Problem-Driven Visualization Through Design Studies

Tamara Munzner

Department of Computer Science University of British Columbia



<u>@tamaramunzner</u>

VINCI 2021 Keynote Sep 7 2021, virtual / Potsdam

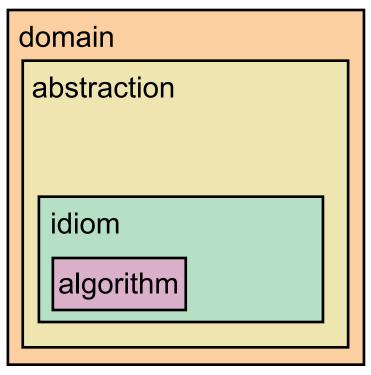






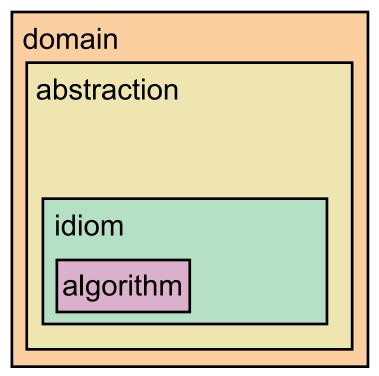


http://www.cs.ubc.ca/~tmm/talks.html#vinci21



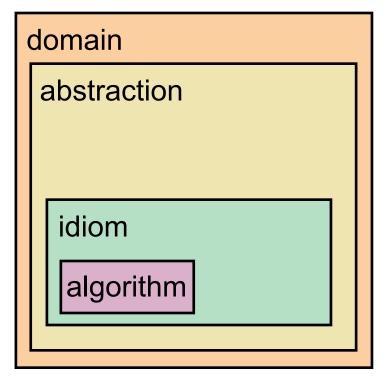
[A Nested Model of Visualization Design and Validation. Munzner. IEEETVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - -who are the target users?



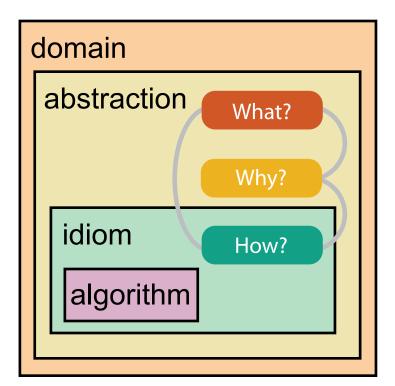
[A Nested Model of Visualization Design and Validation. Munzner. IEEETVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis



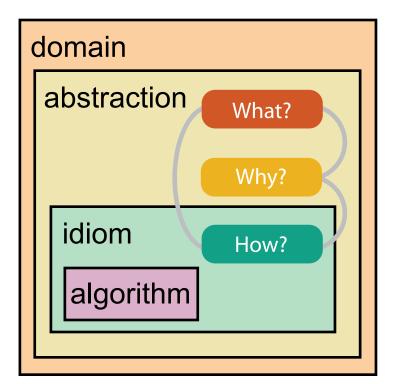
[A Nested Model of Visualization Design and Validation. Munzner. IEEETVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - -who are the target users?
- abstraction
 - translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction



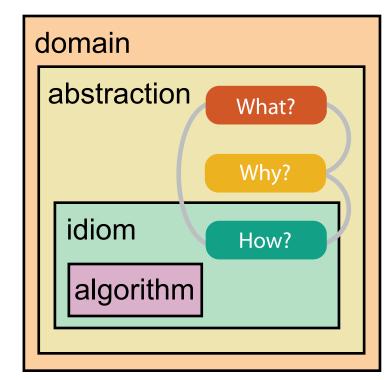
[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form



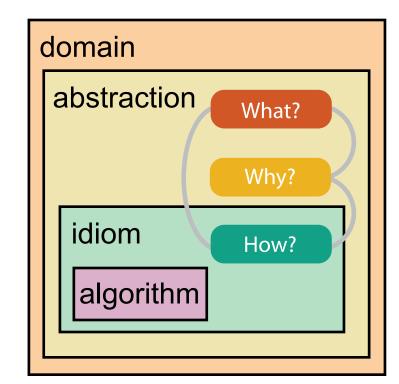
[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form
 - -why is the user looking at it? task abstraction



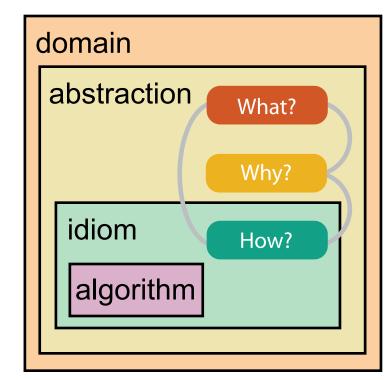
[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form
 - -why is the user looking at it? task abstraction
- idiom
 - how is it shown?



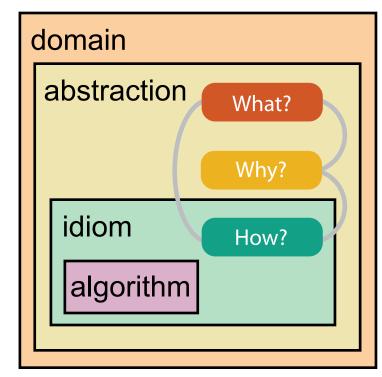
[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form
 - -why is the user looking at it? task abstraction
- idiom
 - how is it shown?
 - visual encoding idiom: how to draw



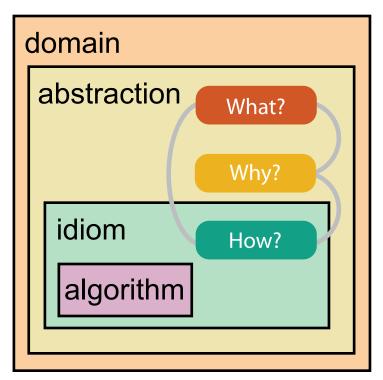
[A Nested Model of Visualization Design and Validation. Munzner. IEEETVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form
 - -why is the user looking at it? task abstraction
- idiom
 - how is it shown?
 - visual encoding idiom: how to draw
 - interaction idiom: how to manipulate



[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

- domain situation
 - who are the target users?
- abstraction
 - -translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - often don't just draw what you're given: transform to new form
 - -why is the user looking at it? task abstraction
- idiom
 - how is it shown?
 - visual encoding idiom: how to draw
 - interaction idiom: how to manipulate
- algorithm
 - -efficient computation



[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

different ways to get it wrong at each level

different ways to get it wrong at each level



Domain situation

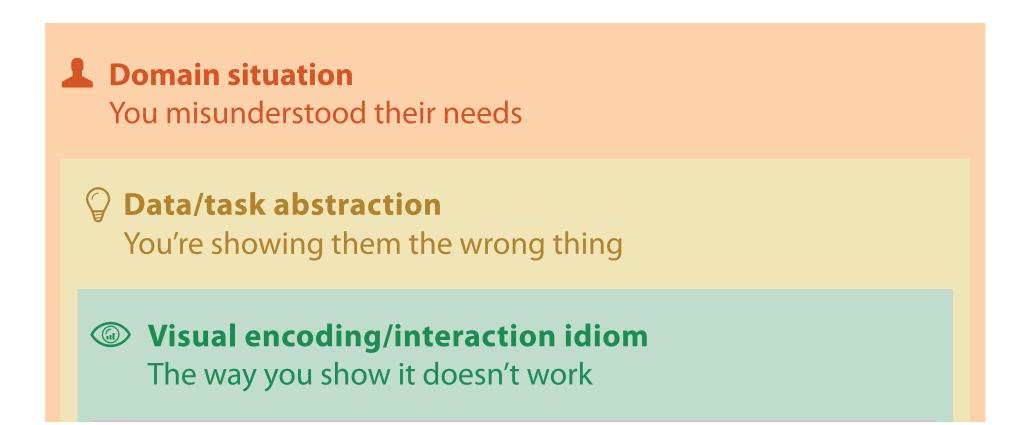
You misunderstood their needs

different ways to get it wrong at each level



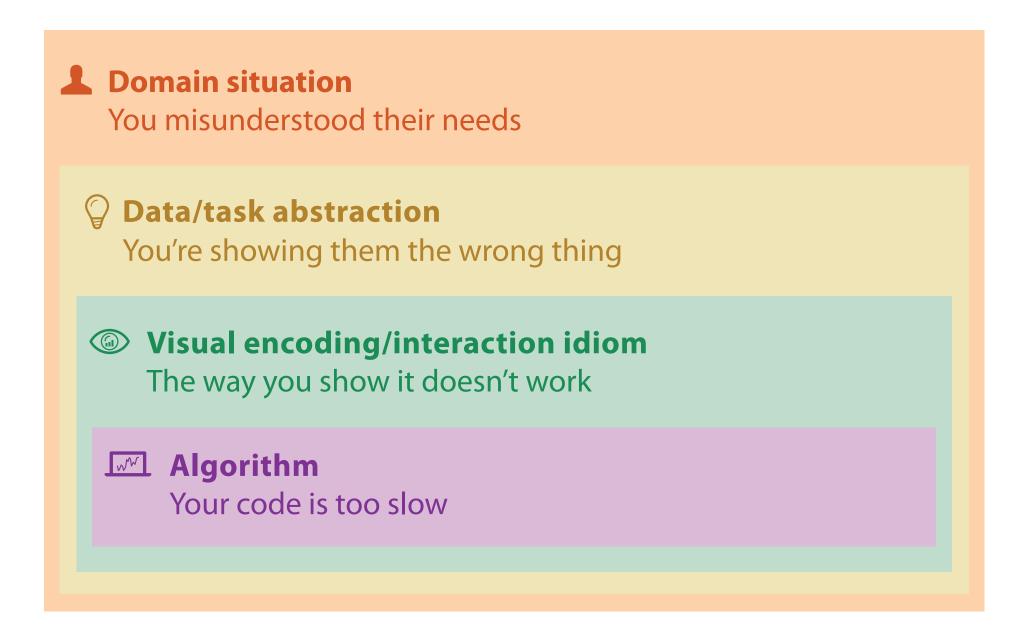
Data/task abstraction
 You're showing them the wrong thing

different ways to get it wrong at each level



[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

different ways to get it wrong at each level



[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



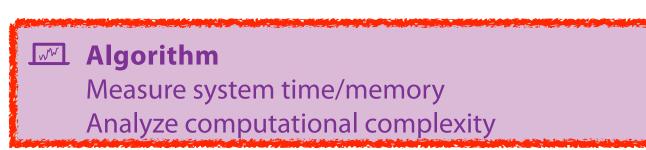
computer science



Algorithm

Measure system time/memory
Analyze computational complexity

computer science



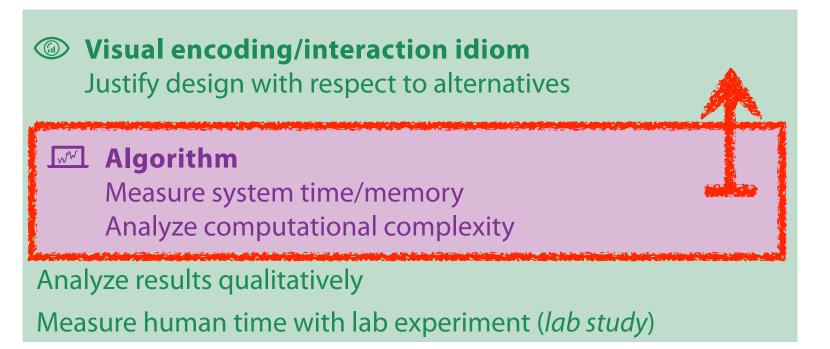


[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

design

computer science

cognitive psychology



technique-driven work

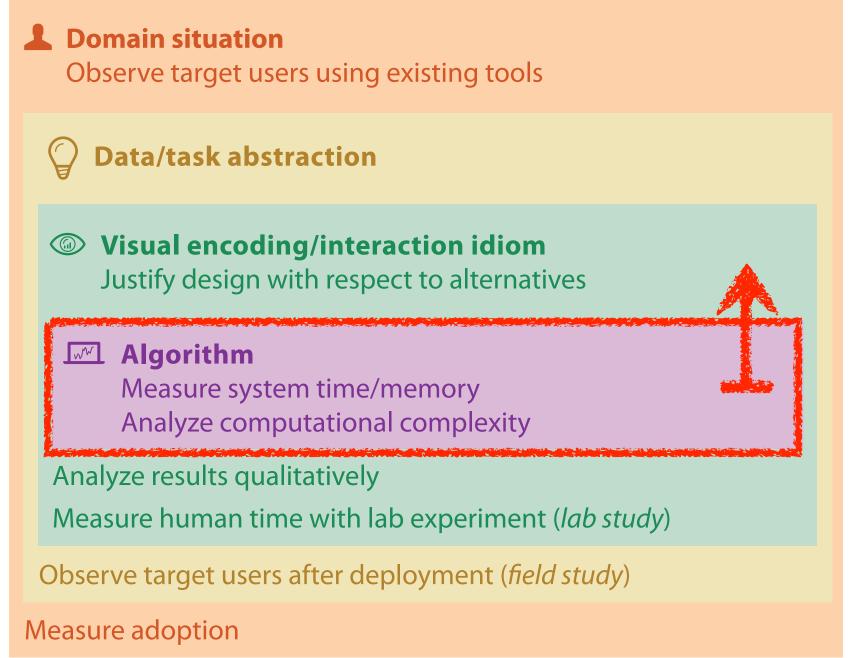
anthropology/ ethnography

design

computer science

cognitive psychology

anthropology/ ethnography



technique-driven work

Domain situation anthropology/ Observe target users using existing tools ethnography Data/task abstraction Wisual encoding/interaction idiom design Justify design with respect to alternatives **Algorithm** computer Measure system time/memory science Analyze computational complexity cognitive Analyze results qualitatively psychology Measure human time with lab experiment (*lab study*) Observe target users after deployment (*field study*) anthropology/ ethnography Measure adoption

problem-driven technique-driven work

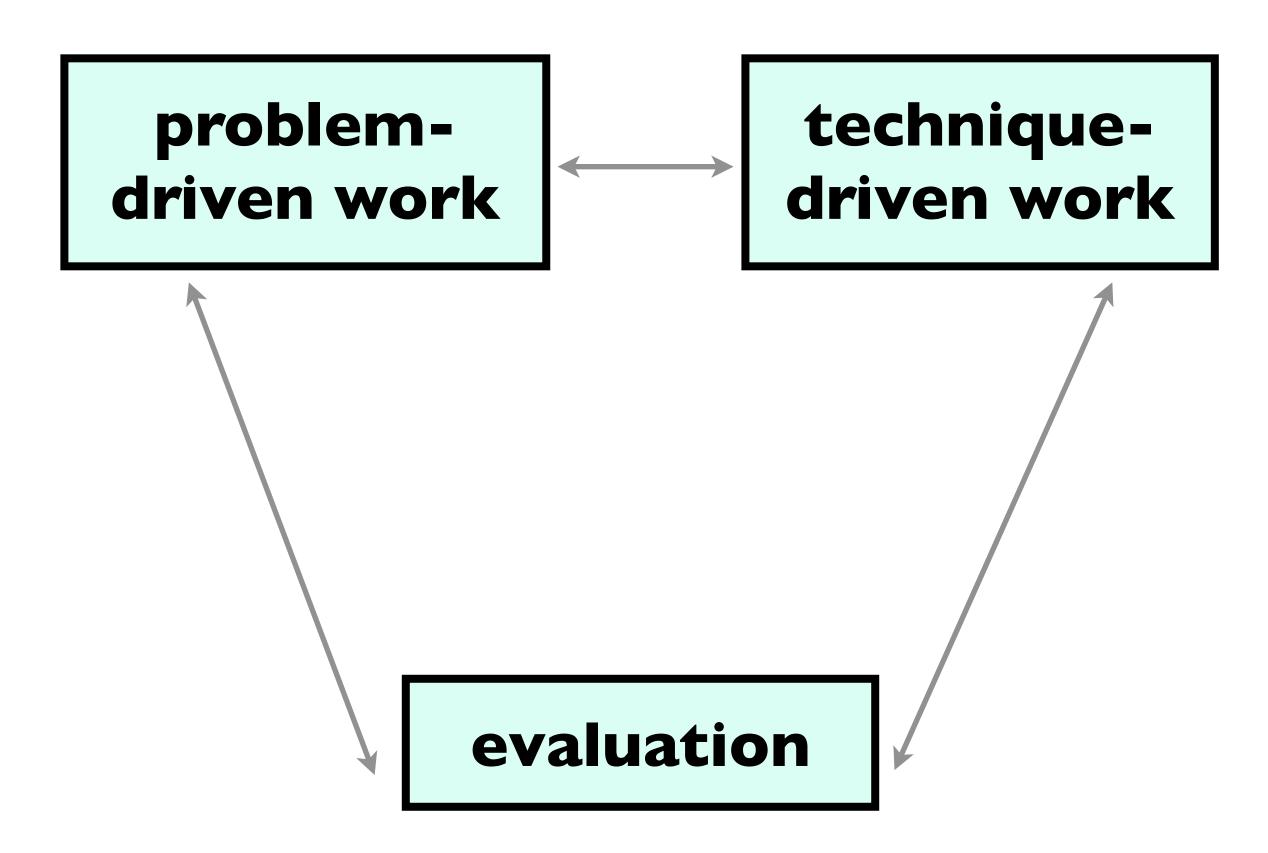
avoid mismatches between level and validation

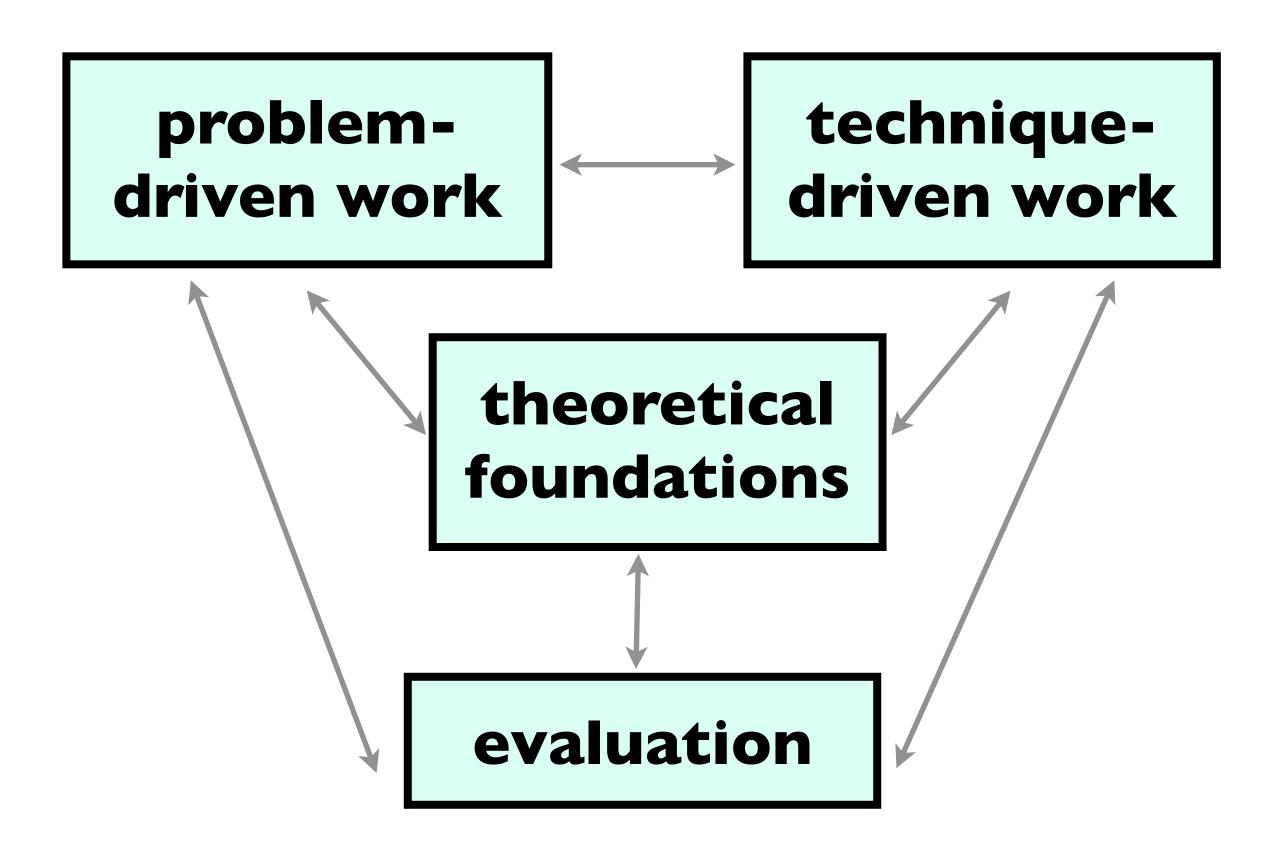
Domain situation anthropology/ Observe target users using existing tools ethnography Data/task abstraction Wisual encoding/interaction idiom design Justify design with respect to alternatives **Algorithm** computer Measure system time/memory science work Analyze computational complexity cognitive Analyze results qualitatively psychology Measure human time with lab experiment (*lab study*) Observe target users after deployment (*field study*) anthropology/ ethnography Measure adoption

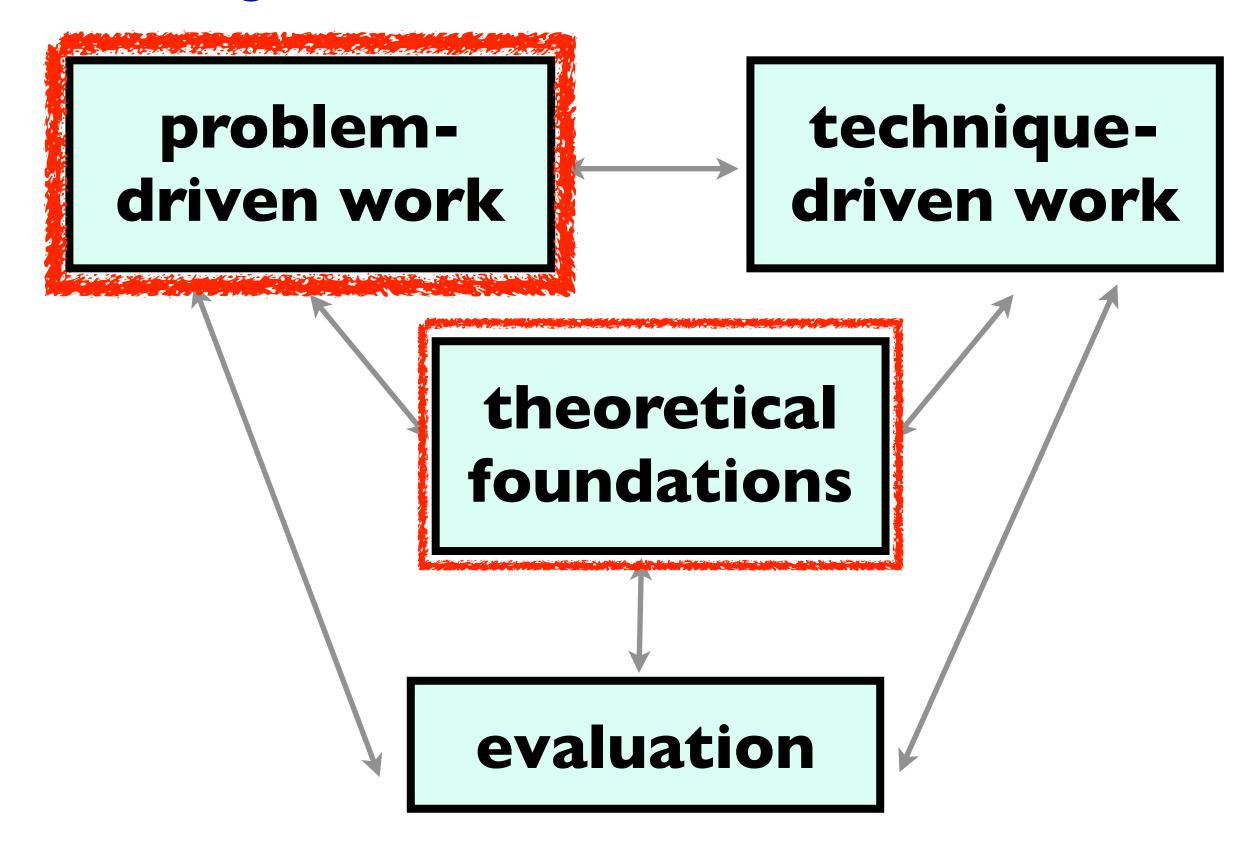
problem-driven technique-driven

problemdriven work

problemdriven work driven work







Problem-driven visualization: Design studies

Problem driven visualization: Design studies

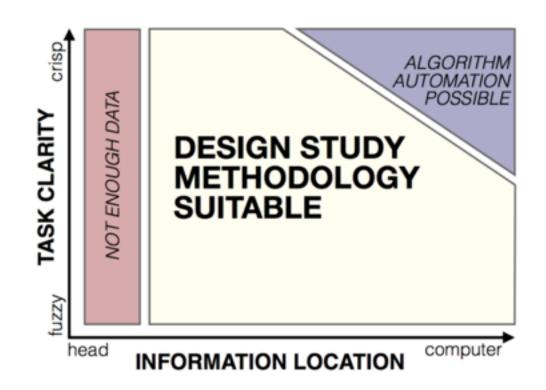
"A design study is a project in which visualization researchers analyze a specific real-world problem faced by domain experts..."

