

# **TRAFFIC CONGESTION DETECTION WITH CROWDSOURCING DATA: A NETWORK-BASED SPATIAL-TEMPORAL CLUSTERING APPROACH**

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# Outline

- Introduction
- Data and Methods
- Results
- Conclusion and Future work

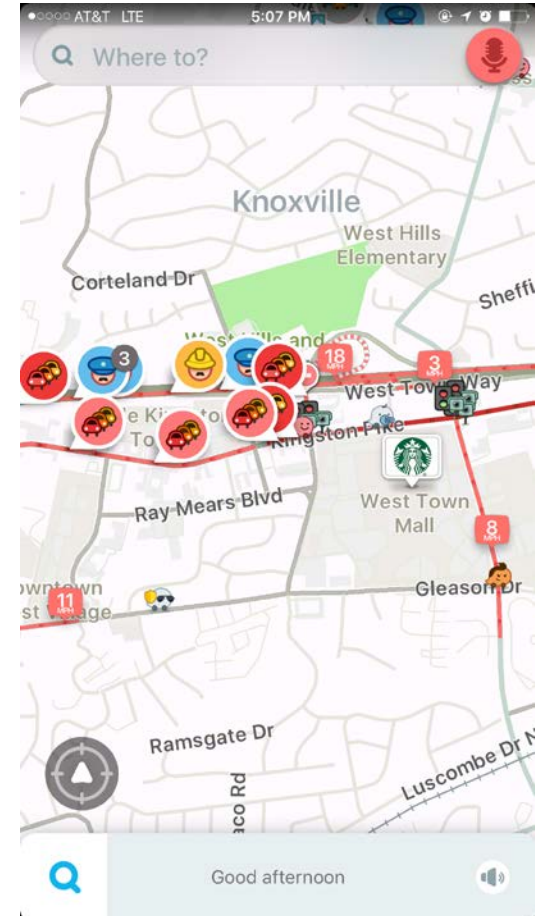
# Introduction

## Traffic Congestion

- Congestion detection, estimation and prediction
- Data limitation: missing value, limited coverage

## Crowdsourcing data, like WAZE

- High coverage of the road network
- Cost-effective
- Reliable



**develop and validate a network-based spatial-temporal clustering approach that supports the accurate detection of traffic congestion with Waze data**



