

An improved area-based guideline for domestic water demand estimation in South Africa

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Abstract

Increased infrastructural development and potable water consumption have highlighted the importance of accurate water-demand estimates for effective municipal water services infrastructure planning and design. In the light of evolving water consumption trends, the current guideline for municipal water demand estimation, published in 1983, needs to be revised. This study investigated, using regression analyses, the combined effect of various socio-economic and climatic parameters on municipal water consumption with the objective of determining the dominant influencing parameters and suggesting a new guideline for water-demand estimation. To this end, an initial database comprising more than 2.5 x10⁶ metered water consumption records extracted from 48 municipal treasury databases, which are located within 5 out of the 7 South African water regions was analysed. Each of the 48 municipal treasury databases spanned a period of at least 12 months. The final amalgamated database, after rigorous cleaning and filtering, comprised 1 091 685 consumption records. Single variable and stepwise multiple variable regression analyses were utilised. Results show that stand area, stand value and geographical location are the dominant parameters influencing municipal water consumption, with stand area and stand value positively correlated to water consumption. In suggesting a new municipal water-demand estimation guideline, these three parameters were considered. Stand value, however, fell away as a reliable parameter for estimating water consumption because of the inconsistent basis for predicting stand values due to the constant fluctuations in the value of property, and municipal valuations that often become outdated. Inland and coastal geographical locations exhibited different consumption patterns, with coastal stands of the same stand area and stand value consistently consuming less water than inland stands. These should therefore be treated separately in any design guideline. Stand area then became the best parameter on which to base water-demand estimations. A single guideline curve is therefore proposed which gives various confidence limits for estimating water demand in South Africa, based on stand area.

Keywords: domestic water demand estimation, area-based guideline

Glossary

Guideline	A document approved and published by the relevant government institution(s) and extensively used by industry with the aim of guiding decisions and criteria in specific areas (in this study, domestic water demand estimation)
Stand	A plot of land
City	A municipal centre incorporated by a province
Town	A built-up area with a name, defined boundaries, and local government, that is larger than a village and generally smaller than a city
Suburb	An outlying residential district of a city
Ward	An administrative division of a city or town, typically represented by a councilor or councilors

Introduction

Background to the study and motivation

In South Africa, the expansion of urban areas, increased infrastructural development and constant need for potable water services,

have brought to light the importance of accurate water-demand estimates in municipal water services planning and design. Water-demand estimates are used to calculate peak water demands and sewer flows and thus determine municipal water and sewer infrastructure requirements. Inaccurate estimates result in a deficiency in basic design information that could lead to inadequate service due to over-, or under-design of water supply infrastructure. A key input therefore in municipal water services planning and design is the estimation of present demand, and the prediction of future water demand. In addition, South Africa is a water-scarce country that constantly strives to apply its available water resources in the most efficient and equitable manner. Different users including industry, domestic, agriculture and the environment, vie for the available resources and have to be awarded an equitable and adequate share. Inaccurate estimates of the water needs of different users may well result in inequitable distribution.

It has been recognised that domestic water-demand estimates should preferably be based on actual water consumption. However, information on actual water consumption is not always readily available and as a consequence, domestic water-demand estimates measured as annual average daily water demand (AADD) are still mostly based on stand area (CSIR, 2003; Jacobs et al., 2004). The estimates obtained are then multiplied by the number of stands and peak-to-average ratios are applied.

The most commonly used South African design guideline for municipal water-demand estimation was first published in 1983 (the so-called *Blue Book*) (Table F6 and Figure F2; DCD,

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1983). The latest version of this document (the so-called *Red Book*) (Tables 9.10, 9.11, 9.14 and Fig. 9.9; CSIR, 2003) contains the same design guideline for municipal water-demand estimation as that published in the *Blue Book* (the colour codes for each book simply refers to the colour of the folder housing the guidelines). This is despite the changes that have occurred in municipal water consumption over the past 25 years. The only significant change that has occurred in these guidelines since publication in 1983 has been to distinguish between domestic water-demand estimation in developing areas and developed areas. For domestic water-demand estimation in developing areas the guideline provides a typical consumption value and a range in $\ell/\text{cap}\cdot\text{d}$ for communal water points, stand pipes and yard taps (Tables 9.10 and 9.11; CSIR, 2003). For domestic water-demand estimation in developed areas, AADD (for single residential stands) is based on stand area (Fig. 9.9; CSIR, 2003). In this figure, an upper and lower limit for domestic demand as a function of stand area is given. The designer is expected to estimate the design demand between these limits, and take into account parameters such as climate, income level and cost of water.

Currently, the *Red Book* AADD guideline is widely used in the Southern Africa region and as mentioned above, does not take into consideration the evolution of municipal water consumption over the past 25 years. A revised guideline is therefore needed. This paper reports on an investigation of metered water consumption over a period of at least 12 months for an initial database of more than 2.5×10^6 consumers from a cross-section of South African towns and cities which are located within 5 out of 7 water regions. After rigorous cleaning and filtering, the final database comprised 1 091 685 consumers. The combined effect of various parameters which influence municipal domestic water consumption is investigated using regression analyses, the dominant parameters influencing water consumption are determined, and an improved guideline for estimating domestic water demand in South Africa is suggested.

Previous studies

The study by Garlipp (1979) was one of the earliest of its kind in South Africa. Garlipp (ibid.) investigated a number of parameters influencing domestic water consumption in Pretoria, Bloemfontein, Cape Town, Port Elizabeth and Durban. Data for this study were sourced from meter readings logged during the study, water meter books from individual customers, and social surveys. Household size was found to be the most significant parameter influencing domestic water consumption with water consumption per capita decreasing with an increase in household size. Other parameters that had a positive influence on domestic consumption were prolonged high temperatures, stand area and income. Domestically, water was found to be largely consumed internally at lower temperatures and externally, at higher temperatures. A limitation of the study was the inadequate and biased social surveys resulting from poor survey responses conducted primarily amongst the engineering fraternity in South Africa.

