Impacts of Ridesourcing – Lyft and Uber – on Transportation including VMT, Mode Replacement, Parking, and Travel Behavior



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Doctoral Dissertation Defense January 19, 2017

Acknowledgments

Committee:

- Dr. Wesley Marshall (Advisor)
- Dr. Bruce Janson (Chair)
- Dr. JoAnn Silverstein
- Dr. Carey McAndrews
- Dr. Debbi Main
- Dr. Kevin Krizek

Funding Institutions

- National Science Foundation
 - IGERT Program Fellowship
 - Bridge to the Doctorate Fellowship
- Dwight Eisenhower Fellowship
- Mountain-Plains Consortium
- ITE Scholarships
- CU Denver Scholarships











People and Organization

- Classmates and Friends
- Transportation Professionals
- Lyft and Uber
- Lyft/Uber Passengers
- FAMILY





Agenda

- I. Introduction
- II. Background
- III. Literature Review
- IV. Research Methods
- V. Data

Agenda

- VI. Driver Perspective
- VII. VMT Impacts
- VIII. Parking Impacts
- IX. Travel Behavior Changes

Agenda

- X. Overall Results
- XI. Summary Conclusions
 - Policy Recommendations
 - Future Applications
 - Future Research

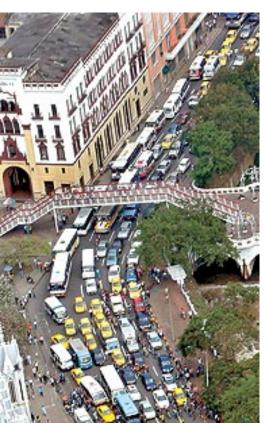
I. Introduction

- Motivation
- Research Needs



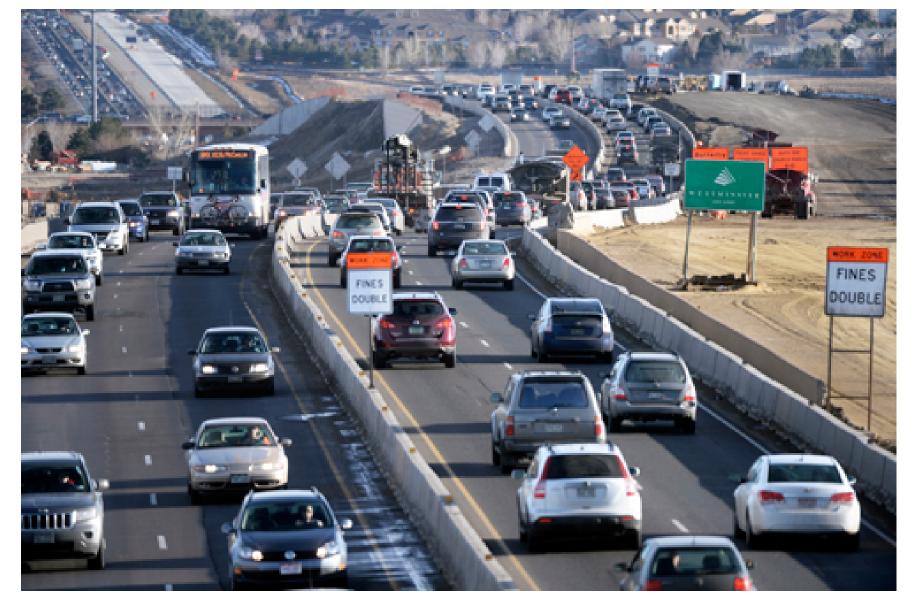












Colorado

Photo Source: Karl Gehring Denver Post





Disrupting Transportation

Many factors, including:

- Social networks
- Real-time information
- Mobile technology

Allow the creation and popularization of on-demand transportation services all over the world.































Ridesourcing

Sourcing of rides from a 'for-fare' driver pool accessible through an app-based platform.





Other names:

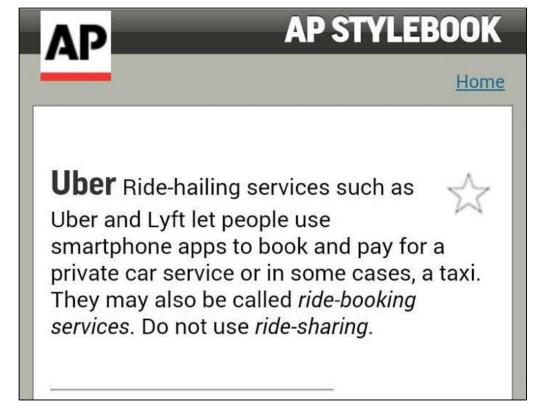
"Transportation Network Companies (TNCs)", "ride-hauling", "ride-booking", "ride-matching", "on-demand-rides", "app-based rides"



Ridesourcing

RIDESOURCING # RIDESHARING

Associated Press





Lyft to go global and take on Uber outside the US

CNBC - Jan 13, 2017

Number two U.S. ride-hailing company, **Lyft**, is growing faster and cutting losses faster than its giant competitor, **Uber**. And this year, the startup ...

Lyft might be eyeing a global market to take on **Uber** Business Insider - Jan 13, 2017

Uber, **Lyft**, transit agencies see potential for partnerships In-Depth - San Francisco Chronicle - Jan 12, 2017



Lyft drivers say they are happier, better paid than Uber drivers

By Carolyn Said, San Francisco Chronicle Updated 3:30 pm, Tuesday, January 17, 2017























Q SEARCH

N.Y. / REGION

Yellow Cab, Long a Fixture of City Life, Is for Many a Thing of the Past

By WINNIE HU JAN. 15, 2017













NYC TRANSPORTATION

Uber and Lyft cars now outnumber yellow cabs in NYC 4 to 1

Who will sing for the yellow cab? BY RACHEL SUGAR | JAN 17, 2017, 12:15PM EST

SF blasts Uber, Lyft for downtown traffic congestion



Ride-hail companies like Uber and Lyft are being blamed by the San Francisco Municipal Transportation Agency for a lack of regulation that has led to increased traffic in The City. (Ekevara Kitpowsong/Special to S.F. Examiner)

By Joe Fitzgerald Rodriguez on December 11, 2016 1:00 am

The potential 45,000 Uber and Lyft drivers circling San Francisco streets for commute fares are gumming up city traffic, according to transit officials.

In a recent state regulatory filing, the San Francisco Municipal Transportation Agency took the California Public Utilities Commission — which is tasked with regulating ride-hail companies — to task for failing to reasonably limit the industry's explosive growth.





The New York Times

SU

N.Y. / REGION

City Hall and Uber Clash in Struggle Over New York Streets

By MATT FLEGENHEIMER and EMMA G. FITZSIMMONS JULY 16, 2015









David Plouffe, a top Uber operative, at Sylvia's restaurant in Harlem on Tuesday, was joined by more than a dozen community leaders, all of them critical of a proposed cap on the company's growth. Bryan R. Smith for The New York Times

For months, the clash has seemed inevitable: the professed disrupters of municipal transportation policy and the chief executive of the country's

STREETS**BLOG** SF

Pedestrian Safety / Bicycling / Muni / Parking / Peninsula / California



Lyft and Uber Won't Release Data to Shed Light on **How They Affect Traffic**

OUR SITES V

By Aaron Bialick Jun 30, 2015

INTRO

s ride-hail services like Lyft and Uber have boomed in San Francisco and other cities, proponents claim they help reduce demand for parking and road space by making it easier for people to own fewer cars. But very little data has been released by the ride-hail companies that would allow experts to assess their impact on streets and traffic.

In a panel discussion yesterday, Lyft's Curtis Rogers emphasized that reducing car ownership is "our end goal that we think we share with the city."

But when Thea Selby of the SF Transit Riders Union pressed Rogers for data to show whether Lyft might be substituting for transit trips more than car trips, he said he couldn't provide it. Rogers insisted, however, that Lyft doesn't want to compete with Muni, walking, or bicycling. "We think we're just one more piece to the puzzle."

