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1 **Factors affecting waste generation and segregation behaviour. An analysis using data from**  
2 **the educated communities in the Western and the North-Western provinces of Sri Lanka**

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11

12

**Abstract**

13 The study intends to identify the factors influencing the waste generation and segregation  
14 behaviour of households. An online survey was conducted with more than 400 residents in the  
15 North-western and Western provinces of Sri Lanka who have obtained a tertiary educational  
16 qualification. It was identified that family income level, local governmental authorities'  
17 involvement in waste collection, age of the household head, and house type significantly affect the  
18 waste segregation behaviour of households of educated communities. Further, age of household  
19 head, family size, attitude, land area, household ownership, family income and involvement of  
20 local governmental authorities in the waste collection were observed to affect the household solid  
21 waste generation. The results could support developing influential policies in the local  
22 governmental authority levels to further strengthen and improve the solid waste management  
23 practices in Sri Lanka. It could also shed light on the formulation of appropriate strategies focussed  
24 on different communities to encourage the practice of waste segregation at source in Sri Lanka.

25

26 **Keywords:** Waste management; waste segregation at source; waste generation; households;  
27 recycling

## 28 **1. Introduction**

29

30 Solid Waste Management (SWM) has become a universal issue owing to the continuous adverse  
31 effects on human health and the environment. The UN-member countries identified SWM as an  
32 essential requirement for achieving Sustainable Development Goals (SDGs) by including target 5  
33 of SDG Goal-12: “substantially reduce waste generation through prevention, reduction, recycling  
34 and reuse (3Rs)”. Per capita waste generation increases with the progress in a country's population  
35 and the movement of populaces from rural areas to urban areas. Further, due to technological  
36 advancement and the expansion of the middle class in developing countries, an emerging trend of  
37 electronic waste could be identified (Nnaji, 2015). There is a significant change in the composition  
38 of solid waste generation due to the global pandemic of COVID 19 as well. For instance, there is  
39 a substantial increase in medical waste and plastics (Yousefi et al., 2021). The plastic waste  
40 generated during the COVID 19 outbreak was estimated at 1.6 million tonnes per day globally,  
41 and approximately 3.4 billion single used face masks were discarded per day (Yousefi et al., 2021).

42 The growing trend in solid waste generation is a critical problem in developing country settings.  
43 Global statistics state that 2.01 billion tonnes of municipal solid waste had been produced in 2016,  
44 and it is projected to be 3.4 billion tonnes by 2050 (Kaza et al., 2018). With rising solid waste  
45 generation, the local administration authorities confer a high expenditure for urban waste  
46 management activities. Therefore, a proper solid waste management system is vital to overcome  
47 the adverse effects due to increased-solid waste. Waste segregation at the source is identified as a  
48 critical process (Chen & Lee, 2020) of proper waste management. In addition, reducing solid waste  
49 generation is a prerequisite in overcoming the solid waste problem. Previously, researchers found  
50 that knowledge, attitude, social norms, situational and socio-economic factors influence the waste

51 segregation behaviour of the households (Alhassan et al., 2018; Ayob et al., 2017; Karim Ghani et  
52 al., 2013; Dai et al., 2017; Owusu et al., 2013). Another strand of literature studied the determinants  
53 of waste generation behaviour of households (Wang & Qiu, 2013; Sarica et al., 2020; Cusano,  
54 2016).

55 The educational level of the residents has been identified as a significant factor influencing waste  
56 management practices in many works of literature. Higher education levels have made individuals  
57 aware of the external effects of their day-to-day behaviours and are involved with greater social  
58 welfare (Meyer, 2015). With the growth in the educational level of the residents, the general  
59 knowledge on waste management practices was found to be augmented, enhancing the  
60 responsibility towards the environment (Babaei, et al., 2015). Moreover, willingness to recycle,  
61 recycling efficiency, and recycling behaviour are revealed to be significantly influenced by the  
62 level of education (Guerin, Crete, & Mercier, 2001; Saphores, Nixon, Ogunseitan, & Shapiro,  
63 2006; Owens, Dickerson, & Macintosh, 2000). Therefore, the study focuses on examining the  
64 educated-communities that have received tertiary education or an educational level above  
65 schooling, particularly university degrees or vocational training courses, etc., and explores the  
66 factors that influence the waste generation and segregation behaviour in their households.

67 In Sri Lanka, approximately 7000 metric tonnes of solid waste is generated per day. More than  
68 50% of the waste is generated in the Western province. The waste generation per person amounts  
69 to 0.4-1.0kg of solid waste per day (Environmental Foundation Ltd, 2017). Inevitably, Sri Lanka  
70 also faces major municipal solid waste management problems in urban areas similar to many other  
71 developing countries. Despite having a multi-level governing structure such as national, provincial  
72 and local government to manage solid waste, many solid waste-related problems are still not  
73 adequately addressed. For instance, Dharmasiri (2020) has identified some issues prevailing in the

74 current waste management system of Sri Lanka. Our focal area of study, the waste segregation  
75 issue, has been highlighted, as a proper system for waste segregation is not prevalent in Sri Lanka.  
76 Furthermore, it was stated that Sri Lanka's waste management system has issues with outdated  
77 waste management practices, inefficient handling, collection, transportation and inefficient  
78 disposal of hospital waste and hazardous waste.

