

cs6630 | October 23 2014

TEXT & SETS

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University of Utah



administrivia . . .

-parallel coordinates due next Thursday

last time . . .

dataset types

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

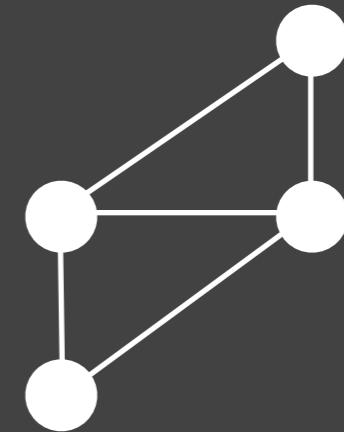
Diagram illustrating dataset types:

- Tables:** Represented by a grid. Labels: "Items (rows)", "Attributes (columns)", "Cell containing value". A 3D cube diagram shows "Key 1", "Key 2", and "Attributes" axes, with "Value in cell" indicated.
- Networks & Trees:** Represented by graphs. Labels: "Items (nodes)", "Links", "Attributes". Diagrams show a network graph with nodes and links, and a tree structure.
- Fields:** Represented by grids. Labels: "Grids", "Positions", "Attributes". Diagrams show a grid of positions and a grid with "Attributes (columns)" and "Value in cell" indicated.
- Geometry:** Represented by spatial data. Labels: "Items", "Positions". Diagrams show a map with red dots representing positions.
- Clusters, Sets, Lists:** Represented by lists. Labels: "Items".

GRAPHS & TREES

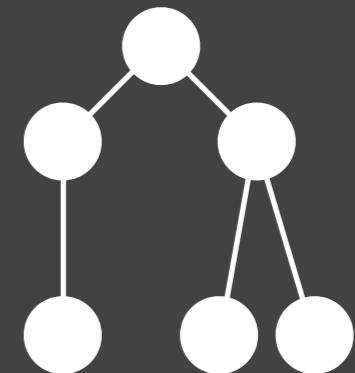
- graphs

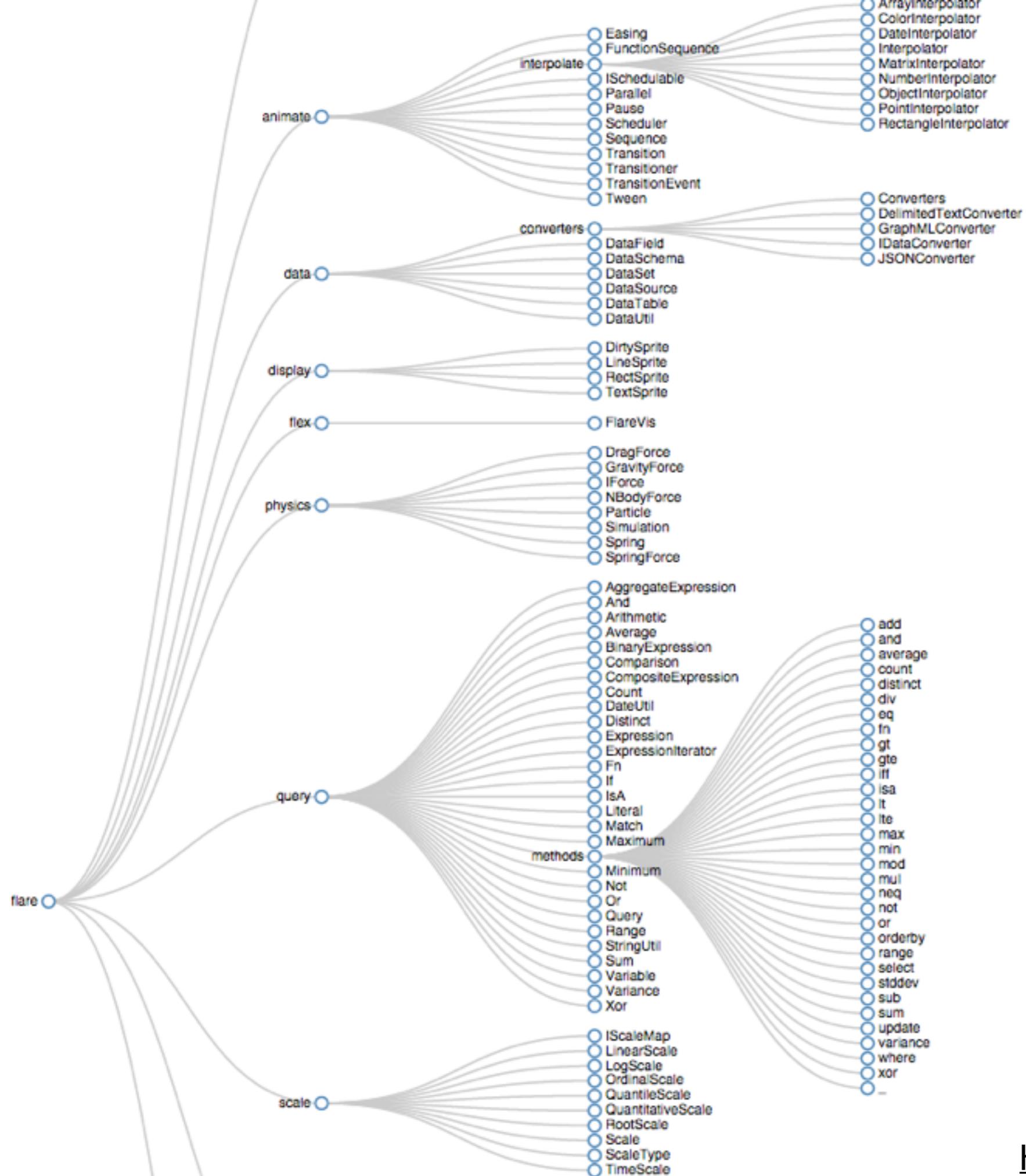
- model relations amount data
- nodes and edges

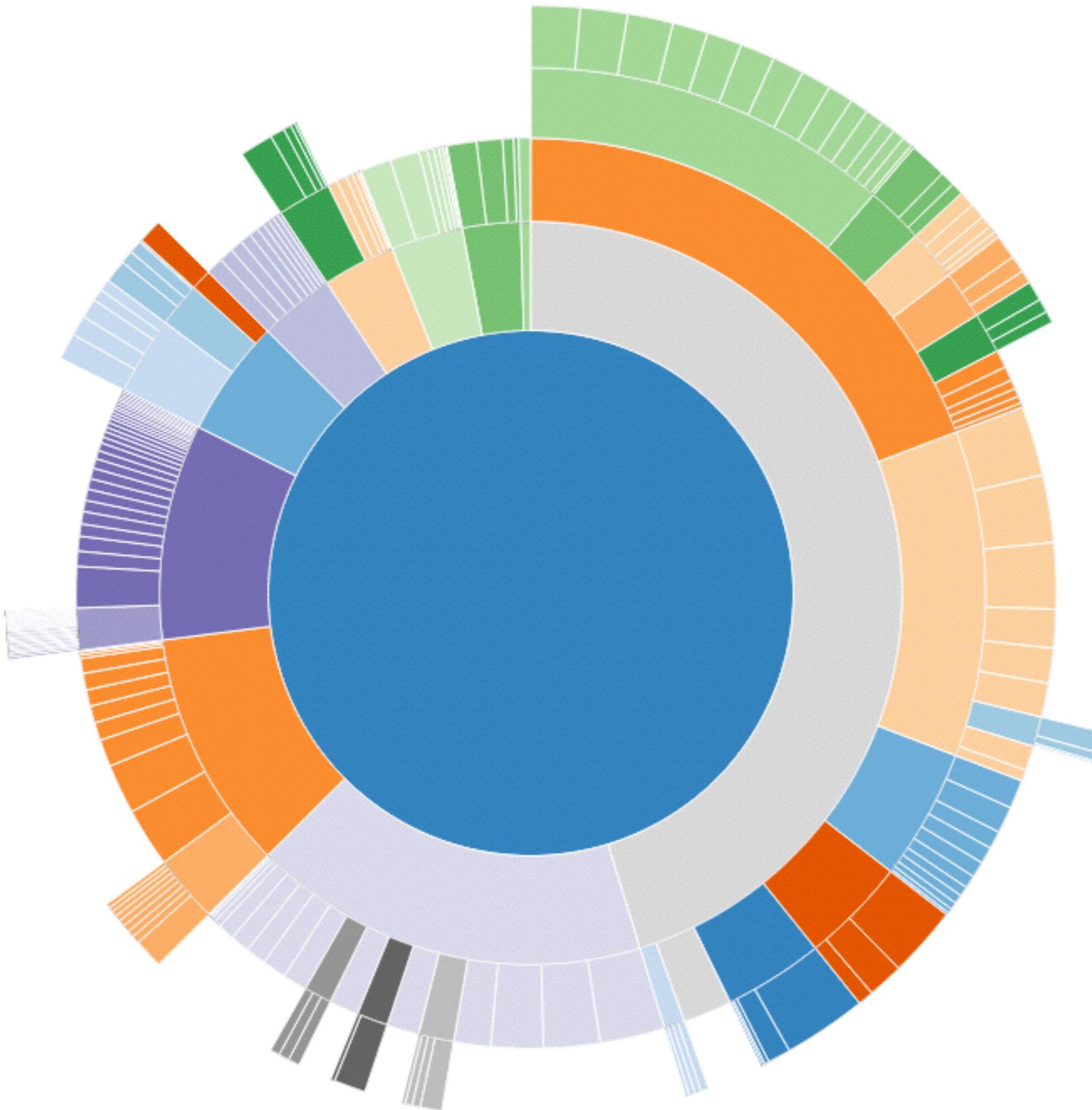


- trees

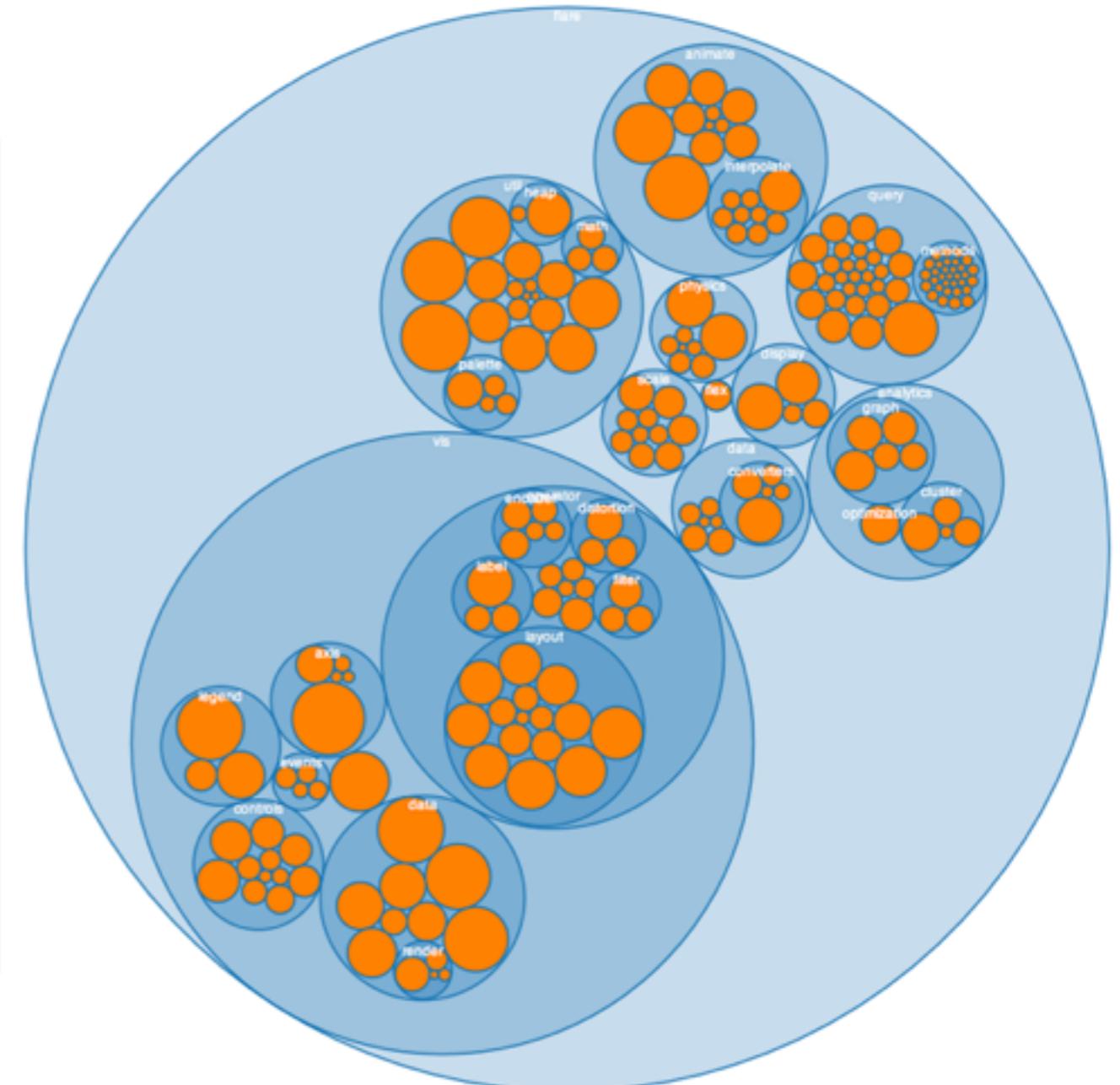
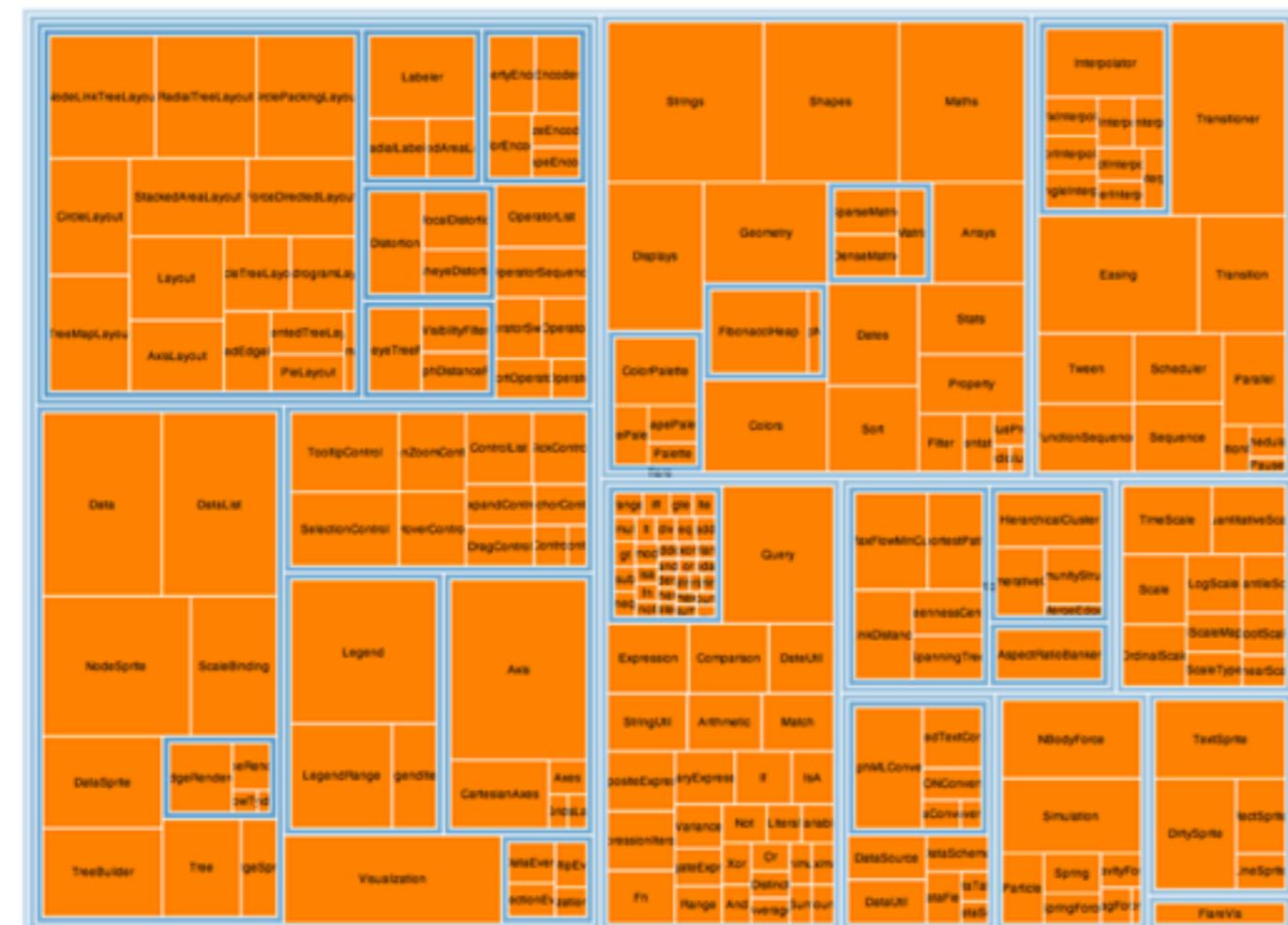
- graphs with hierarchical structure
- nodes as *parents* and *children*







<http://bl.ocks.org/mbostock/raw/4348373/>



VISUALIZING GRAPHS

- node link layouts

- Reingold-Tilford (trees only)
- Sugiyama (directed acyclic graphs)
- Force directed
- Attribute-based

