

# Preference Elicitation & Recommendation

## Part 2

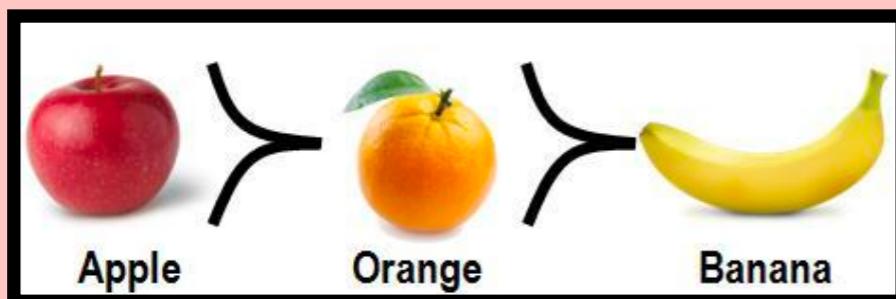
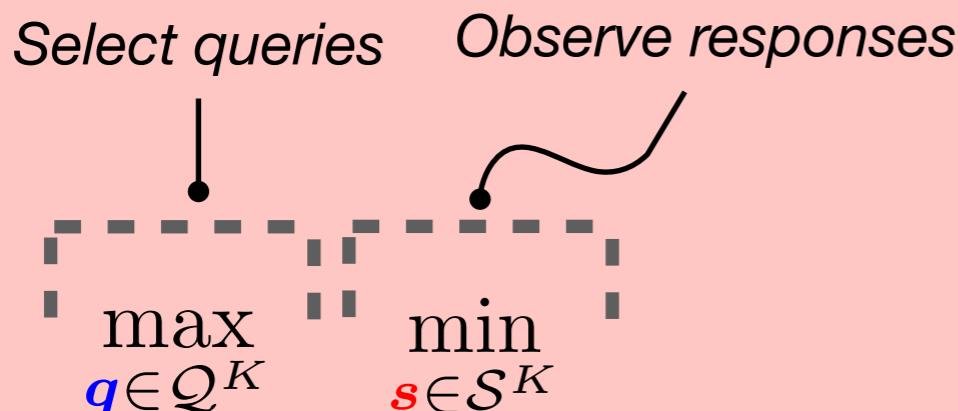
Applied Mechanism Design for Social Good — CMSC828M  
23 April, 2020

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# Elicitation + Recommendation

## Preference Elicitation

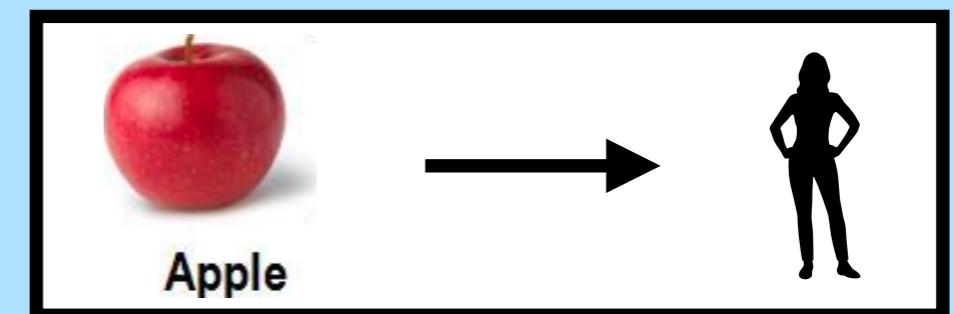
*What does the {agent | customer | user} want?*



## Recommendation

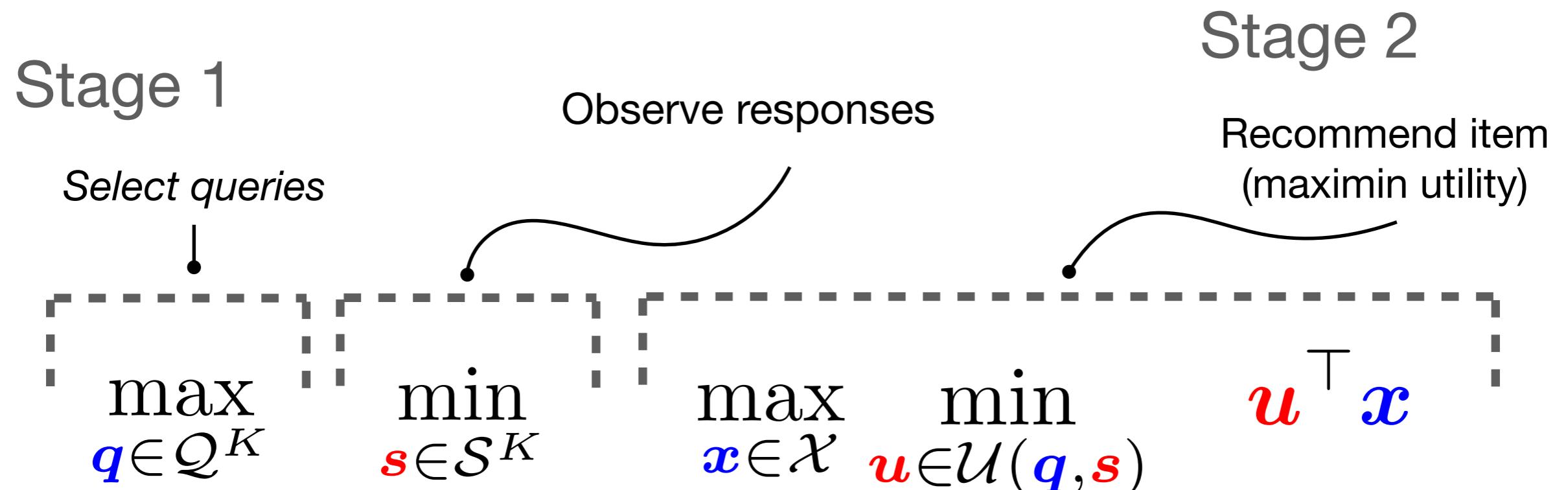
*Which item should we offer?*

$$\mathbf{x}^* \in \arg \max_{\mathbf{x} \in \mathcal{X}} \min_{\mathbf{u} \in \mathcal{U}} \mathbf{u}^\top \mathbf{x}$$



# Elicitation + Recommendation

## Multi-Stage Optimization



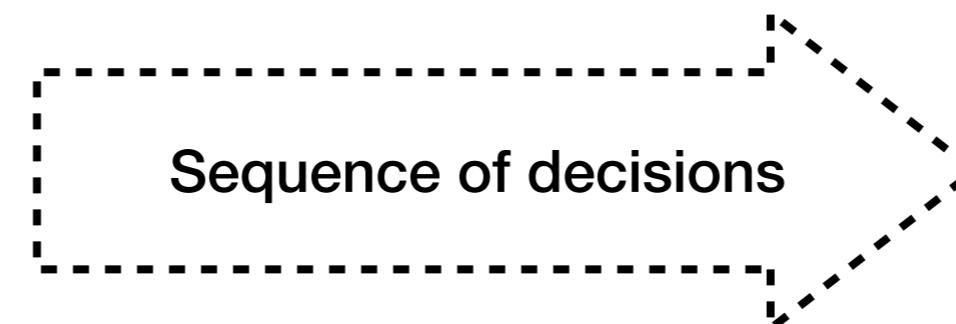
# Elicitation + Recommendation

Multi-Stage Optimization

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

# Elicitation + Recommendation

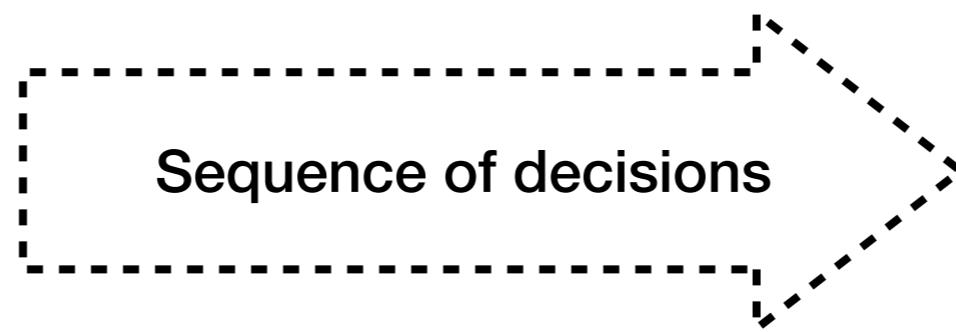
## Multi-Stage Optimization



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# Elicitation + Recommendation

