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Motivation

This work addresses a facility location problem with customer behavior considerations. We develop a Decision-Dependent Distributionally Robust Optimization (DDRO) model that captures customer behavior as a second-stage recourse. Specifically, we utilize a ranking-based choice model informed by machine learning to represent consumer preferences.

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Problem Setting

$$\text{DDRO: } \min \left\{ c^T x + \max_{\mathbb{P}_f \in U(x)} \mathbb{E}[g(x, \tilde{f})] : x \in X \subseteq \{0, 1\}^{|N| \times |H|} \right\}$$

where

$$g(x, \tilde{f}) = \min_{\pi, L \geq 0} \sum_{q \in Q} \sum_{i \in N} \sum_{h \in H} r_{ih} \tilde{f}_q p_q \pi_{qih} + \sum_{i \in N} \sum_{h \in H} l_{ih} L_{ih} + F(\pi)$$

$$\text{s.t. } \pi_q \in \Pi_q(x) \subseteq \{0, 1\}^{|N| \times |H|},$$

$$\sum_{q \in Q} \tilde{f}_q p_q \pi_{qih} \leq C_{ih} x_{ih} + L_{ih}, \quad \forall i \in N, h \in H, q \in Q.$$

- N : Facility location set. • x : First-stage facility location decisions.
- H : Facility type set. • π : Second-stage customer adoption rate decisions.
- Q : Customer set. • L : Second-stage unsatisfied demand decisions.
- \tilde{f}_q : Random demand. • p_q : Average consumption of customer's capacity.
- c : Cost of opening facilities. • l_{ih} : Penalty per unit for unmet demand.
- r_{ih} : Revenue per unit of demand served. • C_{ih} : Maximum capacity for each type facility.
- $F(\pi)$: Penalty for unsatisfied customer preference.

Ambiguity set

$$U(x) = \left\{ \theta : \theta \in \mathbb{R}_+^{|K|}, \sum_{k=1}^K \theta^k = 1, \left| \sum_{k=1}^K \theta^k f^k - \mu_q(x) \right| \leq \epsilon_q^\mu, \forall q \in Q \right\}.$$

- K : Support set. • $\mu_q(x)$: Expected mean with robustness parameter ϵ_q^μ .

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Customer Behavior

$\Pi_q(x) = \{ \pi_q : \pi_{qih} = 1 \text{ if } (i, h) \in \text{argmax}_{(i, h) \in G_q(x)} \{ u_{qih} \}, \text{ and } \pi_{qih} = 0 \text{ otherwise, } \forall i \in N, h \in H \}$

$G_q(x)$: available (i, h) pairs under given x .

For every trip q , using utilities $u_{qih}, \forall i \in N, h \in H$, we can obtain ranking based customer choices (see Section 5)

Preference Ranking

Rankings $\{\sigma^1, \sigma^2, \dots, \sigma^{|Q|}\}$ over $|N| \times |H| + 1$ options indicating each facility and type pair and customers not selecting any service.

Each ranking $\sigma^q : \{(0,0), \dots, (|N| \times |H|)\} \rightarrow \{0, 1, \dots, |N| \times |H|\}$ is a bijection that assigns each option to a rank, where $(0,0)$ denotes the customer choosing no facility.

Then we have $\pi_q \in \Pi_q(x) \subseteq \{0, 1\}^{|N| \times |H|}$.

$$\sum_{i \in N \cup \{0\}} \sum_{h \in H} \pi_{qih} = 1, \pi_{qih} \leq x_{ih}, \sum_{(i', h') : \sigma^q(i', h') > \sigma^q(i, h)} \pi_{qih'} \leq 1 - x_{ih},$$

$$\sum_{(i', h') : \sigma^q(i', h') > \sigma^q(0,0)} \pi_{qih'} = 0, \forall i \in N, q \in Q, h \in H$$

