Shortest path problem from the perspective of duality

Reporter: Maocan Song

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1 Problem statement

Problem statement	Find a least-cost path from origin node to destination node. (Single commodity problem, its quantity equals 1)
Input	 A network with nodes and links Link costs an node costs Origin and destination
Output	The shortest path from origin to destination
Objective function	$\min \sum_{(i,j)\in A} c_{ij} x_{ij}$
Constraints	$\sum_{i:(i,j)\in A} x_{ij} - \sum_{i:(j,i)\in A} x_{ji} = \begin{cases} 1 & i = 0\\ -1 & i = D\\ 0 & o.w. \end{cases}$ $x_{ij} \ge 0, \forall ij \in A$
Problem type	Linear program

2. Lagrangian relaxed problem

$$L(\mu) = \min \sum_{(i,j) \in A} c_{ij} x_{ij} + \mu_O \left(\sum_{O:(O,j) \in A} x_{Oj} - 1 \right) + \sum_{i \in E/\{O,D\}} \mu_i \left(\sum_{i:(i,j) \in A} x_{ij} - \sum_{i:(j,i) \in A} x_{ji} \right) + \mu_D \left(- \sum_{D:(D,j) \in A} x_{Dj} + 1 \right)$$

$$= (\mu_D - \mu_O) + \min \sum_{(i,j) \in A} c_{ij} x_{ij} + \mu_O \sum_{O:(O,j) \in A} x_{Oj} - \mu_D \sum_{D:(D,j) \in A} x_{Dj} + \sum_{i \in E/\{O,D\}} \mu_i \left(\sum_{i:(i,j) \in A} x_{ij} - \sum_{i:(j,i) \in A} x_{ji} \right)$$

$$L_2(\mu) = \min \sum_{(i,j) \in A} c_{ij} x_{ij} + \mu_O \sum_{O:(O,j) \in A} x_{Oj} - \mu_D \sum_{D:(D,j) \in A} x_{Dj} + \sum_{i \in E/\{O,D\}} \mu_i \left(\sum_{i:(i,j) \in A} x_{ij} - \sum_{i:(j,i) \in A} x_{ji} \right)$$

$$L_{2-ij}(\mu) = \min c_{ij} x_{ij} + \mu_i x_{ij} - \mu_j x_{ij} = \min(c_{ij} + \mu_i - \mu_j) x_{ij}$$

$$L(\mu) = L_1(\mu) + L_2(\mu) = L_1(\mu) + \sum_{(i,j) \in A} \min(c_{ij} + \mu_i - \mu_j) x_{ij}, x_{ij} \ge 0$$

$$Case 1: c_{ij} + \mu_i - \mu_j < 0 \ \exists ij \in A, L(\mu) = -\infty \ \Xi \mathcal{F}$$

$$E \boxtimes B \text{ BB} \text{ by } \exists B \text{ by } \exists$$

The relationship with label-correcting algorithm

5.2 OPTIMALITY CONDITIONS

As noted previously, label-correcting algorithms maintain a distance label d(j) for every node $j \in N$. At intermediate stages of computation, the distance label d(j) is an estimate of (an upper bound on) the shortest path distance from the source node s to node j, and at termination it is the shortest path distance. In this section we develop necessary and sufficient conditions for a set of distance labels to represent shortest path distances. Let d(j) for $j \neq s$ denote the length of a shortest path from the source node to the node j [we set d(s) = 0]. If the distance labels are shortest path distances, they must satisfy the following necessary optimality conditions:

$$d(j) \le d(i) + c_{ij}, \quad \text{for all } (i, j) \in A.$$
 (5.1)

These inequalities state that for every arc (i, j) in the network, the length of the shortest path to node j is no greater than the length of the shortest path to node i plus the length of the arc (i, j). For, if not, some arc $(i, j) \in A$ must satisfy the condition $d(j) > d(i) + c_{ij}$; in this case, we could improve the length of the shortest path to node j by passing through node i, thereby contradicting the optimality of distance labels d(j).

These conditions also are sufficient for optimality, in the sense that if each d(j) represents the length of some directed path from the source node to node j and this solution satisfies the conditions (5.1), then it must be optimal. To establish this result, consider any solution d(j) satisfying (5.1). Let $s = i_1 - i_2 - \ldots - i_k = j$ be any directed path P from the source to node j. The conditions (5.1) imply that

$$d(j) = d(i_k) \leq d(i_{k-1}) + c_{i_{k-1}i_k},$$

$$d(i_{k-1}) \leq d(i_{k-2}) + c_{i_{k-2}i_{k-1}},$$

An illustrative example in GAMS

```
variable z;
positive variable x(i,j);
                                                                                       ↑0 k
↓15 l
equations
flow_balance_origin(i)
flow balance destination(i)
flow balance otherwise(i)
obj..z =e= sum(i,sum(j$cost(i,j),cost(i,j)*x(i,j)));
flow balance_origin(i)$origin_node(i).. sum(j$cost(i,j),x(i,j)) =e= 1;
flow_balance_destination(i)$destination_node(i)..sum(j$cost(j,i),x(j,i)) =e= 1;
flow_balance_otherwise(i)$(intermediate_node(i)).. sum(j$cost(i,j),x(i,j))-sum(j$cost(j,i),x(j,i)) =e= 0;
Model SPP /all/;
Solve SPP using LP minimizing z;
Tried aggregator 1 time.
LP Presolve eliminated 8 rows and 10 columns.
 All rows and columns eliminated.
Presolve time =
                           0.00 sec.
LP status(1): optimal
 Optimal solution found.
 Objective :
                              47.000000
```

```
variable z;
variables y(i);
equations
obi
optimal_cost(i,j)
obj.. z = e = y("7") - y("1");
optimal_cost(i,j)$cost(i,j).. y(j)-y(i) =l= cost(i,j);
model SPP /all/;
solve SPP using LP maximizing z;
All rows and columns eliminated.
Presolve time =
                     0.00 sec.
LP status(1): optimal
Optimal solution found.
```

47.000000

Objective :

3. TSP and the dual problem

Primal problem	
Objective function	$\min \sum_{(i,j)\in A} c_{ij} x_{ij}$
Constraints	$\sum_{i:(i,j)\in A} x_{ij} - \sum_{i:(j,i)\in A} x_{ji} = \begin{cases} 1 & i=0\\ -1 & i=D\\ 0 & o.w. \end{cases}$ $\sum_{i:(i,j)\in A} x_{ij} = 1, i \in A_{task}$ $x_{ij} \ge 0, \forall ij \in A$
Dual problem	
Objective function	$\max \mu_D - \mu_O - \sum_{i \in A_{task}} w_i$
Constraints	$\begin{cases} c_{ij} + \mu_i - \mu_j + w_i \ge 0, & i \in A_{task} \\ c_{ij} + \mu_i - \mu_j \ge 0, & o.w. \end{cases}, \forall ij \in A$

Thanks

- TSP-dual problem
- Space-time network
- In State-space-time network?