

Part II All About Data

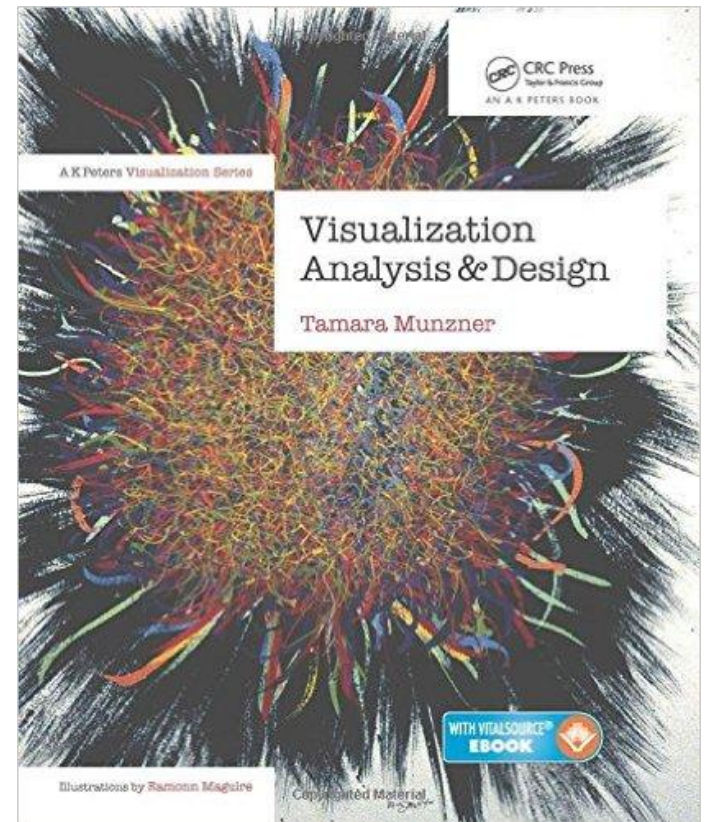
Data Abstraction

The scientist described what is: the engineer creates what never was.

*Theodor von Karman
The father of supersonic flight*

Source of This Unit

- Material of this unit is based on Tamara Munzner, *Visualization Analysis and Design*, AK Peters/CRC Press, 2014.



Overview

- **Topics to be covered in this unit:**

- **Types**

- ❖ **Data Types:** Item, Attribute, Link, Position, Grid, etc.,

- ❖ **Dataset Types:** Table, Network, Field, Geometry, etc.

- ❖ **Attribute Types:** Categorical, Ordered

- **Data Semantics:** Key vs. Value, Temporal, etc.

Data and Dataset: 1/5

- The ***type*** of data is its structural or mathematical interpretation.
 - At the ***data*** level, it can be an item, a link, an attribute, etc.
 - At the ***dataset*** level, it is how these data types are combined into a larger structure such as a table, a tree, a field of values.

Data and Dataset: 2/5

- There are five basic data types:
 - ***Item***: An individual entity that is discrete (e.g., a number, a row of a table, etc.)
 - ***Attribute***: Some measurable property (e.g., salary, price, temperature, etc.)
 - ***Link***: A relationship between items
 - ***Position***: A spatial data (e.g., location, coordinates, etc.)
 - ***Grid***: The strategy for sampling continuous data (e.g., geometric and/or topological) between cells.

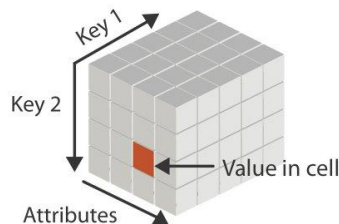
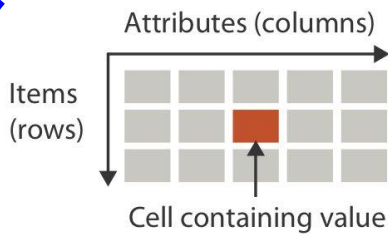
Data and Dataset: 3/5

- A ***dataset*** is a collection of information that is the target of analysis.
- There are four basic types:
 - Tables
 - Networks
 - Fields
 - Geometry
- Complex combinations of multiple data types are commonly seen in real world applications.

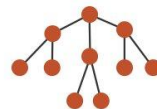
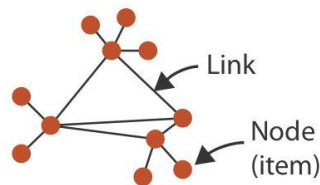
Data and Dataset: 4/5

- There are four basic dataset types: tables, networks, fields and geometry.

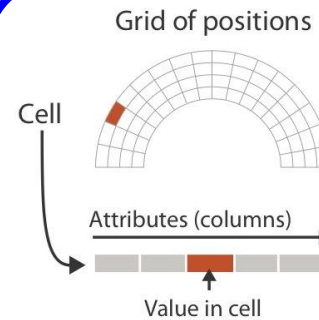
tables



networks



fields



geometry



Data and Dataset: 5/5

- There are four basic dataset types: tables, networks, fields and geometry.
- There are other possible collections of items such as clusters, sets, lists, etc.

Tables	Networks & Trees	Fields	Geometry	Clusters, Sets, Lists
Items	Items (Node)	Grids	Items	Items
Attributes	Links	Positions	Positions	
	Attributes	Attributes		

Dataset Type (Table): 1/2

- Tables are commonly seen type of datasets.
- In a 2D table, each row is an **item**, each column is an **attribute**, and each **cell** has a **value** of a particular item and a particular attribute.

ID	Name	Addr.	City	State	Zip	Income	Phone
31765	John Dow	xxxxxx	Houghton	MI	49931	50K	906-123-4567

item

cell

attribute

Dataset Type (Table): 2/2

- A multidimensional table uses more indices.
- The following 3D table uses three indices to find a cell (i.e., Department, Gender, Status).

