Guest & Research Lectures

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CPSC 547, Information Visualization 7 December 2022

http://www.cs.ubc.ca/~tmm/courses/547-22

Today

- Steve Kasica, UBC
 - -qual study: TableScraps, 15 min
 - -Q&A, 5 min
- Stephen Kobourov, Univ. Arizona
 - -algorithms: Scalable Graph Drawing w/ SGD, 15 min
 - -algorithms: MetroSets, 15 min
 - -Q&A 5-10 min
- Mara Solen, UBC
 - -survey:VisLit, 15 min
 - -Q&A, 5 min
- break, 10 min
- me
 - -design spaces: Timelines Revisited, GEViT, 30 min
 - design studies: Ocupado, Aggregated Dendrograms, 25 min
 - -imperfect models: TimelineCurator, 15 min
 - -Q&A, 10 min

Steve Kasica

Stephen Kobourov

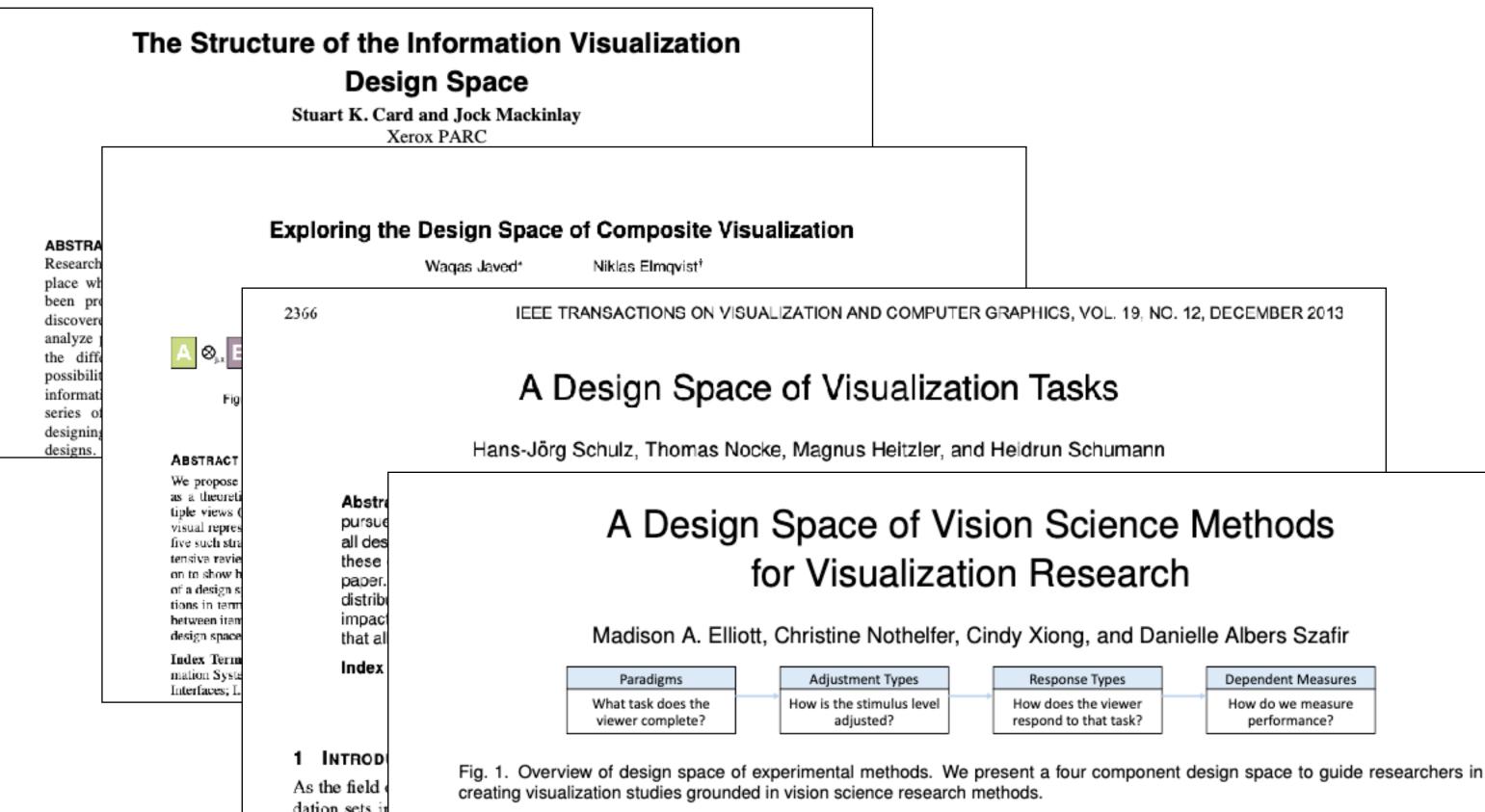


Mara Solen



design spaces

Design spaces: Continuing theme



es	Dependent Measures
ewer task?	 How do we measure performance?

Design spaces: What are they?

- impose systematic structure on set of possibilities for specific problem
 - -to capture the key variables at play
 - -to support reasoning about design choices
- delineate
 - -cross-cutting / independent / orthogonal
 - -axes / dimensions / categories

- many names
 - design spaces, taxonomies, typologies, classifications, frameworks, models, ...
 - space within which to express design patterns [Javed/Elmqvist]

Design spaces: What are they for?

- describe and analyze portions of design space to understand differences among designs & suggest new possibilities [Card & Mackinlay 1997]
- design spaces provide an **actionable** structure for systematically reasoning about solutions [Elliott et al 2020]

- taxonomies increase cognitive efficiency & support inferences [Ralph.Toward Methodological Guidelines for Process Theories & Taxonomies in Software Engineering. **IEEE TSE 2020**]
 - -by grouping similar instances together to facilitate reasoning about classes rather than instances

Design spaces: How to **assess**?

- Michel Beaudoin-Lafon, Designing Interaction, not Interfaces. AVI 2004.
 - -descriptive power: ability to describe significant range of existing examples
 - -evaluative power: ability to help assess multiple design alternatives
 - -generative power: ability to help designers create new designs

Design spaces: How to create?

open coding source material

-grounded theory / thematic analysis / qualitative analysis

• **literature** review

- synthesize across existing theories, compare & contextualize

• personal **reflection**

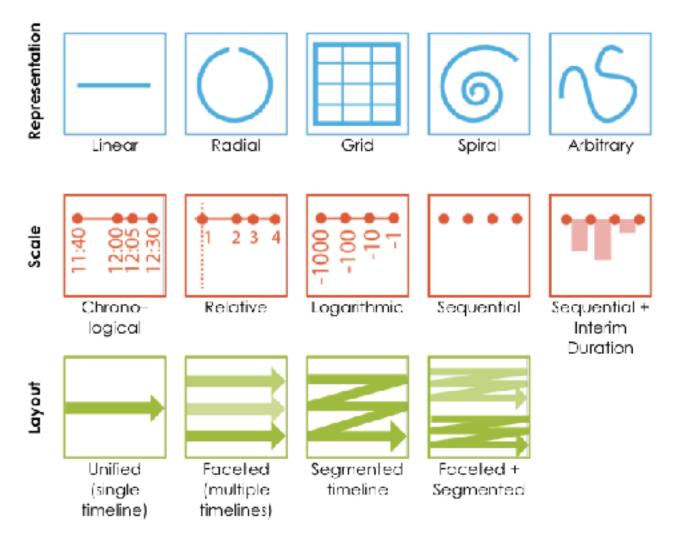
- reflective synthesis

• complex combinations...

Design spaces: Multiple examples

- datatype: temporal, **timeline** visual encoding
- domain: genomic epidemiology, paper figure visual encoding
- domain: journalism, data wrangling activities
- domain agnostic: **abstract tasks**

Timelines



Timelines Revisited

A Design Space and Considerations for Expressive Storytelling

https://timelinesrevisited.github.io/

https://timelinestoryteller.com

Timelines Revisited: A Design Space and Considerations for Expressive Storytelling Brehmer, Lee, Bach, Henry Riche, Munzner. IEEE TVCG 23(9):2151-2164

Matt Brehmer





Bongshin Lee







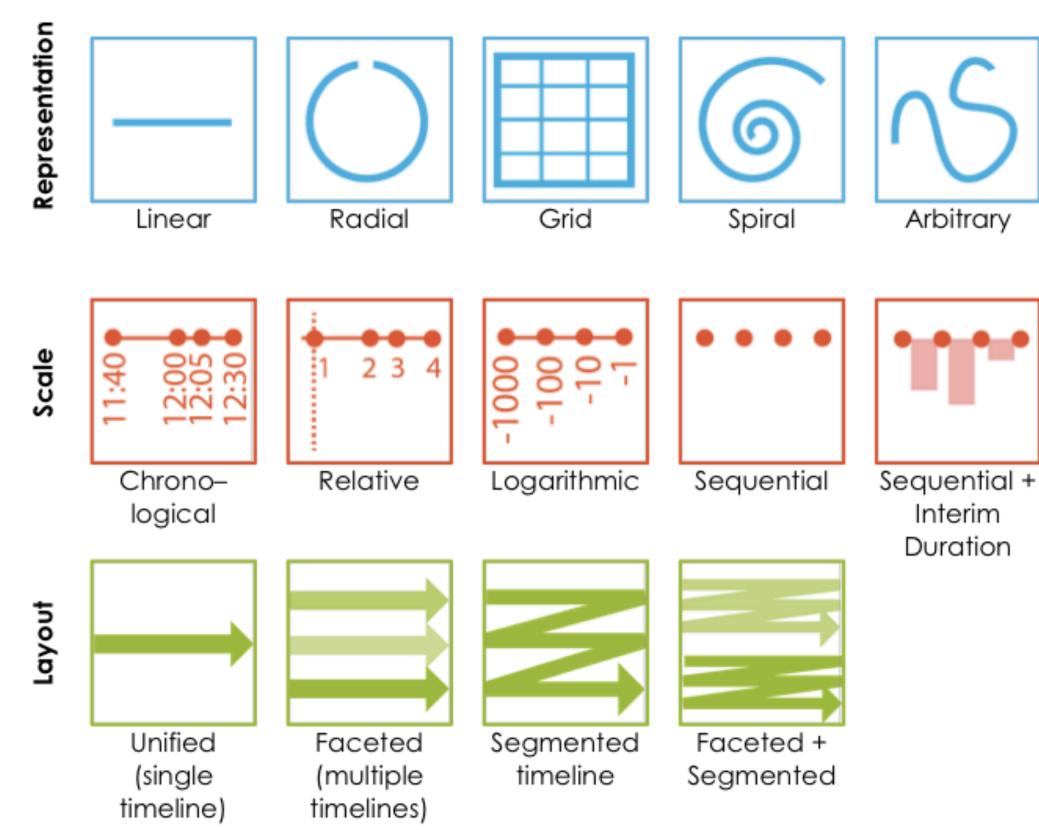


Design space with three axes

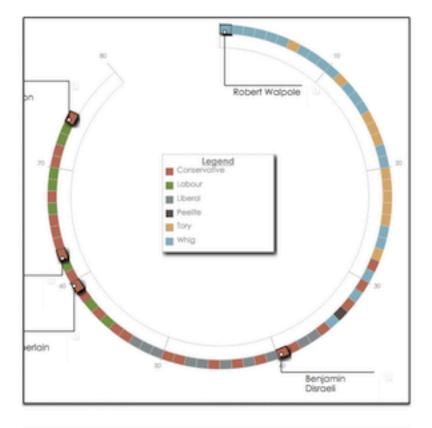


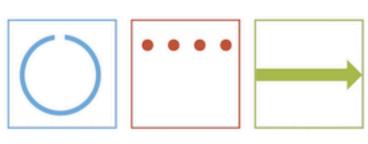
• scale

• layout

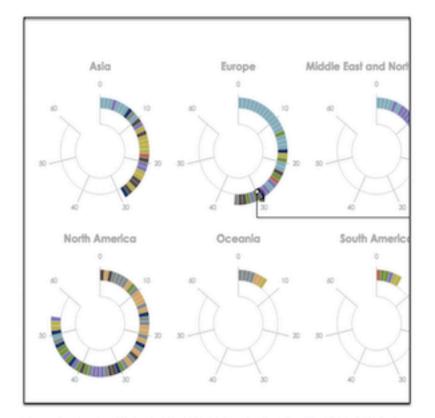


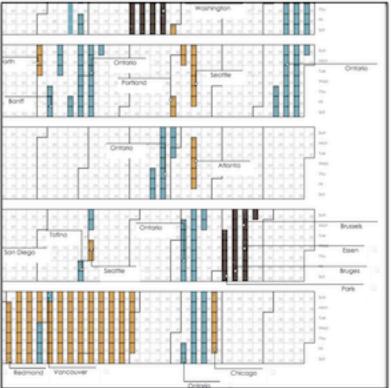
Combinations: Characterize narrative, perceptual

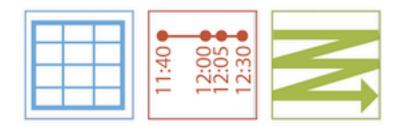




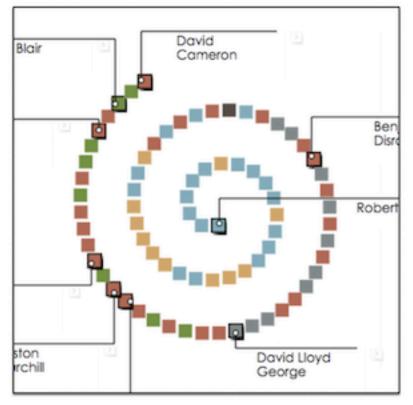
Narrative point: present a sequence of events. Perceptual task: arc position judgments. **Comment:** square aspect ratio.

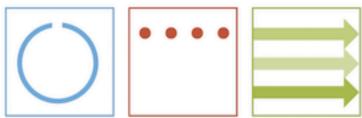




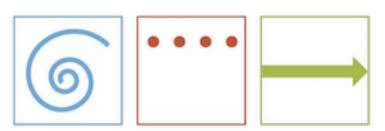


Narrative point: compare chronology, duration, periodicity of events over months, weeks, days. Perceptual task: count and position judgments. **Comment:** only supports consecutive events.





Narrative point: (approximately) compare lengths of sequences between facets. Perceptual task: arc length comparisons.



Narrative point: present a sequence of events. Perceptual task: area judgment. Comment: more compact than radial-sequentialunified timeline.

Viable combinations

- 20 out of 100
- criteria
 - purposeful
 - interpretable
 - -generalizable

	Narrotive point present chienalogy duraten at events. Perceptual task: par length end position judgment. Comment: supported by mest toos.	
	Namelive point: compare relative chicadogy, curation of events between teopts: prosent relative synchronicities. Perceptical task: partiength and position comparisons.	Butayofic aris. "Ufication" -Ter? years -Toriti year -Ter? years -Toriti year ter? years -Toriti year Life Contoler Eadion Toolty pain()
Рода усл Родан 18 20 25 11 20 25 11 20 25 Азал млер него, таслинат	Nanctive point present of sequence of events. Perceptual task: por position (originants. Comment: wide aspect potic.	
	Narrative point present chienalogy duration at events. Perceptual task: and length and position judgments.	
	Narrative point compare relative chicanology, duration of events between facets. Preceptual task, and length and position comparisons.	
	Namelive point: compare chrenology: duration, periodicity of events over centuries, depodes. Preceptual task; count and position judgments. Comment: chly supports corsecutive events.	
6 6	Narrotive point: (approximately) compare lengths of sequences between facets. Perceptual task: area comparisons. Comment: compact.	





chronology of a fimeline with emphasis on recent 1018 9908 OVER'S. Perceptual task; bar sion (sody position judgment. Comment do not stort story with log scale.



Nanalive point: [accurately] compare engins of sequences. between facets. Perceptual task: bar position comparisons. Comment effectively a bar chort.

Narrative point: compare

chronology, duration of

events between facetz

present synchronicities

Perceptual task: barlenath

and position comparisons.

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Nanalive point; preser

[coinciding events].



Nanative point: compare chronology, duration of even's between focets. Perceptual task: ara length and position comparisons.



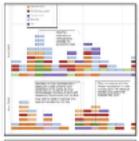
Nanalive point; present a sequence of events. Perceptual task: and position jurgments. Comment: square aspect call or.

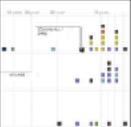


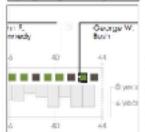
Nanalive point: compare chronology, duration. periodicity of events over months, weeks, doys, Perceptual task: count and position judgments. Comment only supports consecutive events.

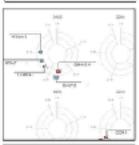


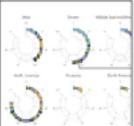
Nonative point: present a sequence of events. Perceptual task: path length judgment. Comment proitrary shape as mnomonia.

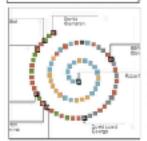














Norative point: compare chronology, curation, potiodicity of events. between segments. Perceptual task: bar length and position comparisons.



