

UNMANNED AERIAL VEHICLES (UAVS)

- UAVs are flying vehicles able to autonomously decide their route (different from drones, that are remotely piloted)
- Historically, used in the military, mainly deployed in hostile territory to reduce pilot losses
- o Now, new applications in civilian and commercial domains:
 - weather monitoring.
 - forest fire detection,
 - traffic control.
 - emergency search and rescue









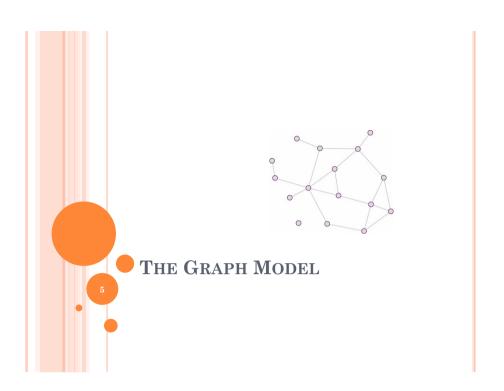


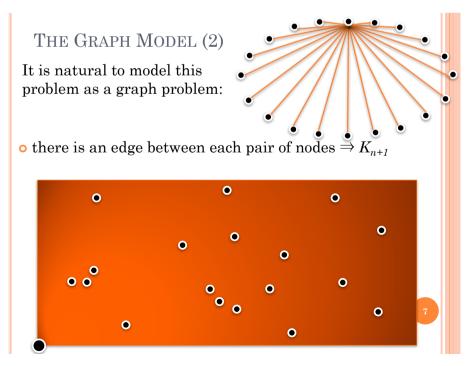


- o Let be given an AoI whose map is known
- we have a fleet of m UAVs leaving from a safe location (v_0) each with a battery B
- in the AoI there is a set $S=\{v_1, ..., v_n\}$ of sites that must be examined (e.g. crumbled buildings after a hearthquacke)
- \circ each site v_i needs a time t_i to be inspected
- each UAV must go back to v_0 in order to recharge its battery when necessary; this takes time R, typically 5-10 times B
- \circ we want to overfly $v_1, ..., v_n$ "as soon as possible" in order to collect data and possibly save people





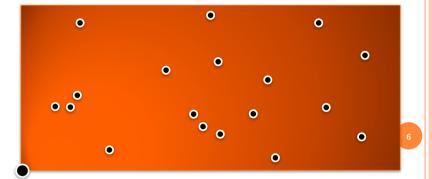




THE GRAPH MODEL (1)

It is natural to model this problem as a graph problem:

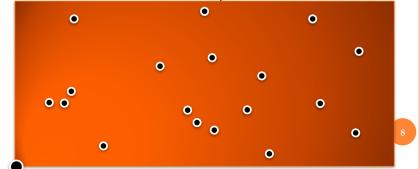
• sites v_1 , ..., v_n + the depot v_0 are the n+1 nodes of the graph



THE GRAPH MODEL (3)

It is natural to model this problem as a graph problem:

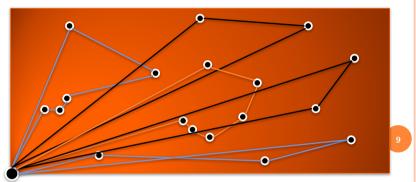
- Each UAV has a flying+inspection time bounded by *B*.
- o for each pair of sites (v_i, v_j) we assume their distance (stored as an edge weight function $w(u_i, u_j)$) as the time a UAV needs to go from u_i to u_j .



THE GRAPH MODEL (4)

- o each UAV is characterized by a different color
- each UAV flies along a cycle (colored with the UAV color) and visits as many sites as it can (w.r.t. its battery constraint *B*), it goes back to the depot to recharge its battery (with time *R*) and it leaves again...

All sites need to be visited in the "shortest time".



THE GRAPH MODEL (6)

Similarities with many problems:

mTSP -multiple Traveling Salesperson

- ullet m salespersons must overall cover n cities,
- o objective: minimize the total length of the path



o no visiting times nor battery constraint



THE GRAPH MODEL (5)

What does it mean that the sites should be visited "as soon as possible"?

