

El Paso, Texas Project Area

Community Forest Assessment

December 2014

Prepared for:

City of El Paso
300 N. Campbell
El Paso, TX 79901
(915) 212-0000

Prepared by:

Davey Resource Group

A Division of The Davey Tree Expert Company

Western Region Office

6005 Capistrano Unit A
Atascadero, CA 93422

Contact:

Vince Mikulanis

Vince.Mikulanis@davey.com

(619) 921-2746

This project was developed in partnership with:



Arizona State Forestry



Davey Resource Group



USDA Forest Service



New Mexico State Forestry



City of Las Cruces



City of El Paso



City of Albuquerque



City of Phoenix



Texas A&M State Forestry

The grant funding for this project originates from the Cooperative Forestry Assistance Grant from the U.S Department of agriculture (USDA) Forest Service under Federal Catalog of Federal Domestic Assistance (CFDA) #10.664 (Grant Number 11-DG-11031600-039)

This report is adapted from the standard i-tree Ecosystem Analysis report that is generated upon submission of i-eco data. Tree Eco (formerly Urban Forest Effects model) was cooperatively developed by USDA Forest Service Northeastern Research Station (NRS), the USDA State and Private Forestry's Urban and Community Forestry Program and Northeastern Area, the Davey Tree Expert Company, and SUNY College of Environmental Science and Forestry.

Table of Contents

Tables and Figures	iii
Executive Summary.....	1
Introduction	2
Methods.....	3
Project Area	3
i-Tree Eco Model and Field Measurements	4
Findings	8
Tree Population Characteristics	8
Species Distribution	8
Species Richness	10
Trees by Land Use Distribution	11
Tree Density	12
Relative Age Distribution	13
Tree Condition.....	14
Tree Species Origin Distribution	15
Cover and Leaf Area.....	16
Importance Value and Leaf Area	16
Groundcover and Canopy	17
Economic and Ecological Benefits.....	19
Structural and Functional Values	19
Relative Tree Effects	19
Air Quality	21
Carbon Storage and Sequestration	22
Oxygen Production.....	23
Avoided Stormwater Runoff	23
Building Energy Use	26
Potential Pathogen Impacts.....	27
Pathogen Proximity and Risk	27
Insect and Pathogen Risk by Tree Species	29
Appendix I. Glossary and Calculations	30
Appendix II. Comparison of Urban Forests	32
Appendix III. General Recommendations for Air Quality Improvement.....	34
Appendix IV. Species Distribution and Botanical Names	35
References	37

Tables and Figures

Table 1. Common Tree Species Composition	8
Table 2. Species Richness.....	10
Table 3. Condition (%) by Land Use	14
Table 4. Trees Listed as Invasive in Texas	15
Table 5. Top 20 Species by Importance Value	16
Table 6. Percent Ground Cover by Land Use	18
Table 7. Top Oxygen Producing Species	23
Table 8. Vegetation NOT Accounted for in Model.....	24
Table 9. Annual Energy Savings Due to Trees Near Residential Buildings.....	26
Table 10. Annual Savings ¹ (\$) in Residential Energy Expenditure.....	26
Table 11. Pathogen Risk by Tree Species.....	29
Table 12. Tree Benefits in Other Areas.....	32
Table 13. Per-Acre Values of Tree Effects in Other Areas.....	33
Table 14. Urban Forest Management Strategies to Improve Air Quality	34
Table 15. Species Distribution and Botanical Names.....	35
Figure 1. Project Area Boundaries, Plot Locations, and City Limits	3
Figure 2. Common Species.....	9
Figure 3. Percent of Trees by Land Use	11
Figure 4. Trees per Acre by Land Use.....	12
Figure 5. Citywide Relative Age Distribution	13
Figure 6. Age Distribution by Land Use.....	13
Figure 7. Percent of Live Trees by Species Origin	15
Figure 8. Ground Cover Type Distribution	17
Figure 9. Annual Pollution Removal (bars) and Associated Value (points).....	21
Figure 10. Top Ten Carbon Sequestering Species.....	22
Figure 11. Rainfall Interception Value (bars) and Number of Trees (points).....	25
Figure 12. Number of Susceptible Trees (Bars) and Structural Value (Points) by Pest.....	27

Definitions for **bold** words are available in the Glossary.
 Monetary values are reported in US dollars throughout the report.

Executive Summary

Understanding an urban forest's structure, function and value can promote management decisions that will improve human health and environmental quality. Different tree species contribute different benefits at varying levels, so a community that wants to manage the urban forest with specific benefits in mind may carefully select species to plant. Tree age and stature also greatly impact benefits, and this report provides an overview of the current relative age distribution and urban forest structure. Finally, managers can use this data to understand pests and diseases present, and not yet found in the area.

In 2013, the New Mexico Energy, Minerals and Natural Resources Department (EMNRD) contracted with Davey Resource Group (DRG) to collect field data and perform an analysis of the ecosystem services and benefits of trees on a landscape level. Data was collected in 201 designated plots which were randomly distributed across the El Paso project area and analyzed using the i-Tree Eco model developed by the U.S. Forest Service, Northern Research Station.

Based on this sample, it is estimated that 1,281,000 trees exist across the sample area which covers 158.2 square miles. Tree canopy is estimated to cover 5.1% of the land area. The most common species found were Italian cypress, Afghan pine, and Mexican fan palm.

The tree population provides valuable benefits to the communities in the El Paso Project Area. The trees are important for air **pollution removal**, intercepting a net 318 **tons** of air pollution annually, valued at \$247,000 dollars. They store 92,800 tons of carbon valued at \$6.61 million and sequester 7,430 tons each year, valued at \$529,000 dollars. **Carbon storage** and **carbon sequestration** values are based on a current market value of \$71.21 per ton. Avoided carbon emissions are valued at \$384,000 annually. The tree population reduces stormwater runoff by 32.9 million cubic feet per year, valued at \$2.19 million dollars. Approximately 14,100 tons of oxygen are produced annually by this resource. The largest monetary value related to the urban forest is the structural values of the trees, which are based on the replacement value of the tree at its present size and condition. These equate to \$1.02 billion dollars.

Predicting emergent pest infestations is more accurately done by local area experts, but the i-Tree Eco model does provide some valuable data about pests that may become a problem. These should be considered in conjunction with the opinions of local pest and disease experts.

El Paso Project Area urban forest managers can use this data to further understand the composition, species and age distribution, benefits and values, and possible risks in the urban forest. Air Quality and Utility managers can use the data to support planting and maintaining appropriate tree species to maximize air quality, stormwater runoff, and energy benefits. This data, unique to the project area, can help managers understand the unique attributes of their communities' urban forests.

Introduction

The urban forest contributes to a healthier, more livable, and prosperous El Paso. This Community Forest Assessment can provide benchmarks for the current amount of canopy, leaf surface area, and structure of the urban forest including both public and private trees. It also provides an overview of the ecosystem services of these trees, providing an important perspective for the city's understanding of the urban forest.



The urban forest contributes to a healthier, more livable, and prosperous El Paso.

The City of El Paso is located in far west Texas on the Rio Grande across the border from Ciudad Juarez, Mexico. The climate is a high desert at 3,800 feet elevation. Nicknamed “Sun City”, the average rainfall is just 8.71 inches (NOAA). In this kind of environment, urban trees must be adapted to the weather conditions, or receive regular irrigation. The climate significantly limits the species palette in the region. Without irrigation, trees rarely survive, and even with irrigation, plant growth rates are typically slow, and small-stature trees are common.

The project area included communities within the city limits of El Paso, Texas. In order to provide a more accurate representation of the trees in the urban forest, the project area did not include some of the large natural areas that were not specifically managed for vegetation. As a result, the total included acreage was 101,238, or 158.1 square miles out of the city's 256.3 square miles of land.

Methods

Project Area

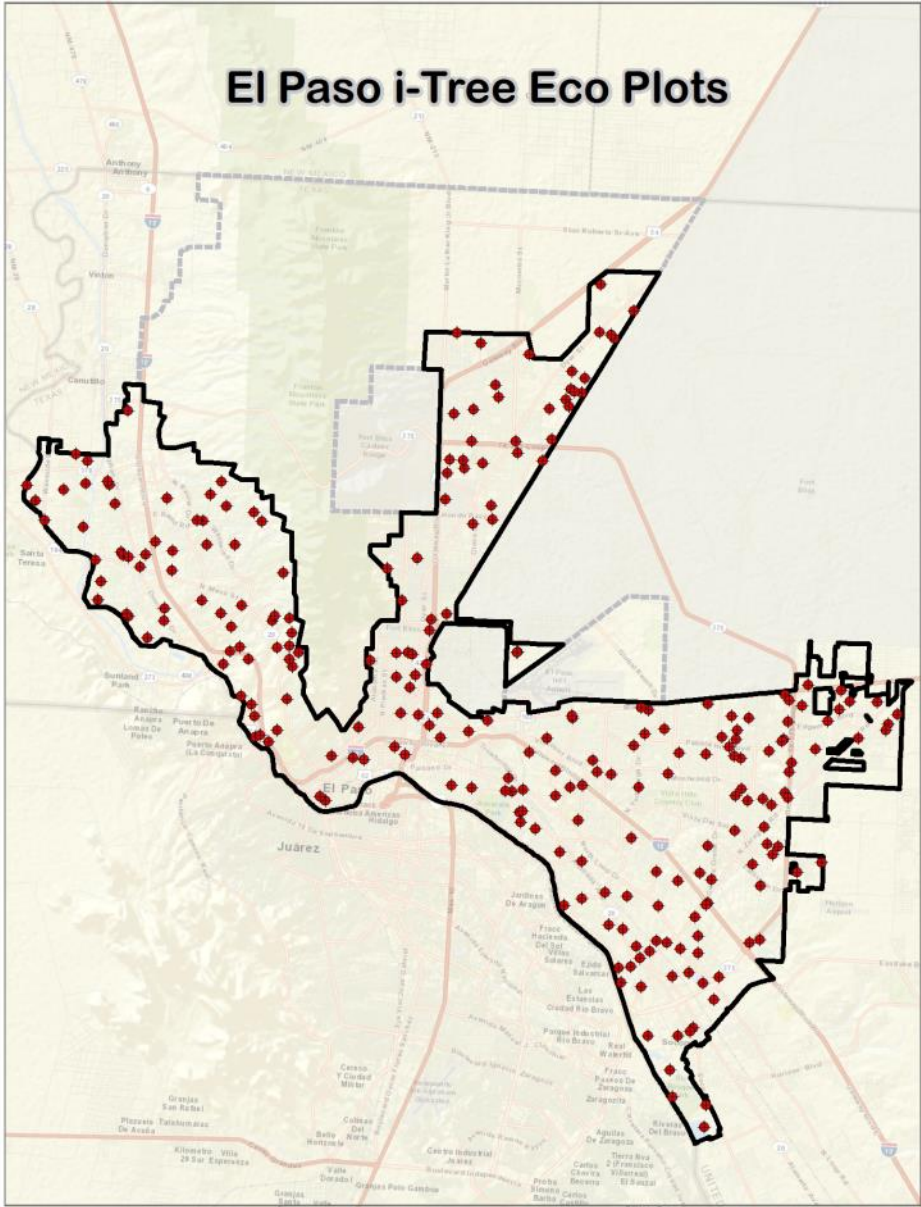


Figure 1. Project Area Boundaries, Plot Locations, and City Limits

The study area includes the 158.2 square miles within the black boundary in Figure 1. The red dots show the random distribution of the 201 measured plots. This area was selected because these are primarily urban areas of the city, and likely more consistent with the i-Tree Eco model. It is expected that the vegetation in the included areas most profoundly influences the urban ecosystem, providing the benefits calculated by the i-Tree Eco model. That is not to say that the trees and shrubs in the excluded areas are not important in providing air quality, stormwater, carbon, and energy benefits, but their influence in the i-Tree Eco model is diminished since they are not in close proximity to urban infrastructure and air conditioned buildings, so their contribution is not likely consistent with the more urban land areas.

