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THE VALUATION OF NON-MONETARY CONSUMPTION IN HOUSEHOLD SURVEYS

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ABSTRACT. Many social indicators are based on household consumption information. The valuation of non-monetary operations is crucial for the analysis of consumption surveys in developing countries because of the importance of own-consumption and transfers in kind. What are the price statistics used in the valuation of consumption indicators? How is the available price information exploited to produce consumption indicators? How can the different steps of the valuation process be analysed? We explore these questions by presenting the valuation method for the consumption used in rural Rwanda for the 1983 consumption survey, and by proposing a general model of valuation algorithm. This is useful not only for improving such algorithms, but also for assessing the impact of the valuation process on economic analyses.

KEY WORDS: consumption analysis, data processing, demand systems, household surveys, poverty analysis, prices, valuation method

INTRODUCTION

22 Many social indicators are based on household consumption
23 information. Consumption surveys (CS) are the main infor-
24 mation source about household consumption. Numerous CS
25 have been implemented since the ninetieth century¹ in Europe
26 and since the beginning of the 20th century² in less developed
27 countries. Producing high quality economic indicators from CS
28 is important for both statisticians and economists³ because
29 biases intervening during the calculus of consumption indica-



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30 tors may be more serious than the ones occurring in data
31 analysis.

32 Although numerous authors have discussed the design of
33 CS⁴, little interest has been devoted to the valuation of non-
34 monetary consumption. Though, this valuation is of consid-
35 erable interest as such consumption can be a significant share
36 of consumption of poor rural households. In many LDCs
37 (Less Developed Countries), especially in rural areas, own-
38 consumption rates (ratios of produced consumption over total
39 consumption) can be very high, and gifts and other transfers
40 in kind very common. For example, the average own-con-
41 sumption rate in 1983 (proportion of consumption coming
42 from own production) in rural Rwanda is above 66% overall
43 and almost 80% for food (from our own estimates using
44 the National Budget-Consumption Survey 1983 in Muller,
45 1992). In these conditions, the valuation method of
46 consumption records is crucial to the measurement of aggre-
47 gate household consumption in LDCs. However, in the
48 absence of precise description of the valuation process, it is
49 hard to understand its nature and its impact on economic
50 analyses.

51 What are the price statistics used for the valuation of
52 consumption in consumer surveys? How can the steps of the
53 valuation process be analysed? Which lessons can be drawn
54 for economic analyses? The aim of this article is to explore
55 these questions first by presenting the valuation method that
56 was used in rural Rwanda for the CS conducted in 1983
57 (denoted CSR83), then by proposing a model of valuation
58 algorithm. This novel approach is useful not only for
59 improving such algorithms, but also for assessing the impact
60 of the valuation process on economic analyses. In second
61 section, we present the valuation algorithm and the
62 price indicators that have been used for CSR83. In third
63 section, we analyse the different stages of a general valua-
64 tion algorithm for a CS. Finally, we conclude in fourth
65 section.

THE VALUATION OF CONSUMPTION IN A CONSUMPTION
SURVEY68 *Origin of the Price Information*

69 It seems a good idea to start with the origin of the price
70 information. The two main sources of prices for valuation
71 algorithms are: price surveys in markets, and the CS itself from
72 the records of value and quantity for purchases or sales. These
73 two sources have different advantages. Price surveys provide
74 price information less dependent on household tastes and
75 purchasing power because of a better control of quality choices.
76 However, since price observations are collected only in selected
77 sites, they may provide inaccurate estimates of the prices with
78 which some households are confronted. Moreover, the wording
79 of the questions, and more generally the collection method of
80 prices, are always debatable in that they constitute artificial
81 observation situations, different from what occurs during ac-
82 tual transactions. Finally, it is never possible to obtain price
83 observations for every good in every selected market or trans-
84 action site. This implies that the treatment of missing values for
85 prices is an important stage of using market price data. Fur-
86 thermore, even when price observations are available, the
87 analyst may not be content to use them if they are isolated.
88 Large samples of price observations are in fact necessary and
89 what is called “market price” in the price file used for the
90 valuation is generally a central tendency of this sample distri-
91 bution, the mean or the median of observed prices.

92 On the other hand, when budget data are used to calculate
93 prices, the information about prices fits household consump-
94 tion patterns more closely. Indeed, goods that are usually
95 consumed in an area appear in records of local purchases or
96 sales, even when they are only consumed in kind (from their
97 own production or received as gift) by some of the households
98 of this area. This significantly contributes towards solving the
99 problem of missing price data. The prices extracted from a
100 budget survey are in fact unit-values, i.e. ratios of values over
101 quantity extracted from observations of the individual trans-



102 actions. The most appropriate transactions for this calculus are
103 the consumption purchases, and the production sales. This
104 leads to elementary unit-value observations that can be inter-
105 preted respectively as “consumption prices” and “production
106 prices”. Other transactions are generally not associated with
107 large enough samples of observations to be of great use.

