

## Mathematical Model of Taxi Routing

**Goal** of taxi routing:

- Determine the **minimum number of taxis needed** to serve all the trips without causing delays (with 20/20 hindsight)

**Input** to the model:

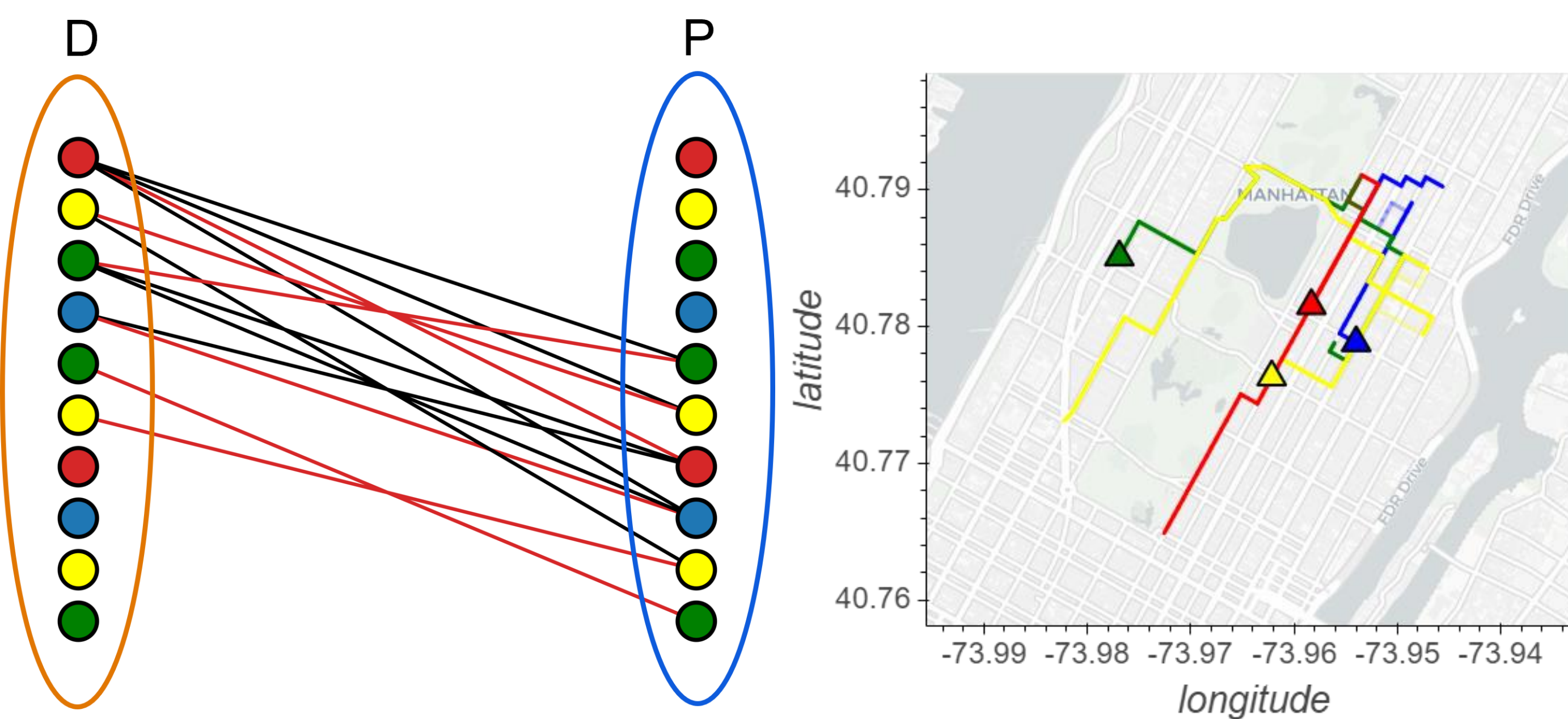
- A collection of  $n$  taxi trips  
 $N = \{(p_1, T_1^p, d_1, T_1^d), \dots, (p_n, T_n^p, d_n, T_n^d)\}$ , where the  $i^{th}$  trip picks up a passenger at time  $T_i^p$  at point  $p_i$  and drops off the fare at time  $T_i^d$  at point  $d_i$ ,  $i \in \{1, \dots, n\}$
- Maximum allowable taxi waiting time  $\delta$  (in minutes)

