

# The $\delta$ -Bounded $\varepsilon$ -Non-Dominated Multi-Objective User Equilibrium ( $\delta$ -EBR-MUE): Model, Theory and Approximation

Your Name

Abstract

This paper introduces a novel multi-objective user equilibrium model under bounded rationality, termed the  $\delta$ -EBR-MUE, which integrates two rationality constraints: (i) weakly  $\varepsilon$ -non-dominated path choice, and (ii) travel time boundedness relative to the shortest path. We provide a complete mathematical formulation, establish existence via a set-valued fixed-point argument, and propose a support function approximation for the non-dominated frontier. Numerical experiments validate the behavioral implications and demonstrate the diversity of solutions beyond classical DUE or BRUE models.

## 1 Notation and Preliminaries

Let  $\mathcal{W}$  denote the set of origin-destination (OD) pairs. For each OD pair  $w \in \mathcal{W}$ , let  $\mathcal{P}^w$  denote the set of all feasible paths, and  $d^w$  the total travel demand.

Let  $\mathbf{f} = (f_p^w)_{p \in \mathcal{P}^w, w \in \mathcal{W}}$  denote the path flow vector. The feasible flow set is defined as:

$$\Lambda := \left\{ \mathbf{f} \in \mathbb{R}_+^{|\mathcal{P}|} : \sum_{p \in \mathcal{P}^w} f_p^w = d^w, \forall w \in \mathcal{W} \right\}.$$

Each path  $p$  is associated with a bi-objective cost vector:

$$\mathbf{C}_p(\mathbf{f}) := (t_p(\mathbf{f}), c_p(\mathbf{f})),$$

where  $t_p(\mathbf{f})$  and  $c_p(\mathbf{f})$  represent travel time and monetary cost under flow  $\mathbf{f}$ .

### Dominance and $\varepsilon$ -Non-Dominance

**Definition 1.1 (Strict Dominance).** Given two paths  $p$  and  $q$  for OD pair  $w$ , path  $q$  strictly dominates path  $p$ , denoted

$$\mathbf{C}_q(\mathbf{f}) \prec \mathbf{C}_p(\mathbf{f}),$$

if and only if:

$$t_q(\mathbf{f}) < t_p(\mathbf{f}) \quad \text{and} \quad c_q(\mathbf{f}) < c_p(\mathbf{f}).$$

**Definition 1.2 ( $\varepsilon$ -Strict Dominance).** For tolerance vector  $\varepsilon = (\varepsilon_t, \varepsilon_m)$ , path  $q$   $\varepsilon$ -strictly dominates path  $p$  if:

$$t_q(\mathbf{f}) + \varepsilon_t < t_p(\mathbf{f}) \quad \text{and} \quad c_q(\mathbf{f}) + \varepsilon_m < c_p(\mathbf{f}).$$

Definition 1.3 ( $\epsilon$ -Weakly Non-Dominated Path). Path  $p$  is  $\epsilon$ -weakly non-dominated if it is not  $\epsilon$ -strictly dominated by any  $q \in \mathcal{P}^w$ , i.e.,

$$\phi_p^\epsilon(\mathbf{f}) := \max_{q \in \mathcal{P}^w} \max \{t_q - t_p + \epsilon_t, c_q - c_p + \epsilon_m\} \leq 0.$$