Nerralive point: comparechronologies with emphasis on moont events between facets, show synchronicity, Perceptual task: bar position comparisons. Comment: do not stort story will log scale.



Narolive point present of sequence of events with distribution of time elepsed belween events.

syear Perceptual lask bar length and position judgments. Comment non-uniform enronological distributions



Norralive point compare chronology, durchen, periodicity of events between segments. Perceptual tasks and kength and position comparisons.



Nortalive point: (approximately) comparelengths of socuences between facets. Percepted task: are length comparisons.



Nanotive point: present of sequence of events. Perceptual task area Judgment Comment: more compact than radial-sequentialunified time inc.

Process

- create design space
 - -assemble source material corpus: 145 timeline visualizations & timeline tools
 - -open code group timelines together, select example for group, sketch alternatives
 - -result: 3-axis design space
- analyze design space
 - -24 unique combinations (of 100) found in corpus
 - -20 we deemed viable

ons & timeline tools group, sketch alternatives

Assessment & adoption

- descriptive power
 - -validated coverage through checking 118 additional timelines ("test set")
 - all timelines can be described (263 total)
 - 253 characterized as viable
- generative power
 - -implemented sandbox authoring software for 20 viable designs
 - & transitions between them
 - -created designs for 28 representative datasets
 - 7 full story videos
- adoption

-open sourced & distributed as Microsoft product

- free browser version at <u>https://timelinestoryteller.com/</u>
- free add-on for PowerBI

Genomic Epidemiology



A systematic method for surveying data visualizations and a resulting genomic epidemiology visualization typology:

GEViT

https://amcrisan.github.io/gevit

A systematic method for surveying data visualizations and a resulting genomic epidemiology visualization typology: GEViT. Crisan, Gardy, Munzner. Oxford Bioinformatics 35(10):1668-1676, 2018.

Anamaria Crisan @amcrisan



Jenn Gardy @jennifergardy



Propose typology creation method: mixed qual and quant

- Analyzed research articles
- Some analyses are automated (¹/₁) and others are manual (¹/₁)



Quantitative Analysis



Use method to develop typology in specific domain

Developed a <u>Genomic Epidemiology Visualization Typology</u> (GEViT)

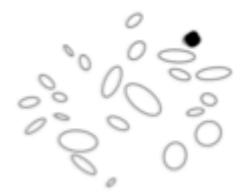


Topic Clusters Sampling Strata

Article Sampling Random stratified sampling

Visualization Analysis

Figure Extraction Sample articles





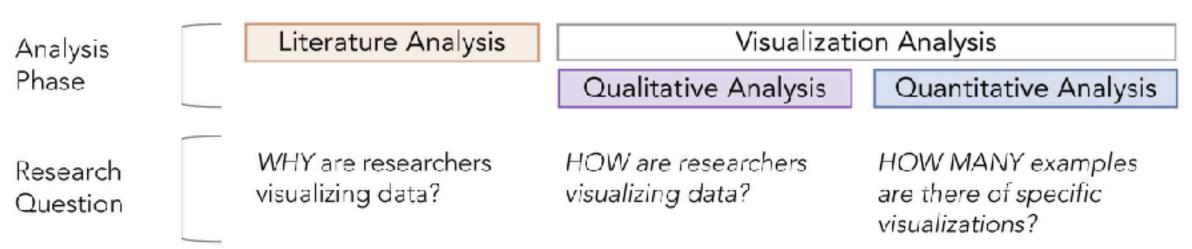


Iterative & Axial Coding Development of GEViT

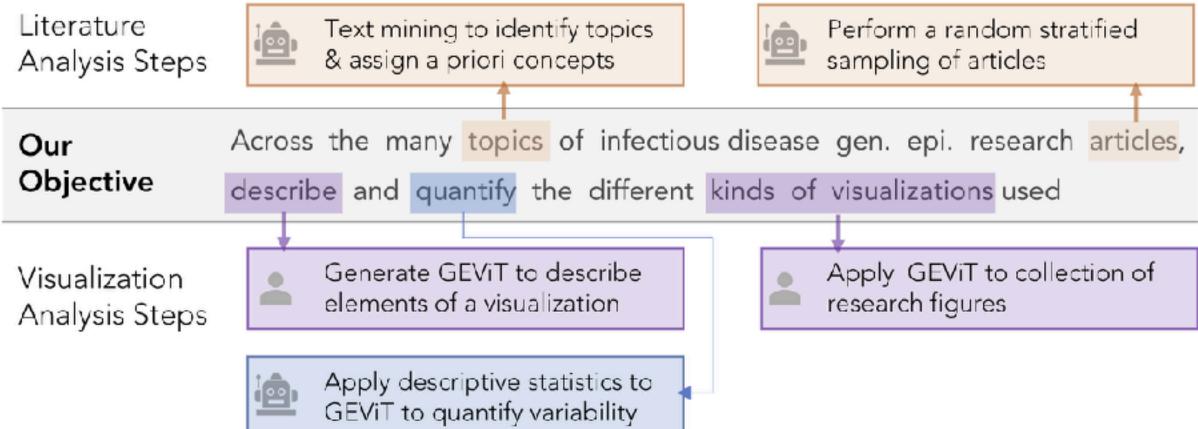
Chart Type Chart Combination Chart Enhancement

Domain prevalence design space

General Method Overview



Application of our Method to Infectious Disease Genomic Epidemiology в



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By the numbers Analysis Step

Number of Articles 17,974 Article acquisition and unsupervised topic clustering Remove articles 15,315 that never cluster Validation and limitation to human pathogens Remove articles with 9,551 non-human pathogens Application of a priori Remove pathogens with 6,350 concepts fewer than 40 articles Sampling Round 1 Reject (No) or Accept (Yes) 276 6074 Yes No 179 97 Sampling Round 2 293 Reject (No) or Accept (Yes) No Yes 186 107 204 Finalization of articles + 17 manually added

221

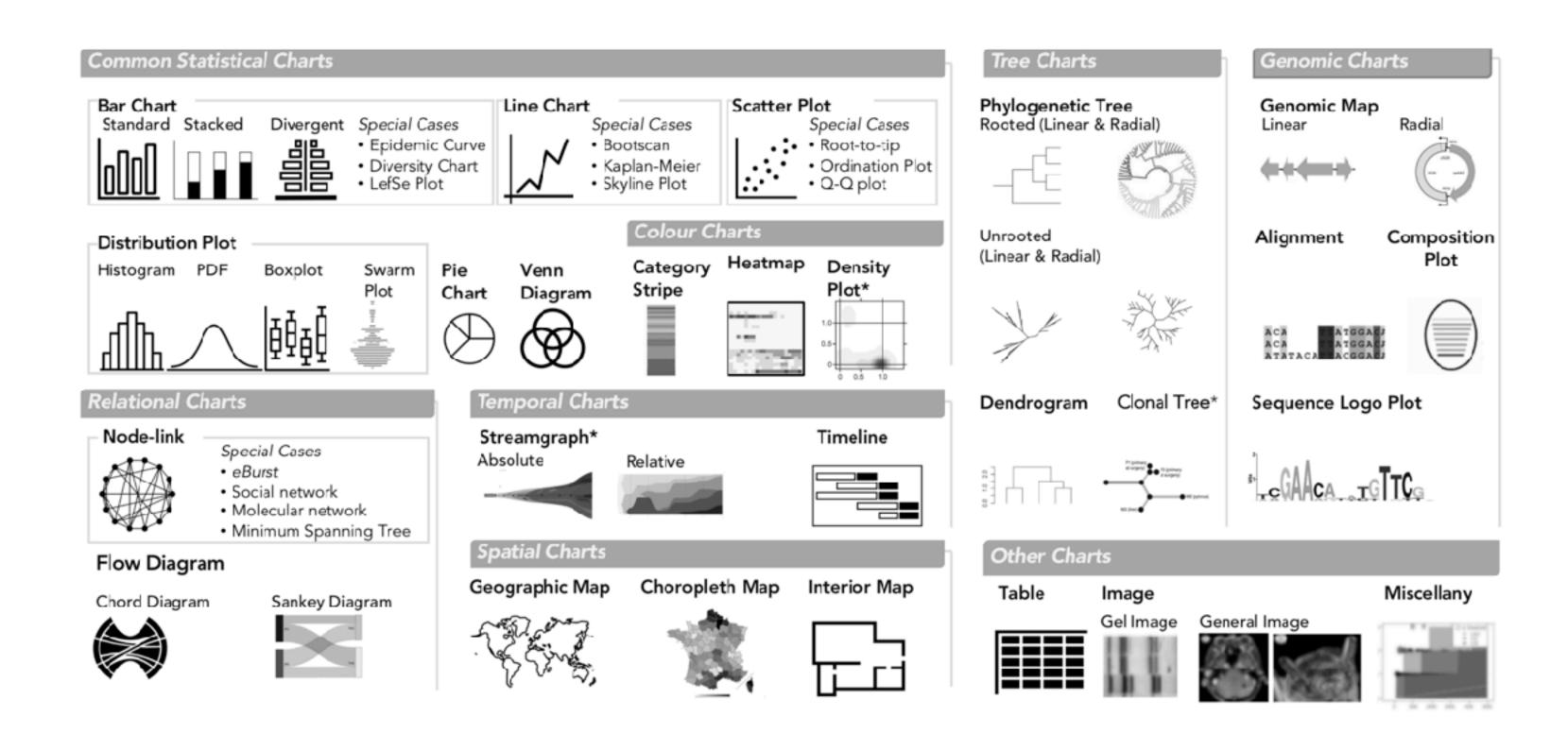
Results

35 topic clusters

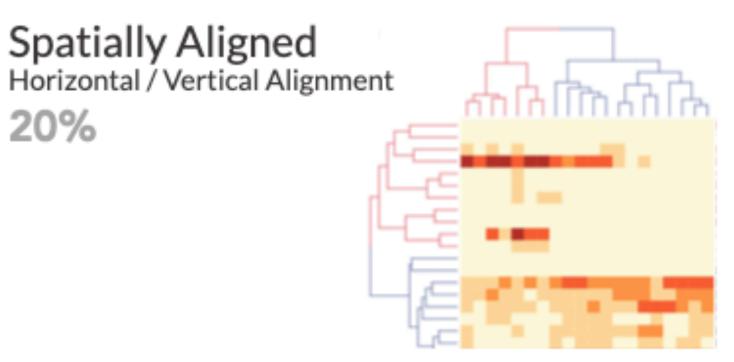
18 pathogen topic clusters 23 a priori concepts

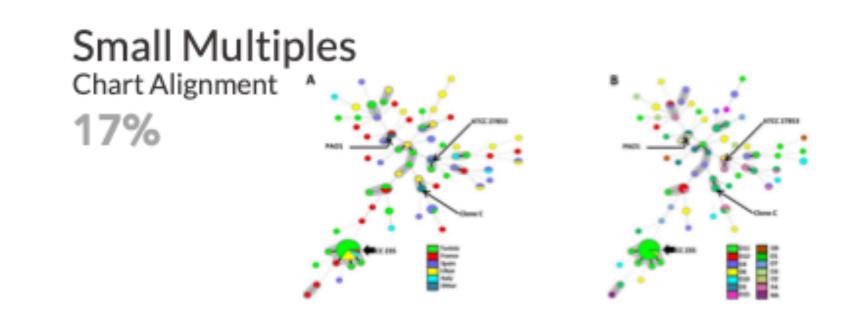
801 Figures 49 Missed Opportunity Tables

Design space axis: Chart types used in genEpi



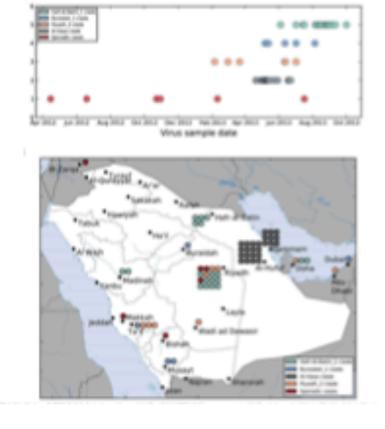
Design space axis: Chart combinations of heterogeneous data



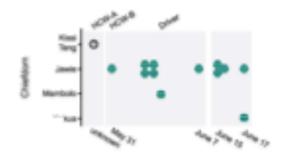


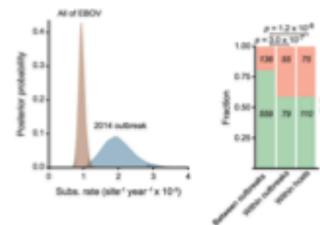
Visually Aligned Colour / Shape Alignment 14%

