# CatCharger: Deploying Wireless Charging Lanes in a Metropolitan Road Network through Categorization and Clustering of Vehicle Traffic

Li Yan, Haiying Shen, Juanjuan Zhao, Chengzhong Xu, Feng Luo and Chenxi Qiu

IEEE INFOCOM Atlanta, US May 2017

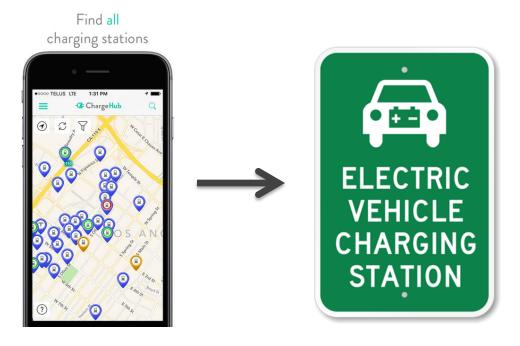




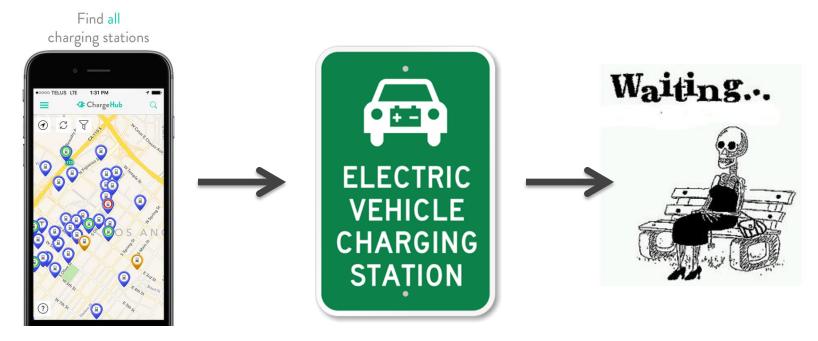
Find all charging stations



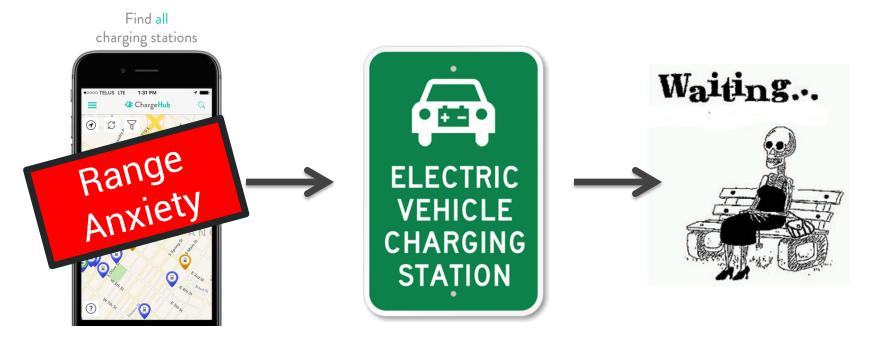




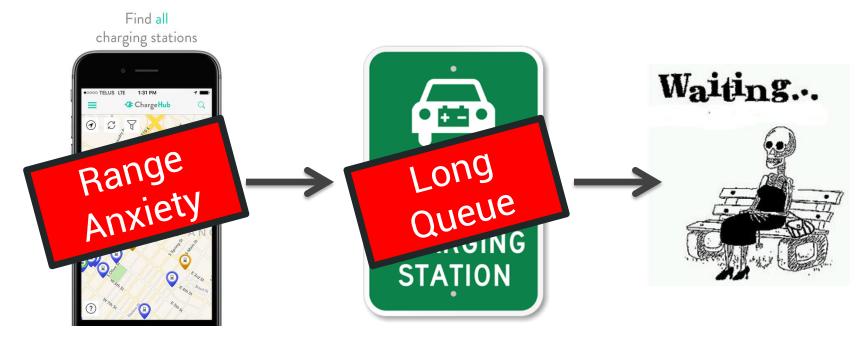




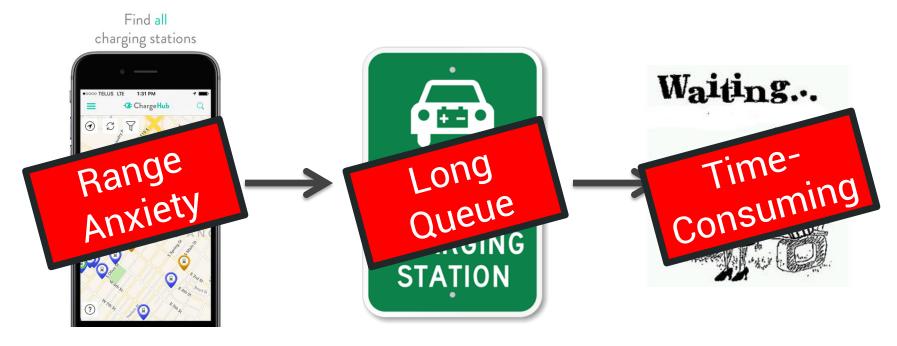




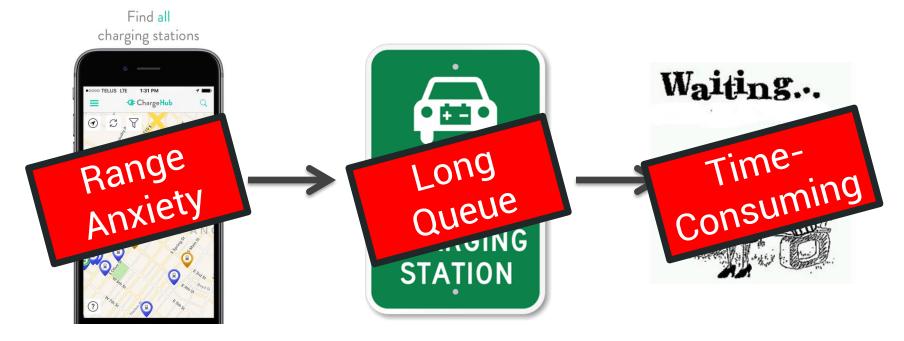








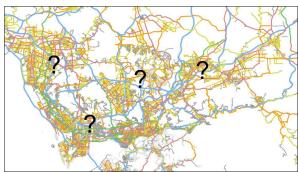




Fail to maintain State-of-Charge (SoC)

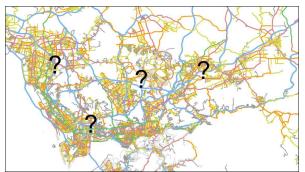










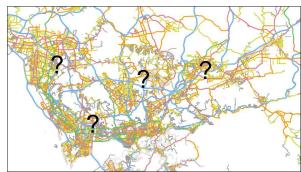


Long Queue









Long Queue

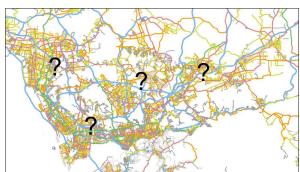


Time-Consuming









Long Queue

Time-Consuming

Range Anxiety

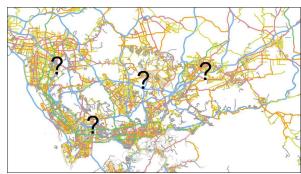












Long Queue

Time-Consuming

Range Anxiety

Maintain SoC



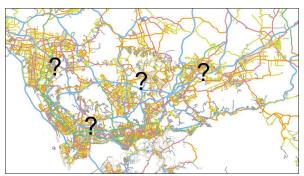












We need a method to schedule the deployment of wireless charging lanes that

- 1. Supports electric vehicles' continuous operability (maintain SoC at any location)
- 2. Minimizes the total deployment cost



### Plug-in charging station

IEEE TSG'12 IEEE TPS'14

IEVC'14 IEEE TSG'14

IEEE TPD'13 IEEE TPS'12

IEEE TPS'14





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IEEE TSG'12 IEEE TPS'14
IEVC'14 IEEE TSG'14
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IEEE TPS'14



Wireless power transfer
Annals of Physics'08
IEEE Systems Journal'16
ICPP'16



Not applicable for dynamic wireless charging

Wireless power transfer
Annals of Physics'08
IEEE Systems Journal'16
ICPP'16





Not applicable for dynamic wireless charging

2

Cannot maintain the SoC of vehicles in a metropolitan road network



# Our Approach:

# CatCharger

Categorization and clustering of multiple sources of vehicle traffic for the deployment of dynamic wireless Chargers in a metropolitan road network





### Outline

Dataset analysis

Design of CatCharger

Performance evaluation

Conclusions





Minimize deployment cost



### Minimize deployment cost

1. Vehicle passing velocity at charging lane matters

$$L_i = \frac{E_{max}}{r} \overline{v}_i$$

The slower the passing velocity, the shorter the charging lane needed



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Charge as many EVs as possible



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Charge as many EVs as possible

Keep the EVs operable (maintain SoC) on any position



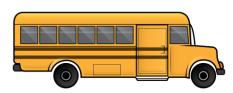


Our datasets (Jul 1~31, 2015) consist of:

