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Information based food policy: is nutritional label the right instrument for everyone?

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Abstract. Individuals' time preference is an important driver for several health-related behaviors, but there is still lack of knowledge about its relationship with specific food-related behaviors. In this paper we investigate if time preference may have a role in affecting consumers' use of food label. The data were collected through a face-to-face survey on a sample of 540 Italian consumers. Time preference was estimated through an index including BMI, smoking behavior and physical activity. The analysis was made estimating a set of three equations. The results confirm our hypothesis that individuals with low time preference, who give more importance to future health outcomes, are more willing to use such information.



1. Introduction

Diet-related chronic diseases constitute a major public concern not only from a medical point of view, but also in economic terms as they are linked to sizable sanitary costs (Rosin, 2008). To challenge these problems and improve public health, economists in the last decades investigated the main determinants of consumers' food choices, and found that food label can play a crucial role in addressing people towards healthier consumptions. The information on label, especially those related to nutritional properties, may lead consumers to more healthful diet choices (Drichoutis *et al.*, 2008; Drichoutis *et al.*, 2009; Barreiro-Hurlè *et al.*, 2010). Nonetheless, the effectiveness of label is strictly linked to the individual willingness to care about this information source. Indeed, there are numerous factors which may encourage (or discourage) people from using it (Cavaliere *et al.*, 2015). Among these, the way consumers value their future health might have a primary role.

The importance attributed to future events depends on the individuals' time preference, that is, the extent at which people are willing to subvert a present gratification for a future utility (Komlos *et al.*, 2004; Smith *et al.*, 2005; Zhang and Rashad, 2008). Consumers with high time preference are more present-concerned and tend to attach more value to the present satisfaction than to the long term utility of being healthy; on the opposite, low time preference characterizes more future-oriented individuals. Therefore, time preference could heavily affect the extent at which consumers engage in health-enhancing activities (Grossman, 1972; Chapman *et al.*, 2001; Smith *et al.*, 2005; Zhang and Rashad, 2008; Brown and Van Der Pol, 2013; Cavaliere *et al.*, 2014).

Although the economic literature is rich in evidence concerning the role of individual time preference on health-related activities (Lawless *et al.*, 2013), there is still lack of evidence concerning how time preference specifically affects food behaviors. Thus, trying to fill this gap, in this paper we analyze which could be the role of individuals' time preference in affecting food label use. The hypothesis is that individuals with low time preference, being more future-concerned, are more willing to make use of labelled information, relative to consumers with higher time preference.

2. Economic framework

Following Grossman's model on the demand for health (1972), we assume that the individual state of health is part of a consumer utility function. A good state of health increases the utility since it allows enjoyment of numerous activities which otherwise could not be undertaken. Suppose that an

individuals' utility is a function of working time (W), free time (F), health (H), and the consumption of a bundle of goods (G) different from health. The utility function can be written as:

$$U=U\left(W,\,F,\,H,\,G\right)\tag{1}$$

In Grossman's model, the endowment of health that people inherit at birth is subject to a depreciation over time, which can be offset through health investments (I_H). These include all the activities and actions aimed at maintaining or improving health status. Therefore, in this model, health is not only demanded, but also produced by individuals. In fact, people are able to affect their own health level to the extent at which they engage in healthy investments. Thus:

$$H=H(I_H,\Omega) \tag{2}$$

where Ω represents a number of exogenous factors independent from the individuals willingness to invest in health. One of the main factors that may affect individuals' willingness to invest in health is time preference (TP) (Grossman, 2006), which represents people preference for present or future utility (Bishai, 2001; Komlos *et al.*, 2004; Smith *et al.*, 2005; Brown and Van Der Pol, 2013). Thereby, the health investment function can be specified as:

$$I_{H}=I_{H}\left(TP,\,\Gamma\right)\tag{3}$$

where TP represents the individuals' time preference, and Γ represents other factors different from TP. In this paper, we focus on the effect of time preference on a specific food-related behavior, that is, consumer use of food labels. More specifically, according to Grossman's model, we consider consumers' label use (L) as an indirect investment in health. Indeed, many studies show that through this active process of information searching consumers can become more informed about nutritional issues. This may in turn lead to more awareness about diet choices, consequently having an impact on health status (Drichoutis *et al.*, 2006; Drichoutis *et al.*, 2008; Drichoutis *et al.*, 2009; NØrgaard and BrunsØ, 2009; Barreiro-Hurlè *et al.*, 2010). Thus, consumers' label use can be explained as a function of time preference and other factors:

$$L=L\left(TP,\,\Gamma\right)\tag{4}$$

Following the recent literature analyzing the main drivers that can affect the use of label, we assume Γ to be composed by consumers' nutritional knowledge (KN), the average time spent choosing a new food product as a proxy of time constraint (T), and some socio-demographic and economic variables (S). Thus, the equation 4 can be rewritten as:

$$L=L(TP, KN, T, S) \tag{5}$$

The nutritional knowledge component included in the label function (5) may represent a source of endogeneity in the equation (Drichoutis *et al.*, 2005; Grunert *et al.*, 2010; Barreiro-Hurlè *et al.*, 2010). To account for the endogeneity issue we define the KN function as:

$$KN = KN(X, E_i)$$
 (6)

where X is a vector of observable individuals' characteristics and E_i represents the unobservable characteristics of nutritional knowledge. Precisely, basing on the main evidence in the economic literature, we include in the vector X the socio-demographic and economic conditions (Grunert *et al.*, 2010; Chang and Nayga, 2011), the time spent choosing a new food product, and some sources of food-related information different from labels (i.e., information from TV, internet, and doctors/nutrition experts) (Drichoutis *et al.*, 2005). Hence, following the approach used by Nayga (2000) and trying to solve the problem of endogeneity, in the empirical model we treat KN as an endogenous variable (as explained in section 3).

3. Methods

3.1 Data collection

The data for our analysis were collected through a consumer survey with personal interviews on a sample of Italian consumers in charge of their grocery shopping. The survey was carried out in Milan, in northern Italy, using an *ad hoc* questionnaire. Consumers were randomly approached outside the grocery stores, both in the central areas and in the suburbs and covering different time bands (early morning, lunch time, and evening). We excluded *a-priori* only consumers younger than 18 year old and we dropped from the final sample those who did not fully complete the survey. The final dataset consisted of 540 observations. Considering the dimension of Milan's population, the relative error is estimated at 3.95% (Mazzocchi, 2008).

3.2 Model specification and variable descriptions

Following eq. (6) in our economic framework, nutritional knowledge is estimated as:

a) KN= KN (GEN, AGE, EDUC, INC, HS, I_{TV}, I_{INTERNET}, I_{DOCTORS}, T)

The predicted value of nutritional knowledge (\widehat{KN}) is then included as a regressor in the following two equations having label use as dependent variable:

- b) L= L (TP, GEN, AGE, INC, HS, \widehat{KN} , T)
- c) L= L (GEN, AGE, INC, HS, \widehat{KN} , T)

In model a) KN is estimated through an index based on a set of questions about protein, fat, and carbohydrate content of food products. The scores obtained are then summed and converted into an index, so that, final low scores are associated with low nutritional knowledge (Table I). The socio-demographic and economic conditions (gender, age, level of education, household income, and household size) are included in the model as described in the economic framework. The variables (I_{TV}), (I_{INTERNET}) and (I_{DOCTORS}) are related to the main sources of information (different from label), that consumers use to get information about food products.

In model b) and c) the label use variable is categorical and is constructed to reflect how frequently consumers use food label. As in Variyam (2008) and in Loureiro *et al.* (2012), label use can assume five values: from 1 corresponding to 'never use', to 5 corresponding to 'always use'. This specification of the dependent variable allows for modelling the intensity of use. So responses can vary by frequency and thereby give a better description of the distribution of uses across the population.

Following the approach employed by Robb *et al.* (2008), we estimate consumers' time preference through a composite measure. This proxy is estimated creating an index with three variables that are shown in the literature to be strictly related to time preference, specifically: BMI (estimated through self-reported height and weight measures), smoking behavior, and physical activity. These variables were coded so that high scores correspond to consumers' high time preference (present orientation), and low scores to low time preference (future orientation). Then, the variables' values were standardized and used to create the proxy. High time preference is associated with high BMI levels, smoking behavior, and scarce physical activity, whereas the opposite is for low time preference.

Model a) is estimated through an ordinary least squares regression. The predicted value of KN obtained with this regression is then used as an explanatory variable in models b) and c), estimated using an ordered logistic regression having label use as dependent variable. Model c) differs from b) because TP is not included among the regressors. Model c) allows to verify if the relationship between label and TP is significant independently of the other variables included in the equation. As a last step in the analysis we compute the marginal effect estimation of model b).

