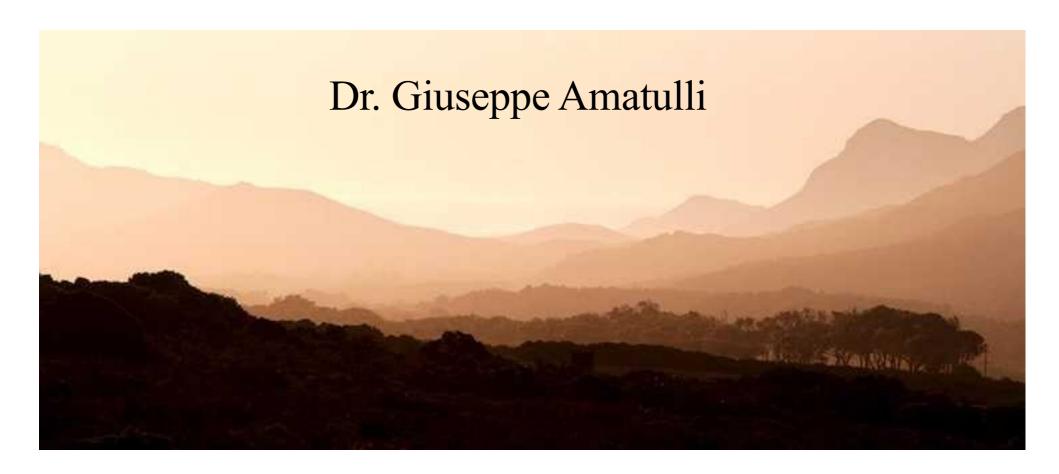
GIS and RS for Biodiversity Biodiversity $\sim f$ (environment)



Objectives Biodiversity $\sim f(\text{environment})$

A whirlwind tour of environmental data:

- Climate (past & future)
- Topography
- *Ecosystem Parameters
- Land Cover



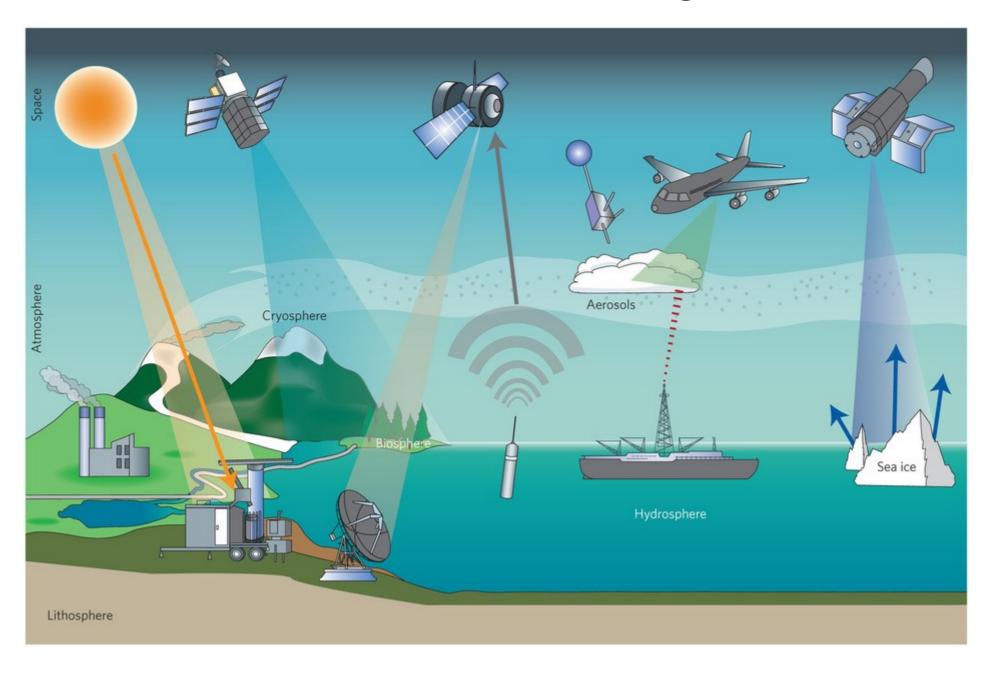




Key take-home messages:

- 1. Data deluge → incredible opportunities
- 2.Data complexity → biogeographers must be vigilant

Remote sensing



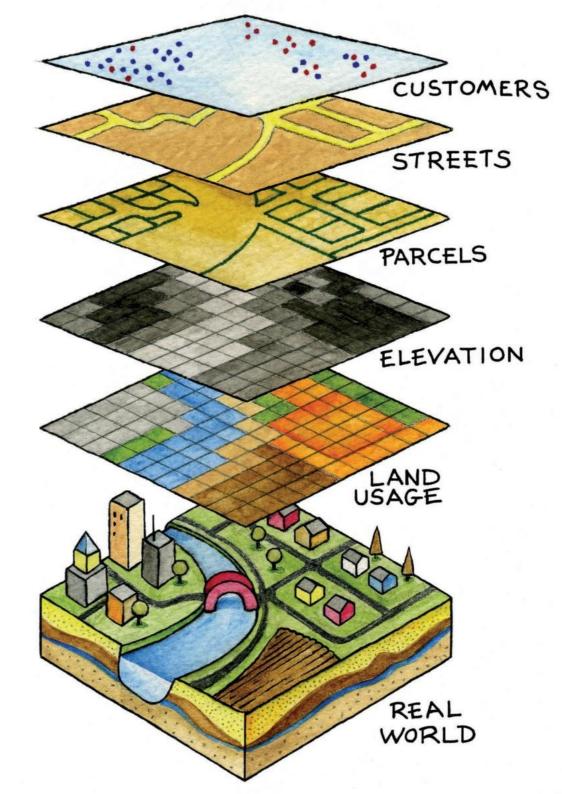
Geographic Information System

Combine:

- satellite information,
- ground true
- gps device data,
- station data



Modeling



Climate: Data Sources

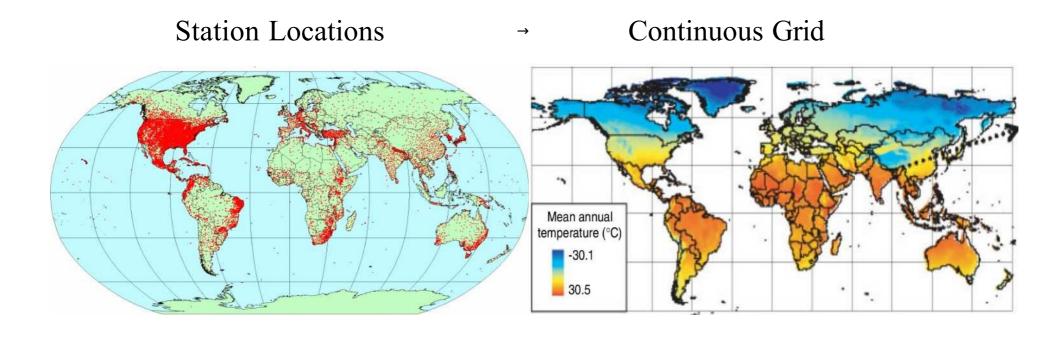
Past

- Pre-historical (<1850): proxies (ice, pollen, trees, etc.)
- Station data (≈1850-Present): National archives, NCDC Global Historical Climate Network
- Gridded Station data (climate or weather): NCDC GHCN, CRU Data, PRISM, WorldClim
- Historical Climate Model data: CMIP5: Last Glacial Maximum 2012
- Sub-daily Gridded 'reanalysis' model data:
 NCEP & NARR & ERA40 reanalysis

Future

- Gridded model output (hourly/daily/monthly):
 Global / Regional Climate Model (GCM/RCM)
- Station data: Downscaled to stations

WorldClim https://worldclim.org Chelsa https://chelsa-climate.org

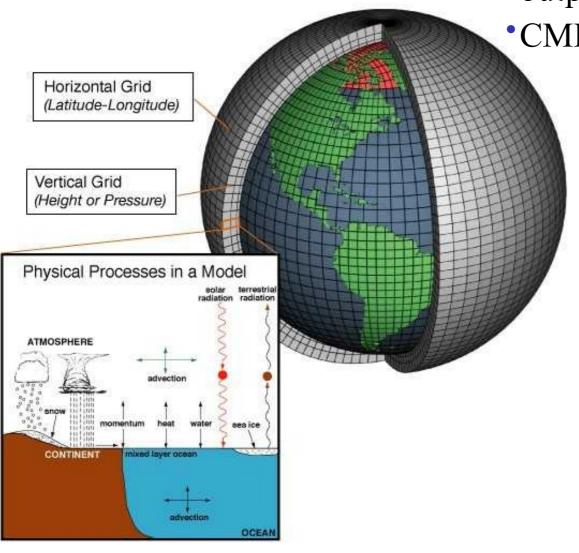


- •Global ≈1km resolution monthly T_{max} , T_{min} , and Precipitation.
- *Thin-plate splines (latitude, longitude, and elevation).
- No pixel-by-pixel metrics of uncertainty

Climate Models

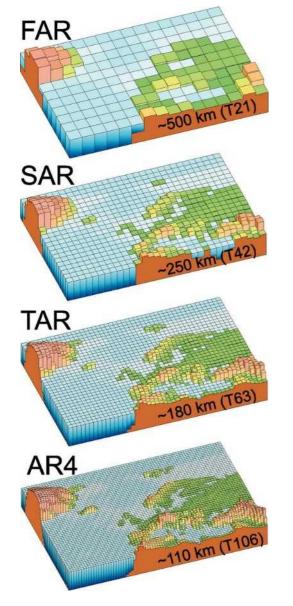
•IPCC/CMIP organize most model output

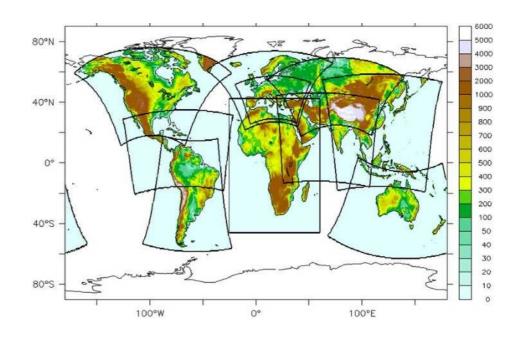
CMIP5 released last year



Climate Models

GCM resolution increasing: CMIP5 atmosphereland time-slices → 25km





Regional Climate Models (AR5 CORDEX): ~50km

Topography complexity

(geomorphometry/geomorphology analysis)







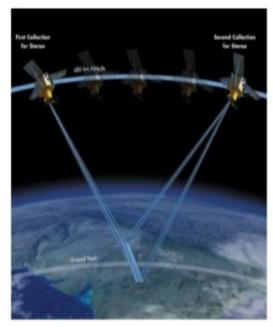


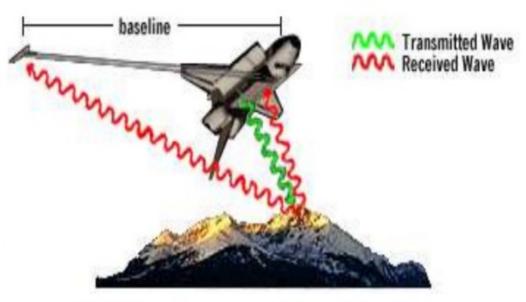




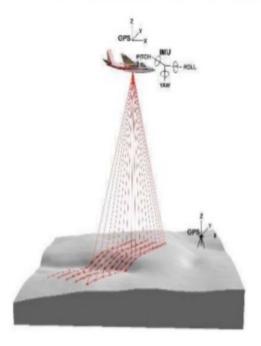
Topographic variation underpins a myriad of patterns and processes in hydrology, climatology, geography and ecology, and is key to understanding the variation of life on the planet (Amatulli et al. 2018).

