

# Appendices

## A. Proofs of Main Theorems

### Proof of Theorem 1.

A brief roadmap of the proof is as follows. We first show that there exist polytopes in the  $0 - 1$  hypercube, parameterized by  $\gamma \in \mathbf{R}^n$ , that correspond to worst-case topologies (see (12)); the remaining of the proof deals with identifying the worst-case polytope within this class, *i.e.*, the worst-case value of the parameter  $\gamma$ , utilizing symmetry and optimization theory arguments.

Geometrically, the  $\alpha$ -fair allocation of any convex utility set in the  $0 - 1$  hypercube lies on its boundary. Consider now the supporting hyperplane at the  $\alpha$ -fair allocation, defined by the gradient of  $W_\alpha$ . Intuitively, any set that is contained in the polytope defined by that supporting hyperplane (and the  $0 - 1$  hypercube) would have the same  $\alpha$ -fair allocation. However, that does not hold true for the utilitarian or max-min allocations. In fact, by considering convex supersets of the original utility set, contained in the described polytope, one could obtain higher values for the utilitarian and/or max-min objectives, while the  $\alpha$ -fair allocation remains constant. As such, one need only consider polytopes of the described form for worst-cases. Note that such an approach can be generalized in a straightforward manner for any similar settings where one considers multiple competing objective functions.

Without loss of generality, we assume that  $U$  is monotone<sup>5</sup>. This is because both schemes we consider, namely utilitarian and  $\alpha$ -fairness yield Pareto optimal allocations. In particular, suppose there exist allocations  $a \in U$  and  $b \notin U$ , with allocation  $a$  dominating allocation  $b$ , *i.e.*,  $0 \leq b \leq a$ . Note that allocation  $b$  can thus not be Pareto optimal. Then, we can equivalently assume that  $b \in U$ , since  $b$  cannot be selected by any of the schemes.

We also assume that the maximum achievable utilities of the players are equal to 1; the proof can be trivially modified otherwise.

By combining the above two assumptions, we get

$$e_j \in U, \quad \forall j = 1, \dots, n, \tag{5}$$

where  $e_j$  is the unit vector in  $\mathbf{R}^n$ , with the  $j$ th component equal to 1.

Fix  $\alpha > 0$  and let  $z = z(\alpha) \in U$  be the unique allocation under the  $\alpha$ -fairness criterion (since  $W_\alpha$  is strictly concave for  $\alpha > 0$ ), and assume, without loss of generality, that

$$z_1 \geq z_2 \geq \dots \geq z_n. \tag{6}$$

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<sup>5</sup>A set  $A \subset \mathbf{R}_+^n$  is called monotone if  $\{b \in \mathbf{R}^n \mid 0 \leq b \leq a\} \subset A, \forall a \in A$ , where the inequality sign notation for vectors is used for componentwise inequality.

The necessary first order condition for the optimality of  $z$  can be expressed as

$$\nabla W_\alpha(z)^T(u - z) \leq 0 \Rightarrow \sum_{j=1}^n z_j^{-\alpha}(u_j - z_j) \leq 0, \quad \forall u \in U,$$

or equivalently

$$\gamma^T u \leq 1, \quad \forall u \in U, \quad (7)$$

where

$$\gamma_j = \frac{z_j^{-\alpha}}{\sum_i z_i^{1-\alpha}}, \quad j = 1, \dots, n. \quad (8)$$

Note that (6) implies

$$\gamma_1 \leq \gamma_2 \leq \dots \leq \gamma_n. \quad (9)$$

Using (5) and (7) we also get

$$\gamma_j = \gamma^T e_j \leq 1, \quad j = 1, \dots, n. \quad (10)$$

We now use (7), and the fact that each player has a maximum achievable utility of 1 to bound the sum of utilities under the utilitarian principle as follows:

$$\begin{aligned} \text{SYSTEM}(U) &= \max \{ \mathbf{1}^T u \mid u \in U \} \\ &\leq \max \{ \mathbf{1}^T u \mid 0 \leq u \leq \mathbf{1}, \gamma^T u \leq 1 \}. \end{aligned} \quad (11)$$

Using the above inequality,

$$\begin{aligned} \text{POF}(U; \alpha) &= \frac{\text{SYSTEM}(U) - \text{FAIR}(U; \alpha)}{\text{SYSTEM}(U)} \\ &= 1 - \frac{\text{FAIR}(U; \alpha)}{\text{SYSTEM}(U)} \\ &= 1 - \frac{\sum_{j=1}^n z_j}{\text{SYSTEM}(U)} \\ &\leq 1 - \frac{\sum_{j=1}^n z_j}{\max \{ \mathbf{1}^T u \mid 0 \leq u \leq \mathbf{1}, \gamma^T u \leq 1 \}}. \end{aligned} \quad (12)$$

The optimization problem in (12) is the linear relaxation of the well-studied knapsack problem, a version of which we review next. Let  $w \in \mathbf{R}_+^n$  be such that  $0 < w_1 \leq \dots \leq w_n \leq 1$  (in particular,  $\gamma$  satisfies those conditions). Then, one can show (see Bertsimas and Tsitsiklis (1997)) that the linear optimization problem

$$\begin{aligned} &\text{maximize} && \mathbf{1}^T y \\ &\text{subject to} && w^T y \leq 1 \\ &&& 0 \leq y \leq \mathbf{1}, \end{aligned} \quad (13)$$

has an optimal value equal to  $\ell(w) + \delta(w)$ , where

$$\ell(w) = \max \left\{ i \mid \sum_{j=1}^i w_j \leq 1, i \leq n-1 \right\} \in \{1, \dots, n-1\} \quad (14)$$

$$\delta(w) = \frac{1 - \sum_{j=1}^{\ell(w)} w_j}{w_{\ell(w)+1}} \in [0, 1]. \quad (15)$$

We can apply the above result to compute the optimal value of the problem in (12),

$$\max \left\{ \mathbf{1}^T u \mid 0 \leq u \leq \mathbf{1}, \gamma^T u \leq 1 \right\} = \ell(\gamma) + \delta(\gamma). \quad (16)$$

The bound from (12) can now be rewritten,

$$\text{POF}(U; \alpha) \leq 1 - \frac{\sum_{j=1}^n z_j}{\ell(\gamma) + \delta(\gamma)}. \quad (17)$$

Consider the set  $S$  in the  $(n+3)$ -dimensional space, defined by the following constraints with variables  $d \in \mathbf{R}$ ,  $\lambda \in \mathbf{N}$  and  $x_1, \dots, x_\lambda, \bar{x}_{\lambda+1}, \underline{x}_{\lambda+1}, x_{\lambda+2}, \dots, x_n \in \mathbf{R}$ . The variables  $d$  and  $\lambda$  correspond to  $\delta$  and  $\lambda$  accordingly, whereas  $x$  corresponds to  $z$ . Note also that we associate two variables,  $\bar{x}_{\lambda+1}$  and  $\underline{x}_{\lambda+1}$ , with  $z_{\lambda+1}$ .

$$0 \leq d \leq 1 \quad (18a)$$

$$1 \leq \lambda \leq n-1 \quad (18b)$$

$$0 \leq x_n \leq \dots \leq x_{\lambda+2} \leq \underline{x}_{\lambda+1} \leq \bar{x}_{\lambda+1} \leq x_\lambda \leq \dots \leq x_1 \leq 1 \quad (18c)$$

$$x_n^{-\alpha} \leq x_1^{1-\alpha} + \dots + x_\lambda^{1-\alpha} + d \bar{x}_{\lambda+1}^{1-\alpha} + (1-d) \underline{x}_{\lambda+1}^{1-\alpha} + x_{\lambda+2}^{1-\alpha} + \dots + x_n^{1-\alpha} \quad (18d)$$

$$x_1^{-\alpha} + \dots + x_\lambda^{-\alpha} + d \bar{x}_{\lambda+1}^{-\alpha} \leq x_1^{1-\alpha} + \dots + x_\lambda^{1-\alpha} + d \bar{x}_{\lambda+1}^{1-\alpha} + (1-d) \underline{x}_{\lambda+1}^{1-\alpha} + x_{\lambda+2}^{1-\alpha} + \dots + x_n^{1-\alpha}. \quad (18e)$$

The introduction of those new variables will allow us to further simplify (17). In particular, we show that

$$\frac{\sum_{j=1}^n z_j}{\ell(\gamma) + \delta(\gamma)} \geq \min_{(d, \lambda, x) \in S} \frac{x_1 + \dots + x_\lambda + d \bar{x}_{\lambda+1} + (1-d) \underline{x}_{\lambda+1} + x_{\lambda+2} + \dots + x_n}{\lambda + d}. \quad (19)$$

