

Measuring Mobility in the 21st Century: What Can We Learn From Mobile Device Location Data?

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Travel Survey



- Historically, personal travel behavior has been tracked through national and local travel surveys
- Typically collect information from a sample of households about:
 - Their socio-demographic characteristics, including work status and work commute
 - The vehicles available to them
 - A diary of all travel undertaken by each member of the household in the last 24 hours

Name			Travel Date MM / DD / YYYY Password				
		Place	Address, Intersection, or Business Name	When did you start your trip?	When did you get there?	Traveled with?	What activitie did you do there
Example	Began the day at: (place at 3AM)	Home					1
Example	Then went to:	Child's School	Luther Jackson Middle School, Falls Church, VA	7:20AM	7:25AM	Chris, Sandy	6
Example	Then went to:	Primary Workplace	ICMA, 777 North Capitol St NE, Washington, DC	7:30AM	8:25AM	Sandy	3, 6
Trip 1	Began the day at: (place at 3AM)						
Trip 2	Then went to:						
Trip 3	Then went to:						
Trip 4	Then went to:						
Trip 5	Then went to:						
Trip 6	Then went to:						
Trip 7	Then went to:						
Trip 8	Then went to:						
Trip 9	Then went to:						
Trip 10	Then went to:						
Trip 11	Then went to:						
Trip 12	Then went to:						
At home act esponsibilities Vork at hom Vork at reg vork locatic Vork-relate sonference, su Yolunteer ac	d activity (e.g., meeting ales call)	y) her g,	 Receive childcare or preschool services Receive adult care services Shop in store (for groceries, clothing, other goods) Eat a meal/have coffee or drink (outside on home or work) Quick stop to pick up food or coffee Fuel vehicle/get gas Receive healthcare services (e.g., medical, dental, etc.) 	18 - Se 19 - Re or m 20 - E 21 - Ge 22 - M ac	ocialize (e.g ecreation (uuseums, vac ercise (e.g., ulk dog) overnmenta all packag etivity	nt (e.g., movi e.g., visit friends e.g., sporting aation) gym, jog/run al, civic, or r e/letter or c el mode (e.g	s/relatives) event, visit bike ride, religious a other pos

Mobile Device Location Data (MDLD)



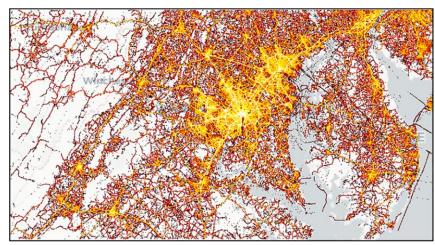
- From cellphone tower, Global Positioning System (GPS), and location-based services (LBS), etc.
- Typically, one location sighting includes an
 - Anonymized device identifier (ID)
 - Latitude and longitude coordinates
 - Time stamps
 - Positioning accuracy



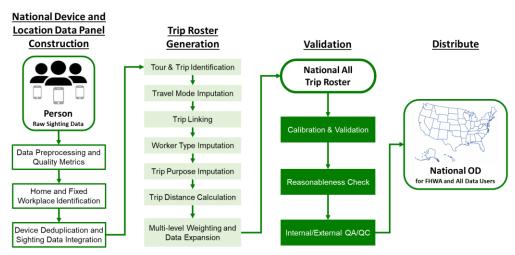
Travel Survey vs. MDLD



	Travel Survey	MDLD		
Pros	 Detailed Information Directly Reported Customization 	 Continuous Monitoring Large-Scale Spatial Accuracy 		
Cons	 Conducted infrequently (every 5-8 years) Covers different households in each wave One-day diaries 	 Limited Behavioral Context Inferential Challenges Computationally expensive 		



Mobile device location data around the State of Maryland



Passenger OD data production flow chart for the Next Generation National Household Travel Survey (NextGen NHTS) OD Data Program

