positive value. So the value of F may increase by at most $h(v) = h(u) - 1 \le 2|V| - 1 - 1 = 2|V| - 2$. From Corollary 6.24 the number of saturated Push operations is at most 2|V|.|E|. So F can increase by at most 2|V|.|E|.(2|V|-2) due to all the saturated Push.

Finally let us consider Relabel operation. Total contribution to F due to all the Relabel operations on a vertex, u, can be at most equal to all the increments to h(u). Since h monotonically increases and its maximum value is 2|V|-1, total contribution to F due to one vertex is 2|V|-1. Hence total increase in F due to all the Relabel operations is at most (|V|-2)(2|V|-1).

Putting all together $0 + 2|V|.|E|.(2|V|-2) + (|V|-2)(2|V|-1) - K \ge 0$ or $K \le 2|V|^2 + 4|V|.|E|(|V|-1) = 4|V|(|E|.(|V|-1) + |V|/2) \le 4|V|.^2|E|.$

Push operation takes O(1) time, Relabel takes O(|V|) time, Insert/Extract takes O(1), and $Insert_Edges$ takes O(|V|) time. After the initialization, each Push may be followed by an $Insert_Edges$. So each of these cost O(|V|). Each Relabel is followed by an $Insert_Edges$ so each Relabel costs O(|V|). So the total cost is at $O(|V|.(4|V|^2.|E|+2|V|.|E|+2|V|^2)) = O(|V|^3.|E|)$.

Theorem 6.26. Algorithm 85 correctly computes a maximum flow in a flow network in $O(|V|^3.|E|)$ time.

6.4.4. The case of multiple Sources and Sinks. Going back to our example of a coal mine and a steel mill let us consider a variant of the problem where there are multiple mines that produce coal and multiple mills that consume it. Further assume that each mine and each mill has a maximum limit for its production and consumption. Figure 6.4.4 shows a sample network. Label b_i at vertex i is to be interpreted as follows. If $b_i > 0$, then i is a source (mine) and it can produce at the rate up to b_i ; if $b_i < 0$, then i is a sink (mill) and can consume at the rate at most $-b_i$; and if $b_i = 0$, then i is an intermediate node that neither produces nor consumes. Formally, the flow must satisfy the condition $0 \le \sum_j f(i,j) \le b_i$ for positive b_i ; $0 \le \sum_j f(j,i) \le -b_i$ for negative b_i ; and $\sum_j f(i,j) = 0$ if $b_i = 0$.

This variant of the problem can be formulated using one-source - one-sink model discussed above. Add two special nodes, s and t, and add one edge from s to each source i and assign capacity b_i to it and add one edge from each sink i to t and assign capacity $-b_i$. See an example in figure 6.4.4.

There is a one to one correspondence between solution of the original network and the new network. A solution of the new network gives a solution, of equal flow magnitude, of the original multi-source multi-sink network with the same flow by simply removing the added edges. Similarly given a solution of the original network add flow to the new edge (s, s_i) (edge (t_i, t)) equal to the flow emerging (sinking) from s_i (into t_i) we can construct a valid flow of the transformed network with same magnitude. Hence we see that single source and single sink model is general enough.

6.5. Minimum Weight Circulation

6.5.1. Weighted Flow. In this section we will discuss a more general flow problem and discuss a classic solution for it. We will also see that the maximum flow problem is a special case of this problem. In this chapter we will associate a cost/weight (or per-unit cost) for each edge in the network. If u units are flowing along an edge e, then total cost of the flow through this edge will be u.w(e) where



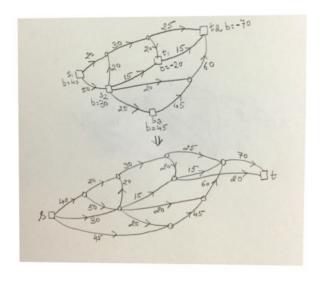


FIGURE 3. An example of a multi source/sink network and reduction to a single source/sink network

w() will be called the weight function. To handle this situation in a general manner, we define a flow network as a directed multi-graph (V, E, c, w) where c is the capacity function and w is the weight function. In this case there can be multiple edges between a pair of vertices. Each edge $e \in E$ will have its own capacity c(e) and weight w(e). The value of c(e) is non-negative but the value of c(e) can be any real number including negative. Since negative value is allowed we choose to call it a weight function rather than a cost function. We will denote the end-vertices of c by c and c where edge is directed from c to c. If two parallel edges, c i.e., c i.e., c and c and c where edge is directed from c to c if two parallel edges, c in the man replace them by a single edge c such that c in c in

