

Storytelling and Visualization: A Survey

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Abstract

Throughout history, storytelling has been an effective way of conveying information and knowledge. In the field of visualization, storytelling is rapidly gaining momentum and evolving cutting-edge techniques that enhance understanding. Many communities have commented on the importance of storytelling in data visualization. Storytellers tend to be integrating complex visualizations into their narratives in growing numbers. In this paper, we present a survey of storytelling literature in visualization and present an overview of the common and important elements in storytelling visualization. We also describe the challenges in this field as well as present a novel classification of the literature on storytelling in visualization. Our classification scheme highlights the open and unsolved problems in this field as well as the more mature storytelling sub-fields. The benefits are a concise overview and starting point into this rapidly evolving research trend.

1. Introduction And Motivation

"We believe in the power of science, exploration, and storytelling to change the world" - Susan Goldberg, Editor in Chief of National Geographic Magazine, from "The Risks of Storytelling", October 2015 [Gol15].

"In a world increasingly saturated with data and information, visualizations are a potent way to break through the clutter, tell your story, and persuade people to action" [Sin14]. -Adam Singer, Clickz.com

Throughout history, storytelling has been an effective way of conveying information and knowledge [LNP*13]. In the field of visualization, storytelling is rapidly developing cutting-edge techniques that enhance understanding. Many communities have commented on the importance of storytelling in data visualization [SH10]. Storytellers tend to be integrating complex visualizations into their narratives in growing numbers. In this paper, we present a survey of storytelling papers in visualization and present an overview of the common and important elements in storytelling visualization. We also describe the challenges in this field as well as present a novel classification of the literature on storytelling in visualization. Our classification highlights both mature and unsolved problems in this area. The benefit is a concise overview and valuable starting point into this rapidly growing and evolving research trend.

1.1. Definition And Storytelling Elements

Story, from the dictionary, means, "a narration of the events in the life of a person or the existence of a thing, or such events as

a subject for narration" [Refa] or "a series of events that are or might be narrated" [Dica]. Storytelling is a popular concept that is used in many fields, such as media [SH10], education [Zip13] and semiotics [Mar97]. Storytelling is a technique to present dynamic relationships between story nodes by interaction. According to Zipes [Zip13], storytelling can involve animation and self-discovery, incorporating models, ethical principles, canons of literature, and social standards. In education, a storyteller can improve and strengthen the literacy of students. Also, the storyteller can engage audiences so they feel a desire to read, write, act, and draw. They learn to express themselves critically and imaginatively with techniques they may learn from the storyteller or teacher.

Martin [Mar97] states that storytelling can consist of three abstract levels: the narrative level which is more general and more abstract. This is the level of story-structure that underlines all discourse. He describes a story or narrative as presenting events and features movement from one state to another by an act of transformation. Constructing narratives is a major theme in our survey.

The figurative level is a surface level of meaning and the level that is the most concrete. It relates to individual characters and to specific events unfolding in a given space and time. In other words, figurative elements are in a context that corresponds to the physical world and that can be apprehended by the five senses: vision, hearing, smell, taste and touch.

The deep level which is also known as the thematic level is the level of the abstract or conceptual. It relates to the inner cognitive world as opposed to the outer physical world of the figurative level. It is the level at which interpretations are articulated the fundamen-

tal values in the text are discussed. Interpretation is one of the big challenges highlighted in this survey.

1.2. Challenges in Storytelling And Visualization

Although storytelling has been developing in other fields for years, in visualization, storytelling is a relatively new subject. It faces many challenges.

- Authorship: Authorship refers to writing or creating a book, article, or document or the creator of a work of art in the Oxford English dictionary [Dicb], especially with reference to an author, creator, producer in dictionary [Refb]. For our purposes, we will adopt a definition of author described by Rodgers [Rod11], "An author is best described as an individual solely responsible for the creation of a unique body of work."
- Data enhancement and derivation: Data enhancement and derivation refers to the process of creating a data set from one or more sources such as books or films through an enhancement process. How can we perform data acquisition from data sources? And how can we store and represent a story at the data level?
- Narratives: Narrative structures include events and visualization of characters. Narrative visuals contain the transition between events. This involves, "Using a tool to visually analyze data, and to generate visualizations via vector graphics or images for presentation," and deciding "how to thread the representations into a compelling yet understandable sequence." [HdR*13].
- Transitions: Visualization of transitions are the focus of visualization and include both dynamic and static which are techniques of presenting visualization. Are static transitions or dynamic transitions more effective for storytelling in visualization?
- Memorability: Memorability is an important goal of storytelling. A good visualization technique will draw the viewer's attention and increase a story's memorability [BMG*10]. Can visualization increase the memorability of data information or knowledge?
- Interpretation: Data interpretation refers to the process of critiquing and determining the significance of important data and information, such as survey results, experimental findings, observations or narrative reports. Does storytelling and visualization aid with data interpretation?

These challenges form the basis of our literature classification on storytelling in visualization.

1.3. Classification of Literature

The Storytelling literature can be divided into various sub-categories: scientific visualization, information visualization, geo-spatial visualization, narratives, static & dynamic transitions, memorability, data enhancement, interpretation & exploration, analysis versus presentation. Here is a description of these categories that we haven't described as part of the challenges:

1. Scientific Visualization, Information Visualization versus Geo-spatial Visualization: Scientific visualization is seen by a user primarily relates to a physical phenomena occurring in nature,

such as flow, volume and surface visualization. Information visualization is more concerned with abstract data with no inherent spatial geometry. While geo-spatial visualization describes objects or phenomena with a specific location on a map [Spe07].

2. Static versus Dynamic: Static visualization requires no action from the user and the encoded data is immediately available. While dynamic visualization refers to those representations that go beyond traditional static such as including animation, interaction and real-time display [Spe07].
3. Visualization Goals: This includes exploration, analysis and presentation. Exploration refers to navigation and browsing undertaken solely with the aim of forming a mental model of part or all of an information space [Spe07]. Analysis refers to a systematic examination and evaluation of data or information by breaking it into its component parts to uncover their interrelationships. And presentation refers to the selection of encoded data and its layout on a display [Spe07].

It is important to note that some papers cover multiple challenges and that sometimes challenges overlap. Thus the classification is not always black and white (or binary).

1.4. Survey Scope

The storytelling visualization papers summarized in this survey include the subjects of scientific visualization, information visualization geo-spatial visualization, and visual analytics. In order to manage the scope of this survey, storytelling papers from other fields are not included, such as:

- Virtual reality and augmented reality: For example, Santiago et al. [SSF14] present "Mogre-storytelling" as a solution of interactive storytelling. This tool provides different functionalities for creating and customization of scenarios in 3D, enables the addition of 3D models from the internet and enables the creation of virtual story using multimedia and storytelling elements.
- Education: For example, Cropper et al. [CLM15] address the extent of how scientific storytelling benefits our communication skills in the sciences, and the connections they establish with the information itself and others in their circle of influence.
- Gaming: Alavesa et al. [AZ13] describe the development of a small scale pervasive game which can take storytelling from camp-fire sites to modern urban environments.
- Multi-media: For example, Chu et al. describe a system to transform any temporal image sequence to a comics-based storytelling visualization [CYW15].
- Language processing: Theune et al. [TMHO06] develop a story generation system. It can create story plots automatically based on the actions of intelligent agents living in a virtual story world. The derived plots are converted to natural language, and presented to the user by an embodied agent that makes use of text-to-speech.

There are other fields that study storytelling as well. In next sections we describe the literature on storytelling in visualization. Our classification is presented in Table 1. Figure 1 shows the visualizations used in the storytelling for data visualization literature.

Table 1: Our classification of the storytelling literature. Both mature areas and unsolved problems are apparent.

	Scientific Visualization	Information Visualization	Geo-spatial Visualization
Authorship	Wohlfart, 2006 [Woh06] Wohlfart et al. 2007 [WH07] Lu and Shen, 2008 [LS08]	Gershon et al. 2001 [GP01] Cruz et al. 2011 [CM11] Kuhn et al. 2012 [KS12] Lee et al. 2013 [LKS13]	Eccles et al. 2007 [EKHW08] Lidal et al. 2012 [LHV12] Lidal et al. 2013 [LNP*13] Lundblad et al. 2013 [LJ13]
Narrative		Viegas et al. 2004 [VBN*04] Akashi et al. 2007 [ASKH07] Fisher et al. 2008 [FHRH08] Segel and Heer, 2010 [SH10] Hullman et al. 2011 [HD11] Hullman et al. 2013 [HDA13] Hullman et al. 2013 [HDR*13] Figueiras, 2014 [Fig14b] Figueiras, 2014 [Fig14a]	
Static Transitions		Robertson, 2008 [RFF*08] Chen et al. 2012 [CLH12] Tanhashi et al. 2012 [TM12] Shixia et al. 2013 [LWW*13]	Ferreira et al. 2013 [FPV*13]
Animated Transitions	Akiba et al. 2010 [AWM10]	Bederson and Boltman, 1999 [BB99] Heer et al. 2007 [HR07]	
Memorability		Bateman et al. 2010 [BMG*10]	Saket et al. 2015 [SSKB15]
Interpretation	Ma et al. 2012 [MLF*12]	Kosara and Mackinlay, 2013 [KM13]	

2. Authorship for storytelling and visualization

Authorship refers to writing or creating a book, article, or document or the creator of a work of art according to The Oxford English dictionary [Dicb] and origin, especially with reference to an author, creator or producer [Refb]. For our purposes, we will adopt a definition of author described by Rodgers [Rod11], "An author is best described as an individual solely responsible for the creation of a unique body of work." Hullman [HDR*13] et al. state, "Story creation involves sequential processes of context definition, information selection, modality selection, and choosing an order to effectively convey the intended narrative".

An author is best described as an individual solely responsible for the creation of a unique body of work [Rod11]. Presenting the findings of a qualitative study of undergraduate writers at The City University of New York, Hullman explores student perspectives on models of authorship, the relationships between these models and student experiences of authorship in different writing situations, and proposes the importance of distinguishing between the multiple models and definitions of authorship and the rhetorical contexts associated with each [Rod11]. Rodger develops a qualitative study of 800 students on the definition of authorship and their rhetorical contexts over a one-hour interview. Students defined authors as "[people] who see writing as being beyond a hobby," and as a term that should be applied only to those individuals for whom writing is "something he or she has to do," "a career," or "an act that will lead to something being published."

What emerged as a key difference between groups was the extent to which each clearly separated the definitions of the terms author and writer. Those students who were comfortable consistently identifying themselves as writers defined the term writer as separate and distinct from the term author, whereas those students who did not clearly separate the two terms, but instead used elements of one definition to inform their definition of the other, were not comfortable consistently identifying themselves as writers.

All papers in this section focus on creating new visualizations for storytelling. Wohlfat and Michael [Woh06] create new volume visualization story for medical applications. Gershon [GP01] and Cruz [CM11] present the general storytelling for information visualization. Kuhn [KS12], Lee [LKS13] and Plowman [PLL*99] all develop unique creator tools for storytelling visualization.

It is important to note that our survey is not simply a list of papers. Individual papers are summarized according to a special methodology [Lar11]. This process connects the papers together such that the relationship to previously published literature is made clear.

