Rainwater Storage Tank Sizing using Raincycle Model:
Case Studies of a Residential, Office and Hospital Blocks

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Abstract - Surface and groundwater extraction has been used to satisfy the rising demand of water in developing countries which is unsustainable. Rainwater harvesting systems is progressively becoming a sustainable water management measure. However, an assessment of their feasibility in small and large buildings is required for considerable investment cost and payback periods. In this study, Raincycle model was used to describe and analyse the different criteria for sizing a rainwater harvesting system. Raincycle model was used to size the storage tank, cost savings of tanks and choose optimum size. Behavioural (or simulation) analysis was included to compute the changes in storage content of a finite reservoir. To apply this method, three case studies were used: a residential apartment, office and hospital blocks. Result reveals a tank size of 4 m³, 10 m³ and 12 m³ could meet the maximum percentage of demand and predicted to save $259, $2,564, and $51,072 over 50 years for case studies 1, 2 and 3 respectively. A payback period of 21 years, 8 years and 1 year was recorded for the three case studies, respectively. This study also shows that the implementation of a rainwater harvesting system in a residential dwelling is not a feasible investment in terms of economic savings as it presents payback time that is not feasible. The office and hospital blocks present a payback time that is considered feasible as payback time is reached during the building’s lifetime.

Keywords: Sustainable water use, Rainwater harvesting, Tank sizing, Economical assessment, Payback time.

I. INTRODUCTION

The world is suffering from severe water scarcity issues as water supply networks are under stresses in many cities [1-2]. Sustainable water use and guaranteed quantity and quality for future generations can be achieved through reduction in demand whilst increasing the efficiency of water use. Rainwater harvesting (RWH) is a technology that should be a precaution for upcoming scarcity situations through reduction in potable water consumption [3].

Rainwater harvesting system (RWHS) have been used in many countries of the world to strengthen urban stormwater management and simultaneously reduce municipal water supply stress and restores natural water cycles in urban areas [4-8]. RWH has been explored globally to reduce potable water consumption. The use of RWHS in a household as led to a savings of up to 40% of potable water use [9]. A research conducted in United Kingdom shows that 36-46% of WC demand can be met, resulting in a payback period of 23 and 7 years respectively [10]. In Sweden, a study in a residential area revealed a saving of more than 60% can be achieved when rainwater is used for WC flushing. In Brazil, the potential for potable water savings was evaluated through use of rainwater for washing vehicles in petrol stations [11].

The design of RWHS has been evaluated using numerous modelling tools and methods including the design of storm approach [12]. The most common approach is long-term continuous simulations which can be used to evaluate the effectiveness of both stormwater management and monetary benefits of RWHS [13-16] and to determine the optimal tank size [17-18]. However, these modelling tools are limited, as a few could be used to assess the water saving and stormwater potential as well as the economic feasibility of RWHS.

A rainwater tank can be considered as a storage reservoir that receives stochastic inflows (effective runoff) over time and is sized to satisfy system demands [19]. The designer controls the tank size, hence some techniques to determine the size that will provide the optimum level of service is required [20]. In another study, the sizing of storage reservoirs was reviewed to identify two general categories of sizing techniques: Moran related methods and critical period methods [21].

Behavioural analysis is a technique more suited to the study; it has several advantages compared to the Moran related and mass curve methods. Numerous researchers investigating RWHS performance have used the technique [22-24]. Several monitoring studies have confirmed the validity of behavioural models [25]. A domestic RWHS installed in a UK property, which was used for WC flushing, was modelled [26]. A behavioural model of the system was created and compared with data collected from the RWHS over 12-months. The actual system performance was accorded with predicted
behaviour and this provides empirical validation of a behavioural approach in a domestic UK context.

This study aims to determine the best criteria to size rainwater storage tank when “behavioural approach” is used. Three case studies were used: a residential apartment, an office block and a hospital block. The results for each case study, such as the tank size and parameters related to the systems performance and the economic feasibility are presented and analysed.

II. CASE STUDIES

2.1 Residential apartment

Case Study 1 is a semi-detached twin bungalow for 12 occupants in Ibadan North Local Government of Oyo State, Nigeria. Table 1 presents the summary of parameters input into the model.

