

Uncertainty

GEOG 5201 – Spring 2022

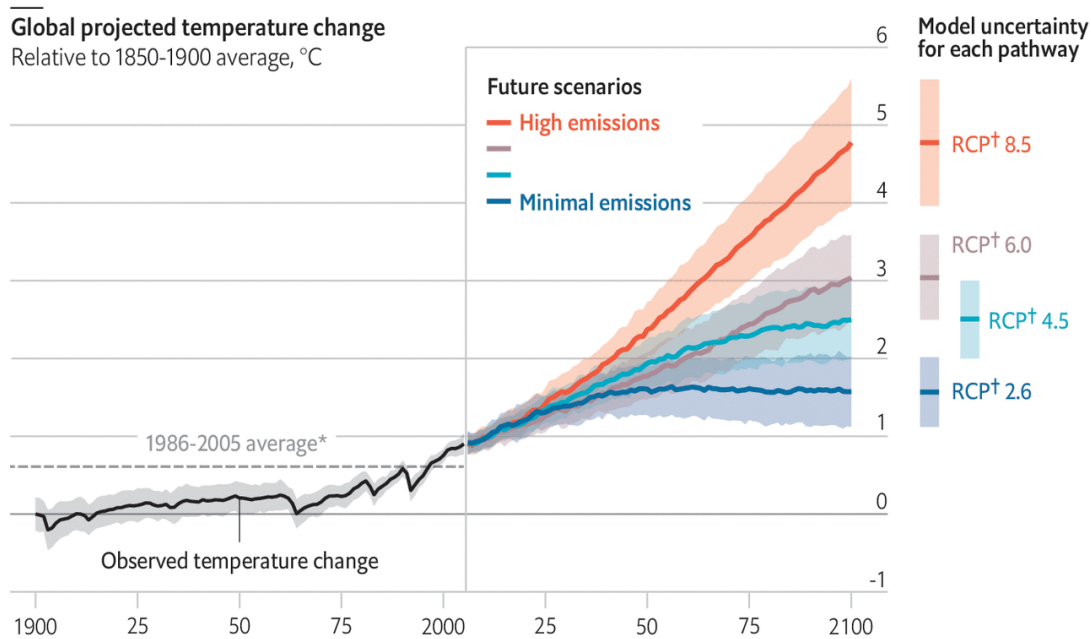
Outline

- Basic elements of uncertainty
 - Sources
 - Uncertainty in the raw data
 - Uncertainty in processing data
 - Uncertainty in the visualization
 - Concepts
 - Uncertainty versus error
 - Uncertainty, reliability, and quality
 - Why uncertainty matters in map making
 - Ethical necessity
 - Decision-making

Uncertainty in Maps

- We often think of maps as truthful representations of reality
 - This may not be correct, because the *truth* is often unknown to us, and the data visualized is very likely to deviate from the truth

→ **Climate models can guide policy even if they are not precise**



Source: IPCC AR5, adjusted to an 1850-1900 baseline
The Economist

*Uncertainties calibrated to 1986-2005, as shown †Representative Concentration Pathway

Sources of Uncertainty

- Uncertainty is commonly used in the literature to describe **the potential variation in values of an attribute** at a spatial location
- Uncertainty emerges during multiple stages of map-making
 - Collection uncertainty -- uncertainty in the raw data
 - Derived uncertainty -- uncertainty in processing data
 - Visualization uncertainty -- uncertainty in the visualization

Sources of Uncertainty: Uncertainty in the Raw Data

- Example: sampling errors in the American Community Survey (ACS)
 - Because the ACS is based on a sample, rather than all housing units and people, ACS estimates have a degree of uncertainty associated with them, known as sampling error
 - The U.S. Census Bureau provides a margin of error (MOE) for each published ACS estimate to help understand the uncertainty

Table 71. Sample Estimates and Margins of Error in American FactFinder: 2015

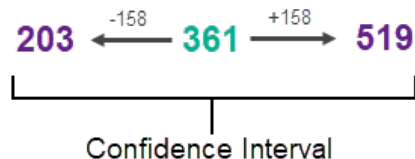
Subject	Colorado			
	Estimate	Margin of Error	Percent	Percent Margin of Error
HOUSEHOLDS BY TYPE				
Total households	2,074,735	+/-7,548	2,074,735	(X)
Family households (families)	1,331,861	+/-11,075	64.2%	+/-0.5
With own children of the householder under 18 years	597,501	+/-8,075	28.8%	+/-0.4
Married-couple family	1,038,040	+/-9,389	50.0%	+/-0.4
With own children of the householder under 18 years	435,028	+/-7,158	21.0%	+/-0.3
Male householder, no wife present, family	93,024	+/-5,026	4.5%	+/-0.2
With own children of the householder under 18 years	48,969	+/-4,185	2.4%	+/-0.2
Female householder, no husband present, family	200,797	+/-6,582	9.7%	+/-0.3
With own children of the householder under 18 years	113,504	+/-4,699	5.5%	+/-0.2
Nonfamily households	742,874	+/-10,127	35.8%	+/-0.5
Householder living alone	564,751	+/-10,127	27.2%	+/-0.5
65 years and over	182,959	+/-5,249	8.8%	+/-0.3
Households with one or more people under 18 years	657,324	+/-8,777	31.7%	+/-0.4
Households with one or more people 65 years and over	497,903	+/-4,124	24.0%	+/-0.2

Source: U.S. Census Bureau, American FactFinder, Table DP02: Selected Social Characteristics in the United States.

