

A hedonic pricing method to estimate value of waterfront in the Gulf of Mexico

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2 **Abstract**

3 Open spaces, including waterfront areas, are publicly-or-privately owned landscapes that
4 provide numerous benefits and services such as opportunities for recreational activities,
5 ecological benefits, and economic development. However, with rapidly growing populations,
6 development pressure on these areas has been increasing, often leading to conflicts between
7 proposed land uses. Information on the monetary value of environmental amenities provided by
8 these spaces would help decision makers account for their importance to quality of life. This
9 study estimated the monetary value associated with waterfronts using the hedonic pricing method
10 (HPM) and real estate sales data for the coastal cities of Mobile and Daphne in Alabama, USA.
11 The price of houses sold during 2001 to 2015 was used as the dependent variable and house
12 structural and neighborhood attributes and presence of environmental amenities served as
13 independent variables. Results showed that coastal residents considered proximity to waterfronts
14 as one of the most important factors when buying a house and paid higher prices for houses
15 located near most waterfront types. In Mobile, marginal implicit prices of proximity to
16 waterfronts ranged from \$2,490 to \$3,530 per km, whereas in Daphne, the price ranged from
17 \$9,250 to \$15,460 per km. Findings can help guide future decisions related to development of
18 coastal areas, land-use planning, urban forestry, and open space preservation by balancing
19 opportunities for urban and commercial development as well as providing public access to open
20 space environmental amenities with close proximity to residential areas.

21 Keywords: Ecosystem services, land-use planning, marginal implicit price, multiple
22 listing service data, real estate, waterfront

23

24 **1.1 Introduction**

25 Open spaces are publicly-or-privately owned landscapes that are partially or completely
26 covered with vegetation or water (Allen Klaiber & Phaneuf, 2010; Bolitzer & Netusil, 2000).
27 Open spaces can be developed (e.g., a city recreational park) or undeveloped (e.g., an exurban
28 woodland). With increases in population and urbanization, damage to and loss of waterfront
29 open spaces have been extensive. Yet, due to their environmental, economic, and cultural
30 benefits, waterfront open spaces remain an integral part of urban landscapes. The challenges city
31 planners and elected officials face in managing urban sprawl and protecting waterfronts often
32 relate to the lack of monetary estimates of waterfront values that would help prioritize
33 preservation versus other land uses. Proponents of open space planning call for preservation of
34 waterfronts based on location, availability, and suitability, including mitigating forest
35 fragmentation through hubs, links, and corridors. Valuation analyses such as this study help
36 identify and gauge the importance of these areas, and may be preferable to the current approach
37 of arbitrarily preserving leftover land, which is not necessarily the open space that provides the
38 highest benefits to the community.

39 The benefits of open space to human welfare have been extensively discussed within
40 diverse research areas. For example, open spaces provide many services including recreational
41 opportunities, scenic views, and mitigation of negative externalities such as pollution and
42 congestion associated with development (Irwin, 2002). In addition, open spaces provide
43 ecological benefits such as wildlife habitat, air and water quality improvement, and urban heat
44 island reduction; economic benefits such as increased real estate values and improved local
45 economies; and other health and socio-cultural benefits such as places to exercise and socialize
46 (Anderson & West, 2006; Brander & Koetse, 2011; Hakim et al., 1999; Lowry, 1967; Nowak,

47 Crane, & Stevens, 2006). Therefore, protection of open spaces should be considered in
48 residential and commercial planning decisions to enhance environmental, cultural, and economic
49 values of adjacent areas.

50 With an increase in population and urbanization, many open space areas have been lost to
51 commercial and residential developments (McDonald, Forman, & Kareiva, 2010; Schuyt, 2005).
52 In terms of U.S. forest land alone, approximately 4.0 million hectares of land were converted to
53 development during 1982 to 1997 and an additional 9.0 million hectares are projected to be
54 developed by 2030 (Alig & Plantinga, 2004). Impacts of such disturbances are not only felt
55 locally where rapid development occurs, but are also transported to distant locations via air and
56 water pollution (Faulkner, 2004). For example, habitat fragmentation as a result of urban sprawl
57 induces edge effects resulting in a loss of connectivity between habitats and, therefore, restricts
58 movement of species, leading some species to a threatened status and others to possible
59 extinction (Faulkner, 2004). Thus, findings showed that loss of open space was highly correlated
60 with population growth.

61 With an increase in population, demand for open space and its benefits has also been
62 increasing, making preservation of open space an important policy and social issue. There have
63 been numerous initiatives at national, state, and local levels to preserve and protect open space
64 (Bengston, Fletcher, & Nelson, 2004). For example, in 2007 the U.S. Forest Service has
65 developed an Open Space Conservation Strategy to help conserve open spaces (USDA, 2007).
66 Tax policies are examples of other initiatives implemented to support open space conservation
67 and promote ecosystem service markets (Bengston, Fletcher, & Nelson, 2004). Similarly, voters
68 in 21 states approved funds of over \$3 billion for open space conservation in 2016 (Trust for
69 Public Land 2016). Such spending suggests the importance of open space to communities.

70 However, this spending presents only a partial estimate of the extent of benefits that open space
71 provides (McConnell & Walls, 2005). Thus, knowing the total economic value of open space
72 will help in formulating public policy to retain waterfront open space and guide future land
73 conservation and development decisions.

74 Estimating a total value of open space is difficult because it considers both market and
75 nonmarket goods and services (McConnell & Walls, 2005). In contrast to crops and timber,
76 nonmarket goods such as clean air and aesthetic values are not directly traded in the market and
77 it is therefore challenging to estimate their monetary value (McConnell & Walls, 2005). If
78 nonmarket goods and services can be quantified in monetary terms, their value can be compared
79 with and possibly outweigh market values such as urban development (Boyer & Polasky, 2004).
80 Monetary valuation helps to identify types of open space that have the greatest importance to
81 residents (i.e., consumers), compare relative values of these open spaces with other land uses,
82 and facilitate more informed policy decisions pertaining to their conservation and development.

83 Various methods have been developed by economists to estimate the monetary value of
84 nonmarket goods and services. The most common methods used in nonmarket valuation include
85 the contingent valuation method (CVM), the travel cost method (TCM), and the hedonic pricing
86 method (HPM). This study employed the HPM, which is a revealed preference method, and
87 utilized real estate market transaction data to estimate the monetary value of waterfront
88 properties. HPM relies on information from property purchase behavior to infer values for
89 environmental amenities. This approach assumes that property price, such as a house price, is a
90 function of structural, neighborhood, and environmental attributes.

