

Volume & Turning Movements Project

Steering Committee Meeting #5

July 27, 2017



Housekeeping Items

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- This is a **virtual meeting experience**
 - Please keep your phone muted until asking a question or speaking (press *6 to mute/unmute individual phone lines)
 - Please do not place call “on hold” as your hold music will be heard by the group
- Speakers will answer questions at the end of their presentation
- The audio from this meeting is being recorded
- All materials & contact information will be available to participants after the webcast

WELCOME!

Denise Markow, I-95 Corridor Coalition



**Welcome to
Volume & Turning
Movements Project**



Speakers



Denise Markow, PE
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I-95 Corridor Coalition
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Stanley Young, PhD, PE
National Renewable Energy Laboratory (NREL)
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Kaveh Sadabadi, PhD
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Yi Hou, PhD
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Shawn Turner, PE
Texas A&M Transportation Institute
(TTI)
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Attendees

Agencies	
Colorado DOT	HERE
District DOT	INRIX
FHWA	TomTom
Maryland DOT/MdSHA	Texas A&M Transportation Institute
MWCOG	NREL
NJTPA	I-95 Corridor Coalition
Pennsylvania DOT	UMD CATT
South Carolina DOT	UMD CATT Lab
Virginia DOT	



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Thank you!



Agenda

	Topic	Speaker
1	Welcome & Project Status Update	Denise Markow, I-95 Corridor Coalition Stan Young, NREL
2	Spotlight Presentation: Real-time Volume Estimation with TomTom Probe Data – Denver Area	Yi Hou, NREL
3	Polling Questions for Steering Committee Input	Steering Committee members
4	Validation Framework	Shawn Turner, TTI
5	Next Steps & Wrap Up	Stan Young, NREL Denise Markow, I-95 Corridor Coalition



Project – Year in Review

- User Survey - Completed
- Preliminary Data Analyses
 - Maryland
 - Rhode Island
 - Florida
 - Colorado
- Calibration (FHWA TMAS)
- Validation (TTI)



Getting Things Done.



Project Goal

- Accelerate the timeframe to a viable volume and turning movement data feed ---
 - Anywhere/anytime on the network
 - Archive and real-time
- Ensure that initial data products meet the I-95 Corridor Coalition members' information needs for operations, performance measurement, and planning.

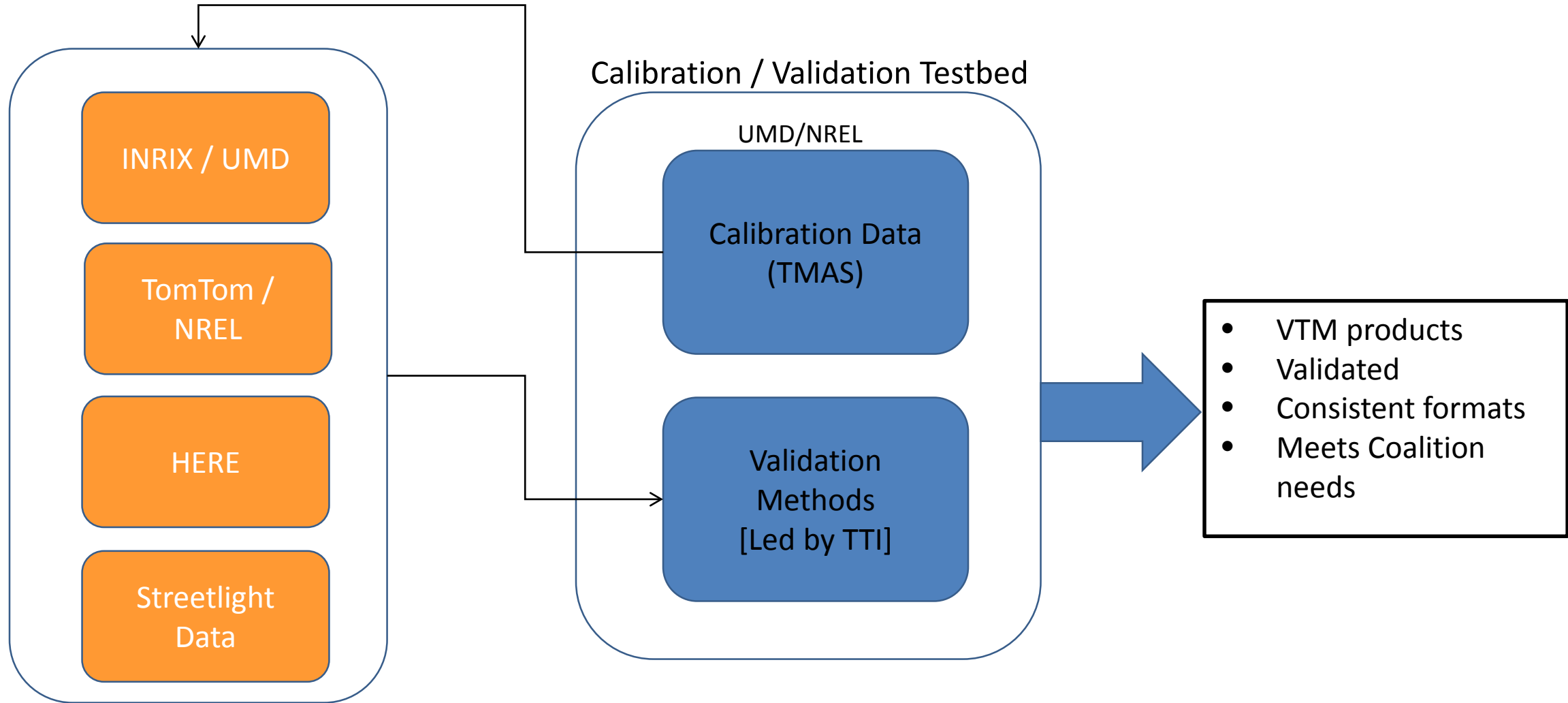


Objectives - Original

- Define a *practical and logistical framework* for the delivery of probe-based volume and turning movement data
- Understand, document, and share *data requirement needs for a variety of DOT applications* requiring such data
- Create a *calibration and validation testbed* to assist vendors' initial development efforts
- Provide *representative data products and set appropriate expectations* for data fidelity, form, granularity, and usability
- Anticipating the need for an ongoing calibration network, estimate *resources needed to maintain/operate* a national calibration/validation testbed



Status of Project



Status

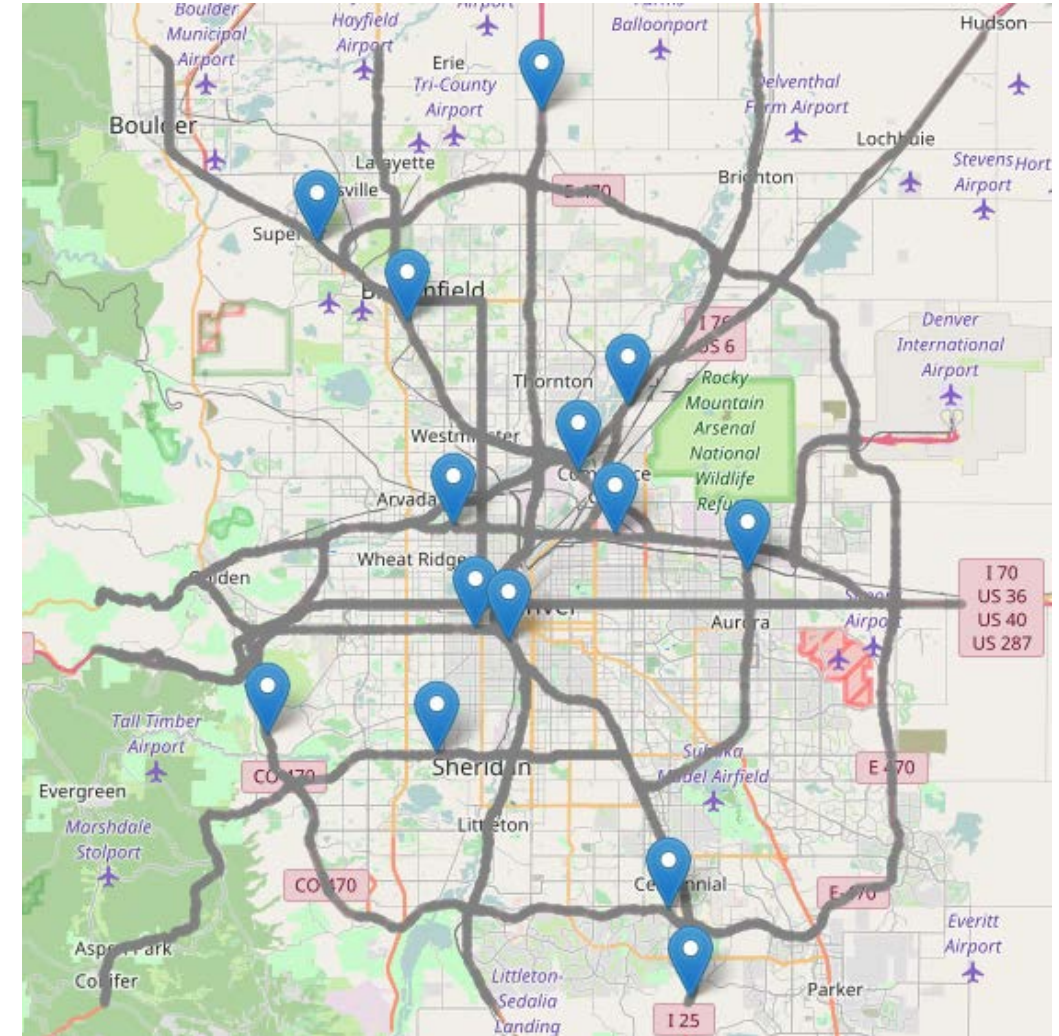
- INRIX/UMD Data – Maryland analysis complete
 - Rhode Island and Florida are the target areas for next analysis
 - Base data: Trips, Origin & Destination, and intervening waypoints
- TomTom/NREL – Archive analysis complete Colorado (presented today)
 - In discussion with TomTom and CDOT for real-time analysis
 - Base data: Probe vehicle counts on segments
- HERE – sample data sets being reviewed
 - Base data: Trips, Origin & Destination
- Streetlight Data – sample data sets being reviewed
 - Base data: AADT estimate for a roadway

**ALL DATA SETS
ARE
DIFFERENT!!**



Today's Agenda

- TomTom/NREL analysis of volume estimates in CO
 - Yi Hou, NREL
 - Parallels the approach used in initial Maryland analysis
- Validation overview/thoughts –
 - Shawn Turner, TTI



Questions





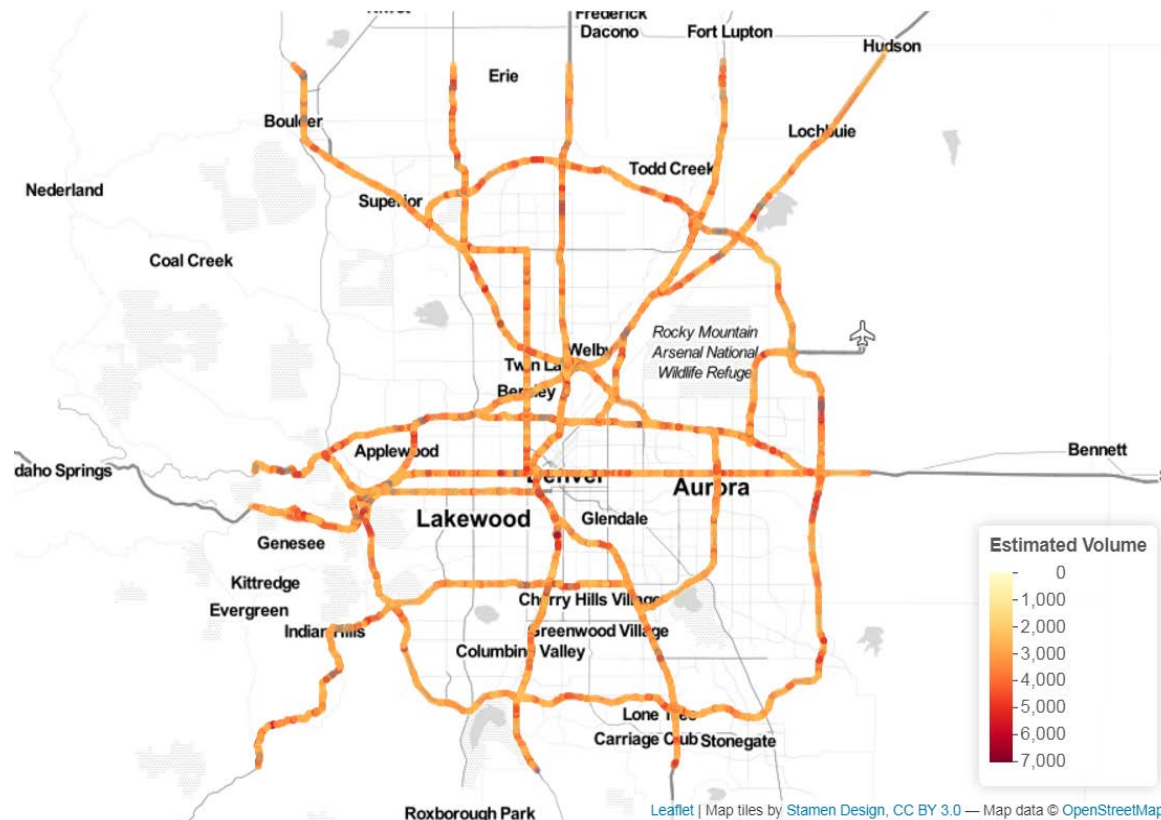
Real-time Volume Estimation with TomTom Probe Data – Denver Area

