

Survey of Surveys (SoS) - Mapping The Landscape of Survey Papers in Information Visualisation – SUPPLEMENTARY MATERIAL

Liam McNabb¹ and Robert S. Laramee¹

¹Department of Computer Science, Swansea University, Wales

Table of Figures

Text visualization taxonomy. <i>Kucher et al.</i> (Figure 1 [KK15])	3
Taxonomy for scientific literature and patents. <i>Federico et al.</i> (Figure 2 [FHKM16])	3
Taxonomy of hierarchical aggregation strategies. <i>Elmqvist and Fekete.</i> (Figure 3 [EF10])	3
Taxonomy of set-type data. <i>Alsallakh et al.</i> (Figure 4 [AMA* 14])	4
Taxonomy of quality metrics for high-d data. <i>Bertini et al.</i> (Figure 5 [BTK11])	5
Taxonomy for high-dimensional data. <i>Liu et al.</i> (Figure 6 [LMW* 15])	6
Taxonomy of parallel coordinate studies. <i>Johansson and Forsell.</i> (Figure 7 [JF16])	6
Taxonomy of graph types. <i>Von Landesberger et al.</i> (Figure 8 [VLKS* 11])	7
Taxonomy of traffic data. <i>Chen et al.</i> (Figure 9 [CGW15])	7
Taxonomy of dynamic graph visualization. <i>Beck et al.</i> (Figure 10 [BBDW14])	8
Taxonomy of temporal graph tasks. <i>Kerracher et al.</i> (Figure 11 [KKC15])	9
Taxonomy of temporal graph techniques. <i>Kerracher et al.</i> (Figure 12 [KKCG15])	9
Taxonomy of matrix reordering algorithms. <i>Behrisch et al.</i> (Figure 13 [BBR* 16])	10
Taxonomy of dynamic visualization techniques. <i>Cottam et al.</i> (Figure 14 [CLW12])	10
Taxonomy of cartograms. <i>Nusrat and Kobourov.</i> (Figure 15 [NK16])	11
Taxonomy of composite visualization techniques. <i>Javed and Elmqvist.</i> (Figure 16 [JE12])	12

Taxonomy of performance visualization techniques. <i>Gao et al.</i> (Figure 17 [GZR* 11])	12
Taxonomy of security visualization systems. <i>Shiravi et al.</i> (Figure 18 [SSG12])	13
Design framework of bicluster visualization. <i>Sun et al.</i> (Figure 19 [SNR14])	14
Taxonomy of infovis techniques. <i>Liu et al.</i> (Figure 20 [LCWL14])	14

Author biographies



Liam McNabb received a masters degree in computing from Swansea University in 2015. He is currently a PhD candidate in Information Visualization at Swansea University with a focus on Smart City Visualization. His interests are found in Information Visualization, Visual Analytics, and Smart City Visualization.



Robert S. Laramee received a bachelors degree in physics, cum laude, from the University of Massachusetts, Amherst (ZooMass) in 1997. In 2000, he received a masters degree in computer science from the University of New Hampshire, Durham. He was awarded a PhD from the Vienna University of Technology (Gruess Gott TUWien), Austria at the Institute of Computer Graphics and Algorithms in 2005. From 2001 to 2006 he was a researcher at the

VRVis Research Center (www.vrvis.at) and a software engineer at AVL (www.avl.com) in the department of Advanced Simulation Technologies. Currently he is an Associate Professor at the Swansea University (Prifysgol Cymru Abertawe), Wales in the Department of Computer Science (Adran Gwyddor Cyfrifiadur). His research interests are in the areas of scientific visualization, information visualization, and visual analytics. He has published more than 100 peer-reviewed papers in scientific journals and conferences.

References

- [AMA*14] ALSALLAKH B., MICALLEF L., AIGNER W., HAUSER H., MIKSCH S., RODGERS P.: Visualizing sets and set-typed data: State-of-the-art and future challenges. In *Eurographics conference on Visualization (EuroVis)–State of The Art Reports* (2014), The Eurographics Association, pp. 1–21. 1, 4
- [APS14] AHN J.-W., PLAISANT C., SHNEIDERMAN B.: A task taxonomy for network evolution analysis. *IEEE transactions on visualization and computer graphics* 20, 3 (2014), 365–376. 22
- [BBDW14] BECK F., BURCH M., DIEHL S., WEISKOPF D.: The state of the art in visualizing dynamic graphs. In *EuroVis - State of the Art Reports* (2014), The Eurographics Association. 1, 8
- [BBR*16] BEHRISCH M., BACH B., RICHE N. H., SCHRECK T., FEKETE J.-D.: Matrix Reordering Methods for Table and Network Visualization. *Computer Graphics Forum* 35, 3 (2016), 693–716. 1, 10
- [BDA*14] BACH B., DRAGICEVIC P., ARCHAMBAULT D., HURTER C., CARPENDALE S.: A review of temporal data visualizations based on space-time cube operations. In *EuroVis 2014 - State of the Art Reports* (2014), The Eurographics Association, pp. 1–21. 23
- [BKC*13] BORGIO R., KEHRER J., CHUNG D. H., MAGUIRE E., LARAMEE R. S., HAUSER H., WARD M., CHEN M.: "glyph-based visualization: Foundations, design guidelines, techniques and applications". *"Eurographics State of the Art Reports"* (may "2013"), "39–63". 19
- [BTK11] BERTINI E., TATU A., KEIM D.: Quality metrics in high-dimensional data visualization: An overview and systematization. *IEEE Transactions on Visualization and Computer Graphics* 17, 12 (2011), 2203–2212. 1, 5
- [CGW15] CHEN W., GUO F., WANG F.-Y.: A survey of traffic data visualization. *IEEE Trans. Intelligent Transportation Systems* 16, 6 (2015), 2970–2984. 1, 7
- [CLW12] COTTAM J. A., LUMSDAINE A., WEAVER C.: Watch this: A taxonomy for dynamic data visualization. In *Visual Analytics Science and Technology (VAST), 2012 IEEE Conference on* (2012), IEEE, pp. 193–202. 1, 10
- [CMS99] CARD S. K., MACKINLAY J. D., SHNEIDERMAN B.: *Readings in information visualization: using vision to think*. Morgan Kaufmann, 1999. 15
- [CZ11] CASERTA P., ZENDRA O.: Visualization of the static aspects of software: A survey. *IEEE transactions on visualization and computer graphics* 17, 7 (2011), 913–933. 24
- [EF10] ELMQVIST N., FEKETE J. D.: Hierarchical aggregation for information visualization: Overview, techniques, and design guidelines. *IEEE Transactions on Visualization and Computer Graphics* 16, 3 (May 2010), 439–454. 1, 3
- [FHKM16] FEDERICO P., HEIMERL F., KOCH S., MIKSCH S.: A survey on visual approaches for analyzing scientific literature and patents. *IEEE Transactions on Visualization and Computer Graphics* (2016). 1, 3
- [FIBK16] FUCHS J., ISENBERG P., BEZERIANOS A., KEIM D.: A systematic review of experimental studies on data glyphs. *IEEE Transactions on Visualization and Computer Graphics* (2016). forthcoming. 20
- [GZR*11] GAO Q., ZHANG X., RAU P.-L. P., MACIEJEWSKI A. A., SIEGEL H. J.: Performance visualization for large-scale computing systems: A literature review. In *Proceedings of the 14th International Conference on Human-computer Interaction: Design and Development Approaches* (2011), HCII'11, Springer-Verlag, pp. 450–460. 1, 12
- [IGJ*14] ISAACS K. E., GIMÉNEZ A., JUSUFI I., GAMBLIN T., BHATELE A., SCHULZ M., HAMANN B., BREMER P.-T.: State of the art of performance visualization. *EuroVis - State of the Art Reports* (2014). 16
- [JE12] JAVED W., ELMQVIST N.: Exploring the design space of composite visualization. In *Visualization Symposium (PacificVis), 2012 IEEE Pacific* (2012), IEEE, pp. 1–8. 1, 12
- [JF16] JOHANSSON J., FORSELL C.: Evaluation of parallel coordinates: Overview, categorization and guidelines for future research. *IEEE transactions on visualization and computer graphics* 22, 1 (2016), 579–588. 1, 6
- [JFCS15] JÄNICKE S., FRANZINI G., CHEEMA M. F., SCHEUERMANN G.: On Close and Distant Reading in Digital Humanities: A Survey and Future Challenges. In *Eurographics Conference on Visualization (EuroVis) - STARS* (2015), pp. 83–103. 17
- [KK15] KUCHER K., KERREN A.: Text visualization techniques: Taxonomy, visual survey, and community insights. In *2015 IEEE Pacific Visualization Symposium (PacificVis)* (2015), IEEE, pp. 117–121. 1, 3
- [KKC15] KERRACHER N., KENNEDY J., CHALMERS K.: A task taxonomy for temporal graph visualisation. *Visualization and Computer Graphics, IEEE Transactions on* 21, 10 (2015), 1160–1172. 1, 9
- [KKCG15] KERRACHER N., KENNEDY J., CHALMERS K., GRAHAM M.: Visual techniques to support exploratory analysis of temporal graph data. In *Eurographics Conference on Visualization (EuroVis) - Short Papers* (2015), The Eurographics Association, pp. 1–21. 1, 9
- [LCWL14] LIU S., CUI W., WU Y., LIU M.: A survey on information visualization: Recent advances and challenges. *The Visual Computer* 30, 12 (2014), 1373–1393. 1, 14
- [LMW*15] LIU S., MALJOVEC D., WANG B., BREMER P.-T., PASCUCCI V.: Visualizing high-dimensional data: Advances in the past decade. In *Eurographics Conference on Visualization (EuroVis) - STARS* (2015), VGTC, The Eurographics Association, pp. 21–40. 1, 6
- [NK16] NUSRAT S., KOBOUROV S.: The state of the art in cartograms. In *Eurographics conference on Visualization (EuroVis)–State of The Art Reports* (2016), vol. 35, The Eurographics Association, pp. 619–642. 1, 11
- [SHS11] SCHULZ H.-J., HADLAK S., SCHUMANN H.: The design space of implicit hierarchy visualization: A survey. *IEEE transactions on visualization and computer graphics* 17, 4 (2011), 393–411. 18
- [SNR14] SUN M., NORTH C., RAMAKRISHNAN N.: A five-level design framework for bicluster visualizations. *IEEE transactions on visualization and computer graphics* 20, 12 (2014), 1713–1722. 1, 14
- [SSG12] SHIRAVI H., SHIRAVI A., GHORBANI A. A.: A survey of visualization systems for network security. *IEEE Transactions on visualization and computer graphics* 18, 8 (2012), 1313–1329. 1, 13
- [STMT12] SEDLMAIR M., TATU A., MUNZNER T., TORY M.: A taxonomy of visual cluster separation factors. In *Computer Graphics Forum* (2012), vol. 31, Wiley Online Library, pp. 1335–1344. 15
- [TGG*16] TOMINSKI C., GLADISCH S., KISTER U., DACHSELT R., SCHUMANN H.: Interactive lenses for visualization: An extended survey. In *Computer Graphics Forum* (2016), Wiley Online Library. 25
- [VBW15] VEHLW C., BECK F., WEISKOPF D.: The state of the art in visualizing group structures in graphs. In *Eurographics Conference on Visualization (EuroVis)-STARS* (2015), VGTC, The Eurographics Association, pp. 21–40. 21
- [VLKS*11] VON LANDESBERGER T., KUIJPER A., SCHRECK T., KOHLHAMMER J., VAN WIJK J. J., FEKETE J.-D., FELLNER D. W.: Visual analysis of large graphs: state-of-the-art and future research challenges. In *Computer graphics forum* (2011), vol. 30, Wiley Online Library, pp. 1719–1749. 1, 7