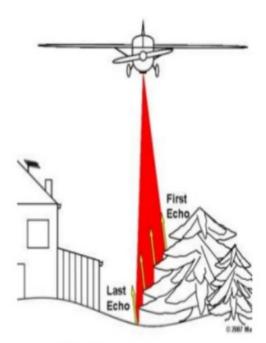
How is a DEM generated?





RADAR Interferometry with 2 receivers, 60m apart





A) From airborne or satellite remote sensing:
Stereo images, RADAR or LIDAR data are used in a semi-automated process to create a DEM.

Airborne Laserscanning or LIDAR

Multi-Error-Removed Improved-Terrain DEM

http://hydro.iis.u-tokyo.ac.jp/~yamadai/MERIT_DEM/





MERIT DEM

Geophysical Research Letters

RESEARCH LETTER

10.1002/2017GL072874

Key Points:

- A high-accuracy global digital elevation model (DEM) was developed by removing multiple height error components from existing DEMs
- Landscape representation was improved, especially in flat regions where height error magnitude was larger than actual topography variation

A high-accuracy map of global terrain elevations

Dai Yamazaki^{1,2} , Daiki Ikeshima³, Ryunosuke Tawatari³, Tomohiro Yamaguchi⁴, Fiachra O'Loughlin⁵, Jeffery C. Neal⁶, Christopher C. Sampson⁷, Shinjiro Kanae³, and Paul D. Bates⁶

¹Department of Integrated Climate Change Projection Research, Japan Agency for Marine-Earth Science and Technology, Yokohama, Japan, ²Institute of Industrial Sciences, University of Tokyo, Tokyo, Japan, ³Department of Civil and Environmental Engineering, Tokyo Institute of Technology, Tokyo, Japan, ⁴Institute of Space and Astronautical Science, Japan Aerospace Exploration Agency, Sagamihara, Japan, ⁵UCD School of Civil Engineering, University College Dublin, Dublin, Ireland, ⁶School of Geographical Sciences, University of Bristol, Bristol, UK, ⁷SSBN, Bristol, UK



MERIT Hydro

Water Resources Research

RESEARCH ARTICLE

10.1029/2019WR024873

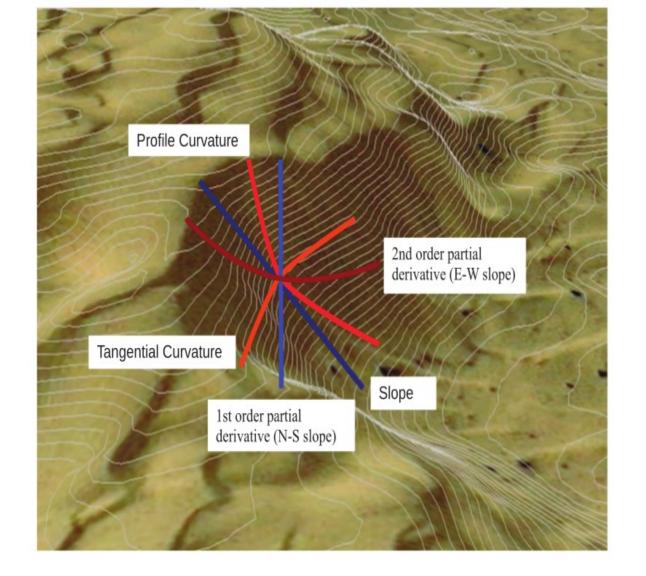
Key Points:

- A global hydrography map was generated using the latest topography dataset
- Near-automatic algorithm applicable for global hydrography delineation was developed
- Adjusted elevation and river width layers consistent with flow direction map are provided

MERIT Hydro: A High-Resolution Global Hydrography Map Based on Latest Topography Dataset

Dai Yamazaki^{1,2}, Daiki Ikeshima², Jeison Sosa³, Paul D. Bates³, George H. Allen⁴, and Tamlin M. Pavelsky⁵

¹Institute of Industrial Science, The University of Tokyo, Tokyo, Japan, ²Department of Civil and Environmental Engineering, Tokyo Institute of Technology, Tokyo, Japan, ³School of Geographical Sciences, University of Bristol, Bristol, UK, ⁴Department of Geography, Texas A&M University, College Station, TX, USA, ⁵Department of Geological Sciences, University of North Carolina, Chapel Hill, NC, USA

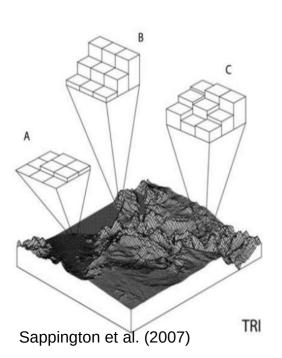


Amatulli *et al.* 2018 DOI: 10.1038/sdata.2018. 40

Figure 2. Graphical representation of landform shapes based on slope and curvature. Slope is the rate of change of elevation in the direction of the steepest descent, whereas the second order partial derivative (N-S slope) is the slope in the North-South direction. The profile and tangential curvatures identify concavity and convexity in the direction of the slope, or perpendicular to the slope. The second order partial derivatives (E-W slope) identify the curvature in the East-West direction.