A study by Stephenson and Turner (1996) investigated different income users in the Gauteng area (242 stands in a high-income residential area, 7 119 stands in 7 middle-income residential areas, 2 370 stands in 2 low-income residential areas/townships and 3 suburbs where the land use was predominantly of a commercial and industrial nature). The study confirmed the *Blue Book* guideline – that stand area exerted the most influence on domestic water consumption. Other parameters that were found to influence water demand included income, population

density, water supply level of service, and housing type. Some limitations of the study included the use of average stand area for all the stands in an area or zone, as this likely led to the misrepresentation of stand area. Also, the guidelines presented in the study for each study area were likely to generate higher AADD values than actual domestic water consumption because the estimates included water losses, and potentially some non-domestic water use (e.g. fire-hydrant water).

Van Vuuren and Van Beek (1997) investigated domestic and non-domestic water consumption for 69 supply areas in the Pretoria vicinity using measured consumption figures for different income categories over about 12 years (March 1982 to October 1994). The study confirmed a strong correlation between domestic consumption and household income with high-income households consuming significantly more water than middle- and low-income households. Also, water consumption patterns of high-income households were found to be more climate (rainfall and temperature) sensitive than in middle- and low-income households, primarily because of the large outdoor water demand in high-income households. For non-domestic users, climate had a negligible effect on water consumption. In the study, all users responded to water restrictions, although high-income households took longer to respond. High-income households however, recovered quicker when the restrictions were lifted. In comparison with the *Red Book* AADD guideline (CSIR, 2003), Van Vuuren and Van Beek (1997) found the domestic AADD for all income categories in the Pretoria vicinity to be lower than estimated by the guidelines. The study had limitations with regard to:

- The accuracy of some of the parameters used (e.g. meter-reading intervals and land-use characteristics)
- Income levels were defined and users categorised by means of property tax information
- The study only considered formal residential developments with potable water connections and water-borne sanitation on each stand, thus excluding informal residential areas.

It is of note that stand areas for the low-income category considered by Van Vuuren and Van Beek (1997) were significantly greater (ranging from 600 to 2 900 m^2) than the stand areas considered by Stephenson and Turner (1996) (ranging from 200 to 1 400 m^2).

Van Zyl et al. (2003) investigated the elasticity of water price, water pressure, household income and stand area for residential water consumption in some Gauteng residential areas (i.e. Alberton, Boksburg, Centurion and Midrand), comprising more than 110 000 users. This pilot study investigated the strengths and weaknesses of end-use modelling as a water-demand predictor in South Africa. The study grouped end-uses into indoor consumption, outdoor consumption and leakage, and provides ranges of elasticity values identified for the modelling parameters (i.e. water price, water pressure, household income and stand area). The potential effects of these elasticity values on water consumption were then evaluated through a sensitivity analysis. Price was found to exert the most significant impact on domestic water consumption patterns. Household income, stand area and water pressure had positive demand elasticities. A case study was used to show the potential for end-use modelling to investigate scenarios of simultaneous changes in different parameters. The merit of this study lies in the fact that typical South African conditions were investigated with regards to suburban and township developments. Potential parameters influencing water demand such as climate, geographic location, level of service and age of infrastructure were not considered in

the analysis. Also, other user categories such as parks, schools, business/commercial sites, etc. were not considered.

New guidelines for domestic water-demand estimation were proposed by Jacobs et al. (2004). The proposed guidelines were based on the analysis of 582 997 domestic users with at least 12 months of data (varying from December 1999 to July 2003) country-wide from municipal treasury data. Stand area was used as the only influencing variable and separate guidelines were proposed for four different geographic regions. Township and suburban areas were considered only in one of the regions. The study analysed consumers using less than 20 kl/d and located on stand areas of between 50 to 2 050 m². The authors concluded the *Red Book* AADD guideline to be conservative. The work by Jacobs et al. (2004) presents a valuable discussion with regards to estimating residential water demand in Southern Africa using a single-coefficient model that relates water demand with stand area. The work presented here is considered to be an improvement over Jacobs et al. (2004) because more data points are utilised, the study is specific to South Africa and several variables influencing water consumption, apart from stand area, are considered in the analysis. In subsequent related studies, household size was identified as being the most significant determinant of indoor water demand, hot water demand and wastewater flow (Jacobs and Haarhoff, 2004; Jacobs, 2007).

Husselmann and Van Zyl (2006) investigated the independent effects of stand area and income (using stand value as a surrogate for income) on water consumption. The study used measured consumption data for 769 393 residential stands in Gauteng with the data divided into 6 stand-area and 6 stand-value categories. The authors found a strong link between water consumption and income, but found that stand value is too variable to be used as a parameter in a design guideline. Hence, they concluded that stand area provides the best basis for a design guideline. A comparison of the results with the *Red Book* AADD guideline showed that the *Red Book* guideline underestimates the AADD in the 300 to 700 m² stand area range and overestimates the AADD for stand areas larger than 700 m². New guideline curves were proposed.

Internationally, one of the longest and most continuous national domestic water use databases (at 5-year intervals from 1950-2000) was generated by the US Geological Survey. Domestic water use included water for drinking, food preparation, bathing, laundry and dishes, flushing toilets, and watering lawns and gardens. Analyses of the database showed a continual increase in water use from 1950-1995 largely due to increases in population. The analyses did not, however, provide any evidence explaining what other factors besides population increase accounted for differences in consumer behaviour (Solley et al., 1998). In the UK, metering and the costs that it incurred for the domestic consumer were shown to cut water demand by about 21% for about 50 000 households in the Isle of Wight. The East Anglian Regional survey of domestic water consumption involved 2 000 properties and showed that per capita consumption could be related to four main variables, i.e. household size, type of house, income level and rateable value of properties (Edwards and Martin, 1995).

