Estimation of Water Demand in Developing Countries - An Overview

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1. Introduction

Water resource planners are facing increasing challenges due to the growing hydrologic variability brought about by climate change. Also, incomes and populations of megacities are both growing, placing new pressures on water resources in many places. Hence, water resource managers need a much better understanding of household demand to manage water systems and we believe a review of what we know and do not know about household water demand in developing countries is timely.

On the one hand, household water demand has been extensively analyzed in developed countries, in particular to provide price and income elasticities measures. In these countries, almost all households get a connection to the water network and tap water is usually the unique source of drinking water, in general of satisfactory quality. These characteristics make the estimation of a water demand function relatively straightforward. The main methodological issue that has been extensively discussed is the one related to the non-linearity of the pricing scheme which may cause endogeneity bias at the estimation stage. On the other hand, estimation of water demand functions in less developed countries (LDCs) remains scarce. One main reason is that the conditions of water access vary in general across households, which makes almost impossible an analysis of water demand based on aggregate data and requires well-designed surveys.1 Households in LDCs usually have the choice among a set of water sources, including piped and non-piped sources with different characteristics and levels of services (price, distance to the source, quality, reliability, etc.). Water is thus a heterogeneous good in these countries, contrary to what is usually the case in developed countries (Mu et al., 1990). Finally, getting water from non-tap sources outside the house involves collection costs that need to be taken into account. Carefully designed household surveys are thus needed, and a significant amount of information on water services has to be collected in order to allow for consistent water demand analysis. See also Mu et al. (1990) for related discussions.

In this paper we discuss empirical issues including data requirement and methodological problems that empirical researchers may face when estimating water demand in developing countries.

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1 Most analyses made in industrialized countries have been based on aggregate consumption data provided by water utilities (usually from records which the water utility maintains for billing purposes).
2. Estimation of water demand – A brief literature overview

Water demand estimation in developed countries has been at the core of many empirical papers, starting with the work of Gottlieb (1963) and Howe and Linaweaver (1967). Studies have been made in a large set of countries including Canada (Kulshreshtha, 1996), Denmark (Hansen, 1996), France (Nauges and Thomas, 2000), Spain (Martínez-Espiñeira, 2002), Sweden (Höglund, 1999), and above all the US (Foster and Beattie, 1979; Agthe and Billings, 1980; Chicoine et al., 1986; Nieswiadomy and Molina, 1989; Hewitt and Hanemann, 1995; Pint, 1999; Renwick and Green, 2000). For comprehensive reviews, see Hanemann (1998), Arbués-Gracia et al. (2003), or Dalhuisen et al. (2003).

In almost all studies, the water demand function is specified as a single demand equation for water provided at the tap. Such an approach implicitly assumes that there is no substitute available for water. Water quality as well as quality of the water supply service are not controlled for, in general (for the main reason that there is not much variation in terms of service quality across distribution units). The focus instead has been on the estimation of price elasticity and the measurement of the impact of socioeconomic characteristics (income mainly) on household water demand. The main methodological issues that have been discussed all along is the choice of marginal price versus average price and price endogeneity when households face a non-linear pricing scheme. If theory advocates the use of marginal price (the price of the last cubic meter), average price (computed as total bill divided by total consumption) has however often been preferred. Authors considering average price argue that households are rarely well informed on the price structure and are thus more likely to react to average price than to marginal price. To control for endogeneity, the instrumental variables (IV) approach has been commonly used in the water demand as well as in the labor supply and energy demand literature (Agthe et al. 1986; Deller et al. 1986; Nieswiadomy and Molina 1988, 1989). The use of instruments for the price variable (in a two-stage least squares framework) allows to get unbiased estimates of the price coefficient in the demand equation. The main drawback of this approach though, is that the interpretation of the price coefficient as a price elasticity is conditional on the household remaining within the observed block of consumption (Olmstead et al., 2007). The (only) theoretically consistent approach so far is the

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2 The only exception is Hansen (1996) who considers water and energy prices in the demand function for water.
two-step approach describing the choice of the block (first step) and the choice of consumption inside the block (second step), see Burtless and Hausman (1978) and Hewitt and Hanemann (1995) for an application to the water sector.³

Household water use in developing countries has also been the focus of numerous articles, but empirical evidence regarding factors driving water demand in these countries is still scarce. Most studies on household water use have been under the form of contingent valuation studies to derive willingness-to-pay for getting a house connection to a piped water network or, more generally, improved water services (see among others Whittington et al. 1990a and Whittington et al. 2002); and hedonic price studies to infer the valuation of a piped connection through observations of house prices (see among others North and Griffin 1993, Daniere 1994, or Komives 2003 for a review). In this article, we will focus our attention on articles which provide estimates of the water demand function for households in developing countries. We will not comment on the earliest investigations of water demand that have basically used the standard approach from the literature on industrialized countries (Inter-American Development Bank, 1985a, b, c; Katzman, 1977; Hubbell, 1977) but instead on studies that have tried to account for the specificities of water demand in developing countries. Our purpose is to discuss the various empirical issues that have been encountered and the way authors have addressed them. We will then review and compare the most important outcomes of this set of articles.