Topography complexity layers

(Surface roughness and terrain forms analysis)



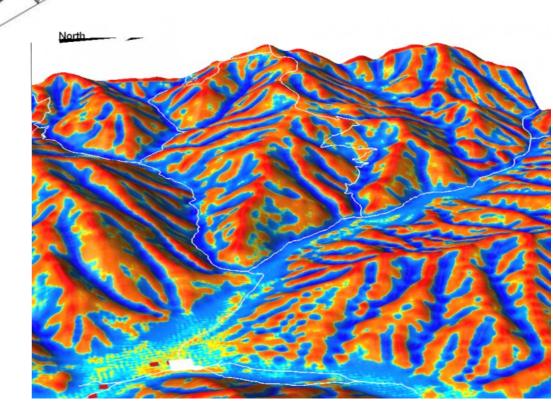
Roughness indices

Roughness - Median / SD
Topographic Position Index - Median / SD
Terrain Ruggedness Index - Median / SD
Vector Ruggedness Measure - Median / SD

Curvature

Profile curvature - Median / SD

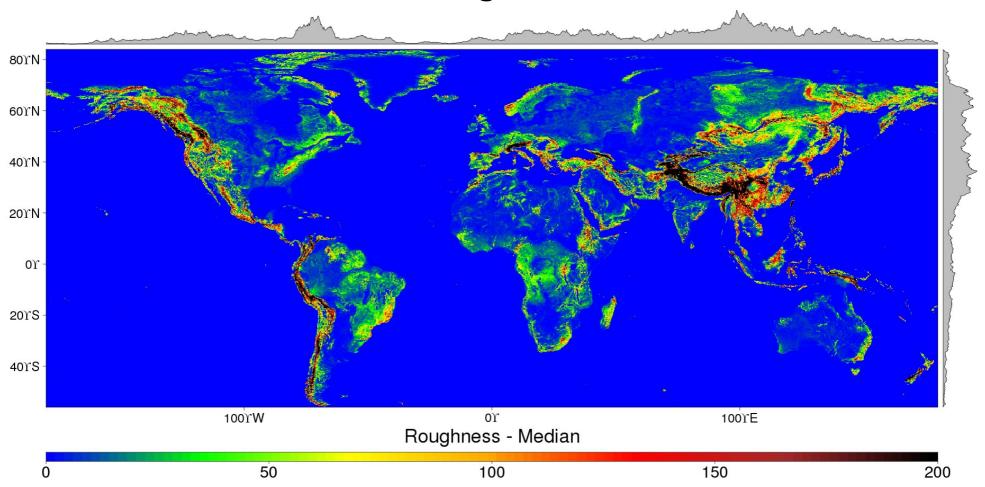
Tangential curvature - Median / SD



Topography complexity layers

(surface roughness and geo-morphology analysis)

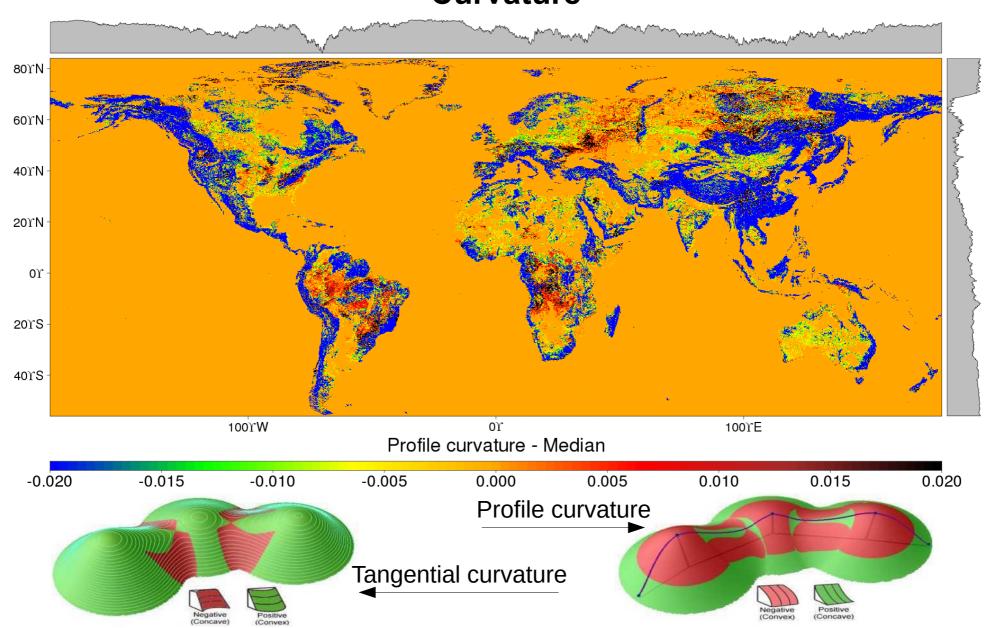




Topography complexity layers

(surface roughness and geo-morphology analysis)

Curvature



Geomorpho90m



Geomorpho90m, empirical DATA DESCRIPTOR evaluation and accuracy assessment of global highresolution geomorphometric layers