2.1. Authorship for Storytelling and Scientific Visualization

Storytelling is a relatively new form of interactive volume visualization presentation [Woh06]. Wohlfat explores the usefulness of storytelling in the context of volume visualization. He presents a story telling model and divides the concept of volumetric storytelling into story authoring and storytelling constituents. Then

	scientific visualization	information visualization	geo-spatial visualization	volume rendering	no-animation	animation	story board	interaction	bar chart	bubble chart	Time line	word clouds	tally chart	line chart	scatter plot	pie chart	map	event graph	table	node-link diagram	color-coding	histogram	photography	Density Heat maps
Wohlfat[Woh06]:Story Telling Aspects in Medical Applications																								
Wohlfat[WH07]:Story Telling for Presentation in Volume Visualization																								
Lu and Shen[LS08]:Interactive Storyboard for Overall Time-Varying Data Visualization																								
Gershon and Page[GP01]: What Storytelling Can Do For Information Visualization																								
Cruz et al[CM11]:Generative Storytelling for Information Visualization																								
Kuhn et al[KS12]:CodeTimeline: Storytelling with Versioning Data																								
Lee et al[LS13]: SketchStory: Telling More Engaging Stories with Data through Freeform Sketching																								
Eccles et al[EKHW08]:Stories in GeoTime																								
Lidai et al[LHV12,LNP13]:GeologicalStorytelling:GraphicallyExploringandCommunicating Geological Sketches																								
Lundblad et al[J13]:Geovisual Analytics and Storytelling Using HTML5																								
Viegas et al[VBN*04]:Digital artifacts for remembering and storytelling: Posthistory and social network fragments																								
Akashi et al[ASKHW07]:Narrative based Topic Visualization for Chronological Data																								
Fisher et al[FHRH08]:Narratives: A Visualization to Track Narrative Events as they Develop																								
Segal and Heer[SH10]:Narrative Visualization: Telling Stories with Data																								
Humallman et al[H11]:Visualization Rhetoric: Framing Effects in Narrative Visualization																								
Humallman et al[HDA13]:Contextifier:AutomaticGenerationofAnnotatedStock visualizations																								
Humallman et al[HDR13]:A Deeper Understanding of Sequence in Narrative Visualization																								
Figueiras[Fig14b]:Narrative Visualization:ACaseStudyofHowtoIncorporate Narrative Elements in Existing Visualizations																								
Figueiras[Fig14a]:How to Tell Stories Using Visualization																								
Rebortson[REF08]:Effectiveness of Animation in Trend Visualization																								
Chen et al[CLH12]:Visual storylines: Semantic visualization of movie sequence																								
Tanhashi et al[TM12]:Design Considerations for Optimizing Storyline Visualizations																								
Shixia et al[LWW*13]:StoryFlow: Tracking the Evolution of Stories																								
Ferreira et al[FPV*13]:Visual Exploration of Big Spatio-Temporal Urban Data: A Study of New York City Taxi Trips																								
Akiba et al[AWM10]:AniVis:A Template-Based Animation Tool for Volume Visualization																								
Bederson and Boltman[BB99]:Does Animation Help Users Build Mental Maps of Spatial Information?																								
Heer et al[HR07]:Animated Transitions in Statistical Data Graphics																								
Bateman et al[BMG*10]:Useful Junk? The Effects of Visual Embellishment on Comprehension and Memorability of Charts																								
Saket et al[SSKB15]:Map-based Visualizations Increase Recall Accuracy of Data																								
Ma et al[MLF*15]:Scientific Storytelling using Visualization																								
Kosara and Mackinlay[KM13]:Storytelling: The Next Step For Visualization																								

Figure 1: A table summarizing the visualization techniques used in each storytelling paper. The papers are ordered corresponding to the order in which they appear in the survey.

he presents a volumetric storytelling prototype application, which is based on the RTVR (real time volume redering) Java library [MH01] for interactive volume rendering. See Figure 2 [Woh06].

1. Storytelling Model: The storytelling model contains a range of hierarchy levels, in top-down order, which are: story node, story transitions, story action group, story action atoms. See Figure 2. The story nodes form the corner marks of the story and store the state of the whole scene. Story nodes are connected by story transitions, each consisting of one or multiple story action groups. Each story action group stores the scene changes relative to its preceding action group (or story node) [Woh06, WH07].
2. Story Authoring Process: The story authoring process contains two steps: a story recording process and a story editing process. The outcome of this recording process is a raw prototype of a story told through volume visualization. The story editing step this raw story is refined until the final story outline is reached [Woh06, WH07].
3. Story Telling Process: This process is to present volume visualization following story telling model. And the key feature is interaction, including viewing interaction, representing interaction and data interaction [Woh06, WH07].
4. Storytelling prototype based on RTVR: Figure 2 shows an image sequence taken from a sample linear volumetric story vi-

sualized with their prototype. The distinct story nodes denoted through (N) refer to the key events in the story, which provide an overview first, then details on specific features in the dataset, and at the end a conclusion made by the story author. The necessary story transitions are represented as orange arrows from one story node to the next and are animated in the prototype application. The story consumer may take over some story parameters (e.g.camera angle) already during playback or at the end of the story to further investigate the dataset [Woh06, WH07].

5. CT-dataset visualization story: This story guides the observers through the visualization, puts the contained visual representations into context to each other and finally introduces them to important features in the data [WH07]. See Figure 3.

Wohlfat is based on previous work of volume visualization [GGSC98, HM90, Vio05] and interactive visualization [Don93] and combine these concepts together to develop a storytelling model for volume visualization.

Lu and Shen propose an approach to reduce the number of time steps that users need to visualize to understand the essential data features by selecting representative datasets. They design a flexible framework for quantifying data differences using multiple dissimilarity matrices [LS08]. Data relationship measurement: Lu and Shen [LS08] present a new visualization approach that filters data analysis results, which is achieved by measuring the degree of data

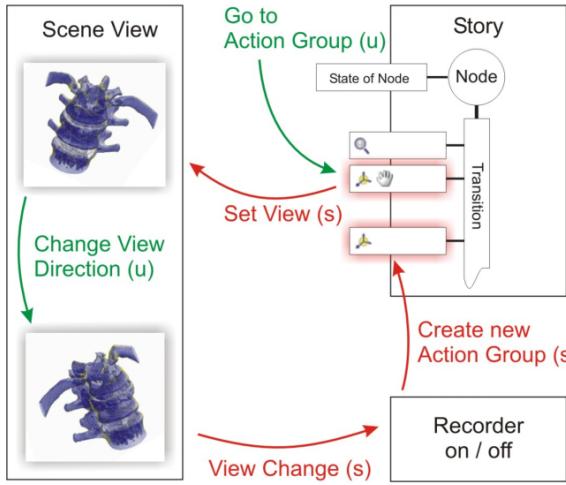


Figure 2: The proposed method to author a story is to record the user's natural interaction with the visualisation software. This image shows the process of the story creation. Green annotations represent user interaction and red annotations refer to internal system processes. As soon as the software starts recording, a new story is created and all interactions are logged [Woh06].

similarity/difference and selecting important datasets that contain essential data features [LS08].

Representative datasets analysis: They interactively select representative datasets that include a significant portion of features of scientific data, whose data distribution requires more analysis than time sequence, reduces the amount of data to necessarily visualize and still keeps the essential data information. This can be used to improve the efficiency of time-varying data visualization [LS08].

Interactive storyboard: An interactive storyboard is used to visualize and explore the overall content of time-varying datasets through composing an appropriate amount of information that can be efficiently understood by users [LS08]. Lu and Shen is based on the previous work of time-vary visualization [HJ11] and design a general method for comparing data dissimilarities. They do not require a dense sampling frequency to capture the object evolution and is not limited to specific feature models, such as geometry or interval volumes, and their attribute designs [LS08].

2.2. Authorship for Storytelling and Information Visualization

Gershon and Page state that storytelling enables visualization to reveal information as effectively and intuitively as if the viewer were watching a movie [GP01]. Gershon et al. introduce concept of storytelling and presents advantages of storytelling [GP01].

1. John Thomas of IBM research: By using storytelling, the repeated information and work detailed using technology in this research can be reduced into short and memorable information [Tho].
2. Enemy positions surround a school: This example presents a situation in which a number of enemy positions surround a school

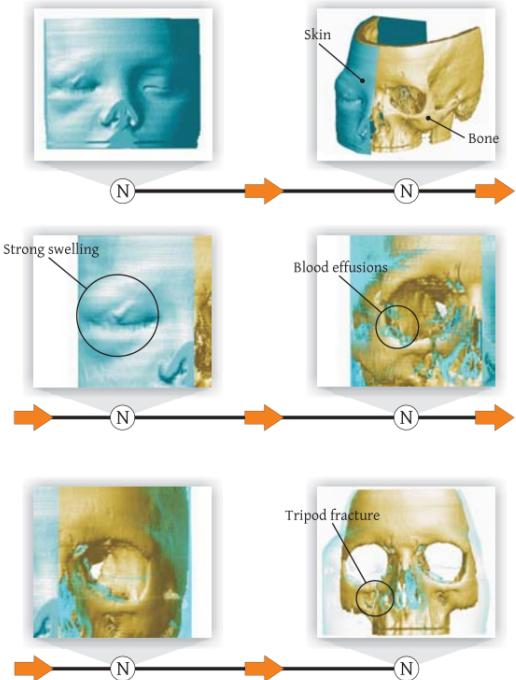


Figure 3: Here the top two images show an overview of the CT scan data. A partial clipping reveals both the skin layer and bone layer, but shows the full set of data. The middle shows a zoomed view that isolates eye swelling in the image (left), and a filtered view that exposes some blood effusions in the swollen region. The bottom offers a comparison of the non-injured eye with the injured one and shows the cause of the swelling which is attributed to a tripod fracture just below the eye. This design offers the user a macro overview as to lay the foundations of a story background then narrows the scope to view the focal point of the image [WH07].

with children trapped inside as de facto hostages as the crossfire fills the space overhead and both sides move toward confrontation [Den01].

Gershon and Page is based on previous work of Denning [Den01] and explain the usage of storytelling in information visualization.

Storytelling, in the context of this article, deals with the core of information visualization by extracting relevant knowledge and enhancing its cognition [CM11]. Cruz et al. present generative storytelling as a conceptual framework for information storytelling. They create stories from data fabulas using computer graphics as a narrative medium. Data fabulas are set of time-ordered events caused or experienced by actors [CM11].

1. Generative-storytelling character: A story is formed by characters. It involves the representation of the fabula's actors and the definition of a temporal structure.
2. The conceptual framework: The engine transforming a fabular into story consists two models. The event model creates a story timeline and an action model create a set of actors behavior.
3. Applying the framework: Visualizing an empire's decline visu-

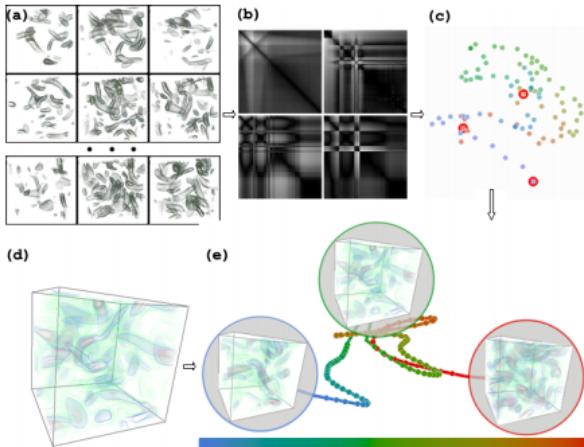


Figure 4: This figure shows the system architecture. It integrates the information of data analysis and a single 3D data visualization method for users to explore and visualize overall time-varying data contents [LS08].