91 HPM is a commonly-used method in quantifying the monetary value of nonmarket
92 benefits, such as those provided by waterfront open spaces (Morancho, 2003). For example,

93 Acharya and Bennett (2001), Irwin (2002), Anderson and West (2006), and Poudyal et al. (2009)
94 used HPM to quantify the monetary value of nonmarket benefits provided by open space in
95 different parts of the U.S. Findings from these studies suggested that proximity to open space
96 was positively associated with property values. The HPM has also been used previously to
97 estimate the value of selected waterfront properties adjacent to lakes and reservoirs, rivers and
98 streams, and oceans (Brown & Pollakowski, 1977; Costanza et al., 2006; Knetsch, 1964; Mahan,
99 Polasky & Adams, 2000; Young & Teti, 1984). For example, Mahan et al. (2000) used HPM to
100 estimate the monetary value of a wetland in Portland, Oregon and reported that wetland
101 characteristics related to its size and distance to an urban area were related with the value of
102 nearby residential properties. The study reported that a one-acre increase in wetland area was
103 associated with an increase of nearby house values by \$24.39. Similarly, decreasing distance
104 between residences and wetlands by 1,000 feet increased property values by \$436.17. A similar
105 study by Bin (2005) found that moving 1,000 feet closer to the nearest river increased property
106 value by \$3,750, whereas Costanza et al. (2006) estimated that sale prices for houses within 300
107 feet of a beach ranged from \$81,000 to \$194,000 higher than for houses located further away,
108 suggesting beach proximity had a positive impact on house values. Some of the other recent
109 studies related to waterfronts are summarized in Table 1.

110 Most previous studies either quantified values of individual waterfront types such as
111 lakes, beaches, and wetlands or treated them as a composite good. However, values may vary
112 across different types of open space (McConnell & Walls, 2005). Valuing different waterfront
113 types as a composite good thus may not be useful to practitioners. To address this limitation, our
114 study estimated monetary values of different waterfront types such as bays, rivers, streams,
115 bayous, and other water bodies in a single HPM model to facilitate the comparison of their

116 monetary values and relative importance to coastal residents. Thus, the study provides useful
117 insights on the monetary value of specific waterfront types and helps to analyze the effect of
118 different land use options. The study used 13,903 house sale records obtained from a multiple
119 listing service (MLS) to estimate marginal implicit prices of proximity to different waterfront
120 types in two coastal cities near the Gulf of Mexico: Mobile and Daphne, Alabama, U.S. Findings
121 have several policy implications related to maximizing net benefits of waterfront access and can
122 be used to facilitate informed planning decisions regarding waterfront open space preservation
123 and alternative development as well as financial and tax decisions related to waterfront open
124 space conservation.

125

126 Table 1. Summary of the recent hedonic studies related to a monetary valuation of waterfronts.

Waterfront type	Author and year	Study site/ location	Analysis method	Finding(s)
Waterbody	Cho et al., 2010	Southern U.S.	Spatial model	Water view was associated with a house price increase of 26%.
Watercourse	MacDonald et al., 2010	Adelaide, South Australia	OLS	A 1-meter increase in distance from a property to a watercourse was associated with a price decrease of AU\$24.
Seashore	Damigos & Anyfantis, 2011	Athens, Greece	Fuzzy Delphi method	Sea view related to dwelling price increase of 21% to 65%, with the most likely price premium of 34%.
Lake	Wen et al, 2012	Hangzhou, China	OLS	A 1% increase in distance to a lake translated to a price decrease of 0.24%.
River	Tapsuwan et al., 2012	Murray-Darling Basin, South Australia	OLS	For an average house located 1 km away from a river, decreasing a distance to a river to 0.50 km was associated with a house price increase of AU\$245,000.
River and coast line	Gibbons et al., 2013	England specifically and Great Britain to a lesser extent	OLS	A 1-km increase in distance to river and coastline related to a house price decrease of 0.93% and 0.14%, respectively.
Stream, canal, and waterbody	Larson & Perrings, 2013	Phoenix, Arizona	OLS and spatial lag model	A 1% increase in distance to stream, canal, and waterbody was associated with a house price decrease of 0.002%, 0.008%, and 0.008% in OLS model and <0.001%, 0.002%, and 0.006% in a spatial lag model, respectively.
Lake	Pandit et al., 2013	Perth, Western Australia	OLS and spatial model	A 1% increase in distance to a park with lake related to a price decrease of 0.03% (OLS) and 0.02% (spatial model).
Waterbody	Kolbe & Wustemann, 2014	Cologne, Germany	OLS	A 1% increase in the size of a waterbody within a 500-meter buffer translated to a price increase of 0.16%.
Waterbody	Cohen et al, 2015	Connecticut	OLS and GWR	Relationship insignificant in OLS, whereas a 1% increase in distance to a waterbody related to a house price decrease of 0.027% in the GWR model.
Surf break (oceanfront)	Scorse et al., 2015	Santa Cruz, California, U.S.	OLS	Ocean view was associated with a house price increase of \$957,000. A house located next to a surf break was valued at \$106,000 more than a house located 1 mile away.
River	Tapsuwan et al., 2015	Murray-Darling Basin, South Australia	OLS and spatial model	Within 3-km buffer, 1-km increase in distance to a river was associated with a price decrease of AU\$2,414.

Note: OLS stands for ordinary least square; GWR stands for geographically weighted regression.

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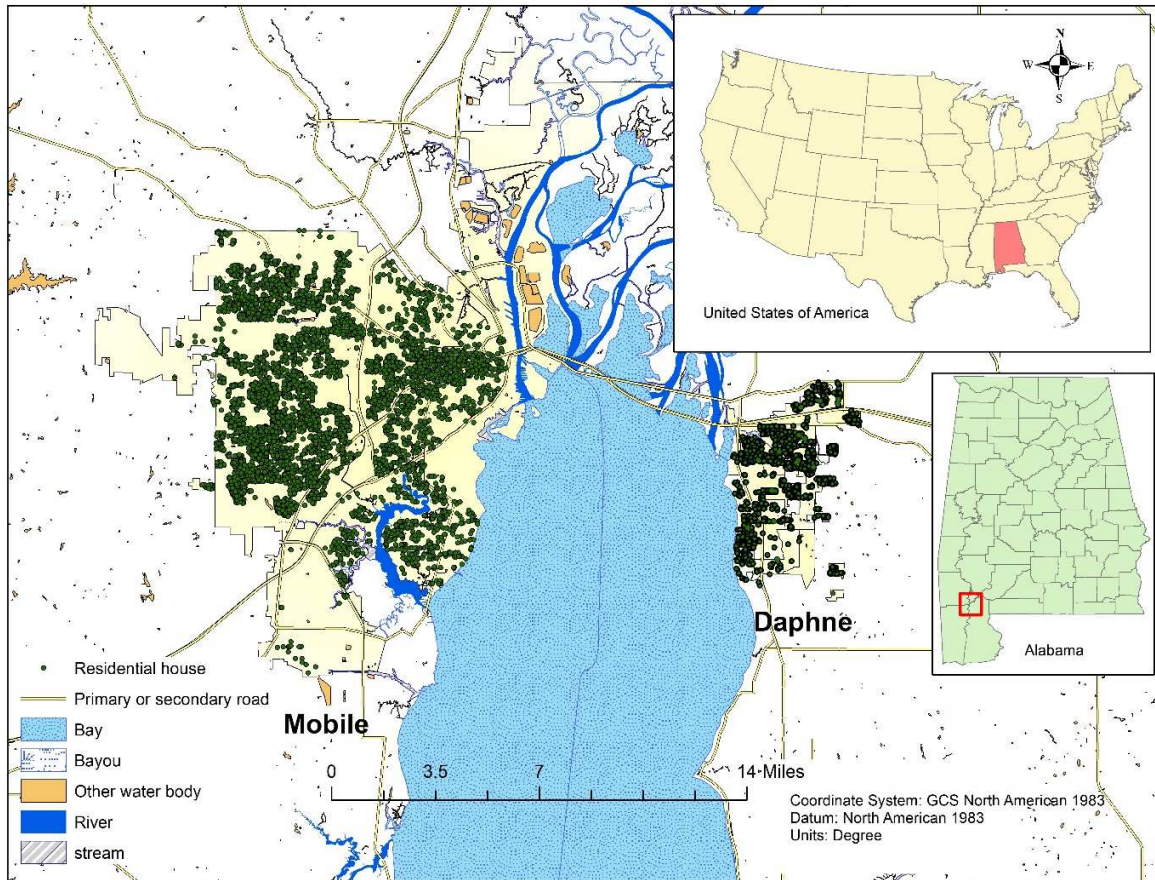
129 **1.2 Material and methods**