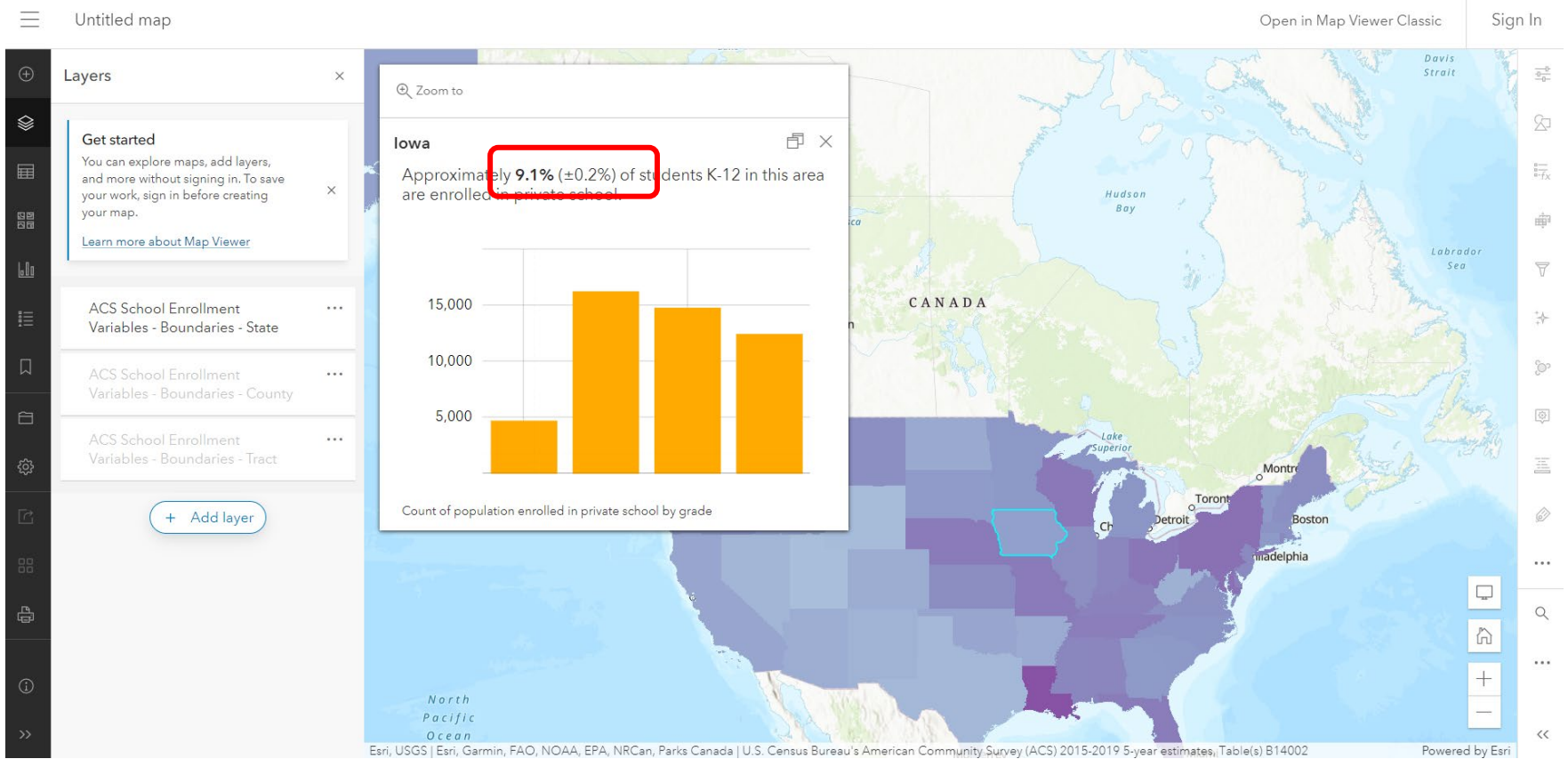
Sources of Uncertainty: Uncertainty in the Raw Data