108 Unfortunately, other difficulties arise. Elementary unit-val-
109 ues are affected by choices of qualities by consumers or sellers.
110 This problem is much less serious than for unit-values calcu-
111 lated from broad categories of consumption (as in Deaton,
112 1988, 1990) where similar goods are aggregated in a common
113 category (for example “fish”). In the latter case, the unit-values
114 calculated from these aggregate values and quantities are
115 clearly different from observed prices in a market (e.g. price of a
116 specific fish). However, quality bias occurring with elementary
117 unit-values would arise from differences in actual quality be-
118 tween two units of the same elementary product, for example
119 “Tilapia”, a species of fish. Even if one expects it to be rela-
120 tively minor, the quality choice problem remains.

121 Another problem is that elementary unit-values may be af-
122 fected by measurement errors occurring in value and quantity
123 observations. In that situation, the measure provided by ele-
124 mentary unit-values may not be accurate enough to validly
125 approximate actual market prices. The endogeneity of unit-
126 values can be treated by using a prediction model for prices or
127 merely by using means of prices, at the cluster or at the regional
128 level. Indeed, providing the size of the unit-value sample is large
129 enough, aggregate means discard most endogeneity problems
130 associated with a specific household that contributes only a
131 negligible fraction of the mean. These procedures do not
132 eliminate endogeneity arising at a regional level, for example
133 high prices reflecting a high quality of consumption related to
134 the general wealth in this region or to regional tastes. However,
135 this difficulty is also present with market prices. Aggregate
136 means of unit-values also provide estimates of prices based on
137 large samples and thus eliminate much noise in price observa-
138 tions and part of missing value problems. The type and the

139 aggregation levels for the price means are to be selected for the
140 valuation algorithm. We now turn to the specific example of
141 CSR83.

142 *An Example of Survey*

143 The Rwandan national budget-consumption survey was con-
144 ducted by the government of Rwanda and the French Coop-
145 eration and Development Ministry, in the rural part of the
146 country from November 1982 to December 1983 (Ministère du
147 Plan, 1986)⁵. The collection of the consumption data was or-
148 ganised in four rounds, roughly corresponding to quarters⁶.
149 During these rounds, 270 households were surveyed about their
150 budget and their consumption. Each household was visited at
151 least once a day during two weeks for every quarter. The
152 consumption was systematically recorded with daily and ret-
153 rospective interviews, and all food was weighed. Every house-
154 hold also had to record budget information in a diary between
155 the quarterly survey rounds.

156 Let us examine in the CSR83, the questionnaires from which
157 price and consumption information can be extracted. They are
158 the following ones.

Q4: Daily transactions. This questionnaire is filled every day during 14 days of every quarter for each household of the core sample (270 households). The collection is based on daily interviews.

Q2: Retrospective transactions. This questionnaire is filled in every quarter for each household of the global sample (1170 households). The collection is based on retrospective interviews and diaries with a three-month recall.

Q5: Food consumption. This questionnaire is filled every day for 7 days of every quarter for each household of the core sample. The collection is based on daily weighing of the food and daily interviews.

Q7: Price survey in markets. This questionnaire is filled every week when households are visited in the same sector. The collection is based on interviews of sellers in markets and in other transaction sites, and weighing of the products.

175 The questionnaires are simultaneously the source of the price
 176 information and the location of consumption records to value.
 177 We shall describe the valuation process for the CSR83 in three
 178 steps: definition of price indicators, comparison of price means,
 179 and valuation algorithm. This illustration based on CSR83 will
 180 suggest the general model in third section.

181 Before examining the price indicator, we need to identify the
 182 records to value. Different types of records are to be valued:
 183 barter, received and offered gifts, food consumption measures
 184 based on direct weighing (purchased, received as gift, taken
 185 from stocks), quantities of food own-consumption weighed
 186 before the meals. The identification of the records that need to
 187 be valued in every questionnaire file indicates the products for
 188 which we need price indicators for the valuation.

