

1 **Multi-level Multi-criteria Analysis of Alternative Fuels for Waste Collection** 2 **Vehicles in the United States**

3 Mousa Maimoun¹, Kaveh Madani², Debra Reinhart³

4 ¹Joyce Engineering, Inc., Charlotte, North Carolina; ²Centre for Environmental Policy, Imperial College London,
5 London, U.K.; ³Department of Civil, Environmental and Construction Engineering, University of Central Florida,
6 Orlando, Florida emails: [1mousamaimoun@knights.ucf.edu](mailto:mousamaimoun@knights.ucf.edu); [2k.madani@imperial.ac.uk](mailto:k.madani@imperial.ac.uk); [3debra.reinhart@ucf.edu](mailto:debra.reinhart@ucf.edu)

7 8 ***Abstract***

9 Historically, the U.S. waste collection fleet was dominated by diesel-fueled waste collection
10 vehicles (WCVs); the growing need for sustainable waste collection has urged decision makers
11 to incorporate economically efficient alternative fuels, while mitigating environmental impacts.
12 The pros and cons of alternative fuels complicate the decisions making process, calling for a
13 comprehensive study that assesses the multiple factors involved. Multi-criteria decision analysis
14 (MCDA) methods allow decision makers to select the best alternatives with respect to selection
15 criteria. In this study, two MCDA methods, Technique for Order Preference by Similarity to
16 Ideal Solution (TOPSIS) and Simple Additive Weighting (SAW), were used to rank fuel
17 alternatives for the U.S. waste collection industry with respect to a multi-level environmental and
18 financial decision matrix. The environmental criteria consisted of life-cycle emissions, tail-pipe
19 emissions, water footprint (WFP), and power density, while the financial criteria comprised of
20 vehicle cost, fuel price, fuel price stability, and fueling station availability. The overall analysis
21 showed that conventional diesel is still the best option, followed by hydraulic-hybrid WCVs,
22 landfill gas (LFG) sourced natural gas, fossil natural gas, and biodiesel. The elimination of the
23 WFP and power density criteria from the environmental criteria ranked biodiesel 100 (BD100) as
24 an environmentally better alternative compared to other fossil fuels (diesel and natural gas). This
25 result showed that considering the WFP and power density as environmental criteria can make a
26 difference in the decision process. The elimination of the fueling station and fuel price stability
27 criteria from the decision matrix ranked fossil natural gas second after LFG-sourced natural gas.

28 This scenario was found to represent the status quo of the waste collection industry. A sensitivity
29 analysis for the status quo scenario showed the overall ranking of diesel and fossil natural gas to
30 be more sensitive to changing fuel prices as compared to other alternatives.

31 **Keywords:** alternative fuels, waste collection, decisions making, multi-criteria analysis

32 **1. Introduction**

33 **1.1 Initial Position**

34 The waste collection industry is driven by the need to reduce costs and emissions while
35 increasing operation efficiency. These challenges encourage the collection industry to explore
36 alternative fuel technologies including compressed natural gas (CNG); liquefied natural gas
37 (LNG); biodiesel (B20, B100), and hydraulic-hybrid (an alternative to conventional diesel trucks,
38 where trucks are able to recapture, store, and reuse braking energy, Bender et al., 2014).

39 Up to 2010, diesel-fueled waste collection vehicles (WCVs) were the backbone of the U.S. waste
40 collection industry with less than one percent of WCVs using alternative fuel (Rogoff et al.,
41 2010). The recent relatively low prices of natural gas compared to high diesel prices have
42 incentivized the industry to consider natural gas as an alternative fuel for their fleets. In 2012,
43 Waste Management Inc., based in Houston, Texas, and a leading provider of comprehensive
44 waste management services in North America, operated the largest natural gas collection
45 vehicles fleet in North America with nearly 1,700 CNG and LNG vehicles. In the next five years,
46 it is anticipated that 80% of the Waste Management new trucks purchased will be fueled by
47 natural gas. The company added 13 CNG fueling stations in the first-half of 2012, which brought
48 their total to 31. Moreover, Waste Management planned to construct another 17 stations by the
49 end of 2012 (Waste Management Inc., 2012). The second major waste hauler in the U.S.,
50 Republic Services, with currently more than 1,000 vehicles running on alternative fuels, plans to

51 add 3,100 natural gas and other alternative-fueled WCVs by the end of 2015 (Republic Services,
52 2012). In 2012, WCV and transfer vehicles accounted for 11 percent of the total U.S. natural gas
53 vehicles (NGVAMERICA, 2012). In contrast, diesel fuel purchases were estimated to consume
54 7.5% of the industry revenues in 2012 (Smith, 2012).

55 Undoubtedly, fuel cost has been the driving factor for the waste industry. A
56 comprehensive decision matrix that considers other factors such as changing policies, future fuel
57 prices, and uncertainty in fuel performance data, has not been developed. In the last three
58 decades, the selection scheme for alternative fuels and energies has changed from a single-
59 criterion cost-based assessment, to a multi-criteria analysis that considers environmental, social,
60 operational, and even political factors (Pohekar and Ramachandran, 2004; Cavallaro, 2005;
61 Wang et al., 2009; Linkov and Moberg , 2012; Read et al., 2013; Hadian and Madani, 2015).
62 A multi-criteria analysis normally involve trade-offs among alternatives. Multi-criteria decision
63 analysis (MCDA) methods allow stakeholders to select an optimal solution for complex
64 problems involving such tradeoffs (Josimovic et al., 2015). The use of MCDA methods allows
65 decision makers to systematically select the best alternative with respect to selection criteria,
66 while understanding the tradeoffs that occur in selecting different alternatives (Linkov and
67 Moberg, 2012).

68 **1.2 Goal and Objectives**

69 The goal of this paper is to determine if the waste collection industry is moving in the right
70 direction toward a more environmental-friendly alternative at a reasonable financial cost. This is
71 done through application of MCDA methods to select the best alternative fuel for the waste
72 collection industry, and to determine trade-offs among environmental and economic aspects of
73 alternatives fuels. MCDA methods have been used to rank alternative fueled buses for public

74 transportation (Tzeng et al., 2005), alternative transportation fuels (Mohamadabadi et al., 2009),
75 electricity generation alternatives (Cristóbal, 2011), municipal solid waste management
76 alternatives (Herva and Roca, 2013), and landfill sites (Şener et al., 2006).

77 In this study, MCDA methods were used to rank alternative fuels for WCVs using a
78 multi-level multi-criteria decision analysis framework (Read et al., 2013) that incorporates
79 environmental and financial criteria, providing insights for better decision-making by the waste
80 industry. Sensitivity analysis will be performed to determine the robustness of fuel rankings to
81 changing policies, selection criteria, and fuel performance data. This will help determine the
82 long-term consequences of selecting a certain fuel for the industry. The initial position of the
83 waste collection industry will be compared to the results of this study.

84 The rest of the paper is outlined as follows. Section 2 presents the MCDA methods and
85 data used to rank alternative fuels. Section 3 ranks alternative fuels for waste collection vehicles.
86 Finally, Section 4 presents the conclusions and recommendations for the waste collection industry.

87 **2. Methods**

88 Alternative fuels were identified based on a literature review. Fuel selection criteria that consider
89 environmental and financial factors were established. The fuel performance data (a quantitative
90 measure of the fuel performance with respect to each selection criteria) were obtained from the
91 literature. Finally, two MCDA methods, Simple Additive Weighting (SAW) (Churchman and
92 Ackoff, 1954) and Technique for Order Preference by Similarity to Ideal Solution (TOPSIS)
93 (Hwang and Yoon, 1981), were used to rank fuel alternatives for the waste collection industry
94 using the multi-level environmental and multi-criteria approach (Read et al., 2013). The selection
95 of these two methods was based on their ability to handle multi-attribute decision making
96 problems. The following sections provide more details about the decision analysis process.

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2.1 Fuel Alternatives for Waste Collection Vehicles

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99 Nine different fuels could be considered for WCVs; gasoline, diesel, natural gas (Gordon et al.,
100 2003), biodiesel (López et al., 2009), liquefied petroleum gas, hydraulic-hybrid (a hydraulic
101 hybrid WCV consists of a typical diesel-fueled WCV components - a diesel engine, a clutch, a
102 transmission system, a differential, and wheels, combined with the hydraulic system elements -
103 an axial piston pump, a clutch, a simple transmission system, used to recapture, store, and reuse
104 braking energy (Bender et al.,2013; Bender et al., 2014;. de Oliveira et al., 2014), hybrid diesel-
105 electric (transfers conventional chassis WCVs into dual power options specifically designed for
106 collection and transportation of the waste, thus reduces tailpipe emissions within cities and
107 neighborhoods, FAUN, 2015), hydrogen gas (FAUN, 2011), ethanol E85, and dimethyl ether
108 (DME) (Tsuchiya and Sato, 2006). Only four fuel technologies were commercially available for
109 WCVs - diesel, natural gas, biodiesel, and hydraulic-hybrid. Diesel-fueled WCVs can operate on
110 fossil diesel or biodiesel (BD) blends (BD20 and BD 100), but may require engine modifications
111 when using biodiesel blends (U.S. EIA, 2015). BD100 is made of 100% biodiesel, while BD20 is
112 a blend of 20% biodiesel and 80% fossil diesel (U.S. EIA, 2015a). In the U.S., biodiesel is
113 produced from a diverse biomass feedstock, led by soybean oil which accounted for more than
114 50% in 2013 (U.S. EIA, 2015b). In this study, two sources of biodiesel were investigated;
115 soybean as a primary source of biodiesel in the US, and algaculture as an alternative future
116 source. Natural gas WCVs can operate either using CNG or LNG, which can be obtained from a
117 fossil or biogenic source. In this study, fossil sources were categorized as North American or
118 Non-North-American. Landfill gas (LFG) sourced natural gas was the only biogenic natural gas
119 source considered in this study. LFG is comprised of mainly methane (50-60%) and carbon
120 dioxide (40-40%) (Shin et al., 2005; U.S. EPA, 2012). It also consists of hundreds of other

121 compounds at lower concentrations such as oxygen, nitrogen, sulfur compounds, water vapor
122 and organic compounds (U.S. EPA, 2000; Shin et al., 2005). In order to use LFG as an
123 alternative vehicular fuel, LFG should be converted to pipeline quality natural gas with high
124 BTU content, through the separation of methane from carbon dioxide and other constituents
125 (Hesson, 2008; U.S. EPA, 2000).

