

Is the private sector more efficient? Big data analytics of construction waste management sectoral efficiency

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Highlights

- Big data can distinguish the public and private disparity with better certainty.
- Overall, public and private sectors have no significant CWM efficiency difference.
- The private sector outperformed their public counterpart in demolition projects.
- The public sector performed superior in foundation and new building projects.

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Abstract

Efficiency disparity between the public and private sectors is a non-trivial issue that concerns fundamental choices of socio-political-economic systems. Waste management academia and industry also wrestle with issues relating to the choice between public and private sectors. To examine the disparity exclusively caused by “sector”, in statistics language, one needs data that is sufficiently big to control many other confounders, e.g., sites, project types, and construction technologies. This paper attempts to ascertain the construction waste management (CWM) efficiency disparity between the public and private sectors by using big data in Hong Kong. The waste disposal records of 132 projects, including 70 public and 62 private projects, were extracted and analysed. By comparing the waste generation flows (WGFs) and accumulative WGFs, it is found that, by and large, there is no significant efficiency disparity in CWM between the two sectors. However, a closer investigation discovered that the private sector outperforms their public counterpart in demolition projects, while the latter performs better in foundation and new building projects. Although there are private projects with higher CWM performance, their divergence between the best and average projects are larger than public ones. Such findings thus reject casual remarks that the private sector is more efficient in CWM. The underlying reasons maybe the waste management index practice promoted in public projects while the private sector is often incentivized to perform better CWM to save waste disposal levies. Future research is recommended to delve into the causes of the efficiency disparity and introduce CWM interventions accordingly.

Keywords: Public-private disparity; economic efficiency; construction waste management; big data; Hong Kong

Introduction

Whether the public sector is more efficient than its private counterpart, or vice versa, is a non-trivial concern in many fundamental choices related to our political, social, and economic systems (Sappington and Stiglitz, 1987; Meier and O'Toole, 2011; Mihaiu et al., 2010). Majority of such research usually conducts comparative analyses, either qualitatively or quantitatively, of the effects obtained in resources used. The comparisons even have an implication to the choice of capitalism and socialism, privatization and nationalization in the institutional economics. However, the debate on comparisons of private and public efficiency has not been concluded. The comparison between the dynamism of private sector and the inefficiency of public sector has been the principal political dogma of the past decades (Busch and Gustafsson, 2002). The fanatic of private sector superiority acclaims that the compared to public sector, private sector is more efficient and dynamic, while the public sector is slower and more wasteful; that a higher proportion of private sector participation will make things better (Simms and Reid, 2013). The neoliberalism also puts superiority in the private sector, arguing it being efficient, dynamic, and modernize. While capitalism adherents are convinced by the public sector with evidences that towering debts have been run up under privatization

71 and tremendous services are being brought back into public ownership. They think
72 privatization *per se* does not guarantee improved efficiency. There are also more and more
73 standpoints disputing that privatized services perform worse according to largest studies. The
74 reasons driving the efficiency disparity between the public and private sectors can be multiplex,
75 ranging from institutional, managerial, technical, and organizational, to information
76 accessibility aspects (Ring and Perry, 1985; Sappington and Stiglitz, 1987; Bretschneider, 1990;
77 Boyne, 2002; Bysted and Hansen, 2015).

78
79 Concerning the efficiency disparity between the public and private sectors, there are a rich vein
80 of works on different aspects. Some economists, from a theoretical perspective, have advanced
81 a strong argument that private firms are more efficient than public firms (Alchian, 1965; De
82 Alessi, 1974; Sheshinski and Lopez-Calva, 2003). They emphasized the importance of non-
83 transferability of ownership and weakening of property rights in public sectors in support of
84 the efficiency disparity between public and private sectors. However, the priori theorized view
85 is weak in empirical confirmation with inconclusive results. Das (2012) confirms that in the
86 mining industry, the participation of the private sector can boost the overall productivity by
87 comparing the extraction efficiency of public and private mining firms. Bhattacharyya et al
88 (1994), on contrary, by analyzing public water utilities practice projects, offer evidence of
89 higher efficiency in public sector although they are more widely scattered between best and
90 worst. More empirical studies reveal that there is no convincing evidence of one form of
91 ownership systematically surpassing over another. For example, Byrnes et al (1986) failed to
92 find any evidence to support the superiority of privately owned utilities over publicly owned
93 ones by measuring efficiency directly in terms of a production function in the water utility
94 industry. The same result goes to Estache and Rossi (2002), who also choose samples of water
95 companies. Karas et al (2010) provide the results that the Russian public banks are not more
96 inefficient than private ones.

97
98 When shifting to the area of waste management, the empirical studies on the public-private
99 argument are limited (Xu et al., 2018). Ichinose et al (2013) find that prefectures where private
100 sector participated in household solid waste collection with a higher proportion are more
101 efficient from the solid waste logistics practices in Japan. By conducting case studies of waste
102 management in Lebanon, Massoud et al (2003) suggest a mixture of private and public sectors
103 without distinguishing the efficiency disparity between the two. Simões et al. (2012) evaluated
104 the productivity and efficiency of the waste sector based on 228 waste collection and treatment
105 utilities and that private sector participation does improve the efficiency in waste collection but
106 benefits ephemerally in waste treatments. Massoud et al. (2003) suggested that, from the cost
107 perspective, private sector provided services were between than public sector provided ones
108 based on the comparison of municipal solid waste collection in two largest cities. Jacobsen et
109 al. (2013) also reached a similar conclusion by the analysis of multiple household waste
110 collection service cases. Lu et al. (2016) applied the Coase Invariant Theorem as a guiding

111 theory to examine potential waste management performance disparity among the public and
112 private construction clients. Therefore, the debate on the efficiency disparity between public
113 and private sectors is far from concluding. Especially, when narrowing down to construction
114 waste management, confident answer is due with the support of empirical research.

115
116 According to previous research, the answer to the debate of efficiency disparity between the
117 public and private sector in waste management with empirical supports is still missing. This
118 paper aims to fill this research gap by answering the question whether the private sector is more
119 efficient by using big empirical data reflecting waste management performance in Hong
120 Kong's construction industry. Empirical big data analytics from real cases of housing
121 development will be adopted as the research method. There are two rationales behind this
122 research method. First, it is to make the classic inquiry more manageable by narrowing the
123 scope down to construction waste management (CWM), which is one of the most oft-examined
124 areas concerning efficiency disparity between sectors. In contrast to previous studies, the
125 research reported here seeks to understand the specific question of efficiency disparity of CWM
126 between the public and private sectors. Second, it is to make good use of a set of big data of
127 CWM in Hong Kong, and other useful information, such as the project location and features,
128 that will guide better business predictions and decision-making (Lu et al., 2015). It is
129 anticipated that big data can help uncover hidden patterns and unknown correlations, control
130 the numerous factoring confounding the efficiency of an economic system to allow the
131 exclusive contribution of "sector" to surface. To this end, this paper aims to contribute to the
132 general question of efficiency disparity between sectors by introducing big data analytics,
133 hoping that big data can open a new avenue, through which the classic question can be brought
134 to a conclusion.

135
136 To be specific, the authors plan to select representative housing construction projects from both
137 the public and private sector, extract their waste treatment data from the open data sets, and
138 conduct comparative analyses of the public and private sector. The detailed research methods
139 will be explained in the Data and methods section. The rest of the paper is organized as follows.
140 After the introductory section is a brief literature review focusing on how big data can be used
141 in applied statistics to examine the classic question of efficiency disparities between sectors.
142 Section 3 presents the big data and the data processing methods, which include standardization
143 of the project data for visualization and statistical analyses. Section 4 is to report the data
144 analyses, results, and findings, followed by an in-depth discussion in Section 5. Conclusions
145 are drawn in Section 6.

