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# LiDAR-Derived Aboveground Biomass and Uncertainty for California Forests, 2005-2014

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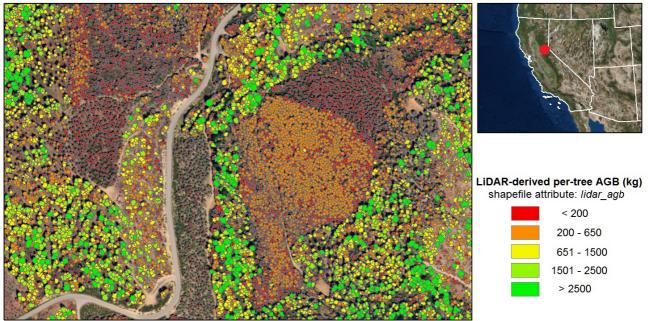
Documentation Revision Date: 2018-08-14

Data Set Version: 1

#### Summary

This dataset provides estimates of aboveground biomass and spatially explicit uncertainty from 53 airborne LiDAR surveys of locations throughout California between 2005 and 2014. Aboveground biomass was estimated by performing individual tree crown detection and applying a customized "remote sensing aware" allometric equation to these individual trees. Aboveground biomass estimates and their uncertainties for each study area are provided in per-tree and gridded format. The canopy height models used for the tree detection and biomass estimation are also provided.

There are 9,504 files in compressed shapefile format (\*.shp contained in \*.zip) and 212 files in GeoTIFF format (\*.tif) included in this dataset.



Blacksmith2012 389389N 12048484W.shp

Figure 1. Per-tree aboveground biomass (kg) derived from LiDAR-measured tree heights for the Blacksmith site in 2012 (from Blacksmith2012\_389389N\_12048484W.zip).

## Citation

Xu, Q., A. Man, M.M. Fredrickson, Z. Hou, J. Pitkanen, B. Wing, C. Ramirez, B. Li, and J. Greenberg. 2018. LiDAR-Derived Aboveground Biomass and Uncertainty for California Forests, 2005-2014. ORNL DAAC, Oak Ridge, Tennessee, USA. https://doi.org/10.3334/ORNLDAAC/1537

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## 1. Data Set Overview

#### Project: Carbon Monitoring System

The NASA Carbon Monitoring System (CMS) is designed to make significant contributions in characterizing, quantifying, understanding, and predicting the evolution of global carbon sources and sinks through improved monitoring of carbon stocks and fluxes. The System will use the full range of NASA satellite observations and modeling/analysis capabilities to establish the accuracy, quantitative uncertainties, and utility of products for supporting national and international policy, regulatory, and management activities. CMS will maintain a global emphasis while providing finer scale regional information, utilizing space-based and surface-based data and will rapidly initiate generation and distribution of products both for user evaluation and to inform near-term policy development and planning.

#### **Related Publication:**

Xu, Q., Man, A., Fredrickson, M., Hou, Z., Pitkänen, J., Wing, B., Ramirez, C., Li, B. & Greenberg, A. J. 2018. Quantification of uncertainty in aboveground biomass estimates derived from small-footprint airborne LiDAR. *Remote Sensing of Environment, 216*, 514-528. .https://doi.org/10.1016/j.rse.2018.07.022

#### Acknowledgements:

The study is funded by NASA Carbon Monitoring System (Proposal number: 14-CMS14-0048; Grant number: NNX14AO80G) under the title of "Reducing uncertainties in estimating California's forest carbon stocks".

## 2. Data Characteristics

Spatial Coverage: Surveyed areas throughout California, see Figure 1.

**Spatial Resolution:** 

Shapfiles: one polygon feature per tree canopy

GeoTIFFs: 30-m (AGB, uncertainty, coefficient of variation) and 25-cm (CHM)

Temporal Coverage: 20050101 - 20141231

Temporal Resolution: Each study site was surveyed once.

Study Area (These coordinates are the approximate locations of the study sites and may not match the extents of the data files. All latitude and longitude given in decimal degrees.)

Site	Westernmost Longitude	Easternmost Longitude	Northernmost Latitude	Southernmost Latitude
Survey areas in California	-123.746	-117.739	41.983	34.051

#### **Data File Information**

There are 9,504 \*.zip files that contain compressed shapefiles (\*.shp) and 212 files in GeoTIFF format (\*.tif) included in this dataset.

- Shapefiles give the tree-level AGB estimates and uncertainties for a survey area. Spatial reference properties for the data files are listed below.
   GeoTIFFs give <u>plot-level aboveground biomass (AGB) estimates</u> and uncertainties for each of the 53 LiDAR survey areas included in the dataset
- as well as a canopy height model.