3. Discussion of empirical issues

3.1. Underlying theoretical model

The water demand function for households is usually specified as an equation linking water consumption \( q \) (the dependent variable) to water price \( p \) and a vector of demand shifters \( x \) (household characteristics, weather conditions, house equipment) to control for heterogeneity of preferences and outside variables affecting water demand:

\[
q = f(p, x) + u. \quad (1)
\]

³ Again, this approach remains theoretically appealing as long as households are aware of the pricing scheme.
The error term \( u \) is added to this relationship to account for unobservables and/or measurement errors in variables. In most cases, function \( f \) is chosen to be linear in the parameters.

When working on data from industrialized countries, authors commonly assume that this demand function derives from the maximisation of household’s utility subject to a budget constraint, under the assumption that water is a homogeneous good that has no direct substitute or complement. In LDCs, the underlying theoretical model is described slightly differently: water demand is usually assumed to be deriving from a model in which the household is considered a joint production and consumption unit (see Berhman and Deolalikar 1998 for description of such demand models).\(^4\) This is for the main reason that demand for water can be regarded as a derived demand for an input to produce health (since water consumption may have health consequences) and, as a consequence, health enters household’s utility, along with consumption goods, leisure time, and other household’s characteristics such as education. This preference function is then maximised subject to a time-income constraint and a set of production functions. For related discussions, see Acharya and Barbier (2002) or Larson et al. (2006).

3.2. Multiple sources

Households in LDCs may face a choice set of sources. This choice set can include water sources as diverse as in-house tap connections, public or private wells, public or (someone else’s) private taps, water vendors or resellers, tank trucks, water provided by neighbours or water collected from rivers, streams or lakes. The choice set as well as the conditions of access can vary significantly across households. We distinguish three main cases: (1) rural areas – where piped distribution networks are often rare; (2) the formal parts of large cities where piped networks are common, but many people may not be connected for a variety of reasons, and (3) urban slums, which are very heterogeneous. Quality, reliability and conditions of access (distance to get to the source, price) usually vary across sources, making water a heterogeneous good.

\(^4\) In this setting, production and consumption are not assumed to be separable (separability condition may not hold if the nature of one’s occupation and one’s productivity directly interact with one health and nutrition).
Households may rely on a unique source or combine water from different sources. The fact that some households utilize more than one source may indicate either that their use of a particular convenient source is rationed (implying that additional water must be taken from an alternative source), or that it is relatively cheap to take some water but not all from a particular source (e.g. the household has limited capacity to haul cheap water from a given source, and obtain the rest more expensively from another source); or that water from different sources are used for different purposes (drinking, bathing, cleaning, etc.).

When households rely on a unique source or when water use comes primarily from one source, a demand equation for water from that particular source can be estimated from data on the sub-sample of households using that source: among other examples, Rizaiza (1991) estimates separately demand equations for Saudi Arabian households with a private connection and for households supplied with tankers; Crane (1994) specifies separate demand equations for Indonesian households supplied by water vendors and for households relying on hydrants; David and Inocencio (1998) on data from the Philippines estimate separate demand equations for households supplied by water vendors and for households owning a private connection; Rietveld et al. (2000) and Basani et al. (2007) estimate the water demand equation for households with a piped connection. In some cases (Crane 1994, David and Inocencio 1998), dummy variables are introduced in single demand equations to control for possible use of extra sources.

Strand and Walker (2005) and Larson et al. (2006) distinguish only between piped and non-piped water, i.e. they consider non-piped water as a homogeneous good, whatever the source it comes from. In Nauges and Strand (2007), a single equation for non-tap water is also estimated for households in Central American cities, but source-specific coefficients (in particular for the price elasticity) are allowed in the demand model.

Finally, estimates of a demand function pooling data from piped and non-piped water sources are reported in Strand and Walker (2005), using data from Central American cities. These authors reject the null hypothesis of poolability of the data for piped and non-piped households. The latter suggests that the homogeneity assumption for water from different sources is likely to be a too strong assumption in most cases, in particular when comparing piped and non-piped water.
The estimation of (single) source-specific demand equations provides insight on variables driving water use from that particular source, such as its own price, quality and accessibility. This approach does not allow to measure substitutability/complementarity between water from different sources though. A system of water demands is a better suited specification when households collect water from different sources. As described in Nauges and van den Berg (2006a), a general system of \( L \) water demand equations (each of them assumed to be linear in the parameters) could read as follows:

\[
\begin{aligned}
q^1 &= \sum_{k=1}^{L} \gamma_{k}^{1} p_{k} + x^{1} \beta^{1} + u^{1} \\
\vdots \\
q^{L} &= \sum_{k=1}^{L} \gamma_{k}^{L} p_{k} + x^{L} \beta^{L} + u^{L}
\end{aligned}
\]

where \( L \) is the total number of water sources used by the household, \( q^{k} \) (\( k = 1, \ldots, L \)) is water consumption from source \( k \), \( p^{k} \) (\( k = 1, \ldots, L \)) is the price for water taken from source \( k \), the \( x \)-vector gathers other water demand drivers, and \( u^{k} \) (\( k = 1, \ldots, L \)) is the usual idiosyncratic error term. If consumptions and prices are measured in logs, then the coefficient \( \gamma_{k}^{k} \) in equation \( k \) will measure “direct” price elasticity of demand for water from source \( k \). The coefficients \( \gamma_{k}^{j} \) for \( j \neq k \) will provide a measure of “cross”-price elasticity, i.e., by how much will consumption of water from source \( j \) change if the price of water from source \( k \) increases.

