

# Revealed Different

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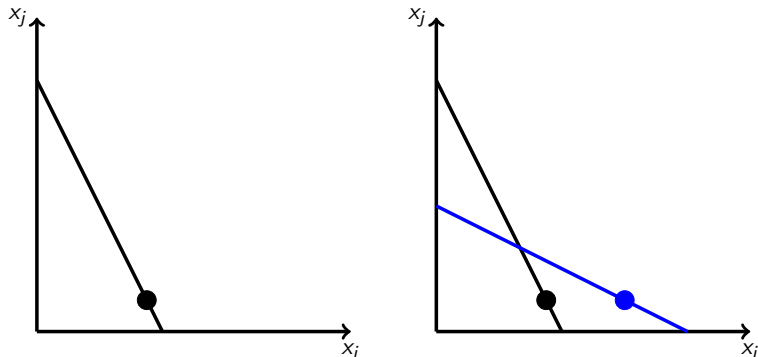
<sup>2</sup>ICES GMU

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# Goal

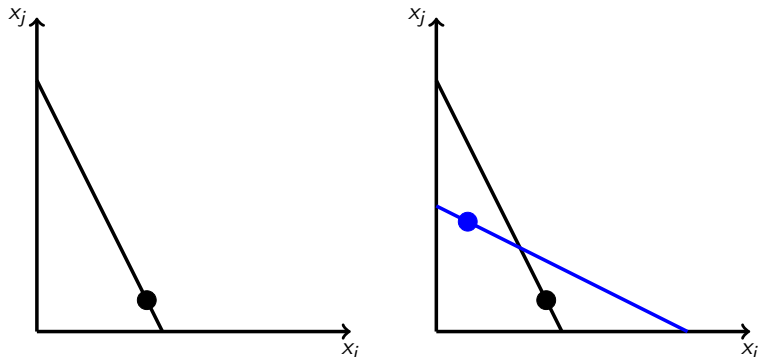
- *Develop a non-parametric method to test if preferences across individuals are different*
- Determine how heterogeneous preferences are and whose preferences are more heterogeneous
- Investigate how non-parametric assumptions on preferences affect the answers to these questions

# Intuition



If rational agents have the same preferences, the combined decisions of both subjects are still rational  
 $\Rightarrow$  They are *not Revealed Different*.

# Intuition



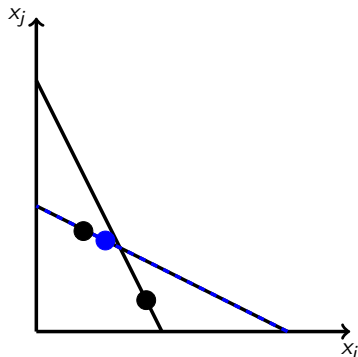
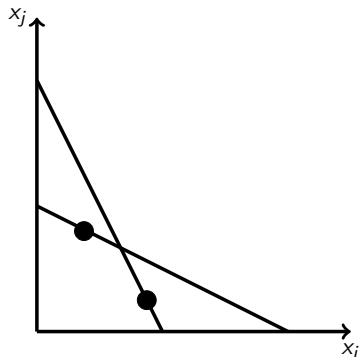
If rational agents have different preferences, the combined decisions of both subjects might not be rational.

⇒ They are *Revealed Different*.

# Challenges

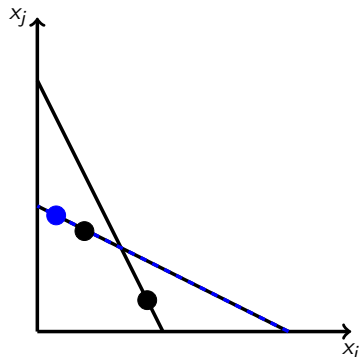
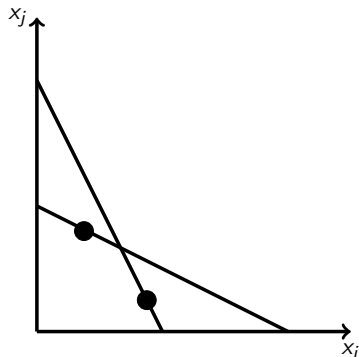
- (!) *Most people are not exactly rational ...*
- (?) How can we detect differences in preferences among not exactly rational agents?