189 The next step consists in creating a price file to organise the
 190 set of price observations. There are three types of prices in the
 191 file: (1) the *production prices* that are ratios of the recorded
 192 value of output sales by the recorded quantity sold; (2) the
 193 *consumption prices* that are ratios of the recorded value of
 194 consumption expenses by the recorded quantity bought; (3) the
 195 direct measurement of *prices in markets*. Let us describe more
 196 precisely how these different price statistics are calculated.

197 The consumption prices (PC1 and PC2) are calculated using
 198 data from questionnaire Q4 (daily transactions), which is the
 199 most accurate source of budget information. Prices PC1 are
 200 calculated using records of consumption purchases, prices PC2
 201 using records of production sales to another person likely to
 202 consume the product of interest. The production prices (PP1
 203 and PP2) are also calculated using data from questionnaire Q4.
 204 Prices PP2 are based on records of production sales, while
 205 prices PP1 are calculated using records of consumption pur-
 206 chases from a peasant selling his own production. Finally, the
 207 market prices (PM) are calculated using data from question-
 208 naire Q7. We use only prices PC1, PP2 and PM for which we
 209 have significant sample sizes. From now we denote PC1 as PC
 210 and PP2 as PP.

211 Based on these three types of prices, the valuation system is
 212 organised in three geographical levels (local, regional, national)

213 and two temporal levels (quarter and year). Rwanda is divided
214 into 10 prefectures and 143 rural communes, although only 90
215 communes had been selected in the sampling scheme. Rwanda
216 is also divided into five agricultural zones, called “regions”
217 here, which are defined on the basis of their agro-ecological
218 characteristics.

219 At all these geographical and temporal levels, we calculate
220 price means for each type of price (PC, PP, PM) and each
221 product. We weigh the price means using the sampling weights
222 and the average consumption (or production) shares of
223 households in the relevant geographical and temporal units.
224 The price means are calculated for the temporal levels: year and
225 quarter, and the geographical levels: the whole country, the
226 prefecture, the region, and the commune. This yields eight
227 possible combinations of levels. Using the regions rather than
228 the prefectures has several advantages. First, the regions have
229 clear economic and agricultural meaning and then the corre-
230 sponding price samples are more likely to be homogeneous.
231 Second, the sizes of price samples on which the means are
232 calculated are larger.

233 The obtained price means for the different aggregate levels
234 are available for more than 100 products. In total almost 24,000
235 price statistics are available. Table I presents the distribution of
236 price mean records across the categories of means. No one
237 category of price means accounts for a substantial part of these
238 records. The main category is composed of the means of PC
239 prices at the levels of commune and quarter, with 17.7% of the
240 price records. The means of PC prices at the levels of prefecture
241 and year, and the means of PM prices at the levels of commune
242 and quarter are also important with respectively 10.6% and
243 10.0% of the records. The majority (54.1 %) of price means are
244 of PC type, the remainder being shared between PM type
245 (27.5%) and PP type (18.4%). About two thirds (65.4%) of the
246 price means are at the quarterly level. Almost half (45.4%) of
247 the price means are at the commune level, while only 7.7% are
248 at the national level. The reason why quarterly and commune
249 levels are not overwhelmingly dominant in this file is that price
250 means have been rejected from the data base when the sample



TABLE I

Distribution of the records of the price database across categories of price means

		N	C	P	Z
PC	A	251 (1.1)	1004 (4.2)	2526 (10.6)	700 (2.9)
	Q	635 (2.7)	4226 (17.7)	2021 (8.5)	1513 (6.4)
PM	A	134 (0.6)	1270 (5.3)	441 (1.9)	324 (1.4)
	Q	331 (1.4)	2374 (10.0)	962 (4.0)	723 (3.0)
PP	A	154 (0.6)	753 (3.2)	390 (1.6)	296 (1.2)
	Q	317 (1.3)	1176 (4.96)	721 (3.0)	570 (2.4)
Total	23812 (100)				
Price type	%	Temporal level	%	Geographical level	%
PC	54.1	A	34.6	N	7.7
PM	27.5	Q	65.4	C	45.4
PP	18.4			P	29.7
				Z	17.3

The first number in each cell is the number of records. The number in parenthesis is the percentage. The categories of the price means are composed of three elements: the type of price (PC, PM, PP); the geographical level (N – national, C – communal, P – prefecture, Z – regional), the temporal level (A – annual, Q – quarterly).

251 of price records at the considered levels was judged too small to
 252 be reliable. At least ten observations of prices of good quality,
 253 and generally more, were required to accept the mean price
 254 calculated at the communal quarterly level. Using data from a
 255 national CS in Tunisia, Ayadi and Muller (2003), show that
 256 measurement errors are important in unit-value data. Relying
 257 on mean or median of several observations of unit-values is a
 258 simple way to overcome most of this issue. This is also a rather
 259 efficient way of getting rid of the influence of quality choices.
 260 This approach is to be opposed with the common use of unit-
 261 values at the household level that have been shown to be
 262 highly sensitive to measurement errors and endogenous quality
 263 choices.