79 With this backdrop, the study intends to identify the factors that influence the waste segregation  
80 decision of the households in Sri Lanka. Educated communities in the Western and North-Western  
81 provinces in Sri Lanka are selected for the study as relatively higher waste generation rates are  
82 recorded in these provinces. This study explores those respondents' waste generation and waste  
83 segregation behaviour by using primary data collected from more than 400 households.

84 The rest of the paper is organised as follows. Section 2 reviews the literature on solid waste  
85 generation and segregation behaviour. Section 3 will describe the methodology used in the paper.  
86 Section 4 and 5 will present the results and discussion on the analysis. Finally, section 6 will  
87 conclude with explaining policy implications.

88

## 89 **2. Literature Review**

90

### 91 **2.1 Solid waste management practices**

92 Solid waste management is a widely acknowledged subject among nations as the solid waste  
93 amount continues to proliferate (Kaza et al., 2018). Numerous research efforts have examined the  
94 extent and nature of SWM practices in different countries and situational settings (Fernando, 2019;  
95 Dharmasiri, 2020; Kumara & Pallegedara, 2020; Yousefi et al., 2021). Significant factors,

96 particularly the administrative factors that influence the successful implementation of solid waste  
97 management practices, have been explored by Fernando (2019). The remuneration to the staff in  
98 the local government authorities, support from the political leadership and contribution of the  
99 business communities and society are found to have a more significant influence on the successful  
100 implementation of solid waste management practices (Fernando, 2019). Furthermore, Kumara and  
101 Pallegedara's study (2020) reveals that Sri Lankan households have transferred to waste collection  
102 arrangements over the past decade rather than burning, dumping and composting. The waste  
103 disposal methodology transfer is shown to create more challenging tasks for the leading service  
104 providers (Kumara & Pallegedara, 2020).

105 With the outbreak of the COVID-19 pandemic situation in recent years, it is revealed by Yousefi  
106 et al. (2021) that the quantity and composition of solid waste had been rapidly changed. Increased  
107 production and consumption of specific equipment such as face masks, disposable gloves and  
108 certain personal protection equipment (PPE) for the community health needs and the behavioural  
109 and lifestyle changes on the created situation have caused the sudden change in the composition  
110 of the solid waste during the pandemic situation. The increased home delivery services during  
111 recent years and home cooking during travel restriction periods have contributed significantly to  
112 increased paper packaging and household waste generation (Yousefi et al., 2021). The authors  
113 identify the waste segregation into infectious and hazardous waste as a better practice to prevent a  
114 further widespread pandemic.

115 Municipal solid waste management is realised to incur higher costs and expenditures in the present  
116 world if the management practices are poor and weakly administered. Unlike developed countries,  
117 the allocation of huge funds for solid waste management is unbearable in the long term  
118 (Balasooriya et al., 2015). Therefore, more sustainable practices should be formulated and

119 implemented by these institutions. Furthermore, local government and national authorities are  
120 requested to use various political and economic instruments to involve the households in  
121 sustainable development (Berglund, 2006). It is also revealed that most municipal solid waste is  
122 generated from households, followed by waste from commercial areas (Dikole & Letshwenyo,  
123 2020; Dharmasiri, 2020). Moreover, a higher proportion of the global solid waste generated goes  
124 to the food waste category or biodegradable category (Nnaji, 2015; Dikole & Letshwenyo, 2020;  
125 Karim Ghani et al., 2013).

126 Food habits, culture and specific traditions, climatic conditions, income level, are identified as  
127 factors influencing the composition of the solid waste generated. Dikole & Letshwenyo (2020)  
128 have conducted a study on the composition of solid waste in Botswana. It was revealed that the  
129 composition of household solid waste was dominated by food waste in the lower, middle- and  
130 high-income level households during the weekdays and weekends. Food, paper and plastic waste  
131 characterised the composition of lower-income households. The middle-income families were  
132 represented with additional glass and tin waste, whereas the high-income household waste  
133 composition consisted of different textile and timber waste (Dikole & Letshwenyo, 2020). In Sri  
134 Lanka, the primary component of solid waste is dominated by food and garden waste, paper and  
135 cardboard and plastic waste (Wijerathne et al., 2012).

136 A few researchers have identified the challenges and opportunities associated with SWM practices.  
137 Nnaji (2015) has identified poor funding, inefficient human resources, inaccessibility to collection  
138 centres, inadequate equipment and dumping sites as significant challenges to the waste  
139 management system in Nigeria. Similarly, Dharmasiri (2020) and Fernando (2019) have identified  
140 the challenges such as lack of institutional setup and a proper regulatory framework, lack of  
141 adequate waste collection and transport facilities, lack of a proper system for waste segregation,

142 lack of appropriate instrument and modern technology as significant challenges affecting the SWM  
143 practices in the Sri Lankan context.

144

## 145 **2.2 Solid waste generation**

146 Knowledge in the waste generation and composition of solid waste could be considered crucial in  
147 planning and formulating continuous monitoring strategies (Dikole & Letshwenyo, 2020;  
148 Wijerathne et al., 2012). Dikole and Letshwenyo (2020) have examined the waste generation rate,  
149 composition and characteristics of solid waste generated in Botswana according to the income  
150 level of the households. It was revealed that the waste generation rate in low-income families'  
151 increases on the weekdays, and the rate decreases in the middle and higher-income homes on the  
152 weekdays.

