

# Multi-attribute Regret-based Dynamic Pricing

Janyl Jumadinova, Raj Dasgupta  
Computer Science Department  
University of Nebraska, Omaha

# Outline

- Problem: Multi-attribute dynamic pricing
- Solution:
  - Preference elicitation using minimax regret
  - Dynamic pricing using minimax regret
- Experimental validation
  - while varying system parameters
  - comparison with other dynamic pricing approaches

# Problem

- Online market with buyers and sellers
- Simplification: Only one type of product or item is sold/purchased
- Each product is differentiated along a finite set of attributes

# Sellers

- Each seller has multiple, infinite number of items in its inventory
- Each seller has a production cost (min threshold) and each buyer has a reservation cost (max threshold)

# Buyer Attribute Preference Model

	A1	A2	A3	A4	A5
Item	0.2	0.15	0.6	0.0	0.05

	Time	Insurance	Seller Repu.	A/S support	Cust. serv.
Item	0.2	0.15	0.6	0.0	0.05

- Each buyer differentiates a product along different attributes using a preference vector of probabilities
- Set of preference vectors is finite
- Buyers can be of different types (finite set of types)
  - each type corresponds to one preference vector

# Market Operation

	Time	Insur- ance	Seller Repu.	A/S support	Cust. serv.
Item	0.7	0.15	0.1	0.0	0.05

Buyer 1:  
Preferred  
attribute  $a_1$



Buyer 2:  
Preferred  
attribute  $a_3$



Get current offer  
from sellers

Select Seller

Select Seller

Get current offer  
from sellers

Seller 1  
<0.8, 0.4, 0.3, 0.5, 0.1>

Seller 2  
<0.85, 0.3, 0.6, 0.7, 0.3>

Seller 3  
<0.7, 0.1, 0.8, 0.1, 0.2>

Seller 4  
<0.6, 0.2, 0.7, 0.4, 0.1>

	Time	Insur- ance	Seller Repu.	A/S support	Cust. serv.
Item	0.2	0.15	0.6	0.0	0.05

<p1, p2, p3, p4, p5> represents seller prices along different product attributes

# Market Operation: Over Time

	Time	Insur- ance	Seller Repu.	A/S support	Cust serv.
Item	0.2	0.65	0.1	0.0	0.05

Buyer 1:  
Preferred  
attribute  $a_2$



Get current offer  
from sellers

Select Seller

Select Seller

Buyer 2:  
Preferred  
attribute  $a_3$



Get current offer  
from sellers

Seller 1  
<0.8, 0.4, 0.3, 0.5, 0.1>

Seller 2  
<0.85, 0.3, 0.6, 0.7, 0.3>

Seller 3  
<0.7, 0.1, 0.8, 0.1, 0.2>

Seller 4  
<0.6, 0.2, 0.7, 0.4, 0.1>

	Time	Insur- ance	Seller Repu.	A/S support	Cust serv.
Item	0.2	0.15	0.6	0.0	0.05

<p1, p2, p3, p4, p5> represents seller prices along different product attributes

# Sellers' Knowledge

Item	Time	Insurance	Seller Repu.	A/S support	Cust serv.
	0.7	0.15	0.1	0.0	0.05

Item	Time	Insurance	Seller Repu.	A/S support	Cust serv.
	0.2	0.65	0.1	0.0	0.05

Buyer 1:  
Preferred attribute  $a_2$



?

Buyer 2:  
Preferred attribute  $a_3$



Seller 1  
~~X~~  $\langle 0.8, 0.4, 0.3, 0.5, 0.1 \rangle$

Seller 2  
~~X~~  $\langle 0.85, 0.3, 0.6, 0.7, 0.3 \rangle$

Seller 3  
~~X~~  $\langle 0.7, 0.1, 0.8, 0.1, 0.2 \rangle$

Seller 4  
 $\langle 0.6, 0.2, 0.7, 0.4, 0.1 \rangle$

Item	Time	Insurance	Seller Repu.	A/S support	Cust serv.
	0.2	0.15	0.6	0.0	0.05

$\langle p_1, p_2, p_3, p_4, p_5 \rangle$  represents seller prices along different product attributes



# Sellers' Knowledge

- A seller knows
  - Set of product attributes
  - Purchase decision of buyer
- A seller does not know
  - How many other sellers are there?
  - What prices other sellers are charging?
  - How many buyers are there?
  - What is the preference distribution of buyers?

