

The USDA Forest Service's Forest Inventory and Analysis Database and the National Register of Champion Trees — A Potentially Symbiotic Relationship

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Abstract

This article shows how probability sampling and citizen science efforts can complement each other, using the USDA Forest Service's Forest Inventory and Analysis (FIA) program and the ongoing search by the National Register of Champion Trees (NRCT) for the largest specimen of each naturally occurring tree species in the United States as an example. We develop a ratio statistic (Z_j) that uses the difference in size of the largest tree of a species from each database to order the tree species according to the assumed ease with which a larger specimen than the current national champion might be found. Our results show ninety-two candidate species that have been recorded by FIA for which there is no national champion and sixty-five species for which a new champion should be easy to find. In a [supplemental](#) table, we show ninety-four species listed as observable by FIA in the NRCT but not recorded in the FIA sample.

Study Implications: An interest in forests and forestry is always accompanied by an interest in trees, especially very big trees. Two very different ways of learning about trees are analyzed concurrently in a way that reveals their complementarity. The two efforts are the probability sample, conducted by the USDA Forest Service's Forest Inventory and Analysis (FIA) Program, and the citizen science effort known as the National Register of Champion Trees (NRCT). We develop a statistic that will help tree sleuths find champion trees and provide FIA practitioners with a quality control measure and an indication of which species would benefit from an increase in sample intensity.

Keywords: champion tree, NFI, FIA, probability sample, citizen science

Since at least the inception of recorded history, humans have been attempting to learn more about the world around them. This has produced many formal and informal methods of acquiring knowledge. Scientists often extol the virtues of formal applications of the scientific method, such as hypothesis testing and probability sampling, and favor the knowledge that has been acquired through these applications. The scientific method, as executed within a probability sample, is well suited to answering the questions asked but less well suited to answering the unasked questions. People in general, however, learn through many different paths and under an assumption that learning is a Markov process, an argument could be made that once someone actually knows something for certain, it hardly matters how it was learned ([Visser et al. 2006](#)). Although the use of citizen scientists in information gathering has a long history (e.g., Audubon's Christmas Bird Count was first conducted in 1900 and the Big Tree Program was conceived in the 1940s), according to Wikipedia (as of June 13, 2023) the term “citizen science,” as it is used today, does not appear until an article by [Kerson \(1989\)](#) in the Massachusetts

Institute of Technology's *Technology Review*. Although the conduct of citizen science is often described as the participation of nonscientists in the scientific process, it is not necessarily constrained to the formal requirements of applications of the scientific method. For instance, there are often no experimental controls or identifiable sampling frames but rather a large number of mostly nonexperts focused on an observational task.

Here we explore how two methods, probability sampling and citizen science, can complement and support each other, with a specific example involving the National Forest Inventory (NFI) in the United States (the USDA Forest Service's [Forest Service] Forest Inventory and Analysis [FIA] program [[Burrill et al. 2018](#)]), and the National Register of Champion Trees (NRCT), ([American Forests 2023](#)). The example involves the ongoing search for the largest specimen of each naturally occurring tree species found in the United States. Credit for the idea that led to this ongoing citizen science effort in the United States (sometimes known as the National Big Tree Program) to identify the largest specimens

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of many of the tree species that are included in today’s NRCT is given by [Leverett and Bertolette \(2014\)](#) to [Stearns \(1940\)](#). Below, we explore how information from this extensive citizen science effort can complement (and be complemented by) information acquired through a probability sample conducted by the FIA program. Although augmented in various ways throughout the country, the areal scale of the base FIA sample is 1:36,000. In [Table 1](#), we list some of the contributory and/or contrasting features of these two knowledge-acquiring efforts with respect to the goal of identifying the largest tree specimen of each species.

Of special note are the reciprocal benefits of the two information-gathering efforts. The citizen science effort that results in the champion tree lists of the various states as well as the NRCT can be, and has been, used to inform the probability sampling efforts of FIA by providing a quality control checkpoint during field data collection. For instance, a data recorder can be programmed to respond with an “Are you sure?” prompt when a height or diameter value is entered that exceeds that of the largest known tree of a species in a state or national champion tree list. Reciprocally, as noted in [May \(1990\)](#), the FIA sample can suggest the degree to which the citizen science effort appears to have been successful in identifying the largest specimen of each species. That is, it can suggest to a citizen scientist tree sleuth the areas within which a larger specimen of any particular species might be found, and it can also suggest which species would require an extreme effort by the citizen scientist to identify a larger specimen. Below, we present an example of a simple data mining task to achieve these reciprocal benefits.

Materials and Methods

We use two major sources of information, FIA’s public-facing database (FIADB) as it existed on April 17, 2023, and the NRCT as it existed on April 11, 2023. These two efforts have quite different primary goals, methods, definitions, cost structures, and funding sources. We do not attempt to enumerate all of those differences here, just those important to our study, but do invite the reader to explore the cited documentation for each program.

Importantly, there are some minor differences between what species are accepted for the champion tree lists of the various states of the United States and the NRCT as well as

the species meeting the definition of a tree between regions in the FIA program. Readers interested in the requirements of a particular state’s champion tree list are referred to [Table A1 in Appendix 1](#), which includes the most recent website information that we found for each state. Readers interested in the regional requirements of FIA are referred to [Burrill et al. 2018](#).

The National Champion Tree Program recommends submission of a large tree’s measurements to the state tree list first, allowing that it may proceed from there to the National Champion Tree list. Most states require (or are transitioning to a requirement for) a tree species to be native or naturalized and listed in [Little \(1979\)](#). This makes sense, given the intermediate goal of finding the largest specimen of each species within a state, as some would argue, it would be hardly noteworthy to find the largest of a non-native species in a state, if only a few specimens existed which had been transplanted from elsewhere. Others might argue that a big tree is a beautiful sight and should be noted no matter how it happened to get to where it is. As there are tree species that are accepted by some states and not others, our data-mining analysis of the FIADB and the NRCT is segmented into species that occur (or could occur) in only one of the data sets and species that are common to both. We also rectified some minor nomenclature and classification differences between the two data sources. The choices made here in this regard are not indicative of a preference on authors’ part or on the part of the authors’ employers but rather reflect an arbitrary choice that had to be made to proceed.

Alternative Measures of Large Trees

The NRCT bases selection of the champion tree for each species (*S*) on the measurements of bole circumference (*C*) in inches at breast height (4.5 ft), tree height (*H*) in feet, and average crown spread or width (*W*) in feet, to obtain the total points for tree (*t*) determined through the formula:

$$PCT_t = C_t + H_t + (0.25 * W_t) \tag{1}$$

Detailed instructions on measurement methods can be found in [Leverett and Bertolette \(2014\)](#). This formula incorporates the essential elements that lead to a perception of large when it comes to tree size. By this formula, 1 in. of bole circumference (at breast height) is worth the same as 12 in.

Table 1. Attributes of Information available from the USDA Forest Inventory and Analysis (FIA) program relative to the National Champion Tree Program.

Program	FIA	Champion Tree Program
Information category	Probability sample	Citizen science
Main goal	Description of forest populations through the sampling and measurement of those populations.	Identification and measurement of the largest tree(s) in each species.
Information sources	Several well-defined and documented sources of information, such as remote sensing layers, sample observations, etc.	Many unrestricted sources of information, such as local lore, incidental observation, friends and family, etc.
Identification of champion trees as a proportion of programmatic focus	much less than 1% - a collateral benefit	100%
Authentication process	Initial measurement conducted by trained, certified professionals whose results are subject to a well-defined, random quality-control process.	Initial measurement is conducted by citizen scientists from diverse backgrounds and education levels whose results are first authenticated by trained professionals at the state level and subsequently by an elite cadre at the national level.