- adjacency matrices

- aggregate views

- Motif Glyphs
- PivotGraph

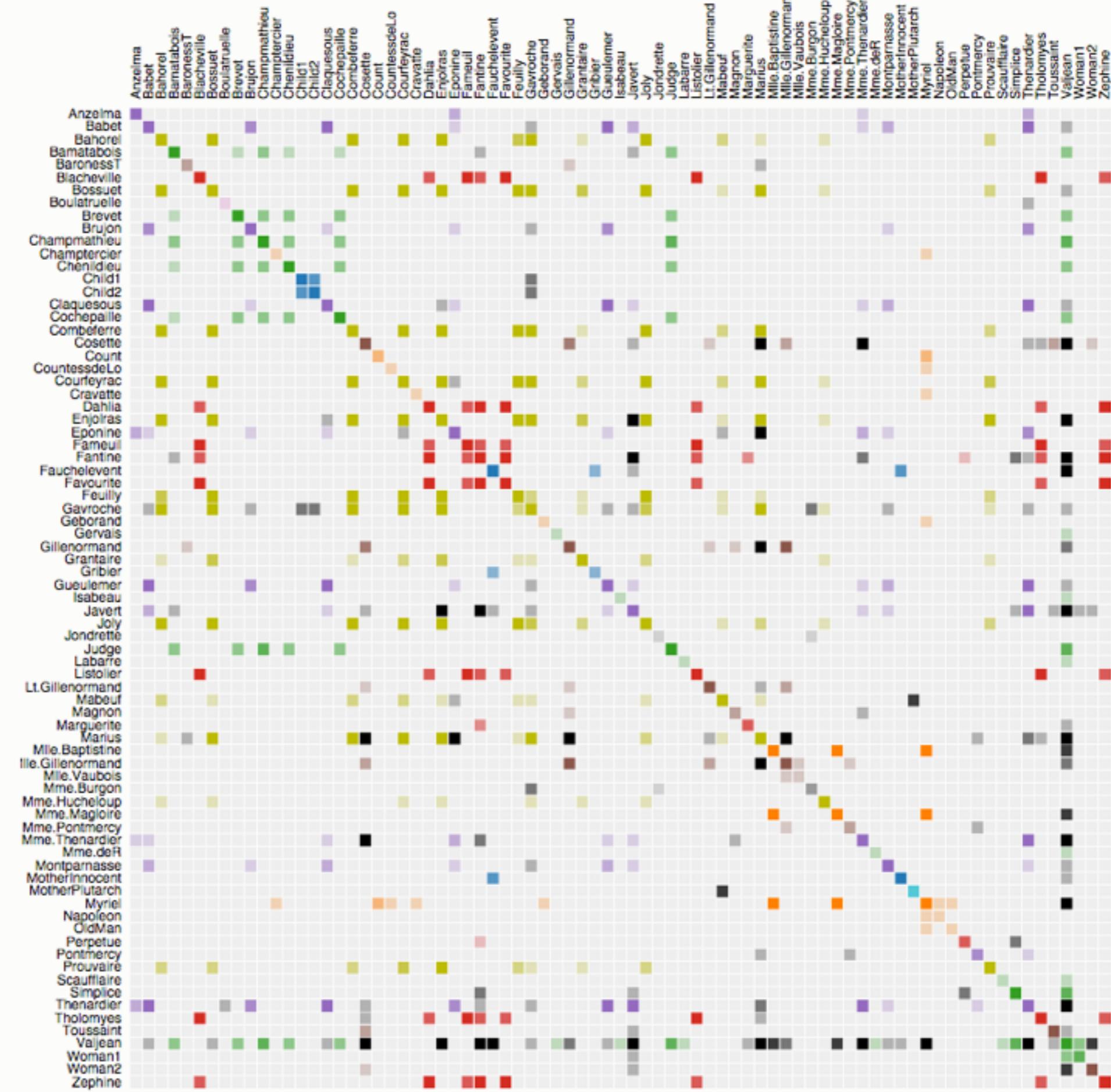
Les Misérables

character co-occurrence

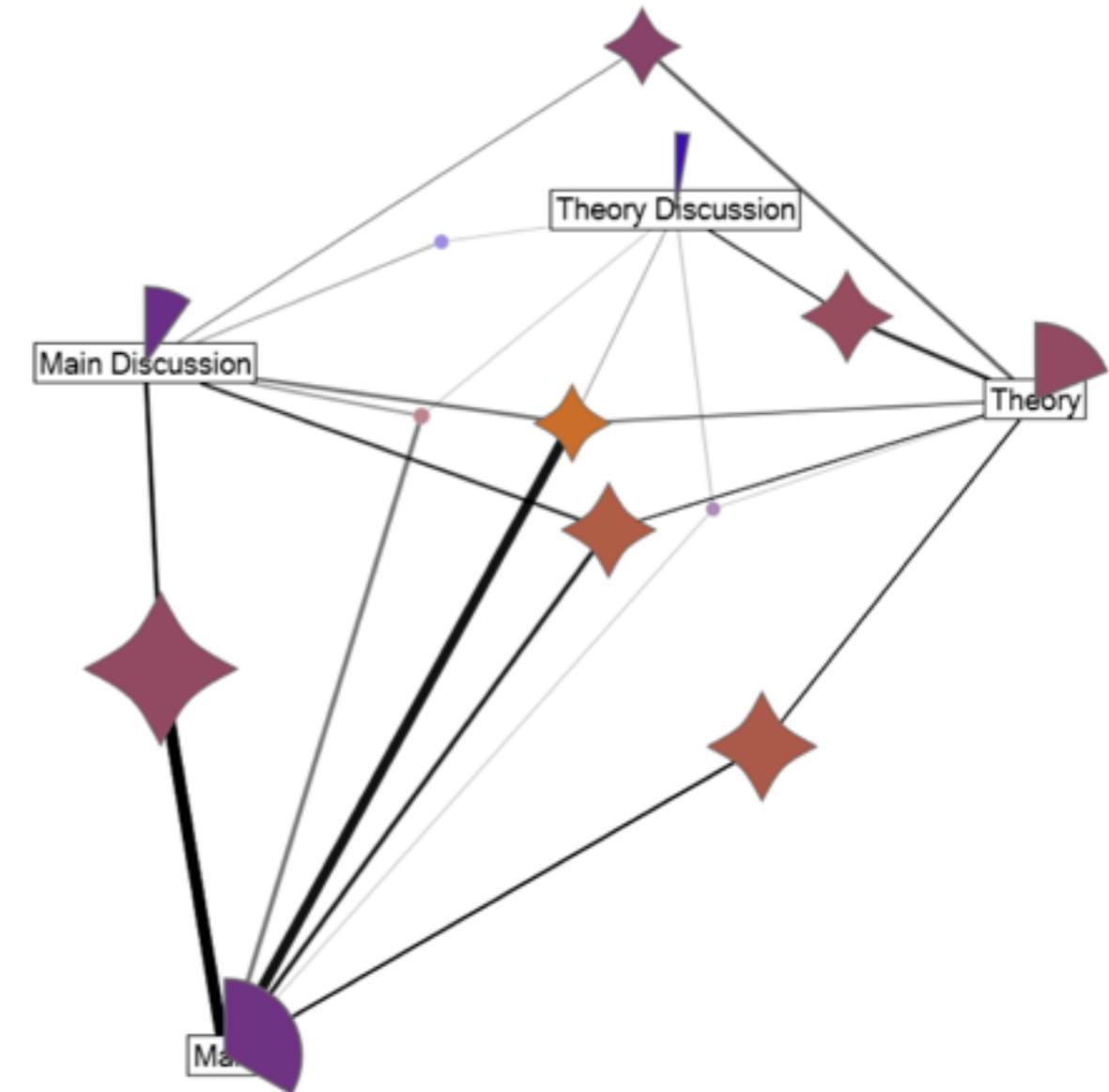
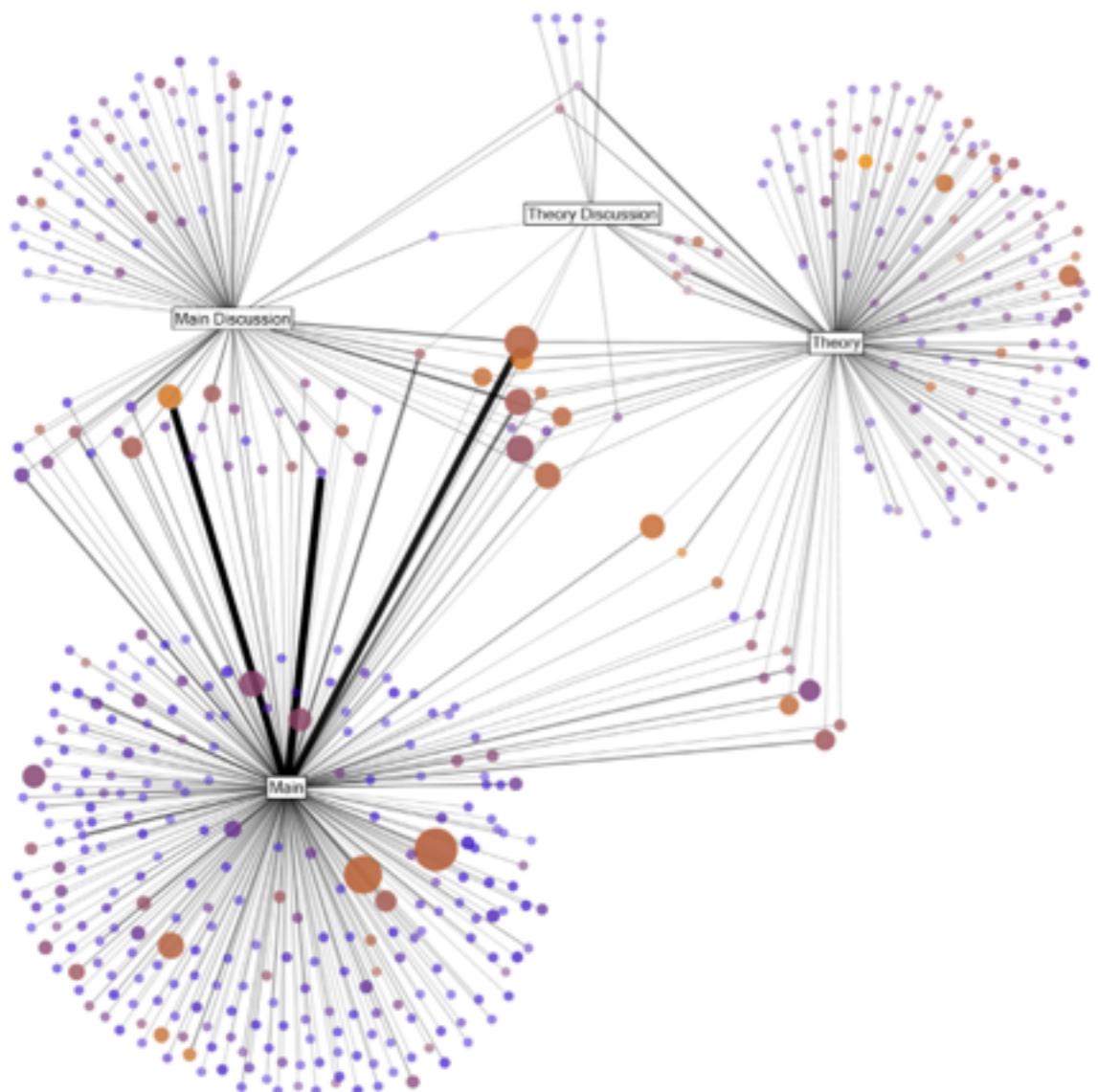


Les Misérables

character
co-occurrence

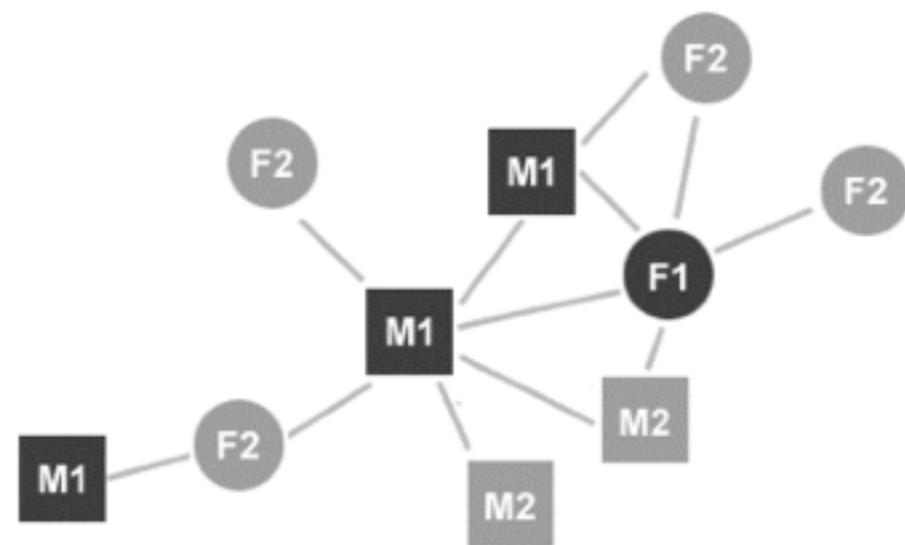


MOTIF GLYPHS

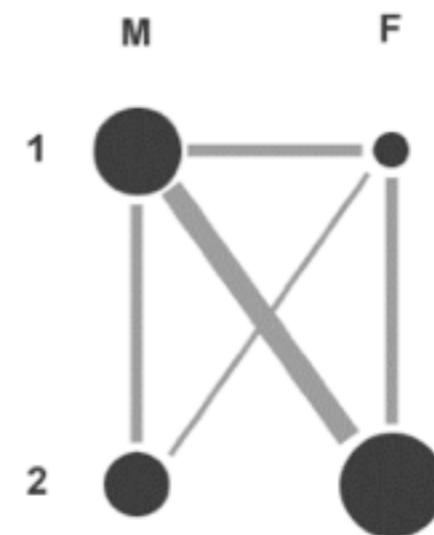


PIVOT GRAPHS

- new graph, derived from categorical node attributes
- 1D or 2D layouts possible
- size of nodes and edges related to number of aggregated original nodes and edges
- scalability through abstraction, not layout algorithm



Node and Link Diagram



PivotGraph Roll-up

today . . .

Tables

Items

Attributes

Networks & Trees

Items (nodes)

Links

Attributes

Fields

Grids

Positions

Attributes

Geometry

Items

Positions

Clusters, Sets, Lists

Items

Tables

Items

Attributes

Networks & Trees

Items (nodes)

Links

Attributes

Fields

Grids

Positions

Attributes

Geometry

Items

Positions

Clusters, Sets, Lists

Items

Text

TEXT

Tables

Items

Attributes

Networks &
Trees

Items (nodes)

Links

Attributes

Fields

Grids

Positions

Attributes

Geometry

Items

Positions

Clusters,
Sets, Lists

Items

Text

?

WHAT DOES IT MEAN TO BE AN "ITEM"?

text data type

- no numbers (implicitly)

- characters: ASCII

- strings

USASCII code chart

b ₇ b ₆ b ₅					0 0 0	0 0 1	0 1 0	0 1 1	1 0 0	1 0 1	1 1 0	1 1 1
b ₄	b ₃	b ₂	b ₁	Column →	0	1	2	3	4	5	6	7
0	0	0	0	Row ↓	NUL	DLE	SP	0	@	P	\	P
0	0	0	1	1	SOH	DC1	!	1	A	Q	a	q
0	0	1	0	2	STX	DC2	"	2	B	R	b	r
0	0	1	1	3	ETX	DC3	#	3	C	S	c	s
0	1	0	0	4	EOT	DC4	\$	4	D	T	d	t
0	1	0	1	5	ENQ	NAK	%	5	E	U	e	u
0	1	1	0	6	ACK	SYN	8	6	F	V	f	v
0	1	1	1	7	BEL	ETB	'	7	G	W	g	w
1	0	0	0	8	BS	CAN	(8	H	X	h	x
1	0	0	1	9	HT	EM)	9	I	Y	i	y
1	0	1	0	10	LF	SUB	*	:	J	Z	j	z
1	0	1	1	11	VT	ESC	+	;	K	[k	{
1	1	0	0	12	FF	FS	,	<	L	\	l	
1	1	0	1	13	CR	GS	-	=	M]	m	}
1	1	1	0	14	SO	RS	.	>	N	^	n	~
1	1	1	1	15	S1	US	/	?	O	-	o	DEL

text data type

- no numbers (implicitly)

- characters: ASCII

- strings

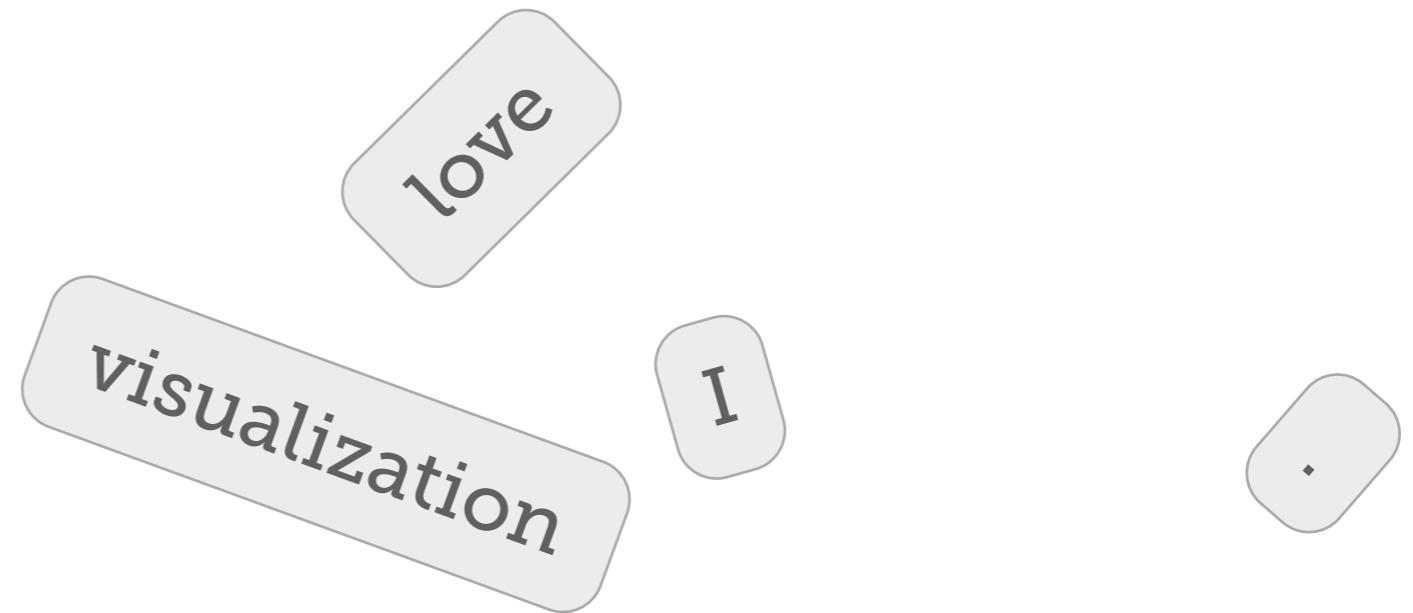
	0	1	2	3	4	5	6	7	8	9	10	11
M	O	n	t	y		P	y	t	h	o	n	

-12 -11 -10 -9 -8 -7 -6 -5 -4 -3 -2 -1
[-12:-7]

USASCII code chart							
b ₇	b ₆	b ₅	b ₄	b ₃	b ₂	b ₁	
0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	NUL
0	0	0	0	1	1	1	DLE
0	0	0	1	1	1	1	SP
0	0	1	0	2	2	2	!
0	0	1	0	2	2	2	SOH
0	0	1	0	2	2	2	DC1
0	0	1	1	3	3	3	DC2
0	0	1	1	3	3	3	ETX
0	0	1	1	3	3	3	DC3
0	1	0	0	4	4	4	EOT
0	1	0	0	4	4	4	DC4
0	1	0	1	5	5	5	ENQ
0	1	0	1	5	5	5	NAK
0	1	1	0	6	6	6	%
0	1	1	0	6	6	6	ACK
0	1	1	0	6	6	6	SYN
0	1	1	1	7	7	7	BEL
0	1	1	1	7	7	7	ETB
1	0	0	0	8	8	8	BS
1	0	0	0	8	8	8	CAN
1	0	0	1	9	9	9	HT
1	0	0	1	9	9	9	EM
B	*	:	J	Z	j	z	
C	+	;	K	[k	{	
S	,	<	L	\	l		
S	-	=	M]	m	}	
S	.	>	N	^	n	~	
S	/	?	O	-	o	DEL	

text data semantics

- words



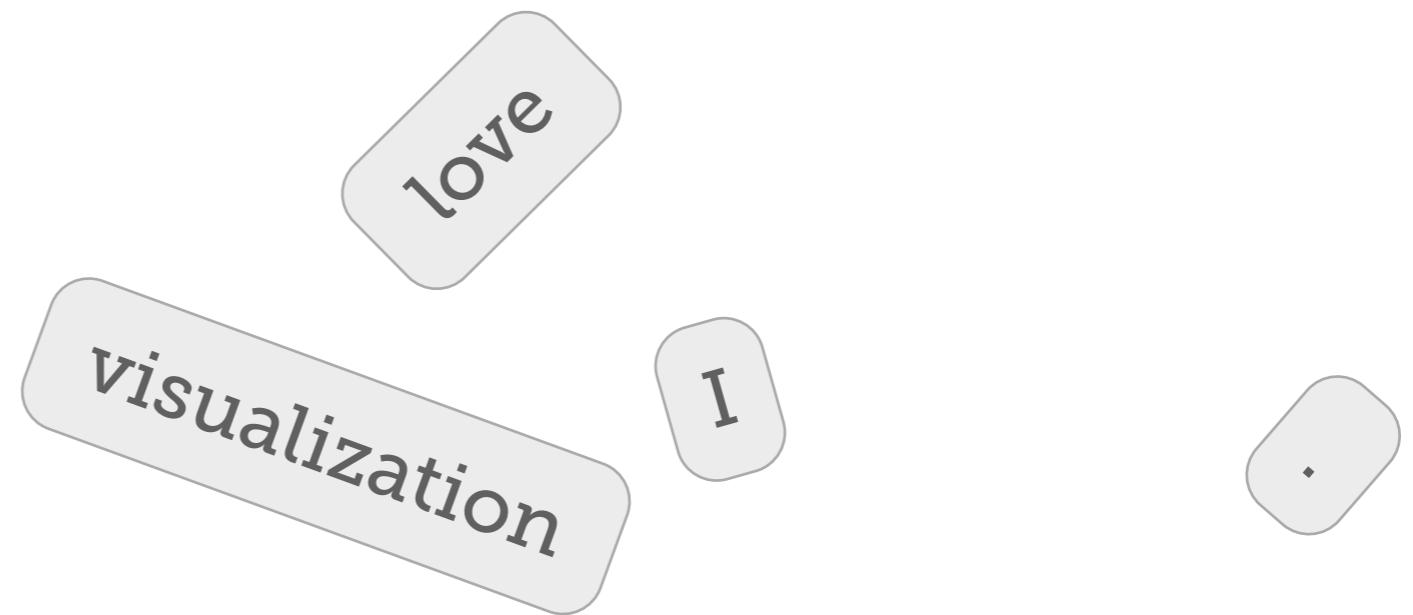
- sentences

- paragraphs
 - chapters

- lines

text data semantics

- words



- sentences

- paragraphs
- chapters

I love visualization.

- lines

text data semantics

- words

- sentences

- paragraphs
- chapters

- lines



```
16 // displays a data set using parallel coordinates
17
18 // dataset info
19 String dataSet = "cars";
20 String fileName = dataSet + ".csv";
21 boolean cluster = true;
22 FloatTable table;
23 float[][] data;
24
25 // row, column info
26 String[] colNames;
27 int col = 0;
28 int colTot;
29 String[] rowNames;
30 int row = 0;
31 int rowTot;
```

text data semantics

- documents
- books
- papers
- webpages
- emails
- twitter post

-corpus: *collection of documents*

Semantics-Based Analysis and Navigation of Heterogeneous Text Corpora: The Purpose News and Blogs Engine

status quo, concepts, and what is available online. This study aims to examine pre-service teachers' search and usage levels by examining their use of online dictionaries as learning tools.

Keywords: text corpus analysis, lexicography

1 Introduction

Many information beyond keywords is information. This is also referred to as "fog" in Google's handling the Web's annotations or the encyclopedias. The next (1991). Both are, information needs a kind of different info.

This situation is in a more serious and of different kinds.

Author's Note: This article is based on a study conducted at the University of Çukurova, Adana, Turkey. The authors would like to thank the anonymous reviewers for their useful comments and suggestions.

Examining the Use of Text Corpora and Online Dictionaries as Learning Tools: Pre-Service Teachers' Perspectives

Sibelcan Eroğluşlu-Grigorian, Garbis Asanjanian,
and Sümeyra Tuncelikli

<p>ICT integration in rescue technology programmes and how they do use them fashion, more as marketers were from or progressive wing organisations, suggest political beliefs. Within classrooms there are, nevertheless, styles, pedagogical patterns and the ways in which teachers relate to content and even philosophies about</p> <p>.....</p> <p>S. Tuckkann-Ong Department of French caustic@optonline.net</p> <p>A. Ameliehane (ed.) DRC 75 10001978-1-6</p>	<p>Searching</p> <p>Thomas G. Christou University of Cyprus christou@hua.ac.cy</p> <p>allegories. The researchers in the case of one project found the teachers to be engaged actively in the process.</p> <p>3. Learning environments</p> <p>Information systems project have been the biggest challenge in the last decade. The Internet revolution has been using IT IN THE classroom. This leads naturally to the question of what is the role of IT in education.</p> <p>Authoring tools based largely on hypermedia and text, have been the primary media including text and this used widespread use of hypertext and multimedia and many other media.</p> <p>The project aimed at developing an environment for learning of the subject which were related to the real world throughout the curriculum.</p> <p>4. Learning environments</p>
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Chapter 3
Automatic Mapping of Social Networks
of Actors from Text Corpora: Time Series
Analysis

James A. Pawowski and Scott Urdzla

Abstract. To test hypotheses about presidential cabinet network centrality and presidential job approval over time and to illustrate automatic social network identification from large volumes of text, this research tested the social networks among the cabinets of Presidents Reagan, G.W. Bush, Clinton, and G.W. Bush based on the members' co-occurrence in news stories. Each administration's data were sliced into time intervals corresponding to the Outag's presidential job approval ratings. It was hypothesized that when the centrality of the president is lower than that of other cabinet members, job approval ratings are higher. This is based on the assumption that news is primarily negative and when the president stands above the other cabinet members in network centrality he or she is more likely to be associated with the negative news coverage in the minds of members of the public. The hypothesis was tested using social network analysis and G.W. Bush having maximum network status during his term. The results support the hypothesis over

**Automatic Analysis of Large Text Corpora –
A Contribution to Structuring Web Communities**

Gerhard Heuer, Univ. Quantenoff, and Christian Wyß

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Computer Science Institute, Natural Language Processing Department
Augsburg University, D-8613 Augsburg
(heuer, quantenoff, wyss)@informatik.uni-augsburg.de

Abstract. This paper describes a corpus linguistic analysis of large text corpora based on webcrawling with the aim of extracting semantic relations from unstructured text. We regard this approach as a suitable method for generating and structuring information about Web communities. Starting from a short description of our corpora as well as our language analysis strategy, we discuss in detail the semantic extraction of relations with the help of examples and also present some first results of our system for extracting semantic relations from Web communities. We conclude with a brief discussion of automatic approaches to semantic analysis.

semantic analysis approaches
is called to map the networks,
situation among nearly 1200
networks of political power
in political economy terms
A, overlapping application of
functional levels has placed

1997 W. Brüderlin & Co., Chicago.
95

the analysis of a corpus (continued).

1. Introduction

Constructing a Large Scale Text Corpus Based on the Grid and Trustworthiness

Poliqin Li^{1,2}, Qiuming Zhu¹, Peiqi Guo¹, and Geoffrey C. Bow³

¹School of Computer Science & Technology, Fudan University, Shanghai 200433, China
²Computer Grid Lab, Institute for Future, Shanghai 200031, China
³IBM T.J. Watson Research Center, Yorktown Heights, NY 10598, USA

Abstract. The construction of a large scale corpus is a hard task. A novel approach is developed to automatically build a large scale text corpus with low cost and short building period based on the trustworthiness. That means users can problems. How can automatically build a large scale text corpus on the Web and how to protect copyright in the corpus. As GCD provides the infrastructure for processing large scale data, our approach uses GCD to collect and process language resources on the Web in this paper. That is made our automatically language material. It is the way according to their own knowledge, and checks them whether they would be used freely or not. Our approach is especially useful for those who have no much time to build a large scale corpus. Therefore it is helpful to build a large scale text corpus.

text data semantics

- documents
- books
- papers
- webpages
- emails
- twitter post

Semantics-Based Analysis and Navigation of Heterogeneous Text Corpora:
The Purpose News and Blogs Engine

Berntz Bonsuett and Daniel Trüppen
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Faculty of Information Technology, Technische Universität Berlin,
D-1058 Berlin, Germany
<http://www.cs.kuleuven.ac.be/~berntz/>

Abstract. More information sites such as news services and search engines offer options beyond keyword search to help people group and identify relevant sources of information. However, these features are often limited to the identification of document titles (theirs or another identifier) or specific criteria that are specific ("Christian documents"). Here, this research focuses on the task of grouping the documents comprising and summarizing different information systems that is related to Information Discovery.

In this paper, we describe *Purpose*, a system that provides users with a navigation interface using an hierarchical, semantic-based classification scheme. This allows users to explore, cluster, and filter news items. Clustering, association, navigation, search and filtering are supported by semantic web technologies and blogs for news items.