In the modified definition of a flow network we need to change the definition of a flow. Previously f(u,v) was equal to -f(v,u) and we had stated the capacity condition as $-c(v,u) \le f(u,v) \le c(u,v)$. Earlier if a unit of flow was in the edge (u,v) and b unit of flow was in (v,u), then we defined f(u,v) = a - b. But in the new definition f assumes non-negative value. It is a function $f: E \to \mathbb{R}^+ \cup \{0\}$ because there may be multiple edges between the same pair of vertices. Given a

flow f in a flow-network, its cost is defined by $\sum_{e \in E} w(e) \cdot f(e)$. In this more general network we will discuss a solution of the following problem.

Minimum Cost Flow Problem: Compute a maximum flow in the network that has minimum weight.

Before we start to discuss the solution of this problem we make a modification in the structure of the flow networks. The new structure is called a *circulation-network*.

6.5.2. Circulation. A circulation-network is similar to a flow network but it does not have any source-node or a sink-node.

DEFINITION 6.27. A weighted circulation-network is a directed multi-graph G = (V, E, c, w), where $c: E \to \mathbb{R}^+$ is a non-negative valued function called capacity-function and $w: E \to \mathbb{R}$ is a real-valued function called weight-function.

Homogeneous structure of the circulation networks (no special vertices as s and t) make them simple to work with as we will see later.

DEFINITION 6.28. A circulation in a circulation-network is a function $f: E \to \mathbb{R}^+$ satisfying following properties:

- $(1) \ 0 \le f(e) \le c(e) \text{ for all } e \in E,$
- (2) $\sum_{e:e^l=x} f(e) = \sum_{e:e^r=x} f(e)$ for all $x \in V$.

The total cost/weight of a circulation is defined by $\sum_{e \in E} w(e) \cdot f(e)$. The minimum cost/weight circulation problem is defined as follows.

Minimum Cost Circulation Problem Given a circulation network (V, E, c, w), compute a circulation for it with minimum $\sum_{e \in E} w(e) f(e)$.

In the remainder of this chapter we will develop an algorithm to solve this problem. But first, let us justify that this problem subsumes the maximum flow problem and the minimum cost flow problem.

Claim 6.29. For a flow network G = (V, E, c) we can define a circulation network G' such that there is a 1 to 1 correspondence between the flows in G and circulations in G'.

PROOF. Modify the given flow network into a circulation network by adding an edge e_0 from t to s with ∞ capacity. Retain the edges of G and their capacities.

Let f be any flow in G. Then define a circulation f' in G' by defining f'(e) = f(e) for all $e \in E$ and define $f'(e_0) = |f|$. Observe that f' is a valid circulation.

Conversely for any circulation f' in G' define f for G as f(e) = f'(e) for all $e \in E$. Then observe that f is a valid flow in G and $|f| = f'(e_0)$.

Using this construct we will show how to reduce the maximum flow problem and minimum cost flow problem into minimum cost circulation problem. Here is a trivial claim.

CLAIM 6.30. Let G = (V, E, c) be an unweighted flow network. Let $G' = (V, E \cup \{e_0\}, c, w')$ be the corresponding circulation network as defined above. Let w'(e) = 0 for all $e \in E$ and $w'(e_0) = -1$. Then G has a flow of magnitude m if and only if the corresponding circulation has cost -m.

This claim shows that the flow in G corresponding to the minimum cost circulation f' is a maximum flow for G. So it shows that the maximum flow problem can be reduced the minimum cost circulation problem. Next consider the following

claim which shows how to reduce the minimum cost flow problem to minimum cost circulation problem.

CLAIM 6.31. Let G = (V, E, c, w) be a weighted flow network. Let $G' = (V, E \cup \{e_0\}, c', w')$ be the corresponding circulation network as defined above. Let w'(e) = w(e) for all $e \in E$ and $w'(e_0) = 0$. Then the cost any flow of G is same as the cost of the corresponding circulation of G'.

DEFINITION 6.32. Given a circulation f in a circulation network G = (V, E, c, w) and a circulation f in it. Then its flow-graph is the directed graph $G^f = (V, \{e \in E | f(e) > 0\})$. If G^f is a simple cycle then f is called a cyclic circulation.