A system of water demands is estimated for piped households combining water from a private connection and water from a private well, and for non-piped households combining water from a private well and water provided by neighbors in Nauges and van den Berg (2006a). A Tobit approach is used to deal with censored observations (since not all piped households complement their water consumption with water from a private well). A similar approach is used in Cheesman et al. (2007) to estimate substitutability/complementarity relationships between water provided through the municipal system and water taken from a private well for households with a private connection in Viet Nam.
All the above discussion has concerned the specification of the demand equation for water from some source, conditional on that source being chosen.

3.3. Simultaneity of choice between source and quantity

The simultaneity between choice of water source and choice of quantity was first acknowledged by Whittington et al. (1987). These authors argue that a complete set of water demand relationships should include models of both water source choice and the quantity of water demanded. If not taken into account, the simultaneity in both decisions could lead to biased estimates of the demand parameters. In particular, if some unobserved variables affect both the choice of water source(s) and the quantity of water used, then estimated parameters could suffer from selection bias (Heckman, 1979). This issue was acknowledged by several authors and two-step Heckman procedure for selection bias correction was applied by, among others, Larson et al. (2006) on data from Madagascar, Nauges and van den Berg (2006a) on data from Sri Lanka, Basani et al. (2007) on data from Cambodia, Cheesman et al. (2007) on data from Viet Nam and Nauges and Strand (2007) on data from Central American cities.

In most cases (Larson et al. 2006, Nauges and van den Berg 2006a, Basani et al. 2007), the first stage involves estimation of a probit model for use of a private connection versus reliance on non-tap sources. In Cheesman et al. (2007), a probit model is estimated to control for piped households combining water from their private connection with water from a private well. In Nauges and Strand (2007), the first stage involves estimation of a multinomial logit model (MNL) for choice of the primary non-tap water source. Selectivity correction terms are computed from the estimation of the discrete choice models and added linearly to the demand equations. Statistical significance of these correction terms indicates presence of selectivity bias.5

The first-step estimation is also interesting in itself since it indicates the variables that affect household’s choice of water source. Mu et al. (1990) argue that the discrete choice model offers a more promising approach for developing predictions of the aggregate number of households which will choose a new source than the traditional (linear) demand model. Some other articles have focused only on household’s choice of water source and will be discussed.

5 See Heckman (1979) [resp. Lee (1983) and Dubin and McFadden (1984)] for computation of correction terms when a probit [resp. a MNL] is used in the first estimation stage.

Authors generally agree that both source attributes (i.e., price, distance to the source, quality, reliability) and household characteristics (income, education, size and composition) should enter the choice model. Source attributes account for heterogeneity in water from different sources while household characteristics account for difference in tastes, opportunity cost of time, and perception of health benefits from improved water.

The most frequent specifications for source choice models are the probit model and the multinomial logit (MNL) model. The latter has been used to describe either the primary source of water chosen by households (Mu et al. 1990, Nauges and Strand 2007) or the water source which is chosen for a specific use such as drinking, bathing or cooking (Madanat and Humplick 1993, Hindman Persson 2002). The MNL model relies on the assumption that alternatives are exclusive and rule out the possibility of combination of sources. In Briand et al. (2006), a bivariate probit model is estimated to describe households’ decision to rely on a private connection or/and household’s decision to use public standpipes.

A complex issue which has not been really addressed in the above-cited articles, is the question of the choice set for each household. In peri-urban areas in particular, it may be the case that all households do not have access to all possible water sources (for example, they may not have the possibility to get a private connection because their living area is not supplied by the municipal water network). Hindman Persson (2002) assumes that the choice of household’s location in the city determines the set of available sources for each household.

3.4. The cost of water

Price

Piped water in the developing world is often charged following an increasing block pricing scheme, thus leading to the same endogeneity issues as the ones discussed in the water

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6 Hindman Persson (2002) estimates a probit model but argues that a nested conditional logit model would be better suited. Madanat and Humplick (1993) estimate a two-level sequential choice model to distinguish between the decision to get a private connection and the choice of non-tap sources.
demand literature in industrialized countries. The same techniques are consequently used to estimate water demand in LDCs under complex pricing: David and Inocencio (1998), Strand and Walker (2005), and Nauges and van den Berg (2006a) implement an IV approach to control for endogeneity of price; Rietveld et al. (2000) apply the discrete/continuous model along the lines of Hewitt and Hanemann (1995). Identification of price elasticity may be difficult though in some cases (or even impossible), in particular when data come from households surveys made in a single city or village. In such cases, there is little cross-sectional variation in the policy-relevant variables such as connection costs, tariff and levels of service. For example, Larson et al. (2006) exclude the price of water from their analysis of water demand in Fianarantsoa (Madagascar) due to no cross-sectional variation in this variable (all surveyed households face the same price schedule).