126 Accordingly, twelve alternative fuels or fuel blends were considered for the WCVs in the
127 U.S. based on fuel type and source; (1) diesel, (2) CNG (North American), (3) CNG (Non-North
128 American), (4) LNG (North American), (5) LNG (Non-North American), (6) hydraulic-hybrid,
129 (7) CNG (LFG sourced), (8) LNG (LFG sourced), (9) BD20 (Algaculture), (10) BD20
130 (soybean), (11) BD100 (Algaculture), and (12) BD100 (soybean).

131 **2.2 Fuel Evaluation Criteria**

132 First, a multi-level fuel selection criteria matrix that considers environmental and financial
133 factors was established (Figure 1). The upper level criteria were then broken down into sub-
134 criterion categories, e.g. tail-pipe emissions (second level environmental criterion) of WCVs
135 were evaluated based on carbon monoxide, carbon dioxide, nitrogen oxides, particulate matters,
136 and total hydrocarbons emissions. Fuel performance data were collected for each alternative with
137 respect to the sub-criterion category, e.g., fuel performance data were used for carbon monoxide,
138 carbon dioxide, nitrogen oxides, particulate matters, and total hydrocarbons emissions (level 3).

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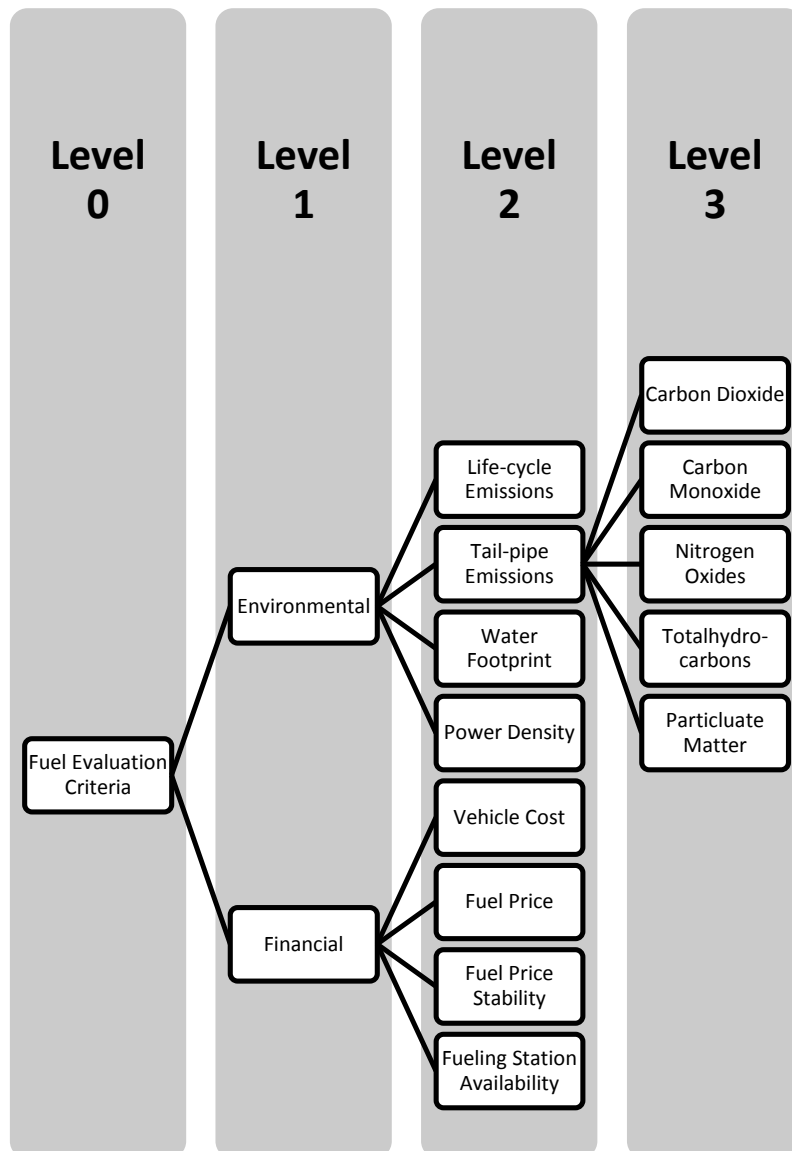


Figure 1: Multi-level multi-criteria decision making matrix.

2.2.1 Environmental Criteria

Four environmental criteria were considered in this study: life-cycle emissions of alternative fuels and fuel blends, tail-pipe emissions of alternative fuel WCVs, water footprint (WFP), and power density of alternative fuel and fuel blends.

148 **2.2.1.1 Life-cycle Emissions of Alternative Fuels**

149 Life-cycle emissions of alternative fuels and fuel blends had been calculated by Maimoun et al.
150 (2013) using the Greenhouse gases, Regulated Emissions, and Energy use in Transportation
151 (GREET) model provided by Argonne National Laboratory (U.S. DOE, 2012a). Life-cycle
152 emissions of alternative fuels and blends represent the total equivalent of greenhouse gas
153 emissions produced during the entire life-cycle of the fuel (fuel production emissions, fuel
154 transportation emissions and tail-pipe emissions at the point of use). The life-cycle emissions
155 associated with diesel, CNG (North American), CNG (Non-North American), LNG (North
156 American), LNG (Non-North American), hydraulic-hybrid, LNG (LFG sourced), CNG (LFG
157 sourced), BD100 (Algaculture), BD20 (Algaculture), BD100 (soybean), and BD20 (soybean)
158 were estimated at 2.85, 3.01, 3.27, 3.14, 3.39, 2.33, 0.62, 0.5, 1.4, 2.52, 0.71 and 2.38 kg CO_{2eq}
159 per collection vehicle kilometer travel (CVkmT), respectively (Maimoun et al., 2013). It should
160 be noted that the fuel consumption and associated emissions are expected to vary significantly
161 depending on vehicle age, driving cycle, weather conditions, traffic pattern, terrain, and other
162 factors. For the purpose of this analysis, average literature values were assumed to be sufficient
163 to conduct the analysis. Moreover, changes in driving cycle and other factors were assumed to
164 the have the same influence on all alternative fuels.

165 **2.2.1.2 Tail-pipe Emissions of Alternative Fuel WCVs**

166 Tail-pipe emissions of WCVs include carbon dioxide (CO₂), carbon monoxide (CO), nitrogen
167 oxides (NO_x), total hydrocarbons (THC) and particulate matter (PM). Tail-pipe emissions for
168 conventional diesel-fueled WCVs were measured by Farzaneh et al. (2009) using two portable
169 emissions measurement systems (PEMS). Emissions from conventional diesel-fueled WCVs
170 were investigated under four different operation modes including (1) urban driving, (2) trash

171 collection, (3) freeway driving, and (4) landfill activities (Farzaneh et al., 2009). For this study, a
172 weighted average was calculated for each pollutant using the average emission factor associated
173 with each driving mode and the fraction of the driving mode with respect to the overall route.
174 The average tail-pipe emissions from conventional diesel-fueled WCVs were estimated to be 2.8
175 kg/km, 17.1 g/km, 17.1 g/km, 0.6 g/km, and 0.06 g/km for CO₂, CO, NO_x, THC, and PM,
176 respectively.

177 A study by Texas Transportation Institute (2009) compared the tail-pipe emissions of CNG
178 fueled WCVs relative to conventional diesel vehicle, e.g. the tail-pipe NO_x emissions of CNG
179 vehicles were found to be 96% less than conventional diesel WCVs (Table 1). Tail-pipe emissions
180 for LNG were assumed to be equal to CNG based on their identical chemical properties. According
181 to the U.S. Environmental Protection Agency (EPA), the use of hydraulic-hybrid diesel WCVs has a
182 potential fuel savings of up to 30%. Therefore, tail-pipe emissions from hydraulic-hybrid WCVs
183 were assumed to be 30% less than conventional diesel-fueled WCVs (Hall, 2010). de Oliveira et al.
184 (2014) also reported 15 to 25% improvement in fuel economy of heavy-duty hydraulic-hybrid WCV
185 compared to conventional diesel-fueled WCVs. Tail-pipe emissions for buses running on BD20 and
186 BD100 showed lower emissions compared to diesel buses, except for NO_x emissions (U.S. EPA,
187 2002). Relative emissions values shown in Table 1 were applied to the weighted average of the
188 conventional diesel tail-pipe emissions to estimate alternative-fueled WCVs tail-pipe emissions.

