

Preference robust optimization for decision making under uncertainty

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(joint work with Benjamin Armbruster** and Jonathan Y. Li***)

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du Canada

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The Diet Selection Problem

(Bertsimas & O'hair, 2013)

- 8.3% of U.S. population (25.8 million) have diabetes which can have serious complications
- Type II diabetes is known to be related to obesity
- Dietary change is among the most effective ways of preventing/controlling the disease
- A study made in the UK showed that only 20% of dieters last more than one month

Nutrition Facts
Serving Size 2/3 cup (55g)
Servings Per Container About 8

Amount Per Serving	
Calories 230	Calories from Fat 40
	% Daily Value*
Total Fat 8g	12%
Saturated Fat 1g	5%
Trans Fat 0g	
Cholesterol 0mg	0%
Sodium 160mg	7%
Total Carbohydrate 37g	12%
Dietary Fiber 4g	16%
Sugars 1g	
Protein 3g	
Vitamin A	10%
Vitamin C	8%
Calcium	20%
Iron	45%

* Percent Daily Values are based on a 2,000 calorie diet.
Your daily value may be higher or lower depending on your calorie needs.

	Calories:	2,000	2,500
Total Fat	Less than	65g	80g
Sat Fat	Less than	20g	25g
Cholesterol	Less than	300mg	300mg
Sodium	Less than	2,400mg	2,400mg
Total Carbohydrate		300g	375g
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How can we identify optimal diets?

RECIPES

MAIN DISHES APPETIZERS DESSERTS INGREDIENTS

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Barley and quinoa Beef Fish Fondue Eggs Legumes Risotto

The Portfolio Selection Problem

- An individual meets with his financial advisor to tell him he wishes to invest in a given industrial sector, country, etc.
- Since uncertain factors affect performance, a « good » portfolio is one where the risks of losses are best justified by the potential gains



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How can we identify optimal investments?



Why are these decisions difficult to make?

- Something important is at stake
 - Someone is held accountable for the decision
 - The performance measure is multi-dimensional
 - The alternatives are numerous
-
- Numerical optimization can only help once the decision maker's subjective preferences have been fully characterized.

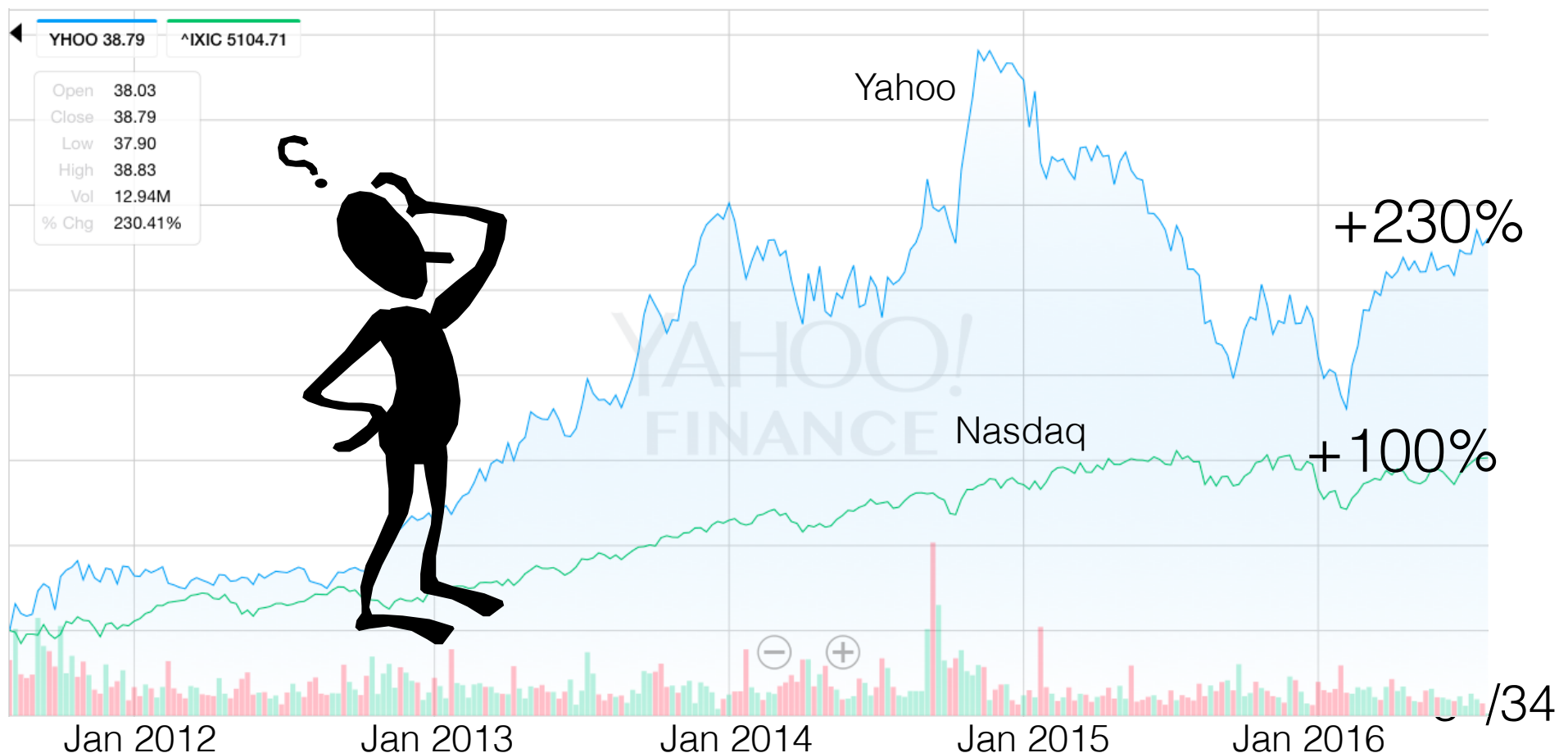
How can we elicit subjective preferences?

- An individual that wishes to control his diet can make pairwise comparisons of meals that he prefers



How can we elicit subjective preferences?