As for non-piped water, the situation varies across places and across sources: water can be distributed free of charge, it can be charged at the bucket, etc. When a price for water exists, then it is usually quite easy to compute a per unit price for each household and each source. If households surveyed get water from different non-tap sources, or from different locations, some cross-sectional variation will likely be observed in the data. The price of non-piped water has been considered exogenous in all studies except in David and Inocencio (1998). These authors argue that the price of vended water in metro Manila (The Philippines) may be endogenous because price is determined by demand and supply factors. More precisely, due to the fragmented nature of the water vending market, household decisions of water demand are likely to influence its price.

Even if free of charge, the collection of water from non-piped sources usually involves costs for hauling water from distant sources.

**Collection costs**

Use of water distributed from non-tap sources usually involves hauling time that comprises time to go to the source, time to wait at the source, and time to haul water. Surveys often gather information on distance or time needed to collect the water, more rarely is information provided on who in the household is in charge of collecting water. The latter is probably the main reason why time cost is usually not translated into a pecuniary collection cost: among other examples, Larson et al. (2006) consider roundtrip walking time to water source and
waiting time at the source; David and Inocencio (1998) use distance from source in meters as an explanatory variable in the demand model; Strand and Walker (2005) consider hauling time per unit of water consumed. In order to convert time cost into monetary cost, one needs to measure the opportunity cost of time of the person in charge of water collection. Conceivably, one could convert collection time into collection costs using a proxy for the unitary cost of time. The latter may be difficult to evaluate though, in particular if one does not know who is in charge of collecting the water. Whittington et al. (1990b) are among the only authors to provide some empirical evidence about the pecuniary cost of collecting water from non-tap sources. These authors develop two approaches based on discrete choice theory to estimate the value of time spent collecting water and illustrate their application using data from Ukunda, Kenya. Their results indicate that the value of time for those households relying on non-tap sources (kiosks, vendors or open wells in this village) is at least 50% of the market wage rate and likely to be near the market wage rate for unskilled labor. Nauges and Strand (2007) are the only authors to transform hauling time into corresponding pecuniary time cost. They use the average hourly wage in the household as the shadow cost of time, but acknowledge that this approximation may overestimate actual costs if the hauling is performed by a child. Finally, Mu et al. (1990) argue that in places where the queue time varies significantly over the course of the day, the collection time could be determined endogeneously.

**Capital investments for coping with unreliable supply**

In response to deficiencies in the water supply system, households may invest in coping strategies, i.e., they may incur fixed costs in the form of investments in alternate supply sources and/or storage facilities. For example, households may buy a storage tank in order to mitigate the reliability and pressure problems that may be associated with private house connections, or they may invest in pumping equipment if relying on well water.

The use of a storage tank or capacity of the water reservoir is controlled for in the demand equation estimated by Crane (1994), Nauges and van den Berg (2006a) and Cheesman et al. (2007). Crane (1994) argues that use of a storage tank (and its capacity) could be endogenous.

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7 These figures were much higher than the ones recommended by the Inter-American Development Bank at the time: for the IADB, time savings should be valued at 50% of the market wage rate for unskilled labor (Whittington et al. 1990b).
in the demand model as the reservoir investment decision was certainly codetermined with the expected need for water.

3.5. Quality of water and quality of service

Since water quality and reliability may vary from one source to another, such variables should be included in the demand models (as well as in models describing source choice). This includes opinion variables about taste, smell, color of water, hours of water availability and potential pressure problems (for piped water). Most of the time these variables are provided by households themselves and may be subject to misreporting. The variables measuring household’s opinion (or perception) about water quality should also be used with caution since they may cause endogeneity in the demand model. For example, households who suffered from water-related diseases in the past may believe that water is less safe than healthy households and hence have different behaviour regarding water use (Nauges and van den Berg 2006b). In order to avoid any such endogeneity bias, one could use average opinion (on water quality) of households living in the same neighbourhood or relying on the same water source, where the average would be computed without considering the opinion of the household under consideration (Briand et al. 2006).

3.6. Summary of data and estimation issues

Data problems

a) Data on water use and price for (metered) households with a private connection are usually recorded from water bills that have to be shown by households. Metered water use data may be poor in some places though because pressures are intermittent, and meters may not provide accurate readings. This is because when the water utility does not supply 24-hour service, air gets into the pipes and the meter can register water passing through when it is really just air.8 The importance of this basic data problem is often not well-appreciated or is simply ignored.

