



CORE
Skills

Delivering Data Science
In Resources & Energy

Special Data Types: Spatial Data

Day 13

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Data Mettle

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Schedule

AWST	AEST	Agenda	Facilitator
07:30	09:30	Q&A, Issues & Announcements	
07:45	09:45	<u>Introduction to spatial data</u>	Ying
09:00	11:00	<i>Morning Tea</i>	
09:15	11:15	Project Share (RTA)	Georgia Hamilton
10:00	12:00	<u>Using spatial data</u>	Ying
11:00	13:00	<i>Lunch</i>	
11:45	13:45	How to pitch a data science project	Jeremy
13:15	15:15	<i>Afternoon Tea</i>	
13:30	15:30	Project & pitch development	All
14:45	16:45	<u>Closeout</u>	



Program Timeline



Q&A, Issues & Announcements

Before we Get Started

- Resources & Tasks
- Reflections, questions and Issues: Teams
- The afternoon is time to work on your **project pitch** and any questions you might have



GitHub Content for Today using Binder

[github.com / core-skills / 13-spatial-data](https://github.com/core-skills/13-spatial-data)

The screenshot shows a GitHub repository page for the 'core-skills/13-spatial-data' repository. The repository has 3 branches (master, gh-pages, releases) and 8 tags. The master branch is selected. There are 40 commits in total, with the most recent commit being a merge from 'gh-pages' to 'master' 2 minutes ago. The commits are listed as follows:

File	Commit Message	Time Ago
data	Added notebook for data exploration exercise	9 months ago
notebooks	Update filer & unified notebook for first exercise	21 minutes ago
program	Updated environment file with minimal package versions and per notebooks	1 hour ago
gh-pages	Update Function Notebooks, gh-pages for processed data	9 months ago
LICENSE	Update License Year	9 months ago
README.md	Point Binder Link to Jupyter	9 months ago
environment.yml	Updated environment file with minimal package versions and per notebooks	1 hour ago

At the bottom of the repository page, there is a section titled 'CORE Skills Program - Day 3 - Zero to Data Science' with a 'launch binder' button.



GitHub – Program Notes

[github.com / core-skills / 13-spatial-data / program / 00_overview.md](https://github.com/core-skills/13-spatial-data/program/00_overview.md)

Overview

[Overview](#) | [Data Culture](#) | [From Here to There](#) | [Data Projects](#) | [Data Exploration](#) | [Closeout](#)

Aim

Provide an overview of a 'typical' data science workflow.

Learning Outcomes

1. To appreciate what data science is
2. To appreciate the fields data science spans
3. Understand the stages of a data science project and define a mental model of it
4. Analyse the opportunity and potential value of data science in your organisation

Schedule

AWST	AEST	Agenda
07:30 - 07:45	09:30 - 09:45	Q&A, Issues & Announcements
07:45 - 09:15	09:45 - 11:15	Creating a Data Culture
09:15 - 09:30	11:15 - 11:30	<i>Morning Tea</i>
09:30 - 11:00	11:30 - 13:00	Getting From Here to There
11:00 - 11:15	12:00 - 12:15	



Binder Backup

Screenshot of a GitHub repository showing a commit history and a "launch binder" button.

The repository has a single commit from "morganejullien" titled "Merge pull request #1 from gh-pages". The commit was pushed 2 minutes ago and includes 42 commits. The commit history shows the following changes:

- data: Added notebook for data exploration exercises 0 months ago
- notebooks: Update filled & unfilled notebooks for first exercises 21 minutes ago
- program: Updated environment file with minimal package versions and per-notebook... 1 hour ago
- gh-pages: Update Function! Notebooks, gh-pages for processed data 0 months ago
- LICENSE: Update License file 0 months ago
- README.md: Point Binder link to data 0 hours ago
- environment.yml: Updated environment file with minimal package versions and per-notebook... 1 hour ago

Below the commit history is a file named "README.md".

CORE Skills Program - Day 3 - Zero to Data Science

[launch binder](#)



Aims & Learning Outcomes

Day 13

Aims

- Extract geospatial features for data science applications.
- Present the practical strategies for using geospatial data.