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195 Table 1: Alternative-fueled waste collection vehicle (WCV) tail-pipe emissions relative to diesel-
 196 fueled vehicles.

Fuel Category	CO ₂	CO	NO _x	THC	PM	Source	Assumption
CNG (Source: American, non-American, LFG)	-27%	+1,200%	-96%	5,700%	--	Texas Transportation Institute (2009)	
LNG (Source: American, non-American, LFG)	-27%	+1200%	-96%	5,700%	--	Texas Transportation Institute (2009)	Tail-pipe emissions from LNG are equal to CNG
hydraulic-hybrid	-30%	-30%	-30%	-30%	-30%	Hall (2010)	Hybrid waste collection vehicles with 30% fuel saving will have 30% less tail-pipe emissions
BD20 (Source: Algaculture, soybean)	--	-11%	+2%	-21%	-10%	U.S. EPA (2002)	Waste collection vehicles and heavy-duty vehicles are similar.
BD100 (Source: Algaculture, soybean)	--	-47%	+10%	-68%	-45%		

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198 2.2.1.3 Water Footprint (WFP) of Alternative Fuels and Fuel Blends

199 The WFP is a measure of both the direct and indirect use of fresh water over the entire process
 200 life cycle (Hoekstra et al., 2009). It consists of three components; blue, accounting for the
 201 consumption of surface and groundwater resources; green, referring to consumption of rainwater
 202 stored in the soil as soil moisture, normally lost through evapotranspiration; and grey, relating to
 203 water pollution and defined as the volume of freshwater that is required to dilute pollutants to
 204 meet existing water quality standards. The total WFP of any process, product, or energy source is
 205 the summation of the blue, green and grey WFPs. The total WFP associated with alternative fuels
 206 was obtained from the literature (Gerbens-Leenes et al., 2008; Singh et al., 2011), except for LFG.
 207 The WFP of LFG source vehicular fuel was not evaluated previously, so the WFP of LFG conversion
 208 to vehicular fuel was calculated and is presented in this section.

209 Currently, commercial methods available to purify LFG include: (1) physical and chemical
 210 sorption of carbon dioxide to materials and solvents, (2) gas cooling separation, and (3) membrane
 211 separation (Läntelä et al., 2012). In this study, the WFP of LFG conversion to vehicular fuel was
 212 calculated for a water scrubber with water recycling to remove carbon dioxide, as it is considered the

213 most cost effective and widely use technology for upgrading LFG to vehicular fuel, particularly when
214 wastewater is reused as an absorbent (Hunter and Oyama, 2000; Rasi et al., 2008). The process
215 consists of absorption, desorption, pumps, compressor, and drying (Läntelä et al., 2012).

216 In order to calculate the WFP of LFG conversion, it was important to set the system
217 boundaries of the process (Madani and Khatami, 2015). The function of any landfill is the disposal
218 of municipal solid waste and LFG is a byproduct of waste landfilling. According to the U.S. EPA
219 (2012), large landfills are required to collect LFG for beneficial use or flaring. As a result, the
220 system boundaries for calculating the WFP of LFG conversion to vehicular fuel excluded landfill
221 construction and operation, LFG collection, and any condensate generated in the process, and only
222 includes (1) water evaporated during the process and need to be replaced, (2) electricity consumption
223 WFP, and (3) the WFP offsets as a result of energy recovered. The functional unit used was cubic
224 meters of water per GJ of vehicular fuel produced.

225 The energy content of methane is 37,700 KJ/Nm³. Therefore, the energy recovered in
226 converting a standard cubic meter of LFG, assuming that 100% of the methane in LFG is recovered,
227 is equal to 18,900-22,600 KJ per Nm³ of LFG. The WFP of fossil natural gas is 110 L per GJ
228 (Gerbens-Leenes et al., 2008), therefore, a WFP offset between -2.1 and -2.5 L per Nm³ of LFG
229 converted is associated with energy recovery from LFG.

230 In a pilot study described by and Rasi et al. (2008) and Läntelä et al. (2012) to convert 7.41
231 Nm³/h of LFG to vehicular fuel using water scrubbers with complete water recycling, Läntelä et al.
232 (2012) estimated that about 1% of circulating water (700 l in total) was evaporated or lost during the
233 upgrade process (3–6 h) and must be replaced. Therefore, it is estimated that the process WFP is
234 approximately 0.21 L per Nm³ LFG processed. The upgrade process electricity consumption was
235 estimated by Läntelä et al., 2012 to be between 0.43-0.55 kwh/Nm³. The WFP of the US electricity
236 was estimated for the 2013 U.S. electric grid energy mix using the WFP of different energy sources

237 compiled by Hadian and Madani (2013). The overall total WFP of the U.S. electric energy mix was
238 calculated at 9 L per kWh. Therefore, it is estimated that the WFP associated with energy
239 consumption is between 3.9 to 4.95 L per Nm³ of gas.

240 The total WFP of converting LFG to vehicular fuel is estimated to be between 4.1-5.2 L per
241 Nm³ of LFG processed, while the net WFP (accounting for offset) is estimated between 1.6-3.1 L per
242 Nm³ or between 0.07-0.16 m³ per GJ of vehicular fuel. The net WFP of LFG sourced vehicular fuel
243 is impacted by the high WFP of the U.S. electric grid and relatively low WFP of fossil natural gas.
244 The WFP of LFG-sourced natural gas is comparable to the WFP of fossil natural gas and it depends
245 on the quality of LFG. Also, the process WFP calculations are based on a pilot study and it was
246 assumed that a full-scale facility will operate with similar demands to the pilot scale. This estimate
247 does not include the WFP of potential contamination or water discharge to the surroundings in case
248 of failure of the water recycling system, and initial construction material, e.g. absorption and
249 desorption columns, where no documentation was found. On average, natural gas has the lowest
250 WFP, followed by LFG-sourced natural gas, while diesel has an average to moderate WFP. The
251 production of biodiesel was found to have the highest WFP, associated with growing and processing
252 of energy crops. The WFP of fuels is presented in Table 2.

253 **2.2.1.4 Power Density of Alternative Fuels and Fuel Blends**

254 The power density was represented by Watts generated per area of land (m²). Biodiesel production,
255 either from algaeculture or soybeans, was found to have a very low power density compared to all
256 other alternative fuels. The power density associated with fuel production is listed in Table 2.

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Table 2: WFP and power density of alternative fuels and fuel blends.

	WFP			Power Density		
	M ³ /GJ	Source	Assumptions	W/m ²	Source	Assumptions
Diesel	1.06	Gerbens-Leenes et al. (2008)	WFP of diesel equals crude oil extraction and processing	10 ³ to 10 ⁴	Smil (2010)	The Power Density of Oil Field was used.
CNG (Fossil Source)	0.11	Gerbens-Leenes et al. (2008)	WFP of CNG equals natural gas extraction and processing	10 ³ to 10 ⁴		Assumed Similar to Diesel
CNG (LFG Source)	0.07-0.16	This Study		10	Amini and Reinhart (2011)	
LNG (Fossil Source)	0.11	Gerbens-Leenes et al. (2008)	Liquefaction of natural gas to LNG consumes water; so it was assumed to be at the high end of CNG WFP	10 ³ to 10 ⁴		Assumed Similar to Diesel
LNG (LFG Source)	0.07-0.16	This Study		10	Amini and Reinhart (2011)	
Biodiesel (Soybean)	383	Singh et al. (2011)	Average WFP of biodiesel	1.32x10 ⁻⁵	Pienkos (2007)	
Biodiesel (Algaculture)	<379	Singh et al. (2011)	Average WFP of biodiesel	3.3x10 ⁻⁴ to 2.75x10 ⁻³	Pienkos (2007)	

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2.2.2 Financial Criteria

265 In this study, four financial criteria were considered; vehicle cost, fuel cost, fuel price stability and
266 fueling station availability. A quantitative measure of each alternative with respect to each criteria is
267 presented in this section.

2.2.2.1 Vehicle Cost of Alternative Fuel Vehicles

269 Vehicle cost is a significant part of the capital cost that is associated with switching to an alternative
270 fuel, therefore it was considered in the selection criteria. The average vehicle cost was reported for
271 each alternative in U.S. dollars per WCV (Table 3).

2.2.2.2 Fuel Cost

273 The relatively low priced natural gas compared to diesel shaped the recent history of vehicle
274 purchases by the waste collection industry, reflecting the significance of fuel prices. In order to
275 estimate the fuel cost, the average fuel mileage was adopted from Maimoun et al. (2013). The fuel

276 mileage was used with the national average fuel price during 2012 (U.S. DOE, 2012b) to estimate the
277 fuel cost in U.S. dollars per collection vehicle kilometer of travel (CVKmT).

278 **2.2.2.3 Fuel Price Stability**

279 Fuel price stability was considered a financial criterion. The fuel price stability was measured by the
280 standard deviation of the U.S. national fuel prices during 2012. The cost of conversion of LFG to
281 vehicular fuel was assumed to be stable over the course of one year (2012).