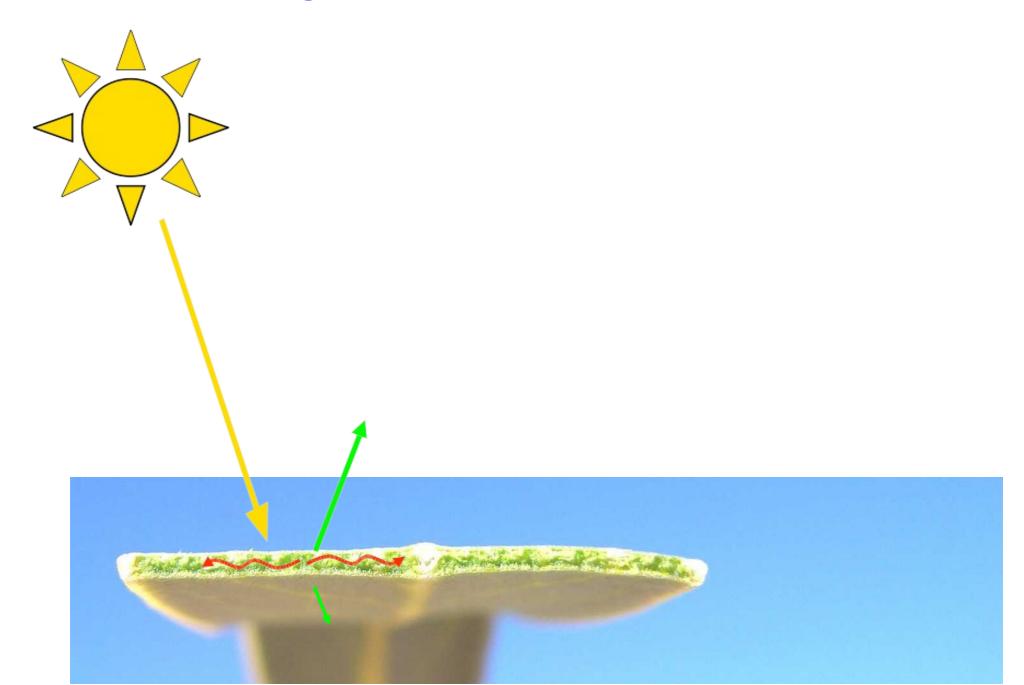
Giuseppe Amatulli 61,2 M, Daniel McInerney3, Tushar Sethi4, Peter Strobl5 & Sami Domisch 6 M

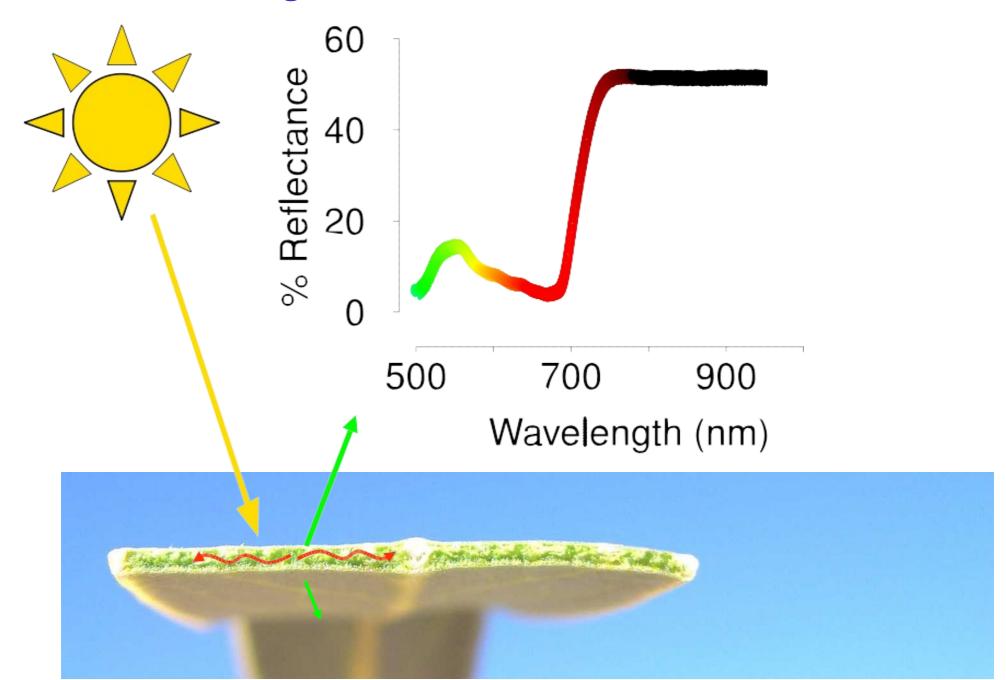
http://spatial-ecology.net/docs/build/html/GEODATA/geomorpho90m/geomorpho90m.html