Problem driven visualization: Design studies

"A design study is a project in which visualization researchers analyze a specific real-world problem faced by domain experts, design a visualization system that supports solving this problem..."

Problem driven visualization: Design studies

"A design study is a project in which visualization researchers analyze a specific real-world problem faced by domain experts, design a visualization system that supports solving this problem, validate the design, and reflect about lessons learned in order to refine visualization design guidelines."



Michael Sedlmair



Miriah Meyer



Design Study Methodology

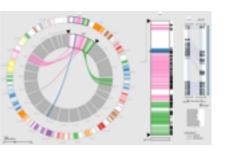
Reflections from the Trenches and from the Stacks

http://www.cs.ubc.ca/labs/imager/tr/2012/dsm/

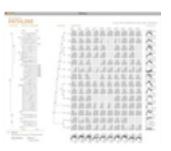
Lessons learned from the trenches: 20+ between us



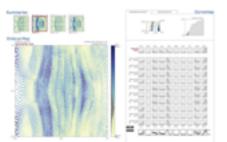
Cerebral genomics



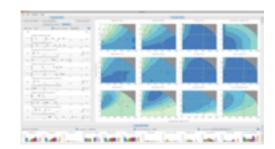
MizBee genomics



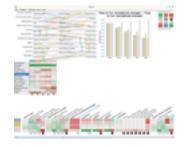
Pathline genomics



MulteeSum genomics



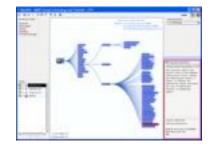
Vismon fisheries management



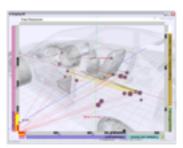
QuestVis sustainability



WiKeVis in-car networks



MostVis in-car networks



Car-X-Ray in-car networks



ProgSpy2010 in-car networks



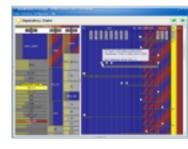
RelEx in-car networks



Cardiogram in-car networks



AutobahnVis in-car networks



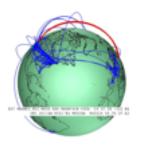
VisTra in-car networks



Constellation linguistics



LibVis cultural heritage



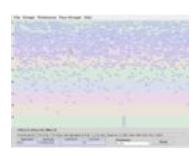
Caidants multicast



SessionViewer web log analysis



LiveRAC server hosting

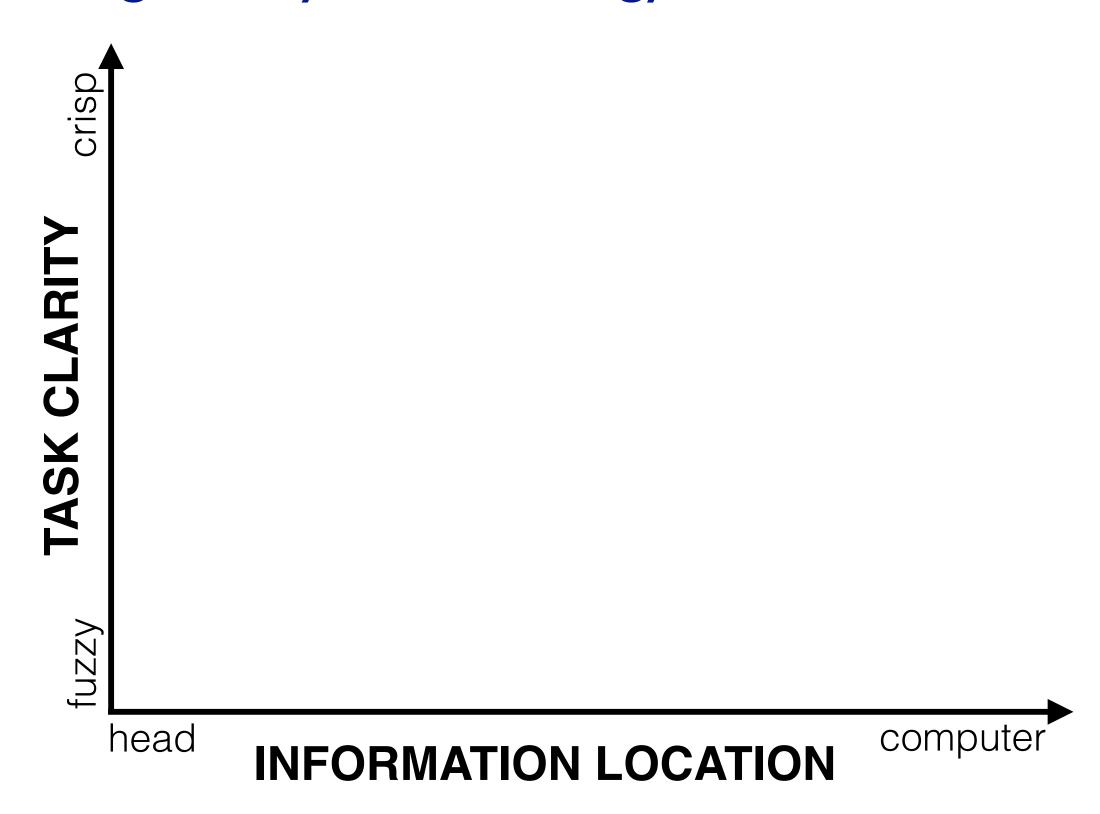


PowerSetViewer data mining

Design study methodology: definitions

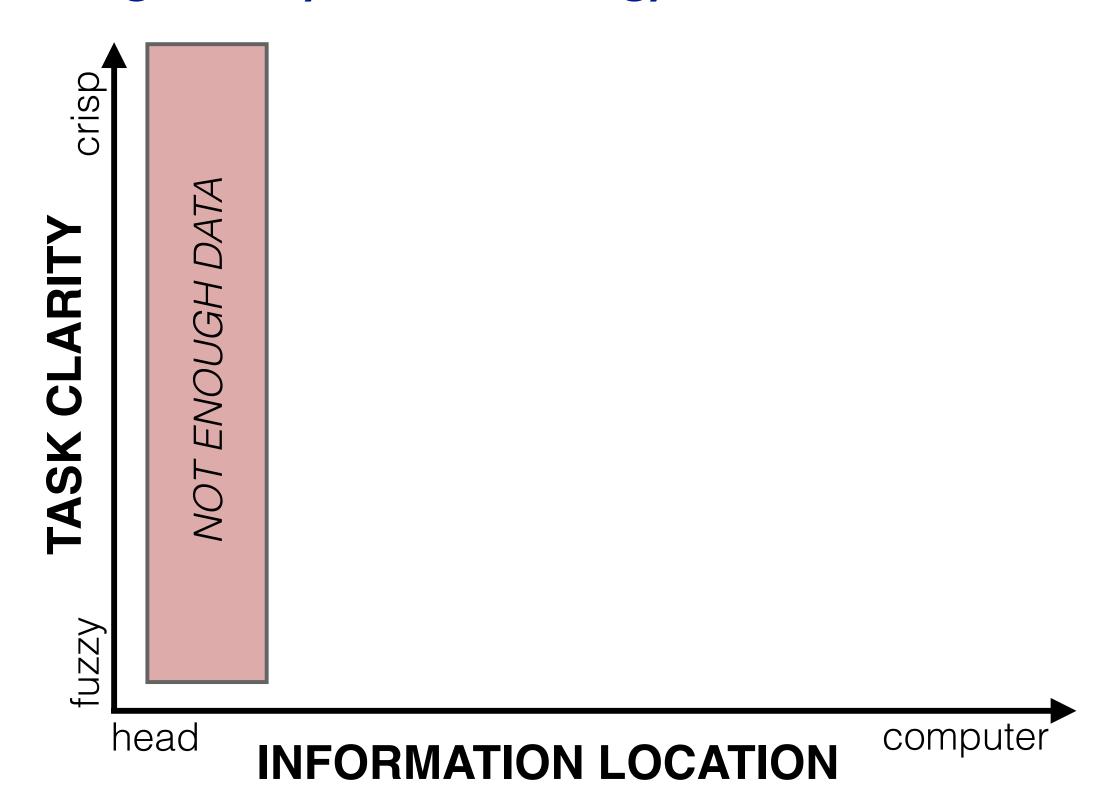


Design study methodology: definitions



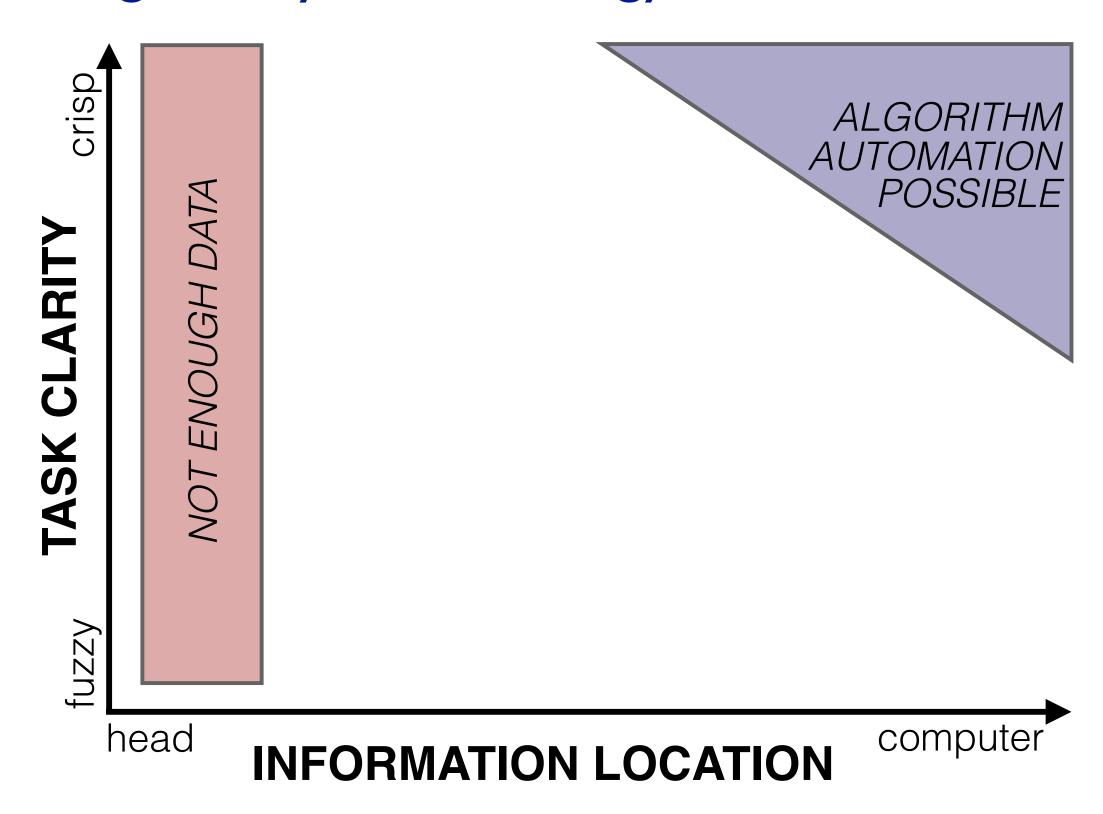
[Design Study Methodology: Reflections from the Trenches and the Stacks. Sedlmair, Meyer & Munzner. IEEE TVCG 18(12): 2431-2440, 2012 (Proc. InfoVis 2012).]

Design study methodology: definitions



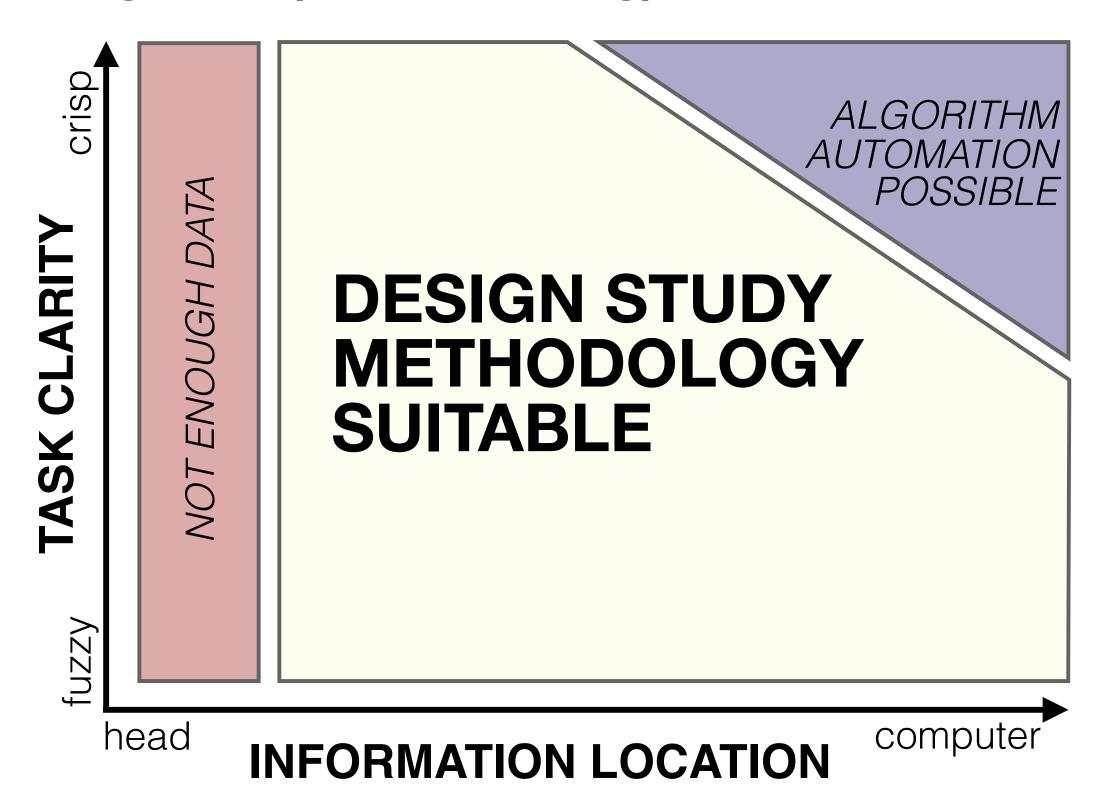
[Design Study Methodology: Reflections from the Trenches and the Stacks. Sedlmair, Meyer & Munzner. IEEE TVCG 18(12): 2431-2440, 2012 (Proc. InfoVis 2012).]

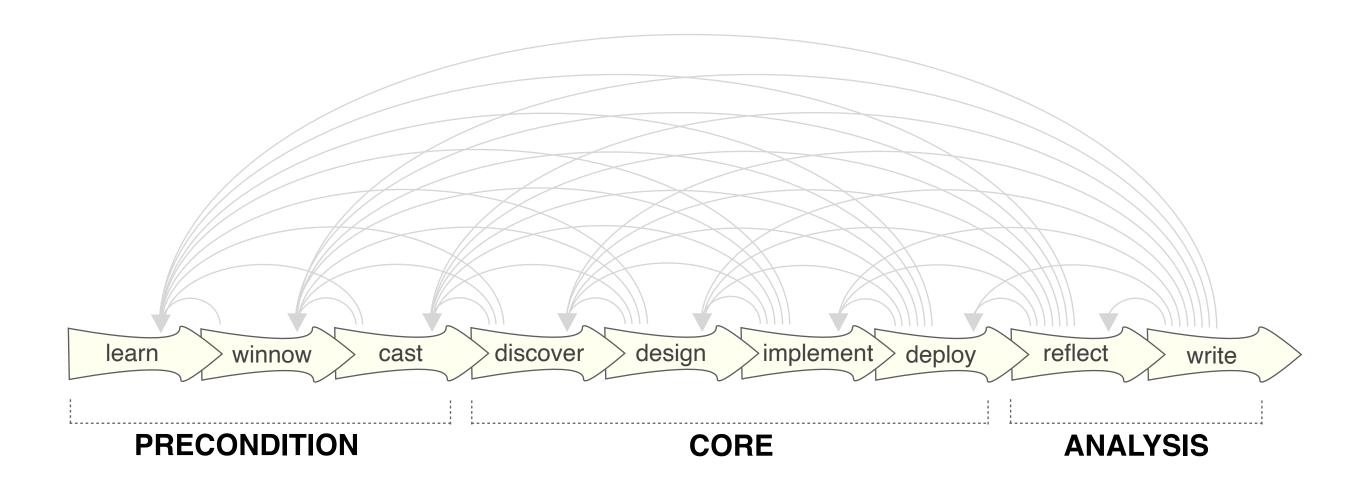
Design study methodology: definitions



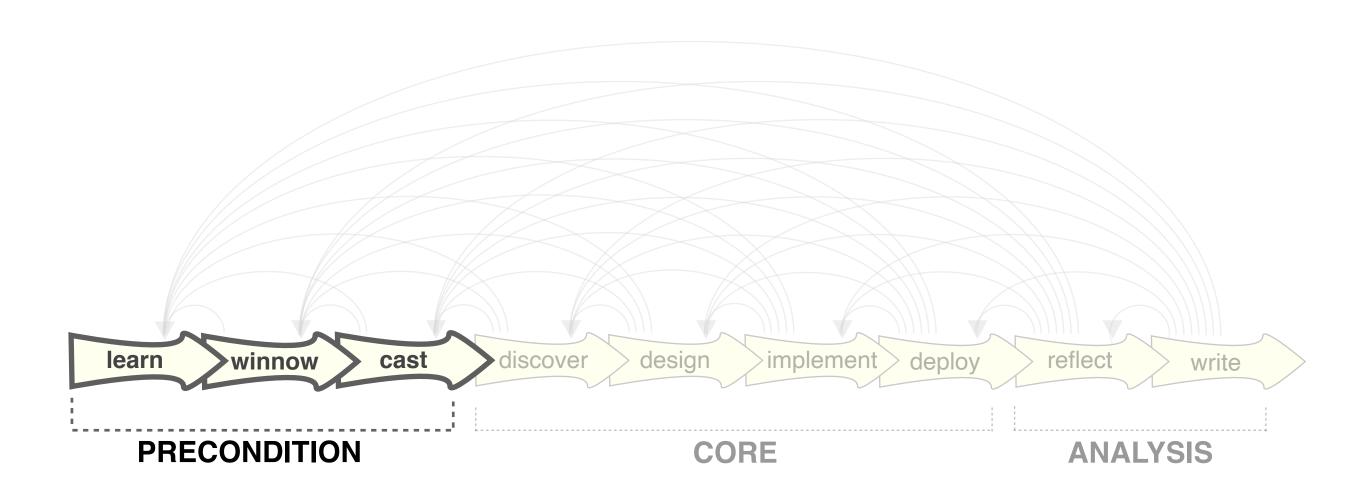
[Design Study Methodology: Reflections from the Trenches and the Stacks. Sedlmair, Meyer & Munzner. IEEE TVCG 18(12): 2431-2440, 2012 (Proc. InfoVis 2012).]

