

SEST-6577

**Geographic Information Systems for Security Studies**

Lecture 04 (Georeferencing and Vectorization)

Yuri M. Zhukov  
Associate Professor  
Georgetown University

September 26, 2024

## What is **georeferencing**?

- assignment of geographic objects to geographic locations
- relation of map image to system of geographic coordinates on the ground



Figure 1: This is georeferencing

## What is **vectorization**?

- generation of vector features from georeferenced raster images
- opposite is called rasterization (which is much easier)

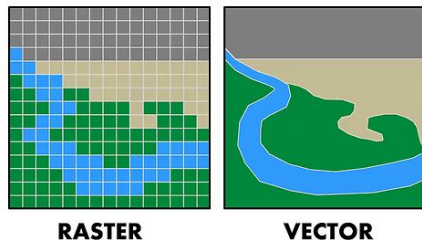


Figure 2: This is vectorization

# Georeferencing



## Why georeference?

- maps contain data you can't find anywhere else
- georeferencing allows us to
  - extract and preserve these data
  - combine map with other types of geospatial data
  - use these data to answer social, economic and political questions



Figure 3: NKVD jurisdictional borders

# Overview

## What is involved?

1. Obtain digital image of map (e.g. scan, web)
2. Select ground control points
3. Transform map to align with chosen coordinate system



Figure 4: Step 1

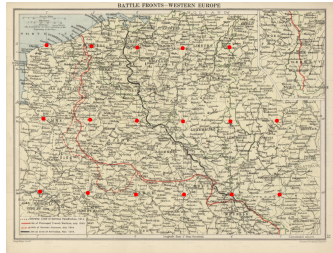


Figure 5: Step 2



Figure 6: Step 3

## What can we georeference?

- historical maps
- satellite and aerial photography
- administrative and military maps
- interesting maps you found online

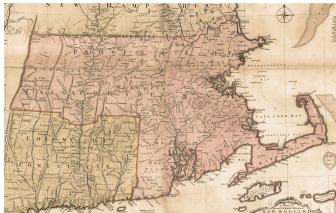


Figure 7: Massachusetts, 1755

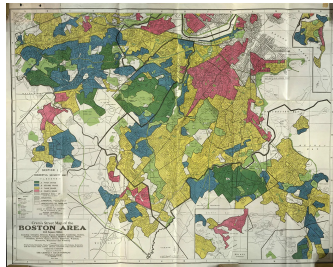


Figure 8: Boston, 1936



Figure 9: Boston, 1955

## Challenges

- projection often unknown
- scale/resolution may be coarse
- distances/angles/shapes may be inaccurate (esp. in older maps)
- impossible to perfectly align historical maps with modern coordinate systems

GEOGRAPHICAL, STATISTICAL, AND HISTORICAL MAP OF MICHIGAN TERRITORY.

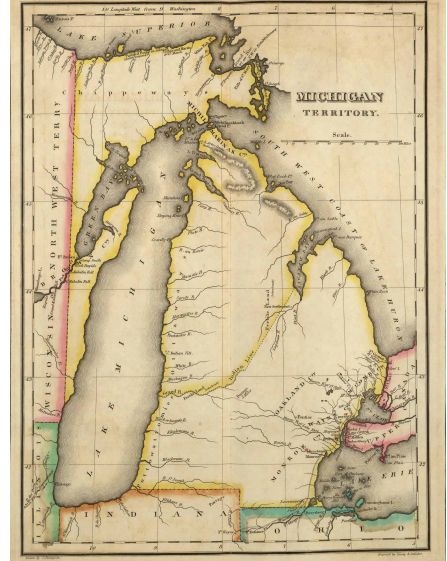


Figure 10: Close, but not quite

## Examples of GCPs

- intersections of graticule lines (most reliable)
- landmarks of known location (e.g. buildings, crossroads, hills, cities)
- distinctive geographic features (e.g. coastal features, curves in rivers, borders)



Figure 11: Graticules

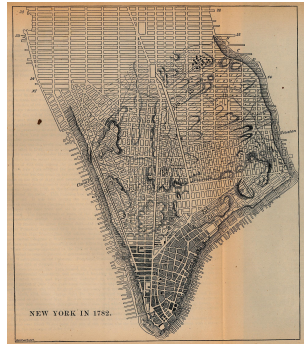


Figure 12: Intersections



Figure 13: Landmarks

# Transformations

## Transformation ("rubber sheeting")

- shift and warp the raster to spatially correct locations in original image
- apply mathematical algorithm to match source control points with target control points
- process changes distances, appearance of lines and shapes