For the remainder of this chapter we will express flow/circulation as a vector in $\mathbb{R}^{|E|}$. A flow f will be denoted by a vector in which e-th component will have value f(e). In addition any simple cycle C will be denoted by a 0-1 vector in which e-th component is 1 if e belongs to the cycle. Otherwise it is 0. In this notation a cyclic circulation can be denoted by $f = |f| \cdot C$ where G^f is the cycle C because all edges of C carry |f| units of flow.

Suppose C_1,C_2,\ldots,C_t are cycles in a circulation network and α_1,\ldots,α_t are positive real numbers. Then $f=\alpha_1.C_1+\alpha_2.C_2+\cdots+\alpha_t.C_t$ defines a vector in which $f(e)=\sum_{j|e\in C_j}\alpha_j$. Observe that f satisfies the Kirchhoff's law at every vertex. So f is a valid circulation if it also satisfies the capacity condition, i.e., $f(e)\leq c(e)$. Following result shows that every circulation can be decomposed into cyclic circulations.

LEMMA 6.33. Given any circulation network and any circulation f in it, there exist cycles C_i in the network graph such that $f = \alpha_1.C_1 + \cdots + \alpha_t.C_t$, where each $\alpha_i > 0$.

PROOF. The proof is by induction on the number of edges in G^f . The claim is trivial if G^f is an empty graph. Suppose the claim holds if G_f has k or less edges. Now suppose G^f has k+1 edges. In a circulation the magnitude of the inflow into a vertex is same as the magnitude of the outflow from that vertex. Hence in G^f , either a vertex is isolated (no incoming or no outgoing edges incident on them) or it has at least one incoming and at least one outgoing edge. Hence the graph must have at least one (directed) cycle. Let $C: x_0x_1 \ldots, x_{r-1}x_0$ be one of the cycles in G^f . So $f(x_0, x_1) > 0$, $f(x_1, x_2) > 0$, ..., $f(x_{r-1}, x_0) > 0$. Let $\alpha = \min_i f(x_i, x_{i+1})$, here index summation is modulo r. Let f' be defined as follows. f'(y, z) = f(y, z) for all (y, z) not in C and $f'(y, z) = f(y, z) - \alpha$ for (y, z) in C. So $f' = f - \alpha .C$. It is easy to verify that f' is also a circulation and $G_{f'}$ has at least one less edge than in G_f . Hence from the induction hypothesis $f' = \alpha_1.C_1 + \cdots + \alpha_{t-1}.C_{t-1}$. Then $f = \alpha_1.C_1 + \cdots + \alpha_{t-1}.C_{t-1} + \alpha .C$.

DEFINITION 6.34. The residual graph of a circulation in a circulation network is defined in the similar fashion as it is defined in the flow network. Let G = (V, E, c, w) be a circulation network and let f be a circulation in it. E^{rev} is set of new edges given by $\{e^{rev}|e \in E\}$. Observe that if e_1 is an edge from x to y and e_2 is an edge from y to x, then e_1^{rev} and e_2^{rev} are new edges which are from y to x and x to y respectively. The residual graph of a circulation f is the circulation network $G_f = (V, \{e \in E | f(e) < c(e)\} \cup \{e^{rev} \in E^{rev} | f(e) > 0\}, c_f, w_f)$ where (i) $c_f(e) = c(e) - f(e)$ for $e \in E$ because we can add c(e) - f(e) flow to e without violating capacity; (ii) $c_f(e^{rev}) = f(e)$ for all $e \in E$ because if α amount is flowing

along $(x,y) \in E$, then we can decrease at most α flow from this edge or increase at most α along the virtual edge (y,x); (iii) If $(x,y) \in E$, then by adding flow along (y,x) we are effectively decreasing "real flow" leading to decrease in the cost at the rate of w(x,y). So w(y,x) = -w(x,y) for all $(y,x) \in E^{rev}$. More generally $w(e^{rev}) = -w(e)$ for all $e \in E \cup E^{rev}$. Thus w_f is independent of f and it is same as w on original edges and -w on the virtual edges. Note that we have used the superscript rev for all edges of $E \cup E^{rev}$, where $(e^{rev})^{rev} = e$.

Following theorem gives a criterion for a minimum cost circulation in any circulation network. It is similar to the flow network case where we had a criterion for maximum flow (having no augmenting path) in the sense that it will lead to an algorithm for computing a minimum cost circulation.