147 **Big data and better certainty**

148 A consensus on the definitions of "big data" is yet to be reached. Researchers are converged to
149 adopt Gartner's three definitional characteristics of big data: volume, variety, and velocity
150 (McAfee et al., 2012). Volume indicates the large quantity of the data; velocity indicates that

151 the data is incoming in a high speed, which requires prompt processing to harness its value;
152 and variety indicates big data must be rich in semantics in the forms of structured, unstructured,
153 semi-structured, or a combination thereof (Russom, 2011; Zaslavsky et al., 2013). In view of
154 the fact that big data does not mean quality data, researchers are increasingly emphasizing
155 ‘veracity’ - how accurate or truthful a data set may be - as the fourth ‘V’ of big data. Big data
156 analytics have been developed to analyze big data in order to discover hidden patterns, non-
157 linear relationships and casual effects that will guide better informed decision-making (Lu et
158 al., 2015). Likewise, according to Agrawal (2016) and WEF (2012), big data can lead to some
159 potential knowledge or guidance information that can be utilized for further decision-makings.
160 Hence, value is advocated as the fifth ‘V’ of big data.

161
162 Lu et al. (2018) critiqued that too often big data analytics is mistakenly associated with ‘pattern
163 detection algorithms’, ‘unsupervised machine learning’, ‘deep learning’, ‘artificial
164 intelligence’, NoSQL database, Hadoop, and other fascinating data mining methods. Instead,
165 they echoed with Leek (2014) that applied statistics should not be left out when harnessing the
166 value of big data. In the history of probability theory, there is a “law of large numbers”, which
167 is a theorem that describes the average value of the results retrieved from a large number of
168 trials should be close to the expected value and will become closer when more trials are
169 conducted (Sen and Singer, 2017). If treating an economic sector as a complex system, it would
170 be legitimate to expect that big data can approach the entirety of the sector as it operates (i.e.,
171 tries naturally), and allow the inquiry of their efficiency disparity, if there is any, to be
172 ascertained.

173
174 The potential of big data to ascertain something in a complex system is further hyped in stories
175 of the psephological analysis of big data about voter behavior, such as Donald Trump’s election
176 campaign in 2016 and Brexit in the same year. These scenarios present a lot of uncertainties,
177 e.g., to vote or not, or to leave or to remain. It has been reported that Cambridge Analytica, a
178 data analytics company, can understand the scenarios with greater certainties with its
179 enormously voluminous and various big data. Actually, they were reported to manipulate the
180 public towards a more certain direction which was wanted by their clients. Inspired by the
181 stories, big data analytics may open a new avenue to examine the classic inquiry on efficiency
182 disparity between the public and private sectors. In statistical language, the data, if big enough,
183 can control the numerous confounders to examine the efficiency disparity that is solely
184 contributed by “sector”. This potential, however, has not been well explored in the literature.
185 It is under the potential that this paper tends to examine the CWM efficiency difference
186 between the public and private construction objects.

Data and methods

Data

The Hong Kong Environmental Protection Department (HKEPD) launched a Construction Waste Disposal Charging Scheme (CWDCS) in 2006. It requires that all construction waste, unless being reused or recycled, must be disposed of at designated government waste disposal facilities such as landfills, public fills, or off-site sorting facilities. Prior to using the facilities, a main contractor who is awarded a contract with more than HK\$ 1 million is mandated to open a billing account in the HKEPD for the contract solely, with basic information of the project including the contract name, client (e.g., public or private), contract sum, site address, type of construction work, etc. When the construction waste is disposed of at the facilities, HKEPD records information on every load of construction and demolition waste, including vehicle number, time, and weight when the vehicle enters and exits, and the billing account number the vehicle uses. Unintentionally, the practice in Hong Kong generates a large data set which makes probing into many aspects of CWM, such as the performance of public and private projects. Figure 1 illustrates the structure of the data sets, which contains the following five types of databases:

- *Project* database, contains basic information (including site addresses, clients, project types, and other details) of all projects with an account to access the above government facilities for waste disposal. There are a total of 27,536 construction projects being recorded in this database.
- *Facility* database, contains basic information (address, capacity) of all government CWM facilities.
- *Vehicle* database, contains 9,863 waste hauling vehicles for construction waste transport.
- *Waste Disposal* database, records detailed information of every truckload of construction waste disposed at the government waste management facilities. There are a total of 7,866,085 disposal records being generated by all the construction projects executed during the eight-year period from 2011 to 2018.
- Moreover, the Hong Kong Buildings Department (HKBD) keeps the *Building* data of all existing and new building projects, including the address of the site, number of blocks and storeys, building type, domestic and non-domestic gross floor area (GFA). The data can also be integrated with the waste disposal data above.

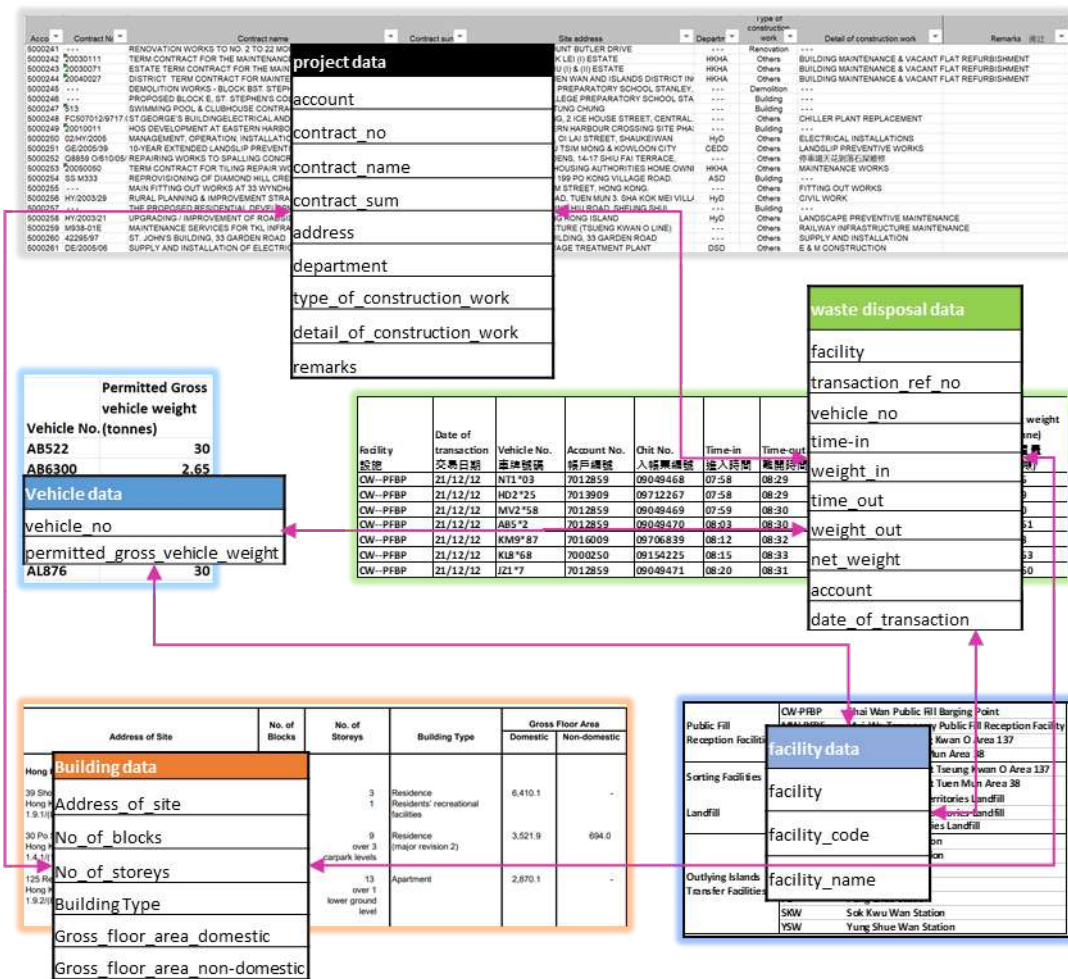


Figure 1. The big data set

It is argued as big data, according to the five 'Vs' criteria as described above, despite the volume is not as big as terra- or zetta-bytes. It is also argued that big data should not be mechanically equated to big volume, but it covers a fuller picture of the subject matter that is not possible provided by small data.