130 **1.2.1 Study area**

131 The study area consists of the city limits of Mobile and Daphne in Alabama, U.S. (Figure
132 1). Both cities are adjacent to Mobile Bay and are less than 65 miles to the Gulf of Mexico. Dog
133 River, Fish River, Fowl River, Mobile River, and Spanish River are some of the largest rivers in
134 the study area, with Mobile Bay and D'Olive Bay serving as popular tourist destinations.

135 Of Mobile's total area (466 km²), 44.74% is commercially and residentially developed,
136 while 19.36% is covered by wetlands (Homer et al. 2015). Despite a 2.05% decrease in
137 population from 2000 to 2010, the number of housing units increased by 3.17% (U.S. Census
138 Bureau 2012). During 2000-2010, the number of households remained relatively flat (+0.65%)
139 and a median household income increased by 17.84%, suggesting an increasing housing demand.

140 Of the 45.24 km² within Daphne's city limits, 3.18 km² are water area, most of which lies
141 in Mobile Bay (U.S. Census Bureau 2012). The city area is 50.56% in residential and
142 commercial development, whereas 8.74% is covered by wetlands (Homer et al. 2015). As of
143 2010, the population of Daphne was 21,570, which represented a 30.09% increase since 2000
144 with a corresponding 40.03% increase in housing units (U.S. Census Bureau 2012). The number
145 of households increased by 35.44%, whereas a median household income increased by 13.76%.
146 Thus, the fast rate of growth in housing units suggested an increasing loss of open space due to
147 land development.



148

149 Figure 1. Study area in the Gulf of Mexico: Mobile (Left) and Daphne (Right), Alabama, U.S.

150

151 1.2.2 Data collection

152 MLS data were obtained from Gulf Coast Multiple Listing Service, Inc. The MLS dataset
 153 included complete information on house prices and their structural characteristics. Data were
 154 drawn for the period 2001 to 2015. Data from 2001 was used because it was the earliest data
 155 available. During the study period, the study area experienced certain events that might affect the
 156 house prices. For example, Mississippi and Alabama experienced devastating losses due to
 157 extreme weather events such as Hurricane Ivan in 2004 and Hurricanes Dennis, Rita, Wilma and
 158 Katrina in 2005. More than 275,000 housing units were lost in Mississippi and Alabama due to

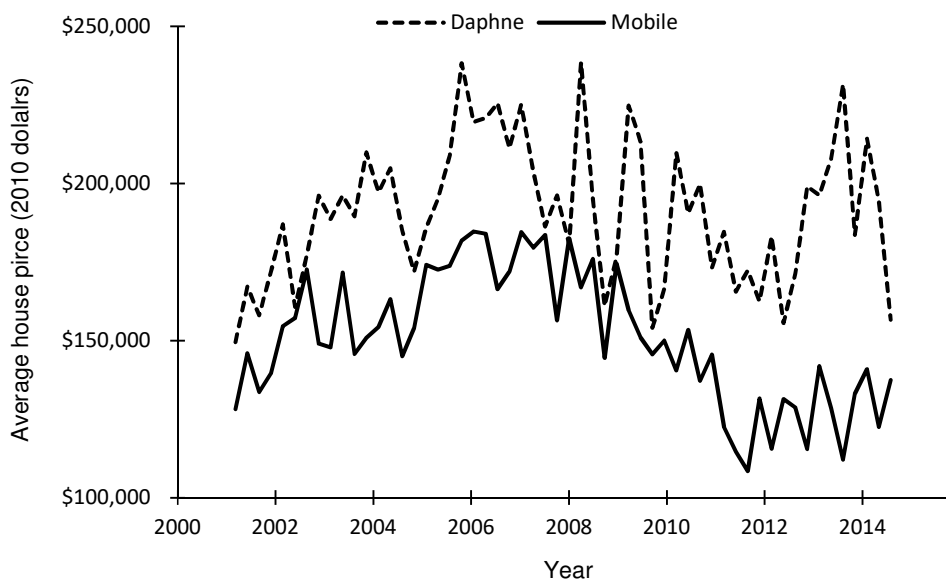
159 hurricanes Katrina, Rita, and Wilma (MASGC, 2012). Thus, using housing data for an extended
160 period helped control for these market fluctuations by averaging their effects.

161

162 **1.2.3 Data preparation**

163 Houses marked as unsold in the MLS dataset were omitted from the analysis. As the
164 house data were not georeferenced, the dataset was transformed to a Geographic Information
165 System (GIS) format to facilitate geospatial analysis. The location of each sold house was
166 geocoded in ArcMap using an Alabama address locator obtained from the TIGER/Line
167 geodatabase maintained by the U.S. Census Bureau. A total of 15,463 house records were
168 matched for Mobile and 3,748 house records for Daphne. However, 5,018 house records from
169 Mobile and 290 from Daphne did not have complete information about house structural
170 characteristics and, therefore, were not used in the study. The final dataset contained 10,445
171 house sale records for Mobile and 3,458 sale records for Daphne, resulting in a total of 13,903
172 records for both cities. House prices were then expressed in 2010 U.S. dollars using housing
173 price index obtained from Federal Housing Finance Agency (FHFA 2017) to control for inflation
174 and real estate market fluctuations, and to be comparable with neighborhood data for the 2010
175 Decennial Census.

176 A trend in housing prices for Mobile and Daphne expressed in 2010 U.S. dollars is
177 illustrated in Figure 2. During 2000-2015, housing prices in Daphne were relatively higher than
178 in Mobile with prices increasing until 2007 and then falling sharply. The global economic
179 recession and sharp decline in housing starts were the major reasons for this downturn (Mian &
180 Sufi, 2010). Mobile followed a similar trend as Daphne in terms of housing prices. An average
181 housing price in Daphne was approximately \$200,000, whereas in Mobile it was \$150,000.