- An investor can indicate what type of wealth evolution he is comfortable with



The strength of utility theory

- In 1954, G. Debreu established that if the preference relation is complete, transitive, and continuous, then there exists a utility mapping such that

$$X \succeq Y \Leftrightarrow u(X) \geq u(Y)$$

where $X, Y \in \mathbb{R}^M$ describe two alternatives

- This implies that any such preference relation can be numerically optimized

$$\max_{x \in \mathcal{X}} u(Z(x))$$

The limitations of utility theory

- Issue #1: One cannot sit down and make all possible comparisons
(e.g. 4657 recipes on ricardocuisine.com makes).



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- Solutions :
 - Make simplifying assumptions about the structure of $u()$ in order to allow interpolation
 - Employ a scheme that handles uncertainty about $u()$

The limitations of utility theory

- Issue #2: One can easily provide false information about his preferences. (Kahneman & Tversky, 1979)

Example (adapted from Kahneman & Tversky, 1979)

We are expecting a new outbreak of the flu virus H1N1 next winter in Canada. Experts believe that 5000 people will die if nothing is done.

Which program should be adopted ?

Drug A (well studied): 4000 among those at risk would be saved

Drug B (experimental): 80% chance that everyone is saved, 20%
that none are saved

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Drug C : 1000 people would end up dying

Drug D : 20% chance that all those at risk die, 80% that none of them die

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Drug C (**same as A**): 1000 people would end up dying

Drug D (**same as B**): 20% chance that all those at risk die, 80%
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Many select Drug A & D but do not mean to be inconsistent.

The limitations of utility theory

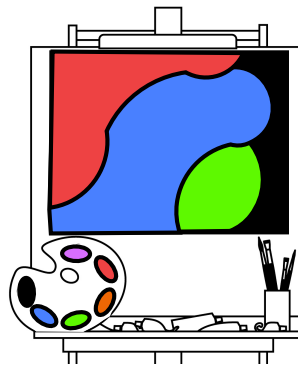
- Issue #2: One can easily provide false information about his preferences. (Kahneman & Tversky, 1979)
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Uncertainty models for preferences

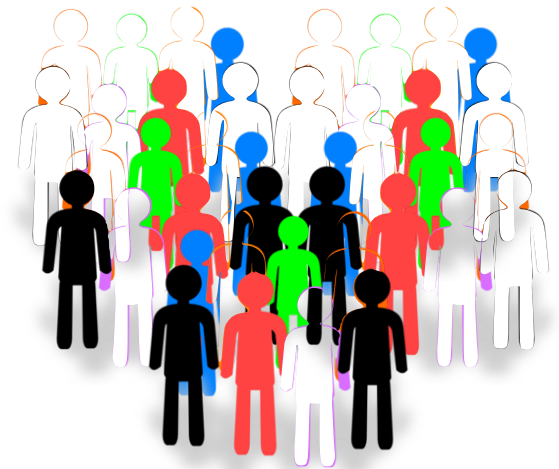
- Stochastic uncertainty: $u \sim F_u$
(e.g. Chajewska et al. 2000)



$$u \sim F_u$$



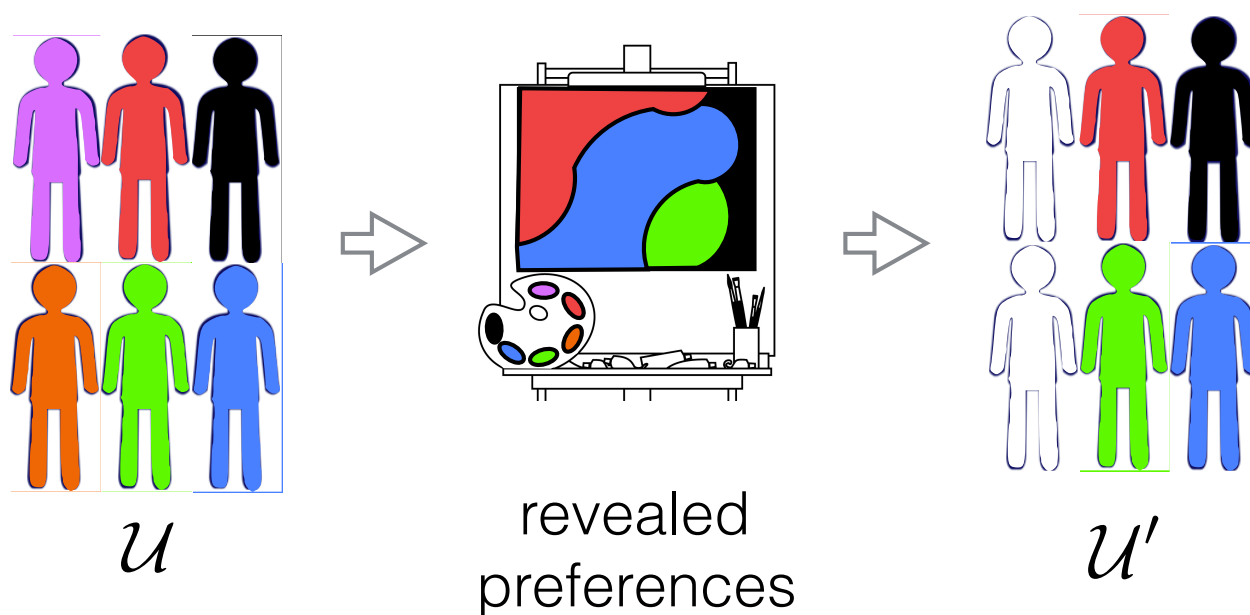
revealed
preferences



$$u \sim F'_u$$

Uncertainty models for preferences

- Stochastic uncertainty: $u \sim F_u$
(e.g. Chajewska et al. 2000)
- Knightian uncertainty: $u \in \mathcal{U}$
(e.g. Hadar & Russell, 1969; Boutilier et al. 2006; Evren & Ok, 2011)



Preference robust optimization

- Max-min utility (Hu & Mehrotra, 2015):

$$\underset{x \in \mathcal{X}}{\text{maximize}} \quad \min_{u \in \mathcal{U}} u(Z(x))$$

- Dominance constraint (Dentcheva & Ruszczyński, 2003; Haskell et al., 2016):

$$\begin{aligned} &\underset{x \in \mathcal{X}}{\text{maximize}} && f(x) \\ &\text{subject to} && u(Z(x)) \geq u(Z_0) , \forall u \in \mathcal{U} \end{aligned}$$