Learning Outcomes

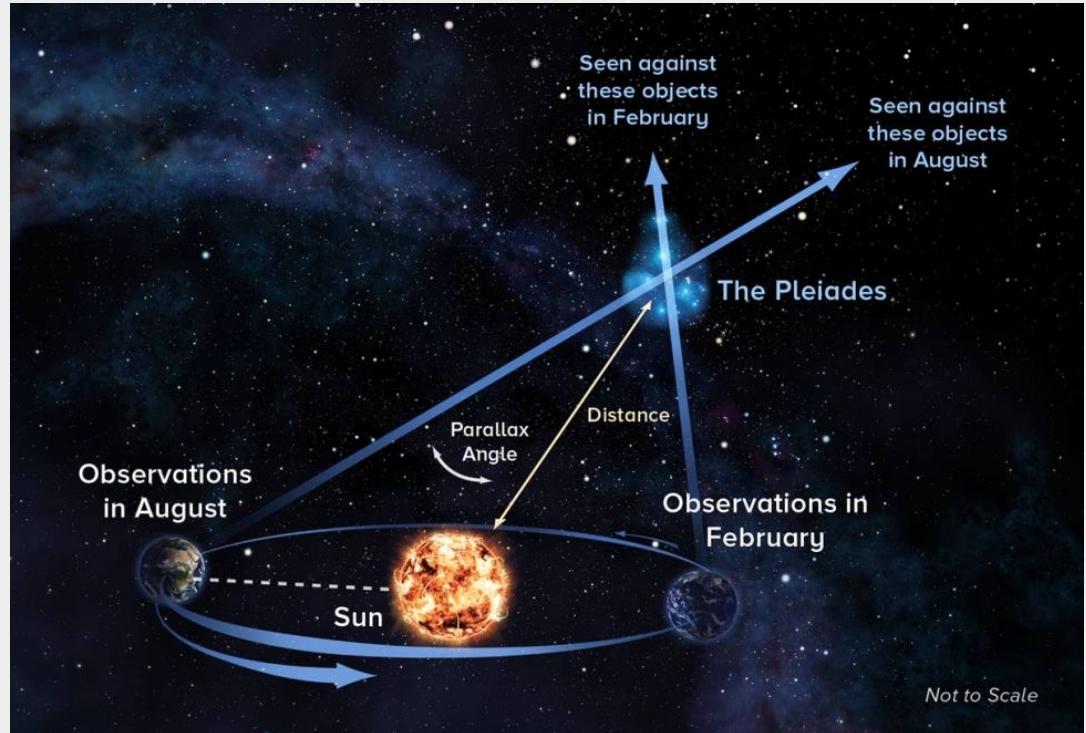
- Understand vector and raster geospatial formats with feature extraction.
- Understand projections and geocoding.
- Understand how to manipulate spatial data sets.

Introduction to spatial data

Fundamental concepts for spatial analysis

Key concepts

- Space
 - Location
 - Distance
-
- Each can be described as:
 - Absolute
 - Relative
 - Cognitive
 - Nominal



Fundamental concepts for spatial analysis

Space

- Absolute space
 - mathematical space, things are given an (x, y, z) tuple and positions are unambiguous.
- Relative space
 - Topological space, can be thought of as relative space. Used to represent connectivity between features of the world
- Cognitive space
 - reflects people's beliefs, experiences and perceptions about places.

Fundamental concepts for spatial analysis

Location

- Absolute
 - an unambiguous descriptor of location, expressed as a coordinate. Can't be confused with any other location.
 - Example: Latitude and longitude.
- Relative
 - Site (physical attributes of location) and situation (location relative to other places).
 - Example: Perth is located far from other cities, near raw materials etc. Or 191 St Georges Terrace
- Cognitive
 - Compiled from personal knowledge, experiences, and impressions
 - Example: traditional knowledge of waterholes as a resource, then a stock route, then a 4WD track. Same physical location but different perception.
- Nominal
 - *Where were you when...?* Linking space and time.
 - Example: Where were you when you heard about 9/11?

Fundamental concepts for spatial analysis

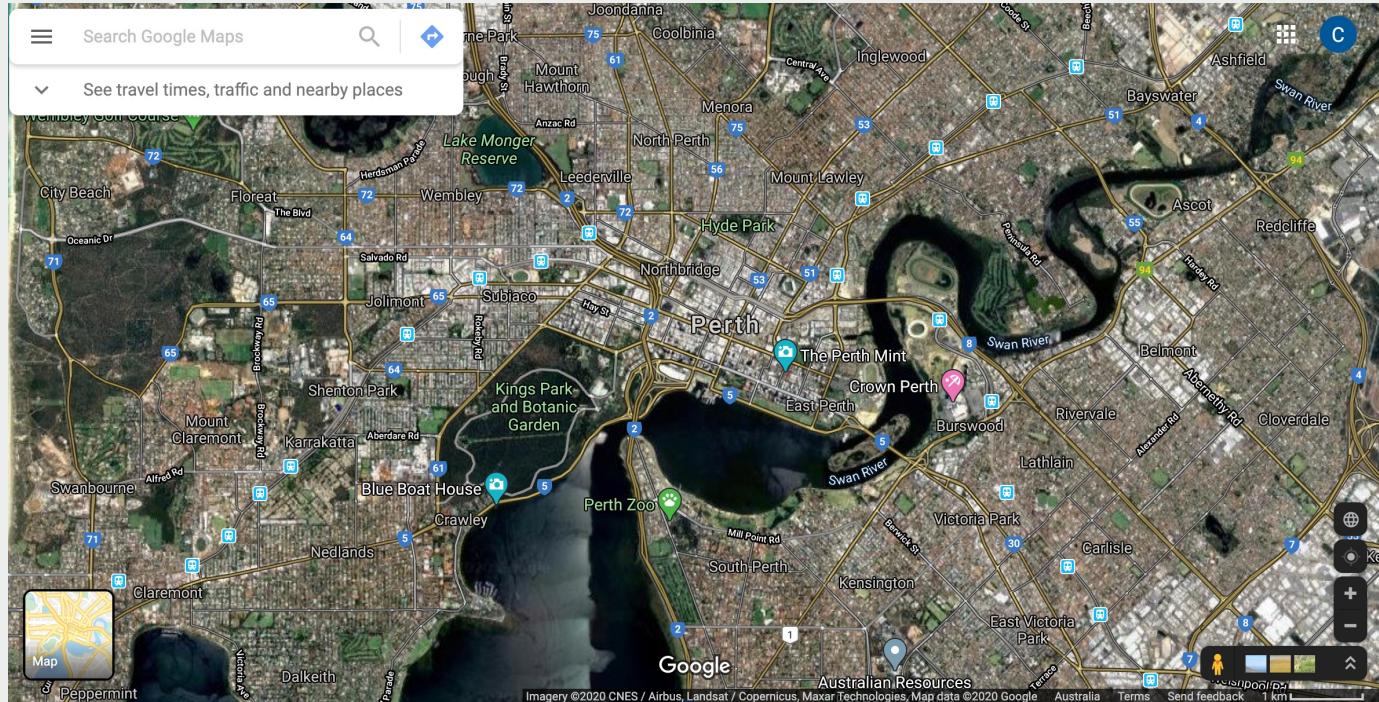
Distance

- Absolute
 - a physical unit of measure Example: kilometers between two assets
- Relative
 - Generally calculated using time, effort or cost (will differ from absolute – what happens when you go over a mountain range?). Example: voting maps sized by number of seats vs area.
- Cognitive
 - Individual perception of distance. Example: driving in NZ vs driving in Australia