Data processing methods

The data processing methods, as illustrated in Figure 2, comprises of eight connected steps ranging from project selection, data extraction, processing, visualization, to detailed comparisons.

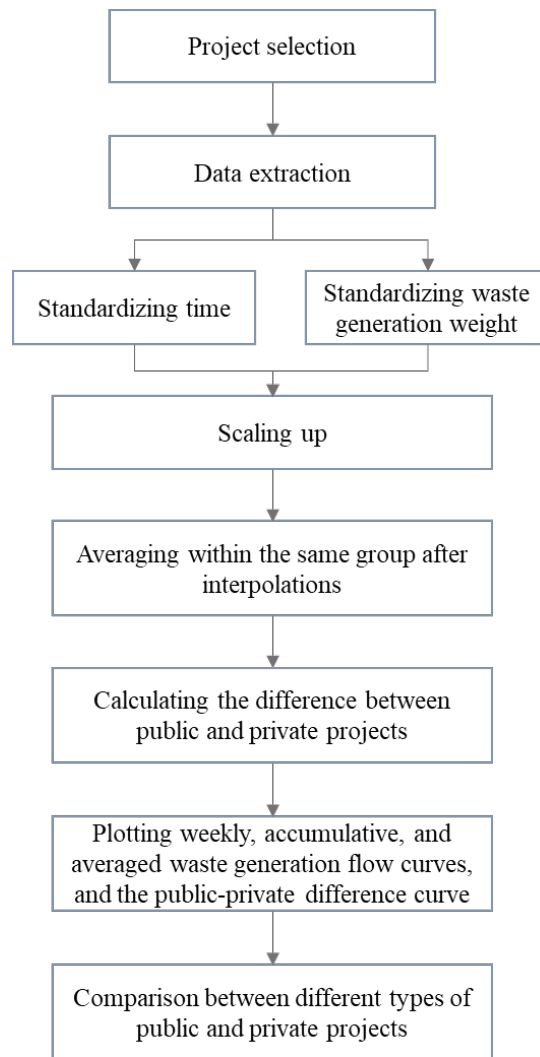


Figure 2. The procedures of data extraction and analysis

Project selection

The first step is to select projects for the analysis. A set of ‘qualified’ projects is selected from the big data set by meeting the following criteria:

- (1) The project must be started after 2011 and finished before 2019 to be reflected completely in the big data set;
- (2) Project type is restricted to demolition, foundation, and new building, as these types of projects are the major source of construction waste; and
- (3) The project must be relatively sizeable, as these projects allow more regular patterns than their smaller counterparts, which are often impacted by random factors.

By applying the three criteria, 132 projects, including 19 demolition, 59 foundation, and 54 new building projects were sourced. They can be further divided into two groups: 70 public and 62 private projects. Basic information of the selected projects is shown in Table 1.

Table 1 Basic information of the selected projects

Project type	No. of projects	No. of raw records	Contract Sum (HK\$)	Project duration (weeks)	Average project duration (weeks)
Public demolition	8	2,734	[1016392, 19008080]	[7, 30]	16.9
Private demolition	11	10,289	[15756600, 55166000]	[28, 155]	55.9
Public foundation	31	238,039	[4404854, 923000000]	[40, 173]	78.8
Private foundation	28	200,075	[94022505, 644652285]	[41, 161]	77.0
Public new building	31	232,393	[339388300, 4711780000]	[109, 441]	220.0
Private new building	23	49,684	[36890955, 3780000000]	[52, 398]	171.1
Total	132	733,214	[1016392, 3780000000]	[7, 441]	122.0

250

251 *Data extraction*

252 With the account number of the selected projects as the reference, the waste disposal records
 253 of the project can be queried and extracted from our database. The query is done by SQL server.
 254 Afterward, all the extracted data are sorted on the project basis.

255

256 *Standardizing time and waste generation weight*

257 Since projects are different from each other in terms of project duration, the first step to process
 258 the data is to standardize the time to make the projects comparable. The time of a project is
 259 counted from its first record of waste disposal. For example, if the first record of Project A is
 260 the 27th week of 2015, then we count from week 27 of 2015 and calculate the following time
 261 based on this baseline. Suppose Project A ends at 10th week of 2017, then the duration of Project
 262 A is 87 weeks taken that one year is 52 weeks. Different projects may start from a different
 263 time, which means their baselines are different. The time is then standardized by percentage,
 264 see formula (2):

$$265 \quad T_i\% = t_i/T * 100\% \quad (2)$$

266 Where t_i is the time point of a project; T is the total time of the project

267 Take Project A as an example, $T = 87$ weeks, the standardized time of the second week is $T_2\%$
 268 $= 2/87*100\% = 2.3\%$. The above standardization method is on a weekly basis. Actually, the
 269 data available allows to do it on a daily basis, but it turned out to be “over-engineered” by
 270 having so many days. It is also too sparse to examine waste generation on a monthly basis,
 271 therefore, the weekly scale is used.

272

273 Meanwhile, waste generation weight is also standardized, as the weight could range radically
 274 from one project to another depending on their size. The approach is to treat the total weight
 275 of waste generation of a project as 100%, and the weekly generation as a certain percentage of
 276 the total waste generation of that specific project. Adding together the construction waste
 277 weight of every vehicle using the same account number at one day, the total waste disposed of
 278 by the corresponding project at that day can be calculated. The construction waste of every
 279 project is further calculated on a weekly basis for easier analysis. The weight of weekly
 280 construction waste was further accumulated to calculate the waste generated until a time point.

281 The ratio of the disposed waste at the corresponding week in the total waste generation of the
282 project is calculated using formula (3):

$$283 \quad r_i\% = \frac{w_i}{W} * 100\% \quad (3)$$

284 where r_i is the percentage of total waste generation at week i , w_i is the waste generated at week
285 i , while W is the total waste generation.

286
287 By using the $T_i\%$ as the x-axis and the $r_i\%$ as the y-axis, the waste generation flow (WGF),
288 which portrays the changing of construction waste generation ratio with the changing of time,
289 can be plotted. The WGF curves can provide a clear and general visualization of the waste
290 generation performance of projects at different stages.

291
292 The accumulative percentage of disposed waste till week j is calculated using formula (4):

$$293 \quad AP_i\% = \sum_1^j r_i * 100\% \quad (4)$$

294 where AP_i is the accumulative percentage of waste generated till week i , p_i is the percentage of
295 total waste generation at week i .

296
297 Similar to WGF, applying the $T_i\%$ as the x-axis and the $AP_i\%$ as the y-axis, the accumulative
298 waste generation flow (accumulative WGF) can be drawn. Different from WGF, an
299 accumulative WGF curve brings a straightforward illustration of how waste is accumulated
300 during the project developing process, serving as another important indicator of construction
301 waste management efficiency.

302 303 *Plotting weekly and accumulative WGF curves*

304 With the standardized time ($T_i\%$) as the x-axis, percentage of total waste generation ($r_i\%$) as
305 the y-axis, the weekly WGF curves can be drawn. Replacing the percentage of total waste
306 generation ($r_i\%$) with accumulative percentage of total waste generation ($AP_i\%$), the
307 accumulative WGF curves can be portrayed.

308 309 *Scaling up*

310 The same approaches of data extraction, time and waste generation weight standardization and
311 WGF curve plotting are applied to all other projects. The data processing, analysis and scaling
312 up take place in Microsoft Excel 2016. Projects of the same type (there are a total of six types
313 of projects, e.g., public demolition, private demolition, public foundation, private foundation,
314 public new building, private new building) are arranged together for the representative WGF
315 curves for comparisons and analyses.