282 **2.2.2.4 Fueling Stations Availability**

283 The limited number of CNG/LNG fueling stations forced waste haulers to invest in building new
284 stations, while gradually switching new vehicles purchases to natural gas as the price of gas
285 plummeted. This demonstrates the significance of fueling station availability to selecting
286 alternative fuels. The number of commercially available fueling stations was reported for each
287 alternative in Table 3. In the case of CNG/LNG from LFG, the number of US landfill gas to
288 vehicular fuel projects was used. In 2008, there were only 20 sites converting LFG to vehicular
289 fuel. However, there were more than 425 landfills in the US of which about 300 are used to
290 generate electricity and 110 commercial/industrial heating fuel (Hesson, 2008). This shows the
291 potential of more landfill sites that can be used to produce vehicular fuel.

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300 Table 3: Financial performance data.

	Vehicle Cost		Fuel Cost				Fuel Price Stability		Fueling Stations	
	U.S. Dollar per Collection Vehicle	Source/ Assumption	Mileage (Km per L diesel equivalent)	Unit Price		Travel Cost (U.S. Dollar per CVKMT)	Standard Deviation of price (2012)	Source/ Assumption	No of Stations in the US	Source
Diesel	160,000-200,000	Gordon et al. (2003)	1.2	\$1.09 per diesel Equivalent L	2012 National Average Price, U.S. DOE (2012b)	0.91	0.24	U.S. DOE (2012b)	128,887	U.S. DOE (2012c)
CNG	200,000-250,000	Gordon et al. (2003)	1.0	\$0.613 per diesel Equivalent L	2012 National Average Price, U.S. DOE (2012b)	0.61	0.66	U.S. DOE (2012b)	1048	U.S. DOE (2012c)
LNG	200,000-250,000	Similar to CNG	0.95	\$0.613 per diesel Equivalent L	LNG price Similar to CNG	0.65	0.66	U.S. DOE (2012b)	53	U.S. DOE (2012c)
Hydraulic Hybrid	260,000-300,000	Danna, 2011	1.5	\$1.08 per diesel Equivalent L	2012 National Average Price, U.S. DOE (2012b)	0.81	0.24	U.S. DOE (2012b)	128,887	U.S. DOE (2012c)
CNG (Source: LFG)	200,000-250,000	Inform, 2006	1.0	\$5 and 8 per MBtu (Average Price of \$6.5 per MBtu was used)	Hesson, 2008	0.22	0	The price of LFG is assumed to be constant	20	Hesson, 2008
LNG (Source: LFG)	200,000-250,000	Similar to CNG	0.95		LNG price Similar to CNG	0.24	0	The price of LFG is assumed to be constant	20	Hesson, 2008
BD100	160,000-200,000	Similar to regular diesel	1.2	\$1.26 per diesel Equivalent L	2012 National Average Price, U.S. DOE (2012b)	1.05	0.52	U.S. DOE (2012b)	621	U.S. DOE (2012c)
BD20	160,000-200,000	Similar to regular diesel	1.2	\$1.12 per diesel Equivalent L	2012 National Average Price, U.S. DOE (2012b)	0.93	0.33	U.S. DOE (2012b)	621	U.S. DOE (2012c)

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302 **2.3 MCDA methods**

303 Two MCDA methods were used to rank alternative fuels with respect to the selected criteria,
 304 SAW (Churchman and Ackoff, 1954) and TOPSIS (Hwang and Yoon, 1981). The selection of
 305 these two methods was based on their ability to handle multi-attribute decision making problems.
 306 SAW (Churchman and Ackoff, 1954) is the most widely known MCDA method and compares
 307 the weighted average of alternative performance data with respect to a selection criteria (Afshari
 308 et al., 2010). TOPSIS is based on choosing a hypothetical ideal solution; the alternative that has
 309 the shortest geometric distance from the positive ideal solution and the longest geometric
 310 distance from the negative solution is the best (optimal) solution. TOPSIS can also accommodate
 311 different criteria weights in ranking alternatives (Hwang and Yoon, 1981).

312 SAW and TOPSIS require a comparable scale for all elements in the decision matrix,
 313 therefore performance values were normalized with respect to each criterion (j). The normalized
 314 performance values were obtained for beneficial criteria (the higher the rating, the better the
 315 performance) using Equation 1 (Nguyen and Gordon-Brown, 2012).

$$316 \quad r_{ij} = \frac{x_{ij} - \min_j}{\max_j - \min_j} \quad \text{Equation 1}$$

317 where:

318 r_{ij} = Normalized value of alternative (i) with respect to criteria (j) (0-1);

319 x_{ij} = Performance value of alternative (i) with respect to criteria (j);

320 \max_j = Maximum performance value with respect to criteria (j); and

321 \min_j = Minimum performance value with respect to criteria (j).

322 For cost criteria (the smaller the rating, the better the performance), the normalized value
 323 was calculated using Equation 2 (Nguyen and Gordon-Brown, 2012).

$$324 \quad r_{ij} = \frac{\max_j - x_{ij}}{\max_j - \min_j} \quad \text{Equation 2}$$

325 **2.3.1 Simple Additive Weighting (SAW)**

326 The SAW method (Churchman and Ackoff, 1954) compares alternatives using the comparison
 327 index (SAW_j) calculated in Equation 3. The higher the index value, the better the performance.

$$328 \quad SAW_j = \sum_{j=1}^n W_j \times r_{ij} \quad \text{Equation 3}$$

329 where:

330 W_j = Entropic weight of each criterion (j)

331

332 The entropic weight (W_j) of each criterion (j) is used to determine the weight of each
 333 criterion based on the dispersion of the performance values (Chan et al., 1999). W_j of each
 334 criterion can be calculated using Equation 4 as described in Madani et al. (2014).

335 $W_j = \frac{d_j}{\sum_{j=1}^n d_j}$ Equation 4

336 where:

337 $d_j = 1 - E_j$; and

338 E_j = The entropy of normalized performances under a given criterion and can be
 339 calculated using Equation 5 as described in Madani et al. (2014).

340 $E_j = -k \sum_{i=1}^m P_{ij} \cdot \ln(P_{ij})$ Equation 5

341 where:

342 m = Total number of alternatives;

343 $k = \frac{1}{\ln(m)}$; and

344 $P_{ij} = \frac{r_{ij}}{\sum_{i=1}^m r_{ij}}$

345 **2.3.2 Technique for Order Performance by Similarity to an Ideal Solution**
 346 **(TOPSIS)**

347 The TOPSIS method (Hwang and Yoon, 1987) selects the alternative that has the minimum
 348 relative performance distance from an ideal solution. The relative distance (CL_i^+) of each
 349 alternative to the ideal solution is calculated using Equation 6 as described in Madani et al.
 350 (2014).

351 $CL_i^+ = \frac{d_i^+}{d_i^+ + d_i^-}$ Equation 6

352 where:

353

354 $d_i^+ = \left[\sum_{j=1}^n (V_{ij} - V_j^+)^2 \right]^{0.5}$

355 $d_i^- = \left[\sum_{j=1}^n (V_{ij} - V_j^-)^2 \right]^{0.5}$

356 The normalized utility (N_{ij}) is used to calculate the weighted normalized performance
357 (V_{ij}) of each alternative under each criterion using Equation 7. The best (V_j^+) and the worst (V_j^-)
358 performance of the alternatives under each criterion are determined, and used to calculate the
359 distance of each alternative from the best and the worst scenario as shown previously in Equation
360 6.

$$361 \quad V_{ij} = N_{ij} \cdot W_{ij} \quad \text{Equation 7}$$

362 where:

$$363 \quad N_{ij} = \frac{r_{ij}}{\sqrt{\sum_{i=1}^m r_{ij}^2}}$$

364 At every level of the decision matrix, SAW and TOPSIS were used to calculate the
365 comparison indices and relative distances of each alternative. The comparison indices (or relative
366 distances) were normalized using Equations 1 and 2, and used as a performance input value for
367 the upper level.

368 **3. Results and discussion**

369 TOPSIS and SAW were used to rank fuel alternatives for the waste collection industry with
370 respect to the multi-level environmental and financial decision matrix (Figures 2 and 3). The
371 overall ranking placed conventional diesel-fueled WCVs as the best option under the decision
372 matrix, followed by hydraulic-hybrid, LFG-sourced natural gas, North American and non-North
373 American natural gas, and biodiesel fuels. The results of the two methods were consistent.
374 Environmentally, WCVs fueled with fossil fuels (diesel and natural) were closer to the ideal than
375 biogenic fuels (BD and LFG); the inclusion of the WFP and power density as environmental
376 measures placed biogenic fuels, biodiesel and LFG, far from being the ideal fuel option.
377 Environmentally, CNG and LNG WCVs fueled by American fossil natural gas had slight
378 advantage over WCVs fueled with non-American natural gas or diesel. Hydraulic-hybrid WCVs

379 were the closest to the optimal solution with respect to the environmental criteria, because fuel
380 savings compared to diesel placed it closer to the optimal environmental option ahead of diesel.
381 Financially, diesel and hydraulic-hybrid ranked closest to the ideal solution under the decision
382 matrix. The vehicle cost of hydraulic-hybrid vehicles averaged \$100,000 more than conventional
383 diesel-fueled WCVs; however, the fuel savings associated with hydraulic hybrid WCVs placed it
384 at a similar distance from the ideal solution as conventional diesel-fueled WCVs. Natural gas
385 (CNG and LNG) and biodiesel were affected by the current lack of fueling stations. The fuel
386 price of biodiesel placed this option far from the ideal solution as it is currently the most
387 expensive alternative. LFG has the cheapest price, however the availability of LFG fueling
388 station impacted the financial and overall performance of this alternative.