Geospatial data types

Objects and attributes

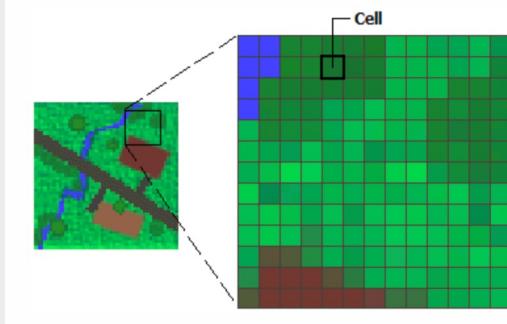
- Raster
- Vector
- Point clouds



Geospatial data types

Raster

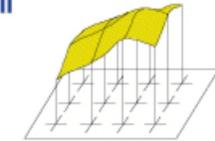
- Regularly defined geographic space, often used for representing continuous phenomena.
Example: orthophotos, remote sensing
- **Example Attribute:** elevation, temperature



Value applies to the center point of the cell

For certain types of data, the cell value represents a measured value at the center point of the cell. An example is a raster of elevation

+	315	+	319	+	321	+	323
+	317	+	323	+	328	+	326
+	313	+	318	+	325	+	323



Value applies to the whole area of the cell

For most data, the cell value represents a sampling of a phenomenon, and the value is presumed to represent the whole cell square.

