

ON MULTI-MARGINAL PARTIAL OPTIMAL TRANSPORT: RECTIFYING INFEASIBLE EXTENSION STRATEGIES AND EFFICIENT PRIMAL-DUAL METHODS

Quang Nguyen Dinh Thien

Dung Hoang Duc

University of Information Technology, Vietnam National University, Ho Chi Minh City, Vietnam

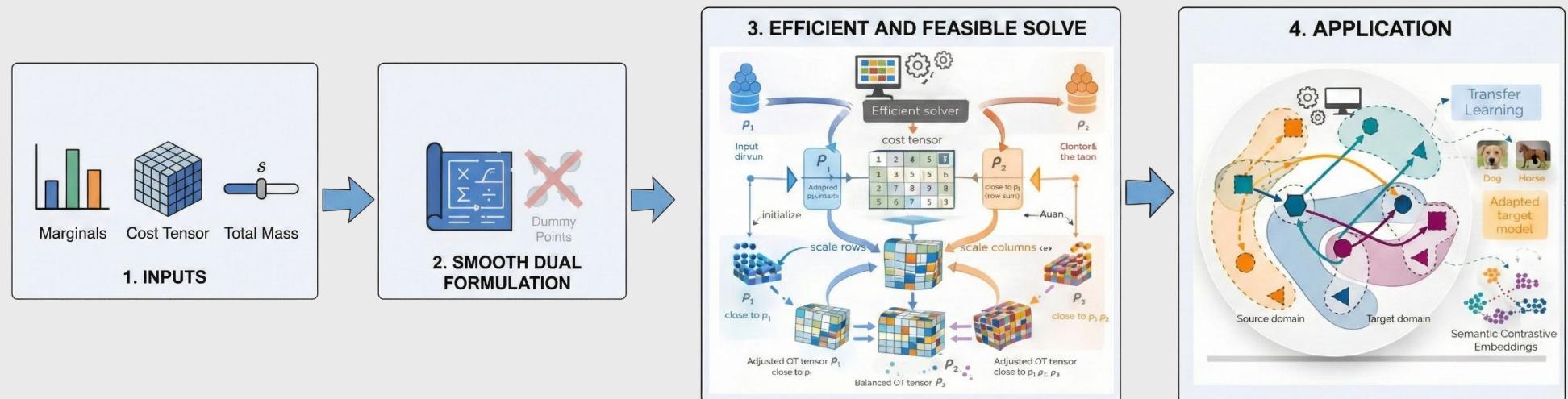
Motivation

- Robustness:** MMPOT handles **noisy and unbalanced data** significantly better than standard Optimal Transport.
- Research Gap:** Current "dummy point" extension strategies are mathematically flawed, leading to **infeasible solutions** that violate total mass constraints.
- Computational Cost:** Previous methods suffer from high complexity $\mathcal{O}(1/\epsilon^4)$, making them impractical for large-scale problems.

Contribution

- Novel Framework:** A new **Primal-Dual formulation** for MMPOT that eliminates the need for dummy point extensions.
- Feasibility:** Theoretically rectifies the infeasibility of previous strategies, ensuring strict adherence to mass constraints.
- Fast Algorithms:** Introduces **GreenkhornMMPOT**, **PDAAM**, and **APDAGD** solvers achieving an optimal convergence rate of $\mathcal{O}(1/\epsilon)$.

Overview



Description

1. The Problem with Previous Methods

- Infeasibility of Extension Strategies:** Current methods rely on adding "dummy points" to balance mass. However, the non-linearity of **Entropic Regularization** mathematically prevents accurate mass adjustment at the marginals.
- Consequence:** This results in **infeasible solutions** that strictly violate the total mass transportation constraint (s).
- Complexity Explosion:** To minimize approximation error, the extended cost entries must be extremely large ($A \approx \mathcal{O}(1/\epsilon)$). This causes the Sinkhorn complexity to surge to $\mathcal{O}(1/\epsilon^4)$ (compared to the standard $\mathcal{O}(1/\epsilon^2)$), or even grow exponentially when marginals $m \geq 4$.

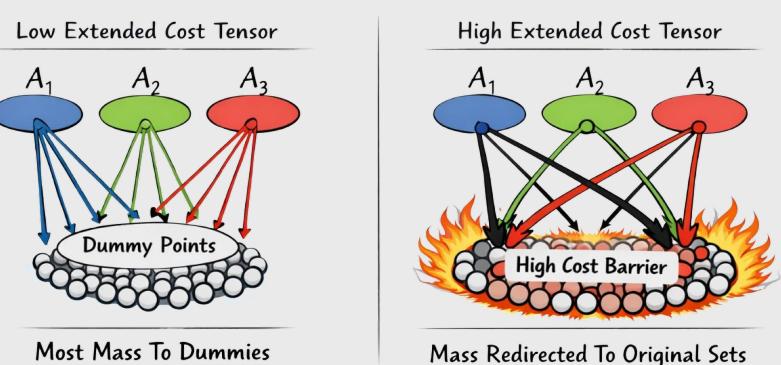


Figure 1. Illustration of mismatch mass behaviour in MMPOT

2. Proposed Algorithms

- Novel Primal-Dual Framework:** Instead of tensor extension, we establish a **smooth dual formulation** directly for MMPOT. This eliminates approximation errors caused by dummy points.
- New Solvers:** We introduce three algorithms:
 - GreenkhornMMPOT:** Greedy coordinate descent, highly efficient for early iterations.
 - PDAAM & APDAGD:** Accelerated algorithms that achieve the optimal theoretical convergence rate of $\mathcal{O}(1/\epsilon)$.
 - Rounding Algorithm:** A novel rounding method utilizing an "Enforcing Procedure" is proposed to project approximate solutions onto the feasible set, ensuring strict adherence to mass constraints.

3. Applications & Results

- Application:** Partial Barycenter computation on noisy datasets (e.g., MNIST). MMPOT is shown to effectively filter outliers by transporting only a specific fraction of the total mass, unlike standard MMOT.
- Performance:** The proposed algorithms (PDAAM, APDAGD) demonstrate significantly **faster convergence** in terms of both runtime and iteration count compared to the Extension Sinkhorn baselines.
- Accuracy:** The solutions achieve lower objective value gaps while strictly satisfying the total mass constraints, verifying the theoretical feasibility analysis.

Algorithm	Complexity	Extended Cost	Optimal n -dep.	Optimal ϵ -dep.	All Mass Transport
First Ext. Sinkhorn	$\mathcal{O}\left(\frac{n^m m^5 \ C\ _\infty^2 \log n}{\epsilon^4}\right)$	✓	✓	✗	✗
Second Ext. Sinkhorn	$\mathcal{O}\left(\frac{n^m m^5 \ C\ _\infty^2 \log n}{\epsilon^4}\right)$	✓	✓	✗	✓
GreenkhornMMPOT	$\mathcal{O}\left(\frac{n^m m^4 \ C\ _\infty \log n}{\epsilon^2}\right)$	✗	✓	✗	✓
PDAAM	$\mathcal{O}\left(\frac{n^{m+0.5} m^3 \ C\ _\infty \log n}{\epsilon}\right)$	✗	✗	✓	✓
APDAGD	$\mathcal{O}\left(\frac{n^{m+0.5} m^3 \ C\ _\infty \log n}{\epsilon}\right)$	✗	✗	✓	✓

Figure 2. Comparison of entropic algorithm solvers for MMPOT