We pick values for  $d$ ,  $\lambda$  and  $x$  that are such that (a) they are feasible for  $S$ , and (b) the function argument of the minimum, if evaluated at  $(d, \lambda, x)$ , is equal to the left-hand side of (19). In

particular, let

$$\begin{aligned} d &= \delta(\gamma), & \lambda &= \ell(\gamma), \\ x_j &= z_j, \quad j \neq \lambda + 1, & \bar{x}_{\lambda+1} &= \underline{x}_{\lambda+1} = z_{\lambda+1}. \end{aligned}$$

Then, (18a), (18b) and (18c) are satisfied because of (15), (14) and (6) respectively. By the definition of  $\gamma$  and the selected value of  $x$ , (18d) can be equivalently expressed as

$$\gamma_n \leq 1,$$

which is implied by (10). Similarly, (18e) is equivalent to

$$\gamma_1 + \dots + \gamma_{\ell(\gamma)} + \delta(\gamma)\gamma_{\ell(\gamma)+1} \leq 1,$$

which again holds true (by (15)). The function argument of the minimum, evaluated at the selected point, is clearly equal to the left-hand side of (19). Finally, the minimum is attained by the Weierstrass Theorem, since the function argument is continuous, and  $S$  is compact. Note that (18d) in conjunction with (18c) bound  $x_n$  away from 0. In particular, if  $\alpha \geq 1$ , we get

$$x_n^{-\alpha} \leq x_1^{1-\alpha} + \dots + x_n^{1-\alpha} \leq nx_n^{1-\alpha} \Rightarrow x_n \geq \frac{1}{n}.$$

Similarly, for  $\alpha < 1$  we get

$$x_n \geq \left(\frac{1}{n}\right)^{\frac{1}{\alpha}}.$$

To evaluate the minimum in (19), one can assume without loss of generality that for a point  $(d', \lambda', x') \in S$  that attains the minimum, we have

$$x'_1 = \dots = x'_\lambda = \bar{x}'_{\lambda+1}, \quad \underline{x}'_{\lambda+1} = x'_{\lambda+2} = \dots = x'_n. \quad (20)$$

Technical details are included in Section C. Using this observation, we can further simplify (19). In particular, consider the set  $T \subset \mathbf{R}^3$ , defined by the following constraints, with variables  $x_1$ ,  $x_2$  and  $y$  (since  $x'_1 = \dots = x'_\lambda = \bar{x}'_{\lambda+1}$ , we associate  $x_1$  with them, and similarly we associate  $x_2$  with the remaining variables of  $x'$ ; variable  $y$  is associated with  $\lambda + d$ ):

$$0 \leq x_2 \leq x_1 \leq 1 \quad (21a)$$

$$1 \leq y \leq n \quad (21b)$$

$$x_2^{-\alpha} \leq yx_1^{1-\alpha} + (n-y)x_2^{1-\alpha} \quad (21c)$$

$$yx_1^{-\alpha} \leq yx_1^{1-\alpha} + (n-y)x_2^{1-\alpha}. \quad (21d)$$

Using similar arguments as in showing (19), one can then show that

$$\min_{(d,\lambda,x) \in S} \frac{x_1 + \dots + x_\lambda + d\bar{x}_{\lambda+1} + (1-d)\underline{x}_{\lambda+1} + x_{\lambda+2} + \dots + x_n}{\lambda + d} \geq \min_{(x_1,x_2,y) \in T} \frac{yx_1 + (n-y)x_2}{y}. \quad (22)$$

If we combine (17), (19), (22) we get

$$\text{POF}(U; \alpha) \leq 1 - \min_{(x_1,x_2,y) \in T} \frac{yx_1 + (n-y)x_2}{y}. \quad (23)$$

The final step is the evaluation of the minimum above. Let  $(x_1^*, x_2^*, y^*) \in T$  be a point that attains the minimum. Then, we have

$$y^* < n, \quad x_2^* < x_1^*. \quad (24)$$

To see this, suppose that  $x_2^* = x_1^*$ . Then, the minimum is equal to  $\frac{nx_1^*}{y^*}$ . But, constraint (21d) yields that  $nx_1^* \geq y^*$ , in which case the minimum is greater than or equal to 1. Then, (23) yields that the price of fairness is always 0, a contradiction. If  $y^* = n$ , (21d) suggests that  $x_1^* = 1$ . Also, the minimum is equal to  $x_1^* = 1$ , a contradiction.

We now show that (21c-21d) are active at  $(x_1^*, x_2^*, y^*)$ . We argue for  $\alpha \geq 1$  and  $\alpha < 1$  separately.

$\alpha \geq 1$ : Suppose that (21c) is inactive. Then, a small enough reduction in the value of  $x_2^*$  preserves feasibility (with respect to  $T$ ), and also yields a strictly lower value for the minimum (since  $y^* < n$ , by (24)), thus contradicting that the point attains the minimum. Similarly, if (21d) is inactive, a small enough reduction in the value of  $x_1^*$  leads to a contradiction.

$\alpha < 1$ : Suppose that (21d) is inactive at  $(x_1^*, x_2^*, y^*)$ . Then, we increase  $y^*$  by a small positive value, such that (21d) and (21b) are still satisfied. Constraint (21c) is then relaxed, since  $(x_1^*)^{1-\alpha} > (x_2^*)^{1-\alpha}$ . The minimum then has a strictly lower value, a contradiction. Hence, (21d) is active at any point that attains the minimum. If we solve for  $y$  and substitute back, the objective of the minimum becomes

$$x_1 + x_2^\alpha(x_1^{-\alpha} - x_1^{1-\alpha}), \quad (25)$$

and the constraints defining the set  $T$  simplify to

$$0 \leq x_2 \leq x_1 \leq 1 \quad (26a)$$

$$x_1^{-\alpha} - x_1^{1-\alpha} + x_2^{1-\alpha} \leq nx_1^{-\alpha}x_2. \quad (26b)$$

In particular, constraint (26b) correspond to constraint (21c). In case (21c) is not active at a minimum, so is (26b). But then, a small enough reduction in the value of  $x_2^*$  leads to a

contradiction.

Since for any point that attains the minimum constraints (21c-21d) are active, we can use the corresponding equations to solve for  $x_1$  and  $x_2$ . We get

$$x_1 = \frac{y^{\frac{1}{\alpha}}}{n - y + y^{\frac{1}{\alpha}}}, \quad (27)$$

$$x_2 = \frac{1}{n - y + y^{\frac{1}{\alpha}}}. \quad (28)$$

If we substitute back to (23), we get

$$\text{POF}(U; \alpha) \leq 1 - \min_{x \in [1, n]} \frac{x^{1+\frac{1}{\alpha}} + n - x}{x^{1+\frac{1}{\alpha}} + (n - x)x}.$$

The asymptotic analysis is included in Section C.

**Proof of Theorem 2.** We follow similar steps to the ones in the proof of Theorem 1. Thus, assume that  $U$  is monotone, the maximum achievable utilities of the players are equal to 1 and that  $z_1 \geq z_2 \geq \dots \geq z_n$  (where  $z = z(\alpha) \in U$  is the unique  $\alpha$ -fair allocation). Then, for the variable  $\gamma$  (defined as in (8)), we similarly have

$$\gamma^T u \leq 1, \quad \forall u \in U,$$

and

$$\gamma_1 \leq \gamma_2 \leq \dots \leq \gamma_n \leq 1.$$

We use the above to bound the maximum value of the fairness metric

$$\max \left\{ \min_{j=1, \dots, n} u_j \mid u \in U \right\} \leq \max \left\{ \min_{j=1, \dots, n} u_j \mid 0 \leq u \leq \mathbf{1}, \gamma^T u \leq 1 \right\} = \frac{1}{\mathbf{1}^T \gamma},$$

where the equality follows from  $z \leq \mathbf{1}$  and  $\mathbf{1}^T \gamma \geq 1$ .