4. Results and discussion

The results confirm our initial hypothesis concerning the role of consumers' time preference in affecting label use (Table II). As expected, the relationship between consumers' TP and label use is negative and significant (-0.144), that is, more future-oriented individuals are more likely to make use

of the information reported on labels. Giving more value to future health outcomes, these consumers might better perceive the benefits deriving from label use, and consider this kind of information as a possible way to improve their health in the long run.

With regard to the socio-demographic and economic characteristics, people aged between 35 and 54 years old are more likely to use labels. This might be due to the fact that in this age category the probability to have little children is higher, thus leading people to be more careful about what they purchase. Also, people with highest income are more likely to make use of food label. This result is in line with previous research which found that high income consumers rely on labelled information more than low income individuals, and generally attribute more value to such information (McLean-Meyinsse, 2001). Moreover, label use decreases as the household size becomes bigger, probably due to the fact that this households may suffer from higher time constraint.

Nutritional knowledge is positively related to label use, meaning that when consumers are more knowledgeable about nutritional properties, they are more willing to use labelled information. This result is in accordance with previous studies highlighting the important role of nutritional knowledge in leading consumers to invest more time in collecting information on food products (Drichoutis *et al.*, 2006).

Finally, the results highlight that the more is the time spent for choosing a new product, the more consumers are likely to use food label. This indicates that when consumers do not experience high time constraint, they are more willing to get information through labels and to select products that are more in line with their preferences.

The last regression, estimated excluding TP from the explanatory variables, shows overall the same results. The signs and statistical significance of all the original regressors in model b) remain unchanged. The coefficient magnitudes of statistically significant variables are unchanged, suggesting that the effect of TP on label use is uncorrelated with the other individual characteristics included in the analysis.

Moreover, the marginal effect computation (Table III) suggests that the effect of low time preference is more evident for those consumers that use labelled information most frequently. Indeed, results highlight a different pattern of signs between consumers who use label with a low-medium frequency (never, rarely, and sometimes) and those who, instead, habitually use it (often, always).

5. Concluding remarks

This preliminary investigation on the role of time preference in affecting consumers' use of food label confirms our hypothesis that people are more prone to use such information when they are more health-oriented. On the opposite, people having scarce interest in food label are those that show a low health orientation in their behaviors and do not care about their future utility. Thus, although food labelling is effective in increasing market transparency, our results suggest that this tool might be not effective to change the consumption patterns of those consumers with a low orientation to health. Indeed, individuals who mostly use food label are those that already engage in other healthy behaviors, as emphasized through our proxy of time preference.

Therefore, different public intervention should be adopted to improve healthy behaviors. Such policies should attempt to decrease consumers' preference for immediate utilities and lead people to rethink about the value of their future health outcomes. This might be achieved through educational and communication campaigns aimed at making consumers more aware about the strong link existing between their eating behavior and diet-related health problems.

The main limitation of our study is related to the label use variable, since more specific information categories, such as nutrition facts or nutritional claims, could be employed in future research. Furthermore, as the survey is based on stated preferences, social desirability bias might occurred. Future studies should explore more extensively the goodness of our time preference proxy, testing it in different contexts.