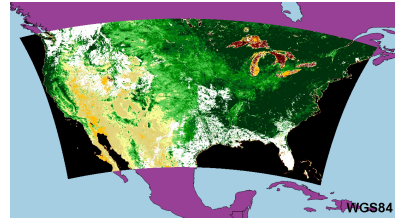


Figure 14: Transformed raster

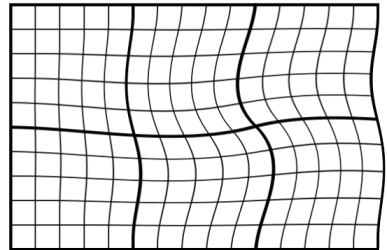


Figure 15: Rubber sheeting



## Challenges

- transformation distorts original map image
- results sensitive to choice of transformation algorithm
- output only reliable within area confined by GCPs

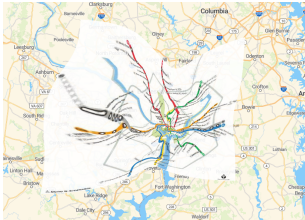


Figure 16: Distortion

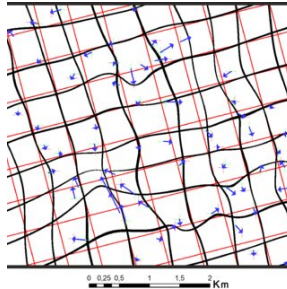


Figure 17: Algorithm

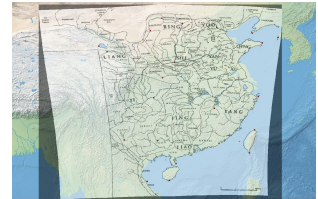


Figure 18: Range

## Polynomial transformations

- uses a polynomial built on control points and a least-squares fitting algorithm
- optimized for global accuracy, not always local accuracy

### Order

- $x_0 + x^1 + x^2 + x^3 + \dots + x^k$ , where  $k$  is *order of polynomial*
- higher order  $\rightarrow$  able to correct more complex distortion
- but rarely need transformations  $> 3$ rd order

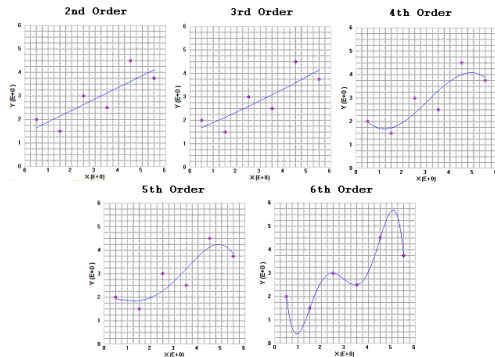


Figure 19: Higher order  $\rightarrow$  closer fit

## 0. Zero-order polynomial

- shifts raster location
- used when raster is already georeferenced, but slightly mis-aligned

## 1. First-order (Affine) polynomial

- shift/scale/rotate a raster
- straight lines on input raster will remain straight
- requires  $\geq 3$  control points

### Graphs of Polynomial Functions:

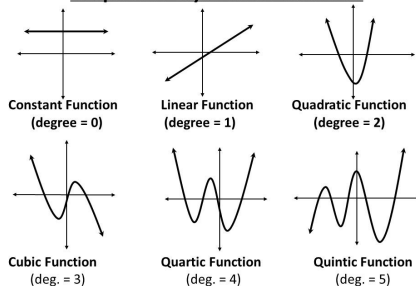


Figure 20: Examples in one dimension

## 2. Second-order polynomial

- applies quadratic formula to calculate raster cell position
- straight lines on input raster will be warped
- requires  $\geq 6$  control points

## 3. Third-order polynomial

- applies cubed formula
- straight lines, margins on input raster will be warped
- corrects more complex distortions
- requires  $\geq 10$  control points

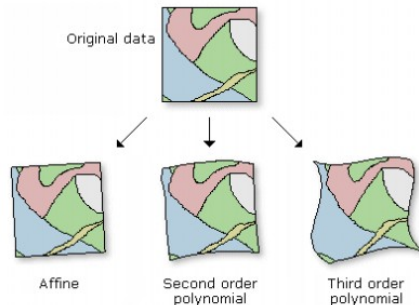


Figure 21: Examples in two dimensions

#### 4. Projective transformation

- linear rotation, translation
- warps lines to keep them straight
- useful for oblique imagery, scanned maps
- requires  $\geq 4$  control points

#### 5. Spline transformation

- uses piecewise polynomial that maintains continuity between adjacent polynomials
- optimized for local accuracy, not always global accuracy
- minimal local error
- requires  $\geq 10$  control points