The excluded areas provide benefits to the community and if they become more developed should be included in future studies. One factor that is not calculated in the study is the urban heat island effect. Vegetation on land outside the study area may mitigate heat associated with buildings and paved surfaces within the study area, and those benefits are not reflected in this model, which is geared toward understanding tree benefits in urbanized areas (Weng et al., 2003).

For example, a tree in an undeveloped area may provide the same carbon storage benefits as its urban counterpart, but because it is not in close proximity to infrastructure, the stormwater benefits are negligible. The pollutant absorption capacity depends on many factors including levels of pollutants, wind and dispersal, and proximity to the source of pollution; thus the capacity of a tree in an undeveloped area to absorb pollution is difficult to calculate with this model, which presumes urban infrastructure and activities are nearby. The tree is also unlikely to provide substantial property value benefits or have a replacement value since wildland trees that fail are not typically replaced. Finally, since the tree is not near buildings, it cannot mitigate the energy use of air-conditioned space. So, while it is fair to say the trees still have value and provide benefits, those benefits do not fit with the attributes in the i-Tree Eco model, and it is reasonable to exclude them from the study.

i-Tree Eco Model and Field Measurements

Model Components

The model selected to calculate urban forest benefits is the i-Tree Eco model. The i-Tree Eco model is designed to use standardized field data from randomly located plots and local hourly air pollution and meteorological data to quantify urban forest structure and its numerous effects [Nowak & Crane, 2000], including:

- Urban forest structure (e.g., species composition, tree health, leaf area, etc.).
- Amount of pollution removed hourly by the urban forest, and its associated percent air quality improvement throughout a year. Pollution removal is calculated for ozone, sulfur dioxide, nitrogen dioxide, carbon monoxide and particulate matter (<2.5 microns and <10 microns).
- Total carbon stored and net carbon annually sequestered by the urban forest.

- Effects of trees on building energy use and consequent effects on carbon dioxide emissions from power plants.
- Structural value of the forest as a replacement cost.
- Potential impact of infestations by pests or pathogens.

In the field, 201 0.1-acre plots were randomly distributed across the study site using the ArcView GIS random point generation tool. All field data was collected during the leaf-on season to properly assess tree canopies. Within each plot, typical data collection includes land use, ground and tree cover, individual tree attributes of species, stem diameter, height, crown width, crown canopy missing and dieback, and distance and direction to residential buildings [Nowak et al., 2005 and Nowak et al., 2008].

The land uses were determined based on the primary use of the land at the sample site. Single Family Residential was assigned to sites where the primary use was housing for four families or fewer, and Multi-Family Residential included sites where structures had more than four residential units. Commercial/Industrial was assigned to buildings and associated landscaped areas and parking lots where the primary use was the sale of goods or services, or manufacturing. Parks included publically-owned land where the primary activities were games and recreation, or the land was protected for conservation purposes. The land use category Other was assigned to sites that didn't fit within the above descriptions.

The i-Tree Eco model uses a local list of invasive plants to determine how many of the trees in the sample are invasive. In this case, the list was developed by the City of Austin for use in Central Texas [WPDR, 2013] These lists are not exhaustive and they cover invasive species of varying degrees of invasiveness and distribution. Tree species that are identified as invasive by the state invasive species list are cross-referenced with native range data. This helps eliminate species that are on the state invasive species list, but are native to the study area.

Urban Tree Benefit and Pathogen and Pest Risk Calculations

To calculate current carbon storage, biomass for each tree was calculated by incorporating measured tree data into equations from the literature. Open-grown, maintained trees tend to have less biomass than predicted by forest-derived biomass equations [Nowak, 1994]. To adjust for this difference, i-Tree Eco multiplies biomass results for open-grown urban trees by 0.8. The i-Tree Eco model converted tree dry-weight biomass to stored carbon by multiplying by 0.5.

To estimate the gross amount of carbon sequestered annually, average diameter growth from the appropriate genera and diameter class and tree condition was added to the existing tree diameter (year x) to estimate tree diameter and carbon storage in year x+1. Carbon storage and carbon sequestration values are based on i-Tree Eco estimated local carbon values.

The amount of oxygen produced is estimated from carbon sequestration based on atomic weights: net O₂ release (kg/yr) = net C sequestration (kg/yr) × 32/12. To estimate the net carbon sequestration rate, the amount of carbon sequestered as a result of tree growth is reduced by the amount lost resulting from tree mortality. Thus, net carbon sequestration and net annual oxygen production of the urban forest account for decomposition [Nowak, Hoehn, & Crane, 2007].

Air pollution removal estimates are derived from calculated hourly tree-canopy resistances for ozone, and sulfur and nitrogen dioxides based on a hybrid of big-leaf and multi-layer canopy deposition models [Baldocchi, 1988 and Baldocchi, Hicks, & Camara, 1987]. As the removal of carbon monoxide and particulate matter by vegetation is not directly related to transpiration, removal rates (deposition velocities) for these pollutants were based on average measured values from the literature [Bidwell & Fraser, 1972 and Lovett, 1994] that were adjusted depending on leaf phenology and leaf area. Removal estimates of particulate matter less than 10 microns incorporated a 50% resuspension rate of particles back to the atmosphere [Zinke, 1967]. Recent updates (2011) to air quality modeling are based on improved leaf area index simulations, weather and pollution processing and interpolation, and updated pollutant monetary values [Hirabayashi, Kroll, & Nowak, 2011, Hirabayashi, Kroll, & Nowak, 2012, and Hirabayashi, 2011].

Air pollution removal value was calculated based on local incidence of adverse health effects and national median externality costs. The number of adverse health effects and associated economic value is calculated for ozone, sulfur dioxide, nitrogen dioxide, and particulate matter <2.5 microns (PM_{2.5}) using the U.S. Environmental Protection Agency's Environmental Benefits Mapping and Analysis Program (BenMAP). The model uses a damage-function approach that is based on the local change in pollution concentration and population [Davidson et al., 2007].

National median externality costs were used to calculate the value of carbon monoxide removal and particulate matter less than 10 microns and greater than 2.5 microns [Murray, Marsh, & Bradford, 1994]. PM₁₀ denotes particulate matter less than 10 microns and greater than 2.5 microns throughout the report. As PM_{2.5} is also estimated, the sum of PM₁₀ and PM_{2.5} provides the total pollution removal and value for particulate matter less than 10 microns.

Annual avoided surface runoff is calculated based on rainfall interception by vegetation, specifically the difference between annual runoff with and without vegetation. Although tree leaves, branches, and bark may intercept precipitation and thus mitigate surface runoff, only the precipitation intercepted by leaves is accounted for in this analysis.

The value of avoided runoff is based on estimated or user-defined local values. The U.S. value of avoided runoff is based on the U.S. Forest Service's Community Tree Guide Series [USFS].

Seasonal effects of trees on residential building energy use were calculated based on procedures described in the literature [McPherson & Simpson, 1999] using distance and direction of trees from

residential structures, tree height and tree condition data. To calculate the monetary value of energy savings, local or custom prices per MWH or MBTU are utilized.

Structural values were based on valuation procedures of the Council of Tree and Landscape Appraisers, which uses tree species, diameter, condition, and location information [Nowak et al., 2002].

Potential pest and pathogen risk is based on their range maps and the known pest and pathogen host species that are likely to experience mortality. Range maps from the Forest Health Technology Enterprise Team (FHTET) [2010] were used to determine the proximity of each pest or pathogen to El Paso County. It was established whether the insect/disease occurs within the county, is within 250 miles of the county edge, is between 250 and 750 miles away, or is greater than 750 miles away.

Findings

Tree Population Characteristics

This section provides an overview of the species, condition, density, geographic origin, and age (size class) of the tree population. These values help provide context for the following sections on canopy cover and leaf area, as well as the ecological and economic benefits of El Paso’s public and private trees.

Species Distribution

The sample identified 50 unique tree species, but the urban forest likely has far greater diversity. Table 1 and Figure 2 show the ten most prevalent species found in the sample. Based on this sample, it is estimated that the urban forest of El Paso has a total of 1,281,000 trees with a tree canopy cover of 5.1%. Because of the sampling method used, the species distribution has very high error rates, and species proportions should not be relied on for management decisions. The i-Tree Streets model is more appropriate for determining species composition in the community if desired.

Table 1. Common Tree Species Composition

Species	# of Trees	Standard Error (+/-)	Error %
Italian cypress	330,880	105,531	32%
Afghan pine	138,655	35,617	26%
Mexican fan palm	93,993	30,013	32%
White mulberry	75,630	24,754	33%
Pecan	63,385	34,626	55%
Velvet ash	59,738	26,780	45%
Siberian elm	41,686	21,227	51%
Tree of heaven	33,088	25,352	77%
Chinaberry	31,377	17,381	55%
Chitalpa	27,273	20,169	74%

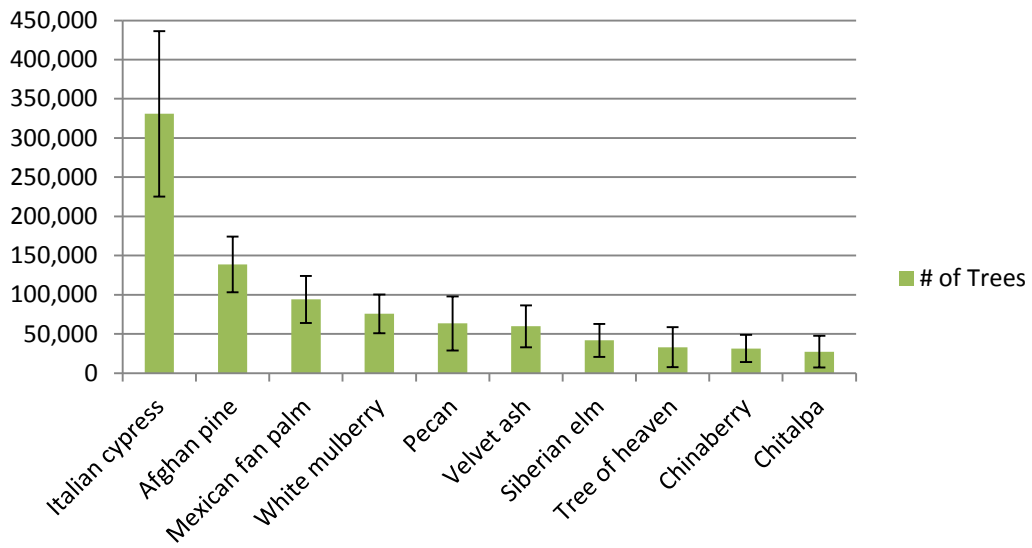


Figure 2. Common Species

Species Richness

Table 2 shows the number of species found in this sample in each Land Use type. This information is provided to show the diversity of trees in the sample, but is not likely a reflection of the full species diversity across the landscape due to the sample size of just 201 plots. The purpose of this plot-based sampling method is to provide a landscape view of the region’s public and private trees. A complete tree inventory can provide a better understanding of species diversity in the project area, but would be prohibitively resource intensive.

