

# Nutritional policy design

*Behavioral insights from the lab to the field*

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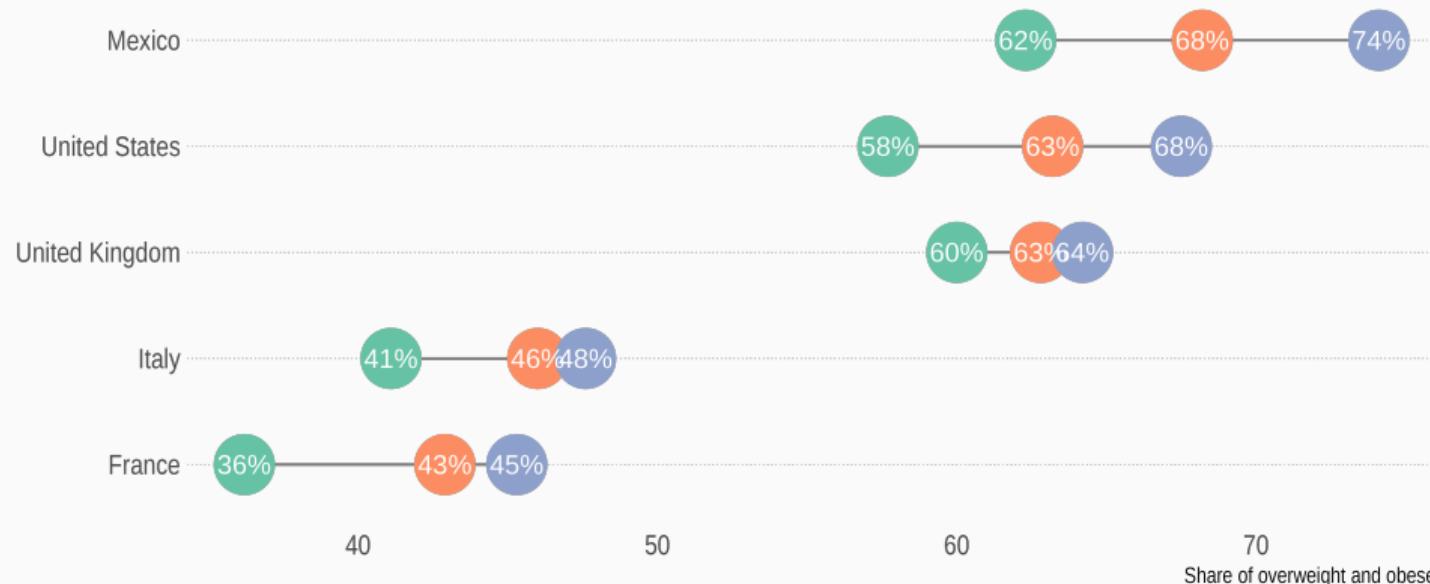
## The problem

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# We're getting fatter

## Evolution of the share of overweight and obese people: 2000, 2010, 2020

Selected OECD countries, all population aged 15+



...and this has huge **costs**

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## Health-related costs

Direct medical costs (2019):

- US: 300bn
- UK: 22bn
- Global: 1 to 3% GDP

Direct + indirect (projected 2030):

- Africa 1% GDP
- Americas 4% GDP
- Middle EAST 5% GDP

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## Health-related costs

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- Middle EAST 5% GDP

## Environment-related costs

- Food: 34% of GHG emissions
- Obesity: 1.4% extra
- 14% more transport emissions
- 140Mt excess consumption

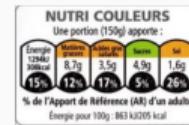
## What **tools** do we have?

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- Regulation
- Information
- Labeling
- Price policies
- Nudges
- ...

# This talk

- Regulation
- Information
- Labeling
- Price policies
- Nudges
- ...



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Fat tax & thin subsidy

## A series of key **policy** questions

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- Do **labels** work?
- Which label design is the **best** to **impact** choice?
- **How much** of an impact labels have?
- Do **price** interventions work?
- Do they work **better** or **worse** than labels?
- How do the two policies **interact**?

## Which role for the lab?

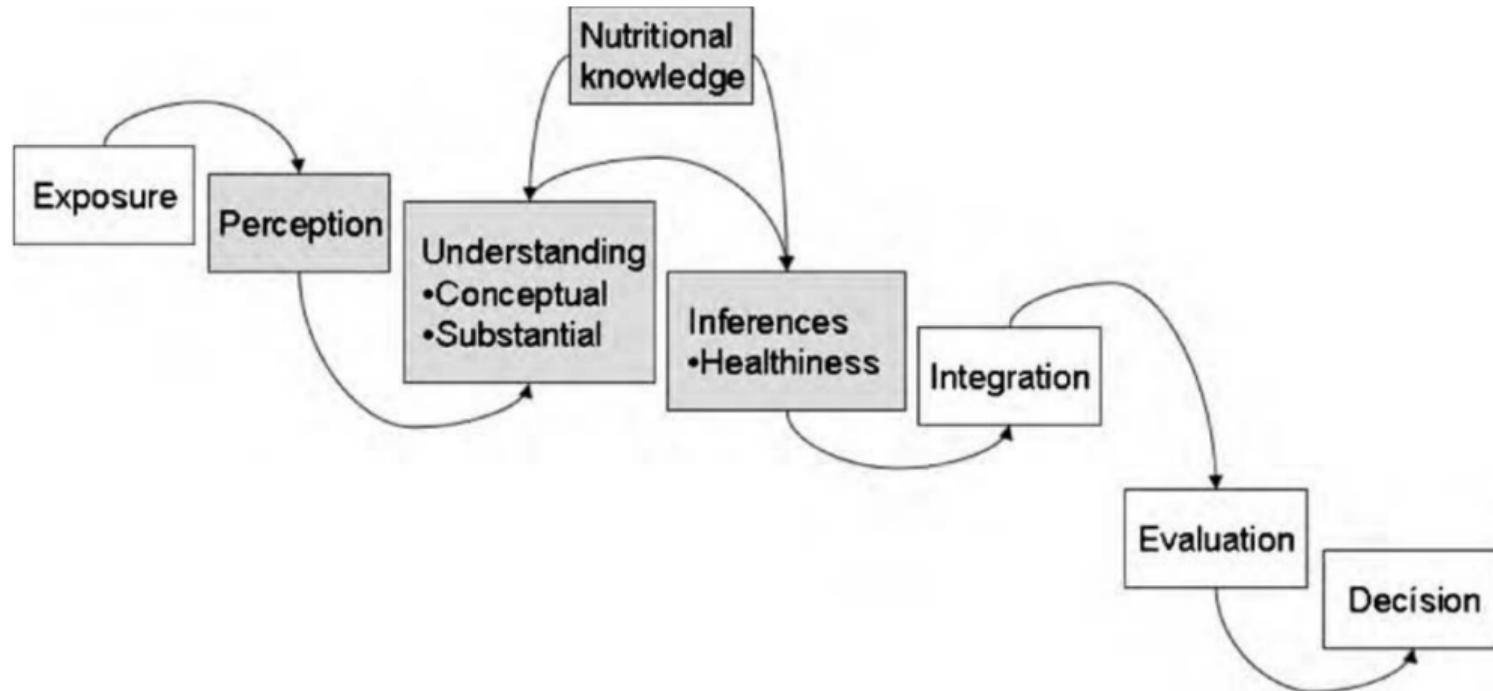
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# What can we learn from the lab?

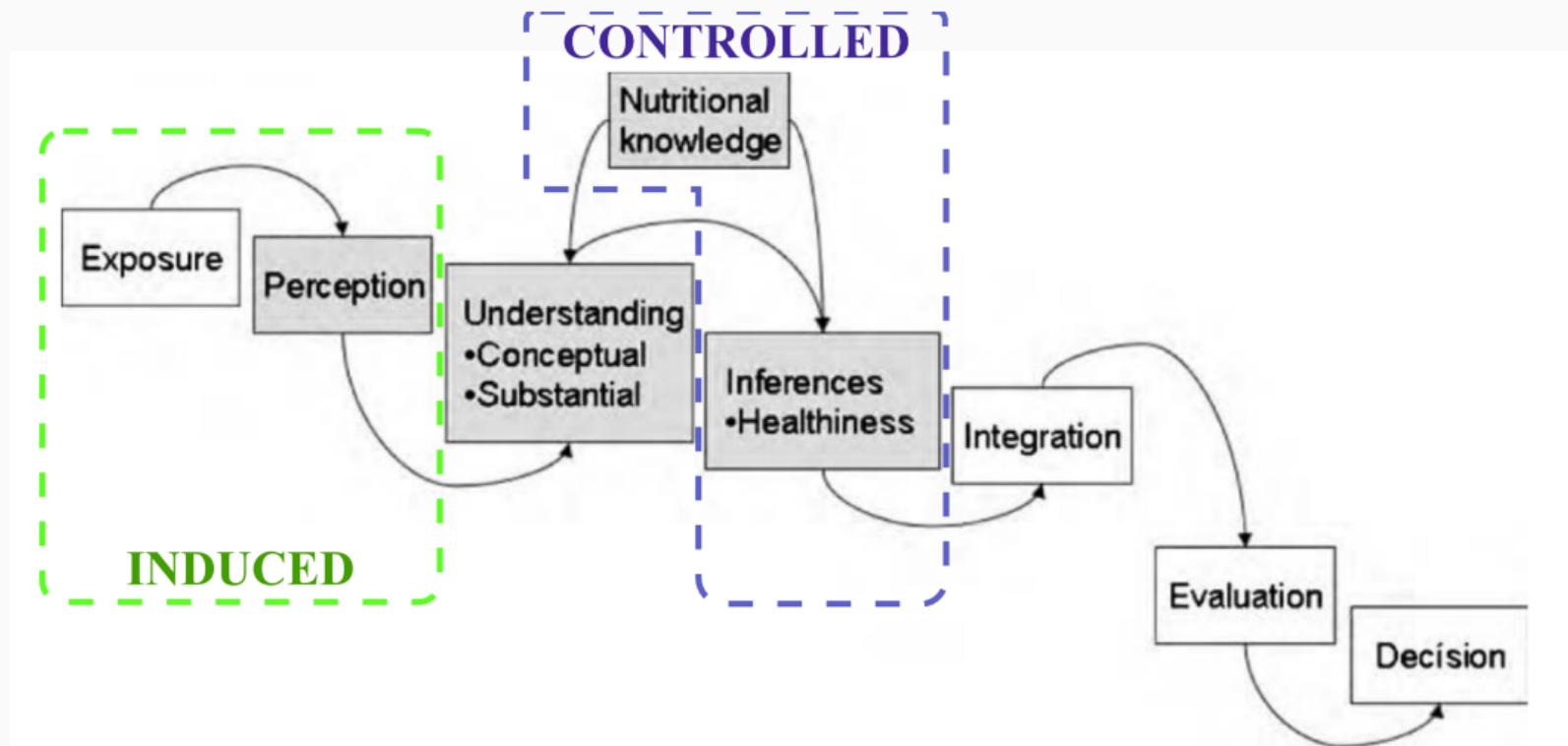
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## A conceptual framework (Grunert)



## A conceptual framework (Grunert)



## Two main roles of the lab in an applied policy context

### Getting into the mind of subjects

- focus on cognitive aspects
- clearly identify mechanisms
- (if needed) sidestep preferences
- heuristics, choice processes