264 Beyond this general picture, it is illuminating to focus on a
 265 few main products to study the orderings of the price means for
 266 the three types of prices. Twenty-two main products have been
 267 selected. The most important ones are: sorghum, sweet cassava,
 268 rice, sweet potatoes, potatoes, beans, aubergines, palm oil, salt,
 269 banana beer, sorghum beer, soap and cigarettes. A specific
 270 variety of the product is used when collecting its price. Table II
 271 shows the distribution of orderings of the types of national
 272 price means for these products by quarter. Only means based
 273 on a large number of observations have been kept, which ex-
 274 plains different totals at different quarters. PP means are always
 275 inferior to the corresponding PC means and this comparison is
 276 not included in the table. PM means are always greater than the
 277 corresponding PP means except on one occasion. The national
 278 production prices can therefore be considered as almost sys-
 279 tematically inferior to consumption prices and market prices,
 280 which seems natural because of transport, transaction and
 281 intermediation costs. Therefore, production prices should be
 282 avoided when possible for the valuation of consumption from a
 283 consumer perspective. The situation is less clear when com-
 284 paring PM and PC means. At the national and quarterly levels,
 285 their ordering is rather indifferent. Further analysis at more
 286 disaggregate levels have shown us that PC and PM prices are
 287 actually very similar. This suggests that the endogeneity prob-
 288 lems related to PC prices, which could arise at the household
 289 level, do not raise serious concern for price means calculated at
 290 the commune level.

291 We now turn to the way we rank these price means before to
 292 incorporate them in the valuation algorithm. The same valua-

TABLE II
 Distribution of the main products by ordering of price means and by quarter

Ordering	Quarter A	Quarter B	Quarter C	Quarter D
PM > PP2	9	8	7	10
PM < PP2	0	1	0	0
PM < PC1	4	7	7	5
PM > PC1	7	4	5	7



293 tion algorithm is used to value barter, non-purchased gifts, and
294 food own-consumption. One must also use the same valuation
295 algorithm for the received gifts and the offered gifts, in order to
296 be able to consistently compare these different flows. To be able
297 to order the different price means for each product, we compare
298 them at different aggregation levels. These investigations mix
299 statistical analyses and expert judgement, the latter based to a
300 large extent on the knowledge of enumerators. The first column
301 in Table III shows the obtained preference ordering for the
302 different types of price means.

303 This ordering of the price means is important because it
304 determines how every consumption record in the questionnaire
305 files is valued. When a desired price mean is not available at a
306 given level for the considered product and geographical and
307 temporal levels, the algorithm searches for a mean price at the
308 next level. We believe that the best price mean, corresponds to
309 the cluster and the quarter where the consumption quantity has
310 been recorded because of the proximity of the actual place and
311 time at which this consumption could have been purchased.
312 However, using this preferred price mean is not always possible
313 because only price means based on large enough samples of
314 observed prices are kept. In a sense, we assume that the accu-
315 racy of the price means for our purpose mostly depends on its
316 temporal and geographical proximity with the consumption
317 observations. Another approach would have been to consider
318 the trade-off between the higher variance of the conceptually
319 correct price (from the same locality and quarter) against the
320 bias of having a price that is from a larger region (but has a
321 lower variance), rather than just using the criterion of the
322 sample size. We did not retain it here because the valuation
323 operations had to be performed by agents not trained in
324 statistics.

325 Once the ranking of the price means is ready, we apply the
326 corresponding algorithm to our consumption survey. Table III
327 shows how the non-monetary consumption records in CSR83
328 are valued. The optimal price means (that of PC type at com-
329 munal quarterly level) are only used for 3% of the non-mon-
330 etary consumption value because for many products the sample

TABLE III
Results of the valuation algorithm for the non-monetary consumption

Type of price mean used for valuation	Percentage of the value of non-monetary consumption	Percentage of food weight of non-monetary food consumption
1	3.87	3.07
2	34.18	29.11
3	15.12	38.50
4	6.41	4.41
5	15.43	6.87
6	18.87	9.96
7	0.01	0.02
8	0.85	0.26
9	2.05	1.49
10	0.49	1.33
11	2.72	4.97

- 1 – Mean consumption price for the same cluster and the same quarter;
 2 – Mean consumption price for the same quarter and the same region;
 3 – Mean market price for the same quarter and the same region;
 4 – Mean consumption price for the same quarter and the same region;
 5 – Mean consumption price for the same quarter and the whole country;
 6 – Mean market price for the whole year and the whole country;
 7 – Mean production price for the whole year and the whole country;
 8 – Mean consumption price for the whole year and the whole country and with a small number of price observations (< 10);
 9 – Mean consumption price for the whole year and the whole country and with a very small number of price observations (< 5);
 10 – Mean market price for the whole year and the whole country and with a very small number of price observations (< 5);
 11 – Price coming from a source external to the survey.