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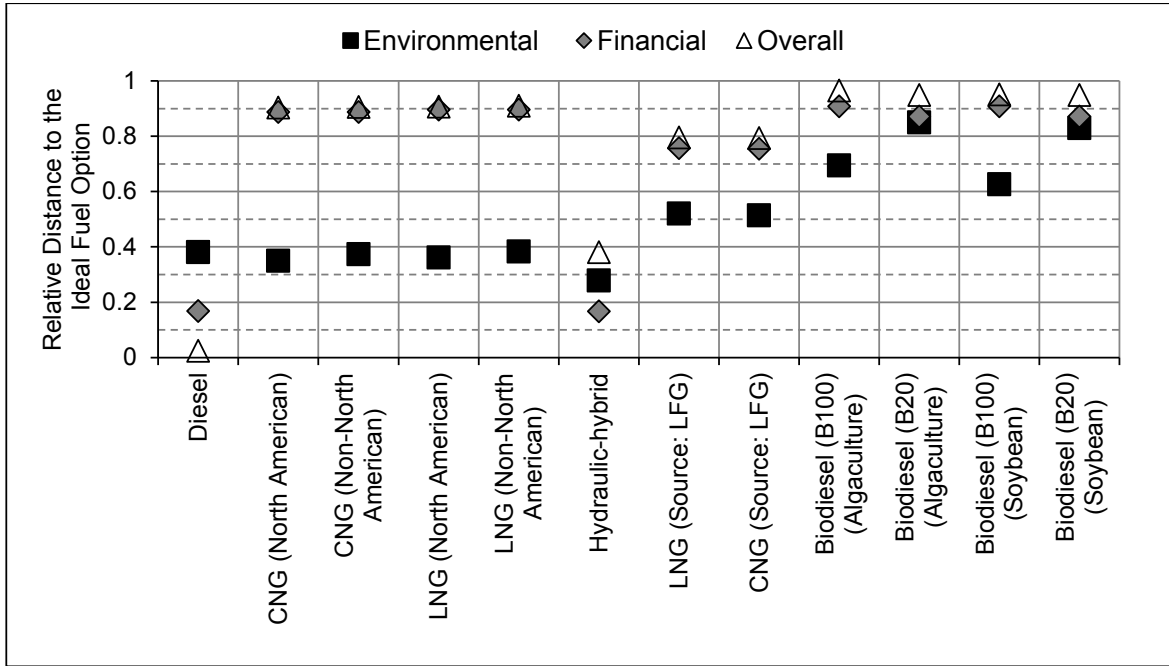
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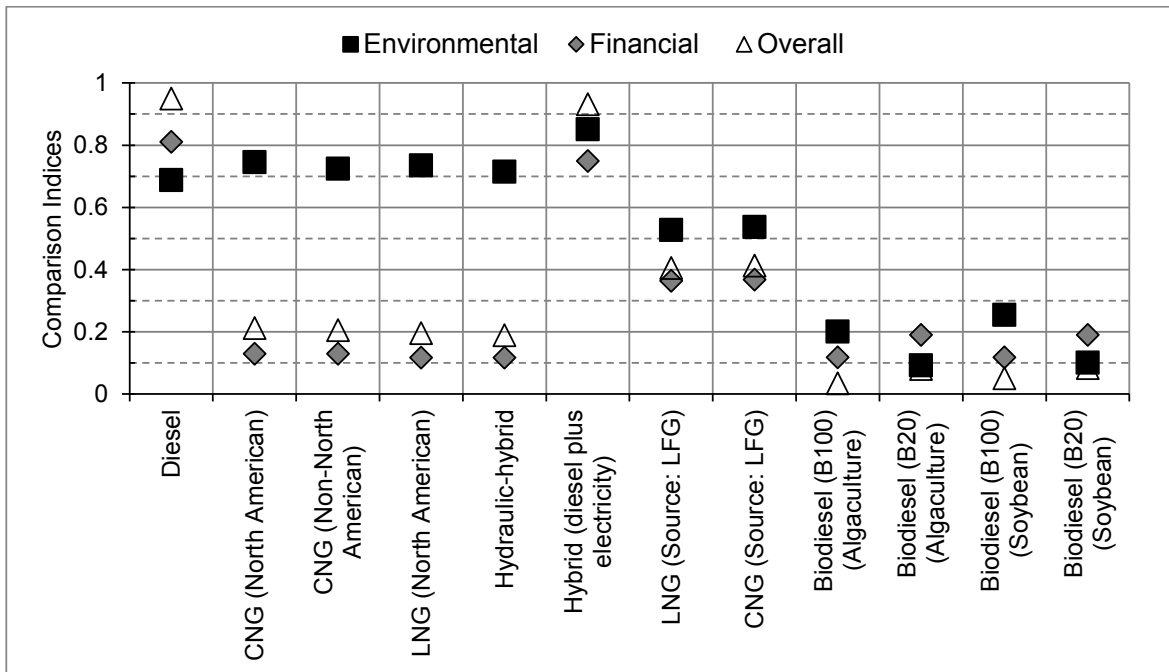
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Figure 2: Relative distances (TOPSIS analysis) of fuel options from the ideal option using the selected decision matrix.



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Figure 3: Comparison indices (SAW analysis) of fuel options using the selected decision-matrix.

406 **3.1 Significance of the Selection Criteria**

407 In the previous analysis, fuel rankings were based on the selected decision matrix; however it is
408 imperative to assess how sensitive the fuel rankings are to the selection criteria considered by
409 decision makers. Therefore, an analysis was conducted by eliminating one or two criteria from
410 the decision matrix, then determining the relative distance of alternatives to the ideal solution
411 (TOPSIS analysis). The following five sensitivity analysis scenarios were considered:

412 **Scenario 1:** Eliminate the water footprint criterion,

413 **Scenario 2:** Eliminate the WFP and the power density criteria,

414 **Scenario 3:** Eliminate the fueling station criterion,

415 **Scenario 4:** Eliminate the fuel price stability criterion,

416 **Scenario 5:** Eliminate the fueling station and fuel price stability criteria.

417 The five sensitivity analysis scenario results are illustrated in Figure 4. For comparison,
418 the results from the original analysis using the complete decision matrix were labeled **Scenario**
419 **0**. In **Scenario 1**, the elimination of the WFP criterion from the decision matrix did not impact
420 the environmental or overall fuel ranking because the ranking of agricultural-based fuel
421 alternatives was also affected by low power density as compared to fossil fuels. Alternative fuels
422 with high WFP are associated with low power density. As a result, the elimination of the WFP
423 alone did not affect the environmental or overall ranking of agriculture-based fuel alternatives. In
424 **Scenario 2**, the elimination of the WFP and the power density from the decision matrix changed
425 the environmental ranking of fuel alternatives so that biofuels (LFG-sourced natural gas and
426 biodiesel) ranked ahead of fossil fuels. Biogenic fuels were considered the best based on life-
427 cycle emissions and some tail-pipe emissions. However, they are associated with high WFP and
428 low power density. LFG-sourced natural gas ranked as the best alternative followed by BD100

429 (soybean then algaculture). The overall ranking of alternatives was slightly affected by removing
430 the WFP and power density criteria from the decision matrix, as LFG-source natural gas ranked
431 third after conventional diesel and hydraulic-hybrid. Biodiesel has favorable life-cycle
432 emissions; however, its production is associated with high WFP and low power density. These
433 results signify the importance of considering the WFP and power density criteria as
434 environmental measures in addition to traditional life-cycle and tail-pipe emissions. It also
435 suggests that the use of different feedstock (e.g. waste) for the production of biodiesel should be
436 considered, which might reduce the WFP and the power density of biodiesel production, making
437 it more favorable.

438 In **Scenario 3**, the fueling station criterion was eliminated from the decision matrix and
439 LFG-sourced natural gas ranked as the best alternative from the financial prospective. Diesel and
440 hydraulic-hybrid were ranked next, followed by BD20, North American, non-North American
441 natural gas, BD100. Therefore, LFG-sourced natural gas is considered as the best option for
442 WCVs when available. In **Scenario 4**, the fuel price stability was eliminated from the decision
443 matrix moving diesel and hydraulic-hybrid to be the optimal financial solution followed by LFG-
444 sourced natural gas, however the overall ranking did not change significantly from Scenario 0. In
445 **Scenario 5**, the elimination of fueling station and fuel price stability criteria ranked LFG-sourced
446 natural gas as the best financial alternative followed by North American fossil natural gas. This
447 scenario was found to represent the status quo of the waste collection industry as the industry is
448 leaning toward fossil natural gas, driven by low natural gas prices. A sensitivity of the fuel
449 ranking to instability of fuel prices was evaluated for the status quo scenario. In the next section,
450 the results of dynamic sensitivity analysis to determine the impact of changing the actual fuel
451 price on the fuel ranking are reported.