We bound the price of efficiency using  $z_1 \geq \dots \geq z_n$ ,  $\gamma_n \leq 1$  and the inequality above as follows:

$$\begin{aligned}
\text{POE}(U; \alpha) &= \frac{\max_{u \in U} \min_{j=1, \dots, n} u_j - \min_{j=1, \dots, n} z_j(\alpha)}{\max_{u \in U} \min_{j=1, \dots, n} u_j} \\
&= 1 - \frac{z_n}{\max_{u \in U} \min_{j=1, \dots, n} u_j} \\
&\leq 1 - z_n \mathbf{1}^T \gamma \\
&= 1 - \frac{z_n (z_1^{-\alpha} + z_2^{-\alpha} + \dots + z_n^{-\alpha})}{z_1^{1-\alpha} + z_2^{1-\alpha} + \dots + z_n^{1-\alpha}} \\
&= 1 - f^*,
\end{aligned}$$

where  $f^*$  is the optimal value of the problem

$$\begin{aligned}
&\text{minimize} && \frac{z_n (z_1^{-\alpha} + z_2^{-\alpha} + \dots + z_n^{-\alpha})}{z_1^{1-\alpha} + z_2^{1-\alpha} + \dots + z_n^{1-\alpha}} \\
&\text{subject to} && 0 \leq z_n \leq z_{n-1} \leq \dots \leq z_1 \leq 1 \\
&&& z_n^{-\alpha} \leq z_1^{1-\alpha} + z_2^{1-\alpha} + \dots + z_n^{1-\alpha}.
\end{aligned} \tag{29}$$

Let  $z^*$  be an optimal solution of (29) (guaranteed to exist by the Weierstrass Theorem). Then, without loss of generality we can assume that (a)  $z_1^* = z_2^* = \dots = z_{n-1}^*$  and (b)  $z_1^* = 1$ . Technical details are included in the Section C. Using those two assumptions,  $f^*$  is then equal to

$$\begin{aligned}
&\text{minimize} && \frac{(n-1)x + x^{1-\alpha}}{n-1 + x^{1-\alpha}} \\
&\text{subject to} && 0 \leq x \leq 1 \\
&&& x^{-\alpha} \leq n-1 + x^{1-\alpha}.
\end{aligned} \tag{30}$$

Finally, note that for  $x \in [0, 1]$  the function  $x^{-\alpha} - x^{1-\alpha} - n - 1$  is strictly decreasing, is positive for  $x$  small and negative for  $x = 1$ . Hence, for  $x \in [0, 1]$  the constraint  $x^{-\alpha} \leq n - 1 + x^{1-\alpha}$  is equivalent to  $x \geq \rho$ . As a result,

$$f^* = \min_{\rho \leq x \leq 1} \frac{(n-1)x + x^{1-\alpha}}{n-1 + x^{1-\alpha}}.$$

The asymptotic analysis is similar to the analysis in Theorem 1 and is omitted.

## B. More on Near Worst-case Examples for the Price of Fairness

We demonstrate how one can construct near worst-case examples, for which the price of fairness is very close to the bounds implied by Theorem 1, for any values of the problem parameters; the

number of players  $n$  and the value of the inequality aversion parameter  $\alpha$ . We then provide details about the bandwidth allocation problem in Section 3.1.1.

For any  $n \in \mathbf{N} \setminus \{0, 1\}$ ,  $\alpha > 0$ , we create a utility set using Procedure 1.

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**Procedure 1** Creation of near worst-case utility set

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**Input:**  $n \in \mathbf{N} \setminus \{0, 1\}$ ,  $\alpha > 0$

**Output:** utility set  $U$

- 1: compute  $y := \operatorname{argmin}_{x \in [1, n]} \frac{x^{1+\frac{1}{\alpha}} + n - x}{x^{1+\frac{1}{\alpha}} + (n-x)x}$
  - 2:  $x_1 \leftarrow \frac{y^{\frac{1}{\alpha}}}{n-y+y^{\frac{1}{\alpha}}}$  (as in (27))
  - 3:  $x_2 \leftarrow \frac{1}{n-y+y^{\frac{1}{\alpha}}}$  (as in (28))
  - 4:  $\ell \leftarrow \min \{\operatorname{round}(y), n-1\}$
  - 5:  $\gamma_i \leftarrow \frac{x_i^{-\alpha}}{yx_1^{1-\alpha} + (n-y)x_2^{1-\alpha}}$  for  $i = 1, 2$
  - 6:  $U \leftarrow \{u \in \mathbf{R}_+^n \mid \gamma_1 u_1 + \dots + \gamma_1 u_\ell + \gamma_2 u_{\ell+1} + \dots + \gamma_2 u_n \leq 1, \quad u \leq \mathbf{1} \forall j\}$
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The following proposition demonstrates why Procedure 1 creates utility sets that achieve a price of fairness very close to the bounds implied by Theorem 1.

**Proposition 1.** *For any  $n \in \mathbf{N} \setminus \{0, 1\}$ ,  $\alpha > 0$ , the output utility set  $U$  of Procedure 1 satisfies the conditions of Theorem 1. If  $y \in \mathbf{N}$ , the output utility set  $U$  satisfies the bound of Theorem 1 with equality.*

**Proof.** The output utility set  $U$  is a bounded polyhedron, hence convex and compact. Boundedness follows from positivity of  $\gamma_1$  and  $\gamma_2$ .

Note that the selection of  $x_1$ ,  $x_2$  and  $y$  in Procedure 1 corresponds to a point that attains the minimum of (23), hence all properties quoted in the proof of Theorem 1 apply. In particular, by (18d) we have  $\gamma_2 \leq 1$  and (21d) is tight,  $y\gamma_1 = 1$ . Moreover, the bound from Theorem 1 can be expressed as

$$\operatorname{POF}(U; \alpha) \leq 1 - \frac{yx_1 + (n-y)x_2}{y}.$$

The maximum achievable utility of the  $j$ th player is equal to 1. To see this, note that the definition of  $U$  includes the constraint  $u_j \leq 1$ , so it suffices to show that  $e_j \in U$ . For  $j \leq \ell$ , we have  $\gamma_1 \leq \gamma_1 y = 1$ . For  $j > \ell$ , we have  $\gamma_2 \leq 1$ . It follows that  $U$  satisfies the conditions of Theorem 1.

Suppose that  $y \in \mathbf{N}$ . By (24) and the choice of  $\ell$  in Procedure 1, we get  $\ell = y$ . Consider the vector  $z \in \mathbf{R}^n$  with  $z_1 = \dots = z_\ell = x_1$  and  $z_{\ell+1} = \dots = z_n = x_2$ . Then, the sufficient first order optimality condition for  $z$  to be the  $\alpha$ -fair allocation of  $U$  is satisfied, as for any  $u \in U$

$$\sum_{j=1}^n z_j^{-\alpha} (u_j - z_j) = x_1^{-\alpha} (u_1 + \dots + u_\ell) + x_2^{-\alpha} (u_{\ell+1} + \dots + u_n) - yx_1^{1-\alpha} - (n-y)x_2^{1-\alpha} \leq 0,$$



since  $\gamma_1(u_1 + \dots + u_\ell) + \gamma_2(u_{\ell+1} + \dots + u_n) \leq 1$ . Hence,

$$\text{FAIR}(U; \alpha) = \mathbf{1}^T z = yx_1 + (n - y)x_2.$$

For the efficiency-maximizing solution, since  $y\gamma_1 = 1$ , we get

$$\text{SYSTEM}(U) = y.$$

Then,

$$\text{POF}(U; \alpha) = 1 - \frac{yx_1 + (n - y)x_2}{y},$$

which is exactly the bound from Theorem 1. ■

The above result demonstrates why one should expect Procedure 1 to generate examples that have a price of fairness very close to the established bounds. In particular, Proposition 1 shows that the source of error between the price of fairness for the utility sets generated by Procedure 1 and the bound is the (potential) non-integrality of  $y$ . In case that error is “large”, one can search in the neighborhood of parameters  $\gamma_1$  and  $\gamma_2$  for an example that achieves a price closer to the bound, for instance by using finite-differencing derivatives and a gradient descent method (respecting feasibility).

### Near worst-case bandwidth allocation

We utilize Proposition 1 and Procedure 1 to construct near worst-case network topologies. In particular, one can show that the line-graph discussed in Section 3.1.1, actually corresponds to a worst-case topology in this setup.

Suppose that we fix the number of players  $n \geq 2$ , the desired inequality aversion parameter  $\alpha > 0$ , and follow Procedure 1. Further suppose that  $y \in \mathbf{N}$ , as in Proposition 1. Consider then a network with  $y$  links of unit capacity, in a line-graph topology: the routes of the first  $y$  flows are disjoint and they all occupy a single (distinct) link. The remaining  $n - y$  flows have routes that utilize all  $y$  links. Each flow derives a utility equal to its assigned nonnegative rate, which we denote  $u_1, \dots, u_n$ . We next show that the price of fairness for this network is equal to the bound of Theorem 1.

