#### Monitoring Roadway Vehicle Miles Traveled using Video Imagery from Transit Buses in Operational Use

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> The Ohio State University Columbus, OH

Smart Mobility Connection Seminar Online

February 11, 2022

# Outline

- Vehicle Miles Traveled
- Video-based Volume Estimates for VMT Determination: "Review"
- Methodological Improvements to Video-based Volume Estimation
- Empirical VMT Results
  - Meaningful empirical patterns
  - Comparison to alternative means of VMT estimation
- Summary and Ongoing Efforts

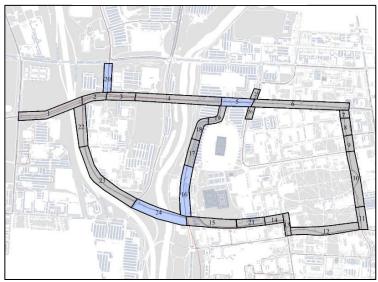
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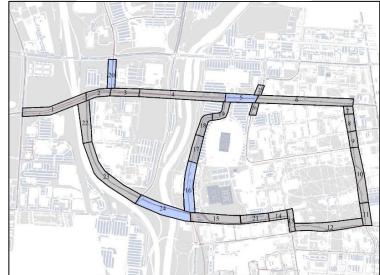
- Most common metric of network-wide travel over a time period
- Used for a variety of monitoring and policy purposes

#### OSU 2019 Network



• Most common metric of network-wide travel over a time period

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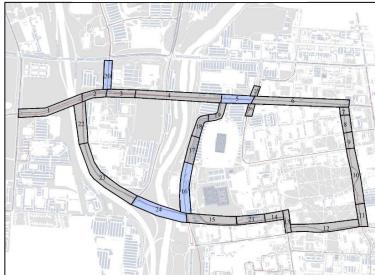


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Sum, over all vehicles, miles traveled by the vehicle on the defined network, during the specified time period

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#### OSU 2019 Network



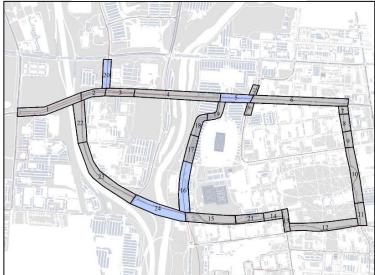
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Sum, over all vehicles, miles traveled by the vehicle on the defined network, during the specified time period

• Usually calculated as mathematical equivalent Sum, over all segments of the defined network, segment length times segment vehicle volume during the specified time period

 Most common metric of network-wide travel over a time period

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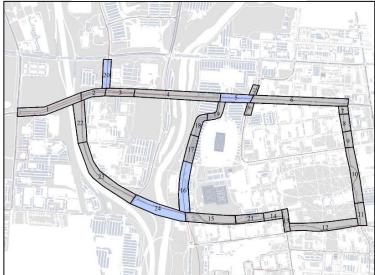
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Sum, over all segments of the defined network, segment **length** times segment vehicle **volume** during the specified time period

- Segment lengths: Straightforward (e.g., GIS)

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- Segment lengths: Straightforward (e.g., GIS)
- Segment volumes: Traditionally from traffic counts

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# **Segment Volumes from Traditional Traffic Counts**

- Volume: Number of vehicles that pass a point on roadway segment over time
- Traditional approach: Go to a point on the roadway, "stay there," and count

Manual Counting



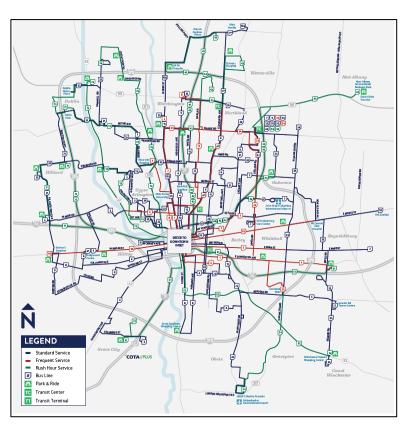
# Traditional traffic studies obtain data to estimate traffic volumes over long time durations but only at *limited locations* and on an *infrequent basis*