8 Also, because water prices are so low in many places, and corruption is high, water utilities have no incentive to keep meters in good working order. So they are not replaced on a timely basis. The end result of all this is that in many cases no one know how much water a household is using – not the utility, not the household, and certainly not the researcher.
b) When surveys are made in a single village or town, there is usually not enough cross-sectional variation in the conditions of access to water (price, quality and reliability of water). This lack of cross-sectional variation may impede identification of parameters of interest such as the price elasticity of water demand. One may overcome this problem by combining revealed and stated preference data, which results in respondents each having multiple observations (and above all multiple water use observations for different (hypothetical) levels of prices). For applications of this approach to household water demand, see Acharya and Barbier (2002) and Cheesman et al. (2007).

c) Data on consumption and price for households relying on non-piped sources are usually based on self-reported information (households are usually asked to report the number of buckets that they get every day), so there is likely to be substantial measurement error. If consumption data are used as dependent variables, random measurement errors leads to imprecise parameter estimates (the measurement error may however not always be random and may depend on the education of the household’s head for example). If price data are used as right-hand-side variables, then this may cause bias in the parameter estimates.

d) Data on consumption, price, service quality are typically recorded over short reference periods (a month), which may not be representative of household’s normal behaviour (in the event of a wedding or ceremony, households may store water). Also, service quality and quality of water may vary across seasons.

e) Information on income may be difficult to gather in some places. Whittington et al. (1990a) stated “the principal investigators and the enumerators had agreed that it was not possible to obtain accurate information on household income through interviews […]. As a substitute, the enumerator recorded a series of observations about the construction of the house itself, such as whether the house was painted, whether the roof was straw or tin, and whether the floor of the house was dirt or cement.” Basani et al. (2007) use household expenditure as a proxy for income, arguing that households surveyed are more likely to undestate their incomes than overstate their expenditure (Deaton, 1997).

Estimation issues
The specificities of water demand in developing countries complicates the estimation of the water demand function:

f) Households usually face a set of possible choices in terms of water sources but this set may vary across households. The choice of which water source to rely on and the quantity of water used are likely to be determined simultaneously. This implies the specification of a model describing both the choice of the source and quantity of water used. Until now, probit and multinomial logit models have been used most often. The probit model has been used to model the decision of households to get or not a tap connection. The multinomial logit has proven useful to describe household’s choice of primary source of water or household’s choice of water source for a specific usage. The multinomial logit considers choices between exclusive alternatives and relies on the assumption of independence from irrelevant alternatives (IIA). If one wants to model household’s choice of water sources, allowing for combination of sources, then multivariate probit or nested logit models should be the preferred alternative. In the multivariate probit setting, households are assumed to make several decisions, each between two alternatives. In the nested logit setting, alternatives are grouped into sub-groups, which allows the variance to differ across the groups while maintaining the IIA assumption within the groups. The nested logit specification can be seen as a two-(or more) level choice problem. See Greene (2003, Ch. 21) for greater details on these models.

g) Households may combine water from different sources. Aggregation of water from different sources and estimation of a single demand equation for water may provide misleading results. Estimation of a system of demand equations is better suited and allows the measurement of direct and cross-price elasticities. To control for censoring of observations in a system of equations, one could refer to the approach developed by Shonkwiler and Yen (1999).

h) Water from different sources usually have different characteristics in terms of price, conditions of access, reliability and quality. These characteristics should be controlled for in the model describing household’s choice of water source and household’s demand for water. Since data on service quality and reliability are based on household’s opinion, they could suffer from endogeneity or be measured with some error.
i) Piped water is commonly sold through a non-linear pricing scheme and price of water may be endogenous in the water demand equation. Estimation techniques like instrumental variables or discrete/continuous choice approaches (Burtless and Hausman 1978, Hewitt and Hanemann 1995) should be used. Again, the latter is appealing as long as households are aware of the pricing scheme.

j) The price of non-piped water may be endogenous in some cases, in particular if water is re-sold by vendors who may change the price depending on demand and supply conditions.

k) Hauling time may be endogenous if the time to queue at the source varies during the day.

l) The decision of households to engage in coping strategies such as buying a storage tank or investing in pumping equipment may be determined simultaneously with the choice of water source and the quantity of water used, making this variable endogenous.

4. Main results from estimation of water demand equations

The main characteristics of each above-cited study (number of households, study area, time period), type of water access of surveyed households, econometric approach used for estimation of water demand, and main estimation results are shown in Table 1.

The studies recorded in the present article have used data from various regions in the world: Central America (El Salvador, Guatemala, Honduras, Nicaragua, Panama, Venezuela), Africa (Kenya, Madagascar) and Asia (Cambodia, Indonesia, The Philippines, Saudi Arabia, Sri Lanka, Viet Nam), and cover a twenty-year time span (the earliest survey dates back to 1985 while the most recent one has been made in 2006). Despite heterogeneity in places and time periods, authors seem to agree on the inelasticity of water demand in LDCs, with most estimates in the range -0.3 to -0.6. Only two studies find evidence of an elastic water demand: David and Inocencio (1998) using data from The Philippines estimate price elasticity for vended water at -2.1 and Rietveld et al. (2000) estimate price elasticity for piped water at -1.2 using data from Indonesia. Interestingly, Rietveld et al. (2000) are the only authors to use the discrete-continuous model first proposed by Burtless and Hausman (1978) and transposed to
the water demand literature by Hewitt and Hanemann (1995). These authors had used this approach to estimate water demand in Texas and price elasticity was estimated at -1.6. This figure was above (in absolute value) most elasticities that had been estimated in developed countries (Espey et al. 1997 report an average of -0.51 for industrialized countries). The price elasticity estimated by David and Inocencio (1998) should be regarded with some caution since alternative estimation techniques used on the same data (by the same authors) seem to provide very different price elasticities. All in all, and based on the existing studies for household water use in LDCs, estimated price elasticity for these households seem to be in the range of price elasticities estimated in industrialized countries.