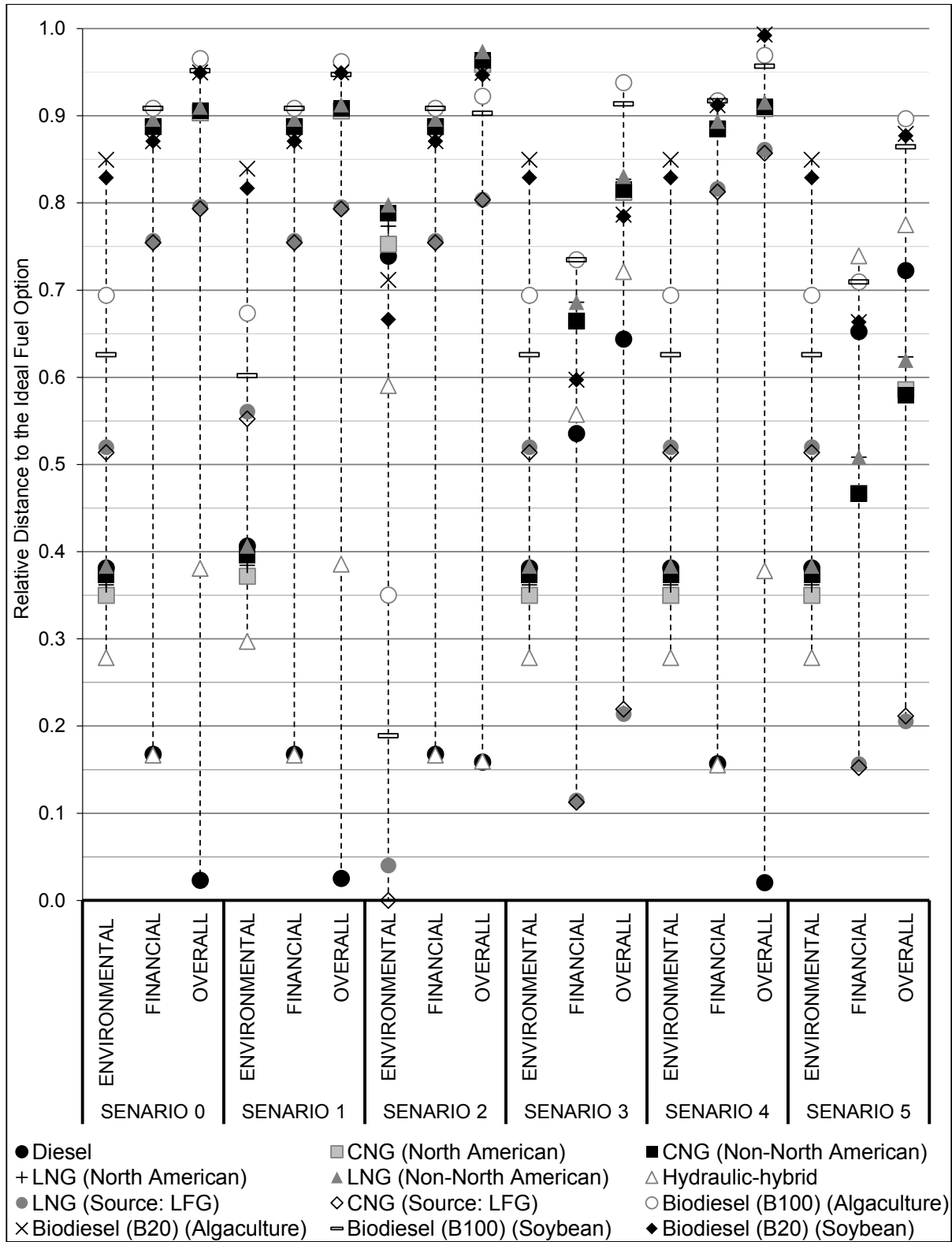


Figure 4: Significance of the selection criteria.

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3.2 Systematic Sensitivity Analysis of Alternative Fuel Price

A systematic sensitivity analysis of alternative fuel price was conducted by evaluating the relative distances (TOPSIS) of each alternative from the ideal financial fuel option (Figure 5) and ideal overall fuel option (Figure 6), using five different price scenarios for diesel, natural gas, LFG, and biodiesel. In the analysis, the relative distances were calculated for each alternative while varying the fuel price of each alternative by -50%, -25%, +25%, and +50% of the current fuel price. The fueling station and fuel price stability criteria were eliminated from the decision matrix during the analysis to illustrate the status quo scenario determined by the sensitivity analysis. The financial criteria consisted of the vehicle cost and fuel price, while environmental criteria included life-cycle emissions, tail-pipe emissions, WFP, and power density. The number of fueling stations gave advantage to some alternatives over the others, while the fuel price stability criterion was excluded as the analysis gauges the sensitivity of ranking to changing fuel prices. The purpose of this analysis was to determine how sensitive the fuel ranking is to changing fuel price, as the industry builds more natural gas fueling stations based on the current natural gas prices.

Financially, CNG and LNG collection vehicles fueled with LFG-sourced natural gas ranked as the best alternatives. However, it was noticed that 50% decrease in diesel fuel price placed diesel in the same rank as LFG-sourced natural gas. Also, a 50% decrease in fossil natural gas prices moved fossil CNG and LNG closer to LFG-sourced natural gas; however, the LFG-sourced natural gas continued to rank as the best alternative. The ranking of diesel and hydraulic-hybrid was found to be more sensitive to fuel price. A drop of diesel price by 25% ranked diesel better than natural gas, while a 50% drop ranked hydraulic-hybrid as favorable as fossil natural gas. On the other hand, a 25% increase in diesel price ranked diesel and hydraulic-hybrid behind

477 all other alternatives. Fossil CNG and LNG ranked behind LFG-sourced natural gas. However,
478 any increase in natural gas prices moved the alternative away from the ideal solution and in the
479 case of a 50% increase, fossil natural gas ranked behind diesel and hydraulic-hybrid. LFG-
480 sourced natural gas continued to rank as the best alternative even at a 50% increase in fuel price.
481 Finally, BD20 and BD100 rankings are sensitive to changing fuel price. A 50% decrease in
482 biodiesel price ranked BD20 and BD100 second after LFG-sourced natural gas, while a 25%
483 ranked BD20 in between fossil CNG and LNG. An increase in biodiesel prices moved diesel
484 toward fossil natural gas, a result of dispersion of fuel prices as biodiesel prices currently are
485 highest.

486 Overall, LFG-sourced natural gas continued to rank as the best alternative with respect to
487 the overall environmental and financial criteria, except at a 50% decrease in diesel prices (Figure
488 6). CNG and LNG collection vehicles fueled with North-American natural gas ranked second
489 after LFG-sourced natural gas. But, any increase in prices could move diesel and hydraulic-
490 hybrid ahead of fossil natural gas (North American or non-North American). Fossil natural gas
491 continued to rank as the second alternative after LFG-sourced natural gas except when natural
492 gas prices increased by 50% or diesel prices dropped by 25 to 50%. The overall ranking of LFG-
493 sourced natural gas, BD 20, or BD100 was not as significantly affected by changing fuel prices.