316 317 *Averaging waste generation proportion with interpolation*

318 To compare between the public and private clients, a representative curve for one type of
319 projects is desired. Since different projects have different time scale, all the project should be
320 averaged on a unified time scale. A 0.1% of scale is used to standardize the timeline to keep

321 the resolution of the data better. For the waste generation percentage at every time point, the
322 percentage covered a period is averaged under the assumption that the waste is evenly
323 generated for that period and then be disposed of once. For example, project B disposes of 15%
324 of waste at 25% of time after its last disposal at 15% of time, then, for every 0.1% of time
325 between 15% and 25% of the completion time, the averaged waste generation proportion is
326 $15\% / (25\% - 15\%) * 0.1\% = 0.1\%$. By this averaging calculation, every project has its averaged
327 waste generation percentage at a 0.1% time point. Averaging the averaged waste generation
328 percentage of same type projects at every time points, an average waste generation percentage
329 value as their representative can be generated. For the accumulative WGF, linear interpolation
330 is used to fill the vacant value for the 0.1% of time scale representation. After acquiring all the
331 points on a 0.1% of time basis, the representative WGF and accumulative WGF curves can be
332 plotted.

333 *Calculating the difference between the public and private sectors*

334 After the interpolation, every type of project has a representative set of WGF and accumulative
335 WGF data set on the 0.1% time scale. Therefore, their difference can be calculated directly by
336 doing minus between the two sectors. Since this paper aims at investigating if private sector
337 outlaws, metrics of the private sector is used to minus that of the public sector and keep their
338 difference as the indicator of their waste management efficiency disparity.

340 *Comparison*

341 After processing the data, the results of public projects and private projects are compared
342 separately. Firstly, the overall waste generation rate (WGR) of all projects will be calculated
343 and displayed to investigate whether their WGRs have a significant difference. Afterward, the
344 WGFs and AWGFs of the public and private sector of demolition projects, foundation projects,
345 and new building projects will be compared one by one to examine their waste management
346 performance disparity.

347 *Microsoft Excel 2016* is used to process and analyze all the data and further draw the figures.
348 It is capable of dealing with big data when the data is structured and can be converted in
349 different worksheets even for a 1-million-row dataset. Our 12,828-row data is structured and
350 can be converted between different worksheets and calculated easily using simple formulas.
351 By drawing the WGFs and accumulative S-curves, how the waste is generated during the whole
352 process of the project can be delineated.

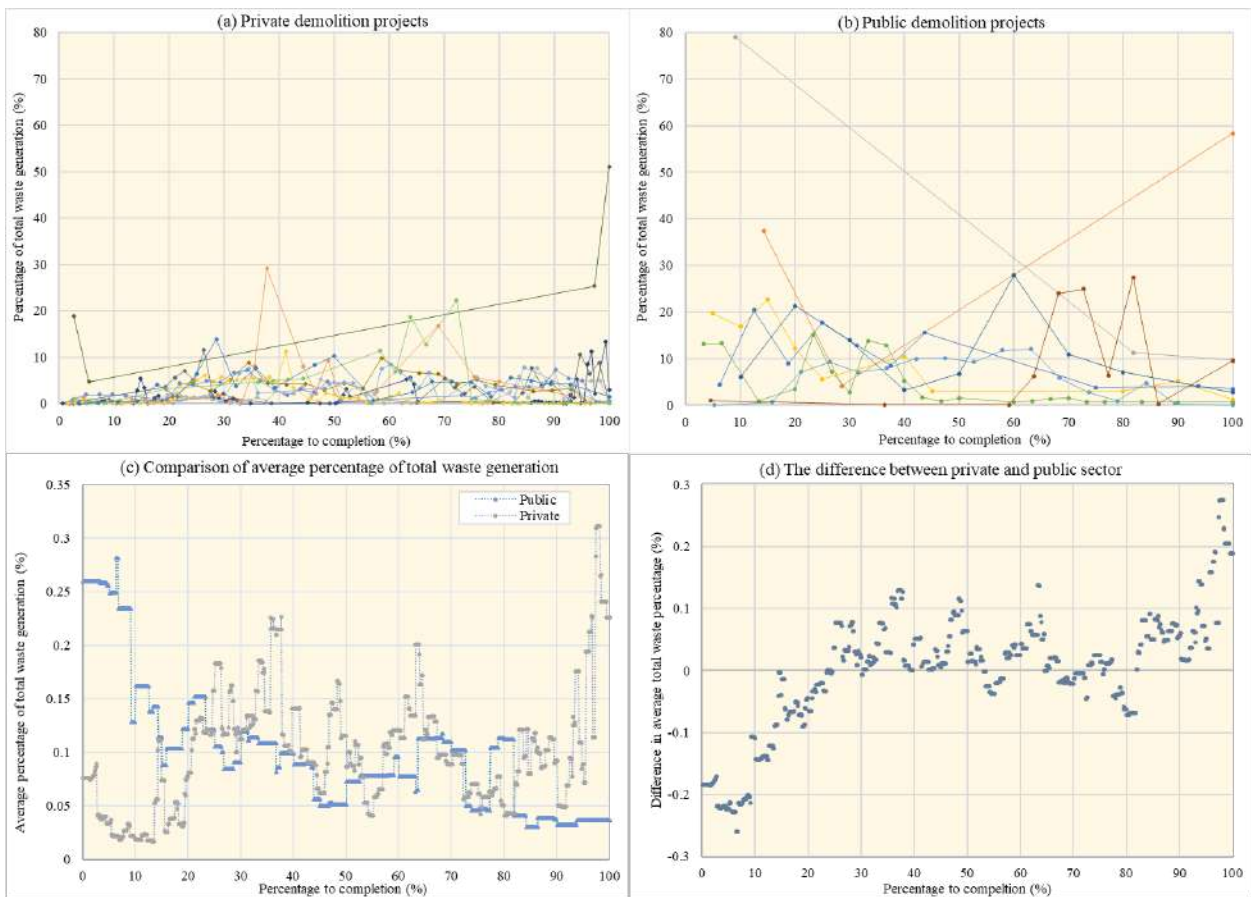
353 **Analyses, results, and findings**

354 *Demolition projects*

355 Figure 3 displays the contrast between the WGFs in the 8 public and 11 private demolition
356 projects. The x-axis is the standardized time, and y-axis the standardized waste generation
357

361 percentages. Both types of projects have their own patterns as shown in Figures 3 (a) and (b):
 362 private projects have evenly distributed proportions except three outliers; public projects have
 363 more randomly distributed ratios with the change of time. By contrast, the weekly waste
 364 generation percentages in public demolition projects are generally larger than private ones and
 365 fluctuate more obviously. It is largely due to the longer duration and larger waste disposal times
 366 of most private projects than the public projects. Figure 3 (c) is the comparison of the average
 367 percentage of total waste generation between public and private demolition projects on a 0.1%-
 368 time basis. On this meticulous time scale, the averaged WGFs of both types of projects fluctuate
 369 dramatically. At the first 15% of the time, public demolition projects generate a much larger
 370 proportion of waste than the private. Afterward, they change alternatively until around 80% of
 371 the time, when private projects produce wastes at a far higher speed. These findings indicate
 372 the better control of waste generation in private projects at early and middle stages.