- Optimized dominance: choosing $\{Z_t\}_{t \in \mathbb{R}}$, $Z_t \succeq Z_{t'} , \forall t \geq t'$
(Armbruster & D., 2015; D. & Li, 2017)

$$\begin{aligned} &\underset{x \in \mathcal{X}, t \in \mathbb{R}}{\text{maximize}} && t \\ &\text{subject to} && u(Z(x)) \geq u(Z_t) , \forall u \in \mathcal{U} \end{aligned}$$

The role of preference robust
optimization in decision
making under uncertainty

Decision making under uncertainty

- Let (Ω, Σ, P) be a probability space with $|\Omega| = M$ and $Z(x) : \Omega \rightarrow \mathbb{R}$ a random variable
- Preference between two R.V. depends on our attitude towards risk

Theorem. *Every decision maker expresses some level of aversion to risk*

Decision making under uncertainty

St. Petersburg paradox

- I offer you to participate in the following game:

1. I flip an unbiased coin
2. First throw: if it falls on « head », I give you 2\$, otherwise I flip again
3. Second throw: If it falls on « head », I give you 4\$, otherwise I flip again
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What if I tell you that the expected winning is infinite ?

$$\mathbb{E}[Z] = \frac{1}{2} \cdot 2 + \frac{1}{4} \cdot 4 + \dots + \left(\frac{1}{2}\right)^k \cdot 2^k + \dots = \infty$$

How can one assess risk tolerance?

[Grable & Lytton, Financial Services Review (1999)]

1. You have just finished saving for a « once-in-a-lifetime » vacation. Three weeks before you plan to leave, you lose your job. You would:
 - A. Cancel the vacation
 - B. Take a much more modest vacation
 - C. Go as scheduled, reasoning that you need the time to prepare for a job search
 - D. Extend your vacation, because this might be your last chance to go first-class

2. You are on a TV game show and can choose one of the following. Which would you take?
 - A. \$1,000 in cash
 - B. A 50% chance at winning \$ 5000
 - C. A 25% chance at winning \$ 10,000
 - D. A 5% chance at winning \$100,000



Axiomatic assumptions

- Monotonicity: $X \geq Y \Rightarrow X \succeq Y$
- Risk Aversion: $X \succeq Y \Rightarrow \theta X + (1 - \theta)Y \succeq Y, \forall \theta \in [0, 1]$
- Law Invariance: $X =_P Y \Rightarrow Y \approx X$
- Independence: $X \succeq Y \Rightarrow X \text{ 🖐️ } Z \succeq Y \text{ 🖐️ } Z, \forall Z$
- Translation Invariance: $X \succeq Y \Rightarrow X + t \succeq Y + t, \forall t$
- M+RA+LI+Indep. = Expected utility for some non-decreasing concave function (von Neumann & Morgenstern, 1944)
- M+RA+LI+TI = Law invariant convex risk measures: (Kusuoka, 2001)
 - Scale Invariance : $X \succeq Y \Rightarrow \alpha X \succeq \alpha Y, \forall \alpha \geq 0$

What are we left with?

- In the case of expected utility, let $\mathcal{S} := \mathbb{R}$:

$$\mathcal{U} := \{u \mid \exists v : \mathcal{S} \rightarrow \mathbb{R}, u(\cdot) = \mathbb{E}[v(\cdot)], v(0) = 0\}$$

$$\text{Monotonicity: } \exists \partial v : \mathcal{S} \rightarrow \mathbb{R}, \partial v(\cdot) \geq 0$$

$$\text{Risk aversion: } v(y) \leq v(x) + (y - x)\partial v(x), \forall x, y \in \mathcal{S}$$

$$\text{Pairwise comparisons: } \mathbb{E}[v(W_k)] \geq \mathbb{E}[v(Y_k)], \forall k \quad \}$$

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Pairwise comparisons: $\mathbb{E}[v(W_k)] \geq \mathbb{E}[v(Y_k)], \forall k \quad \}$

- Fortunately, we can replace $\mathcal{S}$ with finite set $\mathcal{S}'$ that only includes the joint support of W_k, Y_k and Z_t

What are we left with?

- In the case of risk measures, let $\mathcal{S} := \mathbb{R}^M$:

$$\mathcal{U} := \{u \mid \exists \rho : \mathcal{S} \rightarrow \mathbb{R}, u(\cdot) = -\rho(\cdot), \rho(0) = 0$$

$$\text{Monotonicity: } \exists \nabla \rho : \mathcal{S} \rightarrow \mathbb{R}, \nabla \rho(\cdot) \leq 0$$

$$\text{Risk aversion: } \rho(Y) \geq \rho(X) + (Y - X)^T \nabla \rho(X), \forall X, Y \in \mathcal{S}$$

$$\text{Translation invariance: } 1^T \nabla \rho(X) = -1, \forall X \in \mathcal{S}$$

$$\text{Scale invariance: } \rho(X) = X^T \nabla \rho(X), \forall X \in \mathcal{S}$$

Law invariance: (see details in Delage & Li, 2016)

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- Fortunately, we can replace $\mathcal{S}$ with finite set $\mathcal{S}'$ that only includes all W_k , Y_k , and Z_t

How do we optimize ?

- Consider maximizing worst-case utility surplus:

$$\underset{x \in \mathcal{X}}{\text{maximize}} \quad \psi(x) := \min_{u \in \mathcal{U}} u(Z(x)) - u(Z_t)$$

- One can actually show that

$$\psi(x) = \underbrace{\left\{ \begin{array}{l} \min_{u \in \mathcal{U}(\mathcal{S}')} \inf_{\alpha, \beta: \alpha^T X + \beta \leq u(X), \forall X \in \mathcal{S}'} \alpha^T Z(x) + \beta - u(Z_t) \\ \min_{u \in \mathcal{U}(\mathcal{S}')} \mathbb{E} \left[\inf_{\alpha, \beta: \alpha y + \beta \leq u(y), \forall y \in \mathcal{S}'} \alpha Z(x) + \beta - u(Z_t) \right] \end{array} \right\}}_{\text{Expected utility}}$$

Convex risk measure

- Duality can be applied to get a finite dimensional optimization problem

How do we optimize ?