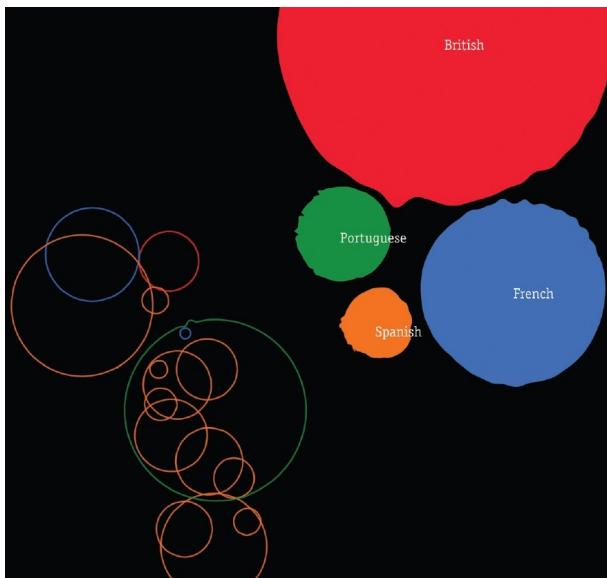


Figure 5: This figure shows the British hegemony and the newly independent South America in 1891. Each empire and independent territory is a circle whose area is proportional to that entity's land area. Former colonies are unfilled circles with rims in the corresponding empires color [CM11].

alizing western empire's decline in the 19th and 20th centuries. See Figure 5.

Cruz et al. is based on previous work of narrative theory [Nar85] and presents generative storytelling as a conceptual framework for information storytelling [CM11].

The CodeTimeline visualization by Kuhn and Stocker [KS12] enables developers who are new to a team to understand the history of the system they are working on. Designed to show a development team's tribal memory, the software offers a partial replacement for exhaustive documentation. See Figure 6. Collaboration View: Collaboration View presents visualizes code ownership and historical patterns in collaboration. Sourcecloud Flow View: Sourcecloud Flow View presents a word cloud of added and removed vocabulary between software releases. Lifetime events can be added by users as a frame of reference in each of the visualizations. This method of linking also enables new users to learn more about the history of the developed software. These events can include anything from email threads to pictures of the team during work.

Prior to Kuhn and Stocker, Ogawa [OM10, OM09] presents "software evolution storylines" and "Code Swarm", which focus on the interactions between developers on projects, but do not focus on telling a story about the software history. Codebook, a concept presented by Begel [BKZ10], outlines a social network that connects software engineers with their shared code base. It encourages interaction with their code and others, enabling a broader understanding of the project they share with other developers.

Lee et al. present SketchStory, a data-enabled digital white board to support real-time storytelling. It enables the presenter to stay focused on a story and interact with charts created during presentation. See Figure 7 [LKS13].

1. Story preparation: The data-based story is recorded in sketch-story as a sequence of charts in XML files. The charts are linked with specific sketch gestures.
2. Sketchstory Interaction: The presenter draws an example icon and then draws a sketch gesture for chart invocation. Sketch-story recognizes the gesture and creates the corresponding chart.

Evaluation: Participating presenters found sketchstory to be good way of storytelling with data, also easy learn and use.

Lee et al. is based on previous work for storytelling of information visualization [GP01, SH10] and sketch-based interaction [LCPT12], and develops sketchstory system to enhance storytelling in a presentation.

2.3. Authorship for Storytelling and Geo-spatial Visualization

Plowman [PLL*99] reports that a narrative specifically refers to "the macro-structure of a document in contrast to the term story which refers to both structure and content." GeoTime events are recorded in X, Y, T, coordinate space. This is used in observation analysis and can make a major contribution to a story telling model. Eccles et al. presents the GeoTime stories prototype that combines a geo-spatial map with narrative events to produce a story framework. See Figure 8 [EKHW08].

1. Annotated Pattern Detection: This system provides functions for simple pattern detection in movement activity. These functions look at possible interactions between people within the narrative, the speed at which they travel, and the type of location that they visit.
2. Narrative text authoring: This enables the analyst to create and present stories found within the data. The story window displays this data as well as discovered patterns.

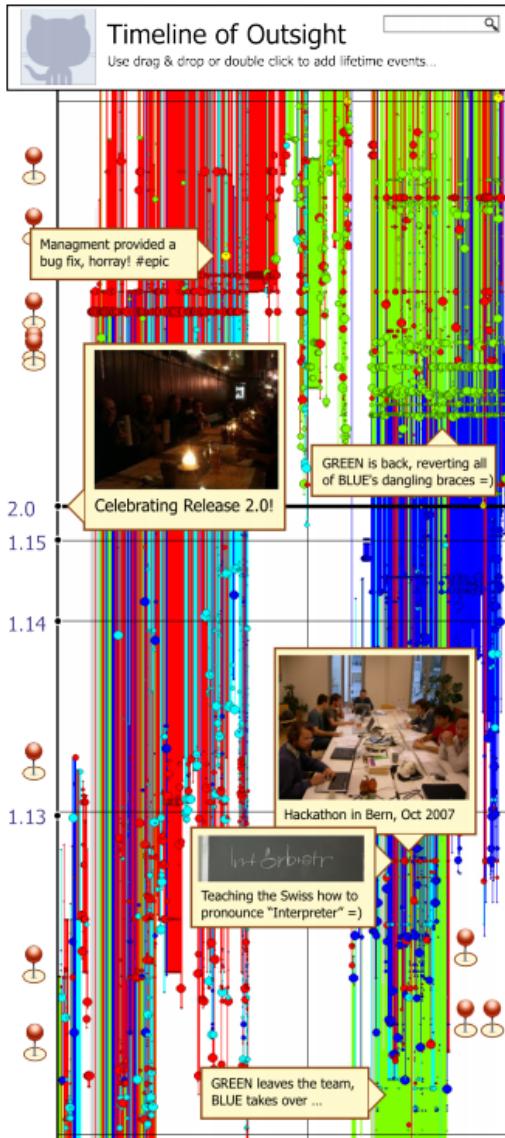


Figure 6: Kuhn and Stoker show the CodeTimeline collaboration view. Colors denote different user contributions and each line represents the life of files in the code. Sticky notes are added so the users can learn the history of the code beyond the file evolution [KS12].

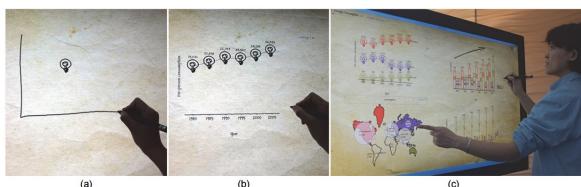


Figure 7: Lee et al. show an example of SketchStory in information visualization presentation [LKS13].

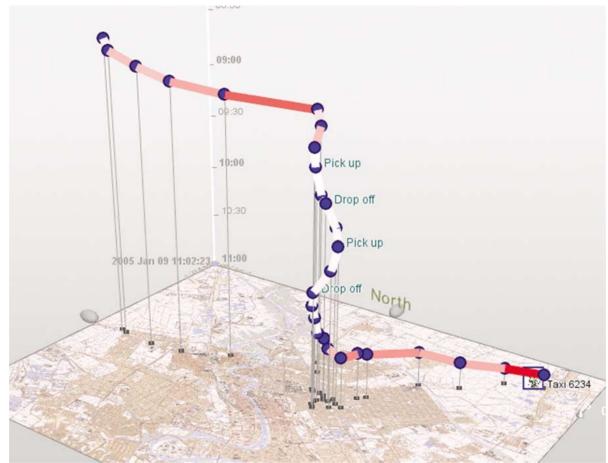


Figure 8: Eccles et al. show a GeoTime visualisation instance. The height of the from the base X,Y plain is linked to the temporal dimension. Here you can see a taxi driver's route over the course of a few hours. Each pick up and drop off is labelled and the route is mapped on the X,Y axis using the map [EHW08].

3. Story Threading: The system allows multiple stories to connect together if they follow a linear flow. Also simultaneous narratives can be shown in a single image for a direct comparison.

This system uses a similar approach to Sense.us [HWW07a]. Instead of using a blog-type discussion workflow for adding text, Geotime is designed for authoring a single story and annotations are integrated into the data itself.

Geological Storytelling is a novel graphical approach for capturing and visualizing the reasoning process that leads to a geological model [LHV12] [LNP*13]. Lidal et al. present a sketch-based interface for rapid modelling and exploration of various geological scenarios. The authors present a concept that handles sketching processed over time and a novel approach for externalizing the mental reasoning process. The process can be presented and evaluated [LHV12] [LNP*13]. The geological storytelling model contains three main parts. See Figure 9.

1. The Canvas: The canvas is a sketch-based interface where the geologist can draw the geological story on a 2D seismic slice backdrop, utilizing a pen and paper interaction style.
2. The StoryTree: This is a tree graph representation of all the geological stories, each with its own subtree of story nodes. Individual story nodes can be selected for editing in the canvas. One or more complete story trees can be selected for playback or comparative visualization in the inspect view.
3. The InspectView: This serves two purposes. First, it is a view where a story can be played and evaluated. In addition, multiple stories can be played synchronized for a side-by-side visual comparison.

Lidal et al. is based on a previous storytelling model [Woh06] scientific for visualization [MLF*12] and develops a storytelling model for geological visualization.

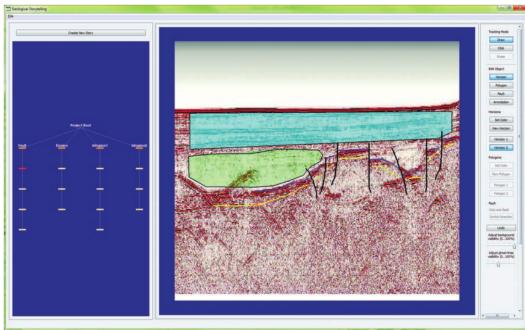


Figure 9: The sketch-based interface is split into two windows. The Story Tree (left) which shows a tree graph representation of all the geological stories, and the Canvas (right) which shows the sketching interface which utilises a pen and paper interaction to record geological sedimentary data. A geological story is built using horizontal lines to separate different geological layers, vertical lines to show fault systems and polygons for highlighting large sedimentary layers. The user can navigate through different geological stories with the story tree and then inspect the geological elements of that story [LHV12] [LNP*13].

Storytelling is one of the most impactful ways to teach, learn, and persuade [LJ13]. Lundblad and Jern present Geovisual analytics software with integrated storytelling. It can be applied to large spatial-temporal and multivariate data through dynamic visual user interfaces.

