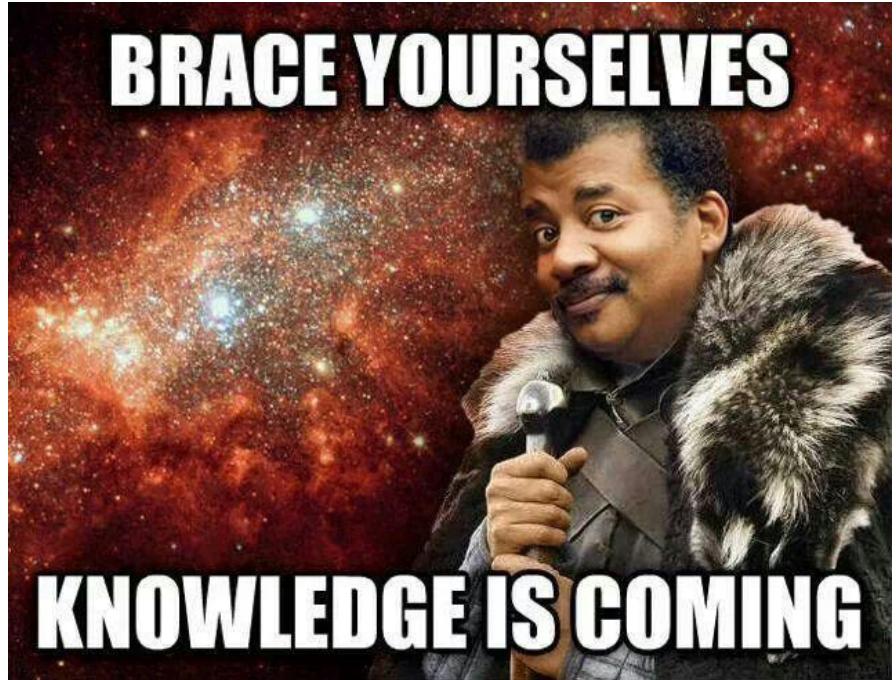


Overview of surface water mapping methods

HYDRAFloods Training
11 October 2021



- Remote Sensing Overview
- Optical water mapping
- SAR water mapping
- Thresholding algorithms
- Advanced algorithms



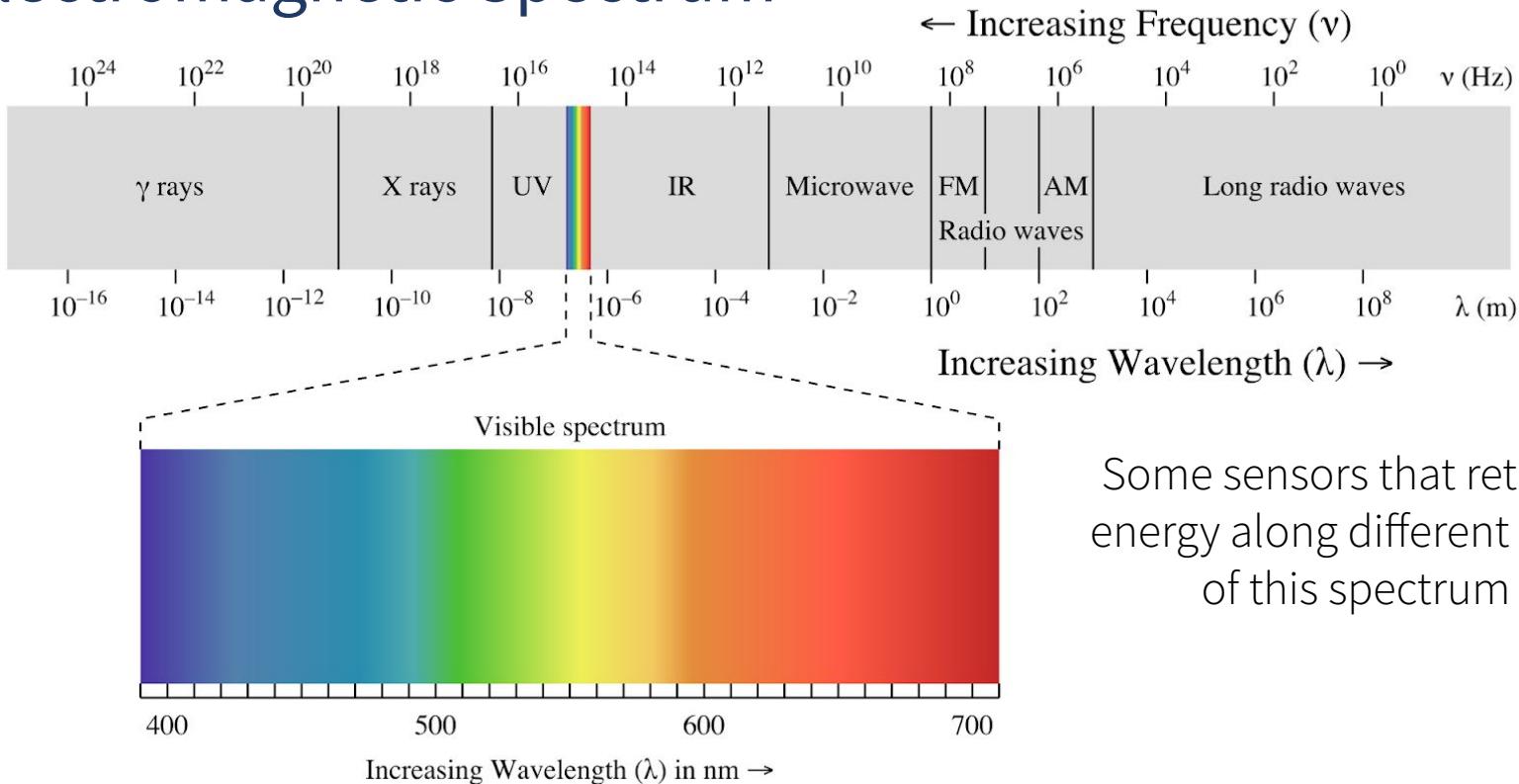
Remote Sensing is...

the practice of deriving information about the Earth's land and water surfaces using images **acquired from an overhead perspective**, by employing **electromagnetic radiation** in one or more regions of the electromagnetic spectrum, **reflected or emitted** from the Earth's surface.

Remote Sensing is...

the practice of deriving information about the Earth's land and water surfaces using images **acquired from an overhead perspective**, by employing **electromagnetic radiation** in one or more regions of the electromagnetic spectrum, **reflected or emitted** from the Earth's surface.