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Acer spicatum</i>	VA	Highland	154	60	Smyth	VA	-1.571
<i>Prosopis pubescens</i>	TX	Llano	56	26	Brewster	TX	-1.161
<i>Pinus strobiformis</i>	AZ	Coconino	217	114	Monroe	NY	-0.902
<i>Clusia rosea</i>	HI	Hawaii	80	49	Broward	FL	-0.638
<i>Quercus buckleyi</i>	TX	Grayson	135	84	Tarrant	TX	-0.611
<i>Prunus pensylvanica</i>	NC	Macon	156	98	Giles	VA	-0.589
<i>Crataegus crus-galli</i>	TN	Lake	83	54	Montgomery	VA	-0.532
<i>Acer pensylvanicum</i>	TN	Carter	143	94	Grayson	VA	-0.524
<i>Quercus ilicifolia</i>	SC	Richland	142	97	Hartford	CT	-0.461
<i>Asimina triloba</i>	TN	Bledsoe	131	96	Montgomery	MD	-0.365
<i>Prosopis glandulosa</i>	TX	Armstrong	326	239	Real	TX	-0.363
<i>Acer leucoderme</i>	TN	Sevier	136	103	Gaston	NC	-0.320
<i>Prunus emarginata</i>	WA	Whatcom	156	118	Marion	OR	-0.318
<i>Juniperus occidentalis</i>	CA	Alpine	388	301	Lake	OR	-0.289
<i>Persea borbonia</i>	GA	Burke	252	198	Montgomery	TX	-0.274
<i>Acer barbatum</i>	AR	Van Buren	310	259	Southampton	VA	-0.198
<i>Fagus grandifolia</i>	NC	Wilkes	373	314	New Kent	VA	-0.188
<i>Sabal palmetto</i>	GA	Camden	153	133	Lafayette	FL	-0.149
<i>Juniperus ashei</i>	TX	Edwards	199	182	Comal	TX	-0.093
<i>Carya texana</i>	TN	Dickson	229	211	Butler	MO	-0.084
<i>Tsuga mertensiana</i>	WA	Grays Harbor	360	335	Mariposa	CA	-0.075
<i>Zanthoxylum clava-herculis</i>	TX	Jasper	72	68	Northampton	VA	-0.052
<i>Pinus resinosa</i>	MN	Clearwater	206	198	Carroll	NH	-0.042
<i>Thrinax morrisii</i>	FL	Monroe	46	44	Miami-Dade	FL	-0.041
<i>Cornus florida</i>	LA	Grant	152	148	Sampson	NC	-0.030
<i>Magnolia macrophylla</i>	TN	Morgan	152	148	Buncombe	NC	-0.030
<i>Sorbus americana</i>	WV	Pocahontas	99	97	Grayson	VA	-0.022
<i>Quercus douglasii</i>	CA	San Joaquin	219	215	Mariposa	CA	-0.017
<i>Pinus elliotii</i>	AL	Monroe	227	224	Portsmouth	VA	-0.012
<i>Pinus monophylla</i>	CA	Alpine	190	189	Washoe	NV	-0.006
<i>Pinus banksiana</i>	MN	St. Louis	158	158	Lake	MN	-0.001
<i>Cercis canadensis</i>	AR	Drew	177	177	Fairfax	VA	0.0016
<i>Populus grandidentata</i>	MA	Hampden	226	227	McKean	PA	0.0054
<i>Tsuga heterophylla</i>	CA	Humboldt	522	533	Jefferson	WA	0.0209
<i>Pinus clausa</i>	FL	Putnam	173	177	Osceola	FL	0.0209
<i>Pinus virginiana</i>	SC	Greenville	194	199	Fairfax	VA	0.0246
<i>Vachellia farnesiana</i>	TX	La Salle	120	123	Maricopa	AZ	0.0249
<i>Pinus serotina</i>	FL	Osceola	208	216	Bladen	NC	0.0352
<i>Quercus rugosa</i>	NM	Sierra	140	146	Santa Cruz	AZ	0.0400
<i>Robinia neomexicana</i>	NM	Otero	103	109	Gila	AZ	0.0550
<i>Pinus glabra</i>	MS	Clarke	238	253	Alachua	FL	0.0607
<i>Ulmus serotina</i>	AR	Crawford	174	187	Hamilton	OH	0.0706
<i>Magnolia virginiana</i>	AL	Tallapoosa	218	236	Jefferson	FL	0.0780
<i>Alnus rubra</i>	WA	Skagit	242	263	Benton	OR	0.0791
<i>Juniperus californica</i>	CA	Fresno	166	182	Yavapai	AZ	0.0852
<i>Diospyros sandwicensis</i>	HI	Hawaii	63	69	Hawai'i	HI	0.0859
<i>Populus tremuloides</i>	MT	Lincoln	214	236	Chippewa	MI	0.0919
<i>Quercus coccinea</i>	TN	Carroll	299	329	Mercer	NJ	0.0921
<i>Castanea pumila</i>	AL	Franklin	64	71	Trinity	TX	0.0927
<i>Halesia carolina</i>	TN	Sevier	177	197	Cheshire	NH	0.0997
<i>Betula populifolia</i>	NY	Saratoga	128	143	Cape May	NJ	0.1024

Table 2. Continued

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Quercus marilandica</i>	MS	Amite	188	210	Chatham	NC	0.1033
<i>Larix laricina</i>	NY	Saratoga	177	200	Carroll	NH	0.1158
<i>Magnolia fraseri</i>	NC	Graham	186	212	Carroll	VA	0.1208
<i>Carpinus caroliniana</i>	AR	Cleveland	207	236	Essex	NJ	0.1213
<i>Betula lenta</i>	VA	Wythe	206	235	Windsor	VT	0.1229
<i>Abies concolor</i>	CA	Madera	417	478	Mariposa	CA	0.1276
<i>Populus angustifolia</i>	CO	Mesa	254	292	Hinsdale	CO	0.1289
<i>Pinus jeffreyi</i>	CA	El Dorado	412	475	Fresno	CA	0.1324
<i>Acer glabrum</i>	WA	Chelan	158	183	Island	WA	0.1339
<i>Nyssa biflora</i>	AL	Greene	277	322	Chesapeake	VA	0.1386
<i>Quercus incana</i>	TX	Houston	134	156	Smith	TX	0.1421
<i>Sideroxylon lanuginosum</i>	MS	Tallahatchie	139	162	Freestone	TX	0.1425
<i>Oxydendrum arboreum</i>	NC	Graham	174	204	Amelia	VA	0.1465
<i>Liquidambar styraciflua</i>	MS	Bolivar	307	360	Burlington	NJ	0.1471

in height or 48 in. of crown spread. Unfortunately, as noted in May (1990) and Oswalt et al. (2010), FIA does not collect data on average crown spread, so to facilitate comparison between the two data sources, we calculated a partial points score (identical to the “bigness” score used in the May [1990] and Oswalt et al. [2010]) for the champion tree of each species (T_s) in the NRCT:

$$PPCT_s = C_{T_s} + H_{T_s} \quad (2)$$

Although possible, given the breakdown of the contribution of W_t to (1), and the fact that the additional weight of very large crowns must be supported by an increase in bole circumference, we suspect that it is unlikely that the use of equation (2) as opposed to (1) would significantly change the ranking of the largest trees in a species. This suspicion is partially supported by the observation that the average contribution of crown width to PCT_t in (1) for the 563 trees in the NRCT is .067. Given that W_t tends to increase with both C_t and H_t , the average additional informational value of W_t is something less than .067 in (1). In subsequent analyses, in the case of co-champions existing in the NRCT, we used the tree with the highest partial points score. We also calculated the partial points score for each live tree (t) in the FIA database measured between 1998 and early 2023 to find the tree with the highest score for each species (S):

$$PPFIA_s = \max_S (C_t + H_t) \quad (3)$$

In a probability sample of a large population, such as the population of trees of concern to FIA, the probability of selecting the tree of maximum value is very small. It stands to reason that a targeted search for that same maximally sized tree, such as that conducted by the National Champion Tree Program, should meet with greater success. Therefore, we devised a simple ratio statistic for each species (S):

$$Z_s = (PPCT_s - PPFIA_s) / PPCT_s \quad (4)$$

Owing to the fact that the National Champion Tree Program targets the largest trees of each species, one would expect Z_s to be positive most of the time. One would also assume that when Z_s is large and positive, the citizen science

effort has been a remarkable success with respect to that species and that when Z_s is positive but small, larger trees (by the NRCT's total point formula) than the NRCT's champion tree of that species, might be found. Furthermore, when Z_s is negative, a larger tree than the NRCT's champion tree for that species has already been found (by the Z_s statistic criterion).

We note here, as was previously noted in both May (1990) and Oswalt et al. (2010), that the exact location of each FIA sample plot is confidential, so the plot coordinates listed in FIADB have been perturbed and cannot be used to pinpoint the exact location of any particular sample tree. However, the Z_s statistic could be used by citizen scientist tree sleuths to identify county-scale search areas for finding alternative national champion trees of particular species.

If a simple local-area search within each county containing the largest trees in the FIADB proves fruitless, quite a few different approaches to defining expanded search areas for champion trees from the information in the two databases could be used. For instance, one might filter the FIADB to find and map the counties of the x largest trees of a species, add the current national champion listed in the NRCT, and then form a search area around their listed counties (or groups of adjacent counties), possibly within ecotypes, for certain species. That is, the search areas could be suggested by the distribution of counties within which these $x + 1$ largest trees were observed. We give one example of this approach applied to a single species (*Pinus virginiana*) in the Results section.

A Note on Viewing the Current Champion Tree as the Maximum Value Observed in a Trend or as an Outlier

In the case of a 100% census, the observation of the largest value in a population will occur. Otherwise, in the case of a probability sample such as FIA, it is usually unlikely that the sample will include the largest tree of a particular species. The estimation of the maximum value from a sample that would be observed in a population has a broad body of literature in statistics, generally referred to as the statistics of extremes (e.g., de Carvalho 2016). Whether or not it makes sense for the interested reader to try to model the size of the

Table 3. The size (in partial points [PPoints]) and Z_s statistic for each species observed by both the USDA Forest Service's Forest Inventory and Analysis program and the National Register of Champion Trees with a Z_s statistic between 0.15 and 0.289.

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Abies grandis</i>	WA	Chelan	427	502	Humboldt	CA	0.1501
<i>Pinus monticola</i>	OR	Lane	394	464	Douglas	NV	0.1502
<i>Populus heterophylla</i>	AR	Arkansas	228	268	Dunklin	MO	0.1503
<i>Abies procera</i>	WA	Pierce	483	568	Cowlitz	WA	0.1503
<i>Picea pungens</i>	WY	Lincoln	279	329	Wasatch	UT	0.1510
<i>Carya aquatica</i>	LA	St. Landry	256	303	Southampton	VA	0.1553
<i>Juniperus osteosperma</i>	UT	Utah	205	243	Mohave	AZ	0.1557
<i>Pinus echinata</i>	FL	Leon	220	262	Smith	TX	0.1591
<i>Prunus serotina</i>	WV	Webster	244	292	Norfolk	VA	0.1627
<i>Picea engelmannii</i>	OR	Lane	333	400	Flathead	MT	0.1674
<i>Gordonia lasianthus</i>	FL	Nassau	196	236	Duval	FL	0.1708
<i>Fraxinus texensis</i>	TX	Collin	124	149	Monroe	NY	0.1708
<i>Fraxinus pennsylvanica</i>	MS	Bolivar	270	327	Hartford	CT	0.1728
<i>Gleditsia aquatica</i>	TN	Dyer	170	206	Montgomery	PA	0.1734
<i>Pinus lambertiana</i>	CA	Tuolumne	498	603	Tuolumne	CA	0.1740
<i>Pinus palustris</i>	AL	Tallapoosa	202	245	Aiken	SC	0.1744
<i>Carya illinoensis</i>	MS	Adams	322	390	Isle of Wight	VA	0.1745
<i>Quercus falcata</i>	MS	Wilkinson	342	414	James City	VA	0.1748
<i>Ulmus crassifolia</i>	LA	Franklin	195	236	Butler	MO	0.1751
<i>Quercus nigra</i>	MS	Amite	283	344	Suffolk	VA	0.1763
<i>Picea mariana</i>	MI	Keweenaw	158	195	Aroostook	ME	0.1905
<i>Pinus muricata</i>	CA	Mendocino	261	324	Marin	CA	0.1948
<i>Quercus palustris</i>	IL	Alexander	281	349	Lake	OH	0.1952
<i>Abies fraseri</i>	TN	Sevier	170	212	Harrisonburg	VA	0.1958
<i>Torreya californica</i>	CA	Sonoma	191	238	Santa Cruz	CA	0.1958
<i>Pinus taeda</i>	TX	Polk	279	349	Richland	SC	0.2015
<i>Carya carolinae-septentrionalis</i>	MS	Noxubee	196	246	Etowah	AL	0.2018
<i>Ilex opaca</i>	MS	Amite	196	246	White	AR	0.2036
<i>Picea rubens</i>	NC	Swain	223	280	Giles	VA	0.2044
<i>Pinus strobus</i>	TN	Unicoi	293	370	Waldo	ME	0.2086
<i>Carya glabra</i>	TN	Jefferson	250	317	Allen	KY	0.2100
<i>Hilairanthus germinans</i>	FL	Collier	109	138	Monroe	FL	0.2103
<i>Arbutus menziesii</i>	CA	Humboldt	272	346	Douglas	OR	0.2136
<i>Quercus similis</i>	AR	Drew	179	228	Red River	TX	0.2138
<i>Juniperus coahuilensis</i>	AZ	Yavapai	135	172	Brewster	TX	0.2154
<i>Pinus flexilis</i>	CA	Madera	285	363	Utah	UT	0.2158
<i>Pinus ponderosa</i>	CA	El Dorado	438	559	El Dorado	CA	0.2170
<i>Diospyros virginiana</i>	GA	Macon	191	244	Scioto	OH	0.2174
<i>Magnolia grandiflora</i>	SC	Hampton	256	327	Sussex	VA	0.2181
<i>Acer saccharum</i>	MI	Wayne	267	342	Giles	VA	0.2197
<i>Tsuga canadensis</i>	NC	Jackson	287	369	Macon	NC	0.2222
<i>Carya tomentosa</i>	AL	Macon	243	313	Caroline	VA	0.2225
<i>Acer nigrum</i>	MI	Eaton	198	255	York	ME	0.2227
<i>Quercus montana</i>	VA	Buchanan	293	381	Ward 3	DC	0.2307
<i>Larix occidentalis</i>	WA	Okanogan	321	421	Missoula	MT	0.2379
<i>Carya ovata</i>	TN	Smith	224	295	Anne Arundel	MD	0.2395
<i>Betula neoalaskana</i>	AK	Matanuska	148	196	Matanuska	AK	0.2459
<i>Malus fusca</i>	OR	Lane	104	138	King	WA	0.2488
<i>Acer negundo</i>	VA	Botetourt	222	298	Essex	VA	0.2537
<i>Ostrya virginiana</i>	SC	Florence	136	183	Washtenaw	MI	0.2548
<i>Pseudotsuga menziesii</i>	OR	Lane	651	875	Grays Harbor	WA	0.2565