Design study methodology: definitions

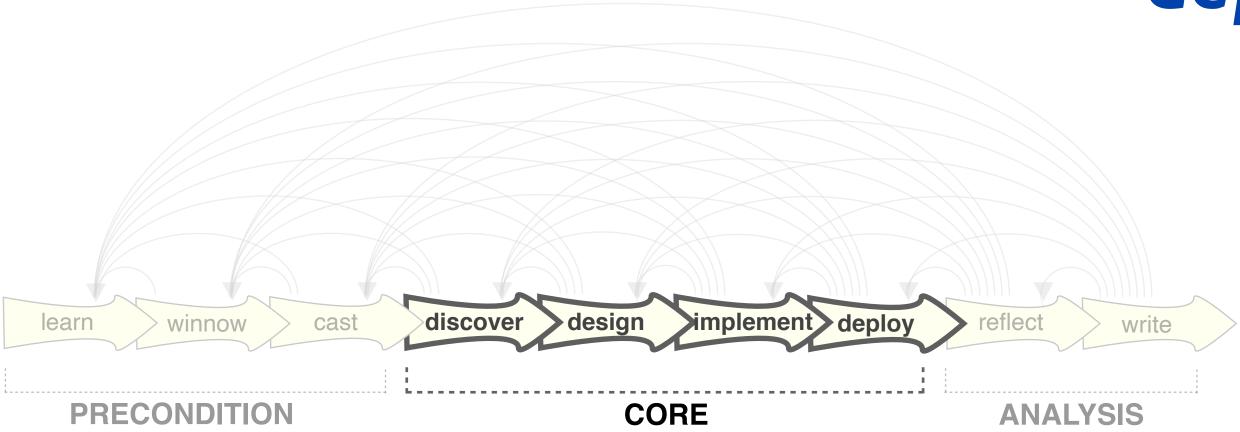




learn winnow cast

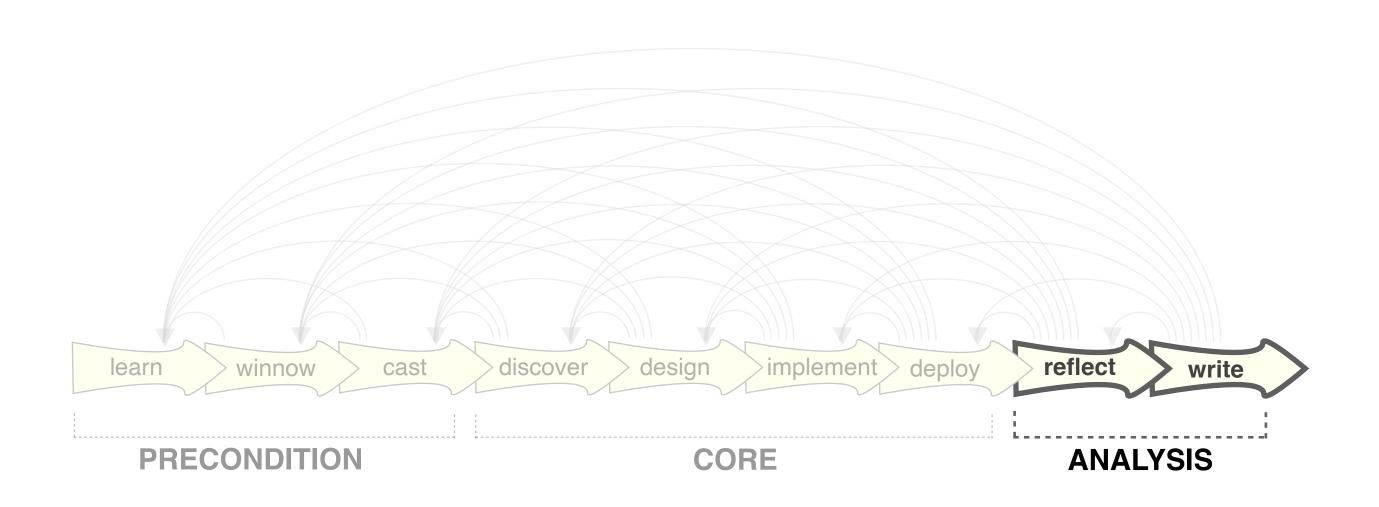


discover design implement deploy

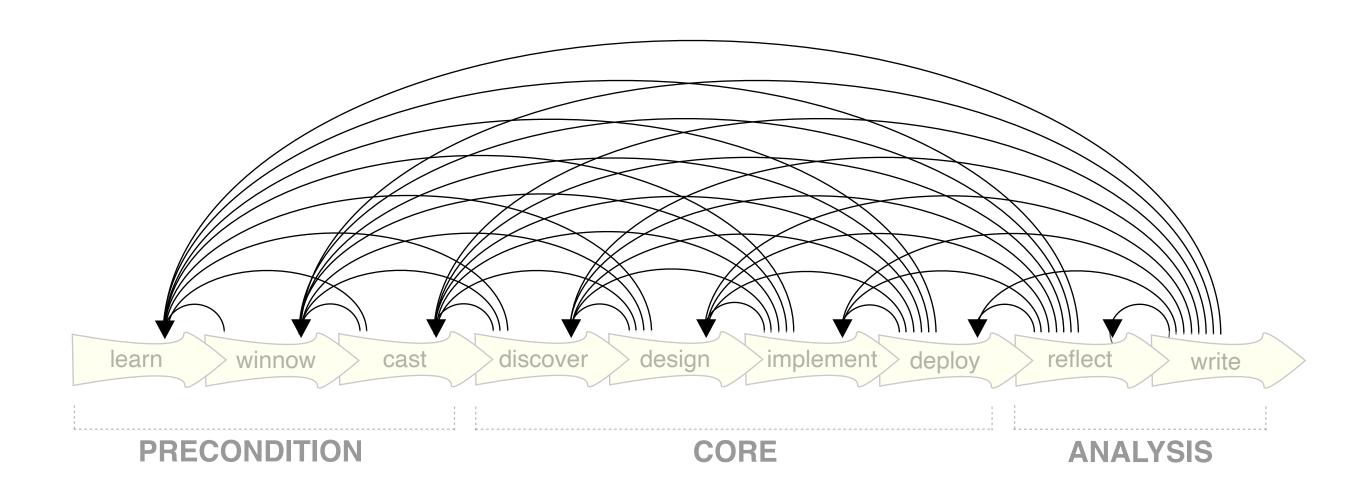


• guidelines: confirm, refine, reject, propose





iterative





		T . 1
PF-1	premature advance: jumping forward over stages	general
PF-2	premature start: insufficient knowledge of vis literature	learn
PF-3	premature commitment: collaboration with wrong people	winnow
PF-4	no real data available (yet)	winnow
PF-5	insufficient time available from potential collaborators	winnow
PF-6	no need for visualization: problem can be automated	winnow
PF-7	researcher expertise does not match domain problem	winnow
PF-8	no need for research: engineering vs. research project	winnow
PF-9	no need for change: existing tools are good enough	winnow
PF-10	no real/important/recurring task	winnow
PF-11	no rapport with collaborators	winnow
PF-12	not identifying front line analyst and gatekeeper before start	cast
PF-13	assuming every project will have the same role distribution	cast
PF-14	mistaking fellow tool builders for real end users	cast



PF-21

PF-22

PF-23

PF-24

PF-25

PF-26

PF-1	premature advance: jumping forward over stages	general
PF-2	premature start: insufficient knowledge of vis literature	learn
PF-3	premature commitment: collaboration with wrong people	winnow
PF-4	no real data available (yet)	winnow
PF-5	insufficient time available from potential collaborators	winnow
PF-6	no need for visualization: problem can be automated	winnow
PF-7	researcher expertise does not match domain problem	winnow
PF-8	no need for research: engineering vs. research project	winnow
PF-9	no need for change: existing tools are good enough	winnow
PF-10	no real/important/recurring task	winnow
PF-11	no rapport with collaborators	winnow
PF-12	not identifying front line analyst and gatekeeper before start	cast
PF-13	assuming every project will have the same role distribution	cast
PF-14	mistaking fellow tool builders for real end users	cast
PF-15	ignoring practices that currently work well	discover
PF-16	expecting just talking or fly on wall to work	discover
PF-17	experts focusing on visualization design vs. domain problem	discover
PF-18	learning their problems/language: too little / too much	discover
PF-19	abstraction: too little	design
PF-20	premature design commitment: consideration space too small	design

design

deploy

deploy

deploy

implement

implement

mistaking technique-driven for problem-driven work

usage study not case study: non-real task/data/user

liking necessary but not sufficient for validation

premature end: insufficient deploy time built into schedule

nonrapid prototyping

usability: too little / too much



PF-1	premature advance: jumping forward over stages	general	PF-21	mistaking technique-driven for problem-driven work	design
PF-2	premature start: insufficient knowledge of vis literature	learn	PF-22	nonrapid prototyping	implement
PF-3	premature commitment: collaboration with wrong people	winnow	PF-23	usability: too little / too much	implement
PF-4	no real data available (yet)	winnow	PF-24	premature end: insufficient deploy time built into schedule	deploy
PF-5	insufficient time available from potential collaborators	winnow	PF-25	usage study not case study: non-real task/data/user	deploy
PF-6	no need for visualization: problem can be automated	winnow	PF-26	liking necessary but not sufficient for validation	deploy
PF-7	researcher expertise does not match domain problem	winnow	PF-27	failing to improve guidelines: confirm, refine, reject, propose	reflect
PF-8	no need for research: engineering vs. research project	winnow	PF-28	insufficient writing time built into schedule	write
PF-9	no need for change: existing tools are good enough	winnow	PF-29	no technique contribution \neq good design study	write
PF-10	no real/important/recurring task	winnow	PF-30	too much domain background in paper	write
PF-11	no rapport with collaborators	winnow	PF-31	story told chronologically vs. focus on final results	write
PF-12	not identifying front line analyst and gatekeeper before start	cast	PF-32	premature end: win race vs. practice music for debut	write
PF-13	assuming every project will have the same role distribution	cast			
PF-14	mistaking fellow tool builders for real end users	cast]		
PF-15	ignoring practices that currently work well	discover	1		
PF-16	expecting just talking or fly on wall to work	discover]		
PF-17	experts focusing on visualization design vs. domain problem	discover	1		

discover

design

design

learning their problems/language: too little / too much

premature design commitment: consideration space too small

abstraction: too little

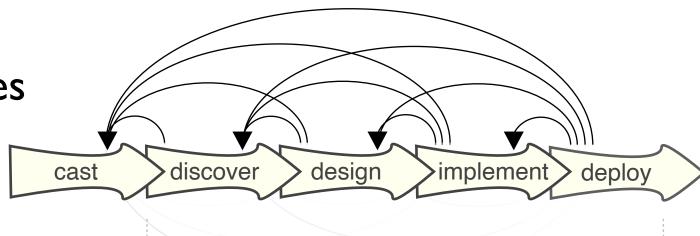
PF-18

PF-19

PF-20

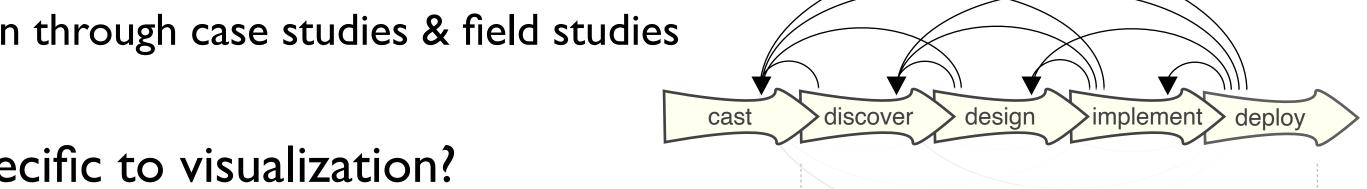
Design studies & user-centered design

- user-centered design: well-known HCl methodology
 - iterative refinement & deployment
 - evaluation through case studies & field studies

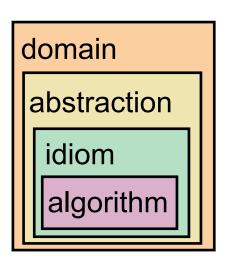


Design studies & user-centered design

- user-centered design: well-known HCl methodology
 - iterative refinement & deployment
 - evaluation through case studies & field studies



- what's specific to visualization?
 - discovering task and data abstractions
 - -designing visual encoding & interaction idioms that map to abstractions



Three case studies of problem-driven work

e-commerce



facilities management



biology



Three case studies of problem-driven work

e-commerce



• facilities management

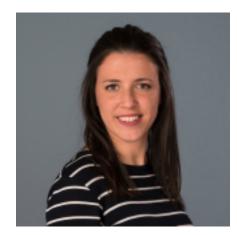


biology





Kim Dextras-Romagnino



Segmentifier

Interactive Refinement of Clickstream Data

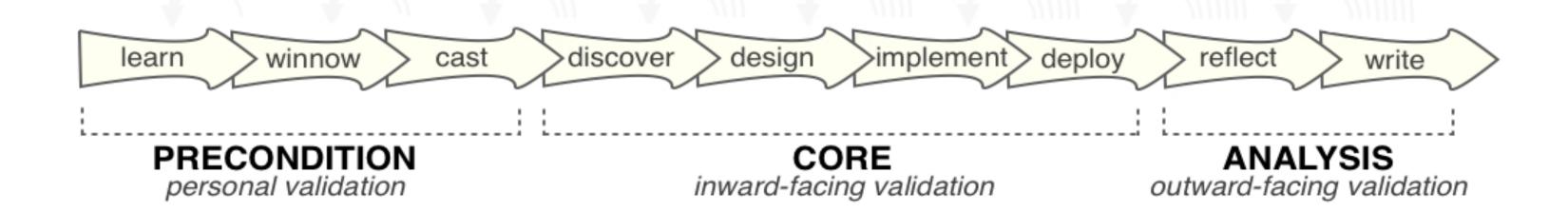
http://www.cs.ubc.ca/labs/imager/tr/2019/segmentifier

E-commerce: mobile apps for large companies



Process: Design Study Methodology

- Precondition Phase (5 months): interviews with 12 employees
- Core Phase (11 months): Iterative design and implementation
- Analysis Phase (3 months): Reflect and write



What are the **Data and Task Abstractions** for *Clickstream Data Analysis?*

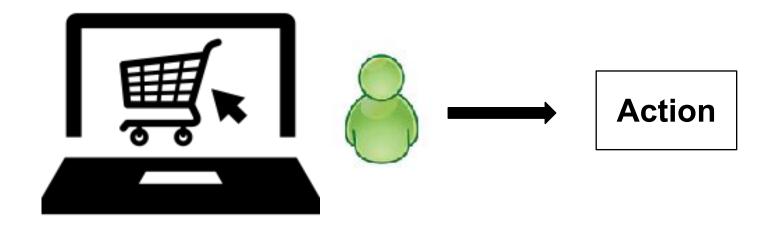
Clickstream Data

Clickstream Analysis Tasks

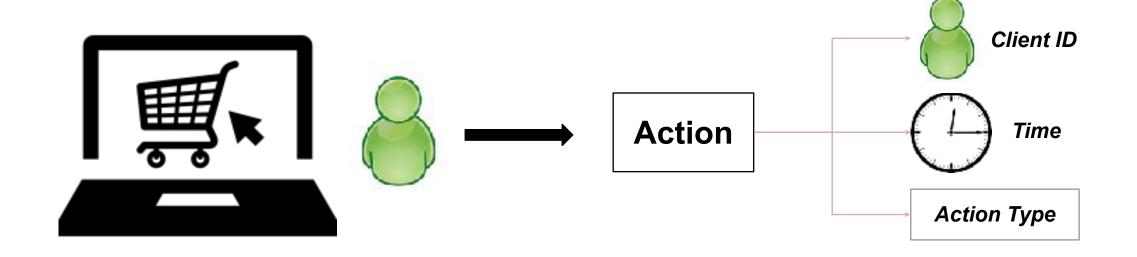
Segmentifier Analysis Model

What is Clickstream Data?