- How to interpret MOEs?
 - At a given confidence level, the MOE, combined with the ACS estimate, give users a **range of values** within which the actual, “real-world” value is likely to fall
 - 90% confidence level is the Census Bureau Standard
 - This range is called a confidence interval
 - Example: percentage of family households is $64.2\% \pm 0.5\%$
 - With 90% confidence, the range 63.7% to 64.7% covers the actual percentage of family households
 - This means that if the survey is conducted 100 times, 90 times the percentage of family households would be within 63.7% and 64.7%, and 10 times the percentage of family households would be either higher than 64.7% or lower than 63.7%

Estimate: **361**
Margin of Error: **158**



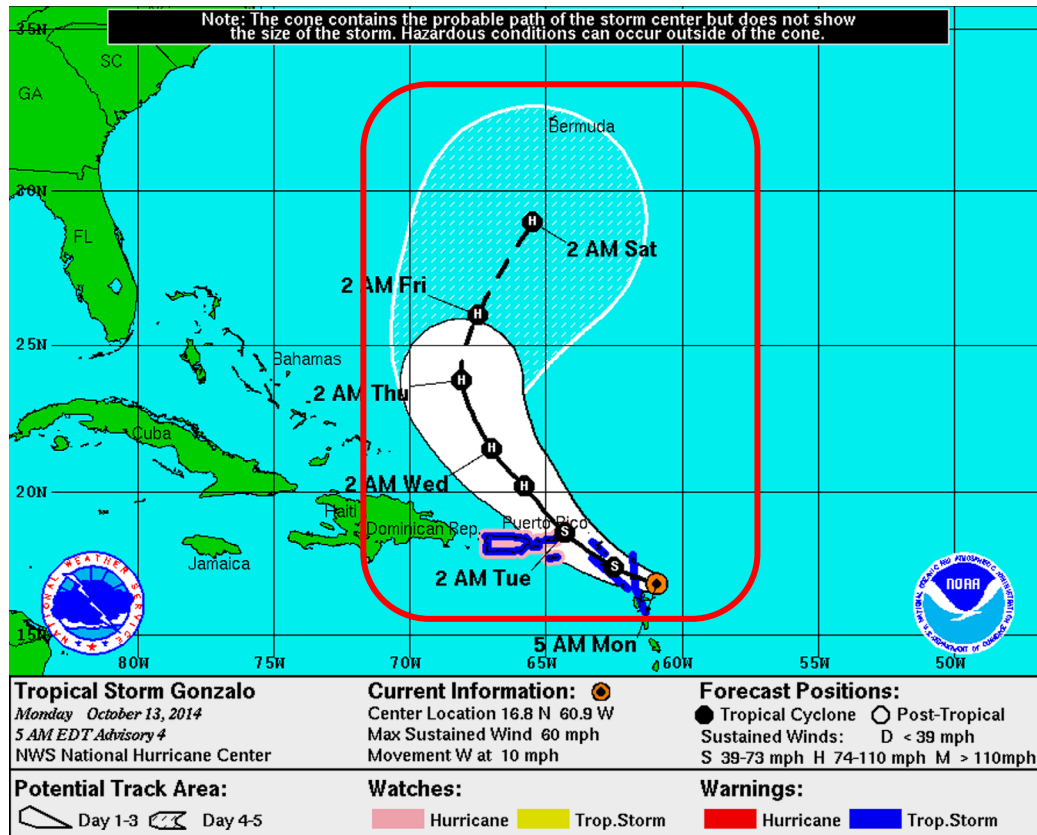
More on ACS uncertainty: [ACS data in ArcGIS Living Atlas](#)



A map showing both estimates of public vs. private school enrollment and associated MOEs

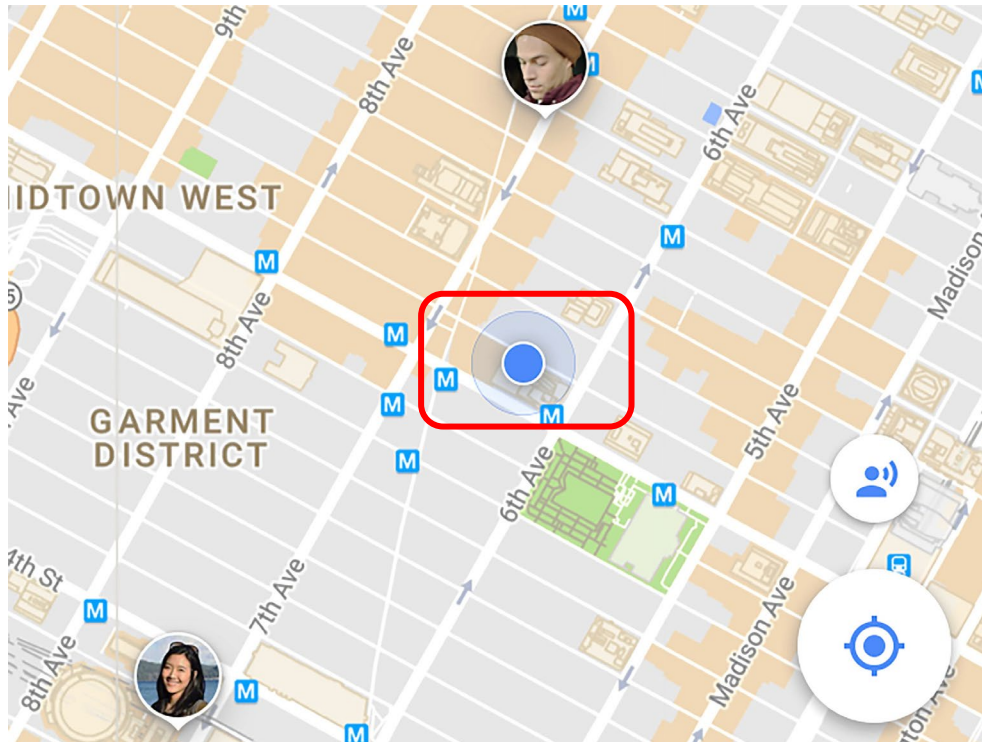
Sources of Uncertainty: Uncertainty in the Raw Data