Table 3. Continued

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Quercus michauxii</i>	NC	Edgecombe	295	399	Virginia Beach	VA	0.2595
<i>Acer rubrum</i>	MI	Manistee	257	348	New Haven	CT	0.2604
<i>Picea glauca</i>	AK	Prince of Wales	191	259	Koochiching	MN	0.2607
<i>Betula alleghaniensis</i>	NC	Jackson	225	305	Kennebec	ME	0.2610
<i>Ilex cassine</i>	FL	Lafayette	93	126	Palm Beach	FL	0.2643
<i>Carya myristiciformis</i>	TN	Van Buren	142	193	Red River	TX	0.2661
<i>Celtis laevigata</i>	SC	Kershaw	224	305	Caroline	VA	0.2671
<i>Diospyros texana</i>	TX	Kendall	61	83	Refugio	TX	0.2673
<i>Nyssa ogeche</i>	FL	Gulf	195	268	Wakulla	FL	0.2709
<i>Pinus contorta</i>	CA	Madera	281	388	San Bernardino	CA	0.2752
<i>Quercus engelmannii</i>	CA	Fresno	164	228	Los Angeles	CA	0.2807
<i>Magnolia tripetala</i>	TN	DeKalb	144	201	New Haven	CT	0.2824
<i>Ulmus alata</i>	MS	Monroe	197	275	Hopewell	VA	0.2847
<i>Pinus cembroides</i>	NM	Grant	93	130	Brewster	TX	0.2859
<i>Pinus balfouriana</i>	CA	Tulare	281	394	Tulare	CA	0.2870
<i>Quercus texana</i>	MS	Sharkey	266	374	Desha	AR	0.2877
<i>Abies lasiocarpa</i>	MT	Ravalli	268	377	Jefferson	WA	0.2887

largest tree to be observed given the current sample depends on at least two things, (1) the willingness to make a set of very strong (and possibly unrealistic) assumptions, beginning with the assumption that the largest observed tree is the maximum observed value of a predictable trend or known distribution, as opposed to an outlier, and (2) having enough data to model the trend. Although it is beyond the scope of this article, an early approach that assumes a uniform distribution, known as the German tank problem (Ruggles and Brodies 1947), might be applied to the problem at hand. For species with many observations, different models could be built under the assumption that there is one tree larger than the largest tree recorded, or two trees larger, or n trees larger, etc. For species with many measured specimens, one could assume a model for the upper tail, possibly as simple as linearity in the upper tail, and then perform a simple linear regression on, say, the ten largest observed trees and assume the next ten larger trees in the population would follow the trend. The Z_s statistic described above could be calculated for the estimated partial points of each projected tree relative to the partial points of the champion tree in the NRCT to obtain a more refined estimate of the likelihood of a larger tree being found.

Results

The largest specimen of each species found in only one of the data sources are given in two supplementary files. Table S1 lists 192 species that occur in the NRCT that have not been observed by FIA on forested sample plots. Of these, ninety-four species are listed as potentially observable by FIA if they would have been found to occur on a sample plot. These ninety-four species, identified in last column of the tables, must be relatively rare to not have been observed at all by FIA, and therefore, finding a larger tree than the one already listed in the NRCT would most likely be difficult. FIA would classify most of the remaining species in this table only to genus, aggregating species within genus.

The FIADB records 217 species, varieties, and subspecies that are not included in the NRCT. The largest specimen of each of these species by partial points is listed in Table S2 online. Ninety-two of the species in this table are listed in Little (1979) and would seem to be valid candidates for the NRCT. These are indicated by a “1” in the last column of the table. A tree sleuth interested in finding a suitable candidate of a particular one of these species for submission to the NRCT could start by looking in and around the county listed in the table for that species.

Tables 2 through 6 list the 299 species common to both the NRCT and the FIADB categorized in order of increasing Z_s . We assume that as the species Z_s statistic increases, the likelihood of finding a larger specimen than the one already listed in the NRCT decreases. Table 2 lists the thirty-one species with negative Z_s statistics and the thirty-four species with very small positive Z_s statistics (less than 0.149), in order of increasing Z_s . A negative Z_s indicates that the tree in the FIADB has more partial points than the tree in the NRCT. Anyone interested in finding candidate trees to propose for inclusion in a state’s champion tree list and eventual replacement of the current national champion could start by looking in and around the counties listed in Table 2 for that species.

Table 3 lists the sixty-eight species with Z_s statistics ranging from 0.15 to 0.289 in order of increasing Z_s . As we pointed out above, the FIADB is not likely to contain the largest tree of any species, so there is some likelihood that a search for a larger tree than the one currently in the NRCT from the species listed in this table, in the vicinity of the associated counties, could be successful.

To facilitate our discussion of the potentially symbiotic relationship of these two data sources, we include Tables 4–6. Table 4 lists the sixty-nine species with Z_s statistics between 0.29 and 0.399 in order of increasing Z_s . Table 5 lists the forty-six species with Z_s between 0.40 and 0.499, and Table 6 list the fifty-one species with Z_s statistics greater than 0.50

Table 4. The size (in partial points [PPoints]) and Z_s statistic for each species observed by both the USDA Forest Service's Forest Inventory and Analysis program and the National Register of Champion Trees with a Z_s statistic between 0.29 and 0.399.

