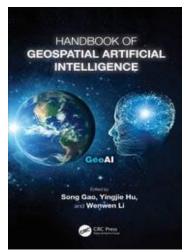
Chapter 5: GeoAl for Spatial Image Processing



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Arundel, S. T., McKeehan, K. G., Li, W., & Gu, Z. (2023). GeoAl for Spatial Image Processing. In *Handbook of Geospatial Artificial Intelligence* (pp. 75-98). CRC Press.

Chapter outline

- Introduction
 - Origins of image processing
 - (Digital) image processing in the spatial realm
 - Image processing in the AI realm
- Image processing in the GeoAl domains
- GeoAl-specific methods and challenges
 - The challenge in training data annotation
 - The challenge of scale
 - The challenge of image resolution
 - The challenge of space and time

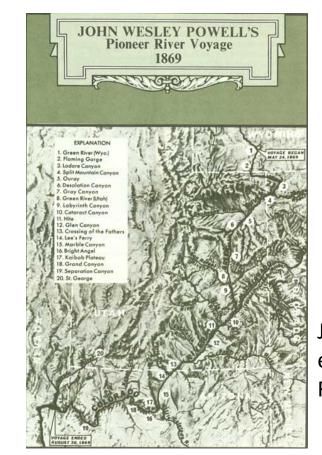
- Future research
 - Geometric algebra
 - Combining GeoAl with physical modeling
 - Knowledge-guided GeoAl
 - Al ethics

Origin for image processing

• 1830s: cameras became essential tools in laboratories and on geographic expeditions



Anna Atkins published her botanical study of algae in the 1840s using cyanotype images



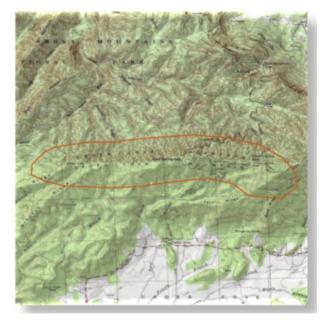
John Wesley Powel 1969 expedition to the Colorado River

Spatial image processing

What is spatial image?



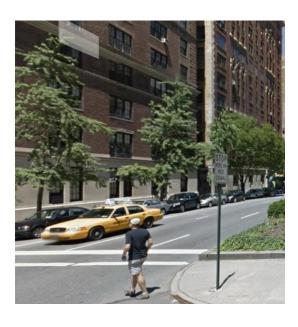
Remote sensing images



Topo maps



Digital Elevation Model (DEM) data



Street view images

Spatial information processing

- Object-based image analysis (OBIA)
- GEOBIA "scientific resolution" create a dynamic multiscale object-based contextual model (Hay and Castilla 2008)

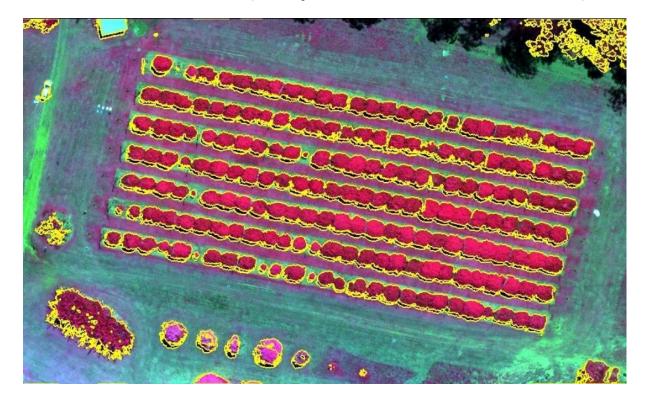
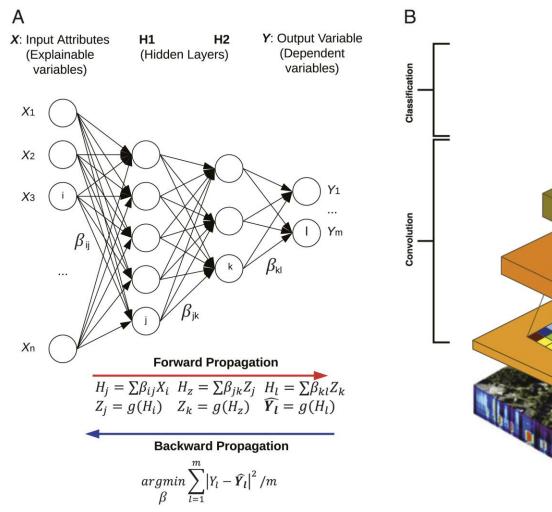
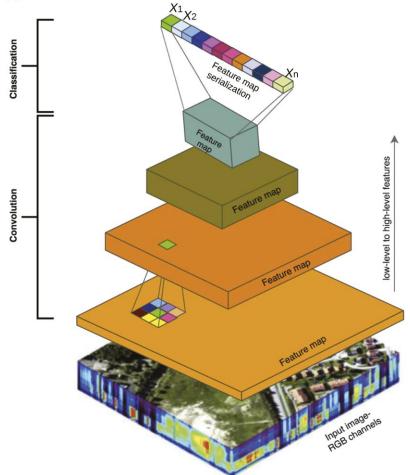


