

Product Labeling for Inattentive Consumers

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Columbia University - Industrial Organization Colloquium

April 25, 2018

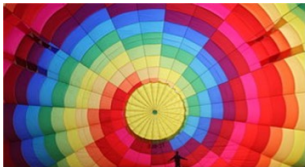
Motivation

- ▶ Labeling as “coarse information” (category) about the product
- ▶ *Reduced fat milk* label: fat content $\leq 2\%$
- ▶ Nutrition facts: Indicates the precise amount
- ▶ Why using labels?
- ▶ No effect on a fully attentive or fully inattentive consumer
- ▶ Possible effect only on partially attentive consumers
 - ▶ Information collection is costly
 - ▶ Decreasing returns of information collection
- ▶ Discrete choice model with cognitively bounded consumers

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Perception problem vs Choice problem



Literature review

► Behavioral and cognitive economics

- Categorization and optimal coding: Krueger and Cement (1994), Arthur (1994), Gennaioli and Shleifer (2010), Manzini and Mariotti (2012), Robson and Whitehead (2017)
- Perceptual bias: Koszegi and Szeidl (2012)
- Rational inattention: Sims (2003), Gabaix (2014)

► Discrete choice models

- Empirical literature: Kiesel (2007), De los Santos et al. (2012), Allcott et al. (2017), Moraga-Gonzales et al. (2017)
- Bounded rational consumers: Mehta et al. (2003), Abaluck (2011), Pires (2015), Clerides and Courty (2015), Pires (2015)
- Rational inattention to discrete choices: Matejka and McKay (2015) [Shannon entropy], Fosgerau et al. (2017) [Generalized entropy], Cheremukhin et al. (2015), Matejka (2016), Aguiar and Riabov (2017), Fosgerau et al. (2018)
- Information design, regulation: Diamond (1989), Mullainathan et al. (2008), Piccione and Spiegler (2012), Hui et al. (2017)

Research proposal

1. How do quality labels affect consumer behavior?

- ▶ Jointly estimate preference and information cost
- ▶ This is my challenge for the 2YP

2. How do quality labels affect producers?

- ▶ Price adjustment under stable product set
- ▶ Quality adjustment
- ▶ Reduced form analysis only

3. Counterfactual: Alternative labeling schemes

- ▶ This is a classic information design problem
- ▶ In this model more (coarse) information is not necessarily better

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Intuition of the model

- ▶ A consumer faces a two-stage choice problem: collect information, then choose a product
- ▶ At the begin (no information) the product space is blurry: all the products (e.g. refrigerators) look identical
- ▶ No information: binary decision buy/no buy, no choice
- ▶ By collecting information the consumer can reduce coarseness
- ▶ Infinite information leads to perfect refinement
- ▶ Information is costly, it takes time to refine the space
- ▶ Intuition: reduce the variance of the T1EV logit error
- ▶ A label delivers immediate, free, yet coarse information

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Intuition of the model

- ▶ Three categories of agents: consumers (C), retailer (R), and an information designer (ID)
- ▶ The information designer decides the minimum requirement for a label that is displayed on all the high-quality products
- ▶ Product quality and label are exogenous for the retailer (monopolist producer) who only sets prices
- ▶ The consumers faces a two-stages discrete choice problem: select information level, then choose a product

What is a good industry for this model?

1. European Union Energy Label

- ▶ 2010 reform: change in the regulation
- ▶ 25 product categories: heterogeneous requirement

2. US Energy Star Label

- ▶ 2011 reform: certification is more demanding
- ▶ product heterogeneity as above

3. US FDA food labeling regulation

- ▶ 2011 Nutrition Keys front-of-pack labeling
- ▶ Additional to previously adopted standard labels

1. European Union Energy Label

- ▶ EU Directive 1992/75/EC established the energy consumption labeling scheme A-G to indicate energy efficiency classes
- ▶ Replaced by Directive 2010/30/EU that introduces A+, A++, and A+++ in the attempt to keep up with advances in energy efficiency
- ▶ 2010 regulation (applied since 31 July 2011) introduces new high-quality levels (A+) and modifies the threshold for the different classes
- ▶ 25 categories, including fridges, dishwashers, air conditioners