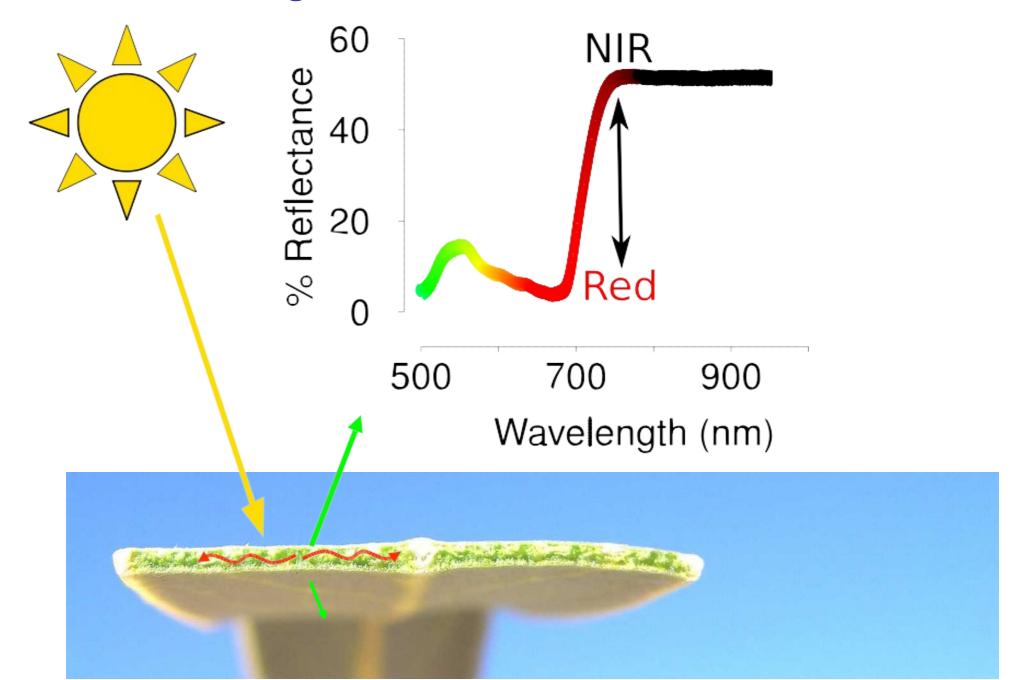


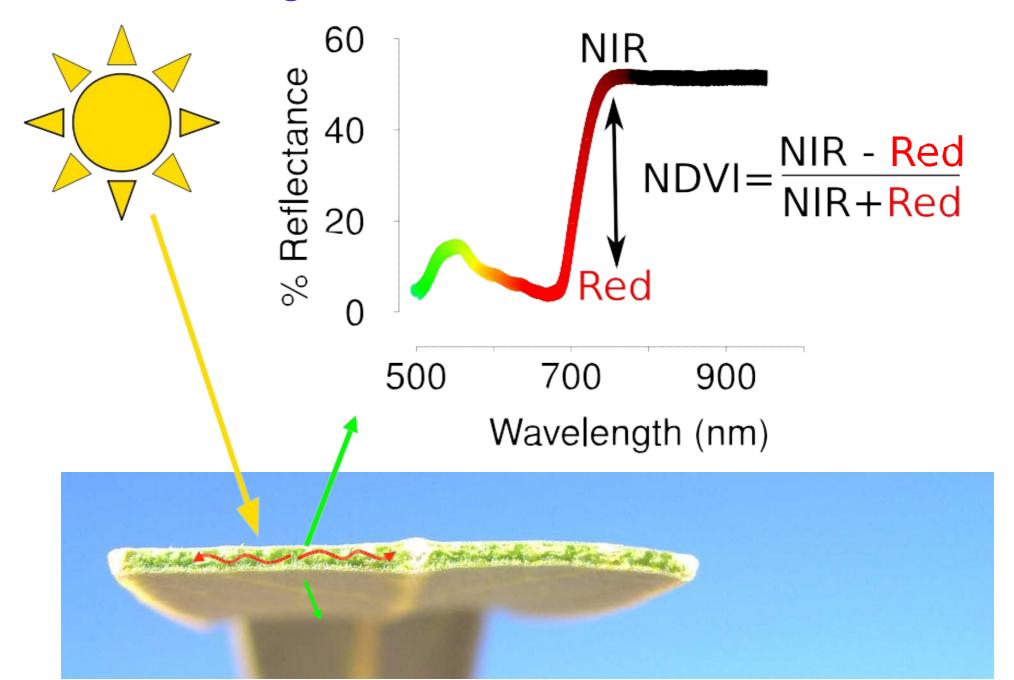
https://www.opentopography.org/

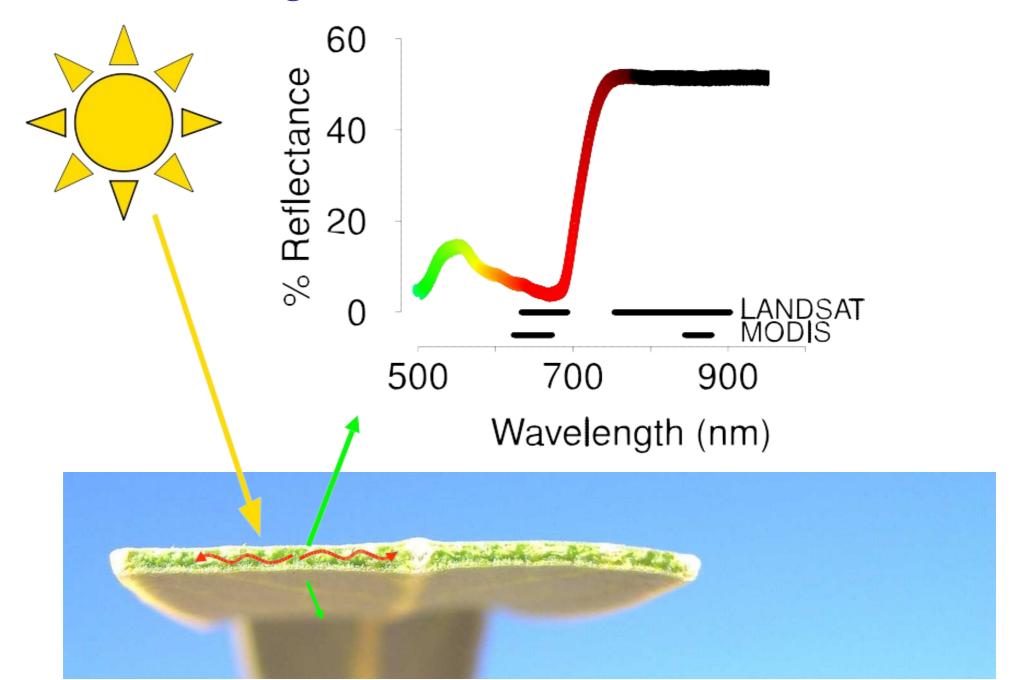
LiDAR data DEM DATA



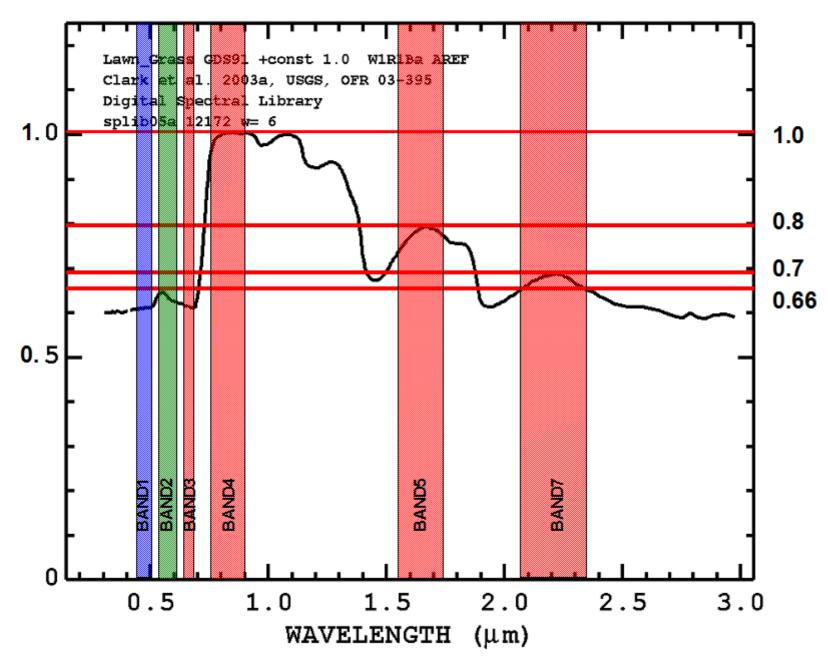








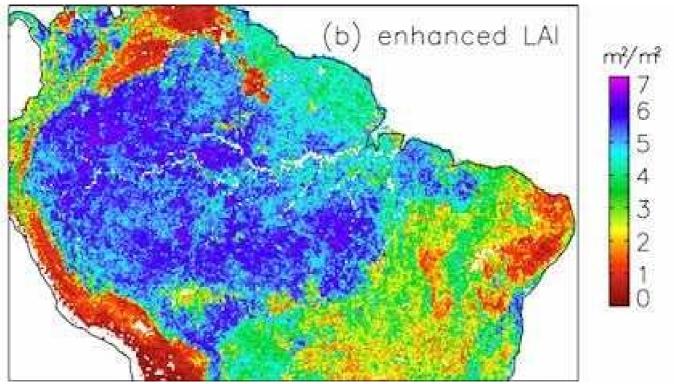
Landsat Multispectral Satellite Image



https://glad.geog.umd.edu/book/glad-landsat-ard-tools

Derived MODIS Metrics: LAI, fPAR, NPP