## Two main roles of the lab in an applied policy context

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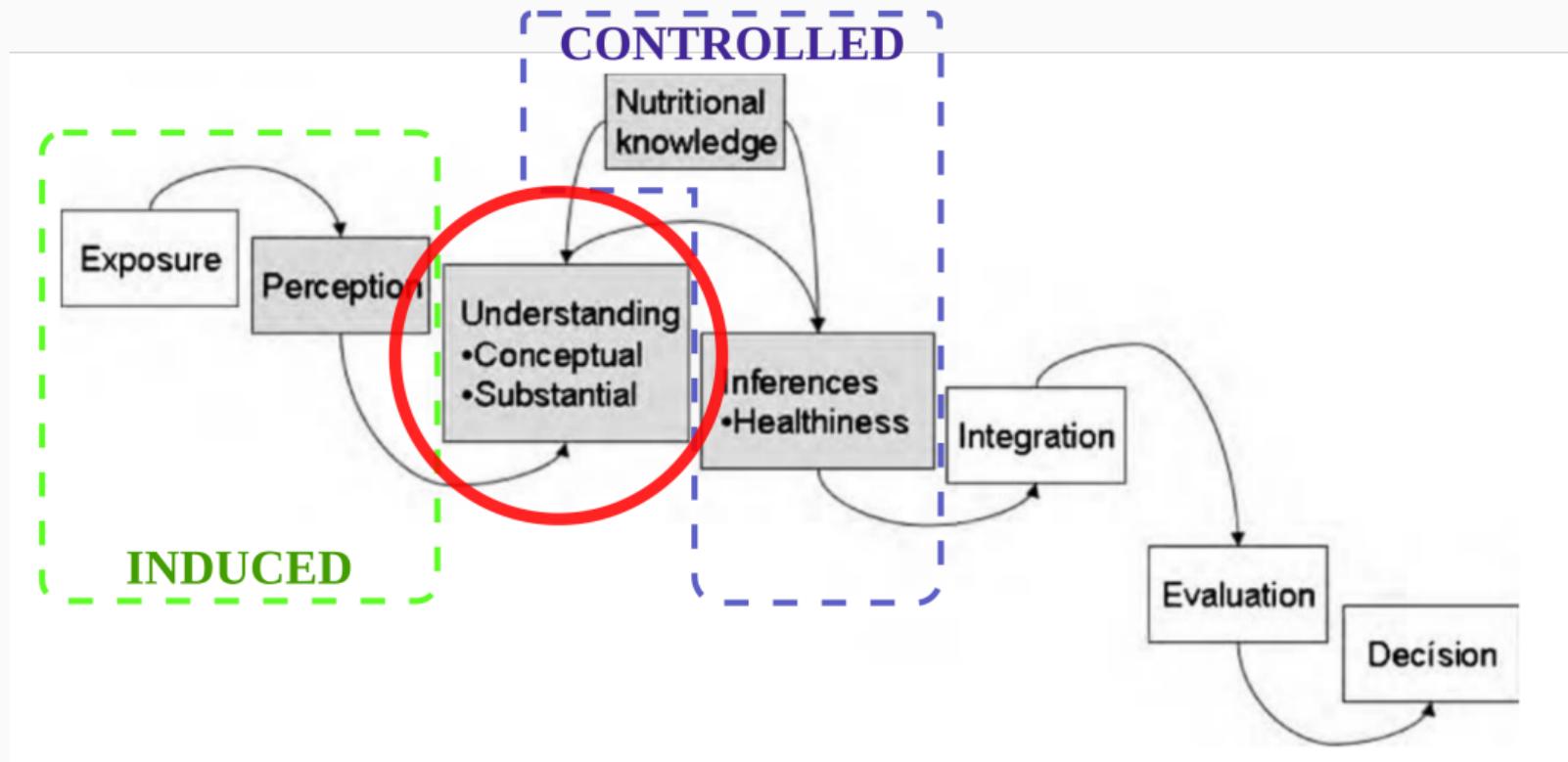
### Building counterfactuals

- explore different scenarios
- preferences, with control
- track macro consequences
- cheaply explore solutions

## **Part 1: into the mind of subjects**

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# Getting into the **mind** of subjects



## The usual design



# The usual design



## **Study 1: building *diets***

**[JoEP 2015 – L. Muller, B. Ruffieux]**

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# What label is better to build a healthy **diet**?

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## Diets

- a diet is a complex object, akin to a **portfolio**
- you won't die for one bad item, but if the overall balance is wrong

## Task

- subject "hired as a nutritionist for a canteen"
- must compose **daily menu** & satisfy nutritional constraints
- subject guided by labels: **numbers, colors, or both.**

## Incentives

- If diet satisfies nutritional constraints  $\Rightarrow$  flat fee (2 euro)
- Several daily diets to build

# Our design: diet-building

Petit déjeuner		Lait frais entier		Nectar de fruits exotiques		Eau gazeuse		Lait aromatisé
		Céréale type All Bran		Pain de mie		Orange		Pain suédois
Déjeuner		Salade fraîche		Pomme de terre à l'huile		Viande des Grisons		Pâté de foie de volaille
		Sandwich crudité fromage		Sandwich crudité rosbif		Pot au feu		Sandwich type libanais (falafel)
Collation		Pamplemousse frais		Fraise		Mousse de fruit		Salade de fruits
		Gâteau de Savoie		Pett-suisse		Meringue		Kiwi
Dîner		Avocat vinaigrette		Laitue		Asperge grande		Rollmops de hareng
		Flageolet		Pâtes complètes		Quinoa		Châtaigne
		Haricot rouge		Pâtes fraîches		Poivron rouge grillé		Carotte

Characteristics:

- no preferences
- incentivized
- "realistic"

We add:

- labels
- constraints

## Dimensions

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Nutrition is **multidimensional**. We consider three cases:

**1-dimension** Kcal only are displayed.

**4-dimension** Kcal + 'bad' nutrients: salt, sugar, fat.

**7-dimension** 4d + 'good' nutrients: vitamin C, fiber, calcium.

## Dimensions

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**4-dimension** Kcal + 'bad' nutrients: salt, sugar, fat.

**7-dimension** 4d + 'good' nutrients: vitamin C, fiber, calcium.

Labels can have **numbers**, or **colors**, or **both**:

**Numbers** modeled on Reference Intakes

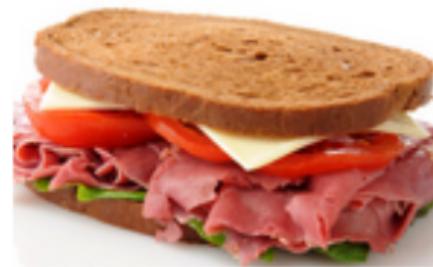
**Colors** modeled on Traffic Lights

**Num+col** both of the above combined

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	Tarte aux poireaux
Energie :	12.3
Sucres :	1.7
Graisses :	46.4
Sel :	19.1

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Sandwich crudités rosbeff

Energie :	14.4
Sucres :	●
Graisses :	●
Sel :	●



## Pêche

Energie :	3	
Sucres :	14	●
Graisses :	0	●
Sel :	0	●
Vitamines :	11	●
Fibres :	12	●
Calcium :	2	●

# A number + colors screen, 7 constraints

Petit déjeuner	Lait frais entier		Energie : 6.3 Sucres : 10.2  Graisses : 21  Sel : 3.8  Vitamines : 8.4  Fibres : 0 	Nectar de fruits exotiques		Energie : 5.5 Sucres : 29.3  Graisses : 0  Sel : 0.3  Vitamines : 7.8  Fibres : 0.8 	Eau gazeuse		Energie : 0 Sucres : 0  Graisses : 0  Sel : 2.9  Vitamines : 0  Fibres : 0 	Lait aromatisé	
	Céréale type All Bran		Energie : 4.2 Sucres : 5.3  Graisses : 1.1  Sel : 10.6  Vitamines : 22.8  Fibres : 32.4 	Pain de mie		Energie : 6.7 Sucres : 1.1  Graisses : 2.4  Sel : 12.5  Vitamines : 4.2  Fibres : 6.6 	Orange		Energie : 2.6 Sucres : 12.5  Graisses : 0.2  Sel : 0.1  Vitamines : 6.4  Fibres : 8.1 	Pain suédois	
Déjeuner	Salade frisée		Energie : 0.2 Sucres : 0.5  Graisses : 0.1  Sel : 0.2  Vitamines : 3.1  Fibres : 1.9 	Pomme de terre à l'huile		Energie : 6.8 Sucres : 3.9  Graisses : 4.9  Sel : 17.3  Vitamines : 8.6  Fibres : 6.8 	Viande des Grisons		Energie : 2.6 Sucres : 0.1  Graisses : 3.2  Sel : 25.9  Vitamines : 13.3  Fibres : 0 	Pâté de foie de volaille	
	Sandwich crudité fromage		Energie : 19 Sucres : 2.8  Graisses : 31.9  Sel : 25.3  Vitamines : 12.7  Fibres : 11.6 	Sandwich crudité rosbif		Energie : 14.4 Sucres : 2.8  Graisses : 4.7  Sel : 23.7  Vitamines : 20.5  Fibres : 11.2 	Pot au feu		Energie : 13.9 Sucres : 11.3  Graisses : 24.6  Sel : 52.3  Vitamines : 28.6  Fibres : 21.4 	Sandwich type libanais (fallafel)	
	Pamplemousse frais		Energie : 3.2 Sucres : 18  Graisses : 0.1  Sel : 0  Vitamines : 13 	Fraise		Energie : 2.2 Sucres : 10.1  Graisses : 0.1  Sel : 0.1  Vitamines : 16.4 	Mousse de fruit		Energie : 5.9 Sucres : 2.5  Graisses : 21.4  Sel : 2  Vitamines : 11.2 	Salade de fruits	

## Two populations, three conditions

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To investigate the role of

- **cognitive resources** and
- **time**

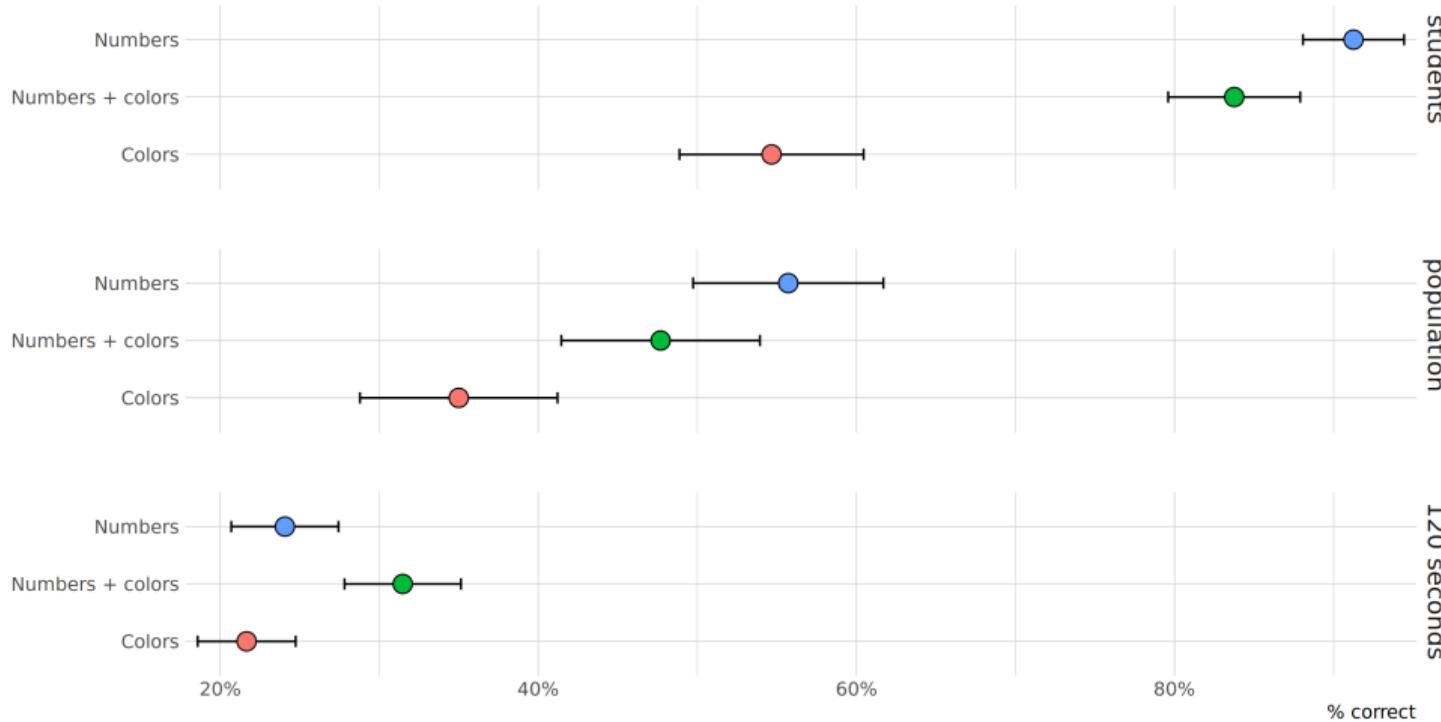
we run three conditions:

**Students** Highly skilled engineering students, no time limit, paper and pencil

**Population** Population at large, no time limit, paper and pencil

**120 seconds** Population at large, 120 seconds, NO paper and pencil

## Performance in the task - 4 constraints



## What do we **learn**?

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Numeric labels **win**, *in a task that was tilted in their favor, only*:

- among a *very* high skilled population
- with plenty of time and computational aid (paper and pencil)

But if *time* and *cognitive resources* are **limited**

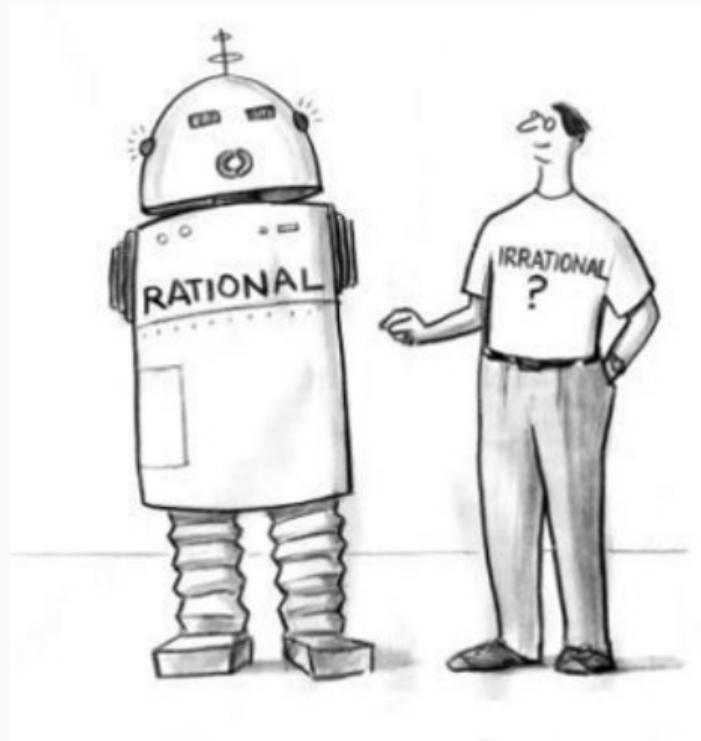
- Numbers and colors equal
- Number + colors overall better
- Dismal performance in all cases

## **Study 2: *fast* vs. *slow* decisions**

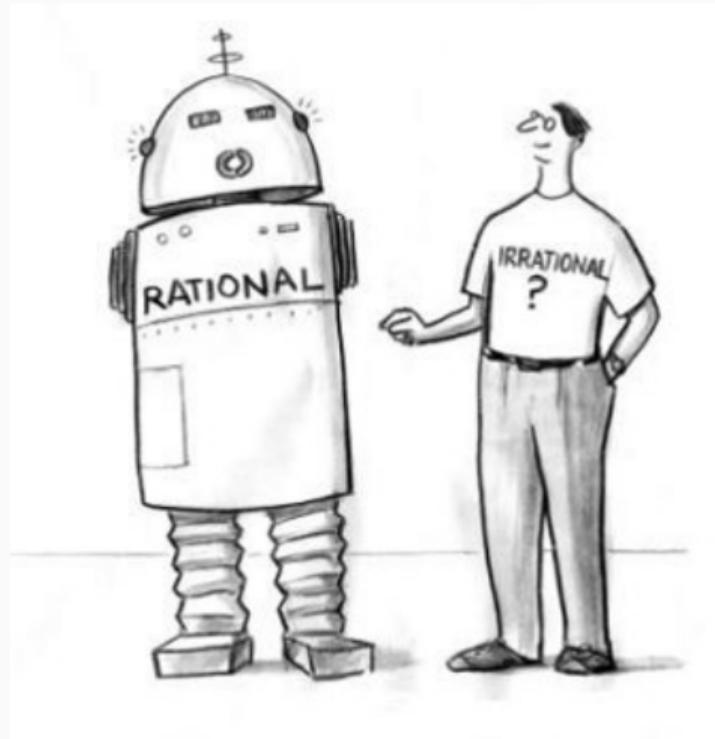
### **[WIP – L. Muller]**

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## Food choice: **fast** or **slow**?



## Food choice: **fast** or **slow**?



Food choice is both fast & slow

**Food choice** : fast

**Health goals** : slow

Labels are both fast & slow

**numbers** : slow

**analytic** : slow

**colors** : fast

**aggregate** : fast

# Nutrition Facts

Serving Size 2 CUPS (30g)

Servings per Container VARIED

## Amount per Serving

**Calories** 150    **Calories from Fat** 70

% Daily Value\*

**Total Fat** 7g    11%

  Saturated Fat 1.5g    6%

**Cholesterol** 0mg    0%

**Sodium** 120mg    5%

**Total Carbohydrate** 20g    7%

  Dietary Fiber 4g    15%

  Sugars 9g

**Protein** 1g

**Vitamin A** 0%    •    **Vitamin C** 0%

**Calcium** 0%    •    **Iron** 2%

\* Percent Daily Values are based on a 2,000 calorie diet. Your daily values 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

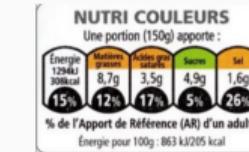
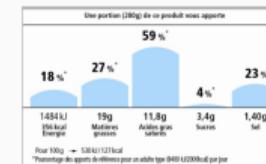
Sodium    Less than 2,400mg    2,400mg

Total Carbohydrate    300g    375g

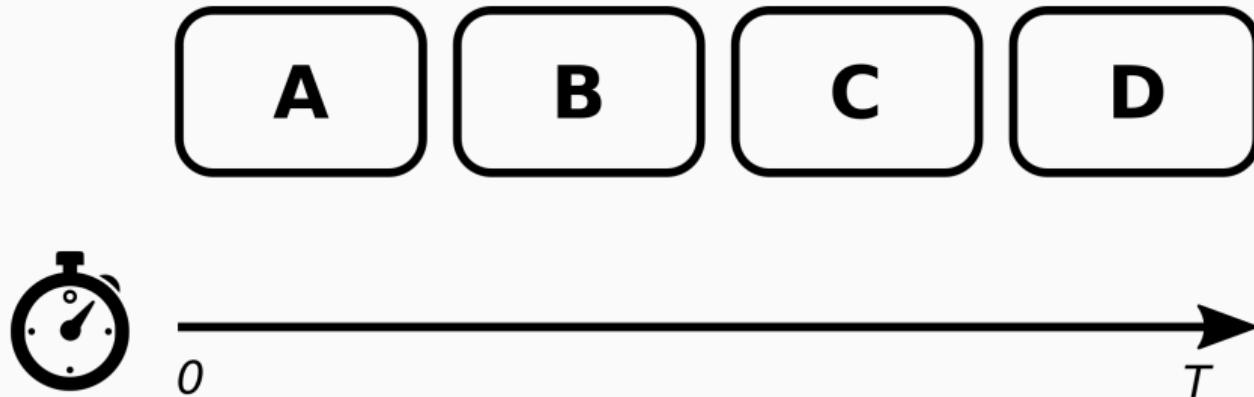
  Dietary Fiber    25g    30g

Calories per gram:

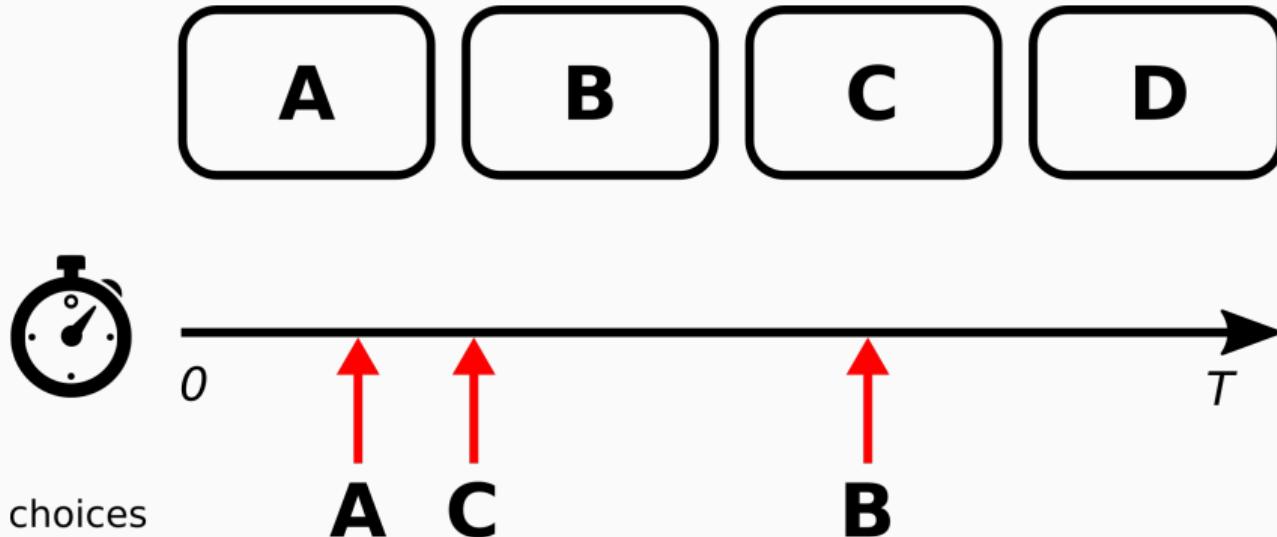
Fat 9    •    Carbohydrate 4    •    Protein 4