Catchment surface area = 196 m²
Water demand: Above average = 258 m³/yr
Average = 214 m³/yr
Below average = 171 m³/yr

Table 1: Data input - summary of parameter values

<table>
<thead>
<tr>
<th>Module</th>
<th>Item</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rainfall profiles</td>
<td>Above average rainfall</td>
<td>1.639 mm/yr</td>
</tr>
<tr>
<td></td>
<td>Average rainfall</td>
<td>1.311 mm/yr</td>
</tr>
<tr>
<td></td>
<td>Below average rainfall</td>
<td>983 mm/yr</td>
</tr>
<tr>
<td>Catchment surface</td>
<td>High runoff coefficient</td>
<td>0.90</td>
</tr>
<tr>
<td></td>
<td>Expected runoff coefficient</td>
<td>0.85</td>
</tr>
<tr>
<td></td>
<td>Low runoff coefficient</td>
<td>0.75</td>
</tr>
<tr>
<td>Rainwater filter</td>
<td>High filter coefficient</td>
<td>0.92</td>
</tr>
<tr>
<td></td>
<td>Expected filter coefficient</td>
<td>0.90</td>
</tr>
<tr>
<td></td>
<td>Low filter coefficient</td>
<td>0.85</td>
</tr>
</tbody>
</table>

2.2 Office block

Case study 2 is Dr Egbogah’s building in the Department of Civil Engineering, University of Ibadan, Nigeria.

Catchment surface area = 566 m²
Water demand: Above average = 458 m³/yr
Average = 441 m³/yr
Below average = 396 m³/yr

2.3 Hospital ward

Case study 3 is Otunba-tunwase Children Outpatient Ward, University College Hospital, Ibadan, Nigeria

Catchment surface area = 8,132 m²
Water demand: Above average = 4,704 m³/yr
Average = 3,920 m³/yr

III. METHODS

For the three case studies presented, Raincycle model was used to size the storage tank, cost savings of tanks and choose optimum size. Behavioural (or simulation) analysis was included [21] to compute the changes in storage content of a finite reservoir (one that can overflow and empty) using the water balance equation shown in equation 1[27]:

\[ V_t = V_{t-1} + Q_t - D_t - \Delta E_t - L_t \]                  (1)

Subject to \( 0 \leq V_t \leq S \)

Where:
\( V_t \) = storage content at time t (m³).
\( V_{t-1} \) = storage content at time t-1 (m³).
\( Q_t \) = flow into the reservoir during time interval (m³).
\( D_t \) = controlled release during time interval t (m³).
\( \Delta E_t \) = net evaporation loss from the reservoir during time interval t (m³).
\( L_t \) = other losses during time interval t, e.g. seepage (m³).
\( S \) = active reservoir capacity (m³).

The most used storage device with regards to contemporary RWHS is the underground tank [28]. These are watertight and airtight, so the net evaporation loss term \( \Delta E_t \), and the other losses term \( L_t \) can be ignored [29]. Equation 1 then becomes:

\[ V_t = V_{t-1} + Q_t - D_t \]                  (2)

Subject to \( 0 \leq V_t \leq S \).

At the end of a prescribed time interval, the water in storage is equal to the volume of water remaining in storage from the previous interval, plus any inflow and less any demand during the time period, provided that the computed volume in store does not exceed slope capacity. Behavioural models thus simulate the operation of a reservoir with respect to time by routing simulated mass flows through an algorithm describing reservoir operation [19].

The generalised YAS algorithm was incorporated into the RainCycle model adopted in this study, with the storage operating parameter \( \theta \) set to zero (YAS) as the default mode of operation. It is acknowledged that a more conservative prediction of system performance would have been achieved with the use of YAS setting than use of YBS. However, research suggested that as long as certain constraints regarding the selected time-step are employed, then YAS models are capable of modelling system performance within ±10% of that predicted by a more accurate hourly time-step
model and this was considered to be an acceptable margin of error [30].