Summary of unresolved issues

Below are some of the research issues unresolved in the previous South African studies:

- Three of the 6 studies indicate that AADD is primarily influenced by stand area. The other studies indicate the most influential parameter to be price, household income and

household size. Consensus on the most dominant parameters is therefore required

- There has been limited investigation on the effect of climate on various categories of users, located in various types of developments (cities vs. towns) and geographical locations (inland vs. coastal)
- Most of the previous studies considered parameters influencing water consumption individually and did not investigate the combined effect of these parameters on water consumption
- Previous studies (Jacobs et al., 2004; Husselmann and Van Zyl, 2004; and Van Vuuren and Van Beek, 1997) indicate that the *Red Book* AADD guideline may be conservative. Hence, this guideline needs to be revised in the subsequent *Red Book* edition.
- Previous guidelines for AADD are limited in scope compared to this study, which is based on a larger database than previously used.

This study reports on an investigation of metered water consumption over a period of at least 12 months (the vast majority for at least 24 months) from an initial database of more than 2.5 x10⁶ consumers from a cross-section of South African towns and cities which are located within 5 out of 7 water regions. The combined effect of various parameters which influence municipal domestic water consumption is investigated using regression analyses, the dominant parameters influencing water consumption are determined, and suggestions are made for improved estimation of domestic water demand in South Africa. Non-domestic demand estimation falls outside the scope of this paper.

Methodology

This study utilised municipal water-meter readings stored in treasury databases as the source of consumption data, as was the case with Jacobs et al. (2004) and Husselmann and Van Zyl (2006). This made it possible to study a large number of consumer records and to conduct rigorous statistical analyses. The relatively large size of the database made it possible to investigate the distribution of the data in greater detail, and to have representative samples for specific consumption characteristics. In South Africa, most users in developed areas are metered and meter readings are generally taken at regular intervals in order to generate monthly water bills. Until recently, actual meter readings were often hidden in complicated databases or could not be directly or easily accessed. However, the past decade has seen significant software developments that now enable the abstraction and analysis of information from treasury databases (Jacobs et al., 2004). One such tool that achieves this end is the *Swift* Software Programme (GLS, 2007). *Swift* allows a user to interrogate and access municipal treasury databases to obtain demographic data, stand characteristics (area and value) and recorded water consumption for individual consumer connections. A large number of municipalities throughout South Africa (e.g. Tshwane Metropolitan, Ekurhuleni, and Johannesburg Water) have implemented *Swift* and the existing treasury databases in these municipalities cover years of consumption data for millions of users. Van Zyl and Geusteyn (2007) undertook an extensive study to develop an archive of water consumption nationally. Data used in this study are a subset of the data compiled in Van Zyl and Geusteyn's (2007) study.

It was assumed for this study that the accuracies of the consumer meters studied are adequate. To address problems like

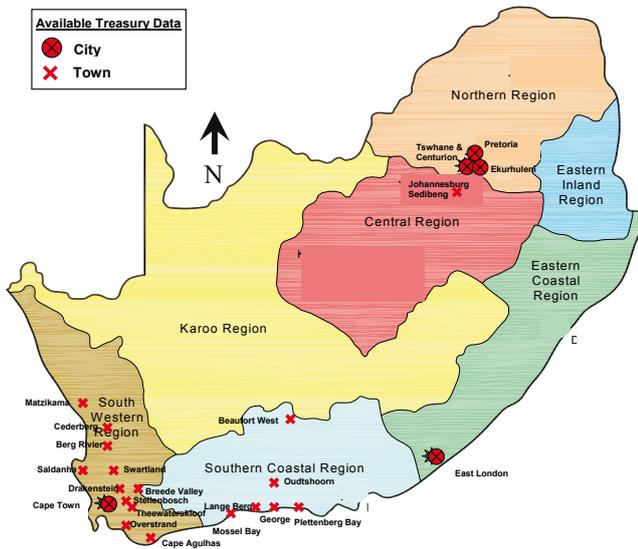


Figure 1

Location of the municipalities used in the study on the map showing the 7 South African water regions

meter clock-overs or replacements, this study made use of the data-cleaning functions contained in *Swift*. In addition, this study also undertook rigorous data-cleaning processes to ensure the integrity of the final database used for analysis.

This study has a number of limitations, including the following:

- Water consumption is an inherently variable process and thus, any metered consumption data will thus include a measure of variability and uncertainty
- Stands using alternative water sources (e.g. groundwater from boreholes, rainwater collected from roofs and on-site reuse of grey water) were not identified in this study. As such, all stands in the database are assumed to utilise only potable municipal supply.

- The length of time for the climatic parameters generated in the study coincides with the years of treasury data available. As such, any long-term climatic effects on water consumption beyond the data length were ignored.

The methodology undertaken in this study therefore involved the following: extracting water consumption data from municipal treasury databases; extracting data on the parameters influencing water consumption from several sources; cleaning the data; and detailed analyses of the data.

Data on water consumption

Forty-eight municipal treasury databases totalling more than 2.5×10^6 records were collected and analysed in this study. This included 4 metros (Johannesburg, Tshwane, Ekurhuleni and Cape Town) and 151 other cities or towns. The location of the municipalities included in the study is indicated on the map (Fig. 1) showing the 7 South African water regions proposed by Basson et al. (1997). These regions were created by grouping together catchments of broadly similar hydro-meteorological characteristics. It is clear from Fig. 1 that the municipalities investigated are fairly representative of the South Western, Southern Coastal and Northern water regions of the country. Only one municipality each in the Eastern Coastal region and Central region was available for this study. Two water regions namely the Eastern Inland and the Karoo had no representation in the database. However, since municipal treasury data were obtained using *Swift*, this study was restricted to municipalities that have implemented *Swift*.

