

Let $A \in \mathcal{V}$ source, $B \in \mathcal{V}$ sink. For the following, we suppose that $Turn_{j-1}$ has just finished and $A = Player(j)$ is currently deciding $Turn_j$. We use the following notation:

$$\begin{aligned} C & \text{ capacity configuration with } c_{Av} = DTr_{A \rightarrow v, j-1} \\ C' & \text{ capacity configuration with } c'_{Av} = DTr_{A \rightarrow v, j} \end{aligned}$$

We will use the following notation for clarity:

$$\begin{aligned} & \begin{cases} X = MaxFlow_{\mathcal{G}_{j-1}}(A, B) \\ X' = MaxFlow_{\mathcal{G}_j}(A, B) \end{cases} , \text{ for some } MaxFlow \text{ execution} \\ & \begin{cases} F = maxFlow_{\mathcal{G}_{j-1}}(A, B) \\ F' = maxFlow_{\mathcal{G}_j}(A, B) \end{cases} \end{aligned}$$

Any subscripts or superscripts applied to X and F refer to the capacity configuration with the exact same subscripts and superscripts.

Furthermore, we suppose an arbitrary ordering of the members of $N^+(A)$. We set $n = |N^+(A)|$. Thus

$$N^+(A) = \{v_1, \dots, v_n\}$$

We use these subscripts to refer to the respective capacities (a.k.a. direct trusts) and flows. Thus

$$\begin{aligned} x_i &= x_{Av_i} , \text{ where } i \in [n] \\ c_i &= c_{Av_i} , \end{aligned}$$

Definition 1 (Trust Reduction).

Trust Reduction on neighbour i is defined as $\delta_i = c_i - c'_i$.

Flow Reduction on neighbour i is defined as $\Delta_i = x_i - c'_i$.

We will also use the standard notation for 1-norm and ∞ -norm:

$$\begin{aligned} ||\delta_i||_1 &= \sum_{i=1}^n \delta_i \\ ||\delta_i||_\infty &= \max_{1 \leq i \leq n} \delta_i \end{aligned}$$

Definition 2 (Restricted Flow).

Let $i \in [n]$. Let $F_{A_i \rightarrow B}$ be x'_i when:

$$\begin{aligned} c'_i &= c_i \text{ and} \\ \forall k \in [n] \setminus \{i\}, c'_k &= 0 . \end{aligned}$$

*This definition can be rephrased equivalently as follows:
Let $v \in N^+(A)$. Let $F_{A_v \rightarrow B}$ be x'_{Av} when:*

$$\begin{aligned} c'_{Av} &= c_{Av} \text{ and} \\ \forall w \in N^+(A) \setminus \{v\}, c'_{Aw} &= 0 . \end{aligned}$$

Let $L \subset [n]$. Let $F_{A_L \rightarrow B}$ be $\sum_{i \in L} x'_i$ when:

$$\begin{aligned} \forall i \in L, c'_i &= c_i \text{ and} \\ \forall i \in [n] \setminus L, c'_i &= 0 . \end{aligned}$$

*The latter definition can be rephrased equivalently as follows:
Let $S \subset N^+(A)$. Let $F_{A_S \rightarrow B}$ be $\sum_{v \in S} x'_{Av}$ when:*

$$\begin{aligned} \forall v \in S, c'_{Av} &= c_{Av} \text{ and} \\ \forall v \in N^+(A) \setminus S, c'_{Av} &= 0 . \end{aligned}$$

The choice of the definition will depend on whether K in $F_{A_K \rightarrow B}$ is a node, an index or a set of nodes or indices.

Theorem 1 (Saturation Theorem).

$$(\forall i \in [n], c'_i \leq x_i) \Rightarrow (\forall i \in [n], x'_i = c'_i)$$

Proof. From the flow definition we know that

$$\forall i \in [n], x'_i \leq c'_i . \tag{1}$$

In turn $j - 1$, there exists some valid flow Y such that

$$\forall i \in [n], y_i = c'_i$$

with a flow value $\sum_{i=1}^n y_i$, which can be created as follows: We start from X and for each (A, v_i) edge we reduce the flow along paths starting from this edge for a total reduction of $x_i - c'_i$ on all those paths. Y is also obviously valid for turn j and, since all capacities c'_i are saturated, there can be no more outgoing flow from the source, thus Y is a maximum flow in \mathcal{G}_j . \square

Theorem 2 (Trust Transfer Theorem).

Let $V \in [0, F]$. We create a C' where

$$\begin{aligned} \forall i \in [n], c'_i &\leq x_i \text{ and} \\ \sum_{i=1}^n c'_i &= F - V . \end{aligned}$$

It then holds that $F' = F - V$.

Proof. From the Saturation theorem (1) we can see that $x'_i = c'_i$. It holds that

$$F' = \sum_{i=1}^n x'_i = \sum_{i=1}^n c'_i = F - V .$$

□

Lemma 1 (Flow Limit Lemma).

$$\forall i \in [n], x_i \leq F_{A_i \rightarrow B}$$

Proof. Suppose a flow where $\exists i \in [n] : x_i > F_{A_i \rightarrow B}$. If for any $k \neq i$ we choose $c'_k < c_k$, then $x'_i \geq x_i$. We set the new capacities as follows:

$$\begin{aligned} \forall k \neq i, c'_k &= 0 \text{ and} \\ c'_i &= c_i . \end{aligned}$$

Then for X' we will have

$$\begin{aligned} \forall k \neq i, x'_k &= 0 \text{ and} \\ x'_i &= x_i , \end{aligned}$$

which is also a valid flow for \mathcal{G}_{j-1} and thus by definition

$$F_{A_i \rightarrow B} = x'_i = x_i > F_{A_i \rightarrow B} ,$$

which is a contradiction. Thus the proposition holds. □

Theorem 3 (Trust Saving Theorem).

Suppose some $i \in [n]$ and two alternative capacities configurations, say C'_1 and C'_2 such that

$$\begin{aligned} c'_{1,i} &= F_{A_i \rightarrow B} , \\ c'_{2,i} &= c_i , \\ \forall k \in [n] \setminus \{i\}, c'_{1,k} &= c'_{2,k} . \end{aligned}$$

Then $F'_1 = F'_2$.