- Without further ado, here is the reformulation for expected utility (with $Z(x) := h(x, \xi(\omega))$):

$$\begin{array}{ll}
 \text{maximize} & \nu_0 \\
 x \in \mathcal{X}, \boldsymbol{\mu}, \boldsymbol{\nu}, \boldsymbol{\lambda}^{(1)}, \boldsymbol{\lambda}^{(2)} & \\
 \text{Linear equalities} & \left\{ \begin{array}{l} \sum_i \mu_{i,j} - (\mathbb{P}(W_0 = \bar{y}_j) - \mathbb{P}(Y_0 = \bar{y}_j))\nu_0 \\ \quad - \sum_k (\mathbb{P}(W_k = \bar{y}_j) - \mathbb{P}(Y_k = \bar{y}_j))\nu_k \\ \quad + (\lambda_j^{(1)} - \lambda_{j-1}^{(1)}) - (\lambda_j^{(2)} - \lambda_{j-1}^{(2)}) = \mathbb{P}(Z_t = \bar{y}_j) \forall j \end{array} \right. \\
 \text{Linear inequalities} & \left\{ \begin{array}{l} \lambda_j^{(2)}(\bar{y}_{j+1} - \bar{y}_j) - \lambda_{j-1}^{(1)}(\bar{y}_j - \bar{y}_{j-1}) \leq 0 \forall j \end{array} \right. \\
 \text{Convex constraints} & \left\{ \begin{array}{l} \sum_j \bar{y}_j \mu_{i,j} \leq \mathbb{P}(\omega_i) h(x, \xi(\omega_i)) \forall i \\ \text{if } h \text{ is concave} \end{array} \right. \\
 \text{Linear equalities} & \left\{ \begin{array}{l} \sum_j \mu_{i,j} = \mathbb{P}(\omega_i) \forall i \end{array} \right. \\
 & \boldsymbol{\mu} \geq 0, \boldsymbol{\nu} \geq 0, \boldsymbol{\lambda}^{(1)} \geq 0, \boldsymbol{\lambda}^{(2)} \geq 0,
 \end{array}$$

How do we optimize ?

- Consider instead the case:

$$\begin{array}{ll} \underset{x \in \mathcal{X}, t \in \mathbb{R}}{\text{maximize}} & t \\ \text{subject to} & u(Z(x)) \geq u(Z_t) , \forall u \in \mathcal{U} \end{array}$$

- If t^* is known to lie in some interval $[t_0, t_1]$, then one can employ the bisection method on t :

$$\max_{t \in [t_0, t_1]} t : \max_{x \in \mathcal{X}} \min_{u \in \mathcal{U}} u(Z(x)) - u(Z_t) \geq 0$$

Value of optimized dominance model

1. Optimized dominance reduces to max-min utility when

$$u(Z_t) = u'(Z_t), \forall u, u' \in \mathcal{U}, \forall t$$

- In expected utility this occurs if $\mathcal{S} \subseteq [z_{\min}, z_{\max}]$

$$Z_t := \begin{cases} z_{\max} & \text{with prob. } t \\ z_{\min} & \text{with prob. } 1 - t \end{cases} \quad \& \quad \forall u \in \mathcal{U}, \begin{cases} u(z_{\max}) = 1 \\ u(z_{\min}) = 0 \end{cases}$$

- With convex risk measures, this occurs if

$$Z_t := t$$

Value of optimized dominance model

2. When $Z_t := t$, the proposed decision is guaranteed to be preferred to the largest available guaranteed return
 - As a consequence, it implements a conservative use of preference information

Value of optimized dominance model

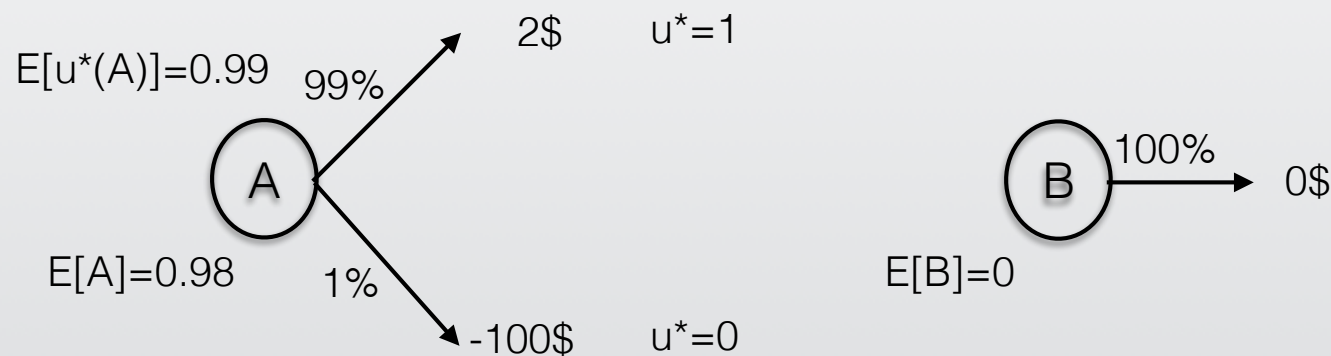
2.

Example:

Consider the following two risky investments.

$$\mathcal{U} := \{u \mid u \text{ concave}, u(-100) = 0, u(2) = 1\}$$

Max-min expected utility would recommend :