Theorem 6.35. A circulation f, in a circulation network G = (V, E, c, w) with finite costs and finite capacities, is a minimum cost circulation if and only if G_f has no (directed) cycle with negative cost (cost of a cycle is the sum of the costs of its edges).

PROOF. All capacities and costs in G are finite so the cost of any circulation must be finite. Consider an arbitrary circulation f in it. Suppose the residual graph G_f has a cycle C with negative cost $-\beta < 0$. Let $\alpha = \min_{e \in C} c_f(e)$. Hence $f' = f + \alpha.C$ is a valid circulation with cost $Cost(f') = Cost(f) + \alpha.(-\beta) < Cost(f)$. Hence f is not a minimum cost circulation.

Conversely suppose f is not a minimum cost circulation and let f' be a minimum cost circulation in the network. Let us define a new circulation f'' in G_f as follows: For all $e \in E$, if $f'(e) \geq f(e)$, then f''(e) = f'(e) - f(e) and if f'(e) < f(e), then $f''(e^{rev}) = f(e) - f'(e)$. To see that f'' satisfies the capacity constraints in G_f we need to consider two cases: (i) If f(e) < f'(e), then $f'(e) - f(e) \leq c(e) - f(e) = c_f(e)$; (ii) If f'(e) < f(e), then $f(e) - f'(e) = c_f(e^{rev}) - f'(e) \leq c_f(e^{rev})$. Besides, it is easy to verify that as a vector f'' = f' - f. Since f and f' satisfy the Kirchhoff's law, f'' will also satisfy it. Hence f'' is also a circulation. We also have the relation w(f'') = w(f') - w(f) < 0 because f' is a minimum cost circulation.

From Lemma 6.33 f'' can be decomposed into cyclic circulations, $f'' = \alpha_1 C_1 + \cdots + \alpha_r C_r$, where each C_i is a cycle in G_f . So we have $w(f'') = \sum_i \alpha_i . w(C) < 0$. This implies that for at least one C_i , $w(C_i) < 0$.

Note that the edges of G_f constitute a subset of $E \cup E^{rev}$ hence the edges of cycles in G_f also belong to $E \cup E^{rev}$. It may be observed that if C is a clockwise cycle in G_f , then for any edge $e \in E$, if e occurs in C, then it must be pointing in the clockwise direction and if e^{rev} occurs in C, then e must be pointing in the anti-clockwise direction.

This theorem gives a method to compute a minimum circulation, like Ford Fulkerson's method computes a maximum flow. Algorithm 86 describes Klein's negative cycle cancellation method. It iteratively selects a cycle having negative weight in the residual graph and augments maximum possible flow along it so that at least one of the edges gets saturated and gets removed from the residual graph, i.e., the cycle breaks.

Theorem 6.35 assures that if this algorithm terminates, then f will be a minimum cost circulation. Klein's algorithm suffers from the same problem that Ford-Fulkerson algorithm did. How to guarantee that the algorithm will terminate? Surprisingly, as we saw in the Edmonds and Karp solution of termination problem

```
\begin{array}{l} \textbf{for } e \in E \ \textbf{do} \\ \mid f(e) := 0; \\ \textbf{end} \\ \textbf{Compute residual graph } G_f; \\ \textbf{while } G_f \ has \ a \ negative \ cost \ cycle \ \textbf{do} \\ \mid \textbf{Find a negative cost } \ cycle \ C \ \text{in } G_f; \\ \alpha := \min_{e \in C} c_f(e); \\ f := f + \alpha.C; \\ \textbf{Compute residual graph } G_f; \\ \textbf{end} \\ \textbf{Return } f; \end{array}
```

Algorithm 80: Klein's Cycle Cancellation Method

in Ford-Fulkerson algorithm, here too there exists a solution for termination problem based on a strategic way to select the negative weight cycles for cancellation. So we need a strategy for the selection of negative cost cycles in Klein's method to ensure termination.