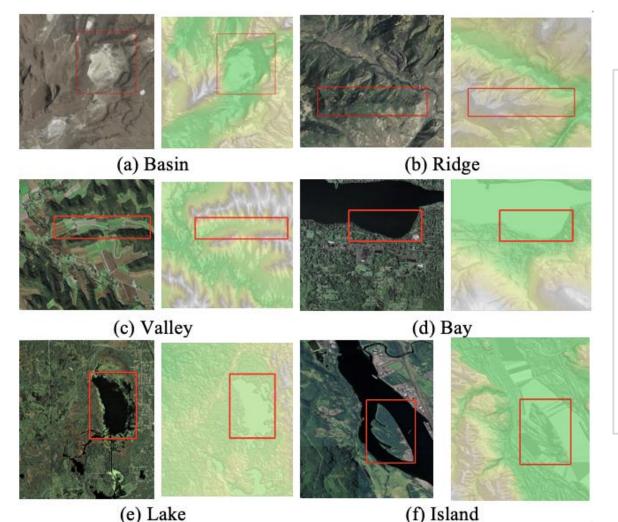
Image credit: https://www.rsrc.org.au/geobia-research

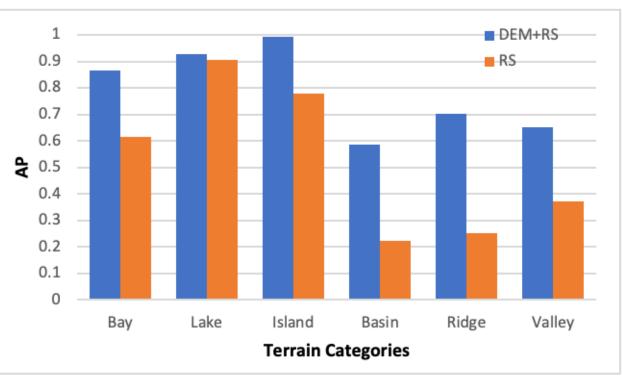
GeoAl and Deep learning for spatial image processing





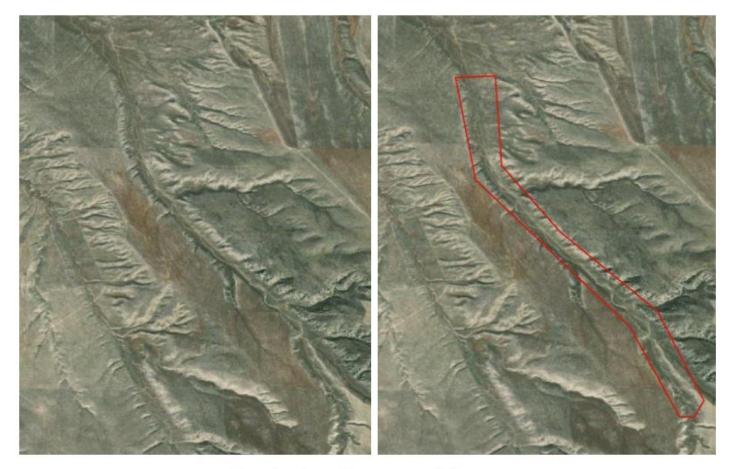
Multi-source GeoAl





Wang and Li (2021), CEUS

Training data annotation



Bachelor Canyon, Montana

• Scale

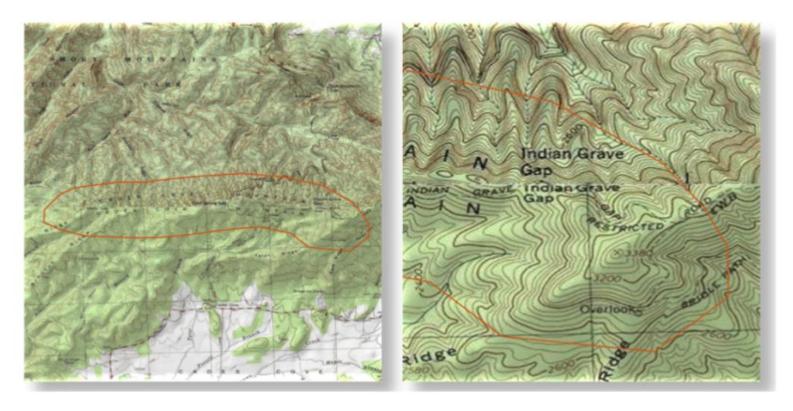
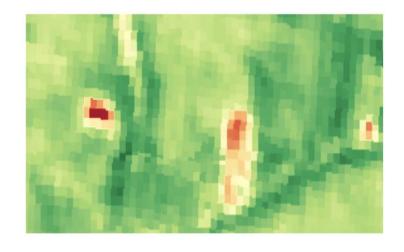


Figure 3. Demonstration of where scale could impact the effective prediction of a landform feature (Codes Cove Mountain, Tennessee), particularly as related to the image resolution at the smaller scale (left).