Department		A	B	C	D	E	F
Male	Accepted	512	353	120	138	53	22
	Rejected	313	207	205	279	138	351
Female	Accepted	89	17	202	131	94	24
	Rejected	19	8	391	244	299	317

Dataset Type (Networks): 1/1

- Networks (or graphs) are useful to represent relationship between several items.
- Here, an ***item*** is a ***node*** and a ***link*** is a relation between two items.
- Each node can have associated attributes (e.g., city size), and each link may also have associated attributes (e.g., distance between two cities).
- A tree is just a special type of networks.

Dataset Type (Fields): 1/8

- A ***field*** contains attribute values associated with cells.
- Each ***cell*** contains measurements or calculations from a ***continuous*** domain.
- Obtaining values from a continuous domain is usually very challenging because the domain is a continuum.
- A good ***sampling*** strategy for taking measurements from discrete positions is needed.

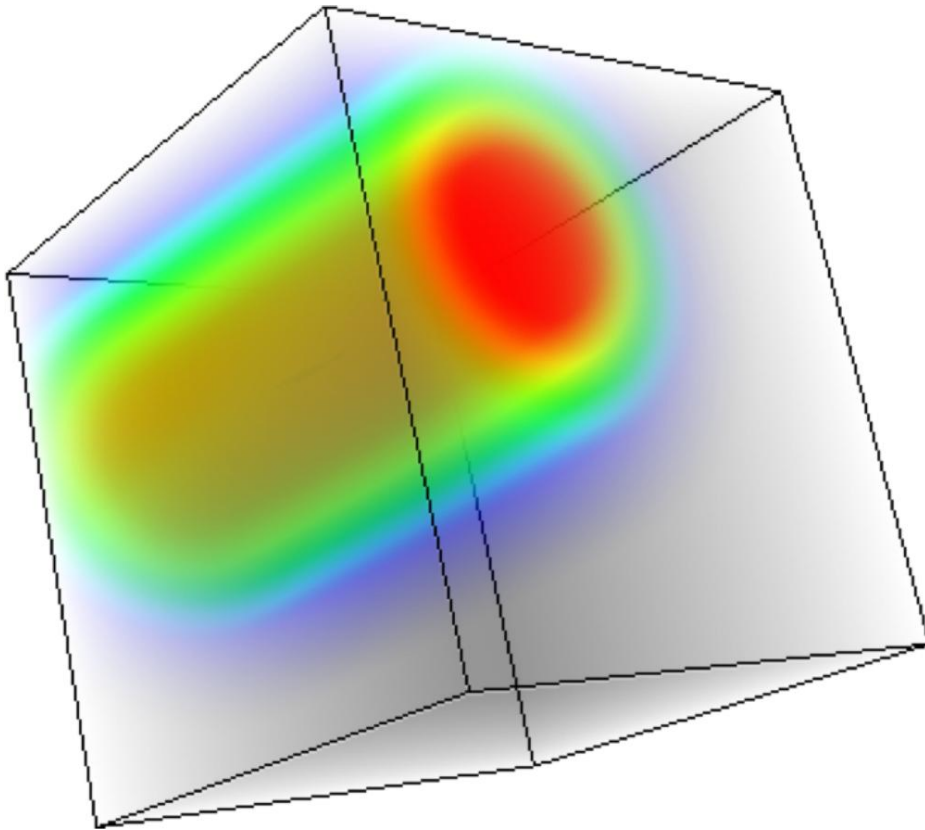
Dataset Type (Fields): 2/8

- Because the number of measurements is only finite, we need to ***interpolate*** those missing measurements (i.e., showing the values between sampled values).
- With a good sample and an appropriate interpolation, the original continuum can be ***reconstructed*** so that the view is faithful to the measured values from an arbitrary viewpoint.
- The handling of continuous domain is always challenging.

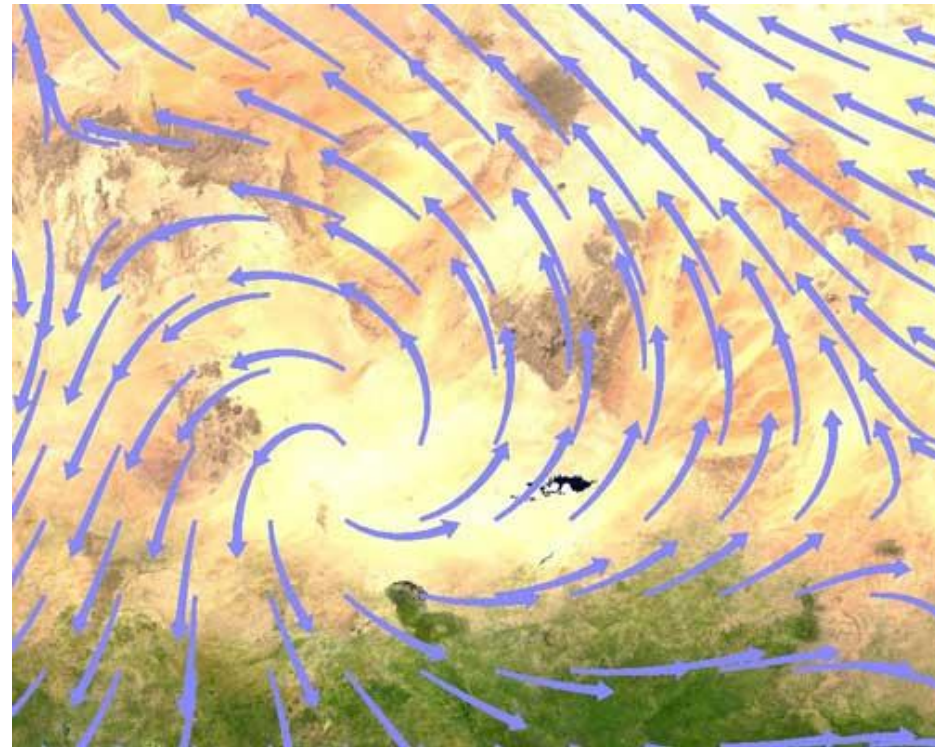
Dataset Type (Fields): 3/8

- The cell structure of a ***spatial field*** is based on sampling at spatial positions.
- ***Example***: We may measure the temperature at a space point and the result is represented by (x,y,z) – a point in space – and the measured temperature t . This is a ***scalar field***, because the measurement is a single value.
- ***Example***: We may also measure the velocity of a flow, and the result is a point (x,y,z) and a vector (i.e., velocity). This is a ***vector field***.