Keywords: text mining, analysis, visualization, classification

1 Introduction

Many information sites beyond keyword search offer options beyond keyword search to help people group and identify relevant sources of information. These include news services and search engines such as Google, handling the Web's annotations or the search engines. The search engines (such as Google, Bing, etc.) look for information needs involving different information systems.

This situation is in a more semantic way of different information systems.

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Nikola Erokhina-Otterup, Gert-Johan van den Berg, and Sander Tijskens
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2 Examining the Use of Online Dictionaries by Pre-Service Teachers

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<http://www.uwo.ca/edcp/primary/afarrell.html>

Chapter 3
Automatic Mapping of Social Networks of Actors from Text Corpora: Time Series Analysis

James A. Obowicki and Noah Capels

Abstract. To test hypotheses about presidential cabinet network centrality and press job approval ratings, this research used social network analysis and time series analysis from large volumes of text. This research tested the social networks among the cabinets of Presidents Reagan, G.W. Bush, Clinton, and G.W. Bush based on the frequency co-occurrence in news stories. Each administration's data was sliced into time intervals corresponding to the Cabinet presidential approval polls to synchronize the social networks with presidential job approval ratings. It was hypothesized that when the centrality of the president is lower than that of other cabinet members, job approval ratings are higher. This is based on the assumption that news is generally negative and when the president stands above the other cabinet members in network centrality he or she is more likely to be associated with the negative news coverage in the minds of members of the public. The hypothesis was supported for G.W. Bush during his first term and G.W. Bush during his second term. This having automatic network analysis tools testing hypotheses over time intervals.

Gottard Heyer, Uwe Quasthoff, and Christian Wylff
University of Leipzig
Computer Science Institute, Natural Language Processing Department
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Chelsea Young @YoungChelsMarie
@EliYoungBand can't wait to see you guys here in Buffalo,NY at the @1065wyrk Taste of Country in June! PS-My daddy's name is James Young! :)

Kat Przybyla @katprz
Who's coming to the @1065wyrk Taste of Country? A trifecta of southern hotties -- @ericchurch @JoeNichols @EliYoungBand [buffalo.com/entertainment/...](http://buffalo.com/entertainment/)

Alica Wiedenbeck @leesh06
@Eric_Church at @1065wyrk Taste of Country this year.... dying. #Cantwait

Michele L. McDaniel @michelemcd13
absolutely thrilled that @ericchurch will be headlining the @1065wyrk taste of country! Only 135 days! (but who's counting?!)

Jaime Lynn @JaimeLynn478
@ClayModen @1065wyrk SOOOOOOO excited for TOC this year...def bringing my 5yr old, we are HUGE Eric Church fans!!!

Jake Rosen @JRosen26
To say @1065wyrk got taste right this year would be an understatement #ericchurch

- corpus: collection of documents

For the analysis of texts
In a case study it will be shown how to build the knowledge base of an English conversational agent for educational purpose from a child story that can answer questions about characters, facts and episodes of the story. A discussion of the main linguistic and methodological issues and further developments is offered in the final part of the chapter.

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© Springer-Verlag Berlin Heidelberg 2013

西遊平景 神奈川縣
葛西





single document

Tag Clouds / Word Clouds

abstract accepted analogue applications applying attuned bar burgeoning challenging chapters chart collections combine communicate conducted convert **data** date difficult discussed earlier effectively end evaluation evocative familiar field focus focused form **general goal** graph highly human hundreds ideas images improve

information innovative insight kinds line makes means

meta-analysis nature new numbers order ost perceive perceptual points positive problems providing purpose range rapidly read reading reasons representations **results** retrieval robust **search** shortciten{chen2000esi} shortcite{larkin1987dsw} shown space

studies successful system table task tasks **text** textual time translate underlying

usability vibrant visual visualization visually web wide widely

<http://www.tagcrowd.com>



Visualizations : definitions of visualization word tree

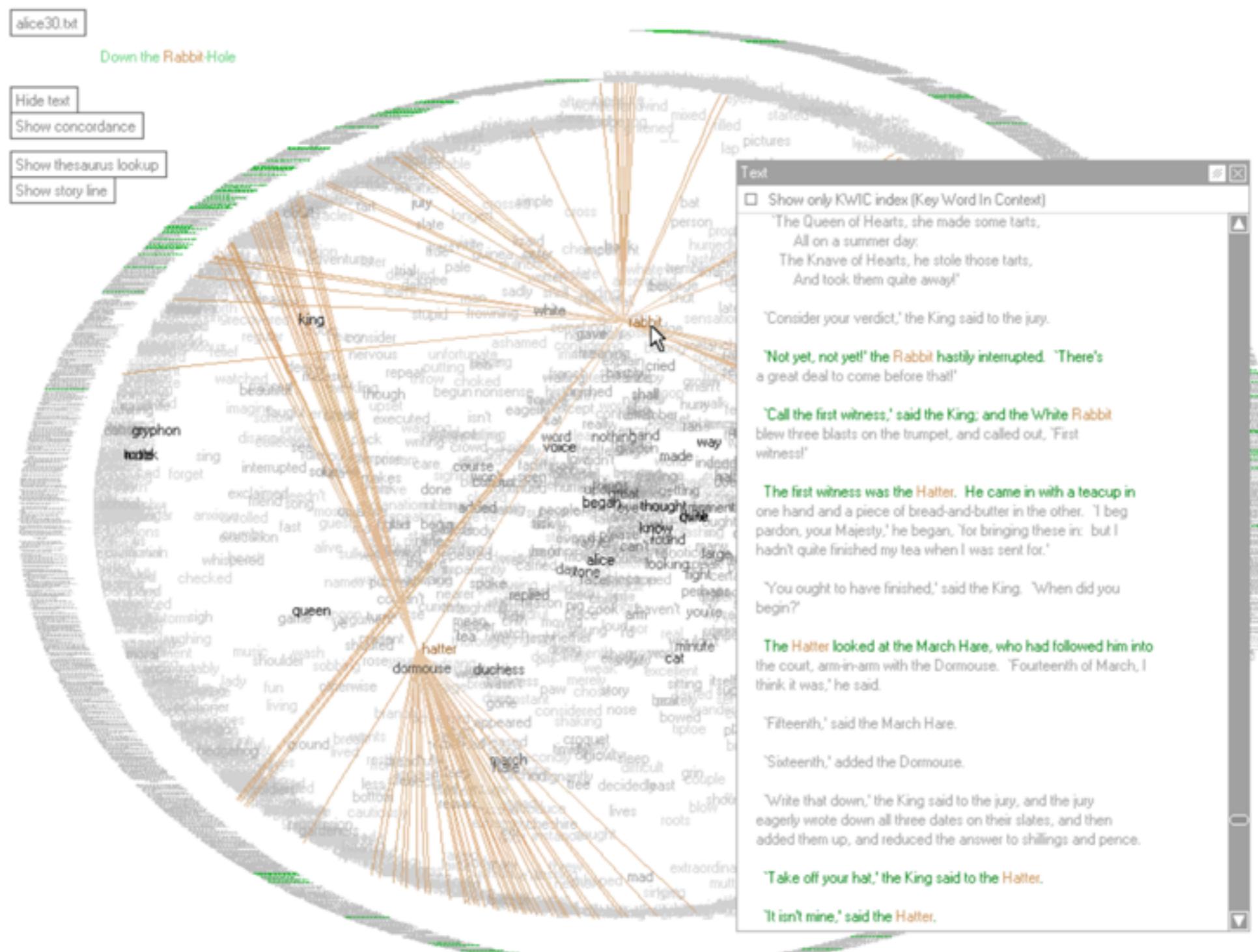
Uploaded by: mhalle

Created at: Wednesday May 21 2008, 11:37 PM

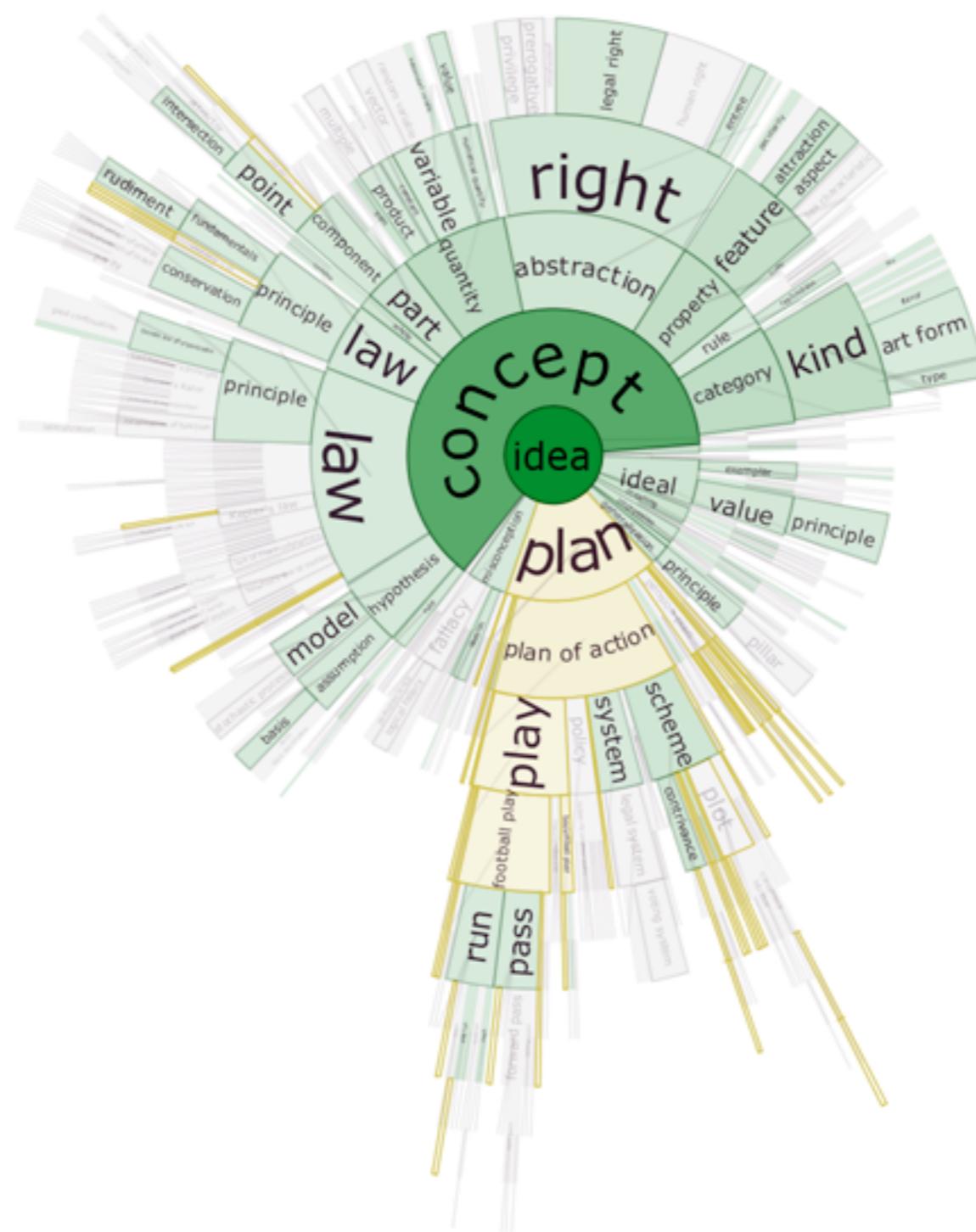
Tags: **text**



Text Arc

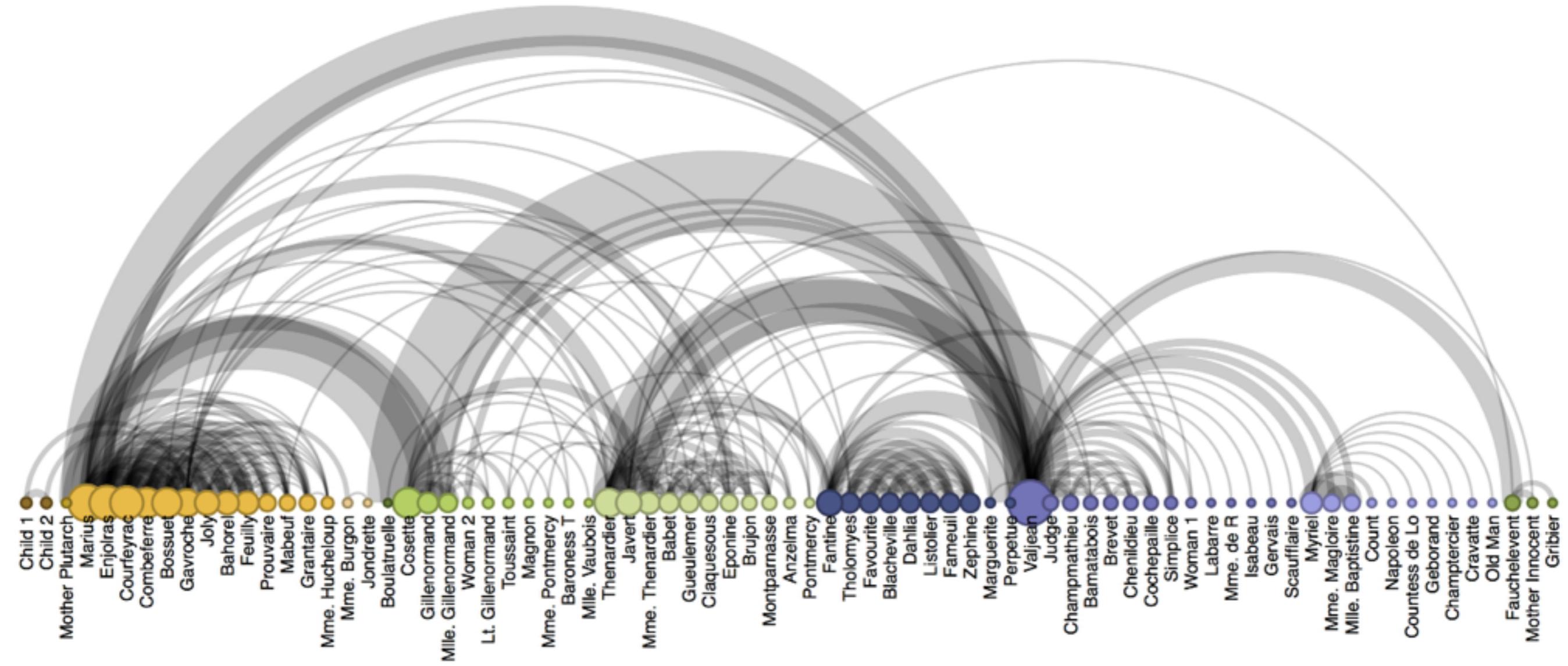


DocuBurst



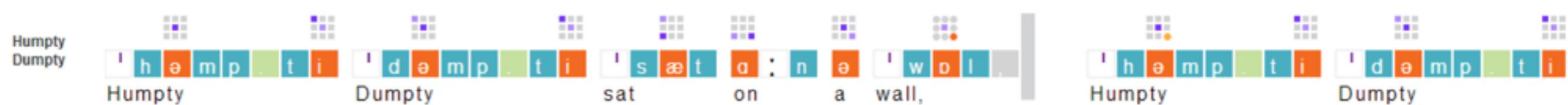
Collins, Carpendale, Penn 2008

Arc Diagrams



Analysis of the Characters from *Les Misérables*: <http://mbostock.github.io/protovis/ex/arc.html>

Rule-Based: Poetry



collection of documents

Parallel Tag Clouds to Explore and Analyze Faceted Text Corpora



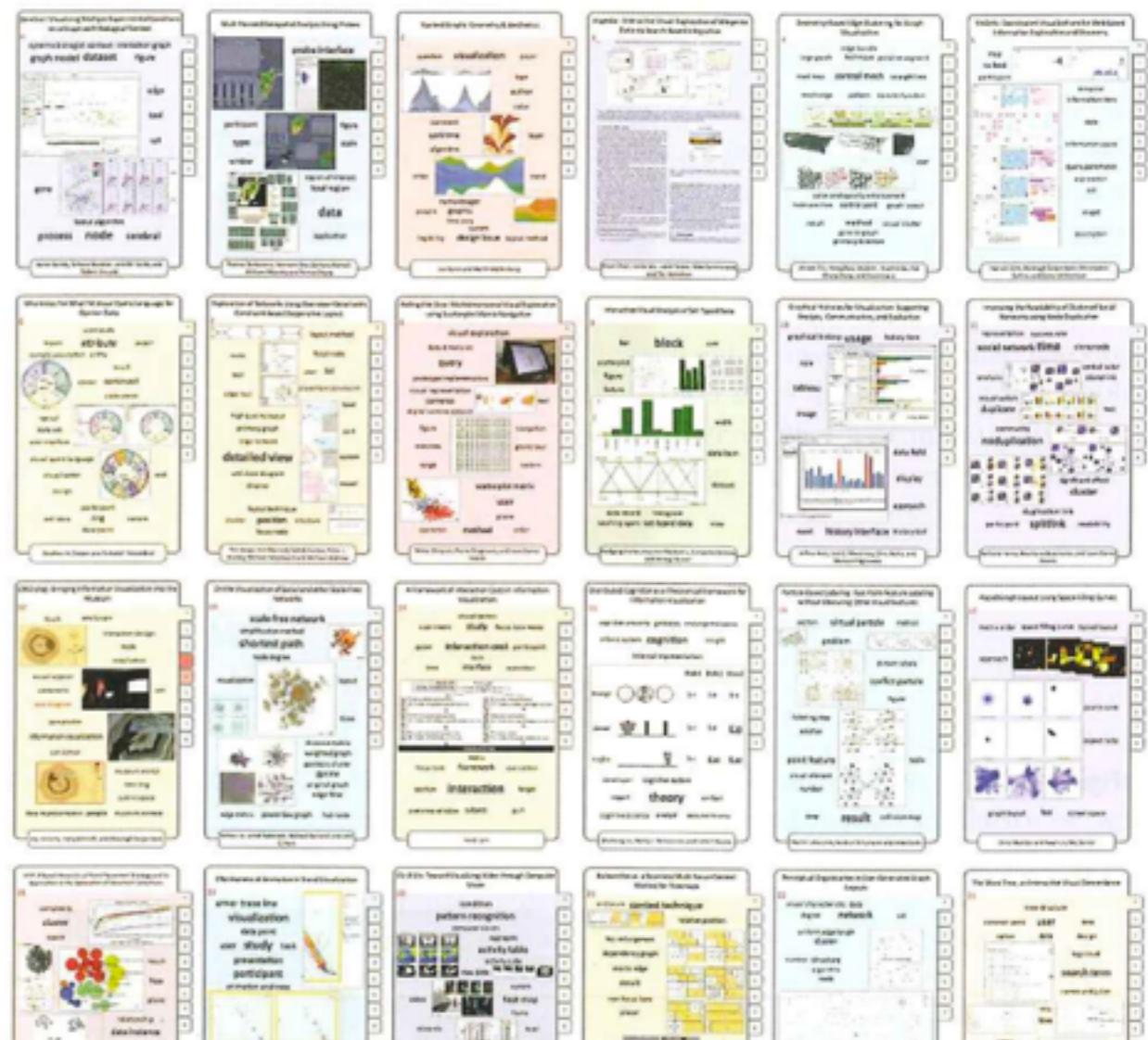
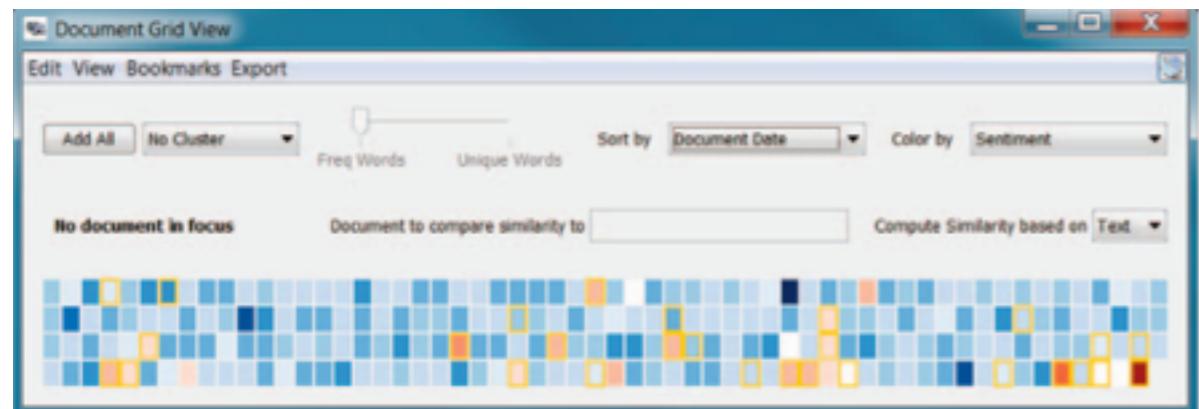
Christopher Collins

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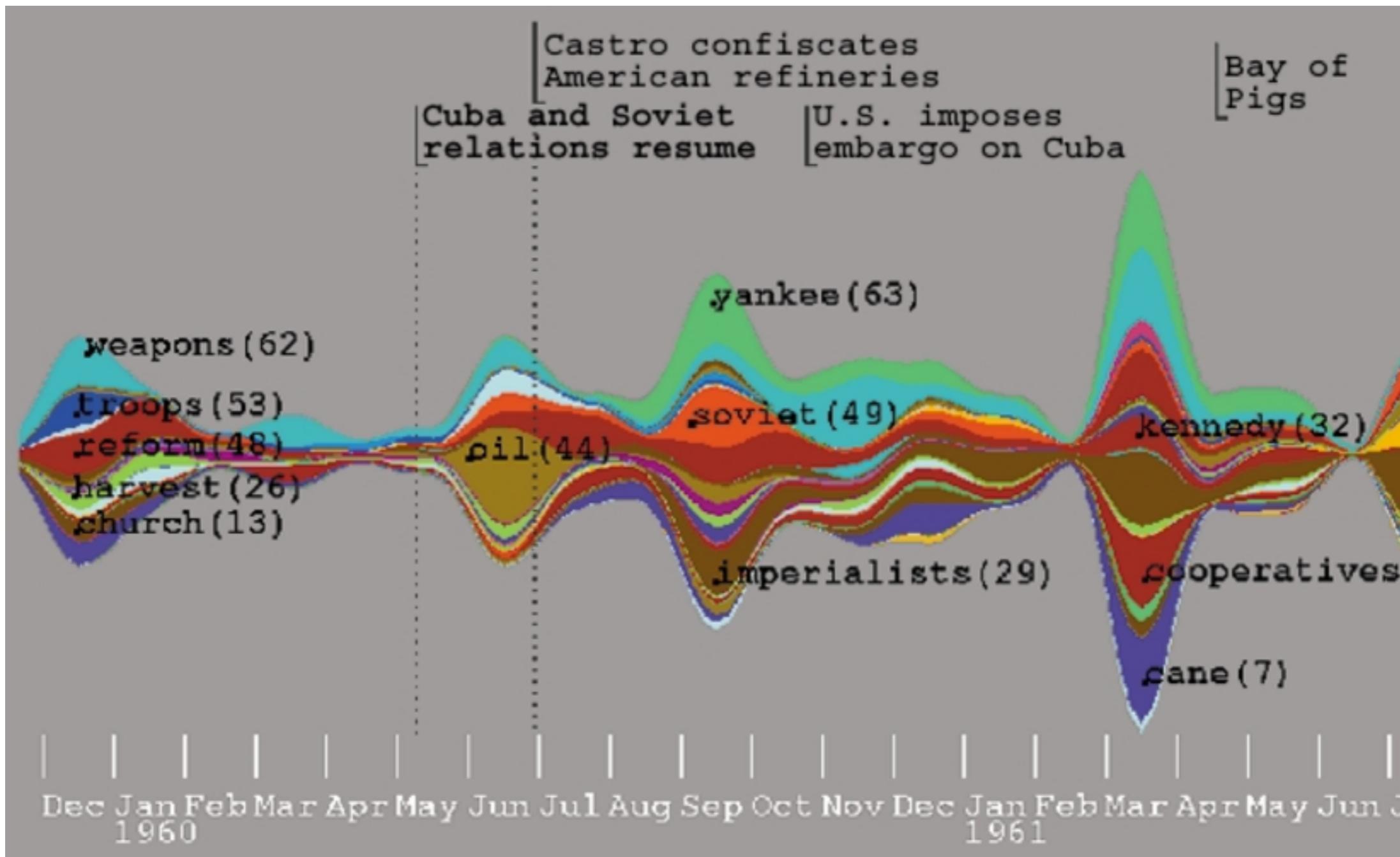
25

2,451

Document Cards (small multiples)



Showing Temporal Relationships: ThemeRiver (Stream Graph)



Jigsaw: Many Linked Views

Visual Analytics Support for Intelligence Analysis Case Study: The 9/11 Report

Carsten Görg

Youn-ah Kang

Zhicheng Liu

John Stasko



Information Interfaces Group
Georgia Institute of Technology

Jigsaw: Many Linked Views

Visual Analytics Support for Intelligence Analysis Case Study: The 9/11 Report

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SETS

Tables

Items

Attributes

Networks & Trees

Items (nodes)

Links

Attributes

Fields

Grids

Positions

Attributes

Geometry

Items

Positions

Clusters, Sets, Lists

Items

Text

?

thought experiment...

