

ASTRO-K: Finding Top-k Sufficiently Distinct Indoor-Outdoor Paths

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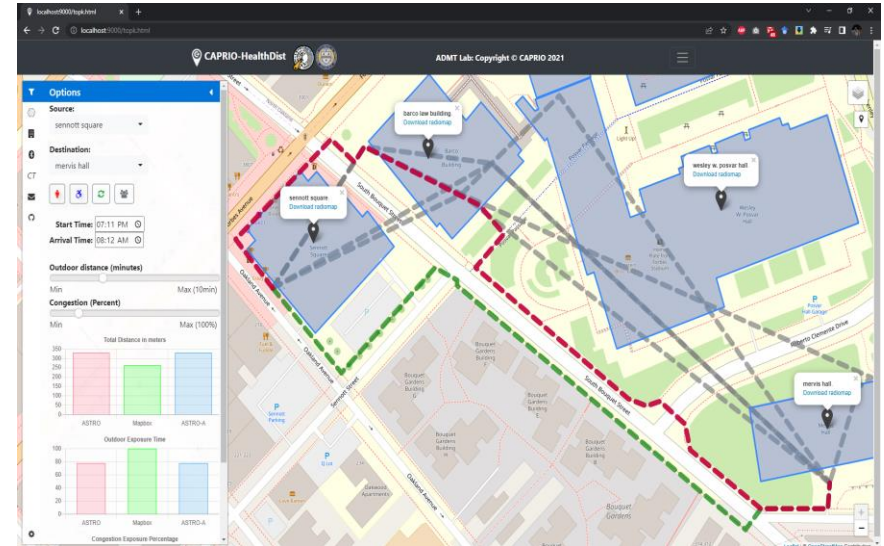
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Outline

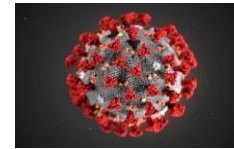
- Motivation
- Problem Formulation
- ASTRO-K
- Experimental Evaluation
- Conclusions & Future Work

Motivation

- CAPRIO: Indoor-outdoor pedestrian path recommendation system that suggests the shortest distance path while satisfying user-mobility constraints
- ASTRO: Constraint-based path-finding algorithm which factors in the predicted congestion of a space when constructing a path
- Observation: ASTRO may inadvertently contribute towards congestion problems by funneling people into a single area



**Congested
Spaces**

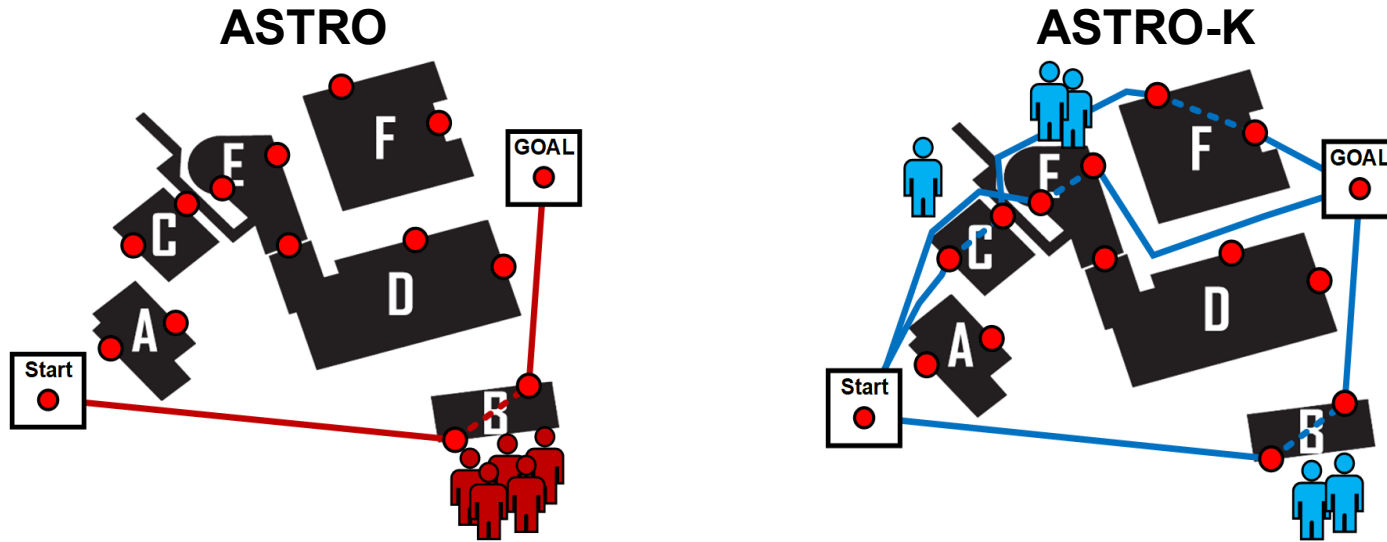


**COVID-19
Distancing**



**Mobility
disability**

Problem & Objective



- **Outdoor Vertices** – comprised of Indoor graphs and represent buildings
- **Indoor Graph** – the path between any two doors
- **Indoor Vertices** – doors allowing people to enter and exit buildings

ASTRO-K: Problem Formulation

- **ASTRO-K:** Given an Indoor-Outdoor graph, a source Outdoor vertex, a terminal Outdoor vertex, a set of constraints, and k ,
- find the **top-k sufficiently distinct paths** from the source Outdoor vertex to the terminal Outdoor vertex which satisfies the given constraints.