The i-Tree Eco model uses established calculations for species diversity indexes, which allow quantitative comparisons of species richness. The Shannon-Wiener Diversity Index assumes that all the species in an area have been sampled, and has a moderate sensitivity to sample size. The Menhinick Index is an indicator of species dominance and has a low sensitivity to sample size and therefore may be more appropriate for comparisons among cities. The Simpson’s Diversity Index is an indicator of species dominance and has a low sensitivity to sample size and is appropriate for comparisons between land-use types.

Table 2. Species Richness

Primary Index	Species	Species/ Acre	Shannon- Wiener Diversity Index	Menhinick Index	Simpson’s Diversity Index
Commercial/Industrial	6	1.28	1.65	1.60	6.50
Multi-Family Residential	12	10.91	2.40	2.50	18.07
Other	3	1.11	0.91	0.83	2.60
Parks	7	3.29	1.84	1.43	7.46
Single-Family Residential	44	5.15	2.79	3.07	7.30
CITY TOTAL	50	2.55	3.11	2.99	11.38

Trees by Land Use Distribution

Based on the sampled plots, about 1.3 million trees are present in the study area on public and private property in El Paso. Trees in single family residential areas make up 79% of the total in this assessment (Figure 3). This Community Forest Assessment is based on stratifying the sampled plots to the land use areas of each type to determine an estimated number of trees by land use, so it helps to understand where most of the sampled trees occurred.

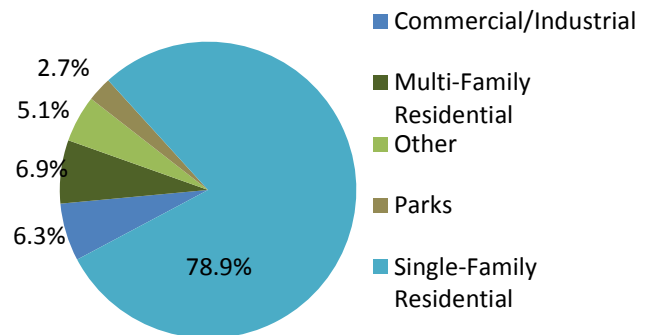


Figure 3. Percent of Trees by Land Use

Tree Density

Another way to consider tree distribution is to analyze the number of trees per acre in each land use type (Figure 4).

Residential land uses typically feature the most trees per acre, and El Paso is no exception. The residential areas had – 20-25 trees per acre, followed by parks with 11 trees per acre. Over all, the tree density in the studied area is 12.7 trees per acre. Appendix II shows comparable values from other cities, including other southwest cities, as reported by i-Tree Eco.



Parks have about 11 trees per acre.

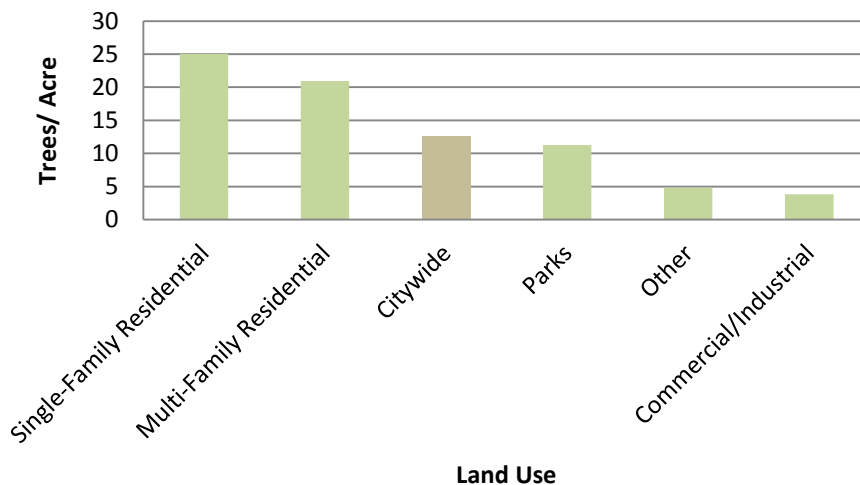


Figure 4. Trees per Acre by Land Use

Relative Age Distribution

For most woody plants, the DBH increases incrementally annually, so it may be used to estimate the age of the population. Based on the relative relationship between age and diameter, the distribution of the sampled trees indicates a young or small-statured population with 53% of trees under 6" DBH (Figure 5).

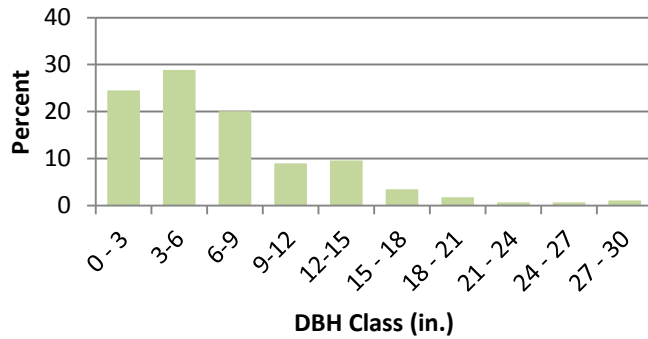


Figure 5. Citywide Relative Age Distribution

Considering the land uses, Figure 6 shows some patterns by land use, for example, the Other land use includes agricultural lands and shows a large portion of trees in the 6-9" DBH range as one might expect in an orchard.

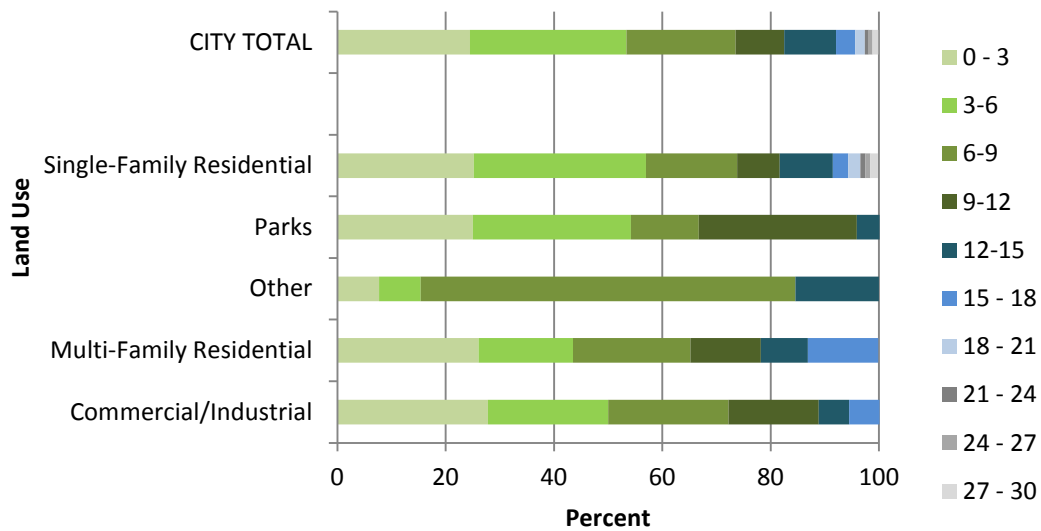


Figure 6. Age Distribution by Land Use

Tree Condition

Tree condition can be related to species fitness, tree age, environmental stressors, and maintenance, and these typically vary with land use. The majority (84%) of trees in the sample are in good to excellent condition. The Commercial/Industrial land use had the highest percent of dead dying and critical trees. While trees sampled in residential areas were largely in excellent to good condition. (Figure 7 and Table 3).

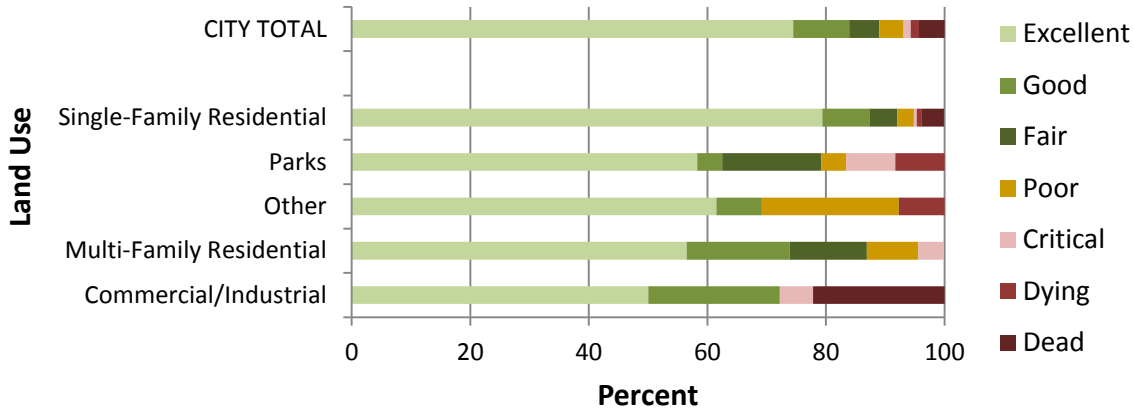


Figure 7. Condition (%) by Land Use

Table 3. Condition (%) by Land Use

Land Use	Excellent		Good		Fair		Poor		Critical		Dying		Dead	
	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)
Commercial/Industrial	50	18.62	22.2	11.02					5.6	4.83			22.2	15.05
Multi-Family Residential	56.5	10.72	17.4	4.3	13	5.77	8.7	6.9	4.3	2.95				
Other	61.5	25.45	7.7	7.18			23.1	16.6			7.7	7.18		
Parks	58.3	9.31	4.2	2.96	16.7	5.59	4.2	2.96	8.3	2.8	8.3	7.3		
Single-Family Residential	79.4	4.48	7.9	2.01	4.7	1.7	2.8	1.14	0.5	0.46	0.9	0.67	3.7	2.02
Citywide	74.5	4.03	9.4	1.8	5.1	1.41	4.1	1.33	1.2	0.52	1.4	0.68	4.3	1.85

Tree Species Origin Distribution

Urban forests are composed of a mixture of native and exotic tree species. Thus, urban forests often have a tree diversity that is higher than surrounding native landscapes. Increased tree diversity can minimize the overall impact or destruction of the urban forest resource by a species-specific pest or pathogen, but it can also pose a risk to native plants if some of the exotic species spread beyond planting sites and aggressively suppress the establishment of native species in both the urban and wildland areas. Those invasive plant species are often characterized by their vigor, ability to adapt, reproductive capacity, and general lack of natural enemies. These abilities enable them to displace native plants and make them a threat to natural areas [USDA, 2011]. .