20%

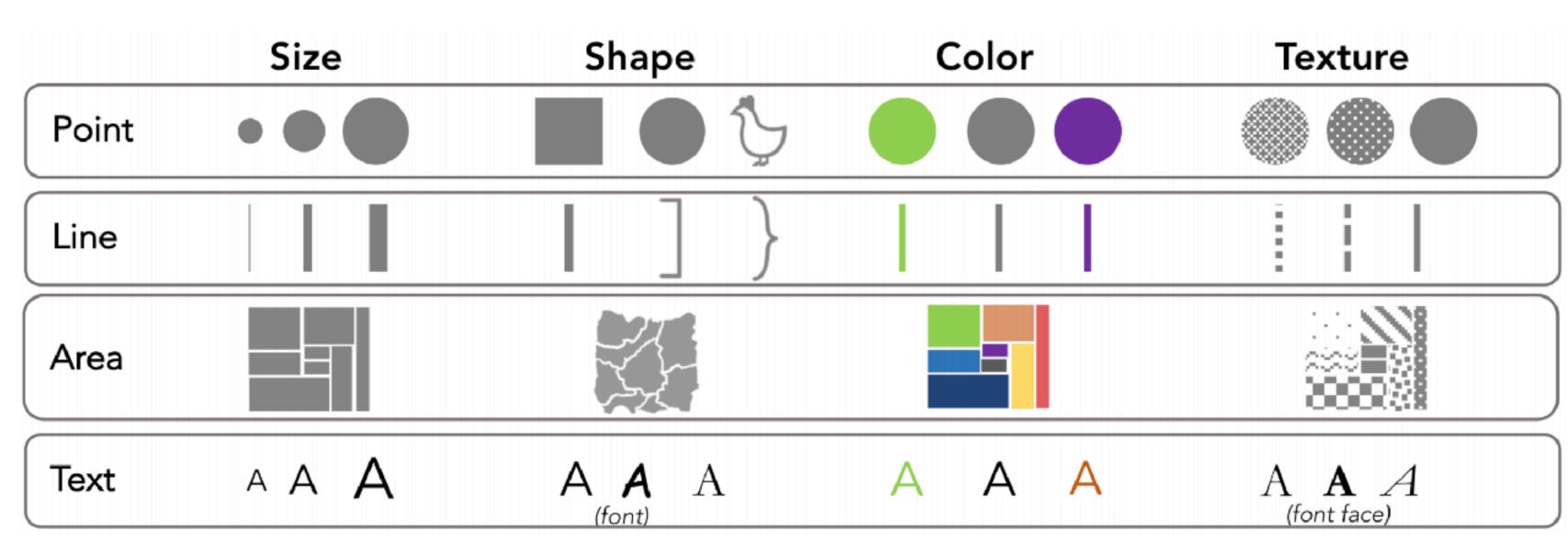


Unaligned 9%

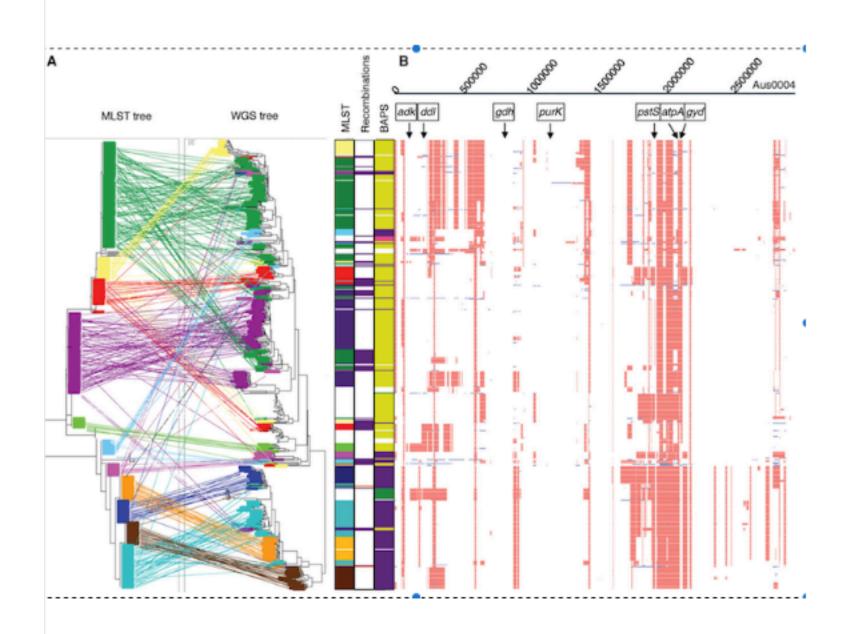


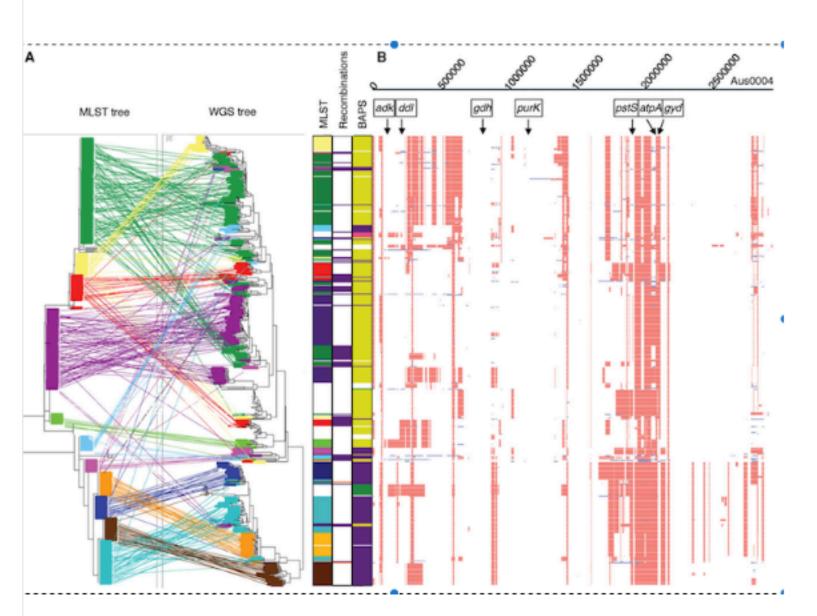


Design space axis: Enhancement choices, atop base chart types



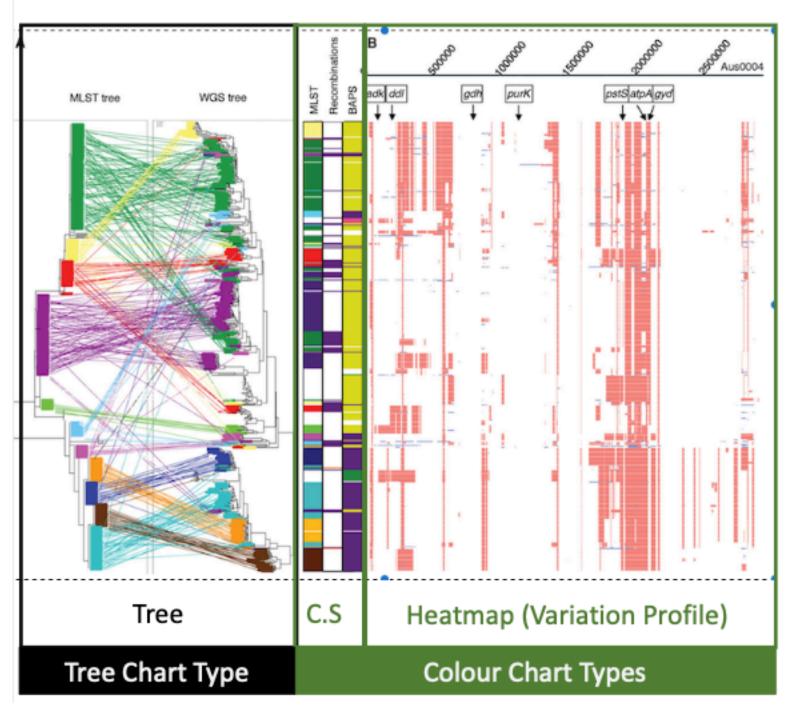
Current Practice >80% of all figures have some enhancement





Visualization Breakdown

Literature Analysis (why) Pathogen: Enterococcus faecium



Visualization Breakdown

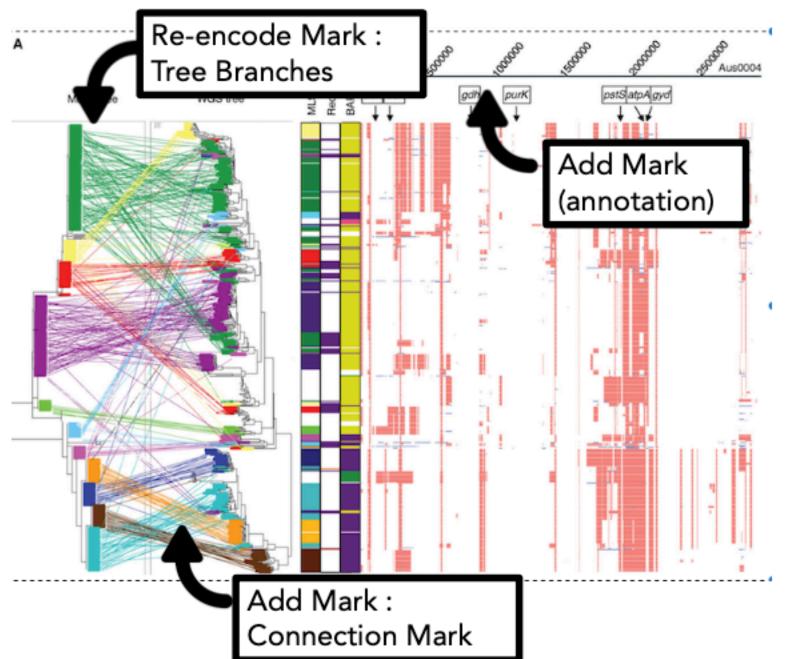
Literature Analysis (why) Pathogen: Enterococcus faecium

Visualization Analysis (how)

Chart Type	Tree (Categ Heatn
Chart Combination	Spatia

(Rooted Phylogenetic Tree) gory Stripe map (Variation Profile)

ally Aligned (horizontal)



Visualization Breakdown

Literature Analysis (why)

Pathogen: Enterococcus faecium

Visualization Analysis (how)

Chart Type	Tree (Rooted Phylogenetic Tree) Category Stripe Heatmap (Variation Profile)		
Chart Combination	Spatially Aligned (horizontal)		
Chart Enhancement	Re-encode Marks	Tree – branches	
	Add Marks	Tree - Connection Marks	
	Add Mark (unstructured)	Heatmap – Textboxes	

Assessment

- descriptive power
 - -provided common language for describing data visualization in genEpi
 - -established gap: unmet tooling needs
 - no existing tool handled full complexity of what people do manually
- evaluative power
 - -revealed shortfalls in practices of some genEpi stakeholders
 - eg overuse of text
- generative power
 - -validated in followup GEViTRec work
 - **build** automatic recommender system using domain prevalence design space

GEViTRec:

Data Reconnaissance Through **Recommendation Using a Domain-Specific** Visualization Prevalence Design Space

https://github.com/amcrisan/GEVitRec

GEViTRec: Data Reconnaissance Through Recommendation Using a Domain-Specific Visualization Prevalence Design Space. Crisan, Fisher, Gardy, Munzner. IEEE TVCG 29(12):4855-4872, 2022.

Anamaria Crisan @amcrisan



Shannah Fisher



Jenn Gardy @jennifergardy



Summary: Multiple design spaces

Design Space	Open Coding Source Material	Sampling Strategy	Reflective Synthesis Timing
timeline visual encoding	standalone timelines	assembled corpus	early
<mark>genEpi</mark> visual encoding	figures from papers	stratified random sampling with topic clusters	-
wrangling activities	software from repos	diversity criteria	late

Vis Research Literature

some source material

terms: light mapping

Summary: Multiple design spaces

Design Space	Descriptive Power	Generative Power	Descriptive vs Generative	Evaluative Power
timeline visual encoding	validated against test set	software implementation of authoring system, used to create example gallery/videos	analysis to characterize viable subset	
genEpi visual encoding	systematic method yields comprehensive coverage	software implementation of automatic recommender (followup)	same (detailed)	
wrangling activities	high precision, gaps / divergence found for domain	concise framework (followup implementation TBD)	develop entirely new framework	

Design spaces: How to assess? Larger context: theory types

- Ben Shneiderman, Designing the User Interface: descriptive, explanatory, prescriptive, predictive
- Paul Ralph, Toward Methodological Guidelines for Process Theories & Taxonomies in Software Engineering, IEEE TSE 2020
 - theory types
 - theories for **understanding**: organizing what is happening into useful categories (taxonomies)
 - **process** theories: how something happens (often taxonomies++)
 - variance theories: why something happens, causal relationships between constructs - predictive
 - relevant criteria for taxonomies
 - **yes**: parsimony, transferability, theoretical saturation
 - **sometimes**: utility, originality, resonance/believability, testability
 - **no**: statistical generalizability, construct validity, internal validity, conclusion validity

design studies

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Two design studies

• facilities management

biology





40



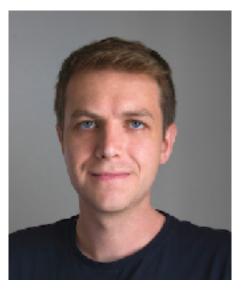
Ocupado Visualizing Location-Based Counts Over Time Across Buildings

http://www.cs.ubc.ca/labs/imager/tr/2020/ocupado/

Ocupado: Visualizing Location-Based Counts Over Time Across Buildings.

Oppermann and Munzner. Computer Graphics Forum (Proc. EuroVis 2020) 39(3):127-138 2020.

Michael Oppermann



Location-Based Counts





Previous measurement required physical counting or installation of additional hardware.



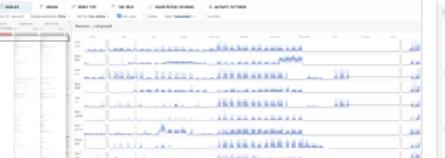
Previous measurement required physical counting or installation of additional hardware.



Previous visualization attempts were limited in space and time.