Construct a **bipartite graph**  $G = (D, P, E)$ :

- $D = \{(d_1, T_1^d), \dots, (d_n, T_n^d)\}$  denotes the set of **drop-off** nodes
- $P = \{(p_1, T_1^p), \dots, (p_n, T_n^p)\}$  denotes the set of **pickup** nodes
- $E = \{(d_i, T_i^d), (p_j, T_j^p) : T_j^p - T_i^d - \delta \leq \text{time}(d_i, p_j) \leq T_j^p - T_i^d\}$
- Elapsed time** between the drop-off time for  $d_i$  and the pickup time for  $p_j$ , that is,  $T_j^p - T_i^d$ , is at least the time needed to travel between these two points  $\text{time}(d_i, p_j)$  and at most  $\text{time}(d_i, p_j) + \delta$ ; hence, the taxi can reach the new pickup in time, and does not need to wait more than  $\delta$  minutes for the pickup to be ready

**Lemma** Given any matching of size  $m$ , there exists a covering of all  $n$  trips with  $n - m$  taxis

- Application: **Maximum cardinality matching** gives rise to the **minimum number of taxis that cover all the trips** [1][2]

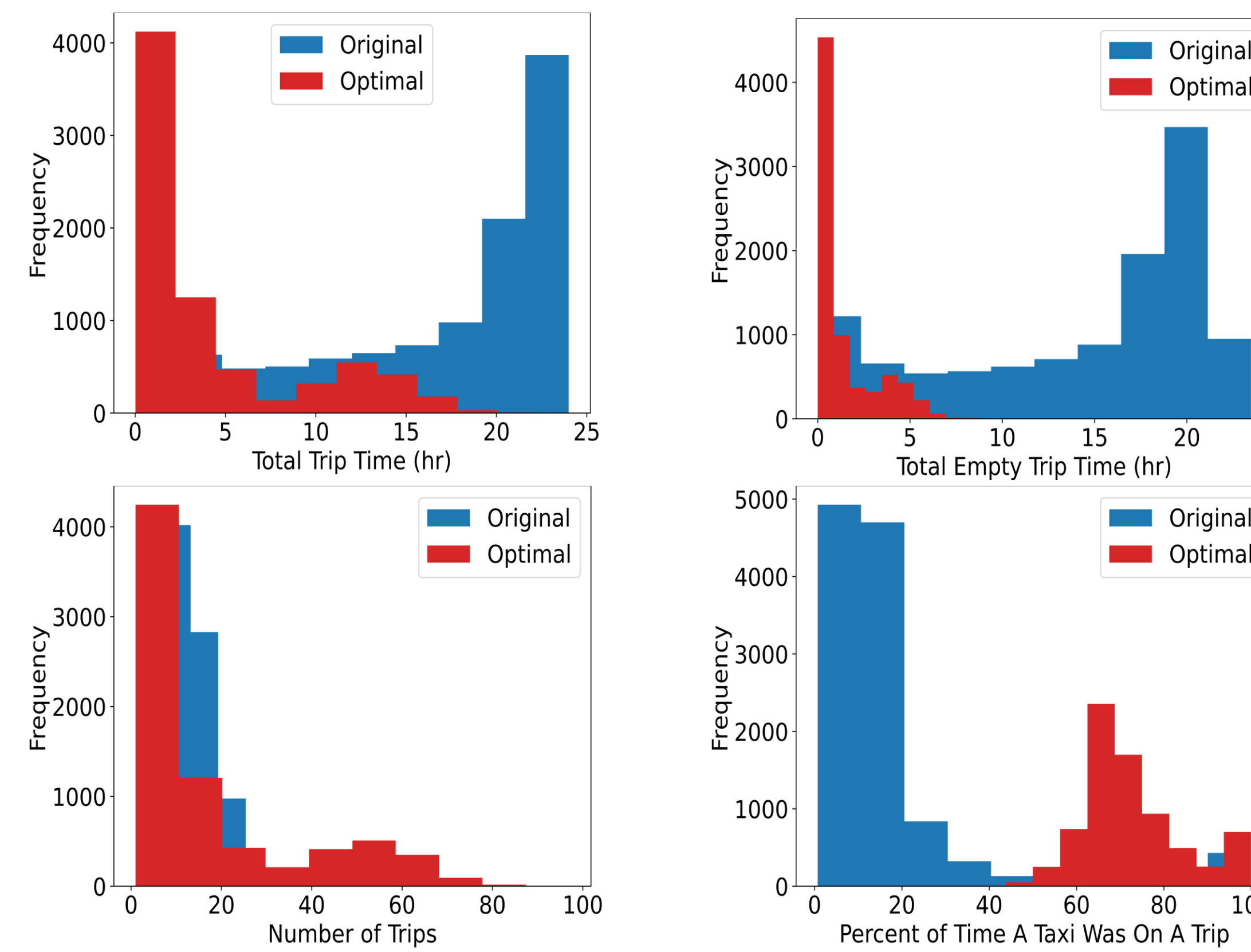


- An example of 10 trips that can be covered by 4 taxis
- Maximum cardinality matching ( $m = 6$ ) is colored in red
- Nodes are colored with respect to their corresponding taxi routes

## Computational Results

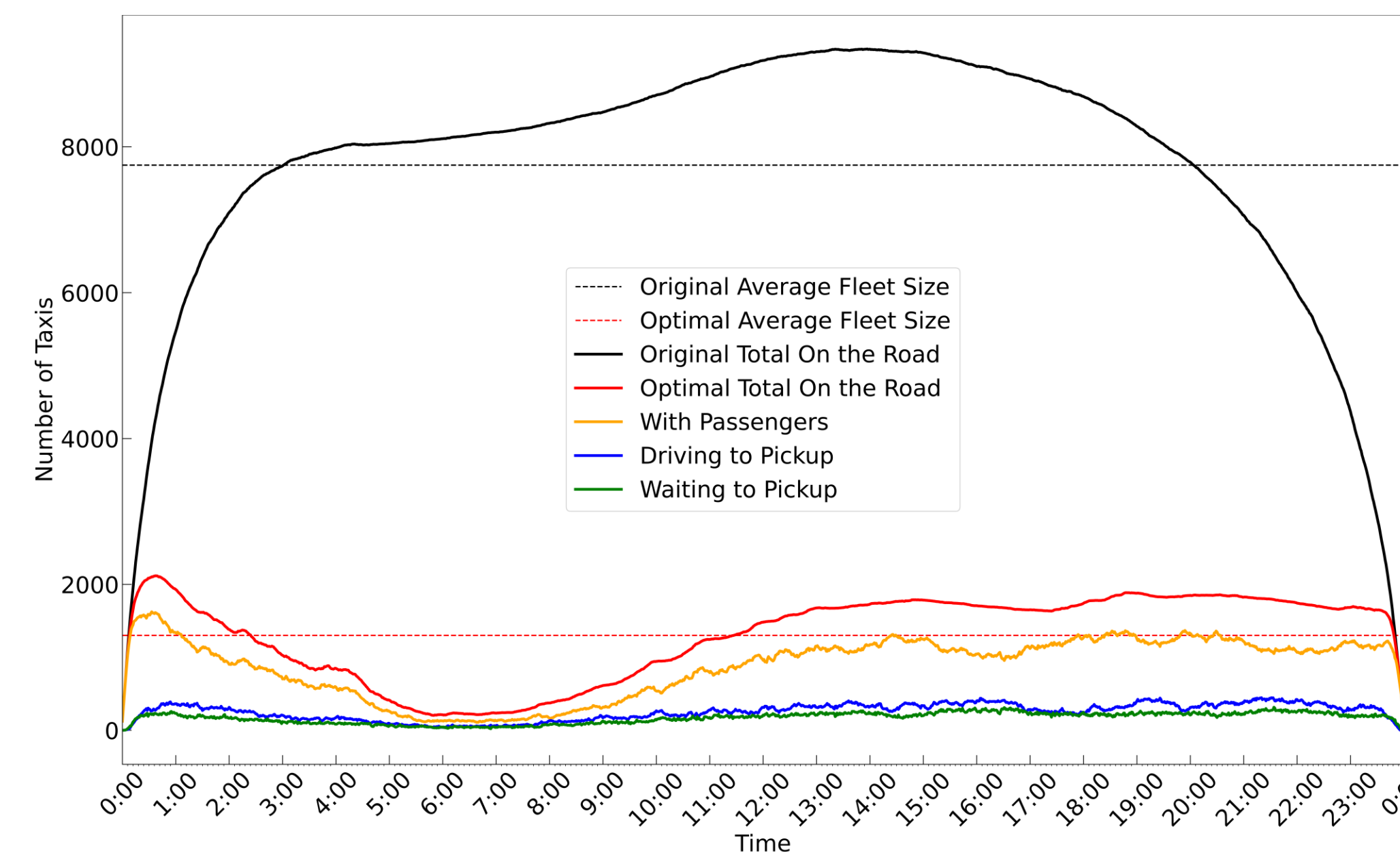
The optimal taxi routing (with  $\delta = 10$ ) is compared with the actual one