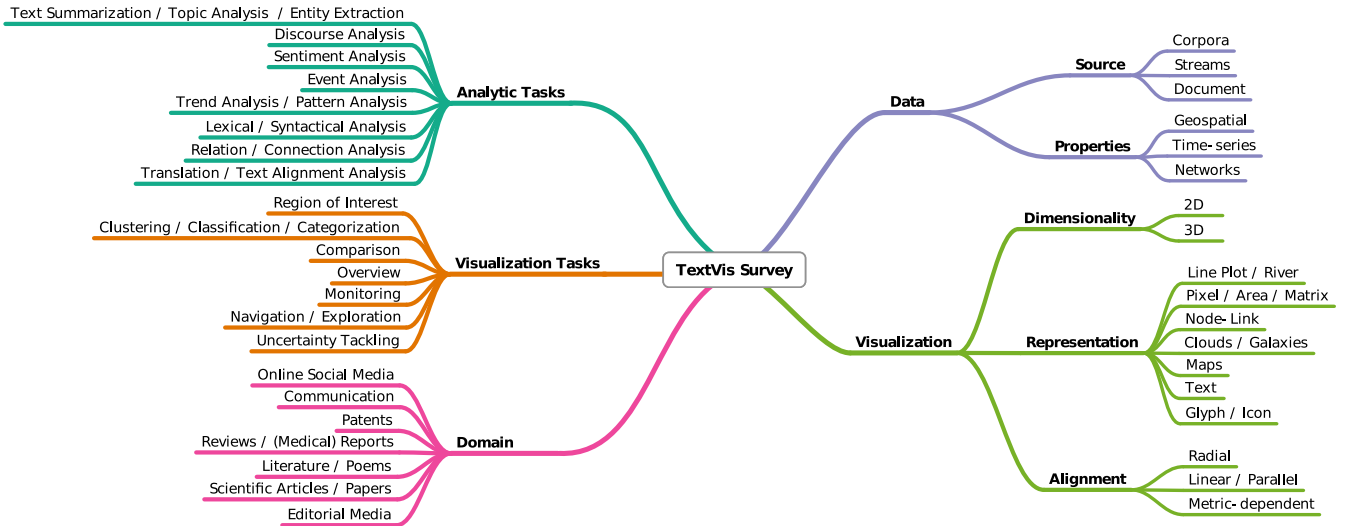


Figure 1: Kucher et al. present a hierarchical taxonomy used to classify text visualisation techniques. Courtesy of Kucher et al. [KK15]

	Elem. lookup & comparison	Elem. relation seeking	Synoptic (Patterns)	Synoptic (Temporal patterns)
Text	8	7	20	5 (6)
Citations	2	7	9	10
Authors	2	1	2	7
Metadata	2	5	8	2 (3)

a Approaches for data types: Text, Citations, Authors, and Metadata (supporting Elementary and Descriptive Synoptic tasks).

	Aggregation	Labelling	Composition	Multiple views	Tight integration
Multiple/Connectonal	11 (6)	4 (3)	3 (1)	12 (3)	5 (2)

b Approaches for data type: Multiple (supporting Connection Discovery tasks), by level of integration.

Figure 2: Table a presents the distribution of papers for each single data category, whilst b contains the distribution papers which look at multiple data-types. Both tables distribute papers based on tasks. Numbers in parentheses are papers identified as a secondary classification. Image courtesy of Federico et al. [FHKM16]

Data structure	Visualization	Type	Aggregation	Visual aggregate	Metadata visualized
multidimensional	scatterplot	O/L	hierarchical clustering	points [19, 74]	average
multidimensional	scatterplot	O/L	hierarchical clustering	boxes [82]	extents (axis-aligned), average
multidimensional	scatterplot	O/L	space-filling subdivision	boxes [80]	extents (axis-aligned), average
multidimensional	scatterplot	O/L	hierarchical clustering	hulls [4]	extents (convex hull), average
multidimensional	scatterplot	O/L	hierarchical clustering	blobs [6, 15, 20, 44]	extents
multidimensional	parallel coordinates	O/L	hierarchical clustering	lines [75]	average
multidimensional	parallel coordinates	O/L	hierarchical clustering	bands [34, 82]	extents, average
multidimensional	parallel coordinates	O/L	hierarchical clustering	color histograms [29, 30]	distribution, extents
multidimensional	parallel coordinates	O/L	hierarchical clustering	beads [4]	distribution, extents
multidimensional	starglyphs	O/L	hierarchical clustering	lines [75]	average
multidimensional	starglyphs	O/L	hierarchical clustering	bands [34, 82]	extents, average
multidimensional	starglyphs	O/L	hierarchical clustering	color histograms [29, 30]	distribution, extents, average
tree	treemap	S/F	existing tree hierarchy	treemap nodes [63]	extents, average
tree	node-link diagram	O/L	existing tree hierarchy	thumbnails [17, 59]	extents, count, depth
graph	node-link diagram	O/L	hierarchical clustering	metanodes [2, 8]	extents, average
graph	node-link diagram	O/L	—	edge bundles [47]	link extents, average
graph	node-link diagram	O/L	data cube aggregation	metanodes [78]	node and link counts
graph	adjacency matrix	S/F	recursive edge merging	edge blocks [1, 28]	distribution, average
spatial	2D/3D geometric	—	recursive data merging	quad/octree blocks [4]	extents, average

Figure 3: A hierarchical classification of aggregation strategies for Information Visualisation techniques. Image courtesy of Elmqvist and Fekete [EF10].

Technique	Element-related Tasks							Set-related Tasks							Attribute-related Tasks					Scalability in # of sets in # of elements							
	A1	A2	A3	A4	A5	A6	A7	B1	B2	B3	B4	B5	B6	B7	B8	B9	B10	B11	B12		B13	B14	C1	C2	C3	C4	C5
Euler diagrams	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 10 hundreds / ∞
ComED	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	10 to 20 hundreds
DupED	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 10 tens
BubbleSets	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 10 tens
LineSets	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	10 to 100 hundreds
Kelp diagrams	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 10 tens
Colored glyphs	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	10 to 20 hundreds
Icon lists	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	large list
Linked lists	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	hundreds
Anchored maps	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	20 to 50 hundreds
PivotPaths	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	50 to 100 hundreds
ConSet	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 100 about 100
PixelLayer	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	tens hundreds
Frequency grids	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	3 to 5 hundreds
Overlap matrix	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	about 100 not applicable
KMVQL	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	4 to 6 not applicable
Mosaic displays	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	up to 4 sets large (agg.)
Double-Decker	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	4 to 6 large (agg)
Sets' o' grams	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	50 to 100 large (agg.)
Radial Sets	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	20 to 30 large (agg.)
Scatter view	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	hundreds not applicable
Cluster view	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	•	hundreds not applicable

C1: Find an element's attribute values
 C2: Attribute distribution in a set / subset
 C3: Compare attribute values between subset
 C4: Set membership for specific attr. values
 C5: Create a set of elements by attributes

B9: Identify the set with largest / smallest number of pair-wise set intersections
 B10: Analyze & compare cardinalities
 B11: Analyze & compare set similarities
 B12: Analyze & compare set exclusiveness
 B13: Highlight specific sets, subsets, etc.
 B14: create a set by set-theoretic operation

B1: Find the number of sets in a family
 B2/B3: Inclusion relations / hierarchies
 B4/B5: Exclusion / intersection relations
 B6: Identify intersections between k sets
 B7: identify sets involved in an overlap
 B8: Identify intersections of a set

A1: Find/Select elements of a specific set
 A2: Find sets containing a specific element
 A3: Find/Select elements by set memberships
 A4: Find/Select elements by their degrees
 A5: Filter out elements by set memberships
 A6: Filter out elements by their degrees
 A7: Create a set out of certain elements

• Task is supported
 ○ Task is partially supported
 • Task requires interaction

Figure 4: A 1-N taxonomy of set-types data showing a comparison between tasks and techniques. Courtesy of Alsallakh et al. [AMA* 14]

Paper Title	Visualization technique		What is measured						Where it is measured				Purpose				Inter-act- ion	
	SP	PC	other	clustering	correlation	outliers	complex patterns	image quality	feature pres.	data	image	space	projection	ordering	abstraction	visual mapping		view optimization
A Projection Pursuit Algorithm for Exploratory Data Analysis - Friedman & Tukey [21]	SP			clustering						data		projection						
A Rank-by-Feature Framework for Unsupervised Multidimensional Data Exploration Using Low Dimensional Projections - Seo & Shneidman [44]	SP		histogram, matrix, list	clustering	correlation	outliers	complex patterns			data		projection						S
Finding and Visualizing Relevant Subspaces for Clustering High-Dimensional Astronomical Data Using Connected Morphological Operators [20]	SP		histogram	clustering						image		projection						T
Graph-Theoretic Scagnostics - Wilkinson et al. [54]	SP			clustering		outliers	complex patterns			image		projection						
Selecting good views of high-dimensional data using class consistency - Spee et al. [48]	SP								class pres.	data		projection						T
Coordinating computational and visual approaches for interactive feature selection and multivariate clustering - Guo [22]			matrix		correlation					data		projection	ordering					
Exploring High-D Spaces with Multifom Matrices and Small Multiples - MacEachern et al. [35]			pixel based vis. matrix, small multiples		correlation					data		projection	ordering					
Improving the Visual Analysis of High-Dimensional Datasets Using Quality Measures - Albuquerque et al. [4]			jigsaw map, matrix, table	clustering	correlation	outliers				data	image	projection	ordering		visual mapping		view optimization	S, T
Interactive Hierarchical Dimension Ordering Spacing and Filtering for Exploration of High Dimensional Datasets - Yang et al. [58]	PC		histogram, star glyphs		correlation	outliers				data		projection	ordering					S, T
Interactive Dimensionality Reduction Through User-defined Combinations of Quality Metrics - Johansson & Johansson [30]	PC			clustering	correlation	outliers				data		projection	ordering					S, T
Paragistics: Image-Space Metrics for Parallel Coordinates - Dasgupta & Kosara [15]	PC			clustering	correlation			image quality			image	projection	ordering					S
Combining automated analysis and visualization techniques for effective exploration of high-dimensional data - Tatu et al. [48]	SP			clustering	correlation		complex patterns		class pres.	data	image	projection	ordering					
High-Dimensional Visual Analytics: Interactive Exploration Guided by Pairwise Views of Point Distributions - Wilkinson et al. [55]	SP			clustering		outliers	complex patterns				image	projection	ordering					
Clutter Reduction in Multi-Dimensional Data Visualization Using Dimension Reordering - Peng et al. [39]	SP		star glyphs, dim. stacking		correlation	outliers		image quality		data	image		ordering					
Similarity Clustering of Dimensions for an Enhanced Visualization of Multidimensional Data - Ankerst et al. [5]	PC		recursive pattern, circle segments		correlation					data			ordering					
Measuring Data Abstraction Quality in Multiresolution Visualizations - Col et al. [14]	SP		histogram						feature pres.	data					abstraction			T
Quality Metrics for 2D Scatterplot Graphics: Automatically Reducing Visual Clutter - Bertini & Santucci [8]	SP			clustering					feature pres.	data	image				abstraction			
A Screen Spaces Quality Method for Data Abstraction - Johansson & Cooper [28]	PC								feature pres.		image				sampling			
Enabling Automatic Clutter Reduction in Parallel Coordinates Plots - Ellis & Dax [12]	PC							image quality			image				sampling			T
Phonostics: Towards measuring the value of visualization - Schneidewind et al. [42]			jigsaw map, pixel bar chart		correlation		complex patterns			data	image				visual mapping			

