Why Do People Take E-scooter Trips? Big Data and Unsupervised Machine Learning Insights on Temporal and Spatial Usage Patterns

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Background

- 86 million e-scooter trips in 2019, which is over a 120% increase in trips compared to 2018 (NACTO, 2020)
- Emerging and disruptive technology
- Policy-level questions
 - Safety
 - Sustainability
 - Parking
 - Mode shift
- When, where, and why people use e-scooters



Photo credit: https://www.irishtimes.com/business/technology/tech-review-electric-scooter-is-fun-portable-but-not-that-light-1.3951480



Research approach

- Survey-based and/or observational studies
 - Pilot project evaluation: City of Austin (2019)
- Micromobility trip data analysis
 - Descriptive statistics: McKenzie (2019) and Liu, Seeder, and Li (2019)
 - Discrete choice model: Reck, Guidon, Haitao, and Axhausen (2020)
 - Regression models: Bai and Jiao (2020) and Younes, Zou, Wu, and Baiocchi (2020)
 - Spatial regression models: Caspi, Smart, and Noland (2020)

Research objective:

To identify e-scooter usage patterns using spatiotemporal analysis of micromobility data



Data

- Study Area: Nashville, Tennessee
- **Trip data source:** Shared Use Mobility Device (SUMD)





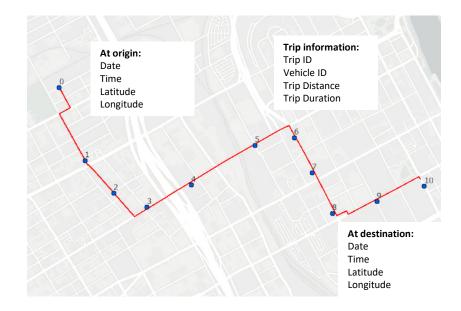


Trip Summary

Device Availability

Parking Compliance

- Study period: September 1, 2018 to August 31, 2019
- Number of trips: 1,072,430

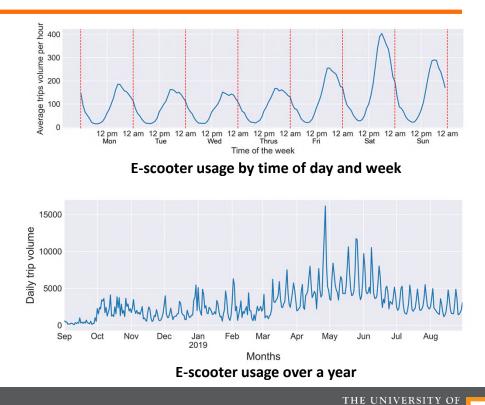


Sample trip summary data from Nashville's SUMD



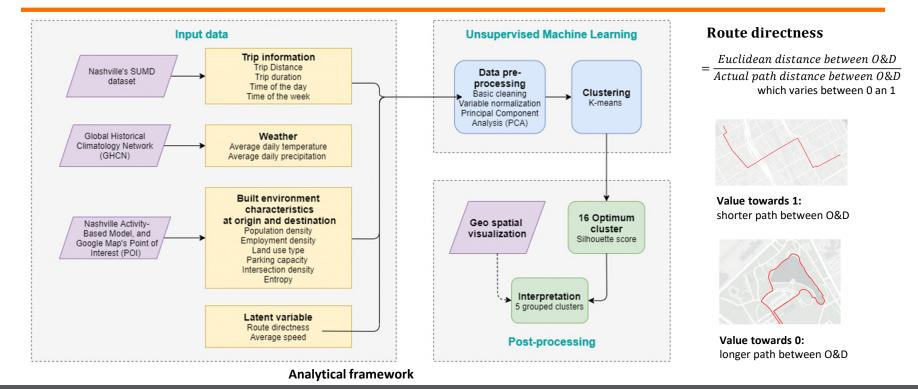
Basic usage information

- One daily peak in the evening
- Weekend (Friday-Sunday) has higher usage than weekdays
- Saturday evening has highest escooter usage in the week
- Summer months have higher escooter usage
- Average trip distance: 0.72 miles
- Average trip duration: 17 minutes



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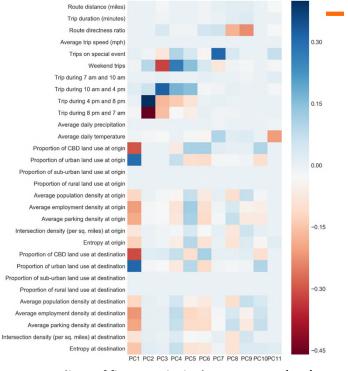
Methodology





Factors influencing e-scooter trips

- Principal Component Analysis (PCA) for dimension reduction
- The color scale in the figure indicates the loadings, which is positive or negative correlation between PCs and variable
- Strong correlation between start time, land use variables, temperature, and route directness



Loadings of first 11 Principal Components (PCs)



Usage-grouped clusters

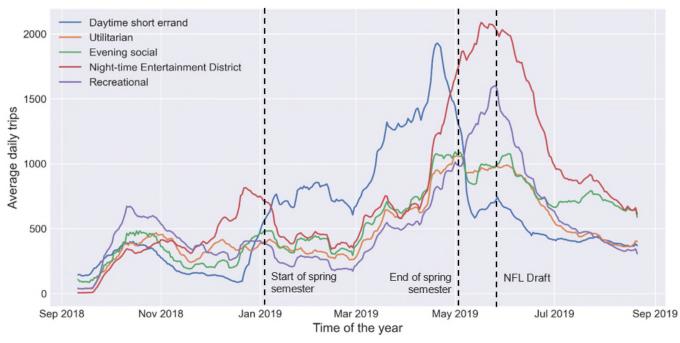
We combined 16 K-means clusters into five usage-grouped clusters based on trip attributes like start time, land use of trip origin and destination, and route directness ratio