Proof. From the Flow Limit lemma (1) we know that $x_i \leq F_{A_i \rightarrow B}$, thus we can see that any increase in c'_i beyond $F_{A_i \rightarrow B}$ will not influence x_i and subsequently will not incur any change on the rest of the flows. \square

Theorem 4 (Invariable Trust Reduction with Naive Algorithms).

If $\forall i \in [n], c'_i \leq x_i$, then $\|\delta_i\|_1$ and $\|\Delta_i\|_1$ are independent of x'_i, c'_i .

Proof. Since $\forall i \in [n], c'_i \leq x_i$, by applying the Saturation theorem (1) we see that $x'_i = c'_i$, thus $\delta_i = c_i - x'_i$ and $\Delta_i = x_i - x'_i$. We know that $\sum_{i=1}^n x'_i = F - V$, so we have

$$\begin{aligned} \|\delta_i\|_1 &= \sum_{i=1}^n \delta_i = \sum_{i=1}^n (c_i - x'_i) = \sum_{i=1}^n c_i - F + V \text{ and} \\ \|\Delta_i\|_1 &= \sum_{i=1}^n \Delta_i = \sum_{i=1}^n (x_i - x'_i) = \sum_{i=1}^n x_i - F + V . \end{aligned}$$

thus $\|\delta_i\|_1, \|\Delta_i\|_1$ are independent from x'_i and c'_i . \square

Until now *MaxFlow* has been viewed purely as an algorithm. This algorithm is not guaranteed to always return the same flow when executed multiple times on the same graph. However, the corresponding flow value, *maxFlow*, is always the same. Thus *maxFlow* can be also viewed as a function from a matrix of capacities to a non-negative real number. Under this perspective, we prove the following theorem. Let \mathcal{C} be the family of all capacity matrices $C = [c_{vw}]_{V(\mathcal{G}) \times V(\mathcal{G})}$.

Theorem 5 (maxFlow Continuity).

Suppose a directed weighted graph with a source A , a drain B and a capacity configuration $C \in \mathcal{C}$. Let also $p \in \mathbb{N} \cup \{\infty\}$. The function $\text{maxFlow} : \mathcal{C} \rightarrow \mathbb{R}^+$ is continuous with respect to the $\|\cdot\|_p$ norm.

Proof. Let $C_0 \in \mathcal{C}$. We want to prove that

$$\forall \epsilon > 0, \exists \delta > 0 : 0 < \|C - C_0\|_p < \delta \Rightarrow |F - F_0| < \epsilon .$$

We will prove it by contradiction. Suppose that

$$\exists \epsilon > 0 : \forall \delta > 0, 0 < \|C - C_0\|_p < \delta \Rightarrow |F - F_0| \geq \epsilon .$$

Let $v_1, u_1 \in V(\mathcal{G})$. Let C such that

$$\begin{aligned} c_{v_1 u_1} &= c_{0, v_1 u_1} + \frac{\epsilon}{2} \\ \forall (v, u) \in E(\mathcal{G}) \setminus \{(v_1, u_1)\}, c_{vu} &= c_{0, vu} . \end{aligned}$$

Due to the construction, for $\delta = \epsilon$ we have

$$0 < \|C - C_0\|_p < \delta . \quad (2)$$

Any valid flow for C_0 is also valid for C , thus

$$F_0 \leq F . \quad (3)$$

Also, it is obvious by the way that C was constructed that

$$F \leq F_0 + \frac{\epsilon}{2} \quad (4)$$

From (3) we have $F_0 \leq F + \frac{\epsilon}{2}$, which, in combination with (4), gives

$$|F - F_0| \leq \frac{\epsilon}{2} < \epsilon ,$$

which, together with (2) contradicts our supposition. Thus *maxFlow* is continuous on C_0 . Since C_0 is arbitrary, the result holds for all $C_0 \in \mathcal{C}$, thus *maxFlow* is continuous with respect to $\|\cdot\|_p$ for any $p \in \mathbb{N} \cup \{\infty\}$. \square

Here we show three naive algorithms for calculating new direct trusts so as to maintain invariable risk when paying a trusted party. Let $F = \sum_{i=1}^n x_i$. To prove the correctness of the algorithms, it suffices to prove that

