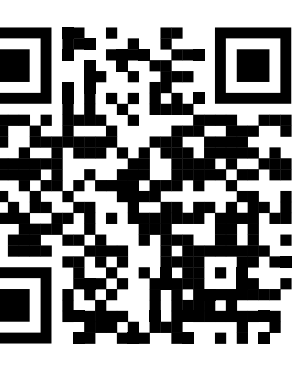




STARS: A Sampling and Threshold Sharing Solution for Network K-function Analytics



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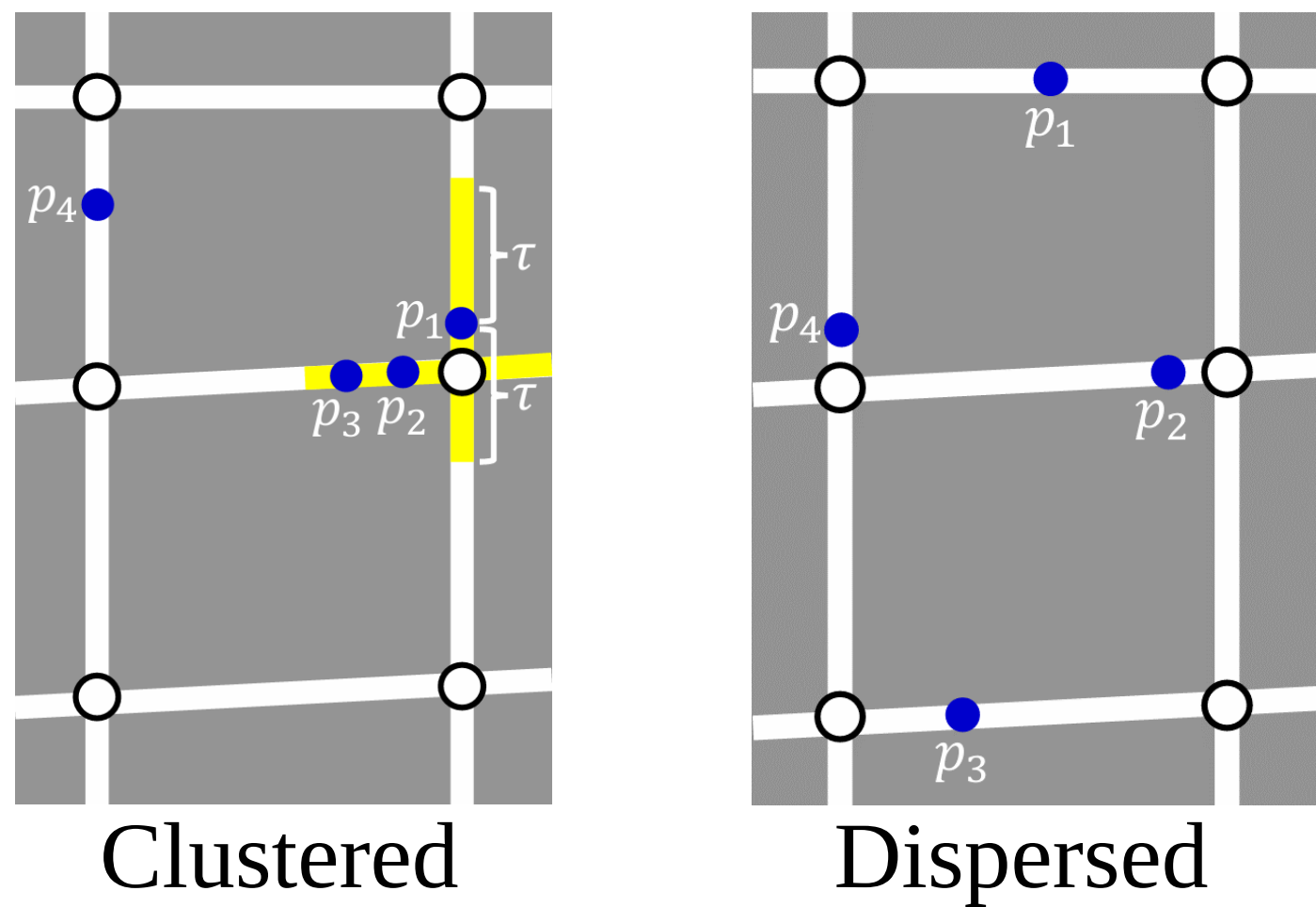
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What is Network K-function Analytics?



- Network K-function** is a fundamental GIS operation for checking whether the clusters/hotspots (discovered by the clustering/hotspot detection algorithms) in the datasets are significant.

$$K_P(\tau) = \frac{1}{n(n-1)} \sum_{\substack{p_i \in P \\ \text{Dataset}}} \sum_{\substack{p_j \in P \\ p_j \neq p_i}} \mathbb{I}(\overbrace{d_G(p_i, p_j)}^{\text{Shortest path distance}} \leq \underbrace{\tau}_{\text{Threshold}})$$

- Domain experts need to (1) determine D thresholds $\tau_1, \tau_2, \dots, \tau_D$, (2) generate L random datasets (with the same size n) $R_1, R_2, \dots, R_L$, and (3) compute the following network K-function values:

- $K_P(\tau_d)$
- $\mathcal{L}(\tau_d) = \min(K_{R_1}(\tau_d), K_{R_2}(\tau_d), \dots, K_{R_L}(\tau_d))$
- $U(\tau_d) = \max(K_{R_1}(\tau_d), K_{R_2}(\tau_d), \dots, K_{R_L}(\tau_d))$

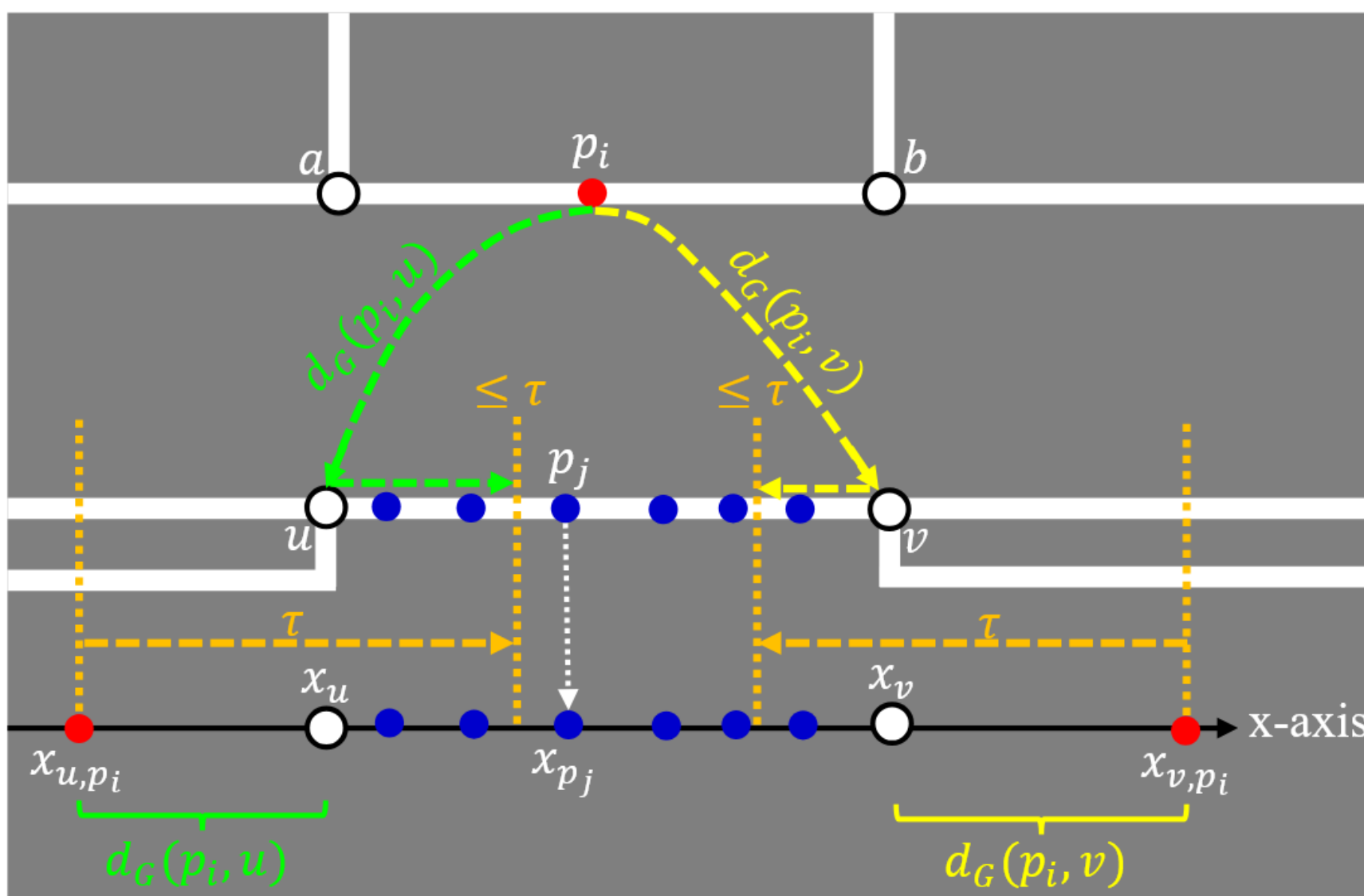
- Existing methods are slow!
Example: >10 hours on 6.05M events and 100 thresholds