- Example: “cone of uncertainty” in hurricane projections
 - The cone contains the probable path of the storm center



Sources of Uncertainty: Uncertainty in the Raw Data

- Example: circle of uncertainty on Google Maps
 - [Google's documentation](#) states that “. . .at times, you may see the dot surrounded by a light blue circle. This indicates that there is some uncertainty about your location.” and “. . .you may be anywhere within it”

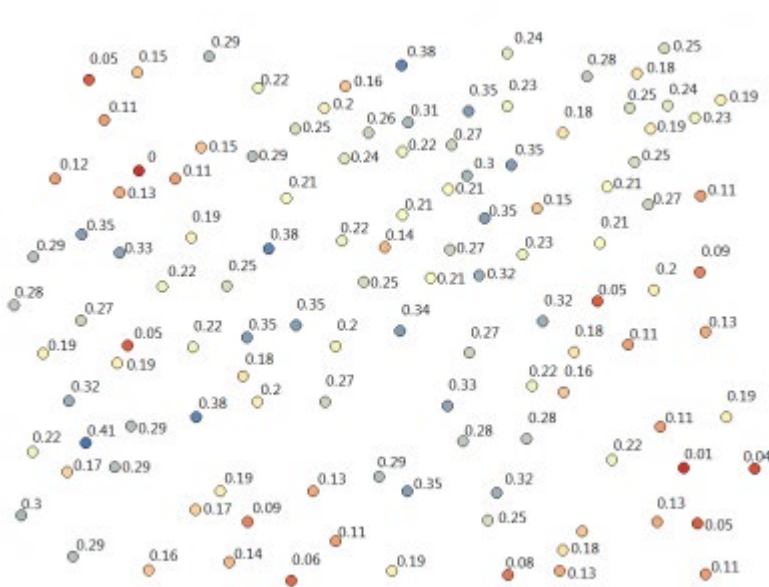


Question 3-1-1

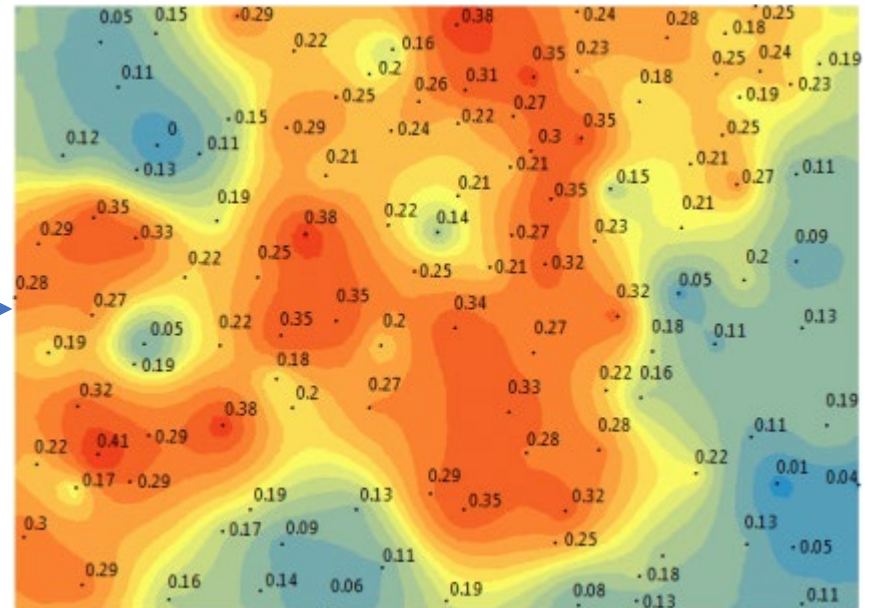
Recall any map you have previously seen or created, either in this class or elsewhere. Explain how uncertainty in the raw data can arise.

