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1       **Development Practices and Ordinances as Predictors of Canopy Coverage in Florida (United**  
2   **States) Cities**

3  
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24

**Abstract:** This study compared the management practices and ordinances enacted by 43 Florida (United States) communities to assess their potential impact on tree canopy coverage. Dot-based canopy analysis was used to assess community level canopy coverage. This information was paired with each community's responses to a 2014 survey of municipal forestry management practices in the United States. Canopy coverage ranged between 17.6% and 63.3% among the communities assessed, with average of 33.7%. Two factors were significant when attempting to predict canopy coverage. Housing density had a negative impact on tree canopy ( $P$ -value = 0.0116). In contrast, ordinance designating and protecting heritage or other trees of significance resulted in a 6.7% increase in canopy coverage ( $P$ -value = 0.0476). Results of this research provide base-level data regarding urban forest cover in a range of Florida communities. More importantly, this research suggests the heritage tree protections afforded to old or large-stature urban trees has a measurable impact on tree canopy retention.

**Keywords:** ecosystem services; environmental policy; heritage trees; tree ordinance; urban forest management; urban tree canopy

1     **1. Introduction**  
2

3     Research has demonstrated that trees in urban environments provide a variety of ecosystem services (e.g.,  
4     creation of habitats) and other benefits (e.g., increased quality of life) (Roy et al., 2012). Many of these  
5     benefits are associated with the healthy leaf area of a tree (Nowak & Greenfield, 2012), making urban tree  
6     canopy cover an important measurement for estimating overall urban forest benefits. Nowak et al. (1996)  
7     defines an urban tree canopy (UTC) as the proportion of area, when viewed from above, occupied by tree  
8     crowns. Environmental benefits derived from an UTC include ecosystem services such as improved air  
9     and water quality (Nowak et al., 2007) and energy use reduction (Akbari, 2002). Economic benefits range  
10    from increased property value (Pandit et al., 2013) to reduced heating and cooling costs (Pandit &  
11    Laband, 2010). Management of urban trees can be conducted with specific UTC cover goals in mind in  
12    order to maximize benefits (Nowak & Crane, 2000; Hill et al., 2010; McPherson et al., 2011; Hauer &  
13    Peterson, 2016).

14  
15    In contrast to these benefits, urban trees have costs associated with their installation, care, and eventual  
16    removal which are shared between residents, businesses, and municipalities (e.g., planting, maintenance,  
17    or infrastructure repair) (Koeser et al., 2016; Vogt et al., 2015). Furthermore, urban trees are considered  
18    public assets in many municipalities, making an UTC subject to municipal tree management practices and  
19    ordinances (e.g., pruning cycles, planting initiatives, preservation ordinances, etc.) created by municipal  
20    staff and external groups (Hauer & Peterson, 2016). Land use ordinances and the degree to which they are  
21    enforced will impact the UTC (Elmendorf et al. 2003; Hill et al., 2010). For instance, a study of the UTC  
22    and associated management practices in the Atlanta Metropolitan Area showed that planning and zoning  
23    regulations aimed at UTC protection and quality growth were associated with an increase in canopy cover  
24    over ten years (Hill et al., 2010). Hill et al. (2010) also point out that UTC management actions should  
25    also extend to private trees, which can make up a large portion of the urban forest. In Tampa, Florida,

26 Landry and Pu (2010) found that tree canopy coverage was greater on private lots developed after the  
27 adoption of a 1974 tree protection ordinance compared to lots developed prior to the ordinance. However,  
28 regulations of private trees can cause political tensions. For example, a tree management bill was  
29 introduced to the Florida Senate in 2018 [Senate Bill (SB) 574: Tree and Vegetation Trimming and  
30 Removal] which, among other provisions, would roll back local governments' abilities to require permits  
31 for the trimming, pruning, removal, or harvesting of trees on private property in certain areas, and to  
32 require mitigation (i.e., replacement) of trees removed or harmed. This proposed bill concerned some  
33 individuals who were aware of the pattern of declining UTC across this United States (Nowak &  
34 Greenfield 2012).

35

36 A comprehensive understanding of a municipality's UTC can provide urban resource managers with  
37 baseline data to set goals, inform key stakeholders of the effects of certain management and development  
38 strategies, and subsequently improve various urban forest functions (Hill et al., 2010). In light of  
39 emerging research on management practices and the UTC, and legislation like the aforementioned Bill,  
40 we were motivated to investigate the effects Florida municipal management actions have on local UTC  
41 and how different municipalities compare. The objectives of this study were to 1. use dot-based spatial  
42 analysis to estimate canopy cover in Florida municipalities, and to 2. investigate the relationships between  
43 canopy cover and variables related to municipal development, tree management practices, and tree-related  
44 municipal ordinances gathered from a survey of municipal forestry managers.