6.5.3. Minimum Mean Cost Strategy. Goldberg and Tarjan showed that the Klein's method will terminate if in each iteration that cycle from G_f is selected which has the minimum mean cost (average cost). If a cycle has length L and its cost is β , then the mean-cost is β/L . If the circulation in not optimum, then from Theorem 6.35 there must exist at least one cycle with negative cost in G_f . The mean-cost of a cycle will be negative if and only if its total cost is negative. Therefore if a circulation is not optimal, then the minimum mean-cost will obviously be negative. In the remainder of this section we will derive an upper-bound on the number of iterations in the Klein's method under this strategy of cycle selection. Let f be a circulation in a circulation network G. We will denote the mean-cost of the minimum mean-cost cycle in G_f by μ_f . Recall that in Chapter 1 we have discussed how to compute a minimum mean-weight cycle.

Let |V| = n and |E| = m. Then Goldberg and Tarjan showed that in 2m iterations of Klein's method the magnitude of the minimum mean-cost will reduce by a factor of 1/n, under minimum mean-weight strategy. That is, if the circulation after r iterations is f and that after r + 2m iterations is f', then $-\mu_{f'} \leq (1 - 2/n)(-\mu_f)$.

The minimum cost circulation problem is trivial if the costs of all the edges are non-negative because in that case f=0 is the minimum cost circulation. Hence we will assume that the cost of at least one edge in G is negative. Assume that $W_0(<0)$ is the smallest edge cost in $E \cup E^{rev}$.

THEOREM 6.36. In a graph with integral costs the Klein's Algorithm computes a minimum cost circulation in $O(m^2 n \log^2 (-W_0.n))$ time if a minimum mean-cost cycle is cancelled (augmented) in every iteration, where W_0 is the smallest edge cost in $E \cup E^{rev}$.

PROOF. If f is a circulation such that $-\mu < 1/n$. Then the total cost of the minimum mean-cost cycle C in G_f will be $w(C) = |C| \cdot \mu > (-1/n)|C| \ge (-1/n) \cdot n = -1$ where |C| denotes the length of the cycle. The cost of the edges

are given to be integral so $w(C) \ge 0$. Therefore there cannot be any negative cost cycle in G_f . Thus the Algorithm 86 must terminate as soon as $-\mu$ decreases below 1/n.

Let W_0 be the smallest edge cost in $E \cup E^{rev}$. So initially $-\mu_i \leq -W_0$. After k.(2m) iterations if the circulation is f, then from Goldberg Tarjan result $-\mu_f \leq \mu_i (1-2/n)^k \leq (-W_0)(1-2/n)^k$. From the previous paragraph we see that the algorithm will terminate after k.(2m) iterations if $(-W_0)(1-2/n)^k < 1/n$ or $k.\log(1-2/n) < -\log(-W_0.n)$. Using $\log(1-x) = -x - x^2/2 - x^3/3 - \ldots$ we get $k > (\log(-W_0.n))/(2/n) = n.\log(-W_0.n)/2$. So we deduce that the algorithm will terminate after at most $(2m).n.\log(-W_0.n) + 1$ iterations.

From Theorem 1.40 we know that μ_f can be computed in $O(n.m.\log_2{(-W_0)})$. By adding μ_f to all the edges of G_f the cycle with minimum mean-wright cycle will reduce to a cycle of weight zero and all other cycles's weights will become positive. This zero weight cycle (minimum weight cycle) can be computed in $O(n^3)$ time, see Claim 1.37. The cost of computing G_f is O(|V| + |E|). Hence the total time cost of one iteration is $O(n+m+n^3+n.m.\log_2{(-W_0)}) = O(n^3+n.m.\log_2{(-W_0)})$. The total time complexity of the algorithm is $O(m^2.n^2(\log(-W_0.n).\log(-W_0)+n^4.m.\log(-W_0.n)))$ if the edge costs are integral.

6.5.4. Goldberg Tarjan Theorem. In the last part of the chapter we will establish the Goldberg Tarjan result. We will show that in the successive iterations $-\mu_f$ never increase and then we will establish the main claim. In order to do this we define a new weight function called reduced weight function which is more "homogeneous". Let $p:V\to\mathbb{R}^+$ be a non-negative function, called a potential function. Then we define a reduced weight function w_p given by $w_p(e)=w(e)+p(x)-p(y)$ for every edge e=(x,y) in $E\cup E^{rev}$. For any edge $e\in E\cup E^{rev}$, e^{rev} will denote the edge (y,x) irrespective of whether e belongs to E or E^{rev} . For any $e=(x,y)\in E\cup E^{rev}$, $w_p(e^{rev})=w(e^{rev})+p(y)-p(x)=-w(e)+p(y)-p(x)=-w_p(e)$.