373
 374



375

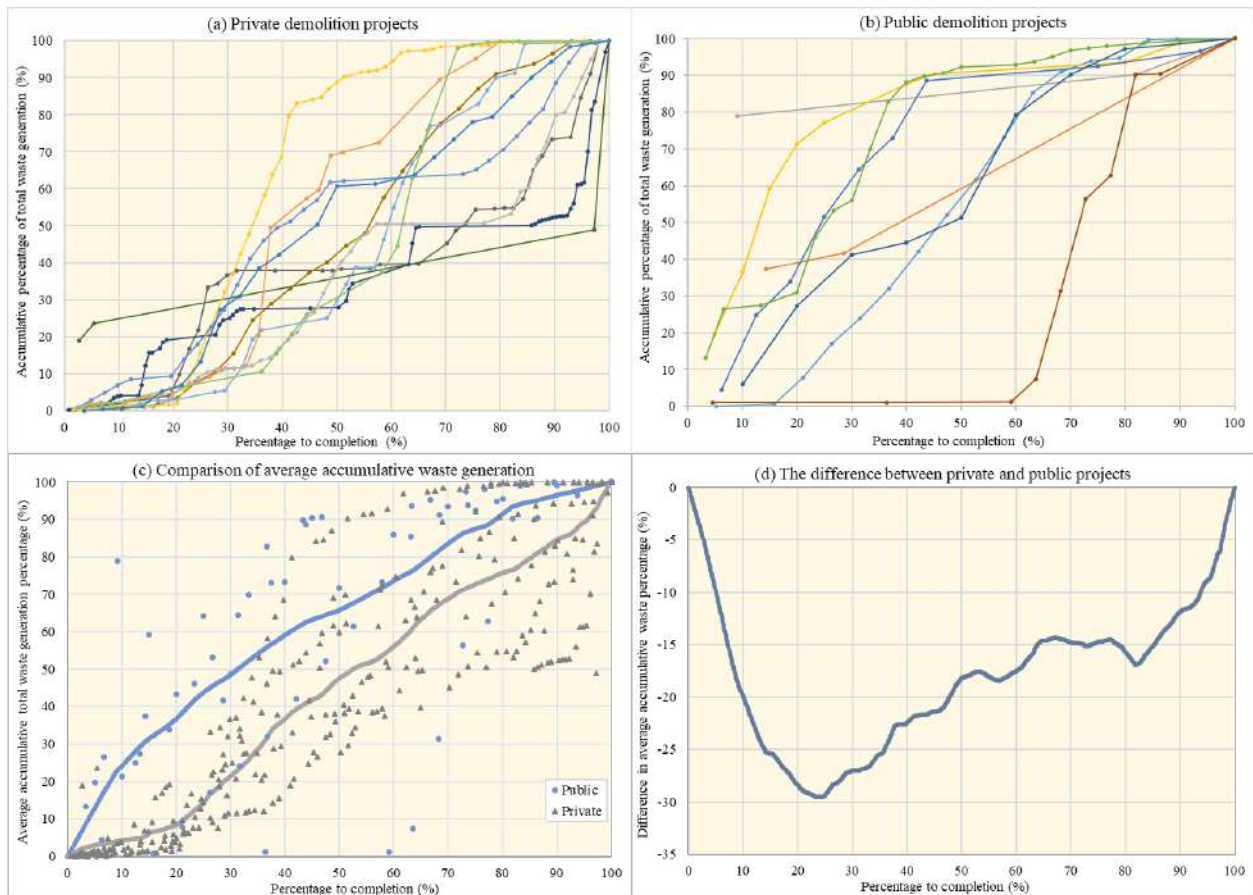
376 Figure 3. Waste generation flow of public and private demolition projects

377

378 Figure 3 (d) further compare the private and public projects by showing their difference with
 379 the changing of time. The y-axis is the difference between private and public average waste
 380 generation percentage. The difference also waves greatly. Generally, it increases rapidly at 25%
 381 of progress, then fluctuates gently to 80% of the time, and increases at high speed again in the
 382 final period. Such results indicate that private projects generate wastes in a slower way at the

383 beginning but faster at the late stage than the public projects. However, the difference is
384 actually small within the range of 0.3% of total waste. A t-test (two samples assuming equal
385 variances) on the average percentage of total waste generation proves that the two sectors
386 perform equally (mean average percentage of total waste generation for public projects is 0.001
387 and private projects 0.000999) with no significant difference ($P=0.971$). From this aspect, the
388 performance of the public and private sector in demolition waste management projects have no
389 substantial dissimilarity.

390
391 The accumulative waste generation curves of public and private demolition projects are shown
392 in Figure 4. Different from Figure 3, the y-axis of Figure 4 is the accumulative waste generation
393 percentage. It is easy to recognize their difference from Figures 4 (a) and (b): the shapes of
394 private curves are more concentrated while public projects vary greatly from each other. To
395 compare their differences, the average accumulative percentage on a 0.1% of time basis of all
396 public demolition projects and all private demolition projects are calculated separately and then
397 plotted in the same figure, see Figure 4 (c). The dots are the actual averaged accumulative waste
398 generation ratios of public and private projects at different time points. Combining the four
399 sections in Figure 3, it can be interpreted that for private projects, the accumulative waste
400 generation is less than 10% in the first 20% of the time for most projects. Afterward, the
401 difference in waste generation speed among projects varies. For public projects, the situation
402 diverges more fiercely. The two curves in Figure 4 (c) is sketched by the averaged accumulative
403 percentage of waste generated after linear interpolation on a 0.1%-time basis. It represents the
404 difference between the public and private projects: the averaged accumulative waste generation
405 of public projects is higher than that of private ones. This difference is rooted in the early stages
406 of the first 20% time and keeps until 80% of the time when the difference is narrowed. Figure
407 4 (d) illustrates the difference in a more straight-forward way. Public projects cumulate an
408 average of nearly 30% more waste than private projects in the first 20% of the time. Afterward,
409 their difference narrowed in ups and downs. These features also echo the findings of WGF
410 comparisons.



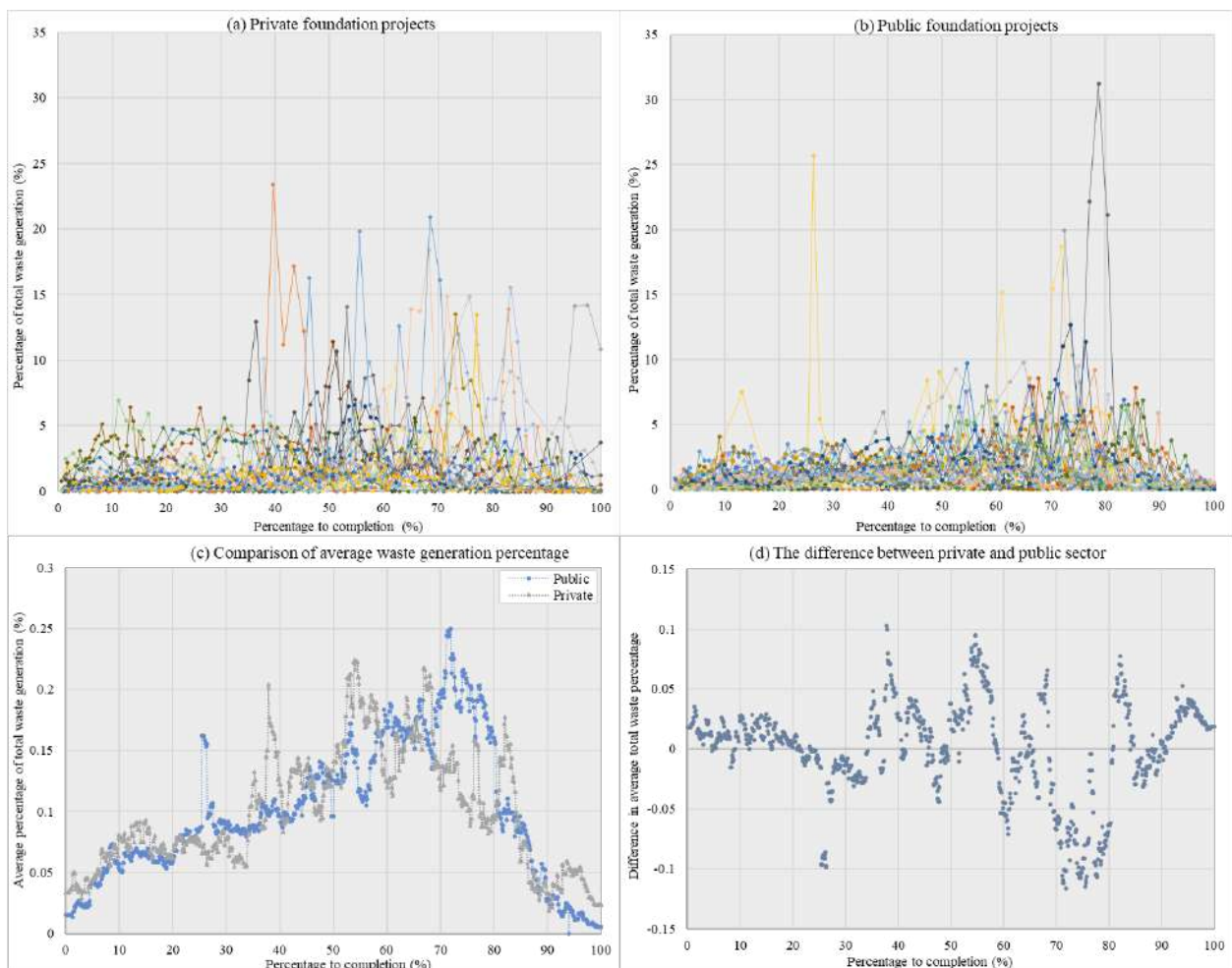
412
413 Figure 4. Comparison of accumulative waste generation flow of public and private demolition
414 projects