Sources of Uncertainty: Uncertainty in Processing Data

- Example: spatial interpolation
 - Use known point values to estimate unknown point values



Known point values (raw data)

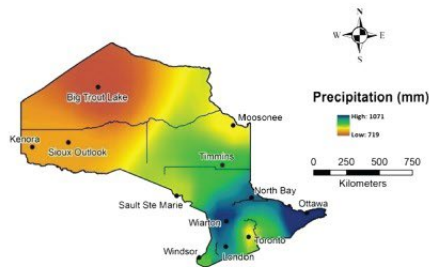


Interpolated surface (processed data)

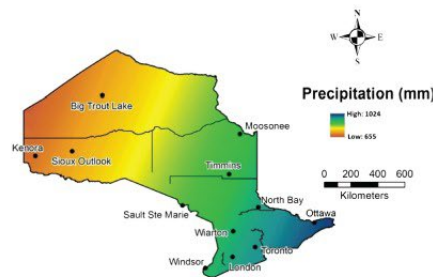
Sources of Uncertainty: Uncertainty in Processing Data

- There are a variety of algorithms, each producing a potentially different set of interpolated values
- For any particular location on a map, we can consider **the set of interpolated values** to be the data's **uncertainty**

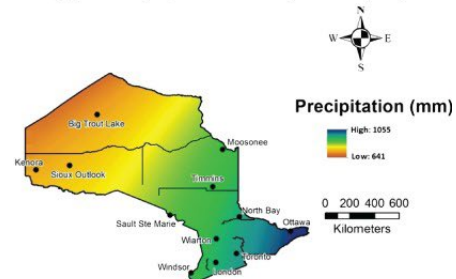
(a) Inverse distance weighting (IDW)



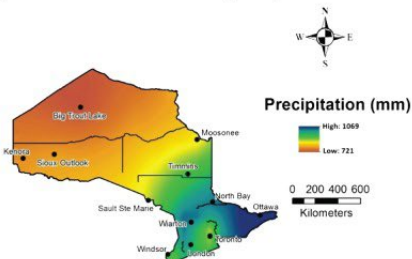
(b) Global polynomial interpolation (GPI)



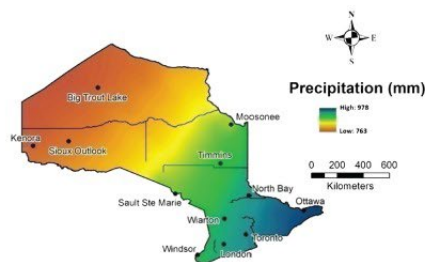
(c) Local polynomial interpolation (LPI)



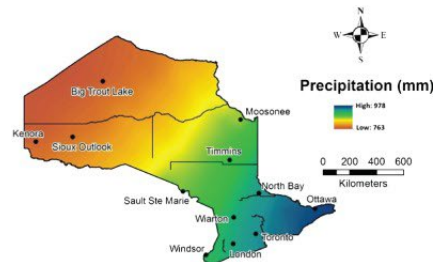
(d) Radial basis functions (RBF)



(e) Ordinary kriging (OK)

