

PyNKDV: An Efficient Network Kernel Density Visualization Library for Geospatial Analytic Systems

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Overview of Network Kernel Density Visualization (NKDV)



(a) 311-call data points

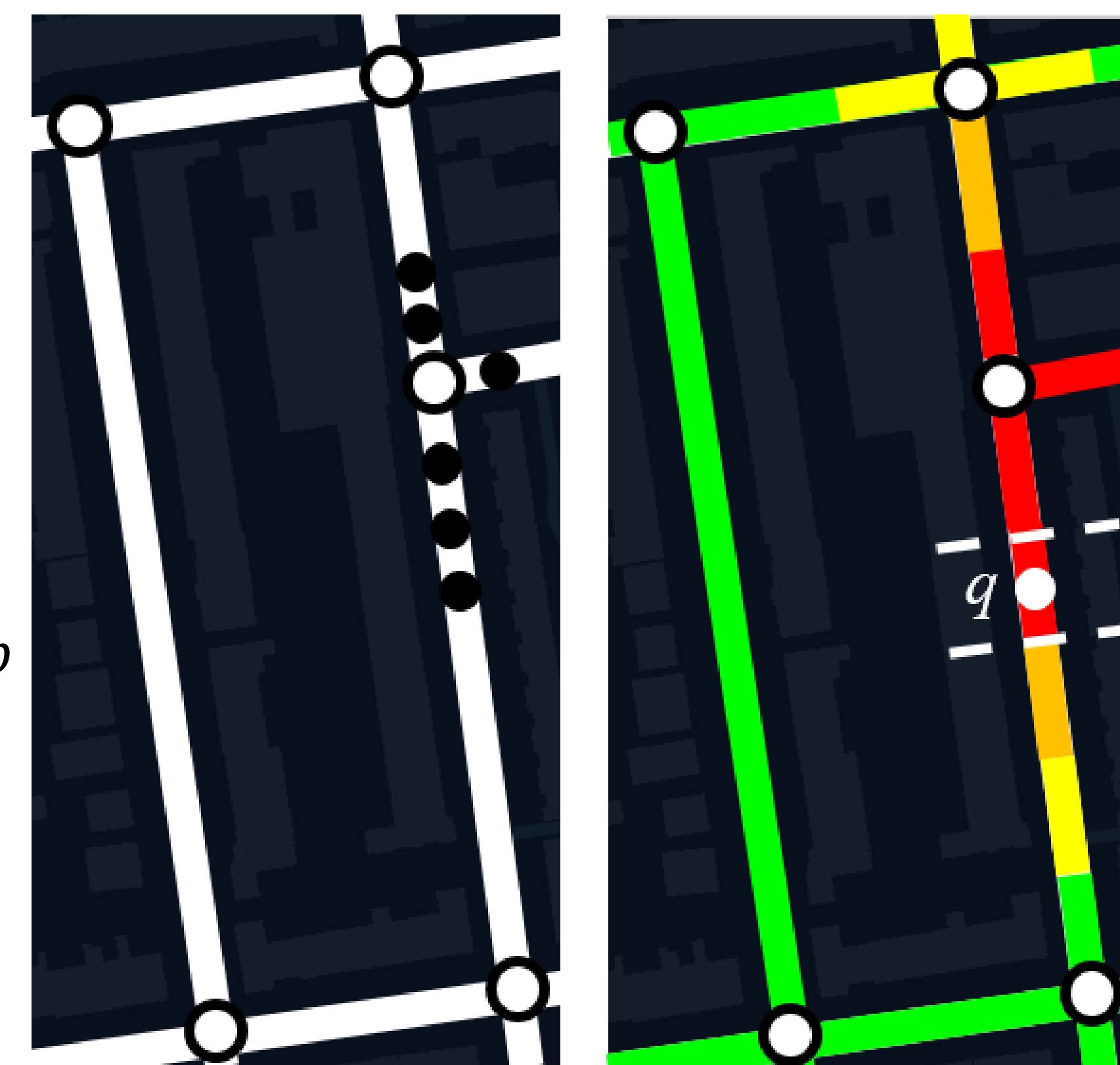
(b) NKDV

- How to generate NKDV?
 - Divide each road into a set of lixels.
 - Color each lixel q based on the network kernel density function $\mathcal{F}_p(q)$.

$$\mathcal{F}_p(q) = \sum_{p \in P} w \cdot \begin{cases} \text{constant} & \text{shortest path distance} \\ 1 - \frac{1}{b^2} d_G(q, p)^2 & \text{if } d_G(q, p) \leq b \\ 0 & \text{otherwise} \end{cases}$$

bandwidth

- The worst-case time complexity of generating NKDV is $O(L(T_{SP} + n))$
 - L is the number of lixels.
 - T_{SP} is the time complexity of the shortest path algorithm.
 - n is the number of data points.