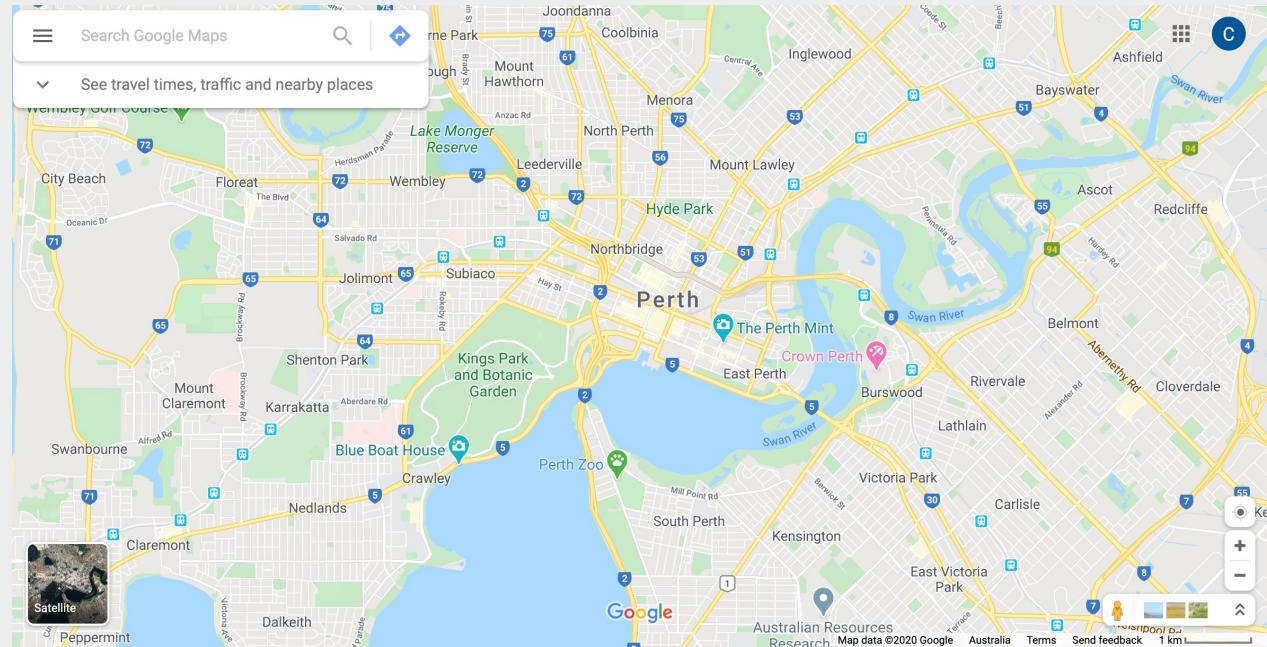
50	45	40	35
35	40	35	25
20	25	30	20



Geospatial data types

Vector

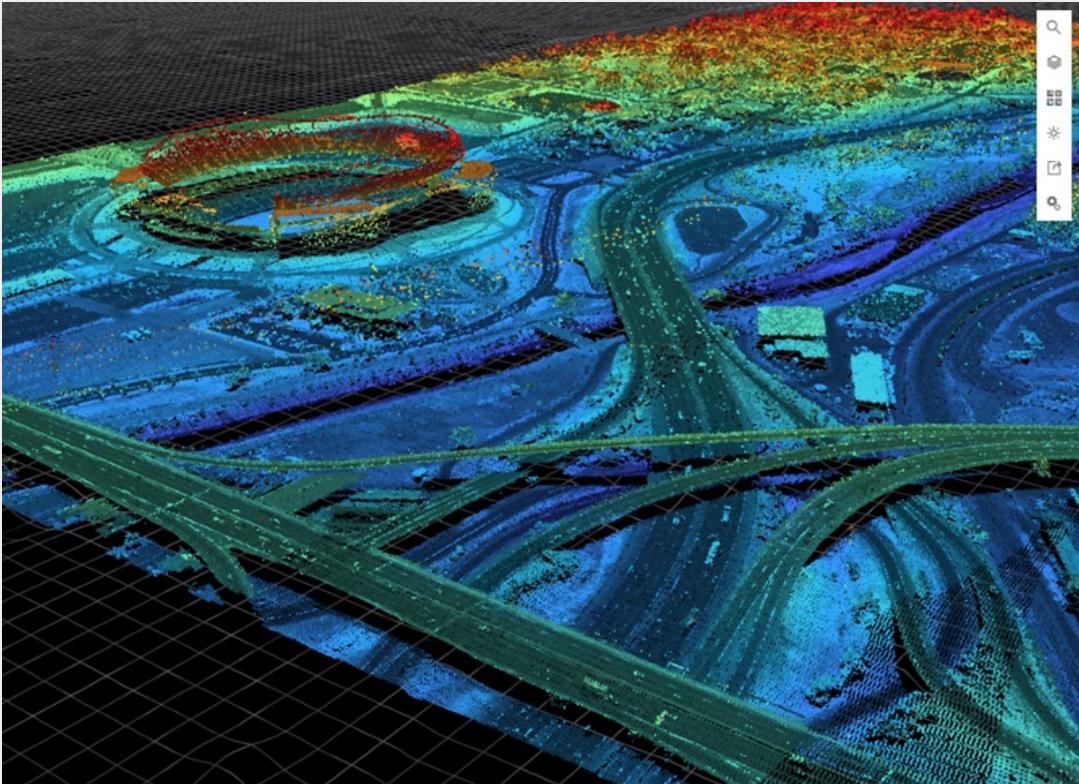
- Defined as point, line, or polygon. Often used for representing discrete phenomena.
Example: River locations
- Example attribute:**
Name of river, pH measurement, date of pH measurement



Geospatial data types

Point clouds

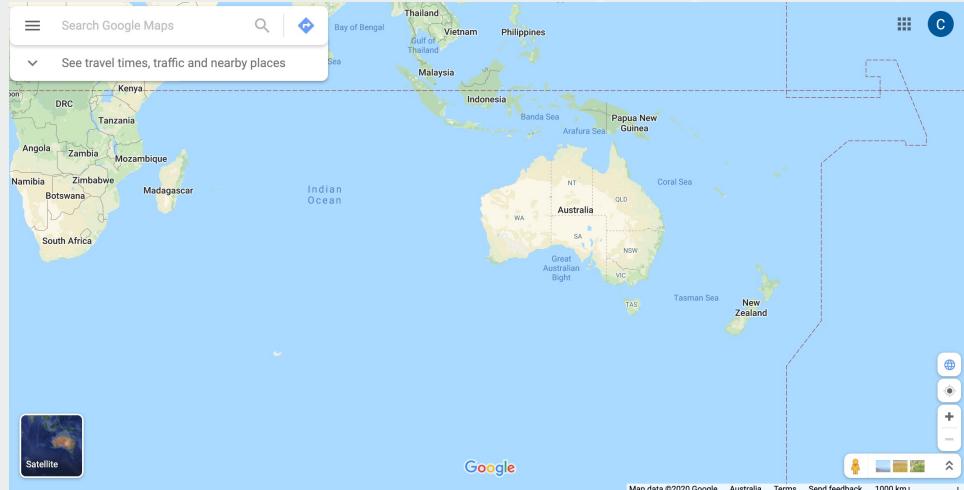
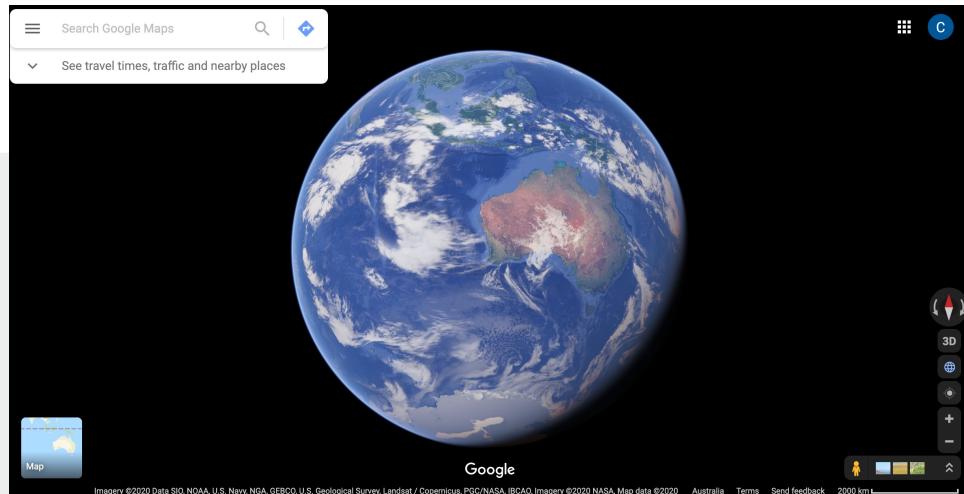
- Vector space representation of a 3D surface or object.
Example: lidar data
- Example attribute: color



Spatial reference frame

A static reference surface for identifying locations on the earth.