182

183 Figure 2. Average house prices in Daphne and Mobile, U.S. during 2001-2015.

184

185 Geospatial data such as location of regional airports, primary and secondary roads,
 186 railroads, parks, waterfront types, shopping centers, hospitals, and schools were obtained from
 187 the TIGER/Line database maintained by the U.S. Census Bureau and open data sources
 188 including usa.com, data.gov, expertGPS, and City of Mobile. Euclidean distances from a sold
 189 house to the nearest geospatial features were computed using a proximity tool of ArcMap 10.3.1.

190

191 **1.2.4 Econometric model**

192 The HPM is a widely used nonmarket valuation technique. According to the model,
 193 goods and services can be viewed as bundles of attributes (McConnell & Walls 2005). For
 194 instance, a house is characterized by many structural attributes affecting its price, such as age,
 195 size, number of bedrooms, bathrooms, and presence of a garage. Additionally, a house price
 196 might be affected by environmental attributes such as proximity to waterfronts and public parks

197 on waterfronts, and neighborhood attributes such as proximity to schools, shopping centers, and
 198 hospitals (McConnell & Walls, 2005). Buyers often value the presence of environmental
 199 amenities and neighborhood attributes and are willing to pay higher prices for houses up to a
 200 point where the marginal cost of access to or being close to such amenities equals their marginal
 201 benefits (Flores, 2003). Even though consumers do not purchase such amenities directly, their
 202 values to consumers are reflected in how much they paid for the houses (Taylor, 2003).
 203 Homebuyers thus buy a bundle of house attributes (structural, neighborhood, and environmental)
 204 when they buy a property (Lancaster 1966). As a result, the total property sale price is a function
 205 of market and nonmarket attributes that characterize the house: $P_h=(Z)$. A particular house can
 206 be described by its structural, locational, and environmental attributes represented as
 207 $Z=(Z_1,Z_2, \dots,Z_n)$, where $i=1, 2\dots n$ represents the types of attributes associated with a property
 208 and P stands for the hedonic price function. Price of a house is thus a function of vector Z .

209 The implicit price of any attribute characterizing the house is conceived as a partial
 210 derivative of the hedonic price function (Morancho, 2003). It is called an implicit price because
 211 the price is implicit in the total house price and represents the marginal price of an attribute
 212 characterizing the house (Taylor, 2003). However, the attribute is not purchased directly; rather,
 213 its monetary value is revealed through the price a buyer pays for the house, of which a particular
 214 attribute is a part. Mathematically, an implicit price of a specific house attribute, keeping other
 215 characteristics constant, can be expressed as:

$$216 \quad P_{Z_i}(Z_i,Z_{i-1}) = \frac{\partial P(Z)}{\partial Z_i} \quad (1)$$

217 where Z_i is vector of house attributes.

218 An ordinary least squares (OLS) regression model was used to estimate marginal implicit
 219 prices of structural, neighborhood, and environmental attributes associated with houses in

220 Mobile and Daphne, U.S. Several model types were tested including simple linear mode,
 221 reciprocal, logarithmic, and logarithmic reciprocal based on goodness of fit (R^2). Finally, a semi-
 222 log model was selected as represented in Equation 2. A separate HPM model was developed for
 223 each city because the housing markets may differ between the cities, while houses in a particular
 224 city may share similar structural and neighborhood attributes. Sales price was used as the
 225 dependent variable. Independent variables included house structural, neighborhood, and
 226 environmental attributes.

$$227 \quad \ln H_i = \beta_0 + \sum \beta_j S_{ij} + \sum \beta_k N_{ik} + \sum \beta_l E_{il} + \varepsilon_i \quad (2)$$

228 where $\ln H_i$ is the natural log of the i^{th} house price (2010 U.S. dollars), S_{ij} represents the j^{th}
 229 house's structural attributes, N_{ik} stands for the k^{th} neighborhood attributes, E_{il} represents the
 230 l^{th} environmental attributes, β 's are the corresponding parameters to be estimated, and ε_i is the
 231 error term.

232 Descriptions of the three groups of independent variables used in the study reflecting
 233 structural, neighborhood, and environmental attributes are presented in Table 2. Structural
 234 attributes included the number of bedrooms and full bathrooms; house age at sale; presence of a
 235 garage, fire-place, and porch; and square footage of the house. Neighborhood attributes were
 236 further categorized into two groups of variables: socioeconomic attributes and
 237 government/municipal/locational services. Socioeconomic attributes included population density,
 238 percentage of families below the poverty line, median income, median resident age, percentage
 239 of vacant houses, and percentage of houses used for recreational or seasonal purposes. These
 240 variables were selected to reflect the level of development, economic condition, and prosperity
 241 of the neighborhood. These data were collected at the Census Block Group level.
 242 Government/municipal/locational services included distance-based variables representing

243 house's proximity to the nearest public school, active railroad, primary or secondary road,
244 hospital, airport, and shopping center.

245 Environmental attributes included variables representing a house's proximity to the
246 nearest waterfront such as river, stream (creek, branch, and fork), bay, bayou, and other water
247 bodies (lake, pond, reservoir, and lagoon). Streams were defined as natural flowing waterways
248 with an intricate network of interlacing channels, artificial waterways constructed to transport
249 water (e.g., to irrigate or drain land), and natural or artificial waterways to connect two or more
250 bodies of water or for watercraft transportation (U.S. Census Bureau, 2018). In Mobile, an
251 average distance to the nearest stream was approximately 3 km. Some of the streams/creeks
252 included Rabbit, Threemile, Chicksaw, and Black Creek. Similarly, in Daphne an average
253 distance to the nearest stream/creek was about 2.5 km. D'Olive Creek is one of the most popular
254 creeks in Daphne. Rivers included natural flowing waterways that did not contain the
255 characteristics of streams (U.S. Census Bureau, 2018). An average distance from a residential
256 property to a river was approximately 6 km. Mobile, Tensaw, Spanish, and Dog River were some
257 of the rivers included in the analysis. Bays included bodies of water partially surrounded by land
258 (U.S. Census Bureau, 2018). An average distance to a bay in Mobile was 9 km, whereas in
259 Daphne it was 3 km. Mobile, Polecat, Delvan, and D'Olive Bay are examples of bays included in
260 the model. Bayous included slowly moving water and marshy land. Environmental attributes also
261 included proximity to the nearest public park. An average distance to a bayou was 8 km and
262 Black, Sara, Big Canot, Catfish, Alligator, Greenwood, Rattlesnake, and Shell bayous are
263 examples of bayous found in Mobile. Other water bodies included lakes, reservoirs, and lagoons.
264 An average distance to nearest other waterbodies in Mobile and Daphne was approximately 1 km
265 (Table 2).