45

## 46 **2. Materials and methods**

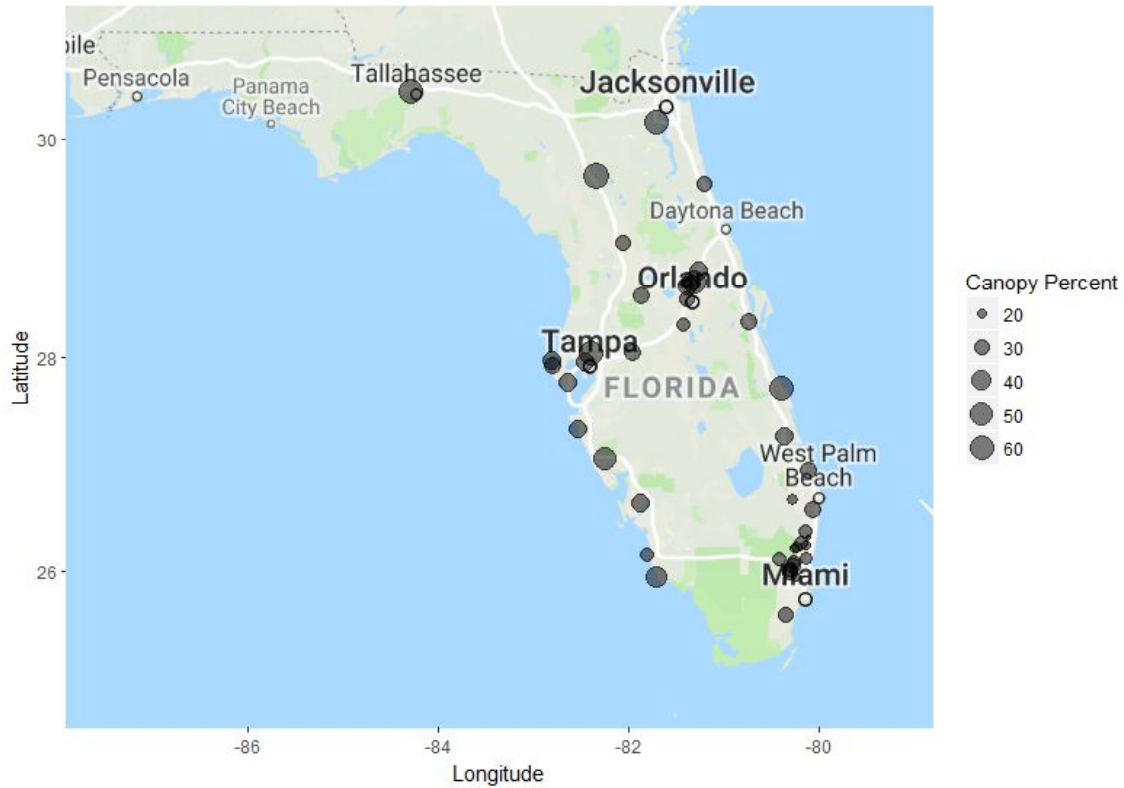
47

### 48 **2.1. Study area**

49

50 The study area consisted of 43 Florida (United States) communities who had returned a comprehensive  
51 survey on urban forest management (Hauer and Peterson, 2016). Locations of these communities are  
52 shown in Fig. 1.

53



54

55 *Figure 1. Locations of the 43 Florida (United States) communities included in this assessment of canopy*  
56 *coverage. The sizes of the points are scaled to reflect relative differences in canopy coverage.*

57

## 58 **2.2. Survey methodology**

59

60 The predictor variables used to model percent canopy coverage were derived in part from the results of a  
61 2014 survey of municipal forestry management practices in the United States (Hauer and Peterson, 2016).

62 The comprehensive, 109-question survey was sent to a stratified sample of 1727 communities in all fifty  
63 states, of which 87 were sent to Florida communities. All communities with populations over 50,000

64 received the survey and a random sample was taken for cities with populations between 25,000 and

65 49,999 (50%) and between 2,500 and 24,999 (10%). More detailed information regarding the generation

66 of the mailing list is available by referencing Koeser et al. (2016). The survey was approved through the

67 [REDACTED] Institutional Review Board (IRB) prior to study recruitment.

68

69 Following the approach outlined by Dillman et al. (2014), all communities on our mailing list received a  
70 pre-notice followed by a printed copy of the survey (i.e., with a cover letter). Non-respondents were also  
71 sent a reminder postcard, followed by a second printed survey (i.e., with a cover letter). A final email  
72 reminder was sent to any remaining non-respondents.

73

74 A second round of surveying was conducted using a truncated, 53-question survey in an attempt to reach  
75 communities that did not complete the larger set of questions. This shortened questionnaire (with cover  
76 letter) was sent non-respondents to the long-form version. A single email reminder was sent to non-  
77 respondents of the truncated survey. In addition to the communities that responded to the survey in 2014,  
78 the authors reached out to the urban foresters from Fort Lauderdale, Naples, Pompano Beach, and Temple  
79 Terrace. The authors knew these individuals personally through their local professional society (though  
80 knew little of their respective programs) and used this connection to increase the sample size for the  
81 canopy analysis. Interviews occurred July of 2018. These four individuals referenced records from 2014  
82 to answer the survey questions relevant to our canopy analysis.

83

### 84 **2.3. Dot-based canopy analysis**

85

86 The aerial imagery assessed in this study was accessed from the National Agricultural Imagery Program  
87 (NAIP; USDA, 2018). Specifically, imagery from 2015 was used to coincide with the timeframe of the  
88 survey mentioned above. Spatial resolution for the 2015 NAIP imagery was 1 m. A random point  
89 sampling method was conducted following the 'i-Tree Canopy' user guidelines  
90 (<https://canopy.itreetools.org/>) and Nowak and Greenfield (2012), which suggests the collection of 500-  
91 1,000 random survey points per community. In an effort to increase measurement confidence, we adopted  
92 the larger, 1,000 point sample size within this suggested range. Boundaries for each community assessed  
93 were provided by the American Community Survey (ACS; United States Census Bureau, 2015).