331 of observed prices on which are based these price means is
 332 judged too small. Then, basing the valuation mostly on PC
 333 means at communal quarterly level may introduce massive data
 334 contamination because the means are often based on isolated
 335 and non-representative observations. The main price means for
 336 the valuation are the quarterly regional PC means that are used
 337 for about one third of the valued non-monetary consumption.
 338 Then, the PM means at the annual national level, the quarterly
 339 national PC means and the quarterly regional PM means are

340 the other major instruments for the valuation. The other price
341 means are less useful. In particular, production prices almost
342 never need to be used. On the whole, almost 60% of the non-
343 monetary consumption are valued with quarterly price indica-
344 tors at local or regional level. The breakdown of the valuation
345 of food non-monetary consumption by food weight provides a
346 similar picture, showing that the price levels of different prod-
347 ucts affect, but do not massively distort the decomposition of
348 the valuation. The results imply that for this survey the valu-
349 ation of non-monetary consumption can be reasonably accu-
350 rate. Such result was not obvious a priori and may not be
351 verified for all consumption surveys. Moreover, the valuation
352 process appears to be more complex than is generally expected
353 in consumer studies. It does not correspond to the simple pic-
354 ture presented in economic theory, or even in applied economic
355 studies.

356 Naturally, examples based on other consumption surveys
357 would deliver different results. However, the valuation proce-
358 dures in such surveys are of similar nature and therefore raise
359 the same type of issues. Let us now draw the lessons of the
360 CSR83 by systematising the valuation process in a general
361 algorithm valid for most consumption surveys.

A GENERAL VALUATION ALGORITHM

363 What we propose is a conceptual pattern for the valuation
364 process. This pattern is useful on several grounds: (1) it eases
365 the replication of valuation operations in different CS; (2) it
366 facilitates the design of valuation algorithms; (3) it clarifies the
367 resources to allot to this operation; (4) it distinguishes the tasks
368 involved in the design and the execution of the valuation; and
369 finally (5) it guides the specification of optimality criteria for the
370 valuation operations.

371 The conceptual framework will be all the more useful if the
372 different stages of the valuation algorithm can be incorporated
373 in the design of a data base management system. In that case, a
374 change in the valuation rules or in the sample of price means

375 could be easily incorporated so as to update the consumption
 376 indicators. Moreover, the price data set could be included in the
 377 survey database. This would enable analysts to use it for aims
 378 other than the consumption valuation. In this regard, the
 379 sorting variables used for the price data set and for the algo-
 380 rithm must be carefully designed. For each product, three
 381 natural sorting variables are (a) the type of prices (PC, PP or
 382 PM in the example); (b) the geographical unit; (c) the temporal
 383 unit. We are now ready to describe the algorithm.

384 We propose a model of valuation algorithm in seven steps:
 385 (1) Selection of records to value in the questionnaire files; (2)
 386 Confection of samples of elementary price indicators; (3) Cal-
 387 culus of price means at the different geographical and temporal
 388 levels and by type of price; (4) Choice of the optimal price mean
 389 indicator; (5) Establishment of a proximity map for the differ-
 390 ent price mean indicators; (6) Ordering of the different price
 391 mean indicators; (7) Execution of the valuation.

392 We now discuss the decision rules applied to each of these
 393 steps.

394 *Step 1 (records to value)*. The rules are semantic here and
 395 must correspond to the economic operations for which the
 396 valuation is desired. In our example, we select in the ques-
 397 tionnaire files the records that are own-consumption, gifts re-
 398 ceived in kind, consumed results of barter, and consumption
 399 taken from stocks. The selection rule for a questionnaire
 400 therefore corresponds to a product of sets ($A_1 \times \dots \times A_s$), each
 401 set A_i describing a list of codes for a selection variable. For
 402 example, for some records one can use the following product of
 403 code sets: {codes of uses corresponding to consumption of the
 404 transacted product} \times {codes of transaction types that corre-
 405 spond to transfer received in kind (among other non-monetary
 406 consumptions)} \times {codes of the products that are consum-
 407 able} \times {missing value for the recorded value of the transaction
 408 (indicating that it remains to value)}.