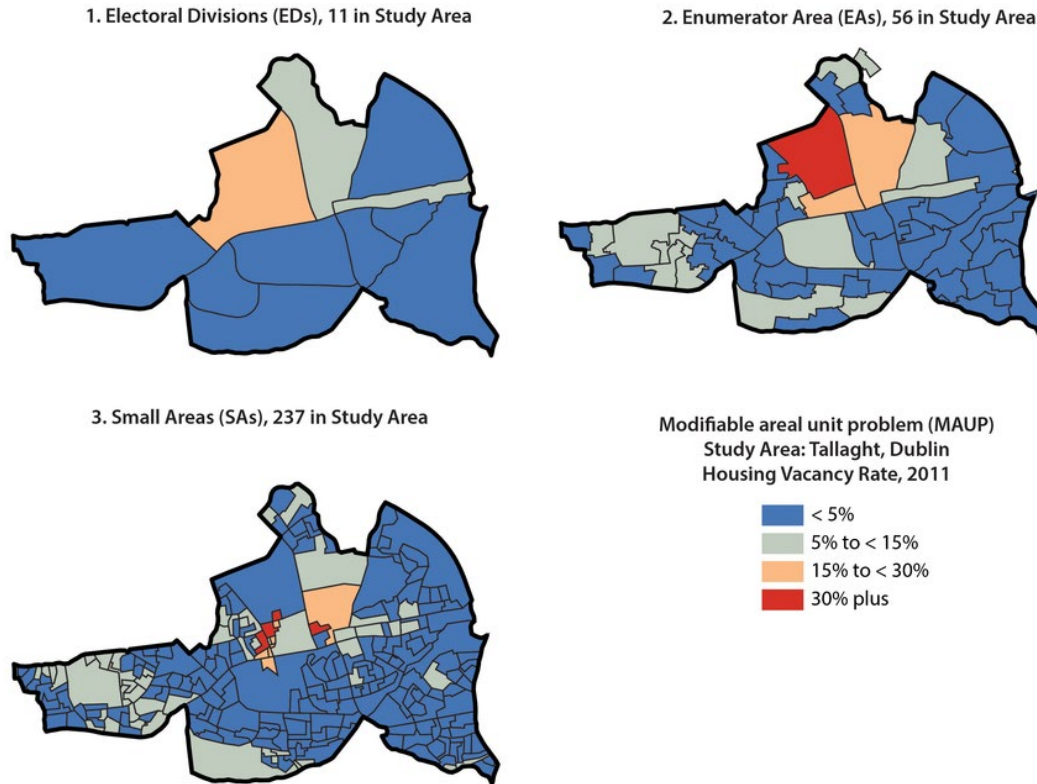


(f) Universal kriging (UK)



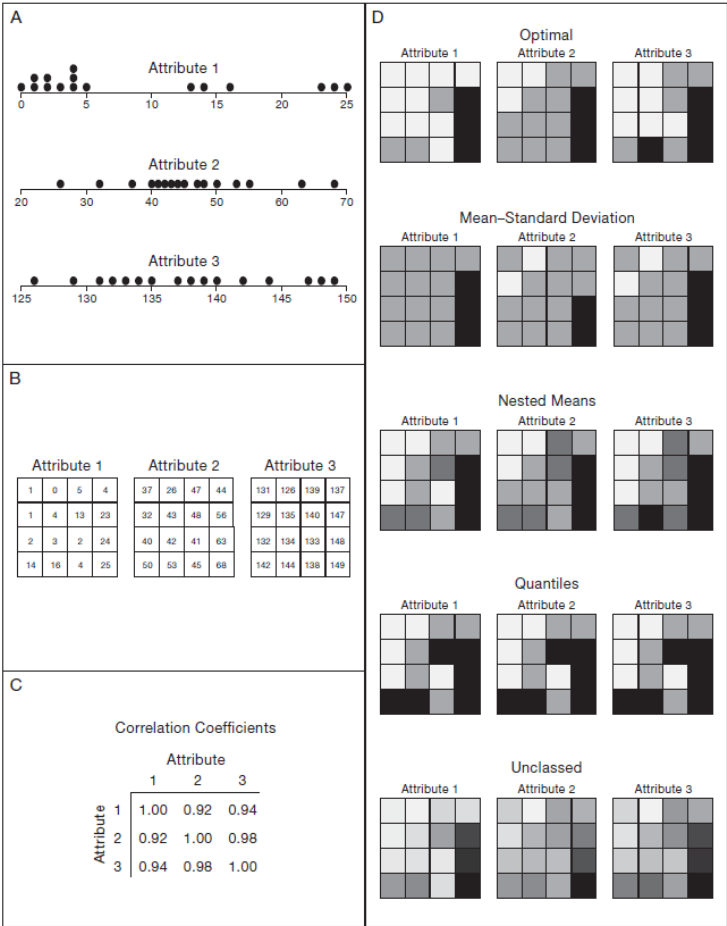
Sources of Uncertainty: Uncertainty in Processing Data

- Example: the modifiable areal unit problem (MAUP)
 - The same basic data yield different results when aggregated in different ways



Sources of Uncertainty: Uncertainty in the Visualization

- Example: different classifications lead to different visualization results



Concepts of Uncertainty: Uncertainty versus Error

- Recall that uncertainty is the potential variation in values of an attribute at a spatial location
 - True values typically unknown
- Error is the **difference between the measured value and the true value of an attribute** at a spatial location
 - True values are known objectively
- Uncertainty covers a broader range of doubt than error alone

Concepts of Uncertainty: Uncertainty, Reliability, and Quality

- The terms “reliability” and “quality” are also used
 - *Uncertainty* is equated with *unreliability* and *poor quality*
- The [U.S. Federal Information Processing Standard 173](#) lists 5 categories for assessing data quality
 - Lineage: history of data, including sources, data processing and transformations
 - **Position accuracy**: location accuracy of geographic features (recall the hurricane projection and Google Maps examples)
 - **Attribute accuracy**: accuracy of features found at particular locations (recall the ACS example)
 - Logical consistency: extent to which objects within the dataset agree; topological correctness
 - Completeness: extent to which data is comprehensive

Question 3-1-2

Explain the difference between uncertainty and error using any of the examples above (p.5 – p.14).