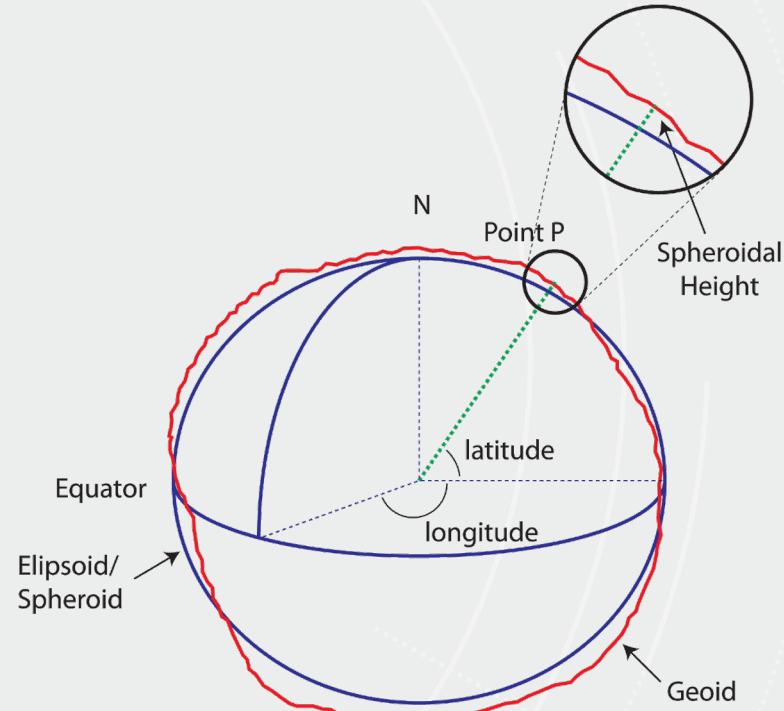
- Datum
- Projection



Spatial reference frame

Datum

- A datum identifies how locations are specified on earth's surface.
- A spheroid (or ellipsoid) is an approximation for the earth's surface, and acts as the reference for the datum.

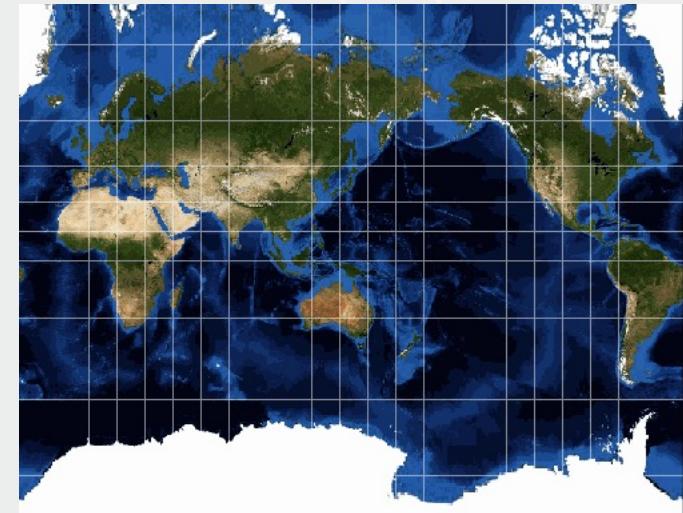
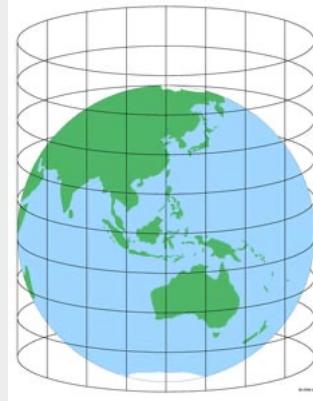


From: <https://www.icsm.gov.au/education/fundamentals-mapping/datums/datums-explained-more-detail>

Spatial reference frame

Projection

- A projection specifies how to represent the features from a spheroidal surface on a flat plane.
Types of projection techniques include:
 - azimuthal
 - cylindrical
 - conical



Excellent reference:

[https://www.icsm.gov.au/education/
fundamentals-mapping/projections](https://www.icsm.gov.au/education/fundamentals-mapping/projections)

From: <https://www.icsm.gov.au/education/fundamentals-mapping/datums/datums-explained-more-detail>

Spatial reference frame

The actual location values (latitude, longitude) will vary depending on the reference frame.

It is important to make sure all spatial data have the appropriate spatial reference information in the metadata.

Introduction to spatial data

[**Open am1_intro_to_spatial_data.ipynb**](#)