Congestion (C) – max congestion (w.r.t the max amount of people) within a $3m^2$ cell during a 5-minute window.

Outdoor Exposure (E) – max amount of time allowed to traverse any given edge.

Time Limit (T) – max time allowed to traverse a path

Path Similarity (θ) – max percentage of total edge distance shared between any two paths in the result set

Components of ASTRO-K

- **ASTRO**: Given an Indoor-Outdoor graph along with a source and a destination point, produce a path with the minimal total time from the source to the destination which satisfies the constraints.

- ASTRO is based on A* Algorithm that at each step it picks the node according to a score 'f' -> $f = g + h$
 - g = the movement cost
 - h = the estimated movement cost

- **ESX**: Given a source and terminal node, find k paths that are sufficiently dissimilar to each other with respect to a similarity threshold θ , and as short as possible.

- ESX is a performance-oriented heuristic algorithm that removes edges from the graph incrementally and then computing the shortest path on the updated graph

ASTRO-K Algorithm

Algorithm 2 *ASTRO-K*

Input: s : source, t : terminal, O : Outdoor Vertices, Π : Constraints;

Output: P : Best Paths

```
1: init List  $P$ , Priority Queue  $PQ$ , Set  $E_{DNR}$ , Set  $E_R$ 
2:  $P.append(ASTRO(s, t, O, E_R, \Pi))$ 
3: for  $e \in P.tail()$  do
4:    $PQ.push(e)$ 
5: while  $|P| < k$  and  $PQ$  not empty do
6:    $p \leftarrow P.tail()$ 
7:   while  $PQ$  not empty and  $Sim(p, P) > \theta$  do
8:      $e \leftarrow PQ.pop()$ 
9:     if  $e \in E_{DNR}$  then
10:      continue
11:      $E_R \leftarrow E_R \cup \{e\}$ 
12:      $p \leftarrow ASTRO(s, t, O, E_R, \Pi)$ 
13:     if  $p = NULL$  then
14:        $E_R \leftarrow E_R - \{e\}$ 
15:        $E_{DNR} \leftarrow E_{DNR} \cup \{e\}$ 
16:   if  $Sim(p, P) < \theta$  then
17:      $P.append(p)$ 
18:      $PQ.clear()$ 
19:     for  $e \in p$  do
20:        $PQ.push(e)$ 
21: return  $P$ 
```

Algorithm 1 Modified ASTRO used in Algorithm 2

Input: s : source, t : terminal, O : Outdoor Vertices, E_R : Edges Removed, Π : Constraints;

Output: p : Best Path;

```
1: init  $PQ$   $OPEN$ , Set  $CLOSED$ 
2: init Outdoor Vertex  $start$  with  $s$ 
3:  $OPEN.push(start)$ 
4: while  $OPEN$  not empty do
5:    $curr \leftarrow OPEN.pop()$ 
6:   if  $curr = NULL$  then
7:     return  $ReconstructPath(curr)$ 
8:   if  $curr \notin CLOSED$  then
9:      $CLOSED \leftarrow CLOSED \cup \{curr\}$ 
10:    for  $o_i \in \{O - CLOSED\}$  do
11:       $IndoorGraph \leftarrow o_i[IndoorGraph] - E_R$ 
12:      for  $in_i \in IndoorGraph$  do
13:         $\vec{s}_1 \leftarrow Status(curr[out], in_i, curr[g])$ 
14:        for  $out_i \in \{IndoorGraph - \{in_i\}\}$  do
15:           $now = curr[g] + \vec{s}_1[total]$ 
16:           $\vec{s}_2 \leftarrow Status(in_i, out_i, now)$ 
17:           $\vec{s} \leftarrow \vec{s}_1 + \vec{s}_2$ 
18:           $g \leftarrow curr[g] + \vec{s}[total]$ 
19:          if  $g < o_i[g]$  and  $Check(\Pi, \vec{s})$  then
20:            init Outdoor Vertex  $toAdd$ 
21:             $OPEN.push(toAdd)$ 
```

Experimental Methodology

- **Datasets:**

- **PITT:** University of Pittsburgh campus and consists of 9 buildings, 2-6 doors per building
- **UCY:** University of Cyprus campus and consists of 9 buildings, 2-7 doors per building