266 The HPM usually suffers from heteroscedasticity and multicollinearity problems
267 (Poudyal et al., 2009; H. A. Sander & Polasky, 2009; H. Sander, Polasky, & Haight, 2010) and,
268 therefore, this model was tested for heteroscedasticity and multicollinearity. The presence of
269 heteroscedasticity was tested for using White's test at a 5% significance level (Greene, 2012). If
270 heteroscedasticity existed, robust standard errors (heteroscedasticity-consistent standard errors)
271 were used to test parameter significance. Variance inflation factor (VIF) was used to test the
272 model for the presence of multicollinearity (Greene, 2012). Variables with a VIF value greater
273 than 10 were regarded as problematic and were omitted from the model.

274 Some of the independent variables were transformed to logarithmic, square, and product
275 forms as some of the variables required non-linear specification (Taylor, 2003). Thus, the final
276 model consisted of logarithmically-transformed variables for house price, square footage,
277 population density, income, and all distance-related variables, whereas all other variables were
278 not transformed. Effects of environmental attributes on house prices were measured by six
279 variables in Mobile and four variables in Daphne because Daphne lacked rivers and bayous
280 within the city limits. In addition, Daphne lacked some neighborhood attributes such as airports
281 and hospitals within the city limits. Thus, the final models consisted of 24 variables for Mobile
282 and 20 variables for Daphne.

283 Table 2. Definition and descriptive statistics of variables used to quantify the monetary value of
 284 waterfront in Mobile and Daphne, Alabama, U.S.

Variables	Definitions	Mobile		Daphne	
		Mean	Std. Dev.	Mean	Std. Dev.
Dependent variable					
House price ^a	House sales price (thousand U.S.\$)	155.74	134.48	196.06	137.74
Structural attributes					
Bedrooms	Number of bedrooms	3.26	0.73	3.43	0.64
Full bathrooms	Number of full bathrooms	1.98	0.68	2.27	0.58
Stories	Number of stories	1.20	0.39	1.18	0.57
Garage	Dummy variable: 1 if the house had a garage, 0 otherwise	0.63	0.48	0.93	0.26
Fireplace	Dummy variable: 1 if the house had a fireplace, 0 otherwise	0.65	0.48	0.84	0.36
Porch	Dummy variable: 1 if the house had porch, 0 otherwise	0.80	0.40	0.90	0.30
Area ^a	Square footage of the house (thousand)	1.94	0.81	2.16	0.70
House age	House age at the date of sale	72.01	270.22	15.62	13.13
Neighborhood attributes					
Population density ^{ab}	Number of people per square mile (thousand)	2.77	1.44	1.31	0.97
Poverty ^b	Percentage of families below the poverty line	11.50	11.45	6.52	4.96
Vacancy ^b	Percentage of vacant houses	8.56	4.79	10.94	2.97
Recreational ^b	Percentage of houses used for seasonal, recreational, or occasional purposes	0.41	0.42	1.44	1.07
Median age ^b	Median age of residents	38.26	5.97	39.96	4.37
Income ^{ab}	Median household income (thousand \$)	52.49	21.25	70.26	22.92
Airport ^a	Distance to the nearest airport	5.53	1.87	NA	NA
Hospitals ^a	Distance to the nearest hospital	4.15	2.50	NA	NA
Railroad ^a	Distance to the nearest active railroad	3.08	2.12	14.81	1.53
Road ^a	Distance to the nearest primary or secondary road	2.18	1.80	1.06	0.74
School ^a	Distance to the nearest public school	1.41	0.96	1.61	0.71
Shopping ^a	Distance to the nearest shopping center	1.01	0.74	2.74	1.31
Environmental amenities					
Park ^a	Distance to the nearest public park	0.87	0.62	1.79	1.15
Stream ^a	Distance to the nearest stream	2.77	1.37	2.45	1.75
River ^a	Distance to the nearest river	6.02	3.28	NA	NA
Bay ^a	Distance to the nearest bay	9.29	3.94	3.01	1.62
Bayou ^a	Distance to the nearest bayou	8.09	3.21	NA	NA
Water ^a	Distance to the nearest other water body	1.22	0.60	1.57	0.73

285 Note: ^a represents a log-transformed variable. ^b represents a value reported at a census block group level.
 286 Distance-related variables were measured in thousand meters (m) from the house location to the nearest
 287 corresponding feature. NA indicates that a variable was not applicable.

288 **1.3 Results**

289 White's test indicated the presence of heteroscedasticity at a 5% significance level and,
290 therefore, robust standard errors were used in the analysis. The VIFs for variables in both models
291 were not greater than 10, indicating multicollinearity was not a major concern in the model,
292 except for house age and its squared term. However, these two variables were not omitted from
293 the models because using a squared term permits a nonlinear relation between age and house
294 price and these two variables have been commonly used in most previous HPM studies (e.g.
295 Troy & Grove, 2008; Poudyal et al., 2009; Nilsson, 2014). In addition, multicollinearity resulting
296 from the use of a squared term does not affect the probability values of the variable and does not
297 have adverse effect on the model (Allison, 2012; Morrissey & House, 2018; Pokharel 2019). The
298 Housman test indicated that the model did not suffer from endogeneity. Coefficients of
299 determination (R^2) were 0.67 and 0.78 for Mobile and Daphne models, respectively (Table 3).
300 These findings indicated that independent variables included in the models accounted for 67.00%
301 and 78.00% of the variation in house prices for Mobile and Daphne, respectively. The F-statistics
302 for the models were 840.3 ($p < 0.001$) (Mobile) and 569.19 ($p < 0.000$) (Daphne) suggesting that
303 the models fit the data better than models with an intercept only.

304 Table 3. Estimates from hedonic price method (HPM) models used to estimate the monetary
 305 value of waterfront in Mobile and Daphne, Alabama, U.S.

Variables	Mobile			Daphne		
	Parameter estimates	White Std. error	VIF	Parameter estimates	White Std. error	VIF
Intercept	0.564*	0.337	0.000	4.844***	0.525	0.000
Bedrooms	-0.075***	0.021	1.764	-0.023**	0.010	2.046
Full Bathrooms	0.138***	0.011	2.504	0.087***	0.013	2.172
Garage	0.103***	0.009	1.127	0.150***	0.027	1.151
Fireplace	0.100***	0.012	1.428	0.067***	0.014	1.237
Porch	0.108***	0.012	1.097	0.058***	0.017	1.032
ln(Area)	1.043***	0.032	3.519	0.871**	0.030	3.140
House age	-0.002***	0.000	26.165	-0.006**	0.002	1.860
House age squared	0.000***	0.000	25.845	0.000	0.000	1.306
ln(Population density)	0.026**	0.011	1.811	-0.083***	0.016	8.395
Poverty	-0.005***	0.001	2.386	0.011***	0.002	1.981
Vacancy	-0.006***	0.002	1.981	-0.020***	0.003	4.544
Recreation	0.131***	0.013	1.458	-0.004	0.010	6.374
Median age	-0.004***	0.001	1.585	NA	NA	NA
ln(Income)	0.291***	0.020	2.713	0.129***	0.034	4.068
ln(Airport)	0.104***	0.016	1.936	NA	NA	NA
ln(Hospital)	-0.106***	0.012	2.495	NA	NA	NA
ln(Road)	0.022***	0.006	2.272	-0.042***	0.007	1.835
ln(School)	0.020**	0.008	1.702	0.032***	0.010	1.773
ln(Shopping centers)	-0.007	0.007	1.298	0.119***	0.018	4.431
ln(park)	-0.003	0.005	1.269	0.000	0.018	3.349
ln(Stream)	-0.026***	0.009	1.708	0.028***	0.010	4.227
ln(River)	-0.036***	0.013	4.265	NA	NA	NA
ln(Bay)	-0.036*	0.019	4.215	-0.127***	0.017	4.748
ln(Bayou)	0.070***	0.015	3.519	NA	NA	NA
ln(Water)	0.002	0.007	1.309	-0.076***	0.010	2.047
F value	840.300***			569.190***		
R ²	0.669			0.777		
Adj. R ²	0.668			0.776		
N	10445.000			3287.000		