Table 1 provides a summary of the database used in this study according to water region and municipality. Each stand has a calculated AADD value based on at least 12 months of consumption data (see Fig. 2). Only single residential stands were considered in this study, i.e. residential properties each consisting of a single dwelling on a stand. Group housing (i.e. flats, townhouses) was not considered in this study.

Water region	Municipality	Initial No. of domestic and non-domestic stands	Initial No. of domestic stands	No. of domestic stands after cleaning and filtering
Central	Sedibeng	170 126	144 135	35 436
Eastern Coastal	Buffalo City (East London)	119 748	102 665	69 581
Northern	Ekurhuleni, Johannesburg Water, Randfontein, Tshwane	1 629 636	1 377 457	840 639
South Western	Berg River, Blaauwberg, Breede River, Breede Valley, Cape Agulhas, Cederberg, Drakenstein, Helderberg, Matzikama, Oostenberg, Overstrand, Saldanha Bay, Stellenbosch, Swartland, Theewaterskloof, Tygerberg	557 671	457 613	359 018
Southern Coastal	Beaufort West, George, Langeberg, Mossel Bay, Oudtshoorn, Plettenberg Bay	111 825	68 685	49 119
Eastern Inland	-	-	-	-
Karoo	-	-	-	-
Sub-total		2 589 006	2 150 555	1 353 793
Additional filtering (outliers)				- 262 108
TOTAL				1 091 685

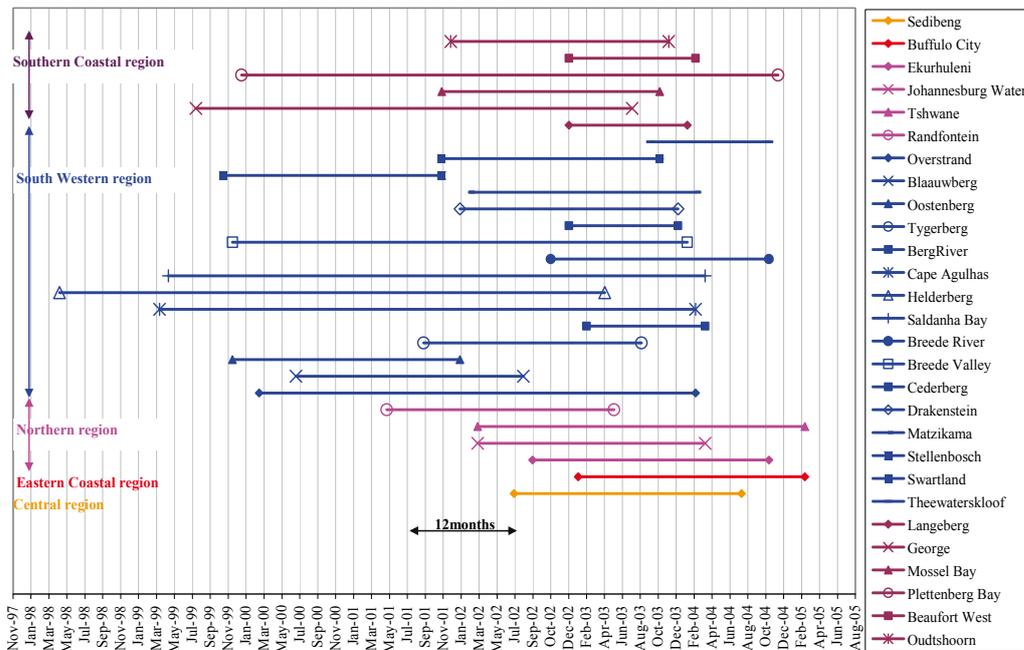


Figure 2
Spread and duration of each municipal database

Data on parameters influencing water consumption

Information was collected from various sources and linked to the consumption data to allow correlations to be analysed. Table 2 shows the parameters used in this study and their sources. The MAP (mean annual precipitation), MAE (mean annual evaporation), and temperatures were linked to the water consumption data on the basis of municipality and, where higher resolution was possible, at a suburb level. Linking the socio-economic data to the consumption data was a complicated process as the former was only available on the political ward level. It was not ideal to use socio-economic data provided on the political ward level since different socio-economic groupings, and thus different water consumption groupings, may occur in the same ward. Reconciliation of the two data sets was made possible using a GIS interface. A single database was created after linking the data.

Data cleaning

To ensure the integrity of the final database, three data-cleaning phases were implemented. In the primary data-cleaning phase, *Swift* adjustment codes (assigned where *Swift* identifies certain anomalies or errors in the data) were used to exclude records with inconsistent readings or dates due to meter replacements or clock-overs, and records with less than 12 months of data. In the secondary data-cleaning phase, records or stands identified as un-metered, vacant, pre-paid, duplicate, including more than

one type of land use, group housing or classified as non-domestic were excluded from the database.

In the tertiary data-cleaning phase, filters were applied to the database to exclude users with unrealistically low or unrealistically high stand areas and stand values. These consumers were omitted not on the basis that the stand areas and values may be wrong, but on the basis that it is not practical to design water-supply infrastructure for consumption values that are unrealistically high or low. Table 1 shows the number of records used in the analyses. The cleaned database contained the following descriptors/values:

- A unique stand identifier and address that can be linked back to the treasury database
- A land-use code and suburb category in which each stand is located
- AADD in $\text{k}\ell/\text{d}$ and the period of calculation of the AADD
- Area (m^2), value, development level (city or town), and geographic location of each stand
- MAP (mm) and MAE (mm) for the area
- Average maximum and minimum temperatures for the water consumption period
- Unique political ward ID as obtained from the South African Demarcation Board
- Percentage of unemployed persons in the economically active population*
- Percentage of households with formal housing*
- Average household size (no. of persons)*, house size (no. of rooms)*, and household income*