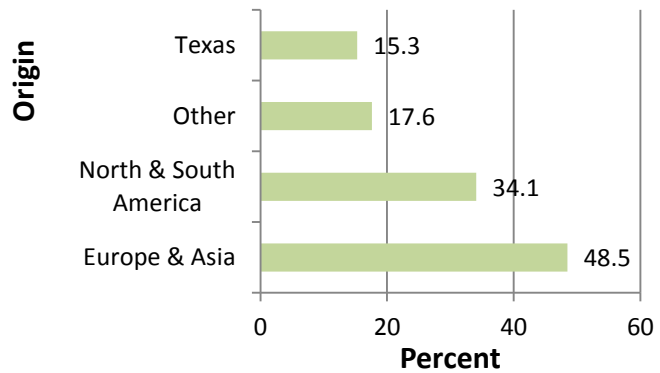


Figure 7. Percent of Live Trees by Species Origin

Figure 8 shows the origin distribution of species found in the sample. In El Paso, about 34% of the trees are species native to North or South America, and 15% are native to Texas (Figure 7). Totals do not sum to 100% due to rounding, and because Texas natives are a subset of North American natives.

Five of the 59 tree species (Table 4) sampled in El Paso are identified as invasive on the state invasive species list [WPDR, 2013]. These invasive species comprise 15% of the tree population though they may only cause a minimal level of impact. The three most common invasive species are white mulberry (6% of population), tree of heaven (2.6%), and chinaberry (2.5%). The model does not calculate the level of impact these trees have on local ecosystems, an assessment best left to the determination of local forest managers.

Table 4. Trees Listed as Invasive in Texas

Species Name	Number of trees	% Tree Number	Leaf Area (mi ²)	% Leaf Area
White mulberry	75,630	5.9	5.71	19.7
Tree of heaven	33,088	2.58	0.21	0.72
Chinaberry	31,377	2.45	0.77	2.67
Mimosa	26,650	2.08	0.07	0.23
TOTAL	192,267	15	7.26	25.03

Cover and Leaf Area

Importance Value and Leaf Area

Many tree benefits equate directly to the amount of healthy leaf surface area of the plant. In the project area, the most impactful species in terms of leaf area and population are Afghan pine, Italian cypress, and white mulberry composing 43% of the population and 40% of the leaf surface area. The 20 most important species are listed in Table 5. Importance values (IV) are calculated as the sum of relative leaf area and relative composition.

Table 5. Top 20 Species by Importance Value

Species	Percent Population	Percent Leaf Area	Importance Value
Afghan pine	10.8	18.1	28.9
Italian cypress	25.8	2.6	28.5
White mulberry	5.9	19.7	25.6
Velvet ash	4.7	12.3	17.0
Pecan	4.9	9.0	13.9
Mexican fan palm	7.3	2.7	10.0
Aleppo pine	1.0	5.9	6.9
Siberian elm	3.3	2.8	6.0
Juniper spp	2.1	3.4	5.5
Chinaberry	2.4	2.7	5.1
Honey mesquite	1.2	2.4	3.6
Tree of heaven	2.6	0.7	3.3
Chitalpa	2.1	0.5	2.6
Red cedar spp	1.0	1.5	2.4
Mimosa	2.1	0.2	2.3
Hollywood juniper	0.4	1.9	2.3
Goodding's willow	0.4	1.9	2.2
Sweet acacia	0.7	1.4	2.1
Chaste tree	0.7	1.3	2.0
Oriental arborvitae	1.1	0.7	1.8
Other species	19.4	8.4	27.8

Groundcover and Canopy

Groundcover types impact stormwater runoff, availability of planting sites, and indicate the degree of urban density. The most dominant ground cover type is bare soil, representing 39.5% of the total city ground cover. The model calculates a percent of “plantable space” based on combining all pervious ground cover. In El Paso, 60% of the city is estimated to be plantable, though irrigation is often required to establish young trees and plants. As a layer above the ground, tree canopy is 5.1% and shrubs cover 1.6%. (Figure 8 and Table 6)

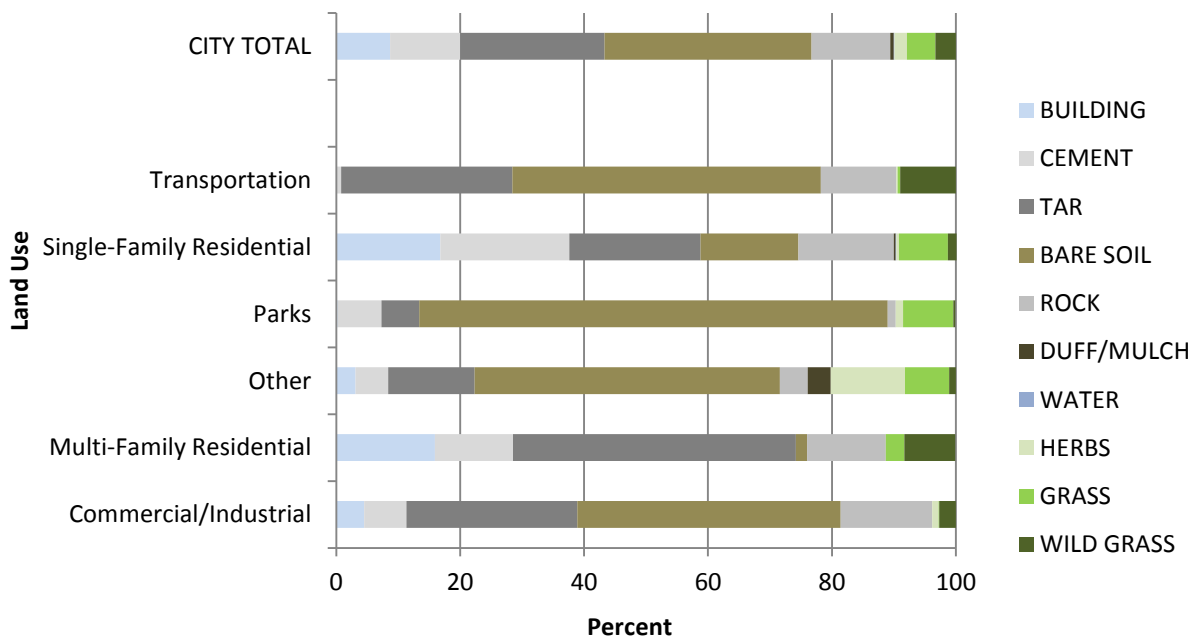


Figure 8. Ground Cover Type Distribution

Table 6. Percent Ground Cover by Land Use

Ground Cover Land Use	BARE SOIL		DUFF/MULCH		WATER		HERBS		GRASS		WILD GRASS	
	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)
Commercial/Industrial	42.5	5.96					1	0.66	0.1	0.06	2.7	0.92
Multi-Family Residential	1.8	0.92							3	2.39	8.2	7.8
Other	49.3	7.37	3.7	3.63			12	5.73	7.1	3.22	1.1	0.57
Parks	75.6	7.13					1.2	0.49	8.1	5.41	0.3	0.32
Single-Family Residential	15.8	2.52	0.3	0.14	0.1	0.11	0.4	0.15	7.9	1.38	1.4	0.48
Transportation	49.8	14.2					0.2	0.18	0.4	0.36	9	8.05
CITY TOTAL	33.4	3.26	0.6	0.49	0.1	0.05	2	0.79	4.6	0.73	3.3	1.57

Ground Cover Land Use	BUILDING		CEMENT		TAR		ROCK	
	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)
Commercial/Industrial	4.5	1.7	6.8	1.45	27.6	5.13	14.8	4.11
Multi-Family Residential	15.9	4.44	12.6	1.92	45.7	7.53	12.7	5.97
Other	3.1	1.91	5.3	2.01	13.9	4.93	4.5	1.78
Parks	0.2	0.23	7.1	2.69	6.1	3.41	1.3	1.13
Single-Family Residential	16.8	1.84	20.8	1.47	21.2	2.51	15.4	1.68
Transportation			0.8	0.52	27.6	13.75	12.2	7.09
CITY TOTAL	8.7	0.87	11.3	0.73	23.3	3.05	12.7	1.75

Economic and Ecological Benefits

Structural and Functional Values

Urban forests have structural values based on the trees themselves (e.g., the cost of having to replace a tree with a similar tree) and functional values (either positive or negative) based on the functions the trees perform (e.g., removing pollution, reducing energy use).

The structural value of an urban forest tends to increase with a rise in the number and size of healthy trees [Nowak, Crane, & Dwyer, 2002]. Annual functional values also tend to increase with increased number and size of healthy trees, and are usually on the order of several million dollars per year. Through proper management, urban forest values can be increased; however, the values and benefits can decrease if the amount of healthy tree cover declines.

Structural values:

- Structural value: \$1.02 billion
- Carbon storage: \$6.61 million

Annual functional values:

- Carbon sequestration: \$529,000
- Pollution removal: \$247,000
- Lower energy costs and carbon emission reductions: \$3.08 million
- Avoided Stormwater Runoff: \$2.19 million

Relative Tree Effects

The urban forest in El Paso provides benefits that include carbon storage and sequestration, and air pollutant removal. To estimate the relative value of these benefits, tree benefits were compared to estimates of average **municipal carbon emissions** [EIA, 2003, and Census.gov, 2003], average **passenger automobile emissions** [EPA, 2002, BTS 2004, and Graham, Wright & Turhollow, 1992], and average **household emissions** [EIA, 2001].

In El Paso, carbon storage is equivalent to:

- **Carbon emissions** in 7 days
- Annual carbon emissions from 55,700 automobiles
- Annual carbon emissions from 28,000 single-family houses

Carbon monoxide removal is equivalent to:

- Annual **carbon monoxide emissions** from 239 automobiles
- Annual carbon monoxide emissions from 95 single-family houses

Nitrogen dioxide removal is equivalent to:

- Annual **nitrogen dioxide emissions** from 1,530 automobiles
- Annual nitrogen dioxide emissions from 1,020 single-family houses

Sulfur dioxide removal is equivalent to:

- Annual **sulfur dioxide emissions** from 6,330 automobiles
- Annual sulfur dioxide emissions from 106 single-family houses

Particulate matter less than 10 micron (PM10) removal is equivalent to:

- Annual **PM10 emissions** from 393,000 automobiles
- Annual PM10 emissions from 37,900 single-family houses

Annual carbon sequestration is equivalent to:

- Amount of carbon emitted in El Paso in 0.5 days
- Annual carbon emissions from 4,500 automobiles
- Annual carbon emissions from 2,200 single-family houses

For definitions and calculations, see Appendix I.

Air Quality

Poor air quality is a common problem in many urban areas. It can lead to decreased human health, damage to trees and shrubs and ecosystem processes, and reduced visibility. The urban forest can help improve air quality by reducing air temperature, directly removing pollutants from the air, and reducing energy consumption in buildings, which consequently reduces air pollutant emissions from power plants. Trees also emit volatile organic compounds that can contribute to ozone formation. Recently, integrative studies have revealed that an increase in tree cover leads to reduced ozone formation [Nowak & Dwyer, 2007].