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Variable name	Description	Ops	Obs Freq Mean	ean SD
Dependent variable				
Label use	Scale, respondents use labels from never $= 1$, to always $= 5$	540		3.27 1.38
Independent variables				
TIME PREFERENCE				
Physical activity	Dummy, respondents practice sport less than once a week = 1; at least once a week = 0	540	237	
Smoking behavior	Dummy, respondents smoke $= 1$; otherwise 0	540	158	
Body Mass Index	Body Mass Index (weight kg/ height squared m²)	540	(1	23.64 3.69
GENDER	Dummy, female = 1, male = 0	540	293	
AGE ₁₈₋₂₄ ^a	Dummy, respondents' age group $18-24=1$; otherwise $=0$	540	2	
AGE 25-34	Dummy, respondents' age group $25-34=1$; otherwise $=0$	540	104	
AGE 35.44	Dummy, respondents' age group $35-44 = 1$; otherwise = 0	540	85	
AGE 45-54	Dummy, respondents' age group $45-54=1$; otherwise $=0$	540	87	
AGE 55-64	Dummy, respondents' age group $55-64 = 1$; otherwise = 0	540	92	
$AGE_{>65}$	Dummy, respondents' age group $>65 = 1$; otherwise = 0	540	108	
EDUCATION Primary school	Dummy, respondents' level of education primary school = 1; otherwise = 0	540	32	
EDUCATION Secondary school	Dummy, respondents' level of education secondary school = 1; otherwise = 1	540	66	
EDUCATION Higher education	Dummy, respondents' level of education higher education $= 1$; otherwise $= 2$	540	233	
EDUCATION Degree	Dummy, respondents' level of education degree = 1; otherwise = 3	540	160	
EDUCATION Post degree	Dummy, respondents' level of education post degree $= 1$; otherwise $= 4$	540	16	
HOUSEHOLD INCOME <80006 ^a	Dummy, household income < 8000 = 1; otherwise = 0	540	33	
HOUSEHOLD INCOME 800-15006	Dunnny, household income $800-1500e = 1$; otherwise = 0	540	139	
HOUSEHOLD INCOME 1500-30006	Dummy, household income $1500-30000 = 1$; otherwise = 0	540	222	
HOUSEHOLD INCOME 3000-50006	Dummy, household income $3000-5000e = 1$; otherwise = 0	540	26	
HOUSEHOLD INCOME >50006	Dummy, household income > 500000 = 1; otherwise = 0	540	49	
HOUSEHOLD SIZE 1 members a	Dummy, household size 1 component = 1; otherwise = 0	540	124	
HOUSEHOLD SIZE 2 members	Dummy, household size 2 components = 1; otherwise = 0	540	168	
HOUSEHOLD SIZE 3 members	Dummy, household size 3 components = 1; otherwise = 0	540	107	
HOUSEHOLD SIZE 4 members	Dummy, household size 4 components = 1; otherwise = 0	540	108	
HOUSEHOLD SIZE more than 4 members	Dummy, household size greater than 4 components = 1; otherwise = 0	540	33	
INFORMATION SOURCE TV	Scale, TV as source of information on food properties from never= 1 to always = 5	540		2.25 1.24
INFORMATION SOURCE Internet	Scale, internet as source of information on food properties from never= 1 to always = 5	540		2.09 1.39
INFORMATION SOURCE Doctors	Scale, doctors and/or nutrition experts as source of information on food properties from never= 1 to always = 5	540		3.28 1.54
NUTRITIONAL KNOWLEDGE				
Knowledge carbohydrates	Dummy, level of knowledge on carbohydrates (knowledgeable $= 1$, not knowledgeable $= 0$)	540	425	
Knowledge far	Dummy, level of knowledge on fat (knowledgeable = 1, not knowledgeable = 0)	540	326	
Knowledge protein	Dunnny, level of knowledge on protein (knowledgeable = 1 , not knowledgeable = 0)	540	424	

TIME SPENT CHOOSING A NEW FOOD PRODUCT

^a Removed for estimation purpose.

540 409

Dummy, more than one minute = 1; less than 1 minute = 0

Table II – Model estimation

	NUTRIT			LABEI (Whit			LABEI (Whitho		
	Coef.	SE		Coef.	SE		Coef.	SE	
TIME PREFERENCE				-0.144	0.083	*			
GENDER	0.180	0.085	**	0.129	0.170		0.159	0.169	
AGE 25-34	-0.113	0.161		0.350	0.298		0.351	0.298	
AGE 35-44	-0.191	0.163		0.665	0.315	*	0.675	0.316 *	
AGE 45-54	-0.074	0.162		0.969	0.310	***	0.938	0.309 **	**
AGE 55-64	0.045	0.169		0.139	0.310		0.071	0.308	
AGE _{>65}	0.050	0.180		0.361	0.314		0.297	0.313	
EDUCATION Secondary school	0.482	0.214	**						
EDUCATION Higher education	0.692	0.218	***						
EDUCATION Degree	1.037	0.237	***						
EDUCATION Post degree	1.254	0.329	***						
HOUSEHOLD INCOME 800-1500€	-0.178	0.200		0.087	0.353		0.097	0.352	
HOUSEHOLD INCOME 1500-3000€	-0.176	0.201		0.509	0.352		0.551	0.350	
HOUSEHOLD INCOME 3000-5000€	-0.045	0.228		0.400	0.408		0.483	0.404	
HOUSEHOLD INCOME >5000€	-0.210	0.247		0.936	0.445	*	1.010	0.441 *	
HOUSEHOLD SIZE 2 members	0.145	0.122		-0.090	0.229		-0.100	0.229	
HOUSEHOLD SIZE 3 members	-0.030	0.137		0.239	0.264		0.211	0.263	
HOUSEHOLD SIZE 4 members	0.195	0.142		-0.044	0.277		-0.019	0.276	
HOUSEHOLD SIZE more than 4 members	0.259	0.199		-0.871	0.379	*	-0.830	0.378 *	
INFORMATION SOURCE TV	-0.010	0.037							
INFORMATION SOURCE Internet	-0.063	0.033	**						
INFORMATION SOURCE Doctors	0.131	0.030	***						
NUTRITIONAL KNOWLEDGE				0.954	0.295	***	0.925	0.294 **	**
TIME SPENT CHOOSING A NEW FOOD PRODUCT	0.195	0.100	**	0.497	0.199	*	0.477	0.199 *	
α_1	-1.154	0.305	***	-0.443	0.451		-0.422	0.450	
α_2				0.214	0.451		0.230	0.450	
α_3				1.222	0.454		1.231	0.454	
α_4				2.560	0.462		2.566	0.462	
R-squared	0.122			0.038			0.038		
N	540			540			540		