153 Extant literature has identified income as an important determinant of the amount of solid waste  
154 generated (Balasooriya et al., 2015; Wijerathne et al., 2012; Wang & Qiu, 2013; Sarica et al.,  
155 2020). Wang and Qiu (2013) reveal that discharge of solid waste is related to the socio-economic  
156 development in China. And the per capita solid waste is found to increase with the income of the  
157 rural households. However, the amount of waste generated deteriorated after rising to a certain  
158 level. After the specified income level, the per capita solid waste generated decreases owing to  
159 residents choosing more environmentally friendly and responsible products. Furthermore, Cusano  
160 (2016) and Sarica et al. (2020) identify income as an important factor for food waste generation.

161 The solid waste generation in the Sri Lankan context has been identified to vary with the  
162 employment, land availability and economic level (Wijerathne et al., 2012; Balasooriya et al.,  
163 2015). In particular, Balasooriya et al. (2015) surveyed Galle and Hambanthota districts in Sri



164 Lanka and discovered that residential solid waste contained a significant proportion of  
165 biodegradable waste and an insignificant proportion of non-biodegradable waste. Commercial  
166 waste was identified to have substantial proportions in both types of waste. Moreover, the amount  
167 of waste generated in Galle district was lower than that in the Hambanthota district, owing to the  
168 high employment and lower land availability in Galle. Perhaps, residents spending more time in  
169 the workplace and having small home gardens have decreased solid waste generated in the Galle  
170 district (Balasooriya et al., 2015).

171

### 172 **2.3 Solid waste segregation**

173 The research related to waste segregation behaviours has been based on the Theory of Planned  
174 Behaviour (TPB). The TPB is a framework that explains human behaviour is triggered by the  
175 intention to change and influenced by attitude, subjective norms and perceived behavioural control  
176 (Ayob et al., 2017). Consequently, the waste segregation behaviour at the source is triggered by  
177 waste segregation intention. The intention would be created based on the attitude to separate waste,  
178 the social group pressures on waste segregation, and the personal confidence in the ability to  
179 separate waste. Considering the aforesaid theory, many other researchers have also found similar  
180 determinants; namely, attitude, subjective norms, and perceived behavioural control, as the  
181 determinants of the waste segregation behaviour (Alhassan et al., 2018; Ayob et al., 2017; Karim  
182 Ghani et al., 2013; Chen & Lee, 2020).

183 Attitude for waste segregation is identified to have a positive and significant effect on the waste  
184 segregation behaviour (Alhassan et al., 2018; Ayob et al., 2017; Karim Ghani et al., 2013; Chen  
185 & Lee, 2020). The research-based TPB collectively suggests a stronger correlation between the

186 attitude factor (whether the segregation is time-saving or wasting, whether to participate or not in  
187 segregation) and the solid waste segregation behaviour. Moreover, some studies identify the  
188 perceived behavioural control factor as a significant influence on waste segregation behaviour  
189 (Ayob et al., 2017; Karim Ghani et al., 2013). Ayob et al. (2017) state that being highly confident  
190 about the waste segregation practices increases waste segregation. However, the results concerning  
191 the influence from subjective norms relating to the pressure from social groups may vary according  
192 to the research settings and country settings. Alhassan et al. (2018), Chen and Lee (2020), Karim  
193 Ghani et al. (2013) have identified subjective norms as a determinant of waste segregation  
194 behaviour. Nonetheless, Ayob et al. (2017) have identified subjective norms as having an  
195 insignificant relationship with waste segregation in the Malaysian context.

196 Besides identifying determinants based on TPB, situational factors (limited space, time,  
197 cooperation, etc.) and socio-economic or demographic factors are identified as determinants for  
198 waste segregation behaviour. Educational level has been identified as a determinant of household  
199 waste segregation behaviour, indicating that people with a higher education qualification engage  
200 more in waste segregation activities (Dai et al., 2017; Alhassan et al., 2018). Saphores et al.(2006)  
201 have concluded that the absence of a college education could decrease the willingness to drop the  
202 waste generated at recycling centres.

203 Furthermore, gender and age are identified to positively relate to the waste segregation behaviour  
204 (Dai et al., 2017; Owusu et al., 2013), indicating that females are more inclined to separate waste  
205 than males, and the older residents are keen on separating waste more than younger residents.

206

### 207 **3. Methodology**

208

209 This study used a quantitative research approach to discover determinants of waste generation and  
210 segregation behaviour at the household level. The educated communities residing in the Western  
211 and North-Western provinces were considered for the analysis. Primary data were collected using  
212 an online questionnaire.

213 With the rapid technological developments, the web user population is observed to have an  
214 increasing trend. However, the usage of internet facilities for informational needs or internet  
215 penetration is still not widespread in the case of developing countries compared to developed  
216 countries (Vasantha Raju & Harinarayana, 2016). A sample of educated communities who have  
217 received tertiary education or an educational qualification above schooling were selected for the  
218 study. The online survey methodology was selected for the focused-group as the internet adaption  
219 is best explained and acknowledged on the educational levels (Billon et al., 2021). In brief, an  
220 online survey tool was employed for the sample data collection since educated communities are  
221 more accessible and have a thorough knowledge of adapting to internet facilities. The sample  
222 consists of 408 households: 57.60% from the Western Province and 42.4% from the North-  
223 Western Province.