Resolution & data quality



Optical image

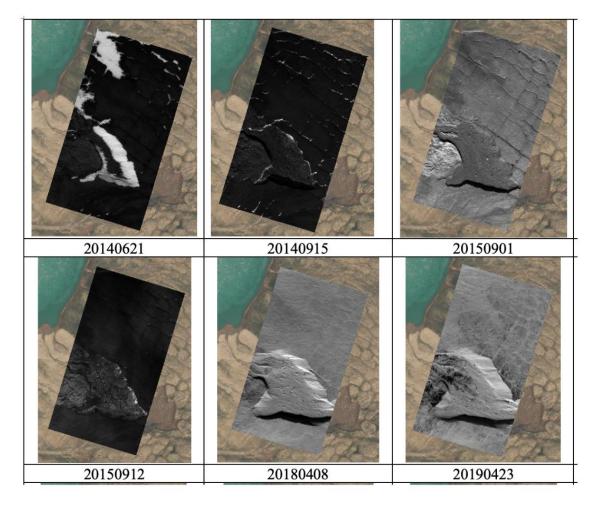


NDVI
(Normalized Difference
Vegetation Index)



DEM

Space and time





Future research

Geometric algebra → feature embedding

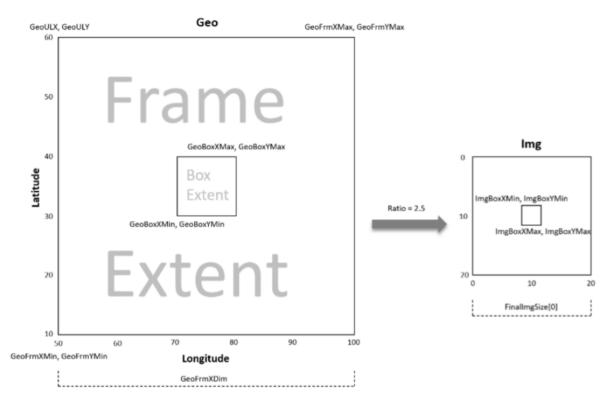
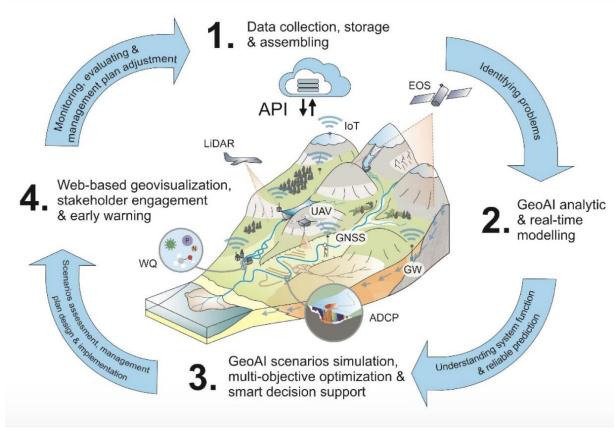


Figure 2. An illustration for coordinate transformation from a geographical extent to an image extent.

Future research -2

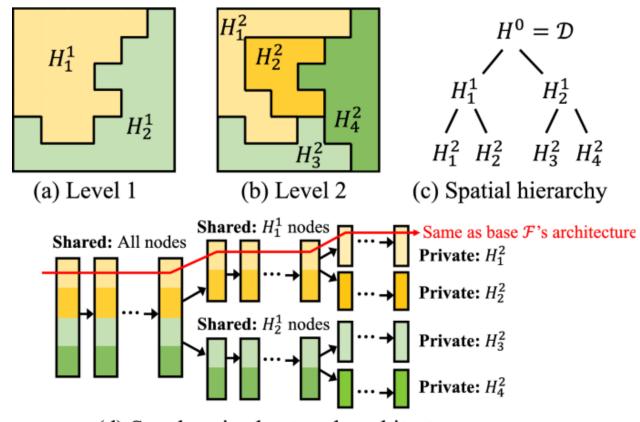
- Integration with physics-based models
 - AI-based calibration
 - Forecasting



Gonzales-Inca et al. 2022

Future research -3

Geo-knowledge based learning – spatial explicit modeling



(d) Synchronized network architecture

Xie et al. 2023

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