Figure 5: A 1-N classification created to systemise quality metrics factors for high-dimensional data. Courtesy of Bertini et al. [BTK11]

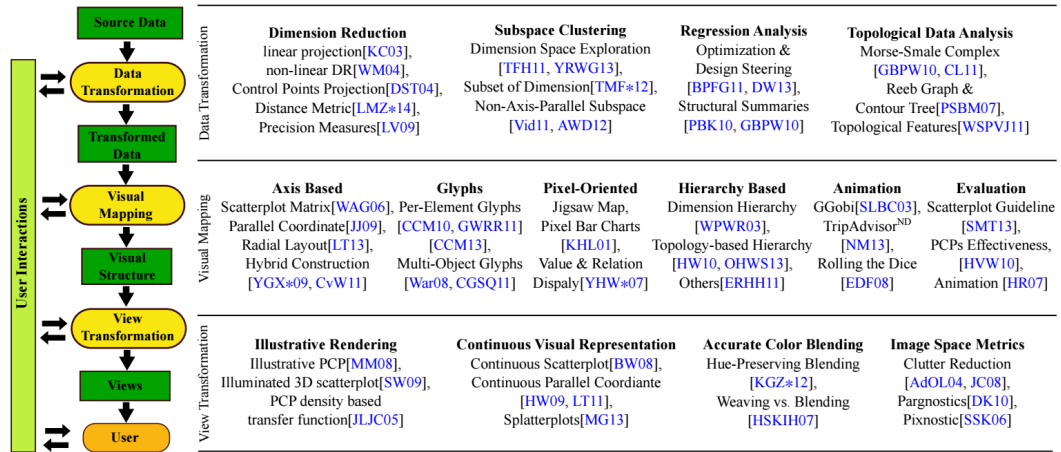


Figure 6: A 2D classification designed using the information visualisation pipeline for the taxonomy of high dimensional data. Courtesy of Liu et al. [LMW*15]

	Analyse Clusters	Analyse Correlations	Find Outliers	Value Retrieval	Detect 1 pattern	Detect N patterns	Trace Lines
2DPC colour [15]	Yellow						
2DPC blending [15]	Yellow						
2DPC colour+blending [15]	Yellow						
2DPC curves [15]	Yellow						
2DPC + Scatter plots [15]	Green						
2DPC random ∇ [15]	Red						
2DPC permutation ∇ [15]	Yellow						
2DPC wobble[15] ∇	Yellow						
3D Multi-relational PC [6, 21]					Yellow	Green	
Bundled PC [12]	Green	Yellow					
Edge-bundled PC [33]	Green	Green					Green
Many-to-many PC [30, 5]					Green		
3D PC [20]					Red	Red	
Progressive PC [38]					Green	Green	
Scatter plots [29, 26, 10]		Green					
Scatter plots, rotated [26]				Red			
Scatter plots, staircase [26]				Red			
Stacked areas [10]		Red					
Stacked lines [10]		Red					
Stacked bars [10]		Red					
Donuts [10]		Red					
Radar Charts [10]		Red					
Line plots [10]		Red					
Ordered line plots [10]		Red					
Tables/lists [2]					Yellow	Red	
Radviz [35]	Green		Red				

Equal to 2DPC
 Better than 2DPC
 Worse than 2DPC
 Not evaluated

Figure 7: A 1-N classification of 26 techniques performed in relation to standard 2D parallel coordinates. Yellow colour indicates no significant difference in performance. Green colour means that the technique outperforms 2DPC for the specific task. Red colour shows the technique performs worse than 2DPC. Light blue colour reveals no evaluation has been found in the literature. ∇ denotes that the technique is based on animation. Courtesy of Johansson and Forsell [JF16].

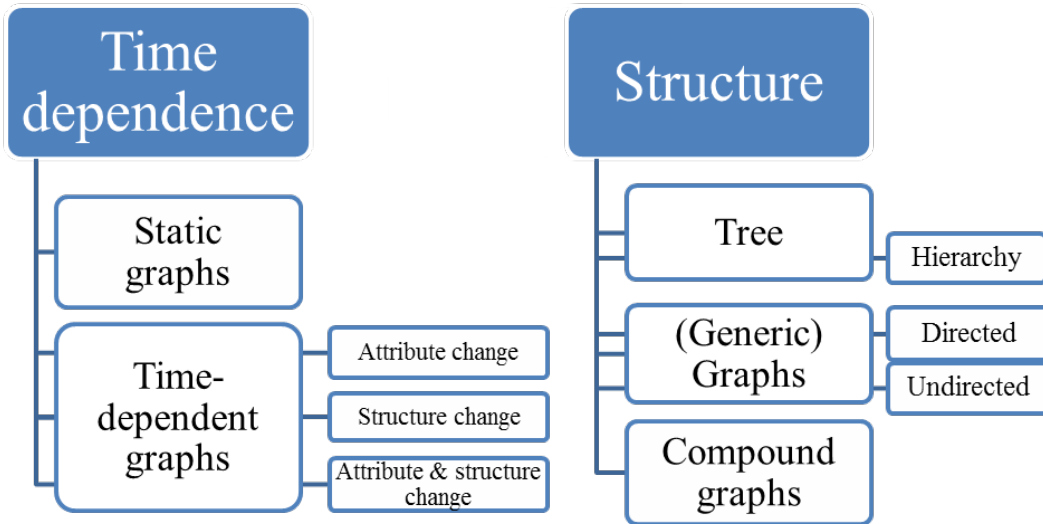


Figure 8: Classification of graphs with respect to the temporal or structural characteristics. Courtesy of Von Landesberger et al. [VLKS* 11]

	Data	Properties	Data Types			Representative Datasets
			N	C	T	
Trajectory	Shipping trajectories	Time	✓			Vessel traffic data [4]
		Location	✓			
		Ship type		✓		
		Destination			✓	
		Velocity	✓			
	Aircraft trajectories	Location	✓			Flight in France [5], Europe 24 [6]
		Flight level	✓			
		Time	✓			
		Velocity	✓			
	Automobile trajectories	Aircraft ID			✓	Taxi GPS data of Beijing [7], [8], Shenzhen [9], Shanghai [10], [11], [12], San Francisco [13], New York City [14], Wuhan [15], [16], and Sweden [17]; Traffic monitoring cells data in Nanjing [18]; GPS data in Louisiana [19]
		Time	✓			
		Location	✓			
		Direction		✓		
		Change of direction		✓		
		Velocity	✓			
Train/Metro trajectories	Acceleration	✓			Train data in France [20], Boston's metro data [21]	
	Pick-up/drop-off		✓			
	Location	✓				
Pedestrian trajectories	Time	✓			Human mobility traces [22]	
	Station			✓		
	Location	✓				
Mixed trajectories	Velocity	✓			Intersection count [23]	
	Object type		✓			
	Position	✓				
	Velocity	✓				
	Direction		✓			
Incident	Tunnel incident	Time	✓			Incident detection system (IDS) data [24]
		Stateful events		✓		
		Stateless events	✓			
	Highway incident	Video				Maryland highway & traffic information [25], Traffic Management Centers Data [26], traffic incident in Singapore [27]
		Location	✓			
		Time of date	✓			
		Weather conditions		✓		
		Vehicle involved			✓	
	Metro incident	Incident type		✓		Metro smartcard records in Shenzhen [28], urban rail transit system data [29]
		Time	✓			
Station				✓		
Check in/out			✓			

Figure 9: The taxonomy displays different data types with their potential properties. These are then categorized into three data types: Numerical; Categorical; or Textual. Examples of related literature are also given. Courtesy of Chen et al. [CGW15]

Taxonomic Category	Illustration	#	Techniques
I. Animation (Time-to-Time Mapping)		33	
I.a. General-Purpose Layout		21	
I.a.1. Online Problem		6	<p>layout adjustment to avoid overlap [MELS95] framework based on Bayesian decision theory [BW97] simulated annealing with customizable weights for optimization criteria [LLY05] efficient algorithm and GPU implementation [FT08b] consider age of nodes to stabilize the layout [GdBG12] more efficient initial positions of nodes [HMHU13]</p>
I.a.2. Offline Problem		7	<p><i>Foresighted Layout (with Tolerance)</i> [DGK01, DG02, GBPD05] <i>GraphAEL</i>: force-directed layout with virtual forces between time steps [EHK*04b, FKN*05] <i>Visione</i>: force-directed layout with additional energy factors between time steps [BS08] user-selected multiple foci [FWSL12]</p>
I.a.3. Transition Problem		8	<p>stepwise animation for navigation based on a spring algorithm [HEW98] <i>Marey</i>: stepwise animation moving (parts of) the graph together [FE01, FE02, FH02, NF02] <i>VisuGraph</i>: using super-graph as intermediate step [LD08] transitions of bundled edges [HEF*13] <i>GraphDiaries</i>: highlight changes in staged transitions [BPF14a]</p>
I.b. Special-Purpose Layout		12	
I.b.1. Compound Graphs		8	<p>force-directed approach preserving the position of clusters [FT04] nested bubbles in 3D [KG06] <i>XLDN</i>: extending <i>Foresighted Layout with Tolerance</i> to dynamic compound graphs [PB08] focused animation collapsing constant parts of the hierarchy [RPD09] <i>ContaxTour</i>: smooth contours of colored clusters [LSCL10] Space-filling maps of colored clusters [MKH12, HKV12] degree-of-interest functions for abstracting and focusing large graphs [AHSS13]</p>
I.b.2. Other		4	<p>online drawing of planar graphs [CDBT*92, CDBTT95] <i>DynaDAG</i>: acyclic graphs based on hierarchical layout [Nor95] stable layout of small world graphs [BFP06]</p>
II. Timeline (Time-to-Space Mapping)		23	
II.a. Node-Link		15	
II.a.1. Juxtaposed		5	<p><i>TimeArcTrees</i>: linearized nodes on vertical axes [GBD09] <i>Parallel Edge Splatting</i>: artificially bipartite, linearized node layout [BVB*11, BBW12] nested circles: partial links in <i>TimeSpiderTrees</i> [BFD10], ego centered graphs [FHQ11]</p>
II.a.2. Superimposed		5	<p>3D stack with fixed positions [BC03, DE02] 3D stack with relaxed positions [EKLND4, GW06] abstracting nodes and links to tubes [GHW09]</p>
II.a.3. Integrated		3	<p>ego network with edges as timelines [Rei10] ego network with ego node as timeline [Sww11] <i>Extended Massive Sequence Views</i>: event-based timeline with parallel edges [vdEHBW13]</p>
II.a.4. Hybrid (Juxt., Super., Int.)		2	<p>juxtaposition as well as 2D and 3D superimposition [FAM*11, ITK10]</p>
II.b. Matrix		8	
II.b.1. Intra-Cell Timelines		4	<p>time series as sparkline bar charts [BSW13, YEL10] <i>Gesta@lines</i> encoding three metrics in angles and line lengths [BN11] pixel-based folded timelines [SWS10]</p>
II.b.2. Layered Matrices		4	<p><i>(Layered) TimeFladar Trees</i>: radially layered lists with radial matrix thumbnails [BD08, BHW11] radially banded and layered matrices [VBSW13] <i>Cubix</i>: stacked matrices to a 3D cube and sliced small multiples thereof [BPF14b]</p>
III. Hybrid (Animation, Timeline)		4	<p>in situ integration of small visualizations [HSS11] cluster evolution on a timeline for navigating animated node-link diagrams [SMM12] moving timeline based on <i>Parallel Edge Splatting</i> [BBV*12] <i>DiVAri</i>: combinations of small multiples, difference representations, and animation [RM13]</p>