# Data



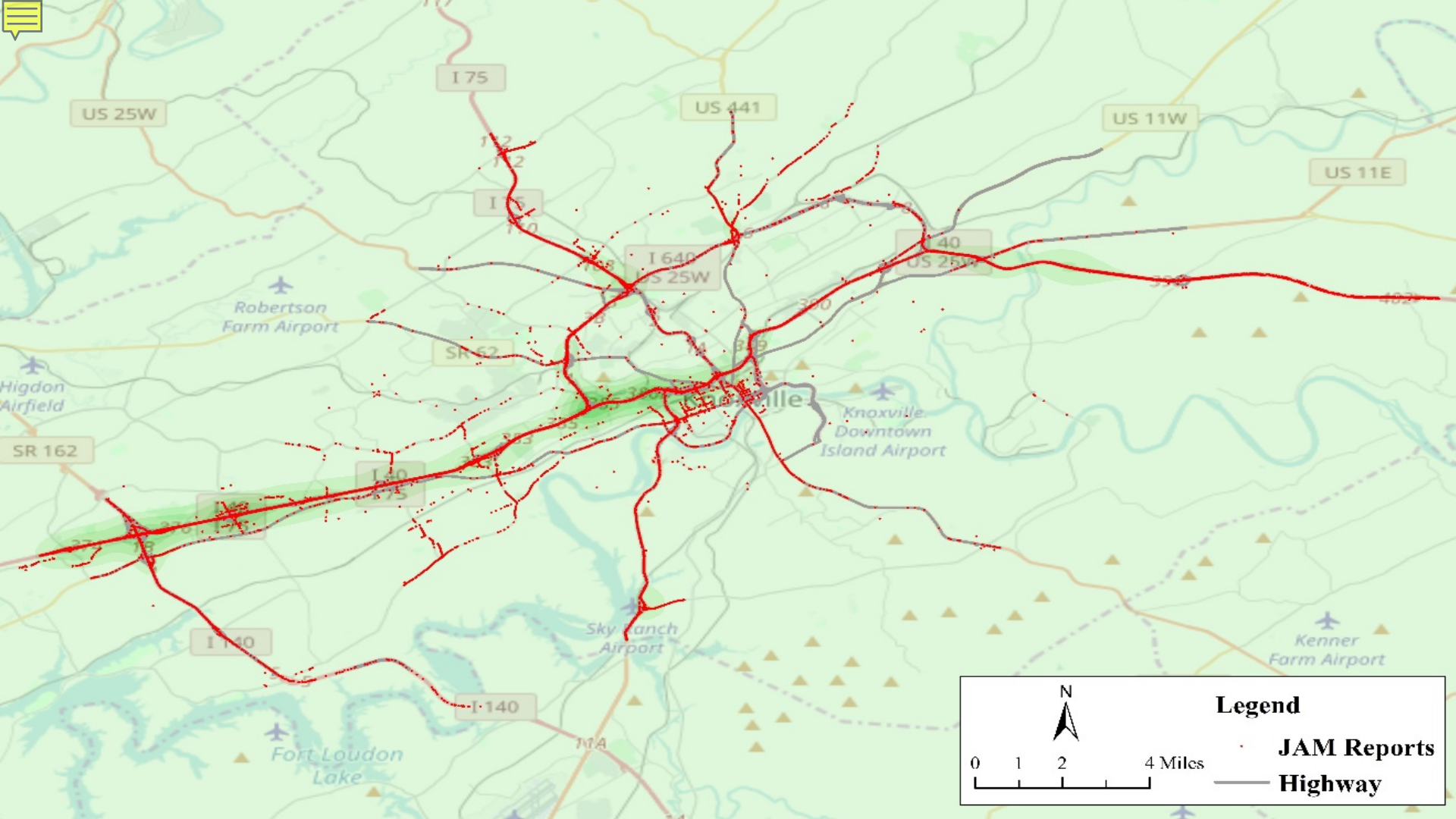
Road network  
(Open Street Map  
road network)

43,158 nodes  
47,054 edges



Waze JAM reports  
for Knoxville  
(July-Sep, 2017)