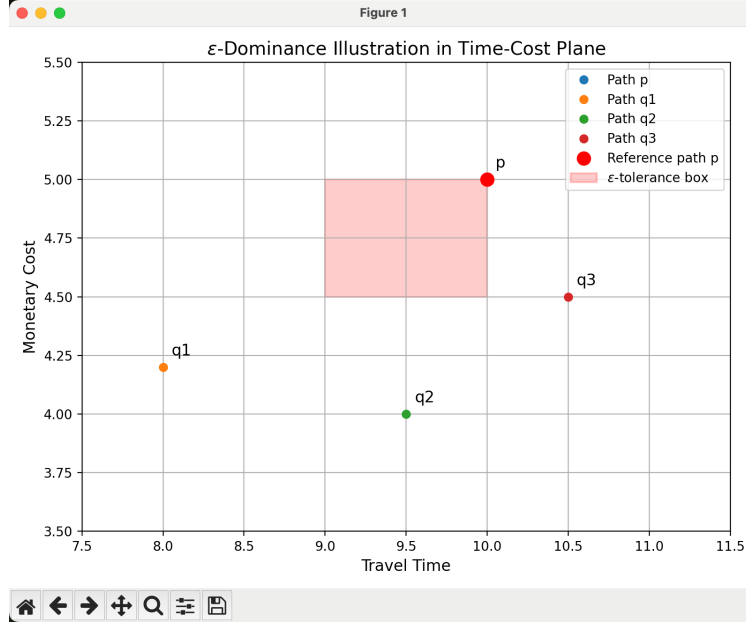


Figure 1: Illustration of  $\epsilon$ -dominance in the time-cost objective space. The red box indicates the  $\epsilon$ -tolerance area around the reference path  $p$ . Paths inside this box are not  $\epsilon$ -strictly dominating  $p$ . Path  $q_1$  strictly dominates  $p$ ,  $q_2$   $\epsilon$ -dominates  $p$ , and  $q_3$  does not dominate  $p$ .

#### Example: $\epsilon$ -Nondominated Flow Region Between Two Paths

Consider two alternative paths  $p$  and  $q$  connecting the same origin-destination pair. Let the scalar flow on path  $p$  be denoted as  $f_p$ , and on path  $q$  as  $f_q$ . The bi-objective cost vectors for these two paths are given by:

$$\mathbf{C}_p(f) = \begin{pmatrix} C_p^1(f) \\ C_p^2(f) \end{pmatrix} = \begin{pmatrix} 2 + f_p \\ 4 + f_p^2 \end{pmatrix},$$

$$\mathbf{C}_q(f) = \begin{pmatrix} C_q^1(f) \\ C_q^2(f) \end{pmatrix} = \begin{pmatrix} 1 + 2f_q \\ 5 + f_q \end{pmatrix}.$$

We consider an  $\epsilon$ -dominance relation with tolerance vector  $\epsilon = (\epsilon^1, \epsilon^2) = (0.5, 0.5)$ . The cost vector  $\mathbf{C}_q(f)$  is said to  $\epsilon$ -dominate  $\mathbf{C}_p(f)$  if:

$$C_q^1(f) \leq C_p^1(f) - \epsilon^1 \quad \text{and} \quad C_q^2(f) \leq C_p^2(f) - \epsilon^2.$$

Therefore, the set of flow pairs  $(f_p, f_q) \in \mathbb{R}_+^2$  under which path  $p$  is not  $\epsilon$ -dominated by path  $q$  is:

$$\mathcal{F}_\varepsilon = \{(f_p, f_q) \in \mathbb{R}_+^2 \mid C_q^1(f) > C_p^1(f) - \varepsilon^1 \quad \text{or} \quad C_q^2(f) > C_p^2(f) - \varepsilon^2\}.$$

This region can be visualised in the  $(f_p, f_q)$ -space as the set where the  $\varepsilon$ -dominance condition does not hold. Due to the nonlinear nature of the cost functions, the set  $\mathcal{F}_\varepsilon$  is generally non-convex.

## 2 Mathematical Formulation of the $\delta$ -EBR-MUE Model

Let  $\mathcal{W}$  be the set of origin-destination (OD) pairs, and  $\mathcal{P}^w$  the set of feasible paths for  $w \in \mathcal{W}$ . For each path  $p \in \mathcal{P}^w$ :

- $f_p^w$ : flow on path  $p$ ;
- $\mathbf{C}_p = (t_p(\mathbf{f}), c_p(\mathbf{f}))$ : travel time and cost;
- $T_w^{\text{UE}}(\mathbf{f}) := \min_{p \in \mathcal{P}^w} t_p(\mathbf{f})$ .

A path  $p$  is said to be  $\varepsilon$ -non-dominated and  $\delta$ -bounded if:

- (i)  $\nexists q \in \mathcal{P}^w : \mathbf{C}_q(\mathbf{f}) \prec \mathbf{C}_p(\mathbf{f}) - \varepsilon$ ,
- (ii)  $t_p(\mathbf{f}) \leq T_w^{\text{UE}}(\mathbf{f}) + \delta$ .