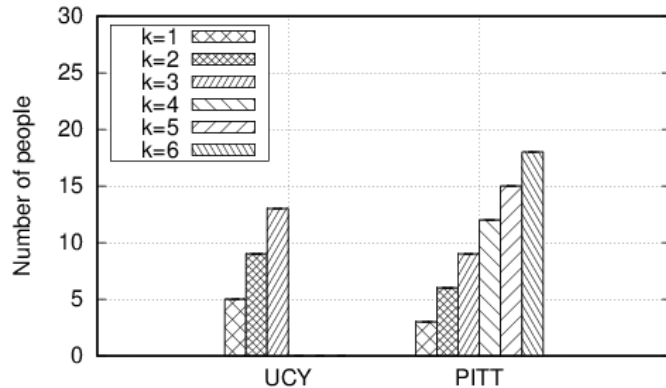
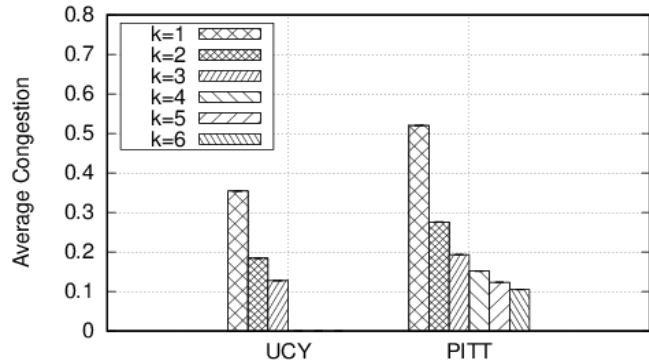
- **Congestion Data:**

- Camera analysis on a 2-hour session and generate congestion data using the procedure EPICGen generator using $\mu_a = 4$, $\sigma_a = 1$, $\mu_d = 1$, $\sigma_d = 0.2$ and a walking speed of 1.4 m/s

- **Metrics:**

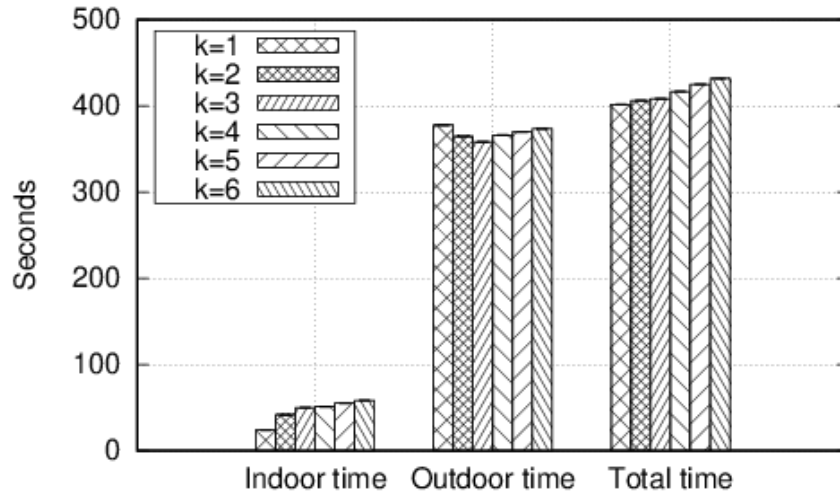
- **Average Congestion (C):** average percentage of congestion (with respect to the maximum amount of people) within a $3m^2$ cell of a building during a 5 minute window.
- **Number of People (P):** amount of people introduced within a 5 minute window.

Experiments



- **Experiment 1:** average congestion of the recommended paths is reduced by 4.5X for the PITT dataset with $K = 6$ and 2.6X for the UCY dataset with $K = 3$.
- **Experiment 2:** the number of people which can follow the recommended paths is increased by 5.9X for the PITT dataset with $K = 6$ and 2.8X for the UCY dataset with $K = 3$.

Experiments (cont.)



- **Experiment 3:** average total time of the recommended paths is increased by 7% for $K = 6$. Furthermore, the average indoor time increases by 2.4X and the average outdoor time decrease by 2% for $K = 6$.

Time	k=1	k=2	k=3	k=4	k=5	k=6
Indoor	23.94	41.62	49.88	50.54	55.09	58.12
Outdoor	377.215	364.52	358.12	365.88	369.60	373.25
Total	401.16	406.14	408.01	416.43	424.70	431.38

Conclusions & Future Work

- Major Contribution:
 - A novel top-k sufficiently distinct path finding algorithm ASTRO-K, which is able to recommend paths to more people without inadvertently congesting an area
- Future Work:
 - Include balancing or scheduling of route recommendations
 - Make Congestion a first-class constraint

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Thank You! Any Questions?

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