Tree Canopy Change Detection in Phoenix, Arizona (2014-2020) Cameron T. Dougal, Mary K. Wright, and David M. Hondula

Introduction

The City of Phoenix, Arizona, United States, is one of many cities across the globe that have made major investments in increasing tree canopy for pedestrian thermal comfort, urban heat reduction, and aesthetic rejuvenation of the City.

To understand the return on investment of past and future tree planting across the City and identify areas where tree canopy has changed, it is vital to have a longitudinal assessment of tree canopy coverage at a high spatial resolution using methods that are comparable across time scales.

To this end, we applied identical methodologies using lidar and highresolution imagery to compare tree canopy change in Phoenix between 2014 and 2020. After significant investments in tree planting that are currently ongoing in the City of Phoenix, the process we used will allow for direct comparison of tree canopy across time.

Step 1. Data processing

- coverage
- (DEM)



Step 2. Data filtering

*These values were selected to limit to plausible tree heights and NDVI values low enough to detect desert trees.

 ey takeaways The citywide tree canopy in both 2014 and 2020 is about 9%. The longitudinal analysis shows that between 2014 and 2020: → The City of Phoenix tree canopy slightly decreased by 0.25%. → Roughly the same amount of tree canopy lost was also gained (3.23% gained vs. 3.49% lost). 		 Key takeaways → Tree canopy coverage varies across City Council Districts and by parcel type (residential vs. commercial). → The largest increase in canopy overall, and in both residential and commer parcels, occurred in District 7 (0.9% overall, 1.2% residential, 0.4% commercial, residential parcels have the highest tree canopy, however, resident parcels saw a larger decrease between 2014 and 2020 (13.7% vs. 12.88% than commercial parcels. Table 2. Tree canopy percentage by City of Phoenix Council District a parcel type 									
District	2014	2020	Change	2014	2020	Change	2014	2020	Change		
1	10.06	9.45	-0.62	12.35	11.20	-1.16	9.21	8.71	-0.50		
litywide canopy	%	2	10.39	9.78	-0.60	13.95	12.80	-1.15	14.48	14.10	-0.38
ree canopy in 2014	9.14	3	11.02	10.39	-0.63	14.43	13.60	-0.83	11.86	11.50	-0.35
ree canopy in 2020	8 89	4	9.92	9.74	-0.18	14.64	14.79	0.15	8.06	7.73	-0.32
	2.22	5	9.64	9.71	0.07	12.30	12.59	0.29	10.88	10.82	-0.05
ree canopy gained	3.23	6	9.26	8.77	-0.49	15.48	14.00	-1.48	17.44	16.09	-1.36
ree canopy lost	3.49	7	6.15	7.05	0.90	8.76	9.95	1.19	7.32	7.75	0.43
ree canopy unchanged	5.65	8	6.40	6.42	0.03	10.63	10.05	-0.58	7.53	7.59	0.06
lo tree canopy in either year	87.62		9.14	8.89	-0.26	13.70	12.88	-0.81 e. c.	10.11	9.88	-0.23

NO LIEE CANOPY IN EILINEI YEAR



Generate Canopy Height Model (CHM) from lidar for all of Phoenix with lidar

Generate Digital Surface Model (DSM) Generate Digital Elevation Model

CHM = DSM - DEM

Calculate NDVI from NAIP aerial imagery

Classify raster pixels as trees if: Height ≥ 1.5 m Height ≤ 40 m NDVI ≥ 0.08

and

*Note that lidar coverage is incomplete in Districts 1 & 2











Methods

Canopy Height Model



Step 3. Longitudinal analysis

- Compare 2014 and 2020
- Tree canopy gained
- Tree canopy lost
- Tree canopy unchanged
- No tree canopy



Data descriptions Geological Survey.

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Next Steps

- Conduct a formal accuracy assessment
 - optimize NDVI cutoff for accuracy
- Tree top detection
- Calculate tree heights
- Tree crown segmentation
- Calculate crown widths

Acknowledgements

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R package version 4.0.3, https://cran.r-project.org/package=lidR. Analysis. R package version 1.7-29, https://cran.r-project.org/package=terra.



Lidar data were obtained for 2014 and 2020 from the United States

The Normalized-Difference Vegetation Index (NDVI) was derived from the United States Department of Agriculture (USDA) National Agriculture Imagery Program (NAIP) for 2015 and 2019.

All data were processed in the R software program ver. 4.2.2. Lidar data were processed into raster heigh models using the 'LAScatalog' processing engine of the 'lidR' package (Roussel 2023)—this engine allows for resolution of edge effects across lidar tiles. Raster data were processed using the terra package (Hijmans et al. 2023).



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- Roussel J, Auty D (2023). Airborne LiDAR Data Manipulation and Visualization for Forestry Applications.
- Hijmans, Robert J., Roger Bivand, Edzer Pebesma, and Michael D. Sumner (2023). terra: Spatial Data