The output utility set of Procedure 1 achieves the bound, by Proposition 1, since  $y \in \mathbf{N}$ . Moreover, we also get that  $y\gamma_1 = 1$  and  $\gamma_2 = 1$ . Hence, the output utility set that achieves the bound can be formulated as

$$U = \{u \geq 0 \mid u_1 + \dots + u_y + y(u_{y+1} + \dots + u_n) \leq y, u \leq \mathbf{1}\}.$$

The utility set corresponding to the line-graph example above can be expressed using the non-negativity constraints of the flow rates, and the capacity constraints on each of the  $y$  links as

follows,

$$\bar{U} = \{u \geq 0 \mid u_j + u_{y+1} + \dots + u_n \leq 1, j = 1, \dots, y\}.$$

Clearly, the maximum sum of utilities under both sets is equal to  $y$ , simply by setting the first  $y$  components of  $u$  to 1. It suffices then to show that the two sets also share the same  $\alpha$ -fair allocation. In particular, by symmetry of  $U$  and strict concavity of  $W_\alpha$ , if  $u^F$  is its  $\alpha$  fair allocation, then  $u_1^F = \dots = u_y^F$ , and  $u_{y+1}^F = \dots = u_n^F$ . As a result, it follows that  $u^F \in \bar{U}$ . Finally, noting that all inequalities in the definition of  $U$  are also valid for  $\bar{U}$ , it follows that  $\bar{U} \subset U$  and that  $u^F$  is also the  $\alpha$ -fair allocation of  $\bar{U}$ .

## C. Auxiliary Results

**Proposition 2.** *For a point  $(d, \lambda, x) \in S$  that attains the minimum of (19),*

(a) *if  $\lambda + 1 < n$ , then without loss of generality*

$$\underline{x}_{\lambda+1} = x_{\lambda+2} = \dots = x_n, \text{ and,}$$

(b) *without loss of generality*

$$x_1 = \dots = x_\lambda = \bar{x}_{\lambda+1}.$$

**Proof.** (a) We drop the underline notation for  $\underline{x}_{\lambda+1}$  to simplify notation. Suppose that  $x_j > x_{j+1}$ , for some index  $j \in \{\lambda + 1, \dots, n - 1\}$ . We will show that there always exists a new point,  $(d, \lambda, x') \in S$ , for which  $x'_i = x_i$ , for all  $i \in \{1, \dots, n\} \setminus \{j, j + 1\}$ , and which either achieves the same objective with  $x'_j = x'_{j+1}$ , or it achieves a strictly lower objective.

If  $j = \lambda + 1$  and  $d = 1$ , we set  $x'_j = x'_{j+1} = x_{j+1}$ . The new point is feasible, and the objective attains the same value.

Otherwise, let  $x'_j = x_j - \epsilon$ , for some  $\epsilon > 0$ . We have two cases.

$\alpha \geq 1$ : Let  $x'_{j+1} = x_{j+1}$  and pick  $\epsilon$  small enough, such that  $x'_j \geq x'_{j+1}$ . Moreover, for the new point (compared to the feasible starting point) the left-hand sides of (18d) and (18e) are unaltered, whereas the right-hand sides are either unaltered (for  $\alpha = 1$ ) or greater, since  $x_j^{1-\alpha} < (x_j - \epsilon)^{1-\alpha}$  for  $\alpha > 1$ . Hence, the new point is feasible. It also achieves a strictly lower objective value.

$\alpha < 1$ : Let  $x'_{j+1} = x_{j+1} + \rho b \epsilon$ , where

$$b = \begin{cases} 1 - d, & \text{if } j = \lambda + 1, \\ 1, & \text{otherwise,} \end{cases}$$

$$\rho \in \left( \frac{x_j^{-\alpha}}{x_{j+1}^{-\alpha}}, 1 \right).$$

For  $\epsilon$  small enough, we have  $x'_j \geq x'_{j+1}$ . For the new point, the left-hand side of (18d) either decreases (if  $j + 1 = n$ ), or remains unaltered. The left-hand side of (18e) remains also unaltered. For the right-hand sides, since the only terms that change are those involving  $x_j$  and  $x_{j+1}$ , we use a first order Taylor series expansion to get

$$\begin{aligned} b \left( x'_j \right)^{1-\alpha} + \left( x'_{j+1} \right)^{1-\alpha} &= b (x_j - \epsilon)^{1-\alpha} + (x_{j+1} + \rho b \epsilon)^{1-\alpha} \\ &= b x_j^{1-\alpha} - b \epsilon (1 - \alpha) x_j^{-\alpha} + x_{j+1}^{1-\alpha} + \rho b \epsilon (1 - \alpha) x_{j+1}^{-\alpha} + O(\epsilon^2) \\ &= \left( b x_j^{1-\alpha} + x_{j+1}^{1-\alpha} \right) + b(1 - \alpha) \left( \rho x_{j+1}^{-\alpha} - x_j^{-\alpha} \right) \epsilon + O(\epsilon^2). \end{aligned}$$

By the selection of  $\rho$ , the coefficient of the first order term (with respect to  $\epsilon$ ) above is positive, and hence, for small enough  $\epsilon$  we get

$$b \left( x'_j \right)^{1-\alpha} + \left( x'_{j+1} \right)^{1-\alpha} > b x_j^{1-\alpha} + x_{j+1}^{1-\alpha}.$$

That shows that the right hand side increases, and the new point is feasible. Finally, the difference in the objective value is  $-b\epsilon + \rho b \epsilon$ , and thus negative.

(b) We drop the overline notation for  $\bar{x}_{\lambda+1}$  to simplify notation. Suppose that  $x_j > x_{j+1}$ , for some index  $j \in \{1, \dots, \lambda\}$ .

We will show that there always exists a new point,  $(d, \lambda, x') \in S$ , for which  $x'_i = x_i$ , for all  $i \in \{1, \dots, n\} \setminus \{j, j + 1\}$ , and which either achieves the same objective with  $x'_j = x'_{j+1}$ , or it achieves a strictly lower objective.

If  $j + 1 = \lambda + 1$  and  $d = 0$ , we set  $x'_j = x'_{j+1} = x_j$ . The new point is feasible, and the objective attains the same value.

Otherwise, let

$$\begin{aligned} x'_j &= x_j - \epsilon \\ x'_{j+1} &= x_{j+1} + \rho c \epsilon, \end{aligned}$$

for some  $\epsilon > 0$ , where

$$\begin{aligned}\rho &\in \left( \frac{x_{j+1}}{x_j}, \frac{x_{j+1}^{-\alpha}}{x_j^{-\alpha}} \right) \\ c &= \frac{x_j^{-\alpha}}{bx_{j+1}^{-\alpha}} \\ b &= \begin{cases} d, & \text{if } j+1 = \lambda+1, \\ 1, & \text{otherwise.} \end{cases}\end{aligned}$$

For  $\epsilon$  small enough, we have  $x'_j \geq x'_{j+1}$ . For the new point, the left-hand side of (18d) remains unaltered. For the left-hand side of (18e) we use a first order Taylor series expansion (similarly as above) to get

$$\begin{aligned}(x'_j)^{-\alpha} + b(x'_{j+1})^{-\alpha} &= (x_j - \epsilon)^{-\alpha} + b(x_{j+1} + \rho c \epsilon)^{-\alpha} \\ &= x_j^{-\alpha} + \epsilon \alpha x_j^{-\alpha-1} + bx_{j+1}^{-\alpha} - b\rho c \epsilon \alpha x_{j+1}^{-\alpha-1} + O(\epsilon^2) \\ &= (x_j^{-\alpha} + bx_{j+1}^{-\alpha}) + \epsilon \alpha x_j^{-\alpha-1} - \rho \epsilon \alpha x_j^{-\alpha} x_{j+1}^{-1} + O(\epsilon^2) \\ &= (x_j^{-\alpha} + bx_{j+1}^{-\alpha}) + \alpha x_j^{-\alpha-1} \left( 1 - \rho \frac{x_j}{x_{j+1}} \right) \epsilon + O(\epsilon^2).\end{aligned}$$

By the selection of  $\rho$ , the coefficient of the first order term (with respect to  $\epsilon$ ) above is negative, and hence, for small enough  $\epsilon$  we get that the left-hand side decreases.