Different possibilities for the optimization function:

- Minimize the Total completion Time
- Minimize the Average Waiting Time
- Minimize the number of cycles
- o ...
- Note: Minimize the Overall Energy Consumption has no meaning

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THE GRAPH MODEL (7)

Similarities with many problems (cntd):

kTRPR -k-Traveling Repairperson Problem with Repairtimes

- given n points, construct k cycles, each starting at a common depot and together covering all the n points calling the *latency* of a point the distance traveled (or the time elapsed) before visiting that point
- o objective: minimize the sum of all latencies
- o no battery constraint



THE GRAPH MODEL (8)

Similarities with many problems (cntd):

mTRPD -multiple Traveling Repairperson Problem with Distance Constraints

- \circ *k* repairpersons have all together to visit all the *n* customers
- they are not allowed to traverse a distance longer than a predetermined limit;
- Objective: minimize the total waiting time of all custemers
- No repairtimes and not trivial to extend a solution by just adding them
- o no. of cycles fixed to k

THE GRAPH MODEL (10)

Similarities with many problems (cntd):

variants of MKP - Multiple Knapsack Problem

- assigns a subset of *n* items to *m* distinct knapsacks,
- Objective: the total profit sum of the selected item is maximised
- o no multiple rounds



THE GRAPH MODEL (9)

Similarities with many problems (cntd):

variants of **VRP** -vehicle routing problem

• Similar to mTRPD but there usually is a constraint on the number of visited customers per vehicle

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THE GRAPH MODEL (11)

Similarities with many problems (cntd):

TOP -team orienteering problem

- equivalent to the first round of our problem
- o Objective: maximize the no. of covered sites
- Repeat many times until all sites have been covereα does not seem a good idea...
- NOTE: From all these similarities we deduce that the problem is NP-hard and we cannot exploit any known result...

MONITORING AN AREA BY UAVS

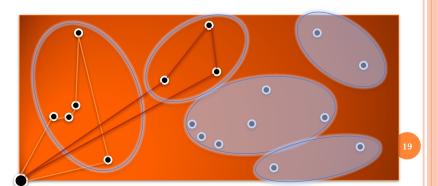
We have to study the problem by itself, going in several possible directions:

- o due to its NP-completeness, approximate algorithms:
 - based on three main pahses:
 - clustering/matroid theory (greedy)
 - o approximating TSP
 - scheduling
- MILP formulation
- o reduction of the dimension of the problem

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MONITORING AN AREA BY UAVS (2)

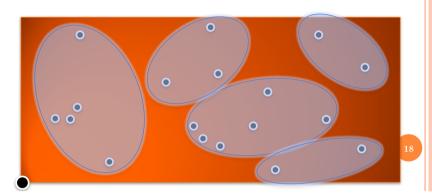
approximating TSP: constructing a cycle covering all sites in each cluster (in fact performed together with the clustering, to guarantee the battery constraint)



MONITORING AN AREA BY UAVS (1)

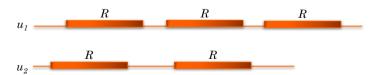
General idea: problem divided into three phases:

clustering: partition the sites so that each set can be covered (traversal+visiting times) within battery B.



MONITORING AN AREA BY UAVS (3)

scheduling: all cycles must be distributed to UAVs so to guarantee min completion time or min latency



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MONITORING AN AREA BY UAVS (4)

Each one of the three phases can be implemented in several ways providing different solutions...

Problem 1: compare all the provided solutions in terms of goodness

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MONITORING AN AREA BY UAVS (6)

Provide a MILP formulation in order to compare optimum solutions of (small) instances with the approximate ones:

- in the graph model, the depot is transformed into two nodes, v_0^s and v_0^t and the graph is oriented
- o define a variable family x_{ij}^{k} , i,j=0,...,n, k=1,...,c such that it is =1 iff edge (i,j) is used by cycle k
- o define a variable family z_i^k , i=0, ..., n, k=1, ..., c such that it is =1 iff cycle k passes through node i

o ...

MONITORING AN AREA BY UAVS (5)

Instead of clustering sites, we can:

- ullet enumerate all possible cycles passing through the depot that can be covered within battery B
- o solve a min set cover.