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Fraxinus nigra</i>	MI	Midland	182	258	Gogebic	MI	0.2932
<i>Planera aquatica</i>	MS	Okribbeha	203	287	Gilchrist	FL	0.2935
<i>Fraxinus profunda</i>	NC	Bladen	212	301	Mississippi	MO	0.2945
<i>Pinus radiata</i>	CA	Santa Cruz	269	381	Humboldt	CA	0.2949
<i>Morella cerifera</i>	SC	Berkeley	51	72	Virginia Beach	VA	0.2957
<i>Chrysolepis chrysophylla</i>	CA	Siskiyou	231	329	Humboldt	CA	0.2970
<i>Betula nigra</i>	GA	Troup	196	279	Hamilton	OH	0.2986
<i>Pinus sabiniana</i>	CA	San Benito	248	354	Nevada	CA	0.2986
<i>Eugenia rhombea</i>	FL	Brevard	32	45	Monroe	FL	0.2991
<i>Ilex anomala</i>	HI	Hawaii	102	146	Hawai'i	HI	0.3003
<i>Quercus imbricaria</i>	IN	Daviess	216	309	Gallia	OH	0.3012
<i>Quercus laevis</i>	FL	Clay	133	190	Moore	NC	0.3016
<i>Aesculus flava</i>	TN	Monroe	263	376	Alleghany	VA	0.3018
<i>Sequoia sempervirens</i>	CA	Del Norte	885	1271	Del Norte	CA	0.3036
<i>Halesia diptera</i>	MS	Wilkinson	96	138	Buncombe	NC	0.3079
<i>Quercus bicolor</i>	KY	Union	258	373	Sussex	NJ	0.3081
<i>Condalia hookeri</i>	TX	Brooks	67	98	Hidalgo	TX	0.3124
<i>Fraxinus americana</i>	TN	Monroe	254	370	Morris	NJ	0.3129
<i>Nyssa sylvatica</i>	NC	Rowan	223	325	Wood	TX	0.3152
<i>Chamaecyparis thyoides</i>	VA	Suffolk city	184	268	Muskingum	OH	0.3152
<i>Quercus pagoda</i>	AL	Lowndes	304	445	Portsmouth	VA	0.3160
<i>Carya pallida</i>	TN	Marion	194	285	Greenville	SC	0.3176
<i>Juglans nigra</i>	MO	Saline	239	350	Westmoreland	VA	0.3180
<i>Quercus muehlenbergii</i>	OH	Mercer	248	364	Fairfield	OH	0.3182
<i>Populus trichocarpa</i>	OR	Clackamas	354	520	Polk	OR	0.3184
<i>Carya ovalis</i>	GA	Floyd	175	258	Caroline	VA	0.3200
<i>Pinus pungens</i>	VA	Montgomery	154	227	Stokes	NC	0.3206
<i>Quercus gambelii</i>	AZ	Navajo	177	261	Coconino	AZ	0.3224
<i>Calocedrus decurrens</i>	CA	Tulare	426	629	Siskiyou	CA	0.3230
<i>Quercus velutina</i>	AR	Phillips	297	440	Hartford	CT	0.3259
<i>Rhizophora mangle</i>	FL	Collier	88	131	Lee	FL	0.3264
<i>Fraxinus caroliniana</i>	GA	Dougherty	169	252	Lake	FL	0.3287
<i>Quercus kelloggii</i>	CA	Placer	283	425	Mariposa	CA	0.3338
<i>Pinus edulis</i>	UT	Grand	145	217	Santa Fe	NM	0.3339
<i>Pinus arizonica</i>	AZ	Cochise	172	258	Pima	AZ	0.3343
<i>Carya cordiformis</i>	TN	Greene	228	343	Russell	VA	0.3353
<i>Quercus alba</i>	NC	Mitchell	280	421	Brunswick	VA	0.3360
<i>Quercus wislizeni</i>	CA	Santa Cruz	228	343	Calaveras	CA	0.3362
<i>Quercus rubra</i>	AL	Montgomery	285	429	Ashtabula	OH	0.3363
<i>Fraxinus quadrangulata</i>	IN	Madison	194	293	Jefferson	KY	0.3385
<i>Quercus macrocarpa</i>	MI	Barry	275	416	Berrien	MI	0.3392
<i>Ulmus americana</i>	MS	Wayne	246	372	Baltimore	MD	0.3394
<i>Aesculus glabra</i>	OH	Hamilton	170	260	Fayette	KY	0.3444
<i>Quercus garryana</i>	CA	Mariposa	252	385	Multnomah	OR	0.3448
<i>Pinus rigida</i>	VA	Alleghany	183	281	Merrimack	NH	0.3473
<i>Syzygium sandwicense</i>	HI	Kauai	71	109	Honolulu	HI	0.3482
<i>Quercus phellos</i>	LA	Madison	281	433	Northampton	VA	0.3511
<i>Quercus laceyi</i>	TX	Bandera	113	175	Comal	TX	0.3525
<i>Quercus laurifolia</i>	LA	St. Landry	260	403	Chesapeake	VA	0.3559
<i>Quercus graciliformis</i>	TX	Brewster	75	116	Brewster	TX	0.3568
<i>Carya laciniosa</i>	MO	Cass	192	299	Culpeper	VA	0.3581

Table 4. Continued

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Pseudotsuga macrocarpa</i>	CA	San Bernardino	276	434	Los Angeles	CA	0.3644
<i>Platanus occidentalis</i>	TN	Lauderdale	353	556	Ashland	OH	0.3648
<i>Salix nigra</i>	MI	St. Clair	257	406	Washington	MN	0.3667
<i>Thuja plicata</i>	WA	Jefferson	576	910	Clallam	WA	0.3668
<i>Quercus agrifolia</i>	CA	Santa Clara	258	408	San Luis Obispo	CA	0.3668
<i>Tilia americana</i>	VA	Augusta	236	378	Fayette	KY	0.3749
<i>Pinus engelmannii</i>	AZ	Cochise	163	262	Cochise	AZ	0.3768
<i>Quercus emoryi</i>	NM	Grant	168	269	Grant	NM	0.3769
<i>Celtis occidentalis</i>	OH	Preble	234	376	Hancock	OH	0.3778
<i>Thuja occidentalis</i>	MI	Marquette	187	302	Nelson	VA	0.3818
<i>Pinus washoensis</i>	CA	Sierra	221	357	Modoc	CA	0.3821
<i>Juniperus deppeana</i>	NM	Catron	232	376	Yavapai	AZ	0.3842
<i>Pinus aristata</i>	NM	Taos	145	237	Huerfano	CO	0.3881
<i>Gleditsia triacanthos</i>	LA	Ascension	214	350	Botetourt	VA	0.3882
<i>Quercus shumardii</i>	AL	Greene	268	441	Powell	KY	0.3918
<i>Styrax grandifolius</i>	LA	Bienville	25	42	Suffolk	VA	0.3953
<i>Abies balsamea</i>	VT	Rutland	160	264	Adams	PA	0.3954
<i>Pinus leiophylla</i>	NM	Grant	128	212	Apache	AZ	0.3982

in order of increasing Z_s . These Z_s statistics indicate, in order of increasing difficulty, the chance of finding a tree larger than the current champion in the NRCT.

For a closer look at one species, Table 7 lists the year and county of observation, height, circumference, and partial points for twelve large specimens of *P. virginiana*, including the current national champion in the NRCT, the largest listed in the FIADB since 1997, and the ten largest listed in the FIADB since 2015. These twelve trees are located in eleven counties in the southeastern United States. The final column in Table 7 groups these eleven counties into four general search areas (I-IV). The locations for the eleven counties in which these twelve trees were observed are depicted in figure 1.

Discussion

In the analysis above, a negative Z_s tells us that the FIA sample has found at least one larger tree than the champion tree of a species, and a small positive Z_s indicates that a larger tree than the current champion could probably be found. The FIA data provide a myriad of ways to narrow the search and increase the probability of success for those searching for champion trees. Our example in figure 1 shows one approach. Other approaches might filter the database by elevation, forest type, owner class, etc. Although some might argue that it would spoil the thrill of the hunt, an obvious initial approach would be to examine aerial photographs or freely available Lidar point clouds (e.g., USGS 3DEP, GEDI) to map canopy heights for potential candidate trees, and then find and measure those candidates in the field.

There are a few possible reasons that a species would have a low Z_s statistic: the species could be rare or not well known or possibly so common that tree sleuths just are not very interested in it. A tree sleuth interested in finding a suitable candidate of a particular one of these species for submission to the NRCT could start by looking in and around the county

listed in the table for any of the ninety-two species listed in Little (1979) and indicated in Table S2 online.

