

Micromobility in Smart
Cities: A Closer Look at
Shared Dockless E-Scooters
via Big Social Data

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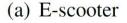
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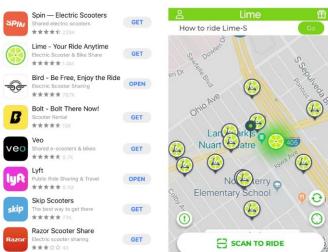
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Background

- Pros: affordability, easy accessibility via an app, and zero emissions.
- **Cons**: traffic rules, public safety, and parking regulations.
- 11% of Paris residents reported using e-scooters either frequently or from time to time [1].
- According to NACTO[2], people took 38.5 million trips on shared escooters in 2018.







(b) E-scooter apps

(c) App UI

Motivation

- A missing research perspective from the riders' comments shared via smartphones, especially via social media apps.
- Social media is a good data source to investigate the e-scooter usages.
 - ☐ **Diversity**: multimodal social data (text, image, timestamps, GPS, emoji, etc.) enables detailed profiles of e-scooter sharing services in various aspects.
 - Scalability: performing a large-scale study of micromobility with social media data is flexible and effortless.
 - ☐ **Transparency**: public social media platforms (e.g., Twitter) addressing potential concerns of research non-reproducibility.

Methodology

Use Twitter to collect data Free APIs to collect tweets of interest The fourth most active social networking platform Users expect the data they post on Twitter to be publicly available Big data analytics Spatio-temporal Visualization Topic discovery - latent Dirichlet allocation (LDA) Optical character recognition (OCR) Logo detection and recognition Sentiment analysis

Data Collection & Bot Detection

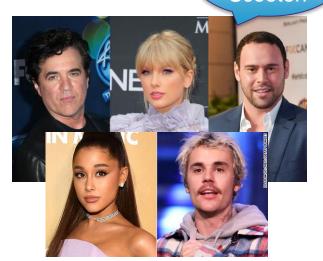
- Twitter's Streaming APIs with keywords of *scooter* and emoji
- ji 🛵

- 5.8 million tweets generated by 2.7 million users
- 178,048 different image URLs (33,851 expired)
- October 6, 2018 to March 14, 2020
- Remove bots [3]
 - Those who posted more than 525 scooter-tagged tweets, i.e., more than one such tweets per day, during the data collection period
 - ☐ Those who posted over 100 scooter-tagged tweets in total and the top three frequent posting intervals covered at least their 90% tweets.

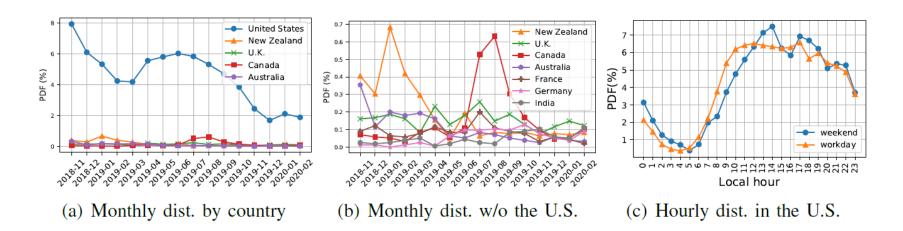
Data Cleanup

- Word co-occurrence analysis
 - Taylor, Swift, Justin, Bieber, Scott, Samuel, Braun, Ariana, Grande, and Borchetta
 - ☐ 1,541,815 related tweets were deleted
- Further reduce false positives
 - ☐ Tweets containing the word of **Share**
 - ☐ Tweets containing shared e-scooter brands including *Bird, Lime, Spin, Bolt, gruv, Lyft, Sherpa, VeoRide, Taxify, Jump, RazorUSA, Scoot Networks*, and *Skip*.

Call me Scooter!

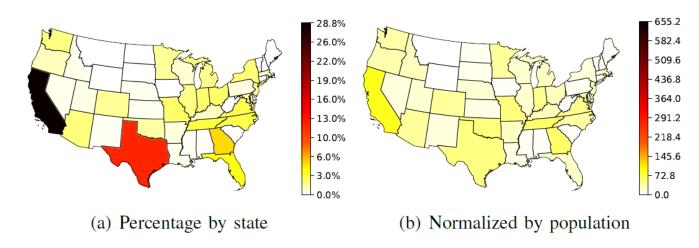


Temporal Distribution



- The United States accounts for more than 82% of all collected tweets.
- Usage peaks in summer and drops in winter.
- New Zealand & Australia ↓, Germany & India ↑, and United Kingdom
 & France →.
- The most active time during weekdays (weekends) is between 10:00 am to 5:00 pm (12:00 pm and 7:00 pm).