#### Road Tubes

OSU Network

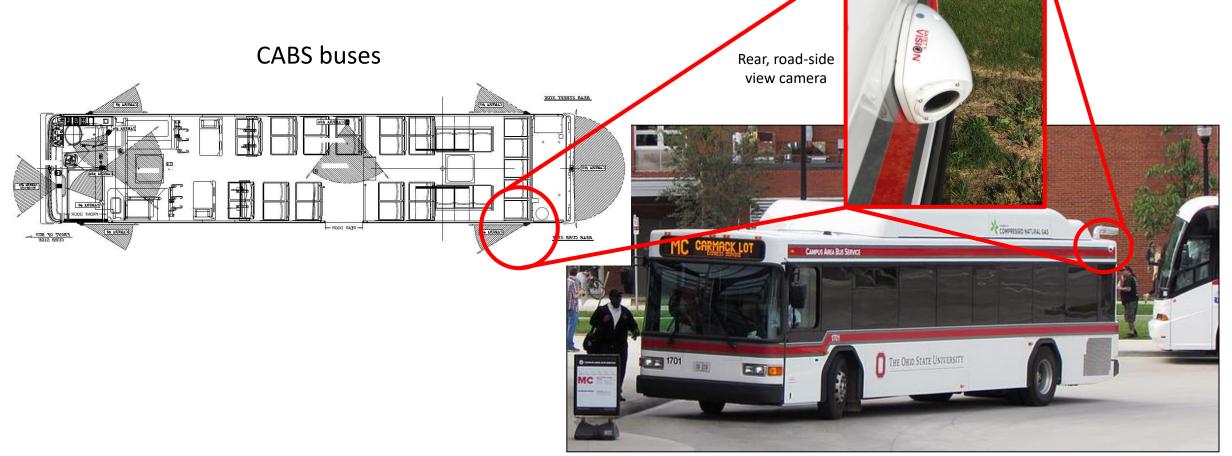
Transit buses cover major roadways across the urban network on a regular, repeated, and ongoing basis

Central Ohio Transit Authority Route Map



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Transit buses are increasingly being equipped with video cameras for safety, security, and liability (i.e., *other*) purposes



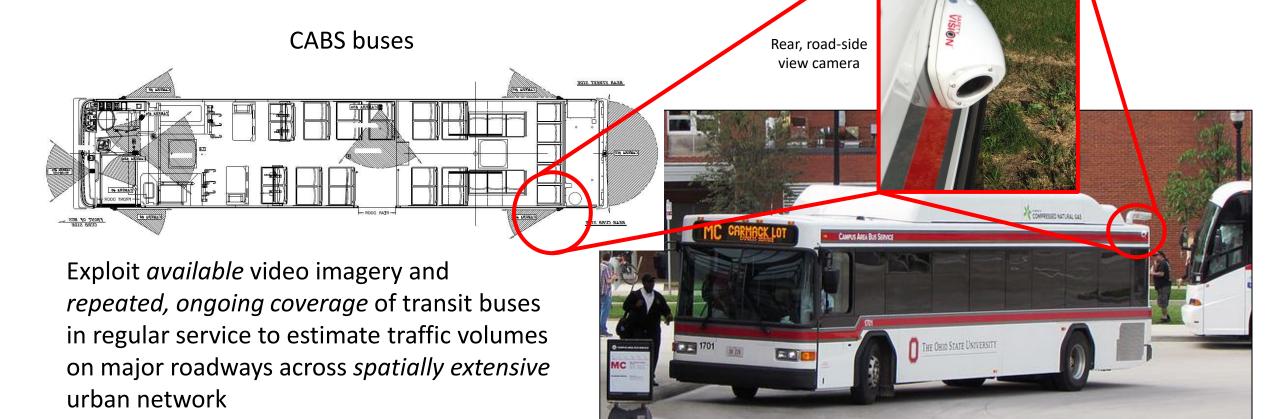
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NICCord, Mishalani, Coifman, Shah, and Charmchi Toosi Smart Mobility Connection