Yi Hou

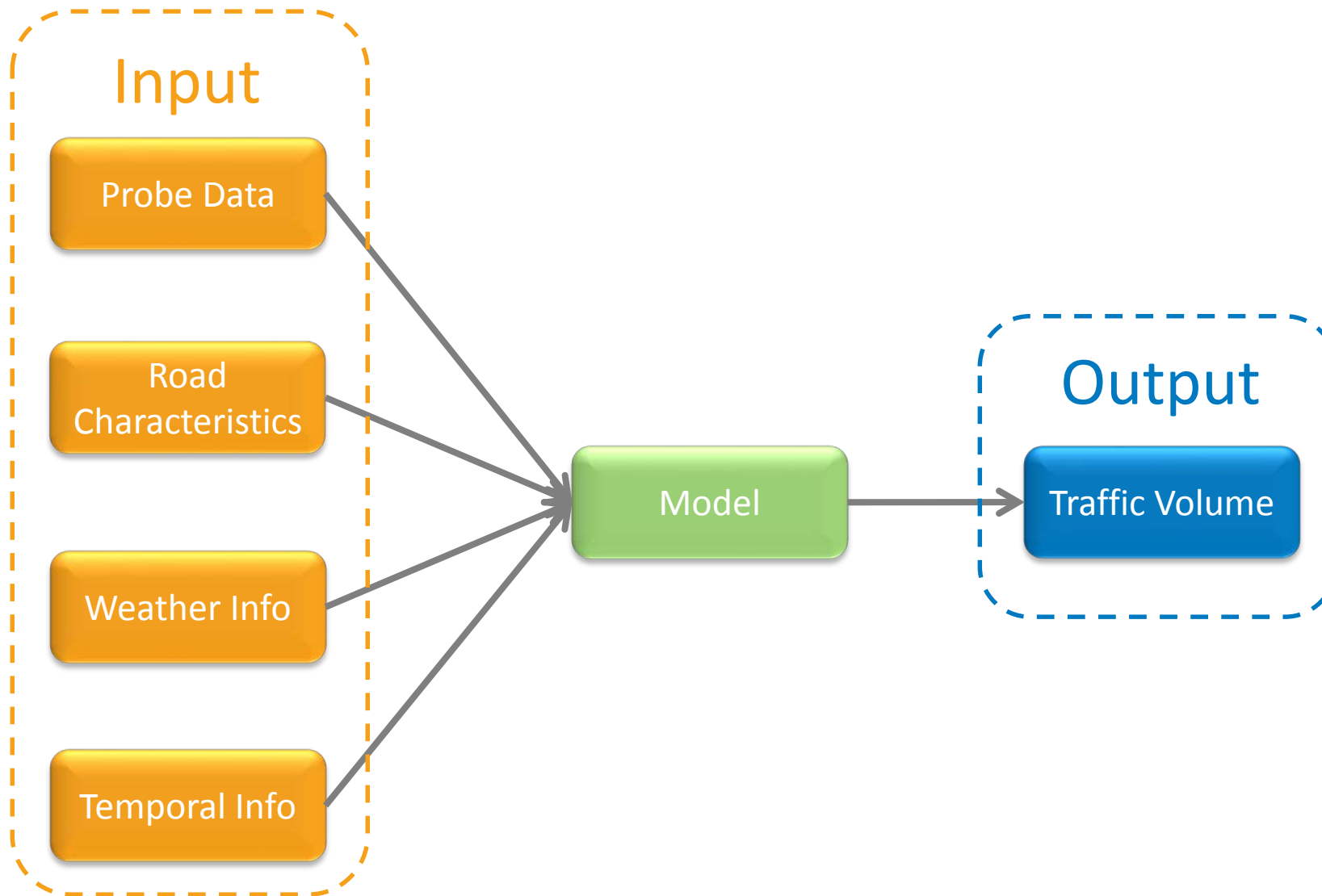
July 27, 2017

Problem Statement

- How to obtain accurate estimate of ubiquitous real-time traffic volume in road networks?



Proposed Solution



- Data Sources
 - FHWA TMAS
 - CDOT ATR Counts
 - Weather
 - TomTom Probe Data
- Model Development
- Summary
- Web App Framework

Data Sources

- FHWA TMAS
- CDOT ATR Counts
- Weather
- TomTom Probe Data



- ATR station information and road characteristics
 - State
 - Station ID
 - Direction
 - Urban/Rural
 - Number of lanes
 - Road sign (interstate, US, state)
 - Road number
 - Latitude
 - Longitude

Data Sources

- FHWA TMAS
- CDOT ATR Counts
- Weather
- TomTom Probe Data



- Temporal and volume information
 - Year
 - Month
 - Day
 - Weekday
 - Hour of day
 - Hourly volume

Data Sources

- FHWA TMAS
- CDOT ATR Counts
- Weather
- TomTom Probe Data



- Weather information
 - Average temperature
 - Visibility
 - Wind
 - Precipitation
 - Snow
 - Fog
 - Rain
 - Thunderstorm

Data Sources

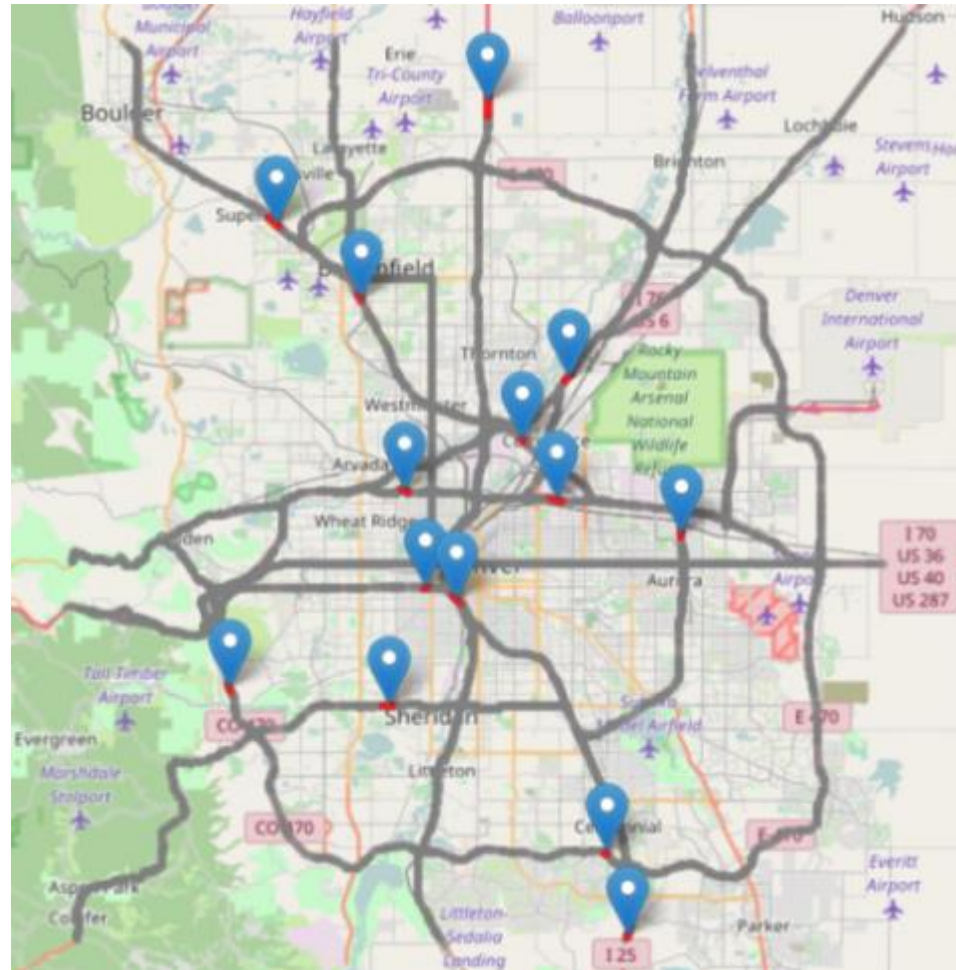
- FHWA TMAS
- CDOT ATR Counts
- Weather
- TomTom Probe Data



- Probe vehicle data
 - Travel time
 - Speed
 - Probe vehicle count
 - Speed limit
 - Street name
 - Segment ID