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Reformulations

Dual form for the second-stage problem:

$$\min_{\beta_q^1, \beta_q^2 \geq 0, \alpha} \left\{ \alpha + \sum_{q \in Q} \beta_q^1 (\mu_q(x) + \epsilon_q^\mu) - \sum_{q \in Q} \beta_q^2 (\mu_q(x) - \epsilon_q^\mu) \mid \alpha + \sum_{q \in Q} (\beta_q^1 f_q^k - \beta_q^2 f_q^k) \geq g(x, f^k), \forall k \right\}$$

Here, $\alpha, \beta_q^1, \beta_q^2$ are the dual variables associate with ambiguity set.

Then, the DDRO full model can be written as the following

$$\begin{aligned} \min \quad & c^T x + \alpha + \sum_{q \in Q} \beta_q^1 (\mu_q(x) + \epsilon_q^\mu) - \sum_{q \in Q} \beta_q^2 (\mu_q(x) - \epsilon_q^\mu) \\ \text{s.t.} \quad & \alpha + \sum_{q \in Q} (\beta_q^1 f_q^k - \beta_q^2 f_q^k) \geq - \sum_{q \in Q} \sum_{i \in N} \sum_{h \in H} r_{ih} f_q^k p_q \pi_{qih}^k + \sum_{i \in N} \sum_{h \in H} l_{ih} L_{ih}^k + F(\pi), \quad \forall k \in K \\ & \sum_{i \in N} \sum_{h \in H} \pi_{qih}^k = 1, \quad \forall q \in Q, k \in K \\ & \pi_{qih}^k \leq x_{ih}, \quad \forall i \in N, q \in Q, h \in H, k \in K \\ & \sum_{(i', h') : \sigma^q(i', h') > \sigma^q(i, h)} \pi_{qih'}^k \leq 1 - x_{ih}, \quad \forall i \in N, q \in Q, h \in H, k \in K \\ & \sum_{(i', h') : \sigma^q(i', h') > \sigma^q(0,0)} \pi_{qih'}^k = 0, \quad \forall q \in Q, k \in K \\ & \sum_{q \in Q} f_q^k p_q \pi_{qih}^k \leq C_{ih} x_{ih} + L_{ih}^k, \quad \forall i \in N, h \in H, k \in K \\ & \sum_{h \in H} x_{ih} \leq 1, \quad \forall i \in N \\ & x_{ih} \in \{0, 1\}, \quad \forall i \in N, h \in H \\ & \pi_{qih}^k \in \{0, 1\}, \forall i \in N, q \in Q, h \in H, k \in K \\ & L_{ih}^k \geq 0, \forall i \in N, h \in H, k \in K \end{aligned}$$

Solution Algorithm

Column-and-Constraint Generation (CCG) method

Algorithm 1 Column-and-Constraint Generation Method (CCG)

- 1: **Input:** scenario set K , maximum iteration T , tolerance $\epsilon > 0$.
- 2: Initialize the master problem with first-stage constraints.
- 3: **for** $t = 0, 1, \dots, T$ **do**
- 4: $(x_t, \alpha_t, \beta_{q(t)}^1, \beta_{q(t)}^2) \leftarrow$ solving the master problem.
- 5: Fix x_t and construct the second-stage solution using closed form.
- 6: Compute $\text{best_val}_t = \max_{k \in K} \{ g(x_t, f^k) - \sum_{q \in Q} (\beta_{q(t)}^1 - \beta_{q(t)}^2) f_q^k \}$.
- 7: Let k_t^* be the scenario attaining best_val_t .
- 8: **if** $\alpha_t + \epsilon \leq \text{best_val}_t$ **then**
- 9: Add $\alpha + \sum_{q \in Q} (\beta_q^1 f_q^{k_t^*} - \beta_q^2 f_q^{k_t^*}) \geq g(x, f^{k_t^*})$ to the master problem.
- 10: Add the second-stage constraints for scenario k_t^* to the master problem.
- 11: **else**
- 12: Break
- 13: **end if**
- 14: **end for**
- 15: Return x_t .