Value of optimized dominance model

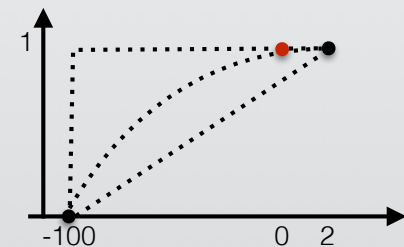
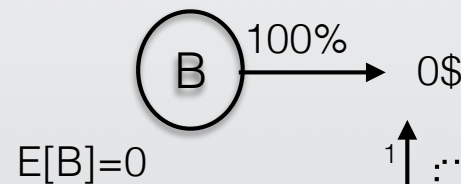
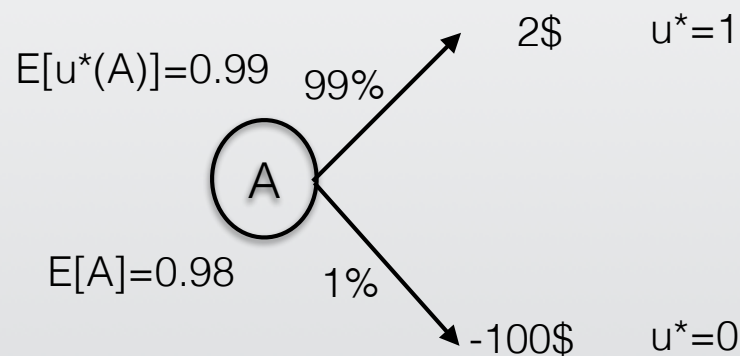
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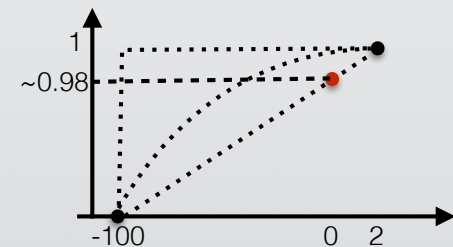
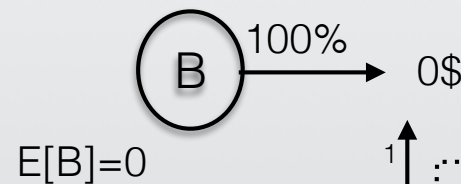
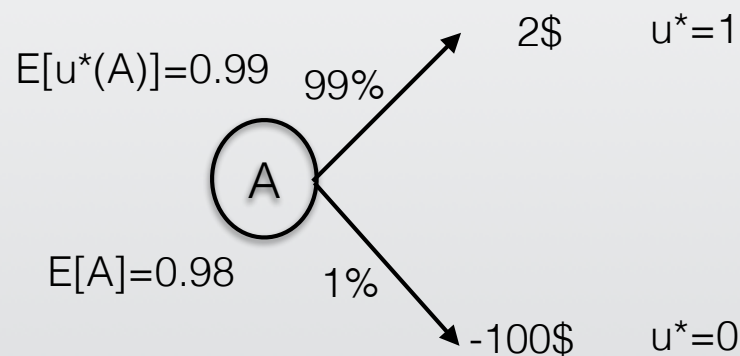
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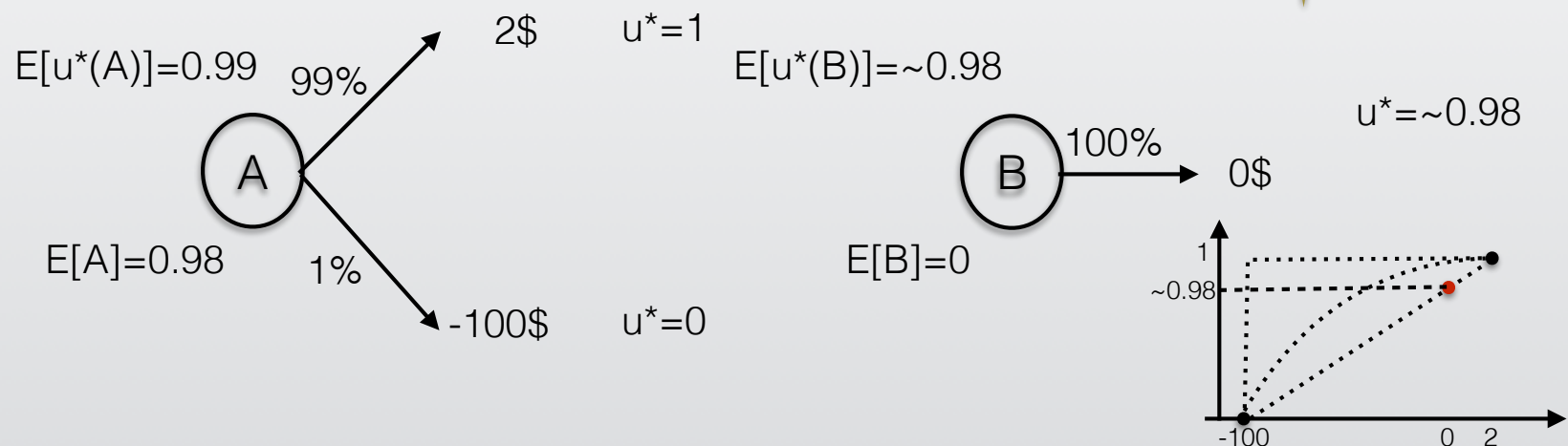
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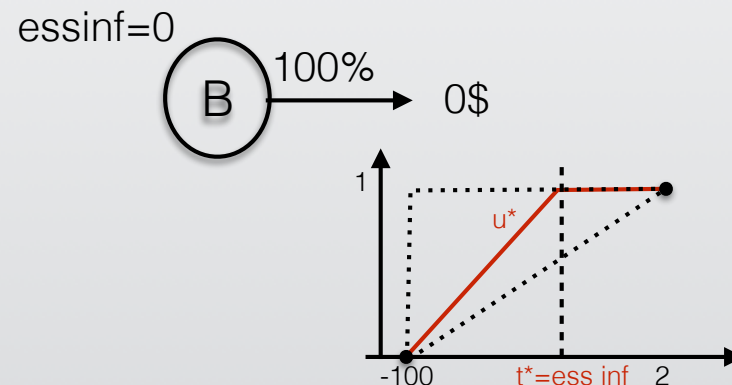
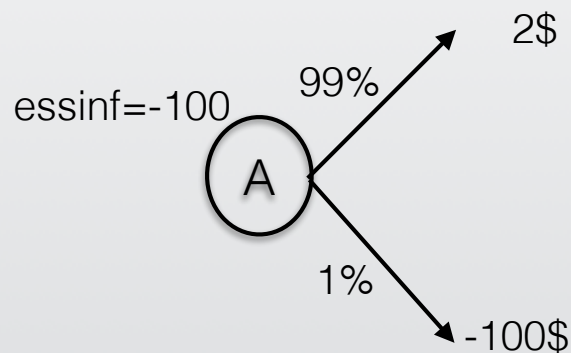
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Optimized dominance with $Z_t := t$ would recommend :