Figure 10: Hierarchical taxonomy of dynamic graph visualisation courtesy of Beck et al. [BBDW14]

		Graph elements (nodes, graph objects)		
		Both constraints	One constraint, one target	Both are targets
Time points	Both constraints	Find connections between elements (comparison) (How) is graph element g_1 connected to graph element g_2 at the given time, t ? $? \lambda : (g_1, t) \wedge (g_2, t)$	Find elements connected in the given way (relation seeking) Find the graph element(s) to which graph element g_1 is connected in the given way at time t : $? g_2 : (g_1, t) \wedge (g_2, t)$	Find elements connected in the given way (relation seeking) Find graph objects which are connected in the given way at the given time: $? g_1, g_2 : (g_1, t) \wedge (g_2, t)$
	Both targets	Hybrid Find the time points at which two given graph objects were connected in the given way: $? t : (g_1, t) \wedge (g_2, t)$	Find elements connected in the given way (relation seeking) Find the graph element(s) to which graph element g_1 is connected and the time(s) at which the connection(s) occur: $? g_2, t : (g_1, t) \wedge (g_2, t)$	Find elements connected in the given way (relation seeking) Find graph objects (and their associated time points) at any time that are connected in the given way: $? g_1, g_2, t : (g_1, t) \wedge (g_2, t)$

Figure 11: Classification of elementary structural task variations. Courtesy of Kerracher et al. [KKC15]

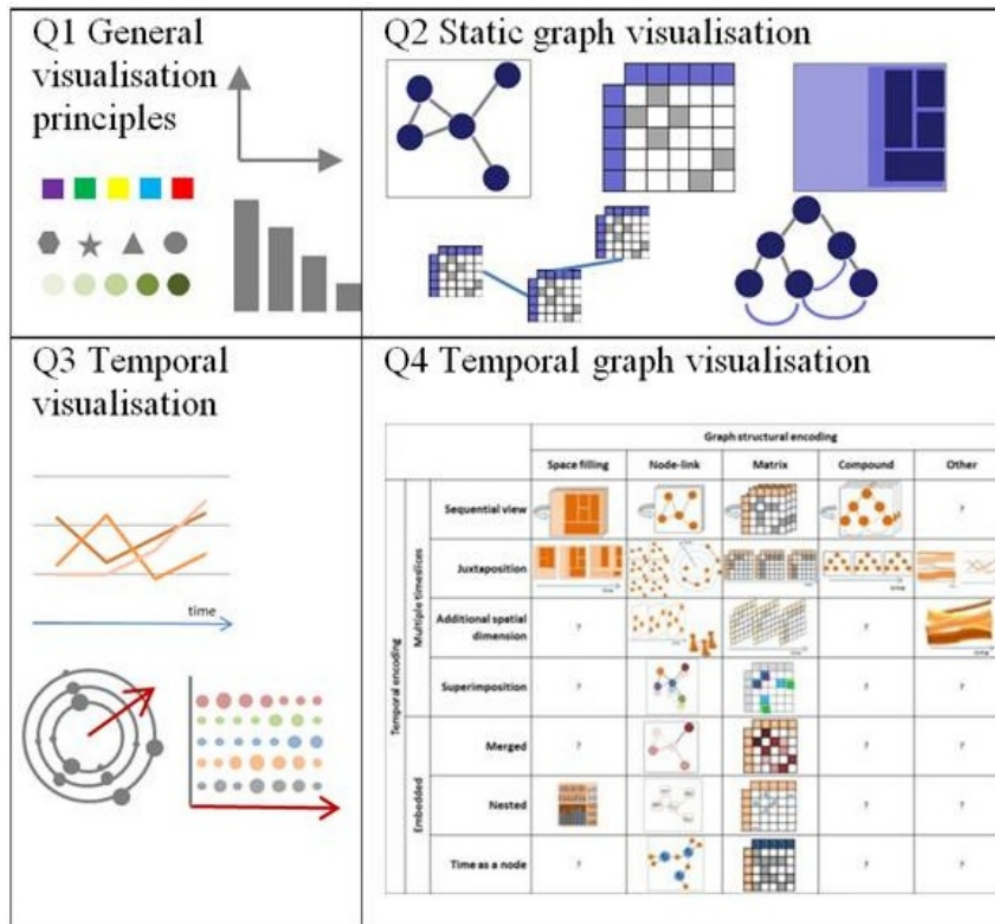


Figure 12: Research areas and techniques associated with data items by quadrant. Courtesy of Kerracher et al. [KKCG15]



Figure 13: Taxonomy presented by Behrlich et al. classifying different matrices reordering algorithms [BBR*16]

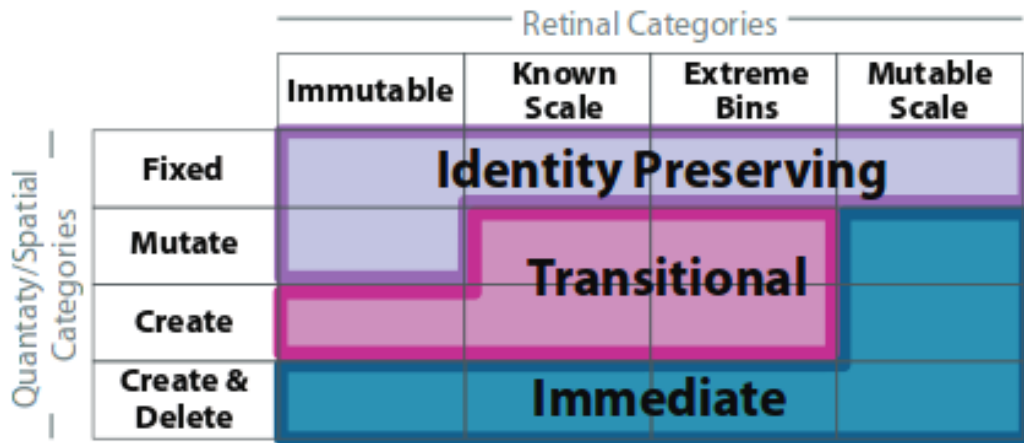


Figure 14: A matrix created by Cottam et al. to classify different dynamic visualisation techniques [CLW12]

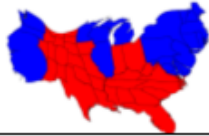


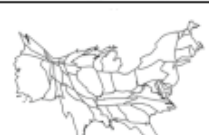


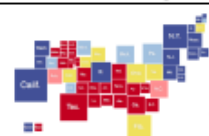
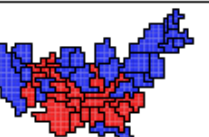

Type	Statistics	Contiguity	Geography	Topology	Example
Diffusion-based cartograms [GN04]	Almost accurate	Contiguous	Distorted	Topology-preserving	
Circular-arc cartograms [KKN13]	Not Accurate	Contiguous	Shape mostly preserved	Topology-preserving	
Optimal rubber sheet method [Sun13b]	Almost accurate	Contiguous	Distorted	Topology-preserving	
Fast, free-form rubber-sheet method [Sun13a]	Almost accurate	Contiguous	Distorted	Topology-preserving	
T-shape cartograms [ABF*13]	Accurate	Contiguous	Shape not preserved	Topology-preserving	
Non-contiguous cartograms [Ols76]	Accurate	Not contiguous	Shape preserved	Topology not preserved	
Demers cartograms [BDC02] (figure from [NYT12])	Accurate	Not contiguous	Shape not preserved (squares)	Topology not preserved	
Mosaic cartograms [CBC*15]	Not accurate	Contiguous	Shape mostly preserved	Topology-preserving	
Table cartograms [EFK*13]	Accurate	Contiguous	Shape not preserved	Topology not preserved	

Figure 15: A 2D systematic overview of different types of cartograms, displayed with their categorisations. Courtesy of Nusrat and Kobourov [NK16].

Technique	Visualization A	Visualization B	Spatial Relation	Data Relation
ComVis [24] (Figure 2)	any	any	juxtapose	none
Improvise [39] (Figure 3)	any	any	juxtapose	none
Jigsaw [36]	any	any	juxtapose	none
Snap-Together [30]	any	any	juxtapose	none
semantic substrates [34] (Figure 4)	node-link	node-link	juxtapose	item-item
VisLink [11] (Figure 5)	radial graph	node-link	juxtapose	item-item
Napoleon's March on Moscow [37]	time line view	area visualization	juxtapose	item-item
Mapgets [38] (Figure 6)	map	text	superimpose	item-item
GeoSpace [22] (Figure 7)	map	bar graph	superimpose	item-item
3D GIS [8]	map	glyphs	superimpose	item-item
Scatter Plots in Parallel Coordinates [45] (Figure 8)	parallel coordinate	scatterplot	overload	item-dimension
Graph links on treemaps [14] (Figure 9)	treemap	node-link	overload	item-item
SparkClouds [21]	tag cloud	line graph	overload	item-item
ZAME [13] (Figure 10)	matrix	glyphs	nested	item-group
NodeTrix [17] (Figure 11)	node-link	matrix	nested	item-group
TimeMatrix [44]	matrix	glyphs	nested	item-group
GPUVis [25]	Scatterplot	glyphs	nested	item-group

Figure 16: Classification of common composite visualisation techniques. Image courtesy of Javed and Elmqvist [JE12].