Let  $\phi_p^\varepsilon(\mathbf{f}) := \max_{q \in \mathcal{P}^w} \max \{t_q - t_p + \varepsilon_t, c_q - c_p + \varepsilon_m\}$  and  $\phi_p^\delta(\mathbf{f}) := t_p(\mathbf{f}) - T_w^{\text{UE}}(\mathbf{f}) - \delta$ . The  $\delta$ -EBR-MUE conditions are:

$$\begin{aligned} f_p^w \cdot \phi_p^\varepsilon(\mathbf{f}) &= 0, & \phi_p^\varepsilon(\mathbf{f}) &\geq 0, \\ f_p^w \cdot \phi_p^\delta(\mathbf{f}) &= 0, & \phi_p^\delta(\mathbf{f}) &\leq 0, \\ \sum_{p \in \mathcal{P}^w} f_p^w &= d^w, & f_p^w &\geq 0. \end{aligned}$$

Interpretation of  $\phi_p^\varepsilon(\mathbf{f})$ . The expression  $\phi_p^\varepsilon(\mathbf{f}) := \max_{q \in \mathcal{P}^w} \max \{t_q - t_p + \varepsilon_t, c_q - c_p + \varepsilon_m\}$  quantifies the worst-case  $\varepsilon$ -dominance violation for path  $p$ .

- The inner max evaluates whether any path  $q$  offers a sufficient improvement in either time or cost to make path  $p$  strictly dominated beyond the tolerance  $\varepsilon$ .
- The outer max scans all such potential dominating paths  $q$ .

If  $\phi_p^\varepsilon(\mathbf{f}) > 0$ , it means there exists a path  $q$  that  $\varepsilon$ -dominates  $p$  in at least one dimension. Therefore, path  $p$  is considered infeasible under the  $\delta$ -EBR-MUE behavioral rule. Conversely,  $\phi_p^\varepsilon(\mathbf{f}) = 0$  implies that  $p$  is weakly  $\varepsilon$ -non-dominated.

## 3 Existence of $\delta$ -EBR-MUE

Let the feasible set be:

$$\Lambda := \left\{ \mathbf{f} \in \mathbb{R}_+^{|\mathcal{P}|} : \sum_{p \in \mathcal{P}^w} f_p^w = d^w, \forall w \in \mathcal{W} \right\}.$$

Define a set-valued map:

$$\Phi(\mathbf{f}) := \left\{ \tilde{\mathbf{f}} \in \Lambda : \tilde{f}_p^w > 0 \Rightarrow p \in \mathcal{A}_w^{\varepsilon, \delta}(\mathbf{f}) \right\},$$

where  $\mathcal{A}_w^{\varepsilon, \delta}(\mathbf{f})$  is the set of all paths satisfying the two criteria above. Under continuity of  $\mathbf{C}_p$ , compactness of  $\Lambda$ , and closedness of  $\Phi$ , Kakutani's fixed-point theorem ensures existence of an equilibrium  $\mathbf{f}^* \in \Phi(\mathbf{f}^*)$ .

## 4 Existence of $\delta$ -EBR-MUE

We now establish the existence of a flow vector  $\mathbf{f}^* \in \Lambda$  satisfying the conditions of the  $\delta$ -EBR-MUE model.

Theorem 4.1 (Existence of  $\delta$ -EBR-MUE). Suppose that:

1. The feasible set of flows  $\Lambda := \left\{ \mathbf{f} \in \mathbb{R}_+^{|\mathcal{P}|} : \sum_{p \in \mathcal{P}^w} f_p^w = d^w, \forall w \in \mathcal{W} \right\}$  is nonempty, compact, and convex;
2. The path cost functions  $t_p(\mathbf{f})$ ,  $c_p(\mathbf{f})$  are continuous with respect to  $\mathbf{f}$ ;
3. The parameters  $\varepsilon_t, \varepsilon_m, \delta \geq 0$  are fixed;

Then there exists a flow vector  $\mathbf{f}^* \in \Lambda$  satisfying the following conditions for all OD pairs  $w \in \mathcal{W}$  and all  $p \in \mathcal{P}^w$ :

$$f_p^{w*} > 0 \Rightarrow \begin{cases} \phi_p^\varepsilon(\mathbf{f}^*) = 0, & (\varepsilon\text{-non-dominance}) \\ t_p(\mathbf{f}^*) \leq T_w^{\text{UE}}(\mathbf{f}^*) + \delta, & (\text{time-bounded}) \end{cases}$$

Proof. We reformulate the equilibrium as a fixed point of a set-valued map.