Denver Road Network

- Match ATR locations with TomTom segments



Snapshot of Merged data

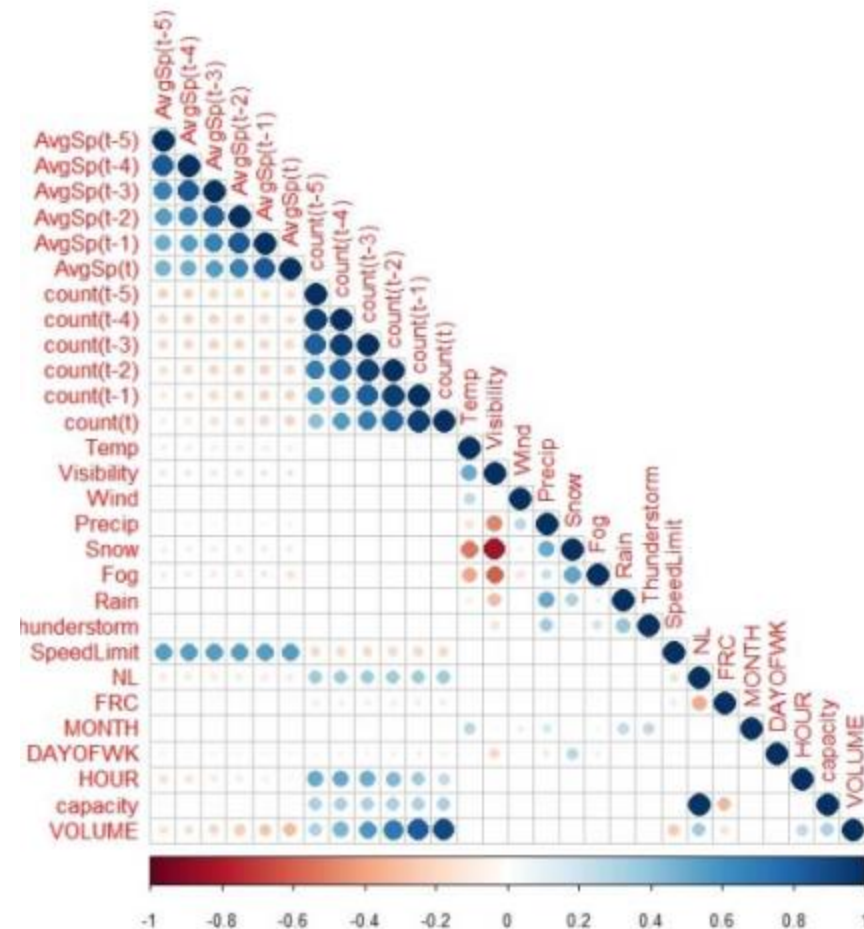
Input Output

Weather Probe Road Info Temporal

Segment.Id	STATIONID	DIRECTION	street_name	date	Temp	Visibility	Wind	Precip	Snow	Fog	Rain	Thunderstorm	AvgSp	count	SpeedLimit	NE	FRC	RS	RN	MONTH	DAYOFWK	HOUR	VOLUME
8.84e+13	107	E	I-70 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	71.67	112	55	3	0	I	70	1	3	16	1726
8.84e+13	107	W	I-70 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	66.32	203	65	3	0	I	70	1	3	16	1856
-8.84e+13	501	N	I-25 N	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	16.76	484	55	5	0	I	25	1	3	16	3492
8.84e+13	501	S	I-25 S	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	49.53	958	55	4	0	I	25	1	3	16	6647
8.84e+13	503	E	US-6 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	59.78	499	55	4	1	US	6	1	3	16	4581
8.84e+13	503	W	US-6 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	45.74	556	55	4	1	US	6	1	3	16	5834
8.84e+13	504	E	US-36 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	62.54	439	65	2	2	US	36	1	3	16	4645
8.84e+13	504	W	US-36 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	66.95	321	65	2	2	US	36	1	3	16	2941
-8.84e+13	506	N	US-285 N	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	62.89	189	65	2	2	US	285	1	3	16	2262
-8.84e+13	506	S	US-285 S	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	54.87	230	65	2	2	US	285	1	3	16	3549
-8.84e+13	507	E	I-270 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	34.00	332	55	2	0	I	270	1	3	16	3211
8.84e+13	507	W	I-270 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	56.13	324	55	2	0	I	270	1	3	16	3336
-8.84e+13	510	E	I-70 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	61.79	391	55	3	0	I	70	1	3	16	3211
8.84e+13	510	W	I-70 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	58.29	389	55	3	0	I	70	1	3	16	4720
8.84e+13	511	E	I-70 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	52.74	536	55	3	0	I	70	1	3	16	3479
8.84e+13	511	W	I-70 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	12.04	249	55	3	0	I	70	1	3	16	2609
-8.84e+13	512	E	CO-470 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	66.06	350	65	3	1	CO	470	1	3	16	4085
-8.84e+13	512	W	CO-470 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	67.46	490	65	3	1	CO	470	1	3	16	3696
8.84e+13	103387	E	I-76 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	64.50	253	65	3	0	I	76	1	3	16	3572
-8.84e+13	103387	W	I-76 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	65.04	246	65	3	0	I	76	1	3	16	2835
8.84e+13	105029	N	I-225 N	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	48.70	530	55	3	0	I	225	1	3	16	5271
-8.84e+13	105029	S	I-225 S	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	50.92	487	55	3	0	I	225	1	3	16	4193
-8.84e+13	105548	E	CO-470 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	65.02	521	65	3	1	CO	470	1	3	16	4123
-8.84e+13	105548	W	CO-470 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	34.24	457	65	4	1	CO	470	1	3	16	4051
8.84e+13	2	E	I-70 E	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	76.94	57	75	2	0	I	70	1	3	17	1210
-8.84e+13	2	W	I-70 W	2017-01-10	39	10	14	0	FALSE	FALSE	FALSE	FALSE	74.48	64	75	2	0	I	70	1	3	17	583

Variable Correlations

- Positively correlated with volume
 - Average hourly probe counts of past 6 hours
 - Number of lanes
 - Capacity
- Negatively correlated with volume
 - Average hourly average speed of past 6 hours
 - Speed limit

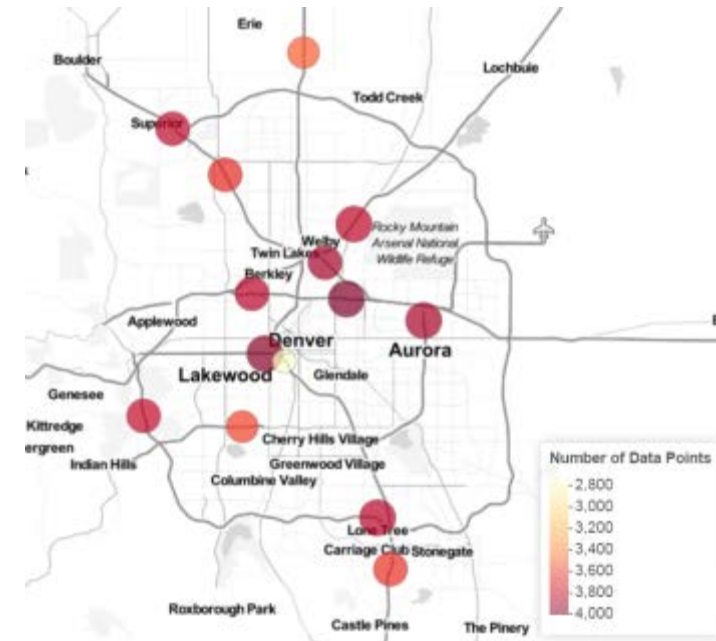
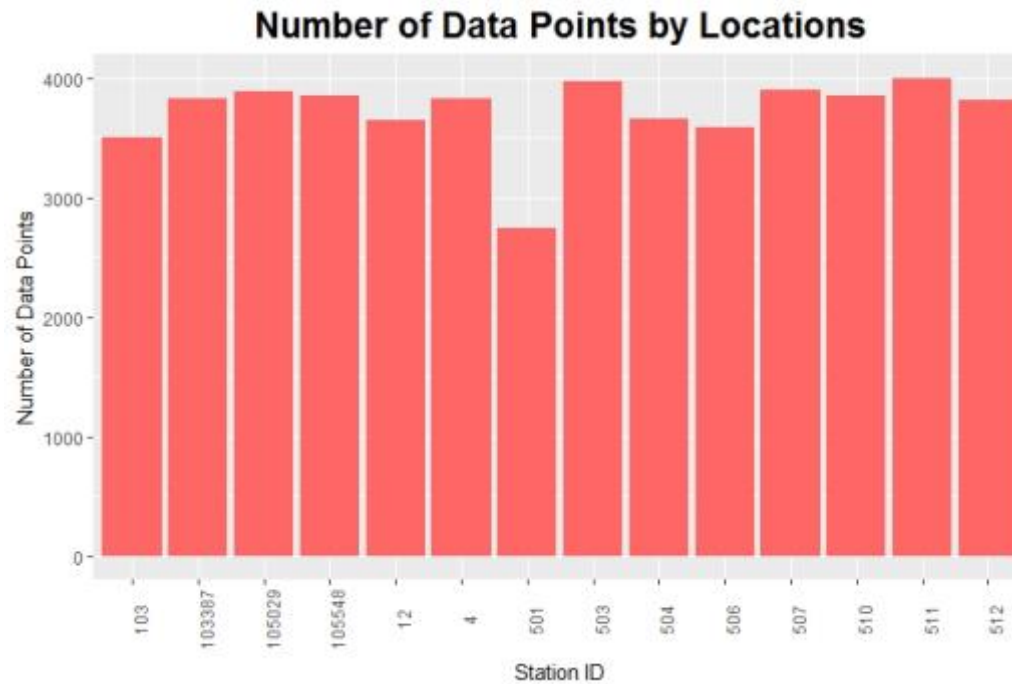


Input Variables

- TomTom GPS data
 - Average speed: average hourly speed of past 6 hours
 - Probe count: average hourly probe counts of past 6 hours
- Weather
 - temperature, visibility
- Road characteristics
 - road type, speed limit, number of lanes, capacity
- Temporal information
 - month, day of week, hour of day

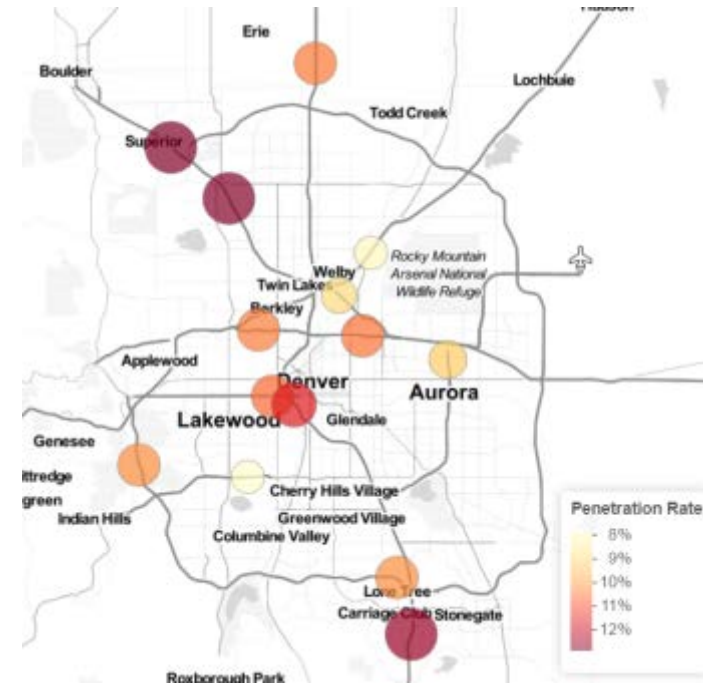
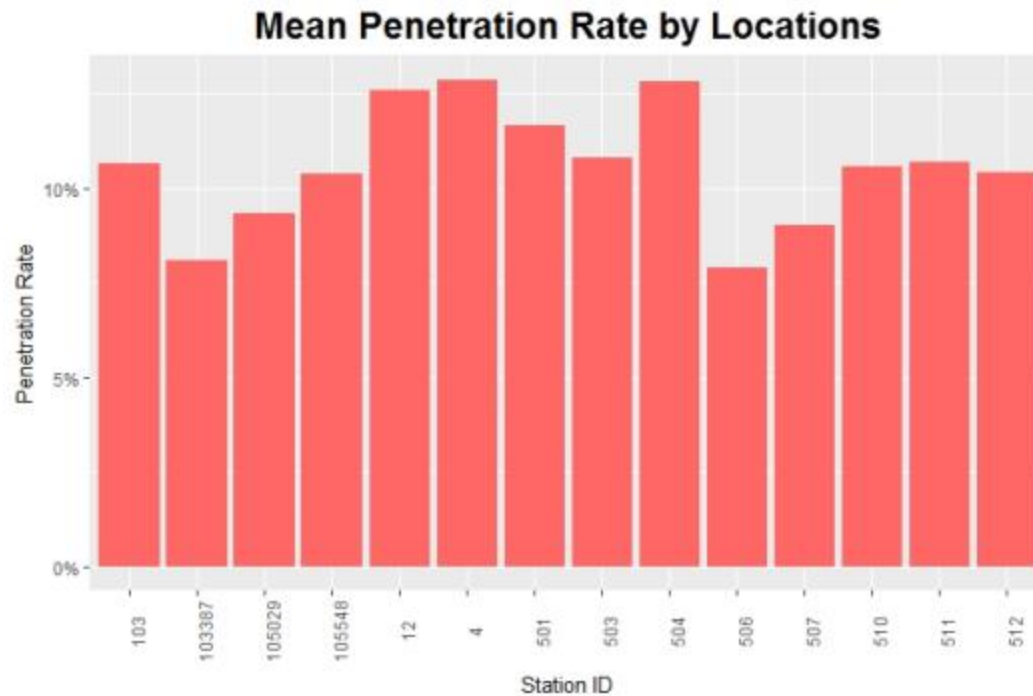
Data Points

- Feb 1, 2017 – April 30, 2017
- Total of 52,092 data points
- Ranges from 2800-4000