415
416 Interpreting the finding from the comparisons of WGF and accumulative WGF between
417 demolition projects, the answer to the question “is the private sector more effective” is that the
418 private and public sector has no significant difference in waste management performance of
419 demolition projects. Yet, the investigations into details indicate that private sector control of
420 WGF during the demolishing process is more effective than the public. The waste management
421 performance here is considered by the control of waste generation. Moderate waste generation
422 is better for the capacity planning of waste disposal facility and the transportation arrangement
423 of contractors. Speedy waste generation in a compact society like Hong Kong will cause
424 pressure to public transportation, environment, as well as the waste disposal facilities. Besides,
425 the variance among private projects is much smaller than public projects. It implies the
426 consistency in waste management performance, which is critical for waste management
427 prediction and planning.

428
429 **Foundation projects**

430 The changes in the weekly percentage of total waste generation over time in foundation projects
431 are displayed in Figure 5. Figures 5 (a) and (b) show that there are some similarities and
432 differences in the WGFs of private and public projects. For most of the recorded points, the
433 percentage of total waste generation is less than 5%. More private projects have a high waste

434 generation proportion at the middle stage than public projects. Also, the peaks of the public
 435 projects come later, one at 26% of the time and the others around 60-80% to completion. Figure
 436 5 (c) further depicts their difference, which changes moderately with some big waves. A more
 437 clear-cut result of the contrast is presented in Figure 5 (d). By and large, public/private projects
 438 exceed the other alternatively within the range of 0.1%. Private projects outpace public ones
 439 on the waste generation speed for a larger proportion of time at early, middle, and later stages.
 440 Even though, a t-test on the average percentage of total waste generation by assuming a null
 441 hypothesis that the two means are equal shows that the two sectors perform the same in
 442 foundation projects on the aspect of waste management with the p-value being 0.97. The mean
 443 percentage of total waste generation is 0.099 for public projects and 0.099 for private projects.
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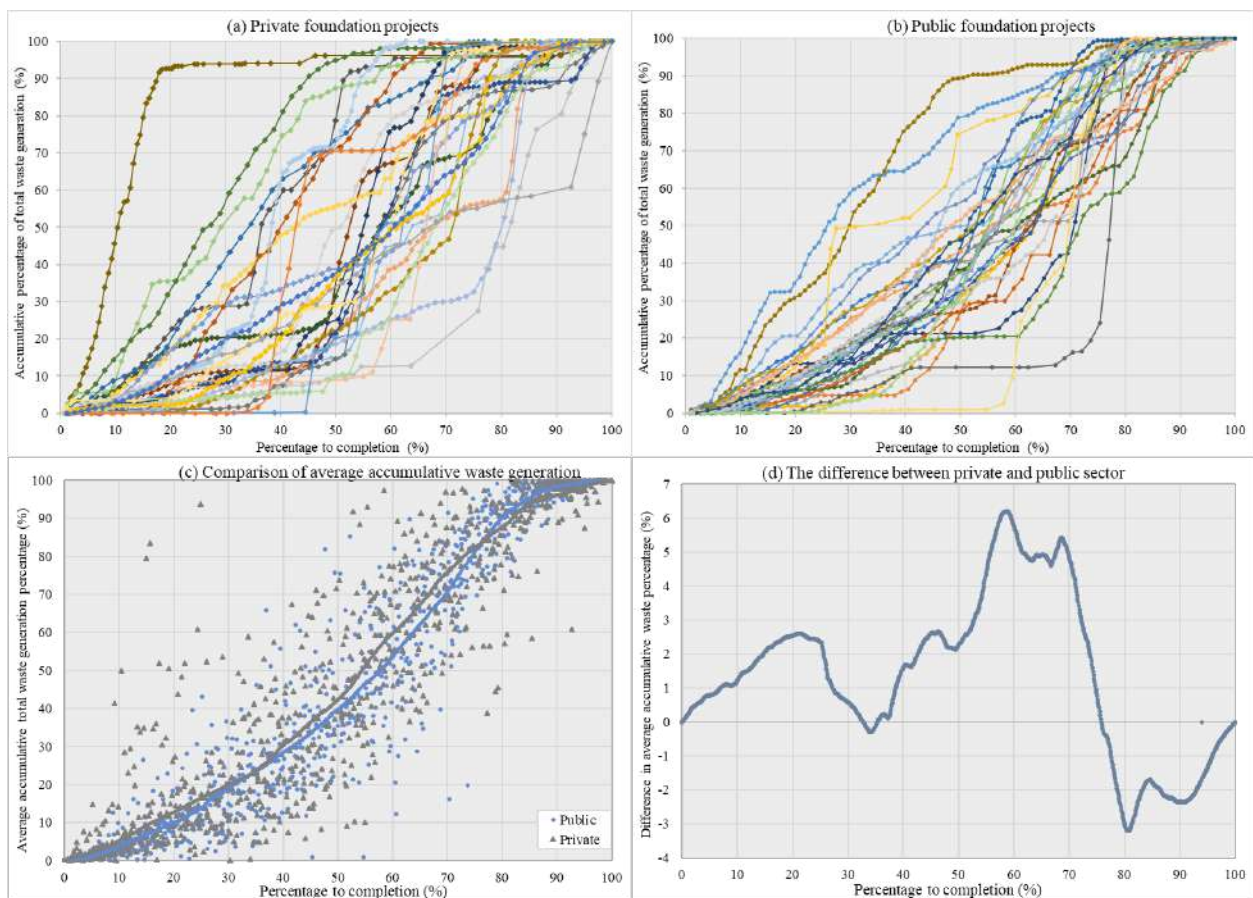
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446 Figure 5. Waste generation flow of public and private foundation projects

447

448 Figure 6 illustrates the accumulative WGFs of public and private foundation projects.
 449 Obviously, the distribution of the S-shape curves of public projects are more focused while the
 450 private ones are sparser, as shown in Figures 6 (a) and (b). This major difference implies that
 451 public projects are more standardized in waste management while the performance of waste
 452 management in private projects staggers with each other. Figure 6 (c) further compare the two
 453 sectors by contrasting their averaged accumulative percentage of total waste generation. The

454 distribution of dots demonstrates that private projects have more scarce distribution in
 455 accumulative WGF, which is also evidenced in Figures 6 (a) and (b). The interpolated average
 456 S-shape curves in Figure 6 (c) advance the comparison in a more pruned way. In general, the
 457 gap between the public and private sector in foundation projects is slight when compared with
 458 demolition projects. The accumulative percentage of total waste in private projects is larger
 459 than that of public projects for most of the time. Figure 6 (d) shows the difference in average
 460 accumulative waste generation percentage in a simple and direct way. At the peak of the curve,
 461 private projects cumulate an average of 6% more waste at the middle stage. Public projects
 462 reverse the trend at 75% of the time and reach the biggest difference of 3% accumulative waste
 463 at 80% of the time.
 464



465
 466 Figure 6. Accumulative waste generation flow of public and private foundation projects
 467