Geospatial Distribution



- California (28.8%) and Texas (11.7%) account for more than 40% of all tweets.
- Six (CA, TX, GA, FL, NC, OH) out of the top ten states with the highest percentages were among the ten most populous states.
- After normalizing by state population, we obtained a relatively smooth distribution.

Topic Discovery

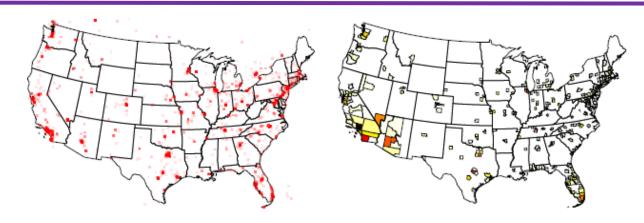
- Data preprocessing
 - Each tweet content is a single document.
 - Corrected misspelled words, removed stop words, tokenized words, and lemmatized words.
 - ☐ Performed the term frequency inverse document frequency (TF-IDF).
- Latent Dirichlet Allocation (LDA)
 - # of topics = 12 achieving the highest coherence score
 - ☐ We clustered the 12 topics into four categories
 - Deployment
 - Stakeholder
 - Operation
 - Emotion

Topic Discovery

TABLE I: The extracted topics using the LDA topic model

	Stakeholder			Emotion		Deployment			Operation				
Rank	Rider	Gig Worker	Company	Positive	Negative	Transport.	City	Regulation	Parking	Transaction	Injury	Product	
1	kid	gig	electr	ride	ride	bike	santa	citi	sidewalk	арр	injuri	balanc	
2	kick	worker	startup	around	one	share	monica	program	park	ride	fire	self	
3	adjust	built	compani	fun	got	transport	st	compani	bike	charg	rider	electr	
4	wheel	contractor	san	day	saw	car	loui	pilot	peopl	use	accid	bo	
5	height	gen	market	love	shit	citi	montreal	electr	road	free	recal	skateboard	
6	amp	economi	via	downtown	time	transit	paul	new	lane	unlock	caus	smart	
7	child	lemon	new	san	fuck	trip	toronto	council	use	code	man	board	
8	light	kmh	launch	time	guy	use	joe	bike	block	tri	injur	inch	
9	adult	stabl	ford	rode	hit	mobil	canal	share	ride	one	via	fold	
10	boy	libbi	tech	one	someon	electr	german	come	pedestrian	minut	report	bluetooth	

Deployment in Cities



- (a) Exact (lat, lon) coordinates
- (b) Geospatial distr. by county
- 3359 exact GPS coordinates located in 579 cities.
- Most extensive scooter deployments occurred in large cities at East Coast and West Coast, and other metropolises.
- Three cities (Los Angeles, San Francisco, San Diego) in California contributed more than 15.6% of all GPS-tagged tweet.

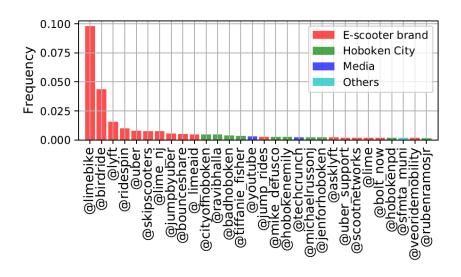
Policies and Regulations

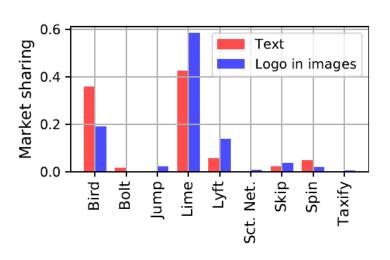
- We surveyed policies on shared e-scooters in ten cities generating the most GPS data.
- Common rules
 - Self-protection requirements, e.g., wearing helmets
 - Riding behaviors, e.g., no phone usage while riding
 - Traffic restrictions, e.g., reasonable max speeds from 15 mph to 30 mph
 - ☐ Parking rules, e.g., not blocking sidewalks

Rider Profiles

- A great gender gap in shared e-scooters with 34.86% identified as female and 65.14% as male, being consistent with a study [4] by Portland State University, which reported 34% riders identified as a woman, 64% as a man, and 2% as transgender or non-binary.
- Only 4.17% riders were recognized as kids.
- 83.51% users did not wear a helmet when riding, which might be one of the most common risky behaviors.