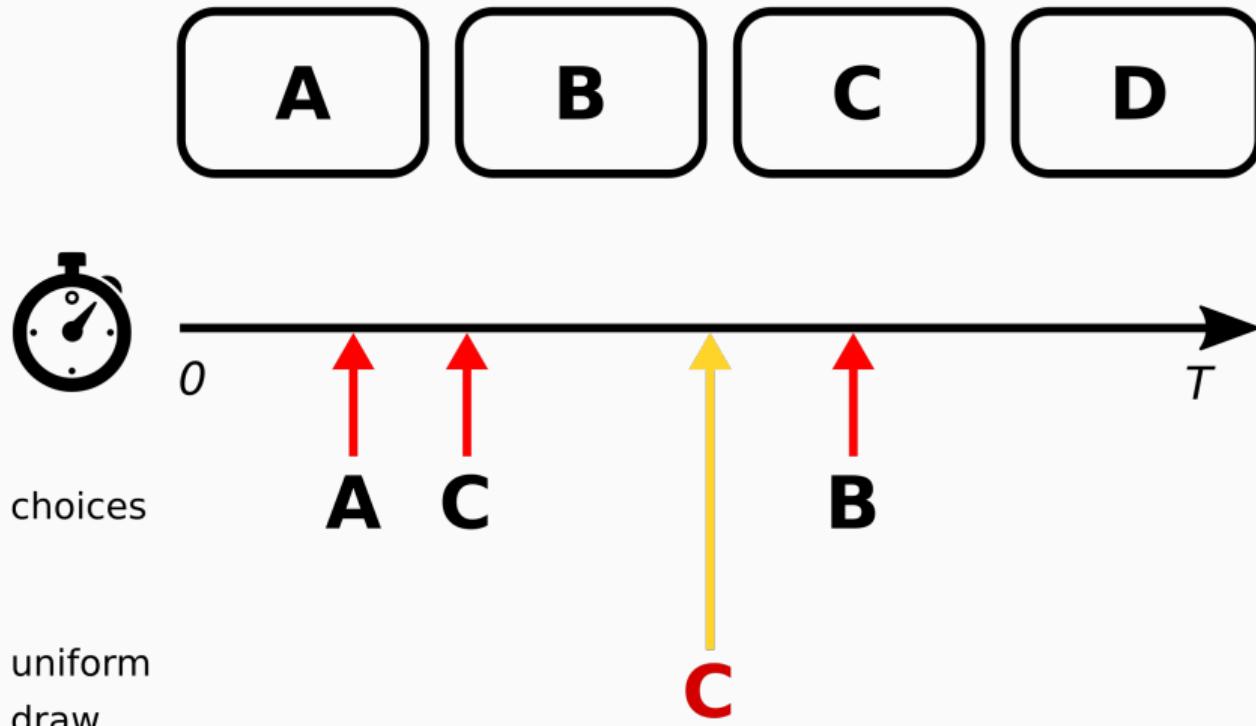
A special input mechanism: don't take your time!



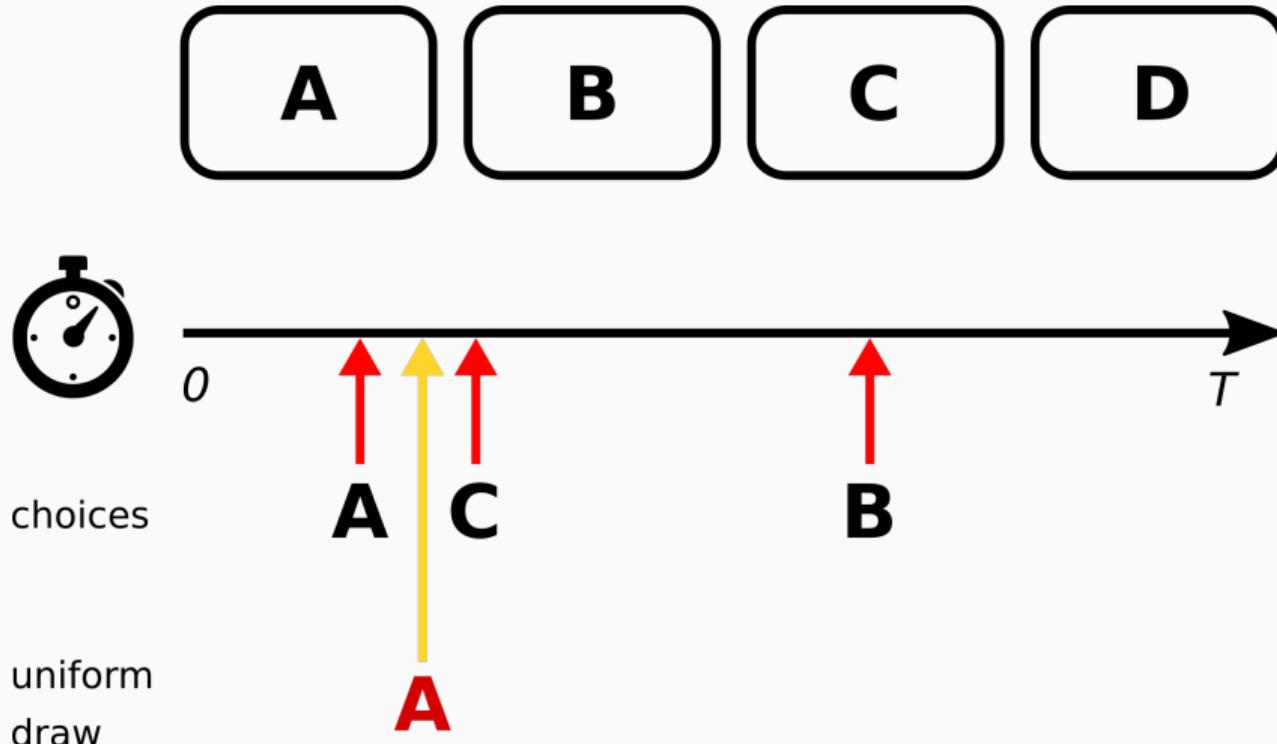
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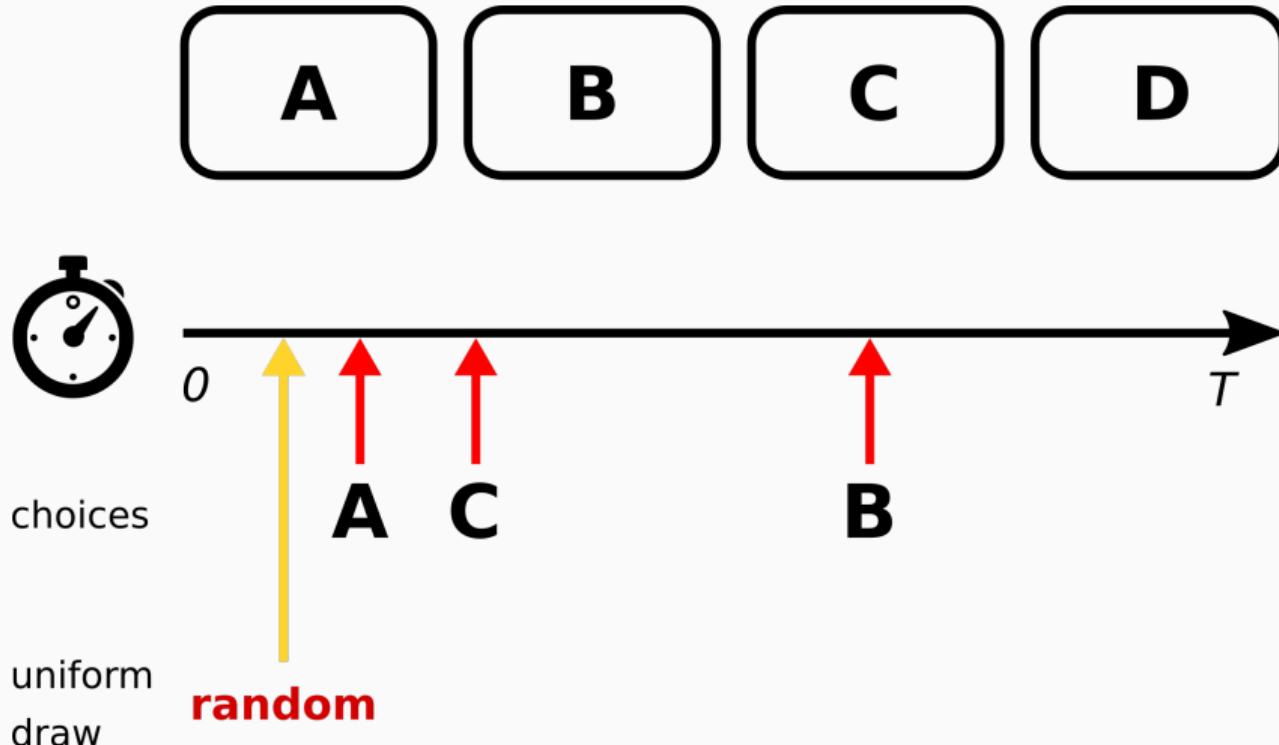
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## A special input mechanism: don't take your time!



## Our design

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- No choice  $\Rightarrow$  random choice: incentive to *fast* reply
- Time is ticking: incentive to *change* first decision upon reflection
- fast to slow *endogenous* (usually: exogenous)
- Data reveal choice *process* (usually: outcome)

This allows us to:

- tell apart how different labels tap on different heuristics
- measure how much *faster* colors are
- assess if numbers do a better job, and when