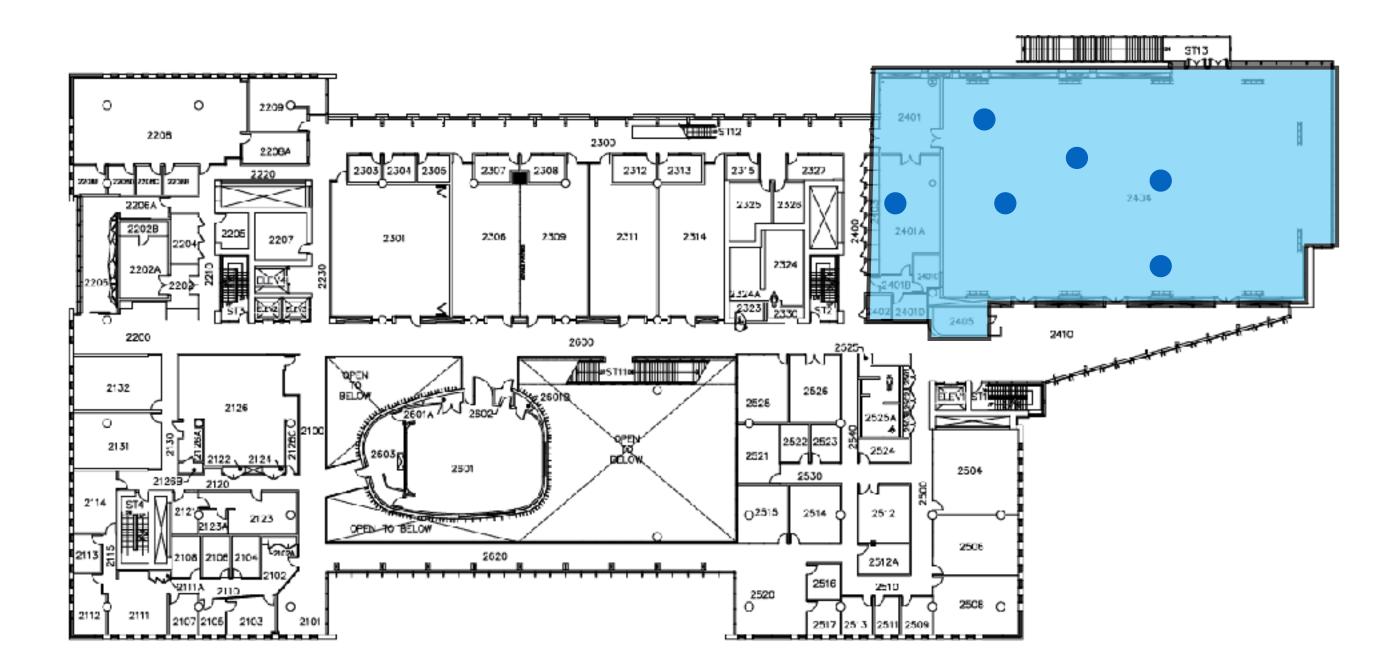


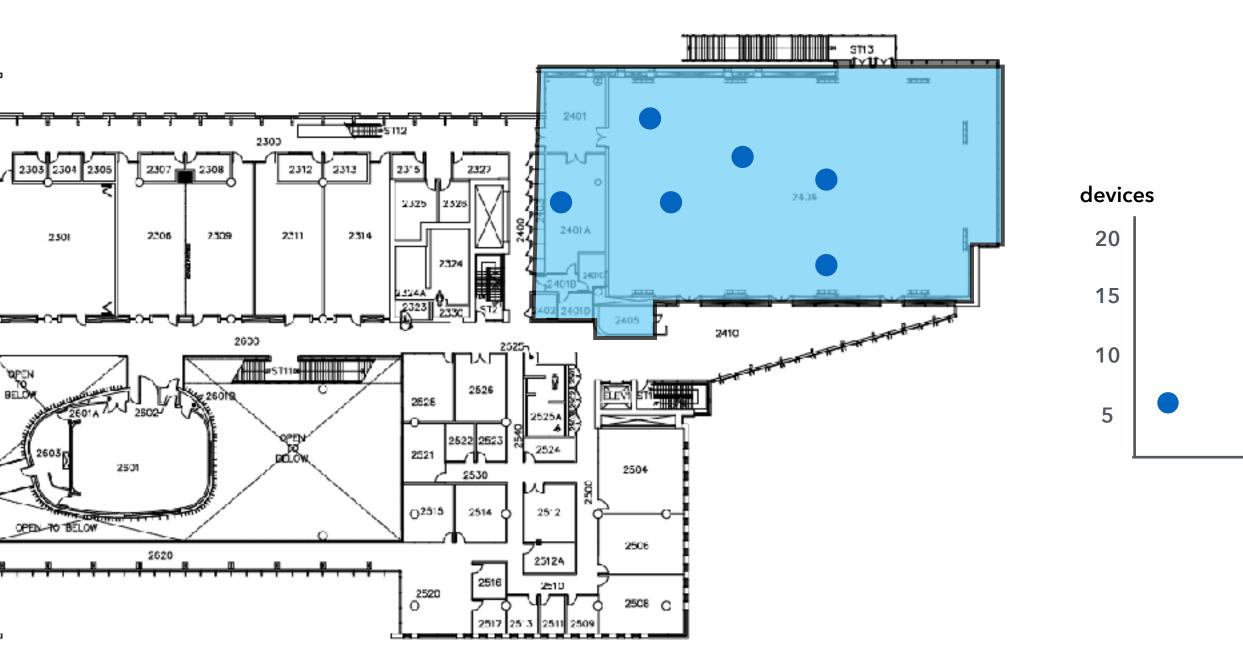




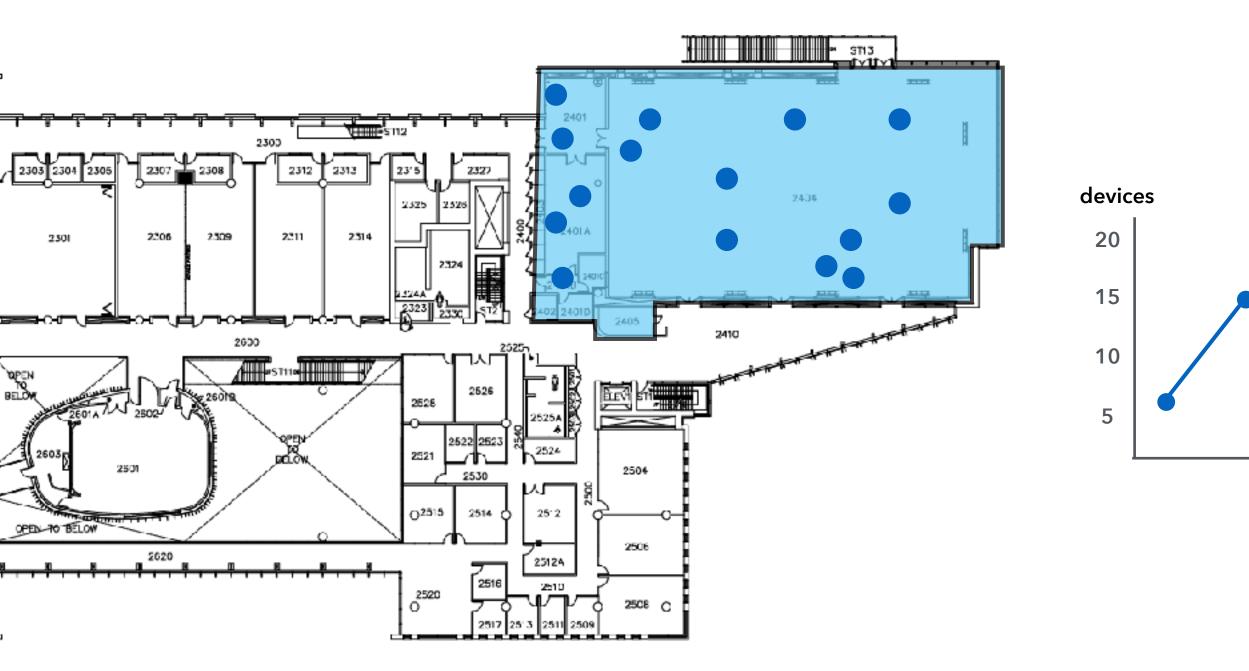








time

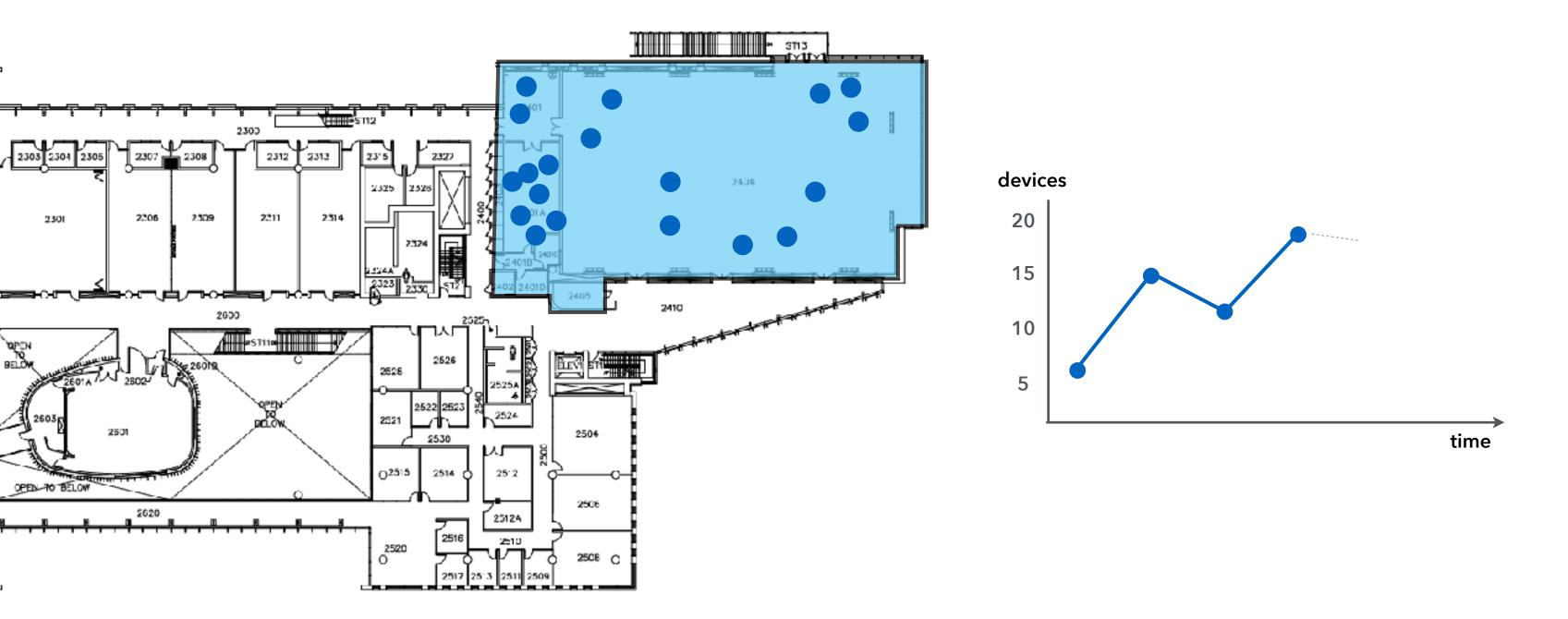


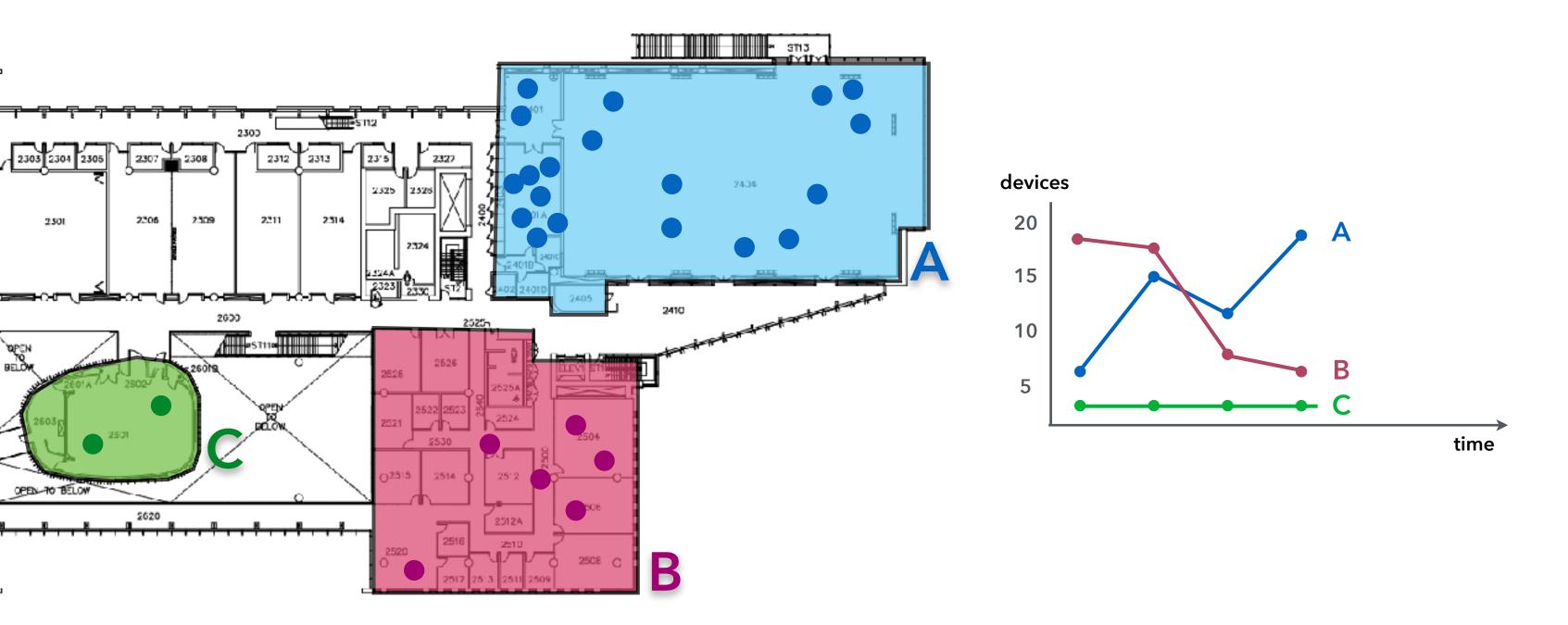














XCOOA-

Location-Based Counts

- Regular intervals (e.g., every 5 minutes)
- → Spatial hierarchy (Zone → Floor → Building → Campus)
- No trajectories or device identifiers are recorded
- Intrinsic privacy advantages

Data

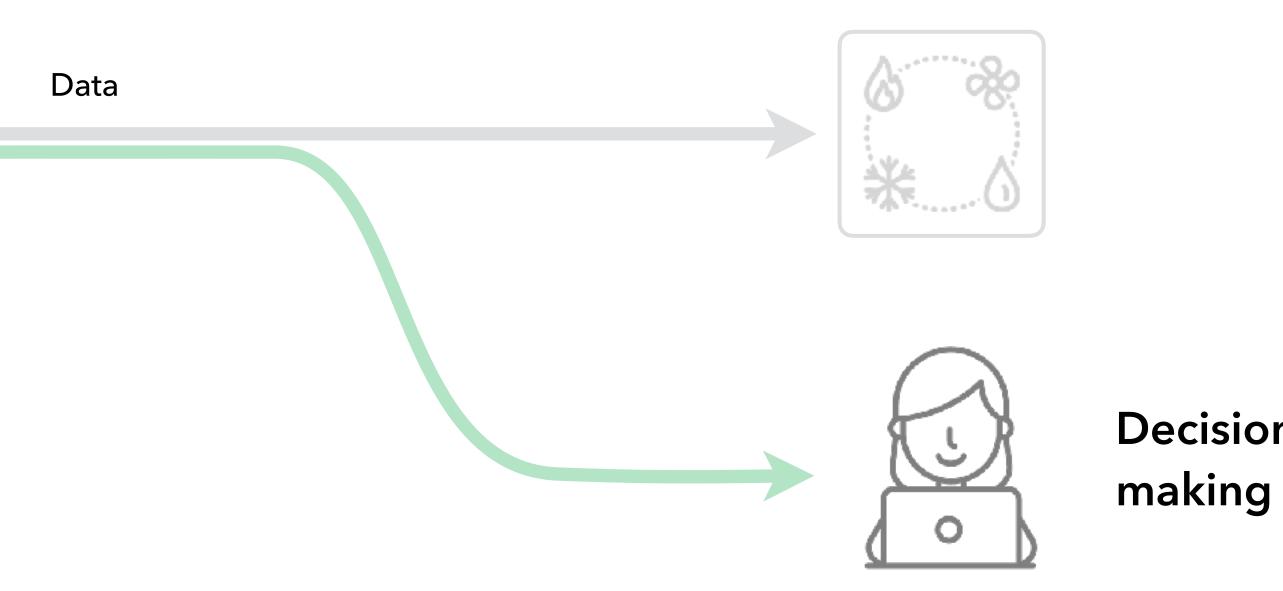


Automated **HVAC** control

Data



55



Decision

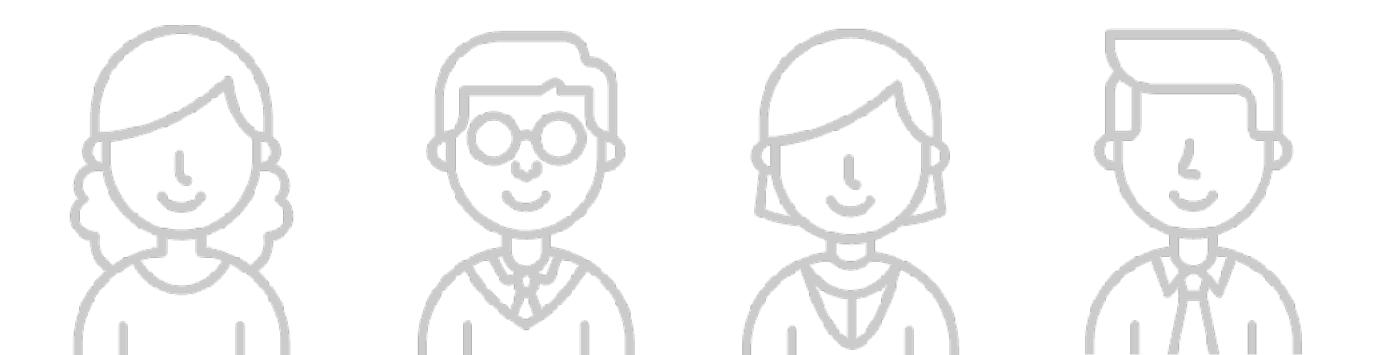


WiFi connections as a proxy for occupancy



WiFi connections as a proxy for occupancy

Interviews with potential stakeholders



Focus Domains

- Space planning
- Building management
- Custodial services
- Classroom management
- Data quality control

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Focus Domains

- Space planning
- Building management
- Custodial services
- Classroom management
- Data quality control



Semi-structured discussions and live demos



Do students occupy room x in evenings or on weekends?

ns. 1 weekends?



Confirm assumptions or previous observations.



Monitor the current/recent utilization rate. Which rooms are empty/busy?



Confirm assumptions or previous observations.



Monitor the current/recent utilization rate.



Communicate space usage and justify decisions. Space usage improved after renovation.



Confirm assumptions or previous observations.



Monitor the current/recent utilization rate.



Communicate space usage and justify decisions.



Validate the data (quality control). Check minimum size of a room that can be captured.

Spatial and Temporal Data Granularities

Visualization Prototypes



Data sketches, static data export

Time

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Visualization Prototypes



Data sketches, static data export

- original plan: different interface for each stakeholder
- realization: task & data abstractions match multiple stakeholders
 - if slice by space & time granularity

Spatial and Temporal Data Granularities

Regions of interest

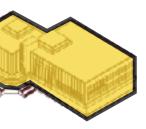




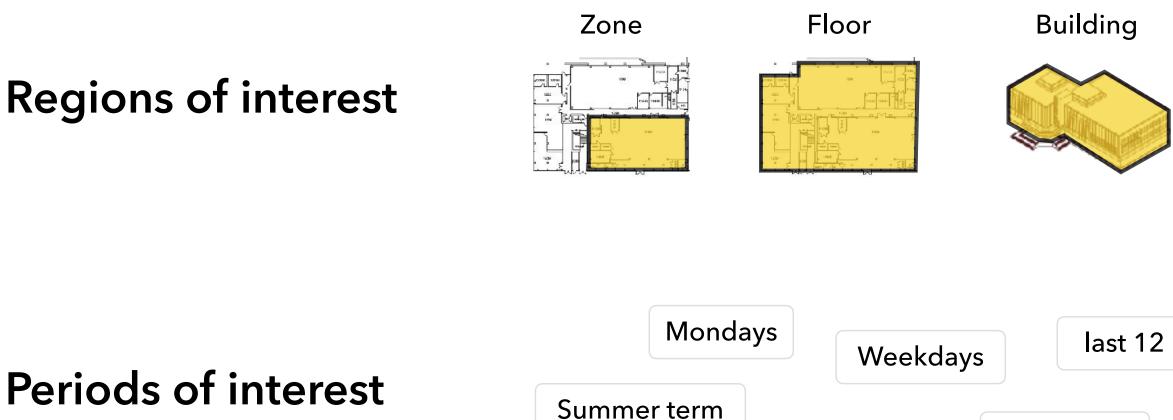




Building



Spatial and Temporal Data Granularities



last 12 hours

Weekends

Fr 8-10am

Visualization Prototypes

Sandbox

Data sketches, static data export

Campus Explorer

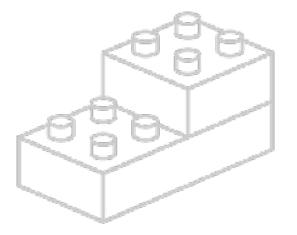
Live-data stream, cross-building analysis

Time

Building Recent

Building Long-term

Region Compare



Reusable Visualization Components

Layout	Visual Encoding	Facet	Comparisor
	Sparkline	Juxtaposition	Repeating (contiguou



ons

g patterns, trends, outliers ous)

Layout	Visual Encoding	Facet	Compariso
	Sparkline	Juxtaposition	Repeating (contiguou
	Box-plot-bars	Juxtaposition	Repeating (non-contig



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Layout	t Visual Encoding	Facet	Compariso
	Sparkline	Juxtaposition	Repeating (contiguou
	Box-plot-bars	Juxtaposition	Repeating (non-contig
	Confidence band line chart	Aggregation	Typical util



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tilization profiles

Layout	Visual Encoding	Facet	Compariso
	Sparkline	Juxtaposition	Repeating (contiguou
	Box-plot-bars	Juxtaposition	Repeating (non-contig
	Confidence band line chart	Aggregation	Typical util
	Superimposed line chart	Superposition	Within-ses