$$\forall i \in [n], c'_i \leq x_i \text{ and} \quad (5)$$

$$\sum_{i=1}^n c'_i = F - V . \quad (6)$$

First Come First Served Trust Transfer

Input : old flows x_i , value V

Output : new capacities c'_i

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1 fcfs( $(x_i)$ ,  $V$ ) :
2    $n = \text{length}(x_i)$ 
3    $F_{cur} = F = \sum_{i=1}^n x_i$ 
4   if ( $F < V$ )
5     return( $\perp$ )
6   for ( $i = 1$  to  $n$ )
7      $c'_i = x_i$ 
8    $i = 1$ 
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9   while (Fcur > F - V)
10    reduce = min(xi, Fcur - (F - V))
11    Fcur = Fcur - reduce
12    c'i = xi - reduce
13    i += 1
14  return( $\bigcup_{i=1}^n \{c'_i\}$ )

```

Proof of correctness for fcfs.

We will first show that at the end of the execution, $i \leq n+1$. Suppose that $i > n+1$ on line 14. This means that $F_{cur,n}$ exists and $F_{cur,n} = F - \sum_{i=1}^n x_i = 0 \leq F - V$ since, according to the condition on line 4, $F - V \geq 0$. This means however that the **while** loop on line 9 will break, thus $F_{cur,n+1}$ cannot exist and $i = n+1$ on line 14, which is a contradiction, thus the first proposition holds. We can also note that, even if $i = n+1$ at the end of the execution, the **while** loop will break right after the last incrementation, thus the algorithm will never try to read or write the nonexistent objects x_{n+1}, c'_{n+1} .

We will now show that $\forall i \in [n], c'_i \leq x_i$, as per the requirement (5). Let $i \in [n]$. In line 7 we can see that $c'_i = x_i$ and the only other occurrence of c'_i is in line 12 where it is never increased ($reduce \geq 0$), thus we see that the requirement (5) is satisfied.

We will finally show that $\sum_{i=1}^n c'_i = F - V$. From line 3 we see that $F_{cur,0} = F$. Let $i \in [n]$ such that $F_{cur,i}$ exists. If $F_{cur,i} \leq F - V$, then $F_{cur,i+1}$ does not exist because the **while** loop (line 9) breaks after calculating $F_{cur,i}$. Else

$$F_{cur,i+1} = F_{cur,i} - \min(x_{i+1}, F_{cur,i} - (F - V)) \quad . \text{ (lines 10- 11)}$$

If $\nexists i \in [n] : \min(x_i, F_{cur,i-1} - (F - V)) = F_{cur,i-1} - (F - V)$, then $\forall i \in [n], \min(x_i, F_{cur,i} - (F - V)) = x_i$, thus from line 12 it will be $\forall i \in [n], c'_i = 0$ and from line 11, $F_{cur,n} = 0$. However, we have

$$\left. \begin{aligned} \min(x_n, F_{cur,n-1} - (F - V)) &\neq F_{cur,n-1} - (F - V) \\ F_{cur,n-1} &= x_n \end{aligned} \right\} \Rightarrow$$

$$\Rightarrow x_n < x_n - (F - V) \Rightarrow F < V$$

which is a contradiction, since if this were the case the algorithm would have failed on lines 4 - 5. Thus

$$\exists i \in [n] : \min(x_{i+1}, F_{cur,i} - (F - V)) = F_{cur,i} - (F - V)$$

That is the only $i \in [n]$ such that $F_{cur,i+1} = F - V$, so

$$\forall 0 < k < i, F_{cur,k} = F_{cur,k-1} - x_k \Rightarrow F_{cur,i} = F - \sum_{k=1}^{i-1} x_k .$$

Furthermore, since $F_{cur,i+1} = F - V$, it is

$$\begin{aligned} c'_{i+1} &= x_{i+1} - F_{cur,i} + F - V = x_i - F + \sum_{k=1}^{i-1} x_k + F - V \Rightarrow \\ &\Rightarrow c'_{i+1} = \sum_{k=1}^i x_k - V , \\ &\forall k \leq i, c'_k = 0 \text{ and} \\ &\forall k > i + 1, c'_k = x_k . \end{aligned}$$

In total, we have

$$\sum_{k=1}^n c'_k = \sum_{k=1}^i x_k - V + \sum_{k=i+1}^n x_k = \sum_{k=1}^n x_k - V \Rightarrow \sum_{k=1}^n c'_k = F - V .$$

Thus the requirement (6) is satisfied. \square

Complexity of fefs. Since i is incremented by 1 on every iteration of the **while** loop (line 13) and $i < n + 1$ at the end of the execution, the complexity of the **while** loop is $O(n)$ in the worst case. The complexity of lines 4 - 5 and 8 is $O(1)$ and the complexity of lines 2 - 3, 6 - 7 and 14 is $O(n)$, thus the total complexity of algorithm is $O(n)$. \square

Note that we choose to calculate the complexity of the **length()** function as $O(n)$. Whereas this complexity is implementation-dependent, even with the most naive approaches it cannot be higher than $O(n)$, thus we use this worst-case complexity in our analysis to cover all cases. This approach will be implicitly used in all subsequent complexity analyses.

Absolute Equality Trust Transfer ($\|\Delta_i\|_\infty$ minimizer)

Input : old flows x_i , value V

Output : new capacities c'_i

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1  abs(( $x_i$ ),  $V$ ) :
2     $n = \text{length}(x_i)$ 
3     $F_{cur} = F = \sum_{i=1}^n x_i$ 
4    if ( $F < V$ )
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5     return( $\perp$ )
6     X = preprocess( $x_i$ )
7     empty = 0
8     reduction = 0
9     while ( $F_{cur} > F - V$ )
10        ( $i, X$ ) = popMin(X)
11         $F_{prov} = F_{cur} - (n - \text{empty}) * (x_i - \text{reduction})$ 
12        if ( $F_{prov} > F - V$ )
13            reduction =  $x_i$ 
14            empty += 1
15             $F_{cur} = F_{prov}$ 
16        else
17            aux = reduction
18            reduction +=  $\frac{F_{cur} - (F - V)}{n - \text{empty}}$ 
19             $F_{cur} -= (n - \text{empty}) * (\text{reduction} - \text{aux})$ 
20            #lines 17 & 19 can be replaced by break. In this
21            #case, the loop (line 9) can become while (TRUE).
22    for ( $i = 1$  to  $n$ )
23         $c'_i = \max(0, x_i - \text{reduction})$ 
24    return( $\bigcup_{i=1}^n \{c'_i\}$ )

```

The function `preprocess(x_i)` returns a data structure `X` containing the set of flows (x_i), such that the corresponding function `popMin(X)` is able to repeatedly return a tuple consisting of the index of the minimum element and a new data structure missing exactly the minimum element. Examples of such pairs of functions are:

$$\begin{aligned}
 & \begin{cases} \text{preprocess} = \text{quickSort} \\ \text{popMin} = (x_1, X \setminus x_1) \end{cases} \quad \text{and} \\
 & \begin{cases} \text{preprocess} = \text{FibonacciHeap} \\ \text{popMin} = (\text{find-min}(X), \text{delete-min}(X)) \end{cases} .
 \end{aligned}$$

Proof of correctness for abs. First of all, we can note that, if $F_{prov} \leq F - V$ in line 12, then the execution will enter the `else` clause of line 16. Therefore, in line 19, F_{cur} will get the value $F - V$, as we can see by executing the lines 17 - 19 by hand. This in turn means that the loop in line 9 will break right after the `else` clause is executed. Furthermore, the assignment in line 15 in combination with the truth of the statement $F_{prov} > F - V$ in line 12 shows that, if the execution enters the `if` clause

of line 12, then the loop of line 9 will be executed at least once more. These two observations amount to the fact that the **else** clause will be executed exactly one time and afterwards the **while** loop will break.