## Multi-Stage Optimization

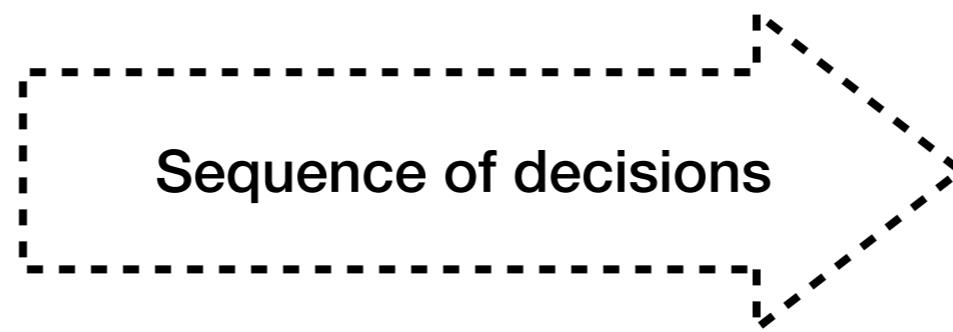


$$\max_{\mathbf{q} \in \mathcal{Q}^K} \left\{ \min_{\mathbf{s} \in \mathcal{S}^K} \max_{\mathbf{x} \in \mathcal{X}} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x} \right\}$$

s.t.  $\mathbf{q}$  is fixed

# Elicitation + Recommendation

## Multi-Stage Optimization



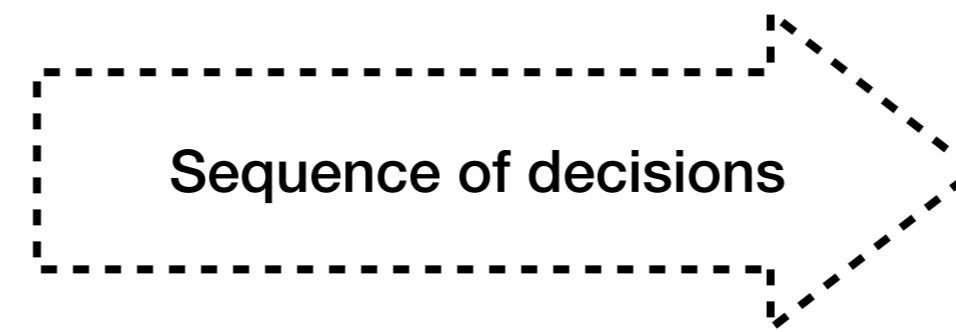
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# Elicitation + Recommendation

## Multi-Stage Optimization



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s.t.  $\mathbf{q}$  is fixed

s.t.  $\mathbf{x}, \mathbf{s}, \mathbf{q}$  are fixed

# Elicitation + Recommendation

## Comment

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \begin{array}{ll} \mathbf{u} \in \mathcal{U}^0 & : \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) > 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ & : \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) = 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 0 \\ & : \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) < 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

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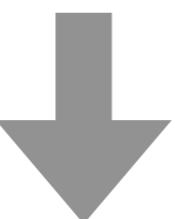
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Indifference response does not impact our optimization problems

# Elicitation + Recommendation

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# Elicitation + Recommendation

Stage 1

Select queries

$$\begin{array}{cccccc} & & \text{Observe responses} & & & \\ \downarrow & & \curvearrowleft & & \curvearrowright & \\ \vdash \max_{\mathbf{q} \in \mathcal{Q}^K} & \vdash \min_{\mathbf{s} \in \mathcal{S}^K} & \vdash \max_{\mathbf{x} \in \mathcal{X}} & \vdash \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} & \mathbf{u}^\top \mathbf{x} & \vdash \end{array}$$

Stage 2

Recommend item  
(maximin utility)

Now, some theory

# Elicitation + Recommendation

Observation

Stage 1

Select queries

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

$$\min_{\mathbf{s} \in \mathcal{S}^K}$$

Observe responses

Stage 2

Recommend item  
(maximin utility)

$$\mathbf{u}^\top \mathbf{x}$$

Suppose we can “design”  
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$$\mathcal{X} = \{\mathbf{x} \in \mathbb{R}^J \mid \mathbf{x}^\top \mathbf{b} \leq c\}$$

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Observation

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Observe responses

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With maximin-utility recommendation, and a  
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### Minimax Theorem: (von Neumann)

- If  $f(\mathbf{x}, \mathbf{y})$  is convex in  $\mathbf{x}$ , concave in  $\mathbf{y}$
- On convex compact sets  $\mathbf{X}$  and  $\mathbf{Y}$ , then:

$$\max_{\mathbf{x} \in \mathcal{X}} \min_{\mathbf{y} \in \mathcal{Y}} f(\mathbf{x}, \mathbf{y}) = \min_{\mathbf{y} \in \mathcal{Y}} \max_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathbf{y})$$

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Convex/concave  
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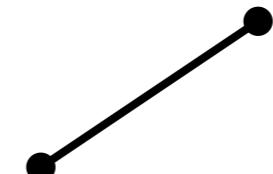
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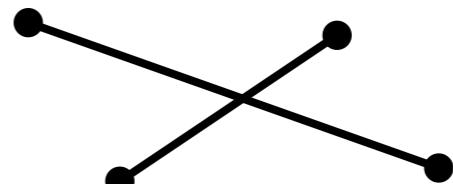
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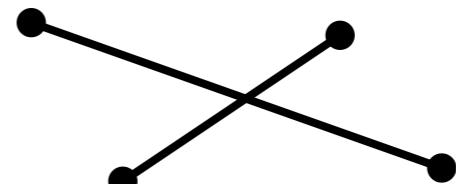
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With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

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Minimizing both  $\mathbf{s} \in \mathcal{S}^K$  and  $\mathbf{u}$  simultaneously means “nature” can choose any  $\mathbf{u}$

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$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$


Minimizing both  $\mathbf{s} \in \mathcal{S}^K$  and  $\mathbf{u}$  simultaneously means “nature” can choose any  $\mathbf{u}$

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

# Elicitation + Recommendation

## Observation (Step 2)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless.**

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \begin{array}{l} \mathbf{u} \in \mathcal{U}^0 : \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

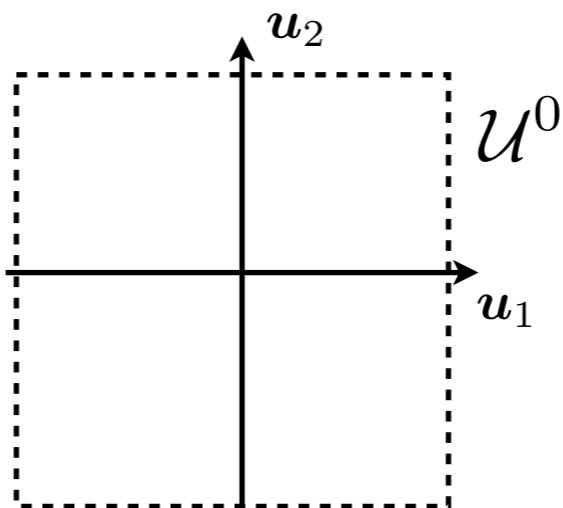
# Elicitation + Recommendation

## Observation (Step 2)

### Observation

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# Elicitation + Recommendation