Define the feasible flow set:

$$\Lambda = \left\{ \mathbf{f} \in \mathbb{R}_+^{|\mathcal{P}|} : \sum_{p \in \mathcal{P}^w} f_p^w = d^w, \forall w \in \mathcal{W} \right\},$$

which is nonempty, convex, and compact.

Define for each OD pair  $w$  the set of acceptable paths under flow  $\mathbf{f}$ :

$$\mathcal{A}_w^{\varepsilon, \delta}(\mathbf{f}) := \left\{ p \in \mathcal{P}^w : \phi_p^\varepsilon(\mathbf{f}) = 0, t_p(\mathbf{f}) \leq T_w^{\text{UE}}(\mathbf{f}) + \delta \right\}.$$

Then define the set-valued mapping:

$$\Phi(\mathbf{f}) := \left\{ \tilde{\mathbf{f}} \in \Lambda : \tilde{f}_p^w > 0 \Rightarrow p \in \mathcal{A}_w^{\varepsilon, \delta}(\mathbf{f}), \forall w \in \mathcal{W} \right\}.$$

We claim that this map  $\Phi : \Lambda \rightrightarrows \Lambda$  satisfies the conditions of Kakutani's fixed-point theorem:

- The domain  $\Lambda$  is compact and convex (from assumption);
- For each  $\mathbf{f}$ , the image  $\Phi(\mathbf{f})$  is convex: since selecting flows on acceptable paths preserves linearity and the feasible set is convex;
- The graph of  $\Phi$  is closed: because both  $\phi_p^\varepsilon(\mathbf{f})$  and  $t_p(\mathbf{f})$  are continuous in  $\mathbf{f}$ , the acceptance set  $\mathcal{A}_w^{\varepsilon, \delta}(\mathbf{f})$  varies upper hemicontinuously with  $\mathbf{f}$ .

By Kakutani's theorem, there exists  $\mathbf{f}^* \in \Lambda$  such that  $\mathbf{f}^* \in \Phi(\mathbf{f}^*)$ , i.e., a  $\delta$ -EBR-MUE exists.  $\square$

## Algorithm: Iterative Solution for $\delta$ -Bounded Rational Multi-Objective Non-Strictly Dominated User Equilibrium ( $\delta$ -EBR-MUE)

Input:

- OD demand set  $\{d_w\}_{w \in W}$
- Path sets  $\{\mathcal{P}_w\}_{w \in W}$  for each OD pair
- Time cost tolerance parameter  $\delta > 0$
- Path cost function  $\mathbf{C}_p(\mathbf{f})$ , including time and other objectives
- Initial feasible path flow  $\mathbf{f}^{(0)}$  satisfying demand conservation
- Maximum number of iterations  $K_{\max}$  and convergence tolerance  $\varepsilon$

Output: Path flow vector  $\mathbf{f}^*$  satisfying the  $\delta$ -EBR-MUE equilibrium conditions.

Algorithm Steps:

1. Initialization: Set iteration counter  $k = 0$  and initialize path flow  $\mathbf{f}^{(0)}$ .
2. Compute Path Costs: For current flow  $\mathbf{f}^{(k)}$ , compute multi-objective path costs:

$$\mathbf{C}_p^{(k)} = \mathbf{C}_p(\mathbf{f}^{(k)}),$$

extracting the time cost component:

$$t_p^{(k)} = \text{time component of } \mathbf{C}_p^{(k)}.$$

3. Determine Minimum Time Cost per OD Pair: For each OD pair  $w$ , find the shortest travel time among all paths:

$$t_w^{*(k)} = \min_{p \in \mathcal{P}_w} t_p^{(k)}.$$

4. Filter Feasible Paths: Construct the feasible path set for each OD pair:

$$\mathcal{P}_w^{(k)} = \left\{ p \in \mathcal{P}_w \mid t_p^{(k)} \leq t_w^{*(k)} + \delta, \quad p \text{ is non-strictly dominated} \right\}.$$

Here, a path  $p$  is non-strictly dominated if there does not exist another path  $p'$  such that:

$$\mathbf{C}_{p'}^{(k)} < \mathbf{C}_p^{(k)} \quad (\text{all objectives strictly less}).$$

5. Flow Reassignment Subproblem: Fixing the feasible path sets  $\{\mathcal{P}_w^{(k)}\}$ , solve the flow assignment problem:

$$\min_{\mathbf{f} \geq 0} \sum_{w \in W} \sum_{p \in \mathcal{P}_w^{(k)}} \int_0^{f_p} C_p(s) ds,$$

subject to:

$$\sum_{p \in \mathcal{P}_w^{(k)}} f_p = d_w, \quad \forall w \in W.$$

Denote the solution by  $\mathbf{f}^{(k+1)}$ .