"We celebrate Muni getting better," said Rogers. "We're well we pulled everyone off of Muni and put them in be going two miles per hour on the road. That's



Photo: Jason A. Staats/Twitter

Research Needs





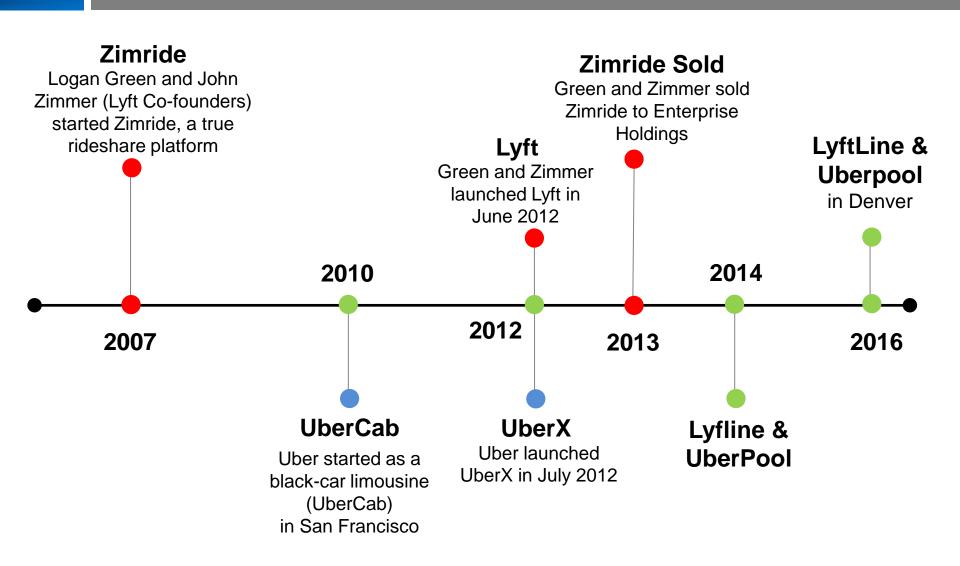


- > DATA
- > DRIVER SIDE
 - Efficiency
 - Earnings
- > VMT IMPACTS
- > PARKING IMPACTS
- > TRAVEL BEHAVIOR
 - Mode Replaced
 - Why?

RIDESOURCING

STUDY

II. Background



Operations

- Uber operates globally (450+ cities)
- Uber completed 2 billion trips in the summer 2016
- First billion rides in 6 years
- Second billion in 6 months
- Lyft so far is only in the U.S.
- Lyft is giving rides at a rate of 17 million U.S. rides per month
- Lyft is estimated to have 20% market-share



Valuation

- Latest Uber valuation: \$62.6 billion
- Lyft: \$5.5. billion dollars
- Valuation without owning vehicles, physical infrastructure, or having to hire drivers as employees

Academic

- Anderson (2014): Interview 20 drivers (Anthropology) about driver strategies and possible VMT impacts
- Cramer & Krueger (2016): Comparison of UberX with Taxis. Hired by Uber to do the study
- Rayle et al. (2016): Intercept survey in San Francisco comparing ridesourcing with taxis. User characteristics, wait times, and trips served

Organizations

- SUMC (2016): Intercept Survey in seven U.S. cities. Higher use of shared modes, the more likely people use transit and own fewer cars.
- FiveThirtyEight (2015): Used data acquired via a Freedom of Information Act request to the city.
 In NY, Uber is taking rides away from taxis and covers a larger area

- Review of carsharing literature
- Help develop research methods for this dissertation
- Each Chapter includes a more detailed Literature Review

- Very limited research studies
- Lack of open data
 - Levitt, Freakonomics (2016). Why Uber Is an Economist's Dream.
- Independent data questionable
- Research design questionable
- Several gaps

Book Chapter

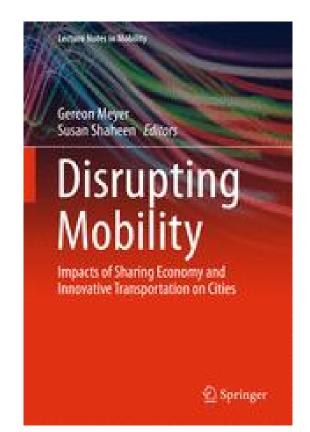
"A Framework for Understanding the Impacts of Ridesourcing on Transportation"

(Henao & Marshall, 2017)

Disrupting Mobility

Impacts of Sharing Economy and Innovative Transportation on Cities

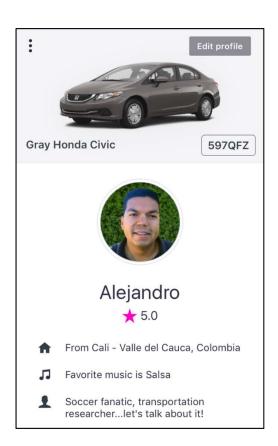
Editors: Gereon Meyer, Susan Shaheen

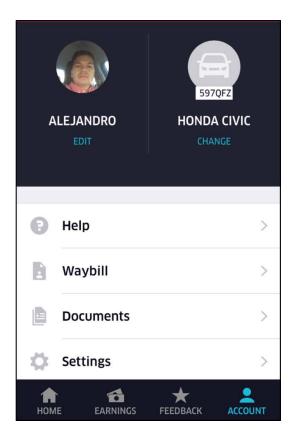




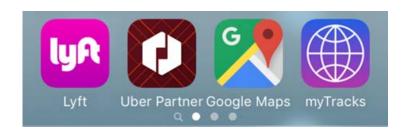
IV. Research Methods

- Innovative approach to collect data
- Became an independent-contractor to drive for both Lyft and Uber and get access to exclusive data
- Exploratory Analysis
- > IRB Approval
- > Two Datasets:
 - 1. Driver Dataset (416 rides)
 - 2. Passenger Dataset: (311 Surveys)





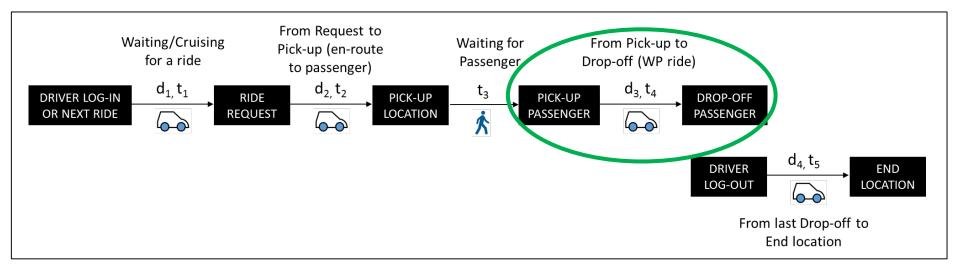
Lyft and Uber Driver Profiles



Smartphone Apps

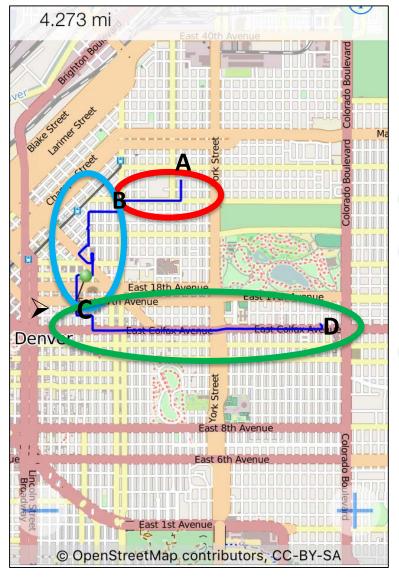


Driver Dataset



Driver Data Collection (e.g. travel attributes)

METHODS



Mileage and Times

- Cruising/Waiting for a ride (A-B)
- En-Route to passenger (B-C)
 - Waiting for Passenger (C)
- With-passenger (WP ride) (C-D)

GPS Tracking of a Lyft/Uber ride

DRIVER DATA COLLECTION								
BREAKS - Mins: _	1	Miles:			Last ride to L 0	Location: OG-OUT time & dist t to End time & dist	t: mins (mi)
	□Foggy □Ra	iny □Sunny □				ool		
Pick-up Location:	:	Waiting/Crui	□SU-	-L □SU-N	1 □SU-H □U-L	for a ride distance:	-MH =U-H =\$	Special
		MyTrac	ks distance:	mi	Time Ride Sta	ne: Req to arts: Drive	er Waiting:	mins
Lyft/Uber est. tir	me: mins	GoogleMaps: MyTrac	mins (ks distance:	_ mi) mi	Time when Ri	□U-ML □U-M □U de Ends: f Lyft/Uber report:	Ride Time: mins (_ mins mi)
	re shown on Ap	p: \$ To Dri	P. Co ver: \$+ ti	ost: \$ ip: \$	C Walk t	ruising time & dist: to dest time & dist: □Chatty □Quiet	mins (mins (mi) mi)

Driver Data Collection Form



Passenger Survey

I interviewed passengers during the ride:



"Hi rider,
I'm a grad student doing
research on transportation.
Would you help me by
doing a short survey (~6
minutes) about this ride?

You can use my tablet or go to this link: www.ride-survey.com.

Thank you!"