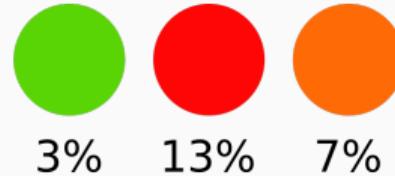
## Labels

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global  
quality



fat sugar salt



fat sugar salt



fat sugar salt



Barres au chocolat au lait  
et aux céréales



Choisir

Barres chocolatées  
fourrées au lait et aux noisettes



Choisir

Barres chocolatées  
au caramel



Choisir

Barres de céréales  
raisins et chocolat au lait



Choisir



Barres au chocolat au lait  
et aux céréales

Sucre



AGS



Sel



Choisir

Barres chocolatées  
fourrées au lait et aux noisettes

Sucre



AGS



Sel



Choisir

Barres chocolatées  
au caramel

Sucre



AGS



Sel



Choisir

Barres de céréales  
raisins et chocolat au lait

Sucre



AGS



Sel



Choisir

Barres au chocolat au lait  
et aux céréales

Sucres	AGS	Sel
55%	110%	5%

Choisir

Barres chocolatées  
fourrées au lait et aux noisettes

Sucres	AGS	Sel
46%	87%	5%

Choisir

Barres chocolatées  
au caramel

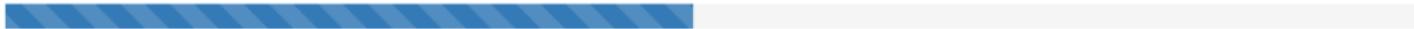
Sucres	AGS	Sel
69%	42%	7%

Choisir

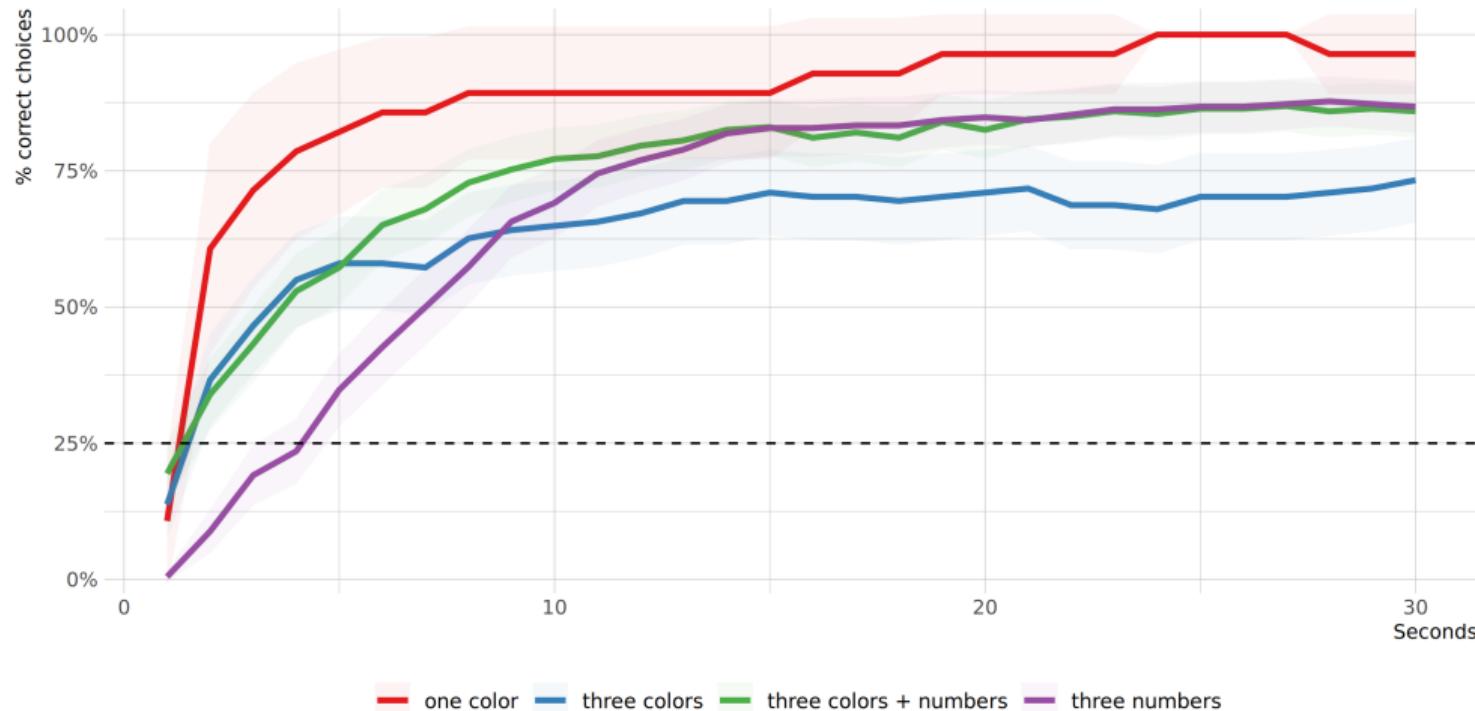
Barres de céréales  
raisins et chocolat au lait

Sucres	AGS	Sel
34%	27%	9%

Choisir



## Share of correct choices in time, by labeling scheme



from Fast&Slow labels, wip

## What do we **learn**?

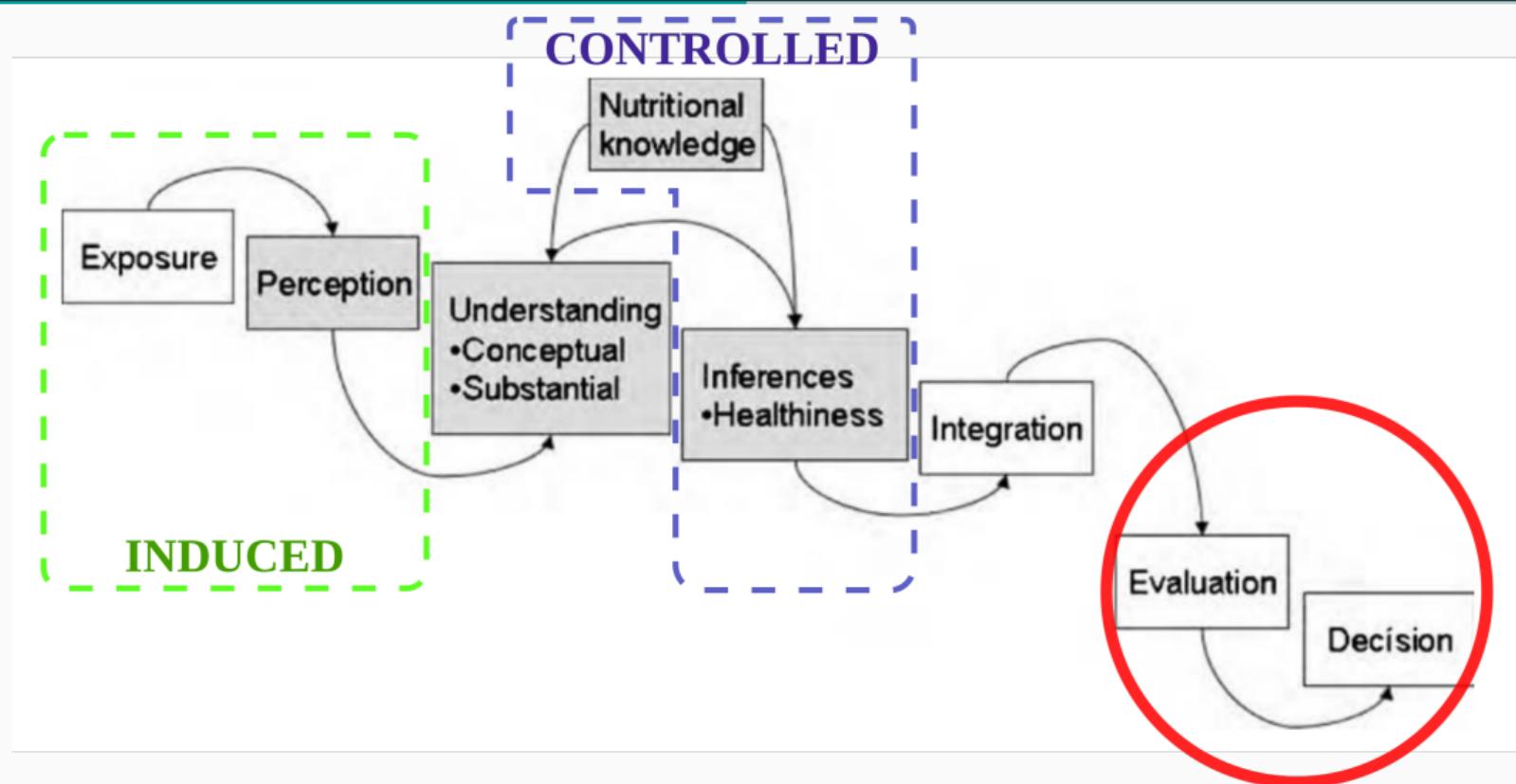
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- Trade off time/accuracy
- Heuristics give way to computation in time
- Indirect evidence of different cognitive processes
- We explicitly measure 'how more intuitive' colors are

## Part 2: building counterfactuals

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# Building counterfactuals



**Study 3: shopping with labels**  
[ERAE 2019 – L. Muller, B. Ruffieux,  
A. Lacroix]

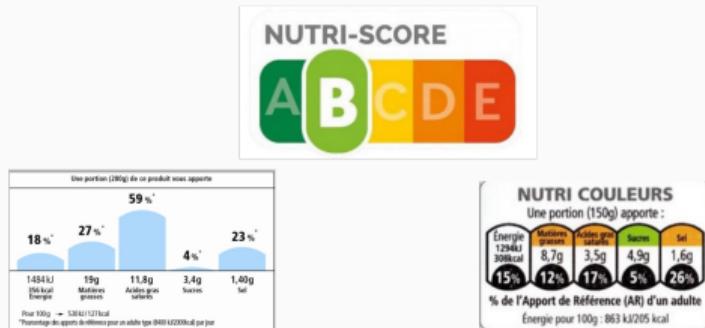
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## Some French context



Fédération du Commerce  
et de la Distribution

# Ex-ante policy evaluations, 2016-17



## French Ministry of Health – 2016

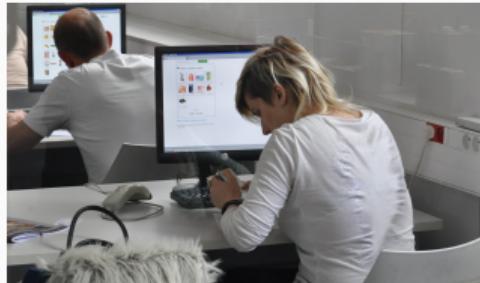
- Which FoPL to choose?
- How large is the effect?
- A RCT in 60 French supermarket
- A large lab experiment (**us!**)

# Our setup

Paper catalog



Computer interface



Real products



# Our setup

Paper catalog



Computer interface



Real products



- Subjects shop for real in the lab
- For two days for their household

- $\sim \frac{1}{4}$  of product supply available
- chosen + we have it  $\Rightarrow$  buy

## A large and **representative** catalog

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- 290 products
- 37 food categories
- custom e-shopping interface
- barcode scanners on the desk
- price, quantity, picture (label) up front
- nutritional table and ingredient list available upon clicking

1138

Chercher

 Votre caddie actuel :

Aucun produit dans ce panier.

Terminer

4,25 €  
260 g  
16,35 €/Kg

- 1 +

Ajouter au caddie

2 Steaks hachés pur boeuf 15% mg



Ingrediénts

Valeurs nutritionnelles

1814

 Chercher

1814

Votre caddie actuel :