Two recent studies (Nauges and van ben Berg 2006a and Cheesman et al. 2007) have shown new insights regarding price elasticity in LDCs. By choosing to estimate system of water demands, they have shown that water from different sources may be used as substitutes. More importantly, these authors report that piped households relying on piped water only are less sensitive to price changes than piped households who complement their water consumption from the tap with water from a private well.

For households relying on non-tap water sources, collection time is found to have a significant negative impact on quantity of water consumed, as expected.

In almost all studies, income elasticity (or expenditure elasticity) is found to be quite low, most often in the range 0.1 – 0.3.

Household size, as expected, is found to be significant in most studies. When the dependent variable is total household consumption, larger households are found to have larger water use. When the dependent variable is per capita consumption, scale effects are confirmed, i.e., per capita consumption decreases with the number of members in the household.

The presence of a storage tank is found to induce higher consumption in two studies (Nauges and van den Berg 2006a and Cheesman et al. 2007). Also, piped water being available for longer hours is found to increase water use by piped households (Nauges and van den Berg 2006). Variables measuring households’ opinion about water quality are not found significant in general.
To account for potential selection bias, authors usually rely on the two-step Heckman approach involving estimation of a discrete-choice model in the first step. As mentioned before, some articles focus only on household’s choice of water source and will be discussed here as well. This includes Madanat and Humplick (1993) on data from Pakistan, Hindman Persson (2002) on data from The Philippines and Briand et al. (2006) on data from Senegal.

The discrete-choice approach has been used to describe household’s choice of primary water source (Mu et al. 1990, Nauges and Strand 2007), or household’s choice of usage-specific water sources (Madanat and Humplick 1993, Hindman Persson 2002), or household’s decision to get or not a private connection (Larson et al. 2006, Nauges and van den Berg 2006a, Basani et al. 2007). Results from these models usually confirm that both source characteristics and household’s characteristics have significant impact on source choice (see Table 2). Source characteristics that are found to be significant drivers of household’s choice are collection time or distance to the source (Mu et al. 1990, Hindman Persson 2002, Briand et al. 2006), water price (Mu et al. 1990), piped water pressure level (Madanat and Humplick 1993), and opinions about taste and reliability of water (Nauges and van den Berg 2006a, Briand et al. 2006). As for household characteristics, almost all studies find evidence that income (or expenditure) and education level (or the ability of household’s head to read and write) drive household’s choice of water source (Madanat and Humplick 1993, Hindman Persson 2002, Briand et al. 2006, Larson et al. 2006, Nauges and van den Berg 2006a, Basani et al. 2007, Nauges and Strand 2007). Mu et al. (1990) and Briand et al. (2006), using data from Kenya and Senegal respectively, find evidence that household’s composition affects choice of water source: in Ukunda (Kenya), households with more women are found to be less likely to use vendors (and more likely to rely on water from wells and kiosks) because more people are available in the household unit to carry water. In Dakar (Senegal), the probability that households use water from the piped system increases if the household’s head is a widow.

5. Conclusion [to be completed]

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9 Even if the empirical evidence is rather limited, it is interesting to mention that no evidence for selection bias due to the choice of water sources was found in Nauges and van den Berg (2006a), Basani et al. (2007) and Cheesman et al. (2007).
This overview of empirical issues has shown that a careful analysis of households’ water demand in LDCs requires a high level of information from the household. Things to remember when designing a survey on water demand are:

a- Surveys should ideally be made in different cities/villages in order to have some cross-sectional variation in conditions of water services, in particular price, connection fee, quality and reliability of services.
b- In most cases, only data on sources that are actually used by the surveyed household are available. Ideally, one should identify the complete set of sources available to the household and gather information on time to walk to the source and time to wait at the source, price of the water, possible rationing or constraints (opening hours, limited availability) and quality of the water from each source (even if not used by the household). This is a prerequisite for consistent estimation of household’s choice of water sources.
c- The measurement of hauling costs is not easy, in particular when one does not have information on who is in charge of collecting water in the household. Information of the persons in charge of collecting the water should be gathered.
d- There is no clear evidence that households who have a piped connection are aware of the water pricing scheme. Whether or not households know the price of water is likely to depend on factors such as the share of water bill in overall expenditure, the complexity of the water pricing scheme, the frequency of billing, and the education level in the household. At the time of the survey, interviewers should test households’ knowledge about their consumption and water expenditure of the last period, and the pricing scheme. This issue has usually been ignored in studies using data from developed countries.10

As a conclusion, we would also like to point out that there are important questions about household water demand behaviour in developing countries that have not yet been addressed or simply cannot be addressed with existing data. First, because most data set are cross-sectional, dynamic analyses of water demand are not doable (by “dynamic” water demand, we mean a water demand function in which consumption of the last period is included in the list of covariates). Estimation of water demand in a dynamic framework is useful though, since it provides measures of households’ responses to a changing environment on the long-run.