Source of data	Parameters
Municipal treasury data	Stand area, stand value and improvements, monthly meter readings
South African Municipal Demarcation Board	Level of unemployment, annual household income, level of sanitation, level of water service, household size, house size/floor area, dwelling type
South African Weather Service	Rainfall, daily temperatures (average maximum and minimum)
Surface Water Resources of South Africa by Midgley et al. (1994)	Mean annual precipitation, mean annual evaporation

- Percentage of houses with potable water house connections*
- Percentage dwellings with waterborne sanitation*

*data only available on ward level

Data analyses and demand estimation

The first step in the analysis was a direct comparison of each suburb's AADD vs. average suburb stand area, with the *Red Book* AADD guideline. The analysis was simply to generate a preliminary idea of how on average, the data compared with the *Red Book* guideline. It was assumed that climatic and socio-economic conditions in each suburb were homogeneous. The average stand area and AADD for each suburb were thus determined and these values superimposed on the *Red Book* guideline. A best-fit line was generated using a single variable regression model with average stand area specified as the independent variable and average AADD, the dependent variable.

The next step in the data analysis was an attempt at quantifying the level of influence the different parameters (climatic and socio-economic) had on domestic water consumption. To this end, 18 smaller databases, comprising 8 stand-area databases (Table 3, Column 1) and 10 stand-value databases (Table 4, Column 1) were created from the single database for efficient data management and analyses. Stand value was used as proxy for household income (Van Zyl et al., 2003 and Husselmann and Van Zyl, 2006). A stepwise multiple variable regression analysis was carried out on the 18 smaller databases with Ln (AADD) as the dependent variable. A stepwise regression for more than one independent variable will always select the variable that correlates best with the dependent variable first and then, the next best correlated variable, and so forth. Thus, multiple variable models were obtained for each of the 18 databases. These models represent the combined effect on water consumption of each of the different parameters. In most models, stand area and stand value influenced water consumption the most, with geographical location in third place (Table 3 and Table 4). Based on this result, a further refinement of the first two parameters was carried out, by distinguishing between inland and coastal geographical areas. Inland and coastal areas represent a broader classification of the different water regions shown in Fig. 1.

Following the outcome of the regression analyses, it was considered constructive to generate one or more curves that could aid design engineers in their estimation of water demand.

Results and discussion

Suburb-level analysis

The effect of average suburb stand area on average suburb domestic water consumption is shown in Fig. 3 with the *Red Book* upper and lower guideline curves.

Of the 1 188 suburbs plotted, 461 (39%) lie below the lower guideline curve and 100 (8%) lie above the upper guideline curve. Thus, the current *Red Book* guideline only accounts for 53% of the suburbs. This gives the strongest indication yet of the need to update the *Red Book* curves.

Of the 92% of suburbs that lie below the upper guideline curve, 57% have stand areas smaller than 800 m². The 100

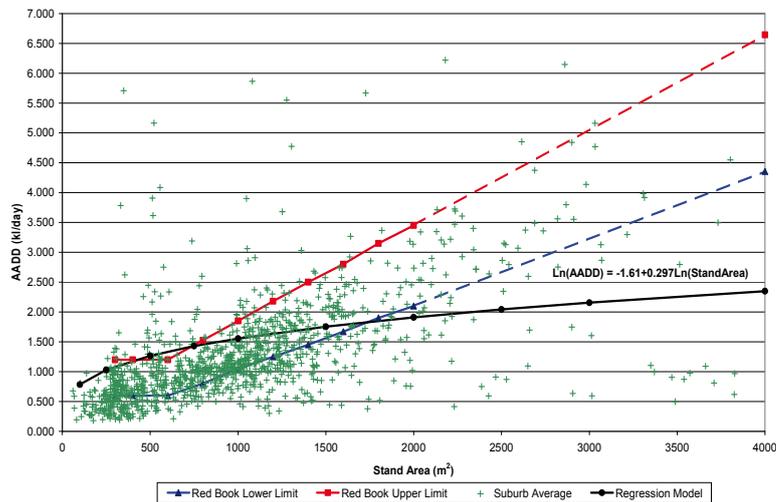


Figure 3
Preliminary assessment of the *Red Book* guideline using suburb AADD vs. average suburb stand area

suburbs that had AADD values above the upper guideline curve were mostly a combination of middle-income, small single-residential developments (such as Cason in Boksburg, and Brixton and Yeoville in Johannesburg), low income or township developments (such as Orange Grove in Johannesburg and Ncala in Germiston), and 23 suburbs were high-income, high-security developments (such as Kyalami Estates, Dainfern, Douglasdale and Sunning Hill in the northern suburbs of Johannesburg).

A single regression analysis was then carried out to determine the equation representing the average AADD as a function of average stand area:

$$\begin{aligned} \text{Ln}(AADD) = & -1.610 + 0.297\text{Ln}(\text{StdArea}) \pm & (1) \\ & 0.860 \sqrt{9.16 \times 10^{-7} + \frac{(\text{LnStdArea} - 6.4124)^2}{666977}} \end{aligned}$$

where:

StdArea = stand area in m²

AADD is in kl/d

The first part of the equation describes the single variable regression curve, and the second part, the 97.5% confidence limits. The regression model generated an adjusted R² value (square of the Pearson product moment correlation coefficient) of 0.218, which implies that 21.8% of the variability in the data can be explained by this equation. This value is acceptable for this analysis considering that an adjusted R² value of more than 20% is considered good when predicting human behaviour. Numerous factors contribute to the variability in the data and given the large database, even greater variation could have resulted. A logarithmic regression model was used to generate the model as this gave the best fit to the data. The *Red Book* guideline curves were extrapolated up to stand areas of 4 000 m² as the database had more than 30 000 residential stands with stand areas greater than 2 000 m².

