

ASSORTMENT OPTIMIZATION

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• (Recall) Choice Model. Given $S \subseteq N = \{1, 2, \dots, n\}$ (products)

- $\pi_j(s) = \mathbb{P}[\text{prod } j \text{ purchased from } S]$

- LCA choice models - $\exists v_i > 0$ s.t

$$\pi_j(s) = \frac{v_j \mathbb{1}_{\{j \in S\}}}{v_0 + v(s)}, \quad v_0 \equiv \text{'attractiveness' of no purchase}$$
$$v(s) = \sum_{i \in S} v_i$$

- Mixture of LCA (mixed MNL)

$$\pi_j(s) = \sum_{g \in G} \alpha^g \left(\frac{v_j^g}{v_0^g + v^g(s)} \right), \quad \sum_{g \in G} \alpha^g = 1$$

- Markov Chain choice model - $\Lambda \equiv \{\lambda_i\}_{i \in N}$
 $P \equiv \{p_{ij}\}_{i, j \in N^2}$

$\mathbb{P}[\text{purchase } j \in S]: \pi_j(s) = \lambda_j + \sum_{i \in \bar{S}} \phi_i(s) p_{ij} \quad \forall j \in S$

$\mathbb{P}[\text{consider } j \notin S]: \phi_j(s) = \lambda_j + \sum_{i \in \bar{S}} \phi_i(s) p_{ij} \quad \forall j \in \bar{S}$

$$\pi_0(s) = 1 - \sum_{j \in S} \pi_j(s)$$

• The assortment optimization problem

- Exogenous prices (profits) $P_j \quad \forall j \in N$

- $R^* = \max_{S \subseteq N} R(s) = \max_{S \subseteq N} \sum_{j \in S} P_j \pi_j(s)$

* Assortment Opt under LCA / MNL

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$$R(s) = \sum_{j \in S} \frac{p_j v_j}{v_0 + v(s)}, \quad R^* = \max_{S \subseteq N} R(s)$$

Thm - Let $p_1 \geq p_2 \geq \dots \geq p_n$

(Nested-by-revenue sets) $E_0 = \emptyset, E_1 = \{1\}, E_2 = \{1, 2\}, \dots, E_n = N$

Then $\exists R^* \in \{0, 1, \dots, n\}$ s.t. $E_{R^*} \in \arg \max_{S \subseteq N} R(s)$

Pf - By definition $R^* \geq \sum_{j \in S} \frac{p_j v_j}{v_0 + v(s)} \quad \forall S \subseteq N$

$$\Rightarrow v_0 R^* \geq \sum_{j \in S} v_j (p_j - R^*) \quad \forall S \subseteq N$$

$\therefore \exists$ some $S \subseteq N$ s.t. $R(s) = R^*$

$$\Rightarrow \arg \max_{S \subseteq N} R(s) = \arg \max_{S \subseteq N} \left\{ \sum_{j \in S} (p_j - R^*) v_j \right\}$$

- Thus we want to find $S \in \arg \max_{S \subseteq N} \left\{ \sum_{j \in S} (p_j - R^*) v_j \right\}$

$$\Rightarrow S^* = \{j \in N; p_j \geq R^*\}$$

- Now even if we do not know R^* , it is clear

- that we only need to consider $S \in \{E_0, E_1, \dots, E_n\}$

LCA with constraints

- Let $x^S \in \{0,1\}^n \equiv$ Indicator of set $S \subseteq N$
(ie, $x_j^S \equiv \mathbb{1}_{\{j \in S\}}$)
- We now want to solve a constrained assortment optⁿ

$$\max_{x \in \{0,1\}^n} \frac{\sum_{j \in N} p_j v_j x_j}{v_0 + \sum_{j \in N} v_j x_j}$$

$$\text{s.t.} \quad \sum_{j \in N} a_{ij} x_j \leq b_i \quad \forall i \in L$$

$$x_j \in \{0,1\} \quad \forall j \in N$$

- Assumption - $A = \{a_{ij}\}$ is totally unimodular, $b_i \in \mathbb{Z}$
(\Rightarrow extreme points of $\{Ax \leq b\}$ are integral)

$$\text{Eg - } \sum_{j \in N} x_j \leq C$$

$$\text{- If } N = \underbrace{S_1 \cup S_2 \cup \dots \cup S_k}_{\text{Partition}}, \quad \sum_{j \in S_i} x_j \in \{b_{s_i}, \dots, B_{s_i}\}$$

- Joint Pricing and assortment optⁿ

• Products $N = \{1, \dots, n\}$, Prices $P = \{p_1, p_2, \dots, p_k\}$

• $v_{ik} \equiv$ attractiveness of product i at price p_k

• Idea - Create virtual products: $x_{ik} \equiv$ product i at price k
- Constraint: at most one $x_{ik} = 1$ for every i

• How do we solve constrained MNL pricing?

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OPT1:
$$\max \sum_{j \in N} \frac{P_j U_j x_j}{U_0 + U^T x}$$

s.t. $Ax \leq b$

$(L \times N) \rightarrow x_j \in \{0, 1\}$

OPT2

$$\max \sum_{j \in N} P_j y_j$$

s.t. $\sum_{j \in N} y_j + y_0 = 1$

$$\sum_{j \in N} \frac{a_{ij}}{U_j} y_j \leq \frac{b_i}{U_0} y_0 \quad \forall i \in L$$

$$0 \leq \frac{y_j}{U_j} \leq \frac{y_0}{U_0} \quad \forall j \in N$$

Thm - The above problems have the same optimal objective.