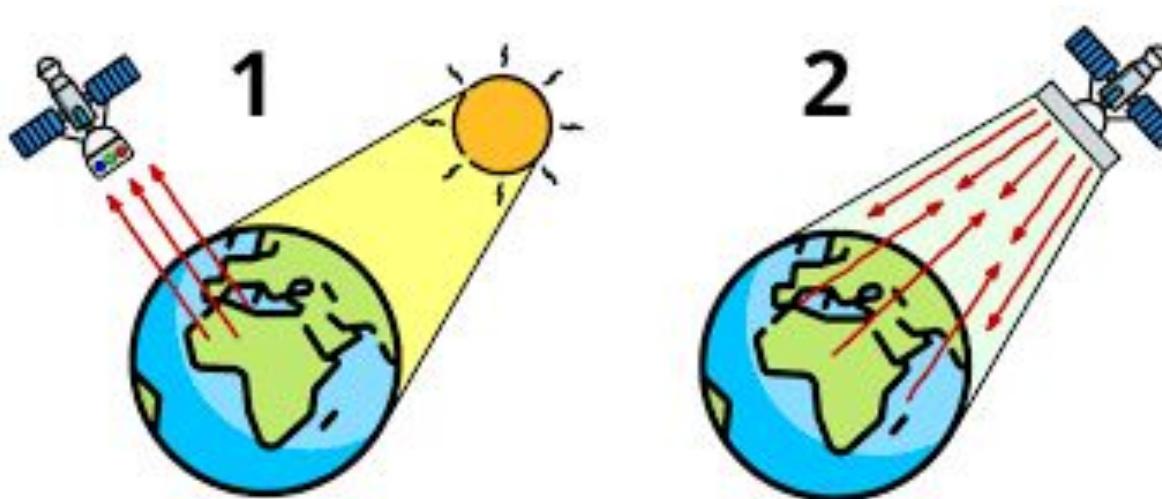
Electromagnetic Spectrum



Some sensors that retrieve energy along different areas of this spectrum

Types of remote sensing

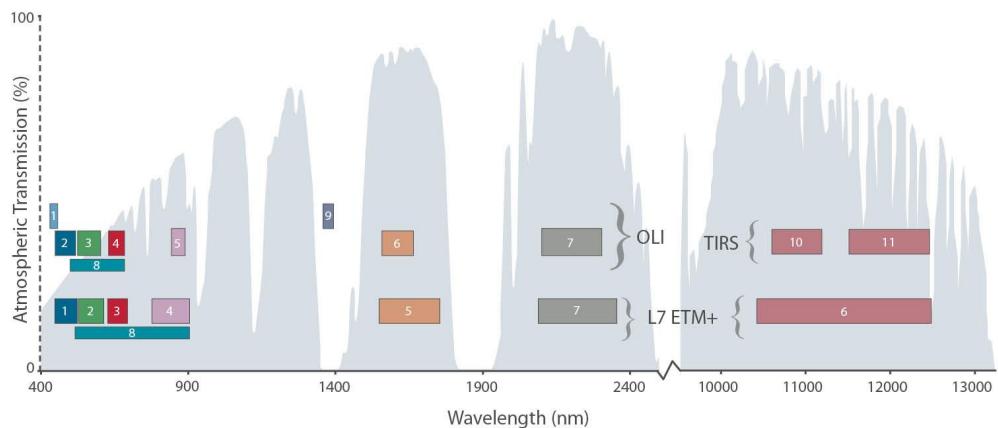
1. Passive Remote Sensing: Energy = Radiation from Earth
2. Active Remote Sensing: Energy = Instrument



<https://paititi.info/research-technology/remote-sensing-from-space/>

Optical Water Mapping

- Measures reflected light from sun in different wavelengths of the EM
- Combinations of bands can highlight different physical properties



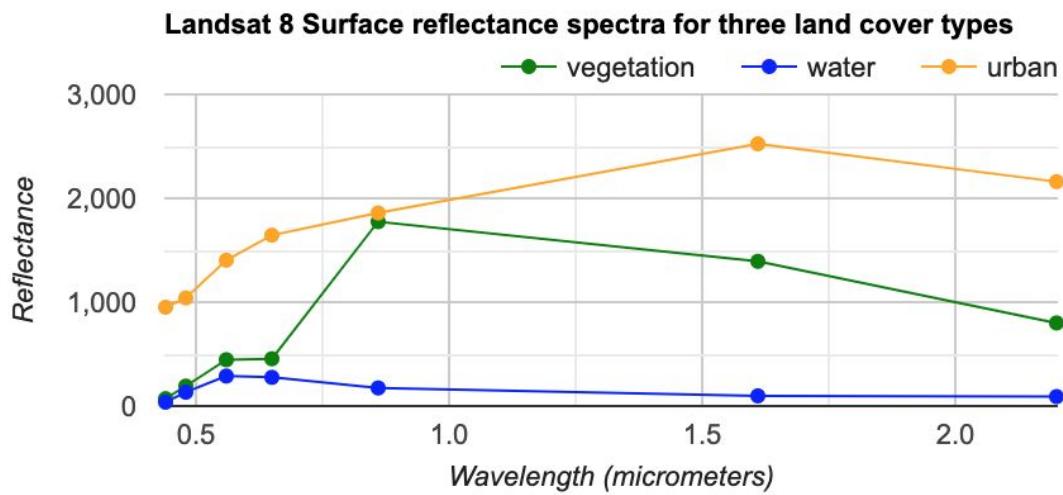
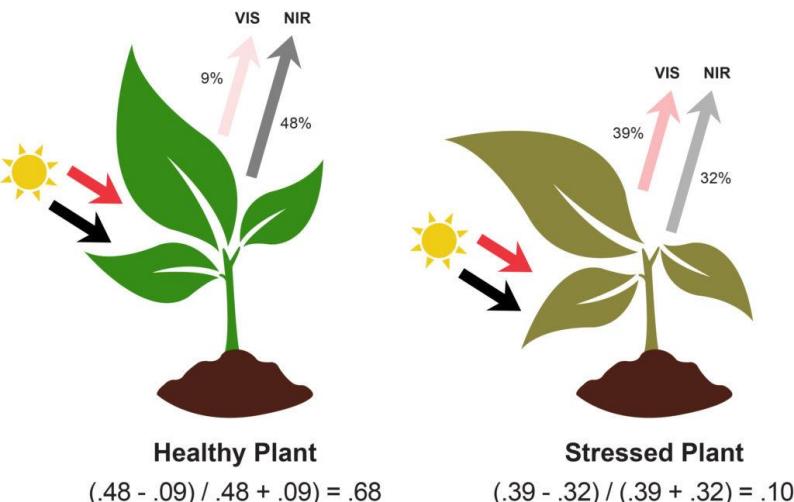
<https://landsat.gsfc.nasa.gov/landsat-8/landsat-8-overview>