224

#### 225 **3.1 Dependent Variables**

226 As the focus of the study is on waste generation and segregation behaviour, the study's dependent  
227 variables are chosen to be the amount of waste generated (in grams) and waste segregation choices  
228 of each household on a daily-basis.

229 Waste generated by each household is the total amount of waste collected in grams in three waste  
230 categories; food waste, polythene waste, and paper waste. Prior research on the household waste  
231 composition and generation has shown that household waste primarily consists of food waste,  
232 polythene and plastic material (Burnley, 2007; Edjabou, et al., 2015; Boer, Jedrczak, Kowalski,  
233 Kulczycka, & Szpadt, 2010). The respondents were given multiple-choice questions to select the  
234 range of weights of each three waste types. The mid-point of the weight range chosen by  
235 respondents was taken as the amount of waste generated under each category, and thus, it is a  
236 continuous variable. Subsequently, the average weights of the three types of wastes, food,  
237 polythene and paper, were summed up to obtain the total waste generated per household.

238 The second dependent variable in the analysis, waste segregation behaviour, takes the form of a  
239 binary variable taking the value of 1 if the residents practise waste segregation in their households  
240 and 0, otherwise.

241

### 242 **3.2 Explanatory variables**

243 Several socio-economic characteristics were taken into consideration. For instance, categorical  
244 variables were used to identify the respondents' characteristics according to age, house type and  
245 ownership status, and family income, while provincial category, local governmental authority  
246 mediation in waste collection and land segmentation status were binary variables. More details of  
247 the categorical variables can be found in Appendix I. The factors such as family size, attitude score,  
248 knowledge score and land area were identified as continuous variables.

249 The attitude of each respondent was measured using the responses received from four attitude-  
250 related sub-questions. Scores were assigned according to the responses based on a Likert scale,

251 where the “strongly agree” choice was given a maximum score of 5 and the “strongly disagree”  
252 response was given a score of 1. Appendix II shows the assigned scores for each question  
253 according to the scale. The scores obtained for each of the four-attitude related questions were  
254 summed up, and the average value was then obtained and multiplied by two to obtain the attitude  
255 score out of ten.

256 Using three related questions, a scoring method was developed to quantify each respondent's waste  
257 segregation-related knowledge score. Based on the responses for the three questions, a maximum  
258 score of ten was assigned if the respondent answered all three questions correctly. If the answers  
259 for two questions were given correctly, the respondent was assigned a knowledge score of six,  
260 while in the case of one correct answer, the respondent was assigned a score of three. If the  
261 respondent gave wrong answers for all three questions, a score of 0 was assigned. Appendix II  
262 provides the questions forwarded to respondents to measure the extent of their knowledge on waste  
263 segregation.

264

### 265 **3.3 Estimation methods**

#### 266 **3.3.1 Determinants of household waste segregation**

267 The waste segregation choice variable is a binary categorical variable. Hence, binary Logistic  
268 Regression Model (LRM) was used rather than using the simple Ordinary Least Square (OLS)  
269 regression as the waste segregation behaviour is a dichotomous (binary) criterion (Cohen, Cohen,  
270 West, & Aiken, 2003). The model specifications are given in equation (1).

$$\text{Log} \left( \frac{P_i}{1 - P_i} \right) = \beta_0 + \beta_1 X_i + e \quad (1)$$

271 Where  $P_i = 1$  if the respondent practises waste segregation at households;  $P_i = 0$ , otherwise;  $X_i$   
 272 representing the explanatory variables as described above;  $\beta_0 =$  constant term;  $\beta_i =$  coefficient of  
 273 explanatory variables;  $e =$  error term and  $i = 1, 2, 3, \dots, n$  which represents the number of independent  
 274 variables.

275 The coefficients in the LRM are estimated through the Maximum Likelihood (ML) method after  
 276 converting the outcome variable to logit variable (Begum, Siwar, Pereira, & Jaafar, 2009; Al-  
 277 Khateeb, Al-Sar, Al-Khatib, & Anayah, 2017) Accordingly, the LRM model used for the analysis  
 278 is as follows;

$$\begin{aligned} \text{Log Odds (waste\_segregation)} = & \beta_0 + \beta_1 \text{age} + \beta_2 \text{familysize} + \beta_3 \text{attitude} + & (2) \\ & \beta_4 \text{knowledge} + \beta_5 \text{lgacollection} + \beta_6 \text{familyincome} + \beta_7 \text{landarea} + \\ & \beta_8 \text{housetype} + \beta_9 \text{houseownership} + \beta_{10} \text{landsegmentation} + \\ & \beta_{11} \text{province} + e, \end{aligned}$$

279 *where,*

280 *waste\\_segregation* = 1, if the respondent practises waste segregation at households; 0 otherwise

281 *age* is the categorical variable denoting the age

282 *familysize* is the number of members in the household

283 *attitude* is the score obtained for the attitude of the respondents on waste management

284 *knowledge* is the score obtained for knowledge of the respondents on waste management

285 *lgacollection* is a binary variable where the local government authorities come to collect waste  
 286 from households or otherwise

287 *familyincome* is the monthly income of the household

288 *landarea* is the total land area of the household  
 289 *housetype* is a categorical variable on the house type  
 290 *houseownership* is a categorical variable on the ownership status of the house  
 291 *Landsegmentation* is a binary categorical variable on whether the land has been segmented  
 292 *province* is a binary categorical variable whether the respondent is from the North-Western or  
 293 Western Province