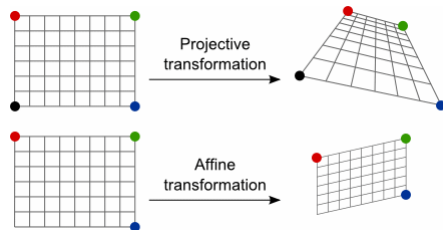


Figure 22: Projective vs. affine

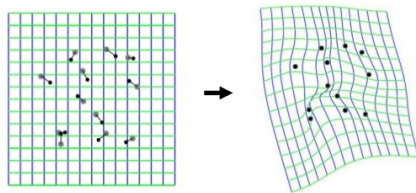


Figure 23: Spline

# Vectorization

## Why vectorize?

- vector is standard data structure for quantities of interest to public policy and social science (e.g. events, roads, administrative zones)
- smaller data size (usually)
- objects can have multiple attributes
- allows more sophisticated spatial analyses
- preserves quality at all scales

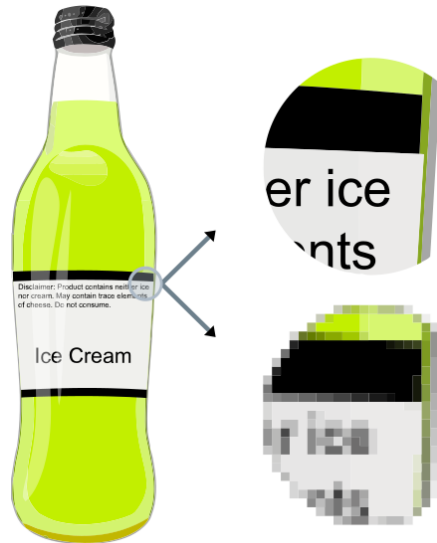


Figure 24: Enhance!

# Options



## *Two ways to identify vector features*

1. Image tracing (manual or automated)
2. Computer vision (machine learning)

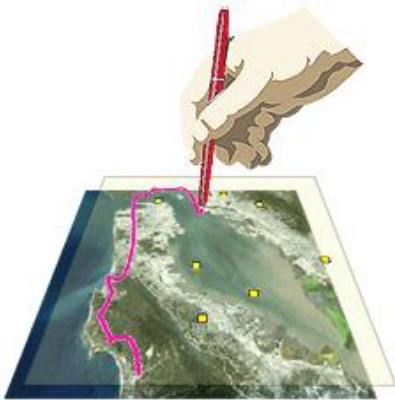


Figure 25: Trace

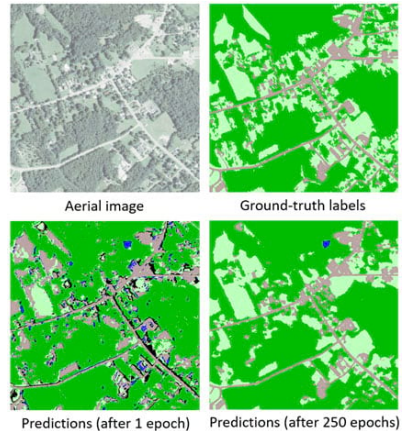


Figure 26: Learn

## 1. Image tracing

- drawing over a raster image with vectors
- *manual tracing*: tracing over the image by hand (using mouse or stylus)
- *automated tracing*: use computer algorithm to detect features, redraw them as vector points, lines, polygons

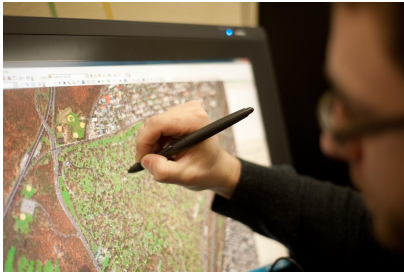


Figure 27: Manual

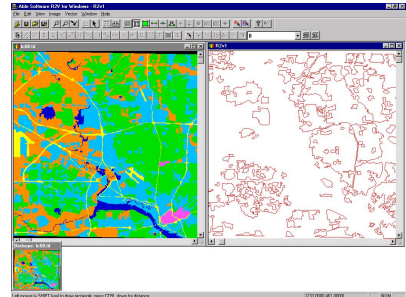


Figure 28: Automated

## Manual tracing

### *Advantages*

- can work with images of any quality
- better understanding of context/meaning
- produces fewer artifacts/superfluous features

### *Disadvantages*

- slow, inefficient
- relatively imprecise/inconsistent
- subject to laziness/fatigue

## Automated tracing

### *Advantages*

- fast and efficient
- output is consistent, replicable

### *Disadvantages*

- more sensitive to image quality
- can require extensive pre-processing/cleanup
- works best with fewer colors

## 2. Computer vision / deep learning

- automated feature detection, extraction
- system “learns” what is/isn’t a feature through training data (e.g. examples of points, lines, polygons in raster images)
- cross-validation of results to improve predictive fit
- examples:
  - convolutional neural networks
  - recurrent neural networks
  - long short term memory models
  - transformer models