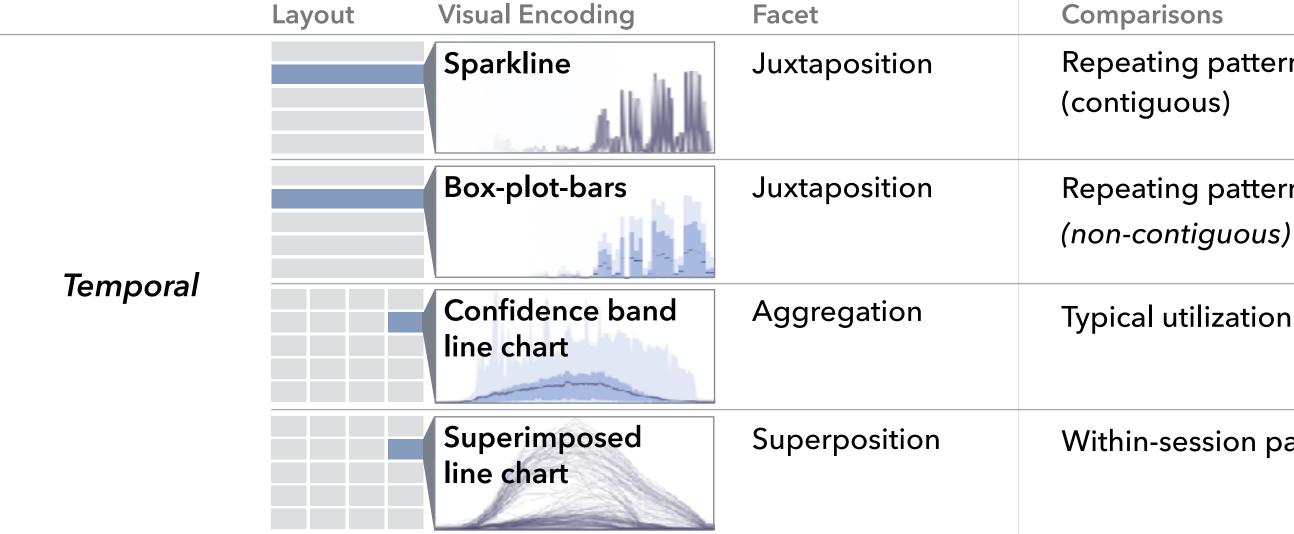
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ession patterns, outliers



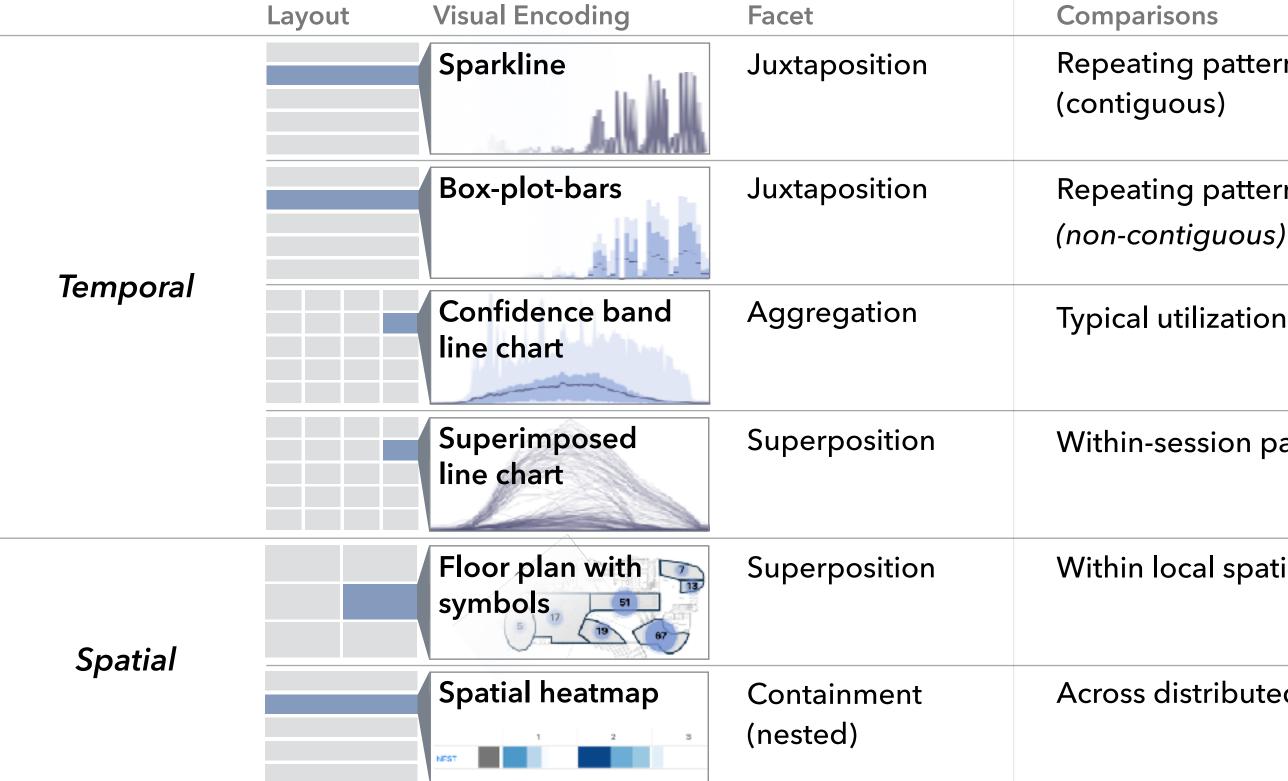


Repeating patterns, trends, outliers

Repeating patterns, trends, outliers

Typical utilization profiles

Within-session patterns, outliers





Repeating patterns, trends, outliers

Repeating patterns, trends, outliers

Typical utilization profiles

Within-session patterns, outliers

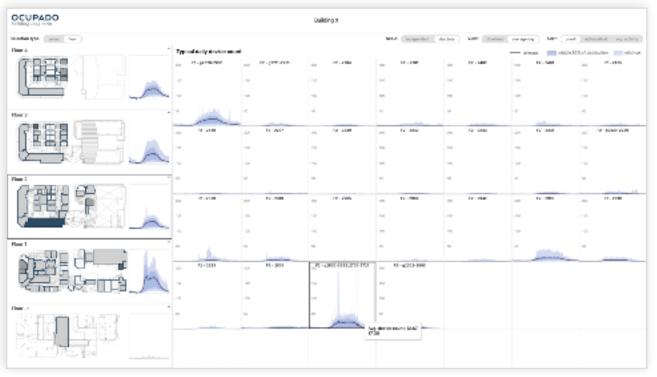
Within local spatial neighborhood

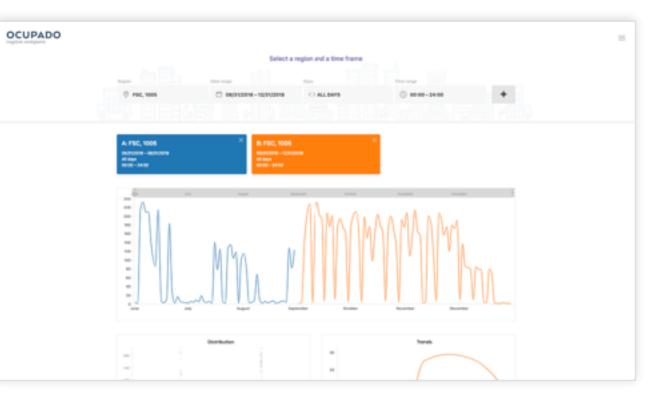
Across distributed regions

Ocupado Interfaces









Campus Explorer

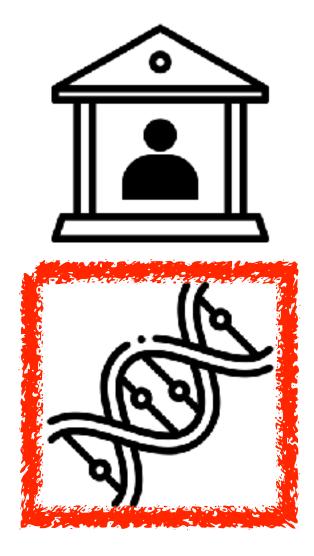
Ocupado Contributions

- Analysis and abstraction of data and tasks for studying space utilization
- Ocupado, a set of visual decision support tools
- Generalizable design choices for visualizing non-trajectory spatiotemporal data relating to large-scale indoor environments

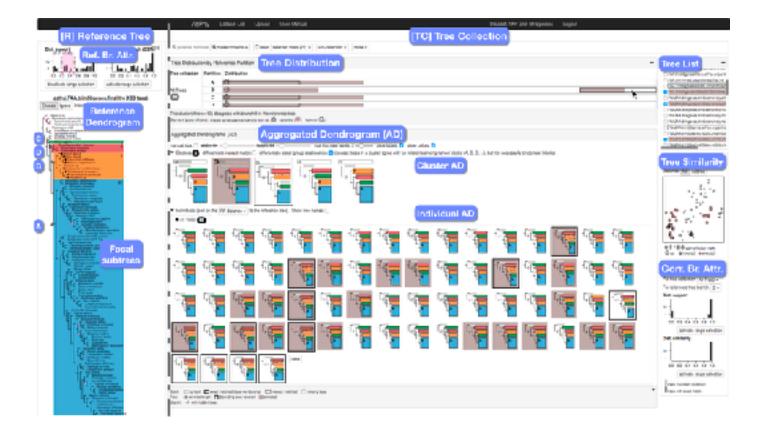
Two design studies

• facilities management

biology



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Aggregated Dendrograms for Visual Comparison Between Many Phylogenetic Trees

http://www.cs.ubc.ca/labs/imager/tr/2019/adview

Aggregated Dendrograms for Visual Comparison Between Many Phylogenetic Trees. Liu, Zhan, Munzner. IEEE Trans. Visualization and Computer Graphics (TVCG) 26(9):2732-2747, 2019. Zipeng Liu