409 *Step 2 (elementary prices)*. The rules are semantic and eco-
 410 nomic. They correspond to selections of records similar to Step
 411 1. However, here the selected records are characterised by

- (A) no missing values for the variables cluster, round, product, measurement unit, quantity, and value.
- (B) specific transactions that correspond to: monetary consumption expenses, monetary production sales, direct collection of prices at transaction sites. Again, the selection rule is a product of sets ($B_1 \times \dots \times B_s$), each set B_i describing a list of codes admissible for a variable. Particularly useful are codes related to the following variables: use of the transacted product, transaction type (expenses, sales, transfers...), transaction partner, transaction location.
- (C) an a priori interval of admissible prices that can be defined for each product. Indeed, enumerators and price analysts are expert enough to be able to eliminate many errors and aberrant price values at this stage. Finally, the considered unit-value (i.e. value divided by quantity) is calculated for each recorded transaction, and is stored in a temporary price data set, with indexation by its level (price type, geographical unit, temporal unit). The observations of the market prices are directly stored.

431 *Step 3 (price means)*. The rules are statistical. Means, stan-
 432 dard deviations and sample sizes of selected sets of price re-
 433 cords in the price data set are calculated for each level. The
 434 statistics may or may not be weighed by using the sampling
 435 scheme and observed consumption (or production) levels. Only
 436 the price means considered close enough to the actual (but
 437 unobserved) price mean are inserted in the price means data-
 438 base. Statistical tests such as Student tests of differences of
 439 means can be used to guide this decision.

440 *Step 4 (optimal price mean)*. The rule is economic. If the
 441 statistical rules applied in Step 3 are judged sufficient to ensure
 442 the overall reliability of the set of price means in the data base,
 443 then it is natural to decide that the optimal price means to use
 444 first are the ones from whichever level is the closest to the
 445 consumption record to value. Since the notion of level is com-
 446 posite, there may be several optimal price mean indicators. In
 447 that case, a loss function can be used to calculate the risk of
 448 different alternatives. In our previous example, the optimal price
 449 mean indicator is the quarterly communal consumption price.

450 *Step 5 (proximity map)*. The rules are topological. A prox-
 451 imity measure is chosen to represent the relative positions of the
 452 different sets of price means. A simple solution is to consider
 453 that the set of price means at the most disaggregate level is a
 454 vector of which dimension is the number of price means at this
 455 level. For example, in CSR83, the vector of the PC means of a
 456 given product at the quarterly communal level, denoted PC_{CQ} ,
 457 would have 4 (the number of quarter) by 90 (the number of
 458 communes), i.e. 360 components. Then, a set of price means of
 459 a more aggregate level can be described by a vector of *same*
 460 *dimension* by replication of the price means in all sub-levels. For
 461 example, the vector of annual regional means of consumption
 462 prices in CSR83, denoted PC_{ZA} , would also have 360 compo-
 463 nents, each couple (commune, quarter) being affected with the
 464 annual price mean for the region to which the considered
 465 commune belongs. Then, one can calculate a distance between
 466 these vectors. For example, the Euclidean distance between
 467 PC_{CQ} and PC_{ZA} can be defined as

$$d(PC_{CQ}, PC_{ZA}) = \left[\sum_i \sum_j \sum_q (PC_{j,q} - PC_{i,A})^2 \right]^{1/2}$$

469 where i denotes the index of regions, j is the index of the
 470 communes in Region i , q the index of quarters, $PC_{j,q}$ the mean
 471 consumption price for Commune j at Quarter q , and $PC_{i,A}$ the
 472 mean consumption price for Region i at annual level. More
 473 sophisticated distance functions involving the empirical vari-
 474 ances and covariances of the price indicators are also interest-
 475 ing. In any case, such approach implies that the observed
 476 proximity is considered to be good indicators of the actual
 477 proximity.

478 *Step 6 (ordering of price means)*. The rules are quasi-order-
 479 ings. Since one optimal set of price means has been chosen in
 480 Step 4, a natural ordering can be deduced from the relative
 481 proximity of sets of price means to this optimal set. The dis-
 482 tance defined in Step 5 can be used for this purpose. However,
 483 other considerations may suggest different orderings. Thus, one
 484 component of the level might be of special importance. For



485 example, one may want to use prices of the same quarter to
 486 avoid incorporating effects of seasonal fluctuations in the cal-
 487 culus of consumption indicators, perhaps because one aim of
 488 the consumption analysis would be to study its seasonality. In
 489 that case, price means with a value of this special component
 490 common with the optimal set will be ordered at a higher level
 491 than other price means. Here, the final use of consumption
 492 indicators is important in determining the ordering of price
 493 means.