As can be seen in Fig. 3, the regression curve falls within the *Red Book* guideline curves for stand areas up to approximately 1 700 m² with the regression line lying very close to the upper limit for stand areas smaller than 750 m². In fact, the model predicts slightly higher AADD values for stand areas between 500 m² and 750 m². For stand areas larger than 1 700 m² the

Category description	No of data points, N	No. of parameters analysed	The three most influential parameters by order of best correlation with Ln(AADD)	Cumulative adjusted R²
≥ 20 m ² and < 500 m ²	433 104	14	Stand area	0.053
			Geographic location	0.066
			Ln (stand value)	0.079
≥ 500 m ² and < 750 m ²	112 443	14	Ln(stand value)	0.037
			Geographic location	0.062
			Ave household income	0.071
≥ 750 m ² and < 1000 m ²	136 019	14	Ln(stand value)	0.067
			Geographic location	0.086
			Ave min temperature	0.101
≥ 1000 m ² and < 1500 m ²	184 497	15	Ln(stand value)	0.092
			Ave household income	0.117
			Ave house size	0.139
≥ 1500 m ² and < 2000 m ²	49 952	11	Ln(stand value)	0.105
			Ave household income	0.132
			Ave house size	0.155
≥ 2000 m ² and < 2500 m ²	17 506	11	Ln(stand value)	0.092
			Geographic location	0.125
			Ave household income	0.136
≥ 2500 m ² and < 3000 m ²	6 884	8	Ln(stand value)	0.161
			Geographic location	0.187
			Stand area	0.194
≥ 3000 m ² and < 4000 m ²	6 168	8	Ln (stand value)	0.164
			Geographic location	0.196
			Stand value	0.207

Category description	No of data points, N	No. of parameters analysed	The three most influential parameters by order of best correlation with Ln(AADD)	Cumulative adjusted R²
≥ R20 000 and < R50 000	211 753	16	Ln(stand value)	0.029
			Ln(stand area)	0.041
			Geographic location	0.050
≥ R50 000 and < R100 000	200 322	14	Ln(stand area)	0.033
			Geographic location	0.049
			Mean annual evaporation	0.055
≥ R100 000 and < R250 000	298 703	15	Ln(Stand area)	0.065
			Geographic location	0.089
			Ave household income	0.099
≥ R250 000 and < R500 000	227 595	16	Ln(stand area)	0.107
			Ave household income	0.134
			Ave house size	0.152
≥ R500 000 and < R750 000	49 807	13	Ln(stand area)	0.102
			Geographic location	0.148
			Ave household income	0.165
≥ R750 000 and < R1 000 000	14 351	13	Geographic location	0.087
			Ln(stand area)	0.133
			Mean annual precipitation	0.139
≥ R1 000 000 and < R2 000 000	8 704	12	Geographic location	0.079
			Ln(stand area)	0.118
			Ave min temperature	0.141
≥ R2 000 000 and < R3 000 000	1 286	9	Ave min temperature	0.261
			% water connection	0.356
			Ave max temperature	0.388
≥ R3 000 000 and < R4 000 000	428	3	Ave min temperature	0.195
			Geographic location	0.291
			Stand area	0.298
≥ R4 000 000	1 275	7	Mean annual evaporation	0.274
			% unemployed	0.366
			Ave min temperature	0.39

TABLE 5 Stepwise regression equations for the most influential parameter in relation to Ln (AADD)		
Category description	Most influential parameter in relation to Ln(AADD)	Regression equation for the most influential parameter
Stand-area categories		
≥ 20 m ² and < 500 m ²	Stand area	$LnAADD = 0.001StdArea - 0.195$
≥ 500 m ² and < 750 m ²	Ln(stand value)	$LnAADD = 0.099 \ln(StdValue) - 0.920$
≥ 750 m ² and < 1 000 m ²	Ln(stand value)	$LnAADD = 0.181 \ln(StdValue) - 1.824$
≥ 1000 m ² and < 1500 m ²	Ln(stand value)	$LnAADD = 0.251 \ln(StdValue) - 2.595$
≥ 1500 m ² and < 2 000 m ²	Ln(stand value)	$LnAADD = 0.295 \ln(StdValue) - 3.050$
≥ 2000 m ² and < 2 500 m ²	Ln(stand value)	$LnAADD = 0.279 \ln(StdValue) - 2.788$
≥ 2500 m ² and < 3 000 m ²	Ln(stand value)	$LnAADD = 0.348 \ln(StdValue) - 3.670$
≥ 3000 m ² and < 4 000 m ²	Ln(stand value)	$LnAADD = 0.331 \ln(StdValue) - 3.453$
Stand value categories		
≥ R20 000 and < R50 000	Ln(stand value)	$LnAADD = 0.266 \ln(StdValue) - 2.736$
≥ R50 000 and < R100 000	Ln(stand area)	$LnAADD = 0.133 \ln(StdArea) - 0.661$
≥ R100 000 and < R250 00	Ln(stand area)	$LnAADD = 0.186 \ln(StdArea) - 0.910$
≥ R250 000 and < R500 000	Ln(stand area)	$LnAADD = 0.286 \ln(StdArea) - 1.455$
≥ R500 000 and < R750 000	Ln(stand area)	$LnAADD = 0.266 \ln(StdArea) - 1.159$
≥ R750 000 and < R1 000 000	Geographic location	$LnAADD = -0.427(GL) + 1.398$
≥ R1 000 000 and < R2 000 000	Geographic location	$LnAADD = -0.446(GL) + 1.534$
≥ R2 000 000 and < R3 000 000	Ave min temp	$LnAADD = -0.172(MinTemp) + 2.890$
≥ R3 000 000 and < R4 000 000	Ave min temp specific	$LnAADD = -0.123(MinTemp) + 2.986$
≥ R4 000 000	MAE	$LnAADD = 0.0027(MAE) - 3.336$

model predicts much lower AADD values. This result corresponds with the comment made while comparing the suburb averages with the *Red Book* guideline, namely that the majority of the stands that exhibited consumption higher than what the guideline predicted were smaller stands with areas of less than 800 m². The 97.5% confidence limits lie very close to the regression line, so much so that the difference can hardly be distinguished. The 97.5% confidence limits indicate the 97.5% probability that the mean AADD of all the suburbs will lie within the upper and lower confidence limits.