Dataset Type (Fields): 4/8



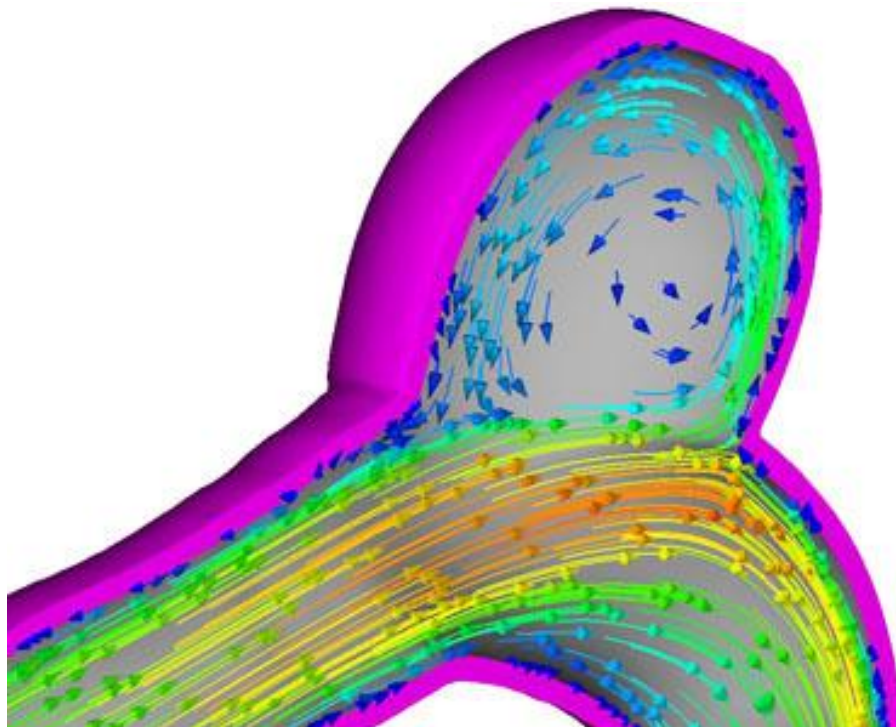
scalar field



vector field

Dataset Type (Fields): 5/8

- The collected dataset may contain spatial information (i.e., the positions). This is a dataset that contains ***spatial data***.

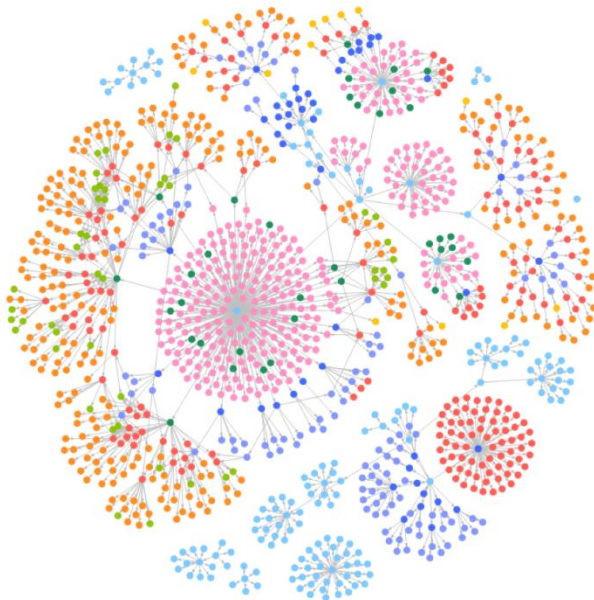


Dataset Type (Fields): 6/8

- In many applications, the spatial information may not be given. This dataset contains ***non-spatial data***.
- Non-spatial data is sometimes also referred to as ***abstract data***.
- In this case, the use of space is chosen by visualization designers.

Dataset Type (Fields): 7/8

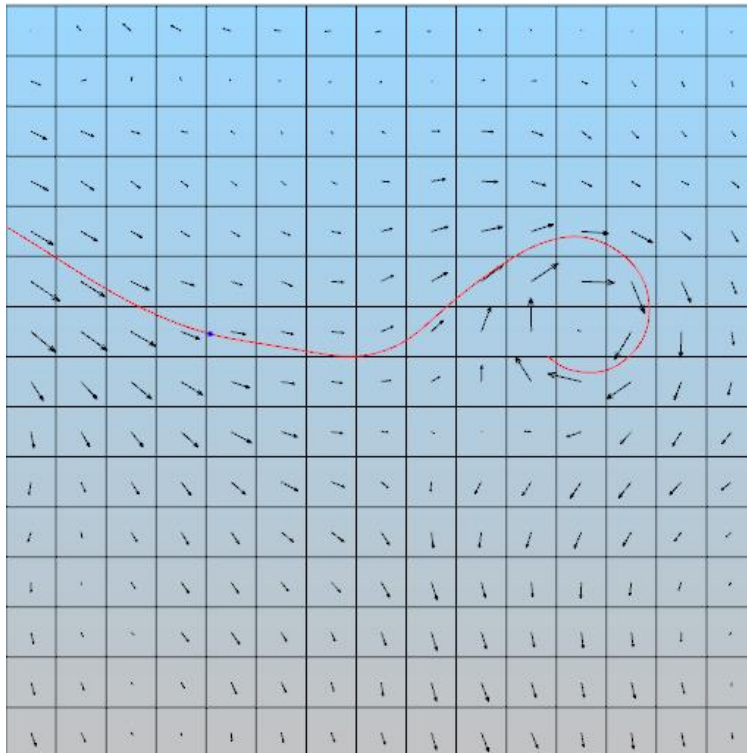
- ***Scientific visualization*** is concerned with situation where spatial position is given, while in ***information visualization*** is concerned with situation where the use of space in a visual encoding is chosen by the designer.



only the nodes and links are provided

Dataset Type (Fields): 8/8

- When a field contains data created by sampling at a completely regular intervals, the cells form a ***uniform grid***.