Given an $\epsilon > 0$, the circulation f is said to be ϵ -optimal with respect to a given potential function p if $w_p(e) \ge -\epsilon$ for all e in G_f . The smallest ϵ , for which there exists a potential function such that f is ϵ -optimal, is denoted by ϵ_f . We will first show that $\epsilon_f = -\mu_f$.

LEMMA 6.37. Let f be any circulation, then $\epsilon_f = -\mu_f$.

PROOF. Let us first show that $\epsilon_f \leq -\mu_f$. Consider the modified weight function $w' = w - \mu_f$. As μ_f is the smallest mean-weight of any cycle in G_f with respect to the weight function w, G_f cannot have any negative weight cycle with respect to w'. Add a new vertex s and edges (s,x) with weight w'(s,x) = 0 for all $x \in V$. Define the potential function $p(x) = dist_{w'}(s,x)$ for all $x \in V$, where $dist_{w'}(s,x)$ is the weight of the minimum weight path from s to x with respect to the weight function w'.

Consider an arbitrary edge e=(x,y) in G_f . Consider a path from s to y which is the shortest path to x followed by the edge e. So $p(y) \leq p(x) + w'(e) = p(x) + w(e) - \mu_f$. So $\mu_f \leq w(e) + p(x) - p(y) = w_p(e)$. So G_f is $(-\mu_f)$ -optimal. Hence by the definition $\epsilon_f \leq -\mu_f$.

To prove the converse consider the cycle C which has the minimum mean weight with respect to w. Suppose it has L edges. Then $L.\mu_f = \sum_{e \in C} w(e)$. Let p be the potential function with respect to which f is ϵ_f -optimal. For any

cycle C', $\sum_{e \in C'} w(e) = \sum_{e \in C'} w_p(e)$ because potential terms get cancelled. In particular, $\sum_{e \in C} w(e) = \sum_{e \in C} w_p(e) \ge \sum_{e \in C} (-\epsilon_f) = L.(-\epsilon_f)$. Therefore we have $\mu \ge -\epsilon_f$.

Next we will show that $-\mu_f$ is a monotonically decreasing function.

Lemma 6.38. $-\mu_{f_i} \ge -\mu_{f_{i+1}}$ for all i, where f_i denotes the circulation after i iterations.

PROOF. Consider an arbitrary iteration i. Suppose C is the minimum meanweight cycle in G_{f_i} which was cancelled in the i+1-st iteration. Let its length be L. Let p be the potential function with respect to which f_i is ϵ_{f_i} -optimal. So $\sum_{e \in C} w_p(e) = \sum_{e \in C} w(e) = L.\mu_{f_i}$. By the definition of ϵ_{f_i} -optimality $w_p(e) \ge -\epsilon_{f_i} = \mu_{f_i}$ for all e. So $w_p(e) = \mu_{f_i}$ for all $e \in C$. In this sense w_p is a homogeneous weight function.

We will now show that $w_p(e) \ge \mu_{f_i}$ for every e in $G_{f_{i+1}}$. This will imply that $\mu_{f_{i+1}} \ge \mu_{f_i}$ because there might a better choice of function p giving a tighter lower bound for $w_p(e)$.

Observe that the edges of $E_{f_{i+1}} \setminus E_{f_i}$ are such edges which were not in G_{f_i} (i.e., were saturated) and their reverse appeared in C. Let p be the optimal potential function for G_{f_i} . If $e \in E_{f_{i+1}} \cap E_{f_i}$, then $w_p(e) \ge -\epsilon_f = \mu_{f_i}$.

function for G_{f_i} . If $e \in E_{f_{i+1}} \cap E_{f_i}$, then $w_p(e) \ge -\epsilon_f = \mu_{f_i}$. If $e = (x, y) \in E_{f_{i+1}} \setminus E_{f_i}$ then $e^{rev} \in C$ and $w_p(e^{rev}) = \mu_{f_i}$. So $w_p(e) = w(e) + p(x) - p(y) = -w(e^{rev}) - p(y) + p(x) = -w_p(e^{rev}) = -\mu_{f_i}$. So $w_p(e) = -\mu_{f_i} > \mu_{f_i}$ because μ_{f_i} is negative. Therefore $w_p(e') \ge \mu_{f_i}$ for all $e' \in E_{f_{i+1}}$. Thus $\mu_{f_{i+1}} \ge \mu_{f_i}$.