17, 283 JAM  
reports



# Methods



## Map matching algorithm

Project the data onto road



## ST-DBSCAN

Density-based clustering

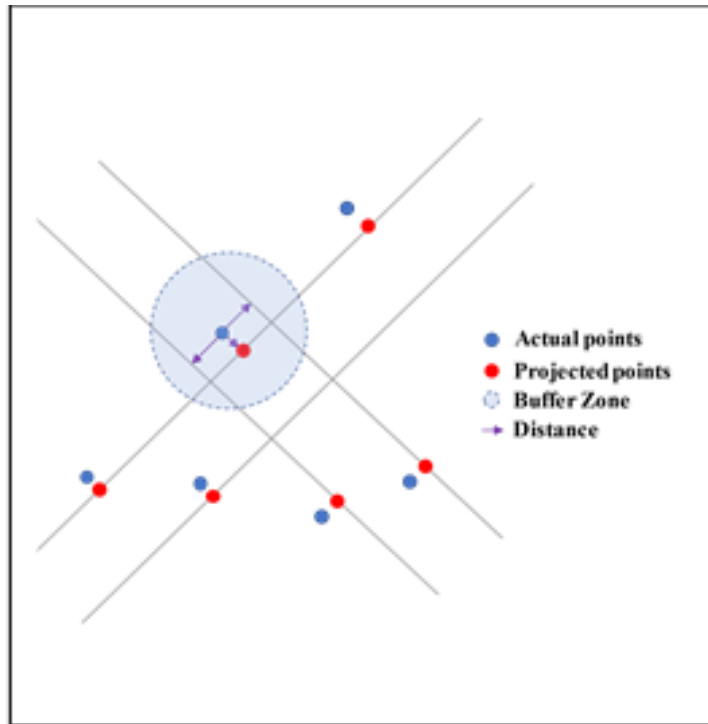


## Dijkstra algorithm

Short-distance finding algorithm

# Map matching algorithm

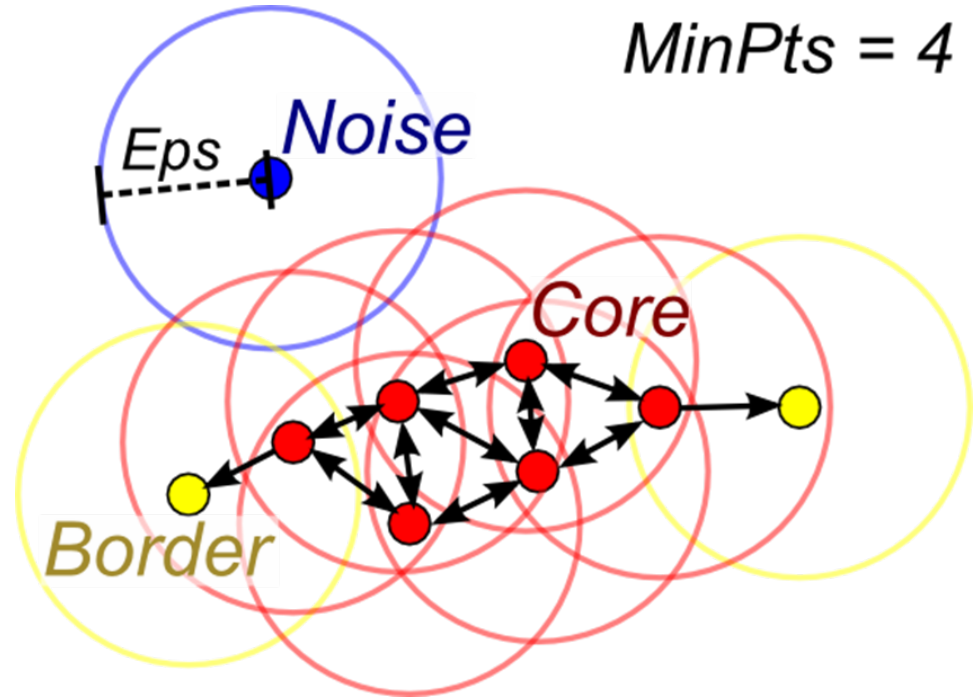
```
MapMatching ( $D, d$ )
   $maxDist = \text{infinity}$ 
   $minDist = 0$ 
   $nearSegment = \text{none}$ 
  for each point  $p$  in dataset  $D$ :
    create a buffer zone  $C$  with a distance  $d$ 
    find the segments  $S$  intersects with  $C$ 
    if  $len(C) == 0$ :
      continue
    for each segment  $s$  in  $S$ :
      if  $d[p, s] < maxDist$ :
         $minDist = d[p, s]$ 
         $nearSegment = s$ 
    project  $p$  onto the segment  $s$ 
```





# ST-DBSCAN

- Parameters
  - Distance threshold ( $\epsilon$ )
  - Time threshold ( $t$ )
  - Minimum points ( $minPts$ )



# Dijkstra algorithm

```
Modified_Dijkstra(G, s,  $\epsilon$ ){
  d[s]  $\leftarrow$  0
  Q = {s}
  neighbor_list = {}
  while Q is not empty:
    u  $\leftarrow$  Q[0]
    Q  $\leftarrow$  Q - {u}
    for each neighbor v of u:
      if v is not visited and d[u]+e(u,v) < d[v]:
        d[v] = d[u]+e(u,v)
        Q  $\leftarrow$  Q  $\cup$  {v}
  sort Q in ascending order based on the distance values of the elements
  if d[Q[0]] >  $\epsilon$  :
    return neighbor_list
  else:
    neighbor_list  $\leftarrow$  neighbor_list  $\cup$  Q[0]
    mark Q[0] as visited
}
```