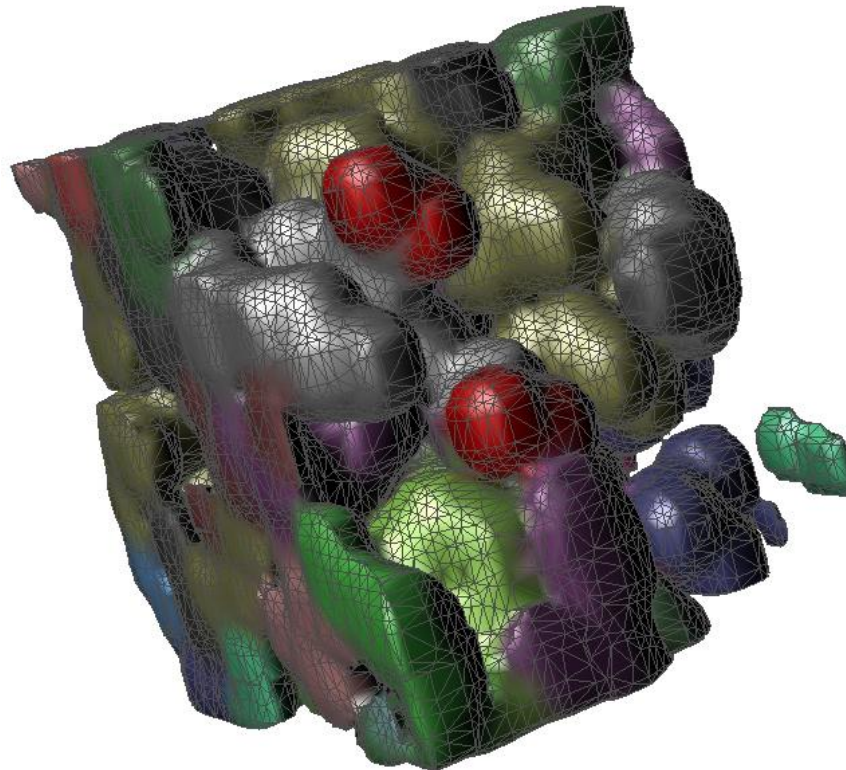


Geometry: 1/4

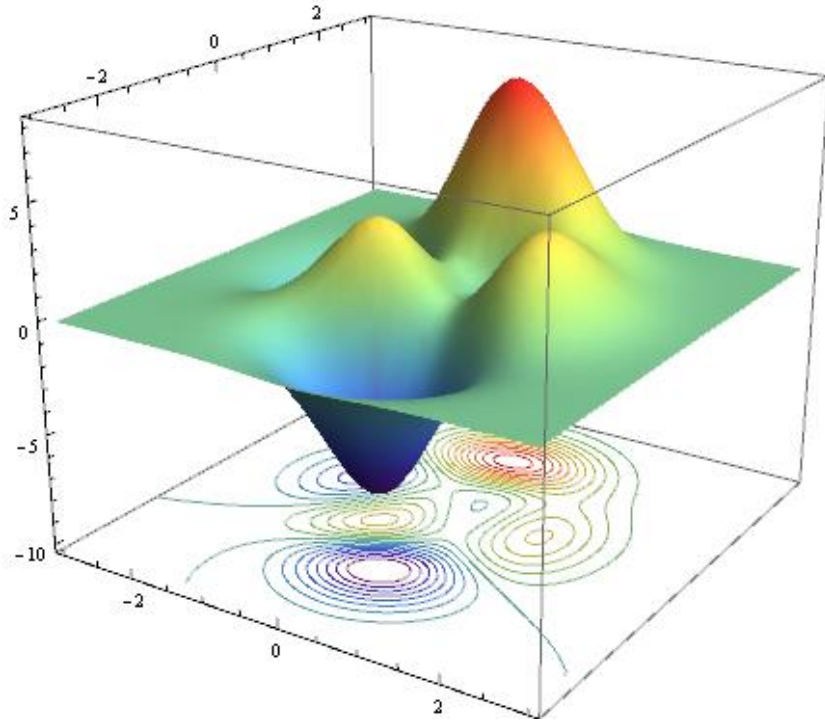
- The ***geometry*** dataset type specifies information about the shape of items with explicit spatial information.
- The items could be points, lines/curves, 2D surfaces/regions, 3D volumes, or even higher dimensional data.
- Therefore, geometry datasets are spatial, and typically occur in the context of tasks that require shape understanding.