CC BY-SA 3.0, https://commons.wikimedia.org/wiki/File:Satellite_image_of_Italy_in_March_2003.jpg



Optical Water Mapping

- Spectral indices are combinations bands from two or more wavelengths that indicate the relative abundance of features of interest

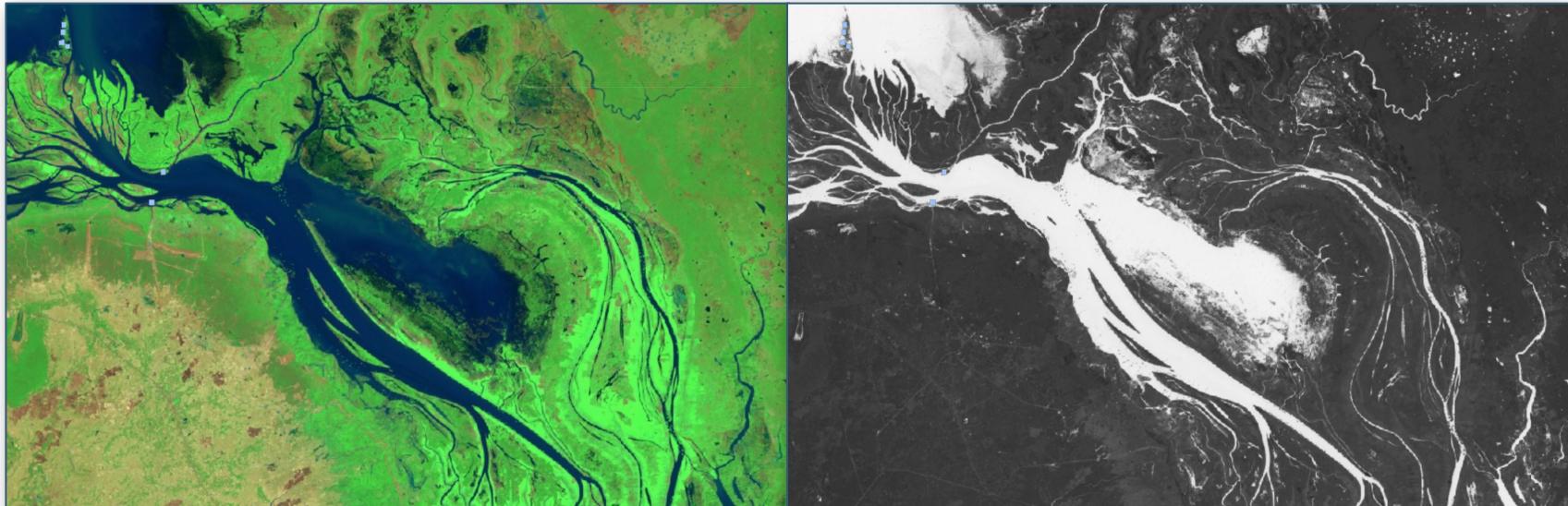


Band Combinations



Water Indices

- Used to transform multispectral imagery to highlight water
- Rely heavily on shortwave infrared band



Water Indices

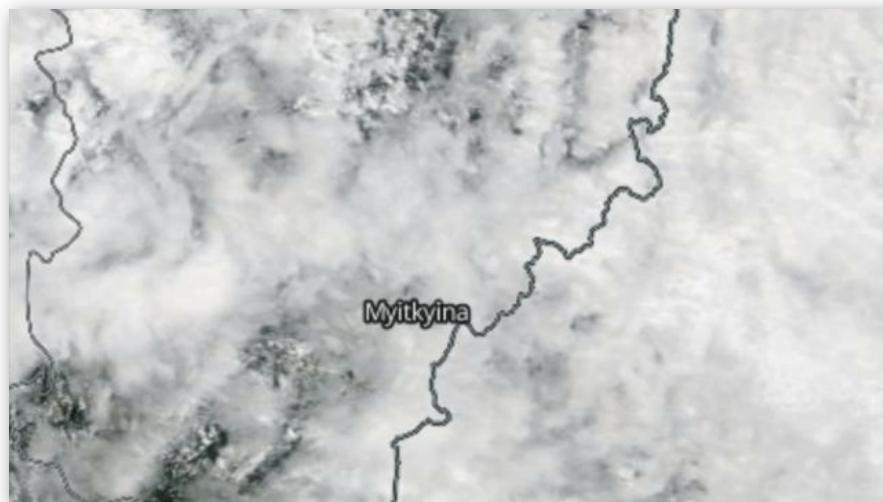
Index Name	Equation	Citation
Normalized Difference Water Index	$(\text{green} - \text{nir}) / (\text{green} + \text{nir})$	https://doi.org/10.1080/01431169608948714
Modified Normalized Difference Water Index	$(\text{green} - \text{swir1}) / (\text{green} + \text{swir1})$	https://doi.org/10.1080/01431160600589179
Automated Water Extraction Index (no shadow)	$4.0 * (\text{green} - \text{swir1}) - ((0.25 * \text{nir}) + (2.75 * \text{swir2}))$	https://doi.org/10.1016/j.rse.2013.08.029
Automated Water Extraction Index (shadow)	$\text{blue} + 2.5 * \text{green} - 1.5 * (\text{nir} + \text{swir1}) - 0.25 * \text{swir2}$	https://doi.org/10.1016/j.rse.2013.08.029
Water Ratio Index	$(\text{green} + \text{red}) / (\text{nir} + \text{swir1})$	https://doi.org/10.1109/GEOINFORMATIC.S.2010.5567762
Land Surface Water Index	$(\text{nir} - \text{swir1}) / (\text{nir} + \text{swir1})$	https://doi.org/10.1080/01431160802575653