Regression analyses

To analyse the various parameters influencing domestic water consumption, a stepwise multiple regression analysis was performed on each of the 18 smaller databases. Detailed regression analyses results for stand area and stand value databases are shown in Tables 3, 4 and 5. For each database, the number of data points and parameters analysed, the three independent parameters influencing water consumption by order of best correlation with Ln (AADD), as well as the adjusted R² value for each regression step are shown (Tables 3 and 4). The regression equations for the most influential parameters are shown in Table 5.

Stand area, stand value (as proxy for household income) and geographic location emerged as the top three parameters influencing domestic water consumption - for stand-area databases, stand value influences water consumption the most; for stand-value databases, stand area influences water consumption the most; for 6 out of the 8 stand-area categories, and 4 out of the 10 stand-value categories, geographic location stands out as the second most influential parameter. The small magnitudes of the adjusted R² values are indicative of the large sample sizes and the inherent variability and uncertainty in the data. The regression analyses are, however, very useful in identifying the most important parameters influencing water consumption.

Following the multiple variable analysis, a single variable analysis was conducted for each of the 18 databases with each of the parameters that emerged (i.e. stand area, stand value and geographical location) as influencing water consumption the most. Geographic location was considered in the single variable regression analysis by categorising for inland and coastal regions. Superimposed on the *Red Book* guideline, Fig. 4 shows the single variable regression models with stand area as independent variable for various income (using stand value as a proxy for household income) categories for both inland and coastal geographic locations. The regression models predict that for the same stand area, higher income levels will consume

Figure 4
Single variable regression models for AADD vs. stand area for different income levels situated in coastal and inland geographical areas

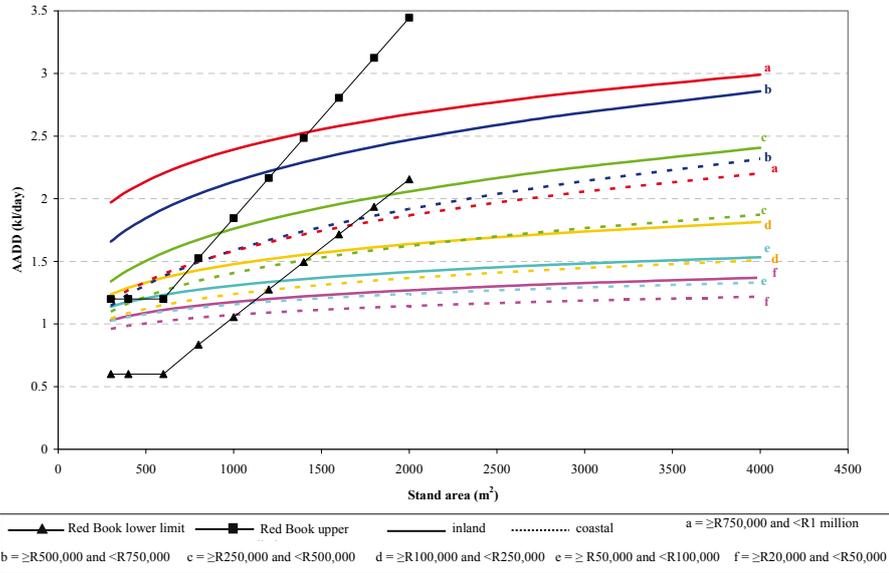
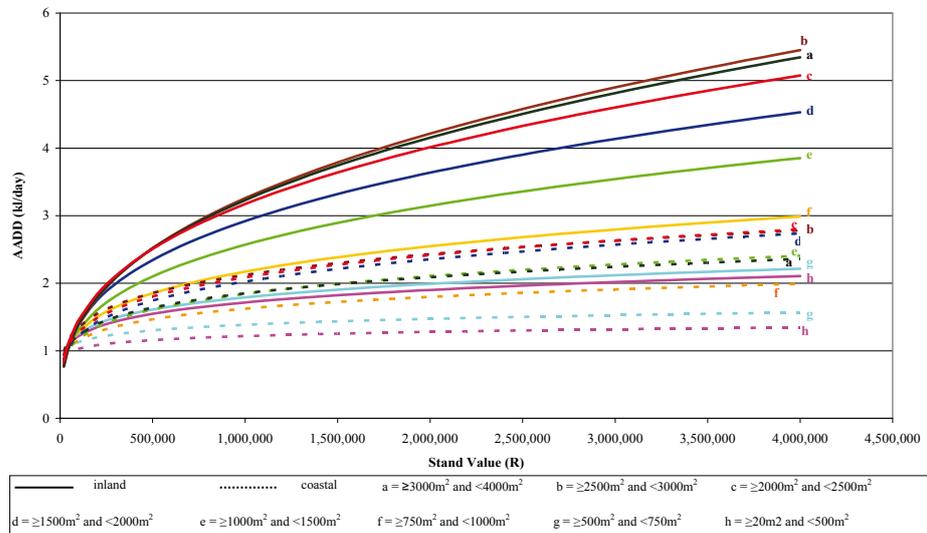


Figure 5
Single variable regression models for AADD vs. stand value for different stand areas situated in coastal and inland geographical areas



more water on average than lower income levels. The regression models also predict that coastal stands with the same stand area and stand value will consistently consume less water than their inland counterparts.

Figure 5 shows the single variable regression models with stand value as the independent variable for various stand-area categories for both inland and coastal geographic locations. These regression models predict that consumers with larger stand areas will generally consume more water than consumers with smaller stand areas for the same stand value (or income) categories. The regression models confirm the previous result that coastal stands with the same stand area and stand value will consistently consume significantly less water than inland stands.