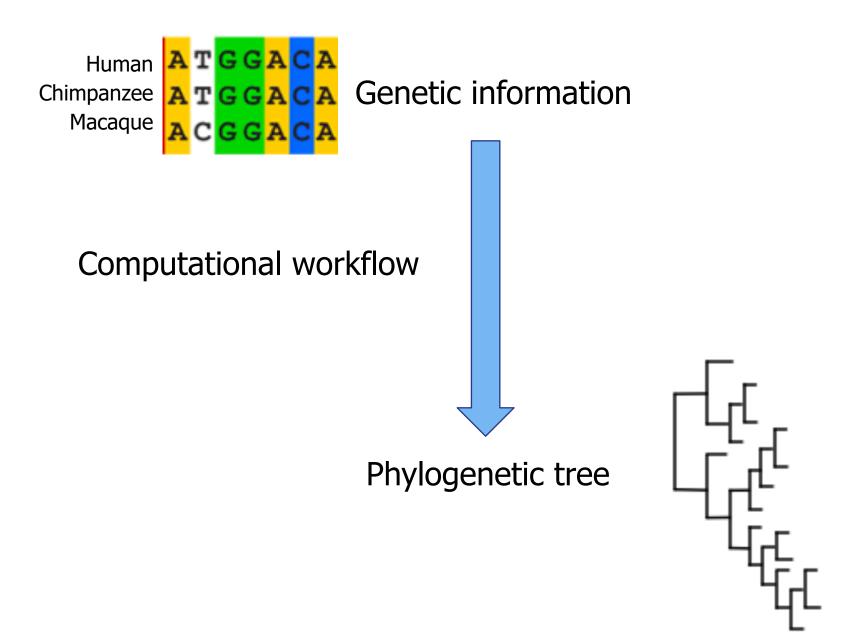


Shing Hei Zhan

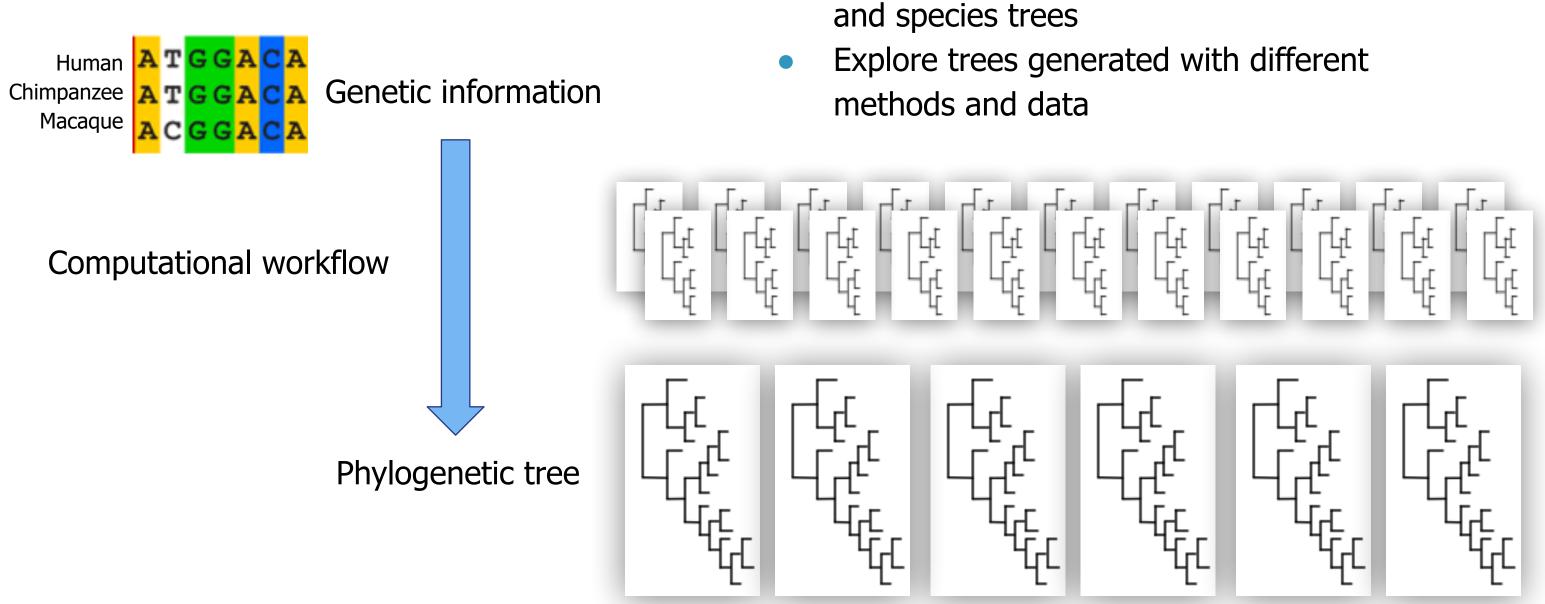


Phylogenetic tree

Evolutionary relationships of organisms



Many phylogenetic trees



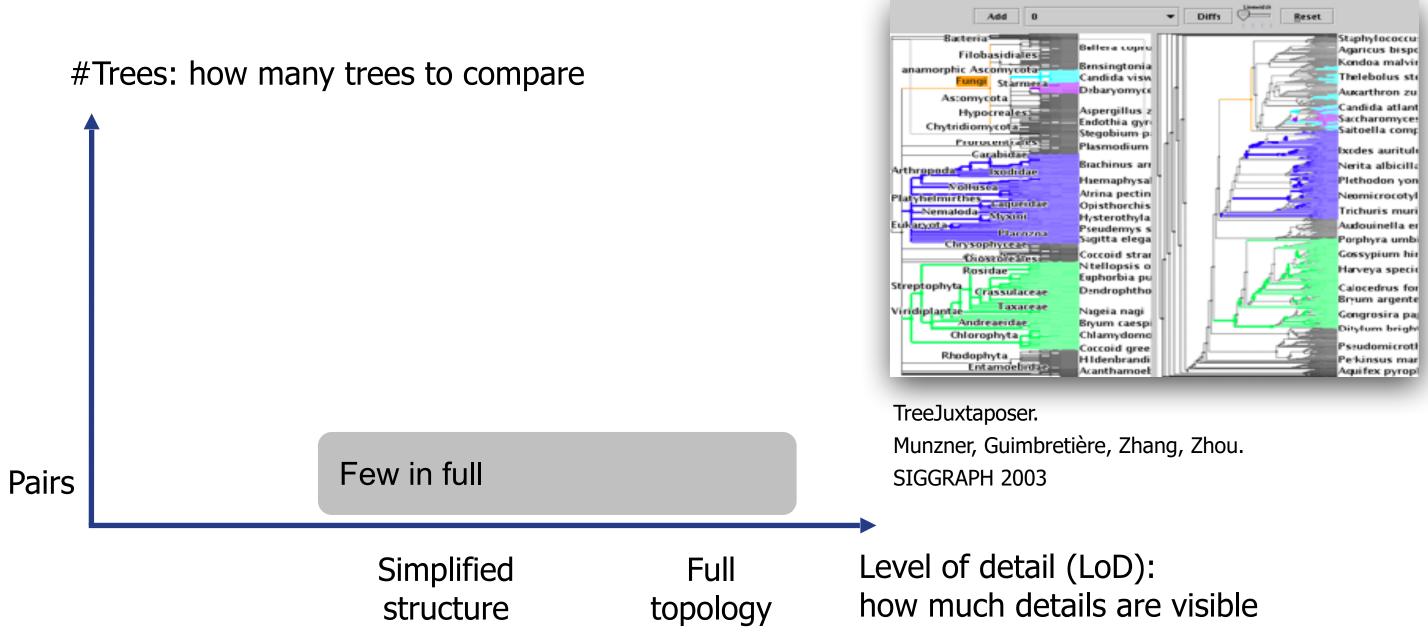
Understand relationships between genes

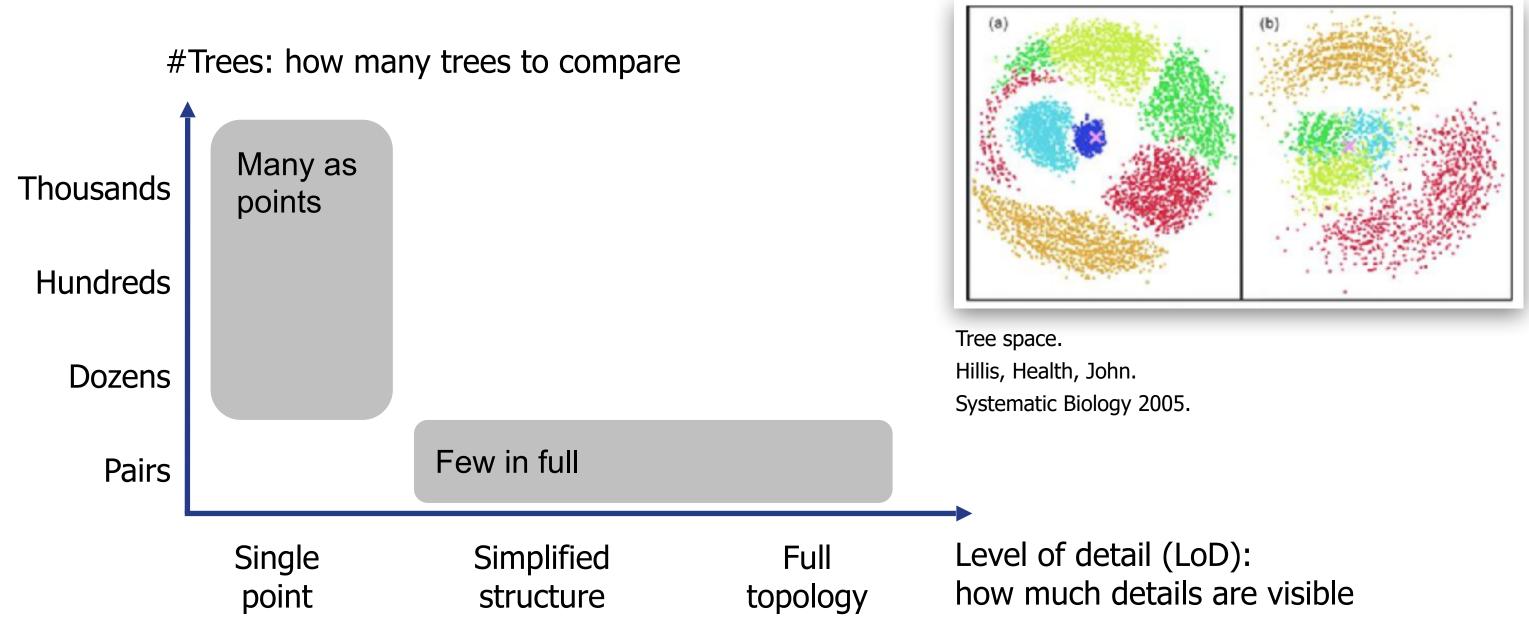
#Trees: how many trees to compare

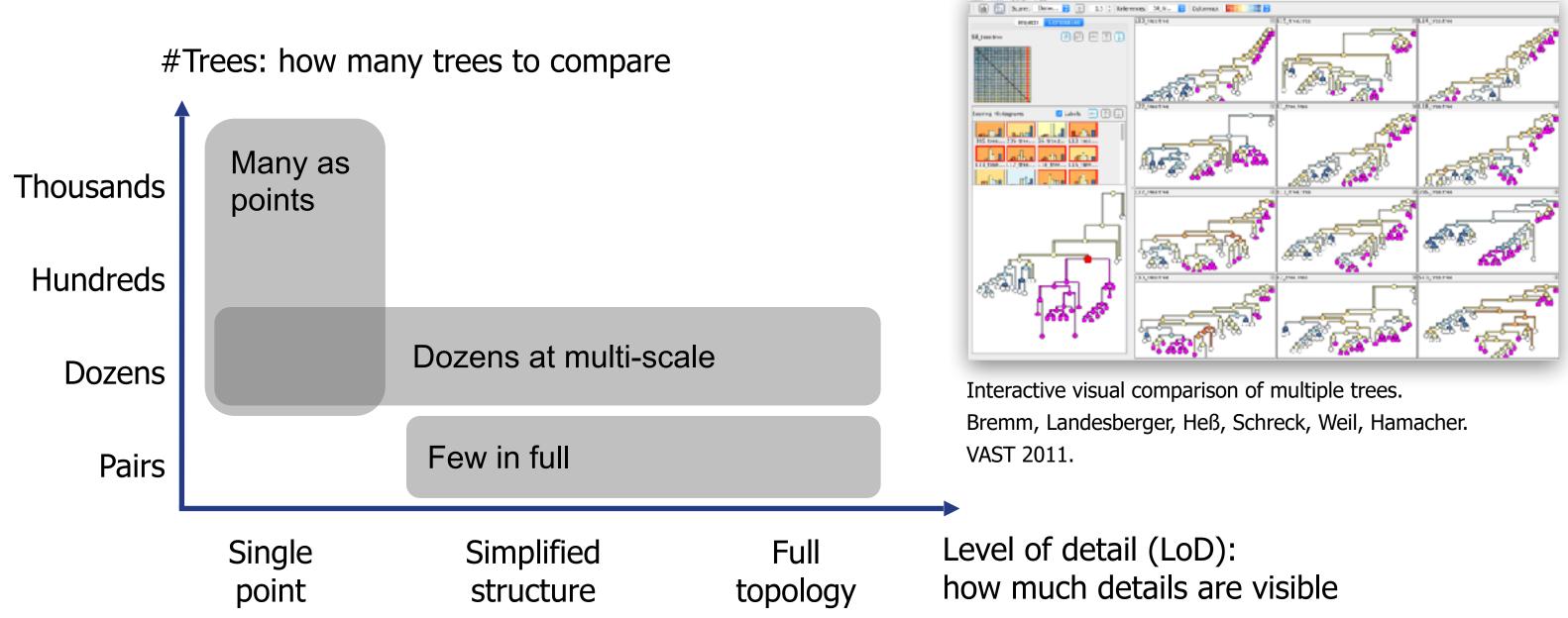


Level of detail (LoD): how much details are visible

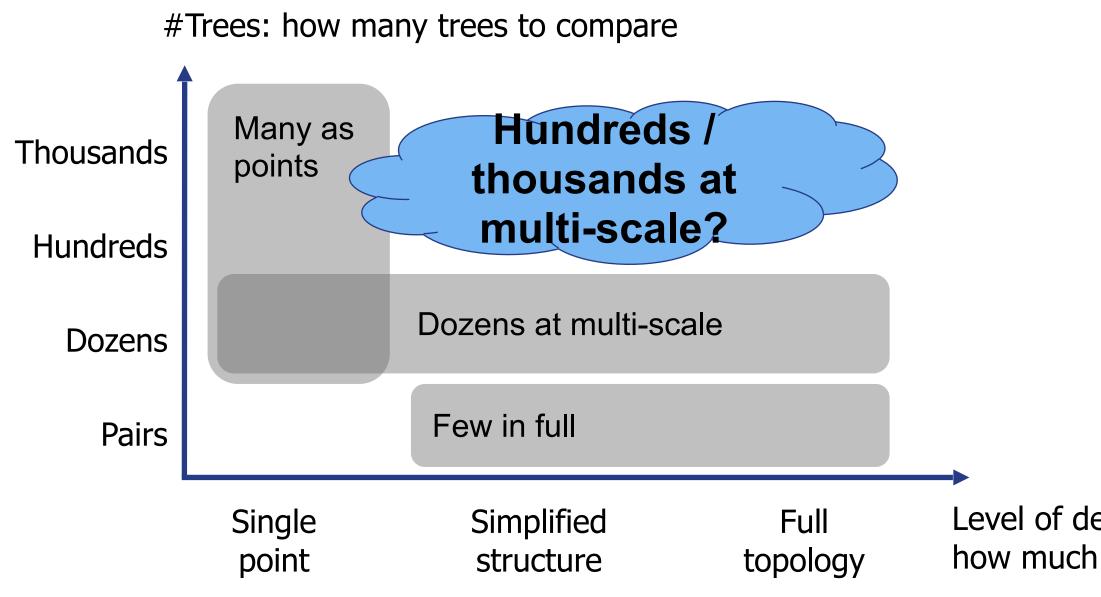








Comparing many phylogenetic trees

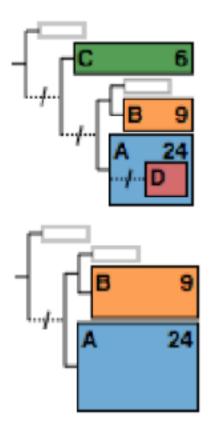


Level of detail (LoD): how much details are visible

Contributions include idiom & algorithm levels • Data and task abstractions for comparison of phylogenetic trees

Contributions include idiom & algorithm levels

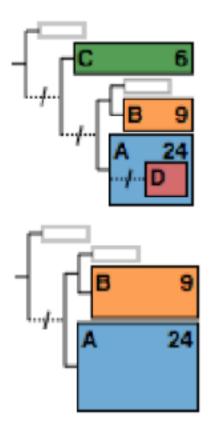
- Data and task abstractions for comparison of phylogenetic trees
- A new visual encoding: **Aggregated Dendrogram**
 - Compact tree representation that focuses on selected subtrees \bigcirc
 - Adapts to available screen space \bigcirc



Contributions include idiom & algorithm levels

- Data and task abstractions for comparison of phylogenetic trees
- A new visual encoding: **Aggregated Dendrogram**
 - Compact tree representation that focuses on selected subtrees \bigcirc
 - Adapts to available screen space \bigcirc
- A multi-view interactive tool: **ADView**
 - Covers multiple levels of details for tree comparison \bigcirc

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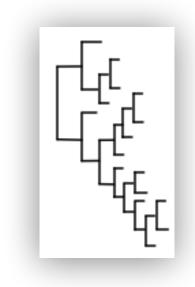


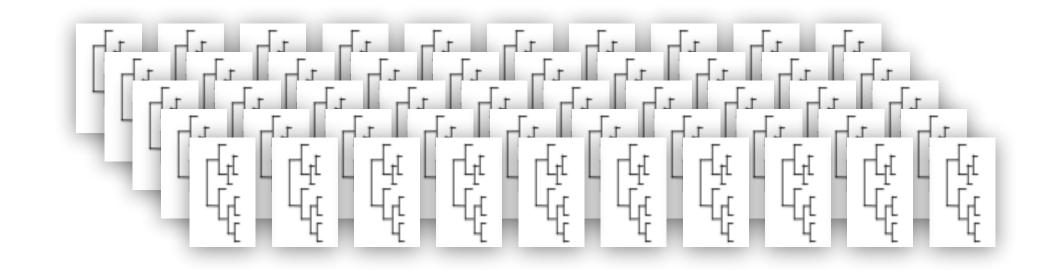
Data & Tasks

- Tree data
- Two crucial tasks



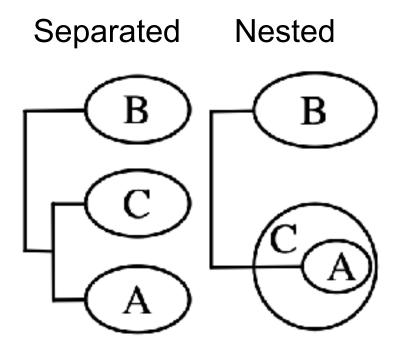
Reference tree vs. Tree collection





Topological relationships between subtrees / leaf nodes

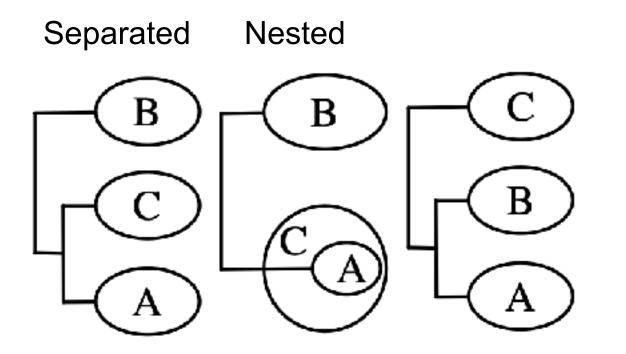
Topological relationships between subtrees / leaf nodes

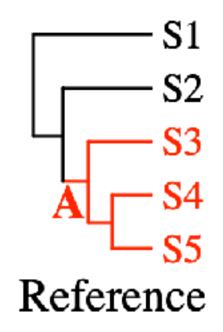


Topological relationships between subtrees / leaf nodes

• Topological distance

Leaf node member reference tree



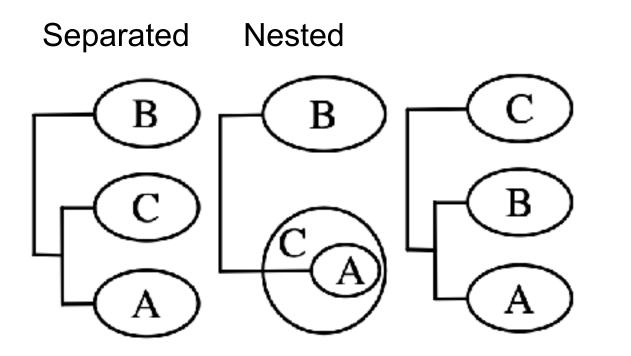


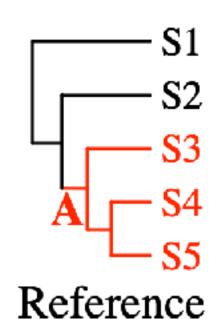
Leaf node memberships compared to

Topological relationships between subtrees / leaf nodes

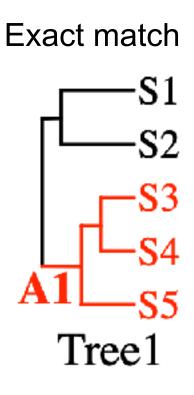
• Topological distance

reference tree





Leaf node memberships compared to



Topological relationships between subtrees / leaf nodes

• Topological distance

Leaf node member reference tree

Separated Nested

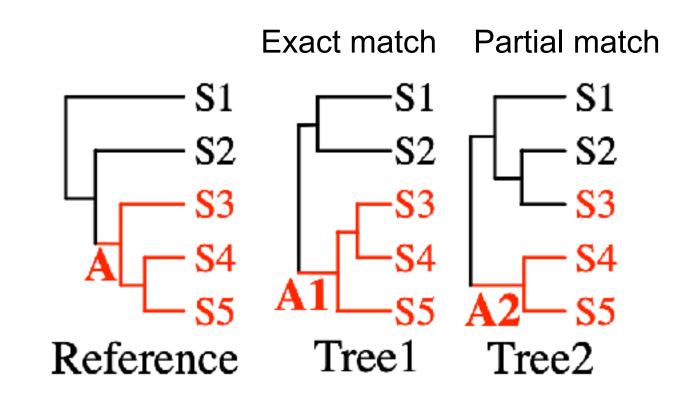
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Leaf node memberships compared to

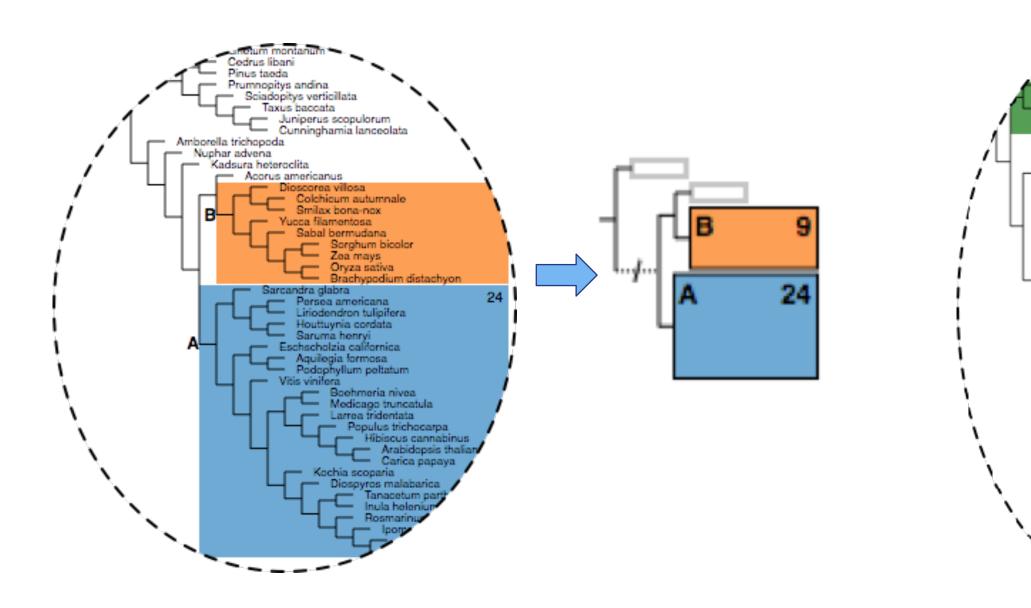
Aggregated Dendrogram (AD)