# Research Question

- How can a seller adjust the prices it charges along different product attributes over time to respond to temporal changes in
  - Buyer demand (Preferences of buyers over different attributes)
  - Competitors' strategies (Prices charged by competing sellers)

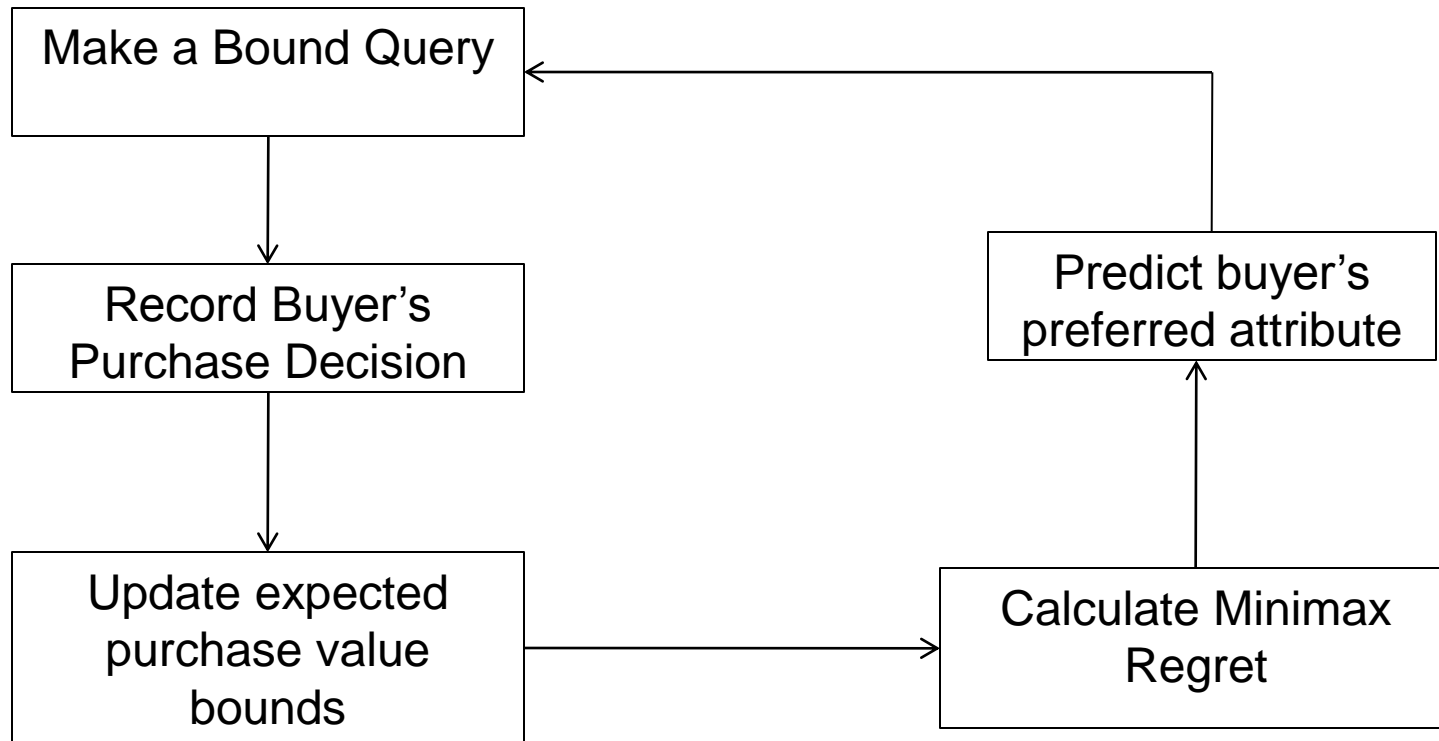
# Minimax Regret-based Attribute Prediction