94

95 A geographic information system (ArcGIS 10; ESRI, Redlands, CA, United States) was used to import  
96 NAIP aerial imagery and generate the random points (at an average minimum distance of 5 meters).  
97 Urban tree canopy was assessed as either tree or non-tree. Each city was assessed at least two interpreters.  
98 Points where the interpreters disagreed were discarded prior to analysis, thus minimizing photointerpreter  
99 bias. Canopy percentage, agreement between/among interpreters, standard error (SE), and 95%  
100 confidence intervals were (Parmehr et al., 2016).

101

102

$$SE = \sqrt{p(1-p)/N}$$

103

Where  $p$  = number of tree points ( $n$ )/the total number of sample points ( $N$ )

104

$$95\% CI = UTC \pm SE \times 1.96$$

105

106

#### 2.4. Data analysis

107

108

A multiple linear regression model was fit using community-wide percent canopy coverage as the

109

dependent variable of interest. This analysis was conducted using the `lm()` function in R (R Core Team,

110

2016). Initially, a maximal model was fit using the explanatory variables listed in Table 1.

111

112



113 **Table 1. Initial set of variables assessed in modelling canopy coverage in Florida (United States)**  
 114 **communities. Mean/counts include data from the survey and data acquired from other sources**  
 115 **(U.S. Census Bureau, 2018; various municipal ordinance websites).**

116

Variable	Definition	Mean (Std. Dev.) or Count
Canopy coverage <sup>z</sup>	Response variable, percent (%) of city covered by tree canopy	34.4% (11.7%)
Housing density <sup>y</sup>	Housing units per square kilometer	520.4 (294.6)
Median home value <sup>y</sup>	Median value of resident-owned housing units	\$221,482 (\$155,976)
House percent since 2010 <sup>y</sup>	Percent of total housing units constructed after 2010	0.6% (0.6%)
House percent since 2000 <sup>y</sup>	Percent of total housing units constructed after 2000	20.3% (14.1%)
House percent since 1990 <sup>y</sup>	Percent of total housing units constructed after 1990	37.5% (18.7%)
Maintains Rights-of-ways (ROW) <sup>x</sup>	Who is responsible for maintaining trees in rights-of-way (e.g., street trees between sidewalk and curb/ alley trees)	Community - 23 Homeowner - 7 Joint ownership - 7 Other - 1
ISA Certified Arborist - yes <sup>x</sup>	Community employs at least one ISA Certified Arborist credential holder	Yes - 31 No - 9
Four year degree - yes <sup>x</sup>	Community employs at least one person with a four year degree related to tree care	Yes - 19 No - 21
Tree board - yes <sup>x</sup>	Community has a government-authorized board to help develop/administer tree management policy	Yes - 28 No - 11
Tree preservation ordinance <sup>xw</sup>	Community has an ordinance requiring the preservation of trees during development	Yes - 32 No - 8
Removal permit ordinance <sup>xw</sup>	Community has an ordinance restricting tree cutting on private property	Yes - 23 No - 18
Heritage tree ordinance <sup>xw</sup>	Community identified and preserves heritage/significant trees	Yes - 26 No - 15

Tree inventory - yes <sup>x</sup>	Community has a record of public trees within its jurisdiction	Yes - 25 No - 14
Canopy goal - yes <sup>x</sup>	Community has a goal for enhancing or maintaining % tree canopy coverage	Developing - 2 Yes - 14 No - 20

117 <sup>z</sup>Source: Dot-based canopy analysis

118 <sup>y</sup>Source: U.S. Census Bureau (2018)

119 <sup>x</sup>Source: Hauer and Peterson (2016)

120 <sup>w</sup>Source: Municipal websites

121

122 As missing data prevented the use of a stepwise deleting function, the `regsubsets()` function from the  
 123 *leaps* package (Lumley and Miller, 2017) was used to run and plot (by  $R^2$  value) the 20 best subsets of  
 124 our predictor variables. This plot (Fig 1.) was used to identify which variables were most commonly  
 125 associated with models having higher predictive power. A second, reduced model with housing density,  
 126 house percent since 2010, maintains ROW, ISA Certified Arborist, tree board, and ordinance: heritage  
 127 trees was run and non-significant explanatory variables were removed one-at-a-time based on P-value  
 128 (highest first). Each reduced model was compared against its preceding model using the `anova()` function  
 129 in R (R core team, 2016) to determine if there was a significant difference in fit between the two iterations  
 130 (Crawley, 2013). All determinations of statistical significance were made at an  $\alpha=0.05$  level of Type 1  
 131 error. Diagnostic plots were referenced to confirm no underlying assumptions associated with the analysis  
 132 were violated.

133

134

135       **3. Results**

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137               **3.1. Survey results for Florida cities**

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139 Overall, 667 communities responded to either the initial survey (n=513) or the truncated survey (n=154)

140 for a total response rate of 38.6% in all 50 states. In Florida, 39 of the 87 communities surveyed

141 responded leading to a statewide response rate of 44.8%, higher than that of the total survey population.