Passenger Survey

- Passengers took survey on the tablet provided
- > On their own devices: www.ride-survey.com
- > In some cases, verbal interview

Passenger survey questions:

- 1. Specific Trip Questions (Q1-Q10)
- 2. General Use Questions (Q11-Q25)
- 3. Demographic Questions (Q26-28)

V. Data

PASSENGER DATASET Travel Attributes: • Travel Times • Travel Distances • Earnings RIDESOURCING DATA PASSENGER DATASET Survey Questions: • Specific Trip (Q1-Q10) • General Use (Q11-Q25) • Demographics (Q26-Q37)

416 Rides

- > 198 Lyft
- > 164 UberX
- ➤ 39 LyftLine
- ➤ 15 UberPool

311 Passenger Surveys

SURVEY RESPONSE

RATE: 87.5%

DATA

Origin-Destination (O-D) Matrix

DESTINATION ORIGIN	Home	Work	School	Shopping/ Errands	Going Out/ Social	Airport	Hotel/ Airbnb	Family/ Friend	Other	Totals
Home	2	36	16	7	34	18	0	4	12	129
Work	21	8	1	1	1	2	6	0	1	41
School	5	0	0	3	0	0	0	2	0	10
Shopping/Errands	11	1	0	3	1	0	0	0	0	16
Going Out/Social	30	1	0	3	10	0	3	3	1	51
Airport	3	0	0	0	0	0	2	0	0	5
Hotel/Airbnb	0	2	0	0	7	4	0	0	4	17
Family/Friend	10	1	0	0	1	1	3	1	2	19
Other	8	3	0	2	2	1	3	1	3	23
Totals	90	52	17	19	56	26	17	11	23	311

	Ridesou	cing	Denver Population ^a		Ridesou	rcing	Denver Population ^a
	Responses	(%)	(%)		Responses	(%)	(%)
Gender				Marital Status			
Female	145	46.9%	50.0%	Single or never married	185	62.7%	41.7%
Male	162	52.4%	50.0%	Married or in a family relationship	80	27.1%	39.2%
Prefer not to answer	2	0.6%		Separated, divorced, or widow	28	9.5%	19.1%
n	309			Other	2	0.7%	
				n	295		
Residency							
Local Resident	254	82.2%		Household size ^b			
Visitor	55	17.8%		1	65	22.3%	
n	309			2	129	44.2%	
				3	56	19.2%	
Age				4	30	10.3%	
18-24 ^b	78	25.2%	10.0%	5+	12	4.1%	
25-34	132	42.7%	21.8%	n	292		
35-44	56	18.1%	15.4%				
45-54	30	9.7%	11.7%	Children in household			
55-64	7	2.3%	10.5%	Yes	47	20.5%	25.1%
65+	6	1.9%	10.7%	No	182	79.5%	74.9%
n	309			n	229		
Race/Etchnicity				Education			
Asian	24	7.8%	3.5%	Less than High School	9	3.0%	13.9%
Black/African American	n 16	5.2%	9.4%	Graduated high school or equiv.	49	16.5%	17.7%
Hispanic or Latino	39	12.7%	30.9%	Some college, no degree	58	19.5%	18.3%
White	206	66.9%	53.1%	Associate or Bachelor's degree	124	41.8%	32.5%
Other	16	5.2%	3.1%	Advanced degree (Master's, PhD)	57	19.2%	17.6%
Prefer not to answer	7	2.3%		n	297		
n	308						
				Employment Status			
Household Income ^c				Working (Full-time or Part-Time)	246	81.7%	70.9%
\$30K or less	34	11.5%	28.3%	Volunteer	1	0.3%	
\$31K - \$45K	56	18.9%	14.0%	Unemployed	15	5.0%	6.3%
\$46K - \$60K	58	19.6%	11.1%	Retired	8	2.7%	
\$61K - \$75K	30	10.1%	10.0%	N/A	31	10.3%	
\$76 - \$100K	40	13.5%	11.9%	n	301		
Over \$100K	50	16.9%	24.9%				
Prefer not to answer	28	9.5%		Student Status			
n	296			Student (Full-time or Part-time)	70	23.3%	34.2%
				Not currently a student	230	76.7%	65.8%
				n	300		

^a 2011-2015 ACS 5-Year Estimates, Denver County ^b Age 1st Range is 15 - 24 for ACS

^c Income Range for ACS slighly different

	Ridesour	cing	Denver Population ^a
	Responses	(%)	(%)
Gender			
Female	145	46.9%	50.0%
Male	162	52.4%	50.0%
Prefer not to answer	2	0.6%	
n	309		
Residency			
Local Resident	254	82.2%	
Visitor	55	17.8%	
n	309		
Age			
18-24 ^b	78	25.2%	10.0%
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n	309		

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	Responses	(%)	(%)
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\$76 - \$100K	40	13.5%	11.9%
Over \$100K	50	16.9%	24.9%
Prefer not to answer	28	9.5%	
n	296		

VI. Driver Perspective

- > Travel times and distances
- > Earnings

Data Analysis

TRAVEL DISTANCES

$$d_{shift} = \left[\sum (d_1 + d_2 + d_3) \right] + d_4$$

$$d_T = \sum d_{shift} = \sum d_1 + \sum d_2 + \sum d_3 + \sum d_4$$

$$VMT_T = \sum d_1 + \sum d_2 + WPMT_T + \sum d_4$$

 $VMT_T = WPMT_T + Additional VMT$

$$Ride sourcing \ Efficiency \ Distance = \frac{\sum d_3}{d_T} = \frac{WPMT_T}{VMT_T}$$

ADDITIONAL PERCENT OF WPMT

$$\frac{Additional\ VMT}{WPMT_T} = \frac{VMT_T}{WPMT_T} - 1$$

$$Total \ \textit{Miles per} \ 100 \ \text{WPMT} \ = \frac{100 * \textit{VMT}_T}{\textit{WPMT}_T}$$

TRAVEL TIMES

$$t_{shift} = \left[\sum (t_1 + t_2 + t_3 + t_4) \right] + t_5$$

$$t_T = \sum t_{shift} = \sum t_1 + \sum t_2 + \sum t_3 + \sum t_4 + \sum t_5$$

$$\textit{Ridesourcing Efficiency Time} = \frac{\sum t_4}{t_T}$$

EARNINGS

$$Gross Earnings (\$/hr) = \frac{\sum Driver Earnings (incl. tip)}{t_T}$$

$$Gross \, Earnings \left(\$/_{mile} \right) = \frac{\sum Driver \, Earnings \, (incl. \, tip)}{d_T}$$

 $Net\ Earnings = Gross\ Earnings - Expenses$

DRIVER STUDY

Travel Times and Distance Summary Statistics

		DRIVER OR NEX	-	-	PICK-UP OCATION t ₃	PASSENIOED	d ₃ , t ₄	DROP-OFF PASSENGER DRIVER LOG-OUT d ₄ , t ₅	END LOCATION
			Waiting/Cruising for a ride	From Request to Pick-up (en-route to passenger)	Waiting for Passenger		ick-up op-off Pride)	From last Drop-off to End Location	Totals (t _T & d _T)
	(š	$Total\left(\Sigma t\right)$	4,965.00	2,511.00	531.00	6,1	06.00	1,416.00	15,529.00
Time	(minutes)	Mean	11.94	6.04	1.28		14.68	21.78*	37.33
Ξ	(min	St. Dev.	15.46	3.65	2.10		10.04	12.27*	20.30
		Median	7.50	5.00	1.00		11.50	20.00*	32.83
		$Total\left(\Sigma d\right)$	635.91	600.56		2,9	29.94	784.29	4,950.69
3)	S	Mean	1.53	1.44			7.04	12.07*	11.90
Distance	(miles)	St. Dev.	3.94	1.44			8.60	7.43*	10.37
	<u> </u>	Median	0.20	1.00			3.55	12.00*	8.30
Ave	rag	e mph		14.35			28.79	33.23	19.13

n=416 (Lyft: 198, LyftLine: 39, UberX:164, UberPool: 15)

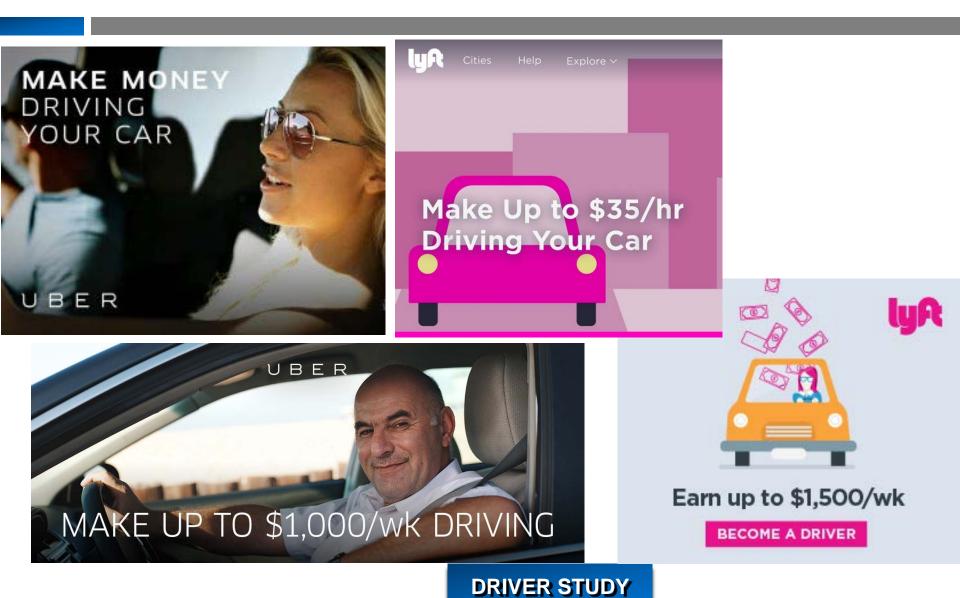


^{*} Commute based on 65 shifts

Times and Distance Efficiency

	WP Ride $(\Sigma d_3 \& \Sigma t_4)$	Total minus Commute at End	Efficiency: WP/(Total minus Commute at End)	Totals (tr. & dr.)	Overall Efficiency (WP/Total)	Percent of	VMT per 0-WPMT
Time (minutes)	6,106.0	14,767.0	41.3%	15,529.0	39.3%		
Distance (miles)	2,929.9	4,482.9	65.4%	4,950.7	59.2%	69.0%	169.0