For the right-hand side of (18d) and (18e), we similarly get that

$$\begin{aligned}(x'_j)^{1-\alpha} + b(x'_{j+1})^{1-\alpha} &= (x_j - \epsilon)^{1-\alpha} + b(x_{j+1} + \rho c \epsilon)^{1-\alpha} \\ &= x_j^{1-\alpha} - \epsilon(1-\alpha)x_j^{-\alpha} + bx_{j+1}^{1-\alpha} + b\rho c \epsilon(1-\alpha)x_{j+1}^{1-\alpha} + O(\epsilon^2) \\ &= (x_j^{1-\alpha} + bx_{j+1}^{1-\alpha}) + (1-\alpha)x_j^{-\alpha}(\rho-1)\epsilon + O(\epsilon^2).\end{aligned}$$

If for  $\alpha > 1$  we pick  $\rho < 1$ , and for  $\alpha < 1$  we pick  $\rho > 1$ , the first order term (with respect to  $\epsilon$ ) above is positive, and hence, for small enough  $\epsilon$  we get that the right-hand side increases for  $\alpha \neq 1$ . For  $\alpha = 1$ , the right-hand side remains unaltered.

In all cases, the new point is feasible, and the difference in the objective value is

$$-\epsilon + \rho c b \epsilon = (\rho c b - 1) \epsilon = \left( \rho \frac{x_j^{-\alpha}}{x_{j+1}^{-\alpha}} - 1 \right) \epsilon,$$

and thus negative (by the selection of  $\rho$ ). ■

**Proposition 3.** Let  $n \in \mathbf{N} \setminus \{0, 1\}$  and  $f : [1, n] \rightarrow \mathbf{R}$  be defined as

$$f(x; \alpha, n) = \frac{x^{1+\frac{1}{\alpha}} + n - x}{x^{1+\frac{1}{\alpha}} + (n-x)x}.$$

For any  $\alpha > 0$ ,

(a)  $-f$  is unimodal over  $[1, n]$ , and thus has a unique minimizer  $\xi^* \in [1, n]$ .

(b)  $\min_{x \in [1, n]} f(x; \alpha, n) = f(\xi^*; \alpha, n) = \Theta\left(n^{-\frac{\alpha}{\alpha+1}}\right)$ .

**Proof.** (a) The derivative of  $f$  is

$$f'(x; \alpha, n) = \frac{g(x)}{\left(x^{1+\frac{1}{\alpha}} + (n-x)x\right)^2},$$

where

$$g(x) = \left(1 - \frac{1}{\alpha}\right)x^{2+\frac{1}{\alpha}} + \frac{n+1}{\alpha}x^{1+\frac{1}{\alpha}} - n\left(1 + \frac{1}{\alpha}\right)x^{\frac{1}{\alpha}} - (x-n)^2.$$

Note that the sign of the derivative is determined by  $g(x)$ , since the denominator is positive for  $1 \leq x \leq n$ , that is,

$$\text{sgn } f'(x; \alpha, n) = \text{sgn } g(x). \quad (31)$$

We will show that  $g$  is strictly increasing over  $[1, n]$ . To this end, we have

$$g'(x) = x^{\frac{1}{\alpha}-1}q(x) + 2(n-x),$$

where

$$q(x) = \left(2 + \frac{1}{\alpha}\right)\left(1 - \frac{1}{\alpha}\right)x^2 + \left(1 + \frac{1}{\alpha}\right)\left(\frac{n+1}{\alpha}\right)x - \frac{n}{\alpha}\left(1 + \frac{1}{\alpha}\right).$$

Since we are interested in the domain  $[1, n]$ , it suffices to show that  $q(x) > 0$  over it. For  $\alpha > 1$ ,  $q$  is a convex quadratic, with its minimizer being equal to

$$-\frac{\left(1 + \frac{1}{\alpha}\right)\left(\frac{n+1}{\alpha}\right)}{2\left(2 + \frac{1}{\alpha}\right)\left(1 - \frac{1}{\alpha}\right)} < 0.$$

Hence,  $q(x) \geq q(1)$  for  $x \in [1, n]$ . Similarly, for  $\alpha < 1$ ,  $q$  is a concave quadratic, and as such, for  $x \in [1, n]$  we have  $q(x) \geq \min\{q(1), q(n)\}$ . For  $\alpha = 1$ ,  $q(x) = 2(n+1)x - 2n$ , which is positive for  $x \geq 1$ . Then,  $q(x) > 0$  in  $[1, n]$  for all  $\alpha > 0$ , if and only if  $q(1) > 0$  and  $q(n) > 0$ . Note that for  $r = 1$ , we get  $q(1) = 2$  and  $q(n) = 2n^2$ , and

$$\frac{dq(1)}{dr} = 2 > 0, \quad \frac{dq(n)}{dr} = 2n^2 > 0,$$

which demonstrates that  $g(1)$  and  $g(n)$  are positive. Furthermore,

$$g(n) = n^{1+\frac{1}{\alpha}}(n-1) > 0.$$

Using the above, the fact that  $g$  is continuous and strictly increasing over  $[1, n]$  and (31), we deduce that if  $g(1) < 0$ , there exists a unique  $m \in (1, n)$  such that

$$\operatorname{sgn} f'(x; \alpha, n) \begin{cases} < 0, & \text{if } 1 \leq x < m, \\ > 0, & \text{if } m < x \leq n. \end{cases}$$

Similarly, if  $g(1) \geq 0$ ,  $f$  is strictly increasing for  $1 \leq x \leq n$ . It follows that  $-f$  is unimodal.

- (b) Let  $\theta_n = n^{\frac{\alpha}{\alpha+1}}$ . Using the mean value Theorem, for every  $n \geq 2$ , there exists a  $\psi_n \in [\theta_n, \xi^*]$  (or  $[\xi^*, \theta_n]$ , depending on if  $\theta_n \leq \xi^*$ ), such that

$$f(\theta_n; \alpha, n) = f(\xi^*; \alpha, n) + f'(\psi_n; \alpha, n)(\theta_n - \xi^*),$$

or, equivalently,

$$\frac{f(\xi^*; \alpha, n)}{f(\theta_n; \alpha, n)} = 1 - \frac{f'(\psi_n; \alpha, n)(\theta_n - \xi^*)}{f(\theta_n; \alpha, n)}.$$

We will show that, for a sufficiently small  $\epsilon > 0$

- (I.)  $f'(\psi_n; \alpha, n) = O\left(n^{-\frac{\min\{1, \alpha\} + 2\alpha}{\alpha+1} + 2\epsilon}\right)$ ,
- (II.)  $\theta_n - \xi^* = O\left(n^{-\frac{\alpha}{\alpha+1} + \epsilon}\right)$ ,
- (III.)  $f(\theta_n; \alpha, n) = \Theta\left(n^{-\frac{\alpha}{\alpha+1}}\right)$ .

Using the above facts, it is easy to see that

$$\frac{f(\xi^*; \alpha, n)}{f(\theta_n; \alpha, n)} = 1 - \frac{f'(\psi_n; \alpha, n)(\theta_n - \xi^*)}{f(\theta_n; \alpha, n)} = 1 - O\left(n^{-\frac{\min\{1, \alpha\}}{\alpha+1} + 3\epsilon}\right) \rightarrow 1,$$

and thus  $f(\xi^*; \alpha, n) = \Theta\left(n^{-\frac{\alpha}{\alpha+1}}\right)$ .

- (I.) We first show that for any sufficiently large  $n$ ,

$$n^{\frac{\alpha}{\alpha+1} - \epsilon} \leq \xi^* \leq n^{\frac{\alpha}{\alpha+1} + \epsilon}. \quad (32)$$

By part (a),  $\xi^*$  is the unique root of  $g$  in the interval  $[1, n]$ . Moreover,  $g$  is strictly increasing.