- Estimate buyer preferences from the buyer's purchase decision
- Minimax regret technique of preference elicitation is used
- Seller makes a decision it would regret the least
  - Which attribute to predict for each buyer

# Minimax Regret-based Attribute Prediction

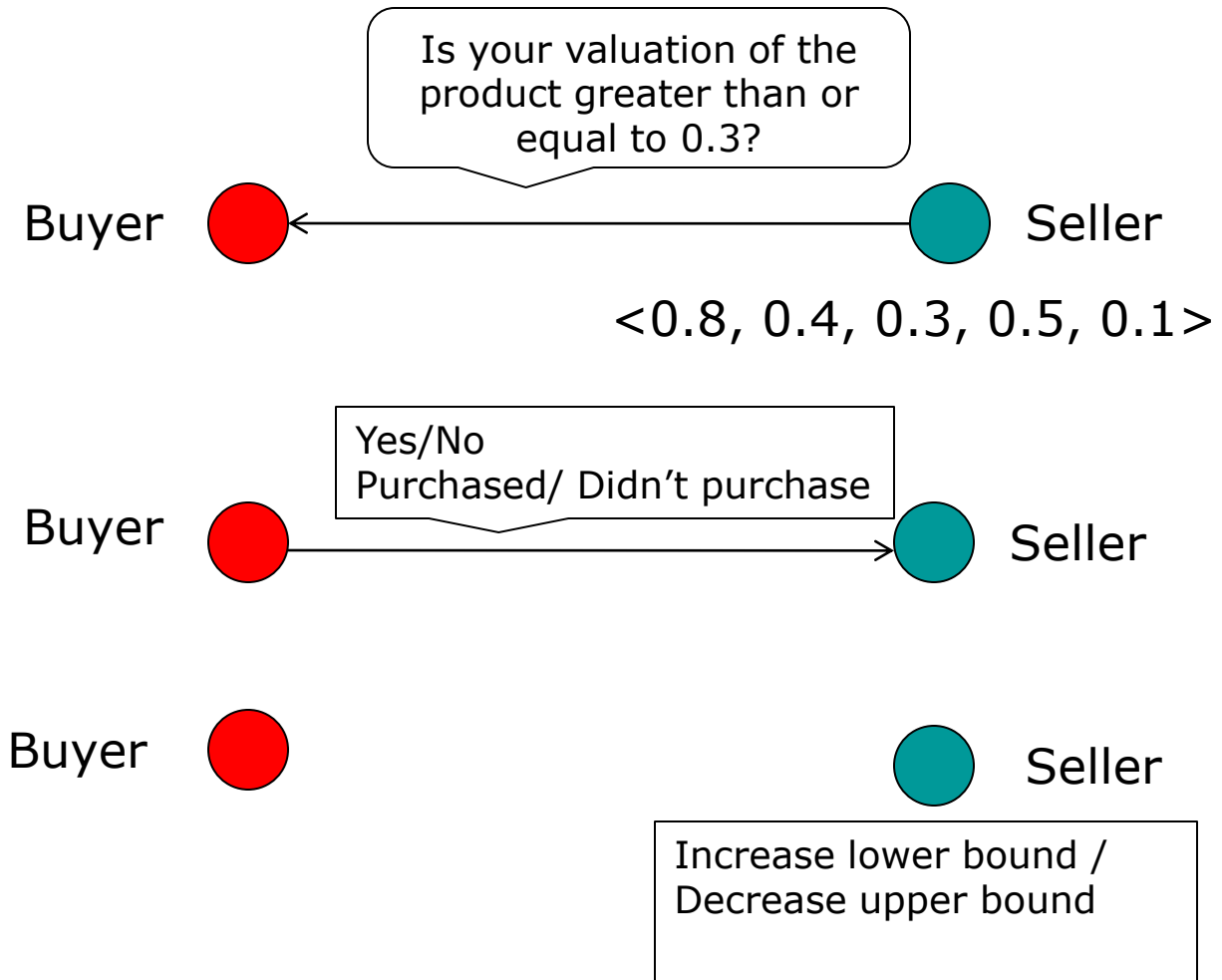
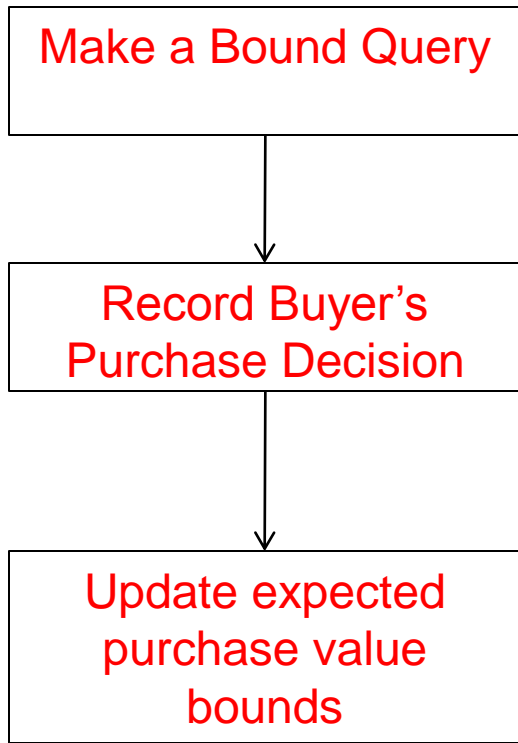
- Sellers keep an upper and lower bounds of each buyer's expected purchase value for each attribute
- Consider buyer-seller interaction as a querying process
- Sellers make an attribute prediction decision at the end of each interval