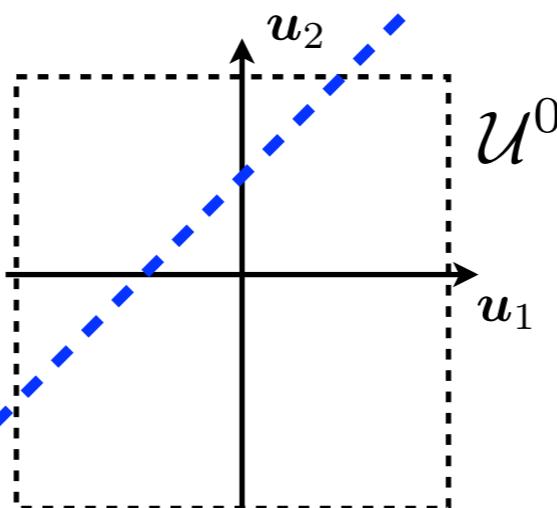
## Observation (Step 2)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

(1)

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$



1) we choose queries

$$\mathbf{x}^{\mathbf{q}_1^A} - \mathbf{x}^{\mathbf{q}_1^B}$$

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

# Elicitation + Recommendation

## Observation (Step 2)

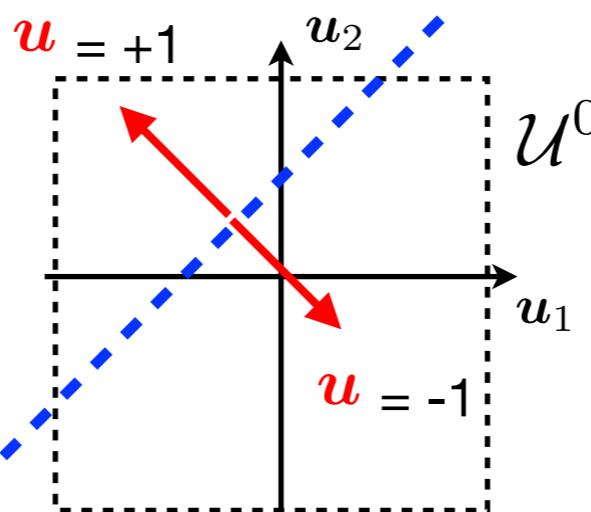
### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\begin{array}{ll} (1) & (2) \\ \max_{\mathbf{q} \in \mathcal{Q}^K} & \min_{\mathbf{s} \in \mathcal{S}^K} \\ & \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \end{array} \quad (\text{Minimax Thm.})$$

1) we choose queries

$$\mathbf{x}^{\mathbf{q}_1^A} - \mathbf{x}^{\mathbf{q}_1^B}$$



2) nature chooses responses

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

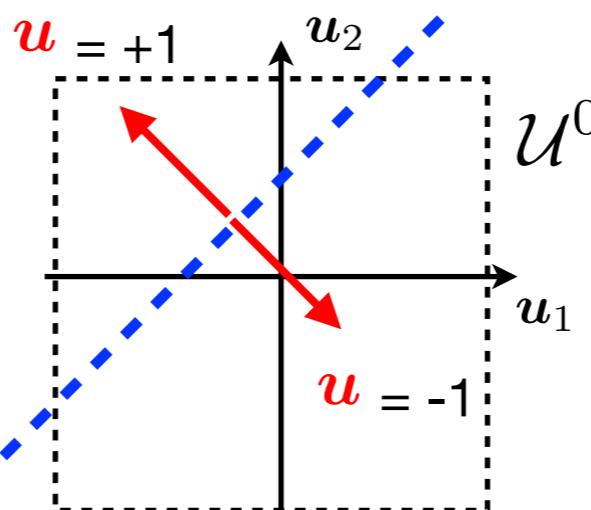
# Elicitation + Recommendation

## Observation (Step 2)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$(1) \quad \max_{\mathbf{q} \in \mathcal{Q}^K} \quad (2) \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad (3) \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$



1) we choose queries

$$\mathbf{x}^{\mathbf{q}_1^A} - \mathbf{x}^{\mathbf{q}_1^B}$$

- 2) nature chooses responses
- 3) nature chooses  $\mathbf{u}$

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

# Elicitation + Recommendation

## Observation (Step 2)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless.**

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$

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# Elicitation + Recommendation

## Observation (Step 2)

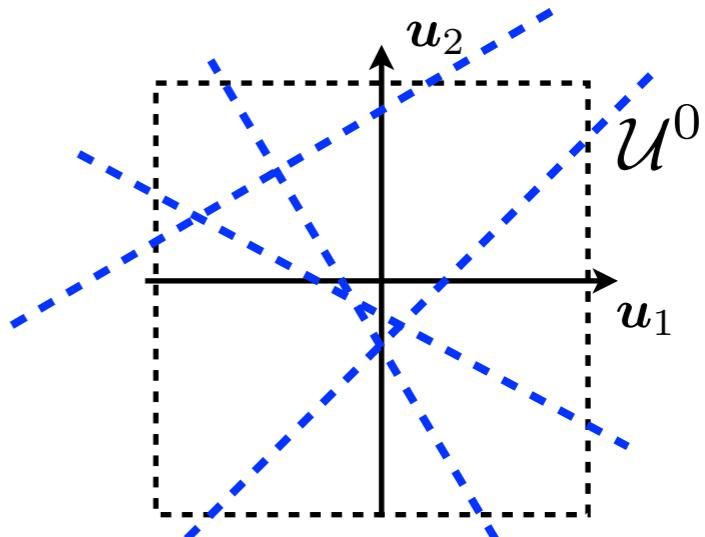
### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$

**Q:** Suppose I ask several **queries**:

Does this restrict **nature** from choosing any  $\mathbf{u} \in \mathcal{U}^0$ ?



$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \begin{array}{l} \mathbf{u} \in \mathcal{U}^0 : \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 \quad \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

# Elicitation + Recommendation

## Observation (Step 2)

### Observation

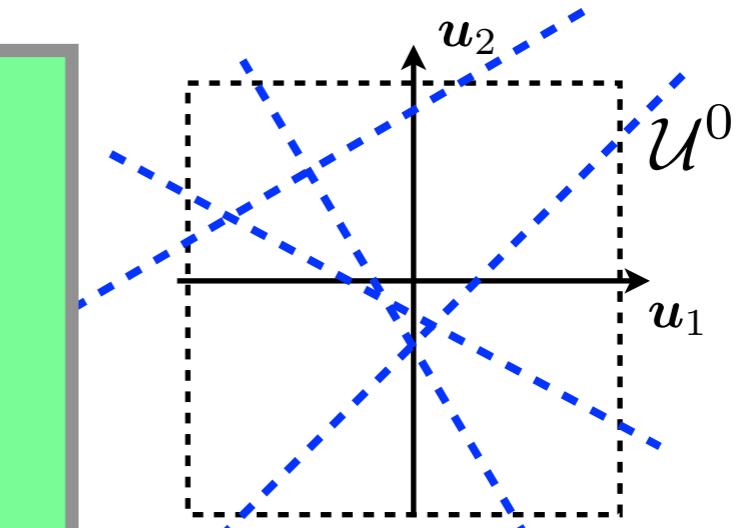
With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \quad (\text{Minimax Thm.})$$

**A: No**, nature can select any

$$\min_{\mathbf{u} \in \bigcup_{\mathbf{s} \in \mathcal{S}^K} \mathcal{U}(\mathbf{q}, \mathbf{s})}$$