The dominant term of

$$g\left(n^{\frac{\alpha}{\alpha+1}-\epsilon}\right) = \left(1 - \frac{1}{\alpha}\right) n^{(2+\frac{1}{\alpha})(\frac{\alpha}{\alpha+1}-\epsilon)} + \frac{1}{\alpha} n^{1-\frac{\alpha+1}{\alpha}\epsilon} + \frac{1}{\alpha} n^{2-\frac{\alpha+1}{\alpha}\epsilon} \\ - \left(1 + \frac{1}{\alpha}\right) n^{1+\frac{1}{\alpha+1}-\frac{1}{\alpha}\epsilon} - n^2 - n^{\frac{2\alpha}{\alpha+1}-2\epsilon} + 2n^{1+\frac{\alpha}{\alpha+1}-\epsilon},$$

is  $-n^2$ , and hence, for sufficiently large  $n$  we have  $g\left(n^{\frac{\alpha}{\alpha+1}-\epsilon}\right) < 0$ . Similarly, the dominant term of  $g\left(n^{\frac{\alpha}{\alpha+1}+\epsilon}\right)$  is  $\frac{1}{\alpha}n^{2+\frac{\alpha+1}{\alpha}\epsilon}$ , and for sufficiently large  $n$  we have  $g\left(n^{\frac{\alpha}{\alpha+1}+\epsilon}\right) > 0$ . The claim then follows. Using the above bound, for sufficiently large  $n$ , we also get that  $\psi_n \geq n^{\frac{\alpha}{\alpha+1}-\epsilon}$ . We now provide a bound for the denominator of  $f'(\psi_n; \alpha, n)$ . In particular, for sufficiently large  $n$ , we get that for  $x \leq n^{\frac{\alpha}{\alpha+1}+\epsilon}$ ,

$$\frac{d}{dx} \left(x^{1+\frac{1}{\alpha}} + nx - x^2\right) = \left(1 + \frac{1}{\alpha}\right) x^{\frac{1}{\alpha}} + n - 2x > 0,$$

which shows that the denominator is strictly increasing. Hence, using the lower bound on  $\psi_n$ ,

$$\frac{1}{\left(\psi_n^{1+\frac{1}{\alpha}} + n\psi_n - \psi_n^2\right)^2} \leq \frac{1}{\left(n^{(\frac{\alpha}{\alpha+1}-\epsilon)(1+\frac{1}{\alpha})} + n^{1+\frac{\alpha}{\alpha+1}-\epsilon} - n^{\frac{2\alpha}{\alpha+1}-2\epsilon}\right)^2} \\ \leq \frac{n^{-2-\frac{2\alpha}{\alpha+1}+2\epsilon}}{\left(n^{-\frac{\alpha}{\alpha+1}-\frac{1}{\alpha}\epsilon} + 1 - n^{-\frac{1}{\alpha+1}}\right)^2} = O\left(n^{-2-\frac{2\alpha}{\alpha+1}+2\epsilon}\right).$$

We now provide a bound for the numerator. Since  $g$  is strictly increasing and  $\xi^*$  is a root, we get

$$|g(\psi_n)| \leq |g(\theta_n)| \\ = \left| \left(1 - \frac{1}{\alpha}\right) \alpha^{\frac{2\alpha+1}{\alpha+1}} n^{-\frac{1}{\alpha+1}+2} + n - \left(1 + \frac{1}{\alpha}\right) \alpha^{\frac{1}{\alpha+1}} n^{-\frac{\alpha}{\alpha+1}+2} - \alpha^{\frac{2\alpha}{\alpha+1}} n^{-\frac{2}{\alpha+1}+2} + 2\alpha^{\frac{\alpha}{\alpha+1}} n^{-\frac{1}{\alpha+1}+2} \right| \\ = O\left(n^{-\frac{\min\{1,\alpha\}}{\alpha+1}+2}\right).$$

If we combine the above results, we get  $f'(\psi_n; \alpha, n) = O\left(n^{-\frac{\min\{1,\alpha\}+2\alpha}{\alpha+1}+2\epsilon}\right)$ .

(II.) Follows from (32).

(III.) We have

$$f(\theta_n; \alpha, n) = \frac{n + n - n^{\frac{\alpha}{\alpha+1}}}{n + n^{1+\frac{\alpha}{\alpha+1}} - n^{\frac{2\alpha}{\alpha+1}}} \\ = \frac{n\left(2 - n^{-\frac{1}{\alpha+1}}\right)}{n^{1+\frac{\alpha}{\alpha+1}}\left(n^{-\frac{\alpha}{\alpha+1}} + 1 - n^{-\frac{1}{\alpha+1}}\right)} = \Theta\left(n^{-\frac{\alpha}{\alpha+1}}\right). \quad \blacksquare$$

**Proposition 4.** *There exists a point  $z \in \mathbf{R}^n$  that attains the minimum of (29), for which*

$$z_1 = \dots = z_{n-1} = 1.$$

**Proof.** For  $\alpha = 1$ , problem (29) is written as

$$\begin{aligned} & \text{minimize} && \frac{1}{n} \left( \frac{z_n}{z_1} + \frac{z_n}{z_2} + \dots + \frac{z_n}{z_{n-1}} + 1 \right) \\ & \text{subject to} && \frac{1}{n} \leq z_n \leq z_{n-1} \leq \dots \leq z_1 \leq 1. \end{aligned}$$

If  $z$  is an optimal solution of the above, then clearly  $z_1 = \dots = z_{n-1} = 1$ .

We now deal with the case of  $\alpha \neq 1$ . We first show that if  $z$  is an optimal solution of (29), then  $z_1 = \dots = z_{n-1}$ . We analyze the cases  $0 < \alpha < 1$  and  $\alpha > 1$  separately.

For  $0 < \alpha < 1$ , the function  $z_1^{1-\alpha} + \dots + z_{n-1}^{1-\alpha}$  is strictly concave, and the function  $z_1^{-\alpha} + \dots + z_{n-1}^{-\alpha}$  is strictly convex. If  $z$  is an optimal solution of (29) for which  $z_1 = \dots = z_{n-1}$  is violated, we construct a point  $\bar{z} \in \mathbf{R}^n$ , such that its first  $n - 1$  components are all equal to the mean of  $z_1, \dots, z_{n-1}$  and  $\bar{z}_n = z_n$ . We show that  $\bar{z}$  is feasible for (29) and it achieves a strictly lower objective value compared to  $z$ , a contradiction. Note that by strict concavity/ convexity we get

$$\bar{z}_1^{1-\alpha} + \dots + \bar{z}_{n-1}^{1-\alpha} > z_1^{1-\alpha} + \dots + z_{n-1}^{1-\alpha},$$

and

$$\bar{z}_1^{-\alpha} + \dots + \bar{z}_{n-1}^{-\alpha} < z_1^{-\alpha} + \dots + z_{n-1}^{-\alpha},$$

respectively. For feasibility,  $0 \leq \bar{z}_n \leq \dots \leq \bar{z}_1 \leq 1$  is immediate and

$$\bar{z}_n^{-\alpha} = z_n^{-\alpha} \leq z_1^{1-\alpha} + \dots + z_{n-1}^{1-\alpha} + z_n^{1-\alpha} < \bar{z}_1^{1-\alpha} + \dots + \bar{z}_{n-1}^{1-\alpha} + \bar{z}_n^{1-\alpha}.$$

Finally, compared to  $z$ , if we evaluate the objective of (29) at  $\bar{z}$ , the numerator strictly decreases and the denominator strictly increases, hence the objective value strictly decreases.

For  $\alpha > 1$ , let  $z$  be an optimal solution of (29) for which  $z_{j+1} < z_j$  for some  $j = 1, \dots, n - 2$ . We similarly construct a feasible point  $\bar{z}$  for (29) that achieves a strictly lower objective value than  $z$ . Let  $\bar{z}_i = z_i$  for all  $i \neq j, j + 1$ ,  $\bar{z}_j = z_j - \epsilon$  and  $\bar{z}_{j+1} = z_{j+1} + \delta\epsilon$ , where  $\epsilon > 0$  and

$$\delta = \frac{z_j^{-\alpha} - \mu}{z_{j+1}^{-\alpha}}, \quad \mu \in \left( 0, z_j^{-\alpha} \left( \frac{z_j - z_{j+1}}{z_j} \right) \right).$$



For small enough  $\epsilon$ ,  $0 \leq \bar{z}_n \leq \dots \leq \bar{z}_1 \leq 1$  is immediate. Using a first order Taylor series expansion,

$$\begin{aligned}\bar{z}_j^{1-\alpha} + \bar{z}_{j+1}^{1-\alpha} &= z_j^{1-\alpha} + z_{j+1}^{1-\alpha} + (z_j^{-\alpha} - \delta z_{j+1}^{-\alpha})(\alpha - 1)\epsilon + O(\epsilon^2) \\ &> z_j^{1-\alpha} + z_{j+1}^{1-\alpha}\end{aligned}$$

for small enough  $\epsilon$ , since  $z_j^{-\alpha} > \delta z_{j+1}^{-\alpha} \Leftrightarrow \mu > 0$ . As a result,

$$\bar{z}_1^{1-\alpha} + \dots + \bar{z}_{n-1}^{1-\alpha} + \bar{z}_n^{1-\alpha} > z_1^{1-\alpha} + \dots + z_{n-1}^{1-\alpha} + z_n^{1-\alpha},$$

and  $\bar{z}$  is feasible. Moreover, the denominator of the objective strictly increases. Thus it suffices to show that the numerator decreases. To this end, we have

$$\begin{aligned}\bar{z}_j^{-\alpha} + \bar{z}_{j+1}^{-\alpha} &= z_j^{-\alpha} + z_{j+1}^{-\alpha} + (z_j^{-\alpha-1} - \delta z_{j+1}^{-\alpha-1})\alpha\epsilon + O(\epsilon^2) \\ &< z_j^{-\alpha} + z_{j+1}^{-\alpha}\end{aligned}$$

for small enough  $\epsilon$ , since  $z_j^{-\alpha-1} < \delta z_{j+1}^{-\alpha-1} \Leftrightarrow \mu < z_j^{-\alpha} \left( \frac{z_j - z_{j+1}}{z_j} \right)$ .