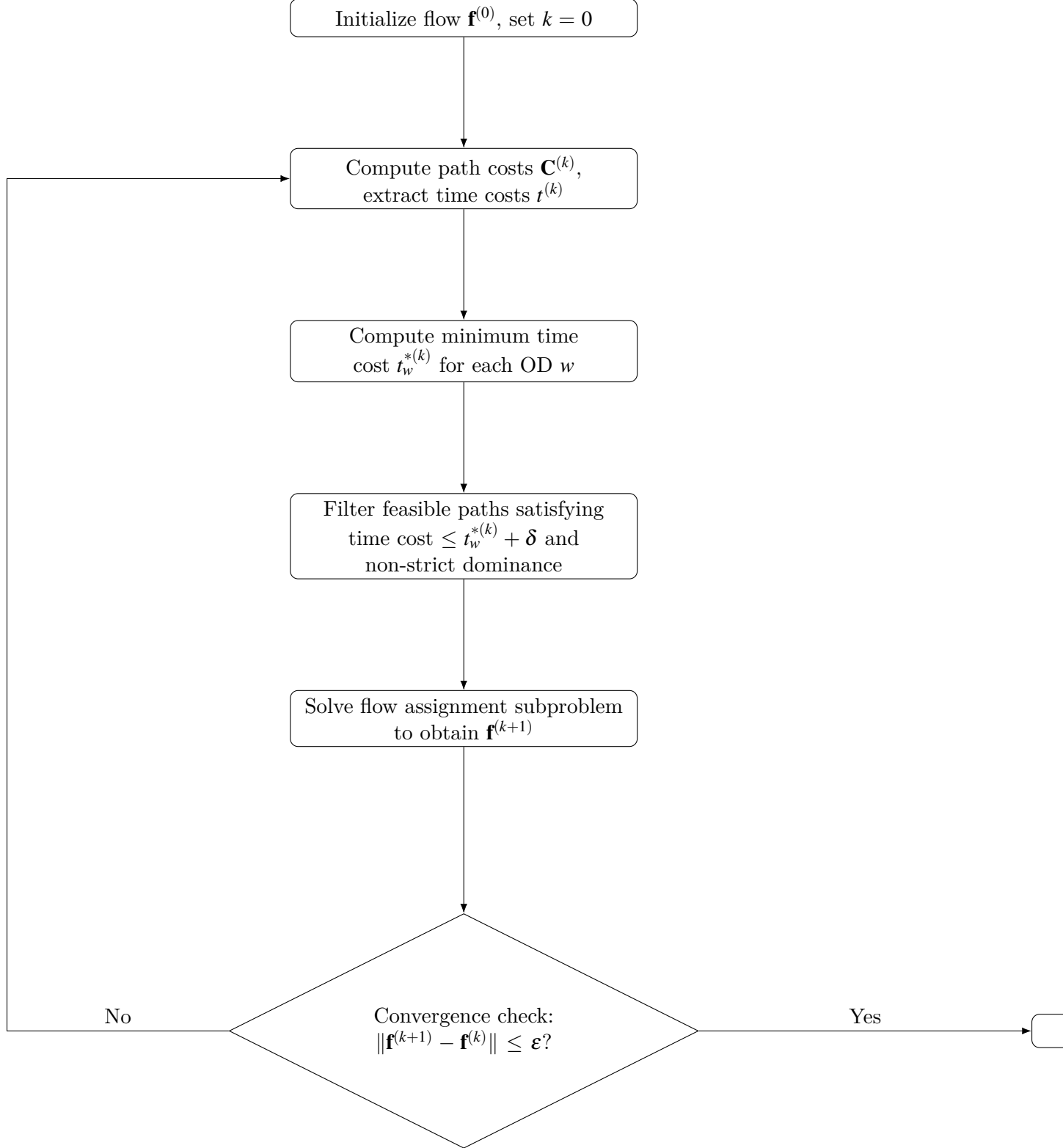
6. Check Convergence: If

$$\|\mathbf{f}^{(k+1)} - \mathbf{f}^{(k)}\| \leq \varepsilon,$$

terminate and set  $\mathbf{f}^* = \mathbf{f}^{(k+1)}$ .