494 *Step 7 (valuation)*. The rules are logical selections. Once the
 495 ordering of sets of price means designed in Step 6 is ready, it is
 496 easy to define the execution process of the valuation.

497 Table IV summarises some possible effects of the different
 498 steps of the valuation process on three types of economic
 499 analyses: aggregate consumption analyses, living standard dis-
 500 tribution analyses, and household behaviour models. These
 501 reflections can be summarised in three levels. Firstly, the val-
 502 uation procedure affects the accuracy of consumption indica-
 503 tors and living standard indicators, at the aggregate level as at
 504 the individual level. Notably, the selection of records to value
 505 determines the coverage of these indicators. Assuming that all
 506 possible consumption quantities have been observed does not
 507 imply that all these records must be valued and used for the
 508 economic analyses, if the available price information is too
 509 mediocre to yield credible numbers. Another danger for the
 510 accuracy of indicators is the occurrence of measurement errors
 511 coming from using imperfect and imprecise combinations of
 512 price type, and geometrical and temporal levels for organising
 513 the price information. Other measurement error for these
 514 indicators, and the macroeconomic variables calculated from
 515 this information arise from the inaccuracy of the proximity
 516 map and of the ordering of price means.

517 Secondly, the valuation process generates inconsisten-
 518 cies with valuation conventions imposed by some eco-
 519 nomic theories. For example: (1) for national
 520 accounting, production is valued at 'producer prices' and
 521 consumption is valued at 'consumer prices'; (2) in welfare
 522 economies, true price indices are often defined as ratios of cost



TABLE IV
Consequences for economic analyses of the steps of the valuation algorithm

Steps of the algorithm	Aggregate consumption analyses	Living standard distribution analyses	Estimates of consumer demand systems and other household behaviour models
1. Selection of records to value	Scope of possible analyses based on values of aggregate variables.	Coverage of living standard variables.	Scope of behaviour studies based on market hypotheses.
2. Samples of elementary prices	Respect of valuation conventions in national accounts (production is valued at producer prices, consumption at consumer prices).	Impact of spatial and temporal distribution of prices on poverty and inequality measures, when using imperfect price indices.	Possibilities of studies of market imperfections (where consumer prices, producer prices and market prices are different).
3. Prices by type, geographical and temporal levels	Aggregation bias and inaccuracy of aggregate variables.	Idem as above.	Unit-value biases.
4. Choice of the optimal price mean indicator	Consistency with valuation conventions for national accounts.	Accuracy of poverty and inequality measures.	Bias and accuracy of agent's responses to price differences.
	Consistency for budget balances.	Consistency with welfare axiomatic (under perfect markets true price indices are ratios of cost functions).	Consistency with hypotheses about agents behaviour and markets (e.g. consumption prices are needed for the basic consumer model).

TABLE IV (Continued)

Steps of the algorithm	Aggregate consumption analyses	Living standard distribution analyses	Estimates of consumer demand systems and other household behaviour models
5. and 6. Proximity map of price means and ordering of price means	Size of bias in measured aggregates as compared with the true unobserved values of these aggregates.	Size and direction of bias in poverty and inequality estimates.	Unit-values bias. Spatial price bias. Temporal price bias.
7. Execution of the valuation	Errors in macroeconomic aggregates.	Errors in poverty and inequality statistics.	Errors in parameter estimates of behaviour models.



523 functions representing the agent preferences; (3) perfect mar-
 524 kets (and therefore a unique price for a given product) are often
 525 assumed in models of consumer demands. Such situation
 526 should be taken into account when proceeding with economic
 527 analyses.

528 Finally, the valuation process is likely to influence the results
 529 of economic analyses. It is for example the case for the measure
 530 of the impact of the spatial and temporal price distribution on
 531 poverty and inequality indicators. Clearly, the available
 532 imperfect price sample for the valuation may influence the re-
 533 sults since price effects may appear at two places: first, in the
 534 calculation of consumption indicators; and second in the cor-
 535 rection of price differences for living standard indicators.
 536 Similarly, when the topic of interest is related to market
 537 imperfections, the used consumption levels may partly result
 538 from such market imperfections and the use of an inappropri-
 539 ate valuation algorithm. However, the most common problem
 540 may be the biases caused in the econometric estimation for (1)
 541 microeconomic behaviour models such as demand systems, (2)
 542 poverty and inequality analyses, (3) macroeconomic models
 543 based on aggregate consumption. Not only the samples of
 544 prices used, but also the way they are organised and the
 545 ordering of price means may affect the estimation results, to an
 546 extent that is unknown. The decomposition of the valuation
 547 process should help in developing analyses of the above effects
 548 and in improving their control. However, further study would
 549 be necessary to analyse the precise effect of the valuation pro-
 550 cess used in a given survey for a specific type of economic
 551 analysis.