Herndon et al., 2019

Zhou et al., 2017

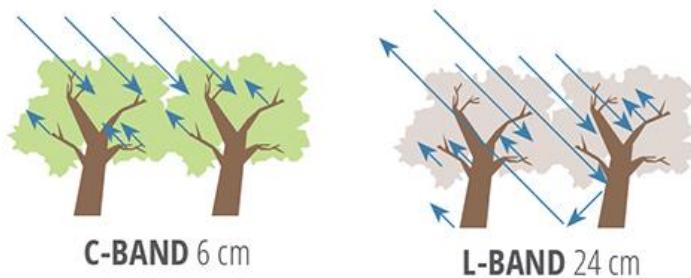
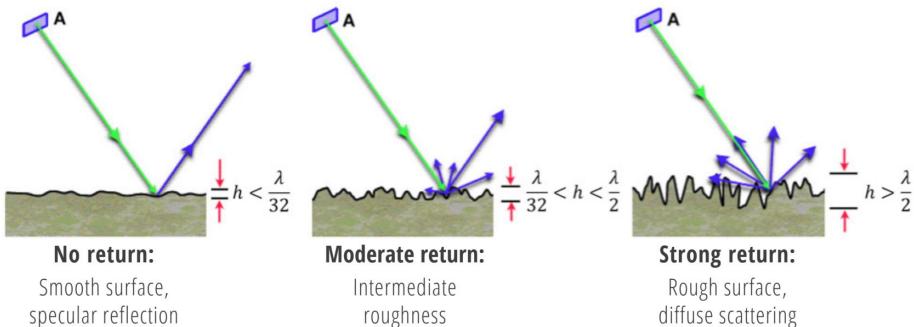
Limitations

- Poor atmospheric conditions can affect water mapping results
- No one best method - best approach to use depends on case
- Clouds!!!



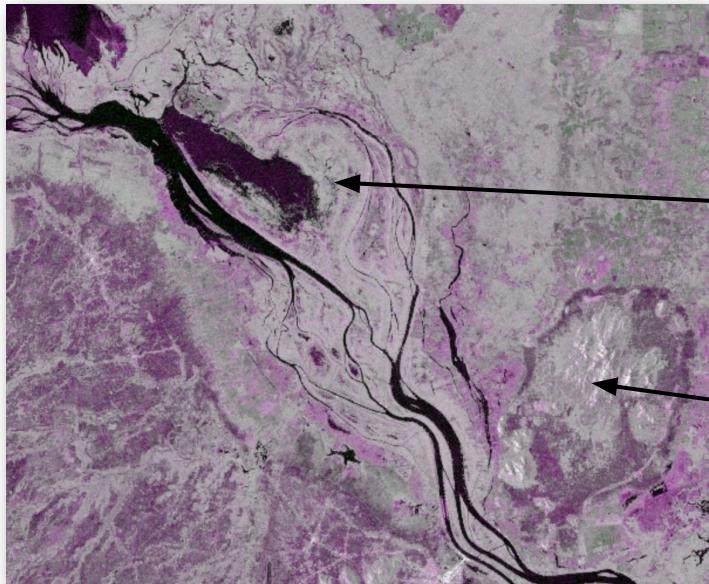
SAR Water Mapping

- **Synthetic Aperture Radar (SAR)** is a type of active remote sensing that is available in all weather conditions
- SAR measures the amount of energy returned to the sensor (backscatter)
- Typically SAR sensors operate in a few frequencies (X, C, L, and P bands)



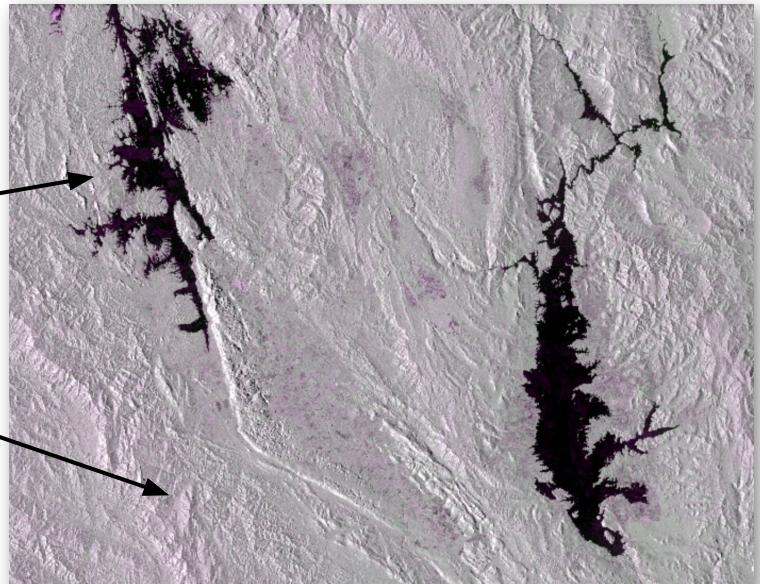
SAR Water Mapping

1. Water - specular reflection (low intensity)
2. Land - scattering by vegetation (high intensity)



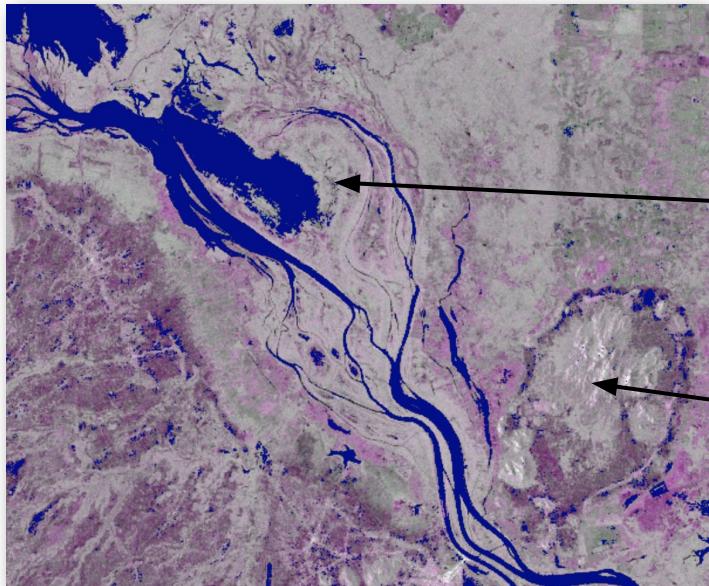
1.

2.