7. Iteration: Otherwise increment  $k \leftarrow k + 1$ . If  $k > K_{\max}$ , terminate with current solution; else return to Step 2.

Flowchart summary:



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Algorithm 1 VI-based Algorithm for Solving  $\delta$ -Bounded  $\varepsilon$ -Non-strictly Dominated User Equilibrium ( $\delta$ -EBR-MUE)

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Require: Initial flow  $\mathbf{f}^{(0)}$ , OD demands  $\{d^w\}$ , path sets  $\{\mathcal{P}_w\}$ , tolerance  $\delta > 0$ ,  $\varepsilon \in \mathbb{R}_+^m$ , maximum iteration  $K_{\max}$

Ensure: Equilibrium flow  $\mathbf{f}^*$

1: Set  $k \leftarrow 0$

2: repeat

3:   Compute the cost vector  $\mathbf{C}_p(\mathbf{f}^{(k)}) = (C_p^1, \dots, C_p^m)$  for all  $p$

4:   for each OD pair  $w \in \mathcal{W}$  do

5:     Find minimum time cost:  $\mathcal{T}_w = \min_{p \in \mathcal{P}_w} C_p^1(\mathbf{f}^{(k)})$

6:     Define feasible path set:

$$\hat{\mathcal{P}}_w := \left\{ p \in \mathcal{P}_w : C_p^1(\mathbf{f}^{(k)}) \leq \mathcal{T}_w + \delta, \nexists q \in \mathcal{P}_w \text{ s.t. } \mathbf{C}_q \leq \mathbf{C}_p - \varepsilon \right\}$$

7:   end for

8:   Define feasible flow set:

$$\mathcal{F}^{\delta, \varepsilon} := \left\{ \mathbf{f} \geq 0 \mid \sum_{p \in \mathcal{P}_w} f_p = d^w, f_p > 0 \Rightarrow p \in \hat{\mathcal{P}}_w, \forall w \right\}$$

9:   Define operator  $F(\mathbf{f}) = (\mathbf{C}_p(\mathbf{f}))_{p \in \cup_w \mathcal{P}_w}$

10:   Solve the following variational inequality (projected method or proximal point method):

$$\text{Find } \mathbf{f}^{(k+1)} \in \mathcal{F}^{\delta, \varepsilon} \text{ such that } \langle F(\mathbf{f}^{(k+1)}), \mathbf{f} - \mathbf{f}^{(k+1)} \rangle \geq 0, \quad \forall \mathbf{f} \in \mathcal{F}^{\delta, \varepsilon}$$

11:    $k \leftarrow k + 1$

12: until  $\|\mathbf{f}^{(k)} - \mathbf{f}^{(k-1)}\| \leq \text{tol}$  or  $k \geq K_{\max}$

13: return  $\mathbf{f}^{(k)}$

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Algorithm 2  $\delta$ -bounded  $\varepsilon$ -Non-dominated User Equilibrium Solver

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Require: Initial flow vector  $\mathbf{f}^0$ , OD demands  $\{d^w\}_{w \in \mathcal{W}}$ , path sets  $\{\mathcal{P}_w\}$ , tolerance parameters  $\delta > 0$ ,  $\varepsilon \in \mathbb{R}_+^m$ , maximum iterations  $K_{\max}$