Geometry: 2/4

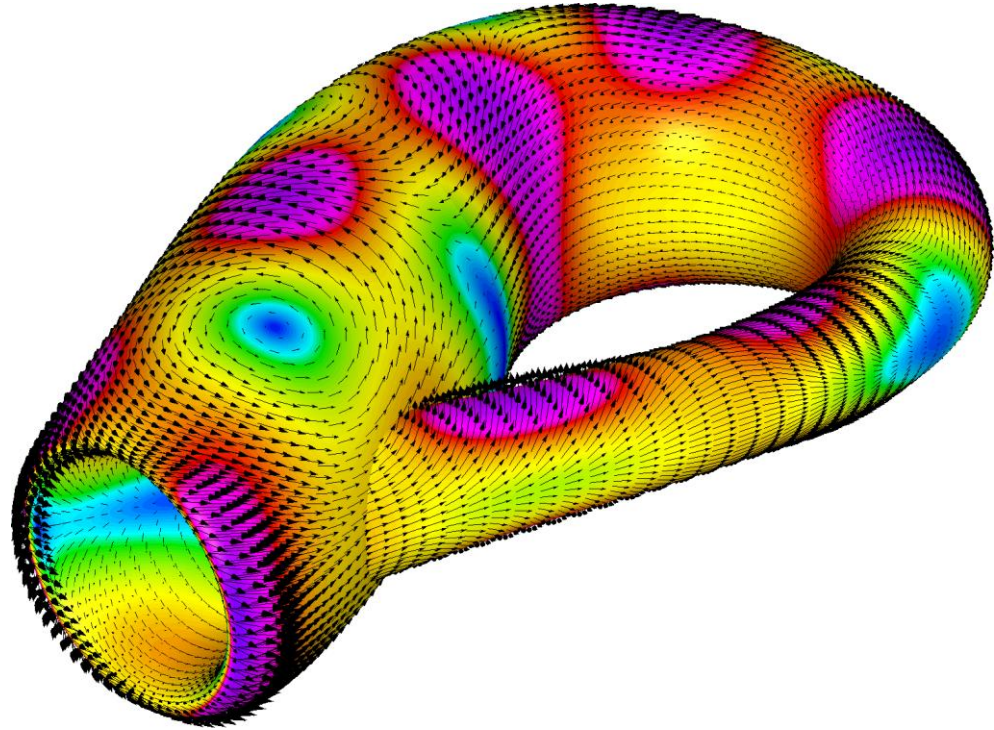
- **Geometry datasets may not have attributes. In situations where we only care about shape understanding, only the positions would be enough.**



Geometry: 3/4

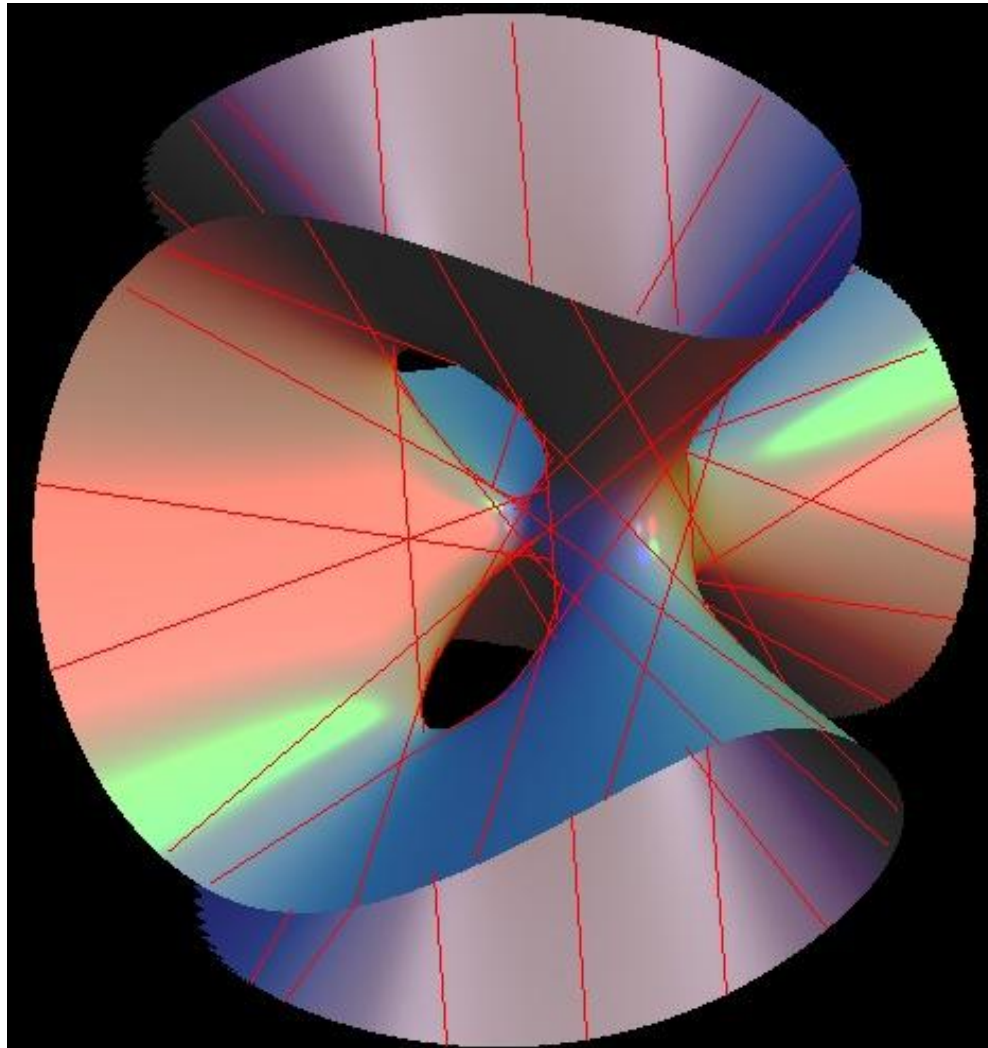


**contours generated from
a spatial field**



**Klein bottle with a vector field
on a regular grid**

Geometry: 4/4



Clebsch cubic surface with 27 lines

Dataset Availability

- There are two kinds of dataset availability: ***static*** or ***dynamic***.
- A ***static file*** means the entire dataset is available all at once.
- Some datasets are only available as ***dynamic streams***, where the dataset information trickles in over the course of the visualization session.
- This dynamic information may mean adding new items, deleting existing ones, or the values of existing items may even change.