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# **Segment Volumes from Bus Video: Methodology Review**

• Step 0: Convert imagery to digital information

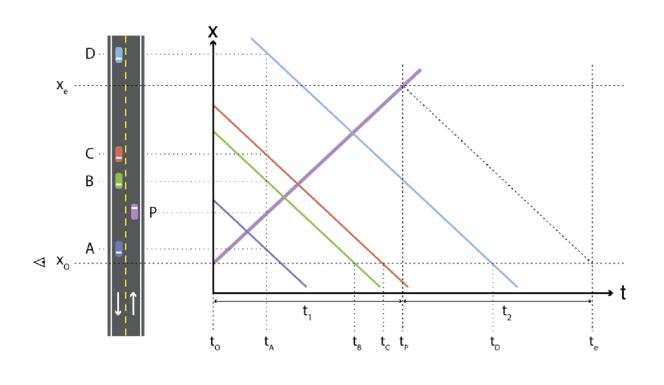
Developed MATLAB-based GUI to digitize vehicle observations, locations, and passage times



# **Segment Volumes from Bus Video: Methodology Review**

- Step 0: Convert imagery to digital information
- Step 1: Estimate volume from an individual bus pass over the segment

Modified the "moving observer" method to account for one-direction bus passes



Hourly Volume Estimate on Bus Pass i

 $V_i = 60 \times n^{veh} / (t_{1.i} + t_2)$ 

# **Segment Volumes from Bus Video: Methodology Review**

- Step 0: Convert imagery to digital information
- Step 1: Estimate volume from an individual bus pass over the segment
- Step 2: Aggregate volumes obtained from multiple bus passes during specified time-ofday period (e.g., hourly volumes)

$$V^h = f(V_1^h, V_2^h, \dots, V_m^h)$$



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• Step 0: Convert imagery to digital information

Automatic Image Identification (we are not there yet) Developed MATLAB-based GUI to digitize vehicle observations, locations, and passage times



- Step 0: Convert imagery to digital information
- Step 1: Estimate volume from an individual bus pass over the segment
- Before aggregating, filter out bus pass volumes
  - Equal to 0



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$$V_i = 60 \times n^{veh} / (t_{1.i} + t_2)$$

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Before aggregating, filter out bus pass volumes

- Equal to 0
- Greater than estimate of capacity (presently using 600 veh/hr/ln)



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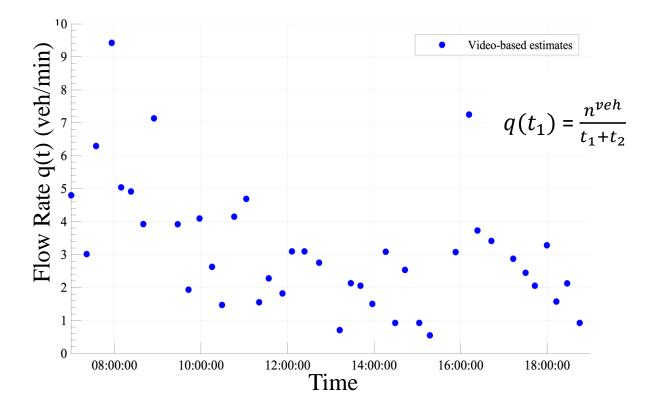
(Investigating other modifications of 0 or excessively large bus pass volumes)