Pollution removal by trees and shrubs in El Paso was estimated using TECQ hourly local air quality data and weather data. It is estimated that trees and shrubs remove a total of 318 tons of air pollution with an associated value of \$247,100 dollars. Figure 9 shows the tons of pollutants removed and their associated values. Pollution removal was greatest for PM 10 at 143 tons and \$382,595 (Figure 9). This estimate is based on estimated local incidence of adverse health effects of the BenMAP model and national median externality costs associated with pollutants [Abdollahi, Ning, & Appeaning, 2000].

The i-Tree Eco model produced an uncommon result for PM_{2.5}, with a negative annual PM_{2.5} removal value in contrast to the positive yearly amount of PM_{2.5} removed. The i-Tree Eco model calculates pollution removal values based on changes in pollution concentration, not overall tons of pollution removed. Trees remove PM_{2.5} when particulate matter is deposited on leaf surfaces, and rain dissolves and transfers the PM_{2.5} to the soil. However, under certain meteorological conditions (e.g., a month with no rain), trees can re-suspend more particles than they remove, thus causing a negative pollution concentration change.

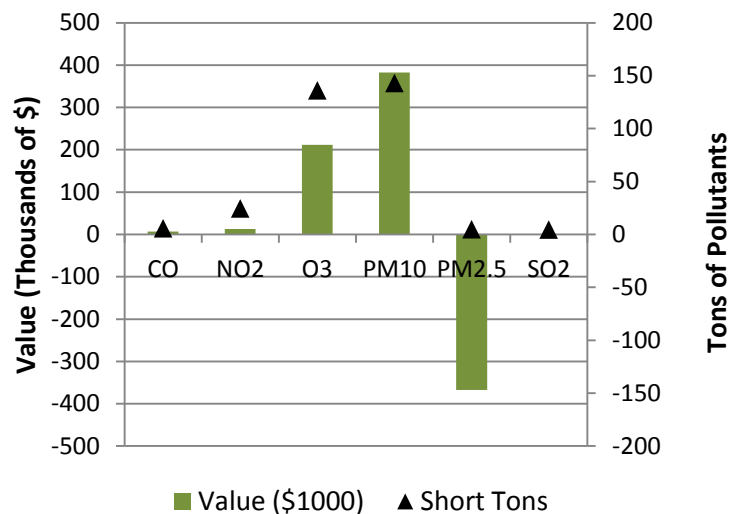


Figure 9. Annual Pollution Removal (bars) and Associated Value (points)

Carbon Storage and Sequestration

Climate change is an issue of global concern. Urban trees can help mitigate climate change by sequestering atmospheric carbon (from carbon dioxide) in tissue, altering energy use in buildings, and consequently altering carbon dioxide emissions from fossil-fuel based power plants [Nowak & Dwyer, 2007].

Trees reduce the amount of carbon in the atmosphere by sequestering carbon in new growth every year. The amount of carbon annually sequestered is increased with the size and health of the trees. The annual sequestration of the project area trees is about 7,430 tons of carbon per year with an associated value of \$529,000. The populations of white mulberry and Afghan pine sequester the greatest amounts of carbon annually, while smaller stature trees, such as chaste trees, have less sequestration capacity. Figure 10 shows the species that sequester the largest amounts of carbon each year.

As trees grow, they store more carbon as wood. As trees die and decay, they release much of the stored carbon back to the atmosphere. Thus, carbon storage is an indication of the amount of carbon that can be lost if trees are allowed to die and decompose. Trees in the project area are estimated to store 92,800 tons of carbon, valued at \$6.61 million. **Carbon storage** and **carbon sequestration** values are calculated based on \$71.21 per ton (see Appendix I for more details).

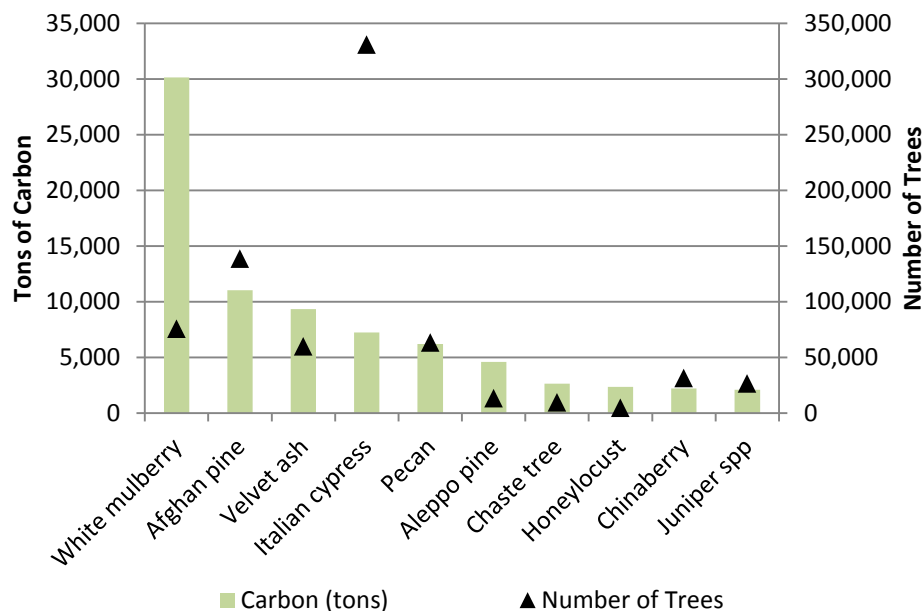


Figure 10. Top Ten Carbon Sequestering Species

Oxygen Production

Oxygen production is one of the most commonly cited benefits of urban trees. The net annual oxygen production of a tree is directly related to the amount of carbon sequestered by the tree, which is tied to the accumulation of tree biomass.

Trees in the project area are estimated to produce 14,100 tons of oxygen per year. Table 7 shows the varying oxygen production of different tree species. This tree benefit is monetarily insignificant because of the large and relatively stable amount of oxygen in the atmosphere and extensive production by aquatic systems. Our atmosphere has an enormous reserve of oxygen [Broecker, 1970]. If all fossil fuel reserves, all trees, and all organic matter in soils were burned, atmospheric oxygen would only drop a few percent, so the monetary value of this contribution is considered negligible.

Table 7. Top Oxygen Producing Species

Species	Oxygen (tons)	Net Carbon Sequestration (tons/yr)	Number of trees	Leaf Area (square miles)
Italian cypress	2,919.8	1,094.9	330,880	0.76
Pecan	1,730.9	649.1	63,385	2.60
Arizona ash	1,479.4	554.8	59,738	3.84
Afghan pine	1,322.1	495.8	138,655	5.24
White mulberry	1,110.0	416.3	75,630	5.71
Chinaberry	805.8	302.2	31,377	0.77
Chaste tree	489.5	183.6	9,454	0.37
Aleppo pine	460.2	172.6	13,325	1.71
Elderberry spp	443.8	166.4	14,181	0.09
Juniper spp	387.8	145.4	26,650	1.00
Chitalpa	354.6	133.0	27,273	0.14
Pear spp	325.8	122.2	14,181	0.15
Cottonwood spp	283.0	106.1	4,727	0.29
Honeylocust	638.9	239.6	14,786	0.44
Tree of heaven	266.0	99.8	33,088	0.21
Siberian elm	239.0	89.6	41,686	0.80
Goodding's willow	211.2	79.2	4,727	0.54
Honey mesquite	167.9	63.0	15,349	0.69
Oriental arborvitae	158.4	59.4	14,181	0.21

Avoided Stormwater Runoff

Surface runoff can be a cause for concern in urban areas, as it can contribute pollution to streams, wetlands, rivers, lakes, and oceans. During precipitation events, some portion of precipitation is intercepted by vegetation (trees, grasses, forbs, and shrubs) while the other portion reaches the ground. The portion of the precipitation that reaches the ground and does not infiltrate into the soil becomes surface runoff. In urban areas, the large extent of impervious surfaces increases the amount of surface runoff, and the cost of infrastructure a community must invest in managing stormwater for the safety of residents and property.

One limitation of the i-Tree Eco model is that grasses and forbs are not specifically accounted for in reporting benefits. In areas such as the desert southwest, these land cover types play a very important role in managing stormwater runoff. Grasses and forbs in the desert southwest may have a proportionately greater role than in other climate types where trees and shrubs are more plentiful. While no specific benefit data is available based on the model, the overall percentage of these land cover types found in this study is substantial. (Table 8). Thus realized stormwater benefits are likely even higher if herbs, grasses, and forbs are considered.

Table 8. Vegetation NOT Accounted for in Model

Ground Cover Land Use	HERBS		GRASS		WILD GRASS		TOTAL %
	%	SE (+/-)	%	SE (+/-)	%	SE (+/-)	
Commercial/Industrial	1	0.66	0.1	0.06	2.7	0.92	3.8
Multi-Family Residential			3	2.39	8.2	7.8	11.2
Other	12	5.73	7.1	3.22	1.1	0.57	20.2
Parks	1.2	0.49	8.1	5.41	0.3	0.32	9.6
Single-Family Residential	0.4	0.15	7.9	1.38	1.4	0.48	9.7
Transportation	0.2	0.18	0.4	0.36	9	8.05	9.6
CITY TOTAL	2	0.79	4.6	0.73	3.3	1.57	9.9

Urban trees are beneficial in reducing surface runoff. Trees intercept precipitation, while their root systems promote infiltration and storage in the soil. The trees throughout the project area help to reduce runoff by an estimated 32.9 million cubic feet a year with an associated value of \$2.19 million dollars. Figure 11 shows the tree species that provide the highest rainfall interception values. This figure demonstrates that population numbers alone do not dictate the interception value, rather, interception is related to leaf surface area which is influenced on tree age, health, species, and stature.

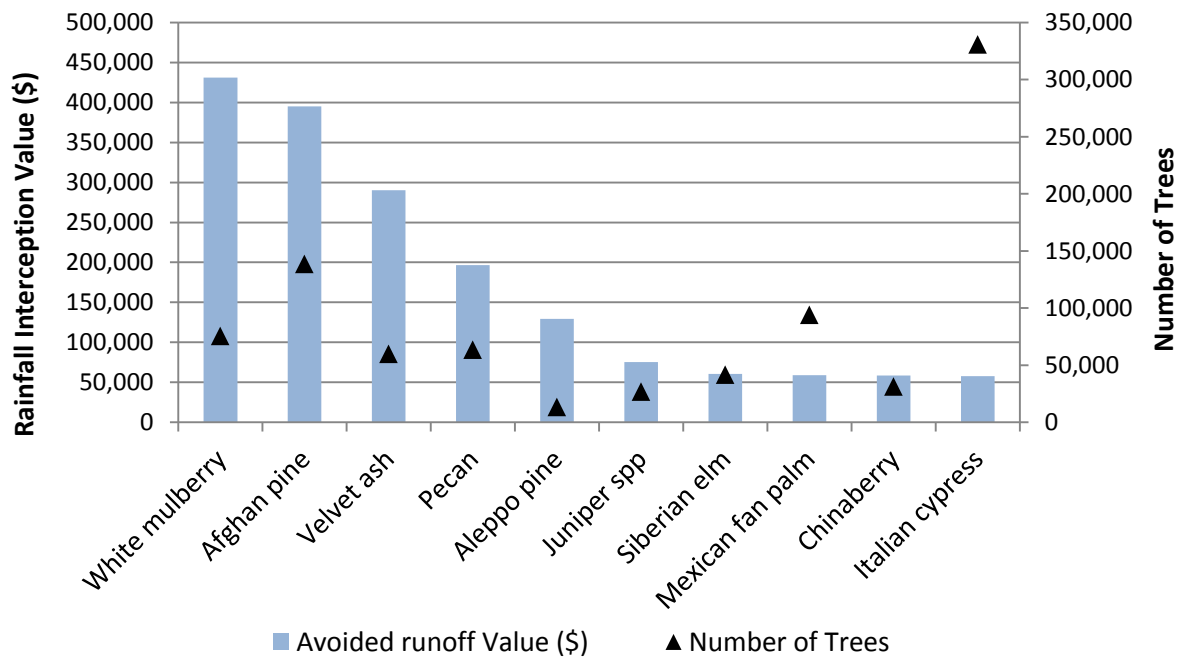


Figure 11. Rainfall Interception Value (bars) and Number of Trees (points)

Building Energy Use

Trees affect energy consumption by shading buildings, providing evaporative cooling, and blocking winter winds. Trees tend to reduce building energy consumption in the summer months and can either increase or decrease building energy use in the winter months, depending on the location of trees around the building. The values for Table 9 were calculated considering savings during heating and cooling seasons. Estimates of tree effects on energy use are based on field measurements of tree distance and direction to air conditioned residential buildings [McPherson & Simpson, 1999].