3.1 Rainfall data used in the adopted RainCycle model

The RainCycle model operates on a daily time-step, and so it requires historic rainfall data with the same temporal scale. It was suggested that a minimum of 10 years’ worth of precipitation records should be used [31]. Since the study is primarily concerned with analysing domestic RWHS in Ibadan, the selection criteria with regards to the selected station was that the location should be representative of typical weather patterns in the City (i.e. the average annual rainfall depth should not significantly deviate from what would be expected for the City as a whole). A continuous 30-year daily rainfall record covering the years 1980-2009 was obtained from Department of Meteorological Station (DMS) Samonda and the International Institute of Tropical Agriculture (IITA), Ibadan. Figure 1 shows annual rainfall depths contained within the data set. Note the extreme marked on the graph, which correspond to the 1980 floods, which affected much of Ibadan.

![Rainfall](image)

**Figure 1:** Ibadan City historic annual rainfall depths 1980-2009 (source:[32])

3.2 Predicting non-potable domestic demand

Per capita consumption varies with household size, type of property, time of the year and ages of household residents [33]. Increases in household demand are primarily driven by population growth, levels of affluence and household occupancy [34]. In modern developments, the UK Code for Sustainable Homes Standard [35] may act as a significant driver for reduction in domestic water use. To achieve the lowest level of compliance, a minimum per capita consumption of 120 litres per day is required for internal water use and this may come to represent the minimum standard for new housing stock.

About 55% of total household demand could be met through domestic RWHS, if used for non-potable applications, such as WC flushing, laundry washing and garden irrigation. Demand forecasting based on rules of thumb or naïve extrapolation is now recognized as inappropriate since estimates obtained this way deviate from reality [36]. A micro-component approach to water demand forecasting is often recommended [34]: that is the study of individual water usage within a household, such as personal washing and WC flushing [33]. Numerous studies have used this approach in predicting future demand [34, 37-38]. However, regarding future water demand, no definite conclusion can be drawn [39], only more or less reasoned and transparent investigations [40]. Considering climate change, assumption that washing machine and WC use are insensitive to long-term variations can be made [39-40].

3.3 Water closet demand

There is no cause to believe that WC usage frequency will increase or decrease significantly, and so existing data based on past monitoring studies was used as an acceptable indicator of future behaviour. The mean of the values equal to 4.59 flushes per person per day. However, it is impossible to flush a toilet 4.59 times, so a per capita usage of 4 times/day was assumed for weekday (Monday-Friday), whilst a per capita usage of 6 times/day was assumed for weekends (Saturday and Sunday). This assumes higher weekend usage, which is reasonable, and gives mean rate of 4.57/person/day, which is close to the actual average of 4.59. An essentially linear relationship was found between household occupancy and frequency of WC flushes [41]. Therefore, an acceptable approach for calculating household usage is to multiply the household occupancy rate by the capita usage frequency. In the current regulations, a maximum flush volume of 6 litres is permitted for single-flush WCs [42]. Dual-flush toilets ranging from 6/4, 6/3, as well as lower volume single flushes, such as 4.5 litres, is available.

To demonstrate the technical lower limit for flush volumes, the last three items were included. But there is no sign of widespread implementation of these methods in the short to medium term. The flush volumes ranging from 6-litre single to 4/2-litre dual were chosen. An assumption of 6/3 dual flush variety was decided for all new houses. There is also the issue of how many uses will involve a full flush and how much only a part flush. It was reported that an assumption of 1:3 or 1:4 is made for the ratio of full to part flush, whilst monitoring trials have shown the actual flush ratio to be in the range of 1:0 (i.e. only full flush used) and 1:2 (1 full to 2 part flushes) [43]. A full flush ratio of 1:2 was been adopted in the RainCycle model.
3.4 Washing machine demand

A range of washing machine (WM) use frequencies is shown in Table 2. Anticipated future per capita use frequencies will probably not differ much from those occurring at present. The mean of domestic washing machine frequencies is 0.21 uses per person per day (~ once every 5 days). This latter figure was used as a standard value for domestic simulations. This is a general relationship between frequency of WM usage and household occupancy [41]. Thus, to determine household usage the per capita frequency can simply be multiplied by the household occupancy rate.