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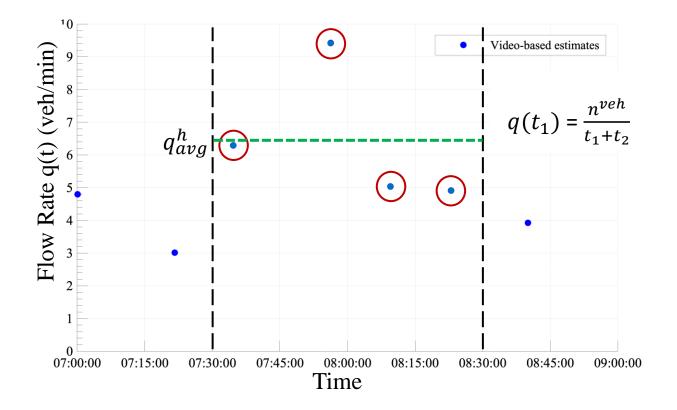
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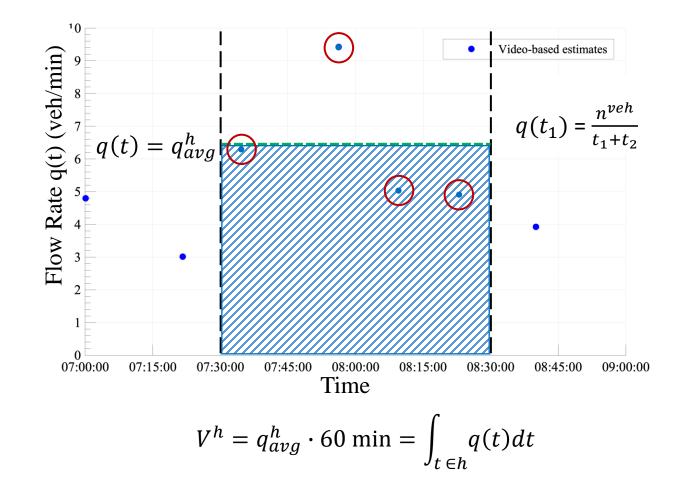
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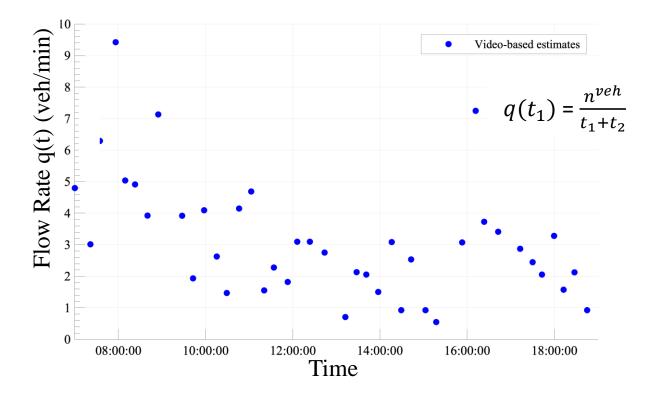
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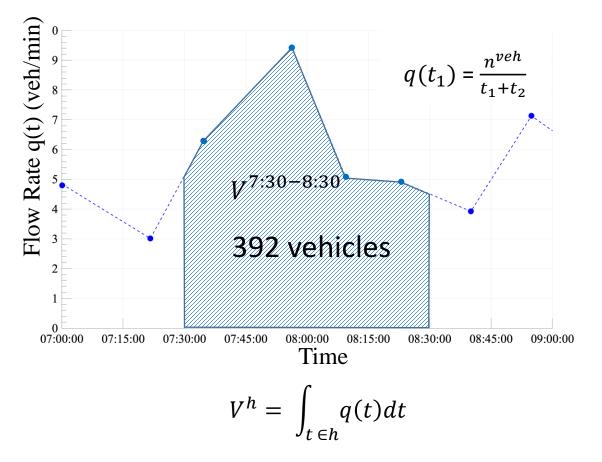
Consider volume as the integral of flow rate q(t) over time during the time interval of interest



