Inferring higher frequency water observations from WOFS & MODIS

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N. Mueller, A. Lewis, D. Roberts, S. Ring, R. Melrose, J. Sixsmith, L. Lymburner, A. McIntyre, P. Tan, S. Curnow, A. Ip, 2016. <u>Water observations from space: Mapping surface water from 25 years of Landsat imagery across Australia</u>. Remote Sensing of Environment 174, 341-352, ISSN 0034-4257.

Motivation

- WOFS (Water Observation from Space)
 - It is the world's first continent-scale map of surface water¹ and provides images and data showing where water has been seen in Australia from 1987 to the present.
 - It is a Landsat derived product
- Daily WOFS observations
 - 25 m grid
 - Infrequent observations (~16 days)

Objective and proposal

• Generate higher temporal frequency water observations from the Water Observation from Space (WOFS) images

Calculate a Water Index (WI) from a higher temporal frequency product (MODIS)

Calibrate the WI to the daily WOFS observations

Generate higher temporal frequency water observations using the calibrated relationship

Data SIO, NOAA, U.S. Navy, NGA, GEBCO Image Landsat// Copernicus Image IBCAO

Google Earth

Methods: Processing Daily WOFS Data



Steps

- 1. Extract daily images (25 m)
- 2. Convert bitcode flags to dry (0) or water (1)
- 3. Calculate % water coverage per MODIS cell (500 m)

Challenges

- Many missing days
- Lots of "no data" values
- Multiple scenes per day
- Working with bit codes
- Spatial aggregation to match the MODIS resolution



Methods: Spatial Aggregation

WOFS 25 m data

- 0 = dry
- 1 = wet

Aggregated 500 m data (matching MODIS)

- 0 = 0% water coverage
- 1 = 100% water coverage





- MODIS (Moderate Resolution Imaging Spectroradiometer)
 - 705 km, TERRA: descending node, AQUA: ascending node.
 - Sun-synchronous, near-polar, circular.











1 Che, X., Feng, M., Jiang, H., Song, J., & Jia, B. (2015). Downscaling MODIS surface reflectance to improve water body extraction. Advances in Meteorology, 2015.



• Challenges

- Two Scenes for our Study Area: reprojection and join the scenes.
- It works individually for each single band
- The images cannot be read as a xarray
- 8-days composite images \rightarrow 46 images per year







Correlating MODIS and WOFS



Steps

- 1. Identify common dates for WOFS and MODIS
- 2. Correlate:
 - WOFS % water coverage
 - MODIS Water Index (WI)
- 3. Predict % water coverage for dates where WOFS images are not available

Challenges & limitations

- MODIS 8-days composite images → WOFS images represent a single day
- Unexpected % water vs WI relationships

Correlating MODIS and WOFS

Initial testing indicated a strong correlation: Pearson's r = 0.91

Visualisation shows that the relationship is not linear

Adopted a threshold approach

- Dry: WI <-.20
- Wet: WI > -0.19
- Partially inundated:

(-0.19 <= WI <= -0.2)



Example Results (Jan-Feb & Nov-Dec 2018)



Water classification from MODIS









Conclusions

- Calibrating water indices derived from MODIS to the WOFS water data is fast & effective
- Results useful for large-scale assessments of inundation area at a more frequent temporal resolution than offered by WOFS

Limitations include:

- 8 day time step insufficient to capture transient events
- Unable to identify rivers and streams during within bank flows

Future Work

Improve Water Index (WI)

- Calculate additional indices from MODIS
- Optimise thresholds (rather than manual adjustment)
- Downscale 500m cells to WOFS 25m cells
 - Establish relationship between inundated cells in WOFS and strength of MODIS WI

Validate on independent data

Thanks to the OZEWex tutors, trainers & staff

Software Environment

Jupyter Notebook

Key packages:

• Xarray

• Digital Earth Australia

Git Hub version control

NCI computing environment