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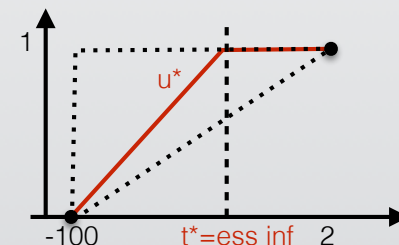
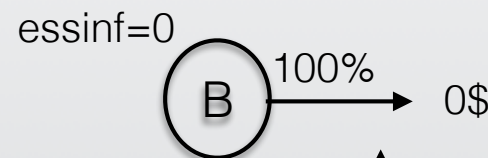
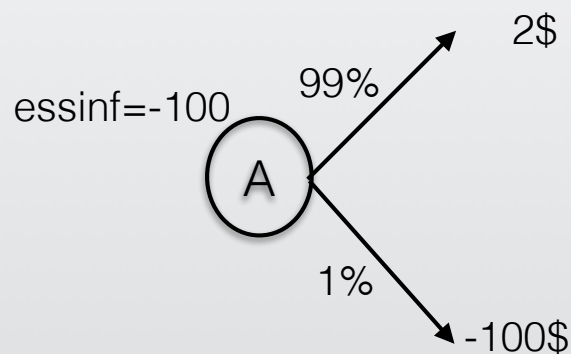
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Optimized dominance with $Z_t := t$ would recommend:

B



Accounting for elicitation errors

- Since even the best intention can lead to making mistakes, we might need to account for mislabeled comparisons



- One can replace the comparison constraint with:

$$\exists \delta \in \mathbb{R}^K, \|\delta\|_1 \leq \Gamma, u(W_k) + \delta_k \geq u(Y_k), \forall k = 1, \dots, K$$

- Bertsimas and O'hair (2015) even propose accounting for some preference reversals with :

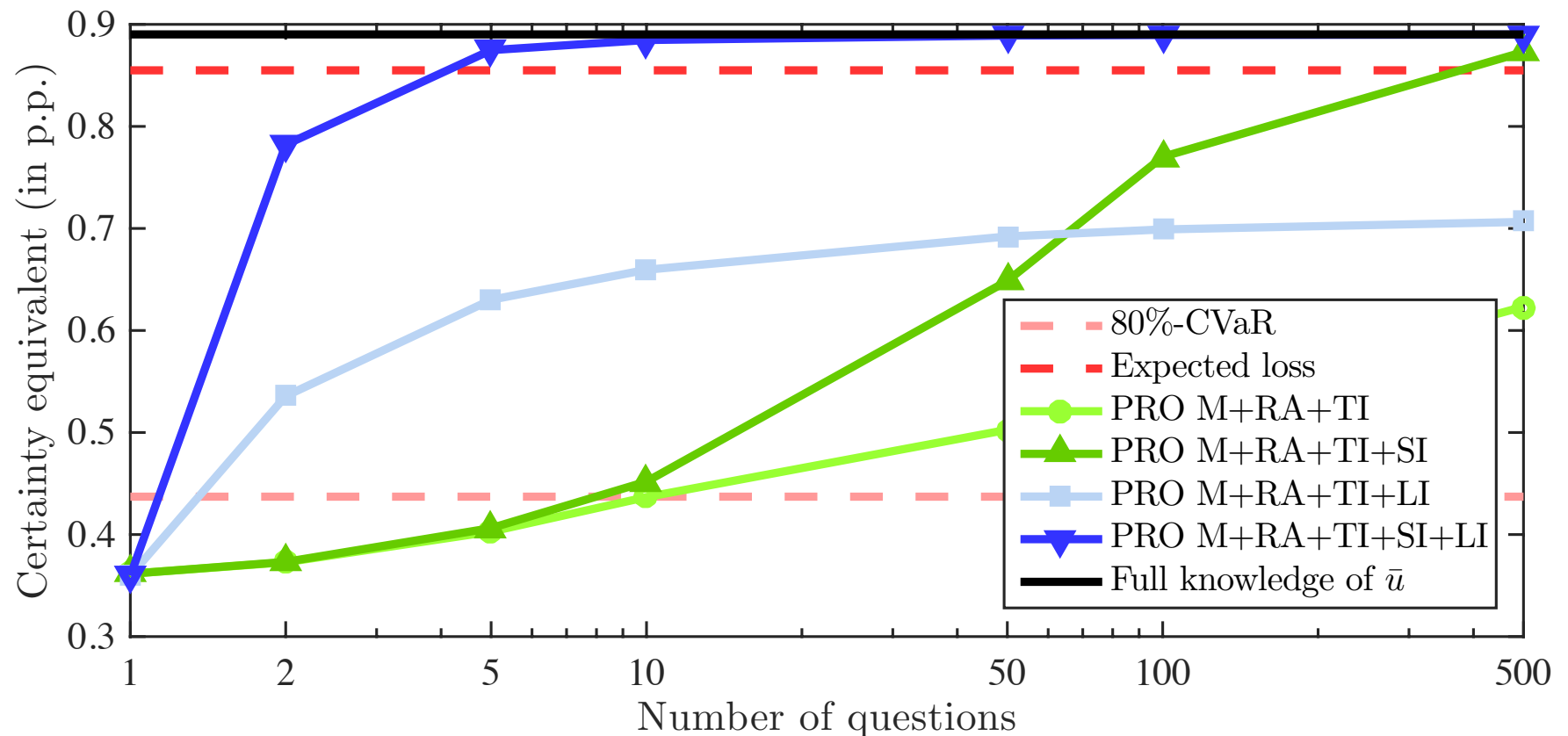
$$\exists z \in \{0, 1\}^K, \sum z_k \leq \Gamma, \left\{ \begin{array}{l} u(W_k) \geq u(Y_k) - Mz_k \\ u(Y_k) \geq u(W_k) - M(1 - z_k) \end{array} \right\}, \forall k$$

A brief survey of our
numerical experiments

Numerical experiments

- Experiments are made using empirical stochastic models based on historical weekly returns from Yahoo Finance
- We create a synthetic decision maker with some choice of $\bar{u}(\cdot)$ which is kept hidden
- Information comes from a number of comparisons made using $\bar{u}(\cdot)$ for sets of randomly picked (W_k, Y_k)
- Results are averaged over a large number of stochastic models and sets of comparisons