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Multinomial Logit (MNL) Choice Model

Utility function: $u_{qih} = v_{qih} + \epsilon_{qih}$ where $v_{qih} = \beta_1 d_{qih} + \beta_2 h$

Here, d_{qih} is the distance between trip q and station (i, h) , β_1, β_2 are parameters to be estimated. We set the error ϵ_{qih} i.i.d. disturbances following the Gumbel distribution.

Then under this assumption, the probability of customer q choosing station i with type h can be written as

$$P_{qih}(S_q) = \frac{\exp(v_{qih})}{\sum_{j \in S_q} \exp(v_{qij})}$$

Where $S_q \subset [|N| \times |H|]$ denotes the choice set available to individual q .

Maximum Likelihood Estimation

$$\max_{\beta_1, \beta_2} \sum_{j=1}^J \log \left(\frac{\exp(\beta_1 d_{jih} + \beta_2 h)}{\sum_{(i', h') \in [|N| \times |H|]} \exp(\beta_1 d_{j i' h'} + \beta_2 h)} \right)$$

J : number of data point

We use the Newton's method (e.g. SLSQP) to solve this problem.

Once we have β_1, β_2 , we can finally define the ranking using utility functions by

$$\mathbb{E}[u_{qih}] > \mathbb{E}[u_{q i' h'}] \text{ if and only if } \sigma^q(i, h) < \sigma^q(i', h')$$

What happens if we consider customer behavior uncertainty (BU) ?

$$\text{DDRO+BU: } \min \left\{ c^T x + \max_{\mathbb{P}_f \in U(x)} \max_{m \in M} \mathbb{E}[g(x, \tilde{f}, \tilde{m})] : x \in X \subseteq \{0, 1\}^{|N| \times |H|} \right\}$$

Here, M can be used to indicate different rankings, based on utility values.

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Computational Study

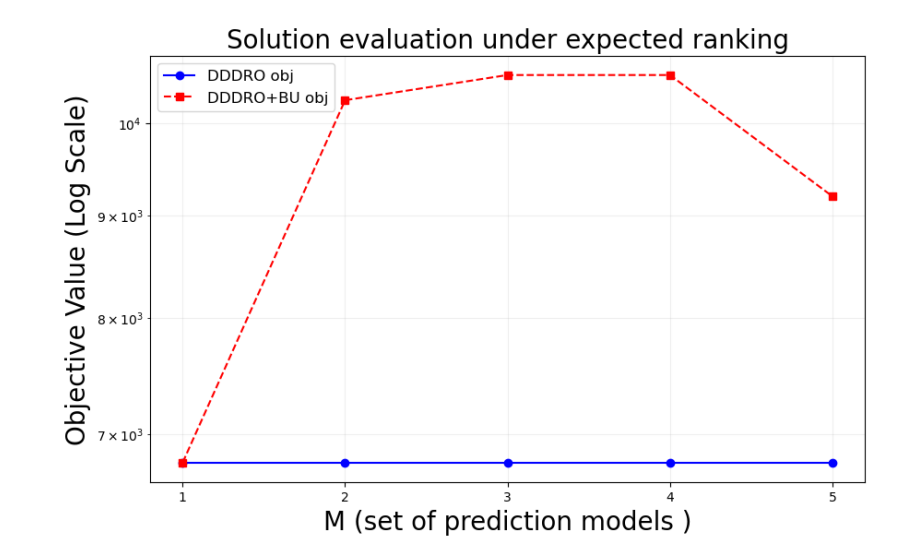
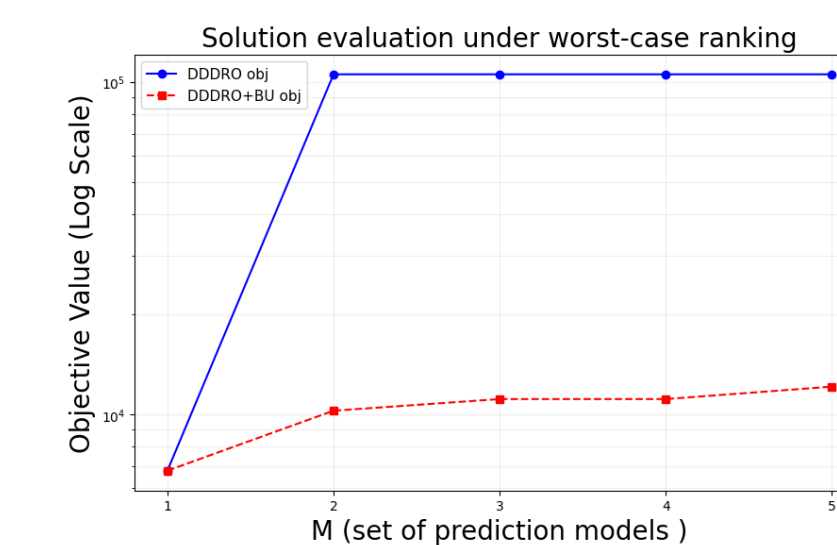
Computational performance under synthetic data, with $M=1$