1. Multivariate Data Analysis: Using scatter matrix gives the analyst a good overview of all correlations between the selected indicators. The analyst can use the scatter plot matrix as an overview and then steer the scatter plot for interesting detailed combinations over time. See Figure 10 [LJ13].
2. European NUTS2 Regions in Dashboard: The distribution plot presents a special visualization technique that displays the variation within individual European countries [LJ13].
3. Motala River Story: The Motala River map is visualized for different stories divided into different layers, such as glyph layer, stream layer, polygon layer and background map layer. It shows the local and total water flow, and water path from source to ocean [LJ13].

Lundblad and Jern is based on the previous work of the storytelling concept and work of web-based geovisual tools, integrates storytelling with geovisual analytics software [LJ13].

3. Narrative Visualization and Storytelling

Narrative structures include events and visualization of characters. An example narrative can be a simple interface that presents trends in keywords over time [FHRH08]. Narrative visuals contain the transition between events. It involves "using a tool to visually analyze data, and to generate visualizations via vector graphics or images for presentation" and decide "how to thread the representations into a compelling yet understandable sequence" [HDR*13]. Plowman [PLL*99, EKHW08] reports that a narrative specifically



Figure 10: This figure shows Vislet aimed at a comparative visualization using linked Scatter Matrix and Scatter Plot to analyze national correlation between 6 indicators during 1960-2010 from the World Databank [LJ13].

refers to the macro-structure of a document in contrast to the term story which refers to both structure and content. This structuring of evidence, combined with the choice of appropriate rhetorical strategies, is referred to as "the art of storytelling" among literary scholars. Research in narrative visualization points to visualization features that afford storytelling including guided emphasis and structures for reader-driven storytelling. It also includes the principles that govern effective structuring of transitions between consecutive visualizations in narrative presentations, and how different tactics for sequencing visualizations are combined into global strategies in formats like slideshow presentations.

Narrative can be seen as a macro-structure which creates global coherence, contributes to local coherence and aids recall through its network of causal links and signposting [PLL*99]. The focus of Plowman's research is how students make sense of their learning with multimedia by constructing their own narratives in conjunction with the narrative guidance [PLL*99]. The design elements presented by the software constitute narrative guidance and can be a combination of features specific to interactive media, such as the need for clear navigational procedures, with features associated with traditional media, such as recognizable narrative and a clear relationship between tasks and the macro-narrative.

Plowman has developed three versions of Galapagos as a research tool, based on extended observation of students using commercially available CD-ROMs in schools [PLL*99].

The linear version is designed in such a way that students are led through the eight sections of the CD-ROM in sequence and it is closest to a traditional narrative as presented in educational television. The linear version presents a high degree of narrative guidance and little opportunity for learners to decide their own narrative path so they have relatively little control. The resource-based version offers no guidance through the CD-ROM sections and leaves students to define their own route. There is very little narrative guidance offered and learners have to construct a narrative by making decisions about sequence, so there is a high degree of user control and heavy use of the menu to decide the route. The guided discovery (GDL) version offers guidance in breaking down the task by providing paths through the material, questions to stimulate enquiries, and direction to specific resources. The GDL version was

designed to offer a balance between narrative guidance and support for narrative construction and this is reflected in a more even balance between user and system control. Learners are able to determine sequence and their course of action but are offered guidance in doing so.

Most sections (seven) are accessed by the GDL users, because the guide encourages them to be interactive in their approach and to use the material to support their response.

All papers in this section develop methods or structure on how to improve narrative storytelling visualization. Viegas et al. [VBN*04] present methods for improving data memorability. Fisher et al. [FHRH08] present ways for tracking narrative events over time. Segal and Heer [SH10] investigate the design of narrative visualizations and identify techniques for telling stories with data. Hullman et al. [HD11, HDA13, HDR*13] design the structure of a visualization to present storytelling. Figueiras [Fig14b, Fig14a] studies how to incorporate narrative elements as storytelling elements.

3.1. Narrative Information Visualization and Storytelling

Viegas et al. summarize two methods of visualizing email archives with the aim of improving memorability of the data. Both focus on the higher level patterns of the user's email habits. The original goal was for these visualizations to uncover social patterns in the archive, but the resulting visualizations caused the user to be more reflective of the data as opposed to analytic. They look at data points and want to recall the story behind it, even share the visualization with friends. See Figure 11 [VBN*04].

PostHistory Components: Visualizing email activity: Two axis stands for time, and the dyadic relationship between user account holder and each human interaction. Pattern recognition: This includes interaction frequency, interaction rhythm, interaction balance, and archive size. Interface: The visualization includes two main panels; the calendar panel, showing email intensity, and the contacts panel, showing the names of the people being interacted with. Interaction: When the user clicks on a day square in the calendar panel, the contact panel highlights the names of the people communicated with that day. A name can be clicked on in the contacts panel and each day where that person had corresponded will be highlighted in the calendar panel. The contact panel can be viewed as an animation transitioning through the year of data.

Social Network Fragment Components: Email header data is used to derive the social context of the communication. Five different relationship types are classified. This can be either directly between correspondents or through mutual recipients in group emails. The Social Network visualization looks at each message and evaluates the role of the user (through the email address used i.e. work, school or personal) and then makes connections regarding the interaction accordingly. This data is visualized as an animation that evolves over time. Each second represents one day in the archive. A clustered word cloud is used to display the data. Previous visualizations of online social interaction data have been focused on unraveling the data from the researchers' perspective, whereas these visualizations are for the benefit of the user [BLRD02, Don95].

Information visualization systems enable users to find patterns,



Figure 11: Viegas et al. show the PostHistory visualisation. On the left is the calendar view, showing 365 squares to represent each day of the year (This image only shows data up until May). Size corresponds to the volume of emails sent on that day. The colour highlights a specific recipient that has been selected in the contact panel (left). The contact panel shows all the contacts the user has been corresponding with over the year. A contact can be selected to highlight their interaction in the calendar view [VBN*04].

relationships, and structures in data and may help users gain knowledge or confirm hypotheses [ASKH07]. The most basic element in a narrative is a character. An event occurs through the interaction of a set of characters. A scene consists of a chunk of events. A story consists of a sequence of scenes. A world model is made up of a set of stories. Akaishi et al. propose several methods visualization of chronological data based on the narrative structure of a document [ASKH07]. Akaishi et al. map each narrative component (world model, story, scene, event, character onto elements of a document, set of stories, sequence of scenes, part of sentence, sets of terms). NANA consists of a decomposition unit and a composition unit. A set of stories is stored in a database by the decomposition unit. In the database, each story is divided into scenes, forming a world model. Appropriate scenes are selected and used by the composition unit to compose a new story. When a user accesses the information, NANA provides the results as a story. The story is presented in various ways.

1. Word Colony: The dependency relationship among terms forms a directed graph, called a Word Colony. In a Word Colony, interdependent terms are embedded into the same node. The strength of term dependence is mapped onto the distance between nodes of terms, and term attractiveness is mapped onto the size of node. To visualize this relationship, Akaishi et al. use a spring model graph, which is a visual overview of a document.
2. Visualization of topic flows: NANA represents the content of a document as a Topic Sequence and Topic Matrix. Topic Sequence is regarded as the graphical plot of a document and Topic Matrix represents the relationships among several topic

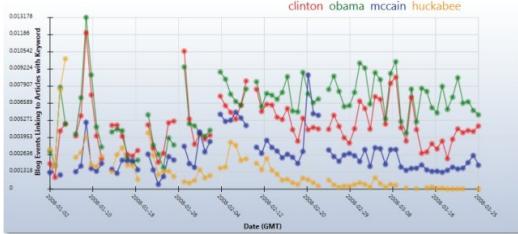


Figure 12: Fisher et al. show daily references to four US presidential candidates from January 1 to March 26, 2008. Time passes along the x axis for each candidate; number of mentions of the term along the y axis [FHRH08].

changes. Akaishi et al. support users' efforts to find desired parts of documents and to guess the context (plot) of the document.

Narrative is a simple interface that straightforwardly presents trends in keywords over time [FHRH08]. Fisher et al. present narrative as a way for presenting temporally dynamic data. In this case, narratives help the user by tracking concepts found in news stories that change over time. Fisher et al. show how to piece together complex of information and examine multiple variables. See Figure 12 [FHRH08]. Design requirements and scenarios: The first step is based on a business analysis task to find trends and public relations. In this case study, the requirement is to find out how a topic has developed over time and to see the evolution of the latest and most interesting stories [FHRH08]. System design: This includes data acquisition, temporal visualization, using other tools for correlation, understanding readership, and adding feature in narratives. The narratives project is based on Microsoft's Live Labs which provides real time data acquisition. Temporal visualization enables us see how a small group of words evolves over time relative to one another. By analysing the form of correspondence and understanding readership, additional features can be added into the narrative project [FHRH08]. Fisher et al. is based on previous work in topic detection and tracking [DKM*07] [SJ00], and temporal visualization [VWVS99], and presents narrative as a new technique in visualization [FHRH08].

Segel and Heer state that storytelling is revealing stories with data and using visualization to function in place of written story [SH10]. The Oxford English Dictionary defines narrative as "an account of a series of events, facts, etc., given in order and with the establishing of connections between them". Heer et al. investigate the design of narrative visualizations and identify techniques for telling stories with data graphics and challenges with the salient dimension of visual storytelling. They describe seven genres of narrative visualization: magazine style, annotated chart, partitioned poster, flow chart, comic strip, slide show, and video. See Figure 13. They also discuss directions for future reader-centric research [HVW07b].

1. New York Times visualization on steroid usage in sports: one larger image and line chart are combined with small images, line charts and bar charts to illustrate the usage of steroids status over 30 years. The visualization incorporates visual highlighting and connecting elements leading viewing order [Bon06]. Years

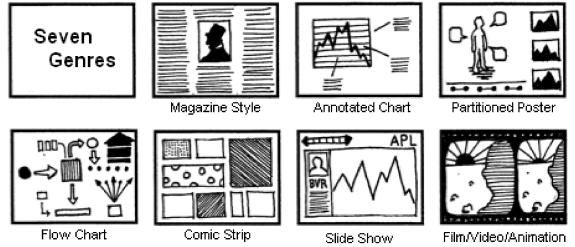


Figure 13: This shows the seven genres of narrative visualization presented by Segal and Heer [SH10]. These vary in terms of the number of frames and the ordering of their visual elements. A video, for example has a strict ordering and high frame number, whereas a 'Magazine Style' poster may have a few frames in one image that are not strictly ordered. These genre elements dictate if a story is author driven or reader driven. Author driven content uses a linear ordering of scenes and has no interactivity. Reader driven content has no prescribed order to scenes and a high level of interactivity with the reader [SH10].

is mapped to the x axis, the amount of steroids is mapped to the y axis, and different colors represent different players.

2. New York Times visualization on budget forecast: a progress bar is used to describe the accuracy of past white house budgets predictions [Cox10]. The time is mapped to x-axis, and budget situation is mapped to y-axis.
3. Afghanistan nation-building development project: This is an interactive geographic visualization with details on-demand sliders that present the status of Afghanistan nation-building development projects [MGBF10]. Opium cultivation is mapped to the color, and contours are shown on the map. Time can be changed from 2005 to 2009 by dragging the control bar.
4. Gapminder, Human Development Index: The Gapminder visualization uses animated bubble charts to show possible detrimental effects on a person's ability to follow trends [Gap05]. Continent is mapped to color, region is mapped to each bubble, and size is mapped to bubble size, position is mapped to average yearly income.
5. Minnesota Employment Explorer: This example shows how mouse-hover provides details-on-demand, double-clicking an industry triggers a drill-down into that sector while an animated transition updates the display to show sub-industry trends [JA07]. Color represents different industries, the x-axis represents the time, and the y-axis represents employment.