294

295 The coefficient ( $\beta$ ) in the logit model represents the effect of a unit increased in the explanatory  
 296 variable on the log odds  $[\ln(\frac{P}{1-P})]$ . For simpler representation in the results through the logit  
 297 model, the marginal effect of each independent variable ( $x_i$ ) could be calculated on the predicted  
 298 probability as follows (Daniels & Minot, 2018);

$$\Delta P / \Delta x_i = \beta_i P(1 - P) \quad (3)$$

299

300 Where the marginal effect of the explanatory variable ( $x_i$ ) on the probability  $P$  that  $Y=1$  is  
 301 equivalent to the multiplicity value of the coefficient of on  $x_i$  and  $P(1 - P)$ .

302 When a model is fitted to the data, how well the model fits must be checked. The Pearson statistic  
 303 and the Hosmer–Lemeshow goodness of fit test was calculated to determine whether the model  
 304 fits. The calculation of the Pearson statistic is given below,

$$\chi^2 = \sum_{j=1}^A \frac{y_j - a_j p_j)^2}{a_j p_j (1 - p_j)} \quad (4)$$

305 Where A is the total number of covariates patterns within B observations, and  $a_j$  is the total  
 306 covariates having the covariate patterns j and  $y_j$  is the total number of positive outcomes in j  
 307 covariate patterns where  $j=1,2,\dots, B$ .  $p_j$  is the probability predicted in j covariate patterns.

308 Hosmer–Lemeshow statistic (Hosmer & Lemeshow, 2000) is calculated similarly, using quantities  
 309 rather than covariate patterns. The model is well fitted if the probability of checking significance  
 310 is higher than the threshold value (0.05) in both tests.

311

### 312 **3.3.2 Determinants of the amount of waste generated**

313 The effect of the potential determinants on the amount of waste generated is estimated by  
 314 employing the following multiple regression model:

$$Y = \beta_0 + \beta_1 X_i + e \quad (5)$$

315 Where Y= the amount of waste generated in the household per day;  $X_i$  Representing the  
 316 explanatory variables as described above;  $\beta_1$ , representing the set of coefficients of the explanatory  
 317 variables;  $\beta_0$  representing the constant term and  $e$ , representing the error term.

318 Specifically, the multiple regression model with potential covariates is as follows:

$$\begin{aligned} \text{waste\_amount} = & \beta_0 + \beta_1 \text{age} + \beta_2 \text{familysize} + \beta_3 \text{attitude} + \\ & \beta_4 \text{knowledge} + \beta_5 \text{lgacollection} + \beta_6 \text{familyincome} + \beta_7 \text{landarea} + \\ & \beta_8 \text{housetype} + \beta_9 \text{houseownership} + \beta_{10} \text{landsegmentation} + \\ & \beta_{11} \text{province} + e \end{aligned} \quad (6)$$



319 Where,  
320 *waste\_amount* is the amount of waste generated per day  
321 *age* is the categorical variable denoting the age  
322 *familysize* is the number of members in the household  
323 *attitude* is the score obtained for the attitude of the respondents on waste management  
324 *knowledge* is the score obtained for knowledge of the respondents on waste management  
325 *lgacollection* is a binary variable where the local government authorities come to collect waste  
326 from households or otherwise  
327 *familyincome* is the monthly income of the household  
328 *landarea* is the total land area of the household  
329 *housetype* is a categorical variable on the house type  
330 *houseownership* is a categorical variable on the ownership status of the house  
331 *landsegmentation* is a binary categorical variable on whether the land has been segmented  
332 *province* is a binary categorical variable whether the respondent is from the North-Western or  
333 Western Province

334

335 The diagnostic tests were conducted to check the precision of the model. The omitted variable tests  
336 (Ramsey, 1976) and multicollinearity tests using centred variance inflation factor (VIF) were  
337 conducted to check the model's goodness of fit, while heteroscedasticity was also controlled in the  
338 study.

339

340 **4. Results**

341

342 *4.1 Characteristics of the respondents of North-Western and Western provinces*

343 According to the descriptive statistics shown in Table 1, most respondents were from the Western  
344 province (57.6%). A higher percentage of respondents belonged to the working-age group aged  
345 25-54 years. The average number of members in the household was four. Most of the respondents  
346 (62.01%) stated that the local governmental authorities do not collect the waste from their  
347 households. A majority of the respondents state that they separate waste in their households  
348 (86.27%), and the average waste collected was 851.4 grams per day in households in the North-  
349 Western and Western provinces.

