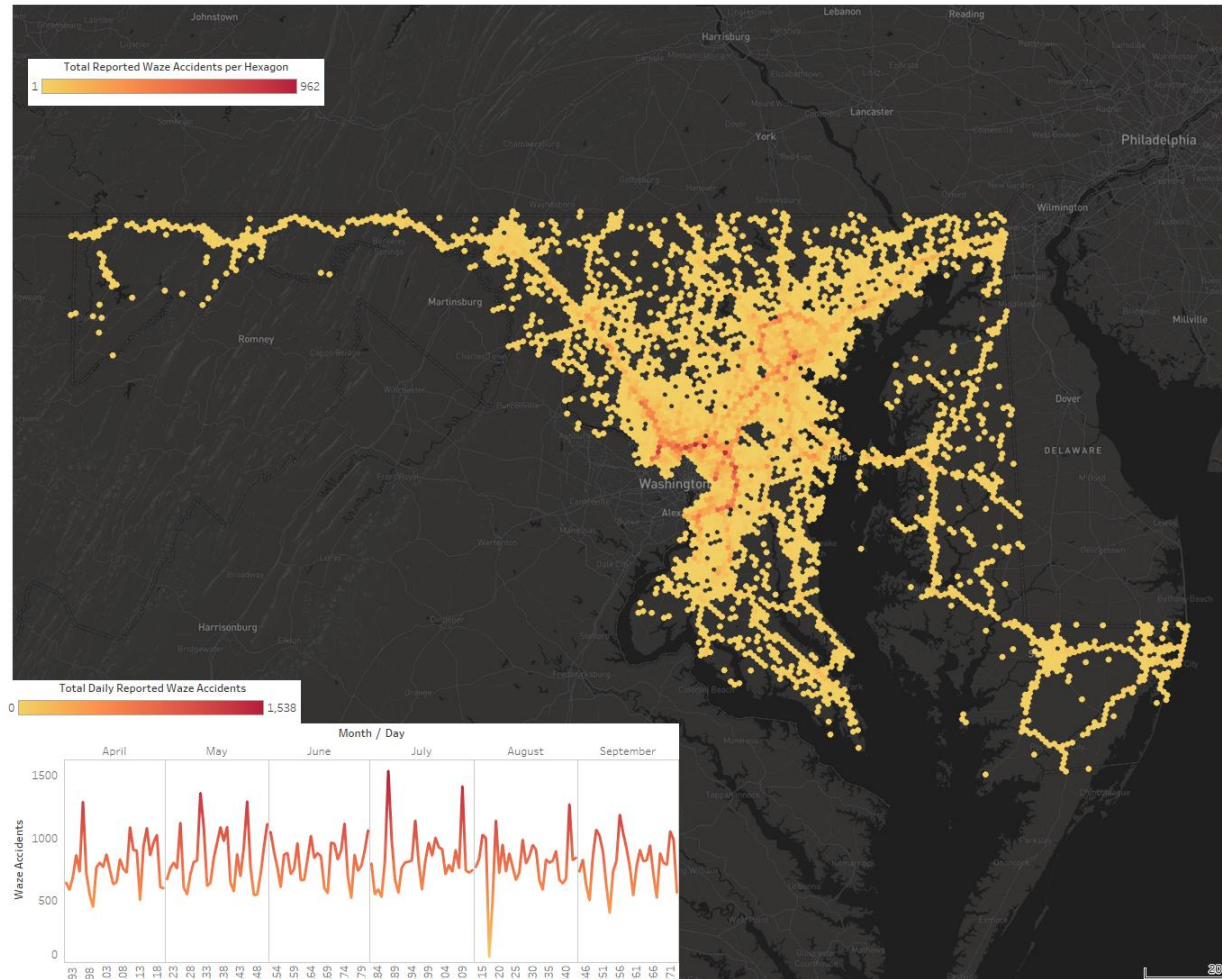
Compare estimated EDT crashes to observed EDT crashes in the test data set to evaluate model performance



Waze Data: Distribution in Space and Time

Six months of geolocated Waze data for Maryland (April - September, 2017)