$$O(|E|T_{sp} + LnD|E| + Ln \log n)$$

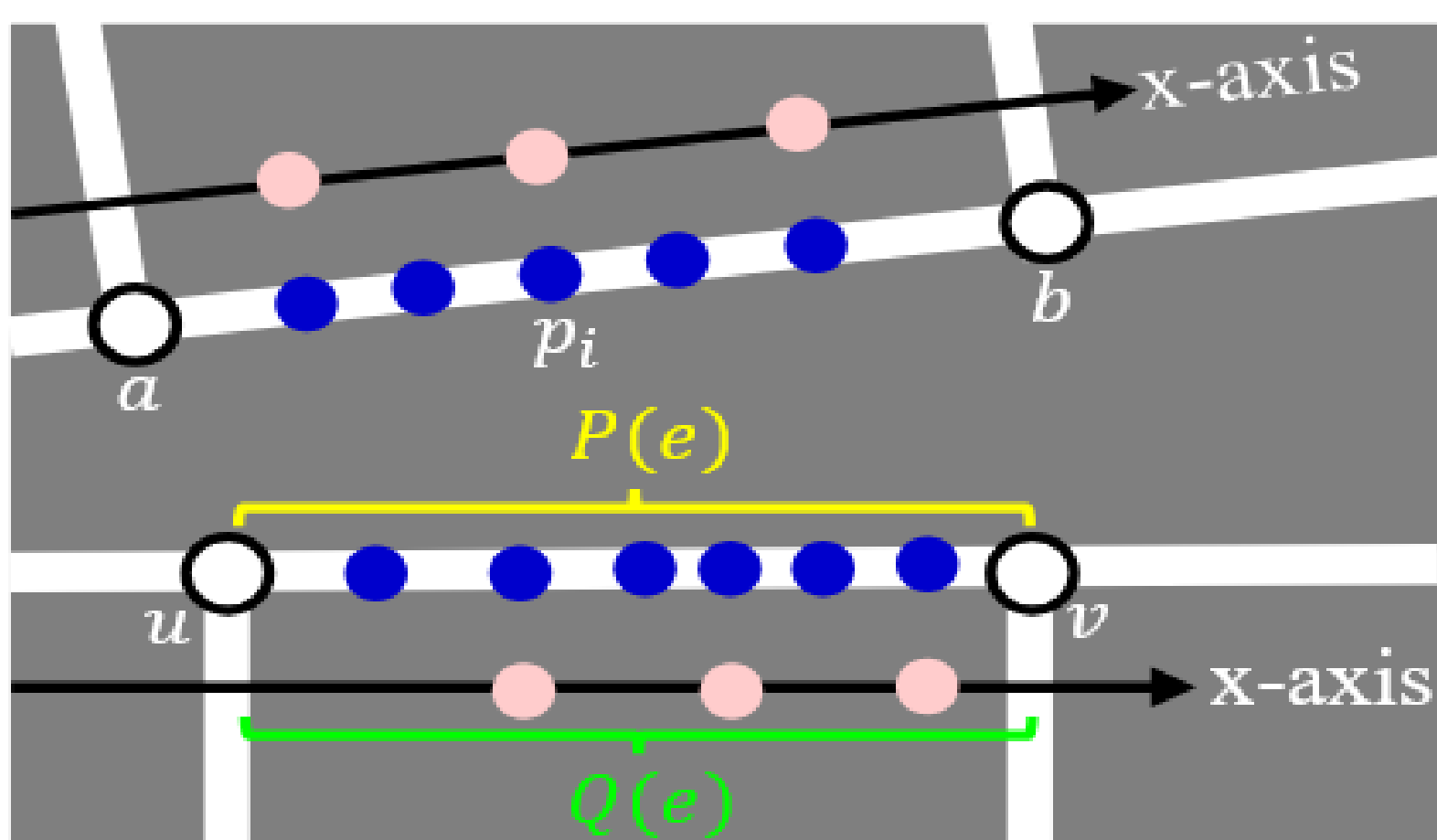
Sampling and Threshold Sharing Solution (STARS)

Sampling

- Reformulate the network K-function as a one-dimensional KDE problem, enabling the use of the optimal Sort Selection (SS) sampling strategy.