- *Leaf Area Index (^m)¹²: Reflectance + Biome Coefficients
- Net Primary Productivity (NPP)¹³: fPAR + LAI + Biome Parameters + Temperature (fPAR = Fraction of Absorbed Photosynthetically Active radiation)



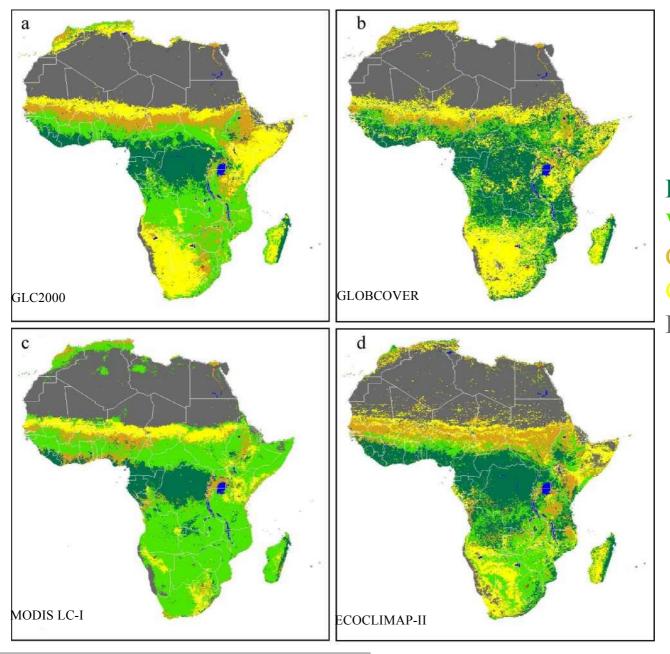
Both LAI and FPAR have been used extensively for calculation of photosynthesis, evaporation and transpiration of water, and Net Primary Productivity (NPP, which estimates how much carbon is taken in by vegetation).