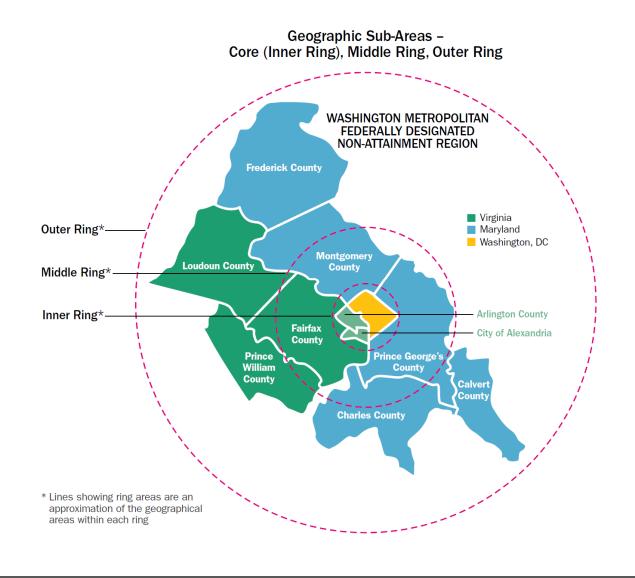
Research Questions



- How do MDLD Vehicle Miles Travelled (VMT) compare with VMT from highquality, more traditional trip diaries/trip rosters?
- Can we form micro-level, longitudinal datasets using the MDLD?
- MDLD VMT...
 - Do they reflect fluctuations and local differences in gasoline prices?
 - Are they good predictors of station-level gasoline prices?

Study Area



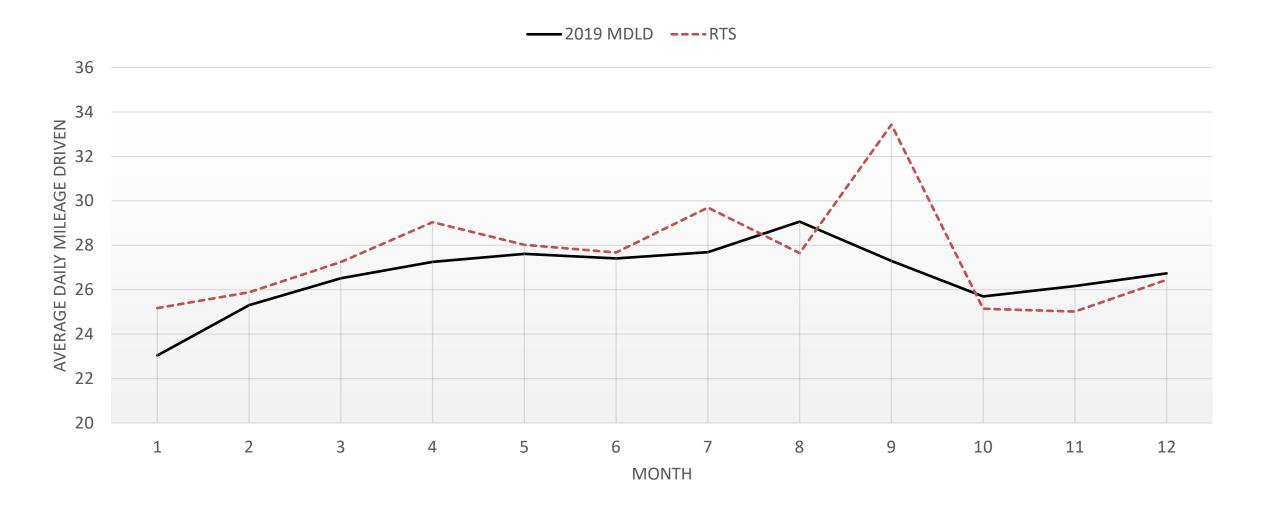


DC + the "Middle Ring" counties

Commuting Patterns:

	2019	2022
DC	35.12%	34.96%
Middle Ring	61.21%	57.71%
Elsewhere or		
unknown	3.66%	7.33%

From Metropolitan Washington Council of Governments' State of the Commute (2019, 2022)



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Leap into Unknown: 2021-2022