We use the notation $F_{cur,0}$, $reduction_0$ and $empty_0$ to refer to the initial values of the corresponding variables, as set in lines 3, 8 and 7 respectively. Furthermore, the notation $empty_j$, $reduction_j$, $F_{cur,j}$ and i_j is used to refer to the values of the corresponding variables after the j -th iteration of the **while** loop. i_j is chosen in line 10. From lines 11, 13, 15, 12 and 17 to 19 we see that

$$F_{cur,j} = \begin{cases} F_{prov,j}, & F_{prov,j} > F - V \\ F - V, & F_{prov,j} \leq F - V \end{cases}, \text{ where}$$

$$F_{prov,j} = F_{cur,j-1} - (n - empty_{j-1}) (x_{i_j} - x_{i_{j-1}}), j \geq 1 \text{ and } x_{i_0} = 0.$$

It is worth noting that the maximum number of iterations is n , or else $j \leq n$. This holds because, if we suppose that $F_{cur,n+1}$ exists, it is

$$F_{cur,n} > F - V \geq 0 \quad (7)$$

However, we can easily see that in this case

$$\begin{aligned} F_{cur,n} &= F_{cur,0} - \sum_{j=1}^n (n - (j - 1)) (x_{i_j} - x_{i_{j-1}}) = \\ &= \sum_{j=1}^n x_{i_j} - \sum_{j=1}^n (n - (j - 1)) x_{i_j} + \sum_{j=1}^{n-1} (n - j) x_{i_j} = \\ &= \sum_{j=1}^n x_{i_j} - \sum_{j=1}^{n-1} [(n - (j - 1)) - (n - j)] x_{i_j} - (n - (n - 1)) x_{i_n} = \\ &= \sum_{j=1}^n x_{i_j} - \sum_{j=1}^{n-1} x_{i_j} - x_{i_n} = 0, \end{aligned}$$

which is a contradiction to (7), thus $F_{cur,n+1}$ does not exist and $j \leq n$. This means that **popMin()** will never fail.

We will now show that $\forall j \in [n], empty_j < n$. At line 7, it is $empty_0 = 0 < n$. $empty$ is again modified in line 14, where it is incremented by at most 1 at each iteration of the **while** loop (line 9). As we saw above, the iterations cannot exceed n and $empty$ is not incremented in the last iteration which consists of the **else** clause, thus $\forall j \in [n], empty_j < n$.

Next, we will show that $\forall i \in [n], c'_i \leq x_i$, as per the requirement (5). From line 23, we see that it suffices to prove that $reduction \geq 0$.

In line 8, *reduction* is initialized to 0. In line 13, *reduction* is set to x_i , which is always a non-negative value. The last line where *reduction* is modified is 18. In this line, it is $F_{cur} > F - V$ or else the **while** loop would have broken before beginning this iteration and $n > empty$ as we previously saw. Thus the non-negative variable *reduction* is increased and the resulting value is always positive.

We will finally show that $\sum_{i=1}^n c'_i = F - V$, which satisfies the requirement (6). Let $k, 0 \leq k \leq n$ be such that at the end of the execution

$$\forall j \leq k, c_{i_j} = 0 \wedge \forall j > k, c_{i_j} > 0 .$$

The following holds:

$$\begin{aligned} \sum_{j=1}^n c'_{i_j} &= \sum_{j=k+1}^n c'_{i_j} = \sum_{j=k+1}^n \left(x_{i_j} - reduction_{k+1} \right) = \\ &\sum_{j=k+1}^n \left(x_{i_j} - \left(x_{i_k} + \frac{F_{cur,k} - (F - V)}{n - k} \right) \right) \\ &\sum_{j=k+1}^n \left(x_{i_j} - \left(x_{i_k} + \frac{F_{cur,0} - \sum_{l=1}^k (n - (l - 1)) (x_{i_l} - x_{i_{l-1}}) - F + V}{n - k} \right) \right) \\ &\sum_{j=k+1}^n \left(x_{i_j} - \left(x_{i_k} + \frac{F - \sum_{l=1}^k (n - l + 1) x_{i_l} + \sum_{l=1}^{k-1} (n - l) x_{i_l} - F + V}{n - k} \right) \right) \\ &\sum_{j=k+1}^n x_{i_j} - (n - k) x_{i_k} - \frac{n - k}{n - k} \left(- \sum_{l=1}^{k-1} x_{i_l} - (n - k + 1) x_{i_k} + V \right) \\ &\sum_{j=k+1}^n x_{i_j} - (n - k) x_{i_k} + \sum_{j=1}^{k-1} x_{i_j} + (n - k + 1) x_{i_k} - V \\ &= \sum_{j=1}^n x_{i_j} - V = F - V , \end{aligned}$$

thus the desired property holds. \square

Complexity of abs. Lines 4 - 5, 7 - 8 and 11 - 19 have a complexity of $O(1)$. Lines 2 - 3 and 22 - 24 have a complexity of $O(n)$. The **while** loop

of line 9 is repeated at most n times, as we saw in the proof of correctness. Thus the total complexity is

$$O(\text{preprocess}) + O(n) O(\text{popMin}) .$$

If the flows are first sorted, it would be

$$\begin{aligned} O(\text{preprocess}) &= O(\text{quicksort}) = O(n \log n) \text{ and} \\ O(\text{popMin}) &= O(1) , \end{aligned}$$

amounting to a total complexity of $O(n \log n)$. In the case a Fibonacci heap is used, it is

$$\begin{aligned} O(\text{preprocess}) &= O(\text{FibonacciHeap}) = O(n) \text{ and} \\ O(\text{popMin}) &= O(\text{find-min}) + O(\text{delete-min}) = O(\log n) , \end{aligned}$$

thus the total complexity is again $O(n \log n)$. \square

Proof that abs minimizes $\|\Delta_i\|_\infty$. Let *reduction* be the final value of the corresponding variable. It holds that

$$\begin{aligned} \forall i \in [n] : c'_i &> 0, \Delta_i = \text{reduction} , \\ \forall i \in [n] : c'_i &= 0, \Delta_i = x_i \text{ and} \\ \forall i \in [n] : c'_i &= 0, \text{reduction} \geq x_i , \end{aligned}$$

thus we deduce that

$$\|\Delta_i\|_\infty = \max_{1 \leq i \leq n} (x_i - c'_i) = \text{reduction} .$$

