

APPENDIX I
PROOF OF LEMMA II.1

(i) Consider the following set

$$\tilde{\mathcal{P}} = \{\mathbf{p} \mid p_{ijk} \in \{0, \bar{p}_{ijk}\}, \mathbf{p} \in \mathcal{P}\}$$

and note that every feasible $\mathbf{p} \in \mathcal{P}$ can be expressed as a convex combination of the points in $\tilde{\mathcal{P}}$. Thus $\text{Conv}(\mathcal{P})$ is finitely generated by the points in $\tilde{\mathcal{P}}$ and is therefore a polyhedron. Clearly, it is also bounded, hence a polytope (i.e., a bounded polyhedron). Since \mathcal{R} is a linear mapping of \mathcal{P} , the same argument establishes that $\text{Conv}(\mathcal{R})$ is a polytope.

(ii) Next, let $\mathbf{r} \in \text{Conv}(\mathcal{R})$ and suppose that the extreme points of $\text{Conv}(\mathcal{R})$ are $\mathbf{r}^1, \dots, \mathbf{r}^L$. These extreme points are also elements of \mathcal{R} , so they can be written as $\mathbf{r}^n = \mathbf{H}\mathbf{p}^n$ for some $\mathbf{p}^n \in \mathcal{P}$ and $n = 1, \dots, L$. Thus, for $\alpha_n \geq 0$ satisfying $\sum_{n=1}^L \alpha_n = 1$,

$$\mathbf{r} = \sum_n \alpha_n \mathbf{r}^n = \sum_n \alpha_n \mathbf{H}\mathbf{p}^n = \mathbf{H}\mathbf{p},$$

where $\mathbf{p} = \sum_n \alpha_n \mathbf{p}^n \in \text{Conv}(\mathcal{P})$. Now, let $\mathbf{p} \in \text{Conv}(\mathcal{P})$ and let $\mathbf{p}^1, \dots, \mathbf{p}^J$ be the extreme points of $\text{Conv}(\mathcal{P})$. For $\beta_i \geq 0$ satisfying $\sum_i \beta_i = 1$, $\mathbf{p} = \sum_i \beta_i \mathbf{p}^i$. These extreme points are also elements of \mathcal{P} . Set $\mathbf{r}^i = \mathbf{H}\mathbf{p}^i$; we have $\mathbf{r}^i \in \mathcal{R}$. It follows

$$\mathbf{H}\mathbf{p} = \sum_i \beta_i \mathbf{H}\mathbf{p}^i = \sum_i \beta_i \mathbf{r}^i \in \text{Conv}(\mathcal{R}).$$

(iii) Let \mathbf{r} be an extreme point of $\text{Conv}(\mathcal{R})$. Then there exists a cost vector $\mathbf{c} \in \mathbb{R}^{(N+M)^2 K}$ such that \mathbf{r} is the unique minimizer of $\mathbf{c}'\mathbf{x}$ over $\mathbf{x} \in \text{Conv}(\mathcal{R})$. Using part (ii), this latter problem is

$$\begin{aligned} \min \quad & \mathbf{c}'\mathbf{x} \\ \text{s.t.} \quad & \mathbf{x} = \mathbf{H}\mathbf{p}, \mathbf{p} \in \text{Conv}(\mathcal{P}). \end{aligned} \quad (28)$$

Let $(\mathbf{r}, \mathbf{p}^*)$ be an optimal solution. Then \mathbf{p}^* is an optimal solution to

$$\begin{aligned} \min \quad & \mathbf{c}'\mathbf{H}\mathbf{p} \\ \text{s.t.} \quad & \mathbf{p} \in \text{Conv}(\mathcal{P}). \end{aligned} \quad (29)$$

(If not, there exists an optimal solution $\tilde{\mathbf{p}}$ of (29) and $\mathbf{H}\tilde{\mathbf{p}} \neq \mathbf{r}$ is optimal for (28), which contradicts the fact that \mathbf{r} uniquely solves (28).) Now, there also exists some extreme point \mathbf{p}^i of $\text{Conv}(\mathcal{P})$ which is optimal for (29). In particular, $\mathbf{c}'\mathbf{H}\mathbf{p}^i = \mathbf{c}'\mathbf{H}\mathbf{p}^* = \mathbf{c}'\mathbf{r}$ and $\mathbf{H}\mathbf{p}^i$ also solves (28). Since \mathbf{r} is a unique solution of (28) we have $\mathbf{r} = \mathbf{H}\mathbf{p}^i$.