- Total number of taxis that covers all trips reduces from 11568 to 7475, a **35%** reduction
- Total trip time** and **empty trip time** are greatly reduced, from 16.1 to 4.2 hours and from 14.3 hours to 1.4 hours, respectively, on average
- To offset the reductions in trip time, the **number of trips** and **on-trip percentage** are greatly increased, from 11.3 to 17.5 and from 17.1% to 72.6% respectively, on average



A comparison of the time distribution of the optimal taxi fleets (with breakdown) and the actual taxi fleets

- On average, **number of circulating taxis** reduces from 7748 to 1300, an **83%** reduction



## References

- M. M. Vazifeh, P. Santi, G. Resta, S. H. Strogatz & C. Ratti, "Addressing the minimum fleet problem in on-demand urban mobility." *Nature* 557.7706 (2018): 534-538.
- X. Zhan, X. Qian & S. Ukkusuri, "A graph-based approach to measuring the efficiency of an urban taxi service system." *IEEE Transactions on Intelligent Transportation Systems* 17.9 (2016): 2479-2489.

## Theoretical Results

Min-max theorem is an important concept in solvable discrete optimization problems; we prove a **min-max theorem** for the taxi routing problem

- Consider a simplified setup:

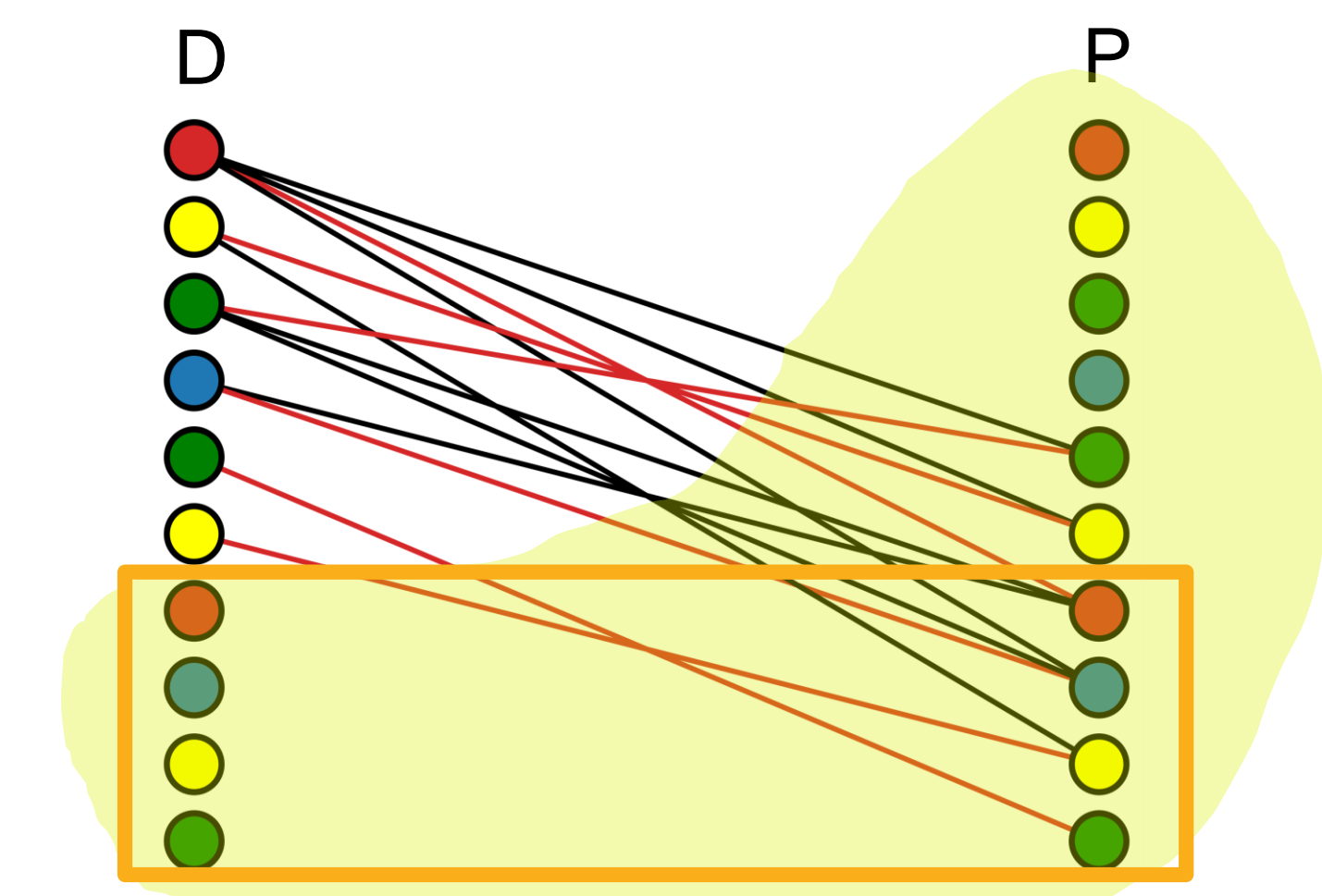
$$E = \{(d_i, T_i^d), (p_j, T_j^p) : \text{time}(d_i, p_j) \leq T_j^p - T_i^d\};$$

i.e., no upper bound on taxi waiting time ( $\delta = \infty$ )

**Definition** Any two trips  $N_i$  and  $N_j$  are considered **compatible** with each other if a taxi can reach from  $d_i$  to  $p_j$  or from  $d_j$  to  $p_i$  in time

**Proposition** Consider two disjoint sets  $P = \{p_1, \dots, p_n\}$  and  $D = \{d_1, \dots, d_n\}$  and a subset  $I \subseteq P \cup D$ , where  $|I| = n + k$ , for some  $k \geq 0$ ; then there exists a subset  $K \subseteq \{1, \dots, n\}$  where  $|K| = k$  and  $S = \cup_{i \in K} \{p_i, d_i\} \subseteq I$