- item: Lego

- attributes



thought experiment...

- item: Lego

- attributes

- color
- height
- width
- length
- shape



dataset: option 1



dataset: option 2

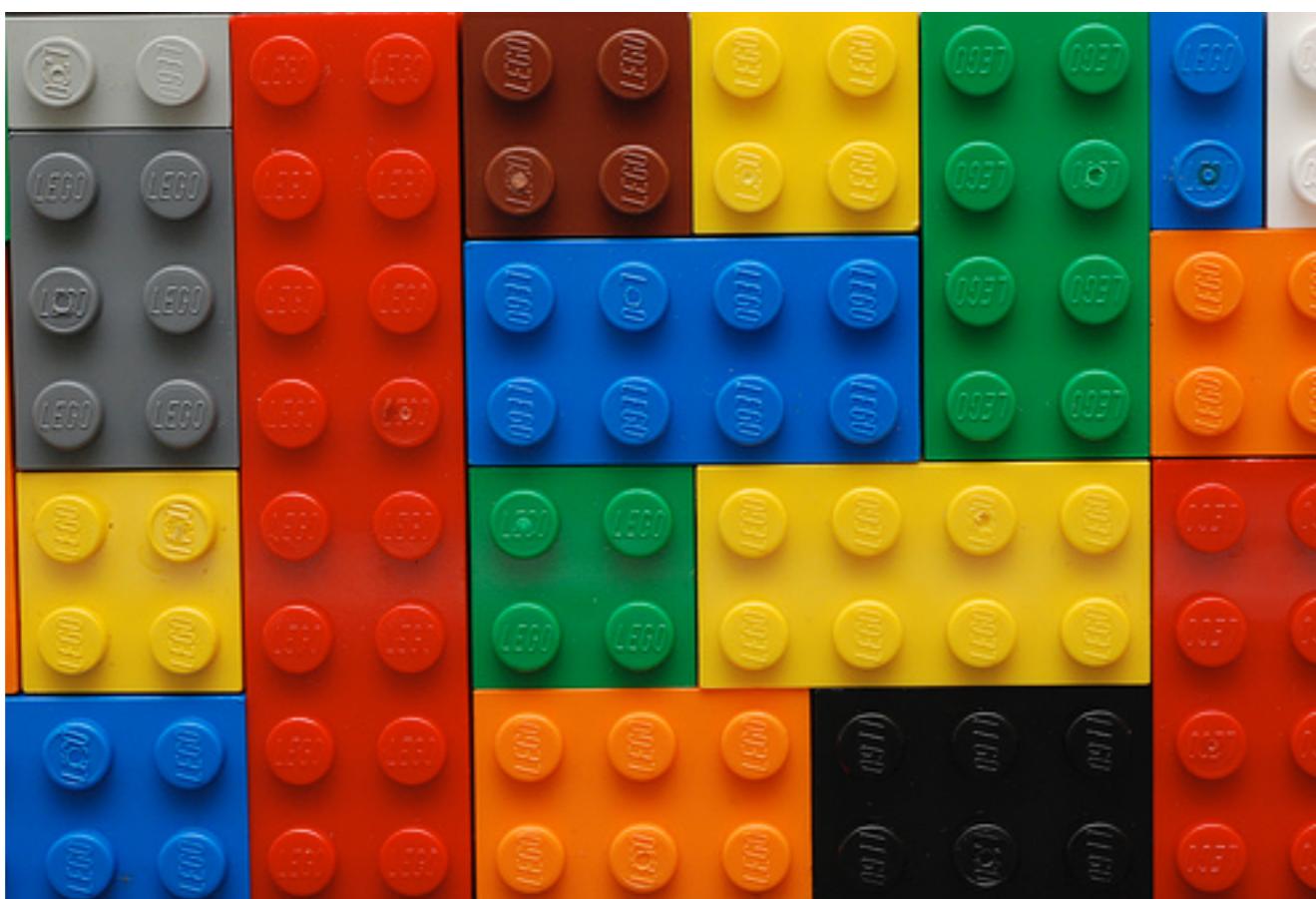


dataset: more realistic



dataset

-where do we start? we need to organize! but, how?



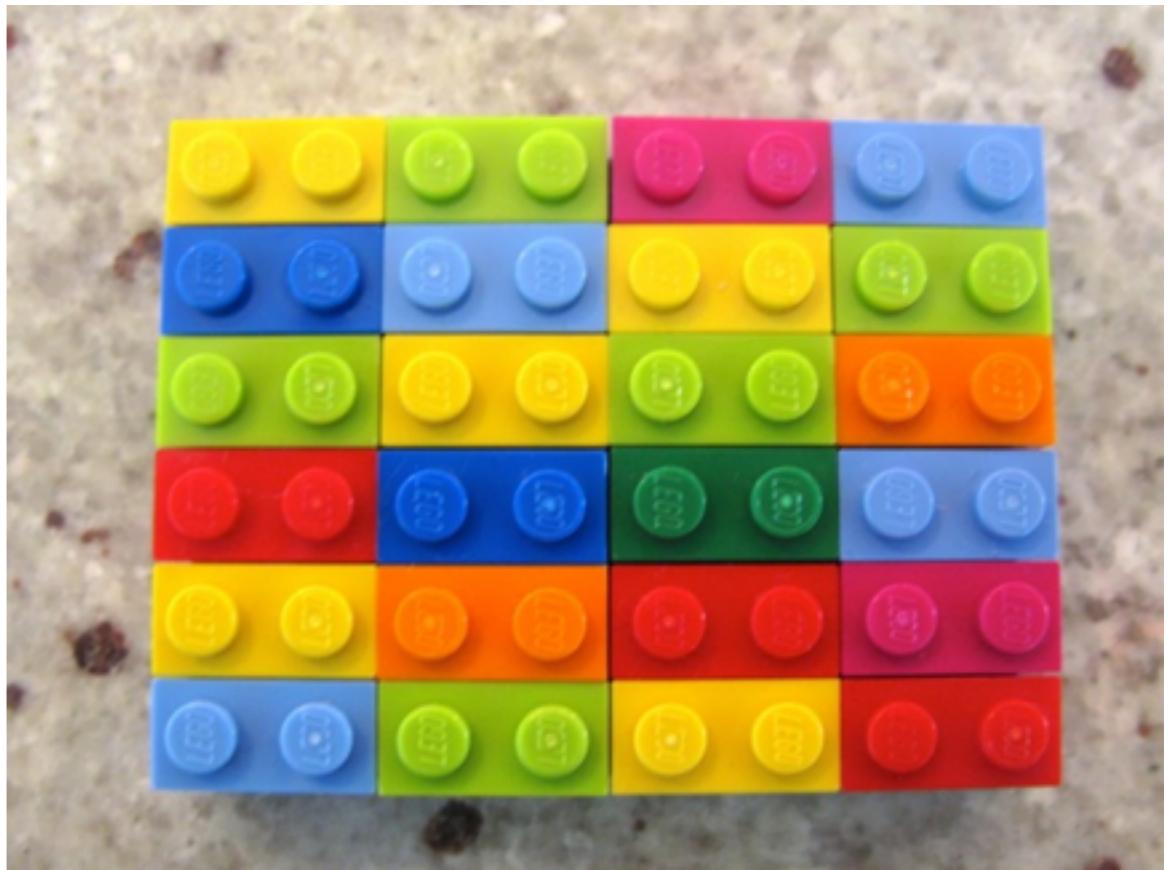
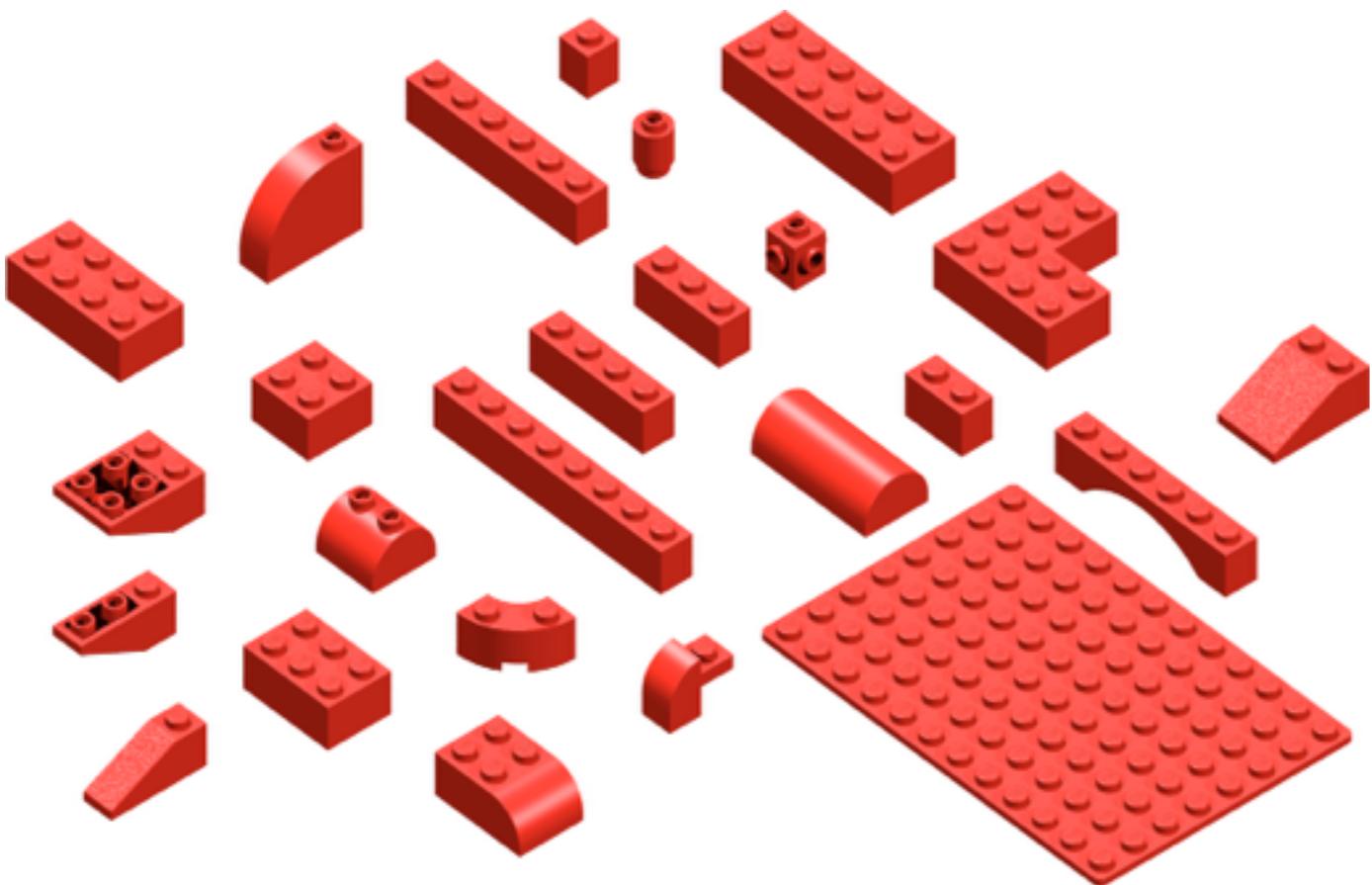
dataset

- sort by color



dataset

- sort by size, shape



dataset

- task: organization

- drawbacks?



dataset

- organization leads us to a set problem
- so what are sets?

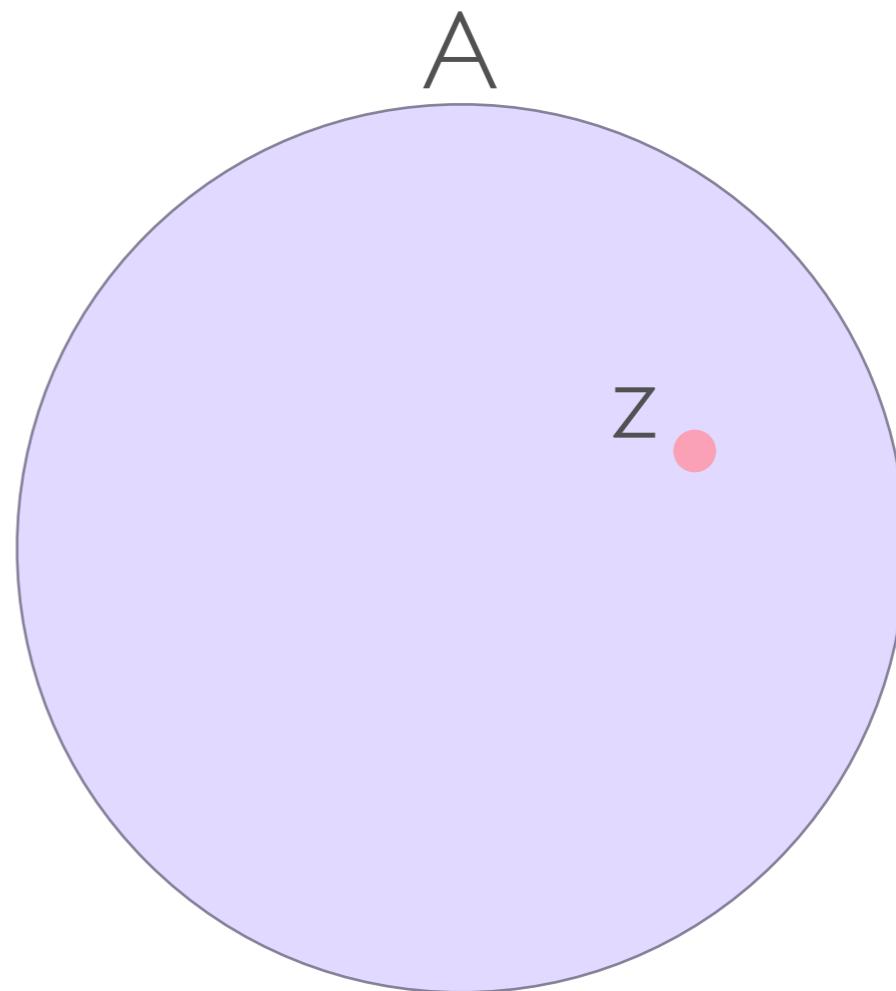
set theory

- **set**

- a collection of objects
- some set: A

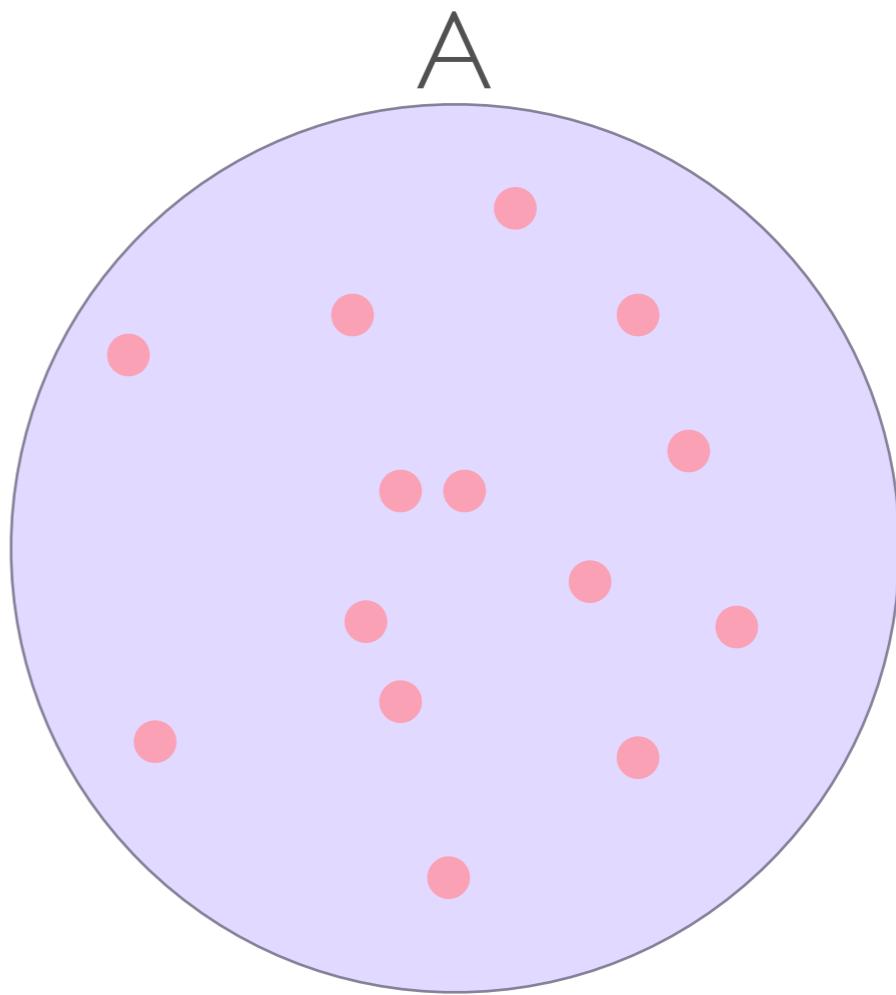
- **object**

- some object: z
- $z \in A$



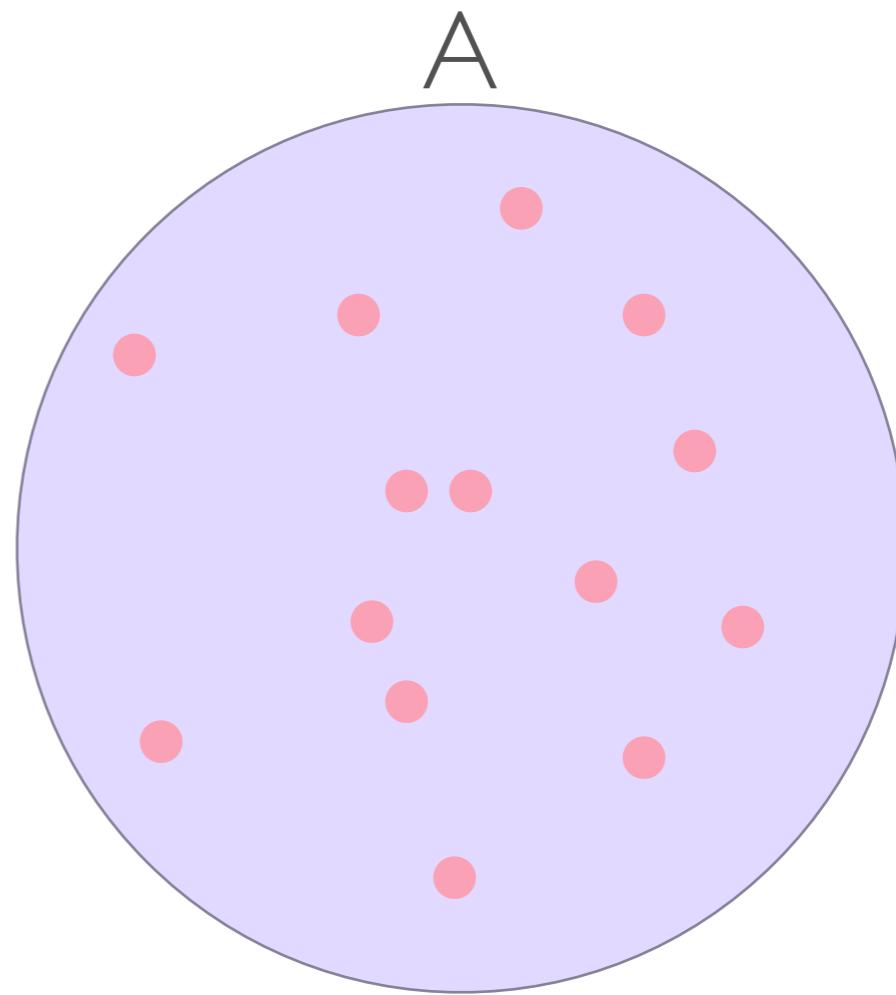
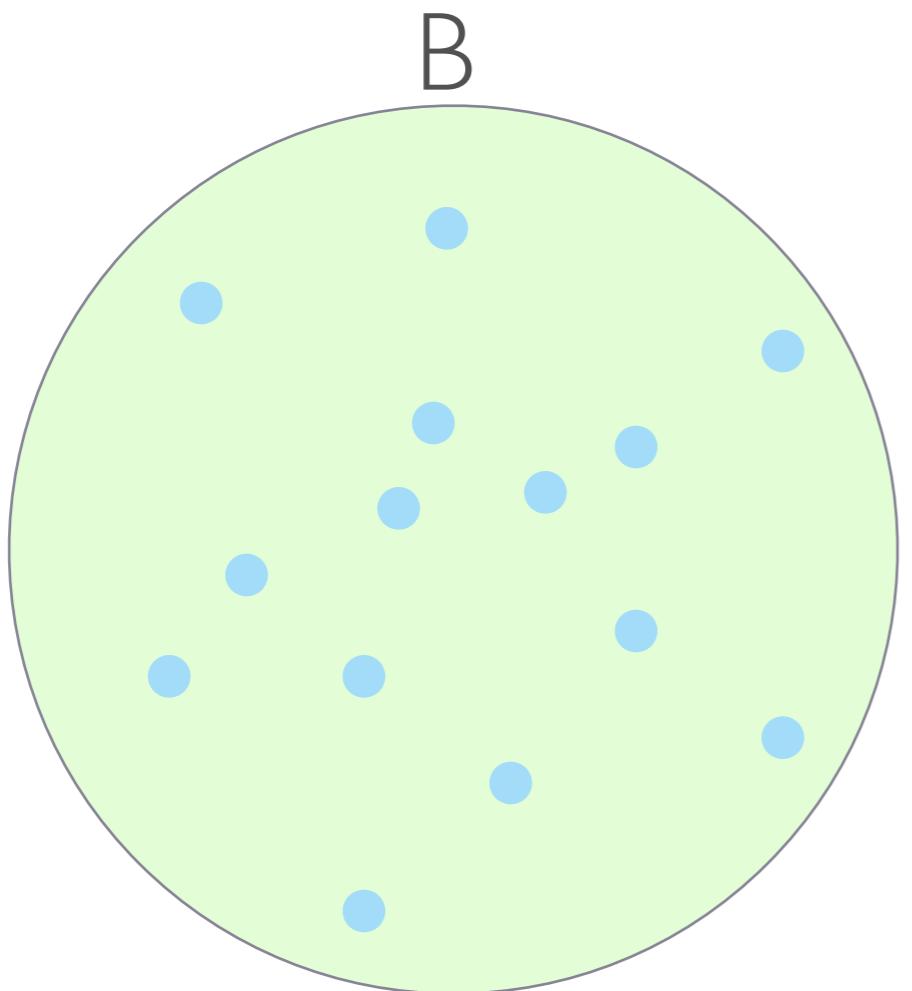
set theory