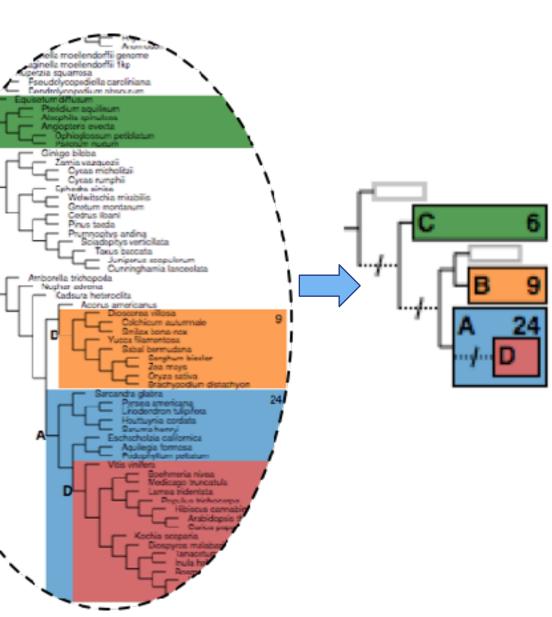
- Intuition
- Visual design

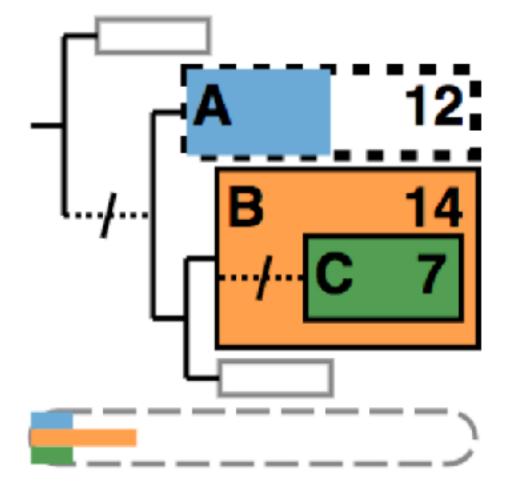


Intuition

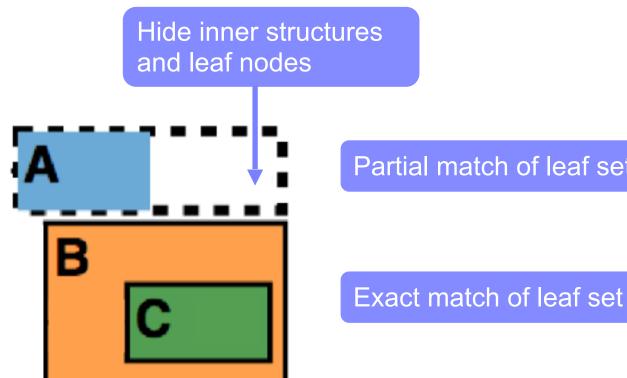
Use glyphs to compress a tree according to user selections







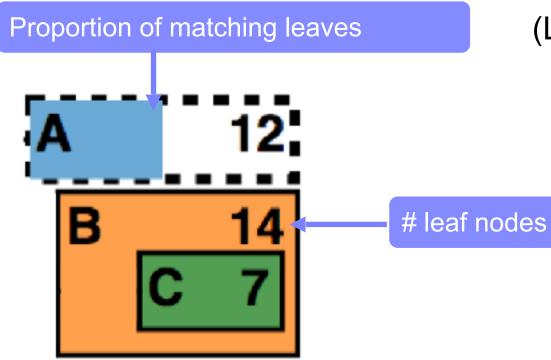
- Focus
 - Selected subtrees \bigcirc



Partial match of leaf set

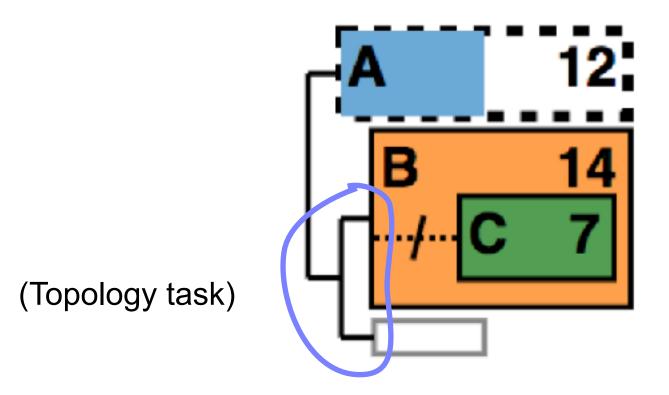
(Leaf task)

- Focus
 - Selected subtrees \bigcirc

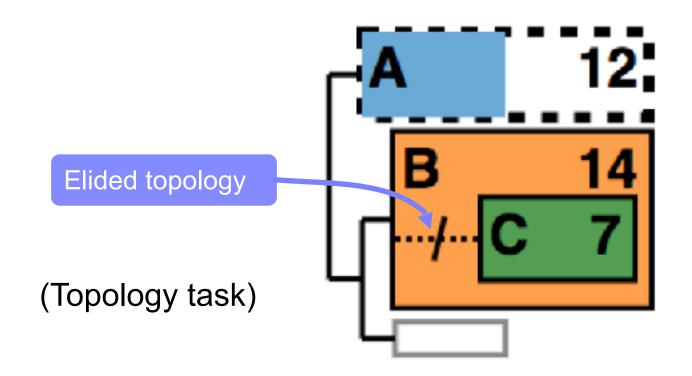


(Leaf task)

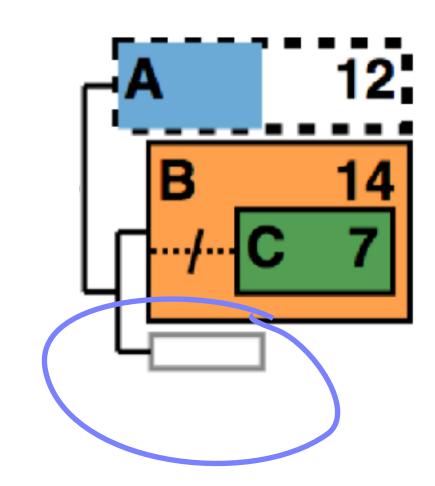
- Focus
 - Selected subtrees
 - Topological relationships between them



- Focus
 - Selected subtrees
 - Topological relationships between them

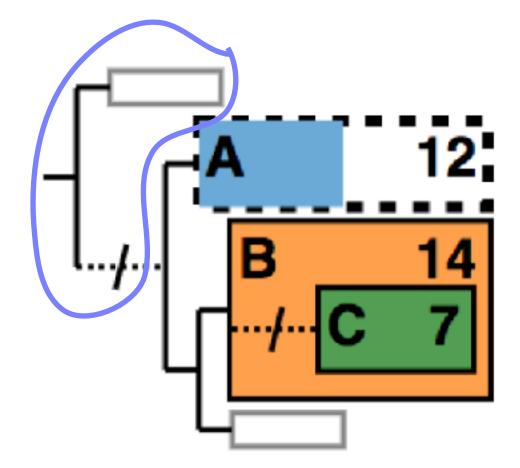


- Focus
 - Selected subtrees
 - Topological relationships between them
- Context
 - Neighboring subtrees



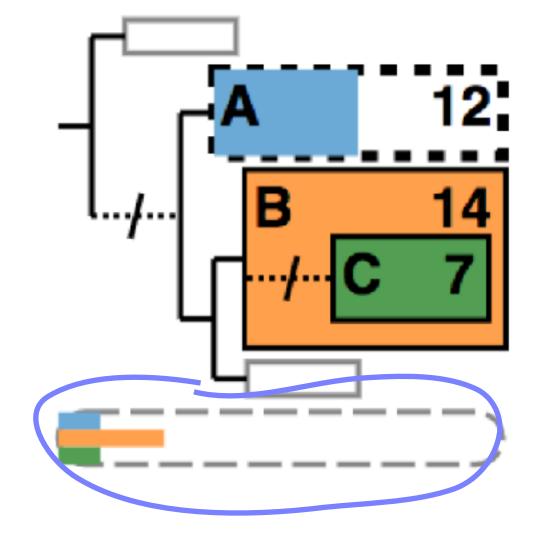
Visual design: focus + context

- Focus
 - Selected subtrees
 - Topological relationships between them
- Context
 - Neighboring subtrees
 - Upstream topology and root



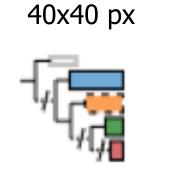
Visual design: focus + context

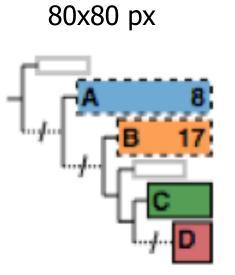
- Focus
 - Selected subtrees
 - Topological relationships between them
- Context
 - Neighboring subtrees
 - Upstream topology and root
 - Missing leaf nodes



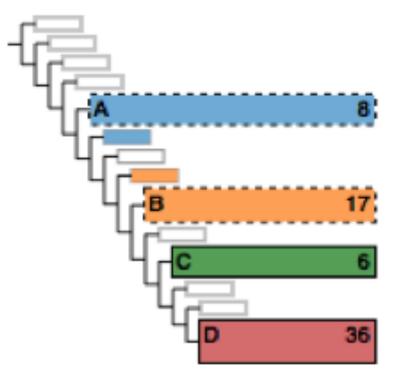
Visual design: algorithm adapts to space

- Show more info when space permitted
 - Labels
 - #leaf nodes
 - Neighboring blocks

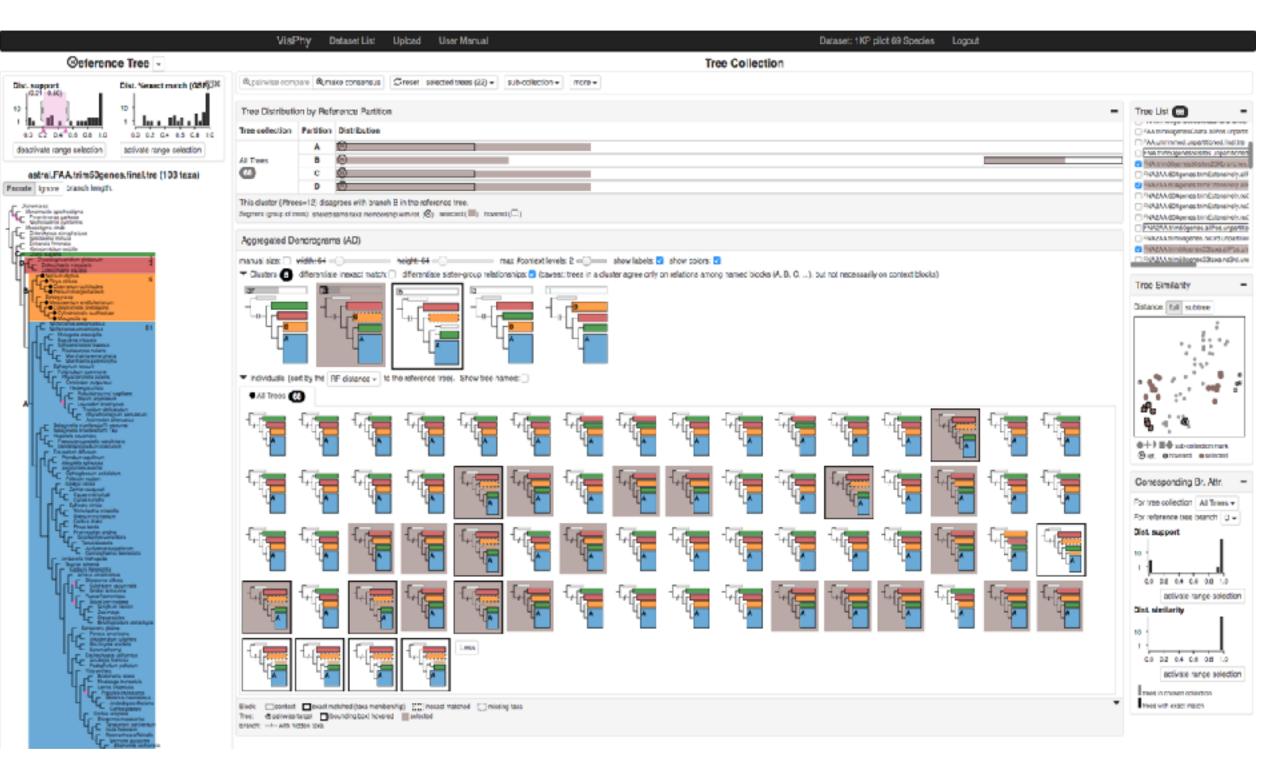




160x160 px

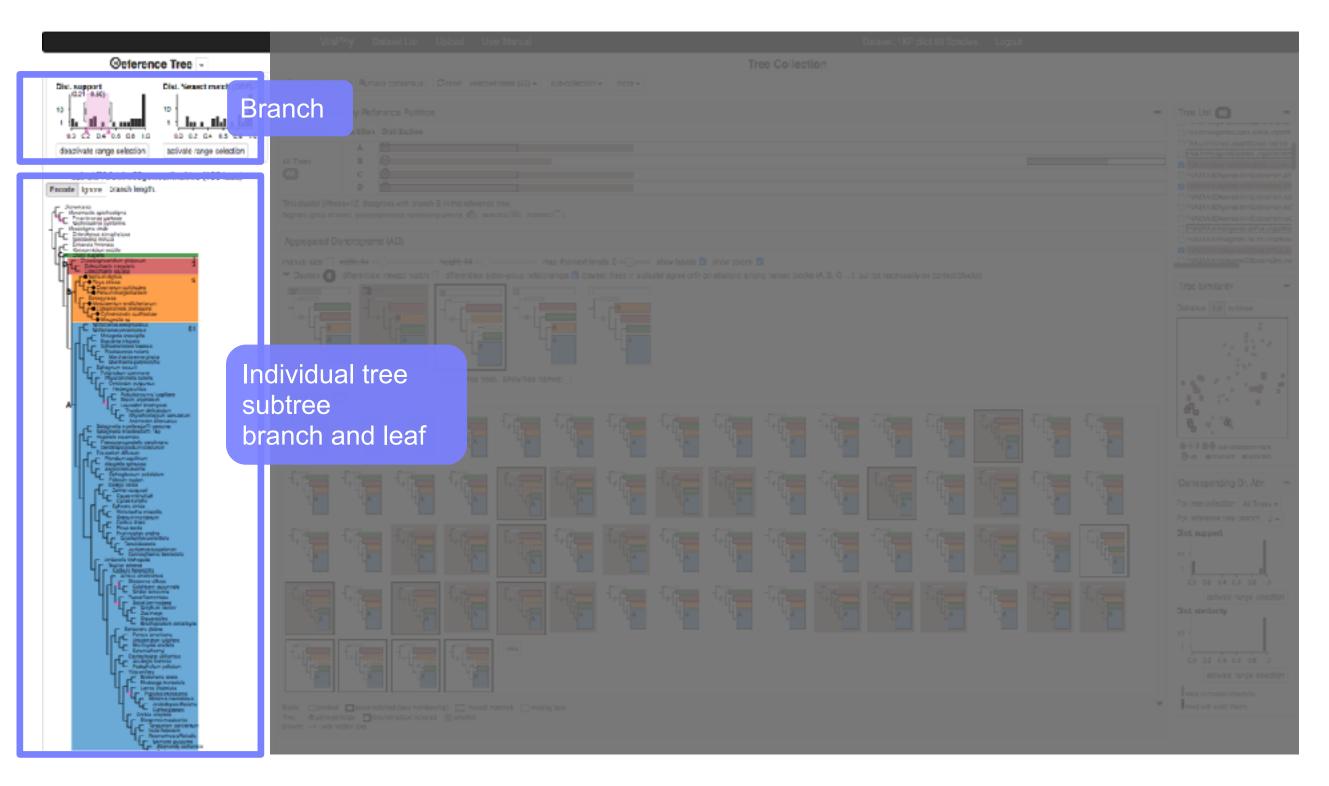


ADView Interface: Multi-level structure across views





Multi-level structure across views



Interface walkthrough: tree collection main views



Interface walkthrough: tree collection aux. views

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* individuals (set by the RF distance - Its the advector tree. Show bee had		Tree c		
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Aggregated Dendrograms for Visual Comparison between Many Phylogenetic Trees

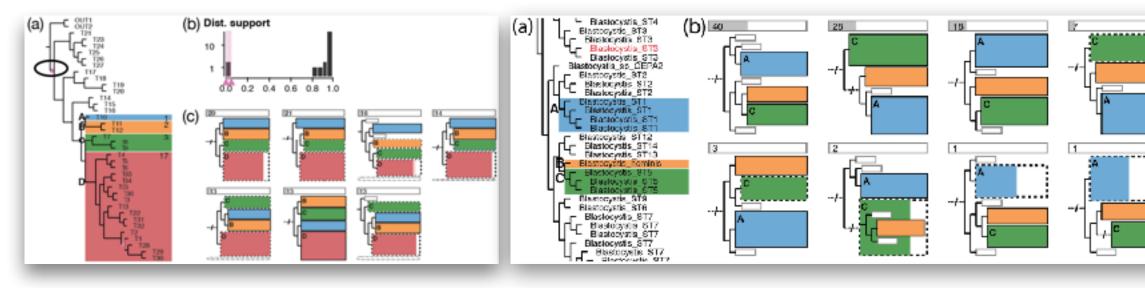
Zipeng Liu, Shing Hei Zhan, Tamara Munzner