Significance: *** p < 0.01; ** p < 0.05; * p < 0.10

Note: the variable education has not been included in models 2 and 3 because of multicollinearity problems.

Table III - Marginal Effect for label use equation

	LABEL USE														
	Ne	Never		Rarely		Sometime			Often			Always			
	Coef.	SE		Coef.	SE		Coef.	SE		Coef.	SE		Coef.	SE	
TIME PREFERENCE	0.019	0.011	*	0.009	0.005	*	0.008	0.005	*	-0.012	0.007	*	-0.024	0.014	*
GENDER	-0.017	0.023		-0.008	0.011		-0.007	0.009		0.011	0.015		0.021	0.028	
AGE 25-34	-0.044	0.035		-0.022	0.018		-0.022	0.021		0.026	0.019		0.061	0.055	
AGE 35-44	-0.076	0.031	*	-0.040	0.018	*	-0.047	0.026	*	0.039	0.012	**	0.123	0.065	*
AGE 45-54	-0.104	0.027	***	-0.056	0.017	***	-0.072	0.028	**	0.044	0.011	***	0.188	0.068	**
AGE 55-64	-0.018	0.039		-0.009	0.019		-0.008	0.019		0.011	0.024		0.023	0.054	
$AGE_{>65}$	-0.045	0.036		-0.022	0.019		-0.023	0.022		0.027	0.020		0.063	0.058	
HOUSEHOLD INCOME 800-1500€	-0.011	0.046		-0.005	0.022		-0.005	0.020		0.007	0.029		0.014	0.059	
HOUSEHOLD INCOME 1500-3000€	-0.066	0.045		-0.031	0.022		-0.029	0.021		0.041	0.026		0.086	0.061	
HOUSEHOLD INCOME 3000-5000€	-0.049	0.046		-0.024	0.025		-0.026	0.030		0.029	0.023		0.070	0.077	
HOUSEHOLD INCOME $_{5000}\varepsilon$	-0.096	0.034	***	-0.054	0.023	**	-0.072	0.041	*	0.036	0.012	**	0.186	0.102	*
HOUSEHOLD SIZE 2 members	0.012	0.031		0.006	0.014		0.005	0.012		-0.008	0.020		-0.015	0.037	
HOUSEHOLD SIZE 3 members	-0.031	0.032		-0.015	0.016		-0.014	0.017		0.019	0.019		0.041	0.047	
HOUSEHOLD SIZE 4 members	0.006	0.038		0.003	0.017		0.002	0.014		-0.004	0.024		-0.007	0.045	
HOUSEHOLD SIZE more than 4 members	0.148	0.078	*	0.047	0.016	**	0.011	0.014		-0.094	0.045	*	-0.113	0.037	**
NUTRITIONAL KNOWLEDGE	-0.128	0.040	***	-0.059	0.020	**	-0.051	0.018	**	0.082	0.028	**	0.157	0.049	**
TIME SPENT CHOOSING A NEW FOOD PRODUCT	-0.073	0.032	*	-0.030	0.012	**	-0.020	0.007	**	0.048	0.021	*	0.076	0.028	**

Significance: *** p < 0.01; ** p < 0.05; * p < 0.10