142 The initial survey, combined with the four additional communities from this study, resulting in a 49.4%

143 response rate. The survey results provided information on different aspects of urban forest management,

144 including the community, staff, management practices, and inventorying activities. When asked who is

145 legally responsible for trees in rights-of-way, 61% said the community was solely responsible. Over

146 three-quarters of the respondents said they had at least one ISA Certified Arborist on the staff, and about

147 half said they had at least one employee with a four-year degree. Over two-thirds of the responding

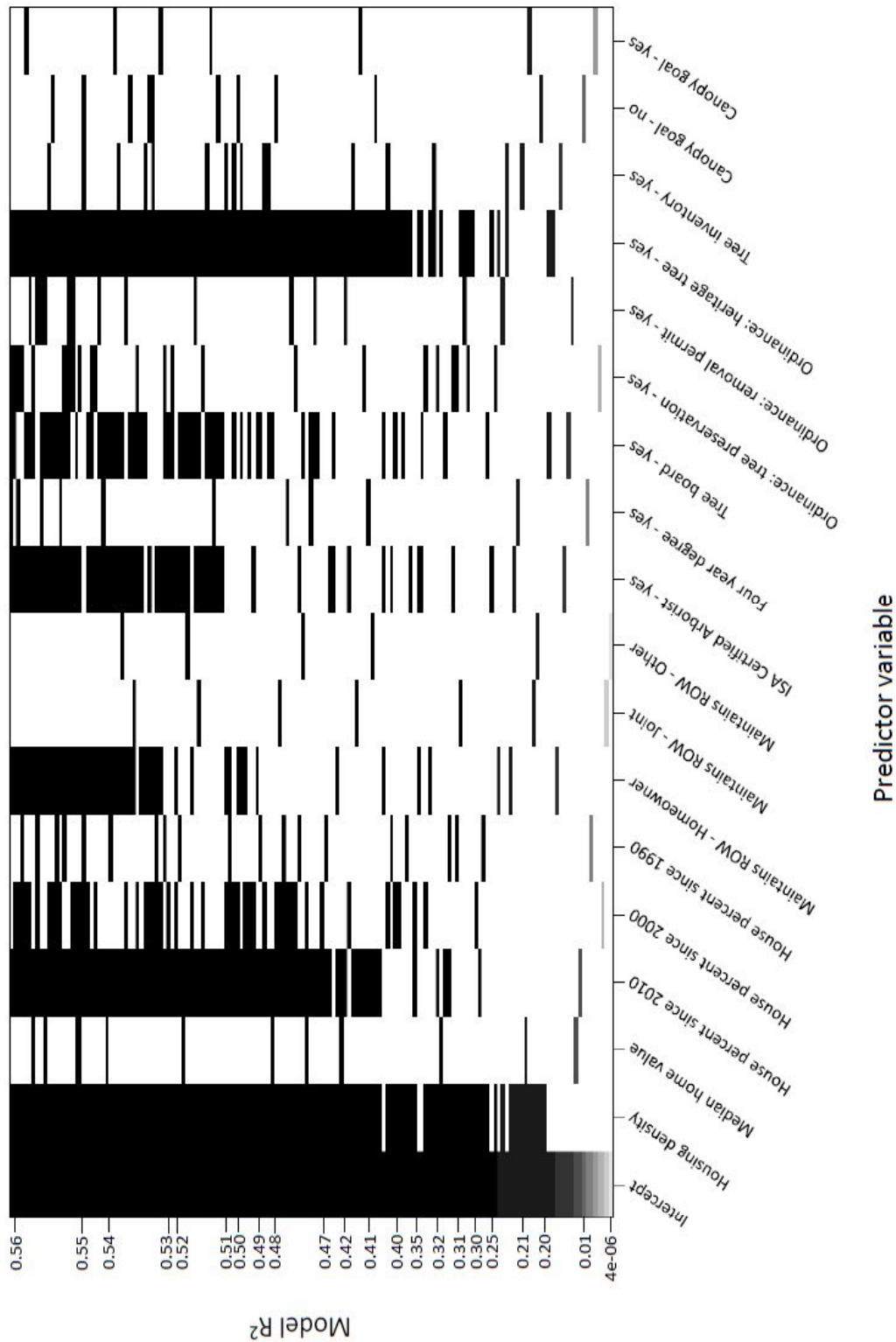
148 communities reporting having a government-organized tree board, over three-quarters had tree

149 preservation ordinances in place, and over half had ordinances that protect heritage or significant trees.

150 About half of the respondents said they had permit requirements that restrict tree cutting on private

151 property. A majority said they had a tree inventory, but only a third of those respondents said the

152 inventory was up-to-date. Finally, less than half of the communities said they had canopy cover goals.



153  
 154 *Figure 2. Coefficient of determination ( $R^2$ ) values for the various combinations of predictor variables*  
 155 *selected for initial testing. Variables most commonly associated with the highest predictive power were*  
 156 *selected for initial model simplification. Figure generated using the leaps package in R (Lumley and*  
 157 *Miller, 2017).*

158 **3.2. Canopy coverage in Florida cities**

159

160 Canopy coverage ranged from 17.6% in Deerfield Beach to 63.3% in Gainesville Florida (Table 2).

161 Average canopy coverage for the 43 assessed communities was 33.7% ( $\pm 1.5\%$ ). Agreement among our

162 interpreters ranged from 94.9% to 99.5%. Average agreement for the 43 cities was 97% (Table 2).

163

164 **Table 2. Population, percent canopy coverage, standard error, 95% confidence intervals, number of**  
 165 **interpreters, and percent agreement associated with the dot-based canopy analysis of 43 Florida**  
 166 **(United States) communities.**