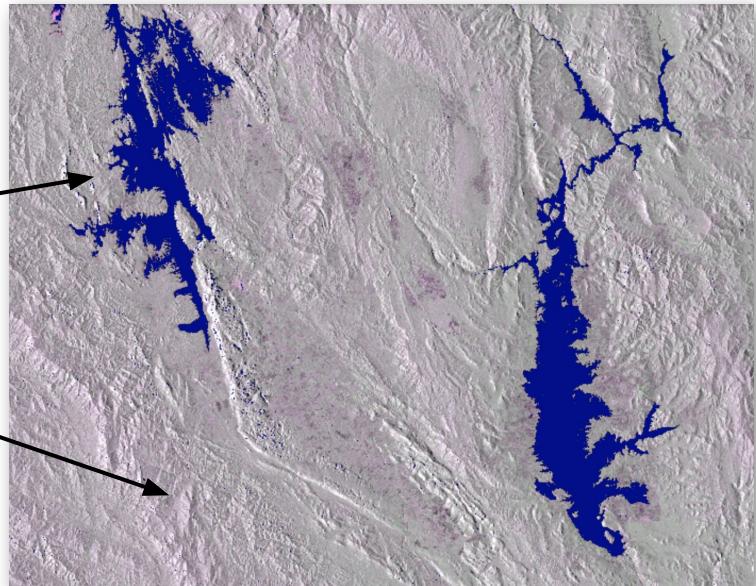
SAR Water Mapping

1. Water - specular reflection (low intensity)
2. Land - scattering by vegetation (high intensity)



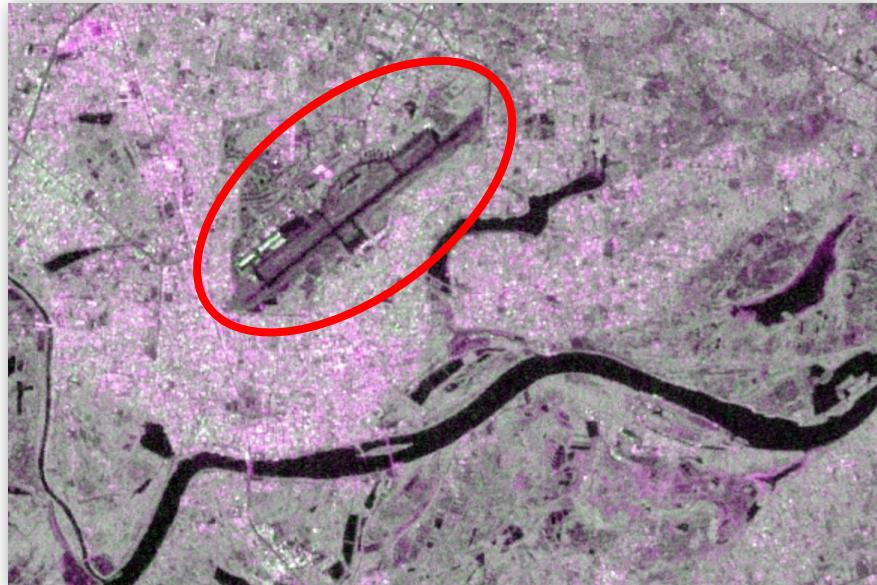
1.

2.



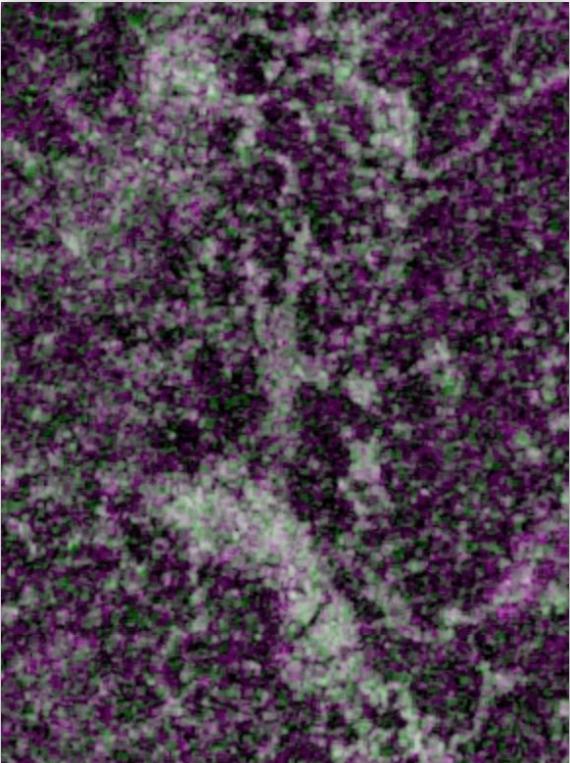
Limitations

- Water mapping with SAR is **based on the specular reflection** of the SAR signal
- Smooth surfaces “look” like water to a SAR sensor



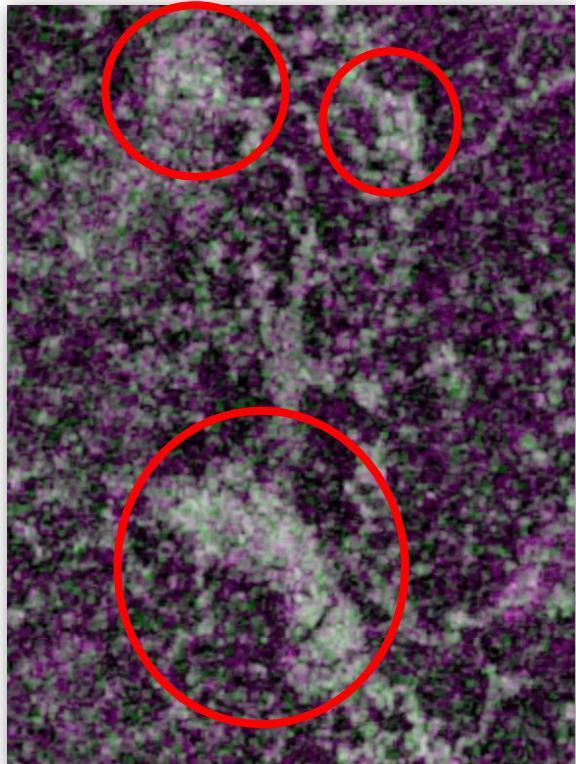
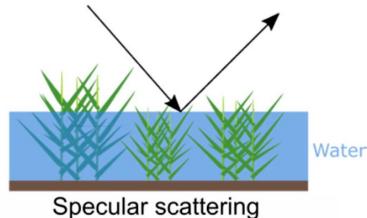
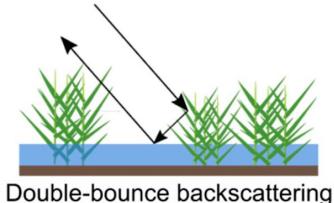
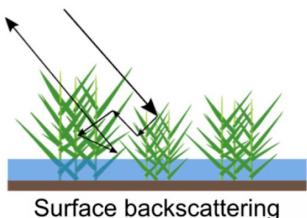
Limitations

- What is water in this image? 