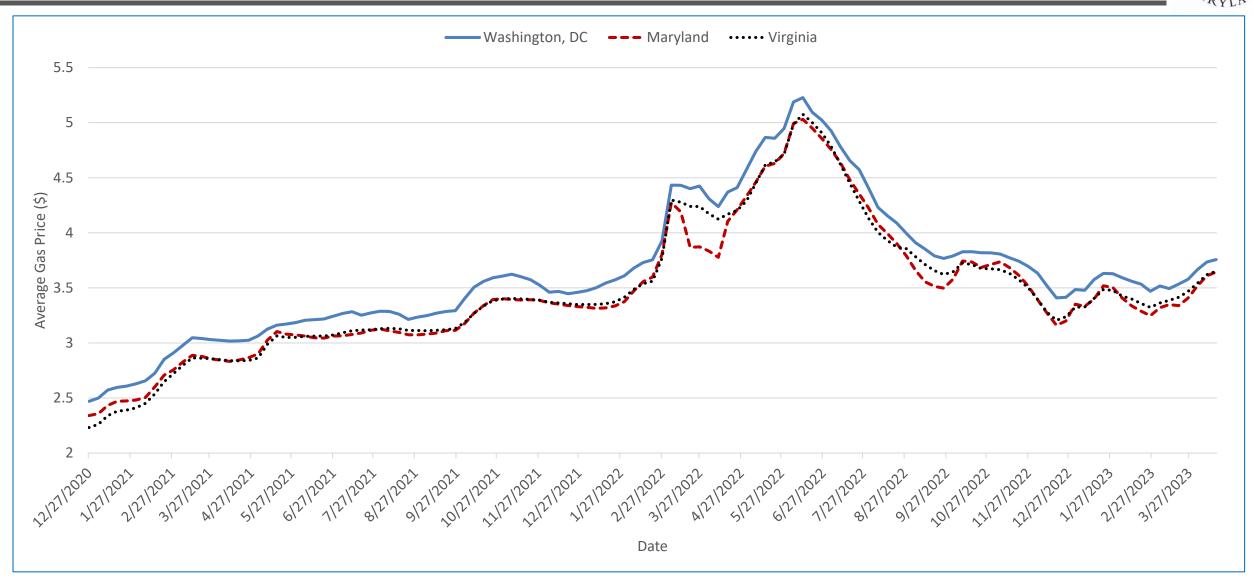
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- No government-run travel surveys were conducted at that time
- With 2021-2022 MDLD:
 - Can we form micro-level, longitudinal datasets using the MDLD?
 - MDLD VMT...
 - Do they reflect fluctuations and local differences in gasoline prices?
 - Are they good predictors of station-level gasoline prices?



	Device Panel	Pseudo Panels		
		#1	#2	
Unit	Device	Census Block Group	Census Tract	
Average daily VMT in each week	By device	By all devices residing in the CBG	By all devices residing in the CT	
N. units	82,866 (10% sample)	817	623	
Avg N. observations (weeks) per unit	7	62	68	
% of Obs with 0 VMT	5%	0.2%	0.02%	

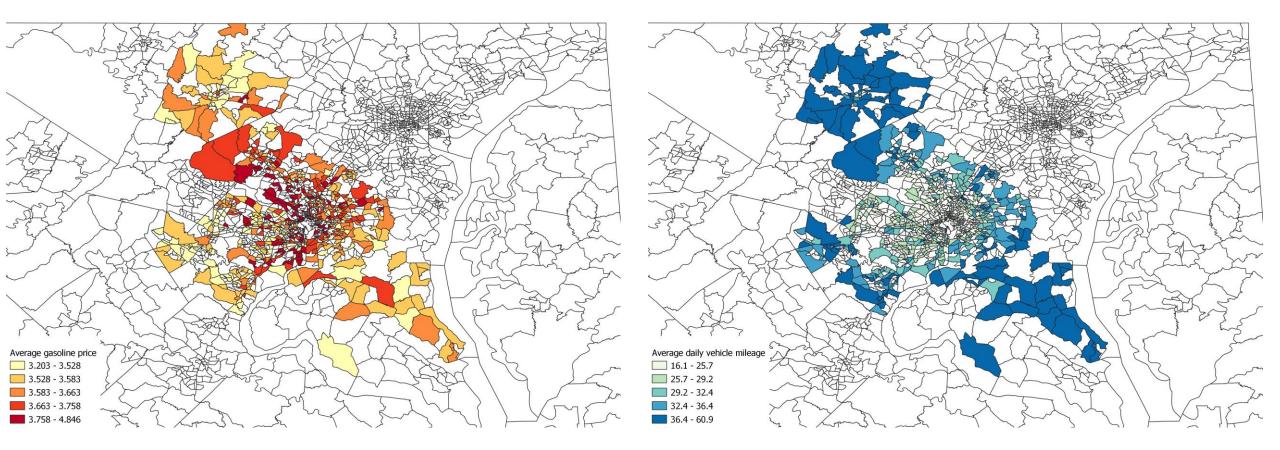
Gasoline Price Variation in Different States