Now we have necessary background to prove the main result.

LEMMA 6.39.
$$-\mu_{f_{q+2m}} \leq (1-2/n)(-\mu_{f_q})$$
 for all q .

PROOF. Our argument to prove this result is very similar to that in the proof of Lemma 6.13 and Theorem 6.15. Consider any 2m+1 consecutive iterations. At least one edge in every augmenting cycle (the cycle to be cancelled) must be critical. Recall that an edge is critical if it gets saturated due to the current augmenting cycle and vanishes from the next residual graph. So in this span at least one edge must be critical twice. Consider arbitrary iteration q. Assume that an edge (x, y) is critical in G_{f_i} and again in G_{f_j} for some $q \leq i < j \leq q + 2m$. Since (x, y) was missing in $G_{f_{i+1}}$ it must have reappeared some time between iterations i+2 and j. For this to happen, edge (y, x) must occur in the augmenting cycle (not necessarily as critical) in G_{f_k} for some $i+1 \leq k \leq j-1$. So we have i < k < j such that an edge (x, y) occurs in the augmenting cycle C_i of G_{f_i} and (y, x) occurs in the augmenting cycle C_k of G_{f_k} .

Let $i \leq r \leq s \leq k$ such that [r,s] is the shortest interval having the property that there is an edge present in augmenting cycle C_r and its reverse is present in the augmenting cycle C_s . As shown above there exists at least one such interval [i,k] so [r,s] exists. Partition the edges of C_s into two sets A and B where A contains those edges which are either not present in G_{f_r} or its reverse is present in the augmenting cycle C_r in G_{f_r} . Thus by the choice of the pair (r,s), A is non-empty. B is the set of the remaining C_s edges. So every B-edge is present in G_{f_r} . Let e be any arbitrary edge in A such that e^{rev} does not occur in C_r . So e does not occur in G_{f_r} . For e to be present in G_{f_s} , e^{rev} must occur in C_q for some r < t < s. Then we have a pair (t,s) such that e^{rev} occurs in C_t and e occurs in C_s and this is a

shorter interval than that of (r, s). This contradicts the choice of the pair (r, s). So we can deduce that reverse of every A-edge occurs in C_r . Also, by the definition of (r, s), A is non-empty.

Let p be the potential function such that $w_p(e) \geq -\epsilon_{f_r}$ for all $e \in G_{f_r}$. So $\sum_{e \in C_s} w(e) = \sum_{e \in C_s} w_p(e) = \sum_{e \in A} w_p(e) + \sum_{e \in B} w_p(e) = -\sum_{e \in A} w_p(e^{rev}) + \sum_{e \in B} w_p(e)$. Every $e \in B$ is also present in G_{f_r} so $w_p(e) \geq -\epsilon_{f_r}$. Every $e \in A$ is such that e^{rev} belongs to C_r . Earlier we have seen that reduced weight of every edge of the augmenting cycle is $-\epsilon_f$. So $w_p(e^{rev}) = -\epsilon_{f_r}$ for all $e \in A$. Therefore $\sum_{e \in C_s} w(e) \geq \epsilon_{f_r} \cdot |A| + (-\epsilon_{f_r}) \cdot |B|$. Let the length of C_s be L. So |A| + |B| = L and we have seen that $|A| \geq 1$. So $\sum_{e \in C_s} w(e) \geq (|L| - 1)(-\epsilon_{f_r}) + \epsilon_{f_r} = -(|L| - 2).\epsilon_{f_r}$.

we have seen that $|A| \geq 1$. So $\sum_{e \in C_s} w(e) \geq (|L| - 1)(-\epsilon_{f_r}) + \epsilon_{f_r} = -(|L| - 2).\epsilon_{f_r}$. We also have $\sum_{e \in C_s} w(e) = |L|.(\mu_{f_s}) = |L|.(-\epsilon_{f_s})$. Hence $|L|.(-\epsilon_{f_s}) \geq -(|L| - 2).\epsilon_{f_r}$ or $\epsilon_{f_s} \leq (1 - 2/|L|)\epsilon_{f_r} \leq (1 - 2/n)\epsilon_{f_r}$. From Lemma 6.38 ϵ_f is a non-increasing function. So $\epsilon_{f_{q+2m}} \leq \epsilon_{f_s} \leq (1 - 2/n)\epsilon_{f_r} \leq (1 - 2/n)\epsilon_{f_q}$.