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10 One exception is Gaudin (2006) which tests, on US data, if differences in the informational content of bills may affect the intensity with which consumers respond to price signals. She finds that price elasticity increases by 30% or more when price information is given on the bill.
Using aggregate data from France, Nauges and Thomas (2003) have shown that the “long-run” price elasticity of water demand (that is the change in consumption following several years of price increase) was significantly higher in magnitude than the “short-term” price elasticity (i.e., price elasticity derived from the estimation of the traditional static demand function). Such analyses have not been performed yet on LDCs.

Second, existing data do not allow us to measure how household water use would respond to the establishment of dual networks (one for drinking and cooking, and the other for uses that do not require such high quality water).

Third, welfare analysis following changes in the conditions of water supply for households in LDCs remains a difficult question, in particular when piped water is charged following a block-pricing scheme and when scenarios involve the connection to the piped network of households that are currently without a connection. As discussed earlier, consistent estimation of water demand under block pricing and the computation of the change in consumption following a change in price is computationally difficult (for details, see Olmstead et al., 2007). The other issue arises from the fact that it is difficult for the researcher to assess water demand for piped water, for households who currently do not have a connection to the piped network. The assumption that households without a piped connection will behave as households who currently have one, is likely to be too strong in most cases, since there is evidence that household’s own characteristics drive both their choice or access to specific water sources and the quantity of water they use.
References


Deaton, 1997


Inter-American Development Bank, Haiti: Second stage of the community health posts and rural drinking water supply program, project report, Washington D.C., 1985b.

Inter-American Development Bank, Honduras: Rural water supply program (III stage), project report, Washington D.C., 1985c.


<table>
<thead>
<tr>
<th>Reference</th>
<th># of households / study region / period</th>
<th>Type of water access</th>
<th>Demand model specification and estimation method</th>
<th>Dependent variable</th>
<th>Significant explanatory variables and estimated elasticities</th>
</tr>
</thead>
</table>
| Mu et al. (1990)        | 69 hh from Ukunda (Kenya) 1986          | - kiosks  
- water vendors  
- open wells  
- hand pumps | Single demand equation with dummy to control for type of water access  
OLS method | Water used per capita per day | - collection time (-)  
- income (+) |
| Rizaiza (1991)          | 563 hh from four major cities in Saudi Arabia 1985 | - private connection  
- tankers | Separate demand equations for hh with a private connection and hh supplied with tankers  
OLS method | Annual water demand per household | - price elasticity ranging from -0.40 (for tankers water) to -0.78 (for piped water)  
- family size (+)  
- income elasticity in the range 0.09 – 0.20  
- average temperature (+)  
- dummy for garden in the property (+) |
| Crane (1994)            | 291 hh from Jakarta (Indonesia) 1991    | - piped system  
- water vendors  
- public hydrants  
- hh resellers  
- neighbors with in-house connection | Separate demand equations for hh supplied by vendors and hh relying on hydrants + dummy to control for use of extra sources  
OLS method | Hh monthly water use | - price elasticity ranging from -0.48 (for vended water) to -0.60 (for hydrant water)  
- time per purchase (- for vended water)  
- capacity of water reservoir (+ for vended water) |
| David and Inocencio (1998) | 506 hh from Manila (The Philippines) 1995 | - piped system  
- tubewell pumps  
- water vendors | Separate demand equations for hh supplied by vendors and for hh with a private connection  
2SLS estimation for correction of price endogeneity | Hh monthly water use | - price elasticity estimated at -2.1 for vended water  
- income elasticity estimated at 0.3 for vended water |
| Rietveld et al. (2000)  | 951 hh from Salatiga (Indonesia) 1994   | - private connection  
- neighbors  
- community water terminal  
- wells  
- rivers | Single demand equation for water from a private connection  
Discrete/continuous approach of Burtless and Hausman (1978) – ML method | Hh monthly water use | - price elasticity: -1.2  
- income elasticity: 0.05  
- household size (+)  
- use of extra sources (-) |
<table>
<thead>
<tr>
<th>Reference</th>
<th># of households / study region / period</th>
<th>Type of water access</th>
<th>Estimation method</th>
<th>Dependent variable</th>
<th>Significant explanatory variables and estimated elasticities</th>
</tr>
</thead>
<tbody>
<tr>
<td>Strand and Walker (2005)</td>
<td>About 3,700 hh from 17 cities in Central America, Surveys made between 1995 and 1998</td>
<td>- private connection - public taps - trucks - wells - rivers / lakes</td>
<td>Separate demand equations for hh with a private connection and for non-tap hh 2SLS estimation for tap water equation and OLS estimation for non-tap water equation</td>
<td>Hh monthly water use</td>
<td>- price elasticity in the range -0.3 (for hh with a private connection) to -0.1 (for non-tap hh)  - income elasticity less than 0.1  - household size (+)  - hauling time (- for non-tap water)</td>
</tr>
<tr>
<td>Larson et al. (2006)</td>
<td>547 hh from Fianarantsoa (Madagascar) 2000</td>
<td>- private connection - public taps - wells - natural sources</td>
<td>Separate demand equations for collecting hh and hh with private connections Two-step Heckman approach to control for use of a private connection</td>
<td>Hh monthly water use</td>
<td>- household size (+)  - roundtrip collection time (- for collecting hh)</td>
</tr>
<tr>
<td>Nauges and van den Berg (2006)</td>
<td>1,800 hh from Sri Lanka 2003</td>
<td>- private connection - public taps - public and private wells - neighbors - surface water</td>
<td>Separate systems of demand equations for piped and non-piped hh Two-step Heckman approach to control for use of a private connection – Tobit model for censored observations</td>
<td>Water use per capita per month (for piped hh) or per day (for non-piped hh)</td>
<td>- price elasticity in the range -0.32 to -0.60 for piped hh  - collection time (- for non-piped water)  - income elasticity: 0.15 for piped hh and in the range 0.07 – 0.22 for non-piped hh  - number of rooms (+)  - hours of piped water availability (+ for piped water)  - household size (- for non-piped hh)  - ethnic group for non-piped water</td>
</tr>
<tr>
<td>Basani et al. (2007)</td>
<td>782 hh from 7 provincial towns in Cambodia</td>
<td>- private connection - rivers / streams - tanks - wells - vendors</td>
<td>Single demand equation for connected hh Two-step Heckman approach to control for use of a private connection</td>
<td>Hh monthly water use</td>
<td>- price elasticity in the range -0.5 to -0.4 (connected hh)  - expenditure elasticity in the range: 0.2 to 0.7 (connected hh)</td>
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### Table 1. Estimation of water demand in developing countries – Overview (cont’d)

