

1. (•) *I understand the course policies.*

Describe and analyze an algorithm to nest the boxes so that the number of visible boxes is as small as possible.

Solution: We assume one box can only contain *exactly* one another box, i.e., two boxes cannot be placed into the same box simultaneously. We reduce this problem to bipartite matching problem, and then use Ford-Fulkerson algorithm to solve it. We construct a bipartite graph G with vertex set $U \cup W$ as follows:

- Each box is a node in both U and W .
- For a pair of node $u \in U$ and $v \in W$, if u can be rotated so that it can be contained in v , we add an edge $e = (u, v)$.

Call two boxes are nested if one contains another. Let N be the maximum number of nested boxes. Then, the number of visible boxes can be simply computed as $n - N$.

Theorem 1. *The maximum number of nested boxes equals to $|M|$, where $|M|$ is the value of the size of the maximum matching in G .*

Proof: If we know the maximum number of nested boxes and how the boxes contain each other, we can transfer to a maximum matching in G as follows: if u contains v , then we select $e = (u, v)$ in G , where $u \in U$ and $v \in W$. Since each box can contain exactly one another box, this is a valid matching. Each time we place a box into another, the number of visible boxes is decreased by one. Let $|M|$ be the final number of edges selected. It is maximized according to the above procedure. Otherwise, if there is a matching that has larger size, there will be a better way to place the boxes. This contradicts with the assumption. By contradiction, M is a maximum matching. Moreover, the number of the nested boxes and the size of the maximum matching is $|M|$.

Conversely, if we know a maximum matching, we can find how to place the boxes into each other by placing u into v for each $e = (u, v)$ in M . Each $u \in U$ is matched exactly once, which means each box can be placed in another box only once. Thus this is a valid solution for the problem. Furthermore, this scheme must has the maximum number of nested boxes. Otherwise, if there is a better way to place the box, we can transform it to a matching that has a larger size. By contradiction, this is the best way to place the box.

Finally, the size of the maximum matching is $|M|$, and equals to the number of nested boxes. \square

There are at most six permutation of width, height and depth. Hence for a pair of boxes, we can determine whether they can contain one another in constant time. Let there be n boxes. Then, there are $2n$ nodes and at most $O(n^2)$ edges in G . Hence, the maximum flow has value at most $O(n)$, and the Ford-Fulkerson algorithm runs in $O(n^3)$ time.

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2. (•) **I understand the course policies.**

Describe an efficient algorithm that either rounds A in this fashion, or reports correctly that no such rounding is possible.

Solution: We can solve this problem by reducing it to maximum flows problem with edge demands. Before we do that, we first check if the sum of each row (or column) is an integer. If not, we report that no such rounding is possible. This is because after rounding, the sum of each row (or column) must be an integer.

Now we assume the sum of each row (or column) is an integer. Let $R[1..m]$ be the sum of each row, and $C[1..n]$ be the sum of each column. We construct a graph G with vertex set $\{s, t\} \cup U \cup V \cup W$ as follows:

- Each element $A[i][j]$ is a node x_{ij} in U .
- Each row i is a node r_i in V .
- Each column j is a node c_j in W .

The edges consists of:

- an edge $e : s \rightarrow x_{ij}$ with capacity $c(e) = \lceil x_{ij} \rceil$ and demand $d(e) = \lfloor x_{ij} \rfloor$ for each $x_{ij} \in U$.
- an edge $e : x_{ij} \rightarrow r_i$ with capacity $c(e) = \infty$ and demand $d(e) = 0$ for each $r_i \in V$.
- an edge $e : x_{ij} \rightarrow c_j$ with capacity $c(e) = \infty$ and demand $d(e) = 0$ for each $c_j \in W$.
- an edge $e : r_i \rightarrow t$ with capacity $c(e) = d(e) = R[i]$ for each $r_i \in V$.
- an edge $e : c_j \rightarrow t$ with capacity $c(e) = d(e) = C[j]$ for each $c_j \in W$.

Theorem 2. *The rounding exists if and only if there is a maximum flow that saturates every edge entering t . The rounding scheme consists of the flow values that leave from s .*

Proof: If we know a correct rounding scheme, we can transform it to a feasible flow in G . For each $x_{ij} = A[i][j]$, we push the same amount of flow to edge $s \rightarrow x_{ij}$. Since x_{ij} can only be rounded to $\lfloor x_{ij} \rfloor$ or $\lceil x_{ij} \rceil$, clearly the values satisfy the edge constraints of these edges. Moreover, since this is a valid rounding scheme, the sum of rows and columns will be exactly the same as the flow value on edges $r_i \rightarrow t$ or $c_j \rightarrow t$, which means these edges are saturated and thus it is a maximum flow.