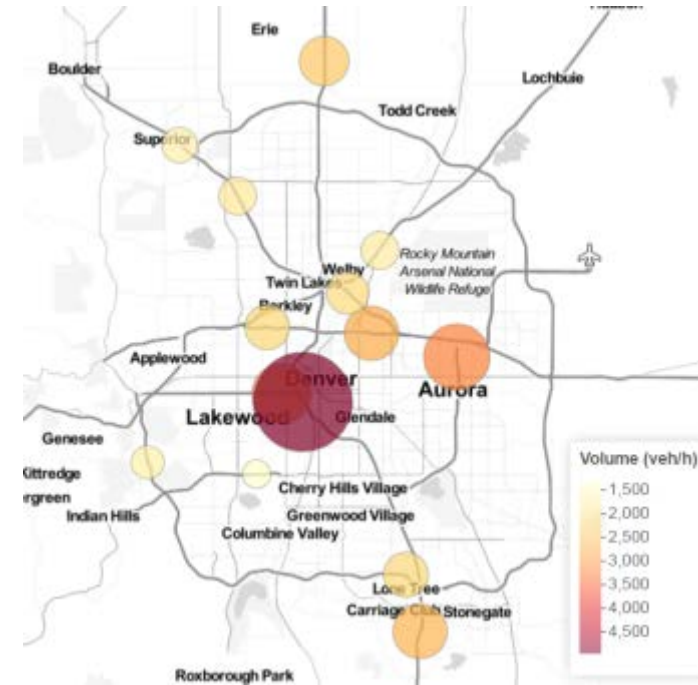
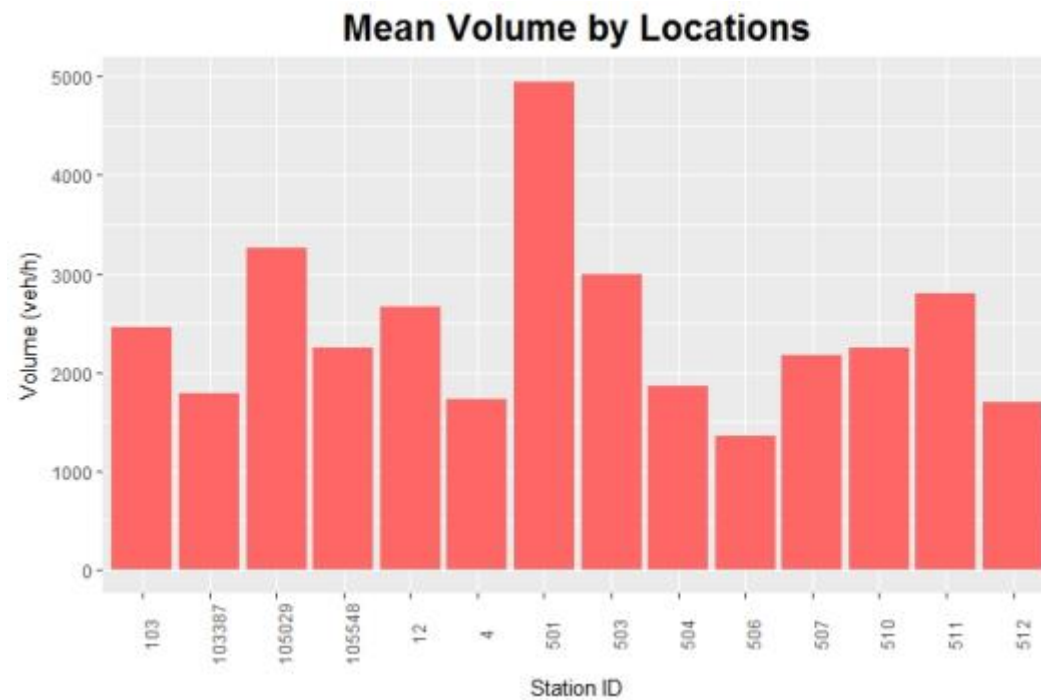
Penetration Rates

- Ranges from 8%-12%



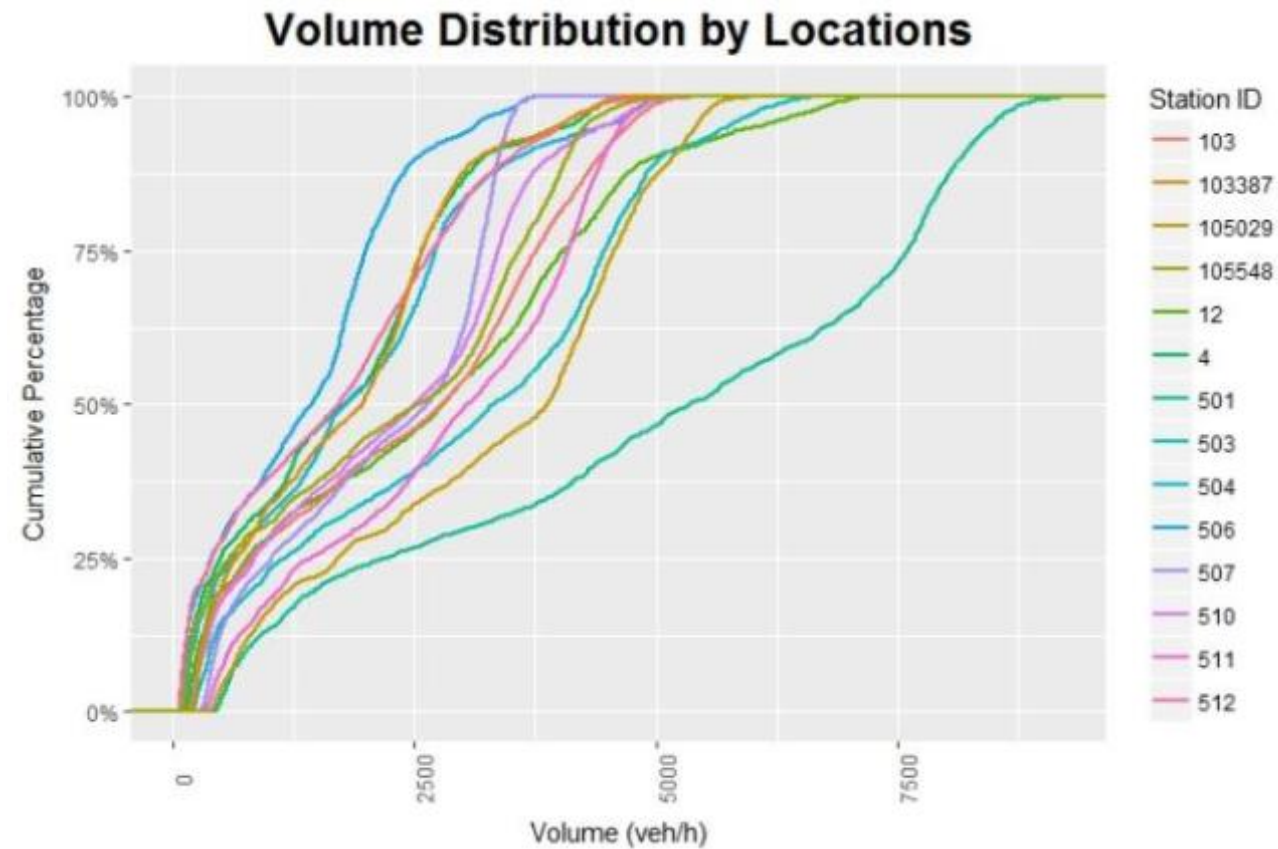
Average Volume

- All Ranges from 1500-3500 veh/hr except station 501
- 501 is an unique case



Volume Distribution by Location

- All ATR stations have similar volume distribution except station 501



- Machine Learning Models
 - Random Forest (RF)
 - Gradient Boost Machine (GBM)
 - Extreme Gradient Boost (XGBoost)
- Advantages of these models
 - Do not require detailed mathematical forms and assumptions on variable distributions
 - Suitable for capturing the underlying relationships among different variables in an environment of uncertainty
 - Fewer parameters to tune and easy to implement

- All 14 ATR stations were used for modeling
 - At each iteration
 - 13 stations are used for training
 - 1 station is used for validation
 - Repeat this 14 times and report validation results for all 14 locations

Model Evaluation Criteria

- Coefficient of Determination: $R^2 = 1 - \frac{(\widehat{V}_i - V_i)^2}{(V_i - \bar{V})^2}$
 - A measure of how well the model can explain data variance
- Mean Absolute Percentage Error: $MAPE = \frac{1}{N} \sum_{i=1}^N \frac{|V_i - \widehat{V}_i|}{V_i}$
 - An error measure when comparing estimated volume with ground truth
- Error to Theoretical Capacity Ratio: $ETCR = \frac{1}{N} \sum_{i=1}^N \frac{|V_i - \widehat{V}_i|}{C_i}$
 - An error measure when comparing estimated volume with roadway capacity

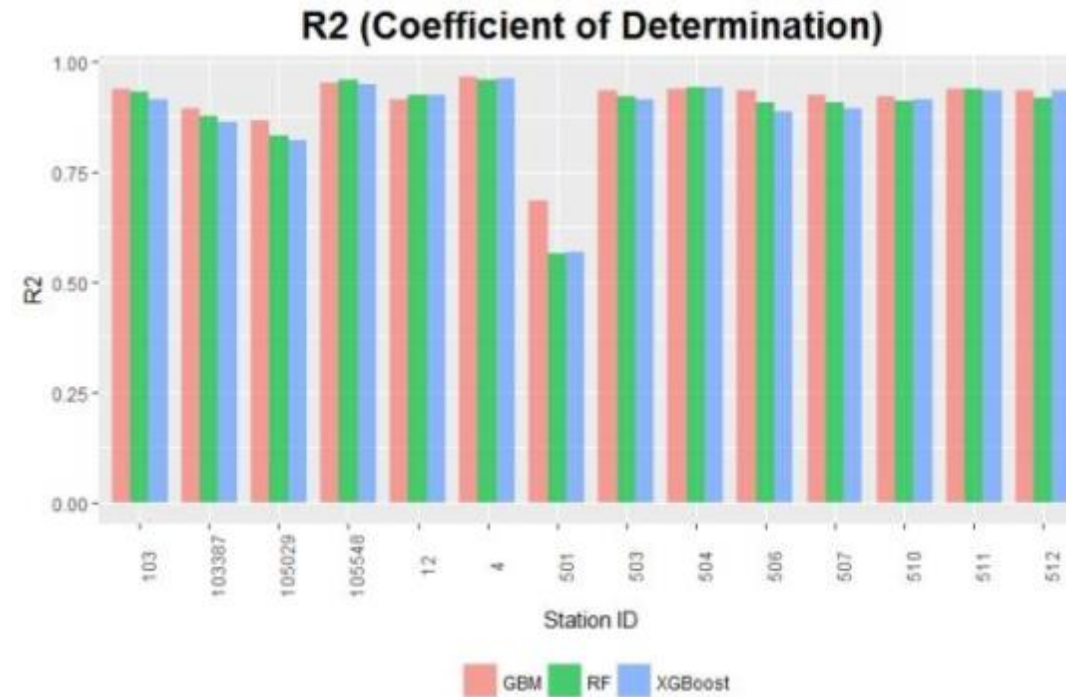
Model Results

- Results exceed the survey expectation: ETCR<10%
- When compared with linear regression:
 - Reduced MAPE by almost 60%
 - Reduced ETCR by almost 30%
 - Increased median R^2 by more than 10%

Model	Overall MAPE	Overall ETCR	Median R2
RF	17.8%	5.2%	0.92
GBM	18.3%	4.8%	0.93
XGBoost	17.7%	5.3%	0.91
Linear Regression	41.8%	7.2%	0.83

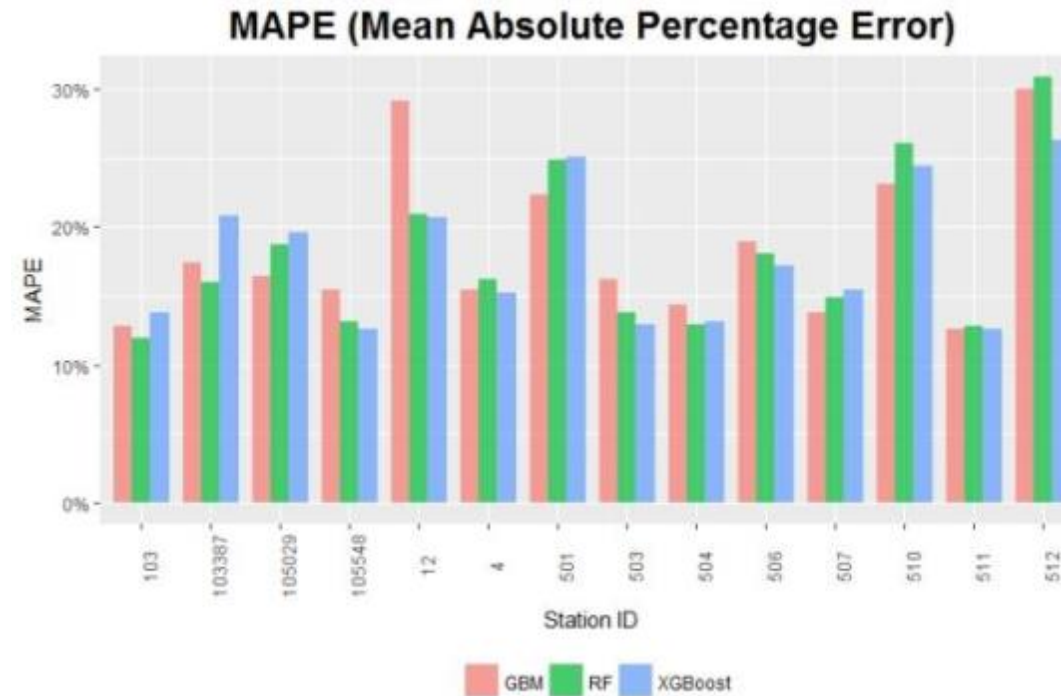
Coefficient of Determination (R^2)