EU Energy Label: Identification strategy

- ▶ Geographical/temporal heterogeneity in prices and choice sets
- ▶ IV price of energy: exogenous wrt product characteristic
- ▶ Consumers differ in preference over observable characteristics (e.g. brand, power, size,...), coarse characteristics (e.g. energy consumption), and price elasticity
- ▶ Quasi-natural experiment: 2010 regulation change
- ▶ Transition phase: from announcement to implementation
- ▶ Post-change phase: after 1 August 2011
- ▶ Stated preference survey data: Heinzle and Wustenhagen 2012

2. US Energy Star Label

- ▶ Energy guide appliance label of qualifying products, aka Energy Star
- ▶ Voluntary program launched by US Environmental Protection Agency (EPA)
- ▶ 75 product categories, it indicates cost-saving energy efficient products
- ▶ Specifications differ across item categories: average refrigerators need 20% savings over the minimum standard, dishwashers need at least 41%
- ▶ Established in 1992, reformed in 2010 because of a scandal (before it was for the most part a self-certification program, vulnerable to fraud and abuse)
- ▶ Since 2011, product are tested in an EPA-recognized laboratory (recognized certification)

3. FDA food labeling regulation

- ▶ 1990: Nutrition Labeling and Education Act (NLEA) implemented in 1994. It requires all packaged foods to bear nutrition labeling and all health claims, labels, and serving size to be standardized.
- ▶ 2011: The Grocery Manufacturers Association announces Nutrition Keys (now Facts Up Front), a voluntary front-of-pack labeling system, just months before the FDA is to issue its guidance to industry on the matter.
- ▶ Coarse labels such as “low calorie” are now less informative as they are combined with other accessible information

Food data: Nielsen HH panel and USDA Nutrient DB

1) Nielsen Consumer Panel Data: longitudinal home scanner data, 40,000 United States households, 2004-2016.

- ▶ Demographic and geographic variables (household)
- ▶ Purchases: date, the UPC code (for each product purchased), quantity, price.
- ▶ Products and product characteristics: all 10 Nielsen food and nonfood departments; UPC code and description, brand, multi-pack, and size, product group, and product module. Some products contain additional characteristics (e.g. flavor).

2) USDA Nutrient Database for Standard Reference

- ▶ Database for branded food products
- ▶ Energy, protein, lipid, fiber, sugars, and minerals (per 100 g)

Model: Framework

- ▶ Products are vertically differentiated and fully represented by the pair of quality and price (x_j, p_j)
- ▶ Three categories of agents
 - ▶ Information designer
 - ▶ Selects the requirement x^* for the high-quality label
 - ▶ Agnostic about the objective (agentic/paternalistic?)
 - ▶ Producers
 - ▶ Set of J distinct products that differ in quality $0 \leq x_j \leq 1$
 - ▶ Visible high-quality label if $x_j \geq x^*$
 - ▶ Maximize joint profits by choosing prices $\{p_j\}_{j=1}^J$
 - ▶ Consumers
 - ▶ Each consumer can buy a single unit, or not buy
 - ▶ Distribution $f(\beta)$ of consumers with pref. over quality $\beta \geq 0$
 - ▶ Constant price elasticity $\alpha = 1$ and information cost $\lambda \geq 0$
 - ▶ Maximize expected utility by choosing info. ρ and product x_j

Model: Two-stages choice problem

1) Choose optimal precision level ρ

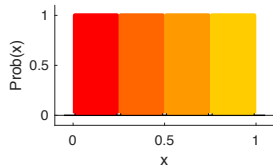
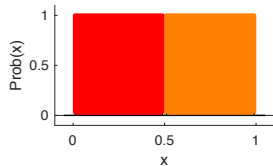
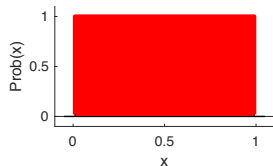
$$\max_{\rho \geq 0} EU(\rho) := EU_{\beta}^*(\rho) - \lambda \cdot \rho$$

2) Given ρ , select a random product from the best partition cell

$$EU_{\beta}^*(\rho) := \max_{c \in [1, \dots, 2^{\rho}] \cup \emptyset} \mathbb{E} \left[\beta \cdot q(x_j) - p_j \mid x_j \in \Pi_{\rho}^c \right]$$