468 Closer probes into the comparison between the projects of the public and private sectors do
 469 disclose some differences. The averaged WGFs and their difference curve illustrate the
 470 divergences in waste generation efficiency between the two sectors: the public sector has a
 471 smaller extent of variation than the private sector. The accumulative WGFs corroborate this
 472 inference. Though averagely, the accumulative WGFs has a minor difference, the variance
 473 among private projects is greater. On this aspect, the public sector has more consistent waste
 474 management performance to a small degree.
 475

New building projects

The WGFs of public and private new building projects are illustrated in Figure 7. Figures 7 (a) and (b) demonstrate that the waste generated in one week is less than 5% of the corresponding total waste amount for most projects, be they public or private. Four private projects produce more than 15% of total waste in one week, but their peaks happen at different time. No public projects generate more than 10% of waste in one week, and their peaks also occur at different time ranges. The difference is that all peaks happen before 72% of public projects progress while randomly in private ones. The reason behind can be the standardization of public projects which will result with controlled waste generation at early and middle stages while the divergent standards among private projects themselves lead to the waste management efficiency disparity. To better detect the difference between public and private projects, the averaged WGFs are plotted on a 0.1% basis, as shown in Figure 7 (c). At the very early and late stages, the difference is obvious: private projects generate waste quicker than their public counterparts. But at middle stages, public projects generate more waste than private ones. Therefore, Figure 7 (d) gives a clear and direct way to expand the comparison. Although for most of the time, their difference fluctuates alternatively, it keeps within 0.1% for most of the time, which is very slight. It is hard to compare the overall performance of the two-sector; thus, a t-test on the average percentage of total waste generation is conducted. It indicates that the two sectors have no significant difference in the average waste generation speed with a two-tailed p-value of 0.892. Their means of average waste generation percentages are 0.998 and 0.996, for public and private sector respectively.

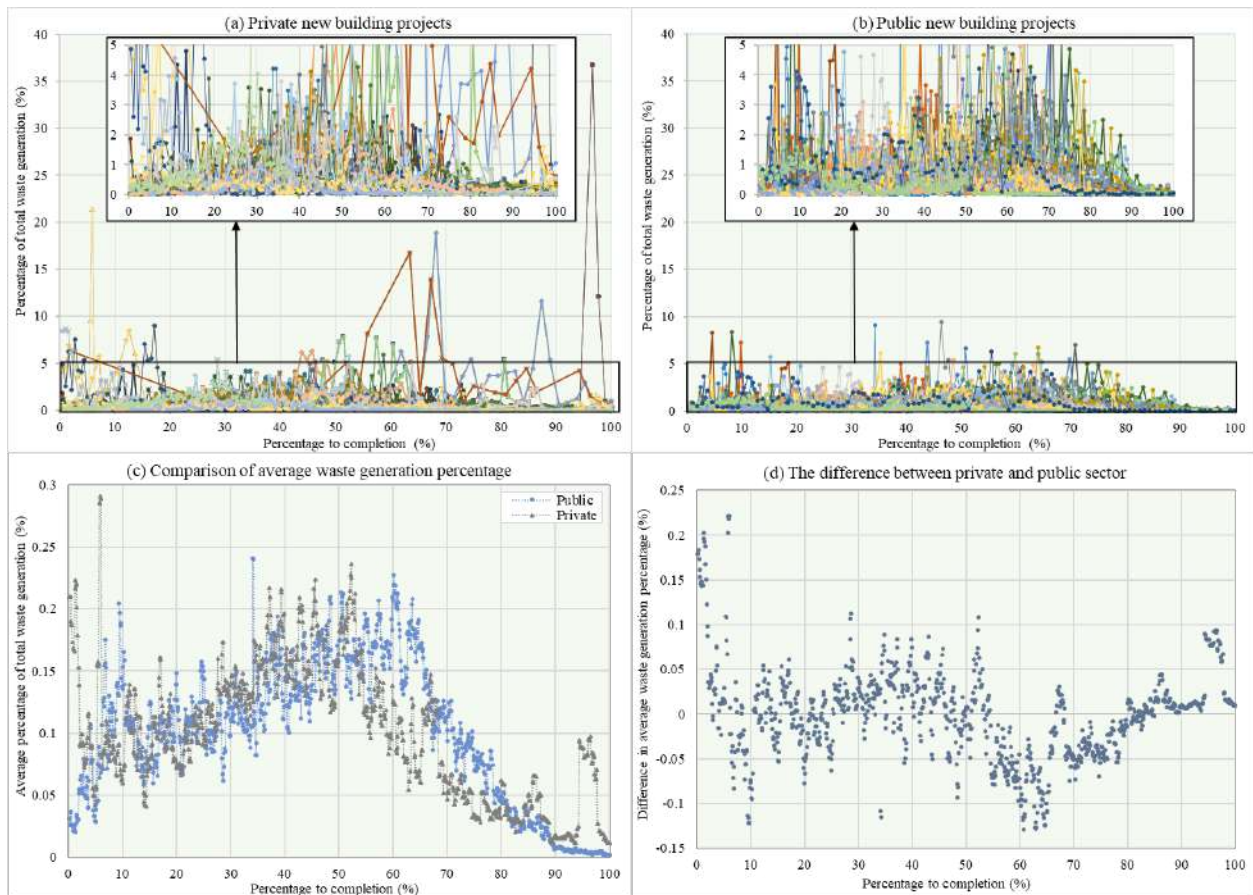
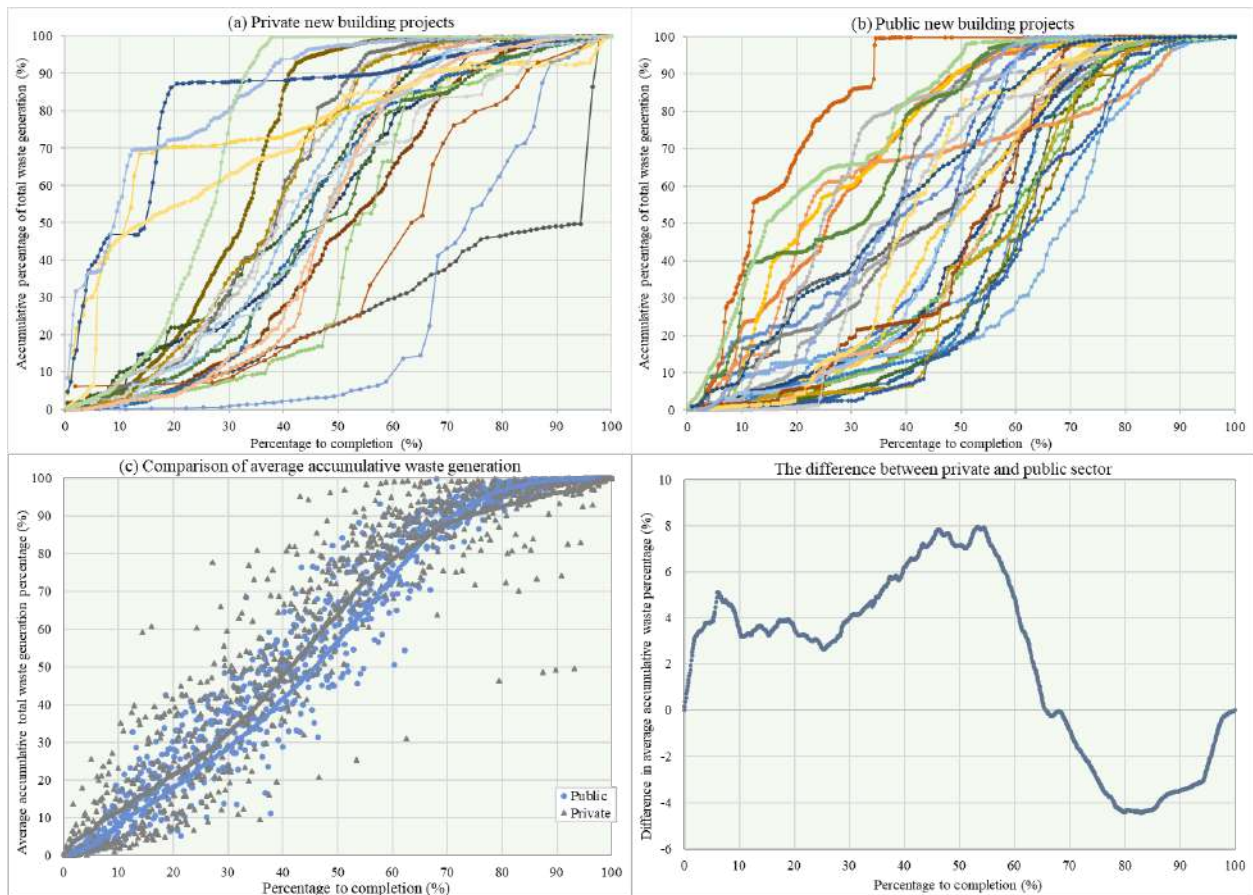