Category	Performance Visualization Techniques	Example applications and studies
Simple visual structures	Pie charts, distribution, box plots, kiviati diagrams	ParaGraph [2], PET [20], SvPablo [16], VAMPIR [21], Devise [22], AIMS [9]
	Timeline views	Paje [23], AIMS [9], Devise [22], AerialVision [24], Paraver [25], SIEVE [14], Virtue [13], utilization and algorithm timeline views in [17]
	Information typologies	SHMAP [26], Vista [4], Voyeur [27], processor and network port display in [28], hierarchical display in [12]
Composed visual structures	Information landscape	Triva [29], Cichild [30]
	Trees & networks	Paradyn [18], Cone Trees [31], Virtue [13], [32]
	Single-axis composition	AIMS [9], Vista [4]
Interactive visual structure	Double-axis composition	Devise [22], AerialVision [24]
	Case composition	Triva [29]
	Interaction through controls (data input, data transformation, visual mapping definition, view operations)	Paje[23], data input, filtering, and view manipulation in [28] and [32]
Focus + context visual structures	Interaction through images (magnifying lens, cascading displays, linking and brushing, direct manipulation of views and objects)	Virtue [13], Cone Trees [31], Devise [22], direct manipulation of the 3D cone and virtual threads in [32]
	Macro-micro composite view	Microscopic profile in [4], PC-Histogram in [24]

Figure 17: A classification of performance visualisation techniques courtesy of Gao et al. [GZR*11]

Visualization System	Visualization Technique(s)	Data Source(s)	Number of Citations	
Host / Server Monitoring				
Erbacher et al. [4][5]	Glyph	Server Logs	106 7	
Tudumi [6]	3D Node Link	Server Logs	38	
NVisionIP [7,8]	Scatter Plot	NetFlows	145 20	Available Online
Portall [9]	Node Link	Packet Traces	21	
HoNe [10]	Node Link	Packet Traces	8	
Perlman et al. [11]	Node Link Glyph	Packet Traces	5	
Radial Traffic [12]	Radial Panel	Packet Traces	23	
Mansmann et al. [13]	Node Link	Packet Traces	2	
Internal/External Monitoring				
VISUAL [14]	Scatter Plot IP Matrix	Packet Traces	93	
VizFlowConnect [15]	Parallel Coordinates	NetFlows	111	Available Online
Erbacher et al. [16]	Radial Panel	Packet Traces	8	
TNV [17]	IP Matrix Color Map	Packet Traces	48	Available Online
Port Activity				
Abdullah et al. [18]	Histogram	Packet Traces	30	
Cube of Doom [19]	3D Scatter Plot	Packet Traces	99	Available Online
PortVis [20]	Scatter Plot	NetFlows	112	
NetBytes Viewer [21]	3D Scatter Plot	NetFlows	7	
Existence Plots [22]	Scatter Plot	Packet Traces	3	
Attack Patterns				
Giardin [29]	Color Map	Packet Traces	60	
NIVA [30]	Node Link Glyph	Intrusion Alerts	51	
Snort View [31]	Scatter Plot Glyph	Intrusion Alerts	67	Available Online
IDGraphs [32]	Scatter Plot	NetFlows	29	
IP Matrix [33]	Scatter Plot Color	Intrusion Alerts	21	
Visual Firewall [34]	Scatter Plot	Packet Traces	24	
IDS Rainstorm [35]	Scatter Plot	Intrusion Alerts	60	
Vizalert [36][37][38]	Radial Panel	Intrusion Alerts	38 35 29	
Rumint [39][40]	Parallel Coordinates	Packet Traces	15 35	Available Online
Ren et al. [41]	Flying Term	DNS Traces	10	
Xiao et al. [42]	Scatter Plot	Packet Traces	23	
Svision [43]	3D Scatter Plot	Packet Traces	9	
Mansmann et al. [44]	Treemap	Packet Traces	20	
SpiralView [45]	Radial Panel	Intrusion Alerts	5	
NFlowVis [46]	Treemap	NetFlows	17	
Avisa [49]	Radial Panel	Intrusion Alerts	2	
Routing Behavior				
BGPlay [50]	Node Link	BGP Traces	22	
Wong et al. [51]	Node Link	BGP Traces	9	
LinkRank [52]	Node Link	BGP Traces	16	
Teoh et al. [53][54][55]	Histogram Node Link	BGP Traces	54 28 35	
BGP Eye [56]	Color Map	BGP Traces	8	

Figure 18: Taxonomy of Security Visualisation Systems, divided into different use-cases. Created by Shiravi et al. [SSG12]

Relations	Major Tasks	Design Choices			Pros	Cons
		Visual Representation	Supplementary Visual Technique	Interaction Design		
Entity Level	1. Show an entity 2. Show a group of entities 3. Show entity level relations (a single case or multiple cases) 4. Show single entity vs. groups of entities level relations (a single case or multiple cases) 5. Find relevant entities for a specific entity 6. Verify relations between some entities 7. Discriminate some entities from others 8. Mark important entities or relations	The Node-Link Diagram	1. Edge bundling 2. Use spatial distance (e.g. the force-directed layout) 3. Use spatial distance + hiding links 4. Color coding to separate nodes of different domains or selected and unselected nodes or links 5. Visual marks (e.g., shapes) to separate nodes and/or links	1. Select nodes/links 2. Highlight nodes/links 3. Drag nodes/links	1. An intuitive way to show either an entity or multiple entities and relations between entities 2. Customizable spatial layout for users 3. Links clearly show specific relations between entities	1. Entities are randomly placed in the space, so it may be difficult to find an entity if there are many entities 2. The number of links exerts much impact on the readability of the diagram 3. Without links, relations between entities cannot be identified easily 4. Color coding and visual marks are not efficient to visually separate domains
		A Simple Matrix	1. A single cell to represent Entity 2. A row or a column to represent Group 3. Use a heatmap	1. Select cells 2. Highlight cells 3. Extract a cell 4. Merge cells	Avoid visual clutters caused by too many links	1. Not as easy as node-link diagrams to perceive 2. Columns or rows rearrangement is the only way to change the layout
		Parallel Coordinates with Two Domains	1. Edge bundling 2. Using curved lines to indicate links	1. Select entities 2. Highlight polylines/entities 3. Brushing 4. Axes rearrangement 5. Entities reposition in axes	1. Place entities of the same group together 2. Relatively easy to find entities 3. Efficiently select multiple entities/polylines	1. The number of links exerts much impact on the readability of the diagram 2. Without links, relations between entities cannot be identified easily 3. Sometimes entity reposition (e.g., moving relevant entities to the top) is necessary to understand grouping
Group Level		Tree Visualizations	1. Icicle 2. Bubble trees 3. Treemaps	1. Select nodes/links 2. Highlight nodes/links 3. Path extraction	Clearly represent hierarchical relations	1. Not all Groups are hierarchical relations 2. Cannot represent biclusters and bicluster-chains
Bicluster Level	1. Show a bicluster 2. Show all biclusters 3. Find biclusters of interest 4. Mark biclusters of interest	Matrices	1. Use a heatmap 2. Reorder rows or columns 3. Repeat rows or columns 4. Color coding the region of a bicluster	1. Reorder rows/columns 2. Select biclusters 3. Highlight biclusters 4. Replicate rows/columns	1. A visual representation that is easy to understand biclusters 2. Efficiently reduce visual clutters caused by many links	1. It is difficult to display all biclusters without replicating rows and/or columns 2. Repeated rows or columns may cause confusion 3. Overlaps may obscure biclusters with less entities
		Parallel Coordinates with Two Domains	1. Edge bundling 2. Use curved lines 3. Wrap entities with polylines 4. Tile-based parallel coordinates	1. Select entities 2. Brushing 3. Highlight polylines/entities/ribbons 4. Axes rearrangement 5. Entities reposition in axes	1. Place entities of the same group together 2. Relatively easy to find entities 3. Efficiently select multiple entities/polylines	1. The number of links exerts much impact on the readability of the diagram 2. Without links, relations between entities cannot be easily identified 3. Sometimes entity reposition (e.g., moving relevant entities to the top) is necessary to understand the relation
		Zoned Node-Link Diagram	1. Wrap nodes of a bicluster in a colored region 2. Use force-directed layout 3. Hide links between nodes	1. Select nodes/links 2. Highlight nodes/links 3. Drag nodes/links	1. Customizable spatial layout for users 2. Links clearly show relations between specific entities 3. Easily find entities that are shared between biclusters	1. Entities are randomly placed in the space, so it may be difficult to find an entity if there are many entities 2. Without links, relations between entities cannot be identified easily 3. Biclusters with less entities may be obscured in the overlapping region
Chain Level	1. Show a chain 2. Show all chains 3. Find chains of interest 4. Mark chains of interest	Node-link Diagram + Matrices	Combine all supplementary visual techniques that the node-link diagram and matrix based visualizations can use and the Bubble Sets technique	Combine all interactions that the node-link diagram and matrix based visualizations can use and path extraction	1. Efficiently reduce the number of links 2. A customizable spatial layout for users 3. Show the overview of the data based on bicluster-chains	1. Entities may replicate many times in multiple matrices 2. Not a trivial visualization for users to understand connections across several biclusters 3. Which bicluster to choose to start a bicluster-chain is a problem
		Parallel Coordinates + Matrices	Combine all supplementary visual techniques that parallel coordinates and matrix based visualizations can use and the Bubble Set technique	Combine all interactions that parallel coordinates and matrix based visualizations can use and path extraction	4. By following links, users can find out how a bicluster-chain is formed	
Schema Level	1. Show the overview of a dataset 2. Guide the exploration of chains or biclusters	The Node-Link Diagram	1. Clutter Map 2. The PivotGraph technique 3. Color coding to indicate different domains 4. Visual marks (e.g., shapes) to separate nodes and/or links 5. Use spatial distance (e.g. force-directed layout) 6. Use spatial distance + hiding links	1. Select nodes/links 2. Highlight nodes/links 3. Dynamic path extraction	1. An intuitive way to show relations between domains 2. The size of nodes and the thickness of links can be used to encode the information of biclusters and/or chains	1. The layout of PivotGraph cannot be easily changed by users 2. Depend on links to perceive relations across several specific domains
		The Chord Diagram	1. Color coding of chords to indicate different domains 2. Use ribbons between chords to indicate connections	1. Select chords/ribbons 2. Highlight chords/ribbons	1. An intuitive way to show relations between domains 2. The length of chords and the thickness of ribbons can be used to encode the information of bicluster and/or chains	1. Not efficient for a dataset with many domains 2. Ribbons inside the diagram may form visual clutters 3. Paths inside the diagram may be obscured by too many crossing ribbons

Figure 19: Design framework associated with bicluster visualisation. Courtesy of Sun et al. [SNR14]

InfoVis techniques	Examples
Empirical methodologies	
Model	[11, 34, 35, 52, 65, 66, 84, 95, 119, 128, 146, 153]
Evaluation	[4, 12, 14, 15, 18, 49, 60, 69, 78, 82, 98, 100, 101, 103, 104, 115, 116, 131, 156]
Interactions	
WIMP interactions	[37, 55, 135]
Post-WIMP interactions	[13, 70, 147]
Frameworks	
Systems and frameworks	[2, 17, 28, 57, 89, 153]
Applications	
Graph visualization	[3, 8, 9, 13, 19, 20, 30, 36, 40, 42, 51, 59, 62, 85, 91, 118, 120, 133, 162, 164, 167, 170]
Text visualization	[1, 5, 22, 31, 32, 83, 92–94, 154, 159, 163, 169]
Map visualization	[1, 44, 71, 102, 106, 117, 125, 136, 144, 148]
Multivariate data visualization	[21, 48, 68, 72, 108, 112, 134, 139, 140]

Figure 20: A taxonomy of InfoVis techniques created by Liu et al. [LCWL14]

Higher-Res Tables

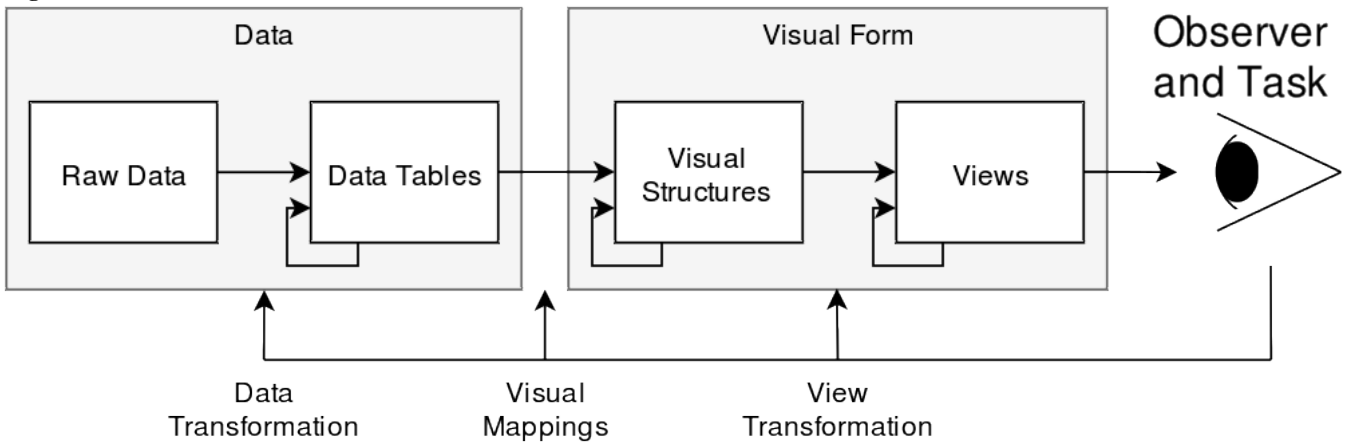


Figure 21: The original Information Visualisation Pipeline model created by Card et al. [CMS99] which we adapt to design our modified classification.