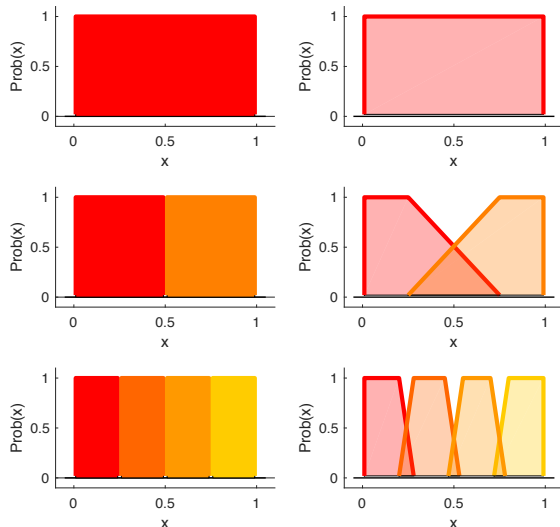
Consumer's individual choice is summarized by the pair (ρ, Π_{ρ}^c) , that is associated with a pdf over products.

Model: Information acquisition



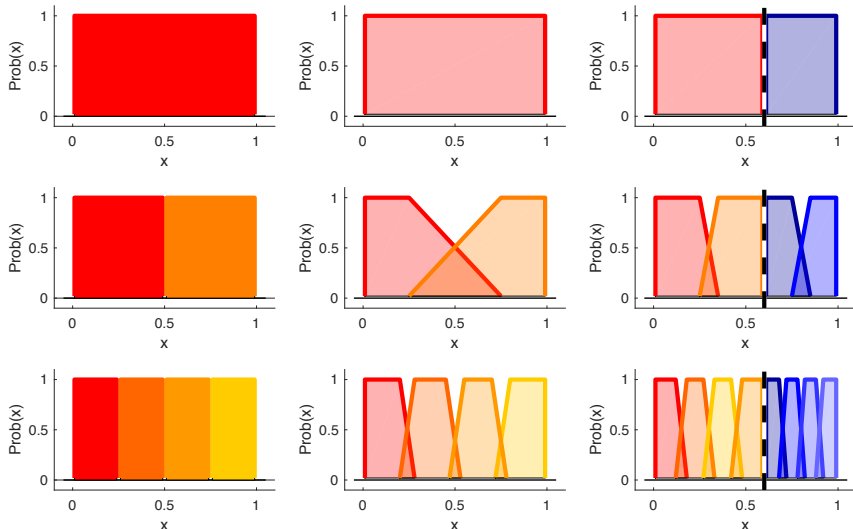
Probability of selecting a product x conditional on partition cell and collected information (top to bottom). Sharp partition (left), fuzzy partition (center), and fuzzy partition with quality label x^* (right).

Model: Information acquisition



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Model: Product space partition

Given a precision level ρ , the interval $[0, 1]$ is divided into 2^ρ disjoint intervals Π of equal length.

When $\rho \rightarrow \infty$ the buyer perfectly distinguishes the products.

Choice probability in the degenerate case (sharp partition)

$$pdf(x_j|\Pi_\rho^c) = \begin{cases} 0 & x \leq \frac{c-1}{2^\rho}, x_j \geq \frac{c}{2^\rho} \\ 2^\rho & \frac{c-1}{2^\rho} \leq x_j \leq \frac{c}{2^\rho} \end{cases}$$

Choice probability for generic trapezoidal pdf (fuzzy partition)

$$pdf(x_j|\Pi_\rho^c; a_\rho^c, b_\rho^c, d_\rho^c, e_\rho^c) = \begin{cases} 0 & x \leq a_\rho^c, x_j \geq d_\rho^c \\ \frac{2(x_j - a_\rho^c)}{(b_\rho^c - a_\rho^c)(e_\rho^c + d_\rho^c - b_\rho^c - a_\rho^c)} & a_\rho^c \leq x_j \leq b_\rho^c \\ \frac{2}{e_\rho^c + d_\rho^c - b_\rho^c - a_\rho^c} & b_\rho^c \leq x_j \leq d_\rho^c \\ \frac{2(d_\rho^c - x_j)}{(e_\rho^c - d_\rho^c)(e_\rho^c + d_\rho^c - b_\rho^c - a_\rho^c)} & d_\rho^c \leq x_j \leq e_\rho^c \end{cases}$$

Model: Consumer

Utility of consumer β from product j after collecting ρ information

$$U_{\beta}(\rho, x_j, p_j) = \beta \cdot q(x_j) - \alpha \cdot p_j - \lambda \cdot \rho$$

Consumers differ in preference intensity β over product quality x , and have homogeneous information cost $\lambda \geq 0$. A consumer needs to pay the cost $\lambda \cdot \rho$ in order to achieve the precision level $\rho \geq 0$.