# Intuition: Dealing With Irrationality



Not-exactly rational agents have the same preferences and their merged decisions are no less rational  
 $\Rightarrow$  They are *not Revealed Different*.

# Intuition: Dealing With Irrationality



Not-exactly rational agents don't have the same preferences and their combined decisions are less rational  
 $\Rightarrow$  They are *Revealed Different*.

# What do we want?

Weitzman (1992)

Existence of a distance function is all that is needed to characterize diversity (i.e. to classify species and value their diversity)

- $d(i, j)$  is a distance function
  - $d(i, i) = 0$
  - $d(i, j) \geq 0; \forall x, y$  (Non-negative)
  - $d(i, j) = d(j, i)$  (Symmetric)



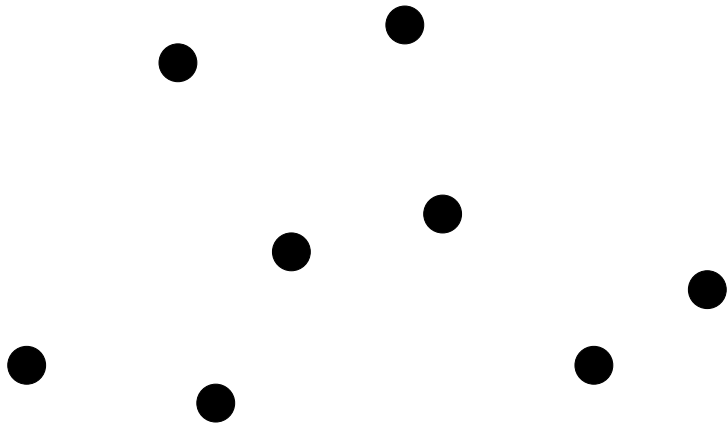
# Being Different

$$RD(x, y) = \begin{cases} 1, & \text{if } \rho(x \cup y) > \max\{\rho(x); \rho(y)\} \\ 0, & \text{otherwise} \end{cases}$$