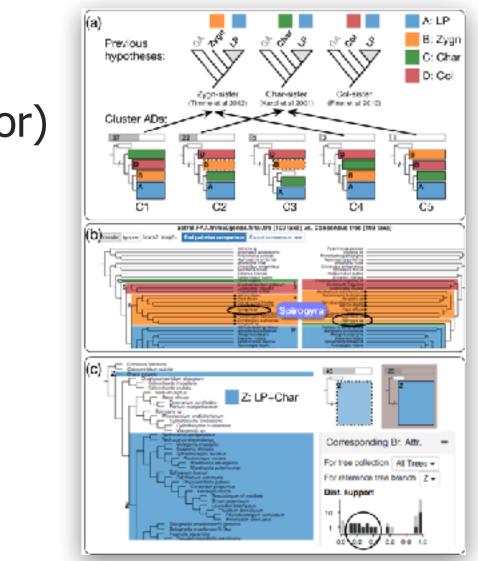
Validation with many biologists

- Work closely with a biology PhD student (second author)
- Demos, interviews and discussions ${ \bullet }$
 - 10 biologists at different times throughout project \bigcirc

Validation with many biologists

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 - 10 biologists at different times throughout project
- User study sessions
 - 5 biologists
 - Using their own datasets

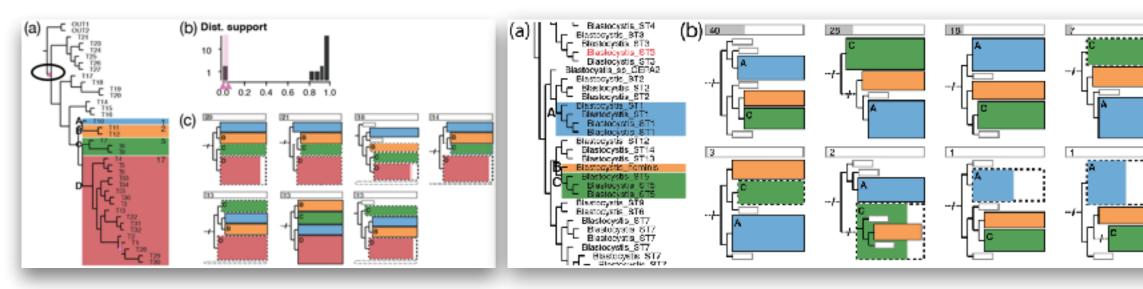


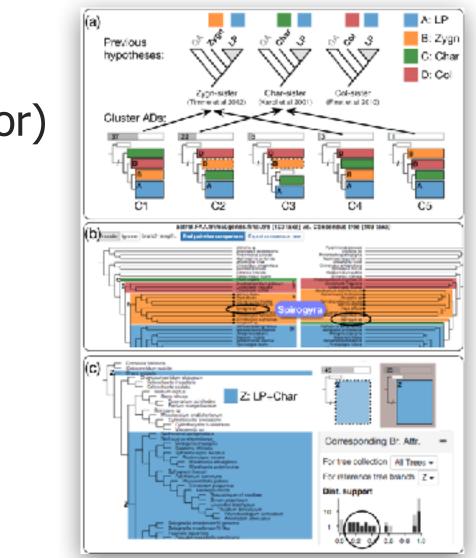


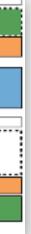


Validation with many biologists

- Work closely with a biology PhD student (second author)
- Demos, interviews and discussions
 - 10 biologists at different times throughout project
- User study sessions
 - 5 biologists
 - Using their own datasets
- Biologists confirmed
 - Validity of data and task abstractions
 - Utility of ADView



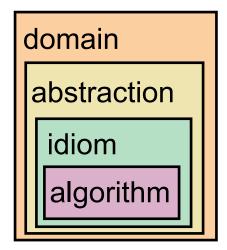




Problem-driven visualization through design studies

- methodology matters
 - -identify abstractions
 - crucial & difficult, iterative process
 - select appropriate idioms
 - or create new ones if necessary

- two examples
 - different domains
 - -different methods

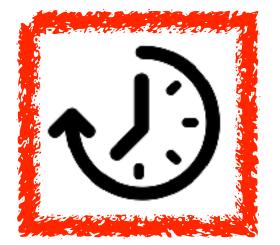




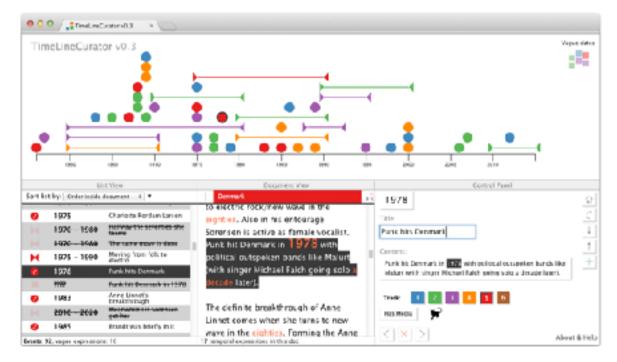
imperfect models

One case study of visualizing imperfect models

• NLP for temporal data







TimeLineCurator Interactive Authoring of Visual Timelines from Unstructured Text

http://about.timelinecurator.org http://timelinecurator.org

TimeLineCurator: Interactive Authoring of Visual Timelines from Unstructured Text. Fulda, Brehmer, Munzner. IEEE Trans. Visualization and Computer Graphics (Proc IEEE VAST 2015) 22(1):300-309, 2015.

Johanna Fulda @jofu_



Matthew Brehmer @mattbrehmer



Tamara Munzner @tamaramunzner

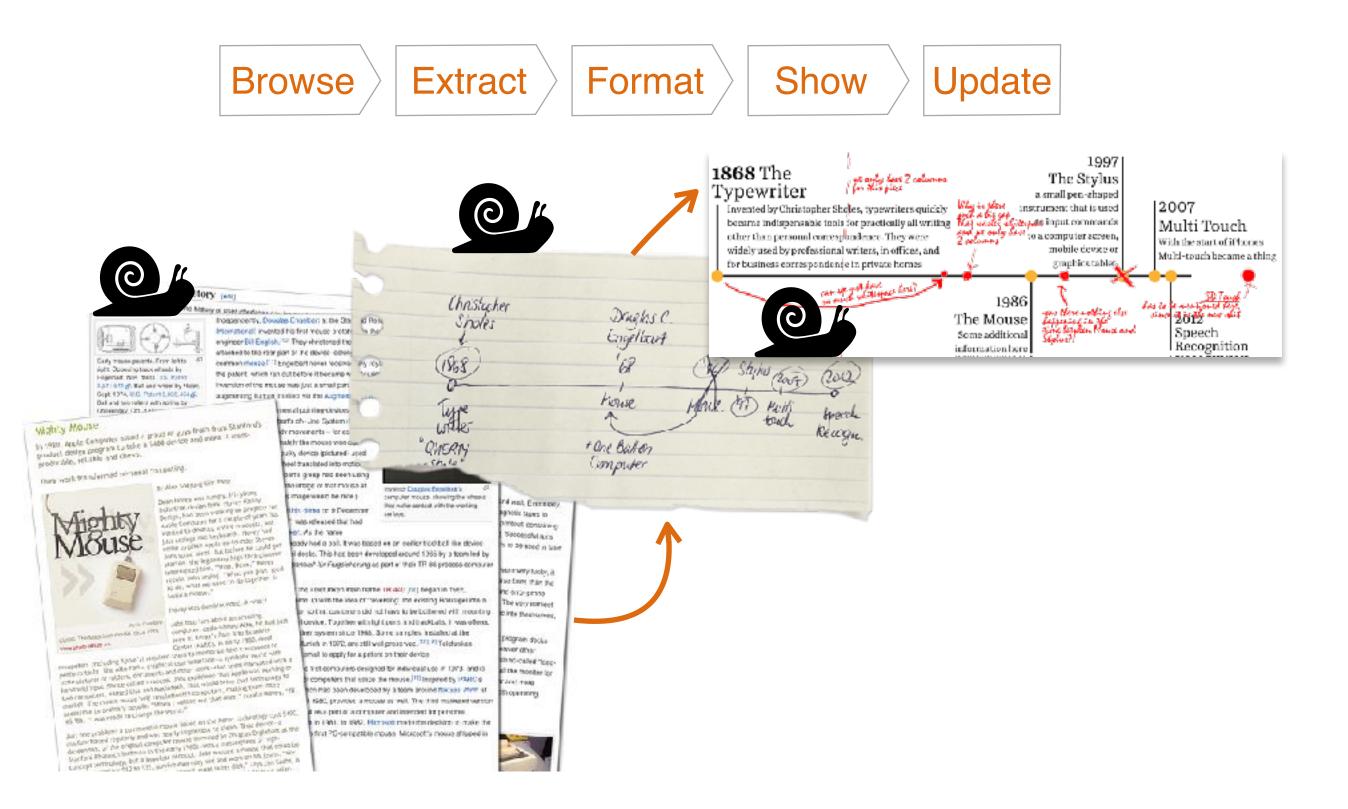


TimeLineCurator al & browser-based

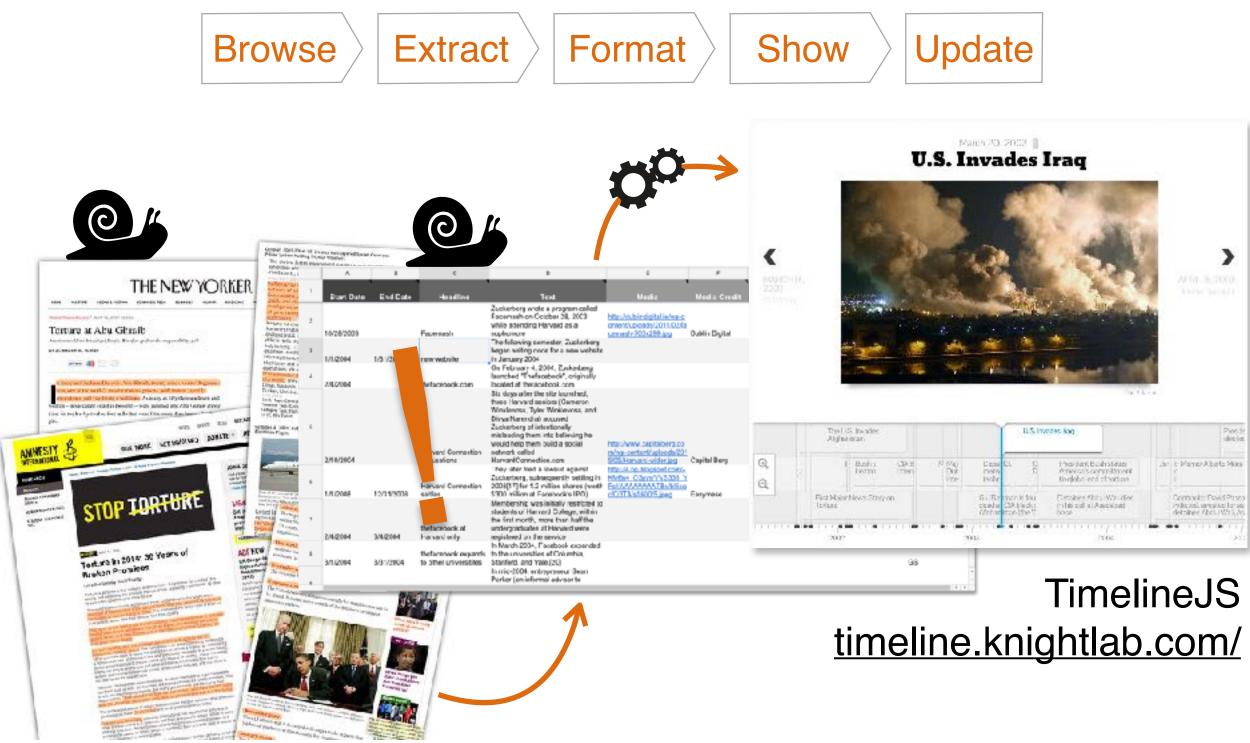




Manual creation process

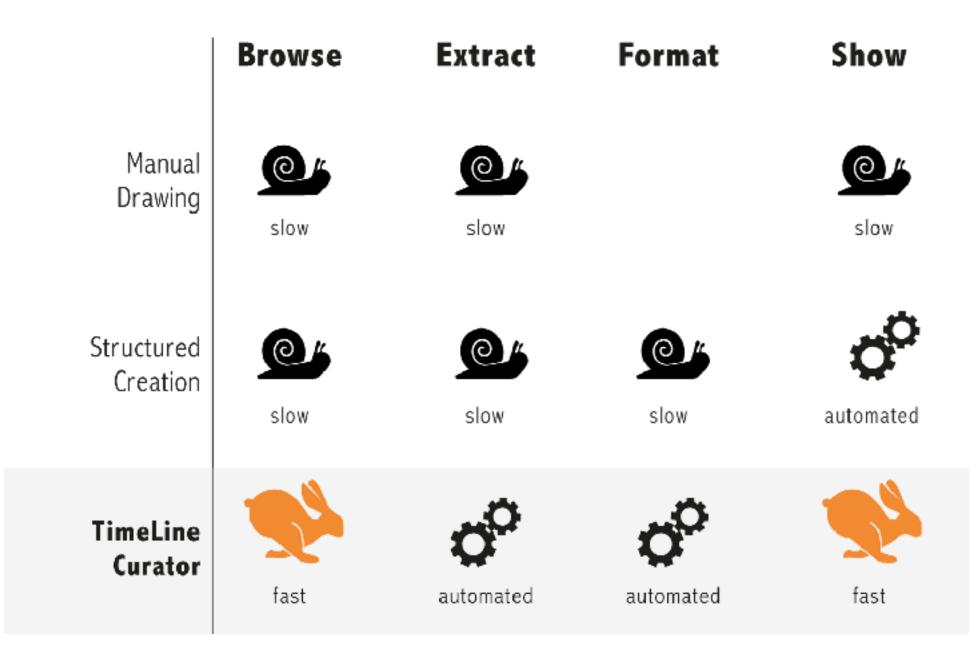


Structured creation process



Timeline authoring model

• time required for each task



Update



slow



fast



fast

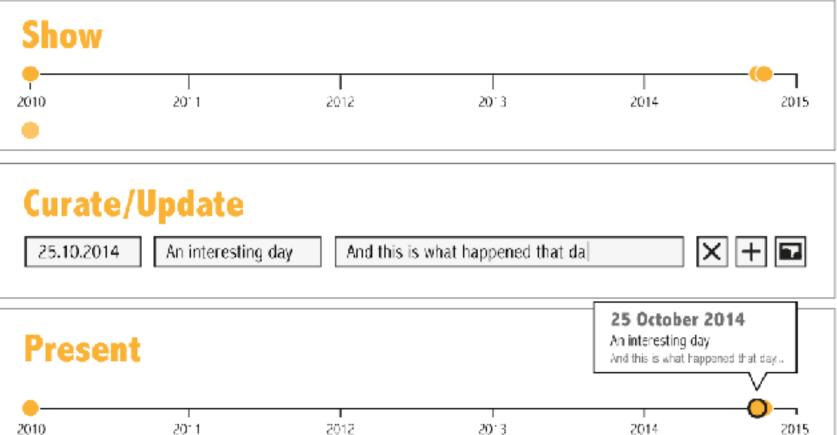
The general case for curation

- build for human in the loop as continuing need
 - -automatic processing to accelerate not replace
 - -assume computational results good but not perfect
 - for the indefinite future!
 - -visual feedback to accelerate



Extract

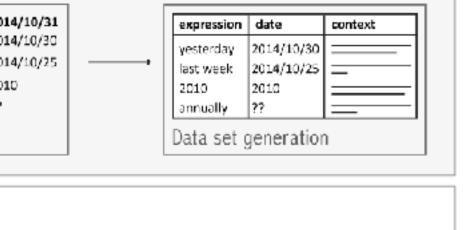
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<u>Architecture</u>

Format



The importance of being brisk

- cool use case: eureka moment
 - -success: enable what was impossible before
 - -vis tools for new insights & discoveries
- workhorse use case: workflow speedup
 - -success: vis tools accelerate your prior workflow
 - sometimes enables the previously infeasible
- TLC use cases
 - -started with speedup use case, for presentation
 - make this doc into a timeline now!
 - -two other use cases nudge towards exploration
 - comparison between multiple timelines
 - speculative browsing



TimeLineCurator: Speculative Browsing





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Come talk!

- encourage meeting with me to get advice/feedback before final present -chance to get feedback while you can still act on it
 - -optional, not mandatory
 - -wise to schedule in advance by email
 - can't meet with all 14 teams in next week office hours, or in last few days!