① Representative Edge Sampling

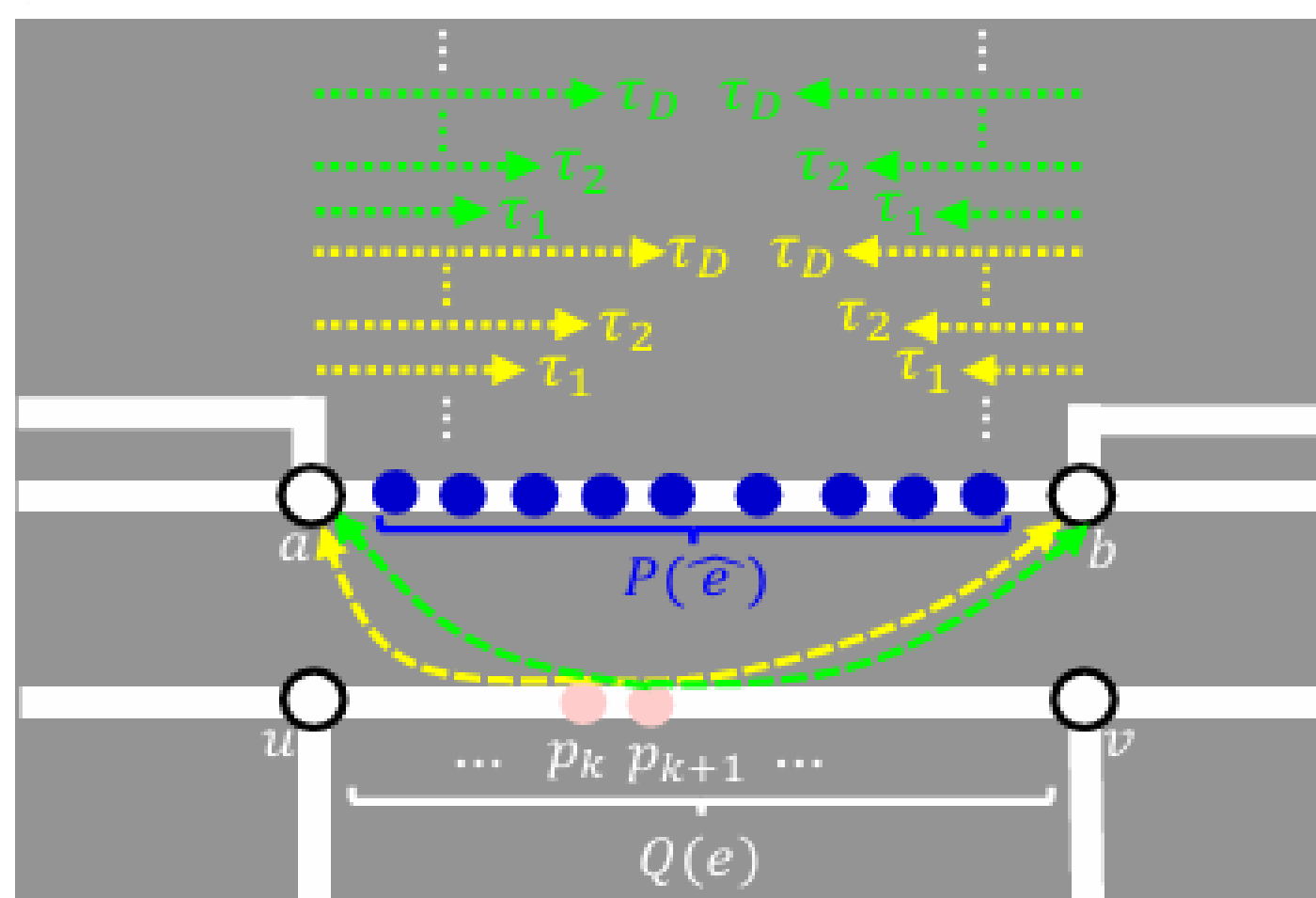


Reduce the evaluated points from $|P(e)|$ to $O(1/\epsilon)$.

Threshold Sharing

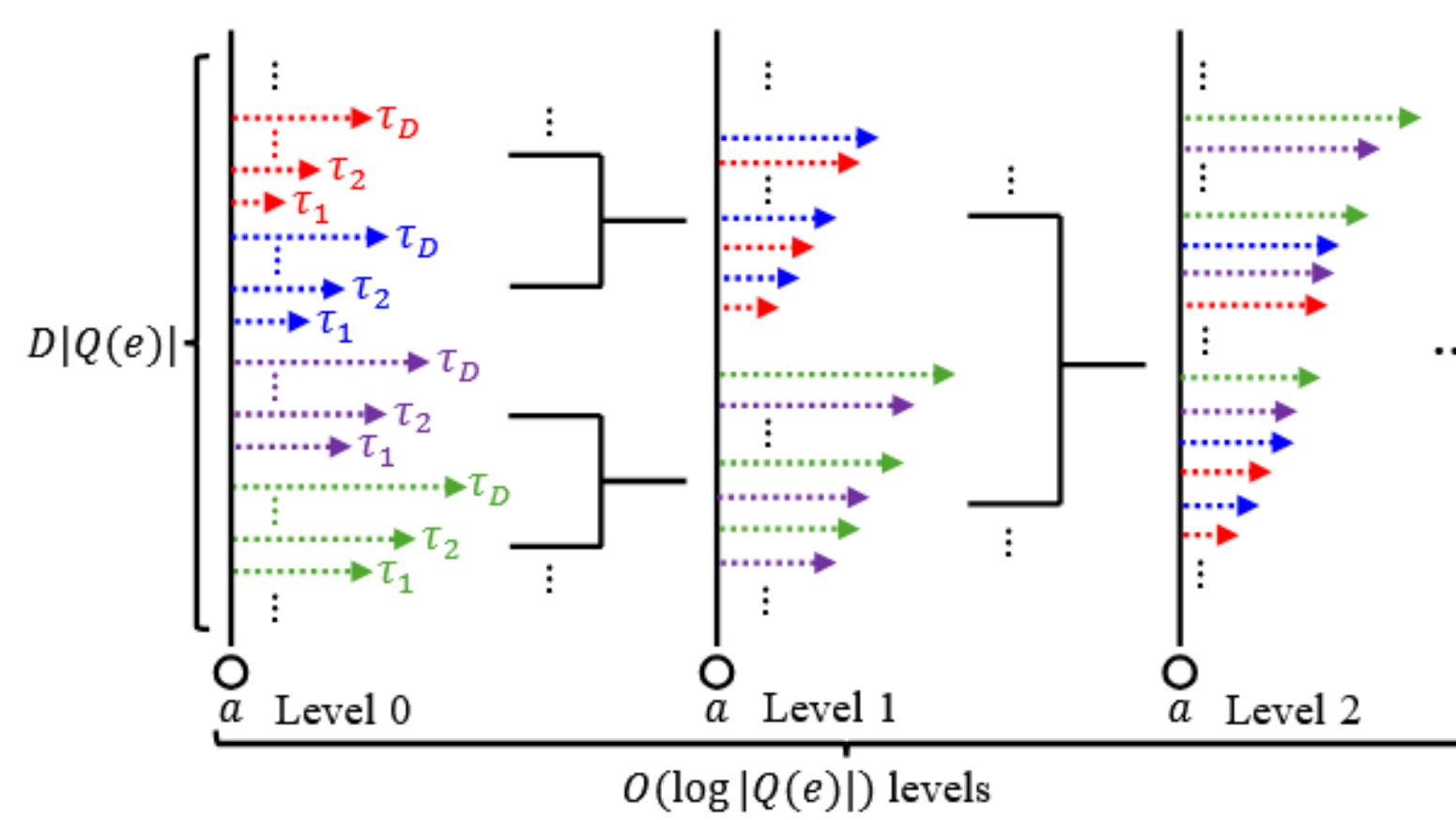
- Merge all threshold boundaries into a single ordered sequence and evaluate all thresholds through one sweep-line pass.