- **set**
 - a collection of objects
 - some set: A



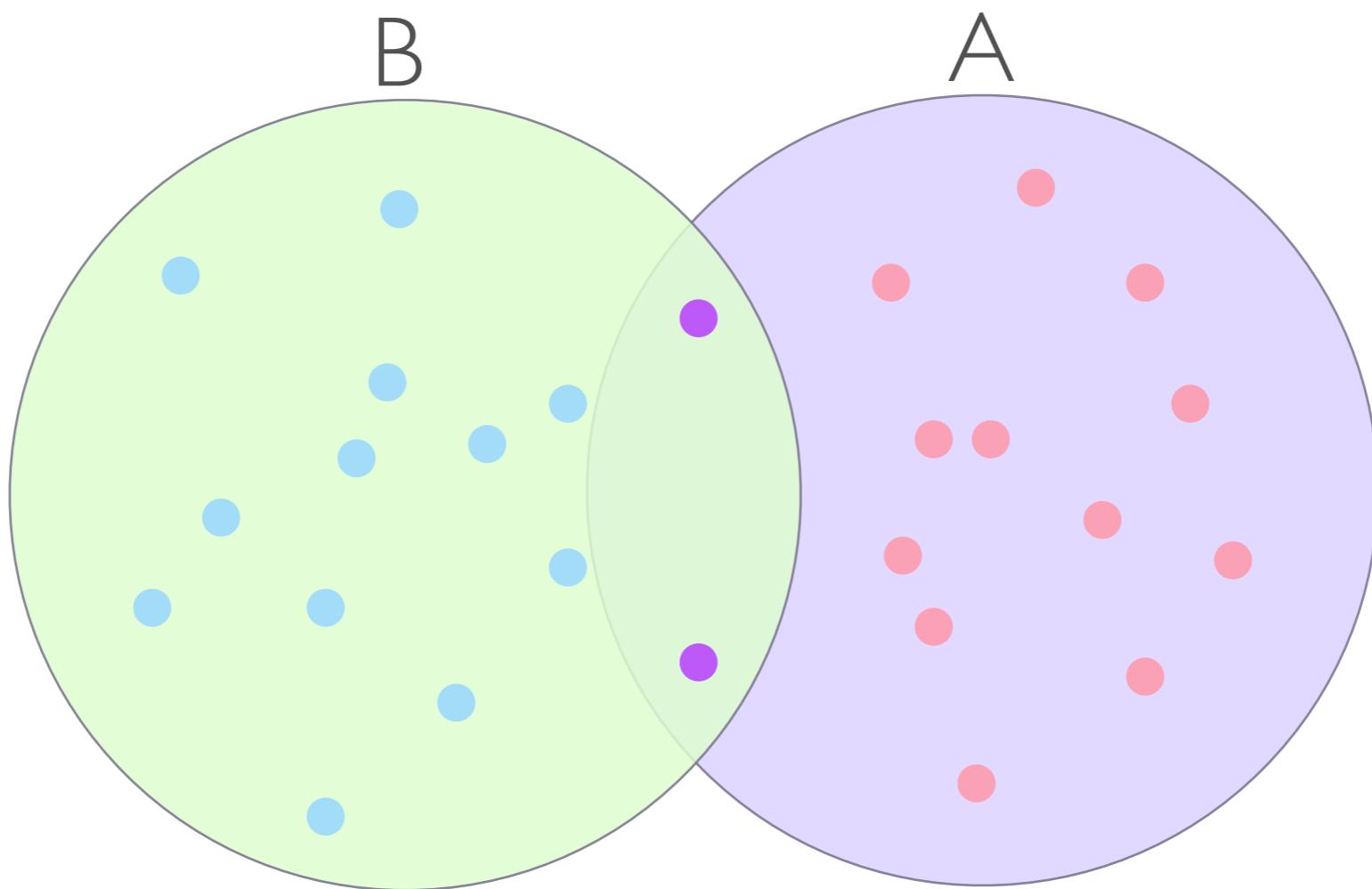
set theory

-multiple sets: A & B



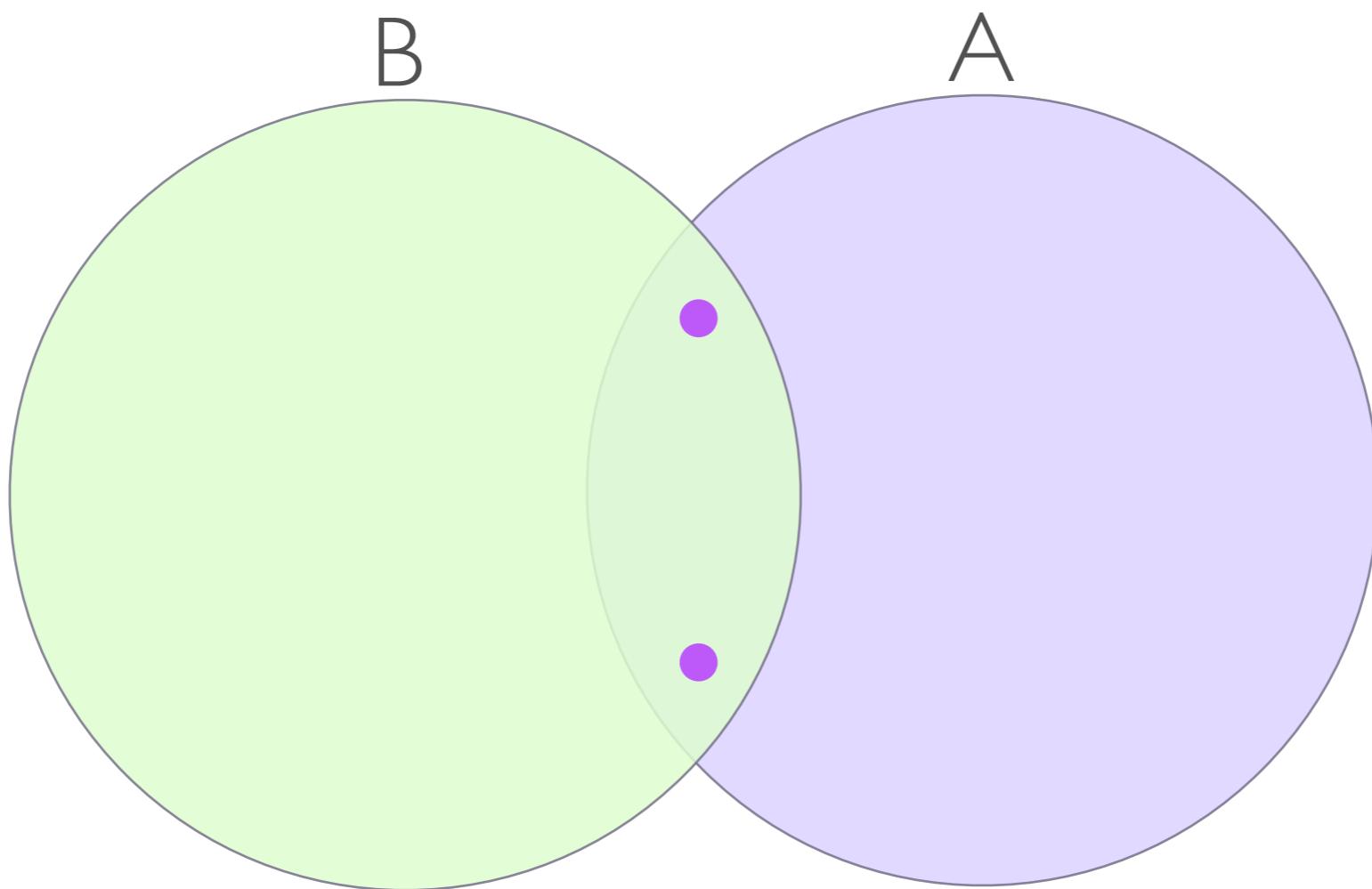
set theory

-union: $A \cup B$



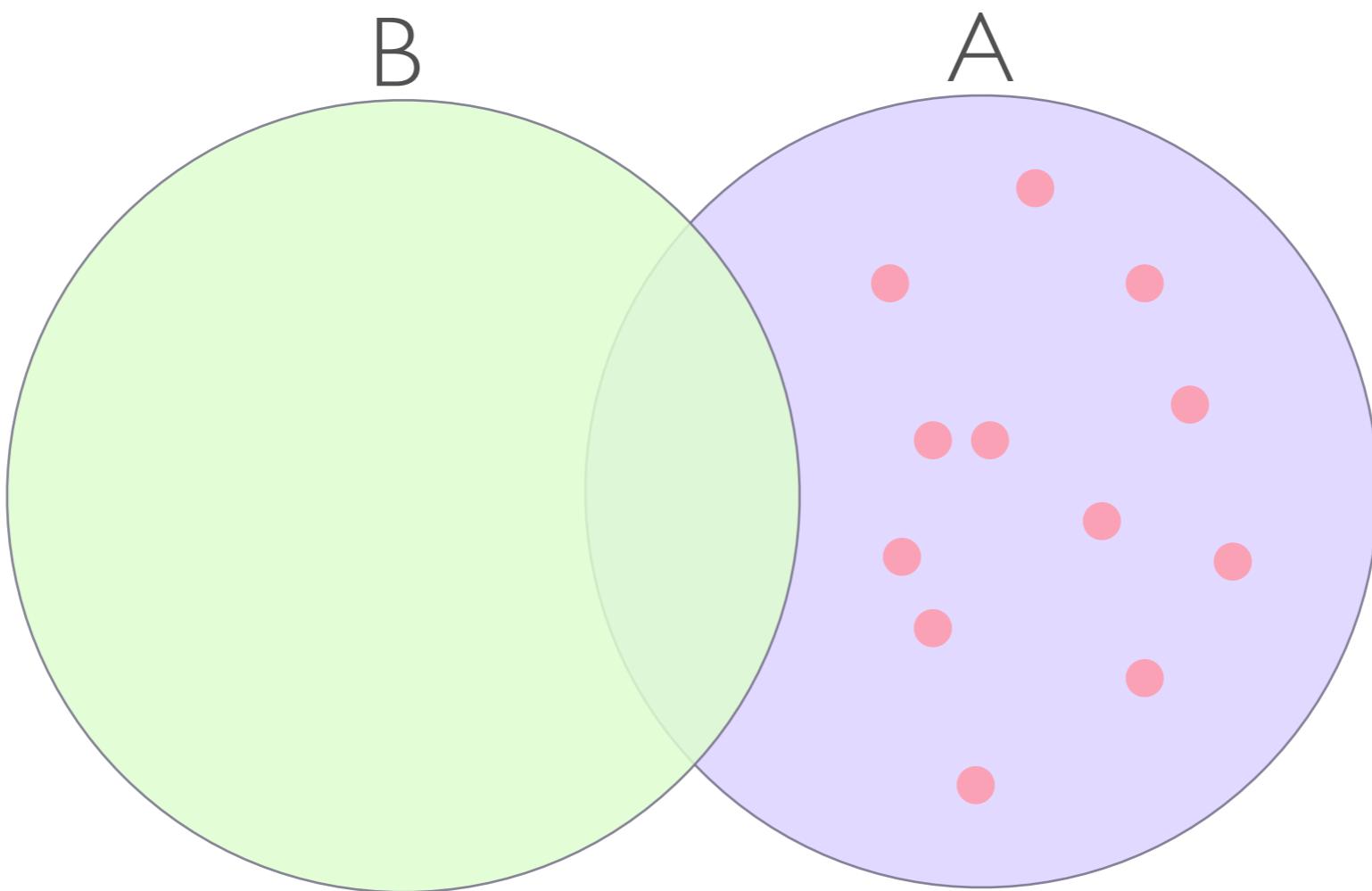
set theory

-intersection: $A \cap B$



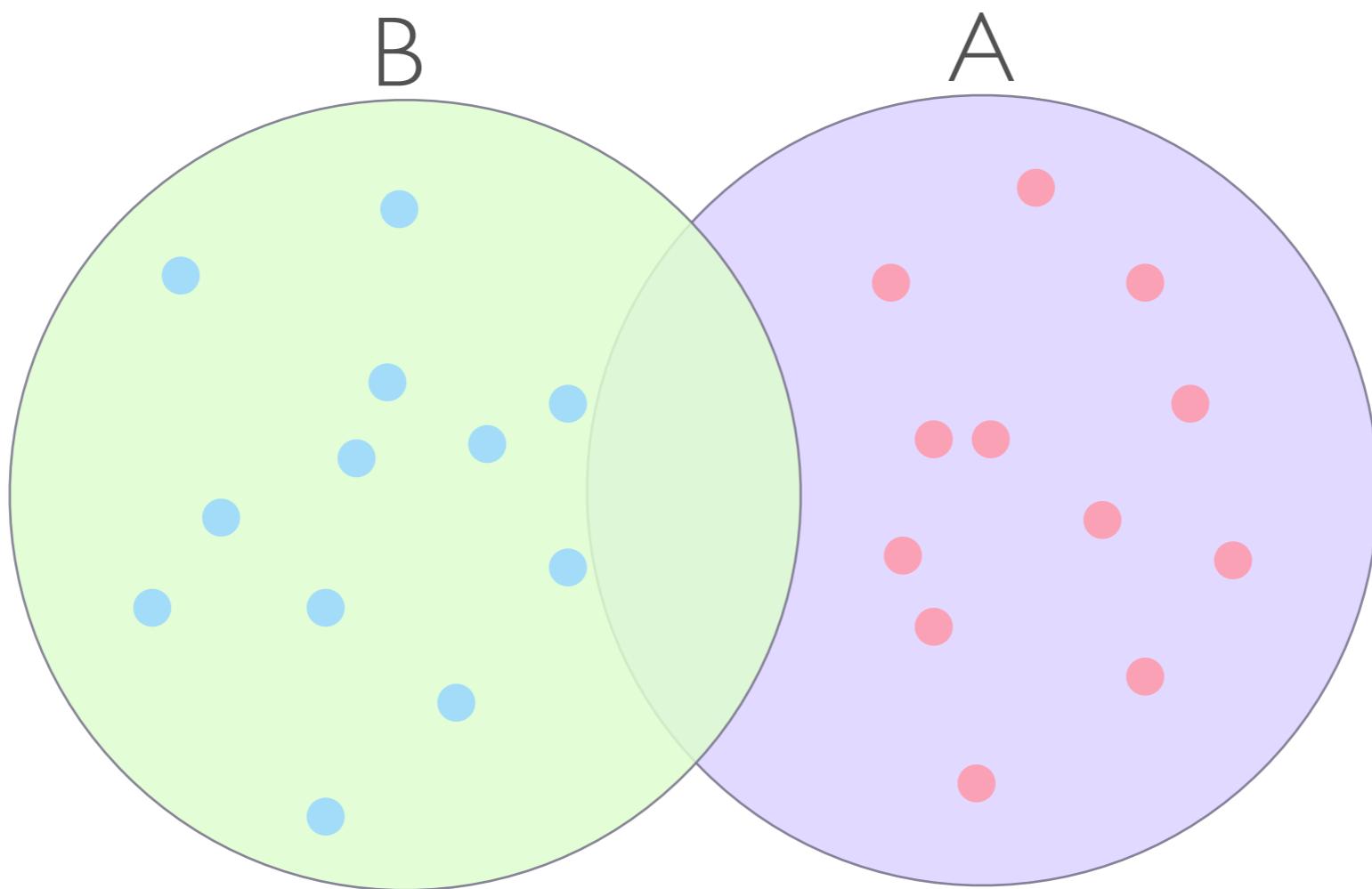
set theory

-set difference: $A \setminus B$



set theory

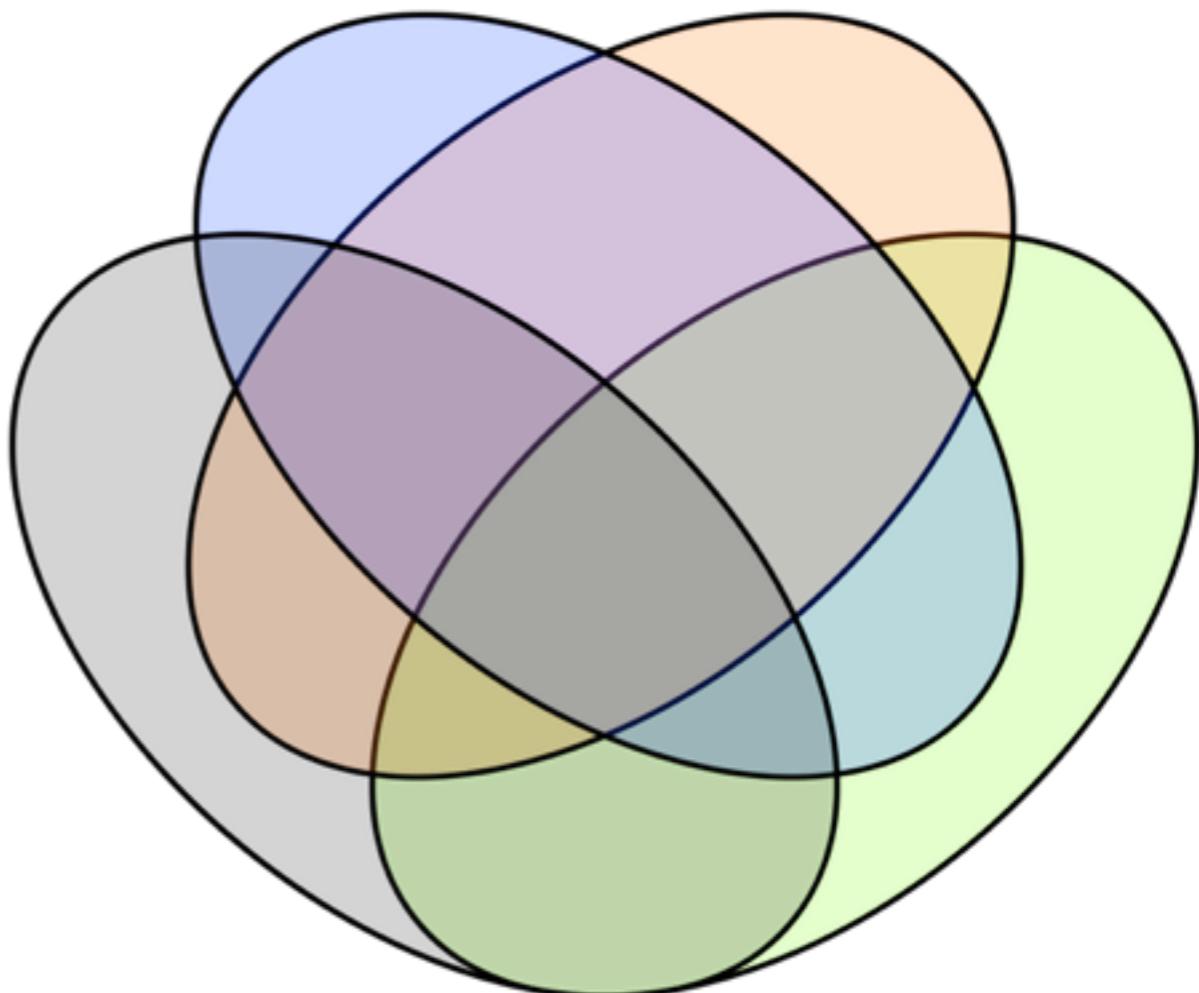
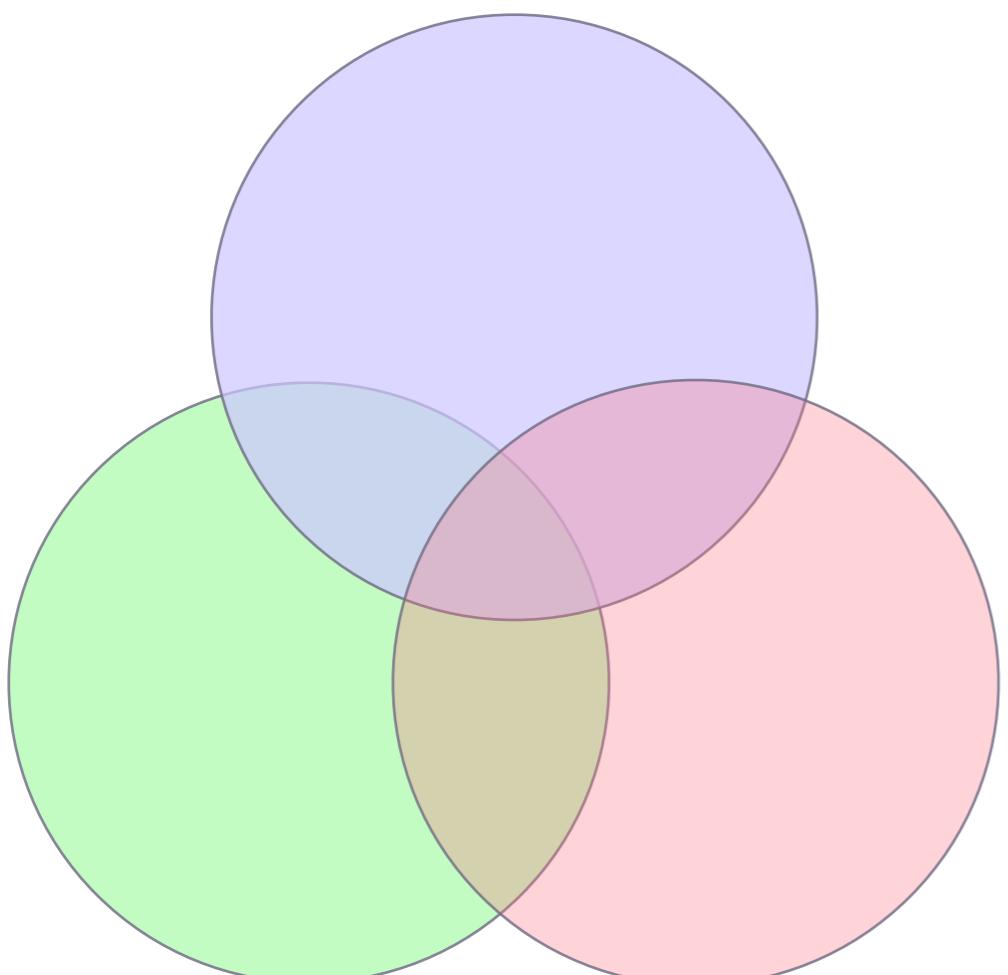
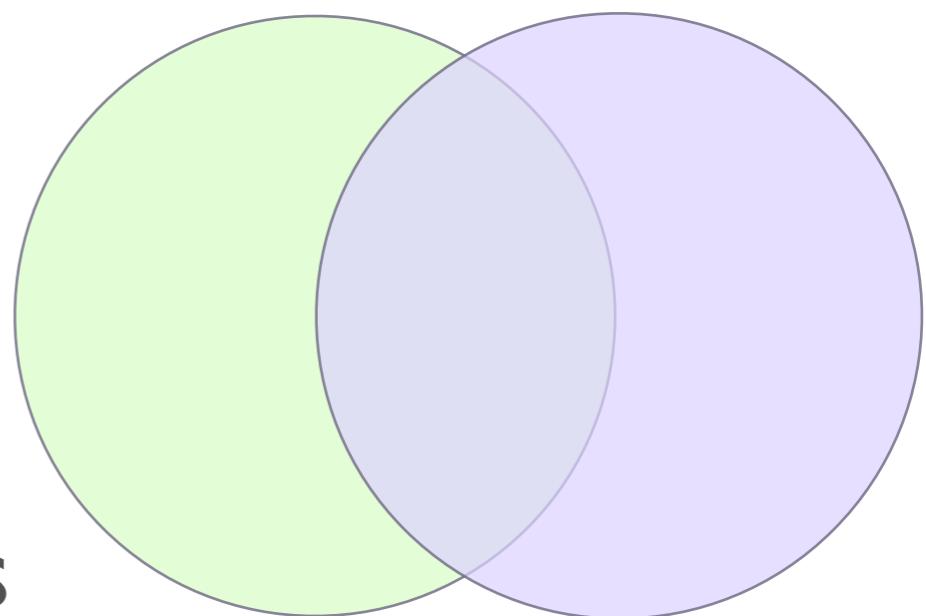
-symmetric difference: $A \ominus B$