15,610 taxicabs







12,386 dada buses





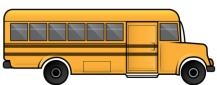


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Road map







Our datasets (Jul 1~31, 2015) consist of:

15,610 taxicabs

14,262 bus<u>es</u>

Generally represent the traffic of public transportation vehicles in Shenzhen



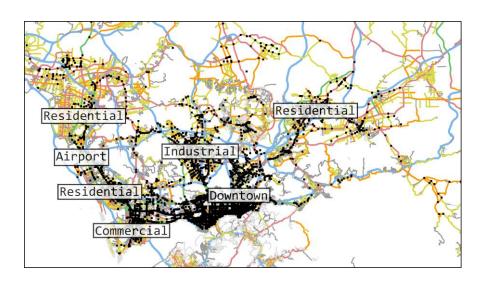


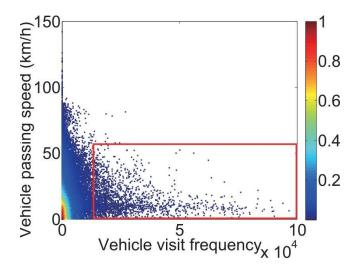


Distribution of potential positions for wireless charging



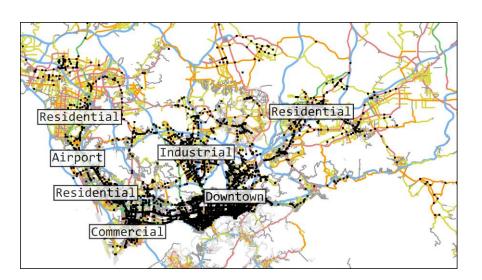
### Distribution of potential positions for wireless charging

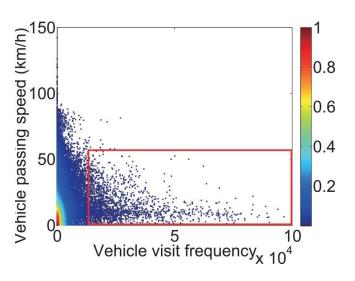






### Distribution of potential positions for wireless charging





Consider vehicle passing speed and vehicle visit frequency



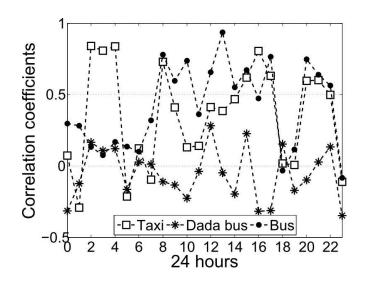
Minimize the cost of a charging lane and the serving capability

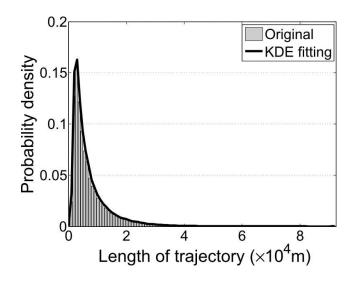


Multiple sources of vehicle traffic should be considered



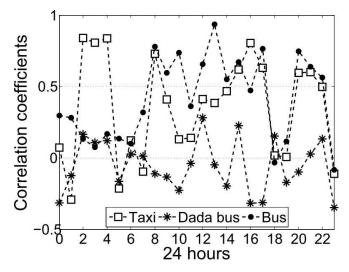
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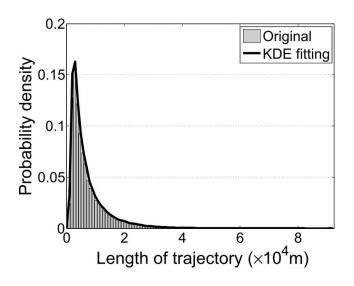






### Multiple sources of vehicle traffic should be considered





Consider multi-source vehicle traffic

Vehicle trip lengths follow certain distribution

Supports metropolitan-scale charging demand (maintain SoC)

# System Design



Vehicle mobility normalization

Charging lane location candidate extraction

-- High visit frequency and low passing speed

Charging lane location determination

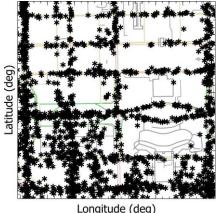
--Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold



Vehicle mobility normalization



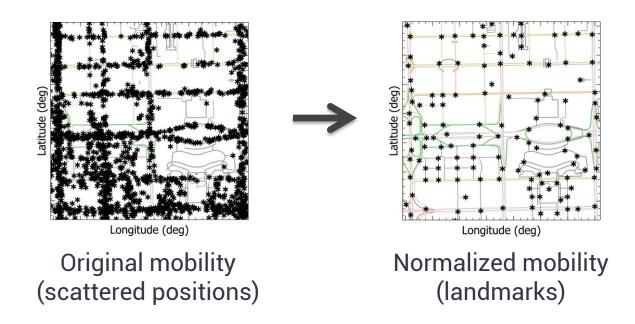
#### Vehicle mobility normalization



**Original mobility** (scattered positions)

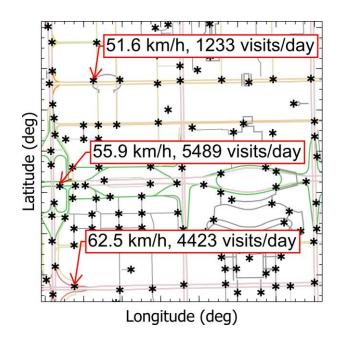


### Vehicle mobility normalization





#### Charging lane location candidate extraction

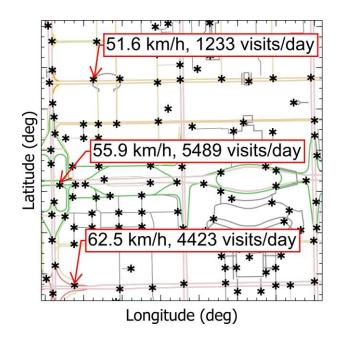


#### FACT

Analysis: consider vehicle passing speed and vehicle visit frequency



#### Charging lane location candidate extraction



#### **FACT**

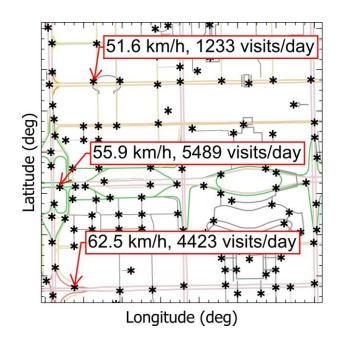
Analysis: consider vehicle passing speed and vehicle visit frequency



Cluster them by attribute values, and select the groups more suitable for deployment



#### Charging lane location candidate extraction



#### **FACT**

Analysis: consider vehicle passing speed and vehicle visit frequency



Cluster them by attribute values, and select the groups more suitable for deployment



How to cluster landmarks with similar attributes?