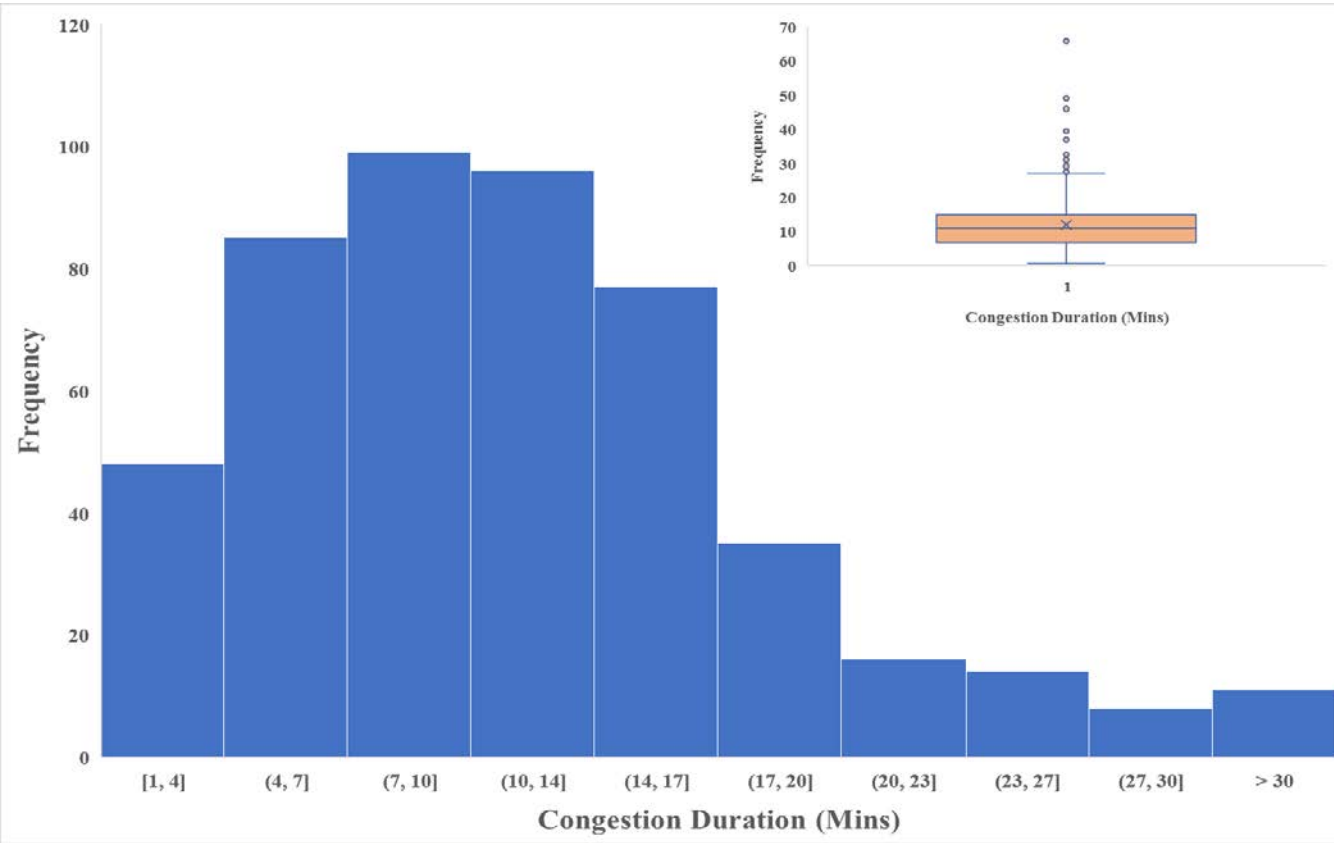
Distance threshold ( $\epsilon$ )

- Get the neighbor points within distance threshold ( $\epsilon$ )