x 1 = 4.25 €



x 1 = 3.95 €



x 1 = 1.17 €



x 1 = 2.83 €



x 3 = 1.83 €

5 Articles

**Total = 14.03 €**

- 1 +

Ajouter au caddie

Frites surgelées pour micro-ondes

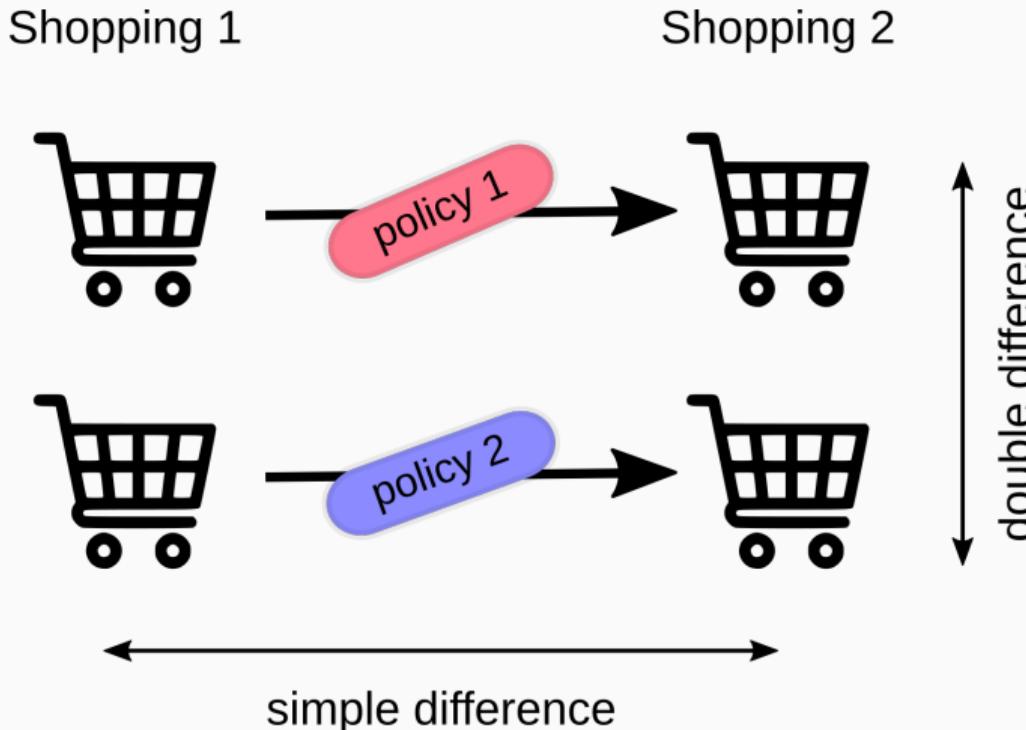


Ingédients

Valeurs nutritionnelles

Terminer

## Experimental design: difference-in-difference



# Treatments

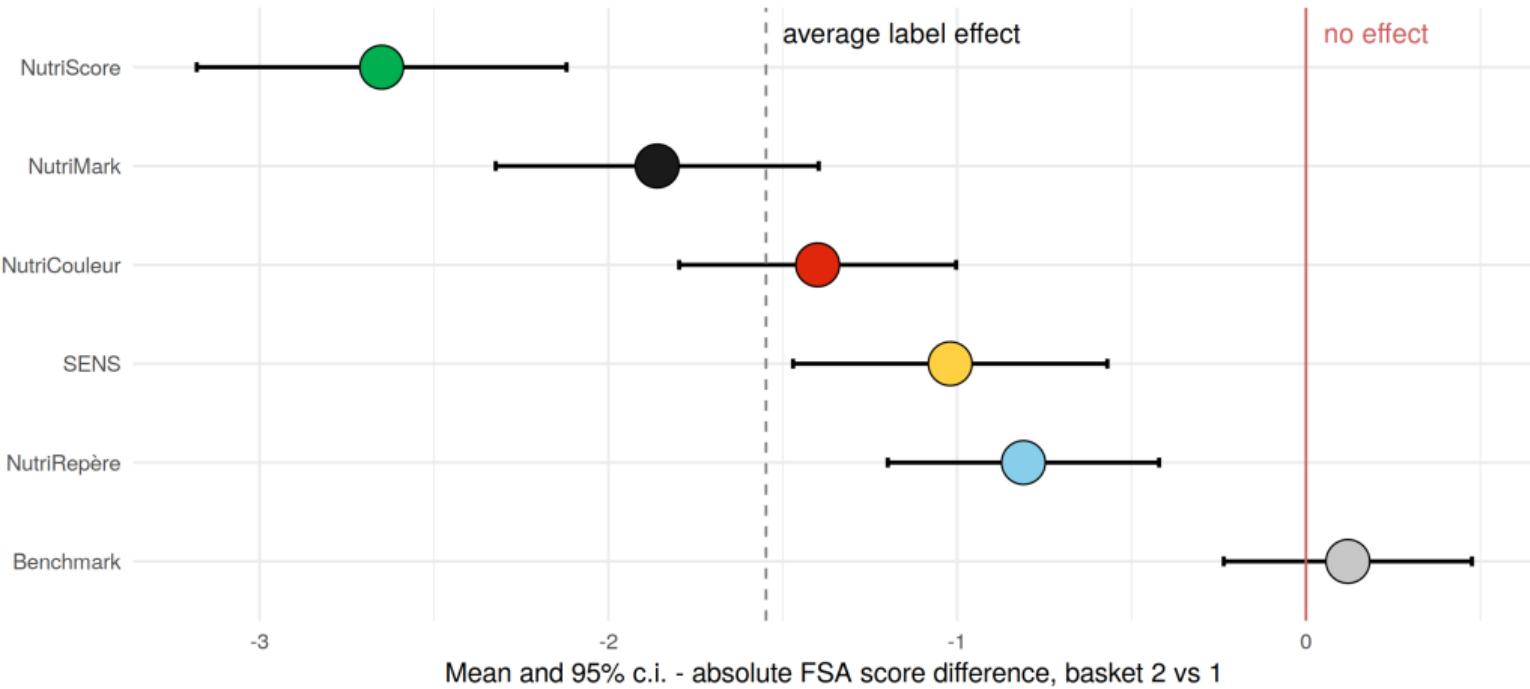


## A large and representative subject pool

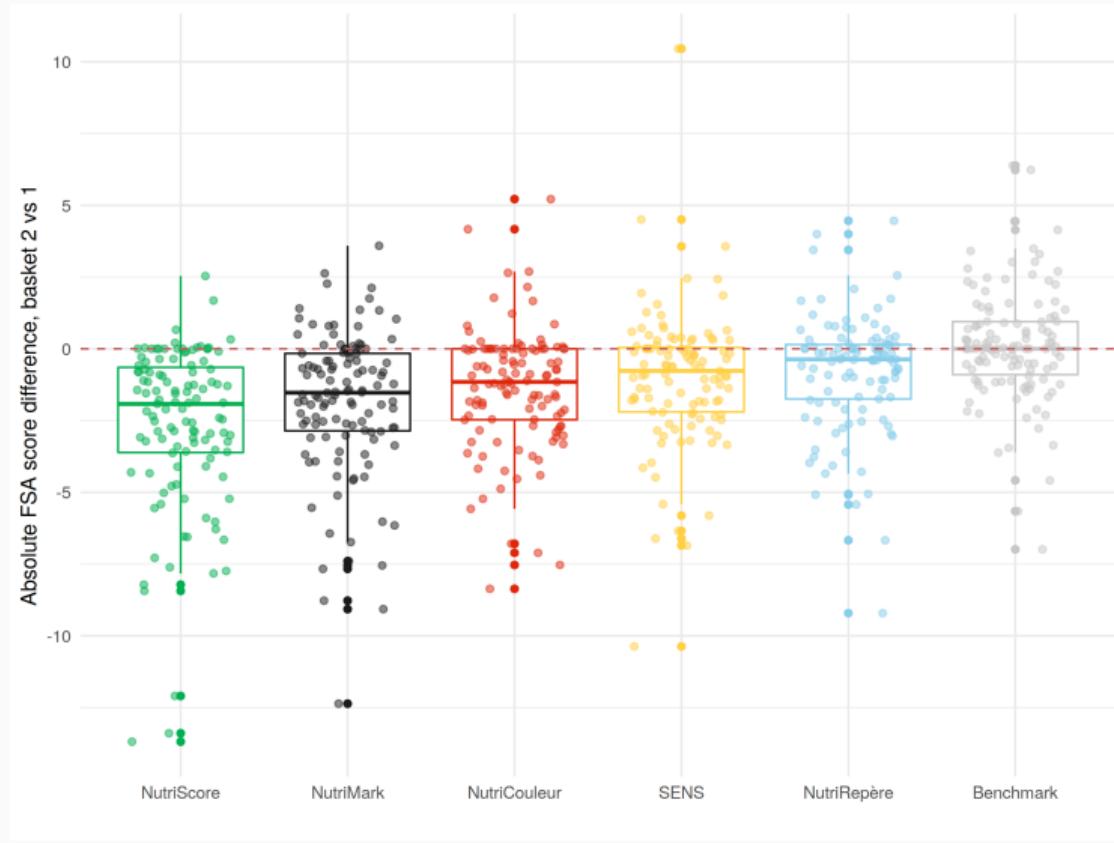
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- 691 subjects
- ~ 110 for each of 6 treatments
- sample issued from the general population
- (recruiting agency boosted our reach into all socio-economic statuses)
- roughly representative

# NutriScore **leads**, by far



## But there is heterogeneity



## What do we **learn**?

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- Color-coded, summary labels perform best
- (but only if they directly relate to quality – not SENS)
- Number-based, analytic labels perform worse

## Policy implications

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- NutriScore officially selected in FR (and ES, BE, DE...)
- Adopted by Auchan, Fleury Michon, Leclerc, Casino, Nestlé...
- The very idea of FOP labels validated
- NutriScore is being proposed by France as EU standard
- Even though *some countries* really do **not** like it

PRESENTI . . . . . 311  
VOTANTI . . . . . 299  
ASTENUTI . . . . . 12  
MAGIORANZA . . . . 150  
FAVOREVOLI . . . . 299  
CONTRARI . . . . . 0

APPROVATO

MARTEDÌ

10

OTTOBRE

## **Study 4: Lab vs. Field [with L. Muller]**

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## Lab vs Field at a glance

	Lab	Field
Location Supermarkets	Grenoble "1"	Paris couronne, Nord, Lyon 60
Task	shop for two days' worth	shop
Real purchases	"yes"	yes
Measure	FSA score for 2000Kcal	FSA score for 2000Kcal
Design	Diff-in-diff	Diff-in-diff
Time frame	2x, same session	5 weeks, 1 year apart
Participants	691	171.827
Products (of which labeled)	290 (all)	3586 (1266)
Food categories	37	4
Purchases	27.882	1.668.301
Manpower needed	8	~ 100
Cost	~100k	~4 million

## Field study: "instructions"



# Field study: product display



## Field study: alerting the subject to the experimentation



# Results: the lab as a magnifying glass

Label	Field	$\Delta$ score FSA	Corr	Zoom
Lab				
		-0.142*	-2.766***	19x
		-0.115	-1.513*	0.88
		-0.062	-1.140	18x
		-0.024	-0.924	38x

## **Study 5: environmental labels?**

**[WIP – P. De Lattre, L. Muller]**

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# The design can be used for environmental labels too