With the capacity configuration C' resulting from **abs**(), it holds that $\sum_{i=1}^n c'_i = F - V$. Suppose that there exists a configuration C_1 that maintains that property:

$$\sum_{i=1}^n c_{1,i} = F - V \tag{8}$$

and furthermore

$$\|\Delta_{1,i}\|_\infty = b < \text{reduction} . \tag{9}$$

Then it must be

$$\forall i \in [n], \Delta_{1,i} \leq b \Rightarrow \forall i \in [n], c_{1,i} \geq x_i - b .$$

Without loss of generality, suppose that x_i are sorted in ascending order. Then

$$\begin{aligned} \exists k' \in [n] \cup \{0\} : & \begin{cases} \forall i \leq k', x_i \leq \text{reduction} \\ \forall i > k', x_i > \text{reduction} \end{cases} \\ \text{and } \exists k_1 \in [n] \cup \{0\} : & \begin{cases} \forall i \leq k_1, x_i \leq b \\ \forall i > k_1, x_i > b \end{cases} . \end{aligned}$$

Since $b < \text{reduction}$, it is $k_1 \leq k'$. It is:

$$\begin{aligned} \forall i \in [k_1], 0 \leq c_{1,i} \leq x_i \text{ and} \\ \forall i \in [n] \setminus [k_1], x_i - b \leq c_{1,i} \leq x_i . \end{aligned}$$

Let all $c_{1,i}$ assume the smallest possible value according to the above restriction. Then

$$\forall i \in [k'] \setminus [k_1], x_i > b \text{ and} \quad (10)$$

$$\begin{aligned} \sum_{i=1}^n c_{1,i} &= \sum_{i=k_1+1}^n (x_i - b) \stackrel{(10)}{\geq} \sum_{i=k'+1}^n (x_i - b) \stackrel{(9)}{>} \\ &> \sum_{i=k'+1}^n (x_i - \text{reduction}) = \sum_{i=1}^n c'_i = F - V . \end{aligned}$$

We see that even with the minimum possible C_1 configuration, the hypothesis (8) is violated, thus the existence of C_1 is a contradiction. We have thus proven the proposition. \square

Proportional Equality Trust Transfer

Input : old flows x_i , value V

Output : new capacities c'_i

```

1 prop( $(x_i)$ ,  $V$ ) :
2    $n = \text{length}(x_i)$ 
3    $F = \sum_{i=1}^n x_i$ 
4   if ( $F < V$ )
5     return( $\perp$ )
6   for ( $i = 1$  to  $n$ )
7      $c'_i = x_i - \frac{V}{F} * x_i$ 
8   return( $\bigcup_{i=1}^n \{c'_i\}$ )

```

Proof of correctness for prop. We will first show that $\forall i \in [n], c'_i \leq x_i$. Let $i \in [n]$. According to line 7, which is the only line where c'_i is modified, it is

$$c'_i = x_i - \frac{V}{F}x_i = x_i \left(1 - \frac{V}{F}\right) . \quad (11)$$

Since $0 < V \leq F$, it is

$$0 < \frac{V}{F} \leq 1 \Rightarrow 0 \leq 1 - \frac{V}{F} < 1 . \quad (12)$$

From (11) and (12), along with the fact that $x_i \geq 0$, it is straightforward to see that $c'_i \leq x_i$, thus the requirement (5) is satisfied.

We will now show that $\sum_{i=1}^n c'_i = F - V$. At the end of the execution it is

$$\sum_{i=1}^n c'_i = \sum_{i=1}^n \left(x_i - \frac{V}{F}x_i\right) = \sum_{i=1}^n x_i - \frac{V}{F} \sum_{i=1}^n x_i \stackrel{\sum_{i=1}^n x_i = F}{=} F - V ,$$

thus the requirement (6) is satisfied. \square

Complexity of prop. The complexity of lines 2 - 3, 6 - 7 and 8 is $O(n)$ and the complexity of lines 4 - 5 is $O(1)$, thus the total complexity of the algorithm is $O(n)$. \square

Naive algorithms result in $c'_i \leq x_i$, thus according to the Invariability Theorem (4), $\|\delta_i\|_1$ is invariable for any of the possible results C' of these algorithms and the resulting norm is not necessarily the minimum. The following algorithms concentrate on finding a configuration C' that achieves $F' = F - V$ while minimizing two δ_i norms, $\|\delta_i\|_\infty$ and $\|\delta_i\|_1$. We start with the $\|\delta_i\|_\infty$ minimizer.

$\|\delta_i\|_\infty$ minimizer

Input : old capacities c_i , source A, drain B, value V, ϵ_1 , ϵ_2

Output : new capacities c'_i

```

1  dinfmin( $c_i$ , V,  $\epsilon_1$ ,  $\epsilon_2$ ) :
2    n = length( $c_i$ )
3    if ( $\epsilon_1 < 0$  or  $\epsilon_2 < 0$ )
4      return( $\perp$ )
5    F = maxFlow(A, B, C)
6    if (F < V)
7      return( $\perp$ )
8     $\delta_{max} = \max(C)$ 
9     $\delta^* = \text{BinSearch}(0, \delta_{max}, F - V, n, A, B, C, \epsilon_1, \epsilon_2)$ 
```

```

10   for (i = 1 to n)
11      $c'_i = \max(0, c_i - \delta^*)$ 
12   return( $\bigcup_{i=1}^n \{c'_i\}$ )

```

Since trust should be considered as a continuous unit and binary search bisects the interval containing the solution on each recursive call, inclusion of the ϵ -parameters in `BinSearch` is necessary for the algorithm to complete in a finite number of steps.