Moreover, given a solution to OPT2, we can construct a solution to OPT1.

Pf - First, as in prev result, we have OPT1 is equiv to

OPT3
$$\max \sum_{j \in N} (P_j - R^*) \frac{U_j}{U_0} x_j$$

s.t. $Ax \leq b, \quad 0 \leq x_j \leq 1$

where $R^* \equiv$ OPT1 objective

This follows from LCA + total unimodularity of A form

Thus we need to show $OPT3 \equiv OPT2$

- (5)
- Let $\{y_j^*\}_{j \in N \cup \{0\}}$ be an optimal soln to OPT2
 - $\{x_j^*\}_{j \in N \cup \{0\}}$ be an optimal soln to OPT3

By defn, $\text{OPT3}(x_j^*) = R^*$

- Now we show $\gamma^* = \text{OPT2}(y_j^*) = R^*$

$$\text{Let } \hat{y}_j = \frac{\sigma_j x_j^*}{\sigma_0 + \sum \sigma_i x_i^*}, \quad \hat{y}_0 = 1 - \sum_{j \in N} \hat{y}_j = \frac{\sigma_0}{\sigma_0 + \sum \sigma_i x_i^*}$$

then $\{\hat{y}_j\}$ satisfies constraints of OPT2

$$- \sum_{j \in N} \frac{a_{ij}}{\sigma_j} \hat{y}_j = \sum_{j \in N} \frac{a_{ij} x_j^*}{\sigma_0 + \sum \sigma_i x_i^*} \leq \frac{b_i}{\sigma_0 + \sum \sigma_i x_i^*} = \frac{b_i \hat{y}_0}{\sigma_0}$$

$$- \frac{\hat{y}_j}{\sigma_j} = \frac{x_j^*}{\sigma_0 + \sum \sigma_i x_i^*} \leq \frac{\hat{y}_0}{\sigma_0} \quad \forall j$$

$\Rightarrow \{\hat{y}_j\}$ is feasible for OPT2

$$\Rightarrow \gamma^* \geq \text{OPT2}(\{\hat{y}_j\}) = \sum_j p_j \frac{x_j^* \sigma_j}{\sigma_0 + \sum \sigma_i x_i^*} = R^*$$

- Now suppose $\gamma^* = \text{OPT2}(y_j^*) > R^*$. Note $y_0^* > 0$

Consider $\hat{x}_j = \frac{y_j^* / \sigma_j}{y_0^* / \sigma_0}$ - Check that \hat{x}_j is feasible for OPT3

$$\begin{aligned} \text{Then } \text{OPT3}(\{\hat{x}_j\}) &= \sum_{j \in N} (p_j - R^*) \frac{\sigma_j}{\sigma_0} \hat{x}_j = \frac{1}{y_0^*} \sum_{j \in N} p_j y_j^* - \frac{R^* (1 - y_0^*)}{y_0^*} \\ &> \frac{R^*}{y_0^*} - \frac{R^* (1 - y_0^*)}{y_0^*} = R^* \Rightarrow \text{contradiction} \end{aligned}$$

* Assortment optⁿ for universal approximators

⑥

- We now consider assortment optimization for choice models which serve as universal approximators
 - First consider the mixture of MNL model
- Q: Does this have an 'easy' (poly-time) algorithm for assortment optⁿ?

A: No!

Consider the problem - 2 classes, $\alpha^a + \alpha^b = 1$
- Prices p_i , ~~etc~~ ^{choice} parameters $\{u_i\}_{i \in N}$

(Note: $\{u_i\}_{i \in N}$ is annotated with $\{a, b\}$ and N in the original image)

Q: Given any K , does there exist $S \subset N$ st

(2-Class Logit)
$$\sum_{i \in N} \left(\frac{\alpha^a u_i^a p_i}{u_i^a + u_i^b(s)} + \frac{\alpha^b u_i^b p_i}{u_i^b + u_i^a(s)} \right) \geq K?$$

Thm (RSTT 13) - 2-Class Logit is NP-complete
(Reduction from set partition)

Eg - (Nested-by-revenue is not optimal)

$$\alpha^a = 0.5, \quad \theta^a = (5, 20, 1)$$

$$\alpha^b = 0.5, \quad \theta^b = (1/5, 10, 10)$$

$$\text{prices} \equiv (8, 4, 3)$$

Then opt for class a $\equiv \{1\}$, $R^* = 20/3$

opt for class b $\equiv \{1, 2\}$, $R^* = 26/7$

opt for mixture $\equiv \{1, 3\}$, $R^* = 4.48$

• What about the Markov Chain choice model

Thm (BGG'16) - Assortment opt under MC model (Λ, P)

is equivalent to the following LP

$$\begin{aligned} \min \quad & \sum_{i \in N} \lambda_i g_i \\ \text{s.t.} \quad & g_i \geq p_i \quad \forall i \in N \\ & g_i \geq \sum_{j \in N} p_{ij} g_j \quad \forall i \in N \\ & g_i \geq 0 \end{aligned}$$

• Here $g_i \equiv$ 'optimal expected revenue starting from i '

• Moreover, $g_i = p_i \Rightarrow i \in S^*$, $g_i = \sum p_{ij} g_j \Rightarrow i \in N \setminus S^*$