Using spatial data



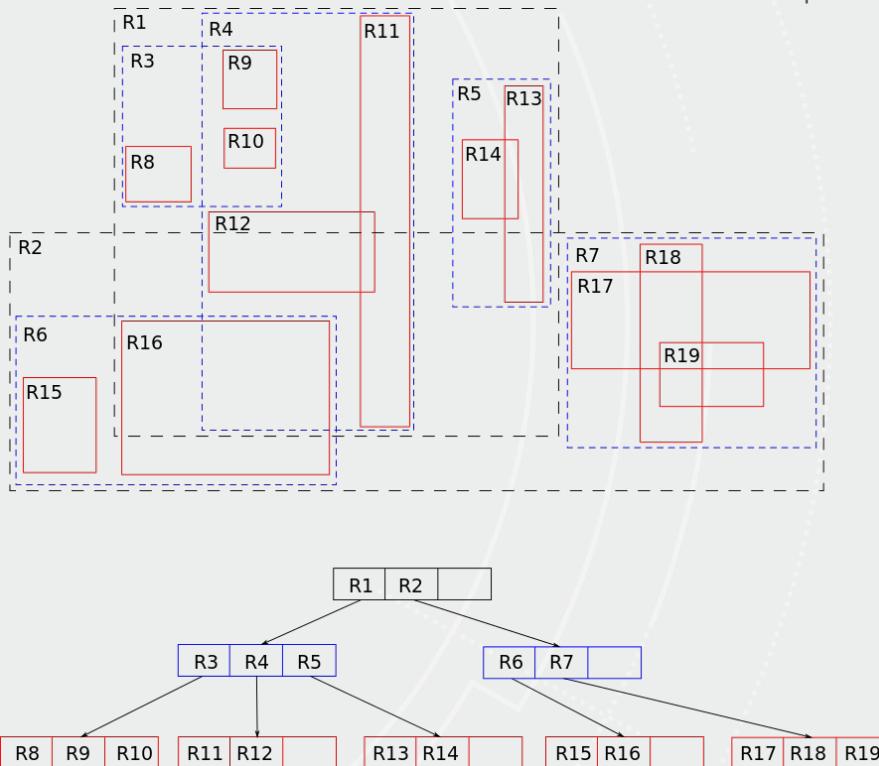


Standard spatial operations

Spatial index and selections

- **R-tree:** hierarchical data structure in which nearby objects are grouped and represented by their minimal bounding rectangle
- Used in spatial data bases for fast querying, e.g., nearest neighbours

Source: Wikipedia

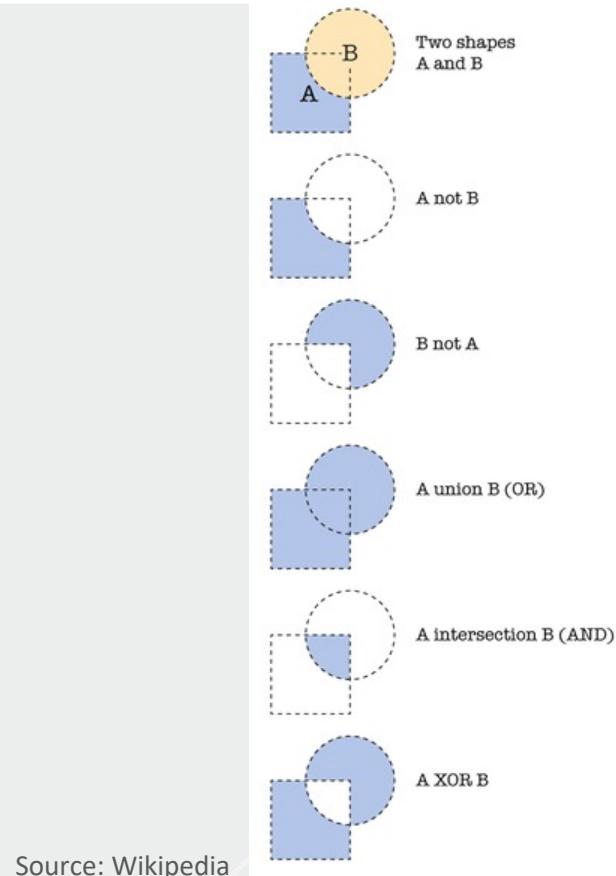




Standard spatial operations

Geometry operations

- Splitting using overlay operations (intersect, union, join)
- Joining/dissolving
- Buffering/ eroding

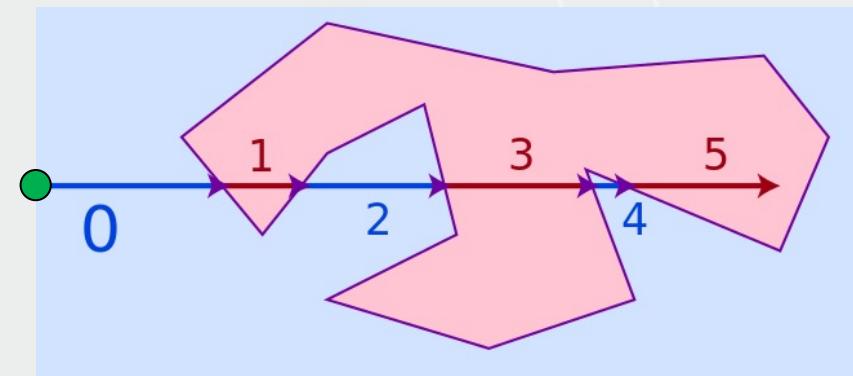




Standard spatial operations

Point in polygon

- = Identifying whether a point lies inside or outside of a polygon
- Several algorithms exist, the simplest being the Ray casting algorithm
- Check how many times a ray starting from the point intersects a polygon's edge:
 - Even = outside
 - Odd = inside