Trees in the project area are estimated to reduce energy-related costs from residential buildings by \$2.7 million annually (Table 10). Trees also provide an additional \$384,101 in value by reducing the amount of carbon released by fossil-fuel based power plants, a reduction of 5,394 tons of carbon emissions (Tables 9 and 10). Negative numbers indicate an increased energy use or carbon emission.

Table 9. Annual Energy Savings Due to Trees Near Residential Buildings

	Heating	Cooling	Total
MBTU ¹	-47,303	n/a	-47,303
MWH ²	-1,194	28,864	27,670
Carbon Avoided (³ t)	-940	6,334	5,394

¹One million British Thermal Units

²Megawatt-hour

³Short ton

Table 10. Annual Savings¹ (\$) in Residential Energy Expenditure

	Heating	Cooling	Total
MBTU ²	-480,145	n/a	-480,145
MWH ³	-137,191	3,316,474	3,179,283
Carbon Avoided ⁴	-66,961	451,061	384,101

¹Based on the prices of \$114.9 per MWH and \$10.15 per MBTU

²One million British Thermal Units

³Megawatt-hour

⁴Carbon avoided value is calculated based on \$71.21 per ton

Potential Urban Forest Health Impacts

Pathogen and Pest Proximity and Risk

Pathogens and pests can infect and infest urban forests, potentially killing trees and reducing the health, value and sustainability of the urban forest. As pathogens and pests have differing tree hosts, the potential damage or risk of each pest will differ among cities. Thirty-one pathogens and pests were analyzed for their potential impact and compared with pest range maps [ForestHealth.info, 2010] for the contiguous United States. In Figure 12, the pests are color coded according to the county's proximity to the pest occurrence in the United States. Red indicates that the pest is within the county; orange indicates that the pest is within 250 miles of the county; and green indicates that the pest is outside of these ranges in 2013.

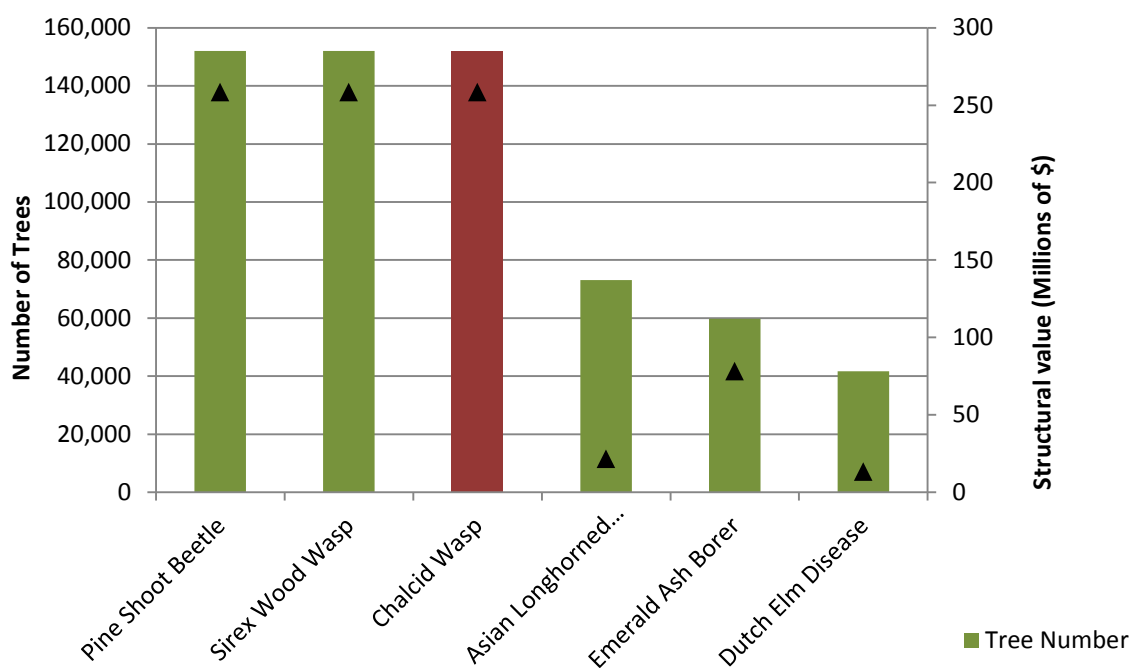


Figure 12. Number of Susceptible Trees (Bars) and Structural Value (Points) by Pest

The pathogens with the largest potential impact on tree populations in the project area are described below. The three pests with the highest possible impact, if they ever migrate to the area, are pine shoot beetle, sirex wood wasp, and chalcid wasp. It should be noted that i-Tree Eco uses the inventory data to calculate the damage potential of a given pathogen to the area of interest. The model does not calculate whether there is a reasonable risk that this pathogen will move there in the foreseeable future. The

model calculates the damage potential, assuming the pathogen will reach the study area and attack the associated tree species.

The following are some of the pests and pathogens identified by the model:

Asian Longhorned Beetle (ALB) [NASPF, 2005] is an insect that bores into and kills a wide range of hardwood species. ALB poses a threat to 5.7 percent of the El Paso urban forest, which represents a potential loss of \$21.4 million in structural value.

American elm, one of the most important street trees in the twentieth century, has been devastated by the Dutch Elm Disease (DED) [NASPF, 1998]. Since first reported in the 1930s, it has killed over 50 percent of the native elm population in the United States. The Siberian elm population in El Paso is somewhat resistant to DED, so this may not become a management problem.

Emerald Ash Borer (EAB) [NASPF, 2005] has killed thousands of ash trees in parts of the United States. EAB has the potential to affect 4.7 percent of the population (\$78.1 million in structural value).

The Pine Shoot Beetle (PSB) [Ciesla, 2001] is a wood borer that attacks various pine species, though Scotch pine is the preferred host in North America, and local experts do not think it's occurrence in El Paso is likely.

The Sirex Wood Wasp (SW) [Haugen, 2005] is a wood borer that primarily attacks pine species. SW poses a threat to 11.9 percent of the El Paso urban forest, which represents a potential loss of \$258 million in structural value.

In addition to the modeled pests, Chalcid Wasp (Family: Eurytomidae) was found in 2008 on Afghan pine. The pest could impact 11.9% of the population (\$258 million in structural value). Local experts also believe mountain pine beetle (*Dendroctonus ponderosae*) may become a problem in the future.

Insect and Pathogen Risk by Tree Species

Based on the host tree species for each pest and the current range of the pest [FHTET, 2009], it is possible to determine what the risk is that each tree species sampled in the urban forest could be attacked by an insect or disease (Table 11). Species that are not listed in the matrix are not known to be hosts to any of the pests analyzed. Species proximities were adjusted based on input from local area experts in El Paso County.

Table 11. Pathogen Risk by Tree Species

Tree Species	Pests						
	ALB	DED	EAB	OW	PSB	SW	CW
Afghan pine					Green	Green	Red
Aleppo pine					Green	Green	
Siberian elm	Green	Red					
Live oak				Orange			
Oak spp				Orange			
Goodding's willow	Green						
Velvet ash			Green				
Mimosa	Green						
Velvet ash			Green				

Species Risk:

- Red indicates that tree species is at risk to at least one pest within county
- Orange indicates that tree species has no risk to pests in county, but has a risk to at least one pest within 250 miles from the county
 - Yellow indicates that tree species has no risk to pests within 250 miles of county, but has a risk to at least one pest that is 250 to 750 miles from the county (No pests fall in this range in El Paso.
 - Green indicates that tree species has no risk to pests within 750 miles of county, but has a risk to at least one pest that is greater than 750 miles from the county

Risk Weight:

Numerical scoring system based on sum of points assigned to pest risks for species. Each pest that could attack tree species is scored as 4 points if red, 3 points if orange, 2 points if yellow and 1 point if green.

Pest Color Codes:

- Red indicates pest is within El Paso County
- Orange indicates pest is within 250 miles of El Paso County
- Yellow indicates pest is within 750 miles of El Paso County (no pests were found in this range)
- Green indicates pest is outside of these ranges

Appendix I. Glossary and Calculations

Carbon dioxide emissions from automobiles assumed six pounds of carbon per gallon of gasoline if energy costs of refinement and transportation are included (Graham, Wright, & Turhollow, 1992)

Carbon emissions Total city carbon emissions were based on 2003 US per capita carbon emissions – calculated as total US emissions (EIA, 2003) divided by the 2003 US total population (Census.gov). This value was multiplied by the population of El Paso (555,417) to estimate total city carbon emissions.

Carbon storage The amount of carbon bound up in the above-ground and below-ground parts of woody vegetation. Carbon storage and carbon sequestration values are calculated based on \$71.21 per ton.

Carbon sequestration The removal of carbon dioxide from the air by plants. Carbon storage and carbon sequestration values are calculated based on \$71.21 per ton.

Diameter at Breast Height (DBH) Is the diameter of the tree measured 4'6" above grade.

Energy saving Value is calculated based on the prices of \$116.9 per MWH and \$11.79 per MBTU.

Household emissions (average) based on average electricity kWh usage, natural gas Btu usage, fuel oil Btu usage, kerosene Btu usage, LPG Btu usage, and wood Btu usage per household (EIA, 2001)
CO₂, SO₂, and NO_x power plant emission per kWh (EPA)
CO emission per kWh assumes 1/3 of one% of C emissions is CO (EIA, 1994)
PM₁₀ emission per kWh (Layton, 2004, 2005)
CO₂, NO_x, SO₂, PM₁₀, and CO emission per Btu for natural gas, propane and butane (average used to represent LPG), Fuel #4 and #6 (average used to represent fuel oil and kerosene) (Abraxas energy Consulting)
CO₂ and fine particle emissions per Btu of wood (Houck et al., 1998)
CO, NO_x and SO_x emission per Btu based on total emissions and wood burning (tons) (www.env.bc.ca, 2005)
Emissions per dry ton of wood converted to emissions per Btu based on average dry weight per cord of wood and average Btu per cord (ianrpubs.unl.edu).