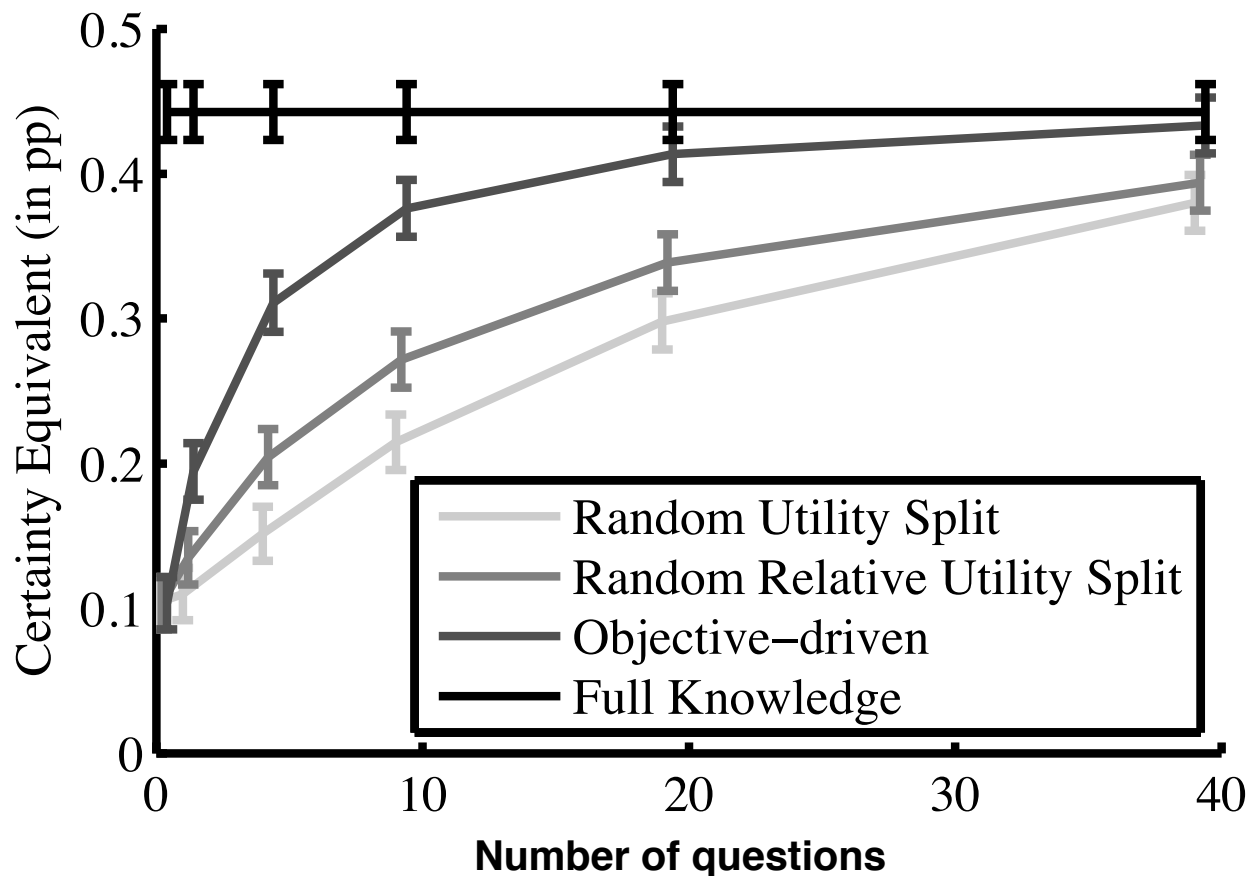
Performance in terms of certainty equivalent*



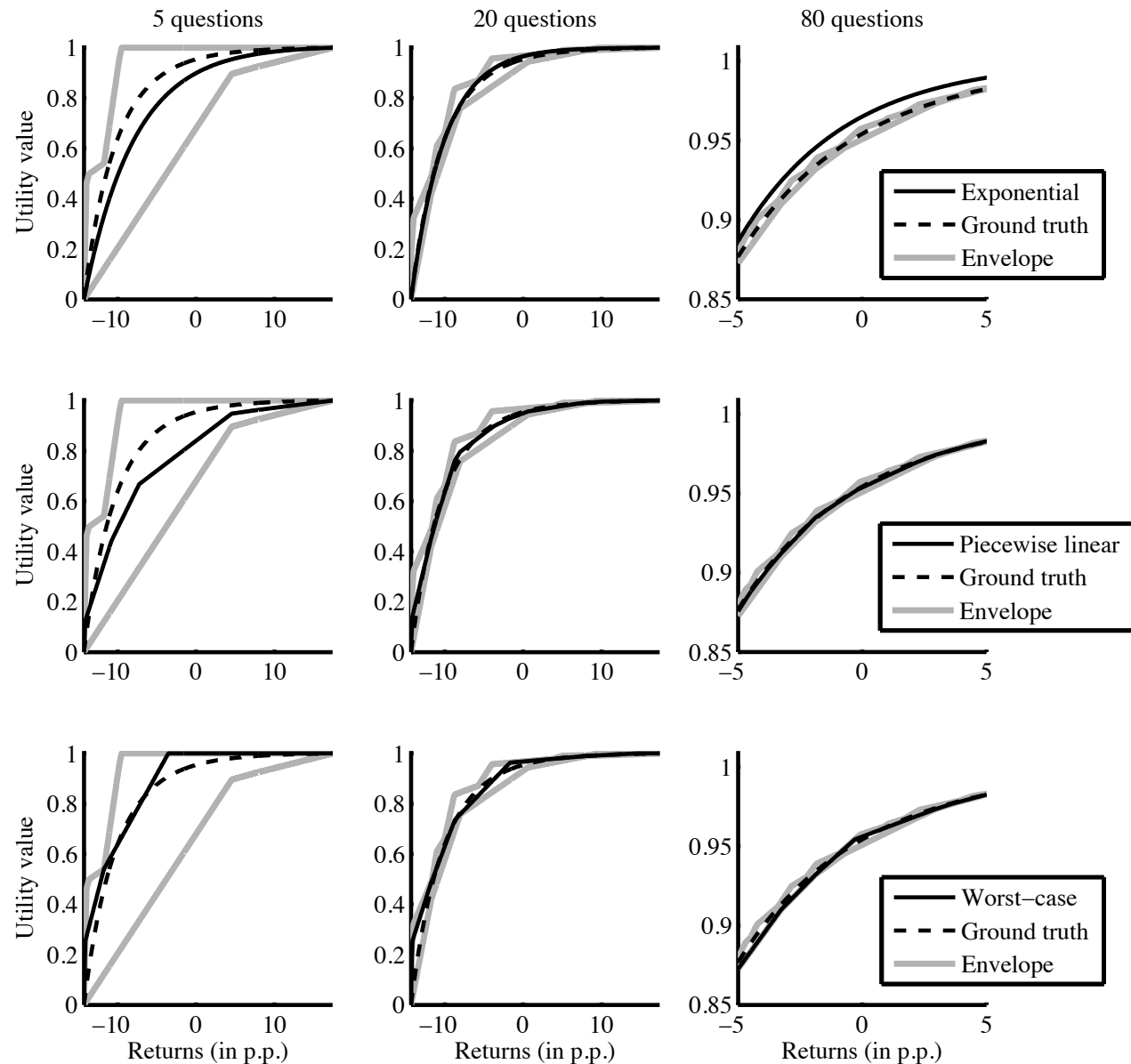
* Certainty equivalent : $\bar{u}(CE[Z(x)]) = \bar{u}(Z(x))$