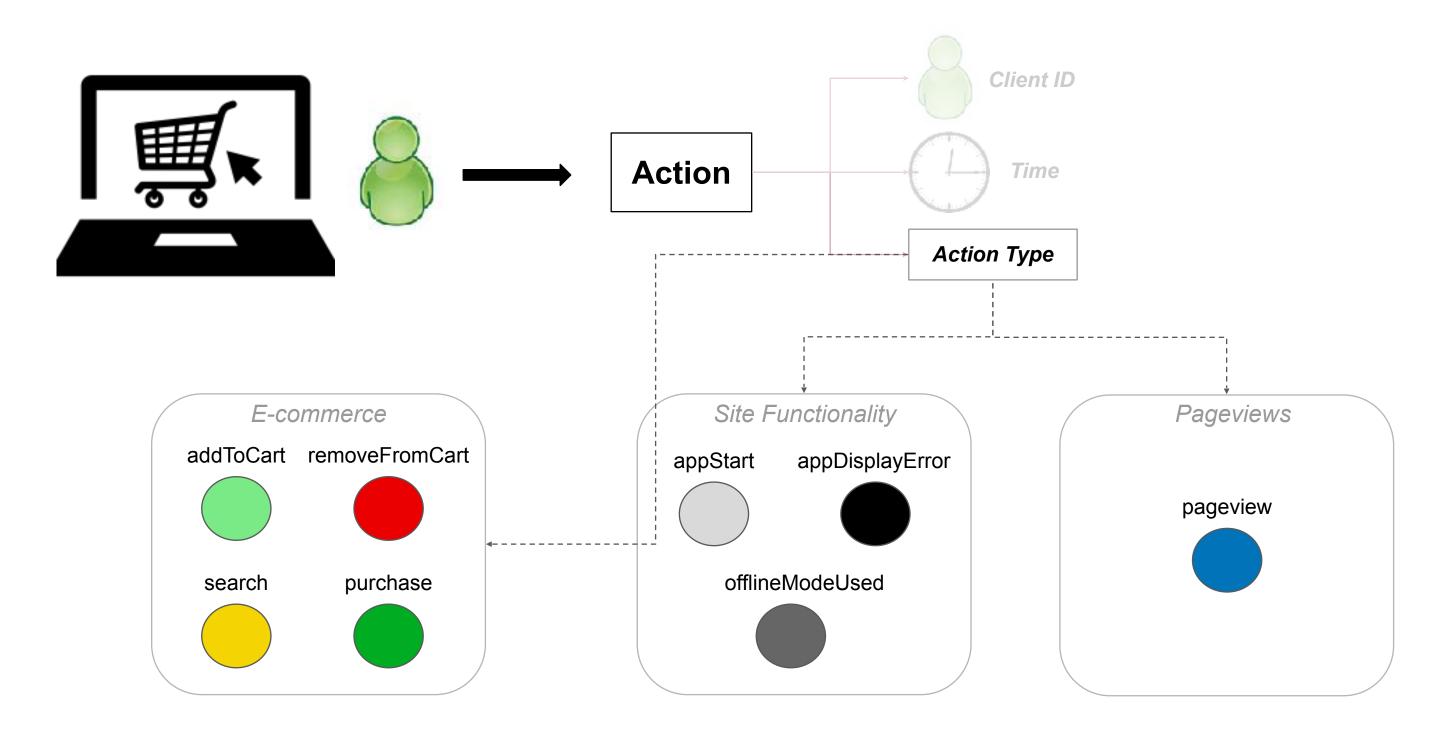
Data: Actions



Data: Action Attributes



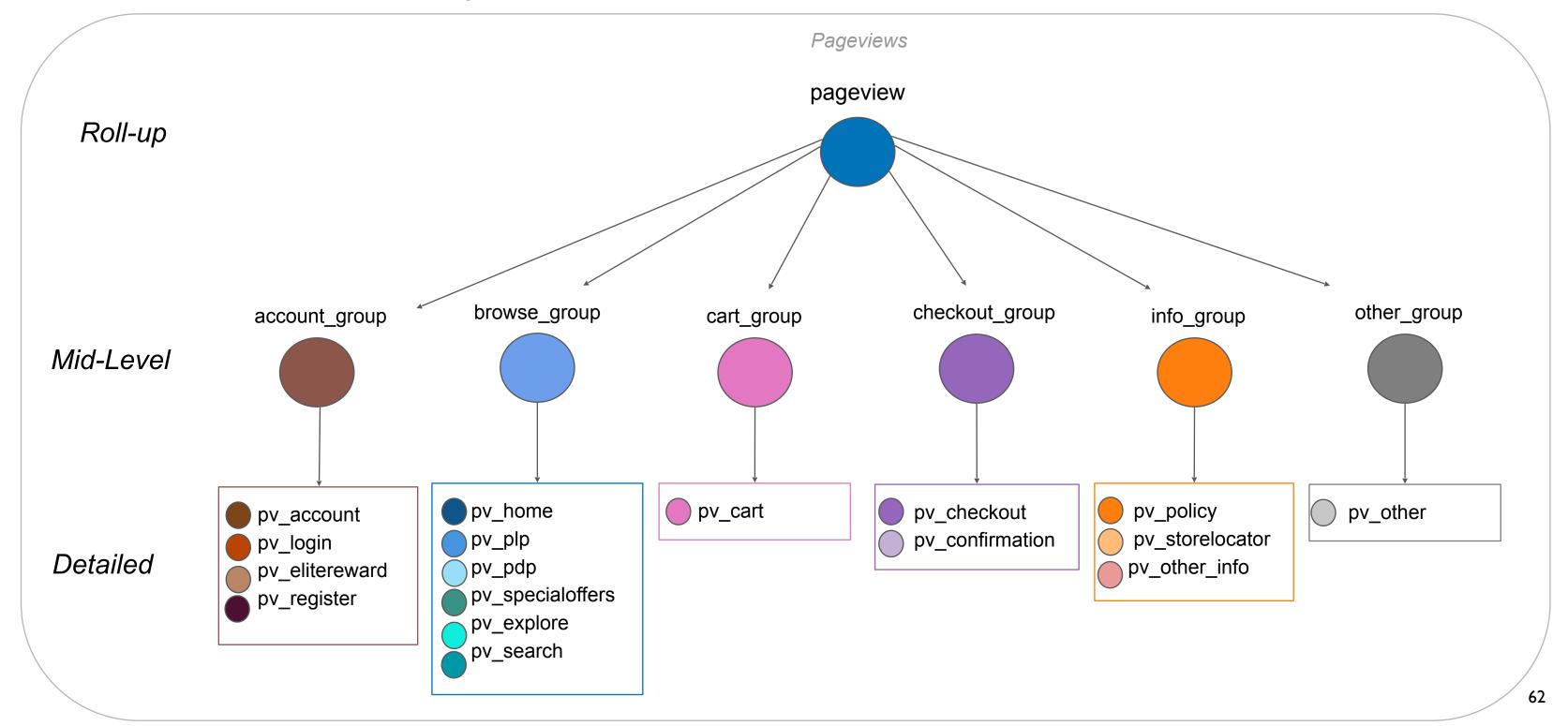
Data: Action Types



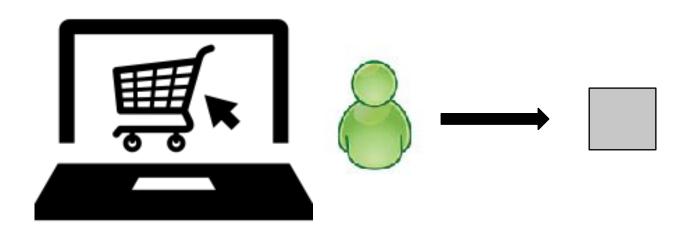
Action Hierarchy

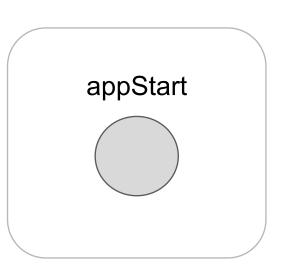


Action Hierarchy

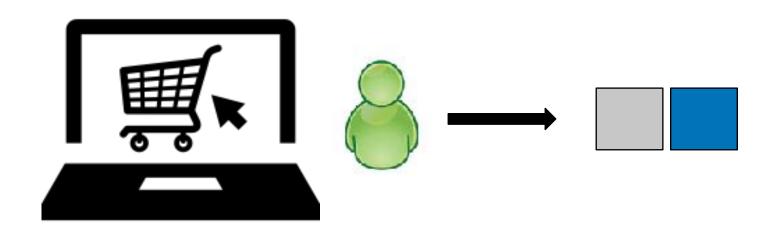


Data: Sequences



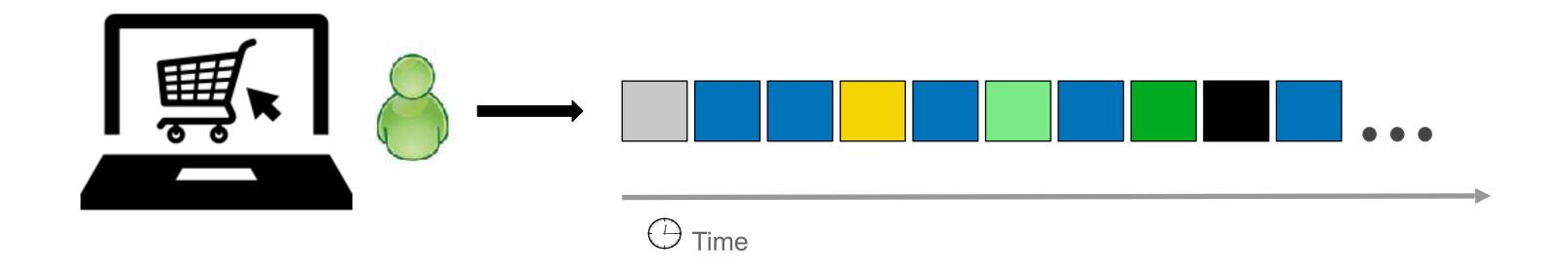


Data: Sequences



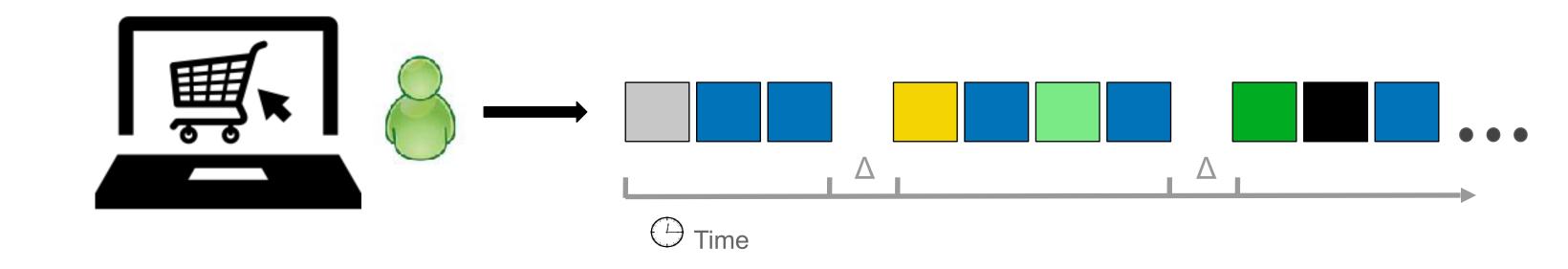


Data: Client Sequences



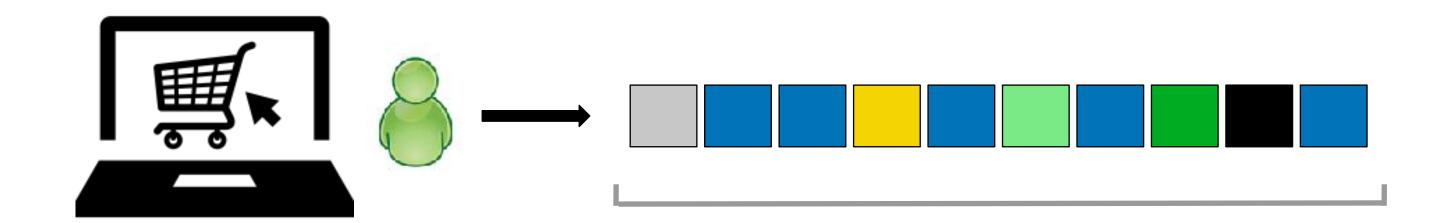
Client Sequences: all actions performed by a single user

Data: Session Sequences



Session Sequences: all actions performed by a single user within a defined amount of time (Δ) from each other. Δ is usually 30 min.

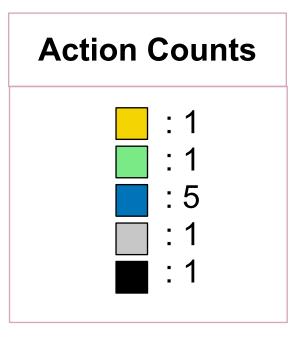
Data: Sequence Attributes



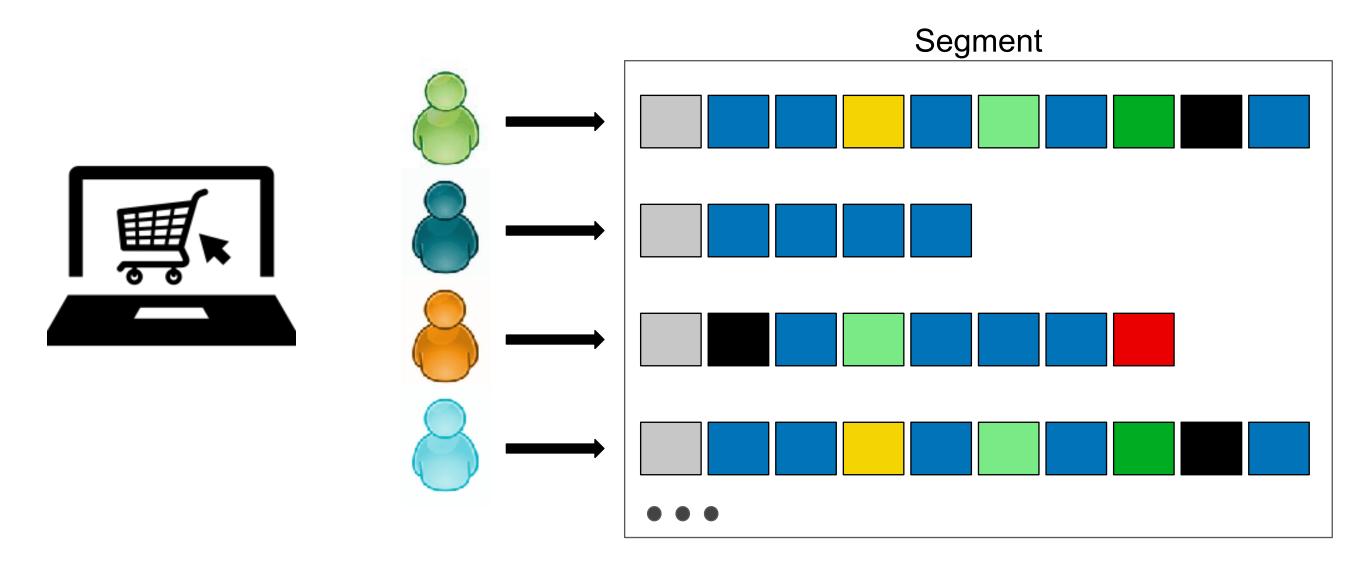
Start time En

End time

Duration

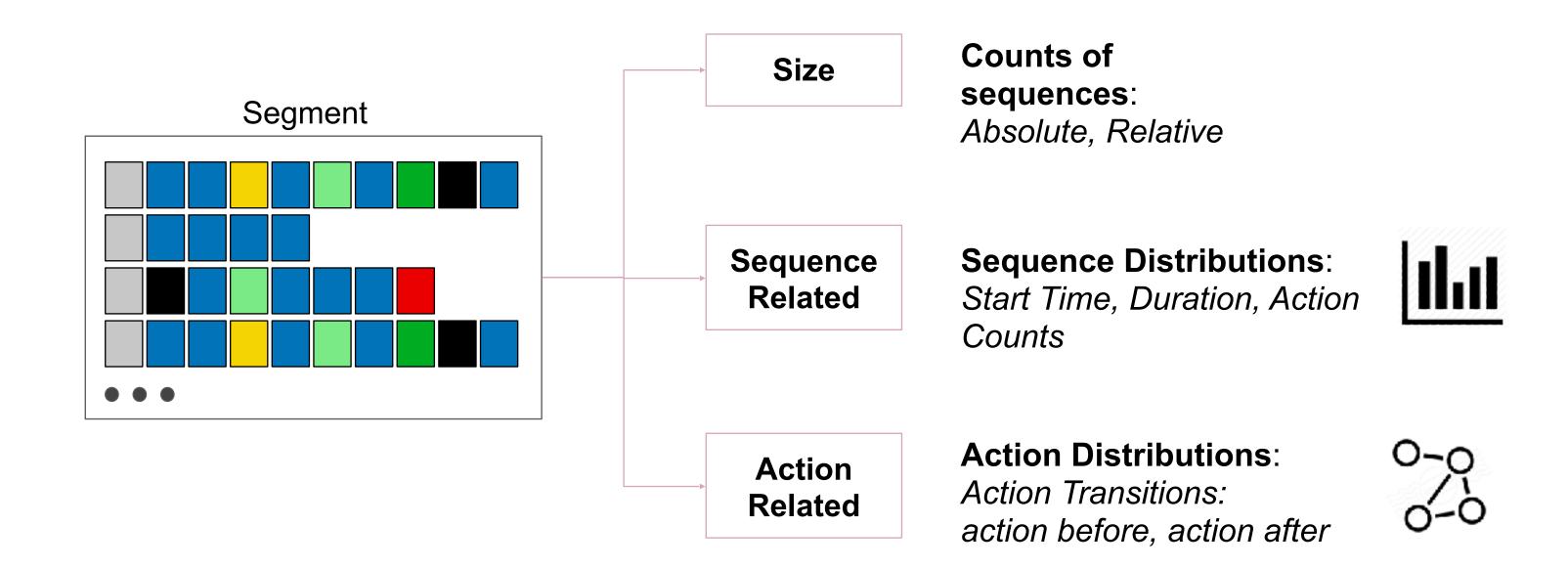


Data: Segments

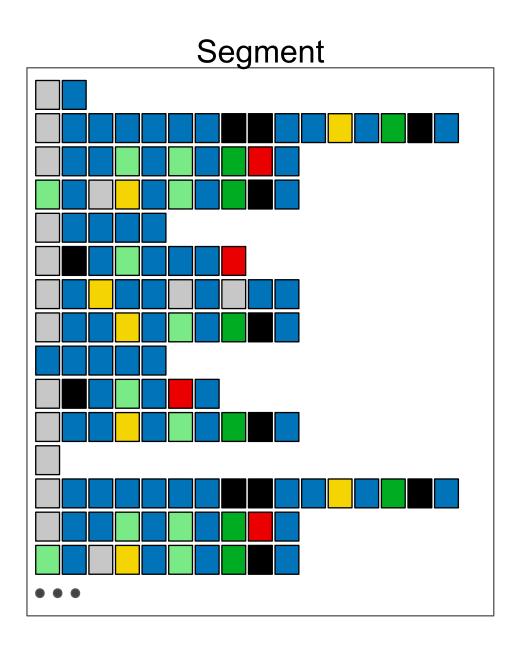


Segment: any set of sequences

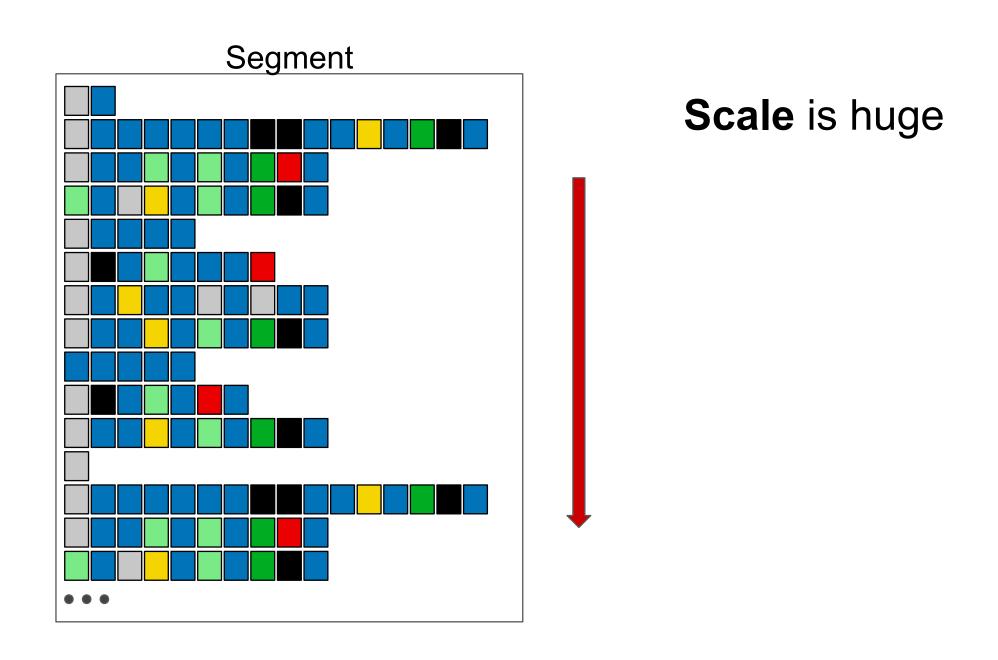
Data: Segment Attributes



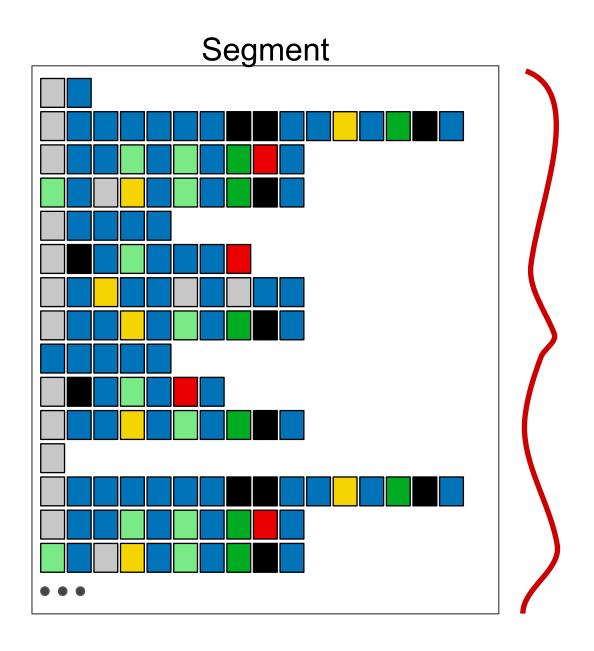
Real-world Clickstream Data



Real-world Clickstream Data



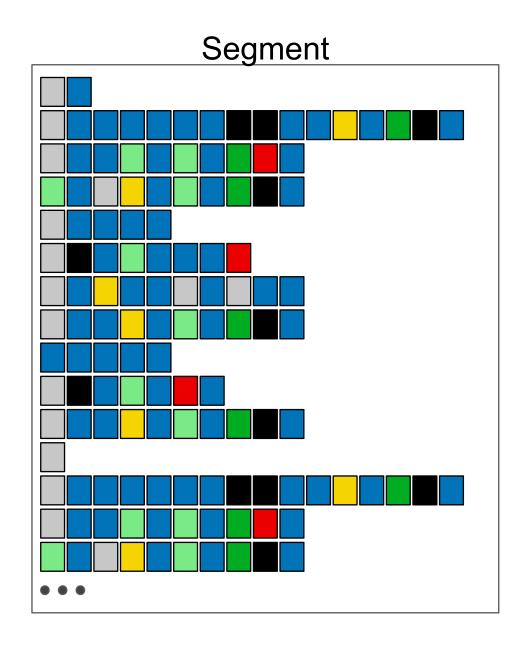
Real-world Clickstream Data



Scale is huge

Variability is high

Real-world Clickstream Data



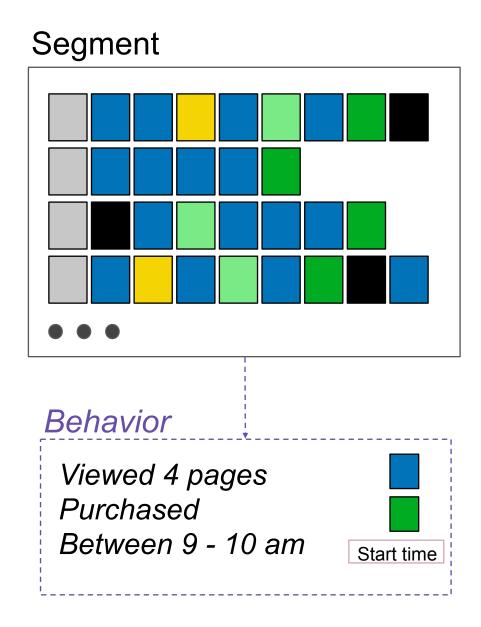
Scale is huge

Variability is high

Most work **fails** when applied to real-world data

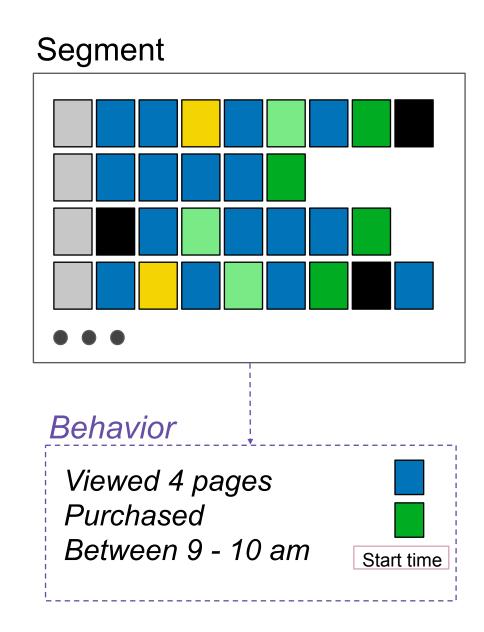
What are Clickstream Data Analysis Tasks?