The consumer selects a precision level ρ , that corresponds to a partition cell Π_{ρ}^c of the product space. Then, she selects the cell with the highest expected value and purchases randomly within the cell.

Outside option (ρ, \emptyset) , with $U_{\beta}(\rho, \emptyset) = -\lambda \cdot \rho$.

Example: Solution of the consumer's problem

- ▶ Assume functional form $q_j := q(x_j) = \log(x_j + 1)$
- ▶ Assume free entry, product price linear in quality $p_j = p(x_j) = x_j$, and distribution $x \sim U[0, 1]$
- ▶ Under perfect information, consumer β selects $x_j = \beta - 1$
- ▶ So every $\beta \geq 1$ purchases a product. Assume $\beta \sim U[1, 2]$
- ▶ Under full coarseness ($\lambda = \infty$), consumers buy a product if
$$\beta \geq \frac{\mathbb{E}[p]}{\mathbb{E}[q]} = \frac{0.5}{2\ln(2)-1} \approx 1.3$$
- ▶ If $q(\cdot)$ is monotonic, the fraction of consumers who do not buy any product is decreasing in the information cost λ

Simulation

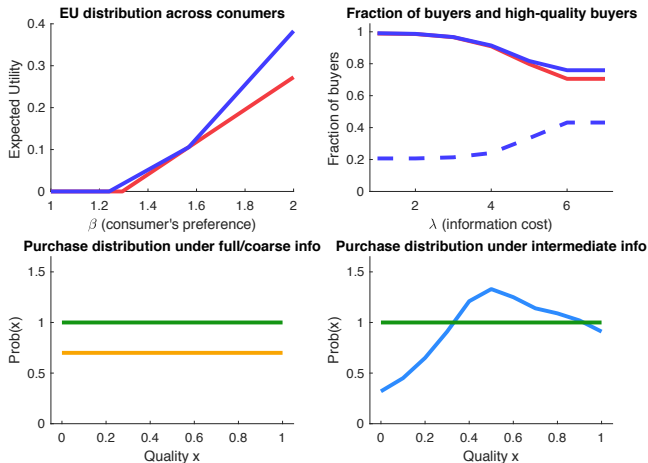


Figure 4: Simulation of the model without label (red) and with label $x^* = 0.8$. 1) Expected utility as a function of β if $\lambda > 1$. 2) Fraction of buyers (full line) and high-quality buyers (dotted line). 3-4) Purchase distribution under coarse (orange), partial (blue), and full info (green).

Simulation

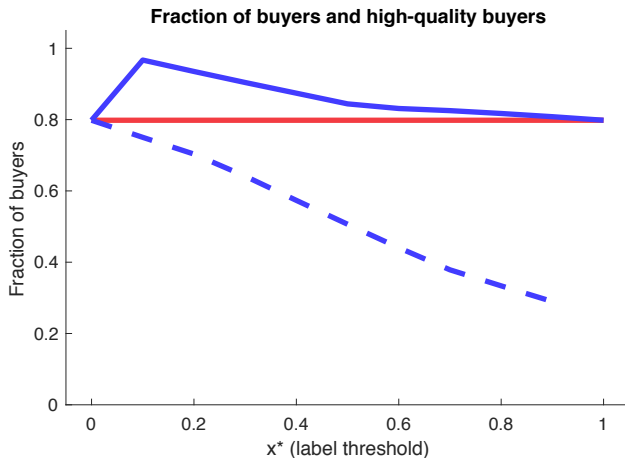


Figure 5: Simulation of the model without label (red) and with label $0 \leq x^* \leq 1$. Fraction of buyers (full line) and high-quality buyers (dotted line) as a function of the label requirement.

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Information theory: Simplicity vs Likelihood

- ▶ Shannon information theory relies on the exact knowledge of the likelihood of different events
- ▶ Generalized entropy relaxes the functional form of the cost but it still relies on information about the event distribution in order to create the signal structure
- ▶ Structural Information Theory (psychology) and Algorithmic Information Theory (computer science) relax this assumption
- ▶ Simplicity principle (e.g. data visualization): minimize the information load in a flexible environment
- ▶ The code allows a perceptual interpretation that is “fairly veridical” in many possible worlds, instead of “highly veridical” in only one world