Segel and Heer is based on previous work of narrative structure, visual narratives and storytelling with data visualization [HVW07b] which observes the storytelling potential of data visualization and drawn parallels to more traditional media. This paper identifies salient design dimensions, clarifies how narrative visualization differs from other storytelling forms and how these differences introduce both opportunities and pitfalls for its narrative potential.

Hullman and Diakopoulos state that narrative information visualizations are a style of visualization that often explores the interplay between aspects of both exploratory and communicative visualization [HD11]. This work contributes to information visualiza-

tion design and theory by providing insight into the types and forms of given rhetorical techniques in narrative visualizations, and the interaction between those techniques and individual and community characteristics of end-users. The authors study how rhetorical techniques are used in visualization. They then investigate the resulting effects of these techniques on user interpretation [HD11]. The authors collect 51 professional narrative visualizations EG from the New York Times and BBC. Each visualization is "coded" using theory form semiotics, statistics, decision theory, and media and communication studies. The visualizations are categorized according to a selection of rhetorics information access, provenance rhetoric, mapping rhetoric, procedural rhetoric, and linguistic rhetoric. Their work provides a taxonomy of specific information presentation manipulations user in narrative visualization.

1. Mapping America visualization: The United States Census represents a nation-wide attempt to provide an objective view of the demographic of the country. Ethnic group is mapped to color, samples are shown on a map and a single ellipse represents 200 people [MA10].
2. Poll visualization: This visualization summarizes the accuracy of political poll predictions from several years and polling agencies in a small multiplies presentation of vertical line graph [D.M]. Colored bars representing the political parties are drawn to connect data points positioned on the y-axis according to the amount of time prior to the election and on the x-axis according to whether the predictions fell over (to the right) or under (to the left) of a centred vertical line representing complete accuracy (or error of zero).

Hullman and Diakopoulos is based on previous work of Segel and Heer [HVV07b] which make an initial step towards highlighting how varying degrees of authorial intention and user interaction are achieved by general design components in narrative visualization. This work examines the design and end-user interpretation of narrative visualizations in order to deepen understanding of how common design techniques represent rhetorical strategies that make certain interpretations more probable.

Hullman et al. describe a system called contextifier, which automatically produces custom, annotated visualizations from a given article [HDA13].

1. System architecture: The contextifier contains four main sections: news corpus consists of a large set of news articles. A query generator identifies the most-relevant company in the article. An annotation selection engine integrates selected features into an annotation. And the graph generator generates line graphs using annotations and series. The flow of information can be seen from Figure 15 [HDA13].

Hullman et al. is based on previous work in storytelling in visualization [SH10] and Kandogan's automatic annotation analytics [Kan12]. It develops a system that can automatically generates custom, annotated visualization from a news article of company. Hullman's work places more emphasis on providing background information or perspective on the data than Kandogan's [Kan12]

Hullman et al. [HDR*13] outline how automatic sequencing (the order in which to present visualizations) can be approached in designing systems to help non designers navigate structuring decisions in creating narrative visualizations. Their focus is on how

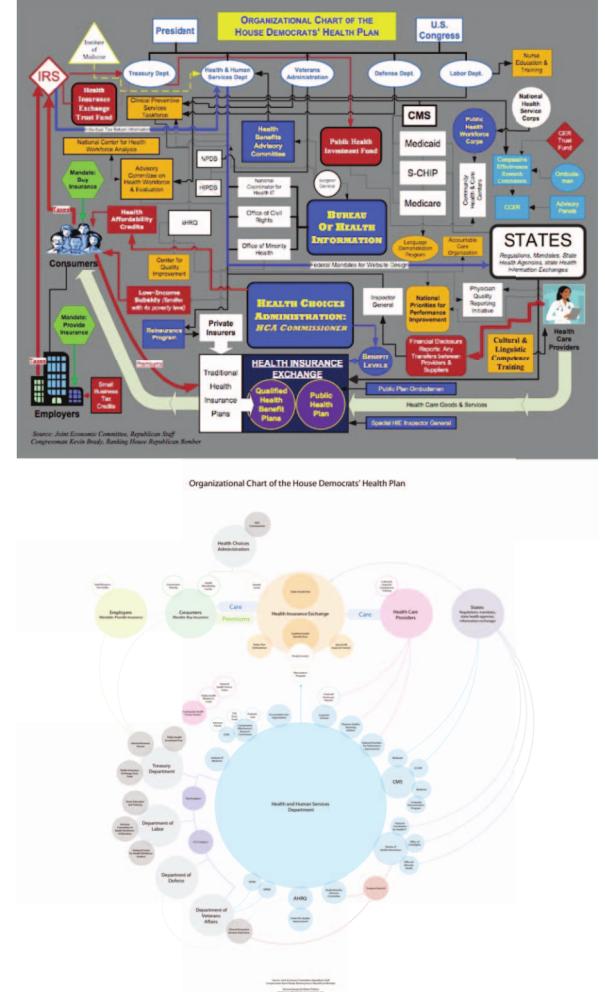


Figure 14: Figure 14 demonstrates how data can be window dressed to change the viewers opinion of it. These two images visualize the same data but each illustrator has different intended outcomes. The top image shows an unstructured, complicated graph of conflicting colors and shapes, clearly intended to confuse and obstruct the data, whereas the bottom lays the data out in a simple fashion using consistent shapes and colors [HD11].

linear, slideshow-style presentation can be optimized using knowledge of sequencing styles and strategies by incorporation.

Motivation: Hullman et al. argue that analysts using narrated data presentations could be aided by tools for identifying effective sequences for visualizations. **Study Design:** They conduct a qualitative analysis of the structural aspects of 42 examples of explicitly-guided professional narrative visualizations. One example is shown in Figure 16. **Algorithm:** They propose a graph-driven approach for finding effective sequences for narrative visualizations informed by their analysis, including defining data attributes for transitions, labelling and maintaining consistency. **Evaluation:** The result suggests a need for more sophisticated global constraints than simply

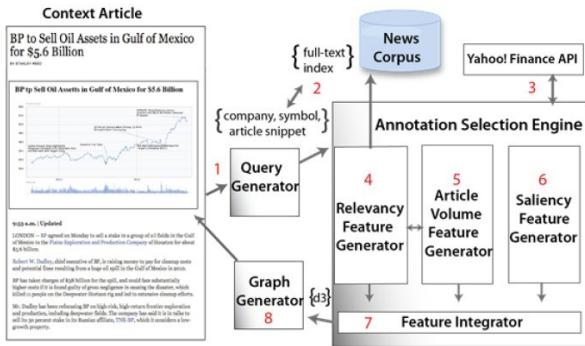


Figure 15: Hullman et al. show the architecture of contextifier [HDA13].

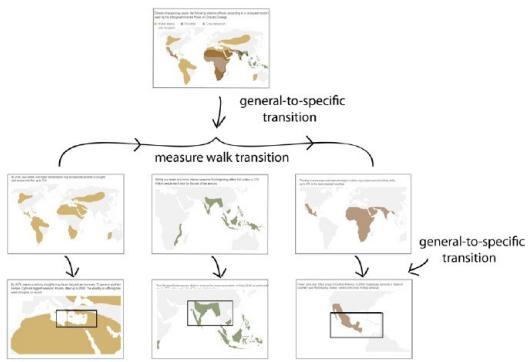


Figure 16: Hullman et al. illustrate Parallelism in sequencing in the NYT Copenhagen [HDR*13].

summing local transition costs to determine the best path through a graph of weighted visualization transitions.

This paper is based on previous work of narrative sequencing [BB79] and narrative visualization [HD11, SH10], and demonstrates that narrative sequencing can be systematically approached in visualization systems.

A visualization with a narrative is set apart from a visualization without through both its structure and its content. A narrative-based visualization attempts to create a natural flow whereby the data has an obvious progression and therefore permits easier understanding and memorability [Fig14b].

Figueiras takes professionally produced visualizations as case studies to analyze how to incorporate narrative elements as storytelling elements. By presenting prototypes of storytelling in selected case studies, Figueiras presents a design study and model for narrative visualization by using storytelling techniques [Fig14b].

1. How many households are like yours: In this figure, users can choose the primary residents and secondary members of a household, then get the number and percentage households. Figueiras [Fig14b] introduces short stories describing different

kinds of families instead of having only one article about types of families. This technique engages the user with a focus on creating empathy.

2. What does China censor online: The figure is a tag cloud that only has a title and text shaped on a map of China. Figueiras [Fig14b] introduce a tooltips pop up when a user clicks on one region, which provides more detailed information. See Figure 17. Tooltips provide additional context in the form of text which help explain the possible reasons for censorship.
3. Death Penalty Statistics, Country by Country: This figure is a static visualization with different size of bubbles representing the number of death sentence rulings. Figueiras [Fig14b] designs an interaction such that when a user chooses a year, a graph displays the number of death sentences handed out that year, which provides extra temporal information and a redesign into a story.

Narrative Strategies:

1. Context: Providing context to a visualization enables the user to make sense of the data using additional information. Without a sufficient amount of context, less meaning can be derived from the data, whereas the addition of context gives the user more information to explore the data and begin to understand features found within it. This is made easier by the development of interactive visualizations and the ability for users to choose what layers of information they see.
2. Empathy: Although not often associated with information visualization, it has been found that emotive/empathetic visualizations are more memorable and more enjoyable for the user [KM13].
3. Time Narrative: Utilizing the temporal nature of data in visualization allows users to mentally map the data by adding a sense of story flow. This improves user memorability and aids in the understanding of the data [KM13].

Figueiras is based on previous work of storytelling [HD11] [MLF*12] [SH10] and narrative visualization [FHRH08], and develops a model to add storytelling in narrative visualization [Fig14b].

Storytelling aims to simplify concepts, create emotional connection, and provides capacity to help retain information [Fig14a]. Figueiras presents the results of a focus group study on collecting information on narrative elements. She then suggests strategies for storytelling in visualization [Fig14a].

Sixteen participants are asked to study 11 information visualizations of different types and different characteristics (Interactive, non-interactive, introductory text, accompanying article, and audio narration). Then they are asked to rate visualizations in terms of comprehension, navigation, and likability. See Figure 18. The participants give high scores to all visualizations, particularly to interactive visualizations. The story suggests that a good storytelling visualization is well-structured and interactive with audience preferences. The results of the user story suggest that interactivity, the option of drilling-down, context, and a sense of relatability and importance for users to feel engaged.

Figueiras is based on previous work of narrative visualization [SH10], and storytelling visualization [KM13, MLF*12]. The au-



Figure 17: Figueiras shows a visualization of Chinese online censorship enhanced with storytelling. An interactive feature is added so that the user can click on an instance of censorship to learn more about it. This supplies context to the user and also may draw an empathetic response from the user [Fig14b].