- RF
 - Range: 0.57-0.96
 - Median: **0.92**
- GBM
 - Range: 0.69-0.96
 - Median: **0.93**
- XGBoost
 - Range: 0.57-0.96
 - Median: **0.91**



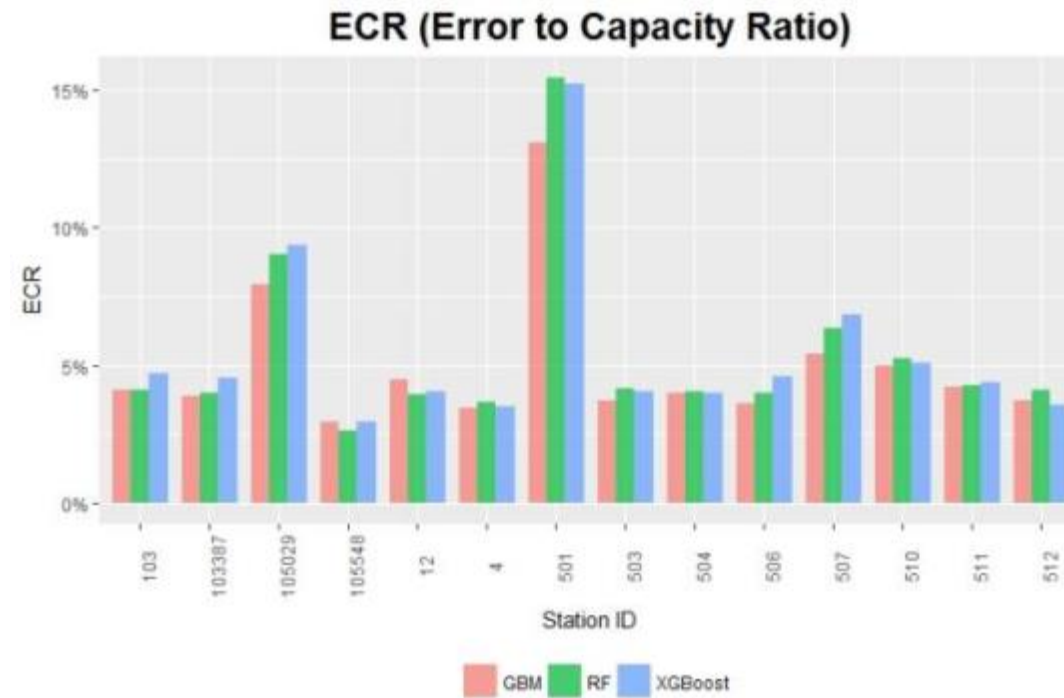
Mean Absolute Percentage Error (MAPE)

- RF
 - Range: 12%-31%
 - Median: **16.1%**
- GBM
 - Range: 13%-29%
 - Median: **16.3%**
- XGBoost
 - Range: 13%-25%
 - Median: **16.3%**



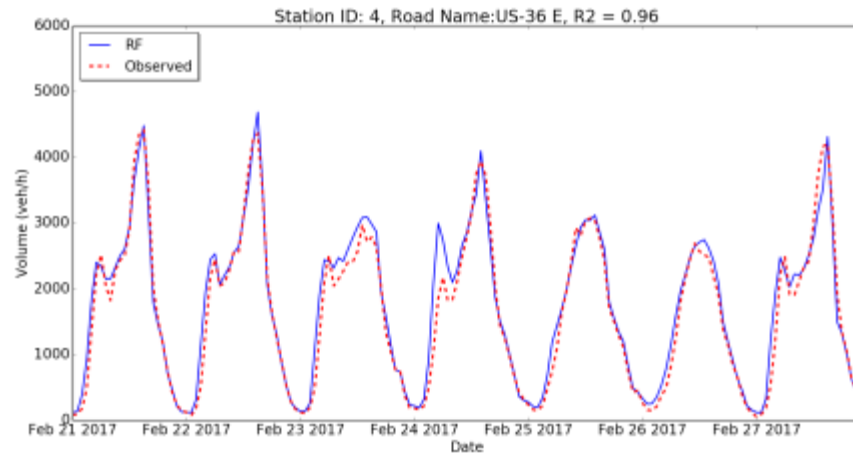
Error to Theoretical Capacity Ratio (ETCR)

- RF
 - Range: 2.6%-15.4%
 - Median: **4.1%**
- GBM
 - Range: 2.9%-13.0%
 - Median: **4.0%**
- XGBoost
 - Range: 3.0%-15.2%
 - Median: **4.5%**

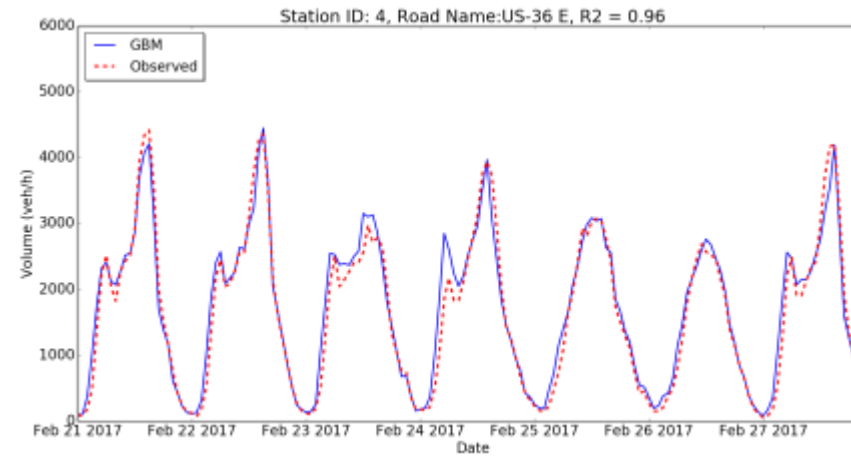


Prediction vs. Observation (Highest R^2)

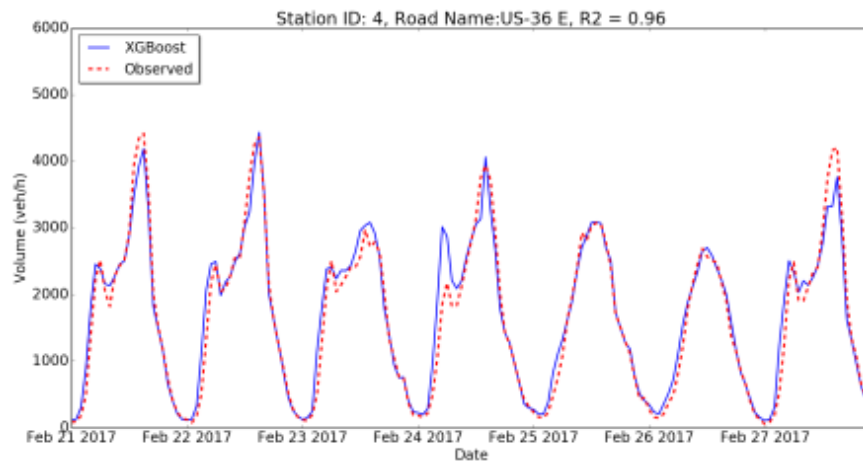
RF



GBM

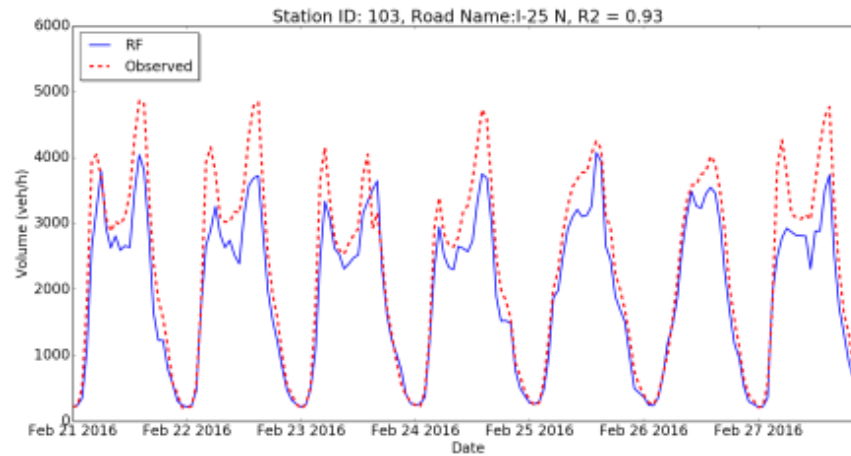


XGBoost

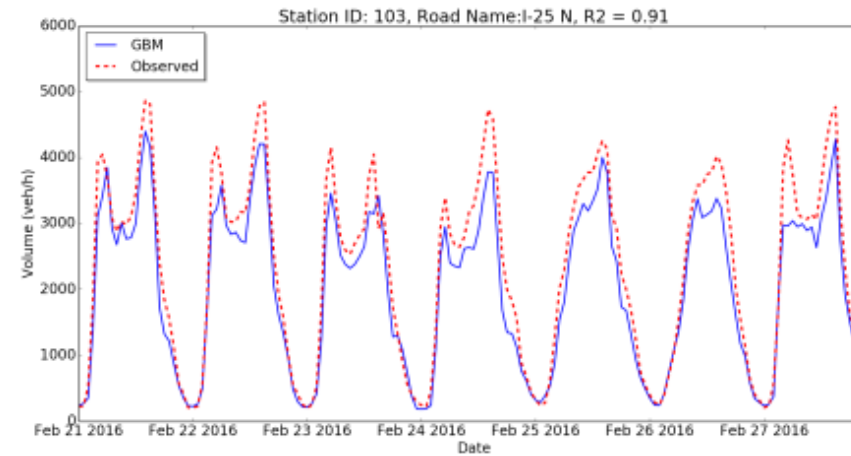


Prediction vs. Observation (Median R^2)

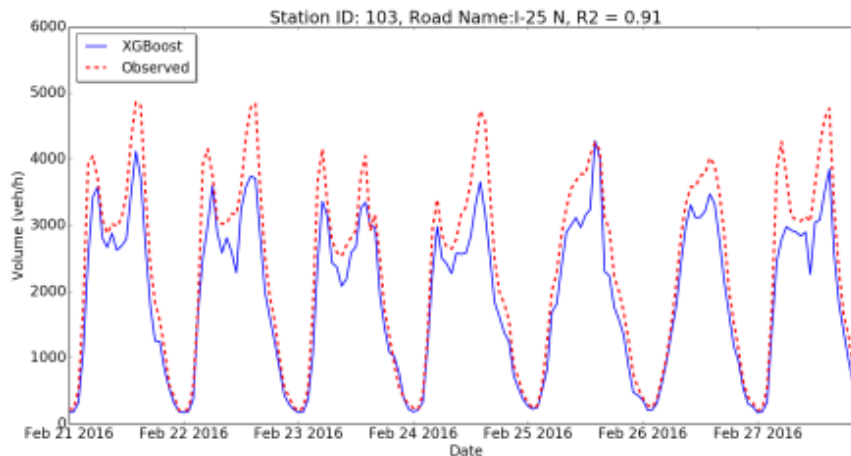
RF



GBM

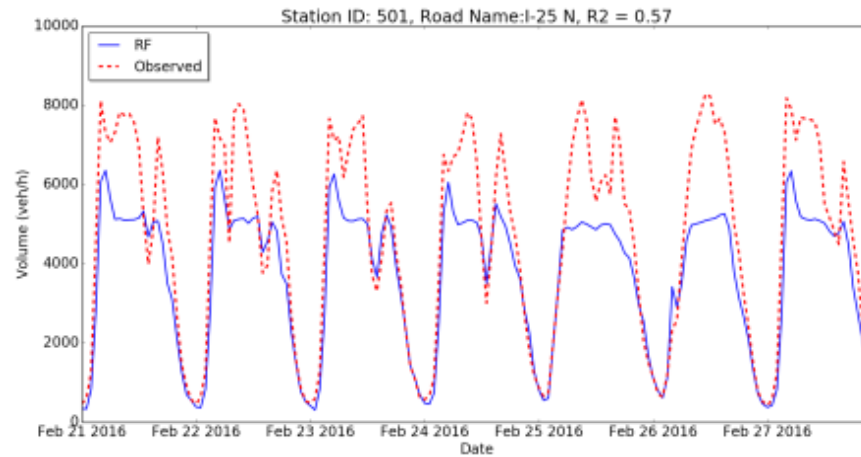


XGBoost

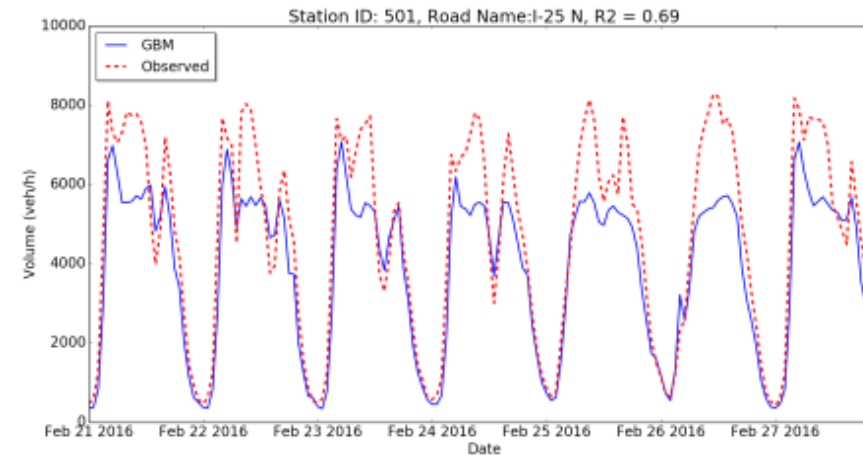


Prediction vs. Observation (Lowest R^2)