167

Community Name	2014 Population	Canopy Coverage (%)	SE (%)	95% CI Lower (%)	95% CI Upper (%)	Interpreters	Agreement (%)
Gainesville	124,354	63.3	1.5	60.4	66.2	2	99.3
Tallahassee	181,376	58.7	1.6	55.6	61.8	2	99.0
Indian River Shores	4,070	57.9	1.6	54.8	61.0	2	99.4
Temple Terrace	25,495	55.6	1.6	52.5	58.7	2	98.8
Orange Park	8,412	55.0	1.6	51.9	58.1	2	97.8
Winter Springs	33,282	54.4	1.6	51.3	57.5	2	96.1
North Port	57,357	51.5	1.6	48.4	54.6	2	98.7
Marco Island	16,413	48.5	1.6	45.4	51.6	2	98.2
Altamonte Springs	42,215	40.2	1.6	37.1	43.3	2	97.2
Clearwater	107,685	37.2	1.6	34.1	40.3	3	95.6
St. Petersburg	244,769	36.9	1.5	34.0	39.8	2	97.5
Fort Myers	62,298	36.1	1.5	33.2	39.0	3	97.4
Port St. Lucie	164,603	36.1	1.5	33.2	39.0	2	97.7
Sarasota	51,917	35.8	1.5	32.9	38.7	2	97.6
Tampa	335,709	35.7	1.5	32.8	38.6	2	98.4
Sanford	53,570	35.1	1.5	32.2	38.0	2	99.1
Casselberry	26,241	34.7	1.5	31.8	37.6	3	97.2
Jupiter	55,156	34.4	1.5	31.5	37.3	2	98.4
Largo	77,648	34.2	1.5	31.3	37.1	2	97.5
Lakeland	97,422	34.0	1.5	31.1	36.9	2	97.4
Rockledge	24,926	32.9	1.5	30.0	35.8	2	98.1
Hypoluxo	2,588	32.2	1.5	29.3	35.1	2	97.4
Groveland	8,729	32.1	1.5	29.2	35.0	2	96.7
Cutler Bay	40,286	30.1	1.5	27.2	33.0	3	96.2
Cooper City	28,547	29.8	1.5	26.9	32.7	2	96.9

Palm Coast	75,180	29.7	1.5	26.8	32.6	2	98.8
Belleview	4,492	29.4	1.5	26.5	32.3	3	98.5
Orlando	238,300	29.4	1.4	26.7	32.1	2	99.3
Miramar	122,041	28.1	1.4	25.4	30.8	2	99.3
Pembroke Pines	154,750	28	1.4	25.3	30.7	2	96.8
Coconut Creek	52,909	26.9	1.4	24.2	29.6	3	94.9
Kissimmee	59,682	26.3	1.4	23.6	29.0	2	98.4
Boca Raton	84,392	26.2	1.4	23.5	28.9	3	95.5
Weston	65,333	25.6	1.4	22.9	28.3	2	97.7
Naples	20,913	25.3	1.4	22.6	28.0	2	98.0
Davie	91,992	25.1	1.4	22.4	27.8	3	97.3
Fort Lauderdale	175,599	24.5	1.4	21.8	27.2	2	99.5
North Lauderdale	41,023	22.6	1.3	20.1	25.1	2	96.8
Wellington	56,508	21.6	1.3	19.1	24.1	2	96.9
Pompano Beach	105,851	20.6	1.3	18.1	23.1	2	98.9
Tamarac	60,427	20.4	1.3	17.9	22.9	2	97.4
Miami Gardens	107,167	19.4	1.3	16.9	21.9	2	99.4
Deerfield Beach	75,018	17.6	1.2	15.2	20.0	3	98.8

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**3.3. Predictors of canopy coverage**

In conducting the model simplification process, two significant predictors of canopy coverage beyond the intercept term were significant. The first significant predictor was *housing density* ( $P < 0.0116$ ) which had a negative relationship with canopy coverage (Table 3). A 1.1% decrease in UTC would occur for each 500 housing units per km<sup>2</sup> compared to a situation with no housing units. The second predictor in our model was having a *heritage tree ordinance* ( $P < 0.0476$ ). For this variable, having some form of heritage or significant tree designation was associated with a 6.7% increase in canopy coverage (Table 3). *Housing density* and *heritage tree ordinance* accounted for approximately a quarter of the variability seen with regard to community canopy coverage (adjusted  $R^2 = 0.24$ ).

183 **Table 3. Final model and regression results in predicting canopy coverage for 43 Florida (United**  
 184 **States) communities with a range of urban forest management strategies and ordinances.**

Variable	Coefficient	Standard Error	<i>P</i> value	95% CI Lower	95% CI Upper
Intercept	37.2696	4.1194	<0.0001	28.9303	45.6089
<i>Housing density</i>	-0.0021	0.0008	0.0116	-0.0038	-0.0005
<i>Heritage tree ordinance</i>	6.7207	3.2827	0.0476	0.0751	13.3664

185

## 186 **4. Discussion**

187

### 188 **4.1. Survey results for Florida cities**

189 The information gleaned from this study creates a picture of urban forest management in Florida. Most  
 190 surveyed communities had tree inventories, despite this very few had tree canopy goals. This might be  
 191 due in part to the fact that most UTC typically falls on private property, complicating a municipality's  
 192 ability to maintain a specific amount of canopy cover (Miller et al., 2015). However, many respondents  
 193 said they had measures in place (i.e., permit requirements) for cutting trees on private property.

194 Additionally, a majority of the Florida communities had tree preservation and heritage tree ordinances  
 195 protecting their public and (in some instances) private trees. Preservation ordinances that protect trees  
 196 during development activities can be significant management practices in rapidly-urbanizing areas such as  
 197 Florida. This is also true of heritage tree preservation ordinances, which protect trees with large stem  
 198 diameters. These significantly-sized trees would likely have greater canopy areas that provide more  
 199 benefits and ecosystem services than smaller trees of the same species (Maco and McPherson, 2003;  
 200 Leibowitz, 2012). Extra attention should be paid to these large trees when examining UTC and creating  
 201 canopy goals.