where  $\rho(\cdot)$  = distance from rationality

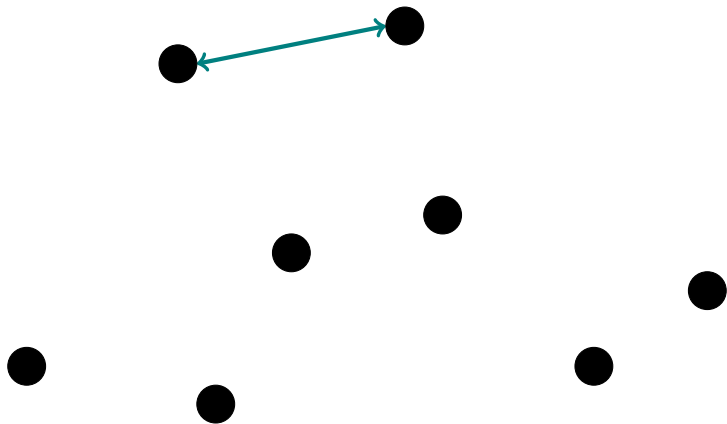
# Characterizing heterogeneity

Dots for people, arrows for *revealed different*



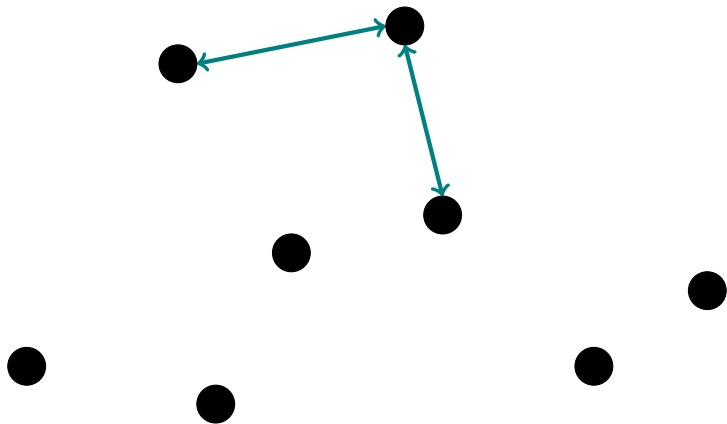
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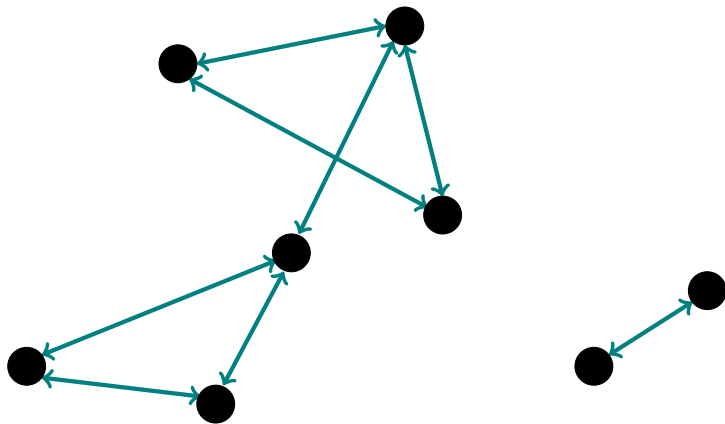
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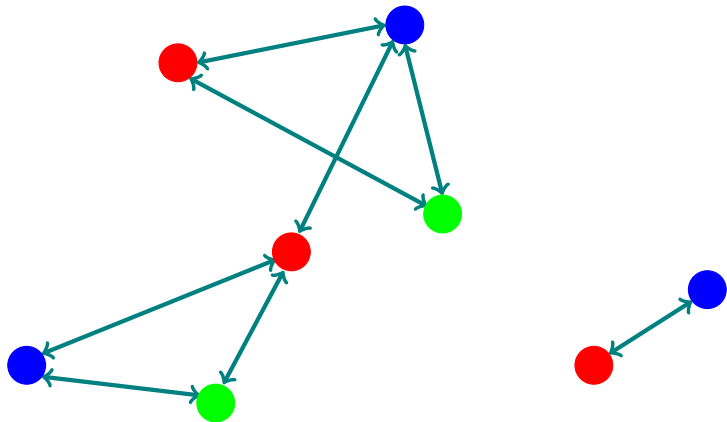


# Characterizing heterogeneity

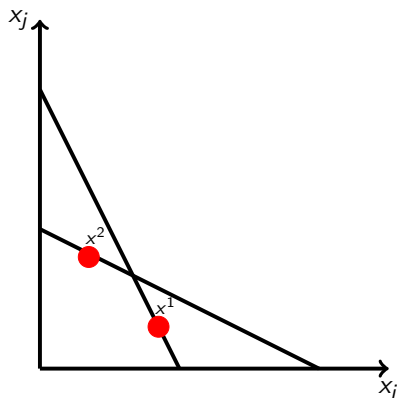
- How many types does the Revealed Different relation contain?
- We group people such that:
  - ⇒ No two agents in a group is different from each other
  - ⇒ The amount of types is minimal

# Coloring Graph (3 types: red, blue and green)

Dots for people, arrows for *revealed different*



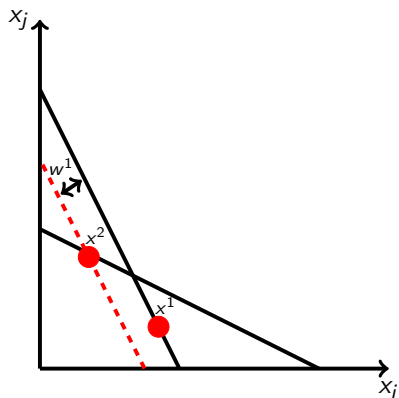
# Critical Cost Efficiency Index (Afriat, 1973)