Earnings



Earnings

Lyft/Uber Fares & Commission

		P	assenger (
	Lyft/Uber Service Fee Fare		Cost per Cost per Minute Mile Fare Fare		Minimum Paid by Passenger (Fee + Fare)	To Driver**	Lyft/Uber Commision**
Lyft	\$2.10	\$0.50	\$0.12	\$1.01	\$7.10	80% Fare	100% Service Fee
UberX	\$1.95	\$0.75	\$0.13	\$1.00	\$6.95	+ 100% Tips	+ 20% Fare

^{*} Rates as of Fall 2016 in U.S. dollars. Rates varied and have been lowered over time

^{** 20%} Commision when first signed-up in 2014. Newer drivers pay a higher commision (25% or more)

Passenger Cost, Driver Earnings, Real Commission

	Passe	nger Cost	1	To Driver		To Lyft/Uber		
	Total Paid (before tip)	Total Cost per WP Mile (before tip)	Total Earned (before tips)	Tips	Total Earned (with tips)	Actual Commision (before tip)	Actual Comission (after tip)	
Lyft (n=237)	\$2,934.58	\$1.87	\$2,059.25	\$276.00	\$2,335.25	29.8%	27.3%	
Uber (n=179)	\$2,505.62	\$1.84	\$1,687.83	\$39.00	\$1,726.83	32.6%	32.1%	
All Trips (n=416)	\$5,440.20	\$1.86	\$3,747.08	\$315.00	\$4,062.08	31.1%	29.4%	

^{*} Earnings include prime and guarantee bonus per hour but does not include initial sign-up bonus.

^{**} Earnings in Year 2016 U.S. dollars

Gross Earnings – Lyft vs Uber

	Gross Earnings (before tip) (\$/hr)	Gross Earnings (with tip) (\$/hr)	Gross Earnings (before tip) (\$/mile)	Gross Earnings (with tip) (\$/mile)
Lyft (n=237)	\$14.38	\$16.31	\$0.77	\$0.87
Uber (n=179)	\$14.60	\$14.93	\$0.75	\$0.76
All Trips (n=416)	\$14.48	\$15.69	\$0.76	\$0.82

^{*} Earnings based in Totals ($t_T \& d_T$)

^{**} Earnings in Year 2016 U.S. dollars

Item	Basic Added Cost 1-15hr/week,	Most Drivers 16-49hr/week,	U.S. Federal Standard Mileage Rate (2016)	Average Mileage Rate
0 11	~11k miles/year	~33K miles/year	Kate (2010)	Nate
<u>Ownership</u>				
Depreciation	\$1,320.00	\$3,960.00		
Finance Charge	-	\$500.00		
License, Registration & Tax	-	\$350.00		
Insurance	-	\$1,500.00		
Operating				
Gas	\$1,015.38	\$3,046.15		
Maintenance	\$589.60	\$1,768.80		
Miscellaneous	\$150.00	\$2,000.00	_	
Total	\$3,074.98	\$13,124.95		
\$/mile	\$0.28	\$0.40	0.54*	\$0.41
\$/hr	\$5.34	\$7.60	\$10.31	\$7.75

Assumptions: Car value: \$18,000; Lifetime mileage: 150,000; Work: 50 weeks/year; Gas price: \$2.40/galon (Average in 2015); Gas efficiency: 26 MPG; Maintenance: 5.36 cents/mile; Miscellaneous include car wash & cleaning, mobile device & data fees, parking & traffic violations, risk of crash or injury

DRIVER STUDY

^{* 2016} U.S. Federal Standard Mileage Rate

Item	Basic Added Cost 1-15hr/week, ~11k miles/year	Most Drivers 16-49hr/week, ~33K miles/year	U.S. Federal Standard Mileage Rate (2016)	Average Mileage Rate
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\$/mile	\$0.28	\$0.40	0.54*	\$0.41
\$/hr	\$5.34	\$7.60	\$10.31	\$7.75

Assumptions: Car value: \$18,000; Lifetime mileage: 150,000; Work: 50 weeks/year; Gas price: \$2.40/galon (Average in 2015); Gas efficiency: 26 MPG; Maintenance: 5.36 cents/mile; Miscellaneous include car wash & cleaning, mobile device & data fees, parking & traffic violations, risk of crash or injury

DRIVER STUDY

^{* 2016} U.S. Federal Standard Mileage Rate

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Operating				
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Maintenance	\$589.60	\$1,768.80		
Miscellaneous	\$150.00	\$2,000.00		
Total	\$3,074.98	\$13,124.95		
\$/mile	\$0.28	\$0.40	0.54*	\$0.41
\$/hr	\$5.34	\$7.60	\$10.31	\$7.75

Assumptions: Car value: \$18,000; Lifetime mileage: 150,000; Work: 50 weeks/year; Gas price: \$2.40/galon (Average in 2015); Gas efficiency: 26 MPG; Maintenance: 5.36 cents/mile; Miscellaneous include car wash & cleaning, mobile device & data fees, parking & traffic violations, risk of crash or injury



^{* 2016} U.S. Federal Standard Mileage Rate

	Basic Added Cost	Most Drivers	U.S. Federal	Average
Item	1-15hr/week,	16-49hr/week,	Standard Mileage	Mileage
	~11k miles/year	~33K miles/year	Rate (2016)	Rate
Ownership				_
Depreciation	\$1,320.00	\$3,960.00		
Finance Charge	-	\$500.00		
License, Registration & Tax	-	\$350.00		
Insurance	-	\$1,500.00		
Operating				
Gas	\$1,015.38	\$3,046.15		
Maintenance	\$589.60	\$1,768.80		
Miscellaneous	\$150.00	\$2,000.00	_	
Total	\$3,074.98	\$13,124.95		
\$/mile	\$0.28	\$0.40	0.54*	\$0.41
\$/hr	\$5.34	\$7.60	\$10.31	\$7.75

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DRIVER STUDY

^{* 2016} U.S. Federal Standard Mileage Rate

Net Earnings (Gross minus expenses)

	Net Earnings	S
	Range (Low to High)	Average
\$/hr	\$5.38 - \$10.36	\$7.94
\$/mile	\$0.28 - \$0.54	\$0.41

n=416. Earnings include tips (Year 2016 U.S. dollars)

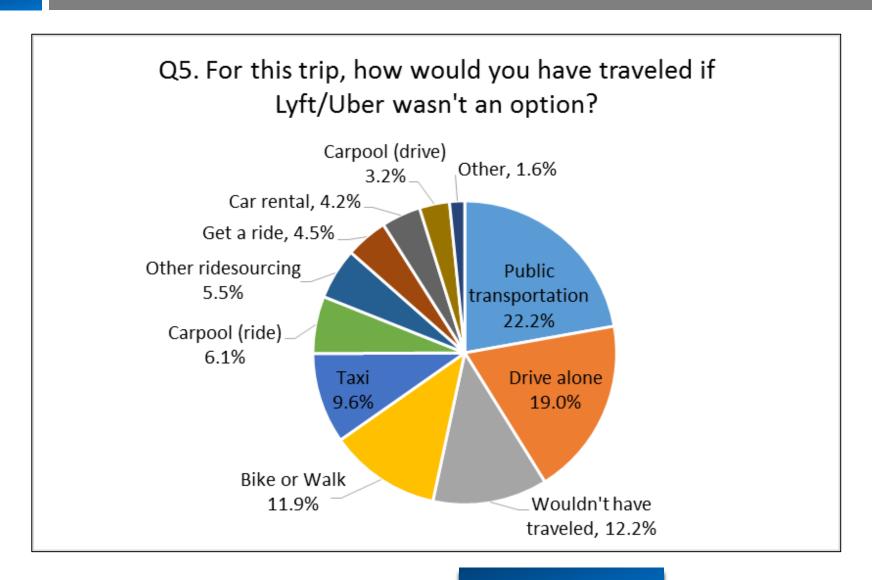
	Net Earnings (before tip) (\$/hr)	Net Earnings (with tip) (\$/hr)	Tip Percent
Lyft (n=237)	\$6.63	\$8.56	29.1%
Uber (n=179)	\$6.85	\$7.18	4.9%

DRIVER STUDY

VII. VMT Study

- ➤ Mode Replacement
- >VMT Impacts

Mode Replacement (Specific Trip)





PMT and **VMT**

- Passenger Miles Traveled (PMT)
- Vehicles Miles Traveled (VMT)

Mode	PMT:VMT	PMT/VMT
Drive (SOV)	1:1	100%
Bike/Walk	1:0	∞
Get a ride	1:2	50%

Ridesourcing?