Spatial Distribution at the Census Tract Level





People may choose to live where housing is more affordable, gas prices tend to be lower, and they drive more

Regression of Average Daily VMT from MDLD



Equation

 $VMT_{it} = \alpha_i + \gamma_t + \beta_1 \cdot GasPrice_{it} + \varepsilon_{it}$

 VMT_{it} - the average daily vehicle miles travelled (VMT) by individual or residents in zone *i* at week *t*

- α_i the fixed effect of individual or zone *i*
- γ_t time fixed effects
- β_1 the coefficient of $GasPrice_{it}$

 $GasPrice_{it}$ - the average gasoline price near individual *i* or inside zone *i* at week *t*

 ε_{it} - the error term

Regression of Average Daily Miles Driven from MDLD



• With local gasoline prices

	Device-level panel, Weekly	Device-level panel, incl. zeros, weekly	Block group averages, weekly	Census tract averages, weekly
Coeff. on gasoline price	-0.6009 *** (0.1647)	-0.3255 * (0.1148)	-2.3679 *** (0.5727)	-2.4099 *** (0.6321)
Price Elasticity of daily VMT	-0.0790 (0.0217)	-0.0451 (0.0159)	-0.2679 (0.0649)	-0.2789 (0.0732)
Fixed effects	Device	Device	Block group	Census tract
Time fixed effects	Month, year	Month, year	Month, year	Month, year
R square	0.03	0.03	0.24	0.41
N. Observations	560,779	590,572	50,435	42,257

Regression of Gasoline Price at the Station Level

Equation

$GasPrice_{it} = \alpha_i + \gamma_t + \beta_1 \cdot VMT_{i,t-1} + \varepsilon_{it}$

 $GasPrice_{it}$ - the average gasoline price of gasoline station i at week t

- α_i the fixed effect of gasoline station *i*
- γ_t time fixed effects
- β_1 the coefficient of $VMT_{i,t-1}$

 $VMT_{i,t-1}$ - the total vehicle miles travelled (VMT) at the previous week t - 1 by individuals residing in the zone where the gasoline station *i* is located

 ε_{it} - the error term



Gas station specific gasoline prices



	#1	#2
Coeff. on total mileage driven in the previous week at Census tract-level, Weekly	-3.12E-08 (2.08E-08)	1.28E-07 *** (3.37E-08)
Coeff. on number of gas stations within 0.25 miles	-0.0172 *** (0.0011)	NA
Other independent variables that are statistically significant	Population density, Income distribution, Age distribution, Education-level distribution, Road-network density, 	NA
Fixed effects	Brand, State	Station, Brand, State
Time fixed effects	Month, Year	Month, Year
R square	0.65	0.67
N. Observations	20,021	20,021

Conclusion



- Average Daily VMT from 2019 MDLD shows a similar monthly trend as VMT from RTS
- Different ways of forming longitudinal datasets with MDLD
 - Device-level panel, with shorter tracking period, and more units
 - Pseudo Panels, aggregated, with longer series, fewer units, and more stable
- MDLD VMT...
 - Average daily VMT negatively correlated with the price of gasoline, reflecting fluctuations and local differences in gasoline prices.
 - Total VMT help explain the price of gasoline in the area and thus serve as a good proxy for demand pressure on fuel prices.



Thank you!

Comments? Questions?

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