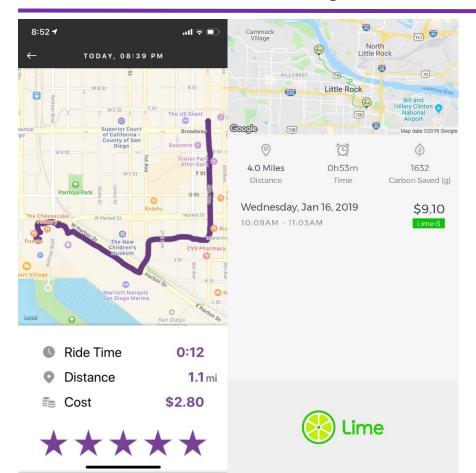
Scooter Companies





- All the top 10 most frequent mentions are e-scooter brands.
- The top 3 of market shares based on tweet text and recognized logos, i.e., Lime, Bird, and Lyft, agree with the top @mentioned accounts

Transaction Analysis



- Optical Character Recognition (OCR) to extract the trip information.
- Regular expressions to extract the payment and trip duration.
- Median payment and median duration were \$3.8 and 15.0 minutes.
- Average payment and duration were \$8.9 and 44.3 minutes with standard deviations of 14.6 and 95.7.

Self-reported Injury

TABLE II: Self-reported injury categories

	Head (22.88%)				Trunk & Hands (27.45%)			Leg & Feet (49.67%)					
Chin 5.23%	•			Others 11.76 %		_						_	Others 10.46%

- Knee protective gears are required because of their highest injury frequency in all body parts.
- **Fingerless gloves** can be a good choice for riders to avoid hand bruises, the second most common injuries, and enable touching smartphone screens at the same time.
- A helmet with chin protection is a must because over half of the reported chin wounds were very serious.

Parking Behaviors

- 37.39% e-scooters were docked at right places properly, and the rest **62.61% were in wrong places**.
- Among those e-scooters parked improperly, 34.78% were parked in the middle of sidewalks; 4.78% were placed indoors; 5.65% were vandalized; and 17.39% were parked in other wrong areas.
- Blocking sidewalks was the most common improper e-scooter parking behavior.

Emotion Analysis

- Emoji: 62.6% positive, 16.2% neutral, and 21.2% negative
 - Positive



Negative









• **Emoticon**: the positive emotions override the negative ones

TABLE III: Emoticons by categories (numbers represent frequency)

Positive |:) 490 :3 467 ;) 137 :-) 87 8) 79 :p 35 ;-) 33 :b 16 *) 9 :* 8 =) 7 :^) 6 x-p 6 :-)) 4 :] 3

Negative |:/ 325 :(321 :\ 245 :-(22 :o 18 :-/ 14 :\$ 10 :c 9 8-0 6 :[3 :{ 2 =\ 2 :-c 0 :-; 0 :; 0

Conclusion

- Leveraged massive volumes of heterogeneous Twitter data, including text, @mentions, GPS data, general photos, screenshots of e-scooter apps, emojis, and emoticons, to study e-scooter ridesharing services on a large scale.
- After performing a comprehensive data preprocessing to remove noise and reduce false positives, we summarized 12 popular topics using the LDA topic model.
- For each of the extracted topics, we reported the profound insights and patterns, such as the popularity in different cities, the gender gap of riders, e-scooter market shares, transaction information, injury types, parking behaviors, and emotions from the public.

Reference

- ODOXA, New modes of travel are disrupting transport in Ile-de-France, 2019. [Online].
 Available: https://shorturl.at/nBLU0
- 2. N. A. of City Transportation, 84 Million Trips Taken on Shared Bikes and Scooters Across the U.S. in 2018, 2020. [Online]. Available: https://nacto.org/shared-micromobility-2018/
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- 4. J. Dill, The e-scooter gender gap, 2020. [Online]. Available: https://jenniferdill.net/2019/02/01/the-e-scooter-gender-gap/



Thank you! Q&A