Source: Wikipedia

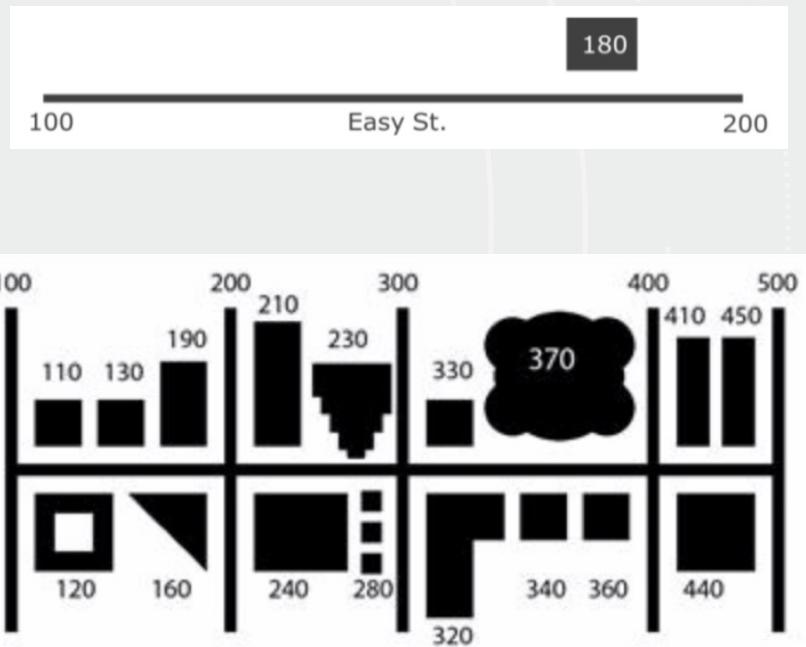


Standard spatial operations

Geocoding

= Turn a physical address (e.g., a street name) into geographical coordinates (e.g., latitude and longitude):

- Linear geocoding (find a point along a line, e.g. 191 St Georges) – estimate from start and end points of the block. Not as accurate but easy to implement and data is commonly available
- Area geocoding (find a point in a polygon) – much more accurate but requires lots more data





Standard spatial operations

Actual use

Generally, we don't just want to do this ourselves
– use an API!

See for instance [scipy.spatial](#), [shapely](#), or [geopy](#) in Python



[Scipy.org](#) [Docs](#) [SciPy v0.14.0 Reference Guide](#)

[index](#) [modules](#) [next](#) [previous](#)

Spatial algorithms and data structures (`scipy.spatial`)

Nearest-neighbor Queries

`KDTree(data[, leafsize])` kd-tree for quick nearest-neighbor lookup
`cKDTree` kd-tree for quick nearest-neighbor lookup
`distance`

Delaunay Triangulation, Convex Hulls and Voronoi

Diagrams

`Delaunay(points[, furthest_site, ...])` Delaunay tessellation in N dimensions.
`ConvexHull(points[, incremental, qhull_options])` Convex hulls in N dimensions.
`Voronoi(points[, furthest_site, ...])` Voronoi diagrams in N dimensions.

Plotting Helpers

`delaunay_plot_2d(tri[, ax])` Plot the given Delaunay triangulation in 2-D
`convex_hull_plot_2d(hull[, ax])` Plot the given convex hull diagram in 2-D
`voronoi_plot_2d(vor[, ax])` Plot the given Voronoi diagram in 2-D

See also:

[Tutorial](#)

Simplex representation

The simplices (triangles, tetrahedra, ...) appearing in the Delaunay tessellation (N-dim

Table Of Contents

- [Spatial algorithms and data structures \(`scipy.spatial`\)](#)
 - [Nearest-neighbor Queries](#)
 - [Delaunay Triangulation, Convex Hulls and Voronoi Diagrams](#)
 - [Plotting Helpers](#)
 - [Simplex representation](#)
 - [Functions](#)

Previous topic

[scipy.sparse.csgraph.minin](#)

Next topic

[scipy.spatial.KDTree](#)



Standard spatial operations

Exercise

Open **am2_geospatial_data_processing.ipynb** and go through the notebook



Spatial predictions

Why do we worry about space at all?

- **Tobler's first law of geography:**

“Everything is related to everything else,
but near things are more related than distant things.”

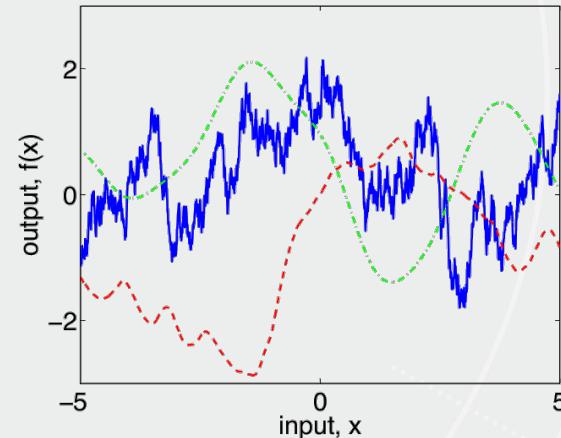
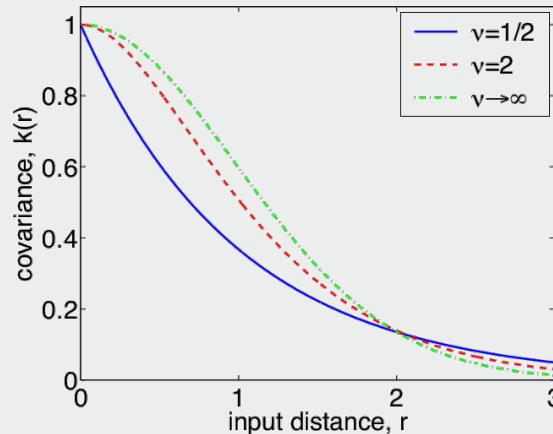
- To predict in space, each location is treated as a variable (same principle in time),
and those variables are dependent.