306 Note: The dependent variable is ln(house price), *p<0.10, **p<0.05, ***p<0.01. NA indicates that a
 307 variable was not applicable to the specific model.
 308

309

310 **1.3.1 House structural and neighborhood attributes**

311 Structural attributes had both positive and negative effects on house prices. The number
312 of bedrooms was negatively related with house prices in both cities. An increase in the number
313 of bedrooms by one was associated with a decrease in house prices of 7.50% ($p=0.000$) in
314 Mobile and 2.30% ($p=0.022$) in Daphne. However, the number of full bathrooms was positively
315 related with house prices. A garage, fireplace, or porch were positively associated with price in
316 both cities. Similarly, square footage of a house was also positively related with house prices and
317 the parameter estimates can be interpreted in terms of elasticity as both independent and
318 dependent variables were log-transformed. In Mobile, a 1% increase in square footage
319 corresponded with a 1.04% ($p<0.001$) increase in house prices, whereas in Daphne the price
320 increase was 0.87% ($p<0.001$). House age was negatively related with house prices. A positive
321 sign for house age squared indicated that house prices had a non-linear relationship with house
322 age such that house prices decreased at a decreasing rate with age ($p<0.001$) in Mobile; however,
323 the inflection point, where houses prices would begin to increase with age, was beyond a general
324 house life span. The relationship between house age and price in Daphne was strictly decreasing.

325 Neighborhood variables demonstrated the relationship between locality and house prices.
326 Population density was positively related with house prices in Mobile and negatively related in
327 Daphne. A 1% increase in population density was associated with a 0.03% ($p=0.013$) increase in
328 house prices in Mobile and a 0.08% ($p<0.000$) decrease in Daphne. The percentage of families
329 with household income below the poverty line was significantly related to house prices in the
330 two cities. In both cities, there was a negative relationship between vacancy house prices. The
331 percentage of houses used for recreational purposes was positively related with house prices in
332 Mobile as an increase in the number of recreational houses by 1% increased house prices by

333 13.10% ($p < 0.001$); no such relationship was found in Daphne ($p > 0.10$). Income had a positive
334 relationship with house prices in both cities where a 1% increase in median household income
335 was associated with 0.29% ($p < 0.001$) and 0.13% ($p < 0.001$) price increases in Mobile and
336 Daphne, respectively. In Mobile, the distance to an airport was positively associated with house
337 prices, whereas the distance to a hospital was negatively associated with house prices. The
338 distance to the nearest primary or secondary road had a positive association with house prices in
339 Mobile and a negative relationship in Daphne. The distance to the nearest public school was
340 positively associated with house prices for both cities.

341

342 **1.3.2 Environmental attributes**

343 The effect of environmental amenities on house prices varied across waterfront types.
344 Distance to the nearest public park was not related to house prices in either city. A 1% increase
345 in distance to the nearest stream, river, or bay in Mobile was associated with a decrease in house
346 prices of 0.03% ($p = 0.003$), 0.04% ($p = 0.004$), and 0.04% ($p = 0.058$), respectively. The distance to
347 the nearest bayou was positively related with house prices in Mobile where a 1% increase in
348 distance reflected an increase in house prices of 0.07% ($p < 0.001$). Distance to other water bodies
349 was not related to house prices in Mobile. As mentioned earlier, there were no rivers and bayous
350 in Daphne. Thus, only distance to bays, streams and other waterbodies were measured. In
351 Daphne, a 1% increase in the distance to the nearest bay and other water body was associated
352 with a decrease in house prices of 0.13% ($p < 0.001$) and 0.08% ($p < 0.001$), respectively.
353 However, a 1% increase in the distance to a stream increased house prices by 0.03% ($p = 0.006$) in
354 Daphne.

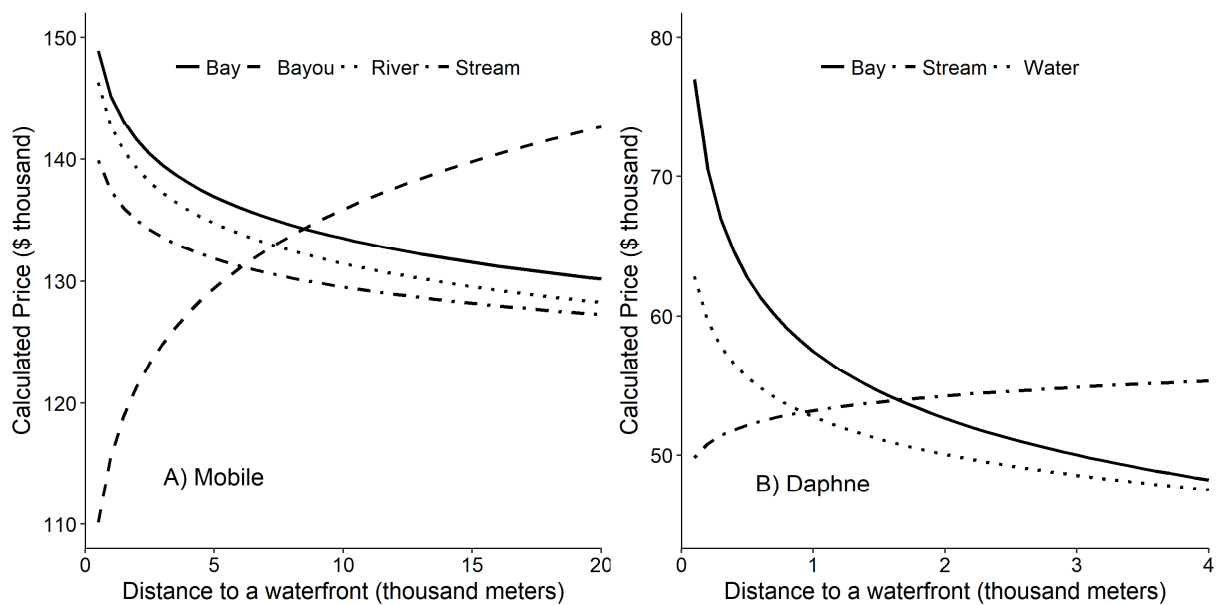
355 The marginal implicit price was calculated by taking a partial derivative of hedonic price
356 function using Equation 1. The marginal implicit price of each statistically significant attribute
357 was evaluated at the mean house prices of \$155,744 and \$196,063 in Mobile and Daphne,
358 respectively and a distance of one mile (1609.34 meters) to each waterfront type (Table 4). The
359 marginal implicit prices of proximity to the nearest stream were -\$2.49 and \$3.47 per meter in
360 Mobile and Daphne, respectively, resulting in a price decrease of \$2,490 and an increase of
361 \$3,470 for each 1-km increase in distance to a stream in Mobile and Daphne, respectively.
362 Similarly, a marginal implicit price of proximity to a river in Mobile was -\$3.46 per meter
363 suggesting a 1-km increase in distance to a river decreased house prices by \$3,460. The marginal
364 implicit price of distance to the nearest bay was -\$3.53 and -\$15.46 per meter in Mobile and
365 Daphne, respectively. This implied that a 1-km increase in distance to a bay was associated with
366 decreases in house prices of \$3,530 and \$15,460 in Mobile and Daphne, respectively. The
367 marginal implicit price of proximity to a bayou was \$6.79 per meter in Mobile with a
368 corresponding increase in house prices of \$6,790 for a 1-km distance increase. Similarly, the
369 marginal implicit price of proximity to the nearest other water body was -\$9.25 per meter in
370 Daphne and corresponded to a \$9,250 decrease in house prices for a 1-km distance increase.