Exploiting the fact that this system is a matroid, a greedy approach guarantees a very good approximation ratio but...

The no. of enumerated cycles is exponential in general...

Problem 2: reduce the space of the cycles so that the approximation ratio does not increase too much

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MONITORING AN AREA BY UAVS (7)

Provide a MILP formulation (cntd) constrains:

- $\sum_{k=1}^{c} z_i^k \le 1$, ∀i = 1, ..., n (every site visited at least once)
- o $\sum_{(v_i,v_j)\in E} w(v_i,v_j) x_{ij}^k + \sum_{i=1}^n t_i z_i^k \leq B, \forall k=1,...,c$ (battery constraint)
- o $x_{ij}^k \le z_i^k \ \forall (v_i, v_j) \in E, \ \forall k = 1, ..., c$ (the k-th cycle passes through site i iff it is in fact assigned to it)
- o $\sum_{i=0}^{n} x_{ij}^{k} = 1 \ \forall j, k \text{ (only 1 edge per cycle enters in } v_i)$
- $\sum_{j=0}^{n} x_{ij}^{k} = 1 \ \forall i, k \text{ (only 1 edge per cycle comes out from } v_{j}$

o ...

MONITORING AN AREA BY UAVS (8)

Provide a MILP formulation (cntd) constrains (cntd):

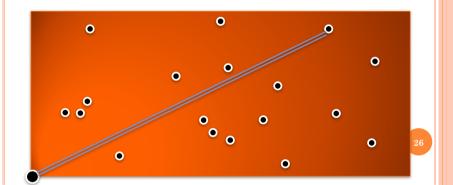
- o $\sum_{j=1}^{n} x_{0j}^{k} = 1$ and $\sum_{i=1}^{n} x_{i0}^{k} = 1 \ \forall k = 1, ..., c$ (all c cycles go out from v_0^s and enter in v_0^t)
- subtour elimination constraints
- Objective: minimize c [usually to avoid...]
- NOTE: this formulation solves only the first two phases: the scheduling is missing...

Problem 3: provide a correct and complete formulation. ...25

MONITORING AN AREA BY UAVS (9)

Reduction of the dimension of the instance:

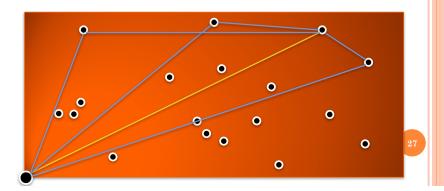
Property 1: if $\exists i$ s.t. 2 $w(v_0, v_i) + t_i = B \Rightarrow$ cycle $v_0 - v_i - v_0$ is in every solution.



MONITORING AN AREA BY UAVS (10)

Reduction of the dimension of the instance (cntd):

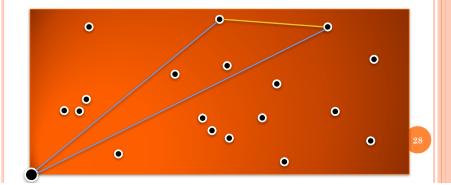
Property 2: if $\forall j$ it holds $w(v_0, v_i) + t_i + w(v_i, v_j) + t_j + w(v_j, v_0) > B$ \Rightarrow cycle $v_0 - v_i - v_0$ is in every solution.



MONITORING AN AREA BY UAVS (11)

Reduction of the dimension of the instance (cntd):

Property 3: if $\exists i,j$ s.t. $w(v_0,v_i)+t_i+w(v_i,v_j)+t_j+w(v_j,v_0)>B \Rightarrow$ edge (v_i,v_i) cannot enter in any solution.



MONITORING AN AREA BY UAVS (12)

Reduction of the dimension of the instance (cntd):

The main idea is that, before solving our problem on the given instance, we can reduce its dimension by forcing to be inside the solution the edges indicated by Properties 1 and 2, and to be outside the solution the edges indicated by Property 3.

Problem 4: given a general (e.g. random, real life, etc.) instance, how much can we expect to reduce its dimension?

OTHER OPEN PROBLEMS

- o determining a tight approx ratio
- o introducing cooperation
- better exploiting UAVs' capabilities
- o ...

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