# Clustering Results





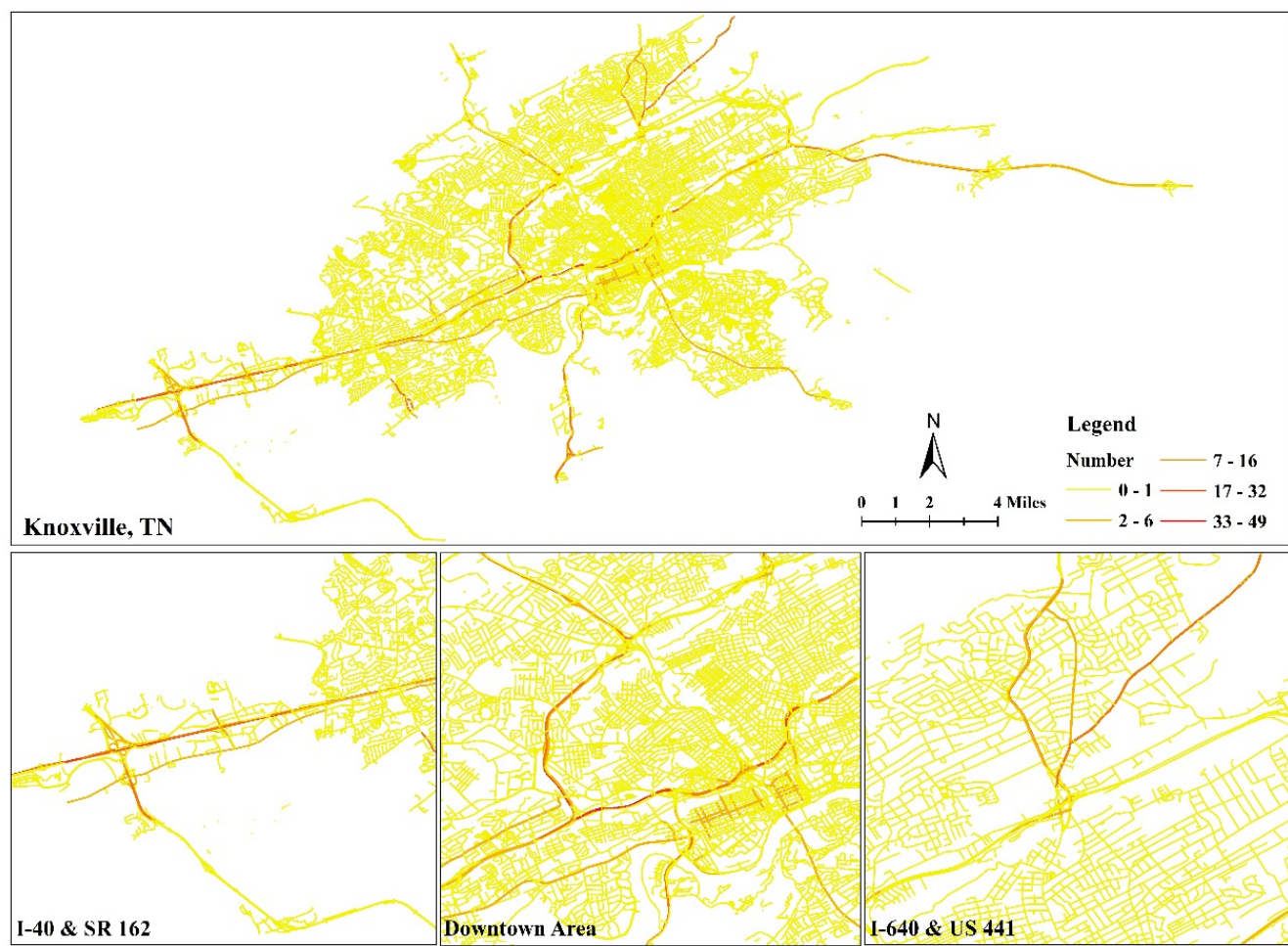
## Congestion duration

- most of the traffic congestions are below 20 mins



## Congestion level

- The congested road segments are mainly on interstate highways, especially at the highway interchanges



# Conclusion and Future work

- Demonstrate the applicability of our algorithm to real world problems with a case study of Knoxville
- Our approach has the capability to identify cluster with any shape and a high routing flexibility.
- Big data application
- real-time application





# Questions?



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