② Generate boundaries



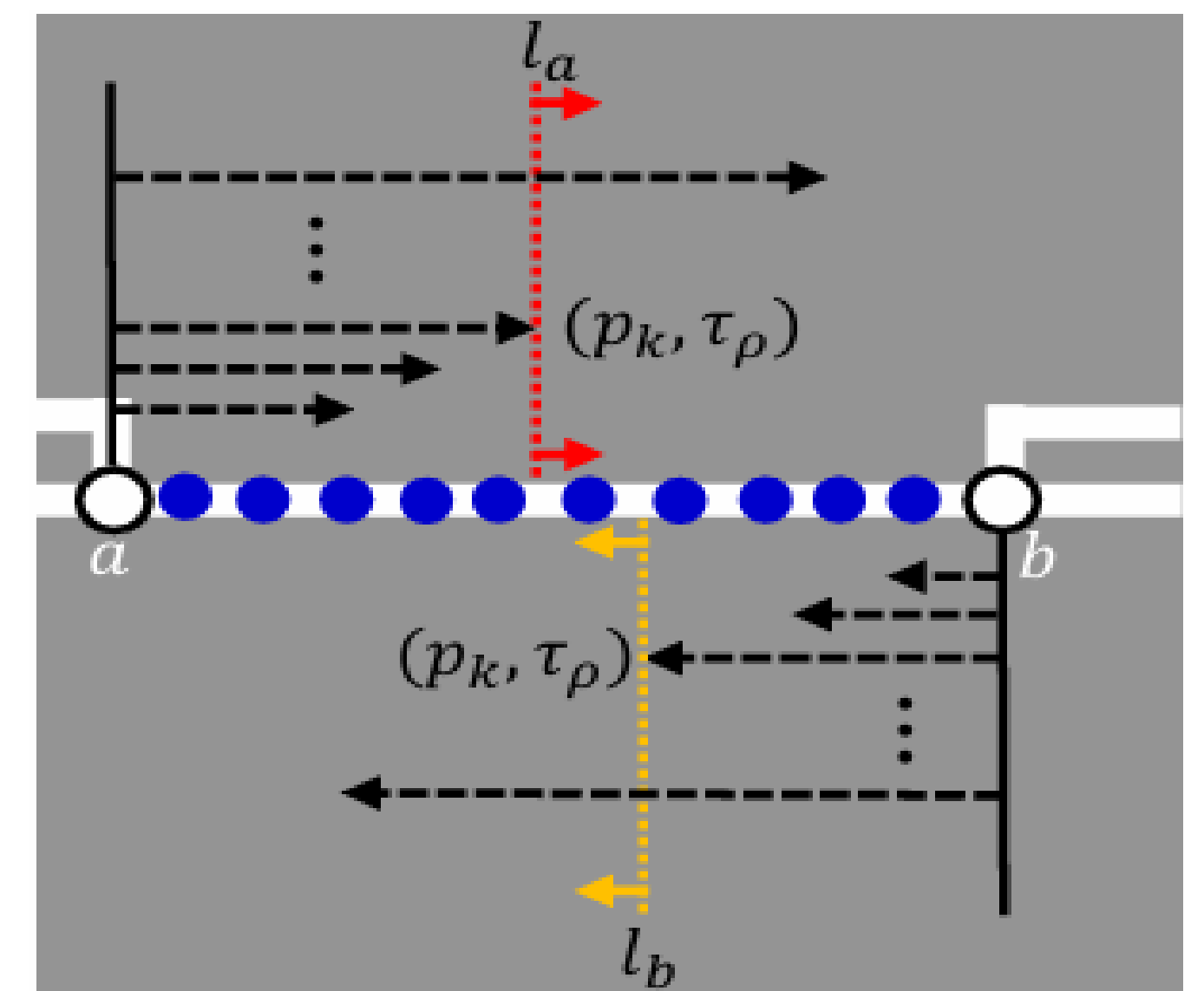
Compute left/right threshold boundaries for every sampled point.

③ Merge & Sort Boundaries



Merge and sort all threshold boundaries into one ordered sequence.

④ Sweep-Line Processing



Sweep once to evaluate all thresholds simultaneously.