Limitations

- Emergent vegetation can affect signal in water
- Bare, dry soils can produce low backscatter signals

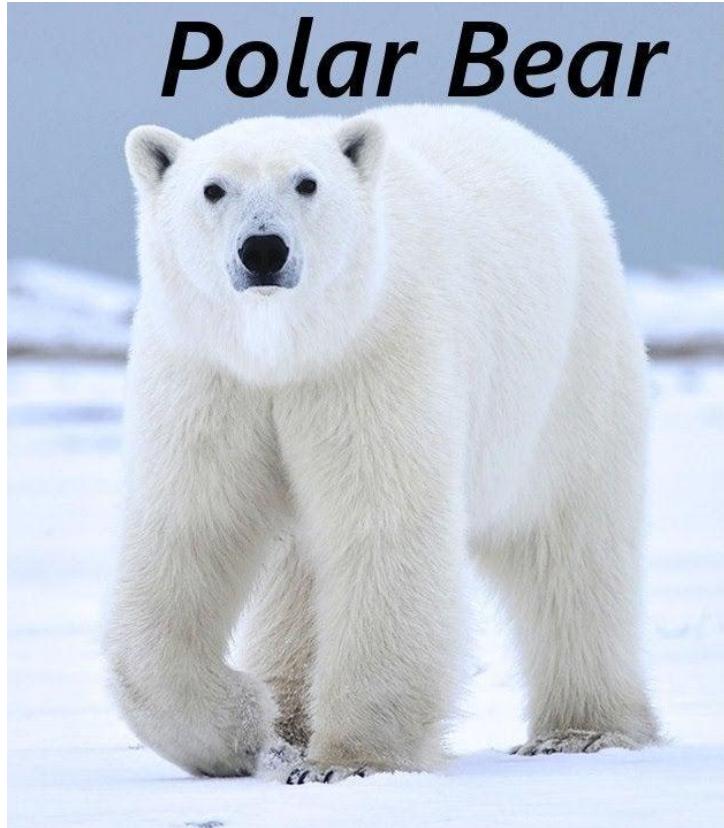


Limitations

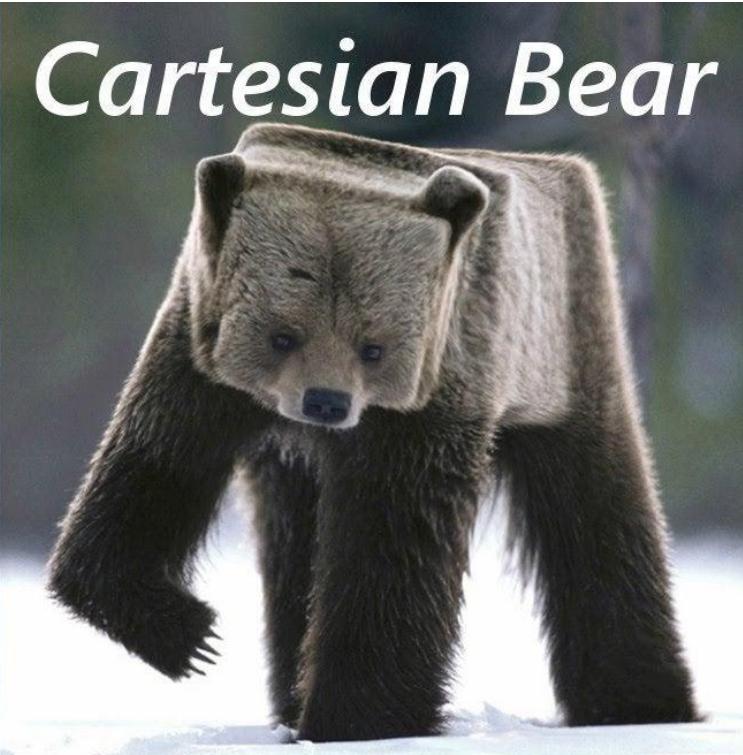
- Good practice to inspect regional conditions before applying workflows



Polar Bear

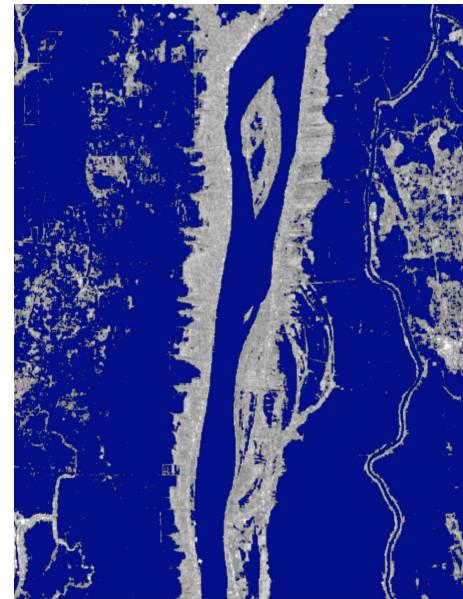
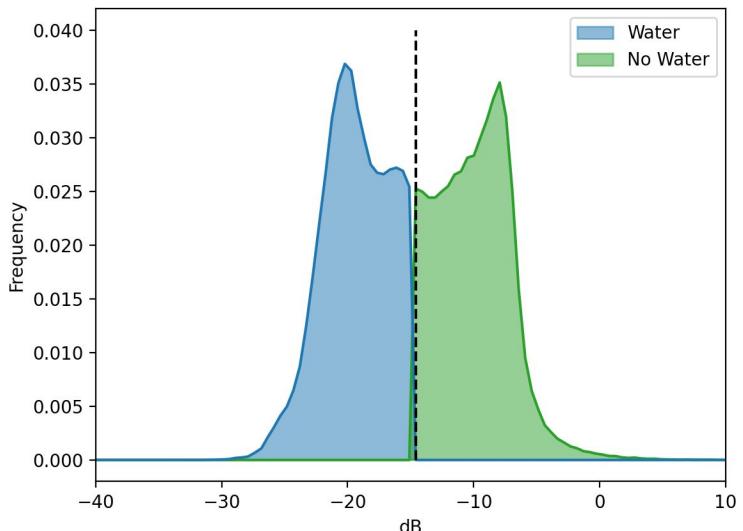
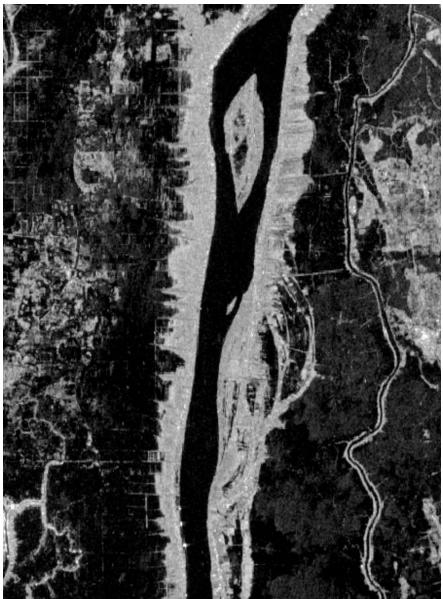