#### Spatial Reference Properties

Geographic Coordinate System: GCS\_North\_American\_1983 Angular Unit: Degree (0.0174532925199433) Prime Meridian: Greenwich (0.0) Datum: D\_North\_American\_1983 Spheroid: GRS\_1980 Semimajor Axis: 6378137.0 Semiminor Axis: 6356752.314140356 Inverse Flattening: 298.257222101

NAD\_1983\_UTM\_zone\_10N WKID: 26910 Authority: EPSG

Projection: Transverse\_Mercator false\_easting: 500000.0 false\_northing: 0.0 central\_meridian: -123.0 scale\_factor: 0.9996 latitude\_of\_origin: 0.0 Linear Unit: Meter (1.0)

NAD\_1983\_UTM\_Zone\_11N WKID: 26911 Authority: EPSG Projection: Transverse\_Mercator false\_easting: 500000.0 false\_northing: 0.0 central\_meridian: -117.0 scale\_factor: 0.9996 latitude\_of\_origin: 0.0 Linear Unit: Meter (1.0)

#### **Tree-level Metrics**

A number of shapefiles with coverages ranging between 1- and 2-km<sup>2</sup> are provided for each of the 53 survey areas. Each shapefile is provided in a zip (\*.*zip*) containing four files (\*.*shp*, \*.*shx*, \*.*dbf*, \*.*prf*).

- Shapefiles are named according to the survey area, year, and upper-left coordinate of the bounding box encompassing the shapefile features, e.g.: <survey area><year>\_#####W.zip.
- The shapefiles contain one feature per tree delineated from the canopy height model. The tree segmentation method is described in Section 5. Spatial data properties for the shapefiles are given below.
- Shapefile feature attributes, units, and descriptions are provided in Table 1.

#### Spatial Data Properties

Spatial Representation Type: Vector Geometry Type: Polygon Geographic Coordinate System: North\_American\_1983 Prime Meridian: Greenwich Angular Unit: Degree

Table 1. Shapefile feature attributes, units, and descriptions.

Attribute	Units	Description	
DN		segment ID	
x_max m		x coordinate of local maximum	
y_max	m	y coordinate of local maximum	
max_h	m	z coordinate of local maximum	
size_mx		number of local maxima	
x_cen	m	x coordinate of segment geometric center	
y_cen	m	y coordinate of segment geometric center	
area	m2	segment area	
diam_mx	m	max diameter of the segment	
diam_prp	m	diameter perpendicular to the maximum diameter	
diam_dr		direction of the max diameter	
mask_rgr		mask region	
D_N_L	m	distance to the nearest segment	
D_mean_L	m	mean distance to all segments in a 30-m pixel cell	
D_25p_L	m	25% percentile of the distances to all segments in a 30-m pixel cell	

D_50p_L Attribute	m Units	50% percentile of the distances to all segments in a 30-m pixel cell <b>Description</b>
-N_T_L		number of segments in a 5-m neighborhood
ellip_L	m2	ellipse area using the two diameters
lidar_agb	kg	LiDAR-derived per-tree AGB
GAE_sd	kg	total standard deviation of the LiDAR-derived AGB
GAE_CV		coefficient of variation of the LiDAR-derived AGB
lidar_comp	kg	standard deviation of the LiDAR-derived AGB associated with LiDAR height measurement
param_comp	kg	standard deviation of the LiDAR-derived AGB associated with model parameters
resid_comp	kg	standard deviation of the LiDAR-derived AGB associated with model residuals

#### **Plot-level Metrics**

Four GeoTIFFs are provided for each of the 53 survey areas.

- Files are named by survey area and year followed by the measurement type.
- Spatial data properties for the GeoTIFFs are given below.
- Example filenames, units, and their descriptions are listed in Table 2.

#### Spatial Data Properties

Spatial Representation Type: Raster Pixel Depth: 64bit Pixel Type: 30-m (AGB) and 25-cm (CHMs) Compression Type: LZW Number of Bands: 1 Raster Format: GeoTIFF No Data Value: -9999 Scale Factor: 1

Table 2. Filenames, units, and descriptions for GeoTIFFs.

Filename		Description
<survey area=""><year>_agb.tif</year></survey>		Aboveground biomass density
<survey area=""><year>_sd_total.tif</year></survey>		Total of allometric and remote sensing associated uncertainties
<survey area=""><year>_coefficient_variance.tif</year></survey>		AGB coefficient of variation
<survey area=""><year>_chm.tif</year></survey>		Canopy height model

## 3. Application and Derivation

This dataset is among the first examples of AGB estimation and uncertainty analysis based on an individual tree detection method. Aboveground biomass is an important metric for assessing carbon sequestration in forests.