Binary Search Function for $\|\delta_i\|_\infty$ minimizer

Input : bot, top, F' , n, source A, drain B, capacities C,

ϵ_1, ϵ_2

Output : δ^*

```

1 BinSearch(bot, top,  $F'$ , n, C,  $\epsilon_1, \epsilon_2$ ) :
2   if (bot == top)
3     return(bot)
4   else
5      $mid = \frac{top + bot}{2}$ 
6     for (i = 1 to n)
7        $c'_i = \max(0, c_i - mid)$ 
8     if ( $\maxFlow(A, B, C') < F' - \epsilon_1$ )
9       return(BinSearch(bot, mid,  $F'$ , n, A, B, C,  $\epsilon_1, \epsilon_2$ ))
10    else if ( $\maxFlow(A, B, C') > F' + \epsilon_2$ )
11      return(BinSearch(mid, top,  $F'$ , n, A, B, C,  $\epsilon_1, \epsilon_2$ ))
12    else
13      return(mid)

```

Let $\delta \in \left[0, \max_{1 \leq i \leq n} \{c_i\}\right]$. Furthermore, let C' such that $\forall i \in [n], c'_i = \max(0, c_i - \delta)$. We define $\maxFlow(\delta) = \maxFlow(A, B, C') = F'$ and $\maxFlow(\delta) = X'$. Conventions similar to F, X and C hold for F', X' and δ as far as subscripts are concerned.

Lemma 2 (maxFlow Monotonicity).

It holds that $\forall \delta \in \left[0, \max_{1 \leq i \leq n} \{c_i\}\right]$ such that $\maxFlow(\delta) < F$, the function $\maxFlow(\delta)$ is strictly decreasing with respect to δ .

Proof. Suppose that

$$\exists \delta_1, \delta_2 : \delta_1 < \delta_2 \wedge \maxFlow(\delta_1) \leq \maxFlow(\delta_2) < F .$$

We choose X'_1, X'_2 such that

$$\forall i \in [n], x'_{1,i} \leq x_i \wedge x'_{2,i} \leq x_i .$$

This is always possible because we can derive the maximum flows X'_1 and X'_2 starting with X and reducing the flow along paths beginning from edges with capacities $c'_{1,i} < x_i$ and $c'_{2,i} < x_i$ respectively, until the flows become valid.

Define $MinCut(\delta)$ as the minimum cut set of the $MaxFlow(\delta)$ configuration. Let

$$S_j = \{i \in [n] : v_i \in N^+(A) \cap MinCut(\delta_j)\}, j \in \{1, 2\} .$$

Suppose that $S_j = \emptyset$. Then

$$MinCut(\delta_j) = MinCut(C) \Rightarrow maxFlow(\delta_j) = F, j \in \{1, 2\} ,$$

which is a contradiction. Thus $S_j \neq \emptyset$. Moreover, it holds that $S_1 \subseteq S_2$, since $\forall i \in [n], c'_{2,i} \leq c'_{1,i}$. More precisely, it is

$$\forall i \in [n] : c'_{2,i} > 0, c'_{2,i} < c'_{1,i} .$$

Every node in the $MinCut(\delta_j)$ is saturated, thus

$$\forall i \in S_1, x'_{j,i} = c'_{j,i}, j \in \{1, 2\} .$$

Thus $\sum_{i \in S_1} x'_{2,i} < \sum_{i \in S_1} x'_{1,i}$ and, since $maxFlow(\delta_1) \leq maxFlow(\delta_2)$, we conclude that for X'_1, X'_2 it is $\sum_{i \in [n] \setminus S_1} x'_{2,i} > \sum_{i \in [n] \setminus S_1} x'_{1,i}$. However, since $x'_{i,j} \leq x_i, j \in \{1, 2\}$, the configuration X'' such that

$$\begin{aligned} \forall i \in S_1, x''_i &= x'_{1,i} \\ \forall i \in [n] \setminus S_1, x''_i &= x'_{2,i} \end{aligned}$$

is a valid flow configuration for C'_1 and then

$$F'_1 \geq \sum_{i \in S_1} x''_i + \sum_{i \in [n] \setminus S_1} x''_i = \sum_{i \in S_1} x'_{1,i} + \sum_{i \in [n] \setminus S_1} x'_{2,i} > maxFlow(\delta_1) ,$$

which is a contradiction because $F'_1 = maxFlow(\delta_1)$ by the hypothesis. Thus $maxFlow(\delta_1) > maxFlow(\delta_2)$ and, since δ_1, δ_2 were chosen arbitrarily with the restriction $\delta_1 < \delta_2$, we deduce that the proposition holds. \square

From the previous lemma we deduce that, given a $V \in (0, F)$, if we determine a δ such that $\text{maxFlow}(\delta) = F - V$, this δ is unique. Let C' such that $\forall i \in [n], c'_i = \max(0, c_i - \delta)$. We will see that $\delta = \min \|\delta_i\|_\infty : F' = F - V$.

Proof of correctness for Binary Search. Suppose that

$$[F' - \epsilon_1, F' + \epsilon_2] \subset [\text{maxFlow}(\text{top}), \text{maxFlow}(\text{bot})] .$$

We will prove that

$$\text{maxFlow}(\delta^*) \in [F' - \epsilon_1, F' + \epsilon_2] .$$

First of all, we should note that if an invocation of **BinSearch** returns without calling **BinSearch** again (line 3 or 13), its return value will be equal to the return value of the initial invocation of **BinSearch**, as we can see on lines 9 and 11, where the return value of the invoked **BinSearch** is returned without any modification. The case where **BinSearch** is called recursively is analyzed next:

If $\text{maxFlow}\left(\frac{\text{top}+\text{bot}}{2}\right) < F' - \epsilon_1$ (line 8) then, by the **maxFlow** monotonicity lemma, $\delta^* \in \left[\text{bot}, \frac{\text{top}+\text{bot}}{2}\right)$. As we see on line 9, the interval $\left(\frac{\text{top}+\text{bot}}{2}, \text{top}\right]$ is discarded when the next **BinSearch** is recursively called. Since $F' + \epsilon_2 \leq \text{maxFlow}(\text{bot})$, we have

$$[F' - \epsilon_1, F' + \epsilon_2] \subset \left[\text{maxFlow}\left(\frac{\text{top} + \text{bot}}{2}\right), \text{maxFlow}(\text{bot})\right]$$

and the length of the available interval is divided by 2.