visualizing sets

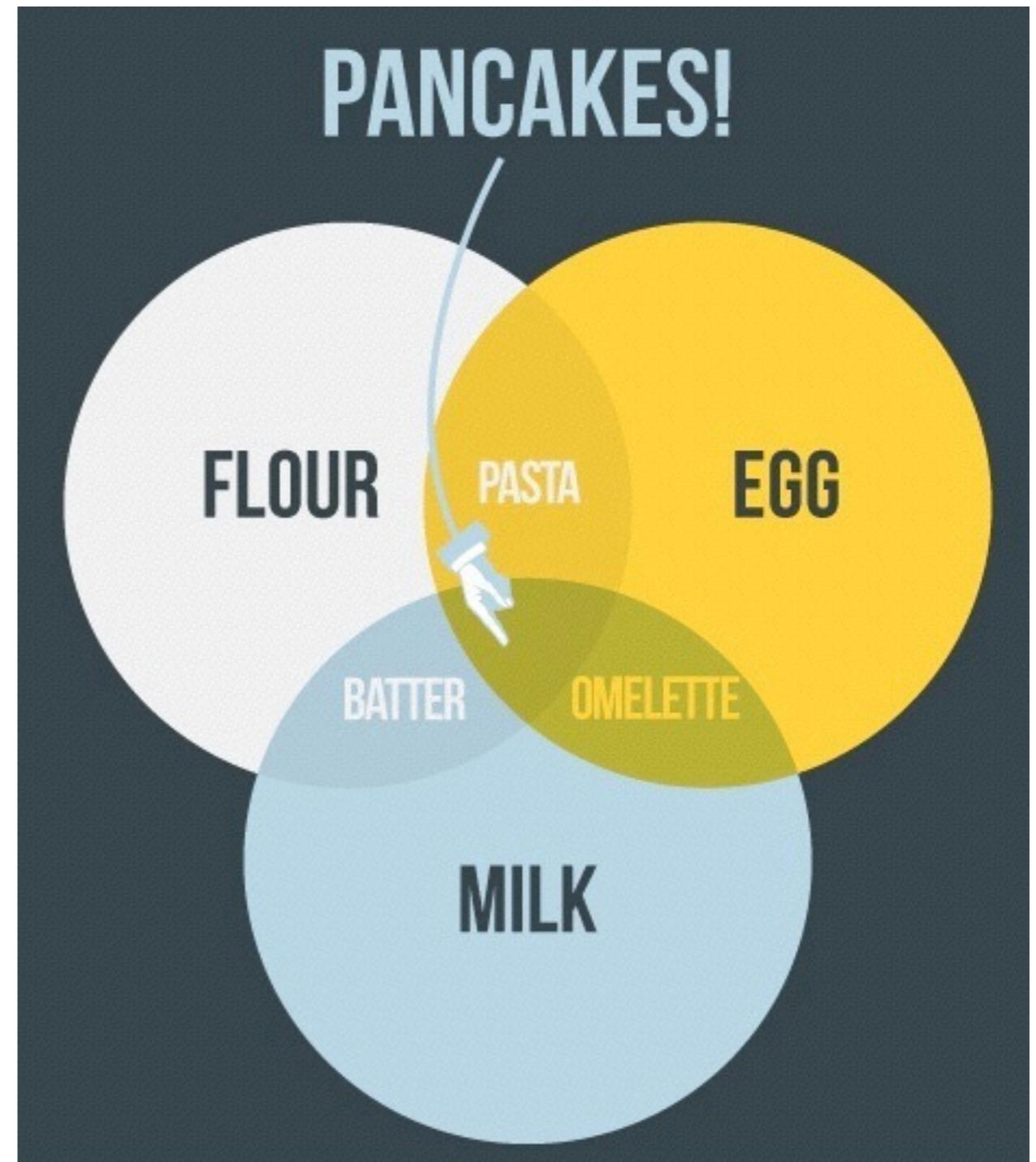
venn diagrams

- show all possible relationships



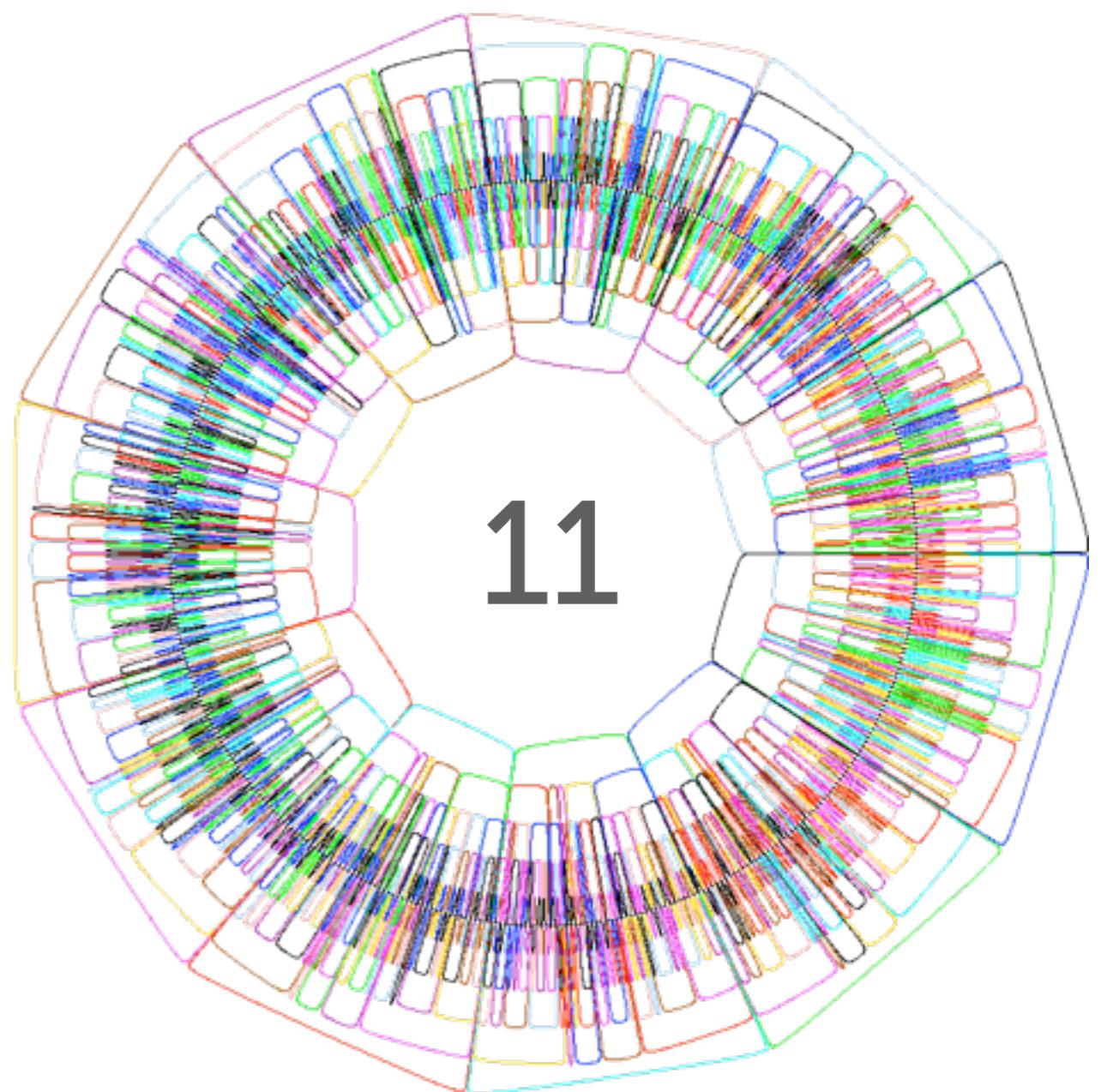
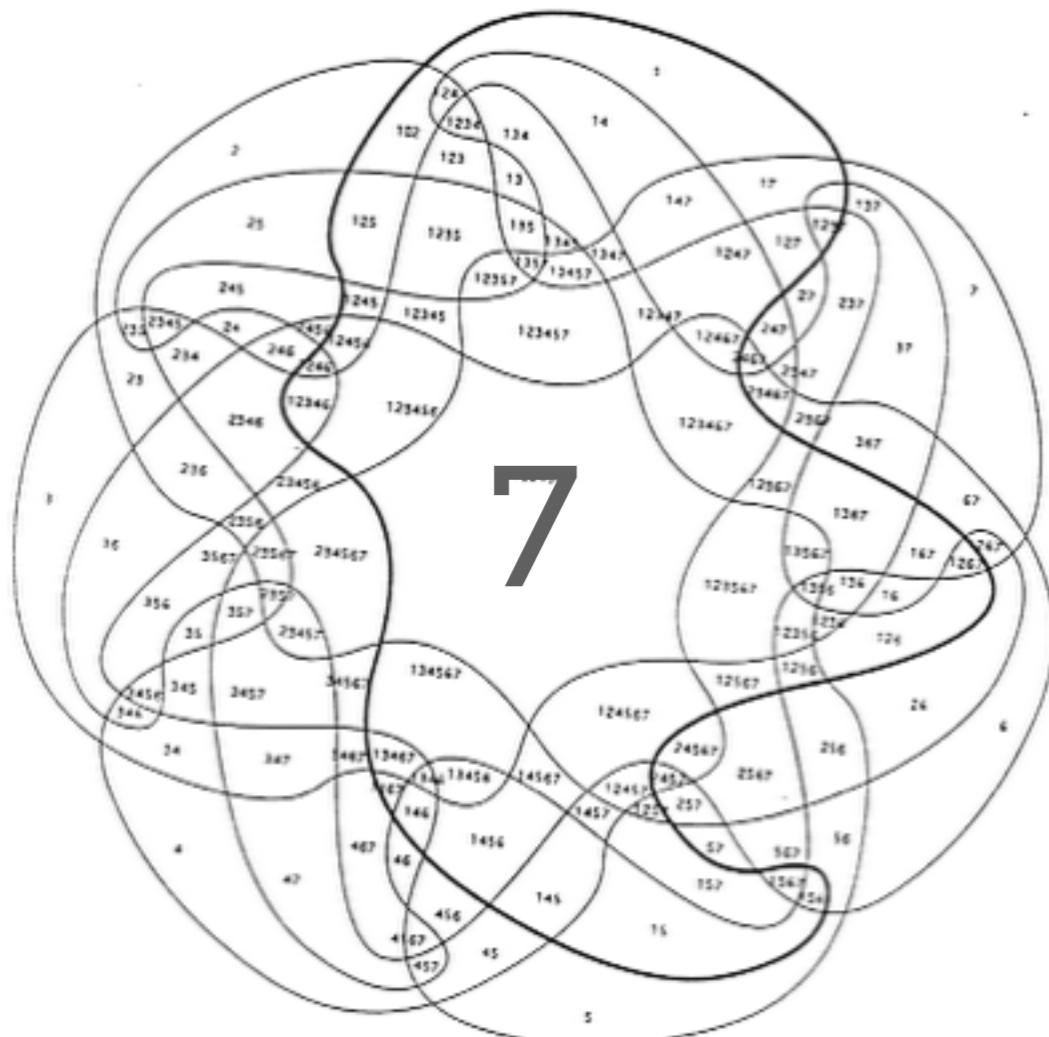
venn diagrams