	Daytime short Errand	Utilitarian	Evening social	Night-time Entertainment District	Recreational
Percentage of trips by count	20.4	16.8	18.9	26.1	17.8
Percentage of trips by Vehicle-Miles Travelled	13.3	20.5	20.9	25.4	20.0
Percentage of trips by travel duration	19.0	16.9	20.5	25.2	18.5
General Attributes (mean)					
Route distance (miles)	0.47	0.88	0.80	0.71	0.81
Trip duration (minutes)	15.2	16.5	17.8	15.9	17.0
Route directness ratio	0.58	0.64	0.53	0.54	0.48
Percentage of weekend trips within the cluster	17	4	56	35	48

Note: Red color indicates lower values while blue color indicates higher values among clusters



Temporal Distribution

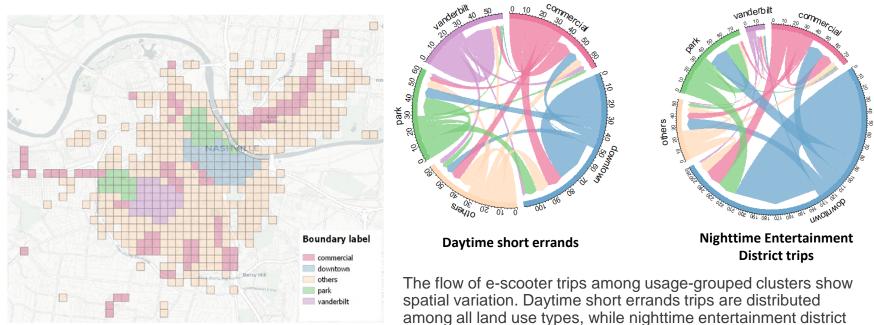


Trip distribution of usage-grouped clusters over months

- Daily average trips increased during the summer months for all usage-grouped clusters
- E-scooter usage was influenced by events in Nashville
- Overall usage of all usage-grouped clusters increased over a year



Spatial Distribution



Boundary for chord diagram

trips are predominantly in downtown area.



Key takeaways

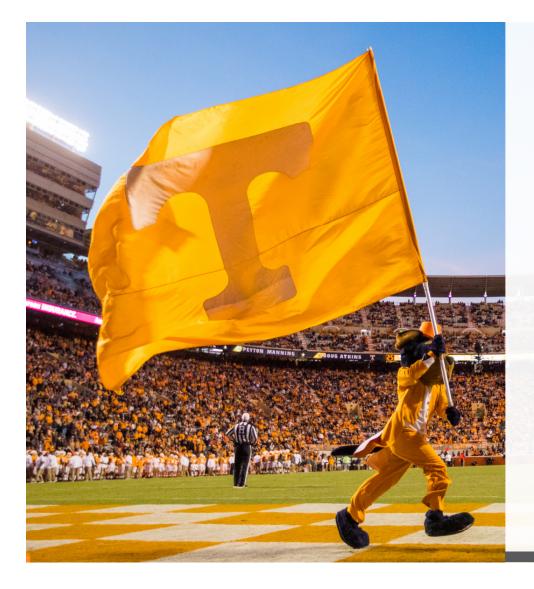
- We identified five distinct purpose-oriented e-scooter usage patterns in Nashville
- ³/₄ of e-scooter trips were NOT tourist-oriented nighttime entertainment trips
- E-scooter usage pattern doesn't resemble commuting pattern in Nashville
- The number of e-scooter trips of each five distinct patterns increased over a year
- E-scooter ridership increases during weekends and summer months in general
- Future research
 - O Validation using survey results; semi-supervised learning methods
 - O Include built environment information at route level for clustering
 - O Comparative analysis of e-scooter usage patterns from other cities



References

- Bai, S., & Jiao, J. (2020). Dockless E-scooter usage patterns and urban built environments: a comparison study of Austin, TX, and Minneapolis, MN. *Travel behaviour and society, 20*, 264-272.
- Caspi, O., Smart, M. J., & Noland, R. B. (2020). Spatial associations of dockless shared e-scooter usage. *Transportation Research Part D: Transport and Environment, 86*, 102396.
- City of Austin. (2019). Dockless Mobility Community Survey Report. Retrieved from <u>https://austintexas.gov/sites/default/files/files/Transportation/Dockless_Mobility_Community_Survey_Report_2-28-19.pdf</u>
- Liu, M., Seeder, S., & Li, H. (2019). Analysis of E-Scooter trips and their temporal usage patterns. Institute of Transportation Engineers. ITE Journal, 89(6), 44-49.
- McKenzie, G. (2019). Spatiotemporal comparative analysis of scooter-share and bike-share usage patterns in Washington, DC. *Journal of Transport Geography, 78*, 19-28.
- NACTO. (2020). *Shared Micromobility in the US: 2019*. Retrieved from <u>https://nacto.org/wp-content/uploads/2020/08/2020bikesharesnapshot.pdf</u>
- Younes, H., Zou, Z., Wu, J., & Baiocchi, G. (2020). Comparing the Temporal Determinants of Dockless Scooter-share and Station-based Bike-share in Washington, DC. *Transportation Research Part A: Policy and Practice*, *134*, 308-320.





Questions?

Contact information: Nitesh Shah <u>nshah12@vols.utk.edu</u> <u>https://www.linkedin.com/in/niteshrajshah/</u>