Ensure: Approximate equilibrium flow  $\mathbf{f}^*$

- 1: Set  $k \leftarrow 0$
- 2: Initialize  $\mathbf{f}^{(0)}$
- 3: repeat
- 4:   Compute cost vectors  $\mathbf{C}_p(\mathbf{f}^{(k)})$  for all paths  $p$
- 5:   for each OD pair  $w \in \mathcal{W}$  do
- 6:     Let  $\mathcal{T}_w^{(k)} = \min_{p \in \mathcal{P}_w} C_p^1(\mathbf{f}^{(k)})$   $\triangleright$  Minimum travel time cost
- 7:     Let  $\hat{\mathcal{P}}_w^{(k)} \leftarrow \emptyset$
- 8:     for each path  $p \in \mathcal{P}_w$  do
- 9:       if  $C_p^1(\mathbf{f}^{(k)}) \leq \mathcal{T}_w^{(k)} + \delta$  and not strictly  $\varepsilon$ -dominated by other paths in  $\mathcal{P}_w$  then
- 10:        Add  $p$  to  $\hat{\mathcal{P}}_w^{(k)}$
- 11:       end if
- 12:     end for
- 13:   end for
- 14:   Assign flow uniformly over  $\hat{\mathcal{P}}_w^{(k)}$  for each  $w$ , respecting demand  $d^w$ , to obtain  $\mathbf{f}^{(k+1)}$
- 15:    $k \leftarrow k + 1$
- 16: until  $\|\mathbf{f}^{(k)} - \mathbf{f}^{(k-1)}\| \leq \text{tol}$  or  $k \geq K_{\max}$
- 17: return  $\mathbf{f}^{(k)}$

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## 5 Support Function Approximation of the $\varepsilon$ -Nondominated Set

Let  $\mathcal{C}^w := \{\mathbf{C}_p : p \in \mathcal{P}^w\}$  and choose  $\mathcal{U}_N \subset \mathbb{S}^1$  directions. Define:

$$\rho_{\mathcal{C}^w}(\mathbf{u}) := \min_{p \in \mathcal{P}^w} \mathbf{u}^\top \mathbf{C}_p.$$

Then:

$$\mathcal{E}_\varepsilon^w := \left\{ \mathbf{C}_p : \forall \mathbf{u} \in \mathcal{U}_N, \mathbf{u}^\top \mathbf{C}_p \leq \rho_{\mathcal{C}^w}(\mathbf{u}) + \varepsilon_{\mathbf{u}} \right\}.$$

As  $N \rightarrow \infty$ ,  $\mathcal{E}_\varepsilon^w$  converges in Hausdorff distance to the true weakly nondominated frontier.

## 6 Limiting Case: $\varepsilon = 0$

We consider the limiting case where the dominance tolerance  $\varepsilon = (0, 0)$ . In this case, the  $\delta$ -EBR-MUE model reduces to a sharper form, where users only accept paths that are not strictly dominated in the cost vector space, while still satisfying the travel time bound.

**Definition 6.1** ( $\delta$ -Bounded Weakly Non-Dominated User Equilibrium ( $\delta$ -WNUE)). A feasible path flow  $\mathbf{f}^*$  is said to satisfy the  $\delta$ -WNUE condition if:

$$f_p^w > 0 \Rightarrow \begin{cases} \mathbf{C}_q(\mathbf{f}^*) \not\prec \mathbf{C}_p(\mathbf{f}^*), & \forall q \in \mathcal{P}^w \\ t_p(\mathbf{f}^*) \leq T_w^{\text{UE}}(\mathbf{f}^*) + \delta, \end{cases}$$

Proposition 6.1. Every  $\delta$ -EBR-MUE with  $\varepsilon = 0$  is a  $\delta$ -WNUE. Conversely, every  $\delta$ -WNUE is a  $\delta$ -EBR-MUE with  $\varepsilon = 0$ .

Proof. Immediate from the definitions: with  $\varepsilon = 0$ , the  $\varepsilon$ -non-dominance condition becomes standard weak non-dominance. The time-bound condition remains unchanged.  $\square$

Remark 6.1. As  $\varepsilon \rightarrow 0$ , the  $\delta$ -EBR-MUE set converges to the set of  $\delta$ -WNUE flows. However, for any  $\varepsilon > 0$ , the equilibrium permits a richer variety of solutions, possibly including near-dominated paths, thereby reflecting behavioural flexibility and tolerance.

## 7 Numerical Example

Consider a simple network with three parallel paths. Let the travel time on each path be  $t_p(f) = a_p + b_p f_p$  and cost  $c_p$  be fixed. Set:

- $a = [3, 5, 8]$ ,  $b = [0.01, 0.01, 0.02]$ ;
- $c = [2, 1.5, 0.5]$  (monetary units);
- demand  $d = 1$ ,  $\varepsilon = (1.5, 0.5)$ ,  $\delta = 2$ .

Compute:

1. User Equilibrium (UE);
2. Bounded Rational UE (BRUE);
3.  $\delta$ -EBR-MUE using fixed-point iteration.

Compare results in terms of used paths, average cost, and Pareto dominance.