Monetary values (\$) are reported in US Dollars throughout the report.

PM₁₀ consists of particulate matter less than 10 microns and greater than 2.5 microns. As PM_{2.5} is also estimated, the sum of PM₁₀ and PM_{2.5} provides the total pollution removal and value for particulate matter less than 10 microns.

Passenger automobile emissions per mile (average) were based on dividing total 2002 pollutant emissions from light-duty gas vehicles (EPA, 2004). Average annual passenger automobile emissions per vehicle were based on dividing total 2002 pollutant emissions from light-duty gas vehicles by total number of passenger cars in 2002 (National Transportation Statistics, 2004).

Pollution removal Value is calculated based on the prices of \$1136 per ton (carbon monoxide), \$1260 per ton (ozone), \$226 per ton (nitrogen dioxide), \$110 per ton (sulfur dioxide), \$5840 per ton (particulate matter less than 10 microns and greater than 2.5 microns), \$17993 per ton (particulate matter less than 2.5 microns).

Trees remove PM2.5 when particulate matter is deposited on leaf surfaces. This deposited PM2.5 can be resuspended to the atmosphere or removed during rain events and dissolved or transferred to the soil. This combination of events can lead to interesting results depending on various atmospheric factors. Generally, pollution removal is positive with positive benefits. However, there are some cases when net removal is negative or resuspended particles lead to increased pollution concentrations and negative values. During some months (e.g., with no rain), trees resuspend more particles than they remove. Resuspension can also lead to increased overall PM2.5 concentrations if the boundary layer conditions are lower during net resuspension periods than during net removal periods. Since the pollution removal value is based on the change in pollution concentration, it is possible to have situations when trees remove PM2.5 but increase concentrations and thus have negative values during periods of positive overall removal. These events are not common, but can happen.

Structural value Value based on the physical resource itself (e.g., the cost of having to replace a tree with a similar tree).

Ton Short ton (U.S.) (2,000 lbs).

Appendix II. Comparison of Urban Forests

Sometimes it is useful to determine how a city compares to other areas. Although comparison among cities should be made with caution as there are many attributes of a city that affect urban forest structure and functions, summary data are provided from other cities analyzed using the i-Tree Eco model. This comparison information is provided by the i-eco model and reporting. Unfortunately additional comparison cities for the desert southwest are unavailable.

Table 12. Tree Benefits in Other Areas

Area	Number of trees	Carbon Storage (tons)	Carbon Sequestration (tons/year)	Pollution Removal (tons/year)
Calgary, Canada	11,889,000	445,000	21,422	326
Atlanta, GA	9,415,000	1,345,000	46,433	1,662
Toronto, Canada	7,542,000	992,000	40,345	1212
New York, NY	5,212,000	1,351,000	42,283	1,677
Phoenix, AZ	3,166,000	305,000	35,400	1770
Baltimore, MD	2,627,000	596,000	16,127	430
Philadelphia, PA	2,113,000	530,000	16,115	576
Washington, DC	1,928,000	523,000	16,148	418
Albuquerque, NM	1,504,000	226,000	9,710	366
El Paso, TX	1,281,000	92,800	7,430	318
Boston, MA	1,183,000	319,000	10,509	284
Woodbridge, NJ	986,000	160,000	5,561	210
Minneapolis, MN	979,000	250,000	8,895	305
Syracuse, NY	876,000	173,000	5,425	109
Morgantown, WV	661,000	94,000	2,940	66
Moorestown, NJ	583,000	117,000	3,758	118
Las Cruces, NM	257,000	17,800	1,580	92
Eastern Colorado	251,000	71,900	2,200	77
Jersey City, NJ	136,000	21,000	890	41
Freehold, NJ	48,000	20,000	545	21

Table 13. Per-Acre Values of Tree Effects in Other Areas

Area	Number of Trees	Carbon Storage (tons)	Carbon Sequestration (tons/year)
Morgantown, WV	119.7	17.0	0.53
Atlanta, GA	111.6	15.9	0.55
Calgary, Canada	66.7	2.5	0.12
Woodbridge, NJ	66.5	10.8	0.38
Moorestown, NJ	62.0	12.5	0.4
Syracuse, NY	54.5	10.8	0.34
Baltimore, MD	50.8	11.5	0.31
Washington, DC	49.0	13.3	0.41
Toronto, Canada	48.3	6.4	0.26
Freehold, NJ	38.5	16.0	0.44
Boston, MA	33.5	9.0	0.3
New York, NY	26.4	6.8	0.21
Minneapolis, MN	26.2	6.7	0.24
Philadelphia, PA	25.0	6.3	0.19
Albuquerque, NM	17.8	2.7	0.11
Jersey City, NJ	14.3	2.2	0.09
Phoenix, AZ	12.9	1.2	0.14
El Paso, TX	12.7	0.9	0.07
Eastern Colorado	12.1	3.5	0.11
Las Cruces, NM	9.1	0.6	0.06

Appendix III. General Recommendations for Air Quality Improvement

Urban vegetation can directly and indirectly affect local and regional air quality by altering the urban atmosphere environment. Four main ways that urban trees affect air quality are [Nowak, 1995]:

- Temperature reduction and other microclimate effects
- Removal of air pollutants
- Emission of volatile organic compounds (VOC) and tree maintenance emissions
- Energy effects on buildings

The cumulative and interactive effects of trees on climate, pollution removal, and VOC and power plant emissions determine the impact of trees on air pollution. Cumulative studies involving urban tree impacts on ozone have revealed that increased urban canopy cover, particularly with low VOC emitting species, leads to reduced ozone concentrations in cities [Nowak, 2000]. Local urban management decisions also can help improve air quality.

Table 14. Urban Forest Management Strategies to Improve Air Quality

Strategy	Result
Increase the number of healthy trees	Increase pollution removal
Sustain existing tree cover	Maintain pollution removal levels
Maximize use of low VOC-emitting trees	Reduces ozone and carbon monoxide formation
Sustain large, healthy trees	Large trees have greatest per-tree effects
Use long-lived trees	Reduce long-term pollutant emissions from planting and removal
Use low maintenance trees	Reduce pollutants emissions from maintenance activities
Reduce fossil fuel use in maintaining vegetation	Reduce pollutant emissions
Plant trees in energy conserving locations	Reduce pollutant emissions from power plants
Plant trees to shade parked cars	Reduce vehicular VOC emissions
Supply ample water to vegetation	Enhance pollution removal and temperature reduction
Plant trees in polluted or heavily populated areas	Maximizes tree air quality benefits
Avoid pollutant-sensitive species	Improve tree health
Utilize evergreen trees for particulate matter	Year-round removal of particles

Appendix IV. Species Distribution and Botanical Names

Table 15. Species Distribution and Botanical Names

Species	Species	Percent Population	Percent Leaf Area	Importance Value
Italian cypress	<i>Cupressus sempervirens</i>	25.8	2.6	28.5
Afghan pine	<i>Pinus eldarica</i>	10.8	18.1	28.9
Mexican fan palm	<i>Washingtonia robusta</i>	7.3	2.7	10.0
White mulberry	<i>Morus alba</i>	5.9	19.7	25.6
Pecan	<i>Carya illinoensis</i>	4.9	9.0	13.9
Velvet ash	<i>Fraxinus velutina</i>	4.7	12.3	17.0
Siberian elm	<i>Ulmus pumila</i>	3.3	2.8	6.0
Tree of heaven	<i>Ailanthus altissima</i>	2.6	0.7	3.3
Chinaberry	<i>Melia azedarach</i>	2.4	2.7	5.1
Juniper spp	<i>Juniperus</i>	2.1	3.4	5.5
Chitalpa	<i>Chitalpa tashkentensis</i>	2.1	0.5	2.6
Mimosa	<i>Albizia julibrissin</i>	2.1	0.2	2.3
Chinese pistache	<i>Pistacia chinensis</i>	1.3	0.5	1.7
Honeylocust	<i>Gleditsia triacanthos</i>	1.2	1.5	2.6
Honey mesquite	<i>Prosopis glandulosa</i>	1.2	2.4	3.6
Oriental arborvitae	<i>Platycladus orientalis</i>	1.1	0.7	1.8
Pear spp	<i>Pyrus</i>	1.1	0.5	1.6
Redbud spp	<i>Cercis</i>	1.1	0.6	1.6
Elderberry spp	<i>Sambucus</i>	1.1	0.3	1.4
Aleppo pine	<i>Pinus halepensis</i>	1.0	5.9	6.9
Red cedar spp	<i>Thuja</i>	1.0	1.5	2.4
Yucca spp	<i>Yucca</i>	1.0	0.4	1.4
Jerusalem thorn	<i>Parkinsonia aculeata</i>	0.9	0.1	1.0
Desertwillow	<i>Chilopsis linearis</i>	0.8	0.5	1.4
Sweet acacia	<i>Acacia farnesiana</i>	0.7	1.4	2.1
Chaste tree	<i>Vitex agnus-castus</i>	0.7	1.3	2.0
Western redcedar	<i>Thuja plicata</i>	0.7	0.7	1.4
Canary island date palm	<i>Phoenix canariensis</i>	0.7	0.6	1.3
Privet spp	<i>Ligustrum</i>	0.7	0.1	0.8
Eucalyptus	<i>Eucalyptus urophylla</i>	0.7	0.1	0.8
Soaptree yucca	<i>Yucca elata</i>	0.7	0.1	0.8
Live oak	<i>Quercus/live virginiana</i>	0.7	0.0	0.8

Species	Species	Percent Population	Percent Leaf Area	Importance Value
Hardwood	<i>Hardwood</i>	0.7	0.0	0.7
Whitethorn acacia	<i>Acacia constricta</i>	0.5	0.2	0.6
Hollywood juniper	<i>Juniperus chinensis 'Torulosa'</i>	0.4	1.9	2.3
Goodding's willow	<i>Salix gooddingii</i>	0.4	1.9	2.2
Cottonwood spp	<i>Populus</i>	0.4	1.0	1.4
Peach	<i>Prunus persica</i>	0.4	0.4	0.8
Black locust	<i>Robinia pseudoacacia</i>	0.4	0.2	0.6
Callery pear	<i>Pyrus calleryana</i>	0.4	0.2	0.5
Torrey yucca	<i>Yucca torreyi</i>	0.4	0.1	0.5
Soapberry spp	<i>Sapindus</i>	0.4	0.1	0.5
Mediterranean fan palm	<i>Chamaerops humilis</i>	0.4	0.1	0.4
Desert museum palo verde	<i>Parkinsonia hybrid Desert Museum</i>	0.4	0.1	0.4
Eve's needle	<i>Yucca faxoniana</i>	0.4	0.1	0.4
Common crapemyrtle	<i>Lagerstroemia indica</i>	0.4	0.1	0.4
Apple spp	<i>Malus</i>	0.4	0.0	0.4
Oak spp	<i>Quercus</i>	0.4	0.0	0.4
Neomexican elderberry	<i>Sambucus caerulea v mexicana</i>	0.4	0.0	0.4
Oleander	<i>Nerium oleander</i>	0.4	0.0	0.4
Screwbean mesquite	<i>Prosopis pubescens</i>	0.2	0.1	0.3