As we noted above, Tables 2 through 6 list the species common to the two databases in order of increasing Z_s and we assume that as the species Z_s increases, the likelihood of finding a larger specimen than the one already listed in the NRCT decreases. Table 2 lists the species that already are included in both the FIADB and the NRCT of greatest promise to tree sleuths. Anyone interested in finding candidate trees to propose for inclusion in a state's champion tree list and eventual submission to the NRCT could start by looking in and around the counties listed in Table 2 for that species. Table 3 is the next most helpful table to tree sleuths, listing the species with small positive Z_s statistics. As we pointed out above, no database resulting from a probability sample is likely to contain the largest tree of any species, so there is some likelihood that a search for a larger tree than the one currently in the NRCT from the species listed in this table, in the vicinity of the associated counties, could be successful.

Tables 4–6 and S1 mostly show which species should be investigated last in the search for champion trees, as they list, in increasing order of difficulty, species for which a larger specimen than the current champion would most likely be difficult to find. However, as indicated in our title, these two data sources offer a potentially symbiotic relationship, and, as a more complete knowledge of all of the tree species in the United States, in addition to other rare events, becomes more desirable, an adaptation to the FIA sample design such as that suggested in Roesch (1993) and Roesch and McWilliams (2007), guided by Tables 4, 5, and especially 6, could be made. The reader should note that the subject species of Roesch and McWilliams (2007), *Castanea dentata*, is found in Table 6.

Table 7 illustrates how one could narrow the search for a champion tree of a particular species armed with only the two databases by first finding the counties within which a small set of the largest members of that species have been observed.

Table 5. The size (in partial points [PPoints]) and Z_s statistic for each species observed by both the USDA Forest Service's Forest Inventory and Analysis program and the National Register of Champion Trees with a Z_s statistic between 0.40 and 0.499.

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Quercus stellata</i>	MS	DeSoto	226	377	Cherokee	AL	0.4009
<i>Liriodendron tulipifera</i>	VA	Russell	300	501	Bedford	VA	0.4014
<i>Populus deltoides</i>	KS	Washington	322	538	Gage	NE	0.4021
<i>Sassafras albidum</i>	MS	Warren	206	345	Daviess	KY	0.4032
<i>Prunus caroliniana</i>	SC	Aiken	69	115	Williamsburg	VA	0.4038
<i>Juniperus virginiana</i>	TN	Rutherford	183	307	Coffee	GA	0.4049
<i>Gymnocladus dioicus</i>	IL	Warren	180	304	Montgomery	MD	0.4076
<i>Quercus lyrata</i>	MS	Lowndes	263	444	Isle of Wight	VA	0.4079
<i>Quercus virginiana</i>	FL	Levy	305	518	Ware	GA	0.4117
<i>Cornus drummondii</i>	LA	Concordia	40	69	Tunica	MS	0.4150
<i>Ulmus rubra</i>	SC	Marion	215	372	Jefferson	KY	0.4209
<i>Morus rubra</i>	SC	Lee	220	380	Ashley	AR	0.4214
<i>Ulmus thomasi</i>	MN	Fillmore	158	275	St. Louis	MO	0.4272
<i>Taxus brevifolia</i>	CA	Humboldt	139	243	Lewis	WA	0.4292
<i>Picea sitchensis</i>	OR	Tillamook	530	931	Grays Harbor	WA	0.4303
<i>Juglans cinerea</i>	MN	Hennepin	187	329	Thurston	WA	0.4308
<i>Prunus virginiana</i>	ME	Oxford	135	237	Baltimore	MD	0.4318
<i>Baccharis halimifolia</i>	GA	Long	32	56	Chesapeake	VA	0.4332
<i>Quercus myrtifolia</i>	FL	Putnam	63	111	Collier	FL	0.4333
<i>Larix lyallii</i>	WA	Okanogan	211	372	Chelan	WA	0.4333
<i>Bursera simaruba</i>	FL	Collier	80	142	Miami-Dade	FL	0.4341
<i>Acer saccharinum</i>	IN	De Kalb	269	478	Houston	MN	0.4382
<i>Quercus arizonica</i>	AZ	Gila	162	290	Santa Cruz	AZ	0.4404
<i>Salix amygdaloides</i>	MN	Benton	161	287	Aitkin	MN	0.4405
<i>Quercus oblongifolia</i>	AZ	Pima	116	209	Lincoln	NM	0.4446
<i>Juniperus scopulorum</i>	NM	Colfax	180	324	Cache	UT	0.4454
<i>Quercus grisea</i>	NM	Sierra	145	263	Hidalgo	NM	0.4482
<i>Acacia koa</i>	HI	Hawaii	252	458	Hawai'i	HI	0.4489
<i>Magnolia acuminata</i>	VA	Tazewell	220	400	Stark	OH	0.4508
<i>Cupressus nootkatensis</i>	WA	Snohomish	316	578	Jefferson	WA	0.4525
<i>Crataegus mollis</i>	NC	Northampton	87	161	Madison	OH	0.4580
<i>Prosopis velutina</i>	AZ	Santa Cruz	131	242	Santa Cruz	AZ	0.4592
<i>Olneya tesota</i>	CA	Riverside	130	241	Maricopa	AZ	0.4611
<i>Ficus aurea</i>	FL	Collier	102	189	Miami-Dade	FL	0.4612
<i>Salix caroliniana</i>	FL	Hillsborough	81	152	Alachua	FL	0.4651
<i>Vaccinium arboreum</i>	GA	Cherokee	48	90	Houston	TX	0.4699
<i>Malus coronaria</i>	PA	Westmoreland	76	144	Orange	VT	0.4699
<i>Catalpa bignonioides</i>	TX	Bowie	172	326	Baltimore	MD	0.4734
<i>Arbutus arizonica</i>	AZ	Cochise	87	166	Santa Cruz	AZ	0.4766
<i>Acer macrophyllum</i>	WA	King	305	582	Lane	OR	0.4766
<i>Maclura pomifera</i>	IN	Montgomery	205	393	Charlotte	VA	0.4781
<i>Carya floridana</i>	FL	Polk	69	133	Lake	FL	0.4785
<i>Quercus lobata</i>	CA	Mendocino	257	501	Mendocino	CA	0.4878
<i>Robinia pseudoacacia</i>	NC	Jackson	217	425	Columbia	NY	0.4892
<i>Persea palustris</i>	NC	Dare	96	190	Virginia Beach	VA	0.4933
<i>Sapindus oahuensis</i>	HI	Honolulu	59	116	O'ahu	HI	0.4947

Although somewhat older, the detailed tables in May (1990) and Oswalt et al. (2010) would also be excellent starting points in the southern United States. The reader should note

that, while our focus is at the national level, these analytical suggestions can be applied at other scales. In addition to the FIADB, there are also many other data sources resulting

Table 6. The size (in partial points [PPoints]) and Z_s statistic for each species observed by both the USDA Forest Service's Forest Inventory and Analysis program and the National Register of Champion Trees with a very large Z_s statistic (≥ 0.50).

