LINFO2266 - Solutions

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1 Lagrangian Relaxation

1.1 Set Covering Problem

1. We introduce one Lagrangian multiplier for each constraint to obtain the Lagrangian relaxation:

$$\min \sum_{i=1}^{m} c_i x_i + \sum_{j=1}^{n} \lambda_j \left(1 - \sum_{\substack{i=1\\j \in S_i}}^{m} x_i \right)$$
 (1)

with $\lambda_j \geq 0, j \in \{1, ..., n\}$. Note how we integrated the constraints: for any feasible solution of the original problem, we will have

$$1 - \sum_{\substack{i=1\\j \in S_i}}^{m} x_i \le 0 \tag{2}$$

which means a *negative* value will be added to the objective function and that we will thus obtain a *lower bound* on the optimal value of the problem.

For clarity, we rewrite eq. (1) as:

$$\min \sum_{i=1}^{m} \left(c_i - \sum_{j \in S_i} \lambda_j \right) x_i + \sum_{j=1}^{n} \lambda_j \tag{3}$$

For fixed values of the Lagrangian multipliers λ_j , we can obtain a lower bound by solving eq. (3). This problem is very simple to solve if we notice that:

- each x_i is multiplied by a fixed coefficient $\left(c_i \sum_{j \in S_i} \lambda_j\right)$
- the second term $\left(\sum_{j=1}^{n} \lambda_j\right)$ is constant

In order to minimize the objective function, it is sufficient to select the sets which have a negative coefficient:

$$x_i = \begin{cases} 1 & \text{if } c_i - \sum_{j \in S_i} \lambda_j \le 0\\ 0 & \text{otherwise.} \end{cases}$$
 (4)

- 2. By applying exactly what we described above, we get a lower bound value of 4.2.
- 3. We adapt the subgradient procedure covered in the lectures for this particular problem, and for multiple Lagrange multipliers in algorithm 1.

Algorithm 1 Subgradient procedure for the Set Covering Problem.

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1: \mathcal{L}^* \leftarrow -\infty, k \leftarrow 1, \mu_0 \leftarrow 1
 2: \lambda_{0,j} \leftarrow 0, 1 \le j \le n
  3: C^* \leftarrow a trivial solution, or none
  4: while \mu_k \geq \epsilon do
          Compute cover C_k with weights \left(c_i - \sum_{j \in S_i} \lambda_{k,j}\right) for 1 \leq i \leq m
          \mathcal{L}_k \leftarrow lagrangianValue(\mathcal{C}_k) (given by eq. (3))
 6:
          if \mathcal{L}_k > \mathcal{L}^* then
 7:
             \mathcal{L}^* \leftarrow \mathcal{L}_k
 8:
          end if
 9:
          if C_k is feasible and value(C_k) < value(C^*) then
10:
             \mathcal{C}^* \leftarrow \mathcal{C}_k
11:
12:
          end if
13:
          if \mathcal{L}^* = value(\mathcal{C}^*) then
             break
14:
15:
          \lambda_{k+1,j} \leftarrow \max\left(0, \lambda_{k,j} + \mu_k \left(1 - \sum_{\substack{i=1\\j \in S_i}}^m x_i\right)\right) \text{ for } 1 \le j \le n
16:
          \mu_{k+1} \leftarrow \frac{1}{k} (or another valid update rule)
          k \leftarrow k + 1
18:
19: end while
20: return \mathcal{L}^* and \mathcal{C}^*
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2 Local search

2.1 Pigment Sequencing Problem

In order to formulate the cost invariants, we use the variables $x_p \in I \cup \{idle\}$ which decide which item is produced at period $p \in [0, p_{max}]$ or if the machine is idle.

If we were to change the value of x_p from i to j and with $i, j \in I \cup \{idle\}$ and $d_i, d_j \geq p$ if $i, j \neq idle$ respectively, the cost of the solution would be increased by a Δ given by:

$$\Delta = \Delta_{stocking} + \Delta_{changeover} \tag{5}$$

where $\Delta_{stocking}$ and $\Delta_{changeover}$ respectively denote the variation of the total stocking and changeover costs.

We first define the individual stocking cost:

$$c_{stocking}(x_p = i) = \begin{cases} 0 & \text{if } i = idle, \\ (d_i - p)h & \text{otherwise.} \end{cases}$$
 (6)

which allows to formulate $\Delta_{stocking}$ easily:

$$\Delta_{stocking} = c_{stocking}(x_p = j) - c_{stocking}(x_p = i) \tag{7}$$

Similarly, we will express the individual changeover cost. To simplify the coming equations, let us define:

$$q(i,j) = \begin{cases} 0 & \text{if } i = idle \text{ or } j = idle, \\ q^{t_i,t_j} & \text{otherwise.} \end{cases}$$
 (8)

Let p', p'' respectively denote the previous and next periods where an item is produced (i.e. the machine is not idle). Formally, we write:

$$p' = \max(\{\pi \mid \pi \in [0, p-1] \text{ and } x_{\pi} \neq idle\})$$
 (9)

$$p' = \max (\{\pi \mid \pi \in [0, p - 1] \text{ and } x_{\pi} \neq idle\})$$

$$p'' = \min (\{\pi \mid \pi \in [p + 1, p_{max}] \text{ and } x_{\pi} \neq idle\})$$
(9)
(10)

The items produced at those periods, or the value *idle* if there is no such period, are given by $k = x_{p'}$ and $l = x_{p''}$.

$$c_{changeover}(x_p = i) = \begin{cases} q(k, l) & \text{if } i = idle, \\ q(k, i) + q(i, l) & \text{otherwise.} \end{cases}$$
 (11)

Finally, we can write:

$$\Delta_{changeover} = c_{changeover}(x_p = j) - c_{changeover}(x_p = i)$$
(12)