Moving  $x^1$  budget

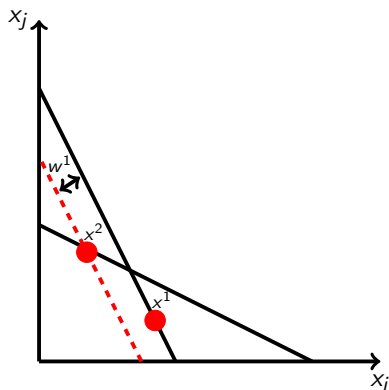


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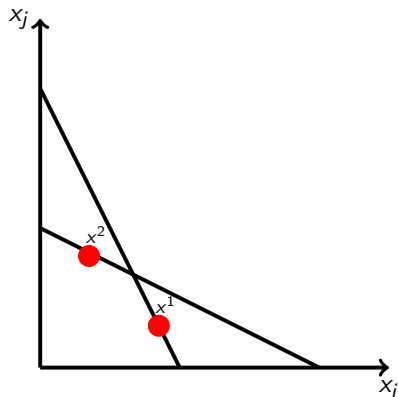


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# Critical Cost Efficiency Index (Afriat, 1973)

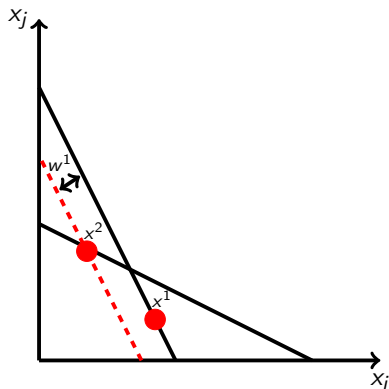


Moving  $x^1$  budget

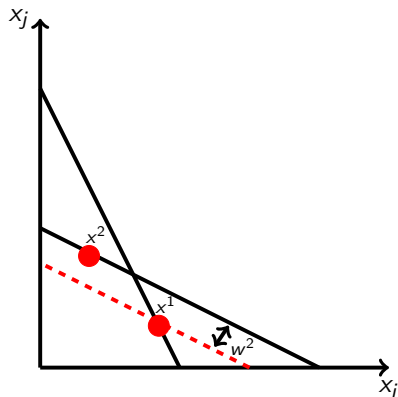


Moving  $x^2$  budget

# Critical Cost Efficiency Index (Afriat, 1973)



Moving  $x^1$  budget



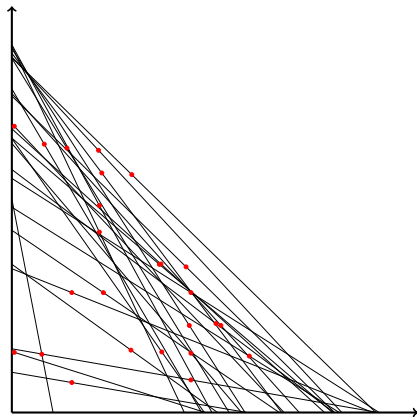
Moving  $x^2$  budget

$$RD(x, y) = \begin{cases} 1, & \text{if } CCEI(x \cup y) < \min\{CCEI(x); CCEI(y)\} \\ 0, & \text{otherwise} \end{cases}$$

Choi et al (2014).

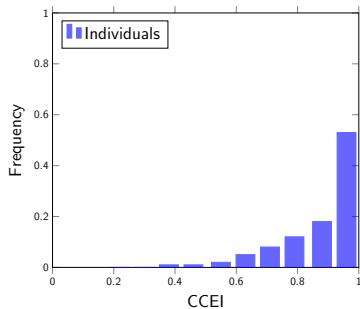
- Participants are a representative sample of the Dutch adult population.
- Each participant faced 25 budgets with two risky assets that obtain with equal probability.
- The cost of each asset were assigned randomly.