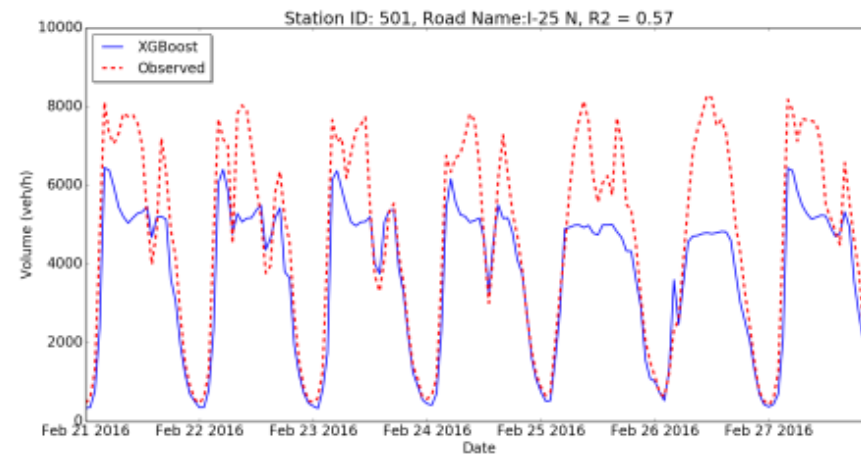
RF



GBM



XGBoost



Contribution of Probe Vehicle Data

- Probe vehicle data has significant impact on volume estimation accuracy

RF

	Overall MAPE	Overall ETCR	Median R ²
All data included	17.8%	5.2%	0.92
With only Probe Data	25.5%	6.4%	0.90
Without Probe Data	35.2%	10.7%	0.70

GBM

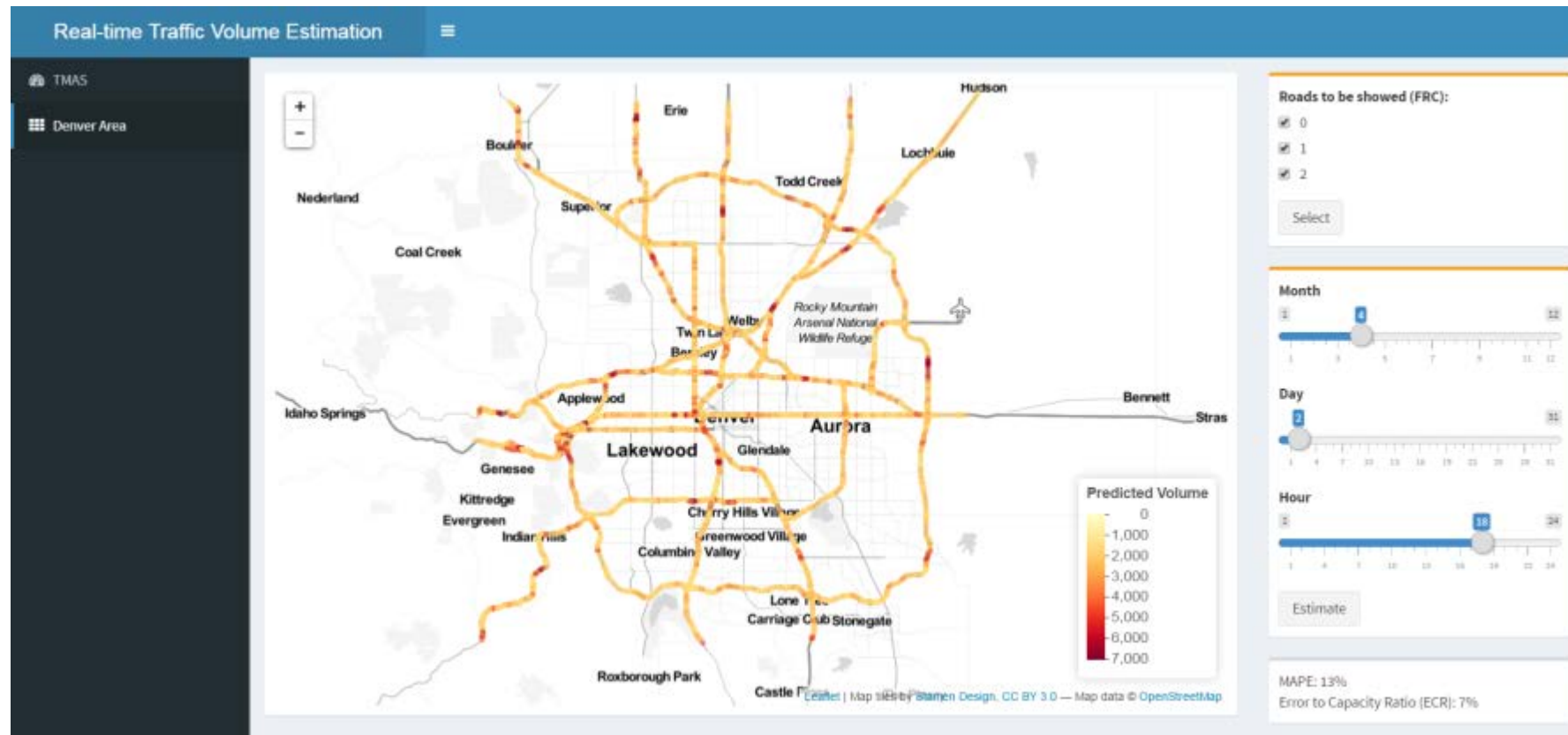
	Overall MAPE	Overall ETCR	Median R ²
All data included	18.3%	4.8%	0.93
With only Probe Data	25.7%	6.4%	0.89
Without Probe Data	40.5%	12.1%	0.64

XGBoost

	Overall MAPE	Overall ETCR	Median R ²
All data included	17.7%	5.3%	0.91
With only Probe Data	24.2%	6.4%	0.89
Without Probe Data	39.4%	12.4%	0.65

- Results satisfy the survey expectation: $ETCR < 10\%$
- Improved MAPE by almost 60% when compared with linear regression
- Probe vehicle data has significant impact on volume estimation accuracy
- Has poor performance on unique cases or outlier
- Accuracy can be further improved with more cases or data
- Next step
 - Will scale up to state level with more data
 - Further improve accuracy by trying deep learning

Web App Framework Prototype



Thank You!

Questions?

www.nrel.gov



Poll

1 Having heard the descriptions of the error measures below, which is the most meaningful to you? (select all that apply)

- ☐ R^2 – Coefficient of Determination
- ☐ MAPE – Mean Absolute Percentage Error
- ☐ ETCR - Error to Theoretical Capacity Ratio, where error is calculated with respect to maximum theoretical/observed capacity
- ☐ None of the above



Poll

2 Having heard the descriptions of the error measures below, should the research team consider other error measures?

- R^2 – Coefficient of Determination
- MAPE – Mean Absolute Percentage Error
- ETCR - Error to Theoretical Capacity Ratio, where error is calculated with respect to maximum theoretical/observed capacity

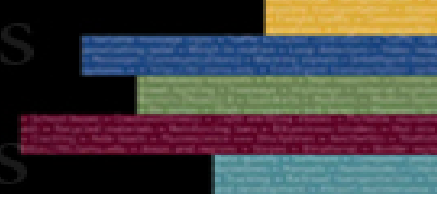
☐ Yes

☐ No



Validation Framework

- Shawn Turner, Texas A&M Transportation Institute (TTI)
 - (979) 845-8829, s-turner@tti.tamu.edu
- Why TTI involvement?
 - Provide 3rd party independent validation of groups that are producing traffic volume estimates (which now include UMD & NREL)
- Smaller-scale MnDOT/TTI research project now finishing up



Validation Framework:

Two-pronged approach - Trust, but Verify

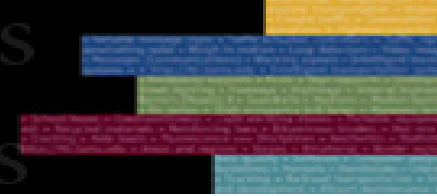
- TRUST
- Self-reported info about model calibration process and results
- High-level description of sample scaling approach (must be non-proprietary and freely distributable)
- Goodness-of-fit or other model calibration statistics (try to standardize)



- VERIFY
- 3rd party independent validation similar to I-95CC travel time validation
- Use existing DOT continuous count stations for benchmark data (that is not yet publicly available)
- Standardized accuracy measures and measure categories (bins in which error results are grouped), similar to I-95CC travel time validation

Trust: Self-Reported Information

- High-level description of sample scaling approach
 - 1-2 pages of text that explains how samples are scaled
 - Build confidence in approach to avoid “black box” perception
 - Avoid overly simplistic marketing text
 - Recognize that details are typically proprietary
- Model calibration information (if applicable)
 - Number of calibration sites
 - Duration/season of calibration data
 - Functional class and geographic representation
 - Frequency distribution of volumes used in calibration
 - Goodness-of-fit statistics (R^2 , others)



Verify: Benchmark Volume Data

- Similar to I-95CC travel time validation, but no extra collection of benchmark data
- DOTs provide most recent volume data that has not yet been made publicly available
 - Ideally, continuous count location from well-maintained site
 - Ideally, location not used in model calibration
- Need to identify DOTs who can provide benchmark volume data shortly after data has been collected (before it's made public through DOT web site or TMAS)

Verify: Accuracy Measures

- Accuracy when it matters:
 - 10% error at high volume is much different than 10% error at low volume
 - For example:
 - 10% error at 200 vphpl = 20 veh
 - 10% error at 2000 vphpl = 200 veh
- Sliding scale – higher % error acceptable at lower volumes
 - MnDOT planning example

AADT Range	Acceptable % Change	
	Decreasing (-)	Increasing (+)
300 - 999	-20%	+25%
1,000 - 4,999	-15%	+20%
5,000 – 49,999	-10%	+15%
50,000+	-10%	+10%

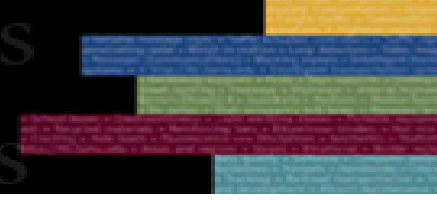
Verify: Accuracy Measures

- Standardized accuracy measures and measure categories
- Comparing notes with UMD/NREL and assessing best approach
- A possibility:

Category	Accuracy Measure #1	Accuracy Measure #2	Accuracy Measure #3
Volume categories (Low, Med, High)	Mean Absolute Percent Error, MAPE (%)	Error-to-Theoretical- Capacity Ratio, ETCR (%)	Mean Signed Error - Bias (Number of Vehicles)
or		or	
% of capacity categories (L, M, H)		Error-to-Maximum-Flow Ratio, EMFR (%)	