#### 202           4.2.    **Canopy coverage in Florida cities**

203

204    Dot-based canopy assessment is a proven, albeit somewhat labor-intensive, method of conducting land  
205    classification (Walton et al., 2008; Jackson et al. 2010). In urban forestry research, dot-based  
206    interpretation is often considered that standard to compare against other manual or more automated  
207    approaches for identifying canopy coverage (Nowak and Greenfield, 2010; Nowak and Greenfield, 2012;  
208    Parmehr et al., 2016). In testing the repeatability of dot-based canopy assessment, Jackson et al (2010),  
209    compared canopy classifications for five locations (e.g., Georgia, Kansas, Michigan, Oregon, and Utah)  
210    across the United States. The authors had two to five assessors interpret a total 208 plots (with 105 dots  
211    per plot) and reported how many plots from each location met a 90% threshold for agreement. With over  
212    70% meeting or surpassing this threshold for all but the Georgia location (which had an errant  
213    interpreter), the authors concluded that the method offered a high level of agreement (Jackson et al.,  
214    2010). In comparison, 100% of the cities assessed by our interpreters met or surpassed the 90% agreement  
215    threshold used by Jackson et al. (2010; Table 2) using the same imagery source, (National Agriculture  
216    Imagery Program or NAIP).

217

218    Canopy coverage for the Florida communities included in this study ranged from 17.6% in Deerfield  
219    Beach to 63.3% in Gainesville. The latter value for percent canopy coverage is among the highest  
220    reported in the literature. In looking at 29 Chicago-area communities, Iverson and Cook (2000) estimated  
221    62.7% canopy coverage in North Barrington, Illinois, United States. Heynen and Lindsey (2003) observed  
222    a maximum canopy coverage of 55.7% in their assessment of 60 central Indiana communities. At 55%  
223    canopy coverage, Nowak et al (1996) identified Baton Rouge, Louisiana as the most treed city in their  
224    aggregation of 68 canopy analyses conducted in the United States.

225

226    Several of the communities assessed for this study had previous assessments of canopy coverage to draw  
227    on for comparison. In 2016, one year after our referenced imagery, researchers estimated Tampa had a



228 total canopy coverage of 32.3% (Landry et al., 2018). In comparing 95% confidence intervals, the lower  
229 bound for our study's estimate of canopy coverage (32.8%) overlapped with the upper bound for the  
230 estimate (33.7%) calculated by Landry et al. (2018). Even greater overlap was noted with canopy  
231 coverage estimates calculated by the City of Fort Lauderdale. The urban forester for this community  
232 related that he had estimated canopy coverage in 2018 at 25.9% with a 95% confidence interval between  
233 23.2 and 28.6 (Mark Williams, personal communication; Table 2). Despite differences in methodology,  
234 canopy estimates from Orlando's 2012 i-Tree Eco analysis (31.4 %; Epke et al., 2012) also fell within our  
235 95% confidence intervals (Table 2).

236

237 Our canopy estimates were less consistent with past estimates when looking at our two most treed  
238 communities - Gainesville and Tallahassee. Using 2013 imagery, Ucer et al. (2016) compared two  
239 different sampling techniques for estimating canopy coverage in Tallahassee. While both methods tested  
240 garnered similar results in their study (44.5% to 49.1% depending on imagery source), their results were  
241 well below our canopy estimate of 58.7% (Table 2). That said, our estimates did align with a 55% canopy  
242 coverage estimate obtained by the City of Tallahassee as part of efforts to develop an Urban Forest  
243 Master Plan (City of Tallahassee, 2018). Similarly, our estimate of tree canopy coverage in Gainesville  
244 (63.3%) was higher than independent estimates (54%) derived from the same 2015 imagery (Andreu et  
245 al., 2017). That noted, our estimate was in line with historic estimates of canopy coverage (59% to 67%)  
246 for the community calculated by Szantoi et al. (2008).

247

#### 248 **4.3. Predictors of canopy coverage**

249

250 While our two-variable reduced model may appear somewhat simplistic compared to other attempts at  
251 predicting canopy coverage (Hill et al., 2010; Landry and Pu, 2010; Kendal et al., 2012; Conway and  
252 Bourne, 2013), it is appropriate for our sample size of 43 communities and likely avoids the generation of  
253 misleading coefficients, *P*-values, and coefficient of determination values associated with overfitting

254 (Minitab Blog Editor, 2015). The negative relationship between housing density and canopy coverage is  
255 both intuitive and in line with findings from past research (Iverson and Cook, 2000; Conway and Bourne,  
256 2013). The importance of housing density also supports the “population density” explanation of tree  
257 canopy distribution (i.e., that people displace trees; Locke et al, 2016). In plotting canopy coverage values  
258 by location (Fig. 1), the relationship is particularly noticeable in the densely populated, Southeastern  
259 portion of the state (e.g., Palm Beach, Broward, and Miami-Dade counties).

260

261 More interesting with regard to our original research questions is the significance of having heritage tree  
262 designations and protections. A 6.7% increase in canopy coverage represents anywhere from 10.5% to  
263 38.1% of the total canopy coverage depending on the community investigated (Table 2). While the other  
264 ordinances noted in our survey did not remain in our reduced model as predictors of canopy coverage,  
265 their absence cannot be taken as evidence that they are not effective. For example, nearly every city  
266 surveyed had ordinances in place requiring the planting of trees for new developments (n=39) and new  
267 parking lots (n=39). As such these were not used as predictors for canopy coverage in any of our models.