-casual infovis



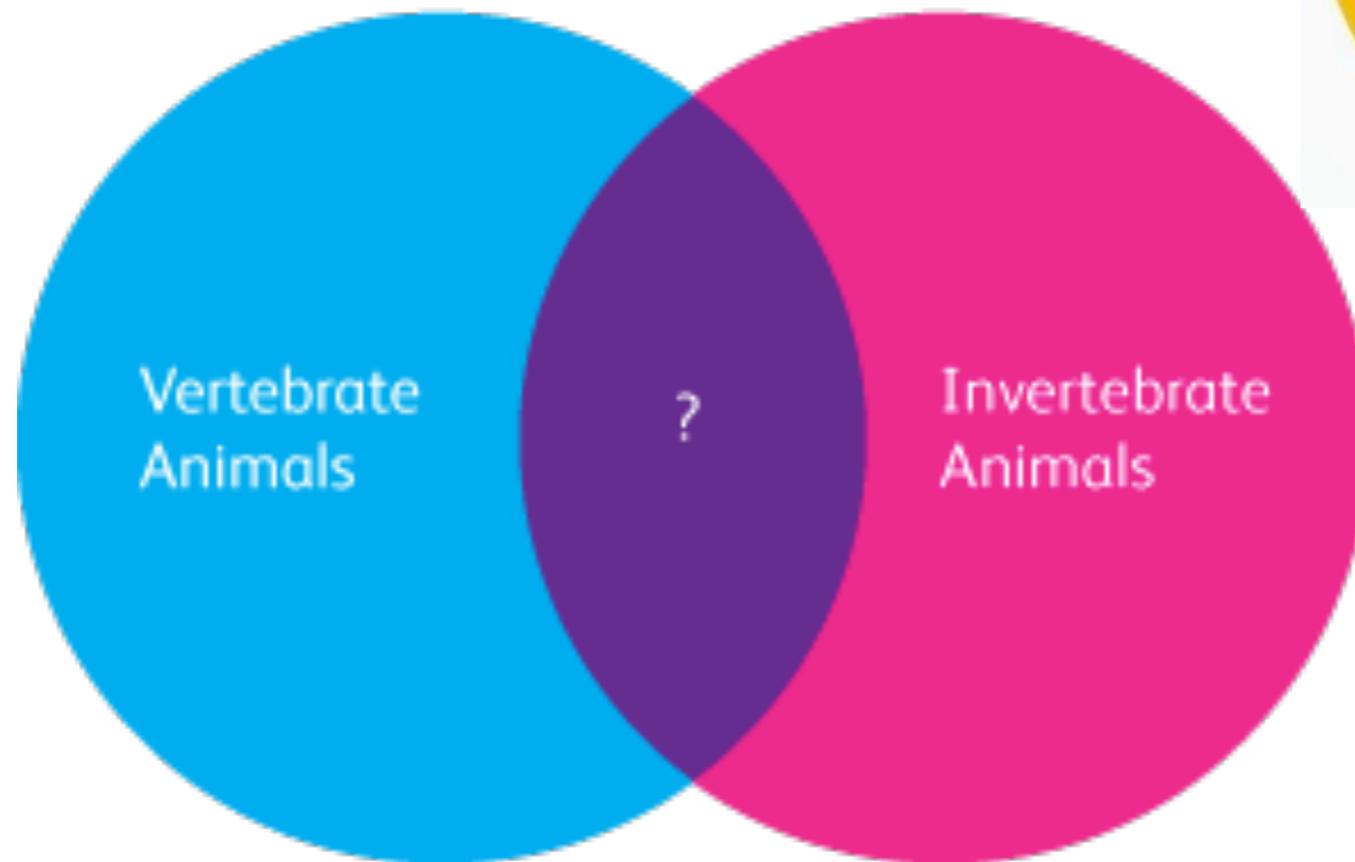
venn diagrams

- get messy fast



venn diagrams

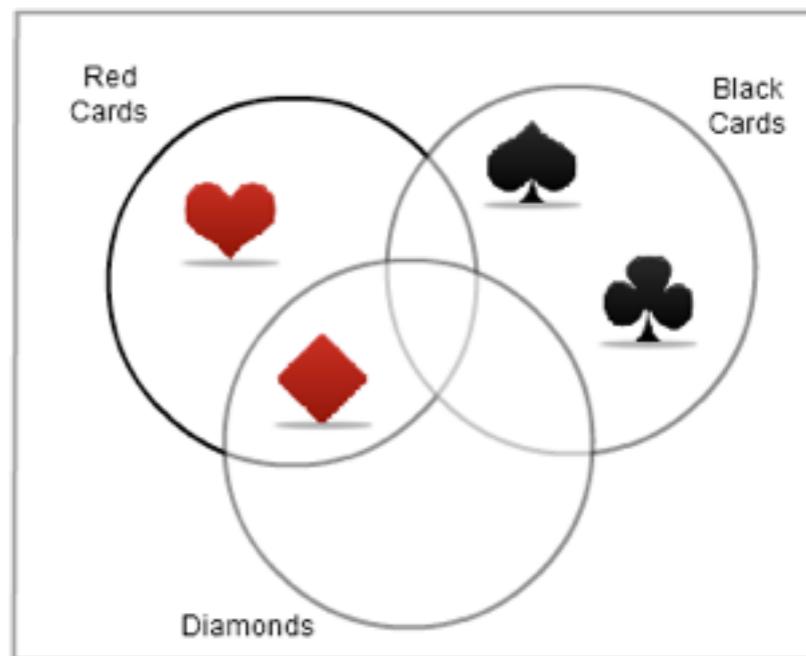
- non-sensical



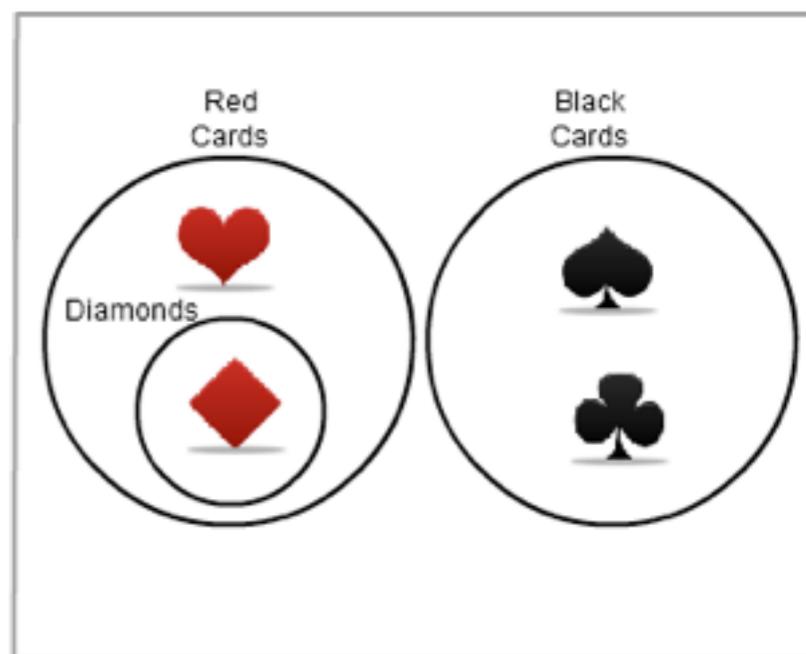
euler diagrams

-show only existing relationships

V
E
N
N

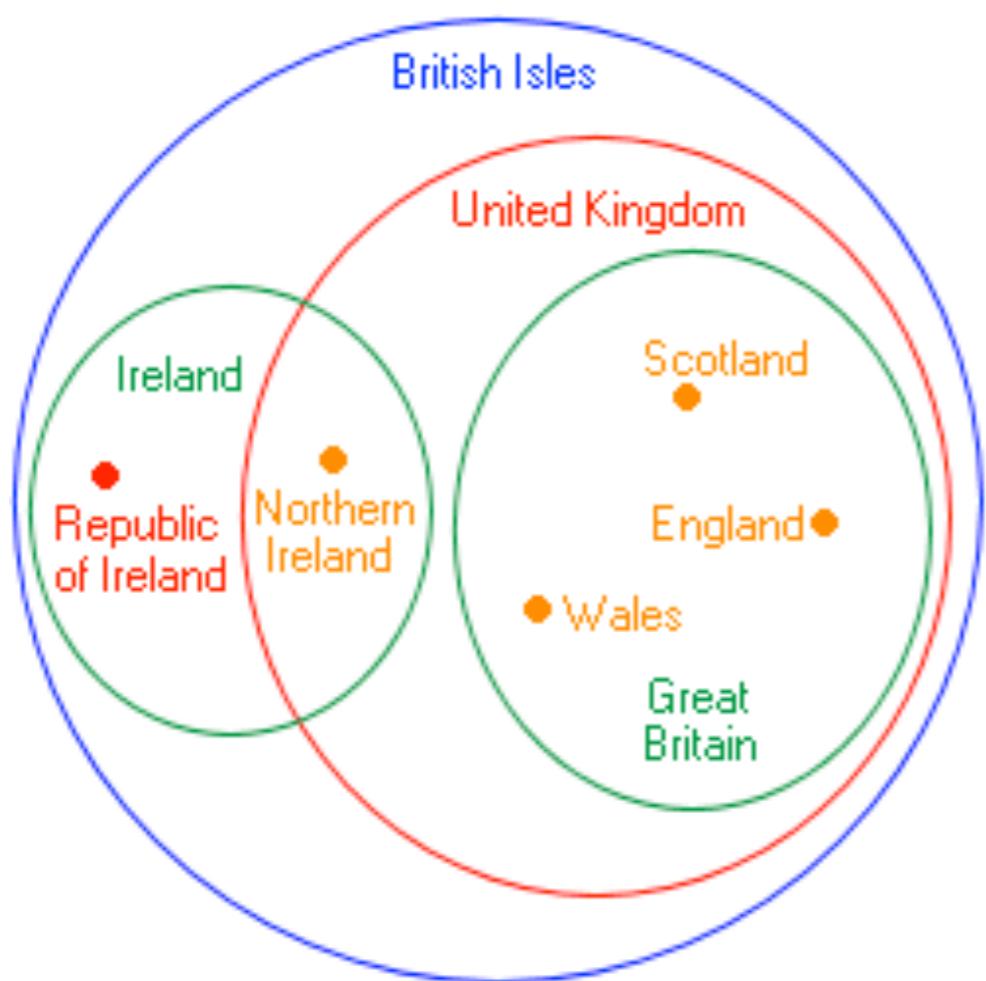


E
U
L
E
R

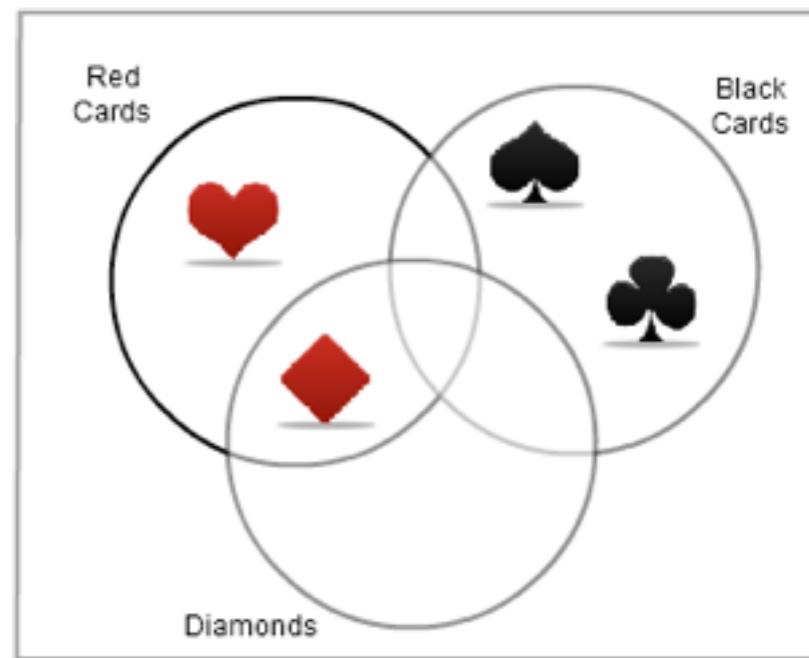


euler diagrams

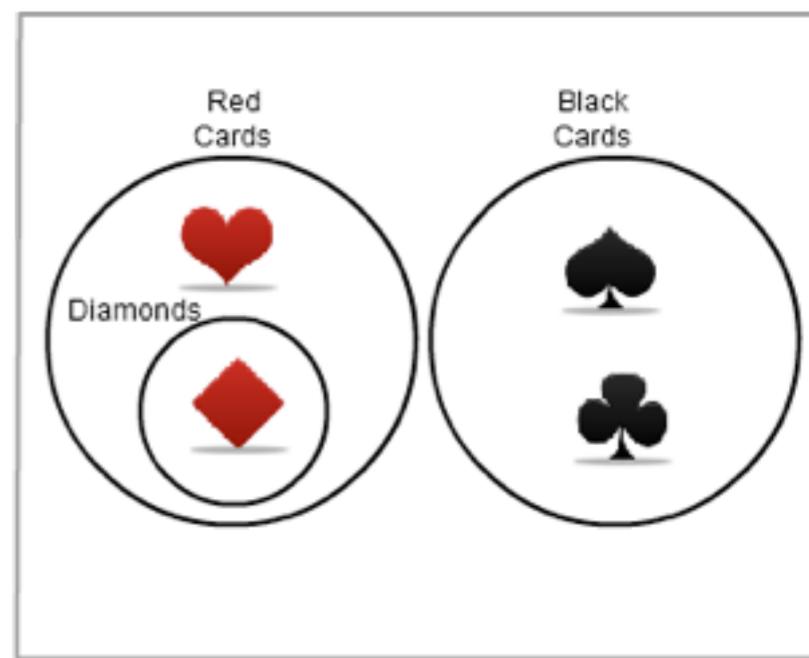
-show only existing relationships



V
E
N
N



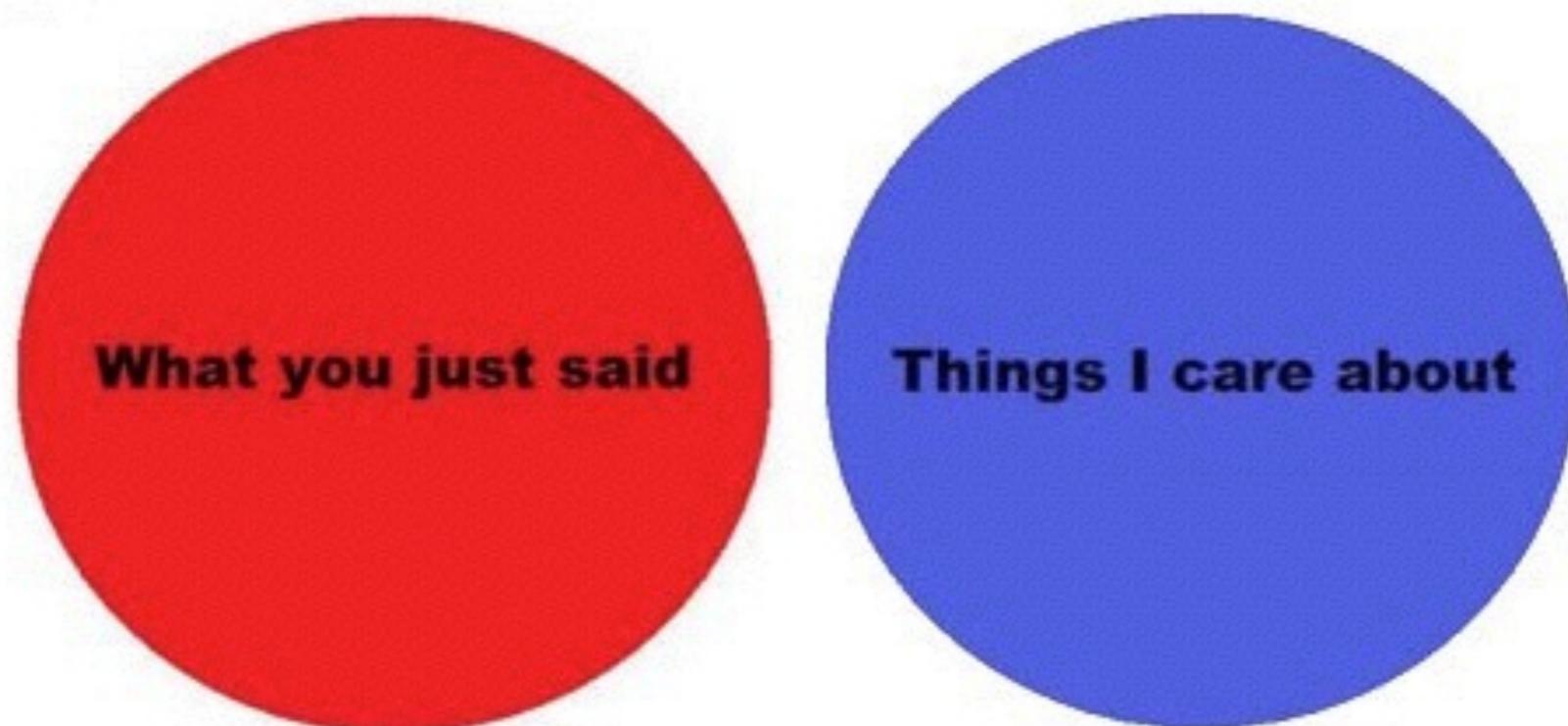
E
U
L
E
R



euler diagrams

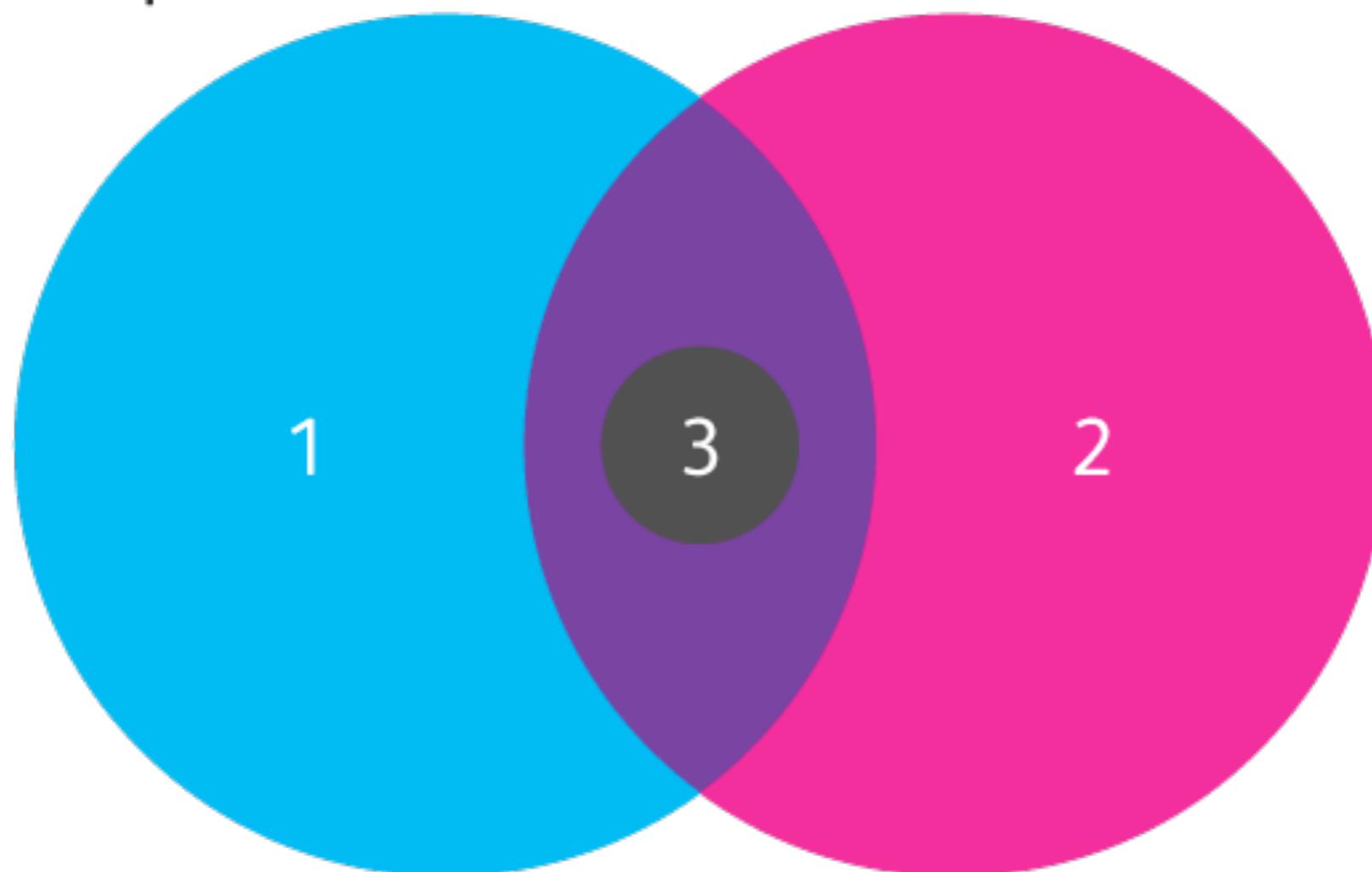
-misunderstood

Maybe this Venn Diagram will explain this better :



euler diagrams

- 1: People who know what a Venn Diagram is.
- 2: People who know what an Euler Diagram is.
- 3: People who know the difference.



venn & euler diagrams

- adjust for area

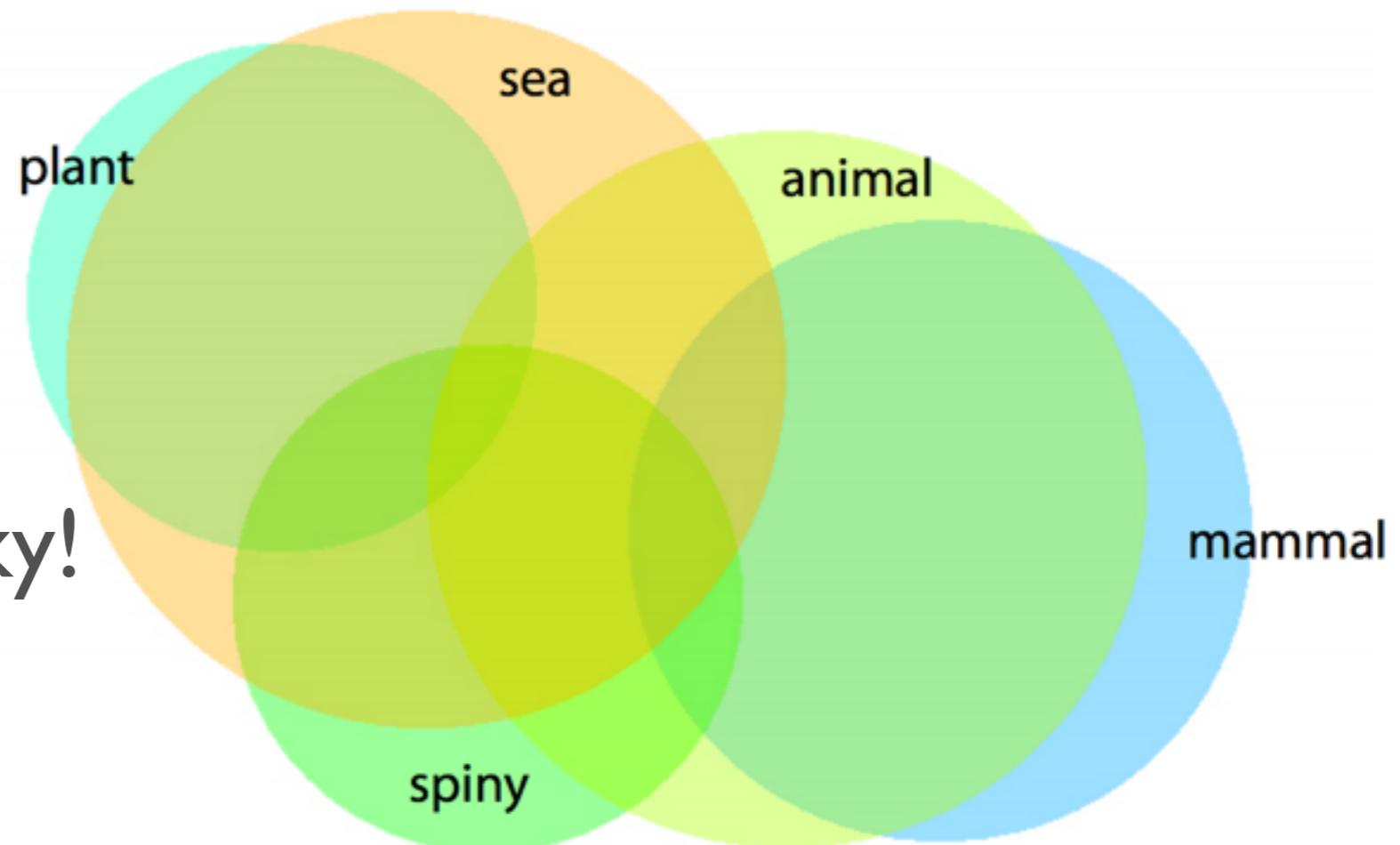
- starts getting tricky!



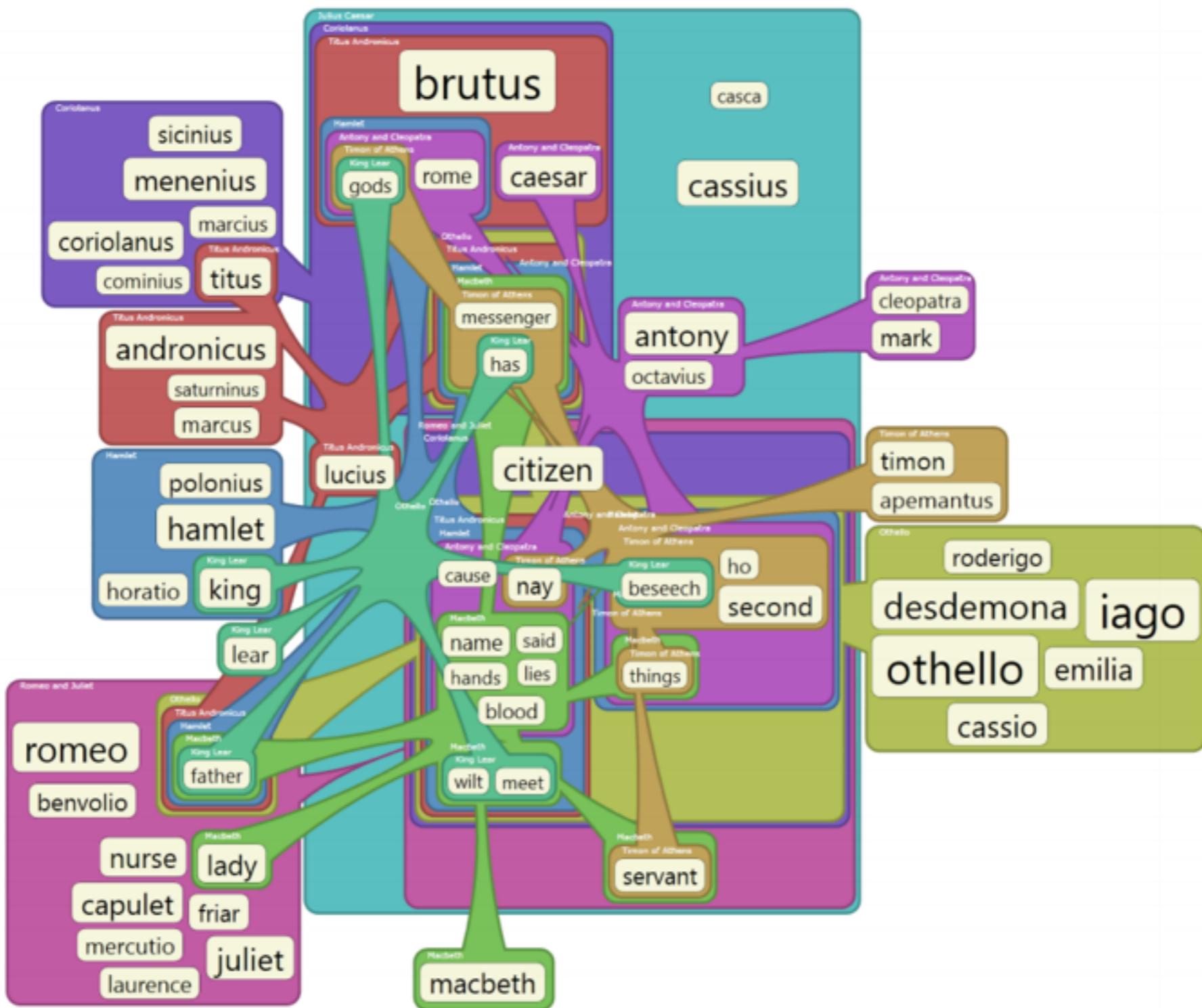
venn & euler diagrams

- adjust for area

- starts getting tricky!

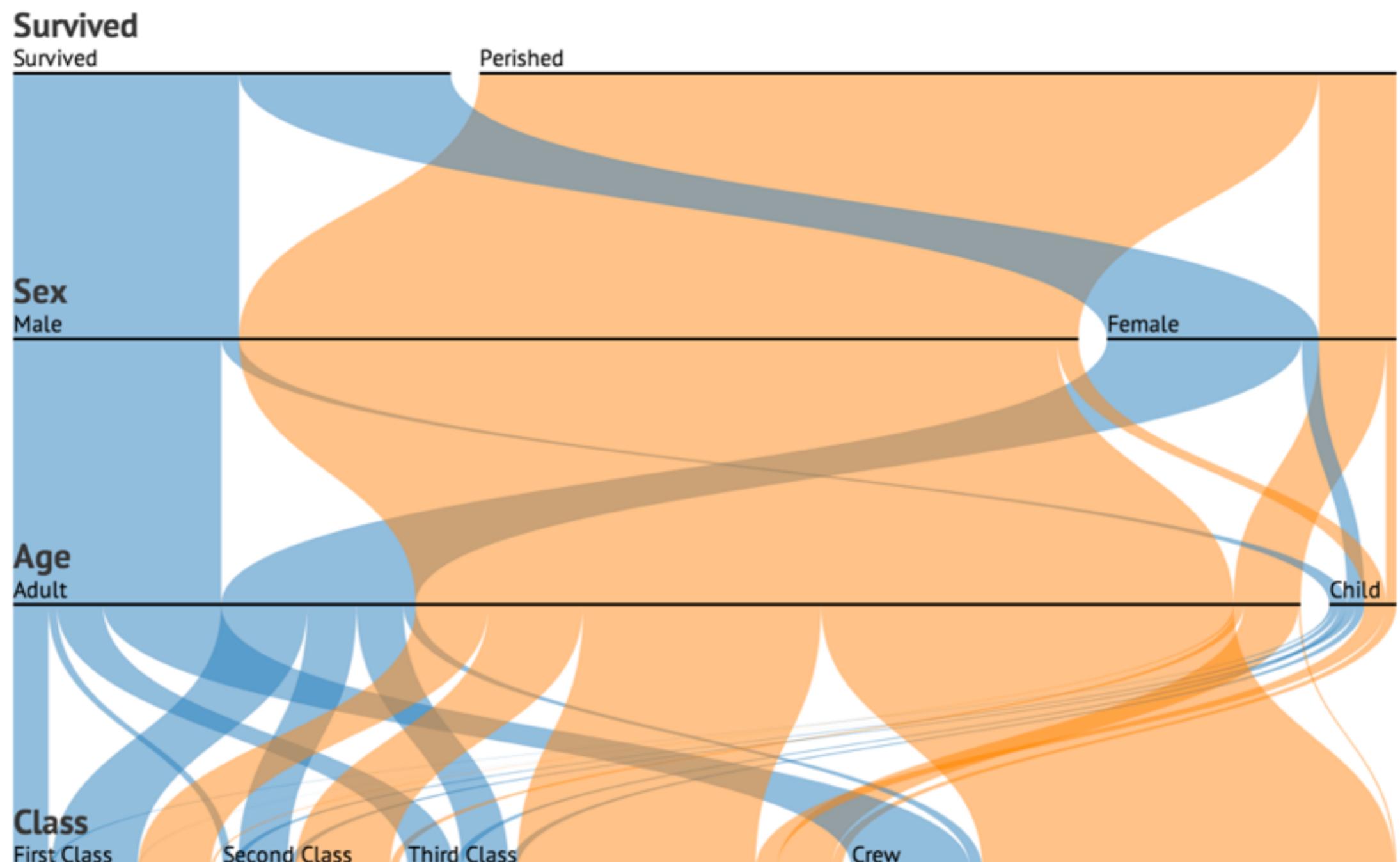


compact euler diagrams



parallel sets

Titanic Survivors

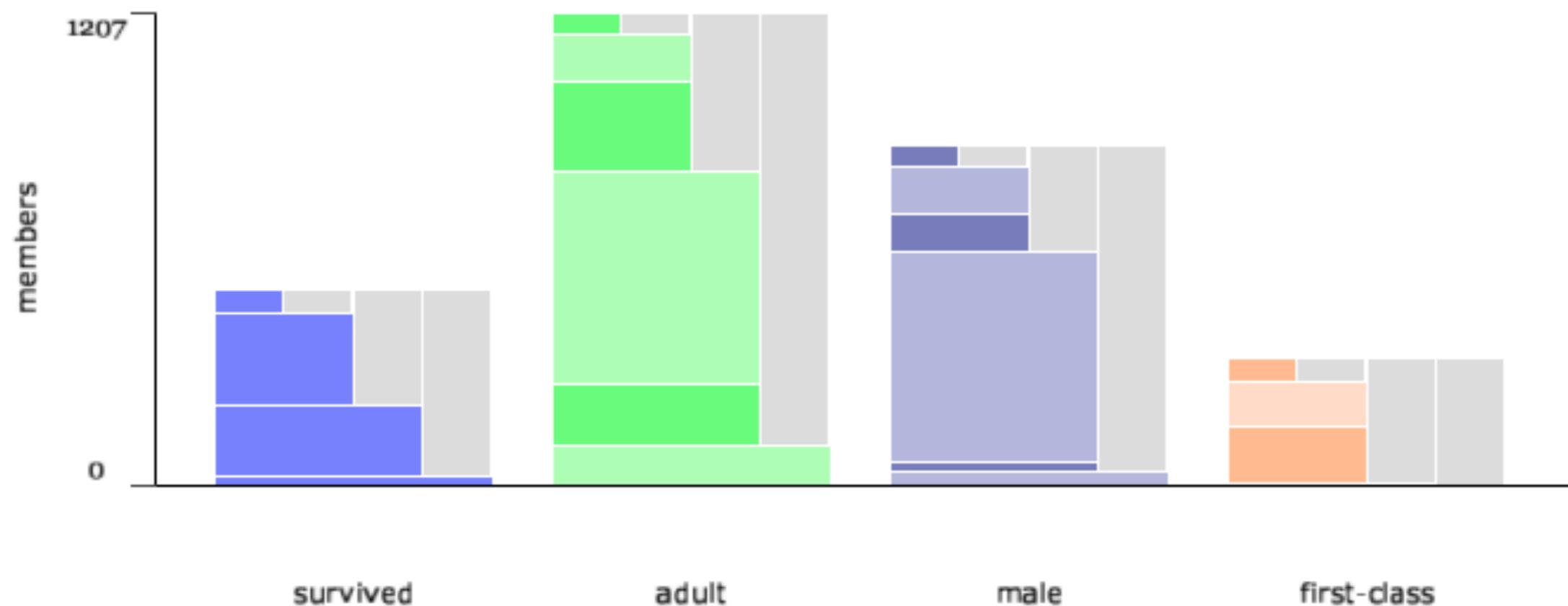


Curves?