Référence



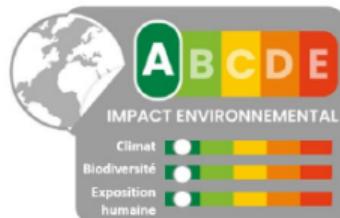
Note



Multiplicateur



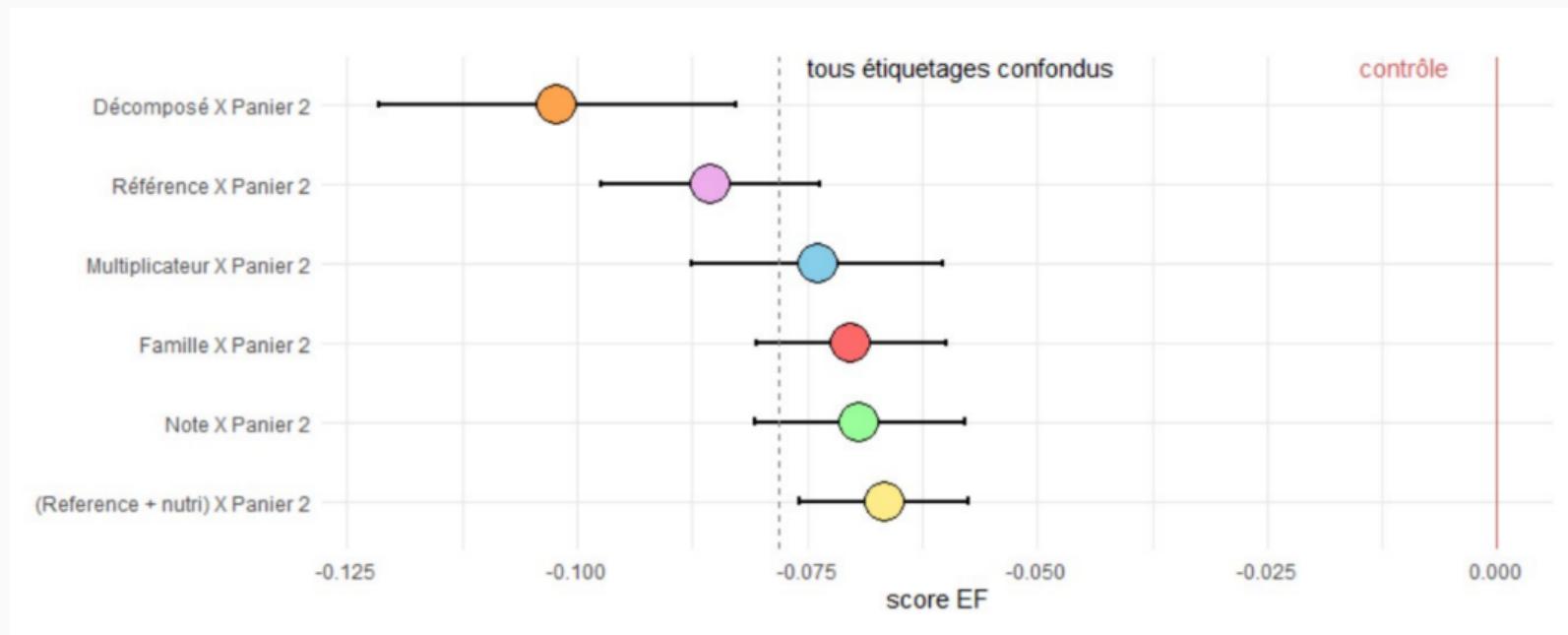
Décomposé



Famille



## The design can be used for environmental labels too



## What do we **learn**?

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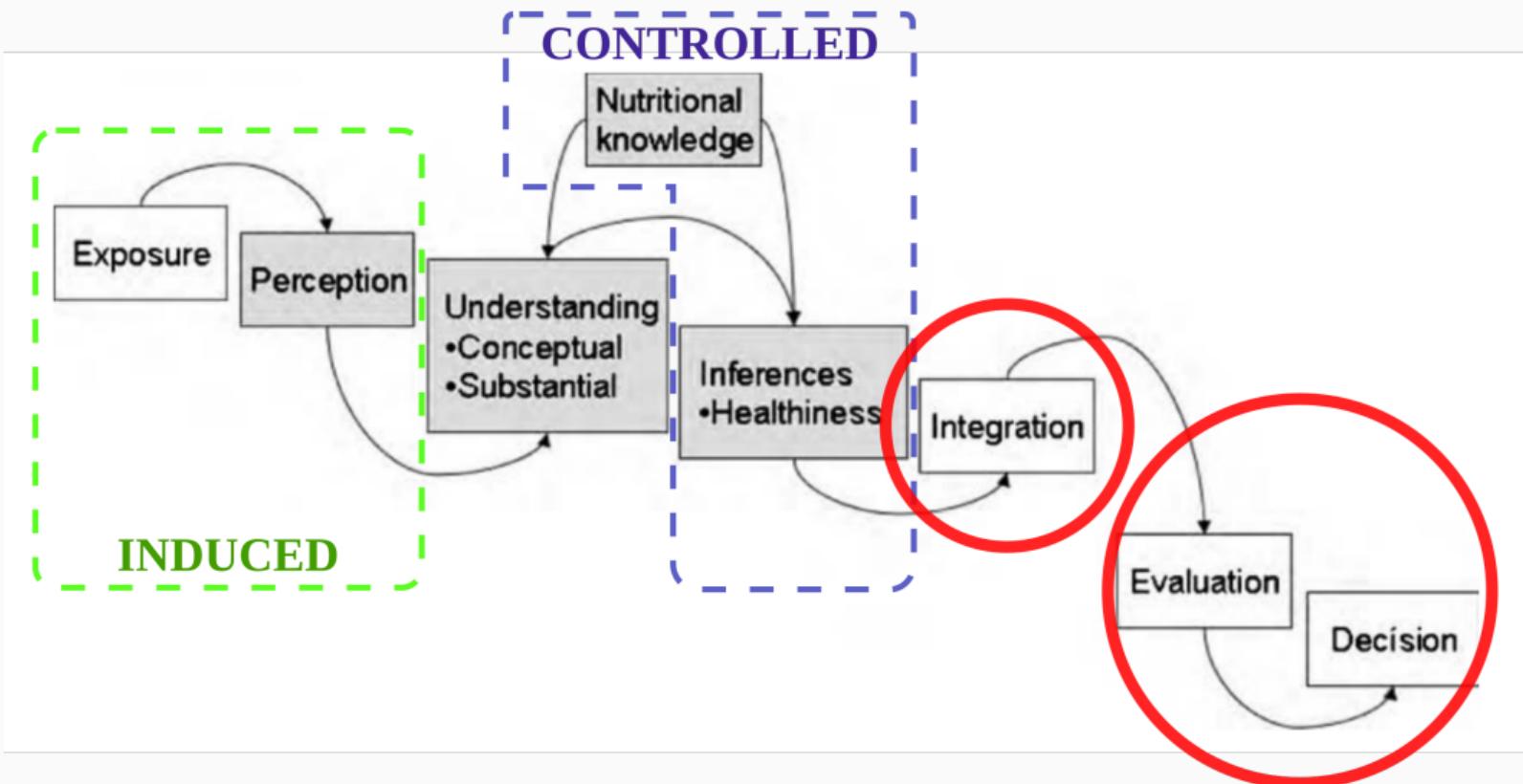
- Consumers seem to take environmental labels into account
- Small effects
- Different formats do not make different impacts

## **Study 6: labels or prices?**

**[JEBO 2023 – L. Muller, B. Ruffieux]**

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## Integrating different policies



## A bit of context

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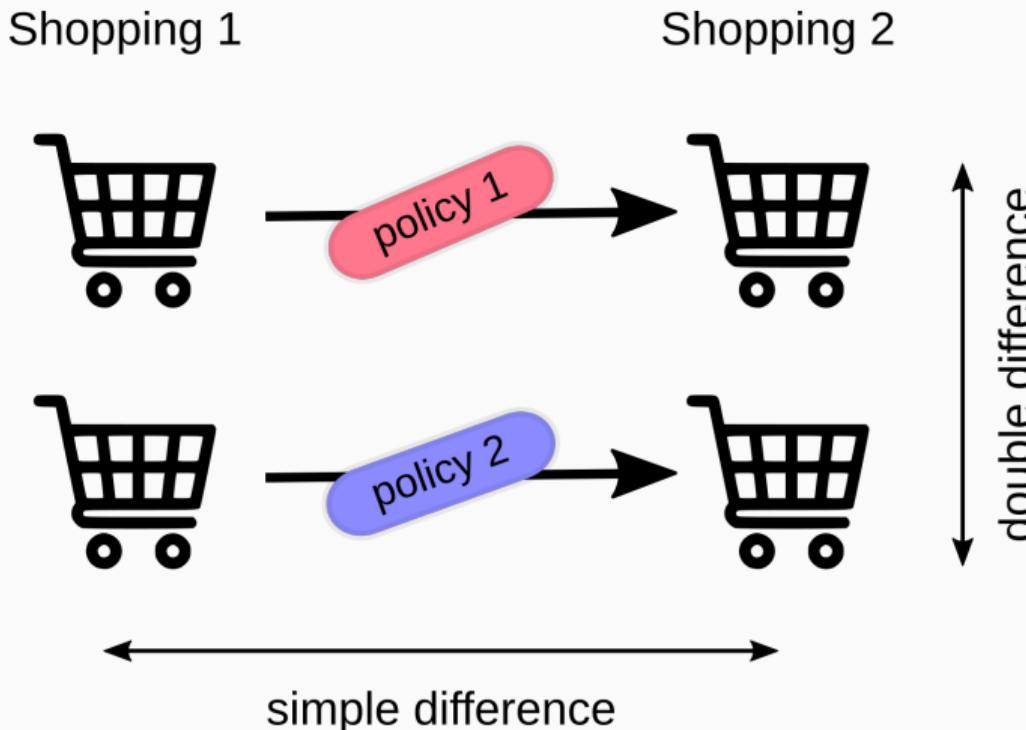
- World Bank: strongly tax unhealthy foods (Shekar and Popkin 2020)
- WHO: introduce dietary taxes on unhealthy food of minimum 20%
- India and Mexico tax unhealthy food & beverages (India : tax of 28%).

## Question and design

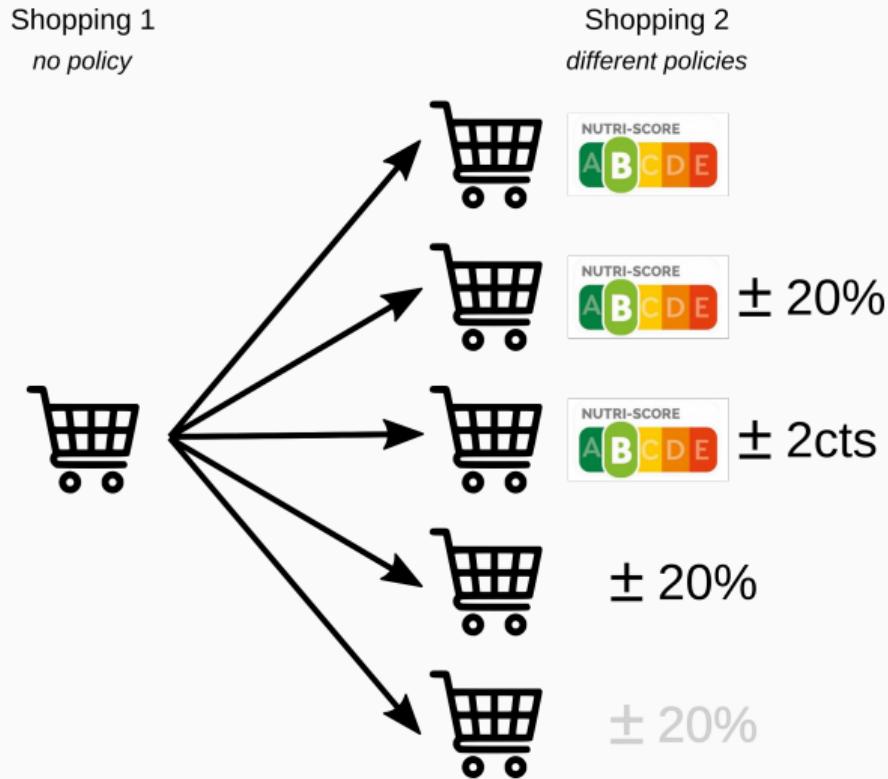
- Suppose we want to couple a **label** with an **incentive** scheme
- e.g. tax unhealthy (soda tax) and subsidize healthy food.
- Does it work? How?
- Will the intervention be (sub/super)additive?
- i.e. label **or** price  $\geq$  label **plus** price?