Since for every optimal solution of (29), we have  $z_1 = \dots = z_{n-1}$ , problem (29) can be written equivalently as

$$\begin{aligned}\text{minimize} \quad & g(z_1, z_2) = \frac{(n-1)z_1^{-\alpha}z_2 + z_2^{1-\alpha}}{(n-1)z_1^{1-\alpha} + z_2^{1-\alpha}} \\ \text{subject to} \quad & 0 \leq z_2 \leq z_1 \leq 1 \\ & z_2^{-\alpha} \leq (n-1)z_1^{1-\alpha} + z_2^{1-\alpha}.\end{aligned}\tag{33}$$

It suffices to show that there exists an optimal solution  $z$  of (33) for which  $z_1 = 1$ .

Let  $z$  be an optimal solution of (33).

If  $0 < \alpha < 1$ , assume that  $z_1 < 1$ . Then, increase  $z_1$  by a small enough amount such that it remains less than 1. The quantity  $z_1^{1-\alpha}$  increases, so the new point we get is feasible. Also, the quantity  $z_1^{-\alpha}$  decreases. Hence, the new point is feasible and achieves a strictly lower objective value, a contradiction.

If  $\alpha > 1$ , the point  $z$  lies on the boundary of the feasible set or is a stationary point of the objective. Suppose that  $z$  is not a stationary point, *i.e.*,  $\nabla g(z_1, z_2) \neq 0$ . If  $z_1 = z_2$ , the objective evaluates to 1 for any such  $z$ , so we can assume  $z_1 = 1$ . We next rule out the possibility of  $z$  lying on the  $z_2^{-\alpha} = (n-1)z_1^{1-\alpha} + z_2^{1-\alpha}$  boundary with  $z_1 < 1$ . Suppose that it does. We will demonstrate that we can always find a feasible direction along which the objective decreases. We have

$$\begin{aligned}\frac{\partial g}{\partial z_1} &= \frac{(n-1)z_1^{-\alpha}z_2}{\left((n-1)z_1^{1-\alpha} + z_2^{1-\alpha}\right)^2} \left( -(n-1)z_1^{-\alpha} - \alpha z_1^{-1}z_2^{1-\alpha} + (\alpha-1)z_2^{-\alpha} \right), \\ \frac{\partial g}{\partial z_2} &= -\frac{z_1}{z_2} \frac{\partial g}{\partial z_1}.\end{aligned}$$

Note that we assumed that  $\nabla g(z) \neq 0$ , hence  $\frac{\partial g}{\partial z_1}(z) \neq 0$ . Suppose that  $\frac{\partial g}{\partial z_1}(z) > 0$ . Then,  $(1, \delta)$  is a direction along which the objective decreases, for large enough  $\delta > 0$ , since

$$\frac{\partial g}{\partial z_1}(z) + \delta \frac{\partial g}{\partial z_2}(z) = \frac{\partial g}{\partial z_1}(z) \left(1 - \delta \frac{z_1}{z_2}\right) < 0.$$

It is also a feasible direction, since for  $\epsilon > 0$  small enough,  $0 \leq z_2 + \delta\epsilon \leq z_1 + \epsilon \leq 1$ , and is also a direction along which  $(n-1)z_1^{1-\alpha} + z_2^{1-\alpha} + z_2^{-\alpha}$  increases, since

$$\begin{aligned} (n-1)z_1^{-\alpha} + \delta \left( (1-\alpha)z_2^{-\alpha} + \alpha z_2^{-\alpha-1} \right) &= (1-\alpha)(z_2^{-\alpha} - z_2^{1-\alpha}) + \delta \left( (1-\alpha)z_2^{-\alpha} + \alpha z_2^{-\alpha-1} \right) \\ &= z_2^{-\alpha} \left( (1-\alpha)(1-z_2) + \delta \left( \frac{\alpha}{z_2} - (\alpha-1) \right) \right) > 0 \end{aligned}$$

for large enough  $\delta$ . Similarly, if  $\frac{\partial g}{\partial z_1}(z) < 0$ , one can show that  $(1, \delta)$  is again a feasible direction along which the objective decreases, for

$$\frac{(\alpha-1)(1-z_2)z_2}{\alpha - (\alpha-1)z_2} < \delta < \frac{z_2}{z_1},$$

if one can select such  $\delta$ . Otherwise, one can show that  $(-1, -\delta)$  is a feasible direction along which the objective decreases, for

$$\frac{z_2}{z_1} < \delta < \frac{(\alpha-1)(1-z_2)z_2}{\alpha - (\alpha-1)z_2}.$$

We have thus established that if  $z$  is not a stationary point, then there also exists an optimal solution for which  $z_1 = 1$ . We next show that the same holds true if  $z$  is a stationary point.

Suppose that  $z$  is a stationary point, *i.e.*,  $\nabla g(z_1, z_2) = 0$ . Then, we have

$$(n-1)z_1^{1-\alpha} + \alpha z_2^{1-\alpha} - (\alpha-1)z_1 z_2^{-\alpha} = 0.$$

Using the above, the objective evaluates to

$$g(z_1, z_2) = \frac{\alpha}{\alpha-1} \frac{z_2}{z_1}.$$

Moreover, if  $z_1 = \lambda z_2$  for some  $\lambda \geq 1$ , the stationarity condition yields

$$(n-1)\lambda^{1-\alpha} - (\alpha-1)\lambda + \alpha = 0,$$

an equation that has a unique solution in  $[1, \infty)$ . Let  $\bar{\lambda}$  be the solution. Then, the problem (33)

constrained on the stationary points of its objective can be expressed as

$$\begin{aligned} & \text{minimize} && \frac{\alpha}{\alpha-1} \frac{z_2}{z_1} \\ & \text{subject to} && z_1 = \bar{\lambda} z_2, \quad z_1 \leq 1 \\ & && z_2^{-\alpha} \leq (n-1)z_1^{1-\alpha} + z_2^{1-\alpha}, \end{aligned}$$

or, equivalently,

$$\begin{aligned} & \text{minimize} && \frac{\alpha}{\alpha-1} \frac{1}{\bar{\lambda}} \\ & \text{subject to} && z_1 = \bar{\lambda} z_2 \\ & && \frac{1}{(\alpha-1)(\bar{\lambda}-1)} \leq z_2 \leq \frac{1}{\bar{\lambda}}. \end{aligned}$$

In case the above problem is feasible, we pick  $z_2 = \frac{1}{\bar{\lambda}}$ , and  $z_1 = 1$  and the proof is complete.  $\blacksquare$

**Proposition 5.** *Consider a resource allocation problem with  $n$  players,  $n \geq 2$ . Let the utility set, denoted by  $U \subset \mathbf{R}^n$ , be compact and convex. If the players have equal maximum achievable utilities (greater than zero),*

$$\text{POF}(U; 1) \leq 1 - \frac{2\sqrt{n} - 1}{n}. \quad (\text{price of proportional fairness})$$

Let  $\{\alpha_k \in \mathbf{R} \mid k \in \mathbf{N}\}$  be a sequence such that  $\alpha_k \rightarrow \infty$  and  $\alpha_k \geq 1, \forall k$ . Then,

$$\limsup_{k \rightarrow \infty} \text{POF}(U; \alpha_k) \leq 1 - \frac{4n}{(n+1)^2}. \quad (\text{price of max-min fairness})$$