PMT/VMT, before and after

PMT	VMT Replaced or VMT _{BEFORE}	Ridesourcing VMT or VMT _{AFTER}	•	Ridesourcing Efficiency
	Total (Σ d	1)	PMT VMT _{BEFORE}	$\frac{PMT}{VMT_{AFTER}}$
2,200.03	1,959.58	3,617.68	112.3%	60.8%

VMT STUDY

PMT and **VMT**

- Passenger Miles Traveled (PMT)
- Vehicles Miles Traveled (VMT)

Mode	PMT:VMT	PMT/VMT
Drive (SOV)	1:1	100%
Bike/Walk	1:0	∞
Get a ride	1:2	50%
Ridesourcing	1:1.6	60.8%

PMT/VMT, before and after

PMT	VMT Replaced or VMT _{BEFORE}	Ridesourcing VMT or VMT _{AFTER}	•	Ridesourcing Efficiency
	Total (Σ d	l)	PMT VMT _{BEFORE}	$\frac{PMT}{VMT_{AFTER}}$
2,200.03	1,959.58	3,617.68	112.3%	60.8%



VMT Impact

Mode Replaced n		PMT		-	VMT Replaced or VMT _{BEFORE}		Ridesourcing VMT or VMT _{AFTER}		VMT _{AFTER}	VMT _{AFTER}
		Total (Σd)	Median	Total (Σd)	Median	Total (Σd)	Median	PMT	PMT	VMT _{BEFORE}
Public transportation	69	419.6	3.50	27.2	0.00	768.9	7.54	0.065	1.832	2826.7%
Drive alone	59	661.3	5.17	661.2	5.17	935.5	10.97	1.000	1.415	141.5%
Wouldn't have traveled	38	194.0	3.67	0.0	0.00	370.2	8.00	0.000	1.908	∞
Bike or Walk	37	74.3	1.65	0.0	0.00	195.9	4.95	0.000	2.638	∞
Taxi	30	364.2	5.77	639.5	14.41	568.3	10.74	1.756	1.560	88.9%
Carpool (ride)	19	132.1	3.87	82.2	1.82	227.7	7.64	0.622	1.724	277.1%
Other ridesourcing	17	52.8	3.00	143.3	7.58	143.3	7.58	2.713	2.713	100.0%
Get a ride	14	132.6	5.67	265.3	11.33	140.5	9.75	2.001	1.060	53.0%
Car rental	13	54.6	3.71	54.6	3.50	119.7	6.52	1.000	2.191	219.1%
Carpool (drive)	10	77.1	2.74	77.1	2.74	93.6	5.51	1.000	1.215	121.5%
Other	5	37.5	2.55	9.2	2.28	54.1	6.09	0.244	1.441	589.8%
Total	311	2200.0	3.50	1959.6	1.82	3617.7	7.56	0.891	1.644	184.6%

VMT Impact

		PMT		VMT Rej VMT _B	•	Ridesourc or VM	O	VMT _{BEFORE}	VMT _{AFTER}	VMT _{AFTER}
Mode Replaced	n	Total (Σd)	Median	Total (Σd)		Total (Σd)		PMT	PMT	VMT _{BEFORE}
Public transportation	69	419.6	3.50	27.2	0.00	768.9	7.54	0.065	1.832	2826.7%
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VMT STUDY

VMT Impact

Mode Replaced n		PMT		VMT Rej VMT _B	•	Ridesourcing VMT or VMT _{AFTER}		VMT _{BEFORE}	VMT _{AFTER}	VMT _{AFTER}
		Total (Σd)	Median	Total (Σd)	Median	Total (Σd)	Median	PMT	PMT	VMT _{BEFORE}
Public transportation	69	419.6	3.50	27.2	0.00	768.9	7.54	0.065	1.832	2826.7%
Drive alone	59	661.3	5.17	661.2	5.17	935.5	10.97	1.000	1.415	141.5%
Wouldn't have traveled	38	194.0	3.67	0.0	0.00	370.2	8.00	0.000	1.908	∞
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Taxi	30	364.2	5.77	639.5	14.41	568.3	10.74	1.756	1.560	88.9%
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Car rental	13	54.6	3.71	54.6	3.50	119.7	6.52	1.000	2.191	219.1%
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Other	5	37.5	2.55	9.2	2.28	54.1	6.09	0.244	1.441	589.8%
Total	311	2200.0	3.50	1959.6	1.82	3617.7	7.56	0.891	1.644	184.6%

Legend:

Worst VMT
Better VMT

VIII. Parking

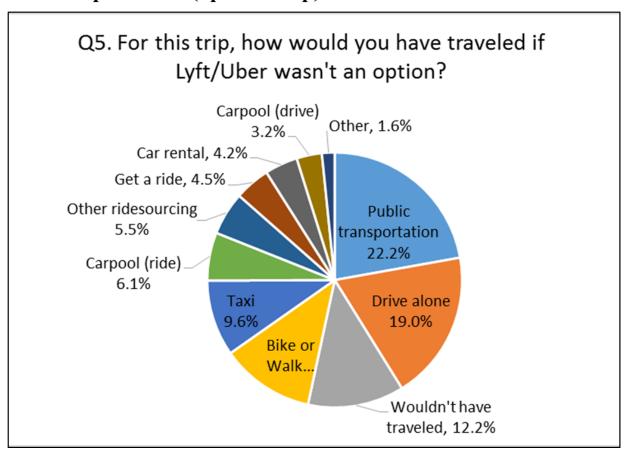
- Parking Demand
- Locations, Trip Purpose, Transit Stations
- Parking as a stated reason

Each theme was explored for:

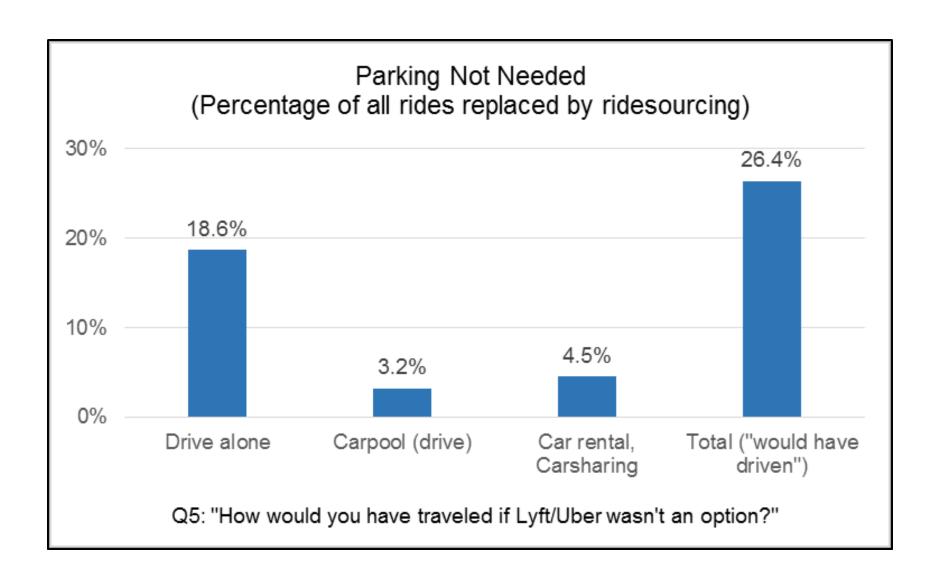
- > Specific trip
- General use

Parking Demand (Specific Trip)

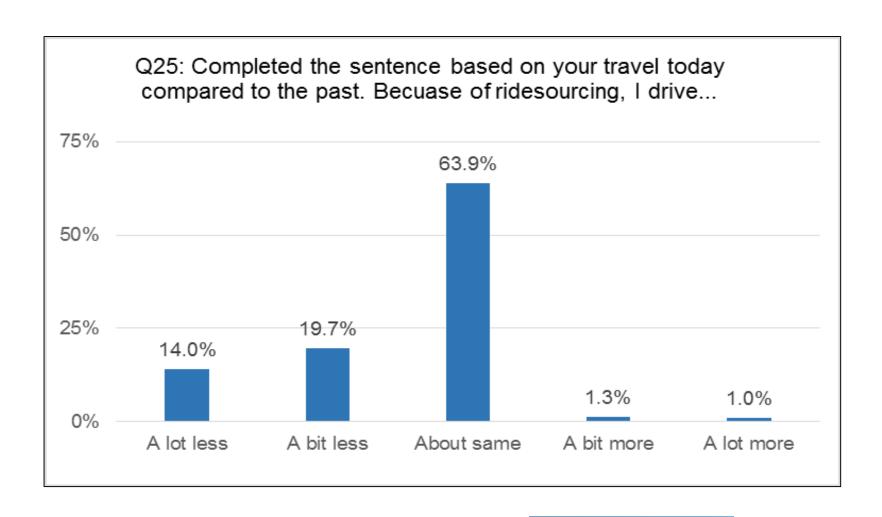
Mode Replacement (Specific Trip)







Parking Demand (General Use)