371 Table 4. A marginal implicit price of proximity to environmental attributes in Mobile and
372 Daphne, Alabama, U.S.

Variables	Parameter estimates		Marginal implicit price (\$ per km)	
	Mobile	Daphne	Mobile	Daphne
Distance to bay	-0.036	-0.127	-3,530	-15,460
Distance to river	-0.036	NA	-3,460	NA
Distance to stream	-0.026	0.028	-2,490	3,470
Distance to other water bodies	NS	-0.076	NS	-9,250
Distance to bayous	0.070	NA	6,790	NA

373 Note: NA indicates that a variable was not applicable; NS indicates that a variable was
374 statistically not significant

375 The relationships between distances to various waterfront types and house prices, holding
 376 all other variables constant, are presented in Figure 3. A downward sloping relationship implied
 377 that houses located nearer to a given waterfront type sold at higher prices than those located
 378 farther away, whereas an upward slope suggested that houses farther away from a specified
 379 waterfront type sold at higher prices. In Mobile, the relationship was downward sloping for bays,
 380 rivers, and streams and upward sloping for bayous. In Daphne, the relationships for streams was
 381 upward sloping, whereas bays and other water bodies were downward sloping. A downward
 382 sloping relation implied that houses in Mobile located near a bay, river, or stream sold at higher
 383 prices than those farther away. Similarly, houses near bays and other water bodies sold at higher
 384 prices in Daphne. However, in the case of bayous in Mobile and streams in Daphne, houses
 385 located farther away sold at higher prices.



386
 387 Figure 3. Relationships between house prices and distance to various waterfront types in
 388 Alabama, U.S.: A) Mobile, B) Daphne.

389 **1.4 Discussion**

390 Structural, neighborhood, and environmental attributes were critical factors considered in
391 residential home purchases. With the exception of number of bedrooms, all structural variables
392 included in the analysis had a positive and statistically significant association with house prices.
393 A negative relationship of the bedroom number with house prices possibly indicated that buyers
394 preferred a less fragmented interior space (Bowman, Thompson, & Colletti, 2009). It could also
395 be that people place more value on the relative size of bedrooms than on number of rooms. The
396 mean numbers of bedrooms for both cities were greater than the average number of bedrooms in
397 the U.S. (U.S. Census Bureau, 2017). Only about 20% of people in the U.S. prefer more than
398 three bedrooms and as a result, additional number of bedrooms had negative impact on house
399 prices. The relationships of other structural variables (full bathroom number; presence of a
400 garage, fireplaces, and porches; house square footage, and age) with house prices were consistent
401 with a number of hedonic studies (e.g., Bolitzer and Netusil 2000, Geoghegan 2002; Poudyal et
402 al. 2009; Sander et al. 2010). While most house structural attributes were associated with higher
403 house prices, consistent with other studies, house age was linked with lower house prices. For
404 example, Bolitzer and Netusil (2000), Poudyal et al. (2009), and Sander et al. (2010) found a
405 drop from 0.2% to 0.7% in house prices for each one year increase in a house age. However, a
406 house price may have a nonlinear relationship with its age. For instance, Sander et al. (2010)
407 reported a decrease in house price up to 88 years and an increase afterwards. Thus, a level of
408 structural improvements such as number of full bathrooms, presence of a garage, and house
409 living area were significant factors in determining a house value.

410 Most neighborhood attributes affected house prices, yet the coefficient signs differed
411 between some Mobile and Daphne attributes. For instance, population density was positively

412 related with house prices in Mobile and negatively related in Daphne. Geoghegan et al. (2003)
413 stated that population density could have two opposite effects on house prices. On one hand,
414 population density can be regarded as a measure of congestion and thus a negative externality.
415 On the other hand, it can serve as a proxy for density of other goods and services that can attract
416 people and therefore increase house prices. The mean population density in Mobile was more
417 than double that of Daphne suggesting a higher demand of other goods and services such as
418 shopping centers and recreationally-used houses. Similarly, distance to the nearest primary or
419 secondary road had a positive association to house prices in Mobile and negative relationship in
420 Daphne. A possible explanation for this relationship might be that Mobile is characterized by
421 substantial industrial and commercial development and, therefore, residents chose to live close to
422 nearby roads for easy and quick access to their work places. However, Daphne is characterized
423 mostly by residential development and people may consider nearby roads as a negative
424 externality because of noise and pollution. The percentage of vacant houses was negatively
425 related with house prices in both cities. Poudyal et al. (2009) and Klaiber and Phaneuf (2010)
426 also reported that house prices decreased as the percentage of vacancies increased. A possible
427 explanation for this trend might be that consumers view increases in the percentage of vacant
428 houses as a safety issue, an unpleasant living environment, and/or a health hazard resulting in a
429 negative impacts on house prices (Heynen, Perkins, & Roy, 2006; Woolle & Rose, 2004). In
430 addition, the presence of vacant houses might be an indication of house oversupply resulting in
431 decreased house prices. Thus, neighborhood attributes played an important role in determining
432 residential property values and required market analysis before arriving at fair market value of
433 any residential property.