Theoretical Guarantees

Sampled dataset size

$$O(|E|/\epsilon)$$

Approximation Guarantees

$$|A_{P,Q}(\tau) - K_P(\tau)| \leq \epsilon$$

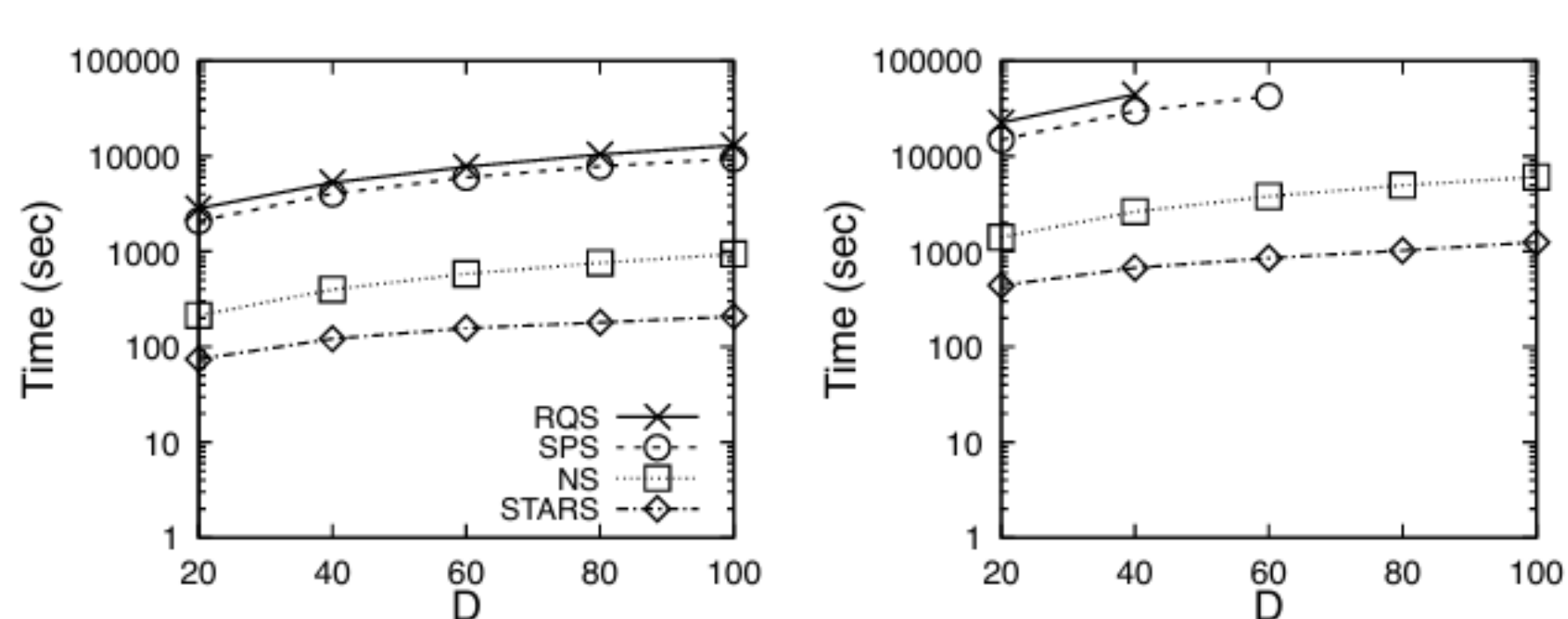
Time Complexity

$$O\left(|E|T_{SP} + \frac{LD|E|^2}{\epsilon} \log\left(\frac{1}{\epsilon}\right) + Ln|E| + Ln \log n\right)$$

Space Complexity

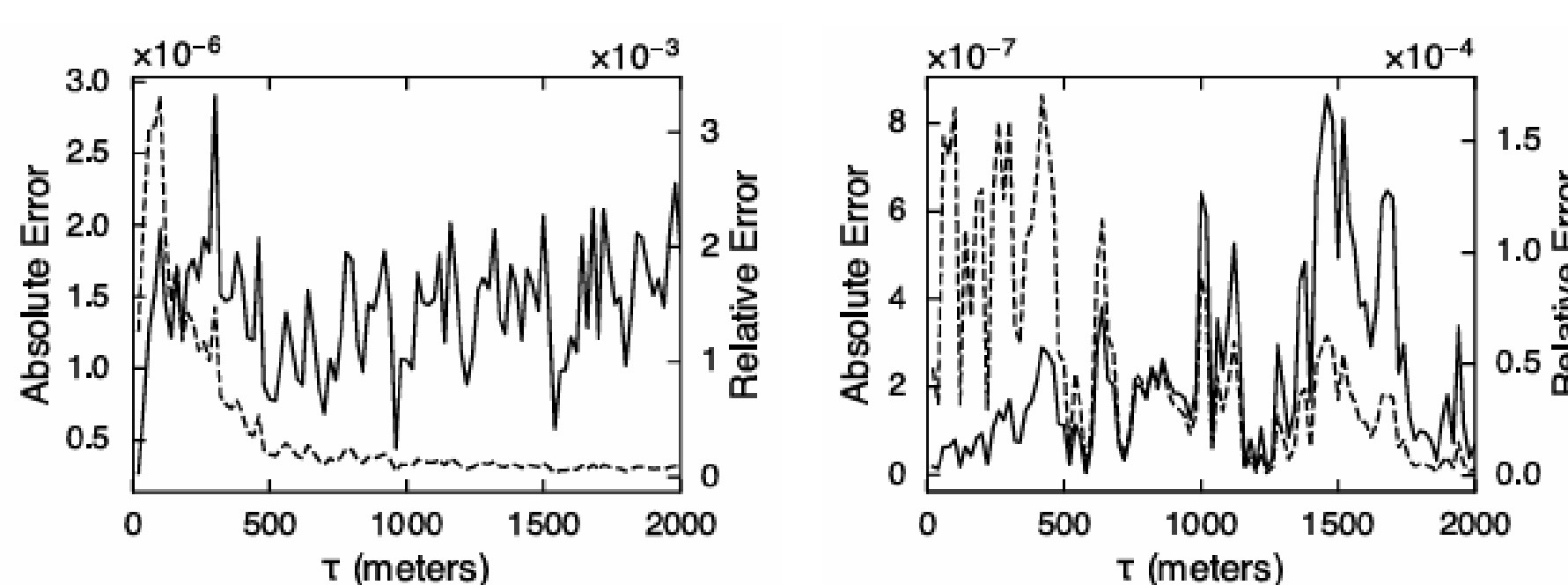
$$O\left(|E| + Ln + \frac{D}{\epsilon} + S_{SP}\right)$$