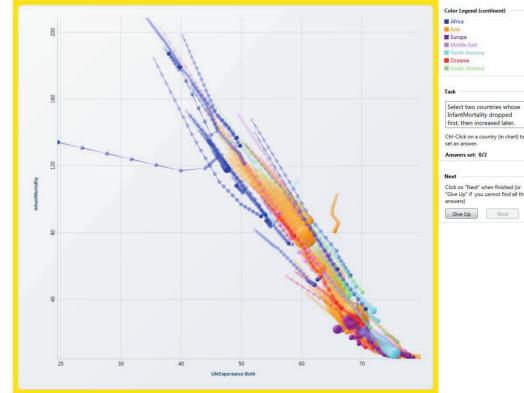


Figure 19: Robertson et al. show the trace lines of the graph animation. The traces visualisation shows bubbles at all x and y locations throughout the time frame. This is a conversion of an animation into a static image [RFF*08].

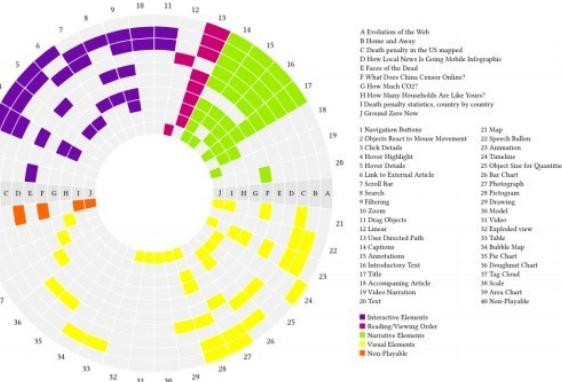


Figure 18: Figueiras shows the visualizations used in the focus group study and the elements that compose them [Fig14a].

thor uses a focus group to examine storytelling effects in information visualization and storytelling visualization.

4. Static Transitions in Storytelling for Visualization

A transition refers to the process or a period of changing from one state or condition to another according to The Oxford English Dictionary [Dicc]. In the visualization literature, transitions may be the focus of visualization and include both dynamic and static which are alternatives of presenting visualization. Static visualizations are those that do not change with time.

In this section, the visualization of transitions is generally static. The authors focus on presenting the trend of data along timelines. Robertson [RFF*08] evaluates three approaches of using bubble charts and attempts to discover which one works best for presentation and analysis. Tanahashi and Ma [TM12] presents a storyline visualization which consists of a series of lines, from left to right along the time-axis. Liu et al. [LWW*13] design a storyline visualization system, StoryFlow, to generate an aesthetically pleasing

and legible storyline visualization. Ferreira [FPV*13] proposes a method of visualizing a large amount of taxi data consisting of both spatial and temporal dimensions.

4.1. Static Transitions in Information Visualization and Storytelling

Robertson et al. define a trend in data as an observed general tendency. The most common way to see a trend in data is to plot a variable's change over time on a line chart or bar chart. If there is a general increase or decrease over time, this is perceived as a trend up or down [RFF*08]. Robertson et al. propose two alternatives to animated bubble charts for visualizing trends in multiple dimensions and describes a user study that evaluates the three approaches for both presentation and analysis. In conclusion, Robertson et al. state that traces and small multiples work best for analysis [RFF*08].

1. Multi-dimensional trends: The gapminder trendalyzer uses a bubble chart to show five dimensions of data, life expency is mapped to the x axis, infant mortality is mapped to the y axis, population is mapped to bubble size and continent is mapped to color [Ros06].
2. Alternative multi-dimensional trend visualization: A trace visualization provides the user with the ability to select particular bubbles such that the animation shows a trace line for the selected bubble as it progresses. See Figure 19 [Ros07].
3. Small multiples visualization: countries can be clustered based on position, size and location. They are further grouped by continent and ordered alphabetically within each group [Tuf90].

Robertson et al. is based on earlier work by Tversky et al. [TMB02] and Baudisch et al. [BTC*06]. Previous work is limited to small data set sizes (200 samples or less). Their work focuses on presentation rather than analysis and relies on animation to show trends over time.

Visual Storylines, by Chen et al. is designed to summarize video storylines in an image composition while preserving the style of the

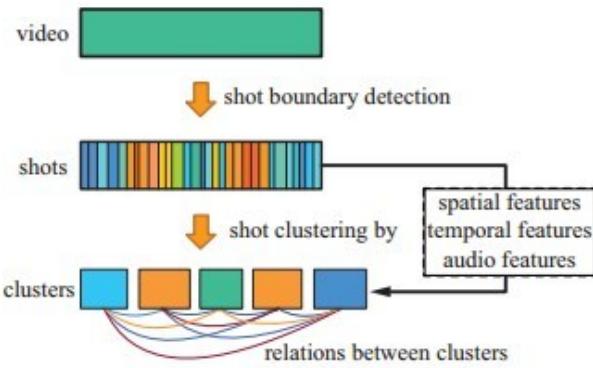


Figure 20: Chen et al. presents video shot clustering algorithm combines both visual and audio features to generate a meaningful storyline [CLH12].

original videos [CLH12]. Chen et al. present a new Visual Storylines method to assist viewers to understand important video contents by revealing essential information about video story units and their relationships. [CLH12]. Automatic storyline extraction: The first step of algorithm is to extract the storylines from a video sequence by segmenting a video into multiple sets of shot sequences and determining their relationships. See Figure 20. Generation of visual storylines: Second step is to visualize a movie sequence in a new type of static visualization by using a multi-level visual storyline approach, which selects and synthesizes important story segments according to their relationships in a storyline.

Chen et al. is based on the work of video summarization [YMH01] and first clusters video shots according to both visual and audio data to form semantic video segments.

Storyline visualization is a technique that portrays the temporal dynamics of social interactions by projecting the timeline of the interaction onto an axis [TM12]. Tanahashi and Ma present a storyline visualization which consists of a series of lines, from left to right along the time-axis, that converge and diverge in the course of their paths [TM12]. The algorithm overview is shown in Figure 21. Layout of interaction sessions: The layout is based on a set of horizontal slots that divide the screen space along the y-axis. Each of these slots has the capacity to accommodate blocks of interaction sessions as long as they do not overlap in time [TM12]. Rearranging lines: This step takes the slot-based layout of interaction sessions derived from a genome and determines the order of the line segments in each interaction session and its alignment in order to reduce unnecessary wiggles and crossovers [TM12]. Removing white spaces: In order to prevent such misleading effects, it is critical for the layout computation to include the removal of unnecessary white space to determine the final layout [TM12]. Tanahashi and Ma [TM12] is based on the idea of XKCD's hand-drawn illusion "Movie Narrative Charts" [Ogi09] and develops an algorithm for general storyline visualization.

Storyline visualizations, aim to illustrate the dynamic relationships between entities in a story [LWW*13]. Liu et al. design a storyline visualization system, StoryFlow, to generate an aesthetically

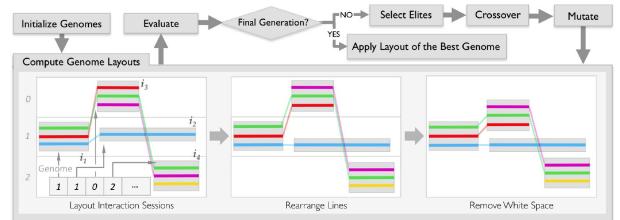


Figure 21: This visualization presents the overview algorithm of generating storyline visualization [TM12].

pleasing and legible storyline visualization. It supports real-time user interaction, hierarchical relationships among entities, and the rendering of a large number of entity lines [LWW*13]. StoryFlow: The layout pipeline consists of four steps: relationship tree generation, session/line ordering, session/line alignment, and layout compaction. In the first step, StoryFlow creates a set of dynamic relationship trees for different time frames. Then the relationship trees are used to order sessions and entity lines. Next, sessions/lines between successive time frames are aligned to maximize the number of straight lines in the layout. Finally, a quadratic optimization algorithm is performed to obtain a compact storyline layout. See Figure 22 [LWW*13].

Liu et al. is based on previous work of Tanahashi et al. [TM12]. Liu et al. add support for real-time interaction, hierarchical relationships, and a large number of entity lines.

4.2. Static Transitions in Geo-spatial Visualization and Storytelling

TaxiVis proposes a method of visualizing a large amount of taxi data consisting of both spatial and temporal dimensions. This approach looks at trends over time as opposed to individual taxi trips, visualizing data from a day in length, up to a year. Seasonal events such as thanksgiving and Christmas can be compared in a like-for-like fashion. See Figure 23 [FPV*13].

1. Time Selection Widget: This allows the user to change the time frame of the visualization.
2. Map: This is the canvas for the visualisation.
3. Data Summary: A graph of the raw data with time plotted to the X axis and frequency of taxi trips on the Y axis.
4. Visualization: To reduce clutter, a density heat map is used. This can either be as points on the map or averaging out the data within regions on the map.

Taxi data is a popular area of research. Among others, Veloso et al. explored patterns and trends in taxi ride data looking at the relationship between pick up and drop off points [VPB11a, VPB*11b]. Liao et al. developed a visual analytics system to error check GPS data streamed from taxis [LYC10].

5. Animated Transitions in Storytelling for Visualization

Gonzalez and Cleotilde define animation as a series of varying images presented dynamically according to user actions in ways that help the user to perceive a continuous change over time and develop

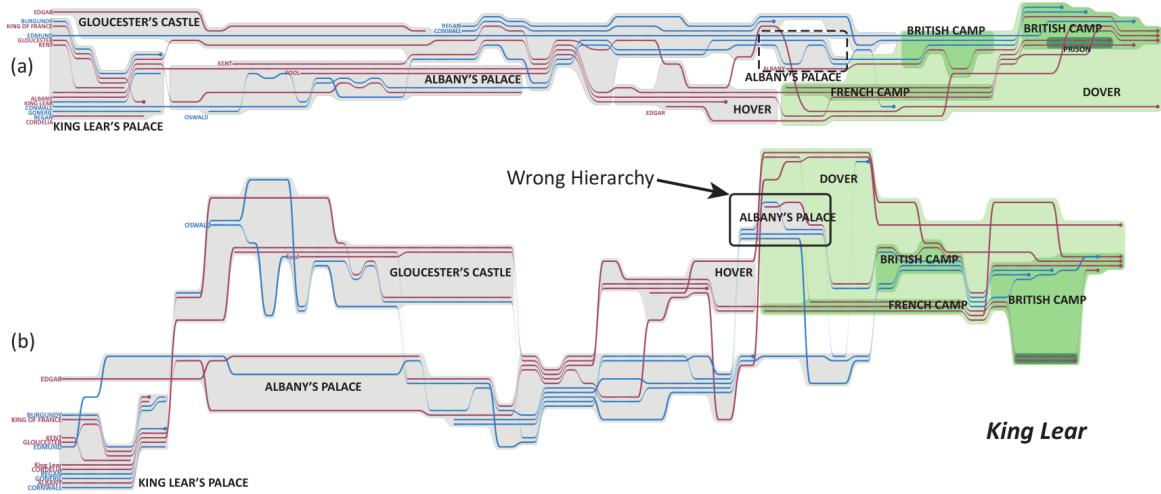


Figure 22: Comparison of King Lear using both methods of layout; (a) - StoryFlow, (b) - previous method by Tanahashi and Ma [TM12]. The StoryFlow layout presented in this paper focuses on minimising white space and efficiently ordering the story lines to ensure the most concise visual representation of a story. Intersecting lines represent interaction between characters and major events in the story are labeled to add clarity to the visualization [LWW* 13].