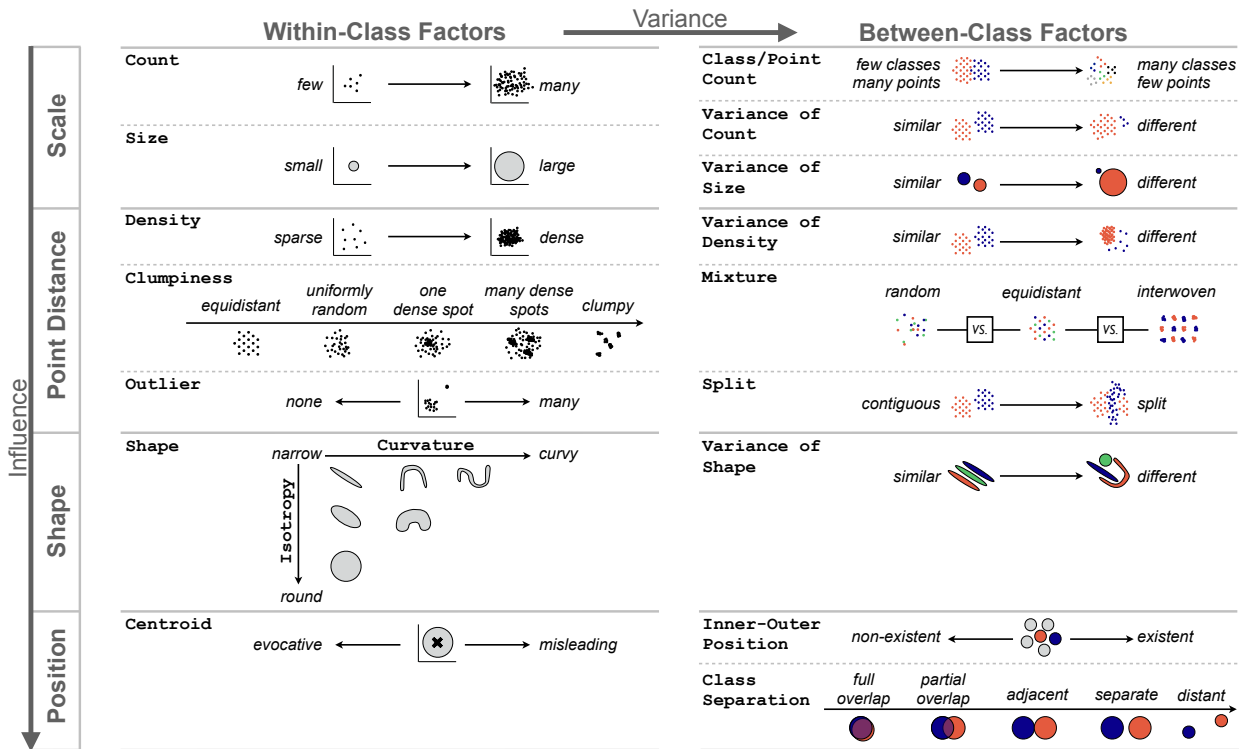


Figure 22: An indirect mapping taxonomy of data characteristics with respect to class separation in scatterplots. Courtesy of Sedlmair et al. [STMT12]

	Papers	Taxonomy				Demonstrated Scale*		Global Compr.		Problem Detection			Diagnosis/ Attribution	
		H	S	T	A	Data	Parallel	Program Structure	Resource Usage	Anomalies	Bottlenecks / Imbalance	Resource Misuse	Software	System
Visualization Techniques	Papers													
Radial Tree	Bhatele et al. [BGI*12]			X		NR	10 ⁴ processes	X	X		X			
Node-Link Graph	Boxfish [LLBT12, ILGT12]	X				NR	10 ⁴ nodes		X		X			X
Radial Tree, Animation	Choudhury and Rosen [CR11]	X	X			10 ⁷ transactions	N/A	X	X		X	X	X	
Layered Node-Link	DOTS [BKS05]	X	X			NR	NR	X	X		X	X	X	
Clustered Node-Link, Animation	Frishman et al. [FT05]	X	X			NR	10 ² objects	X	X	X				
Node-Link Graph	Heapvis [AKGT10]		X			10 ³ nodes		X	X				X	
Radial Tree	Kim et al. [KLJ07]		X	X		NR	10 ³	X		X				X
Node-Link Trees, Indented trees	Lin et al. [LTOB10]		X			NR	NR	X	X					X
Node-Link trees	DeRose et al. [DHJ07]		X	X		NR	10 ² cores				X			X
Node-Link graph, Animation	Sambasivan et al. [SSMG13]		X			NR	NR	X						X
Radial Tree	Sigovan et al. [SMMT13a]	X				10 ¹ resources	10 ³ processes		X	X			X	
Node-Link trees	STAT [AdSLT09]		X	X		NR	10 ⁵ tasks	X			X			X
Clustered Node-Link, Animation/Real Time	Streamsight [DPA09]		X	X		streaming	10 ³ tasks	X	X	X	X			
Layered None-Link	Threadscope [WT10]	X		X		10 ³ events	10 ¹ threads	X	X	X	X			
Node-Link Graph, Treemap	Weidendorfer et al. [WKT04]		X			NR	1	X						X
Timeline, Stacked Graph, Small Multiples	de Pauw et al. [DPWB13]	X	X			streaming	10 ³ tasks	X	X		X			X
Shared Timeline	Muelder et al. [MGM09]			X		NR	10 ⁴ processes	X			X			
Gantt Charts, Timeline, Matrix, Scatterplot	Muelder et al. [MSMT11]	X	X	X		NR	10 ³ cores	X	X		X			
3D Parallel Gantt Chart, Treemap/Force-directed layouts	Triva [SHN10]	X	X			NR	10 ³ processes	X	X		X			
Parallel Gantt Chart, Node-Link Tree, Bar Charts	Zinsight [DPH10]		X	X		10 ⁵ events	10 ² processes	X	X		X			X
1D Color-Coded Array, Histograms	Cheadle and Field [CFAT06]	X	X			10 ¹ memory groups	N/A	X						X
1D Color-Coded Array Stacked By Time	Moreta and Telea [MT07]	X				10 ⁵ allocations	N/A	X				X	X	
Edge Bundling, Gantt Charts, Hierarchies	Extravis [CHZT07]		X			10 ⁵ events	NR	X						X
Parallel Gantt Chart, Indented Trees, Code view	HPCToolkit [ABF*10, TMC*11, LMC13]	X	X	X		10 ¹ gigabytes	10 ⁴ processes	X	X		X	X	X	X
Stacked Barcharts, Stacked Timelines	Lumière [BBH08]	X	X	X		10 ⁶ decisions	NR	X	X	X	X			X
Parallel Gantt Chart, Small multiples, Plots, Ensemble	Projections [KZKL06, LMK08]		X	X		gigabytes	10 ⁴ processes	X	X		X	X	X	
Stacked Barcharts, Scatterplot, Histograms, Code Coloring	TraceVis [RZ05]	X	X			10 ⁷ instructions	NR	X	X		X	X	X	
Icicle Timelines, Coordinated views	Trumper et al. [TBD10]		X	X		10 ⁴ events	10 ¹ threads	X			X	X	X	
Parallel Gantt Chart, Icicle Timeline, Adjacency, Indented Trees, Ensemble Timeline, Plots	Vampir [NAW*96, BW12, ISC*12, VMa13]	X	X	X		terabytes	10 ⁵ processes	X	X		X			
Abstract Diagram	Choudhury et al. [CPP]	X	X			10 ¹ buffers	N/A				X	X	X	
Dot Plot, Bar Charts	Iviz [WYH10]		X	X		10 ⁶ events	2 jobs	X			X			X
Scriptable	ParaProf [SML*12]			X		NR	10 ⁴ processes		X		X			
Indented Trees, Matrix	Scalasca [GWW*10, WG11]		X	X		terabytes	10 ⁵ cores	X	X		X			X
Color-coded 2D matrix, histograms, 3D graph layout	Schulz et al. [SLBT11]	X	X	X		NR	10 ⁴ cores		X	X	X	X		
Bubble Chart, Animation	Sigovan et al. [SMM13]		X			NR	10 ⁴	X	X		X			
City Metaphor	SynchroVis [WWF*13]		X	X		10 ² objects	10 ¹ threads	X		X				X
Icicle Timeline, Bundles	SyncTrace [KTD13]		X	X		10 ⁷ events	10 ² threads	X	X		X	X	X	
Sunburst, Matrix, Dendrogram	Travis [AH10]		X			10 ³ nodes	NR	X						X

Figure 23: Isaacs et al. present a 1-N design space that classifies literature based on the context, scale and goal of each paper. Image courtesy of Isaacs et al. [IG*14]

		Close Reading					Distant Reading						
		Plain	Color	Font size	Glyphs	Connections	Structure	Heat maps	Tag clouds	Maps	Timelines	Graphs	Miscellaneous
Single Text Analysis	enhanced text views	[Pie10], [CGM*12], [Pie13], [GWF14]			x								
		[PSA*06], [CTA*13], [Ben14], [BJ14]		x									
		[ARLC*13]				x	x						
	both	[WMN*14]		x	x								
		[VCPK09], [BGHJ*14], [KJW*14]		x				x		x			
		[WJ13b], [CMLM14], [KZ14]		x			x						
		[Cay05]	x					x					
		[CDP*07]		x					x				
		[WV08]		x									x
	abstract text views	[MFM13]	x										x
[RSDCD*13]		x										x	
[KO07], [FS11], [CTA*13], [OKK13], [Ben14]								x					
[Pie05]							x						
Parallel Text Analysis	section alignments	[PBD14]											x
		[WH11], [HKTK14]		x									
		[Cor13], [WJ13b]		x			x						
		[JRS*09]		x				x					
	sentence alignments	[GCL*13]		x					x				
		[JGBS14b]		x			x		x				x
Corpus Analysis	statistics for textual entities	[BGHE10]		x		x							
		[JGBS14a]			x	x							
		[Bea08], [Bea11], [Bea12], [BJ14]								x			
		[WJ13a], [HCC14]											x
	relationships between texts	[CWG11]		x	x				x				
		[Mur11]		x					x				
		[FKT14]	x						x	x			
		[EX10], [Gal11], [WH11], [Joc12], [CEJ*14], [Ede14]											x
		[RRRG05]							x				
	relationships between textual entities	[OST*10]		x									x
		[Wol13]		x									x
		[RRRG05], [AGL*07], [vHWV09], [KKL*11], [MLSU13], [WJ13a], [Arm14]											x
		[GZ12], [RFH14]		x									x
	social networks	[MH13]		x					x				x
		[AKV*14]		x					x				
		[Cob05], [CSV08], [BDF*10], [RD10], [BHW11], [Kle12], [Boo13], [KOTM13], [Tôt13], [Pet14]											x
	space and time	[KLB14]	x										x
		[JHSS12], [JW13], [DNCM14], [GD MF*14], [ÓML14]										x	x
		[Wea08]							x		x	x	
		[BPBI10]		x							x	x	
	space	[DWS*12]	x							x	x	x	
		[HACQ14]		x						x	x	x	
time	[MBL*06], [DFM*08], [Tra09], [GH11b], [EJ14]									x			
	[KBK11], [ARR*12], [LWW*13]											x	
	[CLT*11], [CLWW14]								x		x		
	[HSC08]	x									x	x	
	[DWS*12]	x							x	x	x		
	[ESK14]								x		x	x	
[HPR14]		x								x			