From these results it is apparent that household income in addition to stand area, have a significant impact on domestic water consumption and that it would be ideal to jointly take these two parameters into account when estimating domestic demand. The single variable regression models confirm the outcome of the multiple variable regression models, i.e. that both stand area and income are positively correlated to domestic water consumption. Geographic location is also an important factor to consider with domestic water-demand estimation, i.e. that coastal stands

consistently consume less water than inland stands for the same stand area and value.

Development of a water-demand estimation guideline

The results of this study confirm that stand area is one of the parameters that influence domestic water consumption the most. However, in this study, stand value (as a proxy for household income) and geographic location were also found to significantly influence domestic water consumption. It was therefore considered constructive to propose a new AADD guideline, given the results of this study.

The 1st possibility investigated was a two-variable model with stand area and stand value specified as independent variables. However, information on stand value is generally difficult to obtain and a designer would not easily have access to such data for design purposes. Also, stand values continuously change and would be significantly different over time from the values used in this study for the exact same suburb. Also, in many instances, stand values are based on perception rather than true values and hence, would distort demand estimates.

The 2nd possibility investigated was to develop three income models (with stand value as a proxy for household income). In

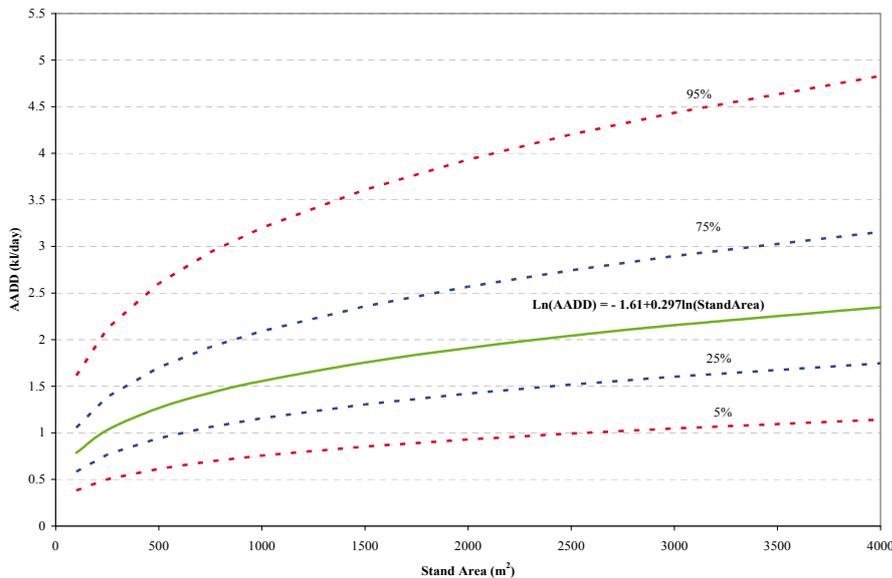


Figure 6
Proposed guideline for
domestic water-demand
estimation

other words, categorise the single database into three income categories namely low-, middle- and high-income, and then generate a regression model for each income group with stand area as the independent variable (Fig. 4 is an example of this possibility). It was felt, however, that the categorisation of stand values into income levels could also be considered too subjective when used for the development of an AADD guideline. A guideline that allows the designer to interpret income levels, knowing that higher income users consume more water, may be a better option.

The 3rd possibility investigated, which is the preferred of the three, is to develop a single guideline curve for AADD vs. stand area using all the AADD values in the single database (similar to Fig. 3). The guideline curve will have envelopes around it based on 5%, 25%, 75% and 95% confidence limits. The confidence limits indicate the boundaries wherein a given percentage of AADD would lie. Figure 6 presents the proposed regression model. The equations for the confidence limits are given below.

95% (+) and 5% (-) confidence limits:

$$\ln(AADD) = -1.610 + 0.297 \ln(StdArea) \pm (1.6449)(0.43865) \sqrt{1 + \frac{(\ln StdArea - 6.4124)^2}{666977}} \quad (2)$$

75% (+) and 25% (-) confidence limits:

$$\ln(AADD) = -1.610 + 0.297 \ln(StdArea) \pm (0.6745)(0.43865) \sqrt{1 + \frac{(\ln StdArea - 6.4124)^2}{666977}} \quad (3)$$

Conclusions

This study investigated the combined effect of various socio-economic and climatic parameters on municipal water consumption with the objective of determining the dominant influencing parameters. The result is a new guideline for AADD estimation. To this end, an initial database comprising over 2.5×10^6 metered water consumption records, extracted from 48 municipal treasury databases, which are located within 5 out of the 7 South African water regions, was analysed. Each of the 48 municipal treasury databases spanned a period of at least 12 months

(the vast majority for at least 24 months). The final amalgamated database, after rigorous cleaning and filtering, comprised 1 091 685 consumption records. Single variable and stepwise multiple variable regression analyses were utilised.

The main conclusions of the study are:

- Water consumption is by nature highly variable and thus it is not possible to develop a single regression model that can predict individual water demands accurately. However, regression models can greatly assist the design engineer by identifying the most important parameters that affect water demand
- Of all the parameters investigated in this study, it was clear that stand area, stand value and geographical location (inland or coastal) are the dominant parameters that affect domestic water consumption
- Inland stands use more water than coastal stands and should be treated separately in any design guideline
- Stand area and stand value are positively correlated with domestic water consumption
- It is not possible to have a consistent basis for predicting stand values due to constant fluctuations in the value of property and municipal valuations that are often outdated. This leaves stand area as the best descriptor of AADD
- The current *Red Book* AADD guideline is not adequate and needs to be revised. This is made clear by the fact that the *Red Book* AADD guideline envelope accounts for only 53% of the suburbs studied
- A guideline curve giving various confidence limits for water demand based on stand area is presented to assist design engineers with selecting appropriate AADD values for new developments.

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