Attribute Types

- **Attribute types are:**
 - **Categorical**
 - **Ordered**
 - **Ordinal**
 - **Quantitative**
 - **The direction of attribute ordering can be:**
 - **Sequential**
 - **Diverging**
 - **Cyclic**

Categorical Type

- ***Categorical*** data does not have an implicit ordering, but often has a hierarchical structure.
- ***Examples***: Gender types, file types, shapes type (e.g., triangles, circles, rectangles, etc.), fruit types (e.g., apples, oranges, bananas, etc.)
- The above examples do not have an “implicit” order imposed to the data.
- However, one may enforce an order to each of the above example. For example, fruit names are arranged in alphabetical order.

Ordered Type

- **Ordered** data has an implicit ordering. There are two ordered types: ordinal and quantitative.
 - **Ordinal**: It has a well-defined ordering but cannot do full-fledged arithmetic.
 - **Example**: shirt size, shoe size, grade (e.g., A, AB, etc.), ranking, zip code, etc.
 - **Quantitative**: This is an ordinal dataset with a well-defined capability to perform arithmetic and comparison.
 - **Example**: weight, height, scores, etc.

Direction of Ordering: 1/2

- An order dataset can be ***sequential*** or ***diverging***.
- A ***sequential*** dataset has a homogeneous range from a minimum to a maximum.
- A ***diverging*** dataset has data measured from a based point and extends to both ends. An elevation dataset is diverging because its measurement start at sea level.

Direction of Ordering: 2/2

- Both ends of a ***diverging*** dataset are sequential.
- A ***cyclic*** dataset has its values wrap around back to a starting point (e.g., the hour of the day, the day of a week, and month of the year, angle measures, etc.).

Hierarchical Attributes

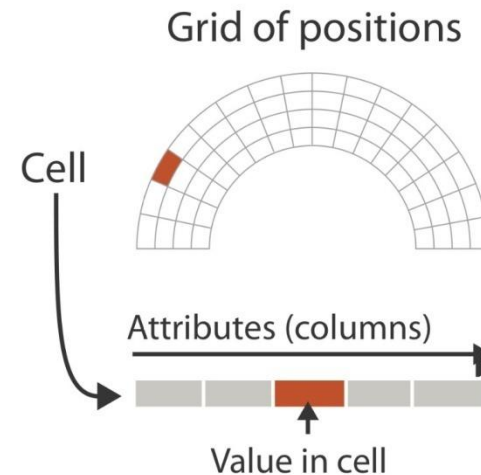
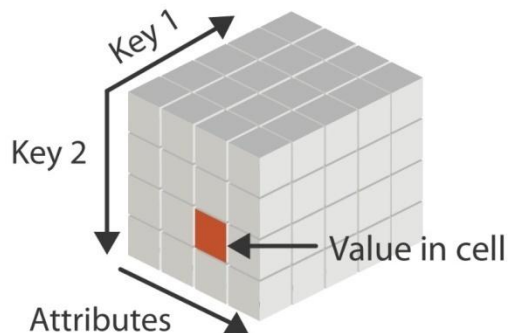
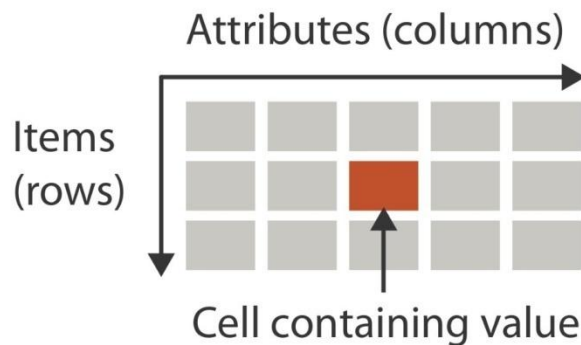
- An attribute or across multiple attributes may have a hierarchical structure.
- ***Example:*** The daily prices of companies collected over the course of a decade is a ***time-series*** dataset.
 - One attribute is time. Moreover, time can be aggregated hierarchically (e.g., weeks, months, years). Each different aggregation may show interesting patterns.
- Geographic attribute of a postal code may be aggregated to the level of cities, states, etc.

Data Semantics

- The ***semantics*** of the data is its real-world meaning.
- What does this number 9400109699937278441167 mean? Is it just a big integer? Is it a USPS tracking number or something else? Is it the number of days since the big bang?
- Is “Johnson” a person’s name, a company’s name, a city name, a password, a program name?

Key vs. Value Semantics: 1/3

- A **key** attribute is an index used to find **value** attributes.



Key vs. Value Semantics: 2/3

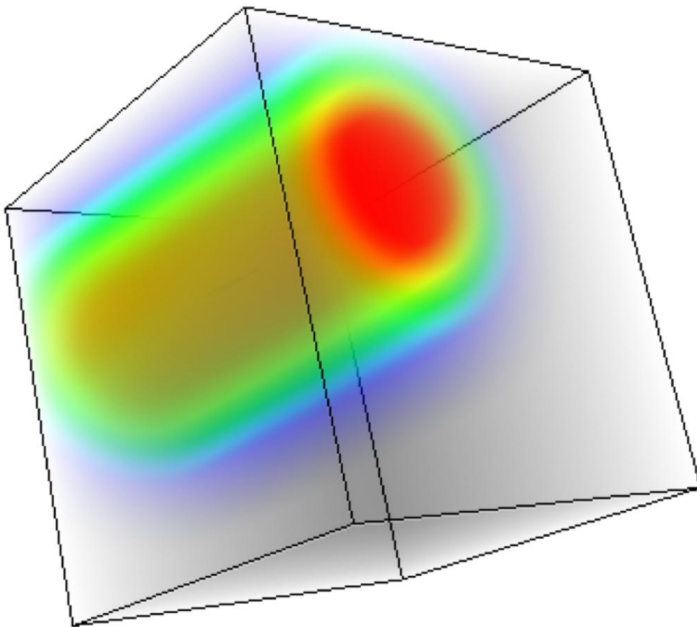
- In a table, the key may just be implicit (e.g., the indices).
- A key attribute can be explicitly given. However, this key attribute must not have duplicate values.
- **Recall this:** Fields are generated through a systematic sampling so that each grid cell is a spanned by a unique range from a continuous domain.