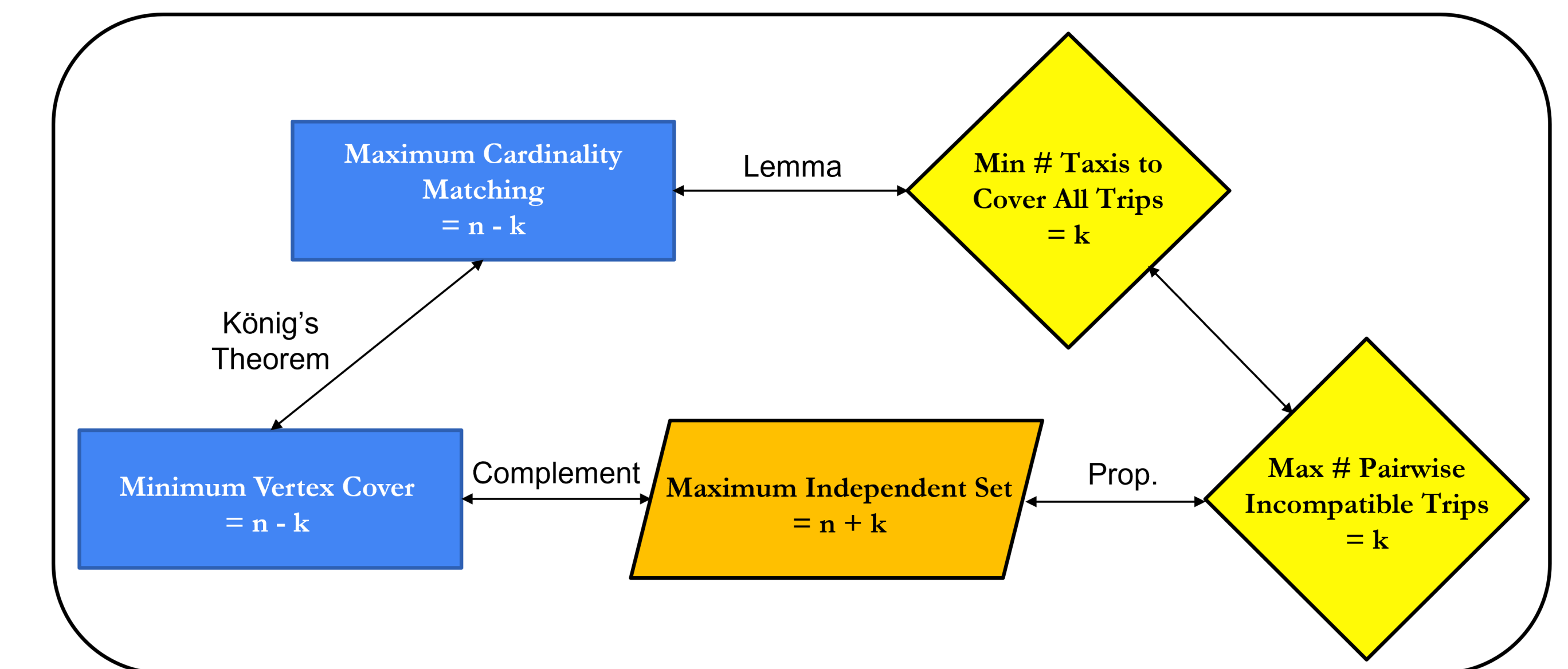
- Lower bound certificate: An **independent set** of size  $n + k$  (nodes in shaded region) leads to  $k$  trips that are **pairwise incompatible** (orange box)



**Theorem**

The **maximum size of a set of trips that are pairwise incompatible** is equal to the minimum number of taxis needed to cover all trips

*Proof.*



## Future Work

- The **dual object** when there's an upper bound ( $\delta$ ) to the waiting time for the new pickup
- An optimal taxi routing that accounts for **secondary objectives** of waiting time (e.g., minimize maximum waiting time)