¹² http://modis.gsfc.nasa.gov/data/dataprod/dataproducts.php?MOD_NUMBER=15

¹³ http://modis.gsfc.nasa.gov/data/dataprod/dataproducts.php?MOD_NUMBER=1

LULC: Products Matter

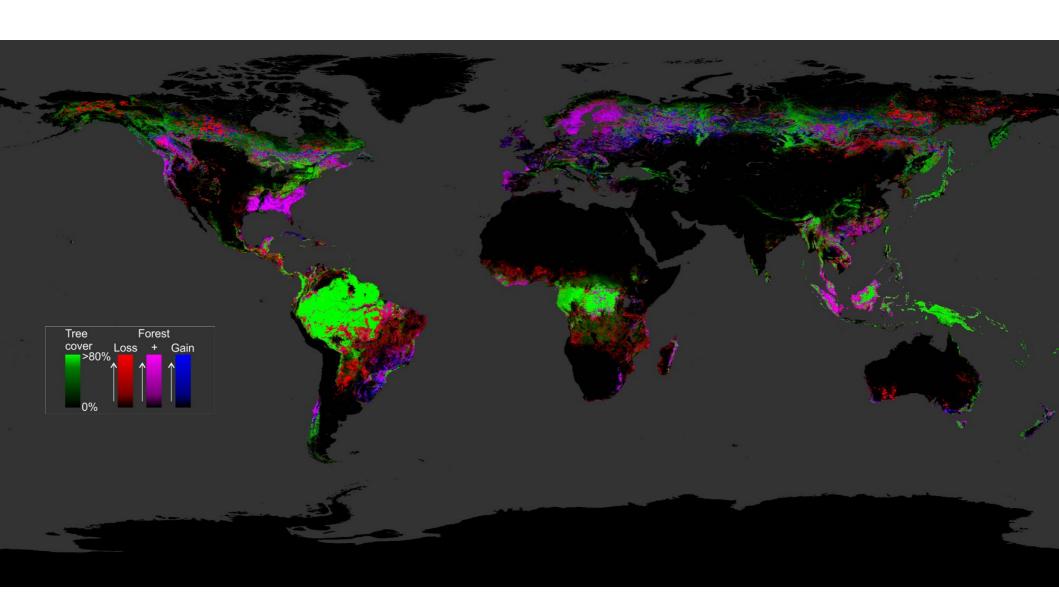
Comparison of four land cover maps for Africa¹⁵



Forest
Woodland/Shrubland
Cropland
Grassland Bare
Land Water Urban

¹⁵Intl J of Appl Earth Observation & Geoinformation (2011) 13(2): 207–219

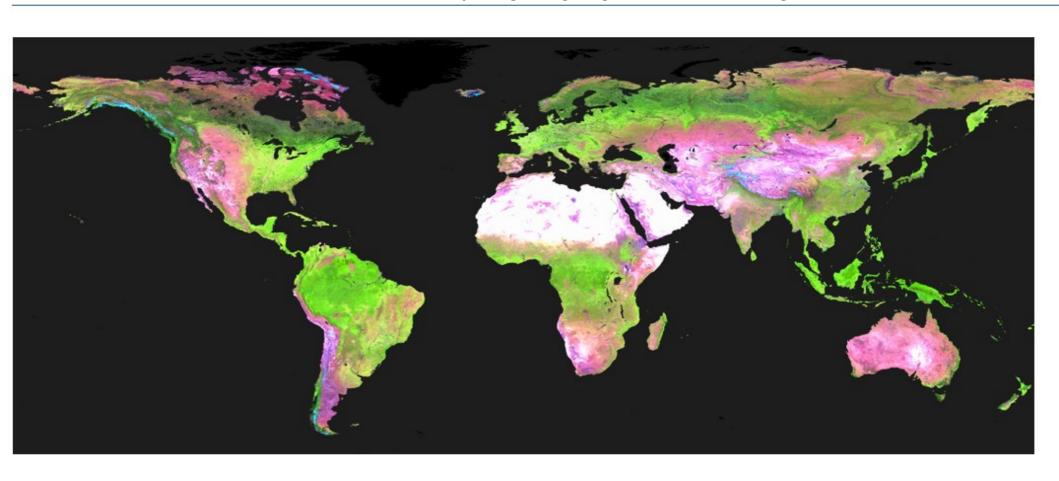
Global Forest Cover Change https://glad.umd.edu/projects/quantifying-global-forest-cover-change



GLAD Landsat ARD Tools

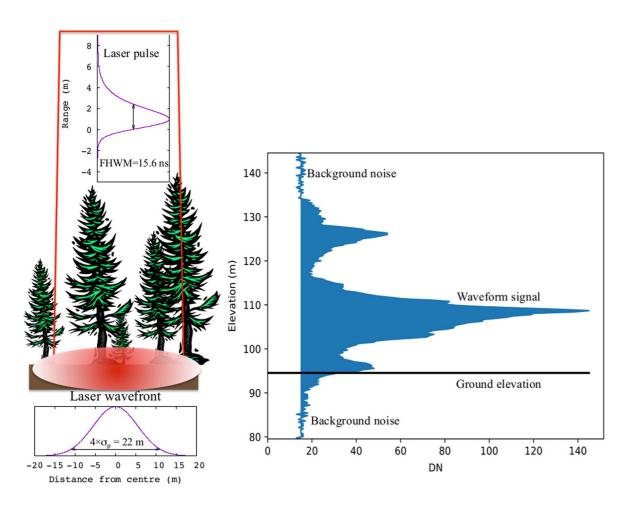
Home Software Download User Registration User Manual License and Disclaimer

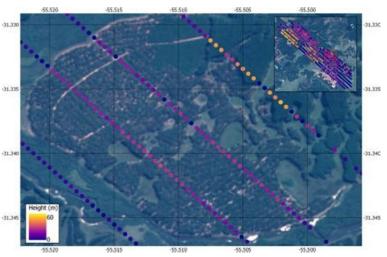
GLAD Landsat ARD Tools V1.1 https://glad.geog.umd.edu/book/glad-landsat-ard-tools





https://gedi.umd.edu/





How to search for geographic data

Raw data (satellite images, LiDAR) Sensor Products (LandCover, DEM)

NASA Products
MODIS Products
LANDSAT Products
Copernicus Products

https://datasetsearch.research.google.com/

Where do we go from here?

- 1. Data deluge → incredible opportunities
- 2.Data complexity → biogeographers must be vigilant

Analysis Steps:

- 3. Clearly define question, then seek data (not vice versa)
- 4. Understand source and limitations
- 5. Propagate uncertainties²⁰!

Why use Open Source Software? Scientific/Technical aspects: data flow

Codes that are easily published > no license constraints Complex work-flows > integrate different data analysis methods