Next Steps

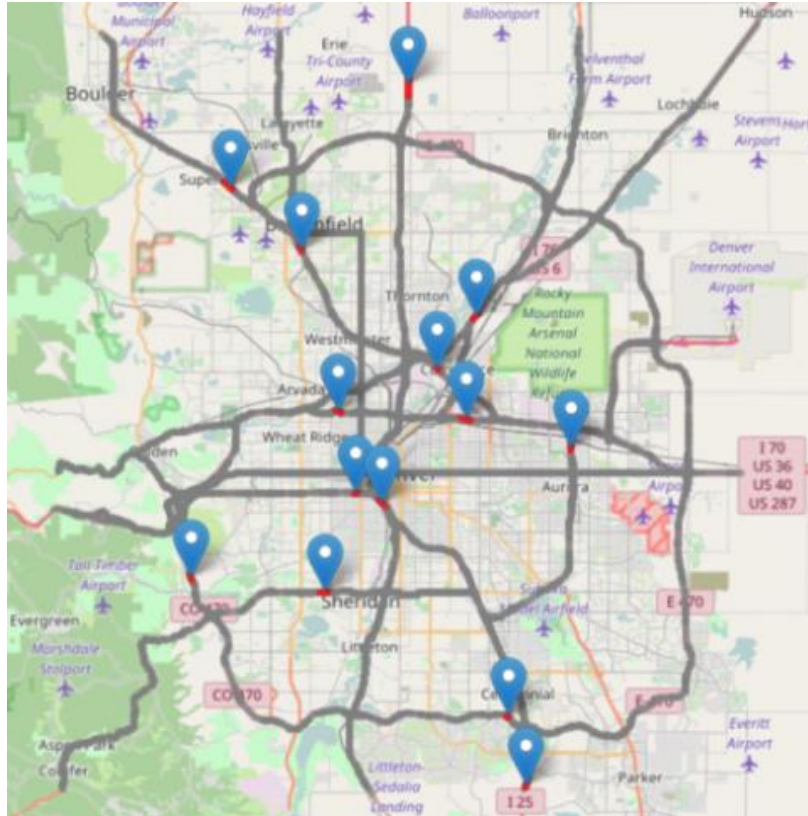
- Gather initial feedback (today)
- Identify available benchmark DOT continuous count data
 - Must be provided to TTI before being made publicly available
- Develop more detail in concept paper
- Gather feedback on concept paper from I-95CC and volume data producers in Sept/Oct
- Develop consensus on validation framework by November



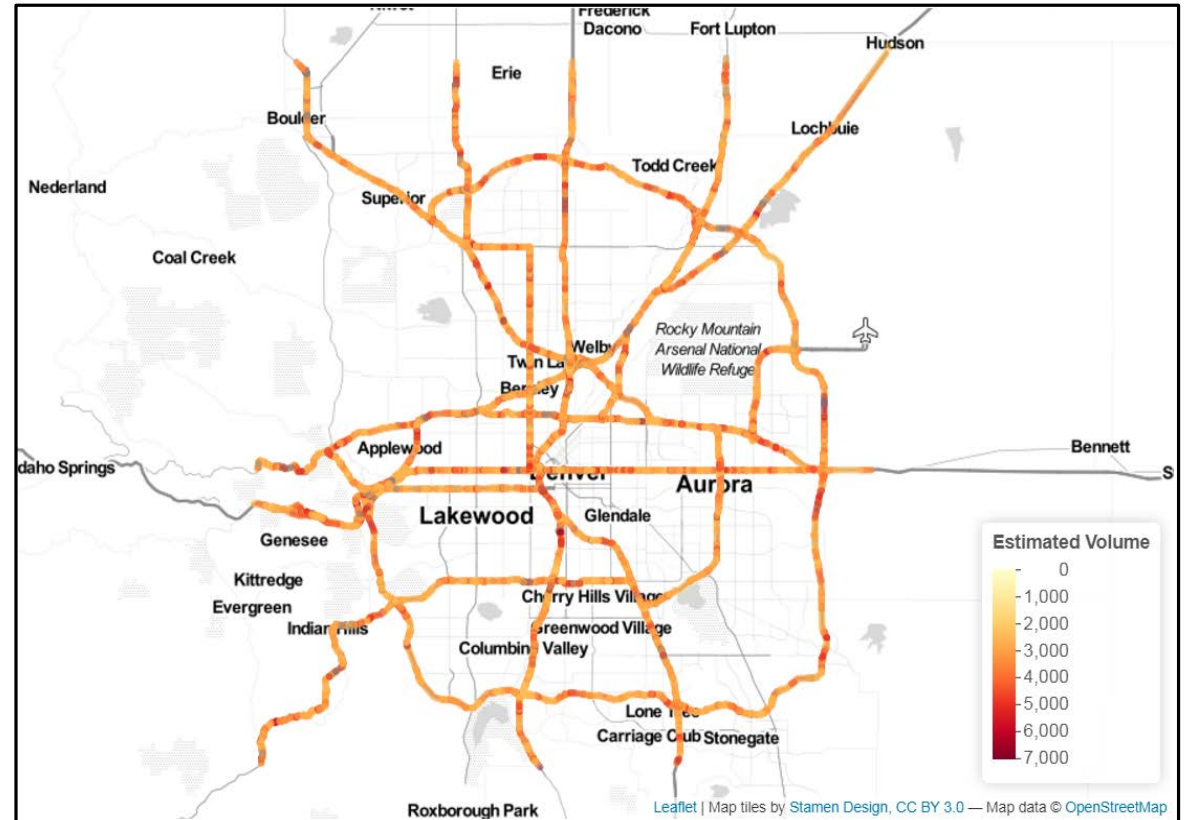
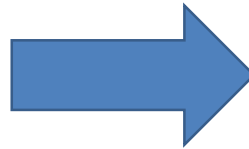
Questions



Wrapping Up



Limited locations, hourly, latent



All locations, all time, historic and real-time



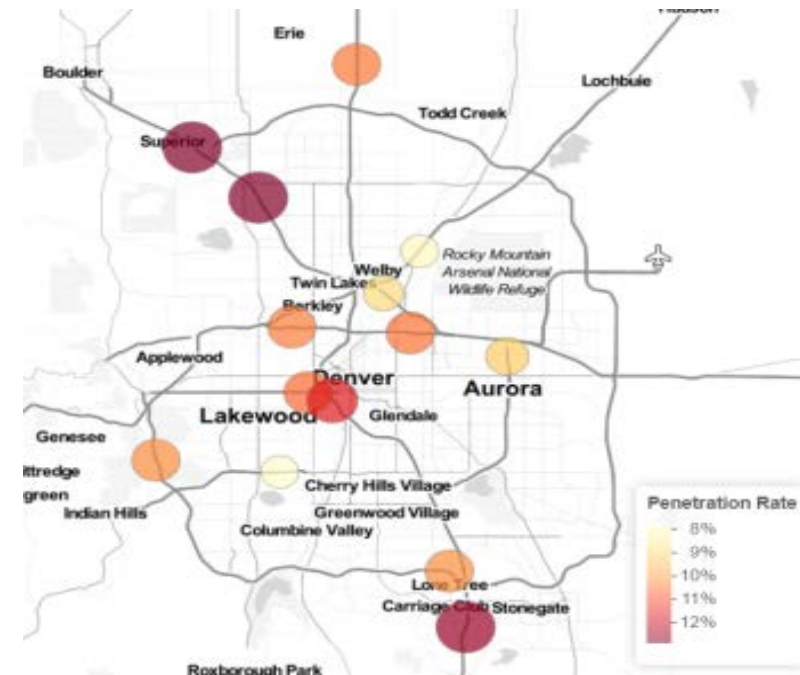
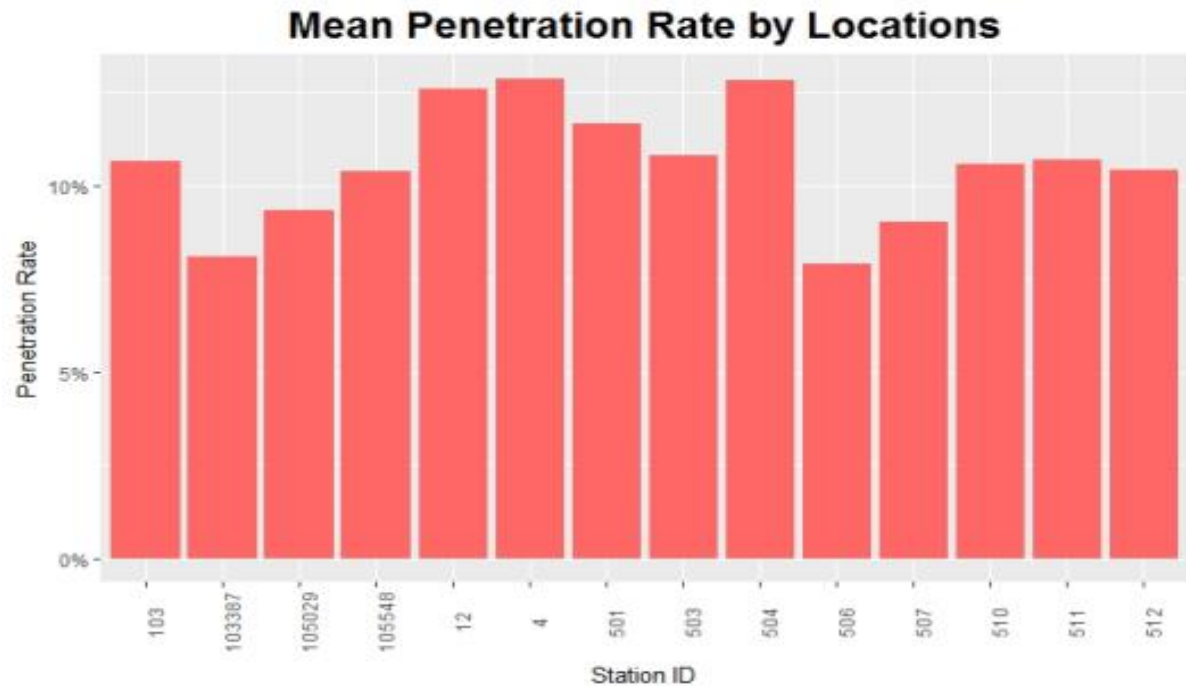
Take Aways – NREL/Colorado Work

- Variation in penetration rates (8% to 12%)
- Large variations in volume, can create problematic outliers
- Volume fluctuation along a roadway a concern
 - Possibly introduce a constrained model