434 Most environmental attributes, which were the focus of this study, were related with
435 house prices. Residents in Mobile and Daphne were willing to pay higher prices for houses in the
436 vicinity of most waterfront types. Coefficient signs for environmental attributes were consistent
437 with previous studies. For instance, a decrease in the distance to a wetland by 1,000 feet was
438 related with an increase in property value of \$436.17 (Bin, 2005). Similarly, Sander and Polasky
439 (2009) reported a relatively small but statistically significant increase in house prices as distance
440 to a lake decreased. Chen and Jim (2010) reported that a 1,000-meter increase in distance to a
441 bay reduced house prices by 0.70%. Distance to a park was not a significant factor for house
442 prices in either city. The reason might be that when both parks and waterfronts were available as
443 open space, coastal residents preferred waterfronts more than parks because of their uniqueness
444 compared to non-coastal areas (also see Mahan et al. 2000).

445 Some of the specific waterfront types, however, had positive coefficients indicating that
446 residents preferred houses located away from these waterfronts. For example, distance to a bayou
447 was positively related to price indicating that houses further away from bayous had higher selling
448 prices. This might be because residents considered bayous as a negative externality due to their
449 specific microclimatic effects and less appealing nature (i.e., marshy character and slowly
450 moving water). Another explanation might be that bayous provide habitat for animals such as
451 alligators and caimans and residents may have considered living in close vicinity to a bayou as
452 unsafe.

453 In terms of waterfront types, distance to a bay had the largest marginal implicit price
454 followed by distance to a river and a stream. This suggest that coastal residents preferred to live
455 near a larger-sized body of water, perhaps because of more opportunities for recreation. In
456 addition, the estimated marginal implicit prices for distance to different waterfront types were

457 larger in Daphne than in Mobile. A possible explanation might be related to the total water
458 coverage in each city. Only 7.04% of the area within the city limits is covered by water in
459 Daphne in contrast to 22.62% in Mobile. Residents of Daphne thus might place a higher value on
460 proximity to waterfronts as a scarce environmental amenity. Furthermore, the mean distance to a
461 bay and other water bodies in Daphne was 3,008 and 1,572 meters, respectively, in comparison
462 to 9,292 and 1,222 meters in Mobile. Thus, the marginal price of proximity to waterfronts was
463 larger in Daphne than in Mobile because of proximity and easier access to waterfronts.

464 In addition, a substantial portion of the area within Mobile city limits fronting Mobile
465 Bay and Mobile River is dedicated to the commercial/industrial activity associated with the Port
466 of Mobile. None of these areas is zoned for residential use, whereas Daphne waterfronts are
467 dedicated for residential use. Therefore, people who prefer waterfront living tend to choose
468 Daphne over Mobile. The coefficient for distance to streams was positively associated with
469 house prices in Daphne indicating that houses located farther away from streams had higher
470 selling values. This result might well be attributed to the presence of small streams, such as
471 Daphne, D'Olive, and Tiawasee Creeks, which flow east-west in the northern section of the city
472 and D'Olive Bay which lies within 1,500 meters from these streams. It is possible that residents
473 were more attracted to the bay than streams, resulting in lower house prices in proximity to
474 streams and higher prices in proximity to the bay. Thus, value of waterfront varied based on its
475 type. Waterfronts that were more appealing, larger sized, and in closer proximity to residential
476 properties, such as bay and river, were valued more than ones that were smaller sized, farther
477 away or having marshy or slowly moving water such as bayou.

478 It should be noted that the implicit prices estimated in this study using HPM represent
479 only a partial component of house prices. These prices are unlikely to capture total value of

480 waterfront because house buyers may not have perceived all the values of waterfronts such as
481 those related to recreation, scenic views, and habitat. The estimates presented in this study reflect
482 only the values and services captured by changes in house prices (Mahan et al., 2000). A study
483 by Dahal et al. (2018) located in the Gulf of Mexico used contingent valuation to measure both
484 use and nonuse values of open space. The authors estimated that coastal residents' average
485 willingness to pay to preserve waterfront open space was \$90.72 per household, which translated
486 to an aggregated value of \$10.84 million. Thus, total value of waterfront might be larger than
487 estimated in this study. If so, then computing the total value of benefits associated with
488 waterfront in coastal areas remains an important area for future research. Total value estimation
489 will be important for identification of resident preferences towards open spaces and
490 incorporation of these preferences into future land-use and urban development decisions as well
491 as design of urban living areas.

492 In addition, this study examined the relationship between proximity to waterfronts by
493 type and house prices only within the city limits. However, house prices might have been
494 affected by proximity to waterfronts beyond the city limits and, thus, restricting the study area to
495 within the city limits might have reduced the quantified implicit prices. Future studies should
496 consider including attributes located outside of administrative boundaries. Doing so would better
497 enable quantification of attributes' impacts on marginal implicit price, and provide a means to
498 determine how far people are willing to travel to enjoy environmental services provided by open
499 spaces. In addition, this study estimated monetary values of waterfront types by developing
500 separate models for each city, assuming houses in a particular city shared structural and
501 neighborhood attributes, and housing market. However, future studies might consider developing
502 a spatial model such as a geographically weighted regression model to estimate how location

503 affects house prices. Finally, future studies might also focus on identifying open space features
504 people value most such as trees, shrubs, water, and grasses as well as their preferences towards
505 natural versus manicured landscapes. This information would facilitate a more precise estimation
506 of implicit prices associated with open space benefits, and help city planners make decisions
507 about balancing preservation of these spaces in face of development desired by local
508 communities.

509

510 **1.5 Conclusions**

511 The study provided insights into how different waterfront types were valued by coastal
512 community residents based on the relationship of residential properties' structural, neighborhood,
513 and environmental attributes with price. Distances to different waterfront types, in addition to
514 house structural and neighborhood attributes, were factors considered by coastal residents when
515 buying a house. With the exception of bayous in Mobile and streams in Daphne, houses near
516 waterfronts sold at higher prices than those located farther away with house price impacts
517 ranging from -\$6,790 to \$15,460 per km depending on waterfront type. Because most waterfronts
518 were highly desired by residents, and they preferred to live in proximity to these open space
519 areas, planners can use information from this study to formulate more effective and transparent
520 conservation and development strategies in budget allocations and city planning.

521 This study has several implications. Results indicated that waterfronts had substantial
522 positive price impacts on residential properties and affected the surrounding land market by
523 creating new and enhancing existing land values that translated into positive economic benefits
524 to coastal communities. Proximity to preserved waterfronts can enhance property value, if the
525 waterfronts are not intensively developed and they are carefully integrated with the

526 neighborhood. Estimates on monetary value will also be useful for federal income tax governing
527 the valuation of conservation easements and other set-asides. In addition, monetary values as
528 estimated by this study can be used in making important policy decisions such as zoning,
529 restrictions on land use, open space through public or private initiatives, and improving local
530 landscape and vegetation ordinances Maintaining visually appealing, usable, and easily
531 accessible waterfronts can be an effective way to make waterfronts lively public destinations that
532 offer not only environmental benefits, but which also create economic opportunities. With proper
533 design and sound policies promoting waterfront open spaces, demand for their benefits will
534 likely increase and lead to higher local tax revenues from properties and economic activities.

535

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