Which is equivalent to any  $\mathbf{u} \in \mathcal{U}^0$



$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

# Elicitation + Recommendation

## Observation (Step 2)

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With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x}$$



$$\min_{\mathbf{u} \in \mathcal{U}^0}$$

(Minimax Thm.)  
(Unrestricted  $\mathbf{u}$ )

# Elicitation + Recommendation

## Observation (Step 3)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless.**

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \min_{\mathbf{u} \in \mathcal{U}^0} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x} \quad \begin{array}{l} \text{(Minimax Thm.)} \\ \text{(Unrestricted } \mathbf{u} \text{)} \end{array}$$

# Elicitation + Recommendation

## Observation (Step 3)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\cancel{\mathbf{q} \in \mathcal{Q}^K}} \min_{\mathbf{u} \in \mathcal{U}^0} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x}$$

(Minimax Thm.)  
(Unrestricted  $\mathbf{u}$ )

Queries don't matter

# Elicitation + Recommendation

## Observation (Step 3)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\cancel{\mathbf{q} \in Q^K}} \min_{\mathbf{u} \in \mathcal{U}^0} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x}$$

(Minimax Thm.)  
(Unrestricted  $\mathbf{u}$ )

Queries don't matter

**Q:** What does this mean?

# Elicitation + Recommendation

## Observation (Step 3)

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With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\cancel{\mathbf{q} \in Q^K}} \min_{\mathbf{u} \in \mathcal{U}^0} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x}$$

(Minimax Thm.)  
(Unrestricted  $\mathbf{u}$ )

Queries don't matter

**Q:** What does this mean?

- don't use convex  $\mathcal{X}$  ?

# Elicitation + Recommendation

## Observation (Step 3)

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With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\cancel{\mathbf{q} \in Q^K}} \min_{\mathbf{u} \in \mathcal{U}^0} \max_{\mathbf{x} \in \mathcal{X}} \mathbf{u}^\top \mathbf{x}$$

(Minimax Thm.)  
(Unrestricted  $\mathbf{u}$ )

Queries don't matter

**Q:** What does this mean?

- don't use convex  $\mathcal{X}$  ?
- robustness is too conservative?

# Elicitation + Recommendation

## Observation (Step 3)

### Observation

With maximin-utility recommendation, and a convex item set  $\mathcal{X}$ , **elicitation is useless**.

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

$$\min_{\mathbf{u} \in \mathcal{U}^0}$$

$$\max_{\mathbf{x} \in \mathcal{X}}$$

$$\mathbf{u}^\top \mathbf{x}$$

(Minimax Thm.)

(Unrestricted  $\mathbf{u}$ )

Queries don't matter

$$\min_{\mathbf{s} \in \mathcal{S}^K}$$

$$\min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})}$$

**Q:** What does this mean?

- don't use convex  $\mathcal{X}$  ?
- robustness is too conservative?

# Elicitation + Recommendation

## MILP Reformulation

Stage 1

Select queries

$$\begin{array}{c} \text{Observe responses} \\ \hline \max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x} \end{array}$$

Stage 2

Recommend item  
(maximin utility)

How do we solve  
this problem?

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

Goal:

Express this problem as a finite-size  
**linear program** (with integer and/or  
continuous variables)

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

Goal:

Express this problem as a finite-size **linear program** (with integer and/or continuous variables)

## Reformulation Tricks

1. “Indexing” discrete variables
2. Epigraph formulation
3. Linearization
4. Duality
5. Decomposition

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\begin{array}{c} \max_{\mathbf{q} \in \mathcal{Q}^K} \\ \text{Fixed} \end{array} \quad \begin{array}{l} \min_{\mathbf{s} \in \mathcal{S}^K} \\ \max_{\mathbf{x} \in \mathcal{X}} \\ \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \end{array} \quad \mathbf{u}^\top \mathbf{x}$$

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})}$$

$$\mathbf{u}^\top \mathbf{x}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

Instead, for each possible agent response scenario  $\mathbf{s}$  ...

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})}$$

$$\mathbf{u}^\top \mathbf{x}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

Instead, for each possible agent response scenario  $\mathbf{s}$  ...

we find the optimal item to recommend,  $\mathbf{x}^{\mathbf{s}}$  ...

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}}$$

$$\min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

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Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}}$$

$$\min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

$$\left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

Instead, for each possible agent response scenario  $\mathbf{s}$  ...

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# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

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Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

Instead, for each possible agent response scenario  $\mathbf{s}$  ...

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# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})}$$

~~$$\mathbf{u}^\top \mathbf{x}$$~~

$$\left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\}$$

$$\mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

Suppose we *don't wait for nature* to select  $\mathbf{s}$  !

Instead, for each possible agent response scenario  $\mathbf{s}$  ...

we find the optimal item to recommend,  $\mathbf{x}^{\mathbf{s}}$  ...

Aside:

In LPs, discrete vars ( $\mathbf{s}$ ) are tricky, so we remove them when possible.

One way to do this is “**indexing**”— rolling out all possible values.

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in \mathcal{Q}^K}$$

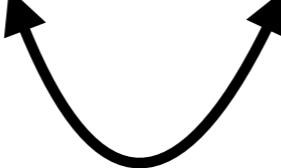
Fixed

$$\min_{\mathbf{s} \in \mathcal{S}^K} \left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

Now, the choice of  $\mathbf{s}$  doesn't impact the choice of each item!

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\begin{array}{c} \max_{\mathbf{q} \in Q^K} \\ \text{Fixed} \end{array} \quad \min_{\mathbf{s} \in S^K} \left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in S^K \end{array} \right\} \quad \min_{\mathbf{u} \in U(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$


Now, the choice of  $\mathbf{s}$  doesn't impact the choice of each item!

# Elicitation-Recommendation Problem

Reformulation (1): Index by  $\mathbf{s}$

$$\max_{\mathbf{q} \in Q^K}$$

Fixed

$$\left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\} \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$
$$\max_{\mathbf{q} \in \mathcal{Q}^K} \left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\} \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

Goal:

Express this problem as a finite-size **linear program** (with integer and/or continuous variables)

## Reformulation Tricks

1. “~~Indexing~~” discrete variables
2. Epigraph formulation
3. Linearization
4. Duality
5. Decomposition

# Elicitation-Recommendation Problem

## Reformulation (2): Epigraph Form

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \left\{ \begin{array}{l} \max_{\mathbf{x}^{\mathbf{s}_1} \in \mathcal{X}} \\ \max_{\mathbf{x}^{\mathbf{s}_2} \in \mathcal{X}} \\ \vdots \\ \forall \mathbf{s} \in \mathcal{S}^K \end{array} \right\} \min_{\mathbf{s} \in \mathcal{S}^K} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}}$$

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**Q:** How can we simplify?

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- Can we select  $\mathbf{q}$  and  $\mathbf{x}^{\mathbf{s}}$  simultaneously?

**A: Yes.** Why?

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**Epigraph Formulation** of a problem  
using aux. variable ( $\mathcal{T}$ )

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$$\max_{z \in \mathcal{Z}} f(z)$$

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$$\begin{array}{ll} \max & \tau \\ \tau \in \mathbb{R}, z \in \mathcal{Z} & \\ \tau \leq f(z) & \end{array}$$

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New Aux.  
variable!



$$\tau \in \mathbb{R}$$

**Epigraph Formulation** of a problem  
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$$\max_{\mathbf{z} \in \mathcal{Z}} f(\mathbf{z}) \quad \leftrightarrow \quad \max_{\tau \in \mathbb{R}, \mathbf{z} \in \mathcal{Z}} \tau \quad \text{subject to} \quad \tau \leq f(\mathbf{z})$$

# Elicitation-Recommendation Problem

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# Elicitation-Recommendation Problem

## Reformulation

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Goal:

Express this problem as a finite-size **linear program** (with integer and/or continuous variables)

## Reformulation Tricks

1. “~~Indexing~~” discrete variables
2. ~~Epigraph formulation~~
3. Linearization
4. Duality
5. Decomposition

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

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$$\tau \in \mathbb{R}$$

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## Reformulation Tricks

Q: How do we solve this?