#### Charging lane location candidate extraction

Categorize original continuous numerical values into respective attribute IDs

```
v: <0, \ 0 \sim 5km/h >, <1, \ 5 \sim 10km/h >, \dots,
f: <0, \ 0 \sim 1000/day >, <1, \ 1000 \sim 2000/day >, \dots
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Each position can be described with two labels. For example,  $\{3 \text{ km/h}, 1500 \text{ visit/day}\} \rightarrow \{0, 1\}$ .



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Start from k starting landmarks  $\longrightarrow$  Landmarks clustered into k groups



In landmark clustering, for each landmark, we measure its "similarity" (entropy) with each group



#### Charging lane location candidate extraction

#### Select landmark groups:

We filter out the groups with passing speed higher than 60 km/h, and vehicle visit frequency lower than 10,000 visits/day

We choose landmarks with slow passing speed and high visit frequency



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#### Select landmarks in each selected group:

Rank the landmarks by their required lane length and visit frequency

$$R(\boldsymbol{lm_i}) = \frac{\log(f_i)}{L_i}$$



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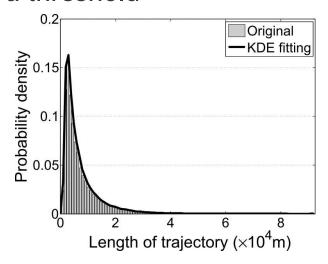
Rank the landmarks by their required lane length and visit frequency

$$R(\boldsymbol{lm_i}) = \frac{\log(f_i)}{L_i}$$

Select the top ranked landmarks (e.g., 10%) from each group as the candidate positions for deploying charging lanes



Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold



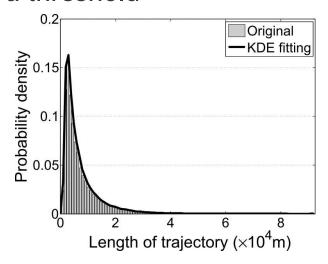
#### FACT

Analysis: Vehicle trip lengths follow certain distribution

The trip lengths for supporting



Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold





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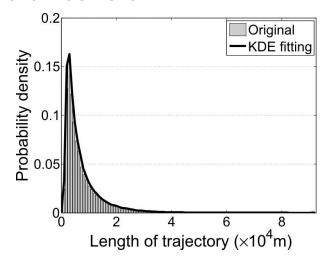


Infer the expected SoC of EVs given the deployed charging lanes in certain landmarks

The trip lengths for supporting



Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold



The trip lengths for supporting

#### FACT

Analysis: Vehicle trip lengths follow certain distribution



Infer the expected SoC of EVs given the deployed charging lanes in certain landmarks



Cannot be described with parametric distribution



Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold

Kernel Density Estimator (KDE)

$$\hat{f}_h(d) = \frac{1}{mh} \sum_{i=0}^{m-1} K(\frac{d-d_i}{h}); -\infty < d < \infty$$
 Probability of driving a certain distance

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 Probability of driving a certain distance

Vehicles energy consumption rate per meter c, minimum battery capacity  $E_{min}$ 

$$SOC(d) = \begin{cases} \frac{E_{min} - cd}{E_{min}}, & \text{if } E_{min} \geqslant cd \\ 0, & \text{otherwise} \end{cases}$$
 SOC estimated from the distance



Ensure expected residual energy (i.e., SoC) at any location is higher than a threshold

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Expected SoC of EVs at landmark  $lm_i$ 

$$\overline{SOC}(\boldsymbol{lm_j}) = \sum_{i=0}^{|\widehat{LM}|-1} \widehat{f}(d_{i,j})SOC(d_{i,j})x_i$$



Formulating optimization problem

Keep the EVs operable (maintain SoC)



Minimize total cost



#### Formulating optimization problem

Keep the EVs operable (maintain SoC)



Minimize total cost

minimize 
$$\sum_{\substack{\boldsymbol{lm_i} \in \widetilde{LM}}} \omega_0 x_i L_i$$
 subject to 
$$\overline{SOC}(\boldsymbol{lm_j}) \geqslant \alpha, \forall \ \boldsymbol{lm_j} \in LM$$
 
$$x_i \in \{0,1\}, \forall \ \boldsymbol{lm_i} \in \widetilde{LM}$$



#### Formulating optimization problem

Keep the EVs operable (maintain SoC)



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Binary Integer Programming problem



#### Comparison methods

Random: randomly deploy the charging lanes

MaxFlow: deploy chargers to maximally cover traffic flows (IEEE TPS'14)



Comparison methods

Random: randomly deploy the charging lanes

MaxFlow: deploy chargers to maximally cover traffic flows (IEEE TPS'14)

**Metrics** 

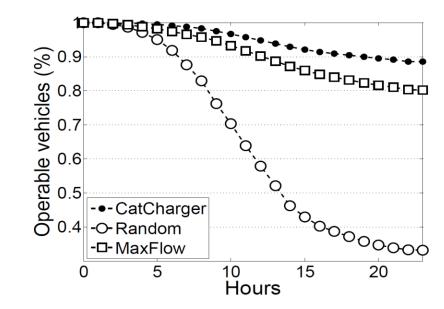
Keep the EVs operable (Maintaining SoC)



Performance in supporting EV charging demand

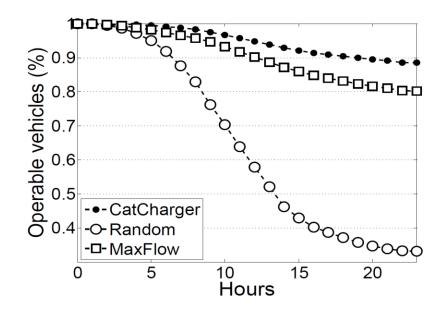


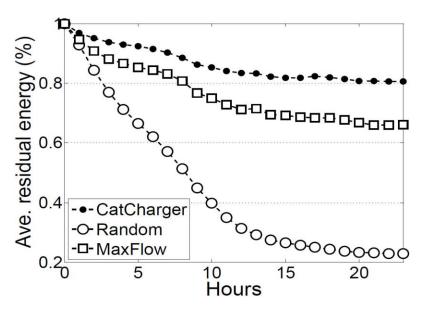
Performance in supporting EV charging demand



Operable vehicles over time

### Performance in supporting EV charging demand





Operable vehicles over time

Average residual energy



#### Conclusions

- 1. We designed a scheme to deploy wireless charging lanes to support metropolitan-scale EV charging demand
- 2. We conducted extensive experiments to verify the effectiveness of CatCharger in supporting the SoC of EVs

3. In the future, we plan to consider the influence of human activities and analyze the after-effect brought by the deployment of charging lanes



## Thank you! Questions & Comments?

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University of Virginia