Spatial predictions

Covariance function

- Mathematical formulation of the first law of geography
- Defines how the correlation between two locations changes with distance



- Variogram in geostatistics, kernel in machine learning

Source: Rasmussen and Williams (2006)

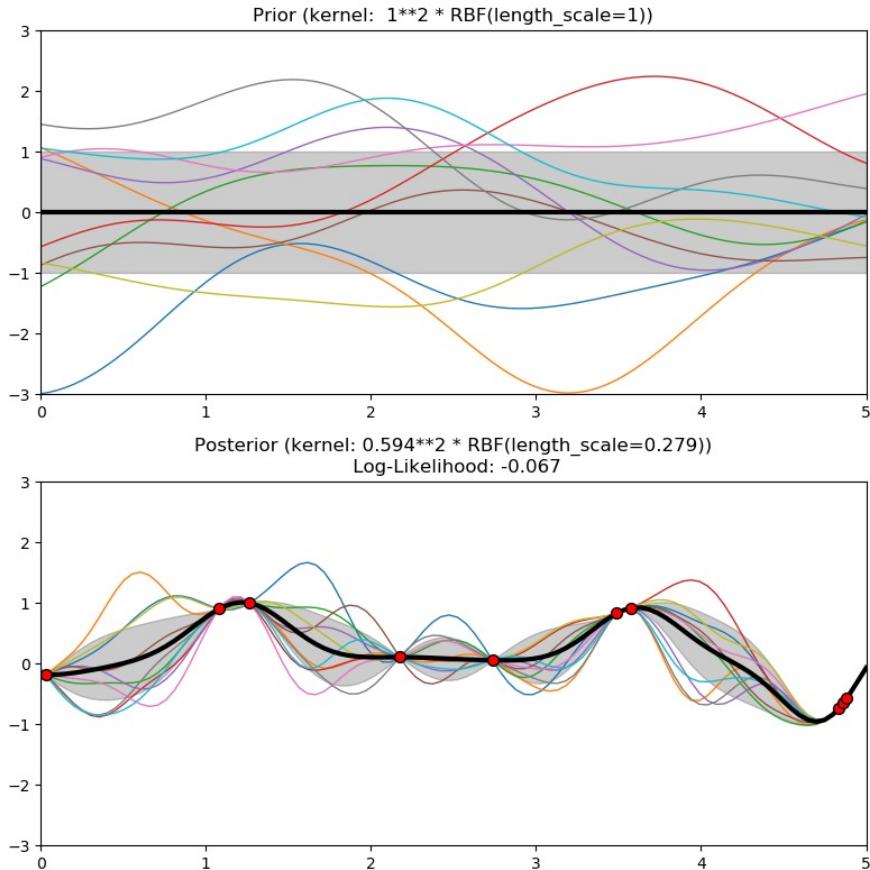


Spatial predictions

Gaussian processes

- Supervised method for regression and classification that uses a covariance function to reproduce spatial continuity
- Can fit data and provide prediction uncertainty
- But computationally demanding
- Equivalent to kriging in geostatistics

Source: Scikit-Learn documentation



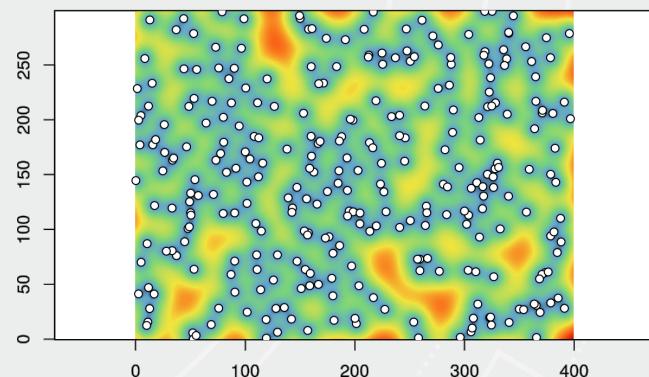
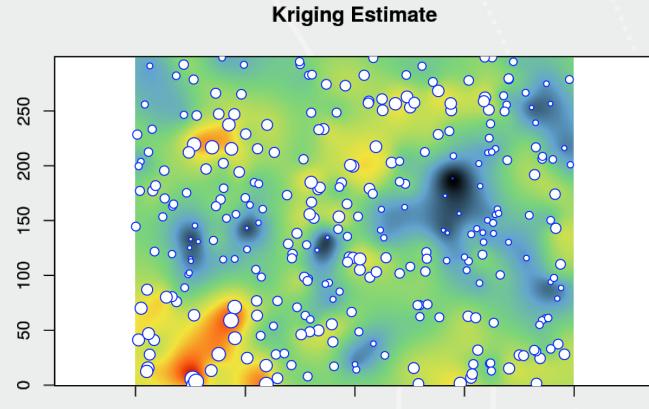


Spatial predictions

Gaussian processes

- Supervised method for regression and classification that uses a covariance function to reproduce spatial continuity
- Can fit data and provide prediction uncertainty
- But computationally demanding
- Equivalent to kriging in geostatistics

Source: Rgeostats documentation



Takeaways



High-level Takeaways From Today

Today we've:

- covered the basic concepts of spatial data: space, location and distance, and the different methods we use to talk about these concepts, including absolute, relative, topological and cognitive measures.
- covered the basics of spatial data analysis, including geometry operations, spatial indices, and geocoding.
- covered how spatial processing can form part of a larger machine learning or data science pipeline, particularly as a feature generator for machine learning algorithms.
- covered how to design an effective pitch that considers the audience, communication styles and content structure.



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