268

269 Of the tree ordinance types used to predict canopy coverage (i.e., *tree preservation ordinance*, *removal*  
270 *permit ordinance*; and *heritage tree ordinance*), only *heritage tree ordinance* was significant (Table 3).  
271 However, 92.3% of the cities that reported having protections enacted for heritage trees also reported  
272 having ordinances regarding the preservation of existing trees during development. Similarly, 69.2% of  
273 cities with heritage tree designations also had ordinances restricting tree cutting on private property. As  
274 such it is not clear if heritage tree protections alone are responsible for the measured increase in canopy.  
275 The increase in canopy may ultimately be the combined impact of all the ordinances in addition to any  
276 special protections communities afford for their large stature trees.

277

278 That noted, *heritage tree ordinance* did play a role in many of the models with the highest predictive  
279 ability derived from our initial set of variables (Fig. 2). Its statistical significance could reflect the stricter

280 protections afforded to trees of noteworthy stature or historical notoriety. Large stature trees in both forest  
281 and urban environments are the genetic and environmental lottery winners of the plant world - defying  
282 odds that weeded out hundreds, if not thousands, of peer trees over the decades (i.e., other trees that did  
283 not reach maturity). While this is the case even in hospitable locales, urban sites can be especially  
284 challenging as the conditions (e.g., adequate space, native soils, protection from neighboring trees) which  
285 allowed heritage trees to grow to their fullest potential may no longer exist in a post-development  
286 environment. Despite their relative rareness, large stature trees provide a disproportionate amount of  
287 environmental and economic benefit. Benefits like shading, air pollution capture, stormwater control, and  
288 carbon sequestration are all tied to tree size and canopy area. Factoring in growth and attrition rates, it  
289 could take decades, dozens of replacement trees, or both to mitigate the loss associated with a single large  
290 stature heritage tree.

291  
292 While tree preservation ordinances and removal permits are intended, in part, to reduce canopy loss,  
293 neither *tree preservation ordinance* nor *removal permit ordinance* made it into our final reduced model  
294 (Table 3). With regard to tree preservation ordinances, there could be several reasons we did not see a  
295 relationship with canopy coverage (beyond the overlap in ordinances noted above). Tree preservation  
296 ordinances do allow the removal of trees to permit the development of a forested site. To offset these  
297 removals, new trees can be planted elsewhere on the property or in the community. Alternatively,  
298 developers are often given the option of paying into a tree mitigation fund if they prefer or if suitable  
299 planting sites are not available. Depending on how the number of replacement trees is calculated, it could  
300 take several decades to regain the canopy lost to development. Additionally, if mitigation funds are not  
301 actively spent to replant trees within a community, the canopy linked to these funds is essentially lost  
302 without replacement. Even with an active replanting program, transplant losses, vandalism, and other  
303 stressors that afflict younger trees could limit canopy replacement efforts - especially if adequate early  
304 care is not provided.

305

306 Research by Landry and Pu (2010) in the Tampa Bay area suggests that protections for trees of a certain  
307 size regardless of ownership (public or private) can lead to higher canopy area. However, tree removal  
308 permits on private land are a potentially contentious issue which residents may see as being at odds with  
309 their individual property rights (Conway and Lue, 2018). In contrast to the findings of Landry and Pu  
310 (2010), our data did not indicate private tree protections had any influence on canopy coverage when  
311 assessment was expanded beyond the City of Tampa, Temple Terrace, and the surrounding county.  
312 Effective private tree protection depends on enforcement and public knowledge of permitting  
313 requirements - both of which may could vary by community (Conway and Lue, 2018). Moreover,  
314 enforcement occurs only after a tree has been cut down and often only after a member of the public has  
315 reported the removal (Conway and Lue, 2018). Finally, permitting generally does not restrict the removal  
316 of trees for development or to reduce tree risk, which likely account for the majority of tree removals  
317 (even unpermitted).

318

#### 319 **4.4. Policy implications**

320

321 As noted, urban canopy coverage directly affects many ecological and economic benefits. However,  
322 maintaining, protecting, and expanding urban tree canopy requires an investment of resources by  
323 communities. Moreover, trees, buildings, and urban infrastructure all compete for limited space -  
324 potentially putting canopy goals at odds with development efforts. Ultimately, it is up to community  
325 leaders and their constituents to decide where this balance best fits their needs and values.

326

327 That noted, this work provides evidence that at least some protection measures currently used in Florida  
328 communities are associated with increases in urban tree canopy. For those questioning the validity of  
329 current ordinances, the data provided for this work hopefully alleviates some of the concerns that existing  
330 regulations are not serving their intended function. Additional work looking at the impacts of urban forest  
331 management over time is the next logical step in this line of inquiry. Additionally, the impacts of urban

332 forest management efforts on storm resiliency (specifically canopy loss) would be a very relevant  
333 question for hurricane-prone areas like Florida.

334

335 **5. Conclusion**

336

337 Many of the ecosystem services urban tree managers calculate when assessing the value of their urban  
338 forest are directly linked to canopy coverage. Large trees contribute more canopy than smaller-sized trees,  
339 but must compete with other aspects of urban infrastructure for above and belowground space. This can  
340 lead to conflict, tree injury, and even tree death if care is not taken during the development and  
341 redevelopment of sites.

342

343 Communities enact ordinances to reduce damage to trees given development efforts and restrict the  
344 removal of healthy, stable trees as a public good. These efforts can be somewhat controversial, especially  
345 when they interfere with private property owner rights. Regardless of one's opinion on this matter, there  
346 appears to be some measurable benefit associated with some tree protection ordinances. In particular, we  
347 observed a significant increase in canopy associated with communities that designated and protected  
348 heritage trees. These findings add much needed empirical evidence to a debate which is playing out in the  
349 study area and beyond.