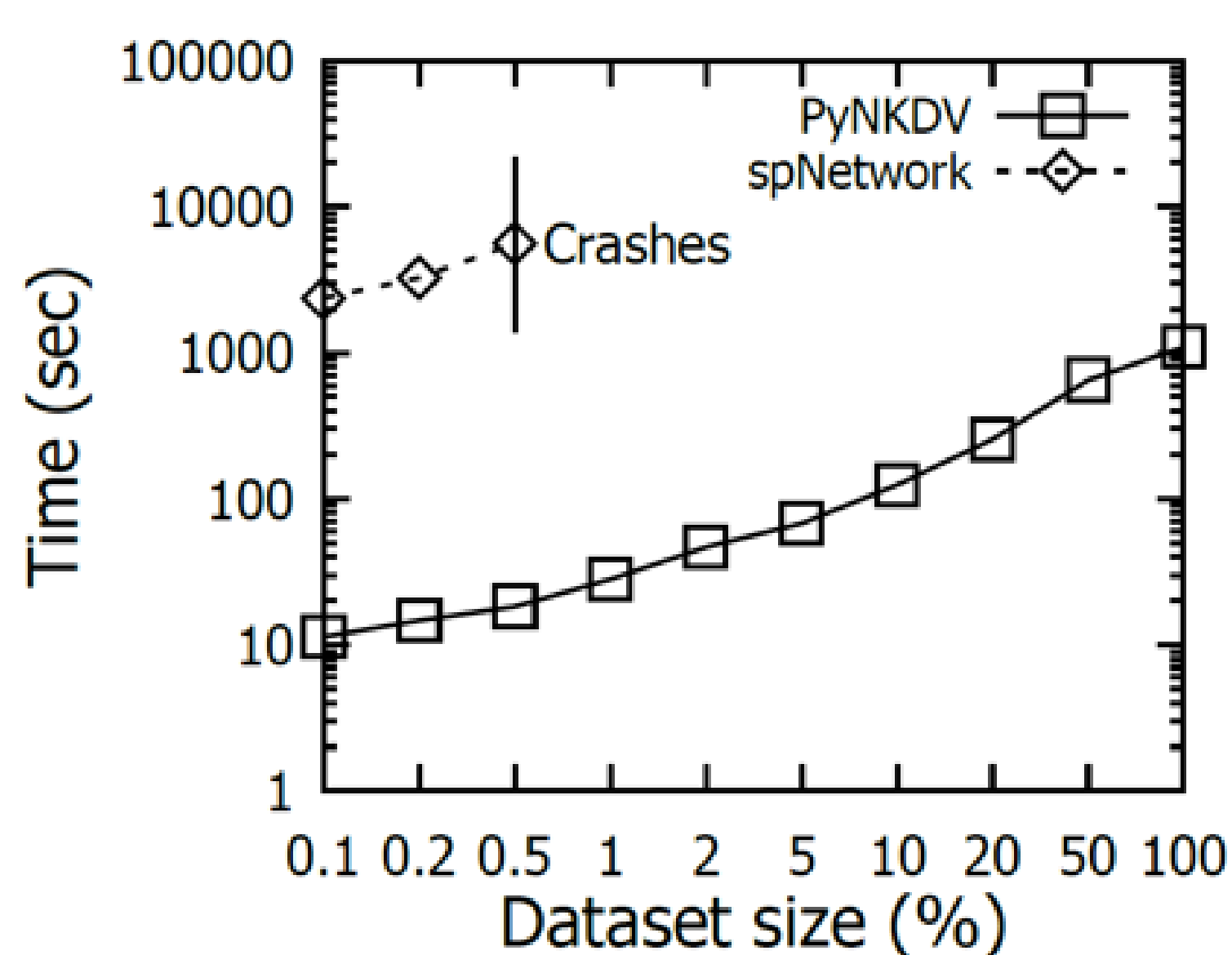
(a) Data points

(b) NKDV

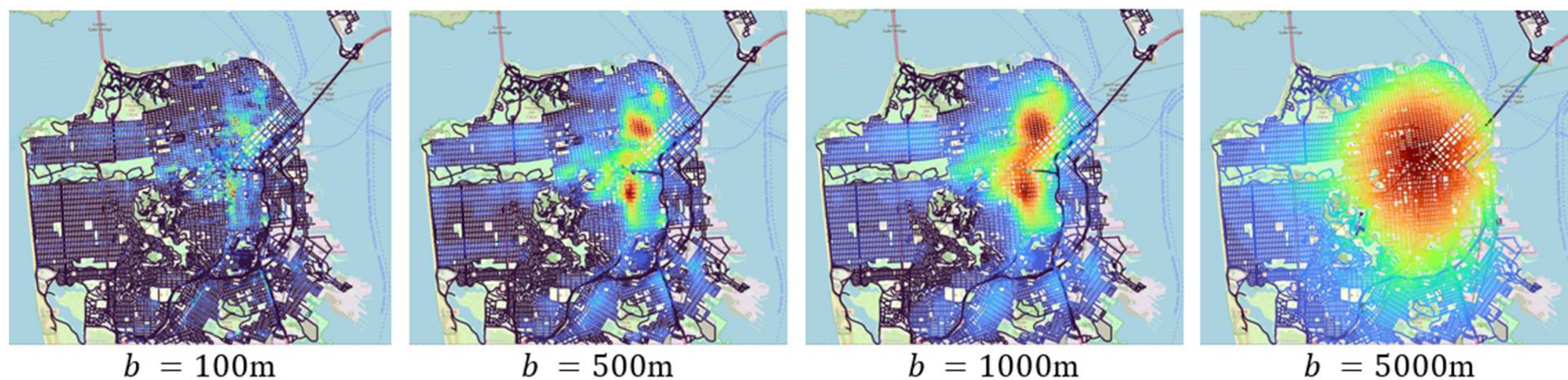
- Generating NKDV is slow ☹️

Why PyNKDV?

- Based on a fast solution, called aggregate distance augmentation (ADA), for generating NKDV 😊



- Efficiently support the bandwidth tuning operation 😊



- Easy to use (Four lines of code) 😊

```
road_data = map_road_network(location_data)
model = PyNKDV(road_data, bandwidth=1000,
               lixel_size=5, num_threads=8)
results = model.compute()
output(results, output_file_name)
```

- The time complexity of ADA is $O(|E|T_{SP} + L|E| \log(\frac{n}{|E|}) + n)$.

Case Study: KDV v.s. NKDV



(a) KDV

(b) NKDV

- KDV tends to overestimate the density values in a road network.
- NKDV can provide more reasonable visualization results.

Github link for PyNKDV