“indexing” discrete variables

bigraph formulation

linearization

4. Duality

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# Elicitation-Recommendation Problem

## Reformulation

$$\begin{array}{ll} \max_{\mathbf{q} \in \mathcal{Q}^K} & \min_{\mathbf{s} \in \mathcal{S}^K} \\ & \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{q} \in \mathcal{Q}^K, \\ & \mathbf{x}^{\mathbf{s}} \in \mathcal{X} : \forall \mathbf{s} \in \mathcal{S}^K \\ & \tau \in \mathbb{R} \\ & \tau \leq \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} \end{array}$$

## Reformulation Tricks

Q: How do we solve this?

- Can we even write this problem down (as a MILP)?

“indexing” discrete variables  
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## Reformulation

$$\begin{array}{ll} \max_{\mathbf{q} \in \mathcal{Q}^K} & \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{q} \in \mathcal{Q}^K, \\ & \mathbf{x}^{\mathbf{s}} \in \mathcal{X} : \forall \mathbf{s} \in \mathcal{S}^K \\ & \tau \in \mathbb{R} \\ & \tau \leq \min_{\mathbf{s} \in \mathcal{S}^K} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} \end{array}$$

## Reformulation Tricks

**Q:** How do we solve this?

- Can we even write this problem down (as a MILP)?

“indexing” discrete variables

bigraph formulation

lazierization

**A: No... ( $\mathbf{s}!!$ )**

# Elicitation-Recommendation Problem

Reformulation (3): Index by  $\textcolor{red}{s}$ ... again!

$$\begin{aligned} & \max \quad \tau \\ \text{s.t.} \quad & \textcolor{blue}{q} \in \mathcal{Q}^K, \\ & \textcolor{blue}{x}^{\textcolor{red}{s}} \in \mathcal{X} : \forall \textcolor{red}{s} \in \mathcal{S}^K \\ & \tau \in \mathbb{R} \\ & \tau \leq \min_{\textcolor{red}{s} \in \mathcal{S}^K} \quad \min_{\textcolor{red}{u} \in \mathcal{U}(\textcolor{blue}{q}, \textcolor{red}{s})} \quad \textcolor{red}{u}^\top \textcolor{blue}{x}^{\textcolor{red}{s}} \end{aligned}$$

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**Aside:**

In LPs, discrete vars ( $\mathbf{S}$ ) are tricky, so we remove them when possible.

One way to do this is  
**“indexing”**— rolling out all possible values.

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## Linearizing a Min constraint

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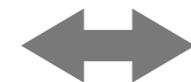
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Comment:

Constraints/vars group naturally by  $\textcolor{red}{s}$

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Reformulation (4): Index by  $\textcolor{red}{s}$ ... again!

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# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\mathbf{s}$ ... again!

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The only remaining problem is the minimization over  $\mathbf{u}$ .

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\mathbf{s}$ ... again!

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The only remaining problem is the minimization over  $\mathbf{u}$ .

**Q:** How do we get rid of this minimization?

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\mathbf{s}$ ... again!

$$\begin{array}{ll} \max & \tau \\ \text{s.t.} & \mathbf{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \mathbf{x}^{\mathbf{s}} \in \mathcal{X} \\ \tau \leq \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} \end{array} \right\} \forall \mathbf{s} \in \mathcal{S}^K \end{array}$$

The only remaining problem is the minimization over  $\mathbf{u}$ .

**Q:** How do we get rid of this minimization?  
... what would happen if this *min* became a *max*?

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The only remaining problem is the minimization over  $\mathbf{u}$ .

Primal & Dual Linear Programs

$$\begin{array}{ll} \min & \mathbf{c}^\top \mathbf{x} \\ \mathbf{x} \in \mathbb{R}^N, \quad & \mathbf{A}\mathbf{x} \geq \mathbf{b} \end{array}$$

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The only remaining problem is the minimization over  $\mathbf{u}$ .

Primal & Dual Linear Programs

$$\min_{\mathbf{x} \in \mathbb{R}^N, \mathbf{A}\mathbf{x} \geq \mathbf{b}} \mathbf{c}^\top \mathbf{x} \quad \longleftrightarrow \quad \max_{\mathbf{y} \in \mathbb{R}^M, \mathbf{A}^\top \mathbf{y} \leq \mathbf{c}} \mathbf{b}^\top \mathbf{y}$$

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$$\min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} \quad \text{"primal"}$$

Primal & Dual Linear Programs

$$\min_{\mathbf{x} \in \mathbb{R}^N, \mathbf{A}\mathbf{x} \geq \mathbf{b}} \mathbf{c}^\top \mathbf{x} \quad \longleftrightarrow \quad \max_{\mathbf{y} \in \mathbb{R}^M, \mathbf{A}^\top \mathbf{y} \leq \mathbf{c}} \mathbf{b}^\top \mathbf{y}$$

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Primal & Dual Linear Programs

$$\mathcal{U}(\mathbf{q}, \mathbf{s}) := \left\{ \mathbf{u} \in \mathcal{U}^0 : \begin{array}{ll} \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \geq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = 1 \\ \mathbf{u}^\top (\mathbf{x}^{\mathbf{q}_\kappa^A} - \mathbf{x}^{\mathbf{q}_\kappa^B}) \leq 0 & \forall \kappa = 1, \dots, K : \mathbf{s}_\kappa = -1 \end{array} \right\}$$

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$$\begin{array}{ll} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} & \text{"primal"} \\ \updownarrow \\ \max_{\mathbf{y} \in \mathcal{Y}} \mathbf{y}^\top \mathbf{b} & \text{"dual"} \end{array}$$

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**Q:** How many different primal/  
dual problems are there?

# Elicitation-Recommendation Problem

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$$\begin{array}{ccc} \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \mathbf{u}^\top \mathbf{x}^{\mathbf{s}} & \text{“primal”} \\ \Updownarrow \\ \max_{\mathbf{y} \in \mathcal{Y}} \mathbf{y}^\top \mathbf{b} & \text{“dual”} \end{array}$$

**Q:** How many different primal/dual problems are there?

**A: One for each  $\mathbf{s} \in \mathcal{S}^K$ ...**

Because we “indexed” over these variables, we need to create different variables for each.

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\mathbf{s}$ ... again!

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**Q:** How many different primal/dual problems are there?

**A: One for each  $\mathbf{s} \in \mathcal{S}^K$ ...**

Because we “indexed” over these variables, we need to create different variables for each.

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\textcolor{red}{s}$ ... again!

$$\begin{aligned} & \max \quad \tau \\ \text{s.t.} \quad & \textcolor{blue}{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \textcolor{blue}{x}^{\textcolor{red}{s}} \in \mathcal{X} \\ \tau \leq \max_{y^{\textcolor{red}{s}} \in \mathcal{Y}(s, \textcolor{blue}{x}^{\textcolor{red}{s}}, \textcolor{blue}{q})} (\boldsymbol{b}^{\textcolor{red}{s}})^{\top} \boldsymbol{y}^{\textcolor{red}{s}} \end{array} \right\} \quad \forall s \in \mathcal{S}^K \end{aligned}$$

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\textcolor{red}{s}$ ... again!

$$\begin{aligned} & \max \quad \tau \\ \text{s.t.} \quad & \mathbf{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \mathbf{x}^{\textcolor{red}{s}} \in \mathcal{X} \\ \tau \leq \max_{\mathbf{y}^{\textcolor{red}{s}} \in \mathcal{Y}(\textcolor{red}{s}, \mathbf{x}^{\textcolor{red}{s}}, \mathbf{q})} (\mathbf{b}^{\textcolor{red}{s}})^{\top} \mathbf{y}^{\textcolor{red}{s}} \end{array} \right\} \quad \forall \textcolor{red}{s} \in \mathcal{S}^K \end{aligned}$$



**Q:** How do we get rid of this?

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\mathbf{s}$ ... again!