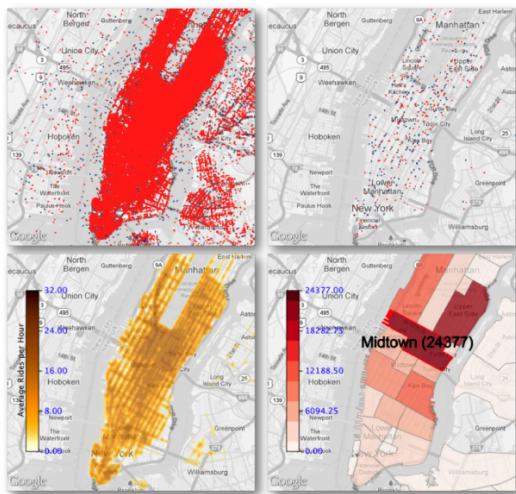


Figure 23: The top left image shows the trips rendered on the map. However the cluttered view can be reduced by employing a level-of-detail approach (top right) which takes a subsample based off the order in which the trips occurred. The bottom left image shows a density heat map of the taxi trips whereas the bottom right image averages out the data in each region to make a regional density heat map [FPV* 13].

a more appropriate mental model of the task [Gon95]. The results of their study show that decision making performance is highly contingent on the properties of the animation user interface such as image realism, transition smoothness, and interactivity style, and also sensitive to the task domain and the user's experience. Values of accuracy, time, ease of use, and enjoyability for the two types of images, transitions, and interactivity styles indicated that realistic images, gradual transitions, and parallel interactivity produced better decisions. Decision making accuracy, time, ease of use, and enjoyability in animated interfaces are influenced by the form of image representation, the transition effects, and the form of interactivity.

This research supports the idea that to be an effective decision support tool, animation must be smooth, simple, interactive, and explicitly account for the appropriateness of the user's mental model of the task.

Gonzalez and Cleotilde reviews selected empirical investigations from the literature in Education, Psychology, and HCI which suggest that animation may make interfaces easier, more enjoyable, and understandable and study the effect of animation on decision making [Gon96a].

5.1. Animated Transitions in Scientific Visualization and Storytelling

Akiba et al. introduces animation tool named: AniVis for scientific visualization exploration and communication. This tool can turn the results of data exploration and visualization into animation content and the users can create a complex animation sequence by combining several simple effects [AWM10].

1. Parameter-space blending: This operator creates intermediate

frames between two instances of frames I1 and I2 by interpolating their respective parameters. If I1 and I2 do not overlap in time, they generate intermediate frames by interpolating the parameters of the last frame of I1 and the first frame of I2. Otherwise, they generate intermediate frames by interpolating the parameters of their corresponding frames [AWM10].

2. Image-space blending: This operator creates the animation content between I1 and I2 by interpolating their respective image frames. Similarly to parameter-space blending, if I1 and I2 don't overlap in time, they generate intermediate frames by blending the last frame of I1 and the first frame of I2. The effect is that the last frame of I1 gradually fades out as the first frame of I2 gradually fades in. If I1 and I2 overlap, they generate intermediate frames by blending [AWM10].
3. Playback: This operator lets users repeatedly loop through one or more consecutive instances of interest [AWM10].

MRI Head Data: This case study focuses on highlighting a brain tumor. The animation is comprised of four pieces of animation content. The first is a spatial overview that rotates the volume data 360 degrees along the y-axis. The second piece is a spatial exploration in which the user customizes the view. The third is a parameter-space blending between a spatial exploration and a slicing, which reveals a tumor's inner structure. The parameter-space blending highlights a tumor by varying the opacity while zooming in on the region of interest. See Figure 24 [AWM10]. **Hurricane Data:** This case study has five components. The first is a caption showing the animation's content, blended with a spatial exploration that zooms in on the data. The second piece is a temporal exploration to show early time steps. The third is a variable overview that browses through three data attributes: vapor, wind speed, and cloud. The fourth piece is a temporal exploration to show later time steps. The fifth is a spatial exploration that zooms in on the hurricane's eye [AWM10].

Akiba et al. is based on previous animation support [CBB*05] and an animation enhanced system [CS05] and develops template-based visualization tools for animation.

5.2. Animated Transitions in Information Visualization and Storytelling

Bederson and Boltman examine how animating a viewpoint change in a spatial information system affects a user's ability to build a mental map of the information in the space. Based on a user-study involving a spatial map of a family tree, animation improved subjects' ability to learn the spatial position of family members within the tree without a speed penalty [BB99].

Family Tree test: two different family trees of nine individuals are presented to two groups people with animation and without animation. The subjects were given three kinds of tasks; navigation of family trees, exploratory family trees and reconstruction of family trees. The speed and accuracy of performance are recorded. In this experiment, there is a statistically significant improvement in accuracy of the reconstruction task than other tasks. Animation resulted in fewer task errors than the static ones. See Figure 25.

Bederson and Boltman is based on Gonzalez [Gon96b] and Donskoy [DK97] addresses the relationship between animation and

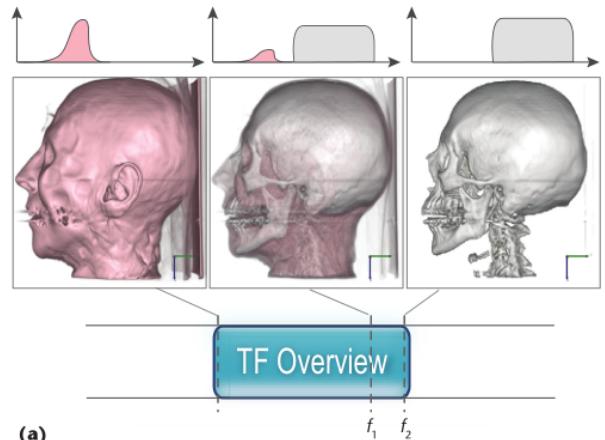


Figure 24: This image shows the AniVis animation tool displaying MRI scan data. By blending the two layers of data together, a new layer of information is revealed (middle image) [AWM10].

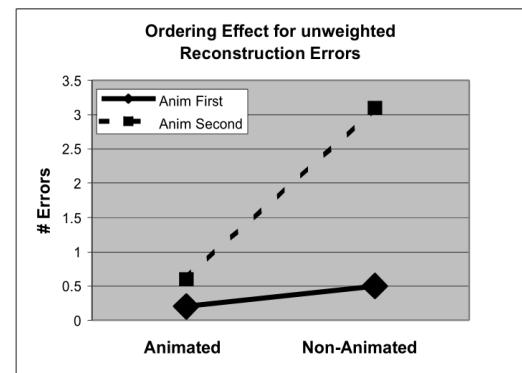


Figure 25: Bederson and Boltman show the ordering effects when presenting an animated and non-animated graphic. If the animated graphic is shown first then there is little difference in recall error, however, if the animation graphic is shown second then the recall error is significantly higher for the non-animated graphic [BB99].

users' understanding. Compared to previous work, Bederson and Boltman focus on animation of the viewpoint. The design of experiment is to change from a single in-between frame to several in-between frames.

Heer and Robertson investigate the effectiveness of animated transitions in traditional statistical data graphs, such as bar charts, pie charts, and scatter plots. A visualisation framework called DynaVis is created to test the effectiveness of animation on the user's preference and information retention. Graph animations are used to keep viewers engaged and to promote creative thinking about the data. See Figure 26 [HR07].

The software displays animated transitions of statistical data graphs. Sorting and filtering animation provide the user insight into the composition of the data. All transitions take place over a time

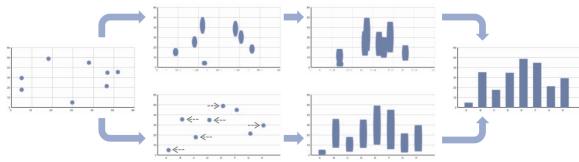


Figure 26: Heer and Robertson show the process of transition for a scatter plot to a bar chart. The top path starts by stretching the points to size and then moving to the right location, whereas the bottom path moves the dots first, then resizes and reshapes them [HR07].

frame rather than instantaneously so the user can see exactly how the visualisation has changed. Animations between different graph types are implemented by morphing the data from one shape and size to another. Statistically significant differences in user preference were found between static graphs and animated graphs. Animated transitions can improve graphical perception. This was reflected in the findings of the user experiments testing recall and understanding. However, not all transition scenarios were found to be significantly different.

Heer and Robertson is based on the previous work of Bederson and Boltman [BB99] but builds upon it by testing different transitional events.

6. Memorability for Storytelling and Visualization

Memory refers to the faculty by which things are remembered; the capacity for retaining, perpetuating, or reviving the thought of things past according to The Oxford English Dictionary [Dicd]. Memorability is an important goal of storytelling. A good visualization technique will draw the viewer's attention and increase a story's memorability [BMG*10].

Borkin et al. [BVB*13] develop an online memorability study over 2000 static visualizations that cover the large variety of visualizations and determine which visualization types and attributes are more memorable. They investigate a domain at the interface between human cognition and visualization design.

Visualization taxonomy: The taxonomy classifies static visualizations according to the underlying data structures, the visual encoding of the data, and the perceptual tasks enabled by these encodings. It features twelve main visualization categories and several popular sub-types for each category.

Memorability experiment: Borkin et al. run memorability tests via Amazon's Mechanical Turk with 261 participants and gather memorability scores.

Experiment result: The result in memorability comparison test demonstrates that there is memorability consistency with scenes, faces, and also visualizations, thus memorability is a generic principle with possibly similar generic, abstract, features. And the result in visualization attribute tests illustrates that higher memorability scores were correlated with visualizations containing pictograms, more color, low data-to-ink ratios, and high visual densities.

Borkin et al. show that visualizations are intrinsically memorable with consistency across people. Visualizations with low data-to-ink ratios and high visual densities (i.e., more chart junk and "clutter") were more memorable than minimal, "clean" visualizations [BVB*13].

All papers in this section evaluate the effects of visualization on memorability. Bateman et al. [BMG*10] explore the effects of embellishment on comprehension and memorability. And Saket et al. [SSKB15] illustrate that map-based visualization can improve accuracy of recall data comparing with node-link visualization.

6.1. Memorability for Storytelling In Information Visualization

Bateman et al. examine whether embellishment is useful for comprehension and memorability of charts. Bateman et al. compare plain and embellished charts, and conclude that a user's accuracy in describing the embellished charts is no worse than for plain charts, and that their recall after a two-to-three week gap is significantly better [BMG*10].

A comparison of plain and embellished charts: 14 embellished charts are selected from Nigel Holmes' book Designer's Guide to Creating Charts and Diagrams [Hol84], and converted to plain charts. See Figure 27. Twenty participants are presented a chart on a slide, alternating between embellished and plain versions. Participants are required to perform two tasks (reading and describing task and recall task) after five-minutes and after 2-3 weeks. The eye-gaze and task performance of participants are recorded for analysis. This study shows that there is no significant difference between plain and embellished versions for interactive interpretation accuracy and recall accuracy after a five-minute gap, but after a long-term gap, recall of both topic and detail of chart (categories and trend) is significantly better for embellished charts. Participants saw the value in message more often in Holmes charts than in the plain charts.