Figure 7. Waste generation flow of public and private new building projects

The accumulative waste generation curves of public and private new building projects are demonstrated in Figure 8. Similar to foundation projects, the distribution of the S-shape curves of public projects are more concentrated while the private ones are more diverse, as shown in Figures 8 (a) and (b). There are eight outliers in the private projects, among which five generates waste very quickly at early stages (more than 65% of waste at 40% of the time) and the other three very slowly at early stages (more than 65% of the time to produce 40% waste). Figure 8 (c) presents a more direct comparison between the public and private sectors. It is observable that private projects have more outlier points while public ones are more compact. This difference might be rooted in the promotion of waste management index (WMI) practice in public new building projects. The curves show that for the first half of the time, the accumulative waste proportion of private projects is slightly larger than public projects at the same progress. The difference between the two sectors keeps at an average of around 5%, as shown in Figure 8 (d), which is moderate.



515 Figure 8. Accumulative waste generation flow of public and private new building projects

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518 Interpreted from Figures 7 and 8, it is safe to draw a conclusion that the overall efficiency
519 difference between public and private sector is not significant although there are indeed some
520 slight differences during the process. For private projects, the performance between individual
521 projects varies more largely than public ones. However, the findings from WGFs and
522 accumulative WGFs is different from the results of the WGR comparison. This can be
523 interpreted that although public projects generate waste more consistently between different
524 projects, their overall waste management efficiency is still lagging compared to their private
525 counterparts. In other word, private projects generate less waste on the same contract sum.
526 Although the performance between one project and another may have a bigger difference, there
527 are some private new building projects that are highly efficient in construction waste
528 management. Therefore, it is necessary to enhance the regulation of private projects from the
529 waste management perspective to enhance the overall efficiency. The successful waste
530 management practices of public projects and some outstanding private projects should be
531 promoted.

532 Discussions

533
534 Both the private and public sectors sponsor construction project development. In Hong Kong,
535 a city evaluated as the world's freest economy for 25 consecutive years in a row (Sum, 2019),
536 building construction clients use the same pool of contractors and sub-contractors, which are

537 all private companies freely competing in the market. Under the same set of environmental
538 regulations and construction codes, they should perform the same without an apparent disparity
539 between their CWM efficiency measured by WGRs. However, the big data analytics in this
540 study discovered that CWM efficiency varies with sectors and different types of construction
541 projects they undertake. According to the comparisons of CWM efficiency of three categories
542 of projects, the public and private sectors have no statistically significant difference in
543 demolition and foundation projects. Surprisingly, in the sampled new building projects, the
544 average WGR of the public sector is significantly larger than the private sector. In summary,
545 the private sector is not consistently more efficient than public, or vice versa, in managing
546 construction waste. Such results reject the significant efficiency disparity between the public
547 and private sectors concerning construction waste management.

548
549 Although under the same set of environmental regulations and construction codes, it is widely
550 considered that the public sector clients are subject to more stringent social scrutiny, therefore
551 tend to care more about CWM efficiency. The public sector clients are promoting waste
552 management index (WMI) practice in Hong Kong, which might be the reason that public
553 foundation and new building projects have much similar WGF and accumulative WGF
554 between each other while the divergence is large in private ones. On the private side, there are
555 often casual remarks in Hong Kong that the private sector outlaws in managing their
556 construction projects. However, the private sector cares more about the savings of material
557 consumption and disposal fees; therefore, better managing construction waste. All these joints
558 forces may explain the finding that no sector is consistently better than another. Nevertheless,
559 CWM performance during the whole process differs observably from one project to another.
560 Although the companies are under the same market conditions, CWM efficiency is determined
561 not only by the client-contractor dyadic relationship, but also by the companies' environmental
562 awareness, corporate social responsibility, and management discretion.

563
564 The big data sourced in this study did provide a fuller picture to examine CWM performance,
565 and their detailed WGFs. The big data analytics helped to gain more statistical confidence,
566 although the analyses did not tell which sector is more efficient than another. Big data opens a
567 window to address the moot question of whether the private sector is more efficient than the
568 public sector, but such data can never be too big to examine a complex system such as CWM.
569 In this study, the data of 132 selected projects, including 70 public projects and 62 private
570 projects, of three categories, i.e., demolition, foundation, and new building are sourced from a
571 big data set. It is bigger than any other data sets one can ever see in the literature. Yet, it appears
572 thin in ascertaining the efficiency disparity between the public and private sectors. Researchers
573 are encouraged to exploit passive, unintentionally left-over big data to examine the classic
574 inquiry of efficiency disparity between the public and private sectors. Even in further future,
575 meta-analyses based on the big-data enabled analyses can bring the inquiry into a conclusion.

577 The methods and findings of this research will also supplement the body of knowledge by
578 adding a new method and perspective to the examination of public and private sectoral
579 efficiency disparity, as well as an answer to the investigation. It is hard to decide which one is
580 better, the better way is to analyze which sector is doing greater in which specific aspect and
581 improve the other sector using the experience and knowledge learnt from the empirical practice
582 comparison. For the field of practice, the answers obtained from empirical case analyses
583 indicate that the private sector needs more improvement than its public counterpart in the aspect
584 of CWM. Private companies should pay more attention to their social environmental
585 responsibilities and learn from the public counterparts. Although the public sector is doing
586 better at a general scale, some private companies are actually doing better, whose experiences
587 are worthy learning by the public sector and other private companies.

588 589 **Conclusion**

590 Efficiency disparity between the public and private sectors is a fundamental scientific inquiry
591 that is closely related to grand challenges such as capitalism vs. socialism, privatization vs.
592 nationalization, or public-private partnership. Over the past decades, it has attracted numerous
593 studies, but the line of inquiry is largely inconclusive. This research aimed to ascertain the
594 difference of construction waste management efficiency between the public and private
595 projects, with a view to understanding the inquiry in a smaller setting and considering big data
596 analytics as a promising solution to a moot question. In this study, the data of 132 selected
597 projects, including 70 public projects and 62 private projects distributed in three categories of
598 projects, including demolition, foundation, and new buildings are extracted and processed for
599 the statistical and visual comparisons. The investigations looking at the waste generation flow
600 (WGF) and accumulative WGF found that there is no statistical significance found in all three
601 types of projects, i.e., demolition, foundation, and new building projects. A comfortable
602 conclusion is that no sector is consistently more efficient than another in managing construction
603 waste, although closer inspections detect the better performance in private demolition projects,
604 public foundation and new building projects. Additionally, CWM performance ranges radically
605 from one project to another, especially in private projects, which suggests the standardization
606 of waste management practice in private projects as a feasible short-term strategy for CWM
607 improvement. Further, improving CWM performance perhaps can be pursued through
608 contractors' individual practices rather than through the project ownership only.

609
610 Big data demonstrated its power to paint a fuller picture of CWM in different projects. It allows
611 us to obtain more statistical confidence, and to probe into detailed WGFs in individual projects.
612 However, the data is still not big enough to allow the examination of many relevant aspects,
613 e.g., contractors' internal CWM policies and onsite practices. By having the data with sufficient
614 variety, it might be possible to ultimately answer the question of efficiency disparity between
615 the public and private sectors. This big, but still not various enough data forms the major
616 limitation of this research.

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