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</tr>
</thead>
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<tr>
<td>Cheesman et al. (2007)</td>
<td>166 hh from Buon Ma Thuot (Viet Nam)</td>
<td>- private connection</td>
<td>Separate estimation for hh using a private connection only (single equation) and hh combining water from a private connection and well water (system) Two-step Heckman approach to control for use of well water</td>
<td>Hh monthly water use</td>
<td>- price elasticity for piped water estimated at -0.06 for hh using a private connection only and at -0.53 for hh using a private connection and well water - income elasticity: 0.14 - household size (+) - use of a storage tank (+) - capital value of pumping equipment (+)</td>
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<tr>
<td></td>
<td>2006</td>
<td>- private wells</td>
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<td></td>
<td></td>
<td>- vendors</td>
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<tr>
<td>Nauges and Strand (2007)</td>
<td>553 non-tap hh from 3 cities in El Salvador and 826 non-tap hh from Tegucigalpa (Honduras)</td>
<td>- private or public well</td>
<td>Single demand equation for non-tap water, allowing for elasticities to water cost varying with type of water access Two-step Heckman approach to control for choice of primary non-tap source</td>
<td>Water use per capita per month</td>
<td>- total water cost (price + hauling cost) elasticity in the range -0.4 to -0.7 - income elasticity in the range 0.2 to 0.3 - household size (-)</td>
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<td></td>
<td>1995 - 97</td>
<td>- someone else’s private tap</td>
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<td>- public tap</td>
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<td></td>
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<td>- trucks</td>
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<td></td>
<td>- rivers/lakes</td>
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<tr>
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<td>Model specification and estimation method</td>
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<td>Mu et al. (1990)</td>
<td>69 hh from Ukunda (Kenya) 1986</td>
<td>- private connection - kiosks - water vendors - open wells - hand pumps</td>
<td>Choice of primary water source</td>
<td>Assumed to be exogenous – all hh face the same choice set</td>
<td>Multinomial Logit model (MNL) – ML approach</td>
</tr>
<tr>
<td>Madanat and Humplick</td>
<td>900 hh from Faisalabad (Pakistan)</td>
<td>- private connection - public piped water - motor/hand pumps - water vendors - wells/ponds/ canals</td>
<td>Choice of usage-specific water source (drinking, bathing, washing, etc.)</td>
<td>Not all the alternatives are available to all hh Focus here on hh who can connect to the water system</td>
<td>Two-level decision model Sequential ML estimation</td>
</tr>
<tr>
<td>Persson (2002)</td>
<td>769 hh in Cebu (The Philippines)</td>
<td>Choice of drinking water source</td>
<td>Suggests that the set of available sources is determined by choice of living areas</td>
<td>Suggests using a nested conditional logit but uses a MNL instead because of too few observations – ML</td>
<td>- annual labor income - walking time to source</td>
</tr>
<tr>
<td>Larson et al. (2006)</td>
<td>547 hh from Fianarantsoa (Madagascar) 2000</td>
<td>- private connection - collecting hh</td>
<td>Decision to get or not a private connection</td>
<td>Assumed to be exogenous – all hh face the same choice set</td>
<td>Probit model – ML</td>
</tr>
<tr>
<td>Briand et al. (2006)</td>
<td>301 hh from Dakar (Senegal)</td>
<td>- private connection - public standposts</td>
<td>Decision to use a private connection and/or a public standpost</td>
<td>Assumed to be exogenous – all hh face the same choice set</td>
<td>Biprobit - ML</td>
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<td>Assumed to be exogenous – all hh face the same choice set</td>
<td>Probit model - ML</td>
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<td>(2006)</td>
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<td>- private or public well - someone else’s private tap - public tap - trucks - rivers/lakes</td>
<td>Choice of primary non-piped water source</td>
<td>Assumed to be exogenous – all hh face the same choice set</td>
<td>MNL model - ML</td>
</tr>
</tbody>
</table>