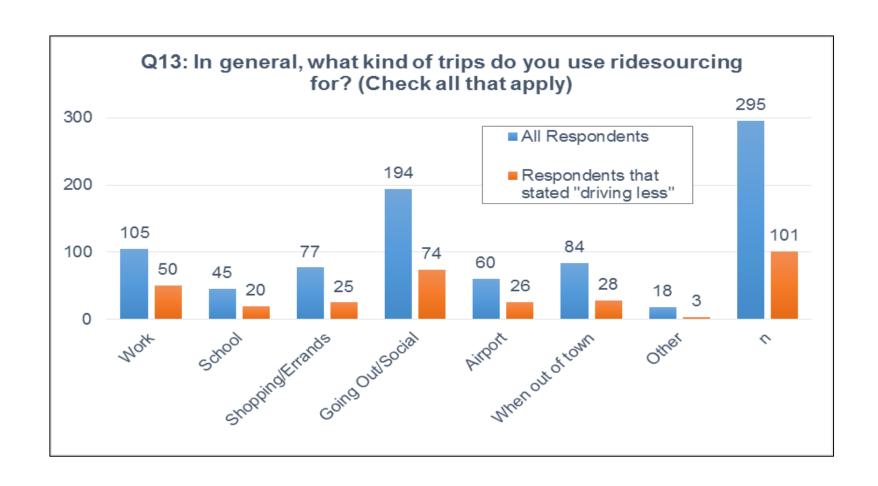


Parking: Locations (Specific Trip)

O-D Matrix (Driving Trips Replaced)

DESTINATION	Home	Work	School	Shopping/	Going Out/	Airport	Hotel/	Family/	Other	Totals
ORIGIN				Errands	Social		Airbnb	Friend		
Home	0	5	1	1	19	13	0	1	2	42
Work	2	2	1	0	0	0	4	0	1	10
School	1	0	0	0	0	0	0	1	0	2
Shopping/Errands	1	0	0	0	0	0	0	0	0	1
Going Out/Social	8	0	0	0	2	0	0	2	1	13
Airport	0	0	0	0	0	0	1	0	0	1
Hotel/Airbnb	0	2	0	0	0	0	0	0	2	4
Family/Friend	2	0	0	0	1	0	1	0	0	4
Other	1	1	0	0	0	1	2	0	0	5
Totals	15	10	2	1	22	14	8	4	6	82

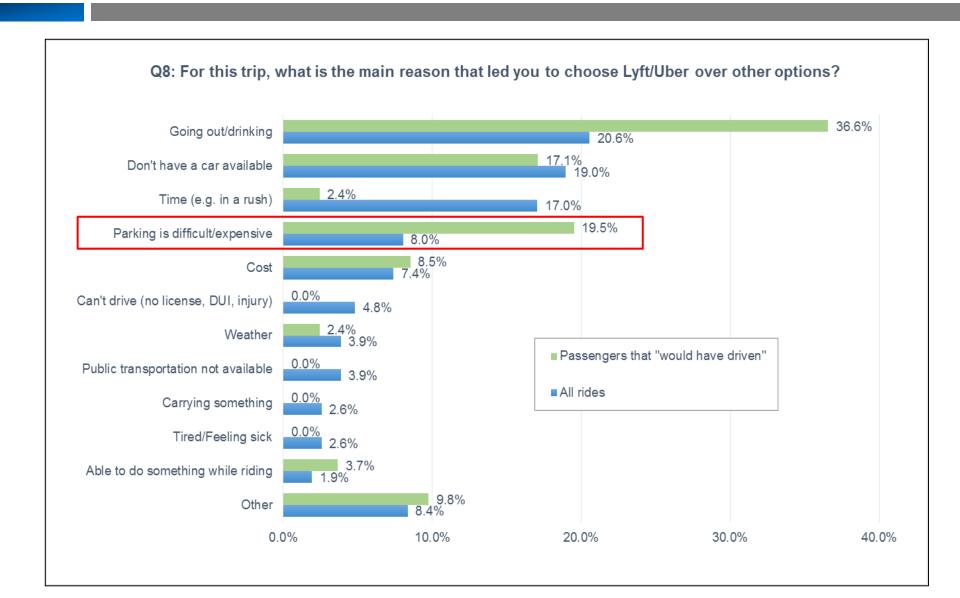
Trip Purpose (General Use)



Parking: Connectivity to Transit

Q9. Ride connecting with other mode (n=311)						
No	294	94.5%				
Yes	17	5.5%				
If yes, number of rides replacing driving and connecting to transit	3	1.0%				
Q22. Have you ever connected with other mode? (n=293)						
No	233	79.5%				
Yes	60	20.5%				
If yes, number of passenger that stated driving less and public transportation (e.g. bus, rail) as the connection mode	21	7.2%				

Stated reason (Specific Trip)



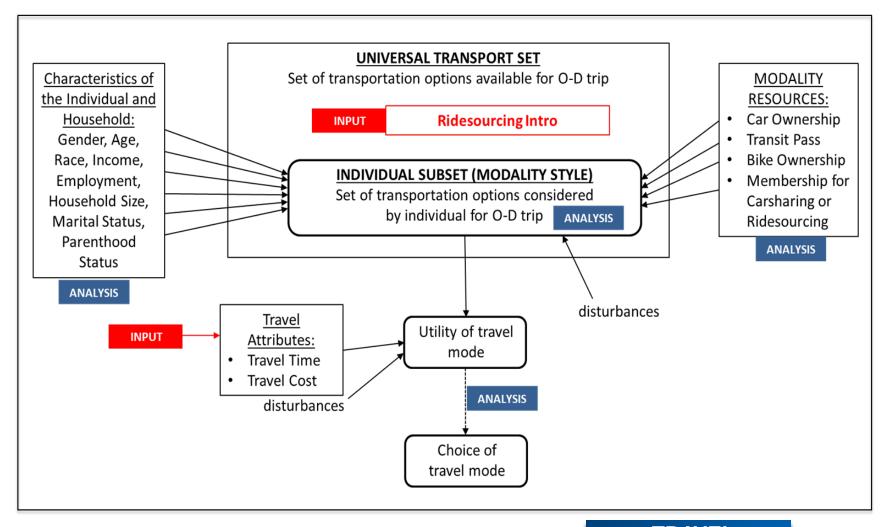


IX. Travel Behavior

- > Travel Demand Framework
- Mode Frequency
- Travel Behavior Changes
- > Trip Purpose
- > Reasons
- Modality Style