References

- Abdollahi, K.K.; Z.H. Ning; and A. Appeaning (eds). 2000. Global climate change and the urban forest. Baton Rouge, LA: GCRCC and Franklin Press. 77p.
- Abraxas energy consulting, <http://www.abraxasenergy.com/emissions/>
- Baldocchi, D. 1988. A multi-layer model for estimating sulfur dioxide deposition to a deciduous oak forest canopy. *Atmospheric Environment*. 22: 869-884.
- Baldocchi, D.D.; Hicks, B.B.; Camara, P. 1987. A canopy stomatal resistance model for gaseous deposition to vegetated surfaces. *Atmospheric Environment*. 21: 91-101.
- Bidwell, R.G.S.; Fraser, D.E. 1972. Carbon monoxide uptake and metabolism by leaves. *Canadian Journal of Botany*. 50: 1435-1439.
- Broecker, W.S. 1970. Man's oxygen reserve. *Science* 168: 1537-1538.
- Ciesla, William M. 2001. *Tomicus piniperda*. Pine Shoot Beetle. North American Forest Commission. Exotic Forest Pest Information System for North America (EXFOR). Can be accessed through: <http://spfnic.fs.fed.us/exfor/data/pestreports.cfm?pestidval=86&langdisplay=english>
- Clark, J.R., Matheny NP, Cross G, Wake V. 1997. A model of urban forest sustainability. *J Arbor* 23(1):17-30.
- Davidson, K., A. Hallberg, D. McCubbin, and B. Hubbell. (2007). Analysis of PM2.5 Using the Environmental Benefits Mapping and Analysis Program (BenMAP). *Journal of Toxicology and Environmental Health, Part A* 70(3): 332-346.
- Energy Information Administration, 2003, Emissions of Greenhouse Gases in the United States 2003. <http://www.eia.doe.gov/oiaf/1605/ggrpt/>
- Energy Information Administration. 1994 Energy Use and Carbon Emissions: Non-OECD Countries DOE/EIA-0579.
- Energy Information Administration. Total Energy Consumption in U.S. Households by Type of Housing Unit, 2001 <http://www.eia.doe.gov/emeu/recs/contents.html>.
- Graham, R.L., Wright, L.L., and Turhollow, A.F. 1992. The potential for short-rotation woody crops to reduce U.S. CO2 Emissions. *Climatic Change* 22:223-238.
- Haugen, Dennis A.; Hoebeke, Richard E. 2005. Sirex woodwasp - *Sirex noctilio* F. (Hymenoptera: Siricidae). Pest Alert. NA-PR-07-05. United States Department of Agriculture, Forest Service, Northern Area State and Private Forestry. Can be accessed through: http://na.fs.fed.us/spfo/pubs/pest_al/sirex_woodwasp/sirex_woodwasp.htm
- Heating with Wood I. Species characteristics and volumes. <http://ianrpubs.unl.edu/forestry/g881.htm>
- Hirabayashi, S. 2011. Urban Forest Effects-Dry Deposition (UFORE-D) Model Enhancements, [http://www.itreetools.org/eco/resources/UFORE-D enhancements.pdf](http://www.itreetools.org/eco/resources/UFORE-D%20enhancements.pdf)
- Hirabayashi, S., C. Kroll, and D. Nowak. 2011. Component-based development and sensitivity analyses of an air pollutant dry deposition model. *Environmental Modeling and Software* 26(6): 804-816.
- Hirabayashi, S., C. Kroll, and D. Nowak. 2012. i-Tree Eco Dry Deposition Model Descriptions V 1.0
- Houck, J.E. Tiegs, P.E, McCrillis, R.C. Keithley, C. and Crouch, J. 1998. Air emissions from residential heating: the wood heating option put into environmental perspective. In: *Proceedings of U.S. EPA and Air Waste Management Association Conference: Living in a Global Environment, V.1:* 373-384.
- Insect/disease proximity to study area was completed using the U.S. Forest Service's Forest Health Technology Enterprise Team (FHTET) database. Data includes distribution of pest by county FIPs code for 2004-2009. FHTET range maps are available at www.foresthealth.info for 2006-2010.

- Kliejunas, John. 2005. *Phytophthora ramorum*. North American Forest Commission. Exotic Forest Pest Information System for North America (EXFOR). Can be accessed through:
<http://spfnic.fs.fed.us/exfor/data/pestreports.cfm?pestidval=62&langdisplay=english>
- Layton, M. 2004. 2005 Electricity Environmental Performance Report: Electricity Generation and Air Emissions. California Energy Commission. http://www.energy.ca.gov/2005_energypolicy/documents/2004-11-15_workshop/2004-11-15_03-A_LAYTON.PDF
- Lovett, G.M. 1994. Atmospheric deposition of nutrients and pollutants in North America: an ecological perspective. *Ecological Applications*. 4: 629-650.
- McPherson, E.G. and J. R. Simpson 1999. Carbon dioxide reduction through urban forestry: guidelines for professional and volunteer tree planters. Gen. Tech. Rep. PSW-171. Albany, CA: U.S. Department of Agriculture, Forest Service, Pacific Southwest Research station 237 p.
http://wcufr.ucdavis.edu/products/cufr_43.pdf
- Murray, F.J.; Marsh L.; Bradford, P.A. 1994. New York State Energy Plan, vol. II: issue reports. Albany, NY: New York State Energy Office.
- National Emission Trends <http://www.epa.gov/ttn/chief/trends/index.html>
- National Transportation Statistics http://www.bts.gov/publications/national_transportation_statistics/2004/.
- Northeastern Area State and Private Forestry. 1998. HOW to identify and manage Dutch Elm Disease. NA-PR-07-98. Newtown Square, PA: U.S. Department of Agriculture, Forest Service, Northeastern Area State and Private Forestry. http://www.na.fs.fed.us/spfo/pubs/howtos/ht_ded/ht_ded.htm
- Northeastern Area State and Private Forestry. 2005. Asian Longhorned Beetle. Newtown Square, PA: U.S. Department of Agriculture, Northeastern Area State and Private Forestry.
<http://www.na.fs.fed.us/spfo/alb/>
- Northeastern Area State and Private Forestry. 2005. Forest health protection emerald ash borer home. Newtown Square, PA: U.S. Department of Agriculture, Forest Service, Northeastern Area State and Private Forestry. <http://www.na.fs.fed.us/spfo/eab/index.html>
- Nowak, D.J. 1994. Atmospheric carbon dioxide reduction by Chicago's urban forest. In: McPherson, E.G.; Nowak, D.J.; Rowntree, R.A., eds. Chicago's urban forest ecosystem: results of the Chicago Urban Forest Climate Project. Gen. Tech. Rep. NE-186. Radnor, PA: U.S. Department of Agriculture, Forest Service, Northeastern Forest Experiment Station: 83-94.
- Nowak, D.J. 1995. Trees pollute? A "TREE" explains it all. In: Proceedings of the 7th National Urban Forestry Conference. Washington, DC: American Forests. Pp. 28-30
- Nowak, D.J. 2000. The interactions between urban forests and global climate change. In: Abdollahi, K.K., Z.H. Ning, and A. Appeaning (Eds). *Global Climate Change and the Urban Forest*. Baton Rouge: GCRCC and Franklin Press. Pp. 31-44.
- Nowak, D.J. and J.F. Dwyer. 2007. Understanding the benefits and costs of urban forest ecosystems. In: Kuser, J. (ed.) *Urban and Community Forestry in the Northeast*. New York: Springer. Pp. 25-46.
- Nowak, D.J., and D.E. Crane. 2000. The Urban Forest Effects (UFORE) Model: quantifying urban forest structure and functions. In: Hansen, M. and T. Burk (Eds.) *Integrated Tools for Natural Resources Inventories in the 21st Century*. Proc. Of the IUFRO Conference. USDA Forest Service General Technical Report NC-212. North Central Research Station, St. Paul, MN. pp. 714-720. See also <http://www.ufore.org>.
- Nowak, D.J., R.E. Hoehn, D.E. Crane, J.C. Stevens, J.T. Walton, and J. Bond. 2008. A ground-based method of assessing urban forest structure and ecosystem services. *Arboriculture and Urban Forestry* 34(6): 347-358.
- Nowak, D.J.; Crane, D.E.; Dwyer, J.F. 2002. Compensatory value of urban trees in the United States. *Journal of Arboriculture*. 28(4): 194 - 199.

- Nowak, D.J.; Crane, D.E.; Stevens, J.C.; Hoehn, R.E. 2005. The urban forest effects (UFORE) model: field data collection manual. V1b. Newtown Square, PA: U.S. Department of Agriculture, Forest Service, Northeastern Research Station, 34 p.
http://www.fs.fed.us/ne/syracuse/Tools/downloads/UFORE_Manual.pdf
- Nowak, D.J.; Crane, D.E.; Stevens, J.C.; Ibarra, M. 2002. Brooklyn's Urban Forest. Gen. Tech. Rep. NE-290. Newtown Square, PA: U.S. Department of Agriculture, Forest Service, Northeastern Research Station. 107 p. Council of Tree and Landscape Appraisers guidelines. For more information, see Nowak, D.J., D.E. Crane, and J.F. Dwyer. 2002. Compensatory value of urban trees in the United States. *Journal of Arboriculture*. 28(4): 194-199.
- Nowak, David J., Hoehn, R., and Crane, D. 2007. Oxygen production by urban trees in the United States. *Arboriculture & Urban Forestry* 33(3):220-226.
- Residential Wood Burning Emissions in British Columbia, 2005.
http://www.env.bc.ca/air/airquality/pdfs/wood_emissions.pdf.
- Rexrode, Charles O.; Brown, H. Daniel 1983. Oak Wilt. Forest Insect & Disease Leaflet 29. United States Department of Agriculture, Forest Service. 6 p. Can be accessed through:
<http://www.fs.fed.us/r6/nr/fid/fidls/fidl-29.pdf>
- Society of American Foresters. Gold Spotted Oak Borer Hitches Ride in Firewood, Kills California Oaks. *Forestry Source*. October 2011 Vol. 16, No.10.
- U.S. Department of Agriculture. National Invasive Species Information Center. 2011.
<http://www.invasivespeciesinfo.gov/plants/main.shtml>
- U.S. Environmental Protection Agency. U.S. Power Plant Emissions Total by Year
www.epa.gov/cleanenergy/egrid/samples.htm.
- U.S. Forest Service. Tree Guides. http://www.fs.fed.us/psw/programs/uesd/uep/tree_guides.php
- United States Census Bureau. United States Population, 2003. Accessed 2013.
- Watershed Protection Development Review. City of Austin. Central Texas Invasive Plants. Volunteer Field Guide. <<http://www.ci.austin.tx.us/growgreen/downloads/invasiveplants.pdf>>
- Weng, Q., Lu, D., Schubring, J., Estimation of Land Surface Temperature-Vegetation Abundance Relationship for Urban Heat Island Studies, *Remote Sensing of Environment*. Elsevier. November, 2003.
- Zinke, P.J. 1967. Forest interception studies in the United States. In: Sopper, W.E.; Lull, H.W., eds. *Forest Hydrology*. Oxford, UK: Pergamon Press: 137-161.