# Data

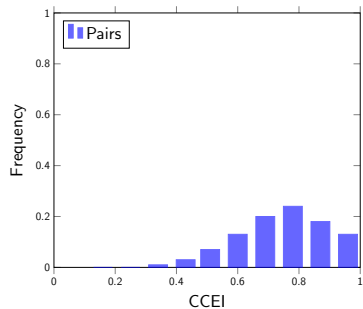


25 Decisions Made by Subject

# Descriptive: CCEI distributions



CCEI of Individuals



CCEI of Pairs

# Coloring Real Graph

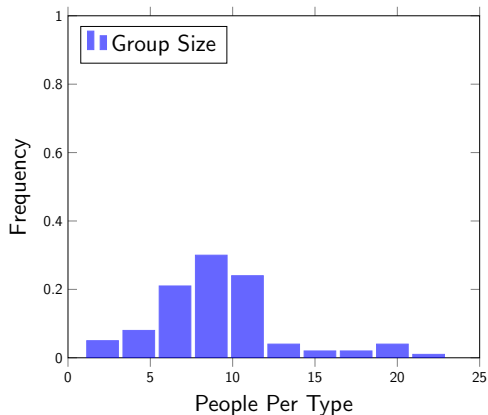


Figure: Distribution of Type Sizes



# Males vs Females: GARP + measurement error

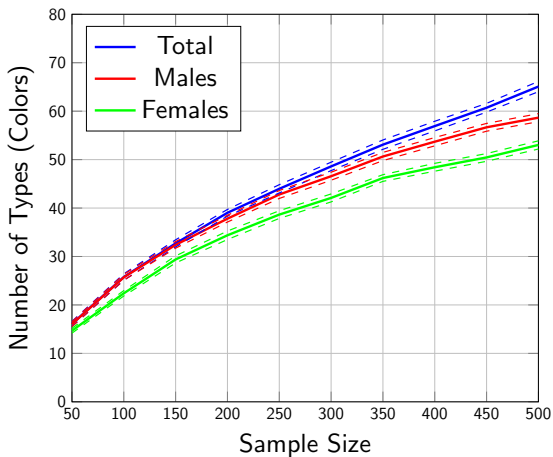


Figure: Number of Types (with respect to GARP)

# Additional Assumptions: GARP vs FGARP

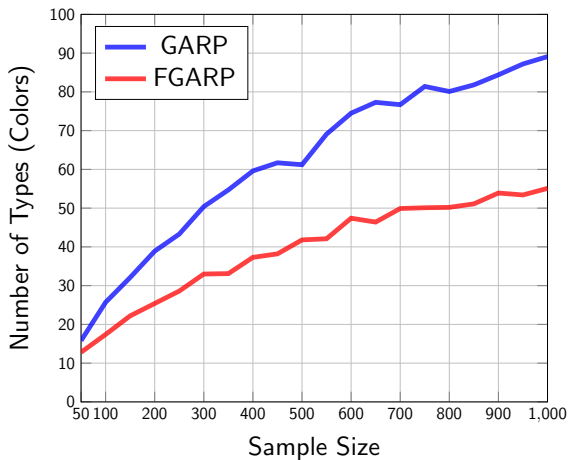


Figure: Number of Types

# Main Findings

- Preferences are quite heterogeneous: The number of different preference types is as large as 135 ( $N=1,182$ )
- Males are different and more heterogeneous than females
- Preferences are idiosyncratic: less than 1% of variation in the *revealed different* relation can be explained by observable differences in individual characteristics
- Additional assumptions on preferences are demanding: When wrong, they add noise and prevent detecting differences in preferences

Thank you !

# Existing Alternatives To Detect Heterogeneity of Preferences

▶ Back

- **Parametric** (e.g. von Gaudecker et al, 2011).  
⇒ Why this function? How much heterogeneity is lost?
- **Non-parametric** (e.g. Moore & Eckel, 2003; Gross, 1995; Crawford & Pendakur, 2013):  
⇒ Concluding that differences in behavior reflect differences in preferences requires assuming that behavior is to some extent rational

# Why do we explore a new approach?

Pollison et al (2015) using experimental data from Choi et al (2007):

- Only **2 %** of subjects are consistent with Expected Utility (32% under 5% error level)
- Only **2 %** of subjects are consistent with Disappointment aversion  $\sim$  cumulative Prospect Theory in 2-goods world (38% under 5% error level)
- Only **17 %** of subjects are consistent with GARP (66 % under 5 % error level)

$\Rightarrow$  Any comparison that invokes these comparison will have little power to detect differences...