Experimental Results and Case Study



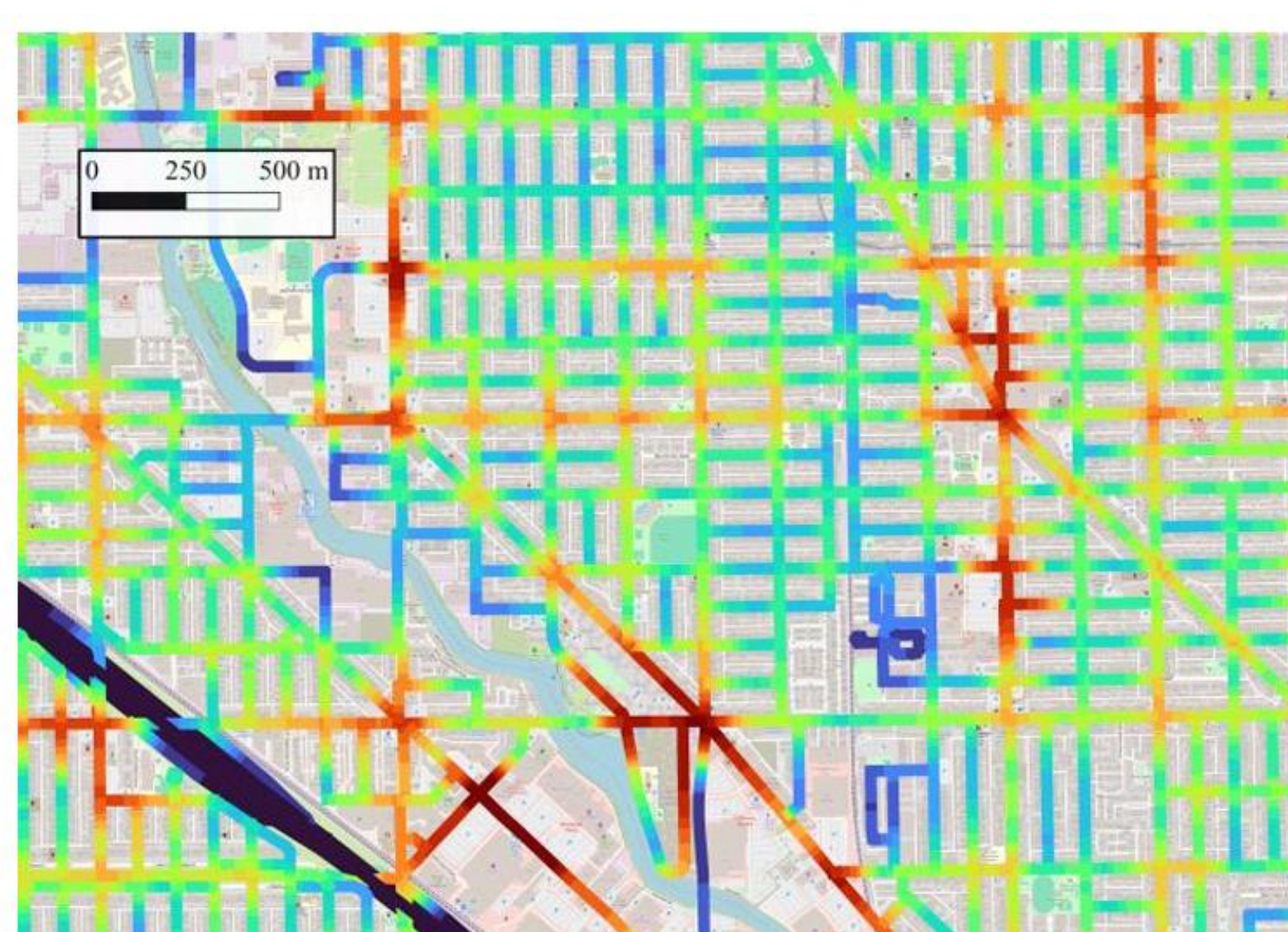
(a) Maryland

(b) New York

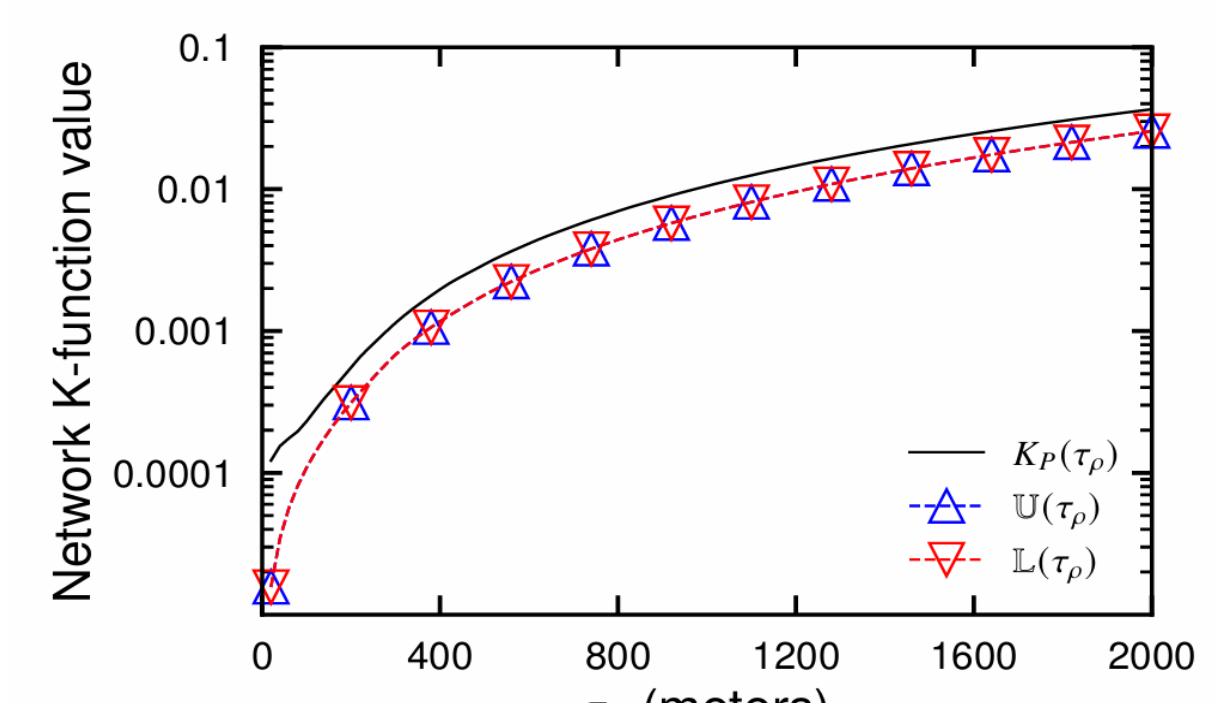


(a) Maryland