350 Table 1: Summary of the variables used in the analysis

Variable	Description	%	Mean	Standard
Type				deviation
categorical	<i>Segregation decision</i>			
independent	Separated	86.27		
variables	Not separated	13.73		
	<i>Province</i>			
	Western	57.6		
	North-Western	42.4		
	<i>Age of the respondent</i>			

15-24 years (Youth)	47.55
25-54years (working-age population)	49.26
55-64 years	2.21
above 65 years	0.98

*LGA waste collection facility*

Yes	37.99
No	62.01

*Family Monthly Income*

0-50,000	38.97
50,000-100,000	23.77
100,000-200,000	20.34
Above 200,000	16.91

*house type*

Single storied	71.57
Two storied or higher	26.96
Other	1.47

*house ownership*

Fully owned	93.38
-------------	-------

	On rental	4.17	
	Other	2.45	
	Land segmented		
	Yes	22.79	
	No	77.21	
Continuous variables	Total waste generated per day (grams)	851.41	559.93
	Family size	4	1.44
	Attitude score	8.53	1.07
	Knowledge Score	8.32	2.60
	Land area (perch)	104.64	614.70

---

351

## 352 **4.2 Determinants of household waste segregation**

353 Table 2 summarises the estimated odd ratios and marginal effects of determinants of waste  
354 segregation behaviour. The working adult age category (25-54 years) and the elderly age category  
355 (above 65 years) were identified to have reduced the probability of household waste segregation  
356 by 8.1 and 48.5 percentage points, respectively, compared with the youth age category (15-24  
357 years). Further, when the local government authorities engaged in the waste collection from  
358 households, the probability of household waste segregation increased by 15.3 per cent than when  
359 the authorities did not collect waste. Residing in a two or more storied house decreased the  
360 probability of household waste segregation by 12.8 per cent compared with living in a single-story

361 house. Obtaining a monthly income of 100,000 - 200,000 LKR by the household increased the  
 362 probability of household waste segregation by 15.1 per cent compared with households with a  
 363 family income of 0 - 50,000 LKR.

364

365

366 Table 2: Logistic Regression Model estimated coefficients for waste segregation

VARIABLES	Odd Ratio	Marginal Effects (dy/dx)
<b>Provincial category (base category: North Western province)</b>		
Western province	0.0842 (0.3458)	0.0106 (0.0436)
<b>Age (base category: 15-24 years)</b>		
25-54 years	-0.5504 (0.3480)	-0.0706 (0.0455)
Above 65 years	-2.6820** (1.1763)	-0.4758** (0.2182)
<b>Family size</b>	0.1169 (0.1235)	0.0147 (0.0155)
<b>Attitude</b>	-0.1110 (0.1628)	-0.0140 (0.0205)
<b>Knowledge</b>	0.0293 (0.0594)	0.0037 (0.0075)

<b>If local government authority involved in the waste collection</b>	1.4958***	0.1571***
(base category: otherwise)	(0.4800)	(0.0405)
<b>Land area</b>	-0.0003	0.0000
	(0.0003)	(0.0000)
<b>House Type (base category: single storied house)</b>		
Two or more stories house	-0.7490	-0.1066
	(0.4755)	(0.0738)
<b>House ownership (base category: Fully owned house)</b>		
On rental	-1.1326	-0.1820
	(0.7785)	(0.1462)
Other	0.1545	0.0183
	(1.1278)	(0.1279)
<b>Monthly family income (base category: 0-50,000LKR)</b>		
50,000 – 100,000LKR	0.3103	0.0449
	(0.3904)	(0.0553)
100,000 – 200,000LKR	1.5899***	0.1611***
	(0.5589)	(0.0465)
<b>If land is segmented ( base category: otherwise)</b>	0.1983	0.0243
	(0.3916)	(0.0466)
Constant	1.5314	
	(1.4953)	
Observations		322

368           \*, \*\*, \*\*\* indicates statistical significance at 10%, 5% and 1% level, respectively.

369           The model is well fitted with reference to the Pearson test ( $p= 0.4894$ ) and Hosmer Lemeshow

370           test ( $p=0.3537$ ) for the goodness of fit.

371

### 372   **4.3 Determinants of the amount of waste generated**

373   Table 3 presents the OLS regression model results for the amount of waste generated. It is evident  
374   that the elements, namely, the age, family size, respondents' attitude towards waste management,  
375   the local government authority mediation in waste collection, total land area, ownership status of  
376   the household and family income, significantly influence the total amount of waste generated at  
377   the household-level.

378   The elderly age category (above 64 years) generated 382 grams less waste than the youth age  
379   category (15-24 years). When the number of family members in a household increased by one unit,  
380   the daily amount of waste generation increased by 38.14 grams daily. Moreover, a positive attitude  
381   with the household on waste management practices decreased the amount of waste generated. Still,  
382   knowledge about waste has not shown a significant relationship with the amount of waste  
383   generated. The waste collected by local government authorities positively influenced the amount  
384   of waste generated. An increase in the land area of the household has shown to decrease the amount  
385   of household waste. In contrast, a decrease of 230 grams of waste was observed in houses operating  
386   on a rental basis, compared with self-owned houses. The households having an income of more  
387   than Rs. 200,000 were observed to generate an additional 202 grams of waste daily compared with  
388   households with a monthly income of less than Rs. 50,000.