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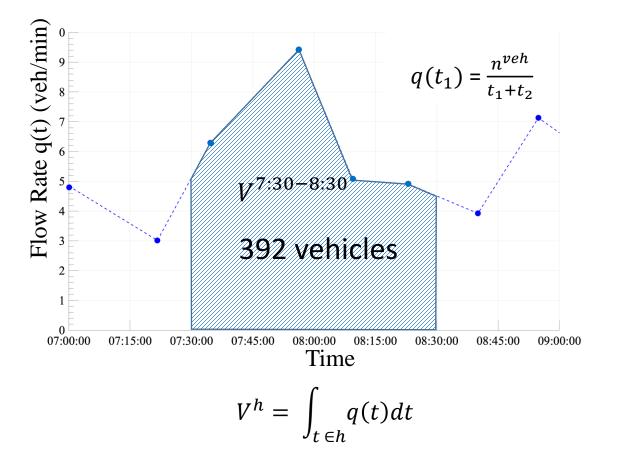
 Estimate q(t) function from individual bus pass estimates



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Consider volume as the integral of flow rate q(t) over time during the time interval of interest

- Estimate q(t) function from individual bus pass estimates
- Presently, linearly interpolate between observations



Compare estimates before and after "improvements" to road tube data considered to be ground truth by segment-hour-direction

Average, across Segment-Direction-Hours of Average Absolute Value of *Difference* between Video-based and Road Tube Hourly Volume

		Before Filtering	After Filtering
Year	No. Seg-Dir-Hr	Average	Average
2018	100*	65.7	56.0
2019	80	116.2	100.0
2020	100	45.7	44.1

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\*97 Segment-Dir-Hr in filtered data

Note: Median differences are much lower than mean differences, but patterns are similar

Compare estimates before and after "improvements" to road tube data considered to be ground truth by segment-hour-direction

Average, across Segment-Direction-Hours of Average Absolute Value of *Relative Difference* between Video-based and Road Tube Hourly Volume, i.e., (Video Vol – Tube Vol)/Tube Vol

		Before Filtering		After Filtering	
Year	No. Seg-Dir-Hr	Average	Integration	Average	Integration
2018	100*	23.4%	21.6%	22.2%	20.8%
2019	80	31.3%	29.3%	26.3%	23.7%
2020	100	19.9%	21.7%	19.3%	20.0%

\*97 Segment-Dir-Hr in filtered data

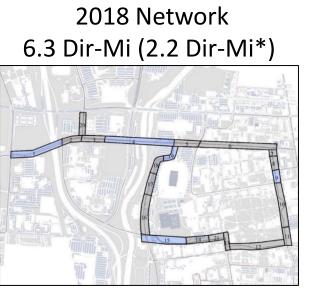
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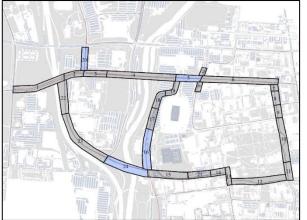
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# **Annual VMT Monitoring on OSU Campus**

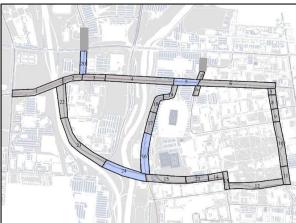
• Extensive data collection held on Thursday, end of October or beginning of November (classes in session) on OSU roadway network over 10-12 hrs



2019 Network 7.7 Dir-Mi (1.3 Dir-Mi\*)



2020/2021\*\* Network 8.0/7.7 Dir-Mi (2.2/0 Dir-Mi\*)



\*Blue shaded segments: Road tube segments (no road tubes in 2021) \*\*Three segments from 2020 network not included in 2021 network

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- Extensive data collection held on Thursday, end of October or beginning of November (classes in session) on OSU roadway network over 10-12 hrs
  - Video-based volumes from OSU CABS buses in operational service



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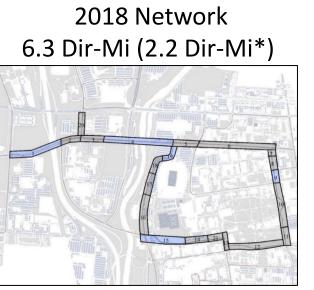
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  - Volumes from Mid-Ohio Regional Planning Commission road tubes