Exact same design as Study 2

# Experimental design: difference-in-difference



# Treatments



## A large price change: $\pm 10\%$ or $20\%$



## A small price change: $\pm 1$ or $2$ cents



# Stimuli

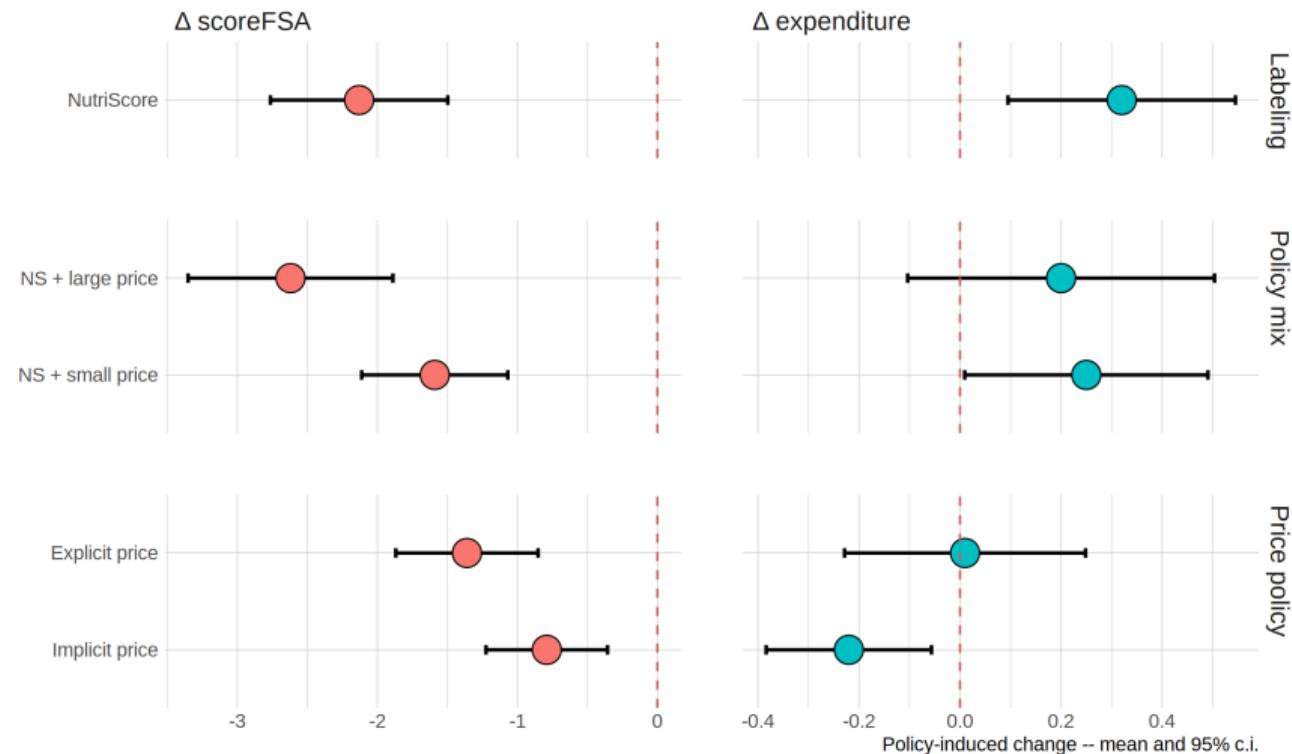
Référence Sans Nutri-Score sans Bonus-Malus	Traitement 1 Nutri-Score sans Bonus-Malus	Traitement 2 Nutri-Score avec Bonus-Malus explicite <i>ad valorem</i> de niveau élevé
<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 5,00€/kg</p> <p>1,00€</p>	<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 5,00€/kg</p> <p>1,00€</p>	<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 4,00€/kg</p> <p>1,00€</p>
<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 4,00€/kg</p> <p>NUTRI-SCORE ABCDE</p> <p>1,00€</p> <p>0,98€</p>	<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 4,00€/kg</p> <p>0,80€</p>	<p>Cacahuètes grillées très pauvres en sel</p>   <p>200g 4,00€/kg</p> <p>1,00€</p>

## A large(ish) and representative subject pool

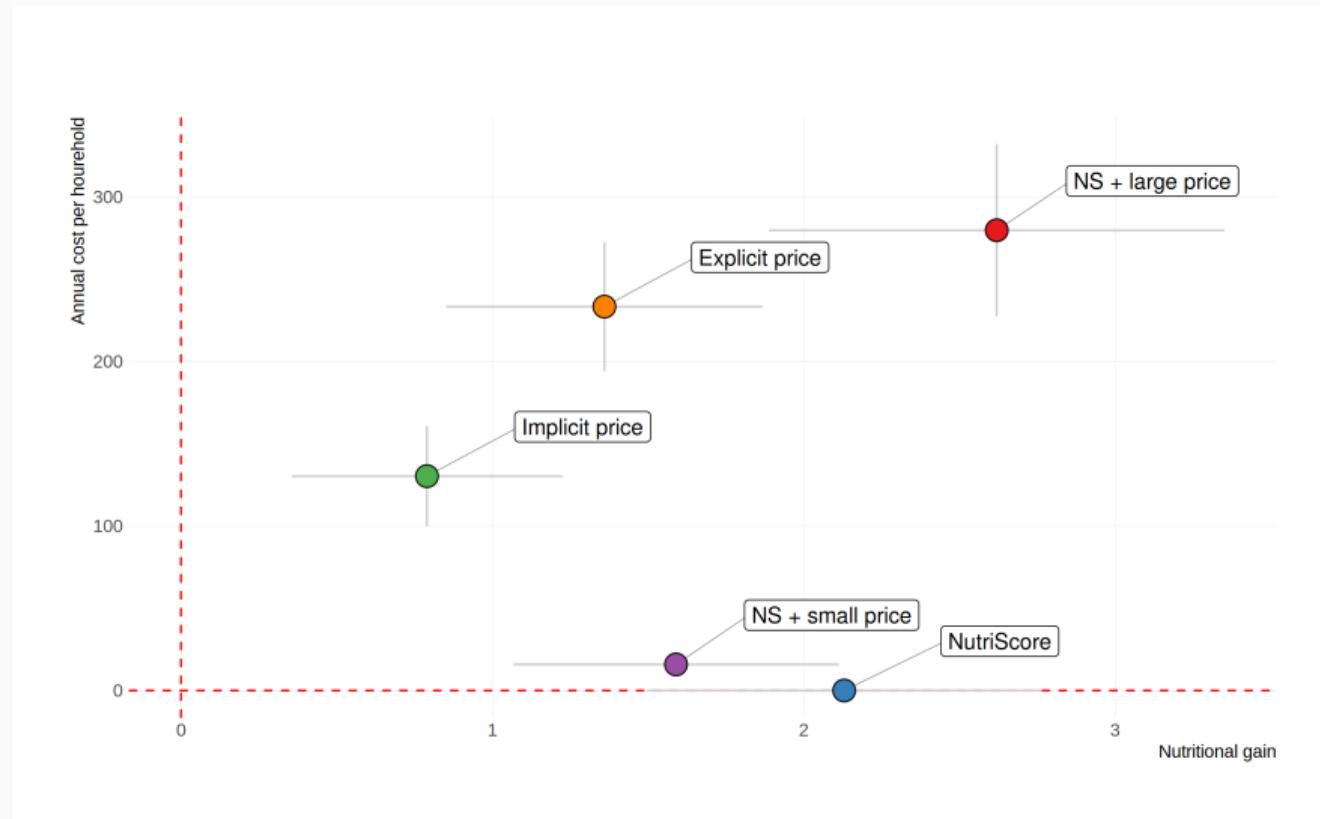
---

- 386 subjects
- $\sim 75$  for each of the 5 treatments
- sample issued from the general population
- roughly representative (++women, ++educated)

# Results



## Welfare analysis: trade-offs



## What do we **learn**?

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- Nutritional policies are subadditive
- *Too small* an incentive reduces the effect (Gneezy & Rustichini)
- Price policies have better be explicit (Chetty et al.)
- Labeling appears as more cost-effective than the policy mix
- ...still, it's just the lab!

## **Part 3: does it matter?**

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## Study 7: epidemiology

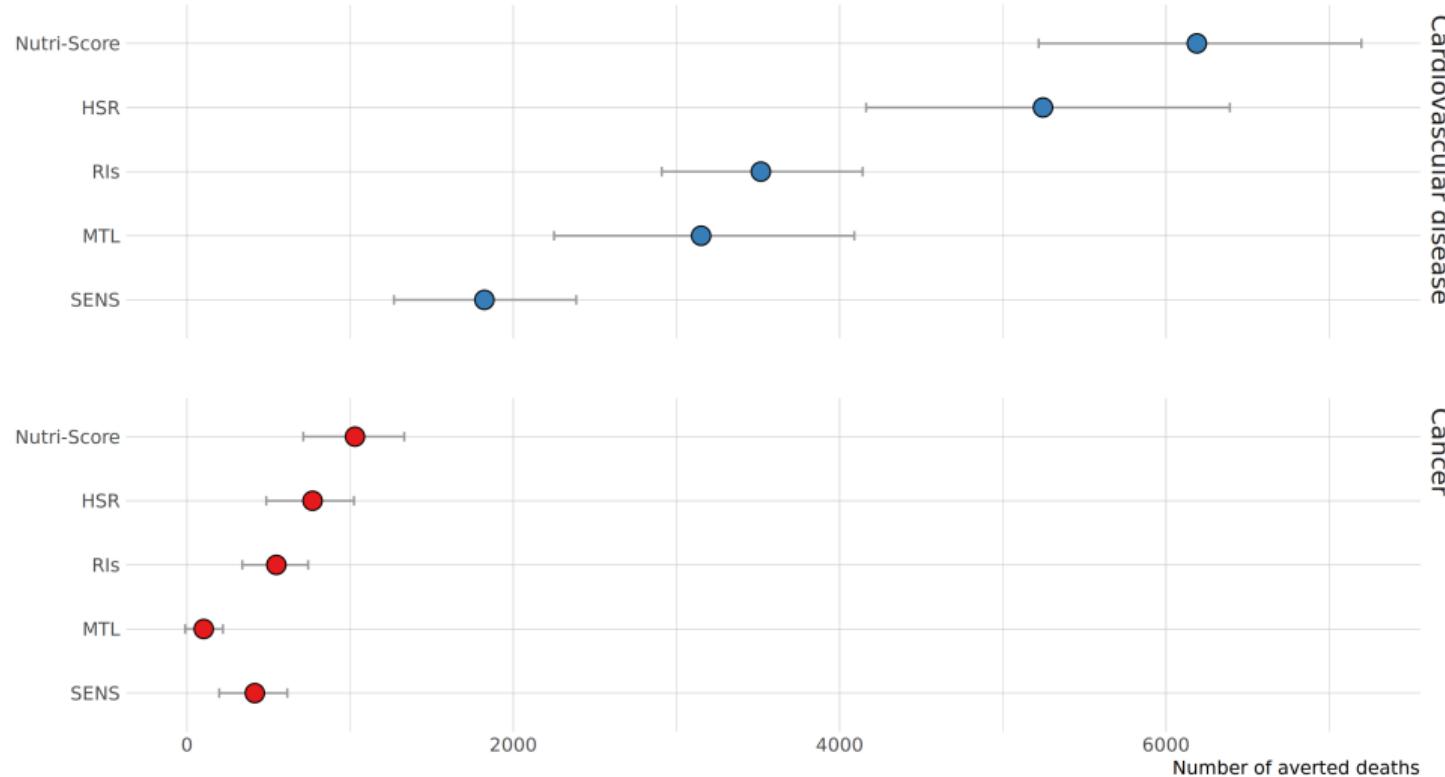
[IJBNPA 2019 – S. Hercberg et al]

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# From micro to macro



# Estimated number of averted deaths, France



## What do we **learn**?

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- Labels have **non-negligible** impacts on mortality
- Results from the lab can be used to feed **macro** models
- Better, intuitive labels are used and save lives.

## Open questions

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## Some open questions: consumers

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### How **externally valid** are our results?

- Integration: can all the effects just be summed up?
- Label proliferation
- Information overload

## Some open questions: **consumers**

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### How **externally valid** are our results?

- Integration: can all the effects just be summed up?
- Label proliferation
- Information overload

### What **other** forces are at play?

- Cultural arena: the battle for label perception
- Nutrition vs tradition
- A contrarian view from Italy

## Some open questions: firms

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### Firms react **strategically**

- Price discrimination
- Multiple labeling
- Labels as anti-competitive devices

## Some open questions: firms

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### Firms react **strategically**

- Price discrimination
- Multiple labeling
- Labels as anti-competitive devices

### Interaction firm/consumers

- Labels working for the *wrong* reasons
- Normative messages
- "*Bisogna che tutto cambi, affinché tutto resti uguale*"

Thank you

## References

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## Links to papers

**Study 1** J Eco Psy 2015 – [Numbers or Colors?](#)

**Study 3** Er Rev Ag Econ 2019 – [5 labels comparison](#)

**Study 6** JEBO 2023 – [Labels or prices?](#)

**Study 7** IJBNPA 2019 – [Tracking macro health consequences](#)