389 Table 3: OLS regression results for the determinants of the amount of waste generated at the  
 390 household level

<b>VARIABLES</b>	<i>Coefficients</i>
<b>Provincial category (base category: North Western province)</b>	
Western province	17.83 (59.11)
<b>Age (base category: 15-24 years)</b>	
25-54 years	-56.39 (81.43)
55-64 years	-122.7 (234.2)
Above 65 years	-382.3*** (119.0)
<b>Family size</b>	41.47** (20.80)
<b>Attitude</b>	-68.97** (32.26)
<b>Knowledge</b>	5.528 (11.19)
<b>If Local government authority involved in the waste collection (base category: otherwise)</b>	170.1** (70.76)
<b>Land Area</b>	-0.0644*** (0.0125)
<b>House Type category (base category: single storied house)</b>	



Two or more stories house	-5.160 (74.88)
Other	454.1* (233.2)
<b>House ownership category (base category: Fully owned house)</b>	
On rental	-230.0* (124.0)
Other	-170.4 (111.7)
<b>Monthly family income category (base category: 0-50,000LKR)</b>	
50,000 – 100,000LKR	98.36 (77.82)
100,000 – 200,000LKR	80.24 (93.02)
Above 200,000 LKR	202.0* (108.3)
<b>If the land is segmented (base category: otherwise)</b>	
Constant	-0.0785 (69.83)
	1,055*** (307.7)
<hr/>	
Observations	398
Probability	0.000
R-squared	0.099
<hr/>	

392 Note: Robust standard errors are given in parentheses  
393 \*, \*\*, \*\*\* indicates statistical significance at 10%, 5% and 1% level, respectively.  
394 The test for omitted variables by Ramsey (1976) to tests for omitted variables ( $p=0.1107$ ) proves that there  
395 is no omitted variable bias  
396 The variance inflation factor (VIF) of all variables was less than 10, suggesting that multicollinearity is not  
397 a serious issue in the present analysis (See Appendix III).

398

## 399 **5. Discussion**

400

### 401 **5.1 Waste segregation behaviour**

402 The present analysis suggests that the age, type of the house, family income and the local  
403 government authority service availability in the waste collection are the significant factors  
404 affecting the waste segregation behaviour among the educated-community in the Western and the  
405 North-Western provinces of Sri Lanka.

406 The categories of the working adult group of 25-54 years, elderly heads above 65 years, and the  
407 youth age category of 15-24 years are identified as the age groups less engaged in household waste  
408 segregation practices. These results contradict the results of Al-Khateeb et al. (2017) and Agwu  
409 (2012). For instance, Al-Khateeb et al. (2017) state that the older generation is more likely to  
410 separate waste than the youth in the case of Palestine, while Agwu (2012) has identified a positive  
411 relationship between age and solid waste management practices in the Nigerian context. However,  
412 a negative relationship was identified between the age and intention for payment for household  
413 waste segregation in Sweden (Berglund, 2006). Thus, it could be justified that the older generation

414 might be less keen on separating waste than the younger generation due to various factors such as  
415 lack of time, awareness and efforts on waste segregation and long-term dissatisfaction caused due  
416 to improper waste management by government authorities.

417 Two or more storied houses are less engaged in waste segregation than single-storey houses. These  
418 results contrast with the findings of Zakianis & Djaja (2017), where they specified that when the  
419 surface area of the household is relatively higher, the waste sorting among those households is  
420 more likely. His study also identified that luxury houses had increased waste sorting practices in  
421 the Indonesian context (Zakianis & Djaja, 2017). Usually, the two-storied dwellings are built in  
422 semi-urban and urban areas due to the problem of space limitation. Our results could be justified  
423 because households are less engaged in waste segregation in two-storied houses as less space is  
424 available for waste segregation activities. A positive relationship is identified between the family  
425 income and the household waste practices, similar to findings of other researchers (Zakianis &  
426 Djaja, 2017; Alhassan, Asante, Oteng-Ababio, & Bawakyillenuo, 2018). Household Income level  
427 is identified as a significant determinant by most studies and has produced mixed results in the  
428 context of different countries and regions.

429 The service of local government authorities in Western and North-Western provinces for waste  
430 collection is identified to have an increased-effect on household waste segregation. As the local  
431 government authorities request the houses to separate the waste when they are handed over to the  
432 compactors and waste collections tractors, the waste segregation likelihood among the households  
433 may have increased compared with the households where authorities have no waste collection  
434 services.

435 The attitude of the household head was not identified as a significant variable affecting the waste  
436 segregation decision in the analysis. These results contradict the prior revelation of attitude

437 influencing positively on the waste segregation intension, where the attitude was found to be a  
438 major factor influencing the intension pertaining to the theory of planned behaviour (Alhassan,,  
439 Asante, Oteng-Ababio, & Bawakyillenuo, 2018; Ayob, Low, Jalil, & Chin , 2017; Chen & Lee,  
440 2020; Al-Khateeb, Al-Sar, Al-Khatib, & Anayah, 2017; Zakianis & Djaja, 2017).

441

## 442 **5.2 Waste generation behaviour**

443 According to the results, age of the household head, attitude on waste management, land area and  
444 house ownership conditions are identified to have a significant negative relationship with the  
445 amount of waste generated, whereas family size, local governmental authority's waste collection  
446 service, and family income factors are identified to have a significant positive impact on the  
447 household waste generation.

448 The older age category of above 65 years is identified to have generated less waste than the youth  
449 age category. Irwan et al. (2013) have stated that the relationship between age and household waste  
450 generation is unclear. The results are similar to the findings of Struk and Soukopová (2016), where  
451 the older age residents are found to generate more solid waste than younger age residents in the  
452 case of Czech municipalities. The older generation's reason for generating more solid waste is  
453 recognised as older people's households are being renovated by refitting and replacing the  
454 equipment and practising with new equipment and methodologies. The younger generation is  
455 found to generate less solid waste with their knowledge in waste recycling and segregation.