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**Q:** How do we get rid of this?

**A:** Add  $\mathbf{y}^{\mathbf{s}} \in \mathcal{Y}(\mathbf{s}, \mathbf{x}^{\mathbf{s}}, \mathbf{q})$  to the domain...

# Elicitation-Recommendation Problem

Reformulation (4): Index by  $\textcolor{red}{s}$ ... again!

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# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

$$\begin{aligned} & \max \quad \tau \\ \text{s.t.} \quad & \mathbf{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \mathbf{x}^{\mathbf{s}} \in \mathcal{X} \\ y^{\mathbf{s}} \in \mathcal{Y}(\mathbf{s}, \mathbf{x}^{\mathbf{s}}, \mathbf{q}) \\ \tau \leq (\mathbf{b}^{\mathbf{s}})^\top \mathbf{y}^{\mathbf{s}} \end{array} \right\} \quad \forall \mathbf{s} \in \mathcal{S}^K \end{aligned}$$

Goal:

Express this problem as a finite-size **linear program** (with integer and/or continuous variables)

## Reformulation Tricks

1. “~~Indexing~~” discrete variables
2. ~~Epigraph formulation~~
3. ~~Linearization~~
4. ~~Duality~~
5. ~~Decomposition~~

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

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- Can we solve this yet???

# Elicitation-Recommendation Problem

## Reformulation

$$\max_{\mathbf{q} \in \mathcal{Q}^K} \quad \min_{\mathbf{s} \in \mathcal{S}^K} \quad \max_{\mathbf{x} \in \mathcal{X}} \quad \min_{\mathbf{u} \in \mathcal{U}(\mathbf{q}, \mathbf{s})} \quad \mathbf{u}^\top \mathbf{x}$$

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- Can we solve this yet???
- The number of response scenarios  $\mathbf{s} \in \mathcal{S}^K$  can be huge

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## Reformulation

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- Can we solve this yet???
- The number of response scenarios  $\mathbf{s} \in \mathcal{S}^K$  can be huge
- We address this with **decomposition**

# Elicitation-Recommendation Problem

## Decomposition

$$\begin{aligned} & \max \quad \tau \\ \text{s.t.} \quad & \mathbf{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \mathbf{x}^{\mathbf{s}} \in \mathcal{X} \\ \mathbf{y}^{\mathbf{s}} \in \mathcal{Y}(\mathbf{s}, \mathbf{x}^{\mathbf{s}}, \mathbf{q}) \\ \tau \leq (\mathbf{b}^{\mathbf{s}})^{\top} \mathbf{y}^{\mathbf{s}} \end{array} \right\} \forall \mathbf{s} \in \mathcal{S}^K \end{aligned}$$

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## Decomposition

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General Idea:

# Elicitation-Recommendation Problem

## Decomposition

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General Idea:

1. start with only a few scenarios:  
 $\mathcal{S}' := \{\mathbf{s}_1, \mathbf{s}_2, \mathbf{s}_3\}$

# Elicitation-Recommendation Problem

## Decomposition

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**Q:** When is a feasible solution to the **reduced problem** also a feasible solution to the “master” problem?

General Idea:

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# Elicitation-Recommendation Problem

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4. **If No:** we’re optimal!

# Elicitation-Recommendation Problem

## Decomposition

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4. **If No:** we're optimal!

# Elicitation-Recommendation Problem

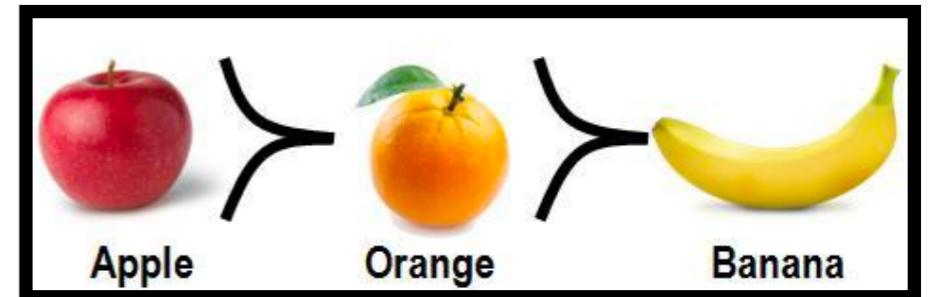
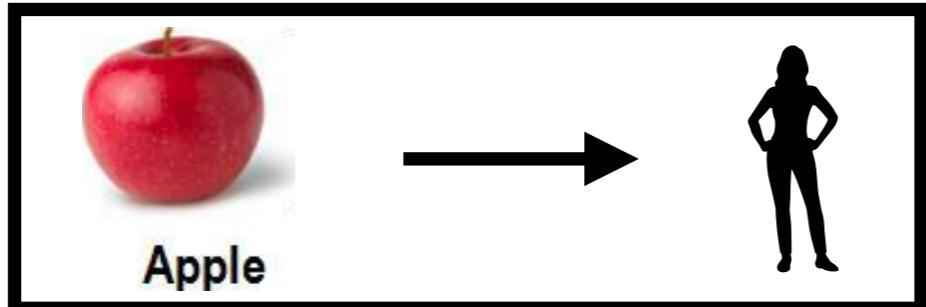
## Decomposition

$$\begin{aligned} \max \quad & \tau \\ \text{s.t.} \quad & \mathbf{q} \in \mathcal{Q}^K, \\ & \tau \in \mathbb{R} \\ & \left\{ \begin{array}{l} \mathbf{x}^{\mathbf{s}} \in \mathcal{X} \\ \mathbf{y}^{\mathbf{s}} \in \mathcal{Y}(\mathbf{s}, \mathbf{x}^{\mathbf{s}}, \mathbf{q}) \\ \tau \leq (\mathbf{b}^{\mathbf{s}})^{\top} \mathbf{y}^{\mathbf{s}} \end{array} \right\} \quad \begin{array}{l} \forall \mathbf{s} \in \mathcal{S}^K \\ \forall \mathbf{s} \in \mathcal{S}' \end{array} \end{aligned}$$

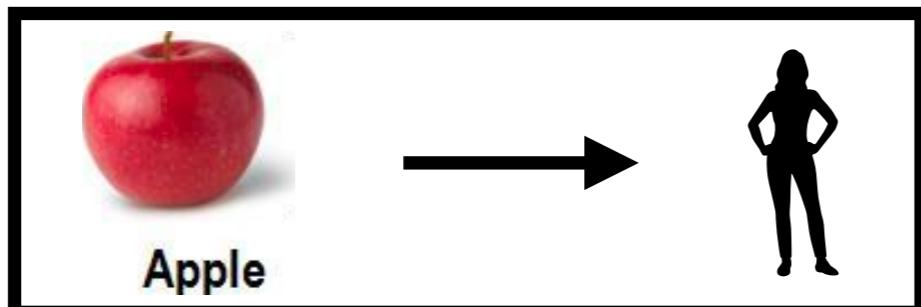
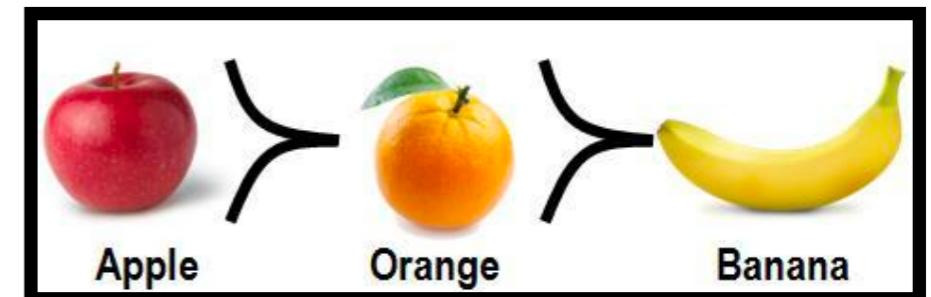
General Idea:

1. start with only a few scenarios:  
 $\mathcal{S}' := \{\mathbf{s}_1, \mathbf{s}_2, \mathbf{s}_3\}$
2. solve this **reduced problem**
3. Are there any violated **constraints** for  $\forall \mathbf{s} \in \mathcal{S}^K$  (is  $\tau$  too large?)
4. **If No:** we're optimal!
5. **If Yes:** Add some violated scenarios to  $\mathcal{S}'$ , repeat

# What's Next?

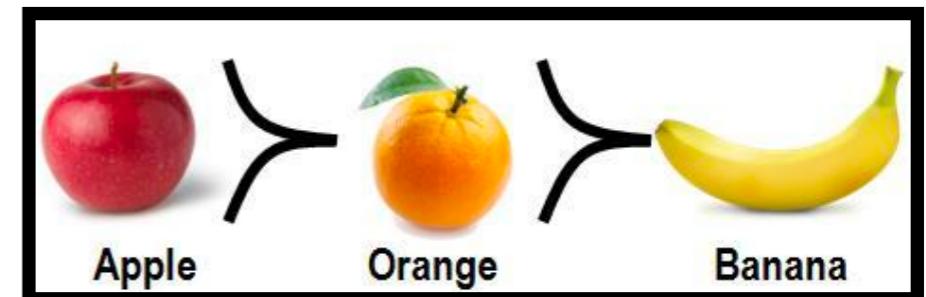


# Potential Research Questions

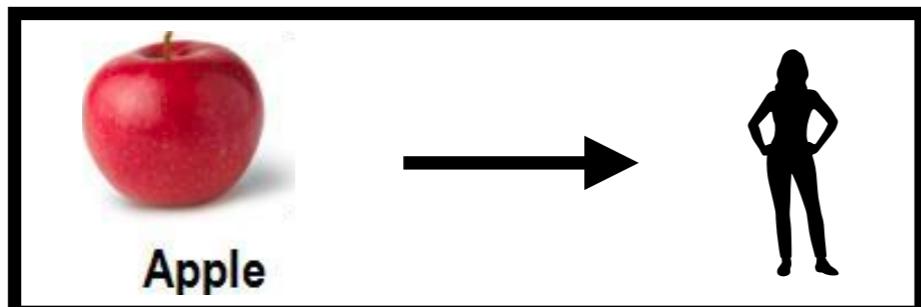


# Potential Research Questions

Theory

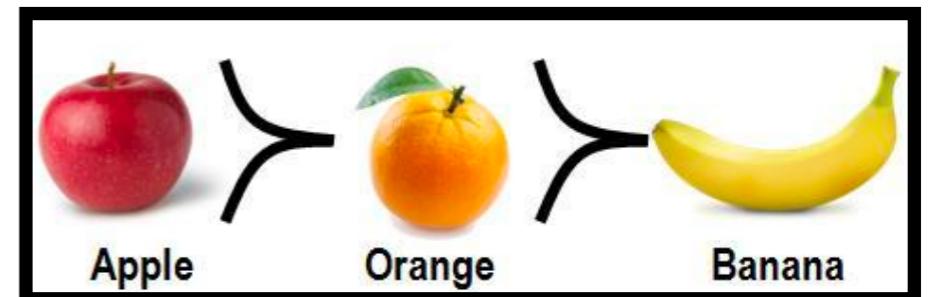
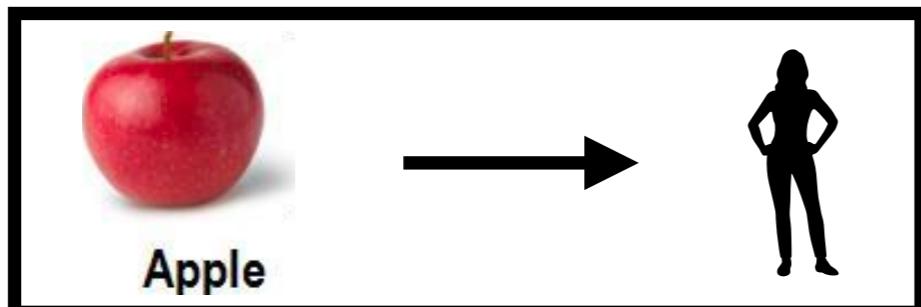


Application



# Potential Research Questions

## Theory

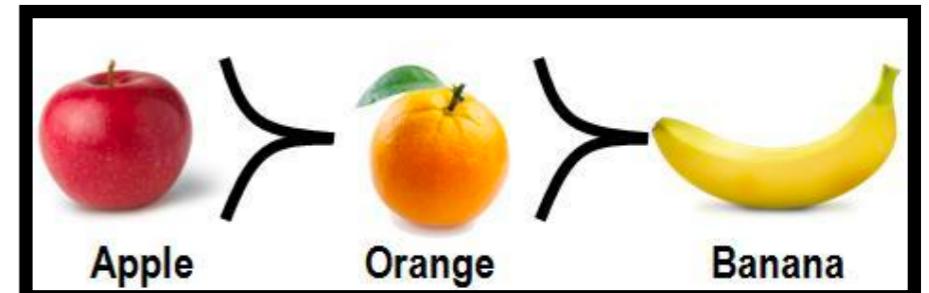
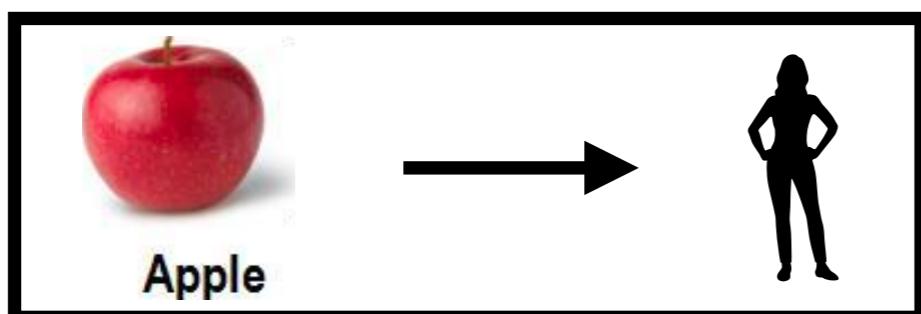


## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)

# Potential Research Questions

## Theory

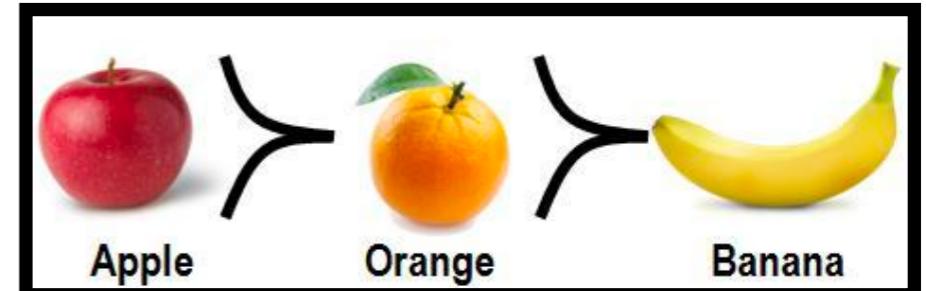
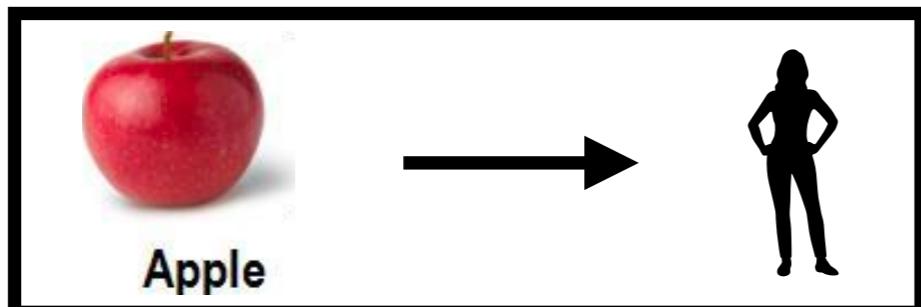


## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)
  
- Q2.** Do our **preference assumptions** hold?
  - transitivity
  - consistency

# Potential Research Questions

## Theory



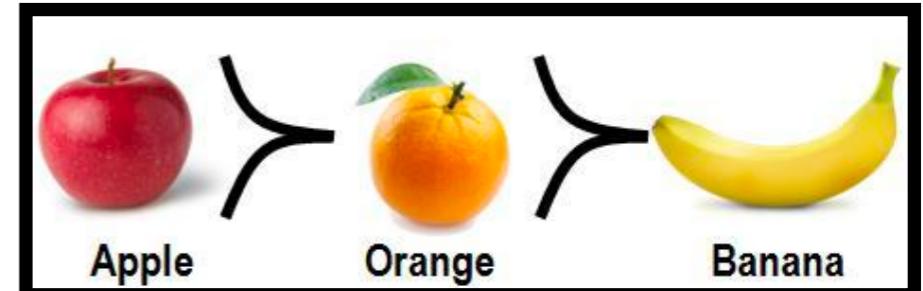
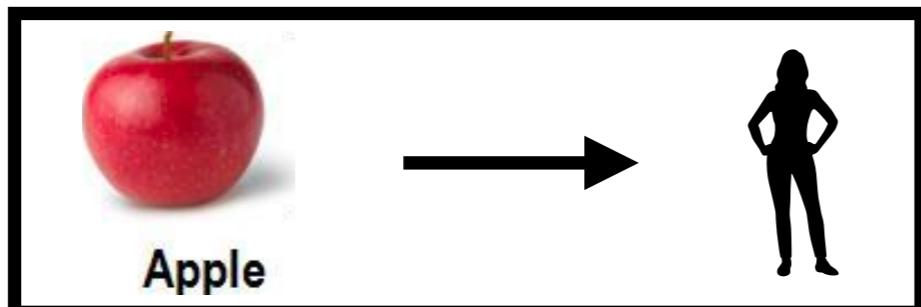
## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)
- Q2.** Do our **preference assumptions** hold?
  - transitivity
  - consistency
- Q3.** Join elicitation with a **decision process**?  
(hiring, resource allocation)

# Potential Research Questions

## Theory

- Q4.** **How many queries** are needed for an “optimal” recommendation?  
(can we ever guarantee rec. quality?)



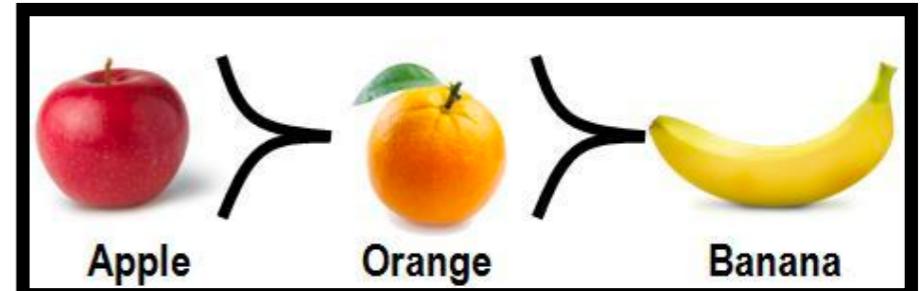
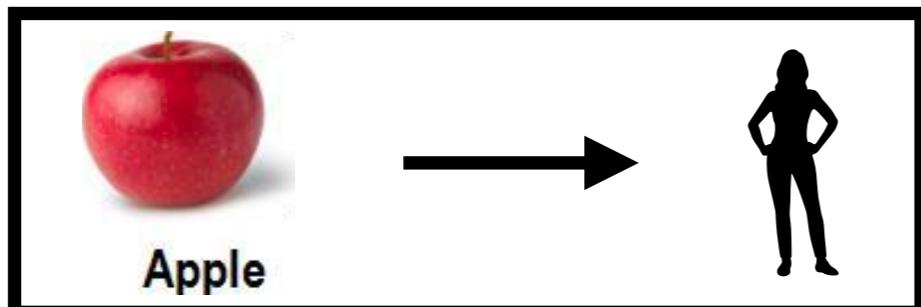
## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)
- Q2.** Do our **preference assumptions** hold?  
- transitivity  
- consistency
- Q3.** Join elicitation with a **decision process**?  
(hiring, resource allocation)

# Potential Research Questions

## Theory

- Q4.** **How many queries** are needed for an “optimal” recommendation?  
(can we ever guarantee rec. quality?)
- Q5.** **Tradeoffs!** Between...  
- utility function complexity & accuracy  
- query complexity and # required queries



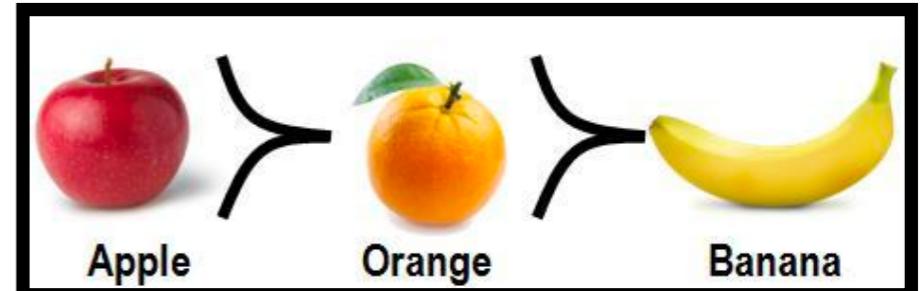
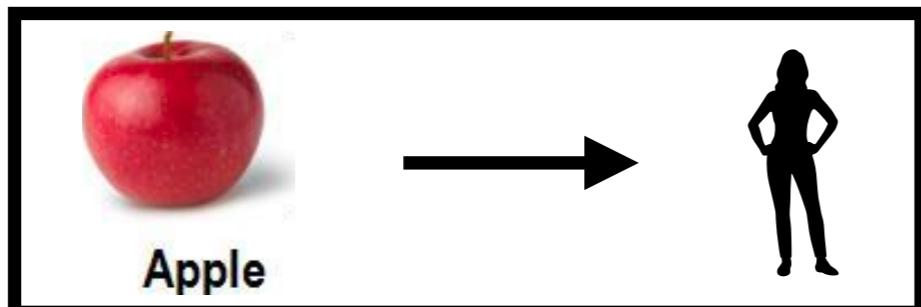
## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)
- Q2.** Do our **preference assumptions** hold?  
- transitivity  
- consistency
- Q3.** Join elicitation with a **decision process**?  
(hiring, resource allocation)

# Potential Research Questions

## Theory

- Q4.** **How many queries** are needed for an “optimal” recommendation?  
(can we ever guarantee rec. quality?)
- Q5.** **Tradeoffs!** Between...
  - utility function complexity & accuracy
  - query complexity and # required queries
- Q6.** What if we have a **group** of agents?  
(cc: social choice?)



## Application

- Q1.** Can we learn preferences in a **lab setting**?  
(prefs for classes? brands? music?)
- Q2.** Do our **preference assumptions** hold?
  - transitivity
  - consistency
- Q3.** Join elicitation with a **decision process**?  
(hiring, resource allocation)