Figure 24: A 1-N Taxonomy by Jänicke et al. to map reading techniques found within different analysis methods. Image courtesy of Janicke et al. [JFCS15]

	2D	3D	Adjacency	Overlap	Inclusion	Rectangles/Cuboids	Squares/Pyramid	Triangles	Polygons	Circles (segments/ Frustums)	Ellipses/ Ellipsoids/ Cylinders	1D Subdivision	2D Subdivision	3D Subdivision	1D Packing	2D Packing	3D Packing	Year Published	Figure in the Appendix
Treemap [10],[11]																		1991	12(c)
Contrast Spiral Treemap [53]																		2007	13(o)
2 1/2D Treemap [65]																		1992	12(d)
Polar Treemap [15]																		1993	12(e)
Jigsaw Map [59]																		2005	13(g)
Generalized Treemap (Pie) [8]																		2006	13(j) - left
Generalized Treemap (Pyramid) [8]																		2006	13(j) - middle
Generalized Treemap (Pie+Pyramid) [8]																		2006	13(j) - right
Tree Cube [30]																		2003	13(c)
3D Treemap [15]																		1993	12(h)
Cushion Treemap [51]																		1999	12(m)
Voronoi Treemap [44]																		2005	13(f)
Circular Partitions [45]																		2008	13(a)
Quantum Treemap [66]																		2001	12(a)
Data Jewelry Box [58]																		2002	12(r)
Cascaded Treemap [19]																		2008	13(p)
Ellimap [48]																		2007	13(k)
Treemaps with Ovals [15]																		1993	12(f)
Pebble Map [36]																		2003	13(b)
CropCircles [37]																		2006	13(h)
Lifted Treemap [68]																		2009	13(r)
3D Nested Treemap [25]																		1999	12(n)
Information Cube [29]																		1993	12(i)
Nested Columns [15]																		1993	12(g)
2D Icicle Plot [12]																		1983	12(b) - left
Castles [17]																		1981	12(a)
Cushioned Icicle Plot [52]																		2007	13(l)
Triangular Aggregated Treemap [48]																		1998	12(l)
Sunburst [40]																		2000	12(p)
Interring [42]																		2002	12(s)
PieTree [38]																		2000	12(o)
Radial Edgeless Tree [46],[47]																		2007	13(m)
Step-tree [20]																		2004	13(d)
3D Icicle Plot [12]																		1983	12(b) - right
Nested Hemispheres [26]																		2004	13(e)
3D Nested Cylinders and Spheres [21]																		2006	13(i)
3D Sunburst [70]																		2007	13(n)
3D Beamtree [23]																		2002	13(a)
Information Pyramids™ [28]																		1997	12(k)
Cheops™ [50]																		1996	12(j)

Figure 25: Design space for implicit hierarchy visualization created by Schulz et al. to compare techniques in the field. Image courtesy of Schulz et al. [SHS11]

Authors / Technique	design guideline													visual channel						
	[DG1] visualization space	[DG2] complexity vs. density	[DG3] hybrid visualizations	[DG4] perceptually uniform properties	[DG5] redundant mapping	[DG6] importance-based mapping	[DG7] view point independence	[DG8] simplicity and symmetry	[DG9] orthogonality and normalization	[DG10] intuitive / semantical mapping	[DG11] balanced glyph placement	[DG12] facilitate 3D depth perception	[DG13] interactive occlusion control	color	shape	size / height / length	orientation	texture	opacity	
Brewer [Bre99]: Color use guidelines																				
Cleveland & McGill [CM84]: Graphical perception	2D/3D																			
Crawfis & Max [CM93]: Vector field visualization	3D	2																		
de Leeuw & van Wijk [dLwV93]: Local flow probe	3D	-3																		
Healey & Enns [HE99]: Combining textures and colors	2.5D	1																		
Healey et al. [HBE96]: Preattentive processing	2D																			
Kindlmann & Westin [KW06]: Glyph packing	3D	2																		
Kindlmann [Kin04]: Superquadric tensor glyphs	2.5D	1.5																		
Kirby et al. [KML99]: Concepts from painting	2D	1																		
Laidlaw et al. [LAK*98]: Stochastic glyph placement	2D	2																		
Li et al. [LMvW10]: Symbol size discrimination	2D																			
Lie et al. [LKH09]: Design aspects of glyph-based 3D visualization	3D	2																		
McGill et al. [MTL78]: Variations of box plots	2D	-3																		
Meyer-Spradow et al. [MSSD*08]: Surface glyphs	2.5D	0																		
Peng et al. [PWR04]: Clutter reduction using dimension reordering	2D	1																		
Pickett & Grinstein [PG88]: Stick figures	2D	3																		
Piringer et al. [PKH04]: Depth perception in 3D scatterplots	3D																			
Rogowitz et al. [RTB96]: How not to lie with visualization	3D																			
Tominski et al. [TSWS05]: Helix glyphs on geographic maps	2.5D	-2																		
Treinisch [Tre99]: Task-specific visualization design	2.5D	-2																		
Ward & Guo [WG11]: Shape space projections	2D	3																		

Figure 26: A 1-N categorization of glyph-based approaches created by Borgo et al. In Design Guideline 2, -3 represents a small amount of complex glyphs with +3 displaying a large number of simple glyphs. Courtesy of Borgo et al. [BKC*13]

		Many-to-One Mapping						One-to-One Mapping		
		Orientation		Color Saturation		Position/Length		Unique		
		Linear	Circular	Linear	Circular	Linear	Circular	Faces	Cars	3D Glyphs
Many-to-One Mapping	Orientation	Linear Stick figure	[72]			[53]	[53]			
		Circular Pie chart								
	Color Saturation	Linear Stripe Calendar		[54], [68], [69], [70]	[24]	[24], [54]	[24]			[71]
		Circular Clock				[24]	[24], [65]	[65]		
One-to-One Mapping	Position/Length	Linear Profile Profile Line glyph			[54], [55]	[24], [47], [48], [49], [50], [51], [52], [53]	[47], [48], [50], [52]		[56]	
		Circular Star glyph Polygon Whisker				[47], [57], [58], [59], [60], [61], [62], [63], [64]*	[47], [48], [50], [52], [61], [65], [66]		[67]	
	Unique	<div style="border: 1px solid black; padding: 5px; display: inline-block;"> Not included in the matrix Planning Line [35] Theme [41] MILSTD2523 [46] Weather vane [36] Shape [38] Rose [40] Arrow [42] Motif [43] Flower [45] </div>								
	<div style="border: 1px solid black; padding: 5px; display: inline-block;"> Faces Chernoff Flury-Riedwyl Kabulov </div>						[17], [50], [61], [73], [74], [75], [76], [77], [78], [79], [80], [81], [82], [83], [21]* [84]* [85]* [86]*	[34]	[87]	
	<div style="border: 1px solid black; padding: 5px; display: inline-block;"> Cars Car glyph </div>									
	<div style="border: 1px solid black; padding: 5px; display: inline-block;"> 3D Glyphs Tender Surface Superquadric Superquadric </div>								[88], [89], [90], [91], [92]	

Figure 27: A 2-Dimensional table showing the classification of the literature in the glyph-based user-study survey. Courtesy of Fuchs et al. [FIBK16]


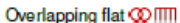
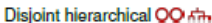



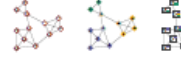

















		Group Structure Taxonomy			
		Disjoint flat 	Overlapping flat 	Disjoint hierarchical 	Overl. hier. 
Visual node attributes	Color  Section 5.1 Figure 1(a)	 1 st [DS13, SKL*14, vHW08] 2 nd [BPF14, CDA*14, EHKP14, ET07, GHK10, HGK10, HKV14, SMM13, vdEW14, VBAW14]	1 st [-] 2 nd [AHRRC11, BT08, BBT08, DvKSW12, DEKB*14, IMMS09, LQB12, LWC*14, NIST12, HRD10, TLTC08, XDC*13]	1 st [-] 2 nd [BD05, BD07, KG08, SBG00]	1 st [-] 2 nd [VRW13]
	Glyph Section 5.1 Figure 3		 1 st [IMMS09, LWC*14, NIST12, TLTC08] 2 nd [ST08, XDC*13]		1 st [-] 2 nd [VRW13]
Group Visualization Taxonomy	Separate Section 5.2.1 Figures 4(a)-(b)	 1 st [SMM13, vdEW14]	 1 st [SJUS08]	 1 st [AKY05, AvHK08, CC07]	
	Juxtaposed Attached Section 5.2.2 Figures 4(c)-(e)			 1 st [AZ13, BPD11, BBV*12, BD13, BD08, BFB10, BVB*11, BHW11, BSW13, GF03, GZ11, GBD09, Hol08, HOw07, NSC05, PwW06, vH03, vHSD09, VBSW13] 2 nd [PMF12]	
Superimposed	Line overlay Section 5.3.1 Figure 5(a)		 1 st [AHRRC11, XDC*13]		
	Contour overlay Section 5.3.2 Figure 5(b)	 1 st [BPF14, EHKP14, ET07, GHK10, HGK10, HKV14] 2 nd [VBAW14]	 1 st [BT08, BBT08, BT09b, DvKSW12, DEKB*14, LQB12, HRD10, ST08]	 1 st [BD05, BD07, DGC*05, Hol08, KG08, SBG00] 2 nd [NSC05]	
	Partitioning Section 5.3.3 Figure 6	 1 st [SKB*14, SA06, ZCCB13]	 1 st [LSKS10]	 1 st [AFH*10, DWS*14, FWD*03, Hol08]	
Embedded	Node-link Section 5.4.1 Figure 7(a)	 1 st [CDA*14, SMER08, VBAW14]	 1 st [RHR*10, SZPM10]	 1 st [ASH14, AMA07a, AMA08, AMA09, AMA11, DM12, DM14a, HN07b, HN07a, RPD09, vHW04]	1 st [VRW13]
	Hybrid Section 5.4.2 Figure 7(b)	 1 st [HFM07]	 1 st [HBF08, MZ11]	 1 st [PMF12]	

Figure 28: Taxonomy table created by Vehlow et al. correlating group visualizations and group structures. Courtesy of Vehlow et al. [VBW15]