Cartesian Bear



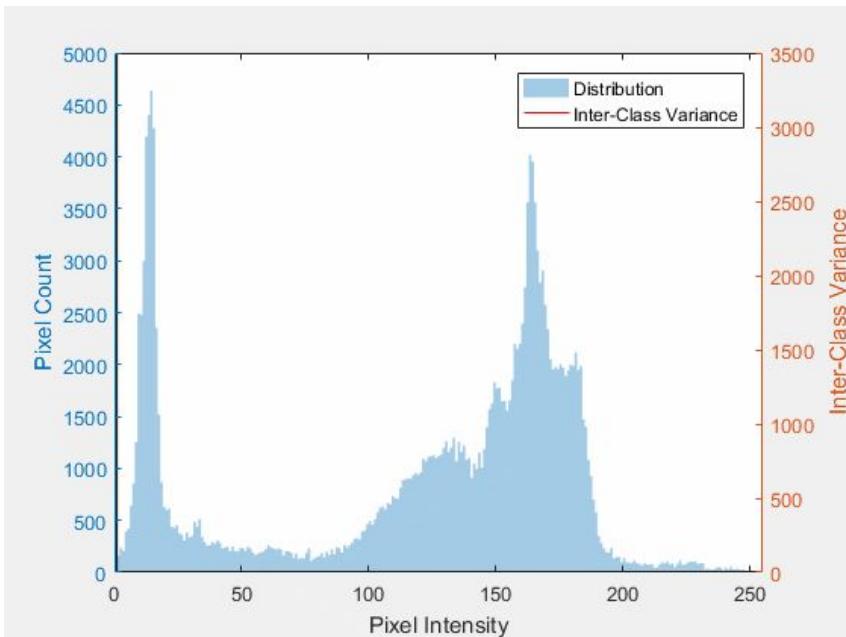
Thresholding Algorithms

- Simplest form of image segmentation
- Uses a single value to classify image into binary classes
- Manual identification of threshold is timely and subjective



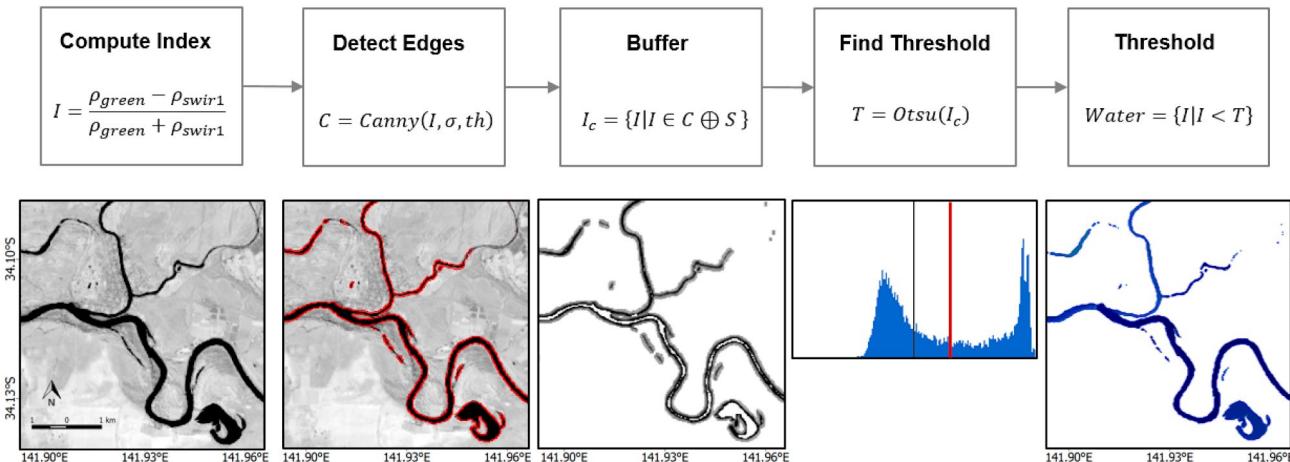
Otsu's method

- Automated histogram-based thresholding approach
- Maximizes inter-class variance between two classes
- **Assumes there are only two classes**, a background and foreground



Constraining histogram sampling

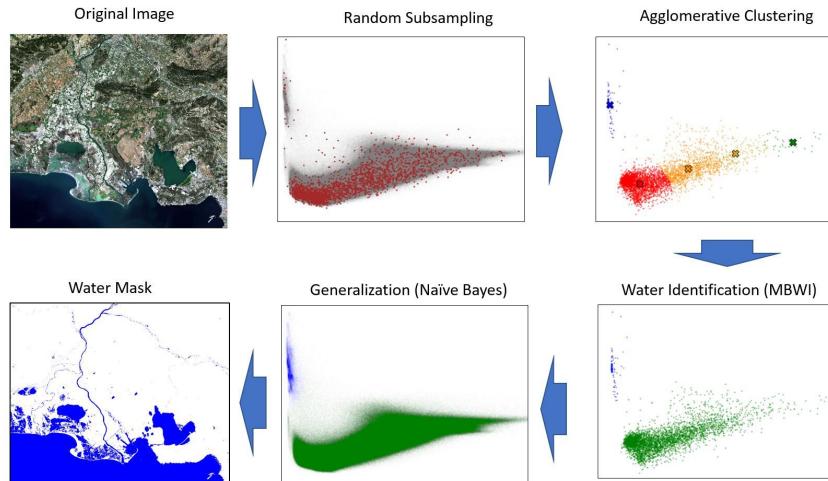
- Ensure that thresholding only has two classes to segment
(Donchyts et al., 2016, Cao et al., 2019, Markert et al., 2020)
- Based on image characteristics (i.e. edges or bimodal areas)