Key vs. Value Semantics: 3/3

- Fields are **multivariate** (resp., **univariate**) if it has **more than** (resp. **only**) one value attributes.
- The **multidimensional** structure of a field depends on the number of keys.
- A field can be multivariate and multidimensional at the same time if it has multiple values and multiple keys.

Scalar Fields

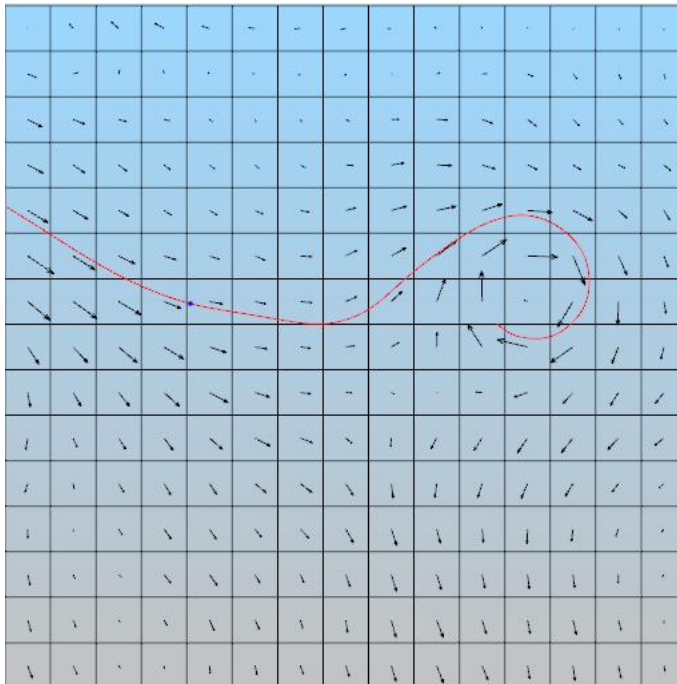
- A scalar field is **univariate**, with a single attribute at each point in space.
- This means we assign a single value (e.g., temperature) to each point in space.



colors are used to show
the values of a scalar field

Vector Fields

- A vector field is ***multivariate***, with a list of multiple attribute values (i.e., a vector) at each point in space.



a vector is assigned to a
representative point of a cell

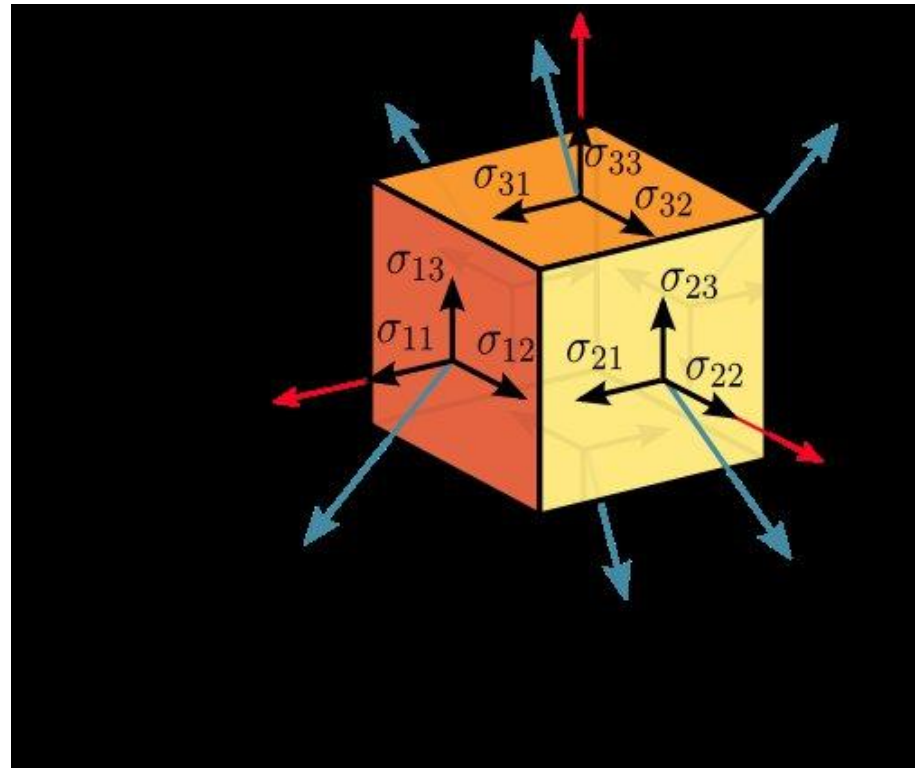
Tensor Fields: 1/5

- A tensor field is ***multivariate*** and has an array of attributes at a point. A rank n tensor in m -dimensional space is a mathematical object that has n indices and m^n components and obeys some transformation rules.
- Tensors are being used in differential geometry, physics, and engineering.

Tensor Fields: 2/5

- **Example:** The Cauchy stress tensor or simply stress tensor that can be represented by a 3×3 real matrix $\sigma = [\sigma_{ij}]_{3 \times 3}$.

Wikipedia



Tensor Fields: 3/5

- ***Example:*** The metric measure on a manifold can also be represented as a tensor.
- The Euclidean metric tensor in the n -dimensional Euclidean space E^n is the $n \times n$ identity matrix.
- The curvature tensor measures the curvature at a point in a space.