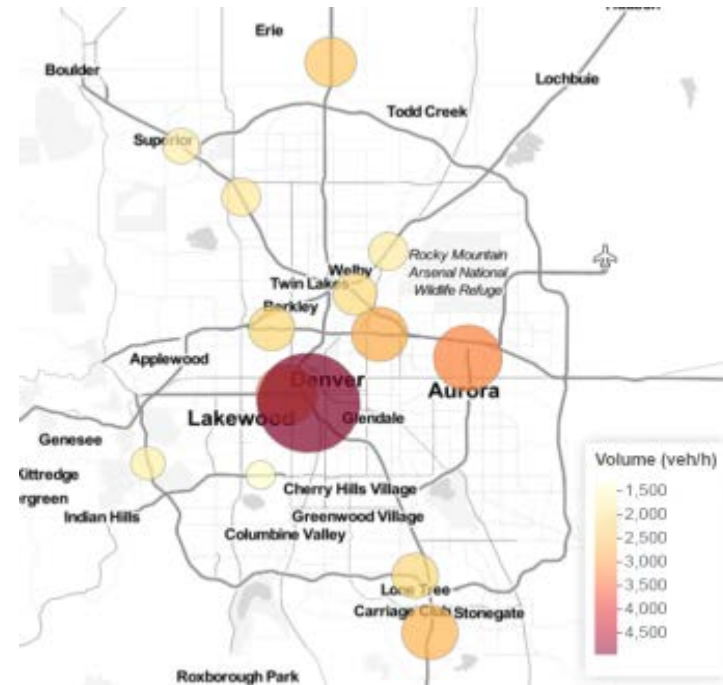
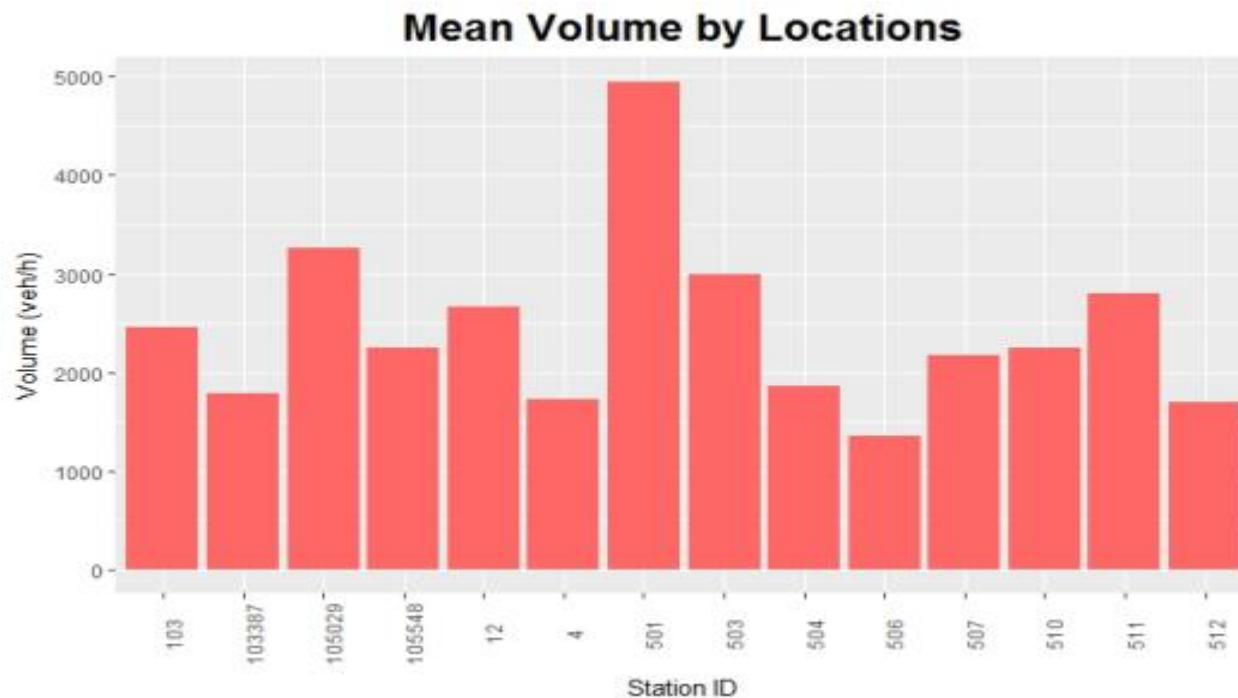
Penetration Rates

- Ranges from 8%-12%



Average Volume

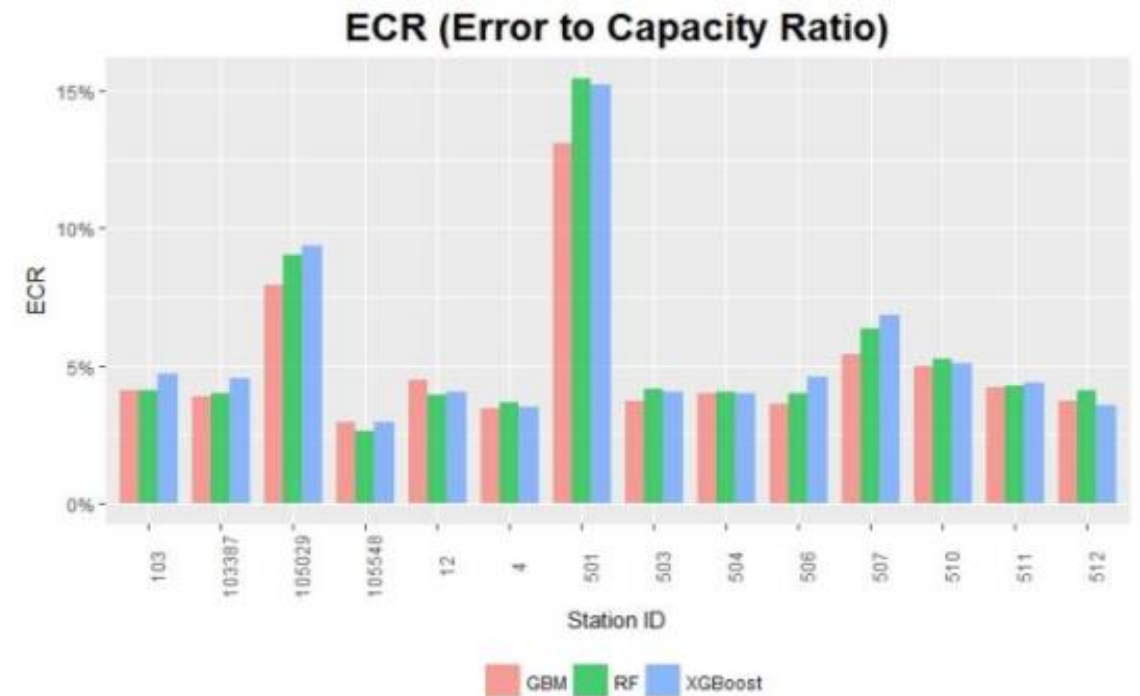
- All Ranges from 1500-3500 veh/hr except 501
- 501 is an unique case



Error to Capacity Ratio (ECR)

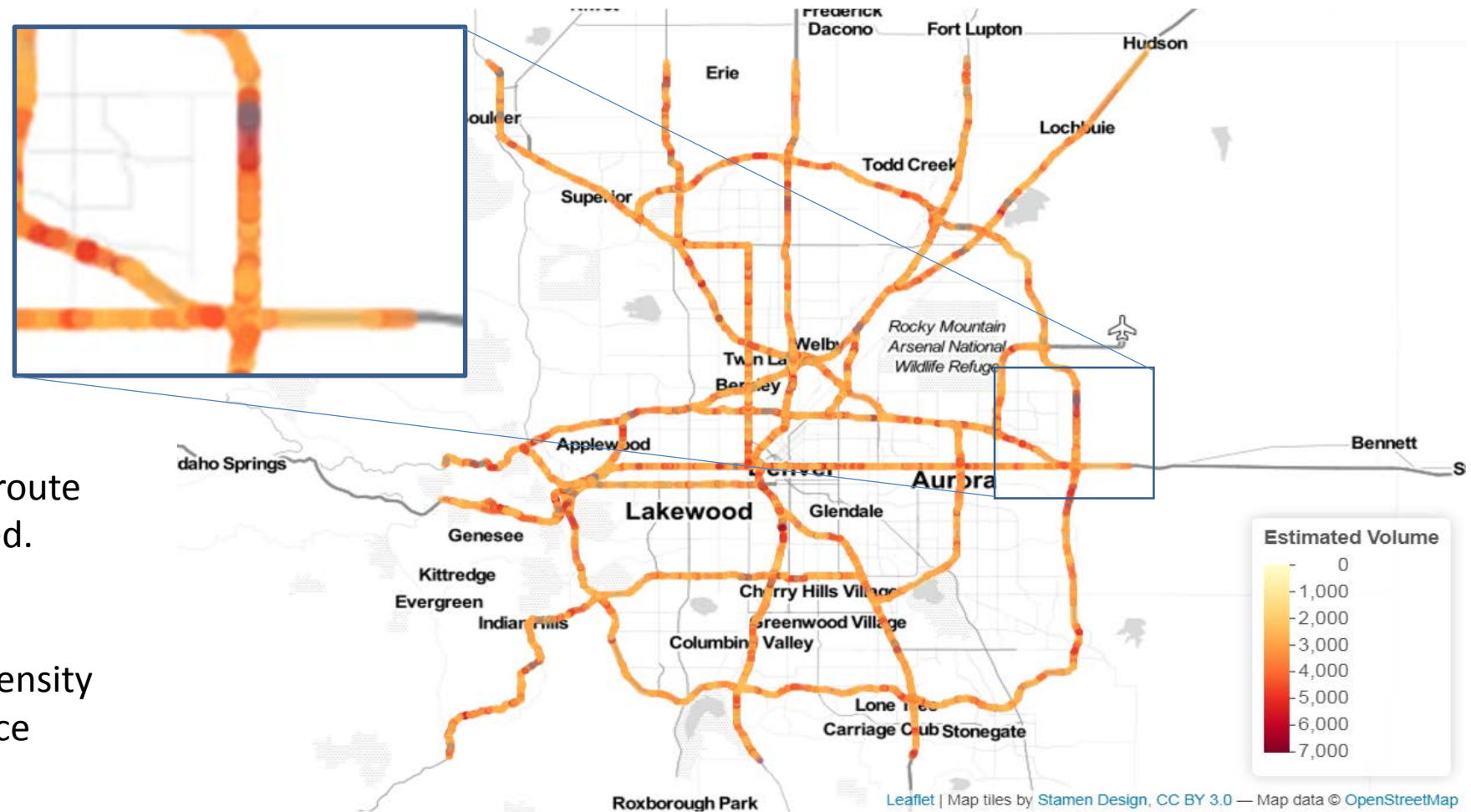
- **Volume Outlier had the highest error**

- RF
 - Range: 2.6%-15.4%
 - Median: **4.1%**
- GBM
 - Range: 2.9%-13.0%
 - Median: **4.0%**
- XGBoost
 - Range: 3.0%-15.2%
 - Median: **4.5%**



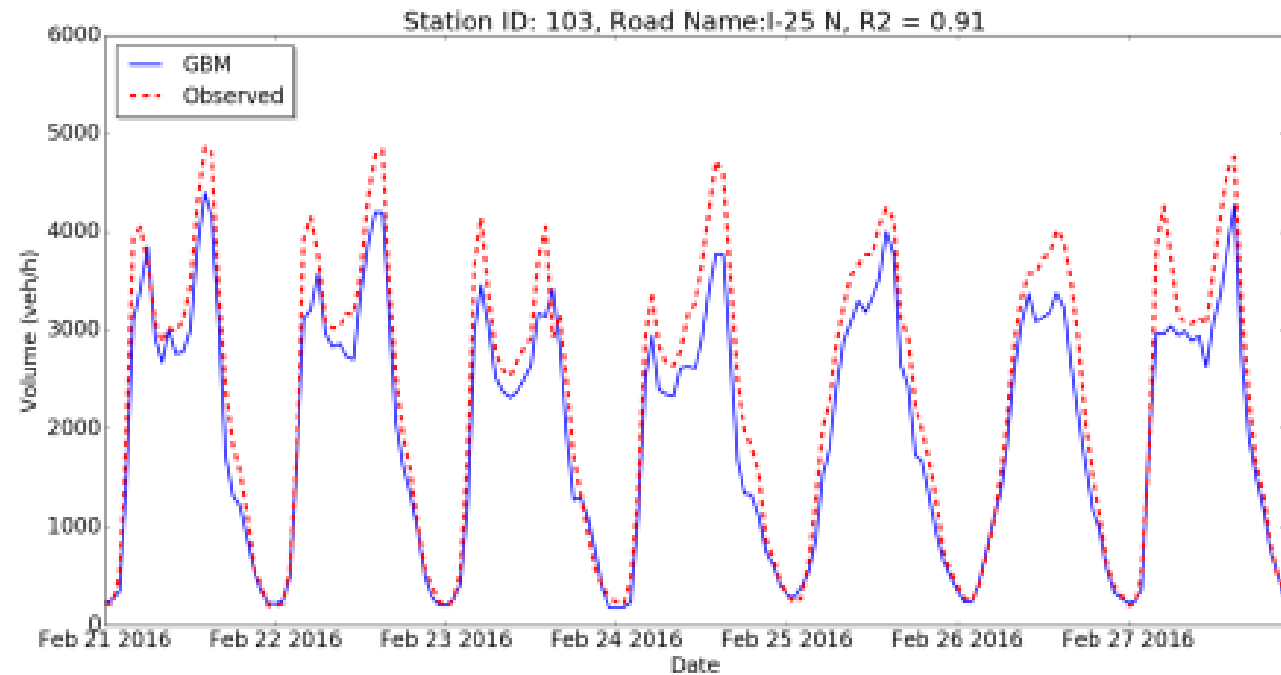
Problem Statement

- Variations along the route need to be considered.
- TMAPS provides no calibration data for sequential volume/density
- May need to introduce constrained model



Final Thoughts

- Larger penetration rates helped
- Volume estimates approached ~5% Error to Capacity ratio (ETCR)
- Moving toward a formalized validation approach



Wrap Up

- Next meeting/webinar
 - Thursday, November 9, 2017
 - 10:30a.m. - 12:00p.m. (EST)



Final Questions



Thank You!

For Questions, please contact:

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