$ K , Q , N , H $	Gurobi		CP		CCG		Speed up	
	RT (s)	Iter	RT (s)	Iter	RT (s)	Iter	Gurobi/CCG	CP/CCG
20, 10, 10, 2	13.49	39.37	4	2.14	3	6.30	18.40	
20, 20, 20, 2	89.09	177.77	3	10.60	3	8.40	16.77	
50, 10, 10, 2	34.75	99.02	4	2.06	3	16.87	48.07	
50, 20, 20, 2	217.49	430.06	3	10.70	3	20.33	40.19	
100, 10, 10, 2	71.80	190.81	4	2.18	3	32.94	87.53	
100, 20, 20, 2	622.30	862.87	3	11.07	3	56.22	77.95	

- CP: Cutting-Plane; RT: Runtime; Iter: Iteration; Speed up: ratio of runtime between different methods.

Computational performance under California EV charging data

		$m=1$	$m=2$	$m=3$	$m=4$	$m=5$	Avg. over rankings
		Average cost	Std. dev.	90%	Average cost	Std. dev.	
DRO	Average cost	2751.14	39277.57	29447.77	43725.42	20123.18	27065.02
	Std. dev.	14.71	308.43	194.74	386.40	179.49	216.75
	90%	2770.40	39678.49	29706.99	44228.66	20351.77	27347.26
DDRO	Average cost	3021.04	35504.76	33922.39	5966.89	18081.13	19299.24
	Std. dev.	16.64	303.40	224.80	24.63	159.24	145.74
	90%	3042.58	39508.02	34221.82	5997.93	18286.68	19491.41
DDRO+BU under $M=5$	Average cost	6504.11	7647.77	7641.29	7075.19	2513.21	6276.31
	Std. dev.	26.99	31.98	32.55	27.46	20.54	27.90
	90%	6539.02	7689.91	7681.93	7109.63	2538.92	6311.88



- The CCG algorithm consistently achieves the fastest convergence, outperforming both Reformulation and CP by several orders of magnitude as problem size increases.
- Under BU with different customer rankings, DDRO+BU obtains the most stable results followed by DDRO and then DRO.
- The first graph shows that DDRO+BU achieves much better solutions than DDRO under the worst-case ranking in the BU uncertainty set.
- The second graph shows that DDRO+BU is more conservative than DDRO under the baseline ranking model, but the gap is much smaller than the worst-case setting.

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References

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- [2] Dimitris Bertsimas and Velibor V. Misić. Exact first-choice product line optimization. *Operations Research*, 67(3):651-670, 2019.