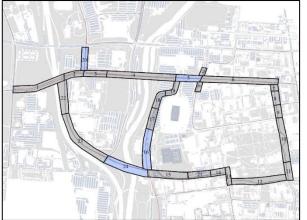


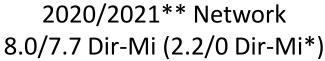


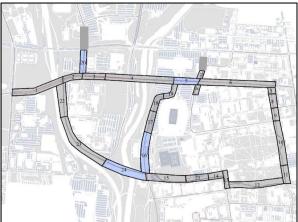
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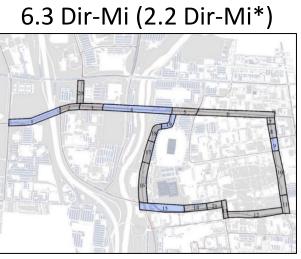




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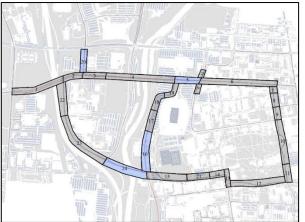
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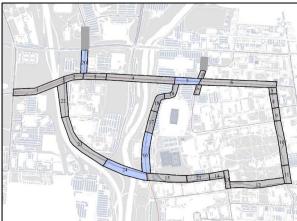


2018 Network

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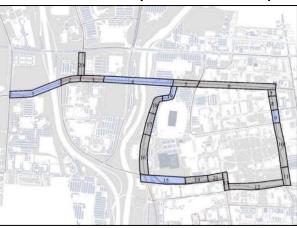


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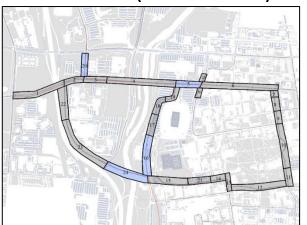
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- Extensive data collection held on Thursday, end of October or beginning of November (classes in session) on OSU roadway network over 10-12 hrs
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- Refined VMT estimates provided to OSU in outreach function

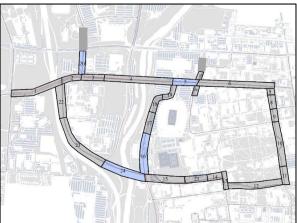
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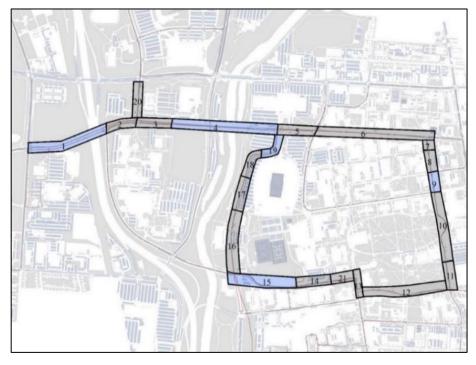


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## Annual VMT Monitoring on OSU Campus: Empirical Results

#### OSU 2018 Network 6.3 Direction-Miles



Vehicle Miles Traveled <sup>1</sup>								
2018 2019		2020	2021					
18,673	19,068	10,455	15,424					
G.F. <sup>2</sup>	1.021	0.559	0.825					