CC-BY 4.0, <https://www.mdpi.com/2072-4292/8/5/386/htm>

Beyond simple thresholding

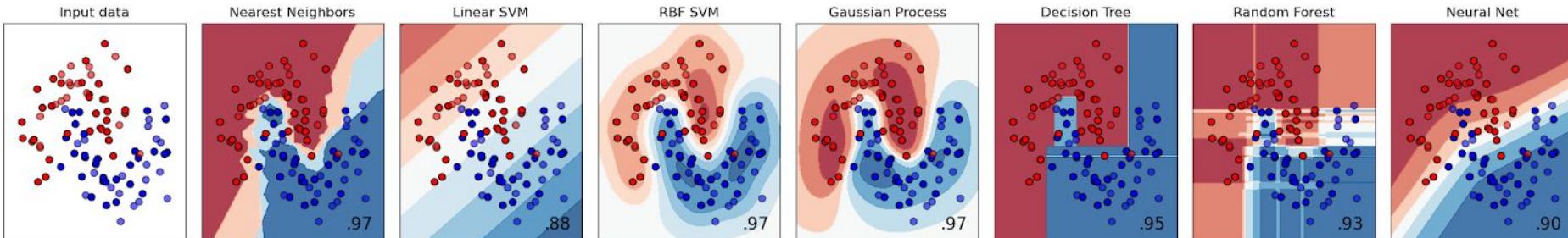
- Clustering algorithms to determine threshold ([Chang et al., 2020](#))
- Clustering then bayesian generalization([Codeiro et al., 2021](#))



<https://github.com/cordmaur/WaterDetect>

Classical machine learning

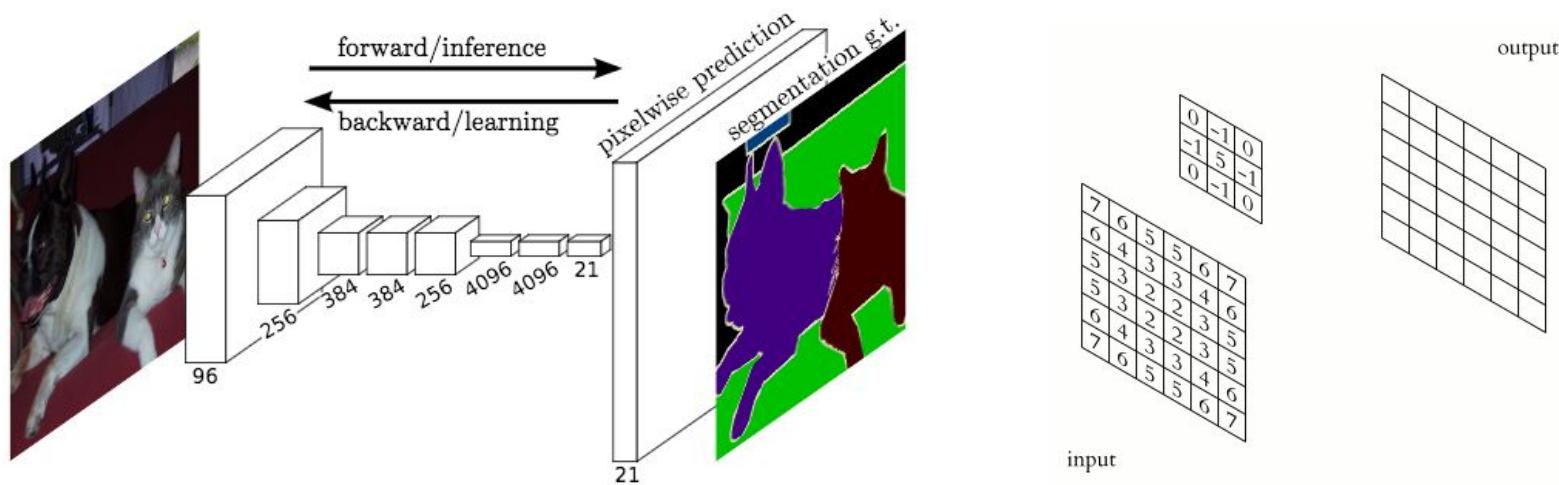
- Take additional dimensionality into account and learn non-linear patterns when mapping water
- Support Vector Machine (SVM) and Random Forest (RF) are popular algorithms and meet many needs



https://scikit-learn.org/stable/auto_examples/classification/plot_classifier_comparison.html#sphx-glr-auto-examples-classification-plot-classifier-comparison-py

Deep Learning

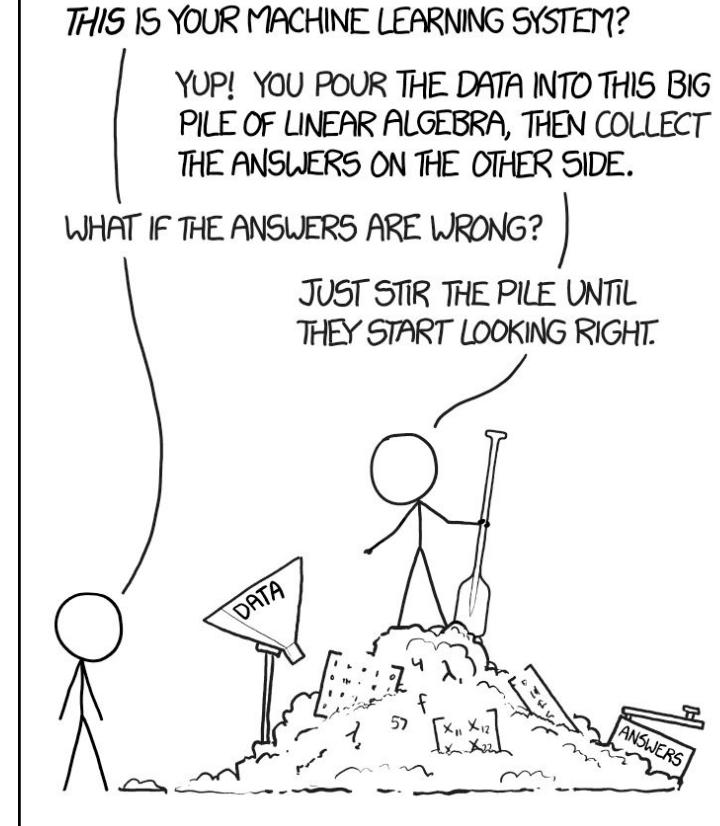
- Use large number (millions) of parameters that the computer “learns” to predict results
- Typically use fully convolutional neural network architecture



<https://arxiv.org/pdf/1411.4038.pdf>

Limitations

- “No Free Lunch” Theorem
- Sensitive to data inputs and requires a lot of training data
- Requires correct model type/architecture for problem
- Computationally expensive to train models



CC BY-NC 2.5, <https://xkcd.com/1838/>



Thank you for your attention!

kel.markert@nasa.gov