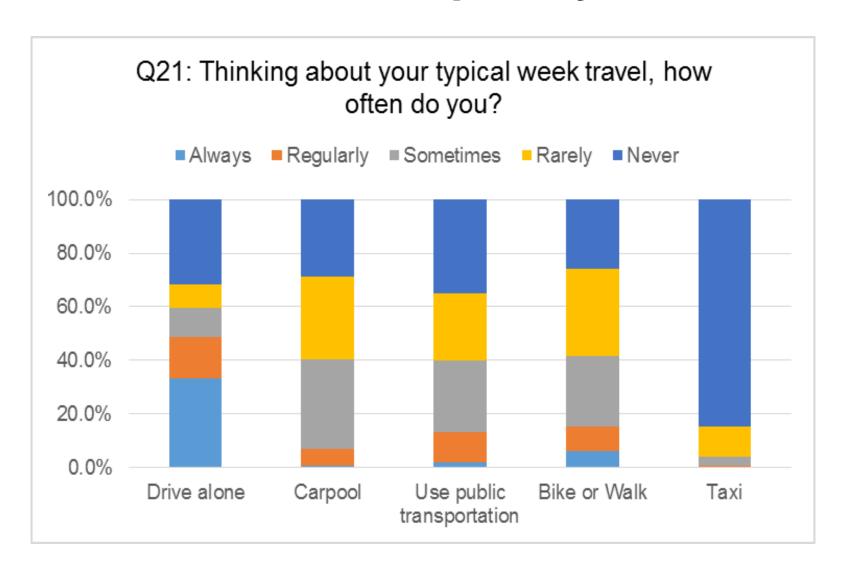


Travel Demand Framework



TRAVEL BEHAVIOR

Mode Frequency





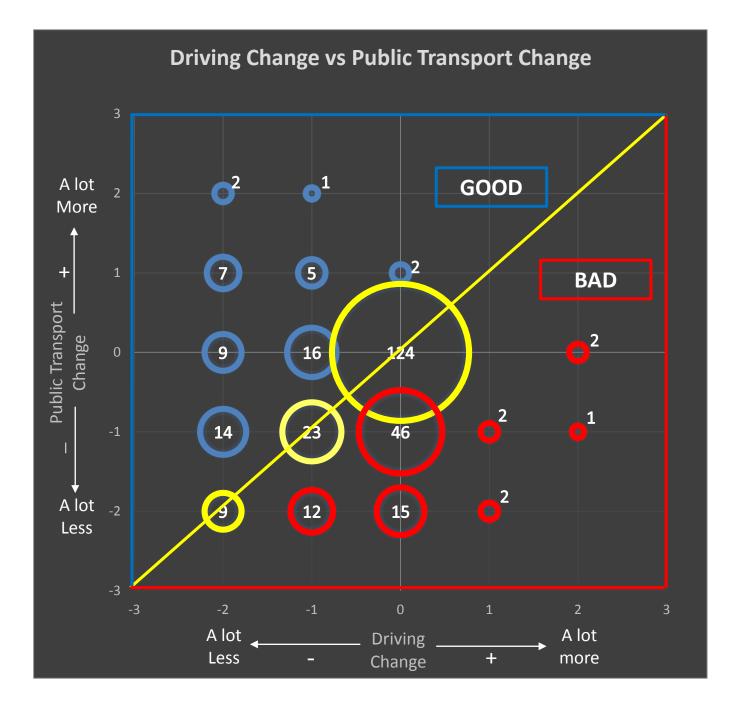
Changes

25. For the next few questions, complete the sentence based on your travel today compared to the past

	A lot less	A bit less	About same	A bit more	A lot more
Because of ridesourcing, I go to places	3	2	144	121	18
	1.0%	0.7%	50.0%	42.0%	6.3%
Because of ridesourcing, I drive	41	57	182	4	3
	14.3%	19.9%	63.4%	1.4%	1.0%
Because of ridesourcing, I use public transport	38	86	146	14	3
	13.2%	30.0%	50.9%	4.9%	1.0%
Because of ridesourcing, I bike or walk	10	77	187	7	7
	3.5%	26.7%	64.9%	2.4%	2.4%
Because of ridesourcing, I take taxis	88	25	165	1	0
	31.5%	9.0%	59.1%	0.4%	0.0%







Driving Frequency and Trip Purpose



Driving Frequency

TRAVEL BEHAVIOR

Driving Frequency and Trip Purpose



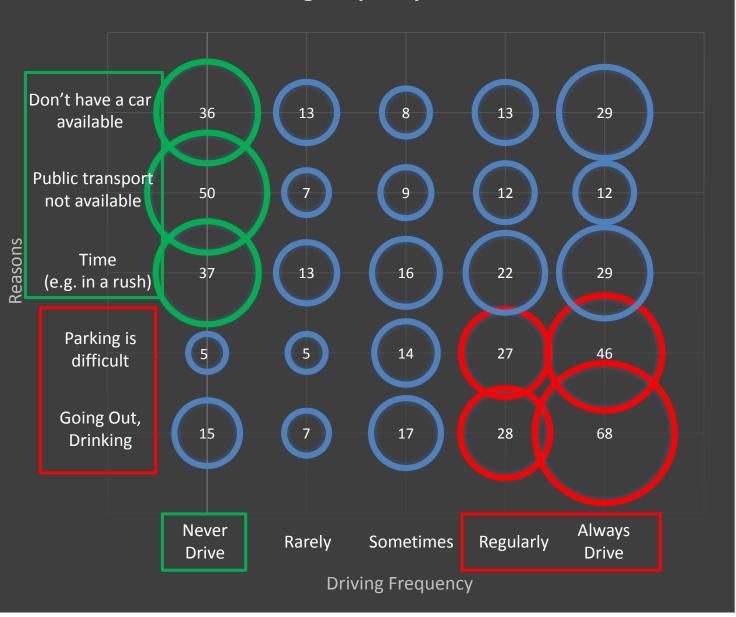
Driving Frequency

TRAVEL BEHAVIOR

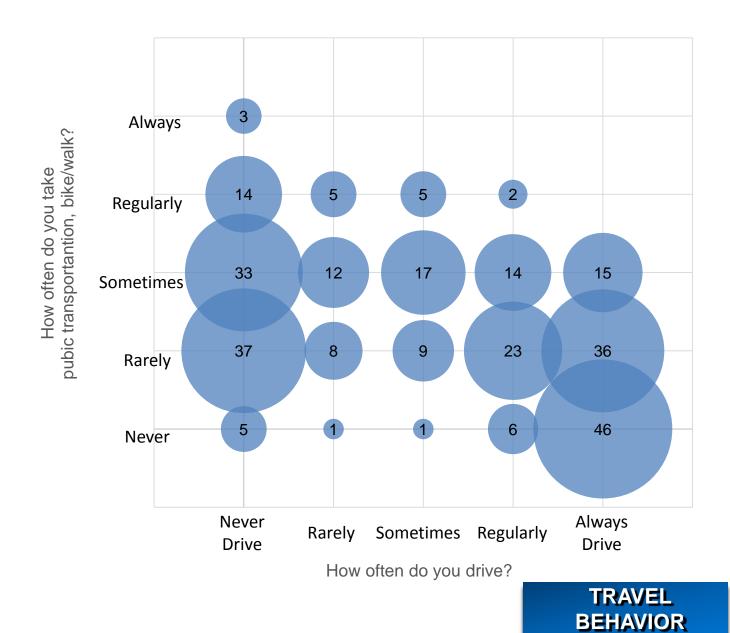
Driving Frequency and Reasons



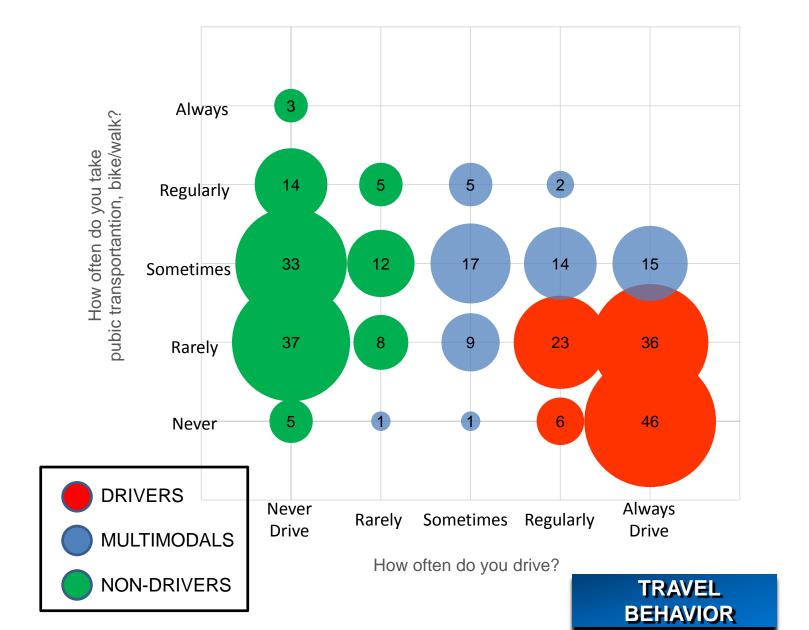
Driving Frequency and Reasons



"Drive Frequency" versus "Public Transportation + Bike/Walk Frequency"



Modality Style



- Driver Perspective
- > VMT
- Parking
- > Travel Behavior

RIDESOURCING TIMES AND DISTANCES

- Overall efficiency rate for the study is 39.3% based on time, and 59.2% based on distance
- In terms of distance, drivers have to travel 69 extra miles in dead-heading for every 100 miles with a passenger

RIDESOURCING EARNINGS

The gross earnings is \$15.69/hour but discounting expenses is less than minimum wage, with an average of **\$7.94/hr** (tips included).

VMT IMPACT

- > Ridesourcing provides more mobility:
 - 12.2% of passengers stated that they "wouldn't have traveled"
- ➤ But PMT/VMT efficiency goes from 112.3% to 60.8%
- ➤ Current ridesourcing VMT is <u>185%</u> what would have been before, which has significant implications for our cities in terms of congestion and environmental concerns

VMT IMPACT

Based on Lyft/Uber current rate of 1 billion rides per year in the U.S. and assuming the results hold true for the country:

Lyft and Uber rides per year in the U.S.	1,000,000,000.00		
$t_{T mean} = (\Sigma d)/ride (Table IV.1)$	11.90		
VMT _{AFTER} = Rides per year * 11.90	11,900,707,268.24		
VMT _{AFTER} /VMT _{BEFORE} (Table V.3)	1.85		
$VMT_{BEFORE} = VMT_{AFT ER} / 1.85$	6,446,228,741.23		
$VMT_{EXTRA} = VMT_{AFTER} - VMT_{BEFORE}$	5,454,478,527.02		

Estimated VMT impact from Lyft/Uber is around 5.5 billion extra miles per year in the U.S.



PARKING

High potential to decrease car dependency

- ➤ Ridesourcing is replacing driving modes, reducing the need for parking
- Parking difficulty/expense is one of the main reasons for passengers to use ridesourcing instead of driving.