Similarly, if $\text{maxFlow}\left(\frac{\text{top}+\text{bot}}{2}\right) > F' + \epsilon_2$ (line 10) then, by the **maxFlow** monotonicity lemma, it holds that $\delta^* \in \left(\frac{\text{top}+\text{bot}}{2}, \text{top}\right]$. According to line 11, the interval $\left[\text{bot}, \frac{\text{top}+\text{bot}}{2}\right)$ is discarded when the next **BinSearch** is recursively called. Since $F' - \epsilon_1 \geq \text{maxFlow}(\text{top})$, we have

$$[F' - \epsilon_1, F' + \epsilon_2] \subset \left[\text{maxFlow}(\text{top}), \text{maxFlow}\left(\frac{\text{top} + \text{bot}}{2}\right)\right]$$

and the length of the available interval is divided by 2.

As we saw, it holds that

$$[F' - \epsilon_1, F' + \epsilon_2] \subset [\text{maxFlow}(\text{top}), \text{maxFlow}(\text{bot})]$$

in every recursive call and $\text{top} - \text{bot}$ is divided by 2 in every call. It is $|[F' - \epsilon_1, F' + \epsilon_2]| = \epsilon_1 + \epsilon_2$. Let $\text{bot}_0, \text{top}_0$ the input values given to the

initial invocation of **BinSearch**, bot_j, top_j the input values given to the j -th recursive call of **BinSearch** and $len_j = |[bot_j, top_j]| = top_j - bot_j$. We have

$$\begin{aligned} \forall j > 0, len_j = top_j - bot_j &= \frac{top_{j-1} - bot_{j-1}}{2} \Rightarrow \\ \forall j > 0, len_j &= \frac{top_0 - bot_0}{2^j} . \end{aligned}$$

We understand that in the worst case

$$len_j = \epsilon_1 + \epsilon_2 \Rightarrow 2^j = \frac{top_0 - bot_0}{\epsilon_1 + \epsilon_2} \Rightarrow j = \log_2\left(\frac{top_0 - bot_0}{\epsilon_1 + \epsilon_2}\right) .$$

Also, as we saw earlier, δ^* is always in the available interval, thus it holds that

$$maxFlow(\delta^*) \in [F' - \epsilon_1, F' + \epsilon_2] .$$

□

Complexity of Binary Search. Lines 2 - 3 have complexity $O(1)$, lines 6 - 7 have complexity $O(n)$, lines 8 - 13 have complexity $O(maxFlow) + O(BinSearch)$. As we saw in the proof of correctness for **BinSearch**, we need at most $\log_2\left(\frac{top-bot}{\epsilon_1+\epsilon_2}\right)$ recursive calls of **BinSearch**. Thus the function **BinSearch** has worst-case complexity

$$O\left((maxFlow + n) \log_2\left(\frac{top - bot}{\epsilon_1 + \epsilon_2}\right)\right) .$$

□

Proof of correctness for $dinfmin$. Let $C' = dinfmin(c_i, V, \epsilon_1, \epsilon_2)$. We will show that

$$F' \in [F - V - \epsilon_1, F - V + \epsilon_2] .$$

We can easily see that

$$\begin{aligned} maxFlow(0) &= F \text{ and} \\ maxFlow\left(\max_{i \in [n]} \{c_i\}\right) &= 0 , \end{aligned}$$

thus $\delta^* \in \max_{i \in [n]} \{c_i\}$. According to the proof of correctness for **BinSearch**, we know that $maxFlow(\delta^*) \in [F - V - \epsilon_1, F - V + \epsilon_2]$, given that ϵ_1, ϵ_2 are chosen so that $F - V - \epsilon_1 \geq 0, F - V + \epsilon_2 \leq F$, so as to satisfy the condition $[F' - \epsilon_1, F' + \epsilon_2] \subset [maxFlow(top), maxFlow(bot)]$. □

Complexity of dinfmin. The complexity of lines 3 - 4 and 6 - 7 is $O(1)$, the complexity of lines 2, 5, 8, 10 - 11 and 12 is $O(n)$ and the complexity of line 9 is $O(\text{BinSearch}) = O\left((\text{maxFlow} + n) \log_2 \left(\frac{\delta_{\text{max}}}{\epsilon_1 + \epsilon_2}\right)\right)$, thus the total complexity of dinfmin is $O\left((\text{maxFlow} + n) \log_2 \left(\frac{\delta_{\text{max}}}{\epsilon_1 + \epsilon_2}\right)\right)$. \square

We will now concentrate on finding a capacity configuration C' such that $F' = F - V$ and that minimizes $\sum_{i=1}^n (c_i - c'_i) = \|\delta_i\|_1$. We treat the flow problem as a linear programming problem. Next we see the formulation of the problem in this form, along with a breakdown of each relevant matrix and vector. In matrix form, the maximum flow problem can be seen as:

$$\begin{aligned} AX &\leq B \\ \max CX \end{aligned}$$

Where:

$$X = \left[\underbrace{c'_{11} \dots c'_{1j} \dots c'_{1n}}_n \quad \underbrace{f'_{11} \dots f'_{1j} \dots f'_{1n}}_n \quad \vdots \quad \underbrace{f'_{i1} \dots f'_{ij} \dots f'_{in}}_n \quad \vdots \quad \underbrace{f'_{n1} \dots f'_{nj} \dots f'_{nn}}_n \right]^T \Bigg\}^n$$

$$X : (n^2 + n) \times 1$$

$$B = \left[\underbrace{c_{11} \dots c_{1j} \dots c_{1n}}_n \right. \\ \left. \begin{array}{c} \underbrace{0 \dots 0}_n \\ \underbrace{c_{21} \dots c_{2j} \dots c_{2n}}_n \\ \vdots \\ c_{i1} \dots c_{ij} \dots c_{in} \\ \vdots \\ c_{n1} \dots c_{nj} \dots c_{nn} \end{array} \right] \left. \vphantom{\begin{array}{c} \underbrace{0 \dots 0}_n \\ \underbrace{c_{21} \dots c_{2j} \dots c_{2n}}_n \\ \vdots \\ c_{i1} \dots c_{ij} \dots c_{in} \\ \vdots \\ c_{n1} \dots c_{nj} \dots c_{nn} \end{array}} \right\} n \\ \underbrace{0 \dots 0}_{n-2} \\ F - V]^\top$$

$$B : (n^2 + 2n - 1) \times 1$$

$$C = [\underbrace{0 \dots 0}_n \underbrace{1 \dots 1}_n \underbrace{0 \dots 0}_{n(n-1)}]$$

$$C : 1 \times (n^2 + n)$$

A is implied by the following constraints.