# Attribute Prediction Process



# Attribute Prediction Process

Assume seller predicted attribute  $a_3$



# Attribute Prediction Process

 Seller

Calculate Minimax  
Regret



Predict buyer's  
preferred attribute

1) Calculate pairwise regret for every attribute

$$R(a_i, a_{-i}) = ub_{a_{-i}} - lb_{a_i}$$

$$R(a_i, a_i) = 0$$

2) Find maximum for each attribute

$$MR_{a_i} = \max R(a_i, a_{-i})$$

3) Choose attribute giving minimum regret

$$a^* = \arg_{a_i} \min MR_{a_i}$$

# Regret-based Dynamic Pricing

- After attribute prediction, sellers calculate profit-maximizing price using:
  - Historical weighted average price
  - Past profits
  - Average bounds on the purchase values across all buyers
  - Normalized number of buyers with preferred attribute  $a_i$  from attribute prediction part



# Regret-based Dynamic Pricing

1) Calculate historical weighted average price over past  $h$  intervals,  $p_{ai}^*$

2) Calculate average regret-based price

$$p_{ai}^{\text{'}} = n_{ai} \cdot ub_{ai} + (1 - n_{ai}) \cdot lb_{ai}$$

3) If the direction of the price movement is the same as the direction of the profit change

$$p_{ai} = \alpha_1 \cdot p_{ai}^{\text{'}} + (1 - \alpha_1) \cdot p_{ai}^*, \text{ with } \alpha_1 > 0.5$$

Otherwise

$$p_{ai} = \text{past\_}p_{ai} + \text{sign} \cdot \varepsilon, \text{ where } \text{sign} \text{ is the sign of the profit difference in the last two intervals}$$

# Simulations

- Number of buyers: 500 or 1000
- Number of sellers: 3 or 5
- Number of product attributes: 5
- Unit production cost: 0.1
- Interval for price updates: 40 quote requests
- Entry price:  $U[0.1, 1]$

# Attribute Prediction

- Buyers randomly select a preference vector upon the entrance to the market
- Buyers change the selected preference vector at different random times
- **Collaborative Filtering** – for comparison
  - Attribute is predicted based on purchase history

# Pricing Comparison Strategies

- Fixed Pricing
  - Price is randomly selected  $U[0.1, 1]$  and is fixed
- Derivative-Follower Pricing
  - Price is determined based on the profits obtained
- Goal-Directed Pricing
  - Price is determined based on the actual and expected number of products sold