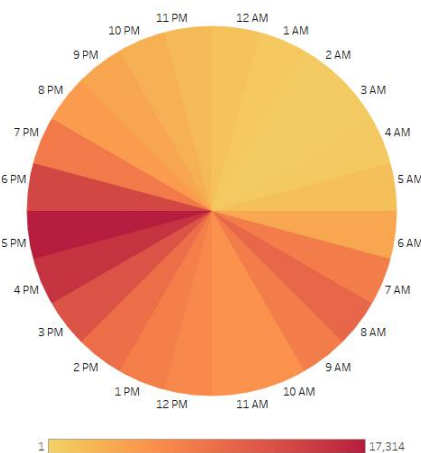
Spatiotemporal Distribution of Reported Waze Accidents in Maryland



Waze Accidents by Day of Week

Day Of Week	Major Accidents	Minor Accidents	Total Accidents
Monday	4,973	8,902	20,212
Tuesday	5,282	10,221	22,180
Wednesday	5,487	10,279	22,412
Thursday	6,128	11,376	25,372
Friday	6,144	10,957	24,960
Saturday	5,274	6,658	18,215
Sunday	4,290	5,207	15,084

Waze Accidents by Hour



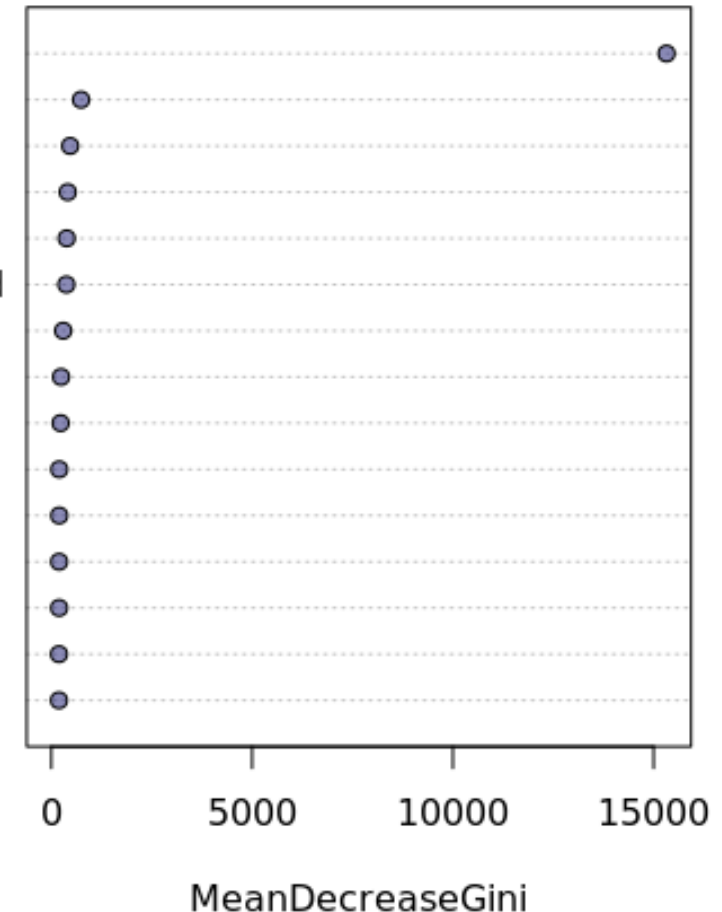
Variable Importance: Waze Accidents (April-Sept)

Mean decrease in Gini impurity:

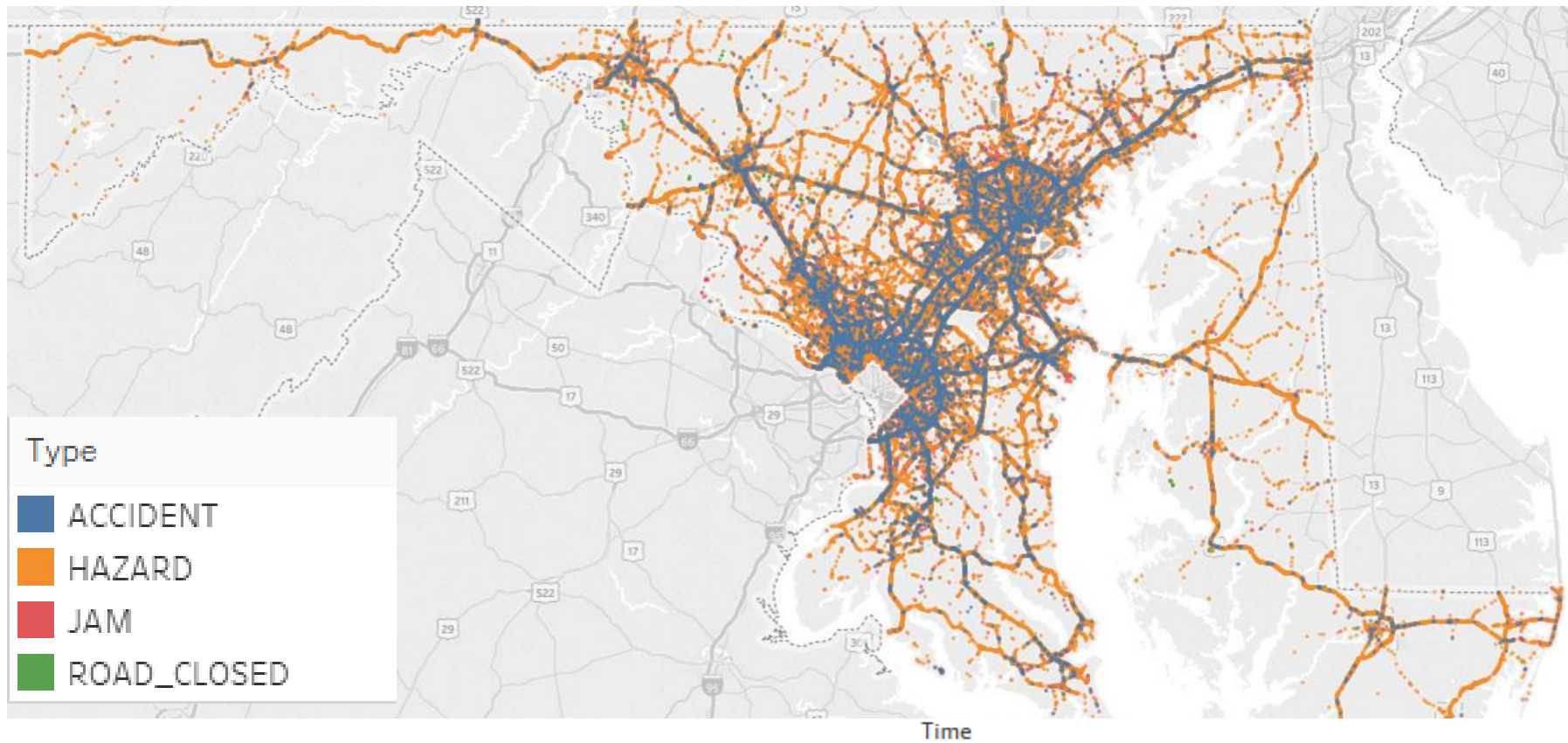
- Variable is useful in separating a node of mixed classes (both 0 and 1 EDT crashes, in our case) into two nodes with pure classes (all 0 or all 1 EDT crashes).
- Across all nodes in all the trees, how much does this variable decrease node impurities, averaged over all trees?

Model 30 Variable Importance

nWazeAccident
nWazeJam
medLastRepRate
nWazeRT6
nWazeRT3
nWazeWeatherOrHazard
medLastConf
medMagVar
MEAN_AADT
DayOfWeek
nWazeRT7
SUM_AADT
medLastReliab
SUM_miles
nMagVar240to360



Waze Data: Jams and Crash Sequence Analysis



April, 2017 MD WazeUniqueCounts

Type	Count
ACCIDENT	15,139
HAZARD	242,787
JAM	180,347
ROAD_CLOSED	1,130

Potential Applications

- Incident Duration
- Clearance Times
- Secondary Crashes