TRAVEL BEHAVIOR

- > Three common groups of ridesoucing:
 - 1. Drivers
 - 2. Multimodals
 - 3. Non-drivers
 - Drivers become bi-modal based on trip purpose
- ➤ For typical drivers, ridesourcing is mostly replacing social trips (e.g. go out), to/from airport, and when out of town
- > For typical non-drivers, it's replacing work/school trips

XI. Summary Conclusions

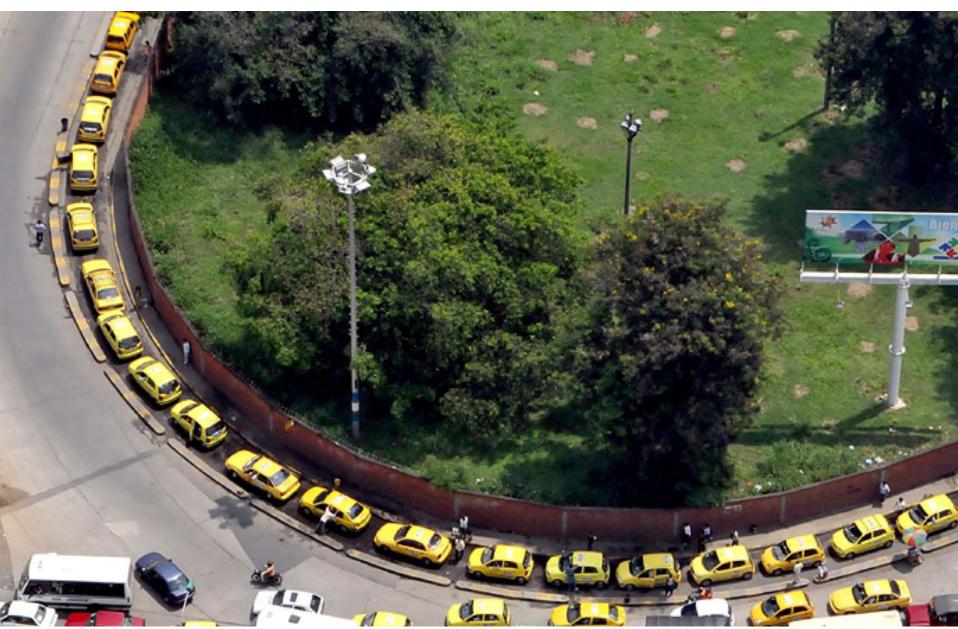
- Opportunities and Barriers
- Policy Recommendations
- Future Applications
- Future Research

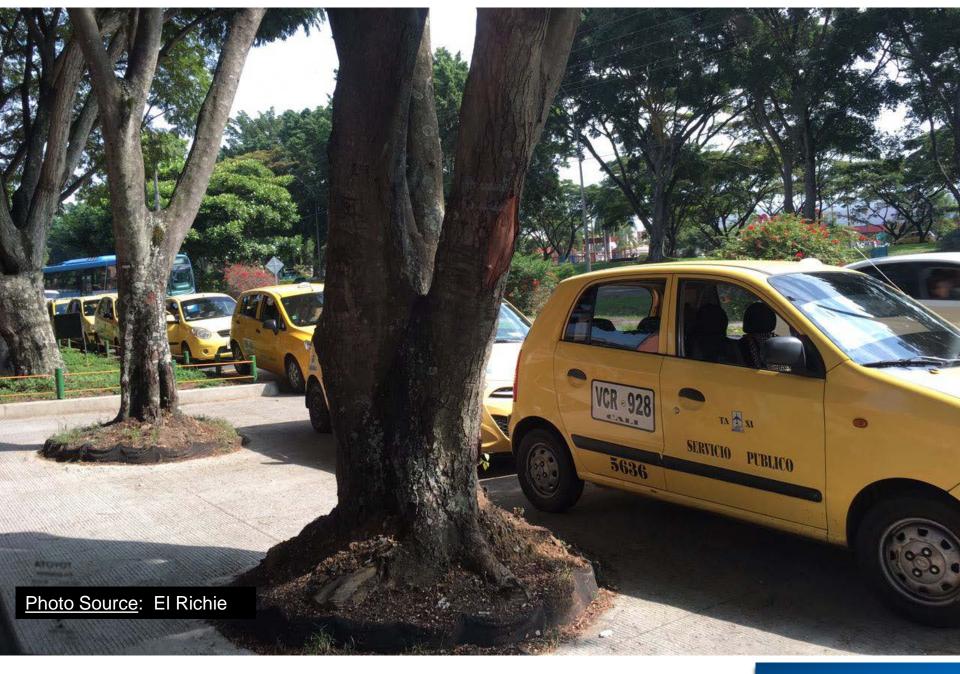
Limitations

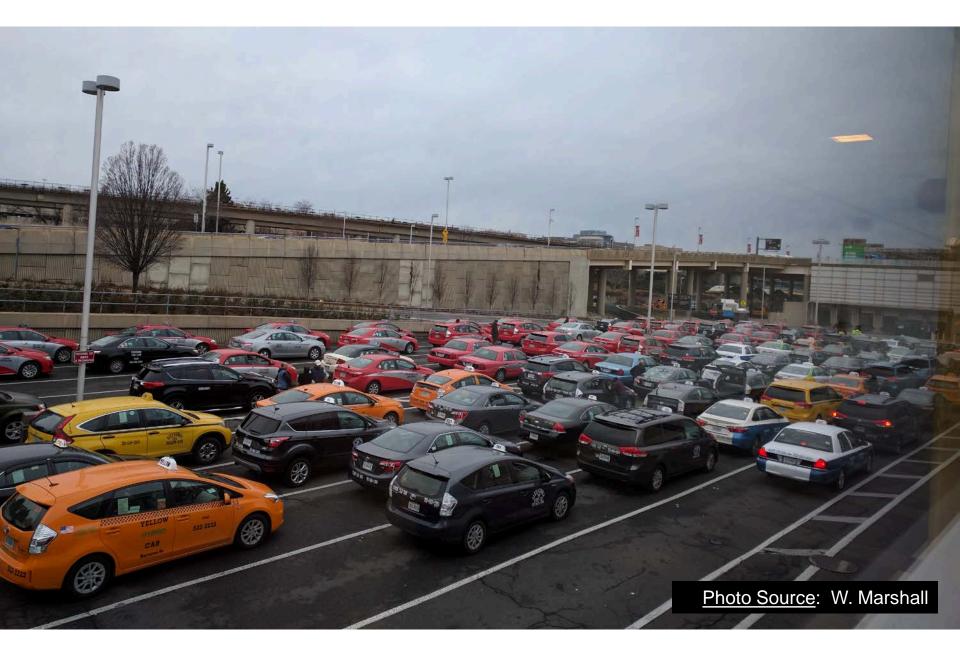
This study doesn't come without limitations:

- > Trip sample size
- Denver metro area
- Driver strategy

Photo Source: ElPais.com.co









WE NEED DATA

- Cities and agencies need data
 - REAL, USEFUL DATA

Highly touted Boston-Uber partnership has not lived up to hype so far



-Robert Galbraith / Reuters

By Adam Vaccaro June 16, 2016

It was hailed as a milestone for both Boston and Uber in January 2015, when the increasingly ubiquitous ride-for-hire service agreed to share data with City Hall on trips conducted in the city.

ALEX DAVIES TRANSPORTATION 01.08.17 4:00 PM

UBER'S MILDLY HELPFUL DATA TOOL COULD HELP CITIES FIX STREETS





POLICY DECISIONS

- Cities and agencies need data
 - REAL, USEFUL DATA
 - BE CAREFUL WITH INFRASTRUCTURE DECISIONS (TRANSIT)

LYFT AND UBER

- Uber and Lyft are great and could be part of the solution for better transportation systems
 - LYFTLINE, UBERPOOL
 - DESTINATION FILTER
 - PARTNERSHIPS
 - CAR-OWNERSHIP
 - EQUITY (PASSENGERS & DRIVERS)



- CAR INDUSTRY
- TAXI INDUSTRY
- STAKEHOLDERS





THE FUTURE

- Autonomous Vehicles
- Infrastructure Changes
- > Transportation as a service



ACADEMIA AND RESEARCH

- NEED MORE EMPIRICAL STUDIES
- BETTER RESEARCH METHODS
- BETTER IMPLEMENTATION IN MODELS
 - Alonso-Mora, J., Samaranayake, S., Wallar, A., Frazzoli, E., & Rus, D. (2017).
 On-demand high-capacity ride-sharing via dynamic trip-vehicle assignment.
 Proceedings of the National Academy of Sciences.

FUTURE STUDIES

- Equity Study using the Uber/Lyft API
 - Hughes & McKenzie (2016): Equity study in Seattle
 - Yanbo Ge et al. (2016): Discrimination study in Seattle and Boston. African American sounding names.
- Deeper analysis of travel demand models
 - Demographics
 - Modality resources
 - Modality Style
 - Mode Choice

FUTURE STUDIES

- > More interesting things on the data
 - Parking (extra time and cost)
 - Passenger side
 - Geographical Variations (e.g. density, urban-suburban)
 - Uber/Lyft Estimated Arrival Time (EAT)
 - LyftLine/UberPool user characteristics
 - Value of Transit increase
 - Value of Time
- ➤ Austin, Texas

Impacts of Ridesourcing – Lyft and Uber – on Transportation including VMT, Mode Replacement, Parking, Equity, and Travel Behavior



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Doctoral Dissertation Defense January 19, 2017