Why Uncertainty Matters in Map Making: Ethical Necessity

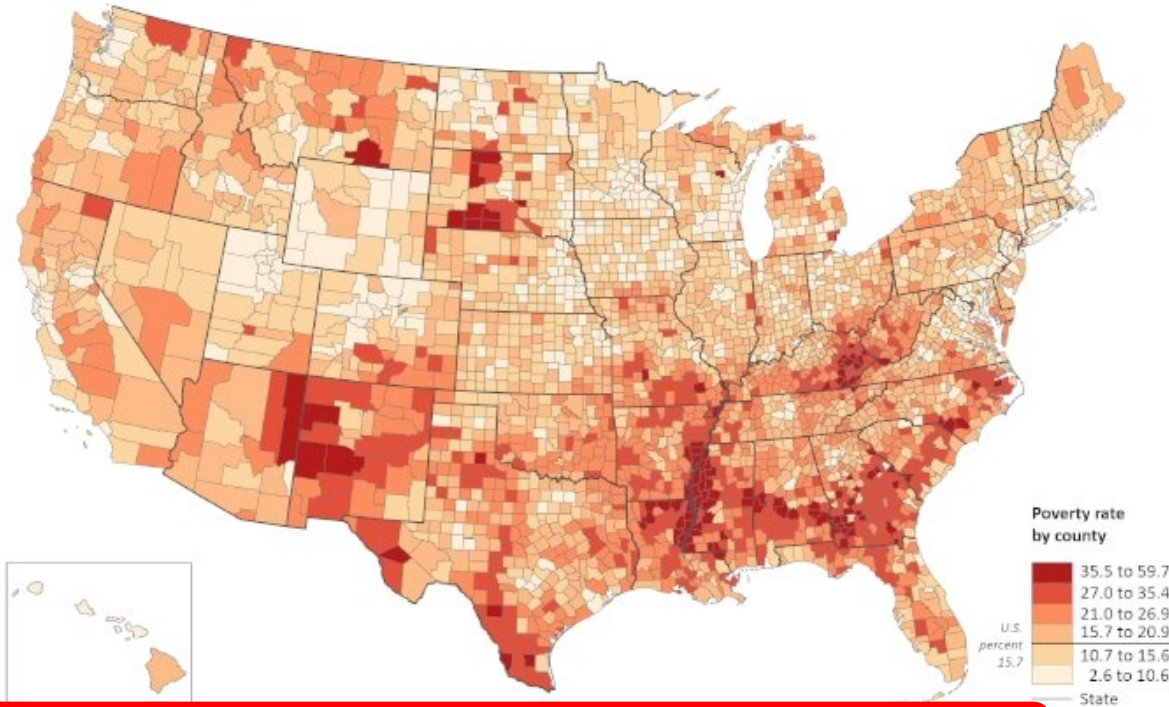
- To indicate the (unavoidable) gap between reality and representation
 - Users need to understand that GIS data and analyses are not necessarily accurate and reliable
 - GIScience researchers, educators and students should know that there is a multitude of reasons that our representations are incomplete and uncertain
- Withholding the uncertainty information from map readers would be misleading
- Good science includes statements of accuracy, and the reliability of results must be understood and communicated



Map of poverty rate of the population under 18 by the U.S. Census Bureau *without uncertainty information presented.*



Poverty Rate of the Population Under 18
by County: 2020

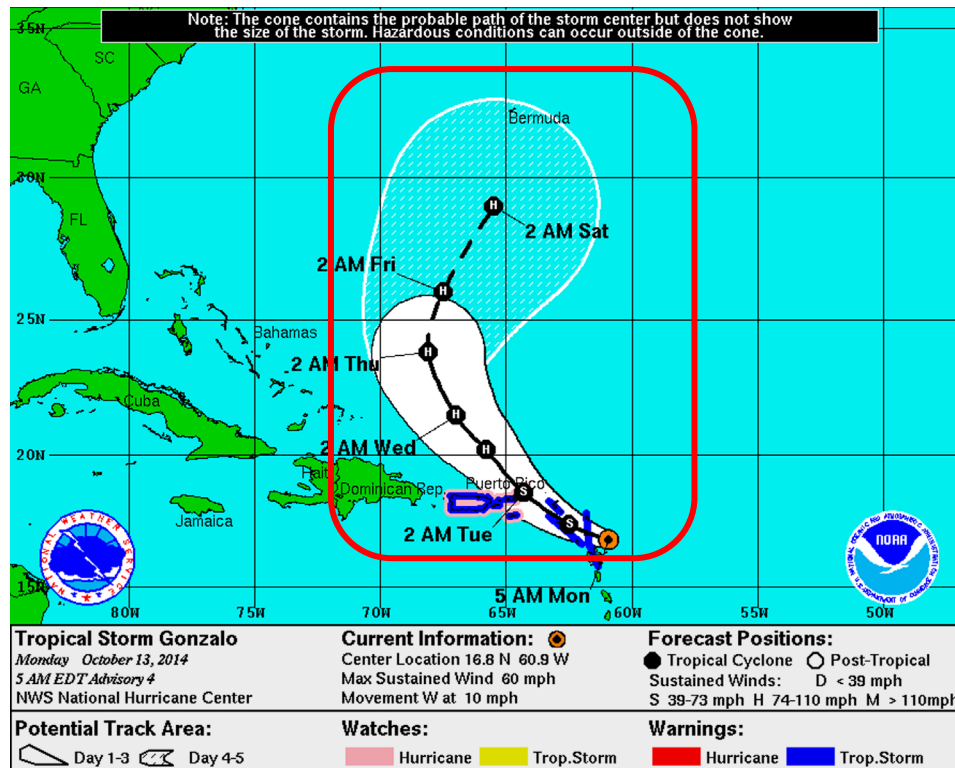


Note: The data provided are indirect estimates produced by statistical model-based methods using sample survey, decennial census, and administrative data sources. The estimates contain error stemming from model error, sampling error, and nonsampling error.
Source: U.S. Census Bureau, Small Area Income and Poverty Estimates (SAIPE) Program, December 2021.