		Entities		
		Node/Link	Group	Network
Temporal Features	Individual Events	Single Occurrences	Observe an entity appears or disappears independently (s1)	
			Examine structural (degree, density, centrality) or domain properties at a time point (s2)	
			Examine the number of node/link or group events (e.g. post, reply, report, invitation, page view) at a time point (s3)	
	Birth/Death	Find when a node/link or a group event appears/disappears (bd1)		
		Find an emergence of a new network structure such as an interaction pattern, or sub-groups (bd2)		
	Replacement	Find if and when a edge direction (e.g. replies) changes [rp1]		
	Shape of Changes	Growth & Contraction	Observe the growth/contraction of entities and their properties [gc1]	
			Observe growth/contraction of structure properties [gc2]	
		Convergence & Divergence	Observe if a structure property converges at a specific time point [cd1]	
Find if a new structure emerges from the convergence [cd2]				
Stability		Find if events or structural properties are stable [st1]		
		Find when the stabilization happen [st2]		
Repetition	Find if events or structural properties change pattern repeats [re1]			
	Identify the pattern of the repetition [re2]			
Peak/Valley	Find if/when events or structural properties show a peak or a valley (pv1)			
	Identify the shape of the peaks/valleys (pv2)			
	Identify when the peaks/valleys appear (pv3)			
Rate of Changes	Fast & Slow	Identify how much changes occur at a given time [fs1]		
	Accelerate & Decelerate	Identify whether a change of events or structural properties is getting faster or slower [ad1]		

Figure 29: Design Space of network temporal evolution tasks courtesy of Ahn et al. [APS14]

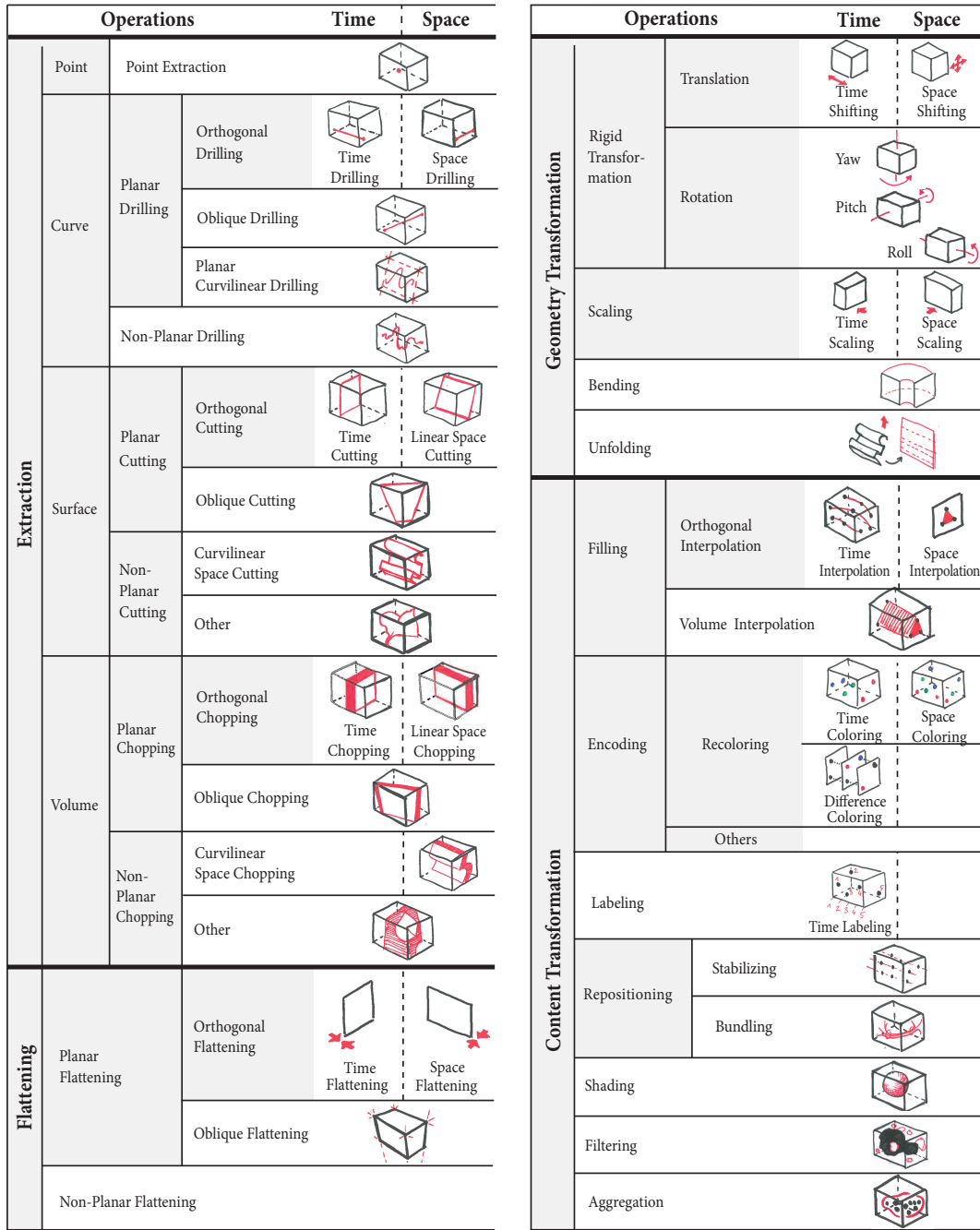


Figure 30: Taxonomy of Space-Time cube operations created by Bach et al. [BDA*14]. Each operation gives a representation of how the operation may work. Bold font indicates complete operations. Gray shading indicates non-leaf nodes. Image courtesy of Bach et al. [BDA*14]

	Level	Focus	Section	Visualization Technique	Representation	References	Year	
Time T Visualization	Line	Line properties	2	Seesoft	2D colored pixel	[1], [2]	1992	
				Sv3d	3D colored cuboid	[3], [4]	2003	
	Class	Functioning, Metrics	3	Class Blueprint	2D layers and graph	[5], [6], [7]	1999	
	Architecture	Organization		4.1	Treemap	2D/3D colored nested boxes	[8], [9], [10]	1991
					Circular Treemap	2D/3D colored nested circles	[8], [11]	1991
					City/Cities	3D city metaphor	[12], [13], [14], [15]	1993
					Sunburst	2D colored radial display	[16], [17], [18]	1998
					Solar System	3D solar system metaphor	[19], [20]	2003
					Voronoi Treemap	2D colored irregular shapes	[21]	2005
		Relationships		4.2	Dependency Structure Matrix	2D table	[22], [23], [24]	1981
					UML	2D diagrams	[25]	1996
					Geon	3D geon diagrams	[26], [27], [28]	1998
					Solar System	3D solar system metaphor	[19], [20]	2003
					Landscape	3D landscape metaphor	[29], [30]	2004
					Hierarchical Edge Bundles	2D graph with bundled edges	[31]	2006
					City/Cities	3D city metaphor with edges	[32], [33], [34]	2007
					3D Clustered Graph	3D clustered graph	[35]	2007
		Metrics		4.3	Polymetric views	2D graph	[5], [36], [37]	1999
					Solar System	3D Solar system metaphor with edges	[19], [20]	2003
					UML MetricView	2D UML diagrams with charts on top	[38]	2005
Treemap metrics					2D nested boxes with color and texture	[39]	2005	
City	3D City metaphor				[40], [41], [42], [43]	2005		
UML Area Of Interest	2D diagrams with area of interest	[44], [45]	2006					
Visualizing Evolution	Line	Changes	5.1	Code Flow	cable-and-plug wiring metaphor	[46], [47]	2007	
	Class		5.2	TimeLine	3D building metaphor	[48]	2008	
			5.3.1	Hierarchical Edge Bundles	2D graph with bundled edges	[49]	2008	
	Archi.	Metrics Evolution		5.3.2	Evolution Matrix	2D matrix	[50], [51]	2001
				5.3.2	RelVis	2D Kiviat diagrams and graph	[52]	2005
				5.3.2	City/Cities	3D city metaphor with animation	[48], [53]	2008

Figure 31: Caserta and Zendra present a table that classifies methods that visualise the static aspects of software, and the associated literature. Image courtesy of Caserta and Zendra [CZ11].

Techniques	Data						Tasks							
	Temporal	Geospatial	Flow	Volume	Multivariate	Graph	Document	Select	Explore	Reconfigure	Encode	Abstract & Elaborate	Filter	Connect
[SB92, SB94] Fisheye Views		•				•						•		
[RM93] Document Lens							•					•		
[CMS94] MagicSphere				•				•			•	•	•	
[RC94] Table Lens					•							•		
[VCWP96] 3D Magic Lenses				•								•	•	
* [FG98] Lenses for Flow Visualization			•								•	•		
[FP99] Excentric Labeling					•				•			•		•
[SHER99] Interactive 3D Lens			•	•	•		•				•	•	•	
[LHJ01] Volume Magnification Lens				•								•		
[SFR01] Time Lens	•								•					
[BCPS02] Fuzzy Lens, Base-Pair Lens, Ring Lens						•				•	•	•		
[MTHG03] 3D Flow Lens			•								•	•		
* [WCG03] EdgeLens						•				•				
* [vHvW04] Graph Abstraction Lens						•						•		
[RHS05] Magic Lenses in Geo-Environments		•									•		•	
* [EBD05, ED06b, ED06a] Sampling Lens					•								•	
[RLE05] Temporal Magic Lens	•							•						
* [WZMK05] The Magic Volume Lens				•								•		
[TAvHS06] Local Edge Lens						•							•	•
[KSW06] ClearView				•								•	•	
[TGBD08] 3D Generalization Lenses		•										•		
* [TAS09] Layout Lens						•				•			•	•
[BRL09] Enhanced Excentric Labeling					•				•			•		•
* [MCH*09] Bring &Go						•				•				•
[ACP10] High-Precision Magnification Lenses	•	•	•	•	•	•	•	•				•		
[JDK10] Network Lens						•					•			
* [KCJ*10] Detail Lenses for Routes		•										•		•
[Kin10] SignalLens	•											•		
* [SNDCl0] PushLens						•				•				
* [STSD10] Tangible Views		•			•	•		•	•	•	•	•	•	•
* [EDF11] Color Lens	•	•	•	•	•	•					•			
[GNBP11] FlowLens			•						•		•	•	•	
[HLTE11] SemLens					•					•		•		•
[HTE11] MoleView		•			•	•				•				
[LWG11] Facet Lens							•				•		•	•
[PBKE11] EdgeAnalyser					•	•					•	•		
[ZCB11] MagicAnalytics Lens	•													•
* [ZCPB11] ChronoLenses	•												•	•
[TSAA12] Time Lens		•							•		•			•
[PPCP12] JellyLens		•										•		
* [KTW*13] TrajectoryLenses		•											•	•
[PPCP13] Gimlenses				•								•	•	
[UvK13] Magic Lenses for Hypergraphs						•							•	
* [CC13] Lens for Querying Documents							•						•	•
[AACP14] RouteLens		•										•		
[BHR14] PhysicLenses		•										•		
* [MW14] Bubble Lens					•			•				•		
[DMC14] VectorLens	•				•			•					•	•
[KRD14] Multi-touch graph Lenses						•				•			•	•
[DSA15] 3DArcLens		•				•				•			•	•

Figure 32: Lens Techniques categorised according to data types and task. Courtesy of Tominski et al. [TGK*16]