**Proof.** Let  $f$  be defined as in Proposition 3. Using Theorem 1 for  $\alpha = 1$  we get

$$\begin{aligned} \text{POF}(U; 1) &\leq 1 - \min_{x \in [1, n]} f(x; 1, n) \\ &= 1 - \min_{x \in [1, n]} \frac{x^2 + n - x}{nx} \\ &= 1 - \frac{2\sqrt{n} - 1}{n}. \end{aligned}$$

Similarly, for any  $k \in \mathbf{N}$  and  $\alpha = \alpha_k$

$$\text{POF}(U; \alpha_k) \leq 1 - \min_{x \in [1, n]} f(x; \alpha_k, n),$$

which implies that

$$\begin{aligned} \limsup_{k \rightarrow \infty} \text{POF}(U; \alpha_k) &\leq \limsup_{k \rightarrow \infty} \left( 1 - \min_{x \in [1, n]} f(x; \alpha_k, n) \right) \\ &\leq 1 - \liminf_{k \rightarrow \infty} \min_{x \in [1, n]} f(x; \alpha_k, n). \end{aligned} \tag{34}$$

Consider the set of (real-valued) functions  $\{f(\cdot; \alpha_k, n) \mid k \in \mathbf{N}\}$  defined over the compact set  $[1, n]$ . We show that the set is equicontinuous, and that the closure of the set  $\{f(x; \alpha_k, n) \mid k \in \mathbf{N}\}$  is bounded for any  $x \in [1, n]$ . Boundedness follows since  $0 \leq f(x; \alpha, n) \leq 1$  for any  $\alpha > 0$  and  $x \in [1, n]$ . The set of functions  $\{f(\cdot; \alpha_k, n) \mid k \in \mathbf{N}\}$  shares the same Lipschitz constant, as for any  $k \in \mathbf{N}$ ,  $\alpha_k \geq 1$  and  $x \in [1, n]$  we have

$$\begin{aligned}
|f'(x; \alpha_k, n)| &= \left| \frac{\left(1 - \frac{1}{\alpha_k}\right) x^{2+\frac{1}{\alpha_k}} + \frac{n+1}{\alpha_k} x^{1+\frac{1}{\alpha_k}} - n \left(1 + \frac{1}{\alpha_k}\right) x^{\frac{1}{\alpha_k}} - (x-n)^2}{\left(x^{1+\frac{1}{\alpha_k}} + (n-x)x\right)^2} \right| \\
&\leq \left| \left(1 - \frac{1}{\alpha_k}\right) x^{2+\frac{1}{\alpha_k}} + \frac{n+1}{\alpha_k} x^{1+\frac{1}{\alpha_k}} - n \left(1 + \frac{1}{\alpha_k}\right) x^{\frac{1}{\alpha_k}} - (x-n)^2 \right| \\
&\leq \left(1 - \frac{1}{\alpha_k}\right) x^{2+\frac{1}{\alpha_k}} + \frac{n+1}{\alpha_k} x^{1+\frac{1}{\alpha_k}} + n \left(1 + \frac{1}{\alpha_k}\right) x^{\frac{1}{\alpha_k}} + (x-n)^2 \\
&\leq n^3 + (n+1)n^2 + 2n^2 + n^2 = 2(n^3 + 2n^2).
\end{aligned}$$

As a result, the set of functions  $\{f(\cdot; \alpha_k, n) \mid k \in \mathbf{N}\}$  is equicontinuous.

Using the above result,

$$\lim_{k \rightarrow \infty} \min_{x \in [1, n]} f(x; \alpha_k, n) = \min_{x \in [1, n]} \lim_{k \rightarrow \infty} f(x; \alpha_k, n).$$

Thus, (34) yields

$$\begin{aligned}
\limsup_{k \rightarrow \infty} \text{POF}(U; \alpha_k) &\leq 1 - \liminf_{k \rightarrow \infty} \min_{x \in [1, n]} f(x; \alpha_k, n) \\
&= 1 - \min_{x \in [1, n]} \lim_{k \rightarrow \infty} f(x; \alpha_k, n) \\
&= 1 - \min_{x \in [1, n]} \lim_{k \rightarrow \infty} \frac{x^{1+\frac{1}{\alpha_k}} + n - x}{x^{1+\frac{1}{\alpha_k}} + (n-x)x} \\
&= 1 - \min_{x \in [1, n]} \frac{n}{x + (n-x)x} \\
&= 1 - \frac{4n}{(n+1)^2}. \quad \blacksquare
\end{aligned}$$

## D. A Model for Air Traffic Flow Management

The following is a model for air traffic flow management due to Bertsimas and Stock-Patterson (1998). Consider a set of flights,  $\mathcal{F} = \{1, \dots, F\}$ , that are operated by airlines over a (discretized) time period in a network of airports, utilizing a capacitated airspace that is divided into sectors. Let  $\mathcal{F}_a \subset \mathcal{F}$  be the set of flights operated by airline  $a \in \mathcal{A}$ , where  $\mathcal{A} = \{1, \dots, A\}$  is the set of airlines. Similarly,  $\mathcal{T} = \{1, \dots, T\}$  is the set of time steps,  $\mathcal{K} = \{1, \dots, K\}$  the set of airports,

and  $\mathcal{J} = \{1, \dots, J\}$  the set of sectors. Flights that are continued are included in a set of pairs,  $\mathcal{C} = \{(f', f) : f' \text{ is continued by flight } f\}$ . The model input data, the main decision variables, and a description of the feasibility set are described below:

**Data.**

$$\begin{aligned}
N_f &= \text{number of sectors in flight } f\text{'s path,} \\
P(f, i) &= \begin{cases} \text{the departure airport, if } i = 1, \\ \text{the } (i - 1)\text{th sector in flight } f\text{'s path, if } 1 < i < N_f, \\ \text{the arrival airport, if } i = N_f, \end{cases} \\
P_f &= (P(f, i) : 1 \leq i \leq N_f), \\
D_k(t) &= \text{departure capacity of airport } k \text{ at time } t, \\
A_k(t) &= \text{arrival capacity of airport } k \text{ at time } t, \\
S_j(t) &= \text{capacity of sector } j \text{ at time } t, \\
d_f &= \text{scheduled departure time of flight } f, \\
r_f &= \text{scheduled arrival time of flight } f, \\
s_f &= \text{turnaround time of an airplane after flight } f, \\
l_{fj} &= \text{number of time steps that flight } f \text{ must spend in sector } j, \\
T_f^j &= \text{set of feasible time steps for flight } f \text{ to arrive to sector } j = \{\underline{T}_f^j, \dots, \overline{T}_f^j\}, \\
\underline{T}_f^j &= \text{first time step in the set } T_f^j, \text{ and} \\
\overline{T}_f^j &= \text{last time step in the set } T_f^j.
\end{aligned}$$

**Decision Variables.**

$$w_{ft}^j = \begin{cases} 1, & \text{if flight } f \text{ arrives at sector } j \text{ by time step } t, \\ 0, & \text{otherwise.} \end{cases}$$

**Feasibility Set.** The variable  $w$  is feasible if it satisfies the constraints:

$$\begin{aligned}
\sum_{f:P(f,1)=k} (w_{ft}^k - w_{f,t-1}^k) &\leq D_k(t) \quad \forall k \in \mathcal{K}, t \in \mathcal{T}, \\
\sum_{f:P(f,N_f)=k} (w_{ft}^k - w_{f,t-1}^k) &\leq A_k(t) \quad \forall k \in \mathcal{K}, t \in \mathcal{T}, \\
\sum_{f:P(f,i)=j, P(f,i+1)=j', i < N_f} (w_{ft}^j - w_{ft}^{j'}) &\leq S_j(t) \quad \forall j \in \mathcal{J}, t \in \mathcal{T}, \\
w_{f,t+l_{fj}}^{j'} - w_{ft}^j &\leq 0 \quad \forall f \in \mathcal{F}, t \in T_f^j, j = P(f, i), j' = P(f, i + 1), i < N_f, \\
w_{ft}^k - w_{f,t-s_f}^k &\leq 0 \quad \forall (f', f) \in \mathcal{C}, t \in T_f^k, k = P(f, i) = P(f', N_f), \\
w_{ft}^j - w_{f,t-1}^j &\geq 0 \quad \forall f \in \mathcal{F}, j \in P_f, t \in T_f^j, \\
w_{ft}^j &\in \{0, 1\} \quad \forall f \in \mathcal{F}, j \in P_f, t \in T_f^j.
\end{aligned}$$

The constraints correspond to capacity constraints for airports and sectors, connectivity between sectors and airports, and connectivity in time (for more details, see Bertsimas and Stock-Patterson (1998)).