Species	FIADB champion tree			National Register of Champion Trees			Zs score
	State	County	PPoints	PPoints	County	State	
<i>Pisonia brunoniana</i>	HI	Hawaii	91	183	Hawai'i	HI	0.5032
<i>Alnus oblongifolia</i>	AZ	Apache	161	327	Socorro	NM	0.5085
<i>Quercus chrysolepis</i>	CA	Shasta	288	597	San Bernardino	CA	0.5168
<i>Taxodium distichum</i>	AR	Arkansas	340	717	West Feliciana	LA	0.5263
<i>Leucaena pulverulenta</i>	TX	Cameron	71	151	Nueces	TX	0.5295
<i>Prunus angustifolia</i>	GA	Telfair	30	63	Alachua	FL	0.5302
<i>Cladrastis kentukea</i>	TN	Cocke	138	301	Morris	NJ	0.5405
<i>Amelanchier arborea</i>	MI	Otsego	70	154	Upshur	WV	0.5424
<i>Sapindus saponaria</i>	OK	Canadian	105	232	Hawai'i	HI	0.5454
<i>Catalpa speciosa</i>	IN	Allen	154	345	Lawrence	OH	0.5528
<i>Chamaecyparis lawsoniana</i>	OR	Curry	341	764	Coos	OR	0.5532
<i>Nyssa aquatica</i>	NC	Hertford	257	576	White	AR	0.5539
<i>Ehretia anacua</i>	TX	Goliad	86	196	Kenedy	TX	0.5632
<i>Acer grandidentatum</i>	AZ	Graham	117	269	Coconino	AZ	0.5639
<i>Quercus hemisphaerica</i>	GA	Clay	153	351	Richmond	VA	0.5652
<i>Quercus gravesii</i>	TX	Jeff Davis	70	160	Brewster	TX	0.5656
<i>Malus angustifolia</i>	GA	Laurens	67	157	Montgomery	MD	0.5753
<i>Coccoloba diversifolia</i>	FL	Dade	50	118	Miami-Dade	FL	0.5800
<i>Pinus longaeva</i>	CA	Mono	209	507	Clark	NV	0.5869
<i>Aesculus californica</i>	CA	Amador	91	222	Santa Cruz	CA	0.5879
<i>Umbellularia californica</i>	CA	Sonoma	287	702	Curry	OR	0.5914
<i>Quercus georgiana</i>	GA	Newton	61	149	Clarke	GA	0.5926
<i>Cotinus obovatus</i>	AL	Madison	72	183	Windham	CT	0.6083
<i>Erythrina sandwicensis</i>	HI	Kauai	88	227	Hawai'i	HI	0.6127
<i>Quercus prinoides</i>	SC	Aiken	31	81	Monroe	NY	0.6205
<i>Populus fremontii</i>	CA	Glenn	250	662	Yavapai	AZ	0.6229
<i>Juniperus flaccida</i>	TX	Brewster	56	152	Brewster	TX	0.6337
<i>Hesperocyparis arizonica</i>	AZ	Pima	119	350	Pima	AZ	0.661
<i>Halesia tetraptera</i>	TN	Carter	73	217	Sevier	TN	0.6640
<i>Castanea dentata</i>	ME	Somerset	110	331	Thurston	WA	0.6675
<i>Aesculus sylvatica</i>	GA	Rabun	37	111	Sussex	VA	0.6685
<i>Swietenia mahagoni</i>	FL	Monroe	84	255	Broward	FL	0.6708
<i>Sequoiadendron giganteum</i>	CA	Tulare	479	1506	Tulare	CA	0.6816
<i>Arbutus xalapensis</i>	TX	Brewster	66	208	Brewster	TX	0.6844
<i>Juglans major</i>	TX	Coryell	108	354	Pima	AZ	0.6958
<i>Fraxinus berlandieriana</i>	TX	Starr	83	276	Cameron	TX	0.6981
<i>Piscidia piscipula</i>	FL	Monroe	54	185	Monroe	FL	0.7095
<i>Ilex vomitoria</i>	TX	Montgomery	37	127	Wayne	NC	0.7103
<i>Conocarpus erectus</i>	FL	Collier	72	264	Monroe	FL	0.7279
<i>Prunus mexicana</i>	TX	Polk	31	113	Harris	TX	0.7292
<i>Sophora chrysophylla</i>	HI	Hawaii	81	319	Hawai'i	HI	0.7453
<i>Cupressus macrocarpa</i>	CA	Humboldt	154	690	San Mateo	CA	0.7767
<i>Cordia boissieri</i>	TX	Hidalgo	34	157	Cameron	TX	0.7820
<i>Ebenopsis ebano</i>	TX	Starr	51	236	Hidalgo	TX	0.7855
<i>Morus microphylla</i>	TX	Bosque	57	276	Yavapai	AZ	0.7919
<i>Quercus geminata</i>	FL	Madison	60	292	Alachua	FL	0.7931
<i>Quercus sinuata</i>	TX	Sutton	52	270	Greene	AL	0.8073
<i>Fraxinus velutina</i>	UT	Washington	63	334	Yavapai	AZ	0.8106
<i>Sideroxylon salicifolium</i>	FL	Martin	20	105	Miami-Dade	FL	0.8123
<i>Hesperocyparis nevadensis</i>	CA	Colusa	35	190	Kern	CA	0.8139
<i>Nestegis sandwicensis</i>	HI	Maui	33	237	Hawai'i	HI	0.8607

Table 7. The year of observation, state, county, height, circumference, partial points (PPoints), map symbol, and search area grouping for twelve large specimens of *Pinus virginiana*; eleven from the FIADB and one from the National Register of Champion Trees.

Symbol	Year	State	County	Height	Circumference	PPoints	Source	Search Area
A	2018	AL	Jackson	91	80	171	FIADB	I
B	2020	GA	Walker	122	70	192	FIADB	I
C	2021	GA	White	105	74	179	FIADB	I
C	2018	GA	White	109	72	181	FIADB	I
D	2018	GA	Habersham	114	53	167	FIADB	I
E	2003	SC	Greenville	114	80	194	FIADB	II
F	2017	TN	Polk	115	54	169	FIADB	I
G	2017	NC	Rockingham	118	46	164	FIADB	II
H	2018	NC	Rowan	102	64	166	FIADB	II
I	2016	KY	Bullitt	103	61	164	FIADB	III
J	2018	VA	Henry	109	57	166	FIADB	II
K	2017	VA	Fairfax	91	108	199	NRCT	IV

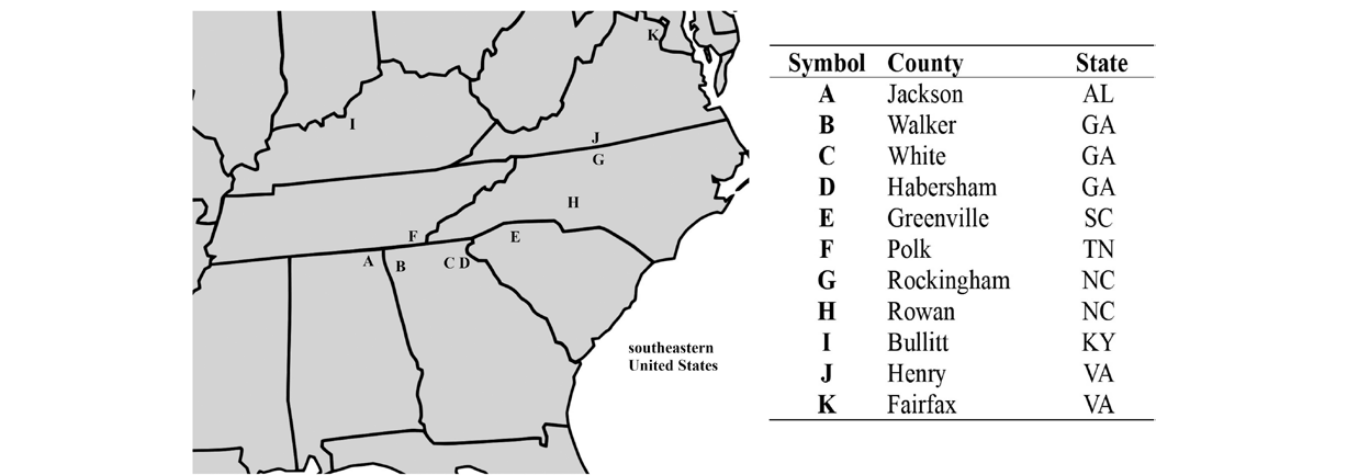


Figure 1 The approximate county locations for twelve large *Pinus virginiana* trees listed in Table 6, including the current national champion in the NRCT, the largest tree measured since 1997 in FIADB (also included in Table 3), and the ten largest trees (by partial points) measured after 2015 in the FIADB.