456 Moreover, a positive attitude towards waste management practices has been shown to have  
457 decreased the amount of waste generated. Allegedly, the households having a better attitude  
458 towards the waste management practices tend to decrease the amount of waste generated. A

459 negative relationship is identified between the land area and the amount of waste generated.  
460 Furthermore, a decrease in the amount of waste generated is also observed with living in places on  
461 a rental basis compared to living in self-owned houses. Our results align with the findings of Kala  
462 et al. (2020). They also conclude that self-owned houses have a positive relationship with  
463 household waste generation in the case of India.

464 Household size or family size is also identified as a positive influencing factor on the amount of  
465 waste generated. Our results are in line with the results of Kala et al. (2020) and Trang et al. (2017)  
466 in the case of Indian and Vietnamese households, respectively. Moreover, the family income level  
467 is shown to positively affect the amount of waste generated within households. The high-income  
468 level with a monthly family income above Rs. 200,000 is identified to have generated increased-  
469 food waste, compared with the households with relatively lower income levels. These results align  
470 with the findings of the following studies (Dikole & Letshwenyo, 2020; Wang & Qiu, 2013;  
471 Sarica, Demircan, Erturk, & Arslantas, 2020; Kala, Kala, & Sushil, 2020). The high-income levels  
472 are found to lead to rapid lifestyle changes. The change in consumption patterns due to the moving  
473 to a more sophisticated lifestyle leads to increased-waste production within households.

474 A significant positive relationship was observed between the local government waste collection  
475 and the amount of waste generated. A reason for the increase in the amount of waste in places  
476 where local governmental authorities come to collect waste could be elaborated as the sense of  
477 relief that households experience due to the disposal of waste by authorities daily or weekly.  
478 Moreover, as the authorities collect the waste from households and are in service in more populated  
479 and urban areas, the urban and semi-urban residents' spending patterns may have led to increased  
480 household waste.

481

## 482 **6. Conclusion**

483

484 Solid waste management practices could be considered one of the grave issues globally, and  
485 household waste is an indispensable aspect of it. Waste segregation is an initial step in managing  
486 the waste at the source, and an integral part of the waste management process. The level of  
487 education is identified to significantly affect the waste management practices as per the prior work  
488 of literature. The objective of the present study was to identify the determinants of solid waste  
489 segregation decision and household solid waste generation in the Sri Lankan context, considering  
490 the educated communities that have received a higher level of education than schooling. Thus, the  
491 educated communities in the North-Western and Western provinces were selected for the study to  
492 find the factors influencing a household waste generation and segregation behaviour.

493 A logistic regression analysis was used to determine the factors influencing the household waste  
494 segregation decision. It was identified that household income levels and the involvement of local  
495 governmental authorities in waste collection positively affect waste segregation behaviour.  
496 Moreover, a significant negative relationship was identified with the age categories and the type  
497 of house. A multiple regression model was estimated to determine the factors influencing the  
498 amount of solid waste generated within the households. The socio-economic factors such as age,  
499 family size, attitude, land area, household ownership, family income, and involvement of local  
500 governmental authorities in the waste collection were considered the main determinants of the  
501 amount of waste generated in households in the Western and North-Western provinces of Sri  
502 Lanka.

503 In Sri Lanka, despite having a national plan on solid waste management practices at the local  
504 governmental authority levels, the existing regulatory framework was fragile (Fernando, 2019).

505 Thus, the improper segregation of solid waste within households lead to garbage mountains in the  
506 dumpsites that cannot be further refined for a helpful state. The findings of this study would aid in  
507 the formulation of strategies for improving the existing solid waste management practices of Sri  
508 Lanka considering specific communities. Waste segregation decisions and waste generation were  
509 influenced by the fact that local government authorities' involvement in the waste collection from  
510 households in the study. This reflects that the decision for waste segregation of educated  
511 communities relies primarily on the government policy regulations and demands of the local  
512 government waste collection services to separate the waste at the source. It was also found in the  
513 study that most households did not get the services from local government waste collection  
514 services in the North-Western and western provinces. Therefore, the local authorities need to  
515 significantly expand the waste collection services to fulfil the demand for household waste  
516 collection and to increase the policy regulations concerning the household's waste segregation  
517 practices.

518 Implementing a proper regulatory framework and policy strategies on waste segregation practises  
519 at the local governmental authority level, focussing on the characteristics of the communities,  
520 would be beneficial in managing the higher amounts of waste generated and further aid in reducing  
521 the costs of waste disposal. Insufficient land for final dumping, recycling and composting, poor  
522 regulatory framework and lower labour productivity and quality were considered as significant  
523 challenges affecting the successful policy implementation in Sri Lanka (Fernando, 2019).  
524 Accordingly, the local government authorities allegedly need the central government's support in  
525 expanding the infrastructure facilities and to provide incentives to implement a proper waste  
526 management mechanism. Further, as implied by the results, it could be anticipated that the  
527 households with a higher family income would be willing to afford the cost of implementing a

528 proper waste collection service by the local government authority level. The local authorities could  
529 mediate in implementing strategies such as executing attractive, profitable models through  
530 recycling and getting the private sector's involvement in the waste collection activities, which will  
531 lead to a better management of the waste generated in the areas under their purview. Further, it  
532 could be suggested to provide them incentives such as waste bins and proper maintenance of the  
533 waste containers and conducting awareness programmes for waste segregation at source at the  
534 local government authority level to improve the municipal solid waste segregation process (Chen  
535 & Lee, 2020).

536

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