Tasks: Segment Behavior



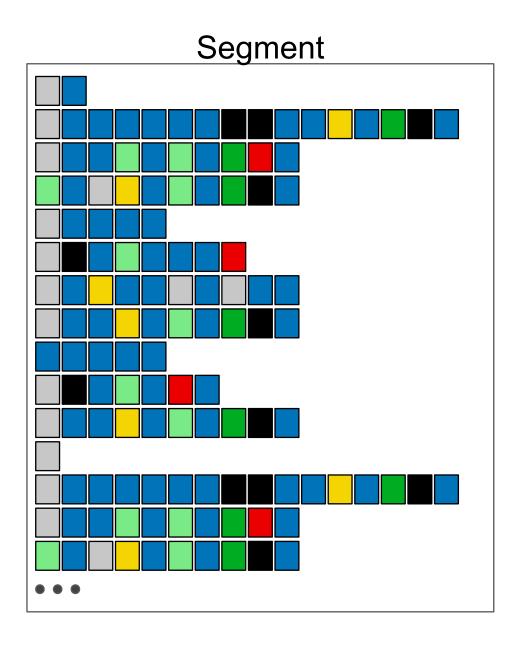
Behavior: set of attribute constraints

Tasks: Segment Behavior

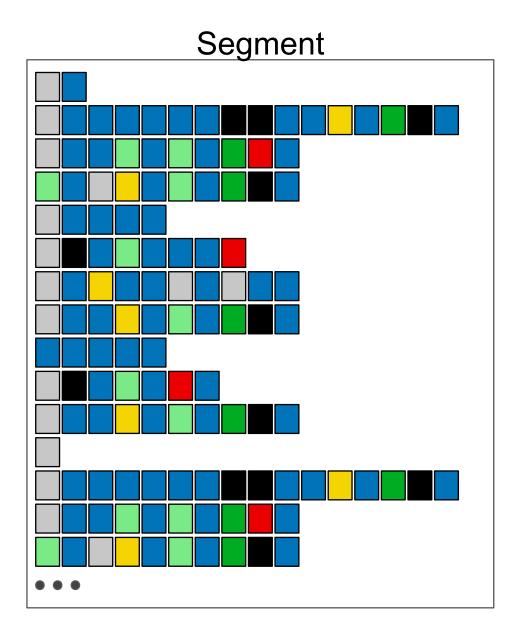


Behavior: set of attribute constraints

- Expected
 Users add to cart before purchasing
- Unexpected
 No purchases on a certain month
- FavorablePurchased
- UnfavorableBounced

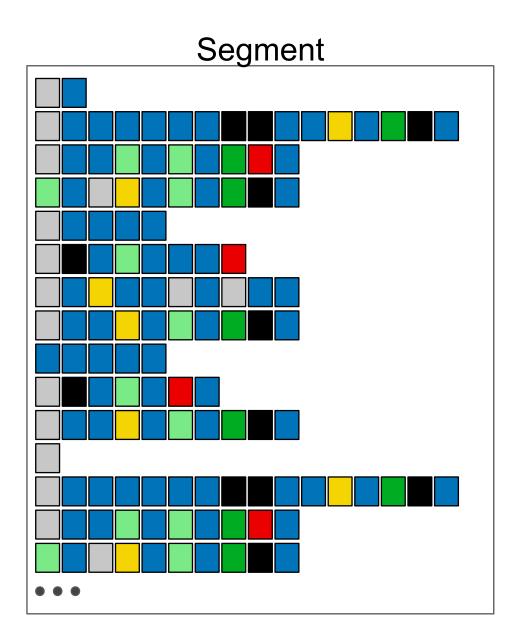


Identify: Find some set of sequences that constitutes interesting behavior



Identify: Find some set of sequences that constitutes interesting behavior

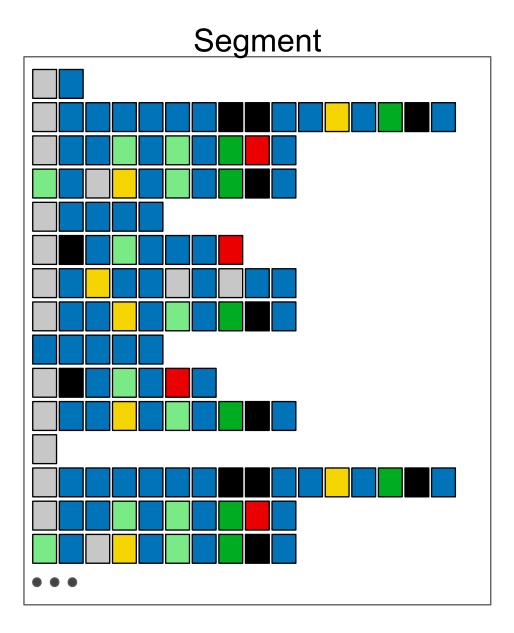
Drilldown: Distinguish more specific *behaviors* to further partition a segment previously defined by looser constraints



Identify: Find some set of sequences that constitutes interesting behavior

Drilldown: Distinguish more specific *behaviors* to further partition a segment previously defined by looser constraints

Frequency: Determine how many sequences are in the segment defined by *behavior*



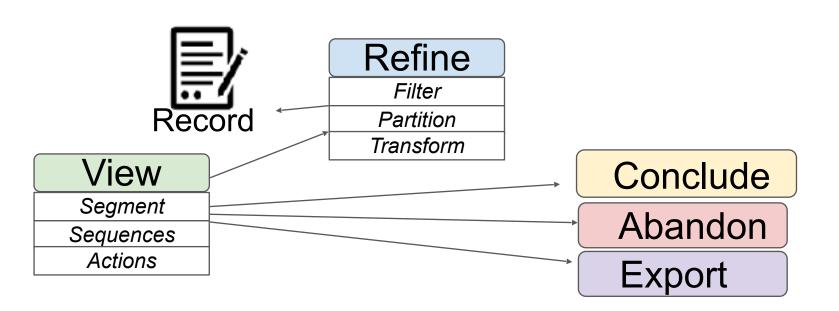
Identify: Find some set of sequences that constitutes interesting behavior

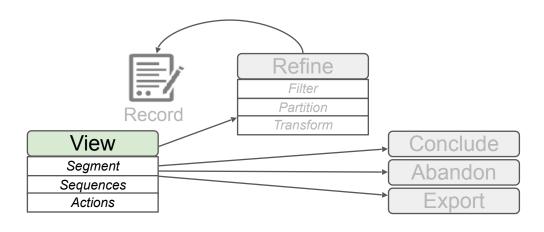
Drilldown: Distinguish more specific *behaviors* to further partition a segment previously defined by looser constraints

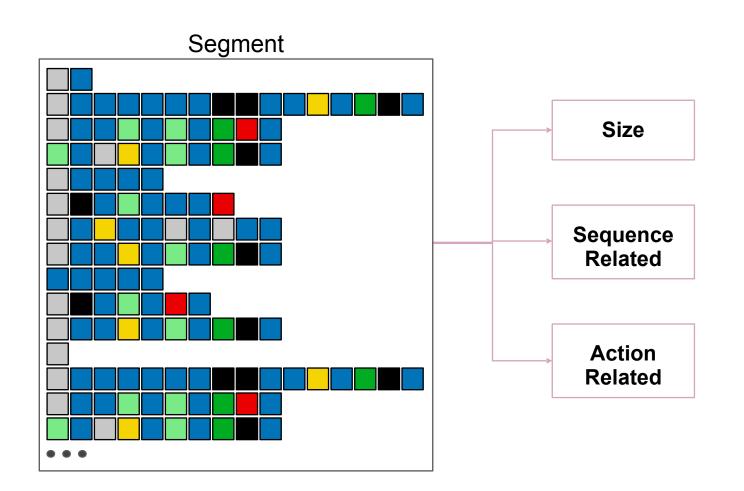
Frequency: Determine how many sequences are in the segment defined by *behavior*

Ordering within sequence: Match if one action subsequence occurs before (or after) another action subsequence in a sequence

- Abstraction above task/data level to provide design rationale
- Take a giant, noisy dataset and refine it into small, clean segments for
 - actionable insights
 - downstream analysis
- Bridge the gap between real-world data and other techniques

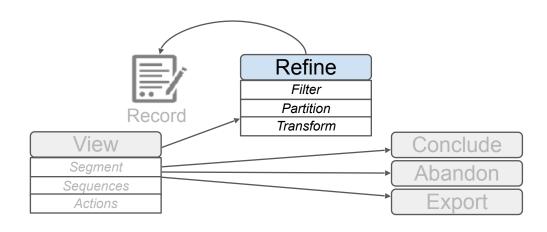


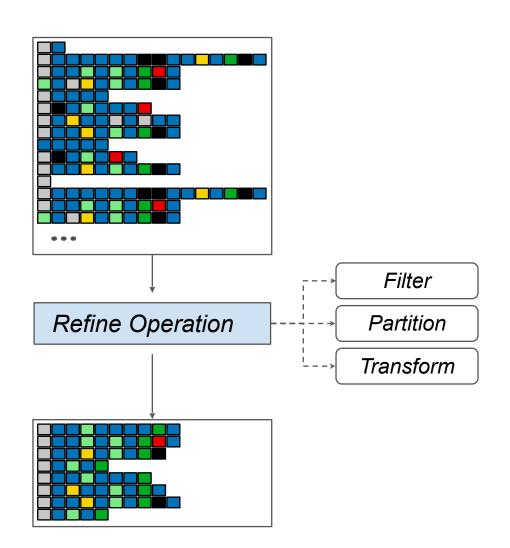




View

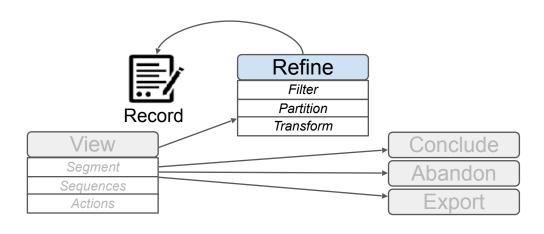
- Gives Insight into underlying data of segment
 - Action Attributes
 - Sequence Attributes
 - Segment Attributes
- Leads to:
 - Insights
 - New ways on how to refine
 - Whether segment should be abandoned
 - Whether segment should be exported

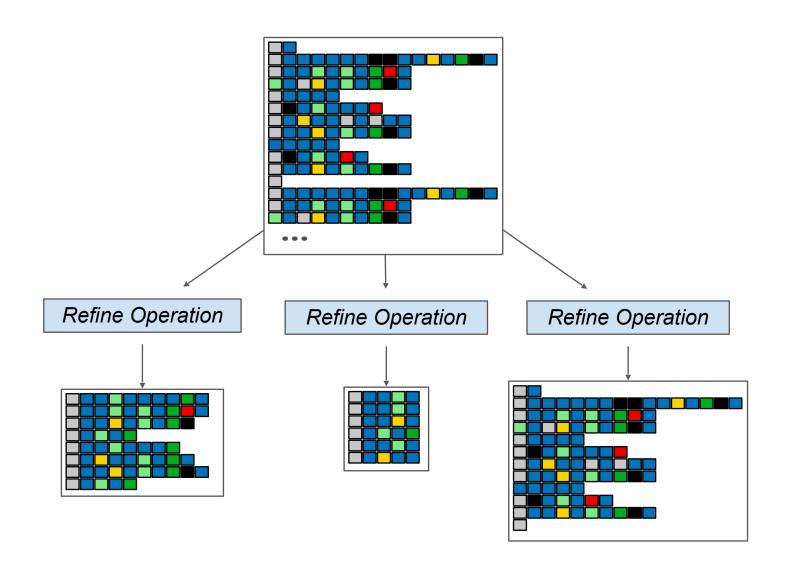


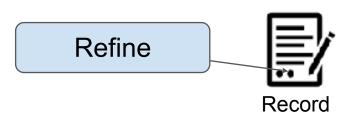


Refine

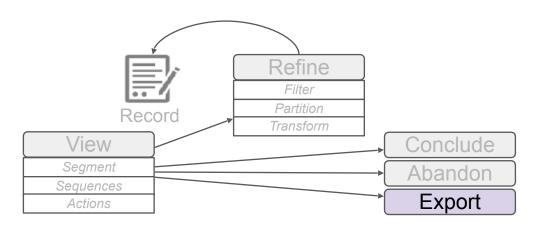
- Apply operation to create new segments
- Type of Refinements
 - o Filter
 - Partition
 - Transform

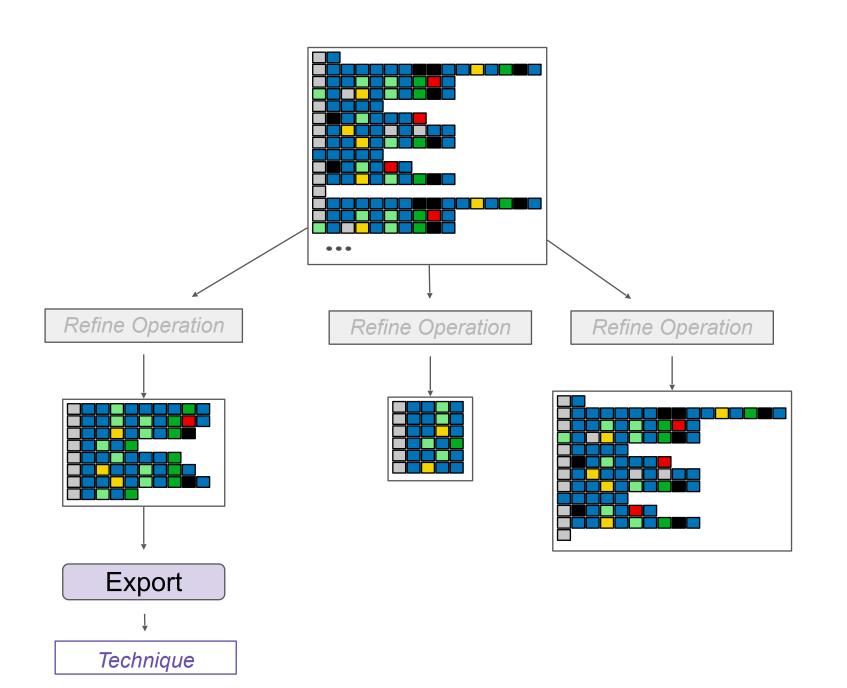






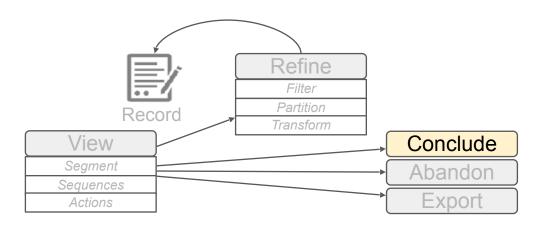
- Record all refinement steps automatically
- Keep track of questions asked and hypotheses tested
- Ability to create and view multiple segments from the same segment

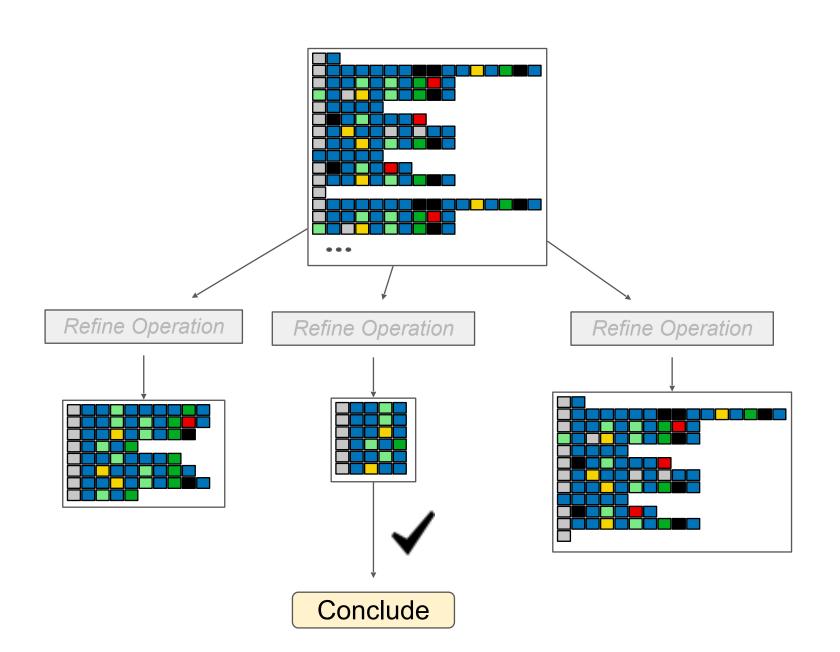




Export

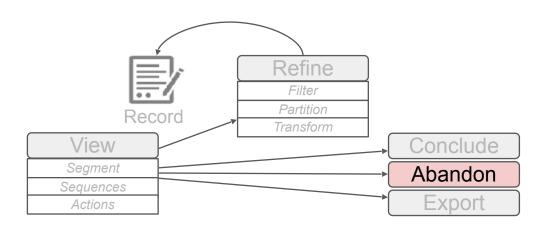
- Export refined segments for further downstream analysis, to more specific tools:
 - Pattern mining
 - Clustering

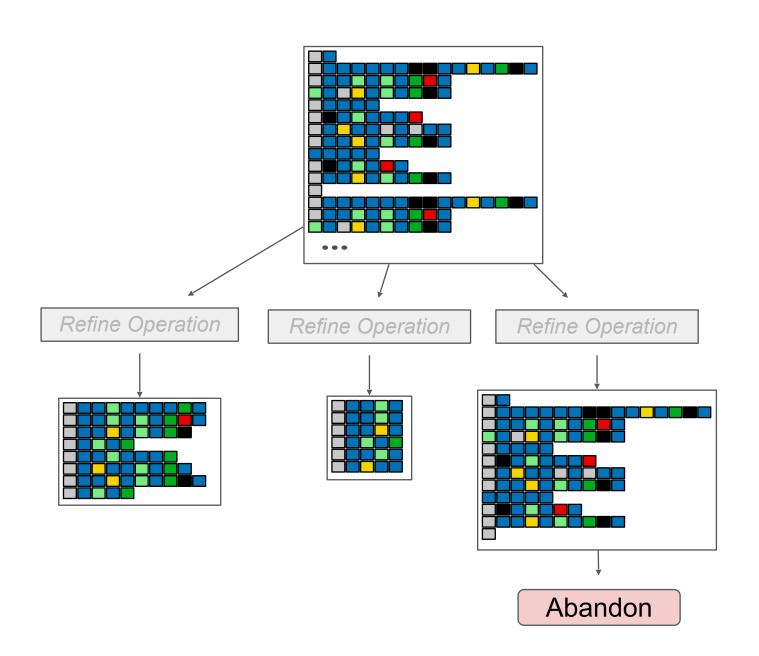




Conclude

Discover actionable insight by viewing segment

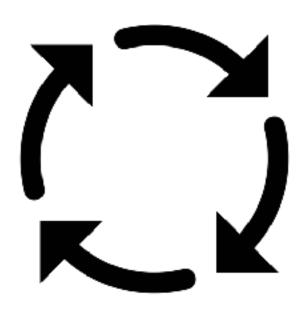




Abandon

- By *viewing* the segment, analyst *abandons* if:
 - No actionable insights
 - No further ways to refine
 - Not suitable for *export*

Why Visual Analytics?



- Automation would be nice...
 - Put data in, actionable results appear
- ... but it is not realistic
 - Many possible questions, data-driven interplay between finding answers and generating new questions
- Human-in-the-loop visual data analysis
 - Integrate computing power of machine with intuition of domain experts

Solution

The Segmentifier Interface

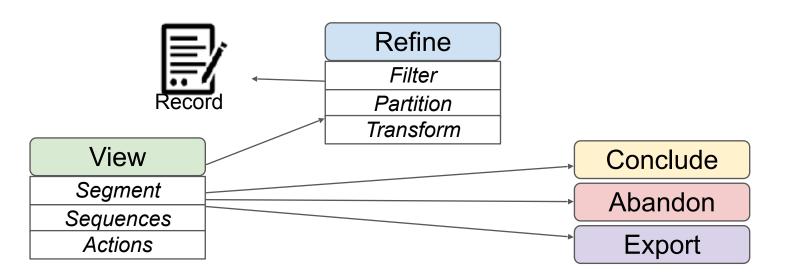


Video



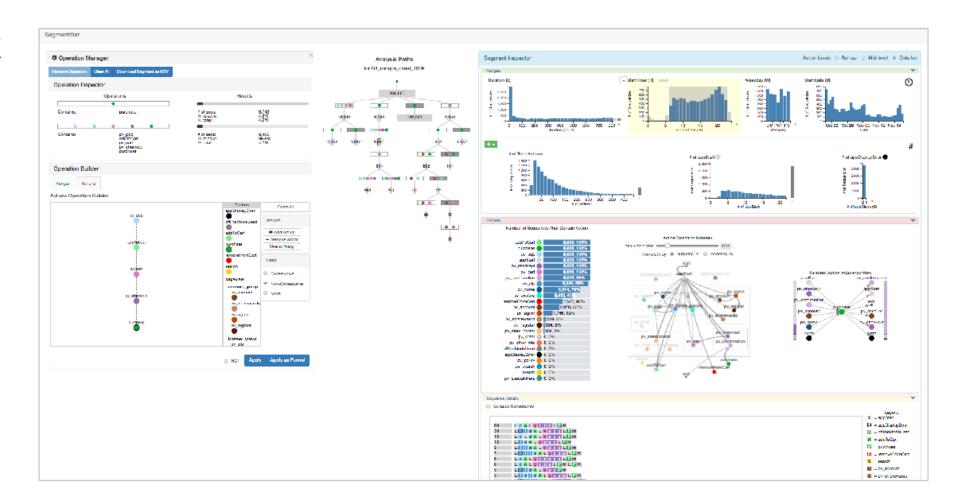
Segmentifier Contributions

Thorough characterization of task and data abstraction for clickstream data analysis



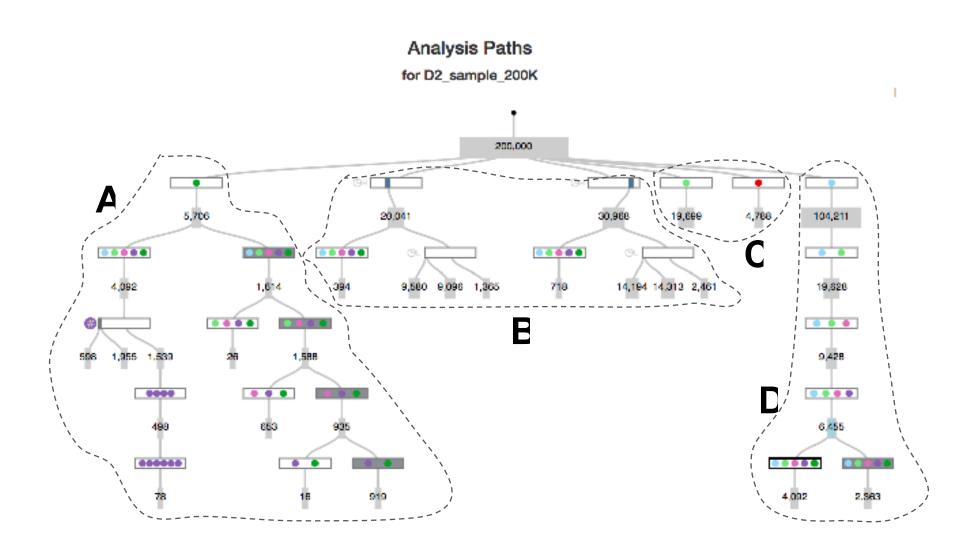
Segmentifier Contributions

- Thorough characterization of task and data abstraction for clickstream data analysis
- Segmentifier: novel analytics interface for refining data segments and viewing characteristics before downstream fine-grained analysis



Segmentifier Contributions

- Thorough characterization of task and data abstraction for clickstream data analysis
- Segmentifier: novel analytics interface for refining data segments and viewing characteristics before downstream fine-grained analysis
- Preliminary evidence of utility



Three case studies of problem-driven work

e-commerce



• facilities management



biology





Michael Oppermann



Ocupado

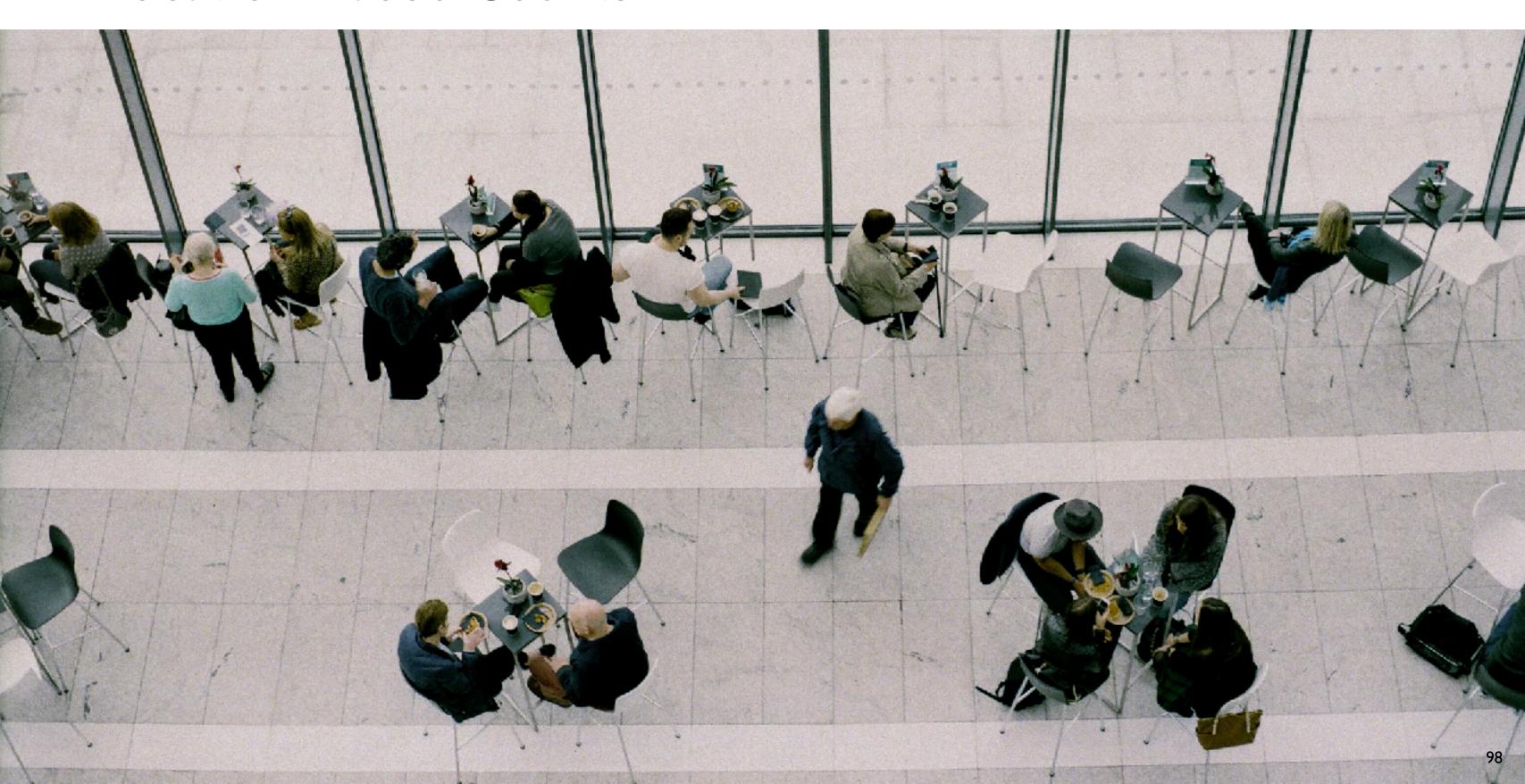
Visualizing Location-Based Counts Over Time Across Buildings

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.