Effect of elicitation strategy

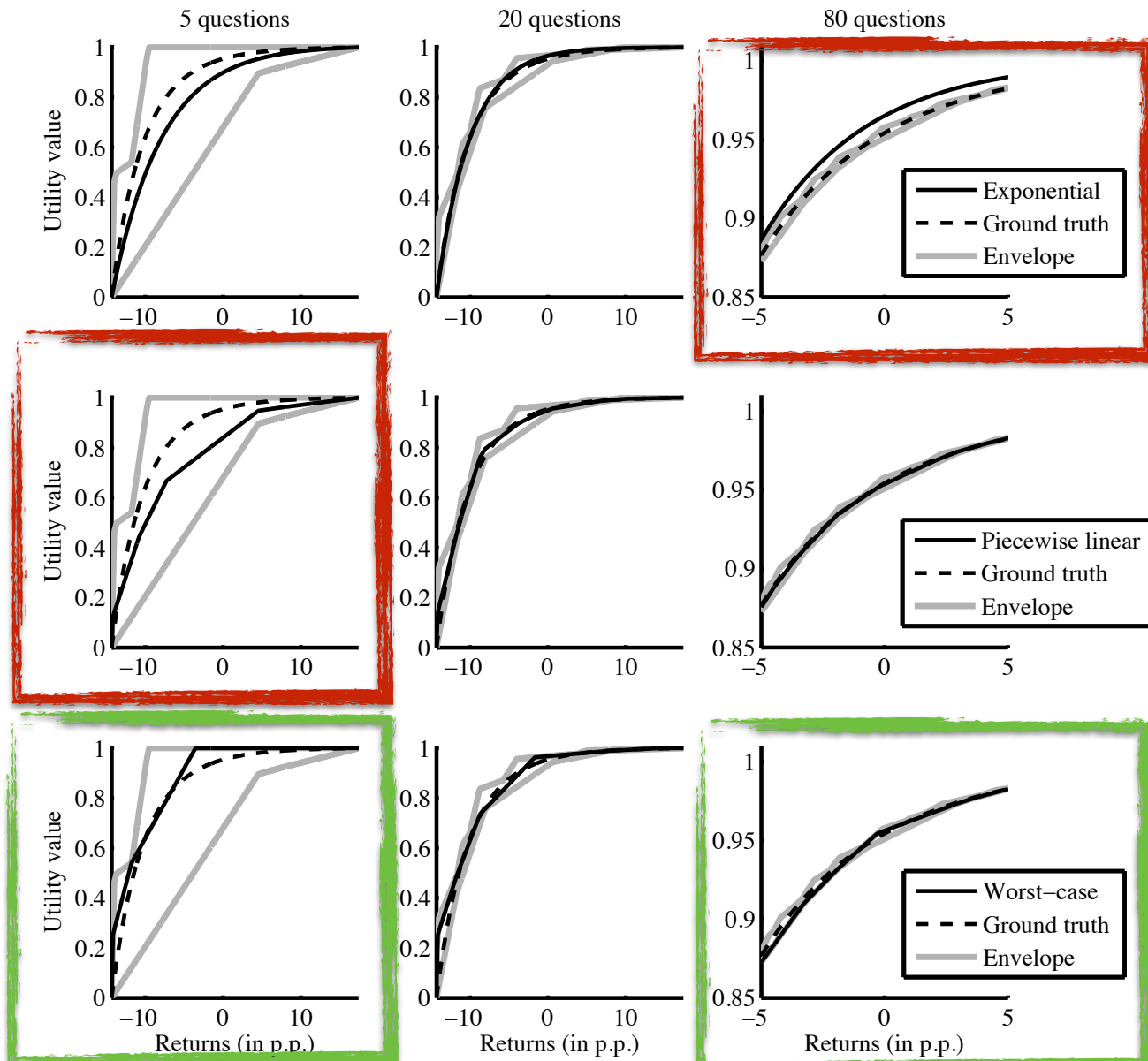
- One can improve convergence rate by designing effective elicitation strategies



Quality of recovered $u()$



Quality of recovered $u()$



- Exponential utility model does not recover true utility function
- Piecewise linear model takes excessive risks when information is limited
- Optimized dominance resolves both issues

Conclusion

- Many optimization problems involve objective functions that need to reflect the decision maker's preferences
- PRO accounts for the limited knowledge about these preferences : axioms + pairwise comparisons
- PRO offers valuable guarantees with respect to the unrevealed preference relation
- PRO preserves difficulty of resolution: LP $\longrightarrow$ LP
- Many potential applications: game theory, revenue management, expert systems, recommendation systems...

Thank you for your
attention



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