CONCLUSION

553 In this paper we first describe the valuation algorithm of the
 554 non-monetary consumption that was used for the consumption
 555 survey of Rwanda in 1983. The majority of the consumption
 556 was valued by using price indicators at local or regional quar-
 557 terly level, although it was not always possible to discard the

558 use of national or yearly price indicators. The analysis of the
559 valuation for this survey shows that compromises are necessary
560 because the samples of available prices are not ideal. It also
561 shows the type of statistical analyses that need to be carried out
562 to design a performing valuation algorithm.

563 Then, extrapolating from this experience, we propose an
564 algorithm that systematises the valuation operations for a
565 consumption survey, and that exhibits the different steps and
566 the decision rules used at each step. Our model reveals the rules
567 for: (a) the identification of consumption records to value; (b)
568 the selection of price records to use for the valuation; (c) the
569 confection of a database of price means; (d) the logical ordering
570 of these sets of price means. This model should ease the repli-
571 cation of valuation operations in consumption surveys, facili-
572 tate the choices for the valuation algorithm, clarify the
573 resources to allot to this operation, help to distinguish the tasks
574 involved in the design and the execution of the valuation, and
575 finally assist in the definition of optimality criteria to judge the
576 quality of the valuation operations.

577 Usually, the valuation method used in a survey is taken
578 for granted by economists. In contrast, our algorithm shows
579 that the consumption indicators used for macroeconomics,
580 estimation of micro-economic demand systems or welfare
581 analysis, are substantially different from the simplicity assumed
582 by the economic theory. This complexity of actual consumption
583 indicators implies that the valuation process may influence the
584 results of microeconomic and macroeconomic analyses.
585 Therefore, the valuation of non-monetary economic operations
586 is a matter to consider seriously. To produce an analytical
587 decomposition of the valuation algorithm should help such
588 control.

589 Naturally, further investigations would be necessary to deal
590 with a given survey and a specific economic analysis in order to
591 isolate what is important in the data generation process related
592 to valuation for the case of interest. Indeed, what is learned for
593 a specific survey cannot always be directly extrapolated to other
594 surveys. The analytical framework that we presented should
595 help the adaptation to different surveys.

596 It is also clear that the valuation is important only if non-
 597 monetary consumption is substantial. This is the case in many
 598 LDCs because of the size of the non-monetary economics in
 599 these countries, but also in more developed countries when
 600 consumption of goods produced domestically, public goods,
 601 externalities, leisure activities and non-market dimensions of
 602 welfare are of interest.

603 Finally, note that there is a comparable problem in Western
 604 societies, the imputation of rent for owned accommodation.
 605 That is, if one owns a house outright and lives in it, taxable
 606 income can be allowed to fall, without a decline in well-being.
 607 More generally, this paper highlights the interplay between
 608 capital ownership (in this case agricultural productive land),
 609 income requirements, and cash expenditure patterns. This is a
 610 near-universal equation.

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NOTES

- 618 ¹ Williams and Zimmerman (1935).
 619 ² Descriptions of typical CS are available in OCDE (1978), Scott, de Andre
 620 and Chander (1980), Booker et al. (1980), Wahab (1980), Grootaert (1985),
 621 Grosh and Munoz (1996).
 622 ³ Bailar (1985), Philipson (1997).
 623 ⁴ Chevy (1962), Winter (1970), Verneuil (1983), United Nations (1983,
 624 1986), Casley and Lury (1987), Morgan (1987), Blaizeau and Dubois (1989),
 625 Biemer et al. (1991), Grosh and Munoz (1996).
 626 ⁵ The main part of the collection was designed with the help of INSEE
 627 (French National Statistical Institute). The author was himself involved in
 628 this project as a technical adviser from the French Cooperation and
 629 Development Ministry.

630 ⁶ Their dates are Round A from 01/11/1982 to 16/01/1983; Round B from
 631 29/01/1983 to 01/05/1983; Round C from 08/05/1983 to 07/08/1983; Round
 632 D from 14/08/1983 to 13/11/1983.

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