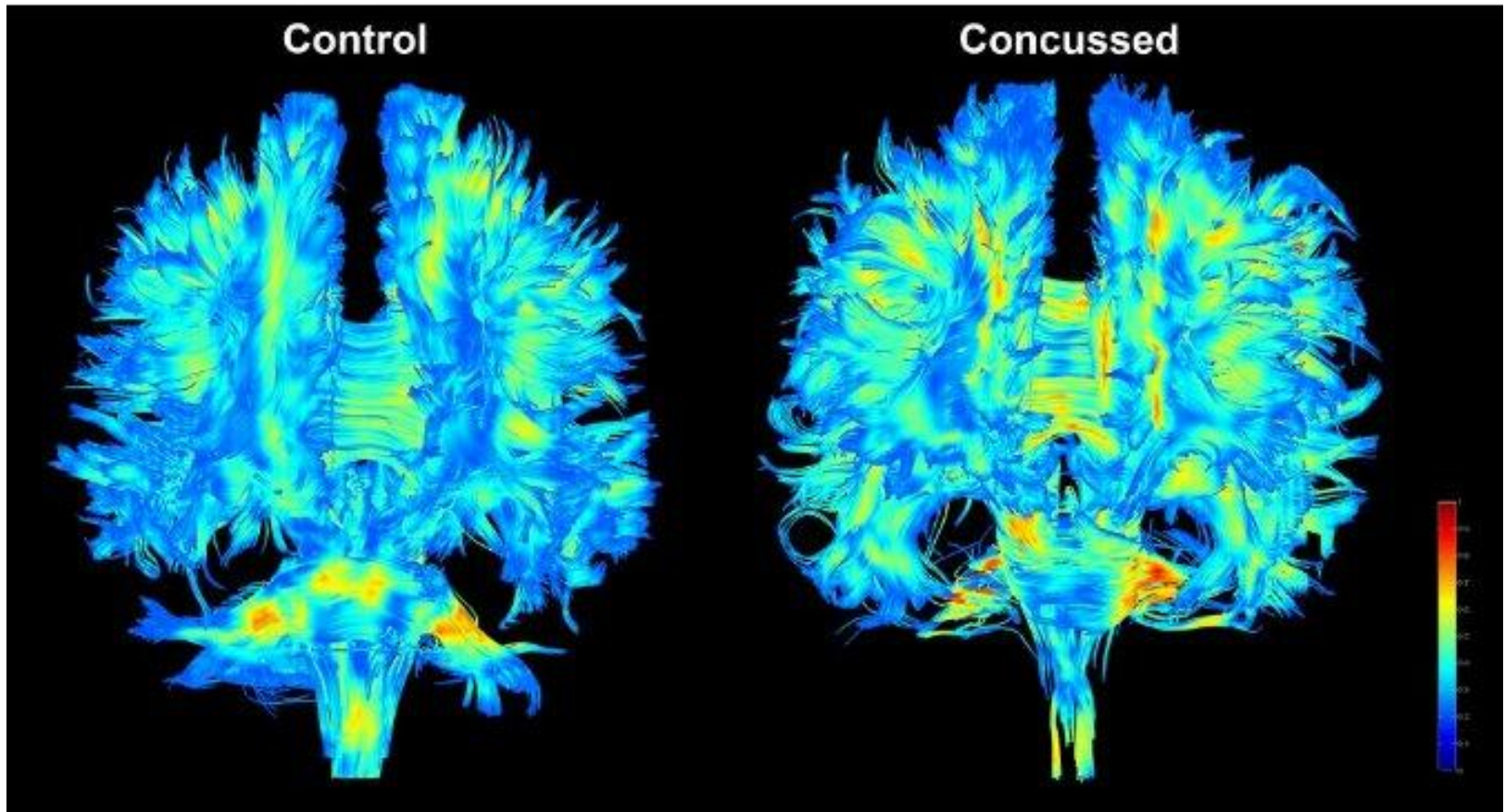
Tensor Fields: 4/5

- The metric tensor of the Minkowski space, used in special relativity, is the one on the left.
- The Schwarzschild metric that describes the space-time around a spherical symmetric body is the one on the right.

$$g = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$$

$$g_{\mu\nu} = \begin{bmatrix} (1 - \frac{2GM}{rc^2}) & 0 & 0 & 0 \\ 0 & -(1 - \frac{2GM}{rc^2})^{-1} & 0 & 0 \\ 0 & 0 & -r^2 & 0 \\ 0 & 0 & 0 & -r^2 \sin^2 \theta \end{bmatrix}$$

Tensor Fields: 5/5



Diffusion tensor visualization: The directionality of water diffusion. Warmer hues represent greater directional preference for water diffusion and cooler hues indicate more random water diffusion. – Michael Borich ⁴¹

Field Semantics

- On previous slides, the categorization of spatial fields requires knowledge of the attribute semantics and cannot be determined from type information alone.
- Thus, if multiple measured values at each spatial point are given without further information, there is no sure way to know its structure.

Temporal Semantics: 1/2

- A ***temporal*** attribute is any kind of information that relates to time.
- Data about time is complicated to handle because of the rich hierarchical structure that can be used to reason about time and the potential for periodic structure.
- The analysis of time usually involves finding or verifying periodicity either at a predetermined scale or at some scale unknown in advance.

Temporal Semantics: 2/2

- It is important to note that there could be multiple ways to visually encoding that data.
- One of these ways may involve animation.
- A temporal key attribute can have either value of key semantics. For example,
 - The day/time (or duration) a transaction happened is a dependent value
 - Time can be an independent key – a MRI scan can have the independent keys of (x,y,z,t) to cover spatial position (x,y,z) and time t .

Time-Varying Data: 1/2

- A dataset has ***time-varying*** semantics when time is one of the key attributes, as opposed to when the temporal attribute is a value rather than a key.
- ***Time-Varying Semantics***: The use of tracking device to track movements. The temporal attribute is an independent key.
- ***Non Time-Varying Semantics***: A train scheduling table contains start time and end time. In this case, the time entries are values.

Time-Varying Data: 2/2

- A commonly seen temporal dataset is a ***time-series*** dataset.
- A time-series dataset usually has an ordered sequence of time-value pairs such as (t, x) , where t and x are time and value.
- Note that the time values may not always be spaced at uniform temporal intervals.
- Typical time-series analysis tasks involve finding trends, correlations, autocorrelations, periods, variations at multiple time scales.

The End