## 4. Quality Assessment

Estimates of uncertainty associated with the plot- and tree-level AGB take into account potential error introduced by the allometry, the remote sensing height measurements and the individual tree detection algorithms.

## 5. Data Acquisition, Materials, and Methods

Estimates of aboveground biomass (AGB) and their uncertainties were established at the plot- and tree-level. A "remote sensing aware" allometric equation relating LiDAR-derived height to AGB was used to estimate AGB and uncertainty at the plot-level. LiDAR-derived height estimates were used to determine and propagate tree-level estimates of AGB.

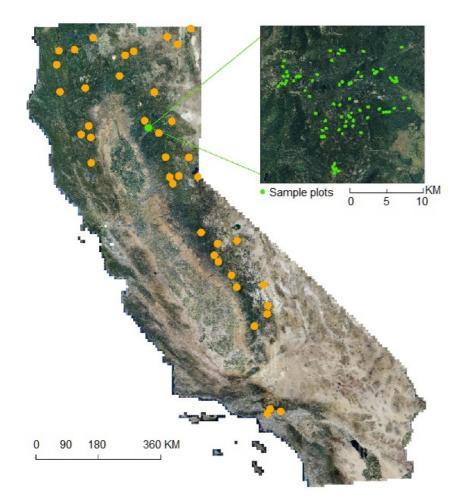


Figure 2. Site locations included in this dataset. The inset map depicts sample plots in the Lassen National Forest. (from Xu et al. 2017)

#### Tree segmentation

Tree heights were measured using the individual tree crown delineation approach outlined in the flow chart in Figure 2. A local maxima-incorporated watershed segmentation algorithm (Gauch 1999; Pitkänen et al. 2004) was applied to the low-pass filtered CHM to identify each individual tree by delineating the crown boundary. Tree height was extracted from the maximum height value around each local maximum. Crown diameters were measured by taking the maximum diameter of each segmented local maximum. Other tree-level attributes were calculated as the distance to the nearest tree and mean distance to all trees in a sample plot, to account for the local stem density.

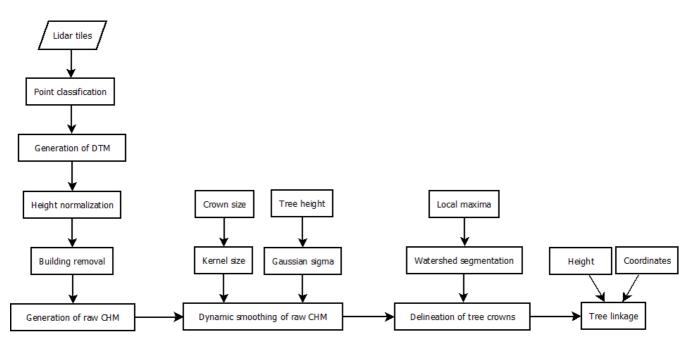


Figure 3. Flowchart of the standardized preprocessing chain and individual tree detection.

The "remote sensing aware" allometric equation relating per-tree heights to AGB in a species-agnostic fashion was developed using a generalized allometric equation approach calibrated using trees in the Forest Inventory Analysis (FIA) database. This approach applies published allometric equations compiled as part of the GlobAllomeTree database (GlobAllomeTree 2013) to estimate a distribution of AGB for trees in the FIA. The result was a database of AGB estimates that was used to generate a novel allometric equation for relating per-tree heights to AGB (Eq. 1; Xu et al. 2017).

Estimates of uncertainty associated with the tree- and plot-level AGB take into account potential error introduced by the allometry and remote sensing height measurements. Detailed tree information for 8313 trees in 146 circular fixed-area sample plots in Lassen National Forest (USDA Forest Service; PNW-FIA Field Manual, 2015) was used to determine the LiDAR height measurement error, omission errors, and commission errors. The equations for deriving per-tree uncertainty (Eqs. 2-5), can be found in Xu et al. (2017).

#### Plot-level aboveground biomass

Plot-level biomass (Mg/ha) was obtained using a new method for predicting biomass density that takes omission and commission errors as a correction to the LiDAR direct detection of AGB. Plot-level AGB and variance are calculated using Equations 6 and 7 from Xu et al. (2017), respectively.

### 6. Data Access

These data are available through the Oak Ridge National Laboratory (ORNL) Distributed Active Archive Center (DAAC).

LiDAR-Derived Aboveground Biomass and Uncertainty for California Forests, 2005-2014

Contact for Data Center Access Information:

- E-mail: uso@daac.ornl.gov
- Telephone: +1 (865) 241-3952

## 7. References

Gauch, J.M. 1999. Image segmentation and analysis via multiscale gradient watershed hierarchies. *IEEE Trans. Image Process.* 8 (1):69–79. doi: https://doi.org/10.1109/83.736688

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