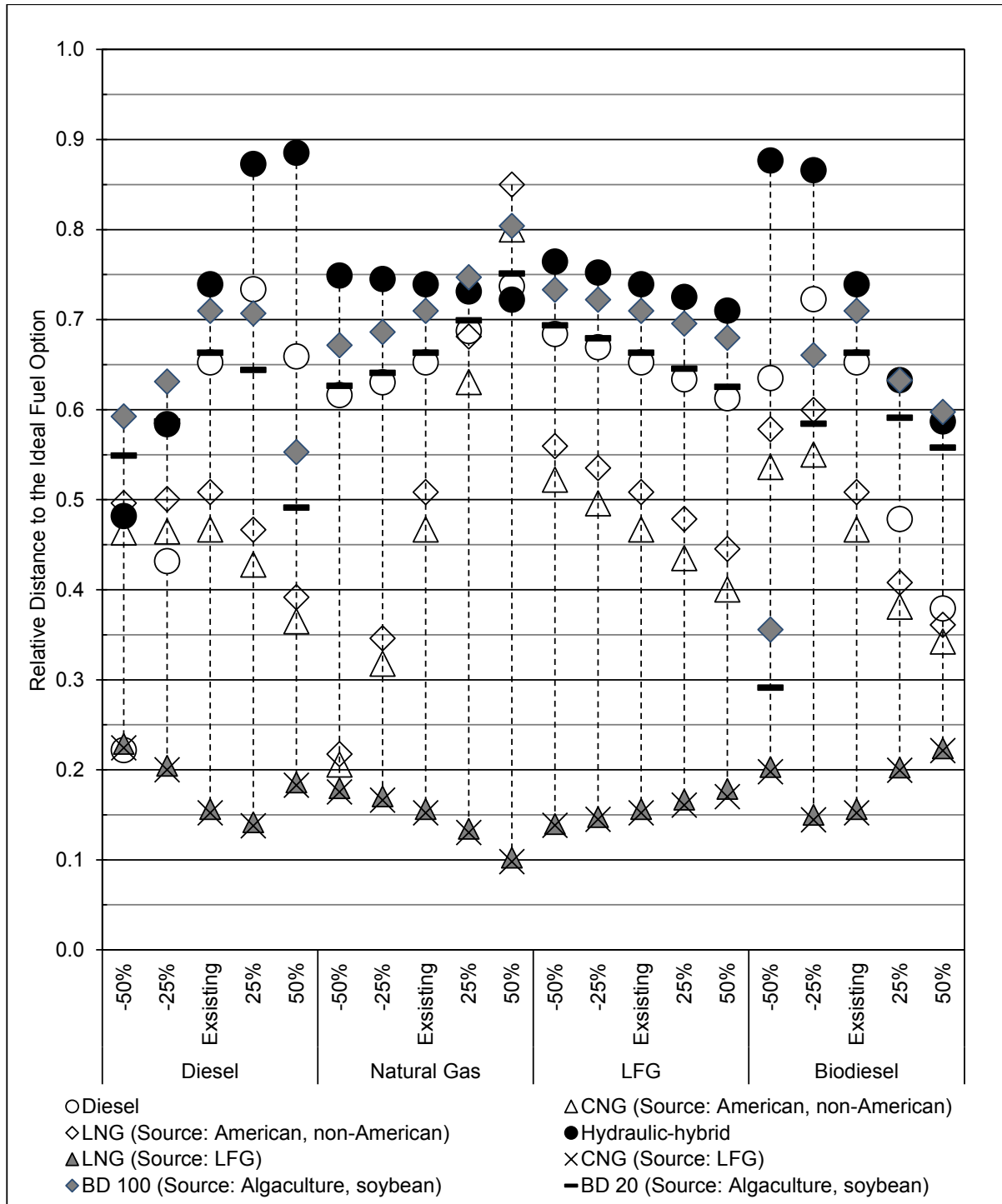
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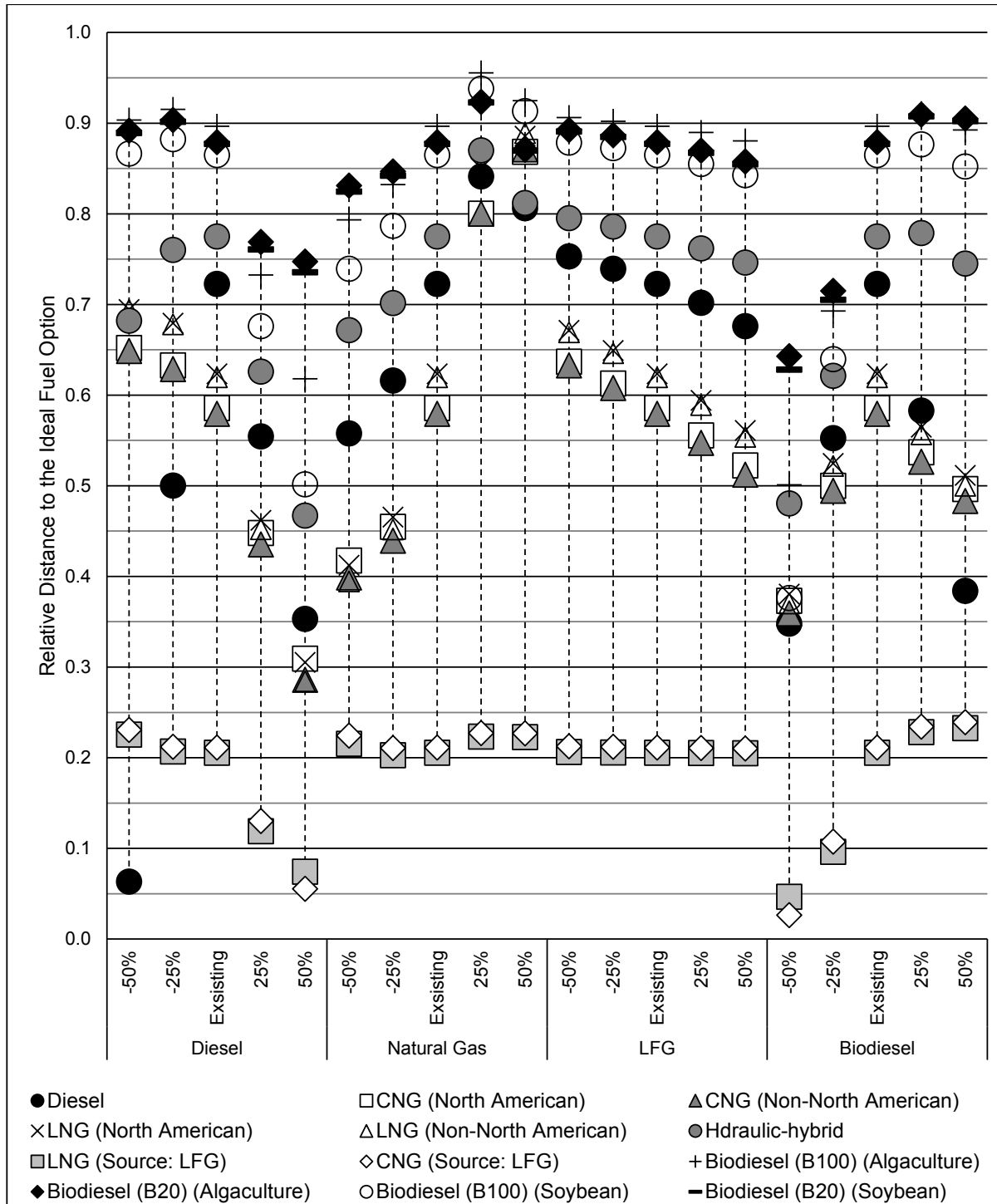
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Figure 5: Systematic sensitivity analysis of the financial performance. (Relative distances (TOPSIS analysis) were calculated for each fuel using five different fuel pricing for each alternative; -50% of the current fuel price, -25% of the current fuel price, existing, +25% of the current fuel price, and +50% of the current fuel price).



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Figure 6: Systematic sensitivity analysis of the overall performance. (Relative distances (TOPSIS analysis) were calculated for each fuel using five different fuel pricing for each alternative; -50% of the current fuel price, -25% of the current fuel price, existing, +25% of the current fuel price, and +50% of the current fuel price).

509 **3.3 Additional Financial Criteria**

510 There are other financial criteria that can influence the selection process; however they were
511 excluded from the initial analysis due to data availability concerns. The maintenance cost of
512 alternative fueled WCVs is a vital component of the running cost and is often considered by
513 decision makers. According to U.S. waste haulers, the cost of maintaining a diesel-fueled WCV
514 average \$8.5 per hour of operation (personal communication with Major Hauler Manager, 2012).
515 The maintenance cost of alternative fueled WCVs is not available for newly acquired fuel
516 technologies, therefore it is not as easily accounted for as conventional diesel-fueled WCVs.
517 Secondly, municipalities and private waste haulers are often interested in retrofitting existing
518 diesel-fueled WCVs to support alternative fuel technologies. In the previous analysis, fuel
519 rankings were based on the assumption that WCVs will be purchased new.

520 Accordingly, this analysis was conducted to determine the impact of maintenance cost
521 and the possibility of vehicle retrofitting to operate an alternative fuel. The status quo scenario
522 “Scenario 5” determined in Section 3.1 was compared to the result drawn from this analysis. In
523 this analysis, the vehicle cost criterion was replaced by the cost of retrofitting an existing diesel-
524 fueled WCV to run on natural gas or hydraulic hybrid. Gordon et al. (2003) reported that the cost
525 of switching an existing WCV to natural gas ranges from \$30,000 to \$100,000. Moreover,
526 Baseley et al. (2007) stated that existing WCVs can be retrofitted with a second hydraulic system
527 easily. The cost of adding a hydraulic system to an existing diesel-fueled WCV was reported to
528 be less than 50,000 (Drozd, 2005; Baseley et al., 2007). For the purpose of this analysis, it was
529 assumed that additional vehicle costs for a municipality or the private hauler to run diesel-fueled
530 WCVs using biodiesel (BD-20 and 100), natural gas (CNG and LNG), and hydraulic hybrid
531 vehicles were \$0, \$65,000, and \$25,000 per vehicle. Moreover, the analysis assumed that the

532 waste haulers can continue to operate their diesel-fueled vehicle at no additional cost (vehicle
533 cost \$0).

534 The results of this analysis (Scenario 5 retrofitted) are shown in Figure 7. The analysis
535 indicated that diesel-fueled WCVs are still the best alternative financially; however if decision
536 makers are interested in switching to an alternative fuel, biodiesel blends can be considered as
537 the second best alternative, followed by hydraulic-hybrid. This is mainly due to the fact that no
538 vehicle cost is associated with switching to diesel or biodiesel blends. The possibility of
539 retrofitting existing diesel-fueled WCVs to support hydraulic hybrid technology ranked hydraulic
540 hybrid as a better financial alternative.

541 For the purpose of recognizing the impact of maintenance cost on the financial ranking of
542 alternative-fueled WCVs, three different hypothetical scenarios were evaluated. The financial
543 analysis of alternative-fueled WCVs was conducted using the financial criteria of the status quo
544 scenario (Scenario no. 5 - vehicle cost and fuel cost) and by adding maintenance cost as a
545 criterion. The assumption made in estimating the maintenance cost of alternative-fueled WCVs
546 are as follows:

547 **Scenario X:** The maintenance cost of running natural gas WCVs is the same as diesel (\$8.5 per
548 hour), while hydraulic hybrid WCVs' maintenance cost is 50% more than diesel;

549 **Scenario Y:** The maintenance cost of hydraulic hybrid WCVs is the same as diesel (\$8.5 per
550 hour), while natural gas WCV's maintenance cost is 50% more than diesel; and

551 **Scenario Z:** The maintenance cost of hydraulic hybrid and natural gas WCV's maintenance cost
552 is 50% more than diesel.