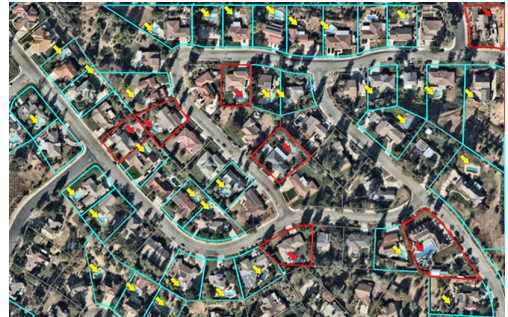


Figure 29: Which houses have pools

## Machine learning

### Advantages

- fast and efficient
- well-suited for large-scale tasks, where fixed rules lead to systematic errors

### Disadvantages

- requires large volume of training data
- requires high-performance computing infrastructure, programming expertise
- same pre-/post-processing issues as automated tracing
- not (yet) available in standard GIS software

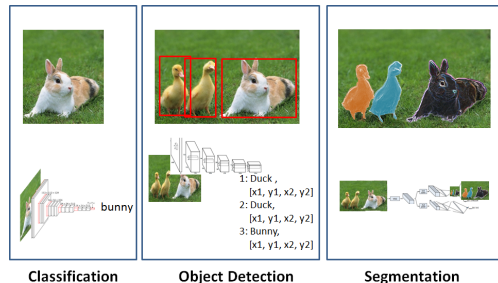


Figure 30: Computer vision tasks

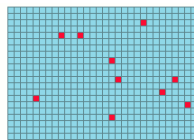
## Sources of error

## Automated vectorization

1. Raster-to-point
  - all non-zero/non-null cells become points
2. Raster-to-line
  - trace positions of non-zero/non-null pixels to identify polyline features
3. Raster-to-polygon
  - use groups of connected pixels with identical values to find areas of a raster
  - determine intersection points of area boundaries, generate lines



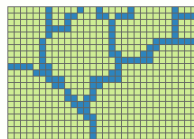
Vector Point Features



Raster Point Features



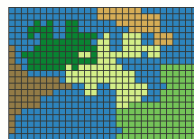
Vector Line Features



Raster Line Features



Vector Polygon Features



Raster Polygon Features

Figure 31: Usually not so seamless

## Types of vectorization errors

### 1. *False positives*

- identification of features where none exist  
(generates small/superfluous vertices that must be removed)

### 2. *False negatives*

- failure to identify features where they exit  
(creates gaps, incomplete features)

## How to reduce errors

### 1. *Pre-processing*

- remove noise, unnecessary elements, colors

### 2. *Post-processing*

- remove superfluous features, fill gaps, improve appearance

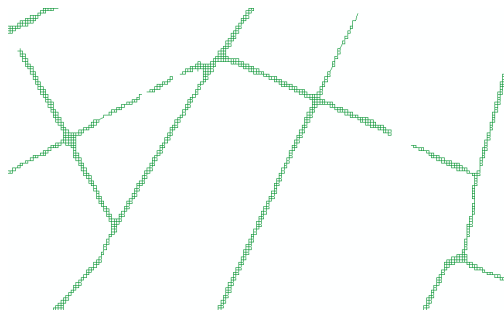


Figure 32: Vectorized “roads”



## 1. Pre-processing of rasters

- *reclassification*: conversion from color/greyscale to binary
- *thinning*: reduce thickness of features to a single, connected lines of pixels

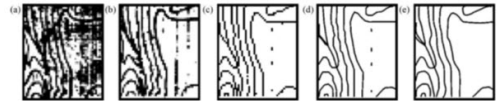


Figure 33: Thinning example

## 2. Post-processing of vector features

- *collapsing*: simplification by removal of spurious nodes, segments, closing of gaps
- *smoothing*: generalization/averaging to smooth pixelated appearance of output vectors

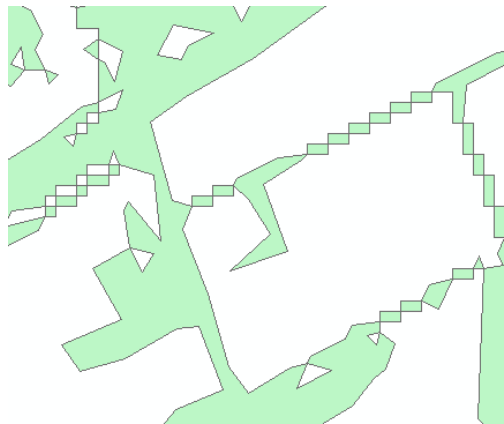


Figure 34: Lots to clean up