Location-Based Counts





Previous measurement required physical counting or installation of additional hardware.

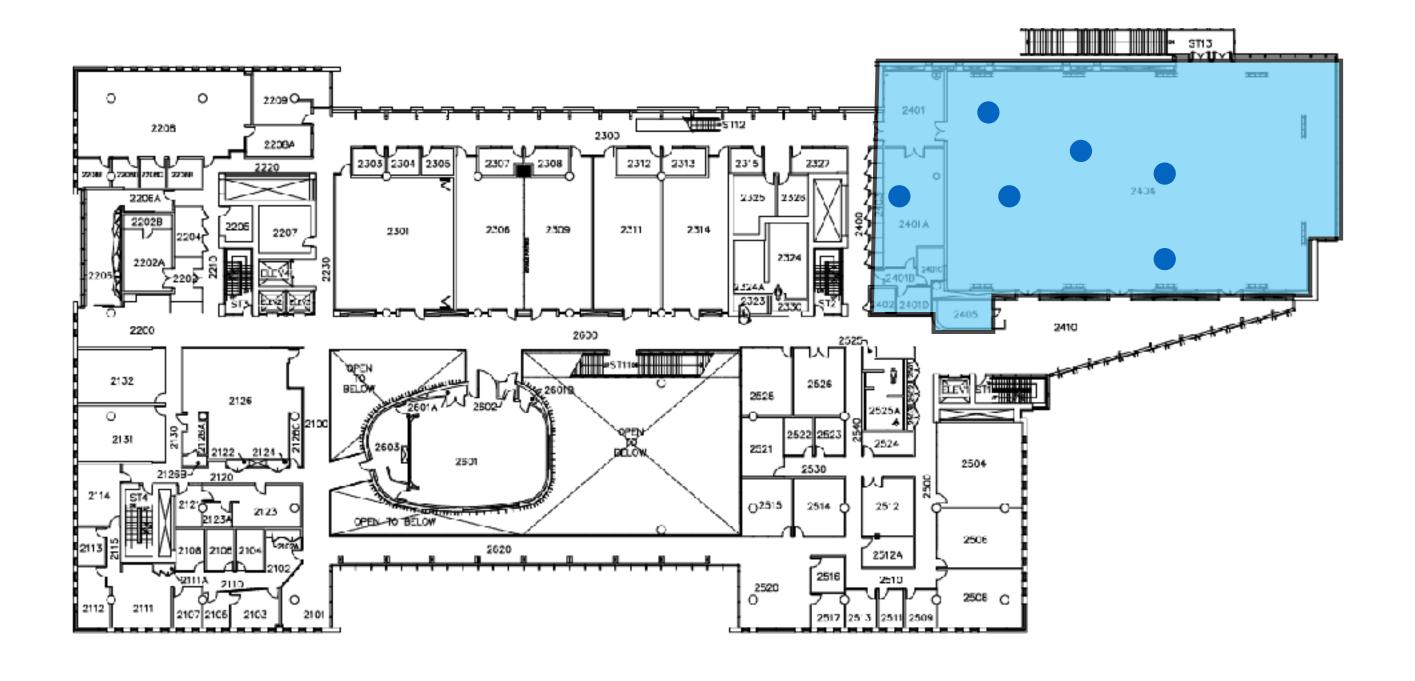


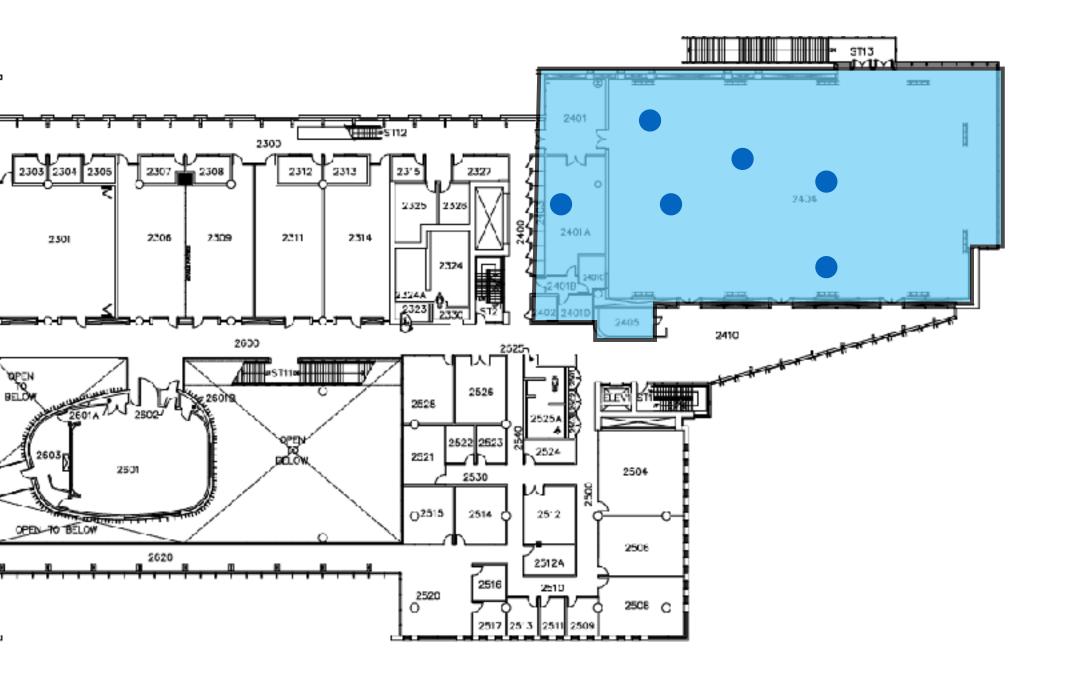
Previous measurement required physical counting or installation of additional hardware.



Previous visualization attempts were limited in space and time.

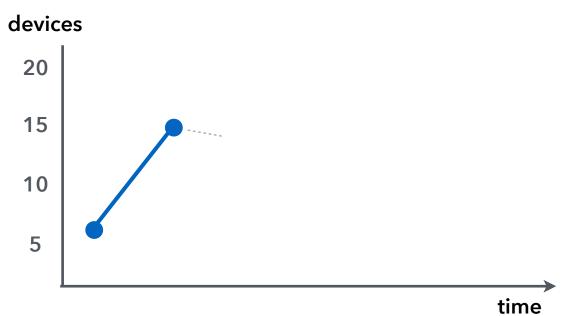


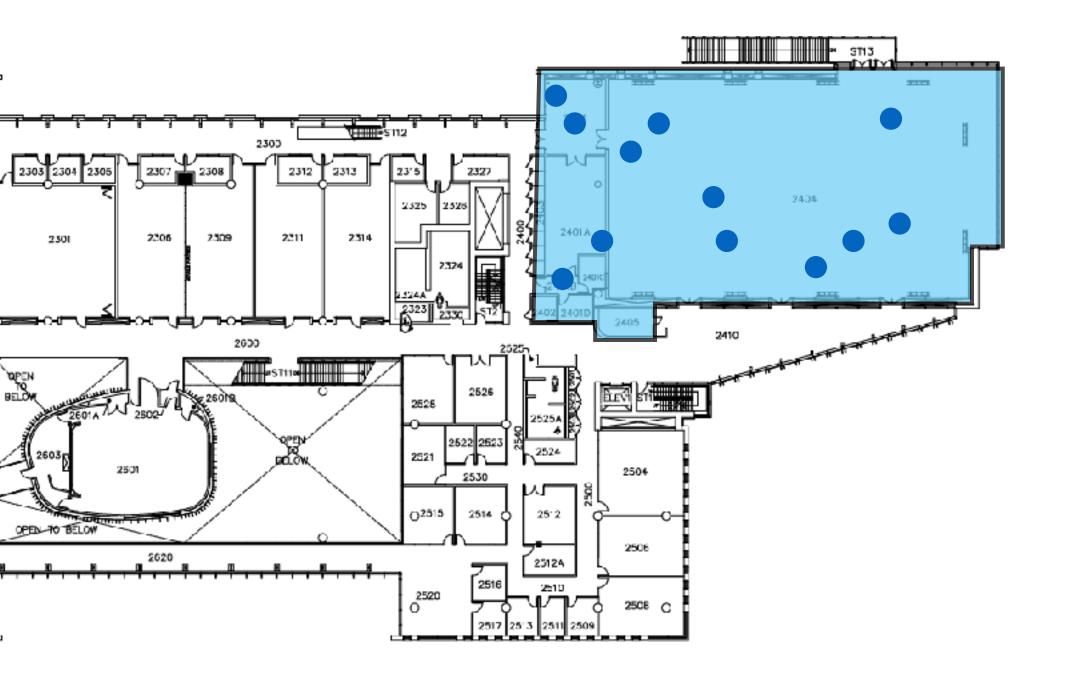


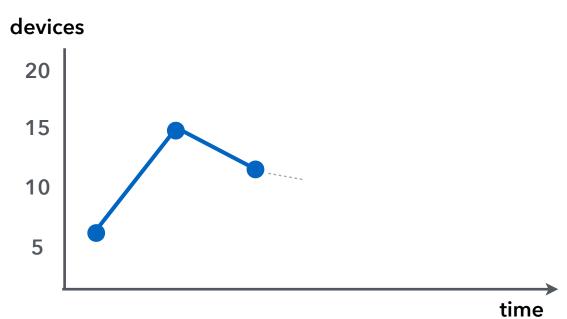


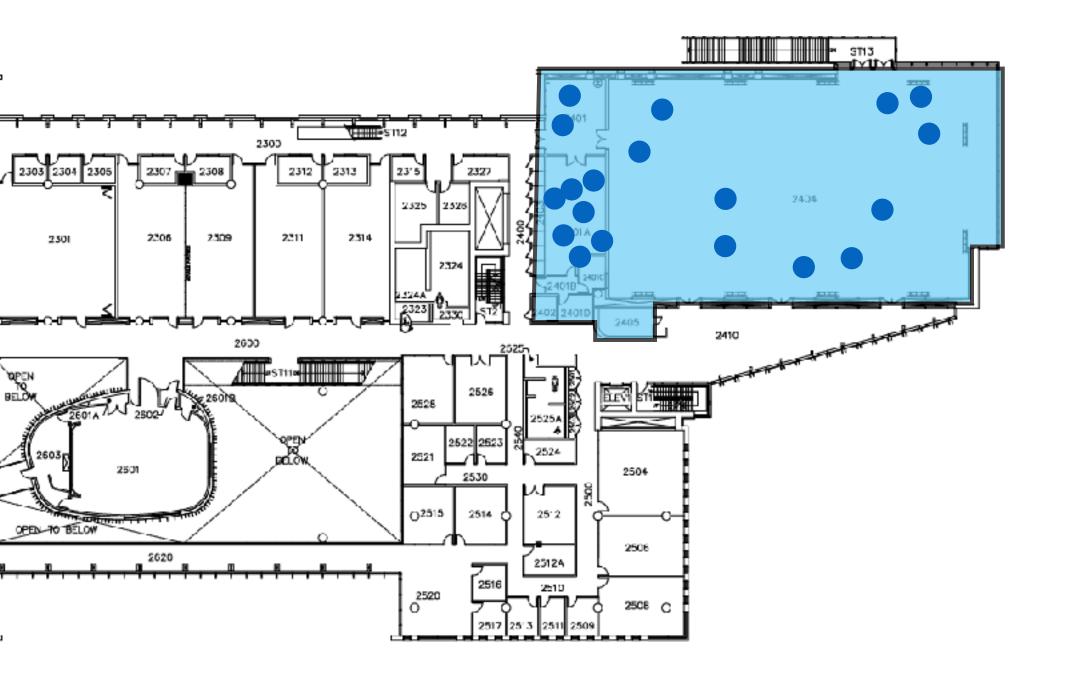


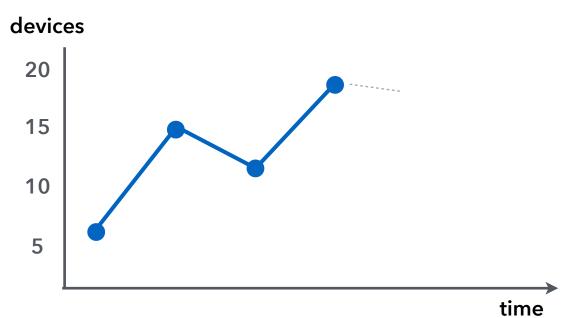


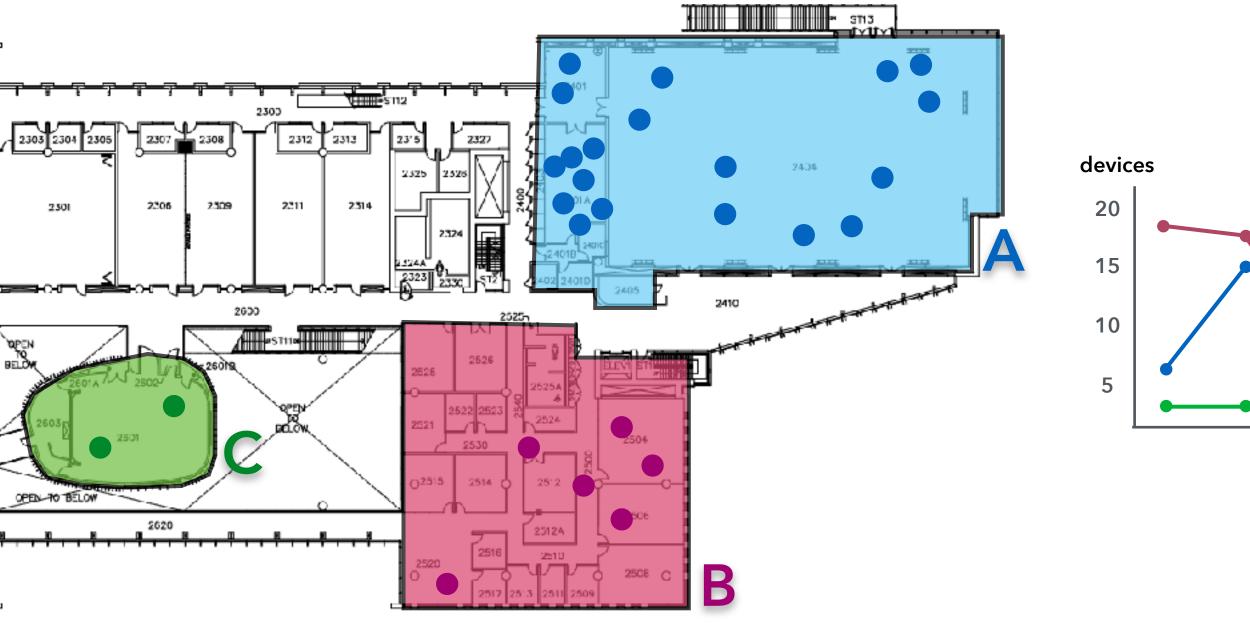


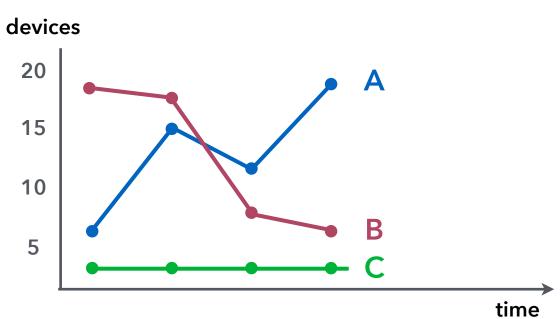














Location-Based Counts

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

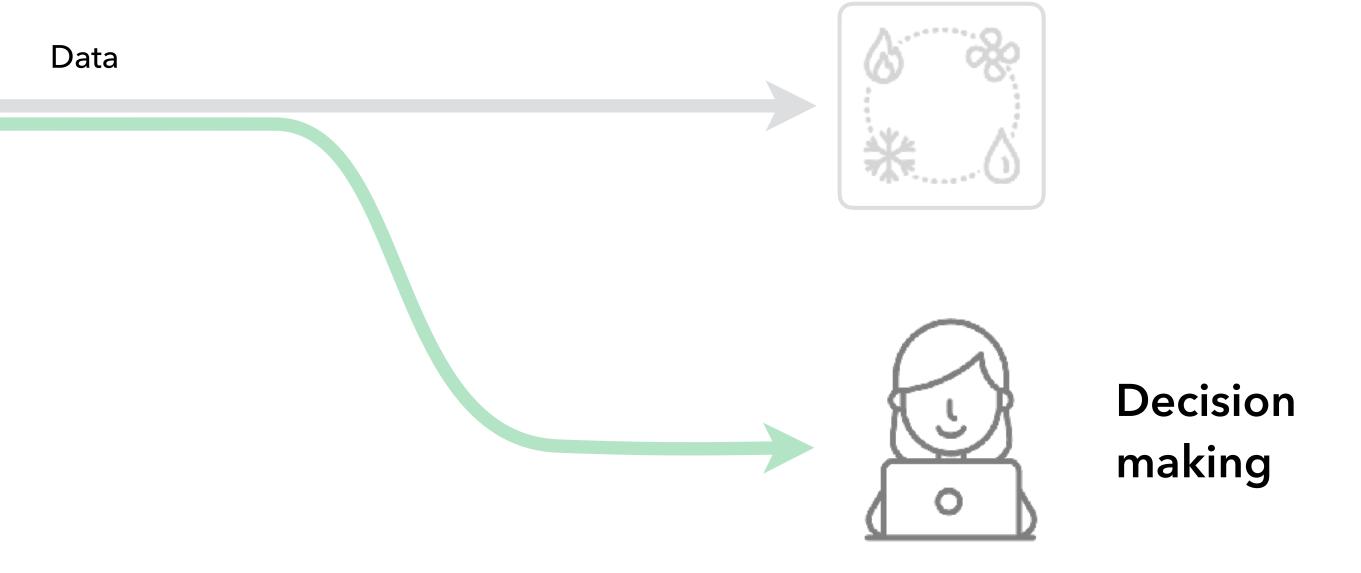
Data



Automated HVAC control

Data





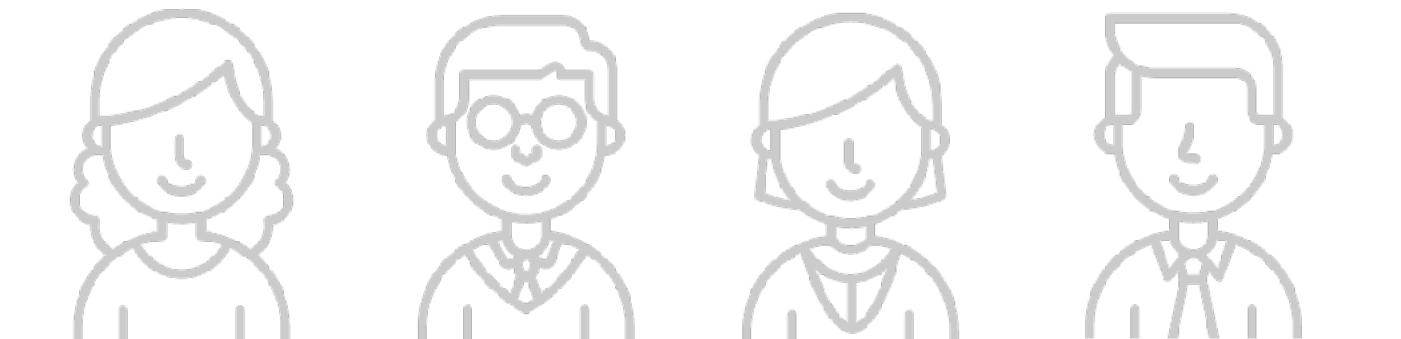


WiFi connections as a proxy for occupancy



WiFi connections as a proxy for occupancy

Interviews with potential stakeholders



Focus Domains

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

Focus Domains

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



Semi-structured discussions and live demos



Confirm assumptions or previous observations.