$$A : (n^2 + 2n - 1) \times (n^2 + n)$$

Let $n = |\mathcal{V}|$, A represented by 1 and B by n . The constraints are:

$$\begin{aligned} \forall j \in [n], \quad c'_{1j} &\leq c_{1j} && (n \text{ constraints}) \\ \forall j \in [n], f'_{1j} - c'_{1j} &\leq 0 && (n \text{ constraints}) \\ \forall i \in [n] \setminus \{1\}, \forall j \in [n], \quad f_{ij} &\leq c_{ij} && (n^2 - n \text{ constraints}) \\ \forall i \in [n] \setminus \{1, n\}, \quad \sum_{j \in [n]} f'_{ij} &= \sum_{j \in [n]} f'_{ji} && (n - 2 \text{ constraints}) \\ \sum_{j \in [n]} f'_{1j} &= F - V && (1 \text{ constraint}) \end{aligned} \tag{13}$$

The constraints are $n^2 + 2n - 1$ in total. The special constraints are:

$$\begin{aligned} \forall j \in [n], c'_{1j} &\geq 0 \\ \forall i, j \in [n], f'_{ij} &\geq 0 \end{aligned}$$

The desired optimization is

$$\max \sum_{j \in [n]} f'_{1j} .$$

We would like to find a solution that, except for maximizing the flow, also minimizes $\|\delta_i\|_\infty$ at the same time. More precisely, we would like to optimize

$$\min \sum_{v \in \mathcal{V}} (c_{Av} - c'_{Av})$$

as well. Since we wish to optimize with regards to two objective functions, we approach the problem as follows: Initially, we ignore the minimization and derive the dual of the previous problem with respect to the maximisation. We then substitute the two problems' optimisations with an additional constraint that equates the two objective functions. Due to the Strong Duality theorem of linear programming [citation needed], this equality can be achieved only by the common optimal solution of the two problems. Next we treat the combination of constraints and variables of the primal and the dual problem, along with the newly introduced constraint and the previously ignored $\|\delta_i\|_1$ minimisation as a new linear problem. For every $j \in [n]$, the solution to this problem contains a c'_{1j} . These capacities will comprise the new configuration that player A requires.

We will now describe the dual problem in detail. In matrix form, the dual problem can be seen as:

$$\begin{aligned} A^T Y &\geq C^T \\ \min B^T Y \end{aligned}$$

Where A^\top, B^\top and C^\top are known and

$$Y = \begin{bmatrix} \underbrace{y_{c11} \dots y_{c1j} \dots y_{c1n}}_n \\ \underbrace{y_{fc11} \dots y_{fc1j} \dots y_{fc1n}}_n \\ \vdots \\ y_{fci1} \dots y_{fcij} \dots y_{fcin} \\ \vdots \\ y_{fcn1} \dots y_{fcn j} \dots y_{fcnn} \\ \underbrace{y_{f2} \dots y_{fi} \dots y_{f(n-1)} y_F}_{n-2} \end{bmatrix}^\top$$

$$Y : (n^2 + 2n - 1) \times 1$$

The constraints of the dual problem are:

$$\begin{aligned} \forall j \in [n], \quad & y_{c1j} - y_{fc1i} \geq 0 \text{ (} n \text{ constraints)} \\ & y_{fc11} + y_F \geq 1 \text{ (1 constraint)} \\ \forall j \in [n] \setminus \{1, n\}, \quad & y_{fc1j} - y_{fj} + y_F \geq 1 \text{ (} n - 2 \text{ constraints)} \\ & y_{fc1n} + y_F \geq 1 \text{ (1 constraint)} \\ \forall i \in [n] \setminus \{1, n\}, \quad & y_{fci1} + y_{fi} \geq 0 \text{ (} n - 2 \text{ constraints)} \\ \forall i, j \in [n] \setminus \{1, n\}, \quad & y_{fcij} + y_{fi} - y_{fj} \geq 0 \text{ (} n^2 - 4n + 4 \text{ constraints)} \\ \forall i \in [n] \setminus \{1, n\}, \quad & y_{fcin} + y_{fi} \geq 0 \text{ (} n - 2 \text{ constraints)} \\ & y_{fcn1} \geq 0 \text{ (1 constraint)} \\ \forall j \in [n] \setminus \{1, n\}, \quad & y_{fcn j} - y_{fj} \geq 0 \text{ (} n - 2 \text{ constraints)} \\ & y_{fcnn} \geq 0 \text{ (1 constraint)} \end{aligned} \tag{14}$$

The constraints are $n^2 + n$ in total. The special constraints are:

$$\begin{aligned} \forall j \in [n], \quad & y_{c1j} \geq 0 \\ \forall i, j \in [n], \quad & y_{fcij} \geq 0 \\ \forall i \in [n] \setminus \{1, n\}, \quad & y_{fi} \in \mathbb{R} \\ & y_F \in \mathbb{R} \end{aligned}$$

The desired optimization is

$$\min \left\{ \sum_{j \in [n]} c_{1j} y_{c1j} + \sum_{i \in [n]} \sum_{j \in [n]} c_{ij} y_{fcij} + (F - V) y_F \right\}.$$

Everything is now in place to define the linear problem whose solution C' yields $\maxFlow(C') = F - V$ and minimizes $\|\delta_i\|_\infty$. The constraints are (13) and (14) supplemented with a constraint that equates the two problems' optimality functions:

$$\sum_{j \in [n]} f'_{1j} = \sum_{j \in [n]} c_{1j} y_{c1j} + \sum_{i \in [n]} \sum_{j \in [n]} c_{ij} y_{fcij} + (F - V) y_F .$$

The desired optimisation is

$$\min \sum_{j \in [n]} (c_{1j} - c'_{1j}) .$$