Previous studies have suggested that minor decoration in charts may not hamper interpretation [BB02], and work in psychology has shown that the use of imagery can affect memorability [GJ93], but there is very little work that looks at how chart imagery can affect the way people view information charts.

6.2. Memorability for Storytelling In Geo-spatial Visualization

Saket et al. illustrate that different visualization designs can effect the recall accuracy of data visualized. Comparing with a node-link diagram visualization, a map-based visualization is more effective [SSKB15].

Two datasets are examined. A book dataset (small) and LastFM dataset (large), are transformed into a node-link visualization and node-link group visualization (map-based). See Figure 28. Three phrases are performed to examine the difference between node-link visualization and map-based visualization. In phase 1 participants examine two kinds of visualization without task with unlimited time. Phase 2 asks participant to study two kinds of visualization with six tasks in a required time. Phase 3 asks participants to recall

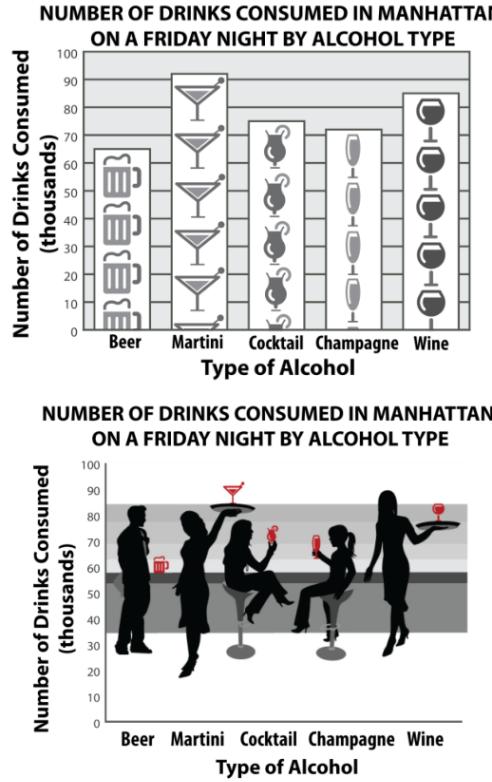


Figure 27: Bateman et al. compare two different levels of graphical embellishment of the same data. The top graph is an embellished image but still retains the recognisable features of a bar chart. The bottom image replaces the bars with a silhouette of a person next to a drink where the height of the drink corresponds to the height of the original bar. This method also uses the addition of color to emphasize the data [BMG*10].

what they read in phase 1 and 2, complete 6 tasks similar to phase 2, and 3 new addition task [SSKB15]. The result of the experiment illustrates that recalling map-based diagrams is more accurate than recalling node-link diagrams, but no faster. The participants spent more time on map-based visualizations than node-link visualizations [SSKB15].

Saket et al. is based on previous work of visualization memorability [BMG*10] and a recalling experiment [IPTO11]

7. Interpretation for Storytelling and Visualization

Interpretation refers to the action of explaining the meaning of something, according to The Oxford English Dictionary [Dice]. Martin [Mar97] refers to interpretation as the process to reach deep level which contains the fundamental values of a story. To find out the fundamental oppositions and transformations underlying the story, interpretation has to reduce all the oppositions found on figurative and narrative level to one or two basic umbrella oppositions and fundamental transformation.

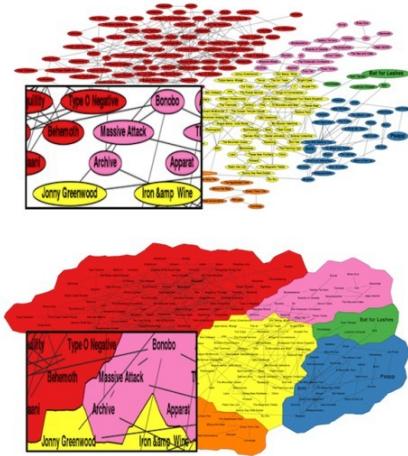


Figure 28: This figure shows two visualization of the same data: node-link diagram and map-based diagram [SSKB15].

7.1. Interpretation for Scientific Visualization

Ma et al. [MLF*12] state that a story that is well paced exhibits deliberate control over the rate at which plot points occur. They present a selection of scientific storytelling visualizations from NASA related work and describes various examples.

1. Visualization at a Scientific Research Center: The Scientific Visualization Studio (SVS) at NASA uses storytelling visualization to investigate observational data collected by instruments and sensors and make it more suitable for consumption by the public [sm14] [svs14].
2. Visualization at a Science Museum: The science museum presents visualization to the public with complex and abstract geographic phenomena at extreme size scales for explanatory animations. The science museums provide further interpretation through labels, videos, and live demonstrations. See Figure 29 [MLF*12].
3. Storytelling using interactive Visualization: Storytelling enables the user to interact with geographic data such as the Earth's climate or the collapse of a star by using a story model, such as story nodes or story transitions [AWM10].

Ma et al. is based on previous scientific visualization work at U.S. NASA, based in the scientific research center and scientific museum and describe how visualization can be used to tell a good story, and tell it well. This is a topic that the scientific visualization research community paid little attention to at that time.

7.2. Interpretation for Information Visualization

Kosara and Mackinlay define a story as an ordered sequence of steps, with a clearly defined path through it [KM13]. This paper presents a selection of previous work on storytelling in visualization and postulates storytelling as a fruitful area of future research [KM13]. The following storytelling examples are described:

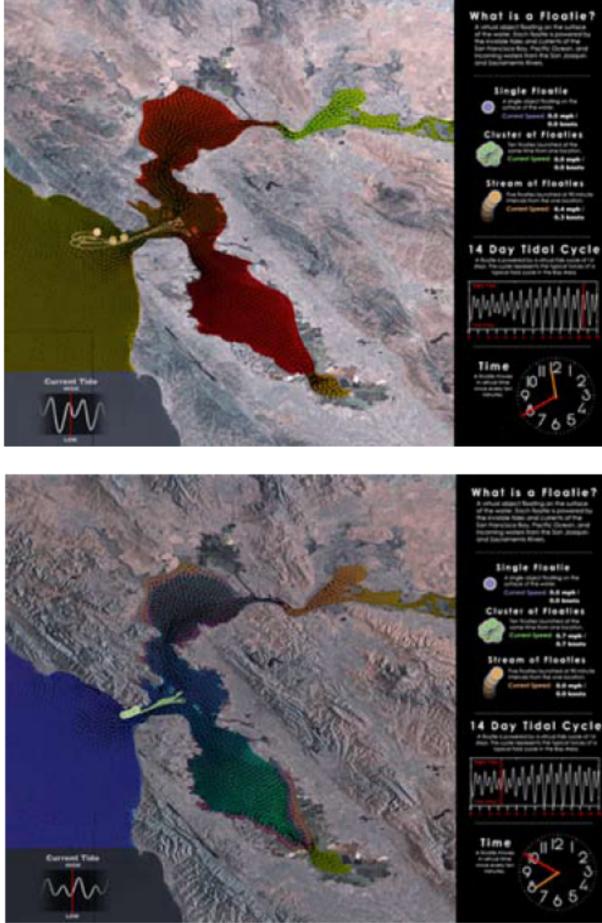


Figure 29: The interactive software used at the Exploratorium in San Francisco. The purpose of this software is to educate users on the process of how tides, currents and rivers combine in the estuary of San Francisco bay. A touch-screen is used to place floats into the virtual water so that the user can see the effects of the current on the float. Users can watch the effects of predicted tide and river flow cycles on the floats trajectory. Other contextual information is provided as an animation alongside the visualization [MLF* 12].

1. Gapminder, Human Development Index: The Gapminder visualization uses animated bubble charts to show possible detrimental effects on a person's ability to follow trends [RFF*08]. A continent is mapped to color, region is mapped to each bubble, population is mapped to bubble size, position is mapped to average yearly income.
2. New York Times, Copenhagen: *The New York Times* visualization illustrates how interaction provides a way of navigating through events by using line charts, which should be supported by a well-constructed story. Color represents different countries and the x-axis represents time and the y-axis represents a standard of industry index.
3. Minard's map of Napoleon's Russian Campaign: The storytelling visualization in this example presents a range of data at-

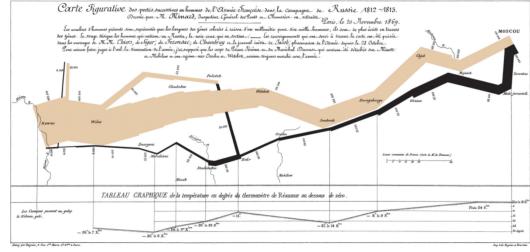


Figure 30: Minard's famous illustration of fatalities caused during Napoleon's attack on Moscow. It depicts the size of the army throughout the campaign, thinning at each battle. Although descriptive, the flow of data only loosely corresponds to time due to the tangents the line take. In addition to the army size line, the temperature is also referenced at the bottom so death rates can be directly compared [KM13].

tributes. The size of army, the path of march and temperature are shown in same graph. The width of the line represents the number of men in the army, the x-axis represents the time and the y-axis represents location and temperature. See Figure 30.

Kosara and Mackinlay is based on the previous history of storytelling, definition and model of Segel and Heer [SH10] and outlines a research program to develop storytelling as a visualization task of equal importance to exploration and analysis. Previous visualization is focused on novel techniques. This gave rise to evaluation papers that compared techniques, and tried to ascertain the perceptual mechanisms behind specific techniques.

8. Unsolved Problems and Conclusion

This survey provides a novel up-to-date overview of storytelling in visualization. The most important recent literature is included and discussed. Since storytelling in visualization is a recently new subject, we expect an increase in research over the coming years. Moreover we believe it will evolve into popular topic in visualization.

Judging by Table 1, the storytelling visualization focuses on information visualization more than scientific visualization, which conveys that more problems are left unsolved. However, by developing a storytelling model in Scientific visualization [WH07], the implementation of storytelling in scientific visualization could increase in the future. We can also see that storytelling in visualization concentrates more on exploration than on presentation. Like Kosara and Mackinlay [KM13] state: visualization techniques address the exploration and analysis of data more than presenting data.

In future work, there are many directions and unsolved problems, such as scientific visualization storytelling for presentation, dynamic storytelling in scientific visualization. And storytelling will gain importance in data presentation and data exploration.

Here is a summary of all unsolved problems in storytelling for visualization.

- Data preparation and enhancement: Virtually no one has addressed the challenge of data preparation and enhancement for storytelling, now is storytelling data best captured or derived from an existing data set or software system?
- Narrative visualization for scientific and geo-spatial visualization: Why has there been such an imbalance of research narrative visualization for information visualization but virtually none for scientific and geo-spatial visualization.
- Transitions for scientific visualization: The benefits of static transition versus dynamic transitions in visualization still remains relatively immature.
- Memorability for visualization: What are the key elements for making a memorable visualization? This is still an open research direction.
- Animated transitions for geo-spatial visualization: Animated transitions for geo-spatial visualization remains an open research question.
- Interpretation for scientific information, and geo-spatial visualization: Although we included two papers that arguably touch upon the topic of effective interpretation of stories.

The classification of literature, we present makes it clear that many future research directions remain open in storytelling and visualization

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