Do students occupy room x in evenings or on 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

Sandbox

Data sketches, static data export

Time

Visualization Prototypes

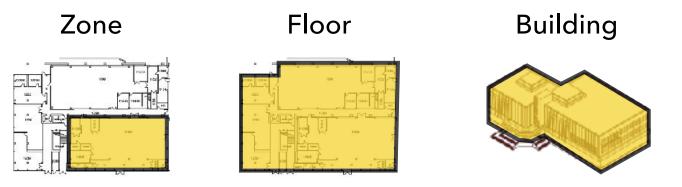
Sandbox

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

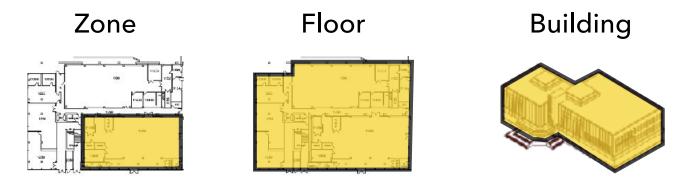
Spatial and Temporal Data Granularities

Regions of interest

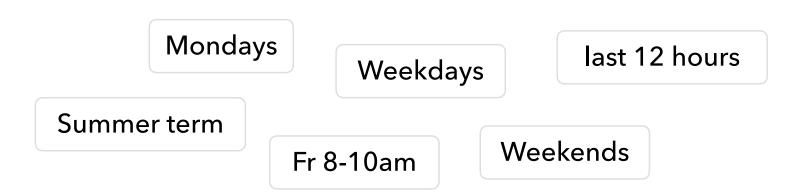


Spatial and Temporal Data Granularities

Regions of interest



Periods of interest



Visualization Prototypes

Sandbox

Data sketches, static data export

Campus Explorer

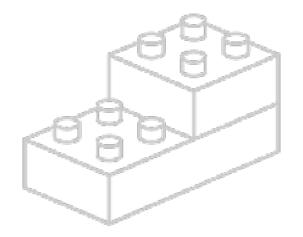
Live-data stream, cross-building analysis

Building Recent

Building Long-term

Region Compare

Time



Layout	Visual Encoding	Facet	Comparisons
	Sparkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)

Layou	t Vis	sual Encoding	Facet	Comparisons
	Sp	parkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)
	В	ox-plot-bars	Juxtaposition	Repeating patterns, trends, outliers (non-contiguous)

Layout	Visual Encoding	Facet	Comparisons
	Sparkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)
	Box-plot-bars	Juxtaposition	Repeating patterns, trends, outliers (non-contiguous)
	Confidence band line chart	Aggregation	Typical utilization profiles

Layout	Visual Encoding	Facet	Comparisons
	Sparkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)
	Box-plot-bars	Juxtaposition	Repeating patterns, trends, outliers (non-contiguous)
	Confidence band line chart	Aggregation	Typical utilization profiles
	Superimposed line chart	Superposition	Within-session patterns, outliers

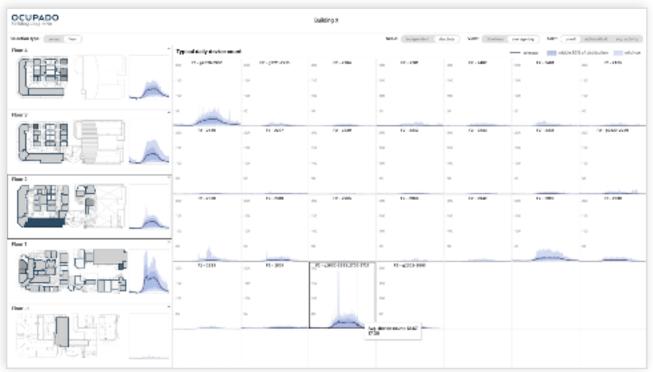
	Layout	Visual Encoding	Facet	Comparisons
Temporal		Sparkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)
		Box-plot-bars	Juxtaposition	Repeating patterns, trends, outliers (non-contiguous)
		Confidence band line chart	Aggregation	Typical utilization profiles
		Superimposed line chart	Superposition	Within-session patterns, outliers

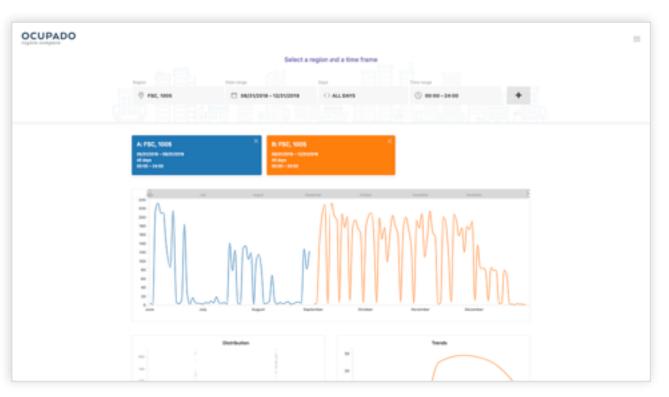
	Layout	Visual Encoding	Facet	Comparisons	
		Sparkline	Juxtaposition	Repeating patterns, trends, outliers (contiguous)	
Temporal		Box-plot-bars	Juxtaposition	Repeating patterns, trends, outliers (non-contiguous)	
		Confidence band line chart	Aggregation	Typical utilization profiles	
		Superimposed line chart	Superposition	Within-session patterns, outliers	
Spatial		Floor plan with symbols	Superposition	Within local spatial neighborhood	
		Spatial heatmap	Containment (nested)	Across distributed regions	134

Ocupado Interfaces







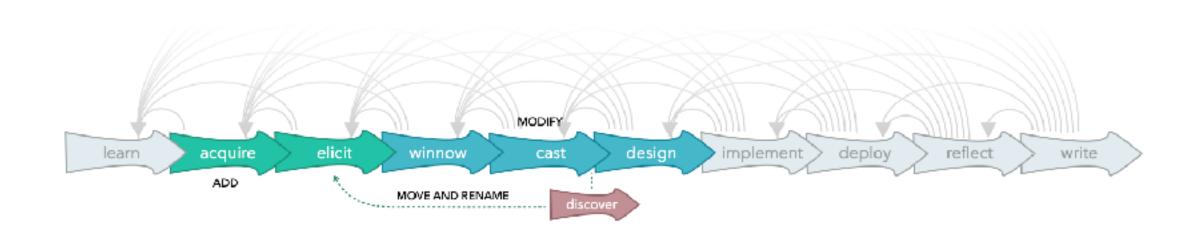


Ocupado Contributions

- 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

Michael Oppermann

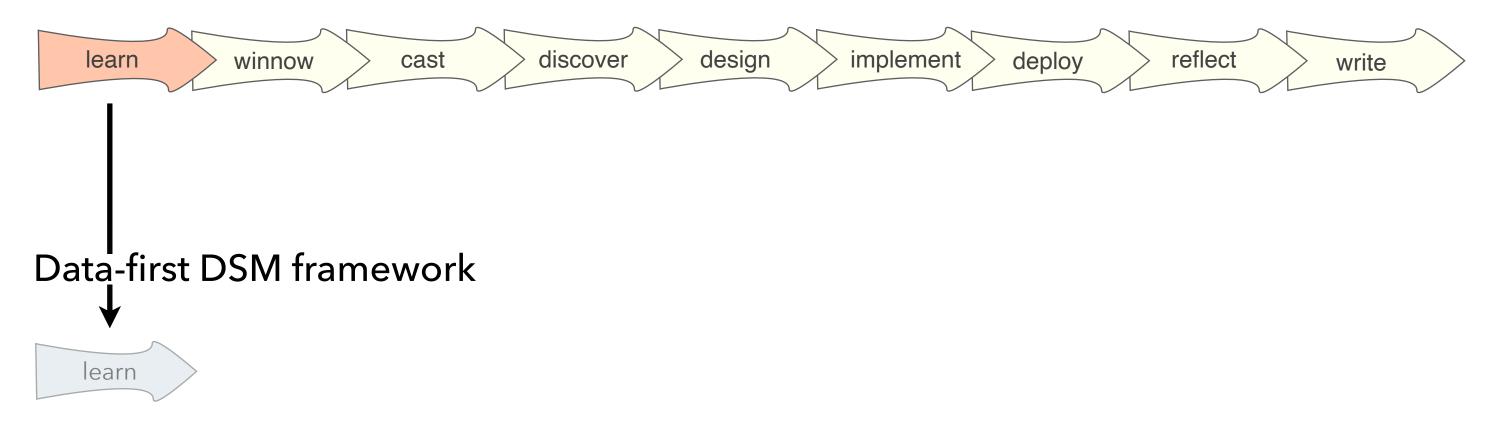




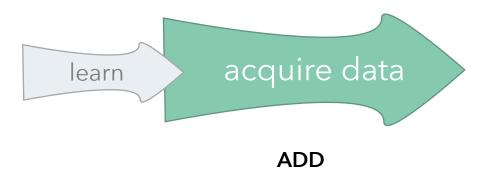
Data-First Design Studies

http://www.cs.ubc.ca/group/infovis/pubs/2020/data-first/

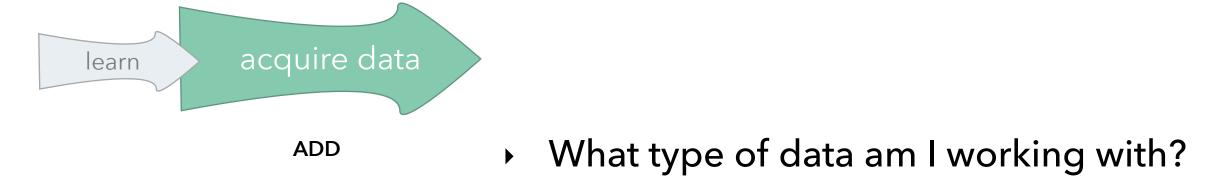










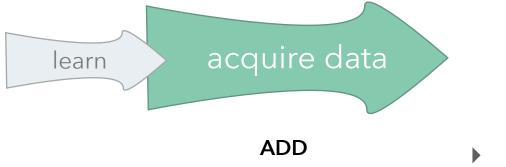






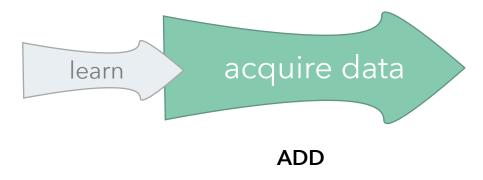
- What type of data am I working with?
- Are there any data quality challenges?



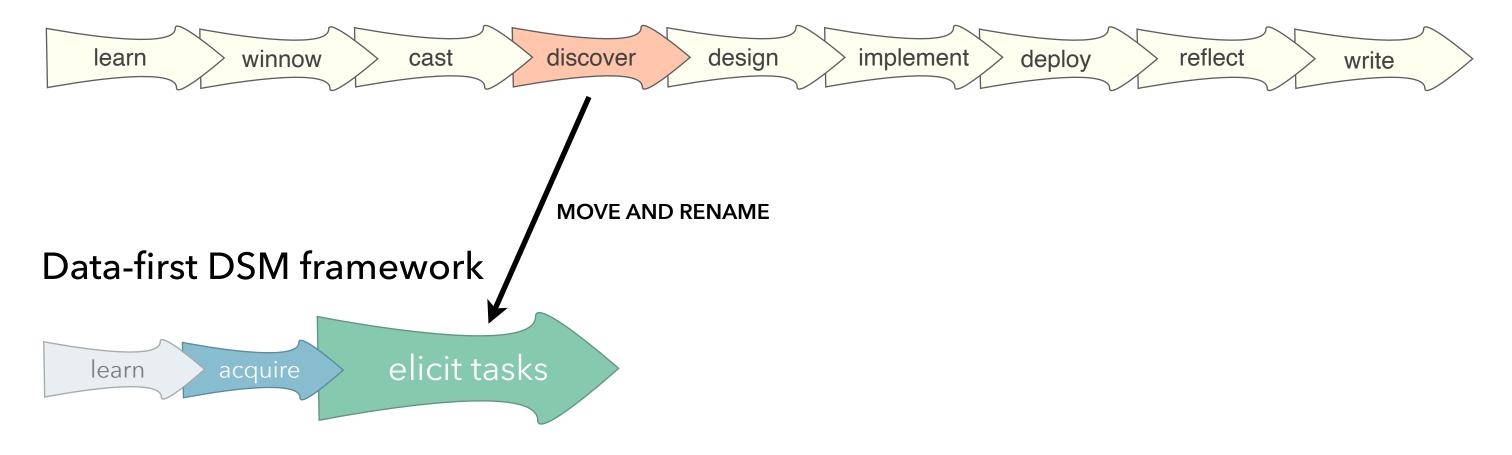


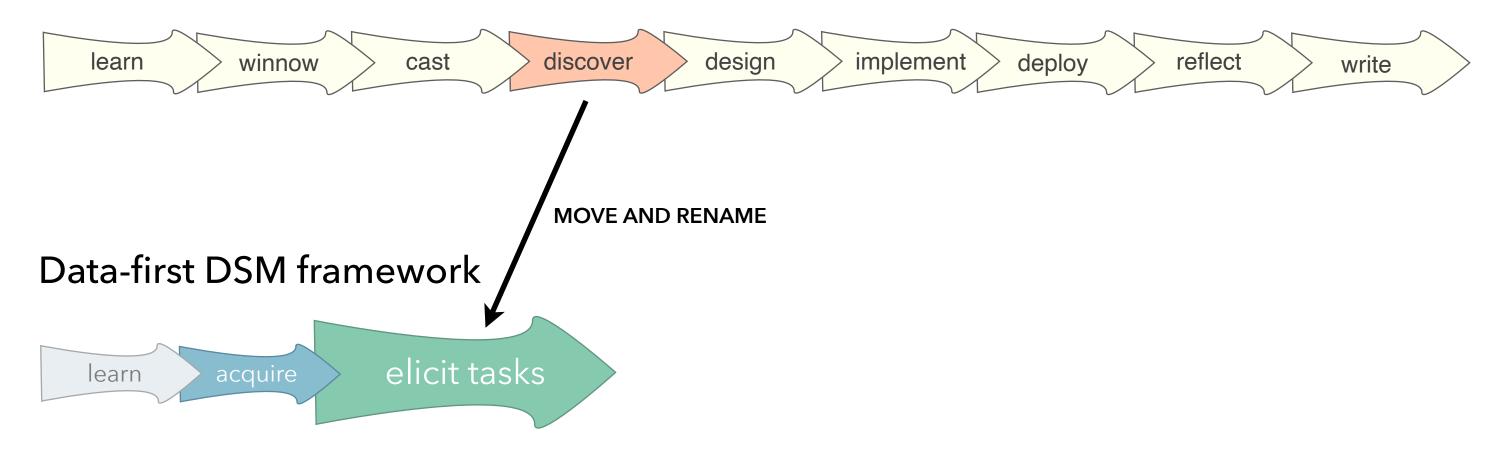
- What type of data am I working with?
- Are there any data quality challenges?
- What is special about this data?



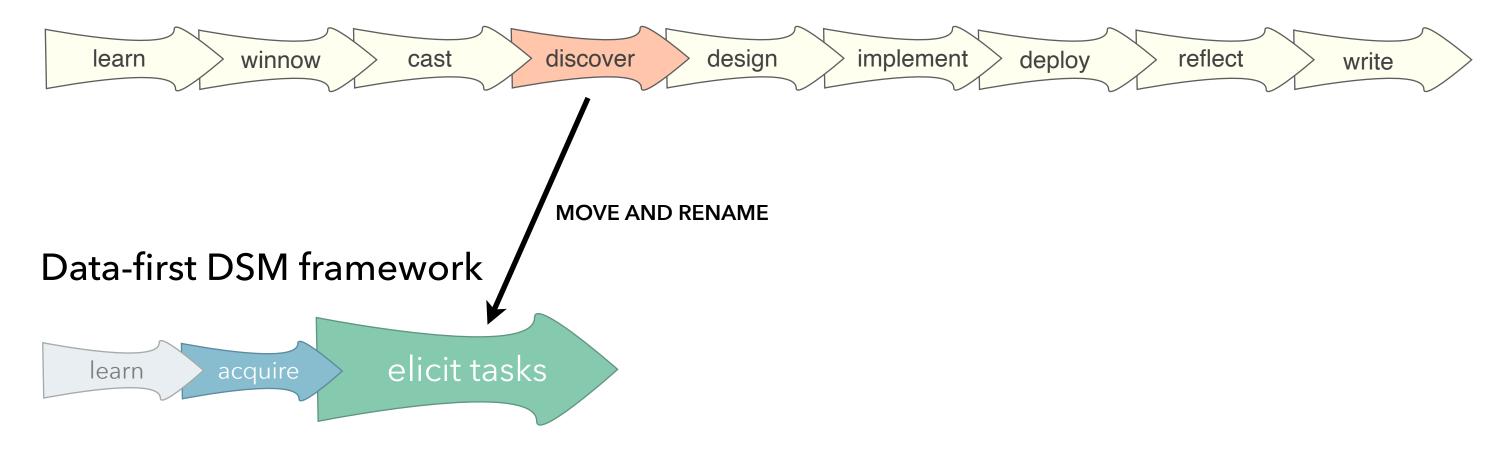


- What type of data am I working with?
- Are there any data quality challenges?
- What is special about this data?
- Who would benefit from seeing and exploring it?