<sup>1</sup>8:00-18:00, after filtering, integration method

## **Annual VMT Monitoring on OSU Campus: Empirical Results**

### • 2018- 2019: Steady traffic

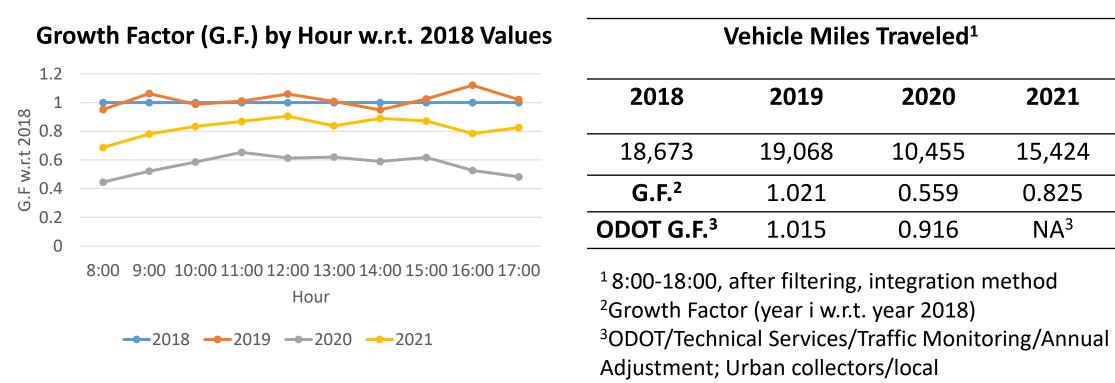
- Reasonable: No change in campus policies or external events
- Consistent with ODOT factor
- 2020: Noticeable traffic decrease
  - Pandemic, Online classes
  - Larger decrease on campus than in . general Ohio travel
- 2021: Noticeable increase w.r.t. 2020
  - Still markedly below 2018, 2019
  - ODOT factors not available yet

2018	2019	2020	2021
18,673	19,068	10,455	15,424
G.F. <sup>2</sup>	1.021	0.559	0.825
ODOT G.F. <sup>3</sup>	1.015	0.916	NA <sup>3</sup>

Vehicle Miles Traveled<sup>1</sup>

<sup>1</sup>8:00-18:00, after filtering, integration method
<sup>2</sup>Growth Factor (year *i* w.r.t. year 2018)
<sup>3</sup>ODOT/Technical Services/Traffic Monitoring/Annual Adjustment; Urban collectors/local
<sup>3</sup>Not yet available

### **Annual VMT Monitoring on OSU Campus: Empirical Results**



### Empirical VMT Comparisons with Results from Popular On-demand LBS Data-based Transportation Information Supplier

- 8:00 18:00 VMT on segments with road tubes calculated from
  - Road tube volumes
  - Bus-based video volumes
  - Data supplier volumes estimates
- Segments with road tubes varied by year

Year	No. Seg-Dir	VMT* [miles]			<b>Relative Difference</b> **	
	Considered	Tube	Video	Supplier	Video	Supplier
2018	10	7,610	7,442	13,445	-2.2%	76.7%
2019	8	5 <i>,</i> 054	5 <i>,</i> 597	6,914	10.7%	36.8%
2020	10	4,929	5,299	11,039	7.5%	124%

\*Different segments are considered in different years; VMT across years are not comparable \*\*Compared to tube-based VMT

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## Summary

- Methodological changes to video-based volume estimation were made and empirically seen to improve performance
- Estimates using transit buses video-based methodology are strikingly more accurate than those provided by a popular on-demand LBS data-based transportation information supplier
- Application of methodology to determine VMT changes over time showed meaningful and otherwise unavailable results

## **Ongoing Activities**

- Continuing efforts towards methodological improvements
  - Using information in passes with zero or very large volumes
  - Smoothing flow rates in integration approach

— . . .

- Applying video-based methodology to multi-day observations
  - Estimating volumes for an "average day"
  - Capturing daily variations
  - Ongoing monitoring: distinguishing daily variations from steady-state levels
- Quantifying uncertainty in volume estimates

## **Ongoing Activities (cont.)**

- Educational activities
  - Continue using concept as class project in transportation data class
  - Continue involvement of graduate and undergraduate students
- Empirical results and outreach activities
  - Continuing to provide OSU campus VMT estimates to campus stakeholders (planners and operators)
  - Investigating impacts of COVID on campus

— . . .

### Acknowledgements

- Other OSU Students: Stephanie Ferzli (presently HDR), Diego Ribeiro de Oliveira Galdino, Marissa McMaster
- OSU Transportation and Traffic Management: Beth Snoke, Tom Holman, Sean Roberts
- Mid-Ohio Regional Planning Commission: Nick Gill, Hwashik Jang, Zhuojun Jiang (presently ODOT)
- ODOT: On-demand LBS data-based supplier of volume estimates