**Authors' contributions**

[Redacted text block containing author contributions]

## Conflicts of interest

The authors declare no conflict of interest

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## Appendix A:

### Select Survey Questions and Summary of Responses

Municipal Tree Care and Management in Florida: *A 2014 Urban & Community Forestry Census of Tree Activities*. The survey was conducted for several municipalities (n=87) in the State of Florida. In looking at the returned survey results, questions with 7 or more non-responses (16.3%) were not considered for inclusion in the regression model.

Section I – Community and Staff

Section II -- Budget

Section III -- Tree Management Profile

Section IV -- Volunteers/Partnerships

Section V -- Contractors

Section VI -- Inventory

Section VII -- Operations Profile

Section VIII -- Assistance Programs

#### Section I -- Community and Staff

Did your community conduct any kind of shade tree/urban & community forestry activities in 2014?

Yes

No

Don't know

Who in your community is primarily (legally) responsible for maintaining trees in municipal rights-of-way, for example street trees between sidewalk and curb or alley trees?

Municipality responsible

Abutting property owner responsible

Jointly responsible (municipality and abutting owner)

Other (*please specify:* \_\_\_\_\_)

Does someone in your community (i.e., employee, volunteer, consultant, etc.) oversee the care of municipal street trees, park trees or other public trees?

Yes

No

Don't Know

How many years has your community had a person responsible for the management of trees?

\_\_\_\_\_ Number of Years

What training and/or credentials are collectively held by the staff responsible for tree activities and/or management of trees?

No specific training or workshops	Yes	No
In-house and/or on-the-job-training	Yes	No
Attend tree care/management workshops	Yes	No
ISA Certified Arborist	Yes	No
ISA Certified Municipal Specialist	Yes	No
Two year degree	Yes	No
Four year degree	Yes	No
Graduate degree	Yes	No

How many public employees, including managers, are involved with the municipal tree management program?

\_\_\_\_\_ # of Total Employees

\_\_\_\_\_ # of Full Time Equivalent (2080 hour base year)

## Section II -- Budget

What is the total municipal budget (excluding school budget) for 2014? (Please include entire amount for all governmental functions, activities, etc.)

\_\_\_\_\_ \$ Total 2014 Municipal Budget

What is the total annual budget of your municipality funded tree care activities and management from all municipal sources? (Include all tree activity expenses; include personnel, overhead, equipment, supplies, tree care and contract payments.)

\_\_\_\_\_ \$ Total 2014 Tree Budget

Is your budget adequate to meet current needs as defined in your work plan or your identified annual urban forestry budget needs? (This includes planting, maintenance, removal, inventory, education, etc.)

Yes

No → If no, \_\_\_\_\_ % below identified need

What percent of the total tree management budget from all sources is used for the following activities?

Tree Removal \_\_\_\_\_

### Section III -- Tree Management Profile

Does your community have a government-authorized tree board, parks board, city department, commission, or similar group that helps develop and/or administer tree management policy?

Yes

No

Does your municipality have one or more municipal ordinances that pertain to trees?

Yes

No

Developing

What topics do your community tree ordinances include?

Requires tree planting in new developments Yes

Requires tree planting around new parking lots Yes

Requires preservation of trees during development Yes

Restricts tree cutting on private property Yes

Identifies preservation of heritage or significant trees Yes

Does your community have a written strategic plan for urban forestry, tree management, open space, green infrastructure, or land use management that includes trees?

Yes

No

Don't Know

### Section IV -- Volunteers/ Partnerships

Does your community work with partners and/or volunteers (individuals or groups not paid for providing services) for tree planting, tree care, or other tree activities on public property?

Yes

No

### Section V-- Contractors

Does your community use paid contractors for any of your tree care activities?

Yes  
No

**Section VI -- Inventory**

Does your community have a tree inventory? (An inventory is any record of public trees in your community.)

Yes  
No

What is the state of your tree inventory? (current = up to date) (CHECK ONE CHOICE)

Current (reflects tree population)  
Developing (in process of making current)  
Not current (missing tree population information)

Does your municipality have a tree canopy goal? (check one)

Yes  
No → (PLEASE GO TO QUESTION 18, PAGE 18)  
Developing → (PLEASE GO TO QUESTION 18, PAGE 18)

What is the total number of publicly owned trees in your community?

\_\_\_\_\_ # of Publicly Owned Trees

**Section VII -- Operations Profile**

Please fill in the number of trees by tree care activity on all municipal properties in 2014 in the appropriate column. (Please enter 0 if no activity type was performed last year.)

# of Trees removed \_\_\_\_\_

# of Trees planted \_\_\_\_\_

What percent of tree care (pruning, pest control, etc.) is done on a systematic (regularly scheduled) cycle and what percent on demand as reactive (complaints, hazardous situations, crisis, post storm etc.)? (Total = 100%)

\_\_\_\_\_ % Systematic (Scheduled)

\_\_\_\_\_ % Reactive (on Demand)

Does your community conduct any of the following urban activities? (Check yes or no for each activity)

Provide technical assistance (information) for tree maintenance on private property?

Yes

No

Provide financial assistance for specific insect or diseased tree removal on private property?

Yes

No

Does your community regularly conduct tree risk management (hazard tree identification)?

Yes

No

Does your community have a written tree risk management policy?

Yes

No

Does your community have an emergency response system which includes trees?

Yes

No

### **Section VIII -- Assistance Programs**

Do municipal staff provide educational presentations to city residents in regard to tree care?

Yes

No

Is your community currently a Tree City USA?

Yes

No

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