APPENDIX II

THE DECOMPOSITION ALGORITHM FOR UTILITY
MAXIMIZATION SUBJECT TO A LIFETIME CONSTRAINT

Suppose we have an extreme point \mathbf{p}^1 of $\text{Conv}(\mathcal{P})$, which belongs to $\{\mathbf{p} \mid \mathbf{H}\mathbf{p} \in \mathcal{S}, \mathbf{C}\mathbf{p} \leq \chi/T\}$ and let $m \in \{1, \dots, J\}$, then the restricted master problem at the m th

iteration is

$$\begin{aligned} \min \quad & -F(\mathbf{H}\mathbf{p}) \\ \text{s.t.} \quad & \mathbf{p} - \sum_{n=1}^m \alpha_n \mathbf{p}^n = 0, \\ & \sum_{n=1}^m \alpha_n = 1, \\ & \sum_{j=1}^{N+M} \frac{p_{ijk} G_{ij} - p_{jik} G_{ji}}{\eta \ln 2} = 0, \quad \forall i \neq s(k), d(k), \forall k, \\ & \mathbf{A}\mathbf{H}\mathbf{p} \leq \mathbf{b}, \\ & \mathbf{C}\mathbf{p} \leq \chi/T, \\ & \alpha_n \geq 0, \quad n = 1, \dots, m, \end{aligned} \quad (30)$$

and the corresponding subproblem is

$$\begin{aligned} \max \quad & \lambda' \mathbf{p} \\ \text{s.t.} \quad & \mathbf{p} \in \mathcal{P}, \end{aligned} \quad (31)$$

with cost vector $\lambda = \lambda^{(m)}$, where as before we have $(\mathbf{p}^{(m)}, \alpha^{(m)}; \lambda^{(m)}, \mu^{(m)}, \nu^{(m)}, \sigma^{(m)}, \xi^{(m)})$ as the optimal primal-dual pair for the restricted master problem (30) with $\xi^{(m)}$ being the optimal dual variable corresponding to the lifetime constraint.

Following the same recipe as described in Fig. 1, we can obtain an optimal solution to problem (24) in a finite number of iterations. The argument is almost identical and is therefore omitted for brevity. Furthermore, problem (31) is still equivalent to the maximum weighted matching problem constructed in a similar way as in Section V, and is solvable in polynomial time. In particular, to solve problem (31), we construct the same undirected graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$. Let $\delta_{ijk} = \lambda_{ijk} \bar{p}_{ijk} \quad \forall i, j, k$, and the weight for each edge $(i, j) \in \mathcal{E}$ is given by

$$w_{ij} = \begin{cases} \max_{k=1, \dots, K} \max\{\delta_{ijk}, \delta_{jik}, 0\}, & \forall i, j \in \mathcal{A}, \\ \max_{k=1, \dots, K} \max\{\delta_{i, N+l, k}, 0\}, & \forall i \in \mathcal{A}, j \in \mathcal{B}_l, \\ \max_{k=1, \dots, K} \max\{\delta_{j, N+l, k}, 0\}, & \forall i \in \mathcal{B}_l, j \in \mathcal{A}, \\ 0 & \text{otherwise.} \end{cases}$$

Furthermore, we construct the set \mathcal{K} as follows. For each $1 \leq i \leq N$, $1 \leq j \leq N + M$, we select only one, if any, k satisfying the conditions

$$k = \begin{cases} \text{argmax}_{t=1, \dots, K} \max\{\delta_{ijt}, \delta_{jit}, 0\}, & \text{if } j \leq N, \\ \text{argmax}_{t=1, \dots, K} \max\{\delta_{ijt}, 0\}, & \text{otherwise,} \end{cases}$$

and

$$\begin{cases} \delta_{ijk} = \max\{\delta_{ijk}, \delta_{jik}, 0\} > 0, & \text{if } j \leq N, \\ \delta_{ijk} > 0, & \text{otherwise,} \end{cases}$$

and let (i, j, k) be an element of \mathcal{K} . Given the graph \mathcal{G} constructed above, we can obtain the optimal solution of (31) in the same way as in Section V.