The data provided are indirect estimates by statistical model-based methods using sample survey, but users may interpret them as population enumeration results

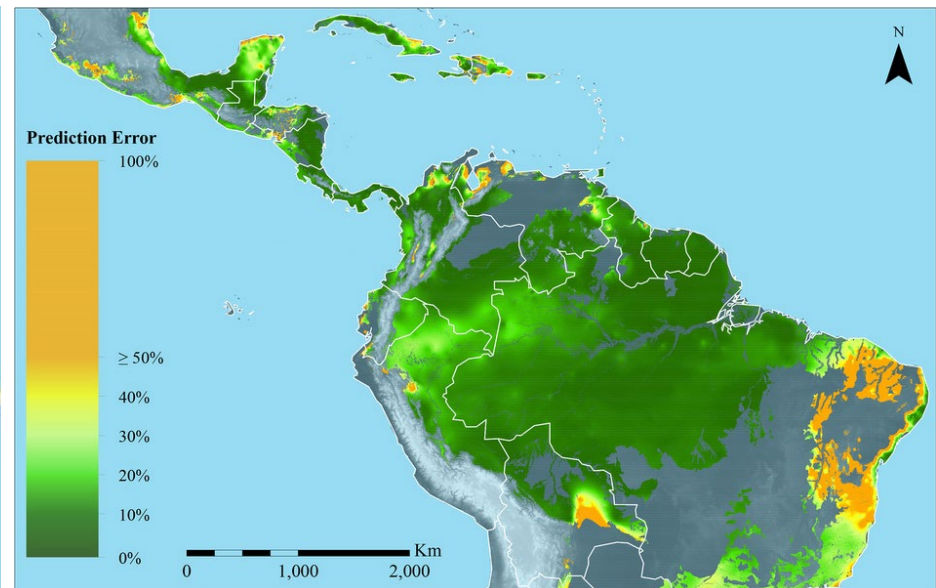
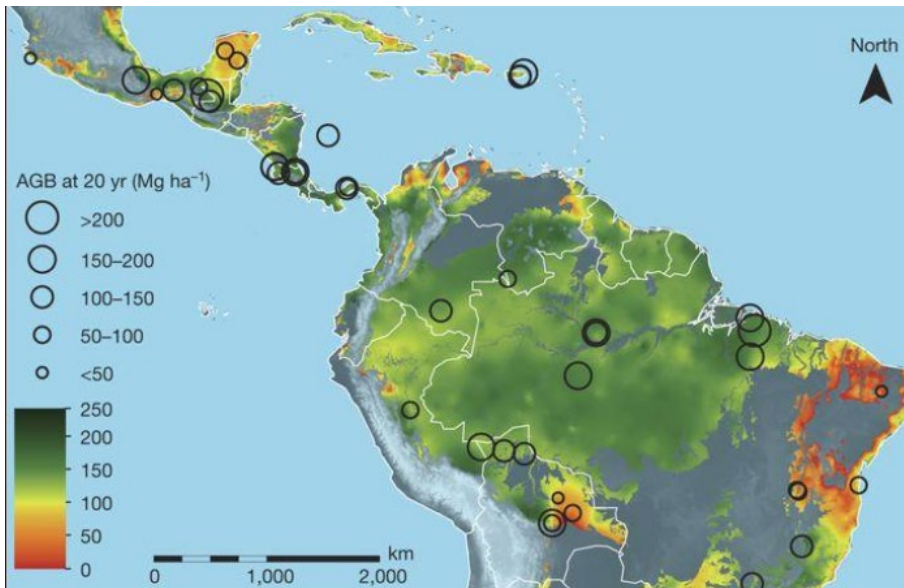
Why Uncertainty Matters in Map Making: Decision-Making

- Uncertainty plays an important role in decision-making
 - Example: the “cone of uncertainty” in maps of predictive hurricane paths often play an important role in decisions made by residents of storm-affected areas



Why Uncertainty Matters in Map Making: Decision-Making

- Professional fields dealing with the natural environment routinely require their practitioners to make decisions using data that can include a range of uncertainties
 - Example: Uncertainty map of potential biomass recovery of Neotropical secondary forests



Question 3-1-3

Can you think of a map where the uncertainty information could be useful? Why do you think uncertainty matters in that map?