Conversely, if we have a maximum flow that saturates every edge entering t , we can round each element $A[i][j] = f(e)$, where e is the edge $s \rightarrow x_{ij}$. The flow value $f(e)$ satisfies $d(e) \leq f(e) \leq c(e)$. Thus, each entry $x_{ij} = A[i][j]$ is rounded correctly to $\lfloor x_{ij} \rfloor$ or $\lceil x_{ij} \rceil$. Finally, the sum of rows and columns are the same as the corresponding capacity (or demand) values in edges $r_i \rightarrow t$ and $c_j \rightarrow t$. Since all these edges are saturated, the sum of entries in any row or column are unchanged. Thus, it is a valid rounding scheme. \square

Clearly, there are $O(mn)$ vertices and $O(mn)$ edges in G . Computing the sums and constructing the graph can be done in $O(mn)$ time. If we use Edmonds-Karp fat-pipe algorithm, we get an overall running time of $O(VE^2) = O(m^3n^3)$. \blacksquare

3. (•) *I understand the course policies.*

Describe and analyze an algorithm to compute a donation schedule, describing how much money each voter should send to each candidate on each day, that guarantees that every candidate gets enough money to win their election. The schedule must obey both Federal laws and individual voters' budget constraints. If no such schedule exists, your algorithm should report that fact.

Solution: Suppose there are k days left from now to the election, there are c candidates and v voters. Let $M[1..c]$ be the money that must be spent on each candidate, $W[1..v][1..c]$ be the amounts that each voter is willing to donate to each candidate $A[1..v]$ be the total amounts that each voter is able to donate to all candidates. We can model this problem as a maximum flow problem. We construct a flow network $G = (V, E)$ with vertices $X \cup Y \cup \{s, t\}$, where each node in X and Y denotes a voter and a candidate, respectively. The edges consist of:

- an edge $s \rightarrow x$ with capacity $\min\{100, A[x]/k\}$ for each $x \in X$.
- an edge $y \rightarrow t$ with capacity $M[y]/k$ for each $y \in Y$.
- an edge $x \rightarrow y$ with capacity $W[x][y]/k$ for each $x \in X$ and $y \in Y$.

Theorem 3. *The donation schedule exists if and only if there is a maximum flow that saturates every edge entering t . The money that each voter should spend to each candidate on every day is denoted by the edges $x \rightarrow y$ for each $x \in X$ and $y \in Y$.*

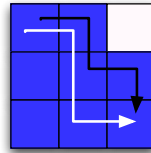
Proof: For each edge $s \rightarrow x$ for each $x \in X$, clearly it satisfies both Federal law and individual voters' budget constraints. For each edge $x \rightarrow y$, it denotes the amount that voter x will send to candidate y each day. For each edge $y \rightarrow t$ for each $y \in Y$, it denotes the amount that every candidate must receive each day to meet the campaign requirement. If the maximum flow we compute saturates all the edges entering t , then following this schedule, after k days each candidate can receive exactly $M[y]$ money for all $y \in Y$. Hence, the maximum flow we compute is the desired schedule. \square

The maximum flow has value at most $S = \sum_y M[y]$. There at most $v + v \cdot c + c = O(vc)$ edges. Thus, Ford-Fulkerson algorithm runs in $O(Svc)$ time. \blacksquare

4. (•) ***I understand the course policies.***

- (a) Prove that this greedy strategy does not always compute an optimal solution.

Solution: We prove by giving a counterexample. As in the following figure, the blue cells are marked. Both of the 2-bend and 3-bend paths contains maximum number of marked cells. However, if we choose the 2-bend path at the first iteration, we need another two monotone path to cover all the cells. On the other hand, choosing the 3-bend path at first can finish marking the cells with only two paths.



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- (b) Describe and analyze an efficient algorithm to compute the smallest set of monotone paths that covers every marked cell. The input to your algorithm is an array $M[1..n, 1..n]$ of booleans, where $M[i, j] = \text{True}$ if and only if cell (i, j) is marked.

Solution: Dynamic programming to construct the graph G . Maximum flow with unit demand and cost on each edge. $d = 1$ means each node must be used. $\$ = 1$ means the more paths are used, the more cost there will be, because all flows are sent from s .

Proof: Minimum set of monotone paths \rightarrow minimum cost flow.

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5. (•) ***I understand the course policies.***

- (a) Prove that if Paul uses a deterministic strategy, and Sally knows his strategy, then Sally can guarantee that she wins.
- (b) Describe a deterministic strategy for Sally that guarantees that she wins when $r \leq M$, no matter what strategy Paul uses.
- (c) Prove that if Sally uses a deterministic strategy, and Paul knows her strategy then Paul can guarantee that he wins when $r < M$.
- (d) Describe a randomized strategy for Sally that guarantees that she wins with probability at least $\min\{r/M, 1\}$, no matter what strategy Paul uses.
- (e) Describe a randomized strategy for Paul that guarantees that he loses with probability at most $\min\{r/M, 1\}$, no matter what strategy Sally uses.