Data: [Robert J. MacG. Dawson](#).

set o'gram

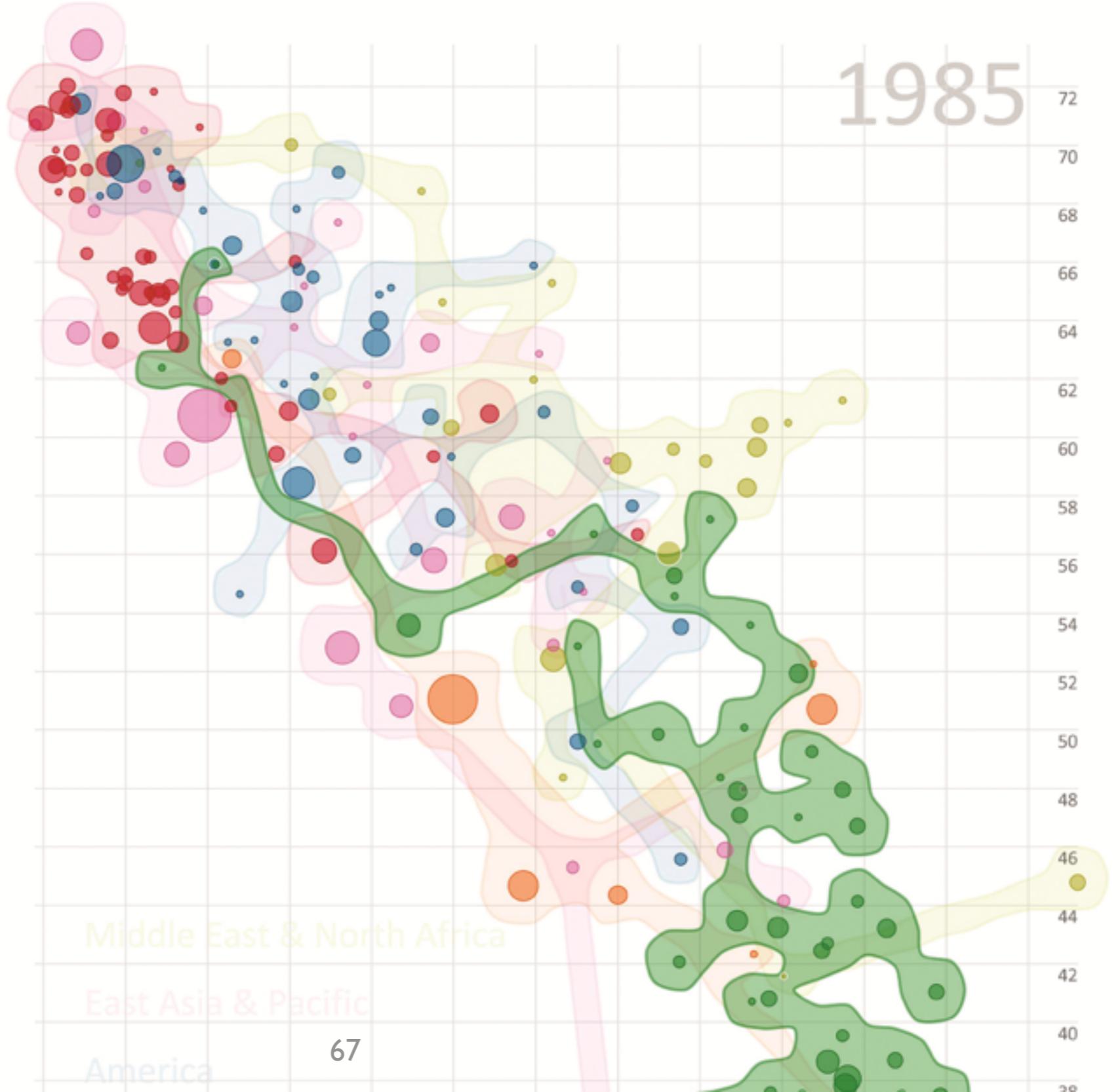
Titanic



visualizing sets with constraints

bubble sets

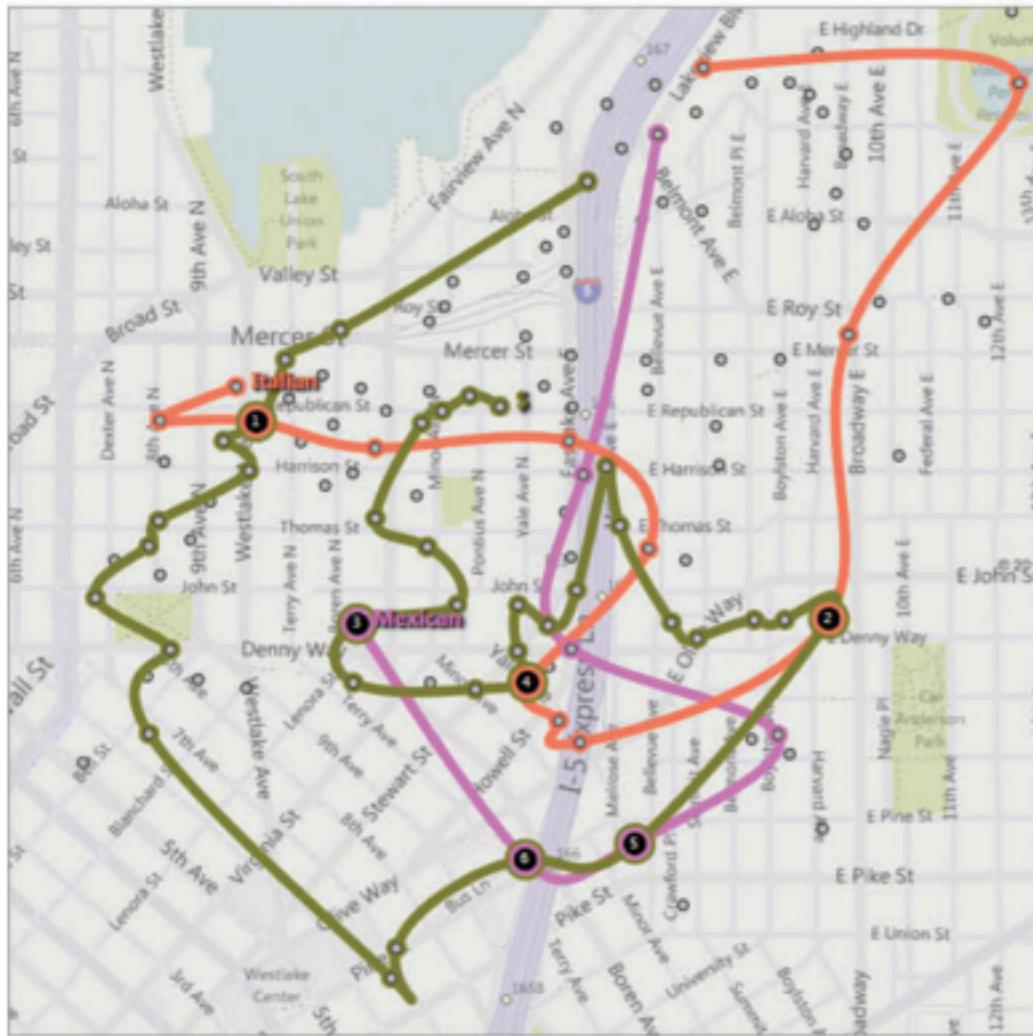
-connect
points



line sets

-restaurants

social communities



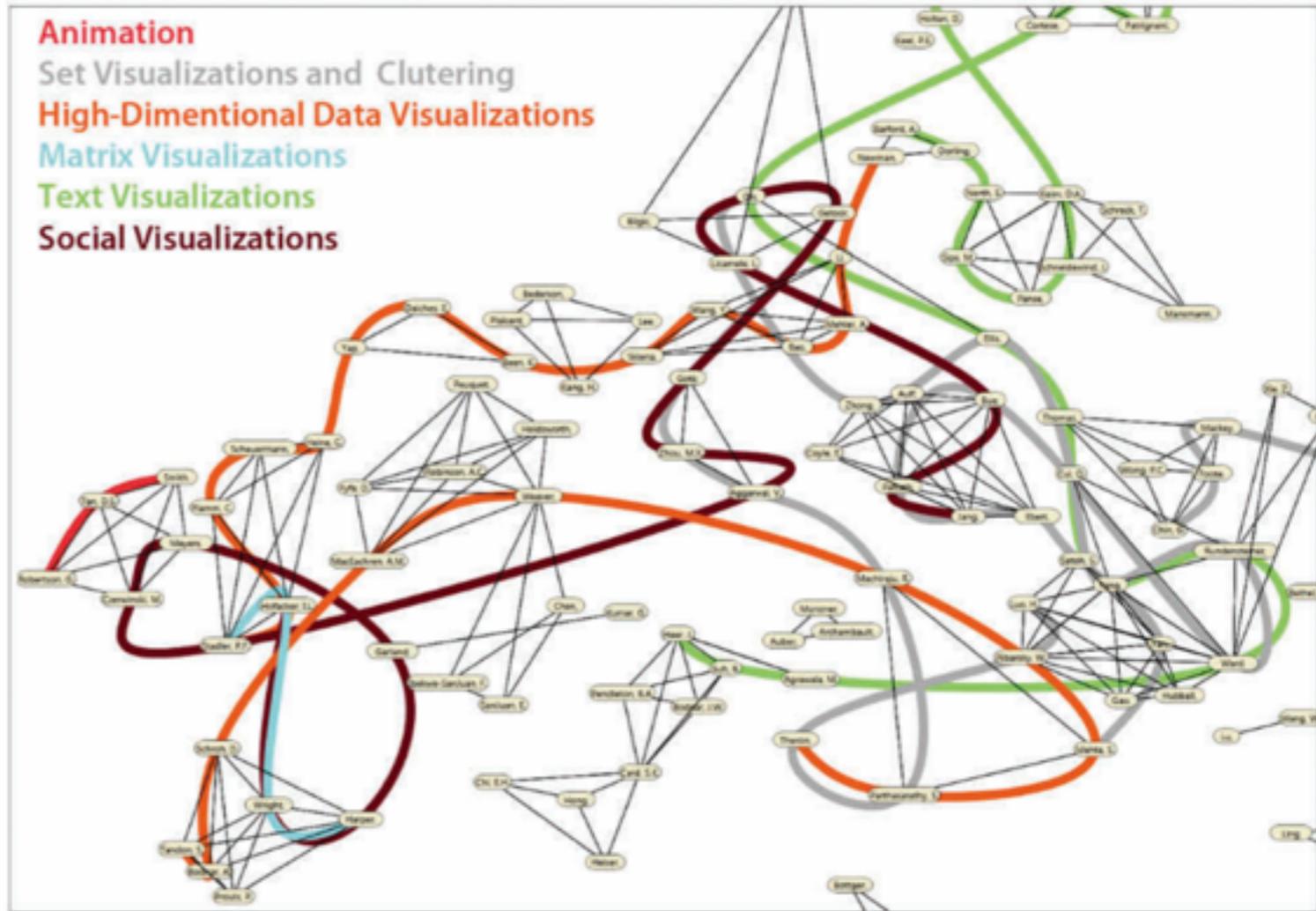
Animation

Set Visualizations and Clustering

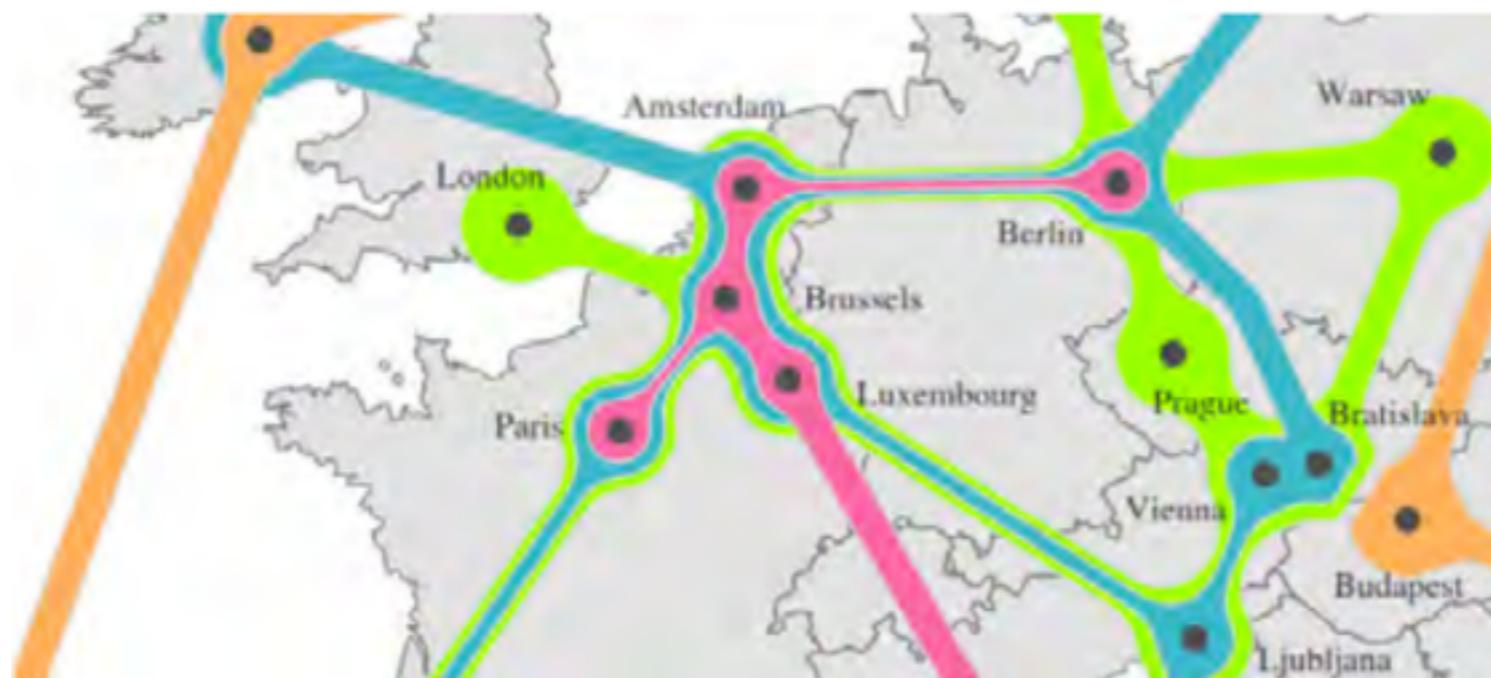
High-Dimensional Data

Matrix Visualization

Text Visualizations

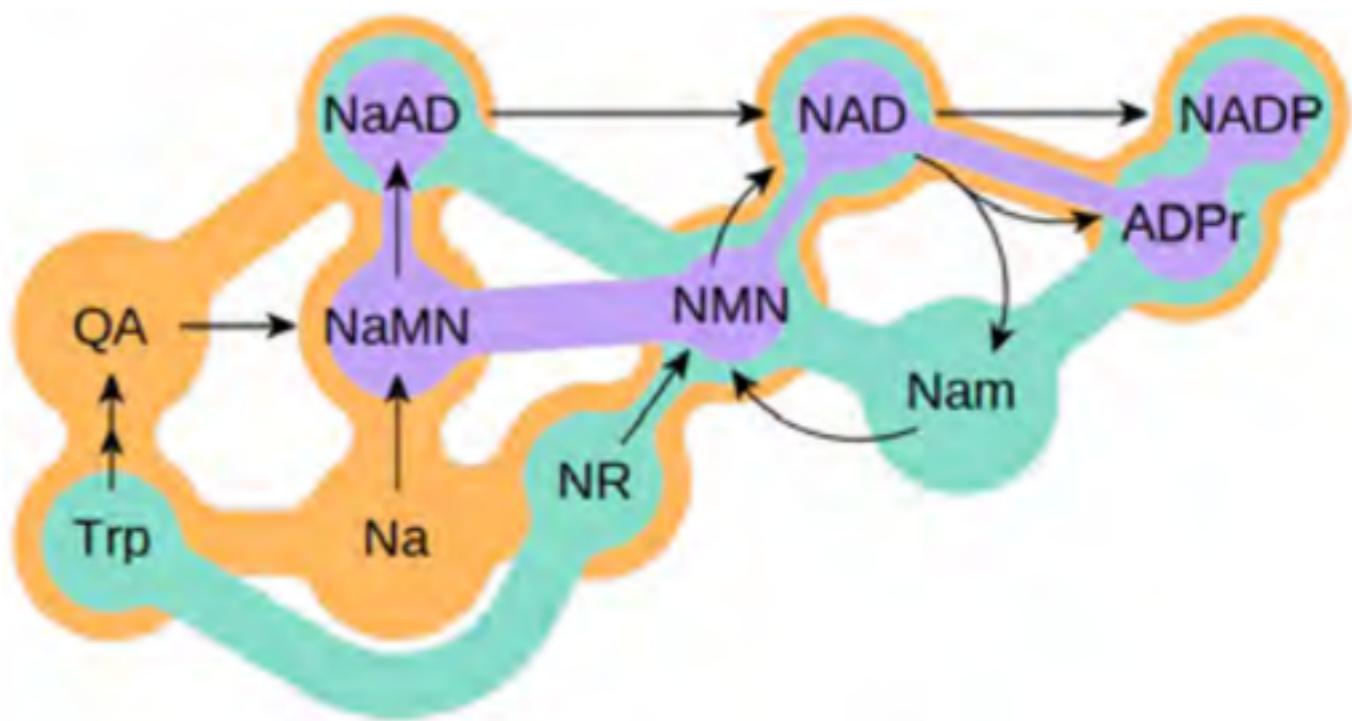


kelp diagrams



- cities on a map

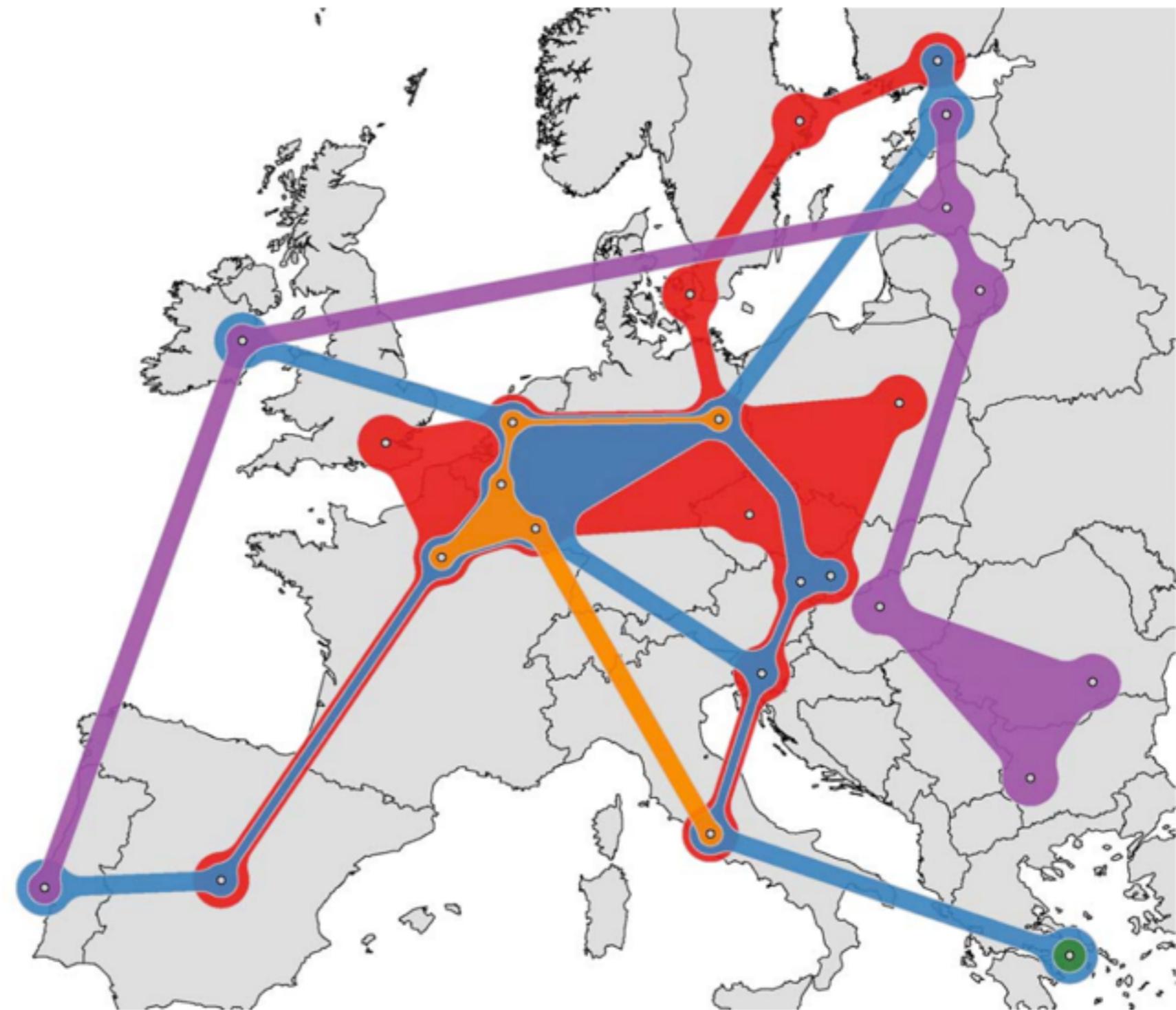
- metabolic network



kelp fusion

- cities on map

- lines & areas



sets

- applies to many datasets
- many combinations may be interesting
- limited numbers of sets

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

L15: Maps

REQUIRED READING

Chapter 8

Arrange Spatial Data

8.1 The Big Picture

For datasets with spatial semantics, the usual choice for `arrange` is to use the given spatial information to guide the layout. In this case, the choices of `express`, `separate`, `order`, and `align` do not apply because the position channel is not available for directly encoding attributes. The two main spatial data types are geometry, where shape information is directly conveyed by spatial elements that do not necessarily have associated attributes, and spatial fields, where attributes are associated with each cell in the field. (See Figure 8.1.) For scalar fields with one attribute at each field cell, the two main visual encoding idiom families are isocontours and direct volume rendering. For both vector and tensor fields, with multiple attributes at each cell, there are four families of encoding idioms: flow glyphs that show local information, geometric approaches that compute derived geometry from a sparse set of seed points, texture approaches that use a dense set of seeds, and feature approaches where data is derived with global computations using information from the entire spatial field.

8.2 Why Use Given?