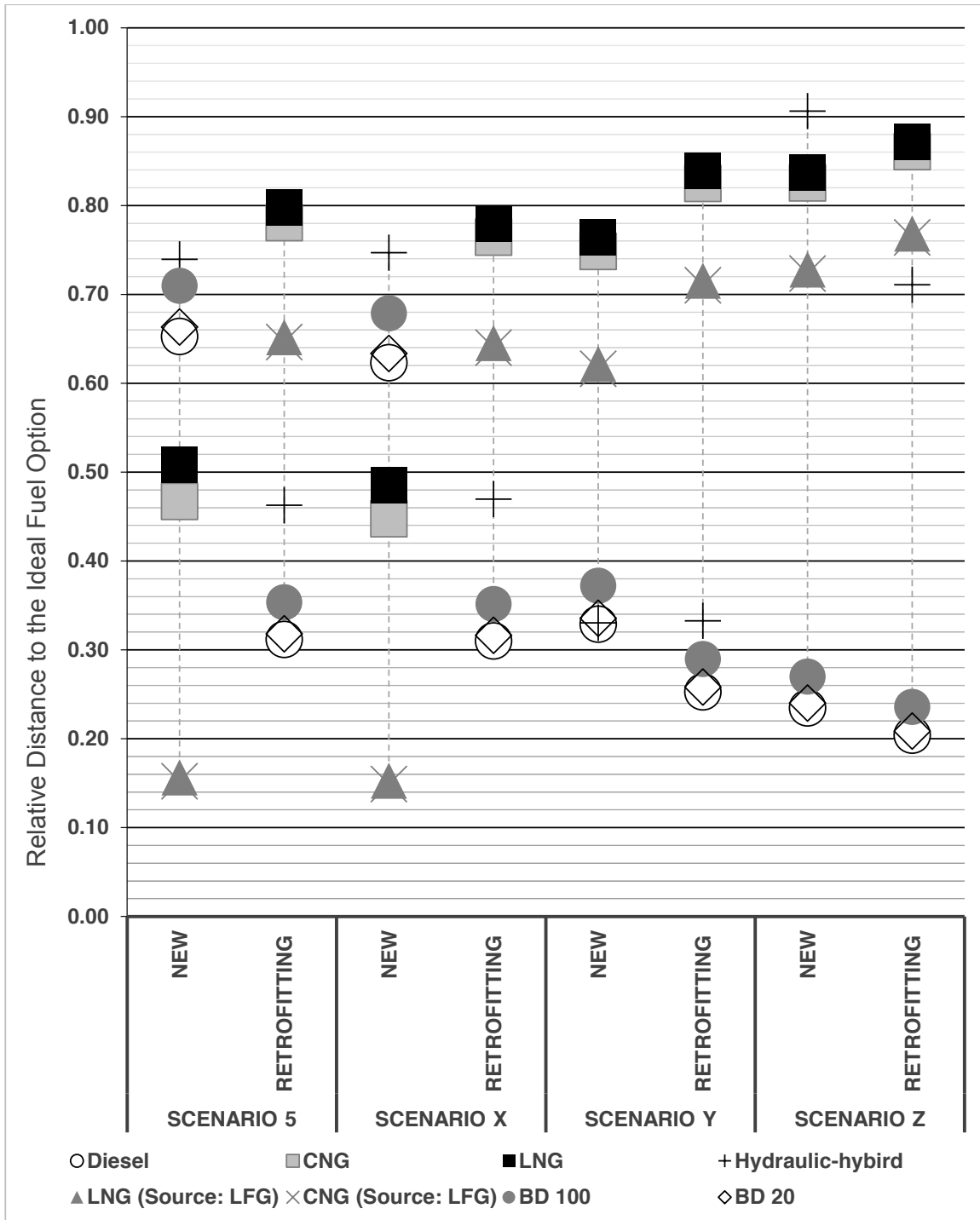
553 It was assumed that the cost of maintaining diesel-fueled WCVs running on biodiesel fuel
554 blends to be the same as vehicles running using fossil diesel. However, it should be noted that

555 waste haulers complain about the use of biodiesel blends especially during cold weather
556 (personal communication with Joseph Grusauskas, 2012). The results of the three scenarios are
557 illustrated in Figure 7. For each scenario, two analyses were performed; one assuming that
558 WCVs will be purchased new, and the second assuming the possibility of retrofitting existing
559 diesel-fueled WCVs to support alternative fuels. It is very clear that adding an additional
560 financial criteria changed the financial ranking of fuel alternatives.

561 In Scenario X, the financial fuel rankings were similar to Scenario 5. However, the
562 hypothetical increased maintenance cost of hydraulic hybrid WCVs pushed the alternative away
563 from the optimal financial solution. In Scenario Y, the hypothetical maintenance cost of natural
564 gas WCVs being 50% more than diesel-fueled WCVs pushed natural gas WCVs, using either
565 fossil natural gas or LFG-sourced natural gas, behind diesel, hydraulic hybrid and biodiesel. This
566 shows that hydraulic hybrid WCVs would be considered better than natural gas WCVs if their
567 maintenance cost were lower than natural gas and closer to conventional diesel WCVs. This also
568 shows the sensitivity of natural gas fueled WCVs financial ranking. Also, biodiesel WCVs
569 ranking moved closer to the optimal solution. This is assuming that the cost of maintaining
570 biodiesel WCVs to be the same as diesel.

571 Scenario Z ranked diesel and biodiesel as the best alternatives when other alternative
572 fuels, e.g. natural gas and hydraulic hybrid, cost 50% more than diesel and biodiesel to maintain.
573 Finally, in almost all cases that involve retrofitting existing WCVs, biodiesel blends are the best
574 alternative to diesel if decision makers are interested in switching away from diesel.

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Figure 7: Sensitivity analysis of the financial performance of alternative fuel technologies, using additional financial criteria. (Relative distances (TOPSIS analysis) were calculated for different Scenarios).

582 **3.4 Operational Issues and Community Acceptance**

583 Decision makers often have to consider other social and operational criteria that may not
584 necessary fit into the economic or environmental criteria described in this study. It is often
585 difficult to quantify the impact of such criteria due to data availability or variability (e.g. changes
586 in social aspects across communities). Operational criteria such as refueling time, vehicle noise
587 level, maintenance complexity, and reliability might be considered. The aforementioned criteria
588 are crucial for waste haulers due to the limited number of replacement vehicles available, long
589 driving distances, and customer service concerns over delayed or missed waste pickups. Public
590 acceptance of an alternative fuel is also vital for waste haulers. In recent years, waste haulers
591 utilized switching to alternative fuels as an advertisement tool to gain public acceptance over
592 environmental friendly “green” infrastructure. This study did not address social aspects of the
593 selection process, however it is recommended that future studies evaluate the influence of public
594 acceptance and other operational criteria on the decision making process.

595 **3.5 Conclusions**

596 MCDA tools were used to rank fuel alternatives for the waste collection industry with respect to
597 a multi-level environmental and financial decision matrix. The environmental criteria consisted
598 of life-cycle emissions, tail-pipe emissions, water footprint, and power density, while the
599 financial criteria included vehicle cost, fuel price, fuel price stability, and fueling station
600 availability. Environmentally, hydraulic-hybrid and fossil natural gas, performed better than
601 conventional-diesel. The vehicle cost of hydraulic-hybrid and lack of fueling stations for natural
602 gas affected the financial ranking, although fuel price savings were observed for both options.
603 The overall analysis using the environmental and financial criteria showed that conventional-
604 diesel and hydraulic-hybrid WCVs are the best alternatives, followed by LFG-sourced natural

605 gas, fossil natural gas, and biodiesel. This fuel ranking changed as different decision matrices
606 were used, signifying the importance of the selection criterion considered by decision makers.
607 The elimination of the WFP and power density criteria from the environmental criteria ranked
608 biodiesel 100 (BD100) as an environmental-friendly alternative compared to other fossil fuels
609 (diesel and natural gas). This result signifies the importance of considering WFP and power
610 density criteria as environmental measures in addition to traditional life-cycle analysis and tail-
611 pipe emissions. The elimination of the fueling station criterion from the financial decision level
612 ranked landfill gas (LFG) sourced natural gas as the best option; suggesting that LFG-sourced
613 natural gas is the best alternative to fuel WCV when accessible. The elimination of the fueling
614 station criterion and fuel price stability criterion from the decision matrix ranked fossil natural
615 gas second after LFG-sourced natural gas. This scenario characterizes the status quo of the
616 industry. The waste collection industry is driven by low natural gas prices compared to other
617 alternatives, and has set investment plans to build natural gas fueling stations. A systematic
618 sensitivity analysis was used to determine the impact of changing fuel prices on decisions. The
619 financial ranking of all alternatives, except LFG-sourced natural gas, was found to be sensitive to
620 changing fuel prices. The overall ranking of diesel and natural gas was found to be more
621 sensitive to changing fuel price as compared to LFG-sourced natural gas, BD20 or BD100.

622

623 **Conflict of Interest**

624 The authors declare that there are no conflicts of interest.

625

626 **Acknowledgement**

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