

# 1 The PC-TSP algorithm approach

The prize-collecting travelling salesperson is a variant of the familiar (NP) problem in which, instead of necessarily seeking a minimum weight cycle through every vertex in a graph (as required in the ordinary TSP), a penalty can be paid instead for missing vertices.

Anderson et al. propose an algorithm for solving KEPs which is inspired by the PC-TSP and which draws on algorithms for solving it. In particular, they suggest that this similarity is reasonable, given that finding longing chains in is useful in both the PC-TSP and KEPs.

## 1.1 Model

As under the Recursive Algorithm, prospective participants in a kidney exchange are associated with the vertices of a directed graph  $\vec{G} = (V, \vec{E})$ . Each donor-recipient pair is represented by a vertex  $v \in P \subseteq V$  and each altruistic donor also by a vertex  $v \in N \subseteq V$ . A directed edge  $e = (v, u)$  connects vertex  $v$  to a vertex  $u$  exactly when there is a compatible match from the donor at vertex  $v$  to the recipient at vertex  $u$ . Each such edge is also assigned a weight  $w(e)$ , intended to capture the importance of the (potential, compatible) transplant it represents.

For each vertex  $v$ , the set  $\text{in}(v)$  is of edges from compatible donors to  $v$ , while the set  $\text{out}(v)$  is of edges from  $v$  to compatible recipients. By construction,  $\text{in}(v)$  is empty for each  $v \in N$ .

The aim, as before, is to find the maximum weight subgraph of  $\vec{G}$  subject to the constraint that cycles are of length at most  $k \in \mathbf{Z}^{\geq 0}$ .

Each edge in the  $\vec{G}$  is associated with a binary decision variable  $y_e$ , which is equal to 1 if edge  $e$  is selected *unless* that edge forms part of a cycle of length no more than  $k$ . Denote by  $\mathcal{C}_k$  the cycles in  $\vec{G}$  of length no greater than  $k$ . Further binary decision variables  $z_C$  are thus created for each  $C \in \mathcal{C}_k$ , with  $z_C = 1$  exactly when the cycle  $C$  is selected. The reason for this distinction between edges used in such cycles and those edges used which are not in cycles in  $\mathcal{C}_k$  becomes apparent in the discussion of the constraints below.

### 1.1.1 Objective function

The objective function is given by

$$\sum_{e \in E} w_e y_e + \sum_{C \in \mathcal{C}_k} w_C z_C,$$

which we seek to maximize.

The weights  $w_C$  are simply calculated as the sum of the weights of the edges which make up the cycle  $C$ .

### 1.1.2 Constraints

Anderson et al. define (for every vertex  $v \in V$ ) the *flow in*  $f_v^i$  and *flow out*  $f_v^o$  by

$$f_v^i = \sum_{e \in \text{in}(v)} y_e$$

and

$$f_v^o = \sum_{e \in \text{in}(v)} y_e.$$

These values are integer and will, in fact, be restricted to be binary.

Note particularly that when  $v$  is part of a cycle  $C \in \mathcal{C}_k$ ,  $f_v^i$  and  $f_v^o$  are both set to 0. These values can therefore in general be interpreted as indicating whether an altruistic donor/donor-recipient pair is involved in kidney donation that is not part of a cycle up to length  $k$ .

The constraint

$$f_v^o + \sum_{C \in \mathcal{C}_k(v)} z_C \leq f_v^i + \sum_{C \in \mathcal{C}_k(v)} z_C \leq 1, \quad \forall v \in P,$$

where  $\mathcal{C}_k(v)$  is the set of cycles of length at most  $k$  which include the vertex  $v$ , serves to ensure that each donor-recipient pair gives and receives at most one kidney and further that no such donation is made unless a kidney is received.

Likewise, imposing the constraint that  $f_v^o \leq 1, \forall v \in N$  ensures that each altruistic donor gives at most one kidney.

The most involved constraint imposed requires that

$$\sum_{e \in \text{in}(S)} y_e \geq f_v^i, \quad \forall S \subseteq P, \quad \forall v \in S, \quad (1)$$

where  $\text{in}(S)$  is the set of directed edges ended at vertices in  $S$ . The authors indicate that this constraint is directly inspired by similar constraints for PC-TSP !INSERT REFERENCE!.

It permits cycles of length up to  $k$  and arbitrarily long chains but will make a proposed solution of the KEP infeasible if it contains a cycle longer than  $k$ :

- If  $v$  is part of a short cycle (i.e. part of  $C \in \mathcal{C}_k$ ), then  $f_v^i = 0$  and the constraint will be satisfied (**for this**  $v$  and any  $S$ ) whatever the values of  $y_e$ .
- If  $f_v^i = 1$ , then either  $v$  is part of a long cycle  $\bar{C}$  or a chain starting from an altruistic donor (since  $v \in P$  cannot start a chain). If  $v$  is in a long cycle, setting  $S = \bar{C}^1$ , we see that the constraint is not satisfied.
- On the other hand, the constraint is not broken by  $v$  in a long chain: whatever selection of  $S \subseteq P$  is made with  $v \in S$ , a vertex  $v$  in a chain requires a flow from an altruistic donor to  $S$ , so  $\sum_{e \in \text{in}(S)} y_e \geq 1$  is guaranteed.

AJM NOTE: I think this might give us a different way to achieve the same thing without looking at every subset of  $P$ . More thought required.

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<sup>1</sup>This satisfies  $S \subseteq P$  as only donor-recipient pairs can be involved in cycles.

## 1.2 Implementation

It is not possible to directly introduce all of the constraints in Equation (1) to the solver, as there are exponentially many such constraints. Instead, a method is required to detect whether these constraints have been satisfied, which Anderson et al. propose achieving by solving a max-flow min-cut problem on an augmented graph.<sup>2</sup>

To form the augmented (weighted, directed) graph, an extra node  $s$  (a ‘source’) is added to the vertex set  $V$  and an edge connects  $s$  to every altruistic donor  $n \in N$ . Weights are associated with all the edges in this augmented graph. In particular, every new edge  $(s, n)$  gets a weight of 1, while the remaining edges are weighted by  $y_e$ .

In this way, the edges involved in short cycles are weighted 1, while the edges of chains and those of the potentially lurking long cycles are weighted 1.

If for any  $v \in P$ , a cut can be found to separate  $s$  from  $v$  which has weight less than  $f_v^i$ , then we have found a long cycle. Notice that we only actually have to check those  $v$  for which  $f_v^i > 0$ .

This approach works because every chain starts at an  $n \in N$  and  $s$  was connected to every  $n \in N$  and all the edges in these ‘superchains’ have weight 1 while the edges of long cycles are also weighted 1 but they are disconnected from  $s$ .

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<sup>2</sup>AJM NOTE: This might actually only be explained in the PNAS paper, and in that case only really explained in the Supplemental Material - I haven’t properly compared with the INFORMS paper just yet.