<table>
<thead>
<tr>
<th>Uses/person/day</th>
<th>References</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.16</td>
<td>[41]</td>
</tr>
<tr>
<td>0.18</td>
<td>[44]</td>
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<tr>
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<td>[34]</td>
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<tr>
<td>0.34</td>
<td>[35]</td>
</tr>
<tr>
<td>0.21</td>
<td>Mean</td>
</tr>
</tbody>
</table>

IV. RESULTS AND DISCUSSIONS

The Raincycle advanced analysis process is used to determine range of suitable tank sizes, cost savings of tanks and choose optimum size (Figures 2 and 3).

4.1 Residential apartment

Optimising tank size results reveals that a tank size of 4 m$^3$ could meet the maximum percentage of demand. Optimise saving analysis showed that there were four tank sizes with a potential long-term profit. The best was 4 m$^3$ tank which was predicted to save $259 over 50 years and had a payback period of 21 years.

The 4 m$^3$ gave acceptable results and so the data for this tank was input into the storage tank module (tank size) and Whole Life Cost (WLC) details module (capital costs) and then the result in the analysis system module were examined. Figures 4 and 5 show the cost comparison graphs for both the long-term and average per-year analyses for this system.

Figure 2: Determining range of suitable tank sizes
Figure 3: Determining cost savings of tanks and choosing optimum size

Cumulative Whole Life Cost Comparison Graph (discounted at 3.5%/yr)

Figure 4: Cumulative long-term analysis cost comparison (residential)

Average Yearly Cost Comparison Graph (discounted - 3.5%)

Figure 5: Average yearly cost comparison (residential)
4.2 Office block

Optimise tank size results reveal that a tank size of 10 m$^3$ could meet the maximum percentage of demand. Optimise saving analysis showed that there were six tank sizes with a potential long-term profit. The best was the 10 m$^3$ tank, which was predicted to save $2,564 over 50 years and had a payback period of 8 years.

The 10 m$^3$ gave acceptable results and so the data for this tank was input into the Storage Tank module (tank size) and WLC Details module and then the result in the Analysis System module examined. Figures 6 and 7 show the cost comparison graphs for both the long-term and average per-year analyses for this system.

The 12 m$^3$ gave acceptable results and so the data for this tank was input into the Storage Tank module (tank size) and WLC Details module and then the result in the Analysis System module examined. Figures 8 and 9 show the cost comparison graphs for both the long-term and average per-year analyses for this system.

4.3 Hospital blocks

Optimise tank size results reveal that a tank size of 12 m$^3$ could meet the maximum percentage of demand. Optimise saving analysis showed that there were seven tank sizes with a potential long-term profit. The best was the 12 m$^3$ tank, which was predicted to save $51,072 over 50 years and had a payback period of 1 year.

The 12 m$^3$ gave acceptable results and so the data for this tank was input into the Storage Tank module (tank size) and WLC Details module and then the result in the Analysis System module examined. Figures 8 and 9 show the cost comparison graphs for both the long-term and average per-year analyses for this system.

Figure 6: Cumulative long-term analysis cost comparison (office)

Figure 7: Average yearly cost comparison (office)

Figure 8: Cumulative long-term analysis cost comparison (hospital)

Figure 9: Average yearly cost comparison (hospital)

V. CONCLUSION

This study focuses on the sizing of rainwater storage tanks using Raincycle model. The aim of this study is to determine the criteria for defining a storage volume and cost savings of tanks when a simulation is made to choose optimum size.

Three case studies were analysed to apply the defined criteria and compare the results of storage volumes and related performance and economic assessments.

Analysis of case study 1 reveals the best tank size was 4 m$^3$ which was predicted to save $259 over 50 years and had a payback period of 21 years.
Analysis of case study 2 reveals the best tank size was 10 m³ which was predicted to save $2,564 over 50 years and had a payback period of 8 years.

Analysis of case study 3 reveals the best tank size was 12 m³ predicted to save $51,072 over 50 years and had a payback period of 1 year.

This study also shows that the implementation of a rainwater harvesting system in a residential dwelling is not a feasible investment in terms of economic savings as it presents payback time that is not feasible. The office and hospital building present a payback time that is considered feasible as payback time is reached during the building’s lifetime.

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