from probability samples that could be used for the same purpose, perhaps after some adaptation to our procedure. For instance, a focused search in a limited number of urban areas could be conducted using the Urban DataMart FIA database (available at <https://apps.fs.usda.gov/fia/datamart/datamart.html>). Although there are many challenges to tree growth in urban environments, such as root restrictions and elevated air pollution levels, (Dervishi et al. 2022; Franceschi et al. 2023; Hurley and Heinrich 2023; Sand et al. 2018; Shi et al. 2023) some studies have found plant growth to sometimes be greater in urban environments (e.g., Briber et al. 2015, and Gregg et al. 2003). Additionally, some urban trees are treasured and protected, allowing them to grow quite large.

The species listed in Table S1 would have a Z_s statistic of 1 if we assumed $PPFIA_s = 0$ for these species, as they have not been observed by FIA on forested sample plots. As noted, ninety-four of these species are listed as potentially observable by FIA so they must be relatively rare on forest land to not have been observed at all by FIA since 1998. An exception to this would be for species that are rare but known to be clustered in small geographic areas. An adaptive sampling design as suggested in Roesch (1993) would be less helpful for

the species listed in Table S1, as such a design does require the first observance of a species. Nevertheless, FIA could intensify its sample in areas where the species has been found and listed in the NRCT and the various state champion tree lists. We note that FIA has intensified its sample to achieve better estimates of attributes of special interest through various mechanisms in the past, most notably through the incorporation of a macro-plot to increase the sample of very large trees in the Pacific Northwest (Gray et al. 2012; Roesch and Van Deusen 2010; Tinkham et al. 2018).

Besides assisting in the quest for the biggest trees, this suite of tables can provide valuable information to the National Forest Inventory (NFI) effort in the United States. We have already mentioned the quality control potential that use of the NRCT has for FIA. Additionally, the NFI will become increasingly concerned with all tree species, which might someday require a change in the sample design to capture more of these “outlier” species as general interest in rare events increases. So far, all we can say about the species listed in Table S1 is that they are too rare on forest land for the FIA sample to have observed them, whereas it is often the extremes of a population that are of the greatest interest. One reason for this

is that changes in the extremes can serve as a harbinger or even as the “canary in the coal mine” as we try to assess the full effects of climate change. That is, most of these species are rare because their environmental requirements are very limited and specific, and minute changes in one or more environmental measures might have a large impact, either positive or negative, on the abundance of these rare species first.

Conclusion

Natural resource professionals are well trained in formal scientific investigatory methods but are often less familiar with the many ways in which citizen science efforts can and do further our understanding of our environment. These efforts are universally beneficial as they keep scientists and their communities engaged with each other. Here, we have shown how two seemingly disparate efforts can complement each other, synergizing the commonalities between the probability sample conducted by the Forest Service’s FIA program and the citizen science effort associated with the NRCT. Although the inputs, attributes, and goals of these efforts differ greatly, the strengths of each can contribute to and further the objectives of the other.

Finally, we note that, with little to no modification, the approach we have used here could be used in many other countries, because many have both a champion tree citizen science effort, often known as a monument tree program, and an NFI. A few examples of the former can be found in Moore and Hughes (2014), Lai et al. (2019), and Sladonja et al. (2023), whereas nationally compatible examples of the latter are found in Gschwantner et al. (2022), Mackenthun (2011), MPIGA (2018), NFI (2021), Pelz and Cunia (1985), and Zeng et al. (2015).

Supplementary Material

Supplementary data is available at *Journal of Forestry* online.

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Conflict of Interest

The authors have no conflicts of interest to declare.

Data availability

The data underlying this article came from two publicly available databases. The Forest Inventory and Analysis Database (FIADB) can be accessed at <https://apps.fs.usda.gov/fia/datamart/datamart.html>, and the National Register of Champion Trees can be accessed at <https://www.americanforests.org/champion-trees/champion-trees-registry/>.

Appendix 1: In this appendix, we include the latest URL found for a website leading to information for each state’s champion tree program.

State	Website URL
Alabama	https://forestry.alabama.gov/Pages/Management/Champion_Tree.aspx
Alaska	Has been managed by Don Bertollette (donbertollette@me.com) Alaska’s Big Tree Coordinator
Arizona	https://dffm.az.gov/arizonas-magnificent-trees
Arkansas	https://www.agriculture.arkansas.gov/forestry/champion-trees/
California	https://californiabigtrees.calpoly.edu/
Colorado	https://www.coloradotrees.org/champion-tree-program
Connecticut	http://oak.conncoll.edu:8080/notabletrees/Champs-ByScientificName.jsp
Delaware	https://news.delaware.gov/2019/05/21/new-big-trees-of-delaware/
Florida	https://www.fdacs.gov/Forest-Wildfire/Our-Forests/Florida-Champion-Trees
Georgia	https://gatrees.org/learn-explore/champion-trees/
Hawaii	https://dlnr.hawaii.gov/forestry/info/big-tree/
Idaho	https://www.uidaho.edu/extension/forestry/big-trees
Illinois	https://univofillinois.maps.arcgis.com/apps/Shortlist/index.html?appid=090997c-75b2e4ca39dd7d35db4328962
Indiana	https://www.in.gov/dnr/forestry/forestry-publications-and-presentations/indiana-big-tree-register/
Iowa	https://www.iowadnr.gov/Portals/1dnr/uploads/forestry/Big%20Trees%20of%20Iowa%20Web-ready.pdf
Kansas	https://www.kansasforests.org/discover_kansas_forests/championtrees.html
Kentucky	https://eec.ky.gov/Natural-Resources/Forestry/ky-champion-trees/Pages/default.aspx
Louisiana	https://www.laforestry.com/champion-trees-in-louisiana
Maine	https://digitalmaine.com/for_docs/265/
Maryland	https://dnr.maryland.gov/forests/Pages/trees/bigtree.aspx
Massachusetts	https://www.mass.gov/guides/massachusetts-legacy-tree-program
Michigan	https://michiganbotanicalsociety.org/big-tree
Minnesota	https://www.dnr.state.mn.us/trees_shrubs/bigtree/big-tree-champions.html
Mississippi	https://www.mfc.ms.gov/2016/02/new-champion-trees-discovered/
Missouri	https://mdc.mo.gov/trees-plants/champion-trees
Montana	https://dnrc.mt.gov/Forestry/Forest-Management/montana-big-trees-program
Nebraska	https://nfs.unl.edu/registry
Nevada	http://forestry.nv.gov/nevadas-big-tree-program
New Hampshire	https://extension.unh.edu/natural-resources/forests-trees/trees/nh-big-trees
New Jersey	https://www.nj.gov/dep/parksandforests/forest/big-trees/treesofinterest.html
New Mexico	https://www.emnrd.nm.gov/sfd/urban-and-community-forestry/find-or-register-a-big-tree/
New York	https://www.dec.ny.gov/animals/5248.html

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