- Gross (1995) used revealed preference to test whether people have similar preferences
  - ⇒ Using Panel Study for Income Dynamics found that women have similar preferences (can be rationalized by same utility function) for childcare
- Crawford and Pendakur (2013) use revealed preference analysis to group people into types
  - ⇒ In a cross-section of Danish household (1 obs per HH), they find that the number of types rationalizing the demand for milk is at most 12





# When does it work? ▶ Back

## Assumption 1

CCEI converges to the real value of CCEI as number of observations goes to infinity:

$$\text{CCEI} \rightarrow \bar{\epsilon}, \text{ when } N \rightarrow \infty$$

## Assumption 2

The number of observations is large enough to elicit real CCEI:  
Observed CCEI  $\approx \bar{\epsilon}$

If Assumption 1 is true  $\Rightarrow$  *Assumption 2 is testable.*



- **Measurement error:**

$RD = 1$ , if  $\min\{CCEI(x); CCEI(y)\} - CCEI(x \cup y) > \epsilon$ ,  
where  $\epsilon > 0$  is computational error;

- **Median difference:**

$RD = 1$ , if  $\min\{CCEI(x); CCEI(y)\} - CCEI(x \cup y) > .01$ ;

- **95% Confidence interval:**

$RD = 1$ , if  $\min\{CCEI(x); CCEI(y)\} - CCEI(x \cup y) > .07$ ;

- **Certain bound:**

$RD = 1$ , if  
 $\min\{CCEI(x); CCEI(y)\} - CCEI(x \cup y) > .1$ .

# Males vs Females: GARP + ( $e = 0.07$ ) [▶ Back](#)

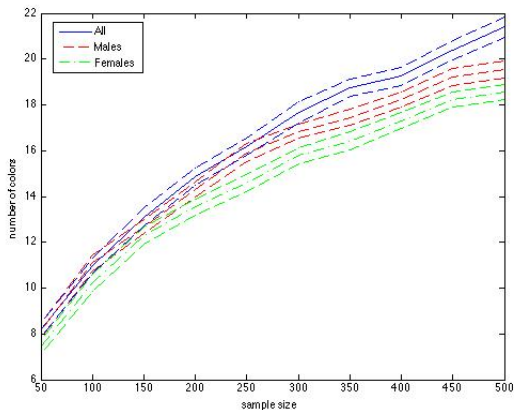


Figure: Number of colors (with respect to GARP) with 95% Confidence Intervals

# How much heterogeneity? [▶ Back](#)

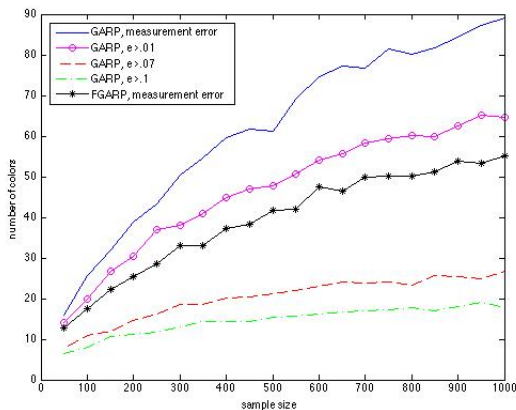


Figure: Number of types (Differences are significant with  $p < 0.001$ )

# How much heterogeneity? [▶ Back](#)

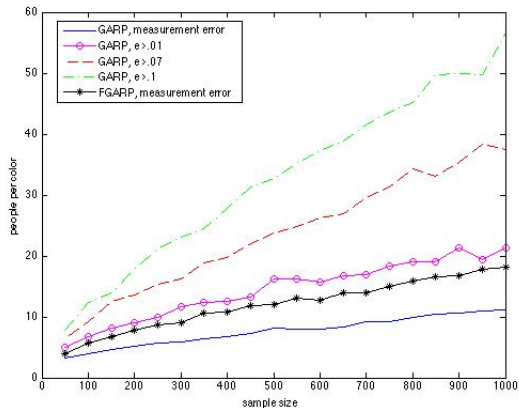


Figure: Number of people per type