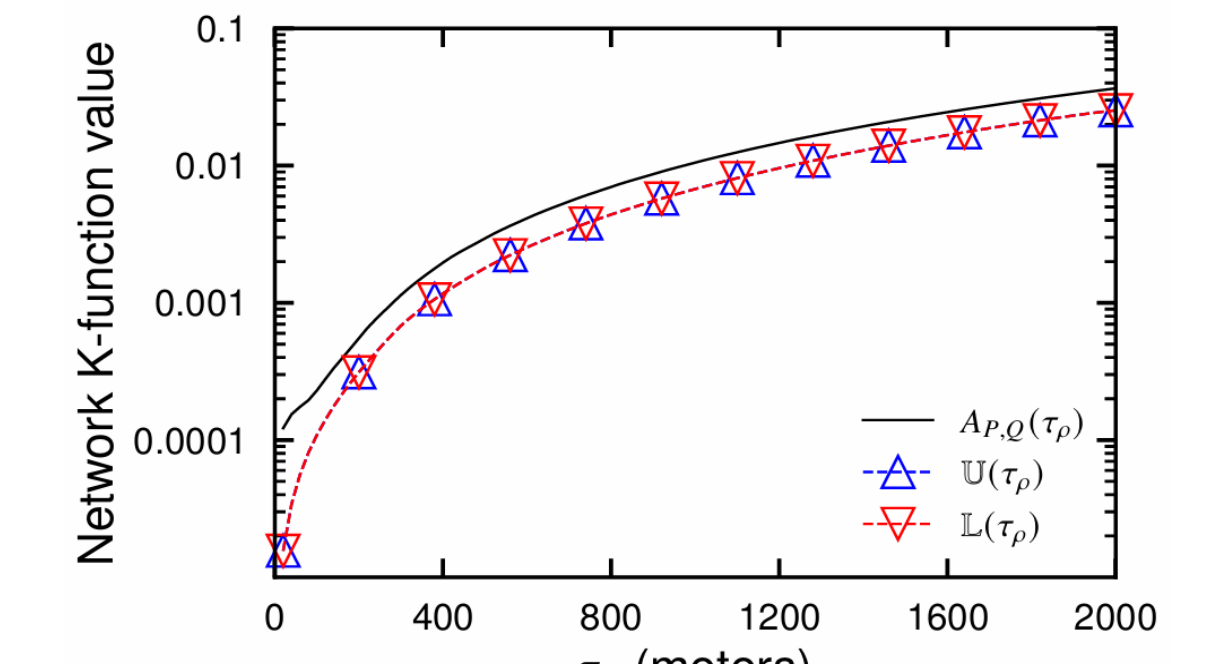
(b) New York



STARS preserves the analytical conclusions of the exact methods while enabling efficient hotspot analysis



Exact



Approximation