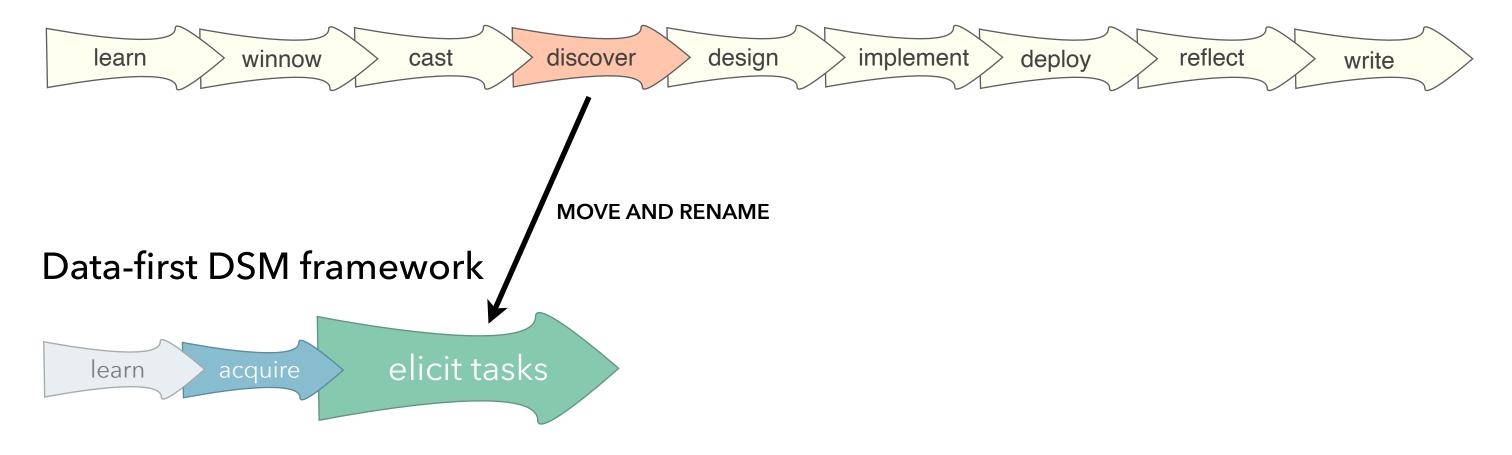




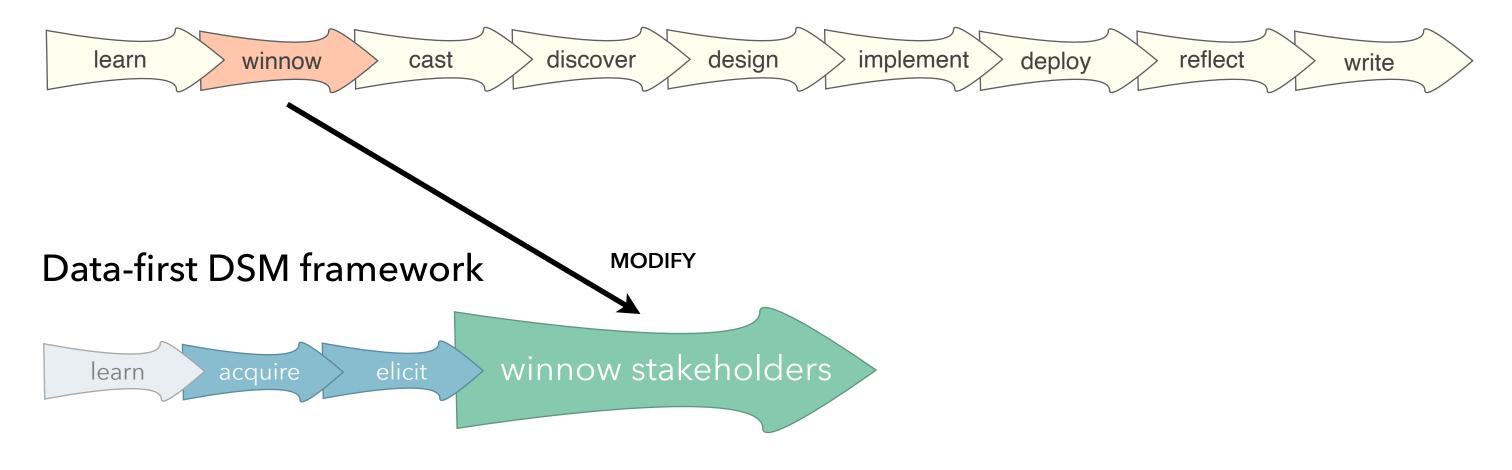
Multiple potential stakeholders

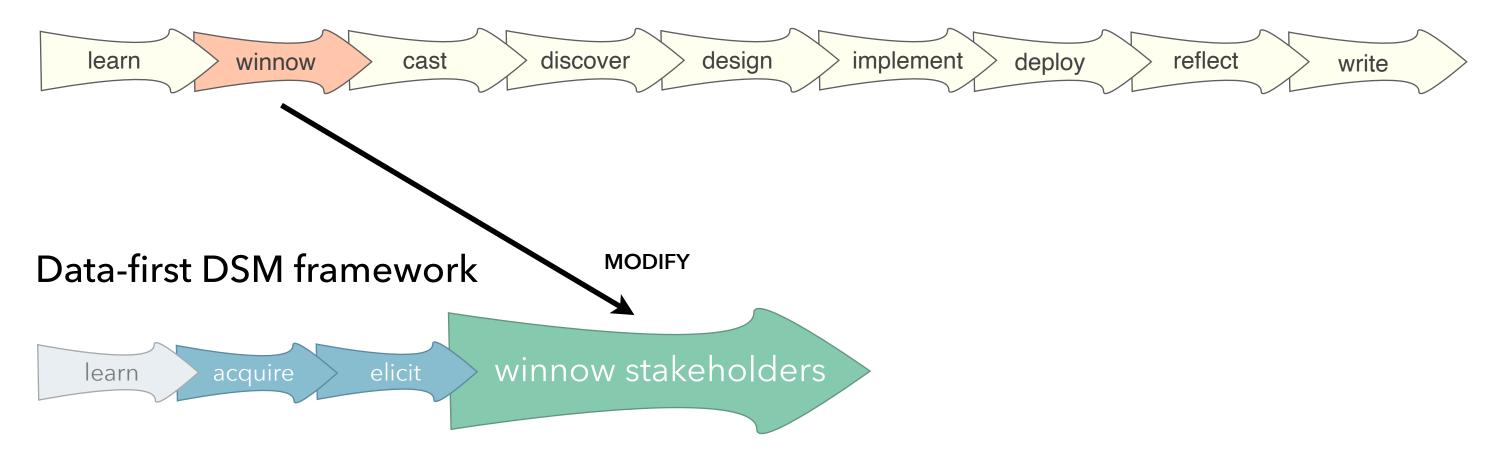


- Multiple potential stakeholders
- Explain initial data abstractions

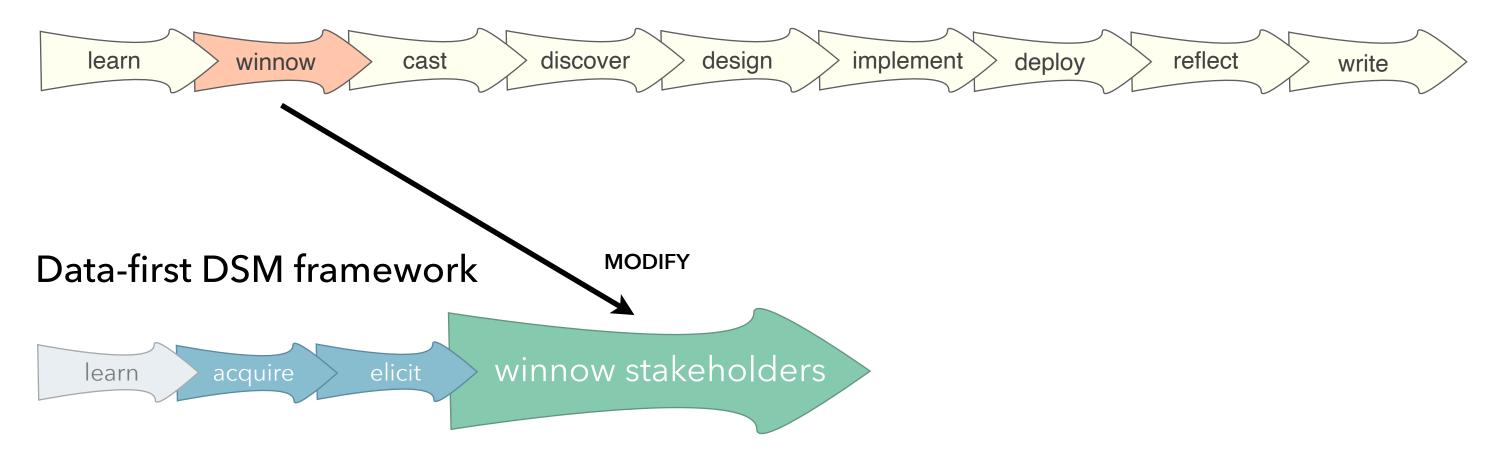


- Multiple potential stakeholders
- Explain initial data abstractions
- Learn about unsolved stakeholder needs

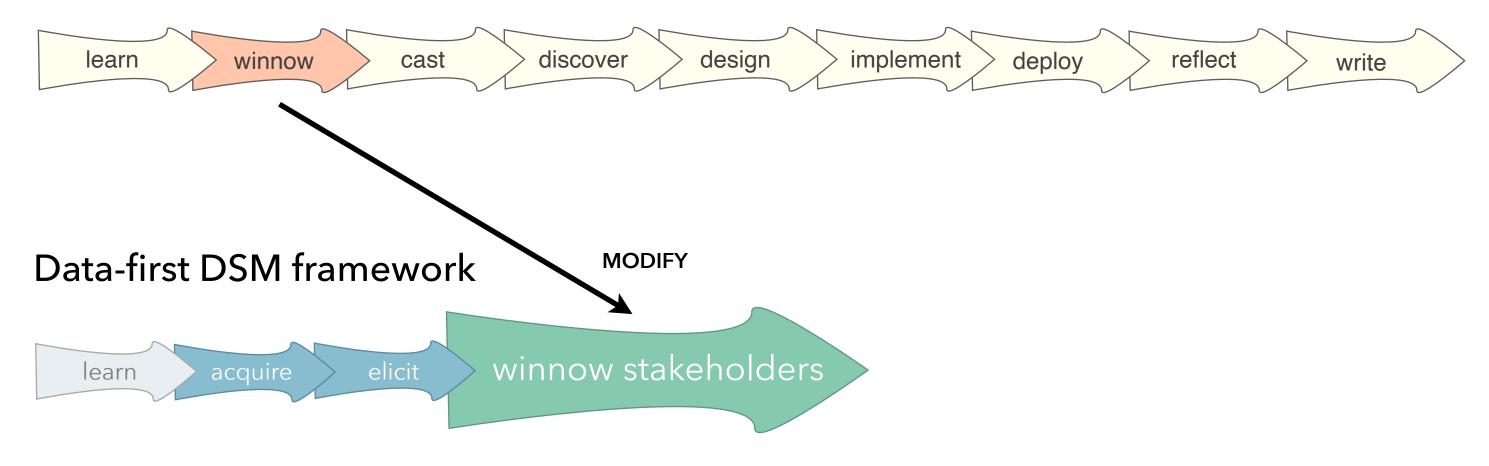




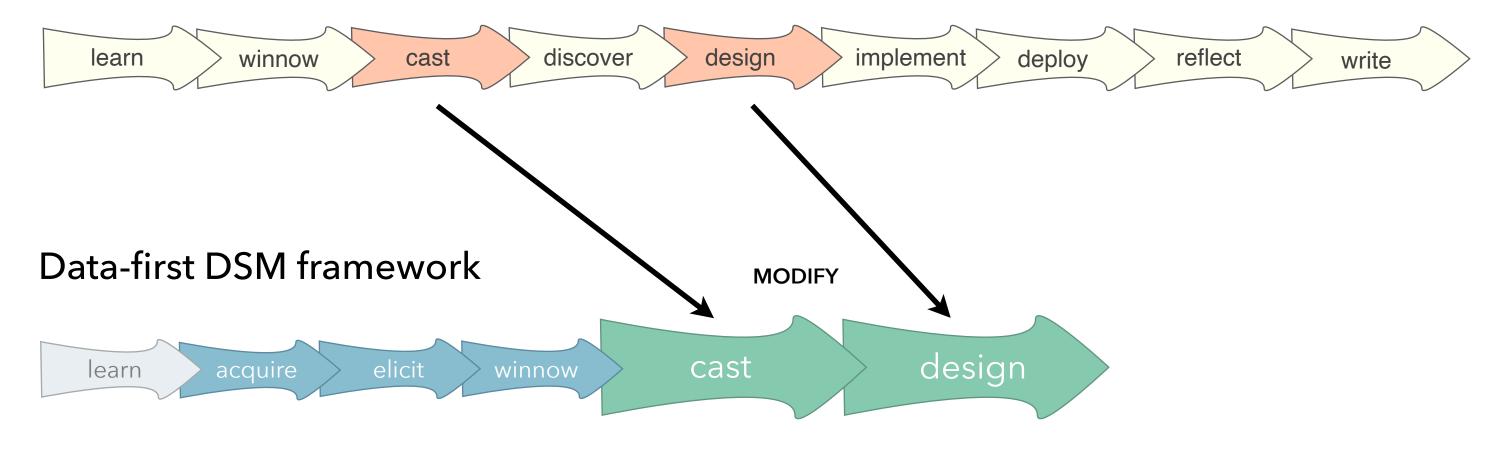
How frequent are their data-relevant tasks?

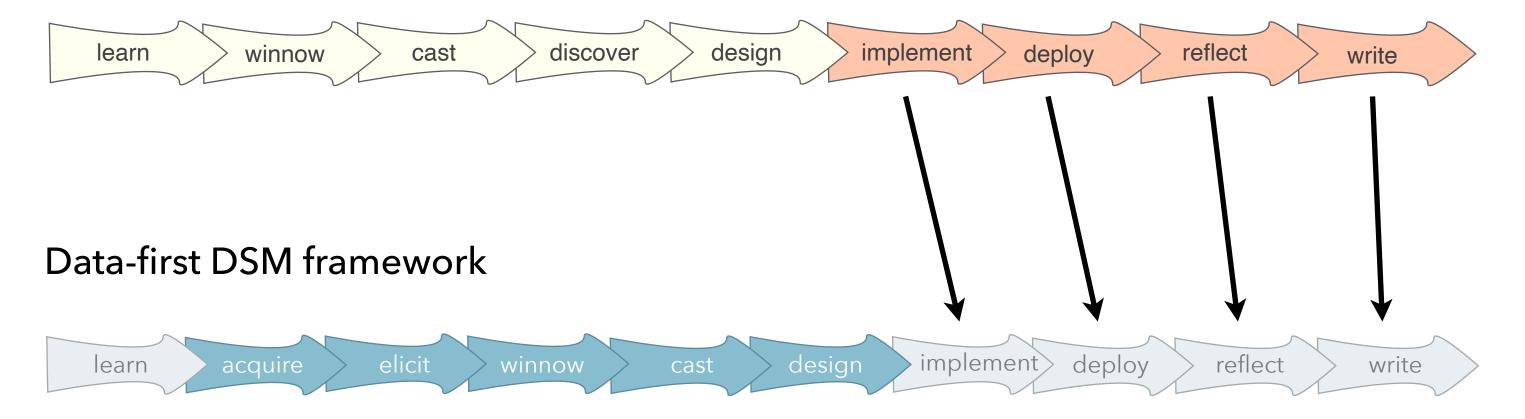


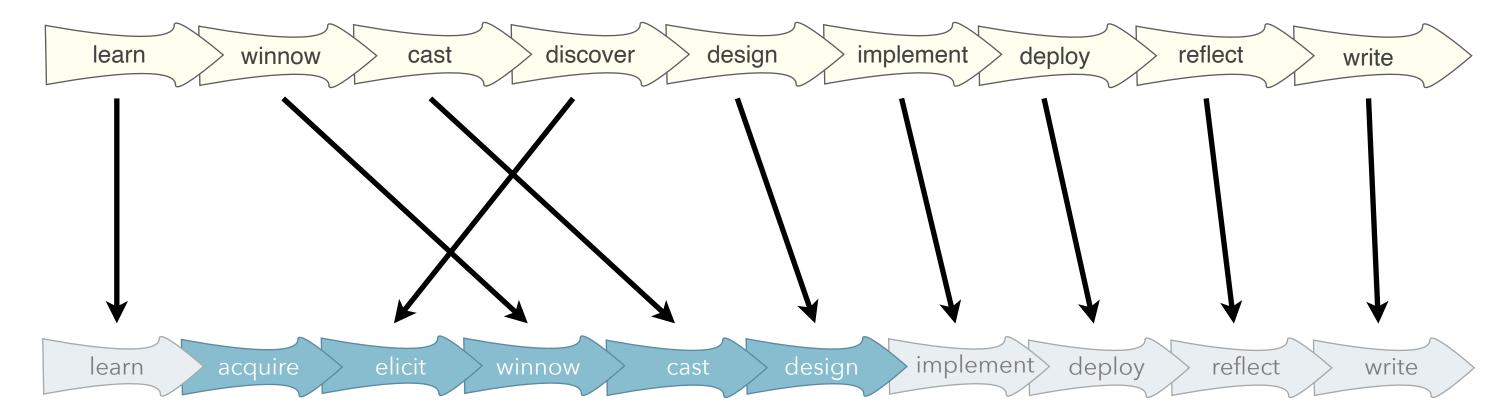
- How frequent are their data-relevant tasks?
- How central are these tasks to the stakeholder's primary mission?



- How frequent are their data-relevant tasks?
- How central are these tasks to the stakeholder's primary mission?
- How many people in the organization deal with these tasks?







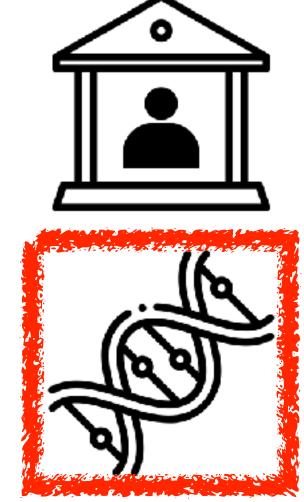
Data-first DSM framework

Three case studies of problem-driven work

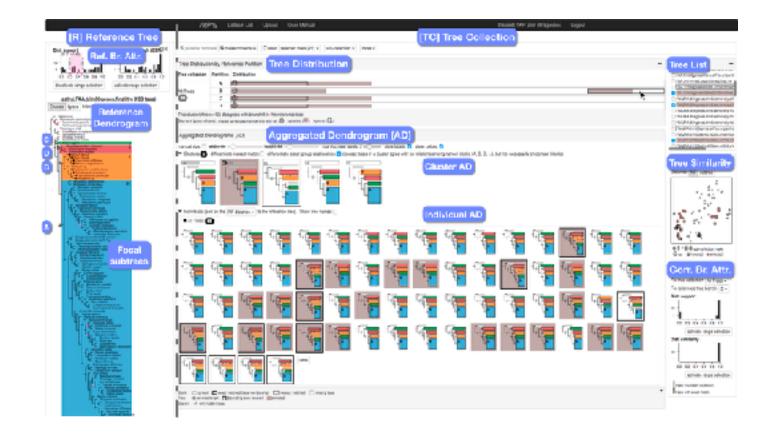
e-commerce



• facilities management



biology



Zipeng Liu



Shing Hei Zhan



Aggregated Dendrograms

for Visual Comparison Between Many Phylogenetic Trees

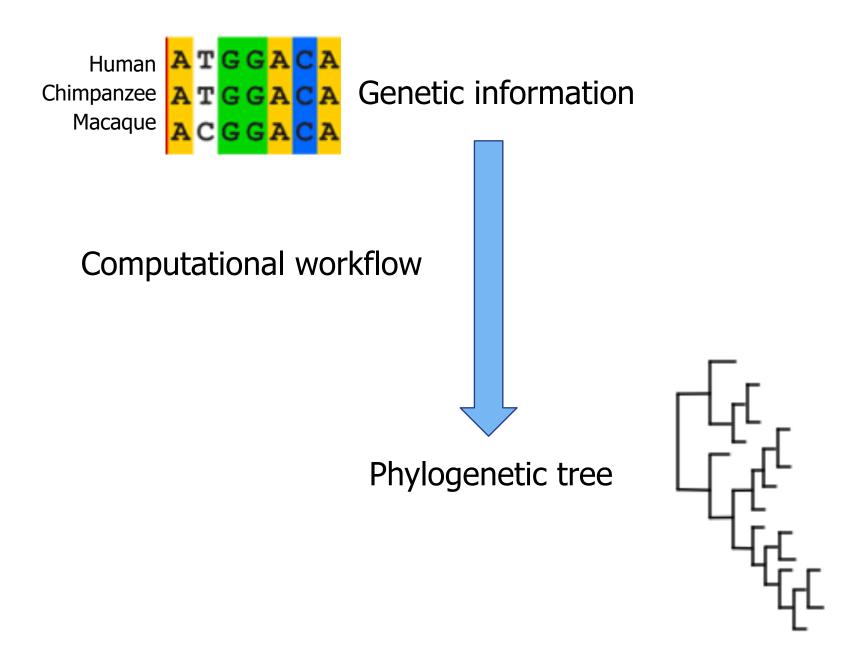
http://www.cs.ubc.ca/labs/imager/tr/2019/adview

Aggregated Dendrograms for Visual Comparison Between Many Phylogenetic Trees.

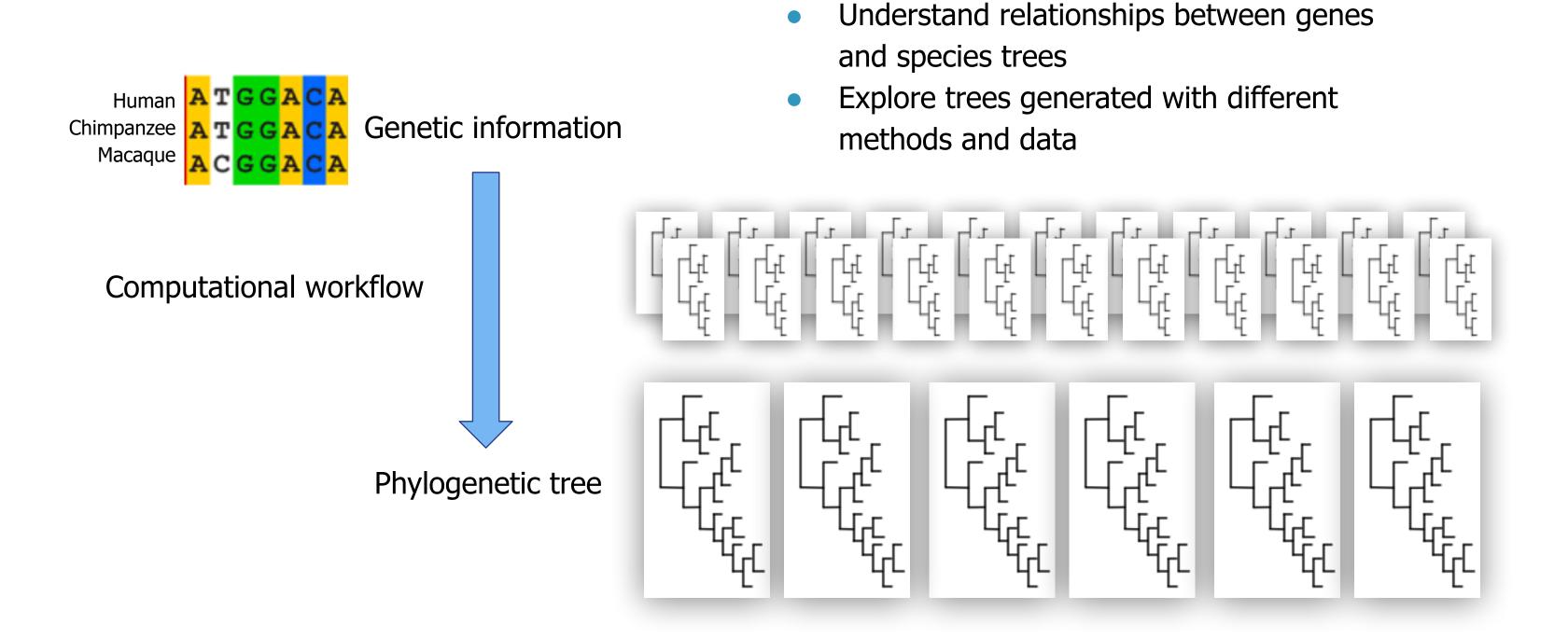
Liu, Zhan, Munzner. IEEE Trans. Visualization and Computer Graphics (TVCG) 26(9):2732-2747, 2019.

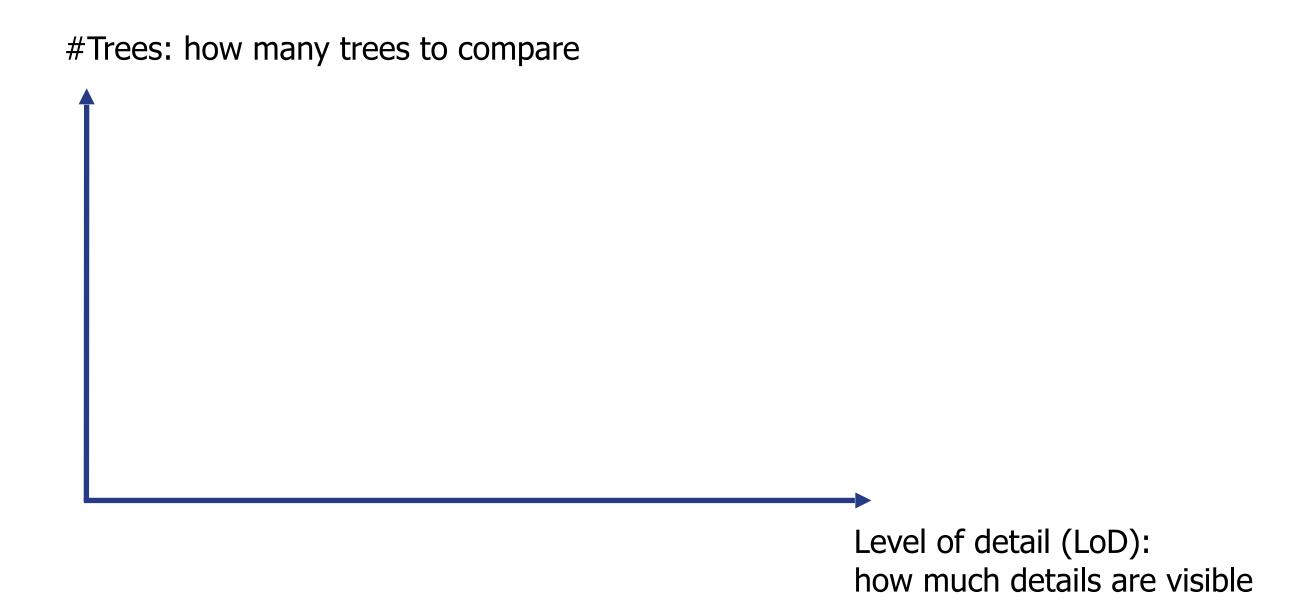
Phylogenetic tree

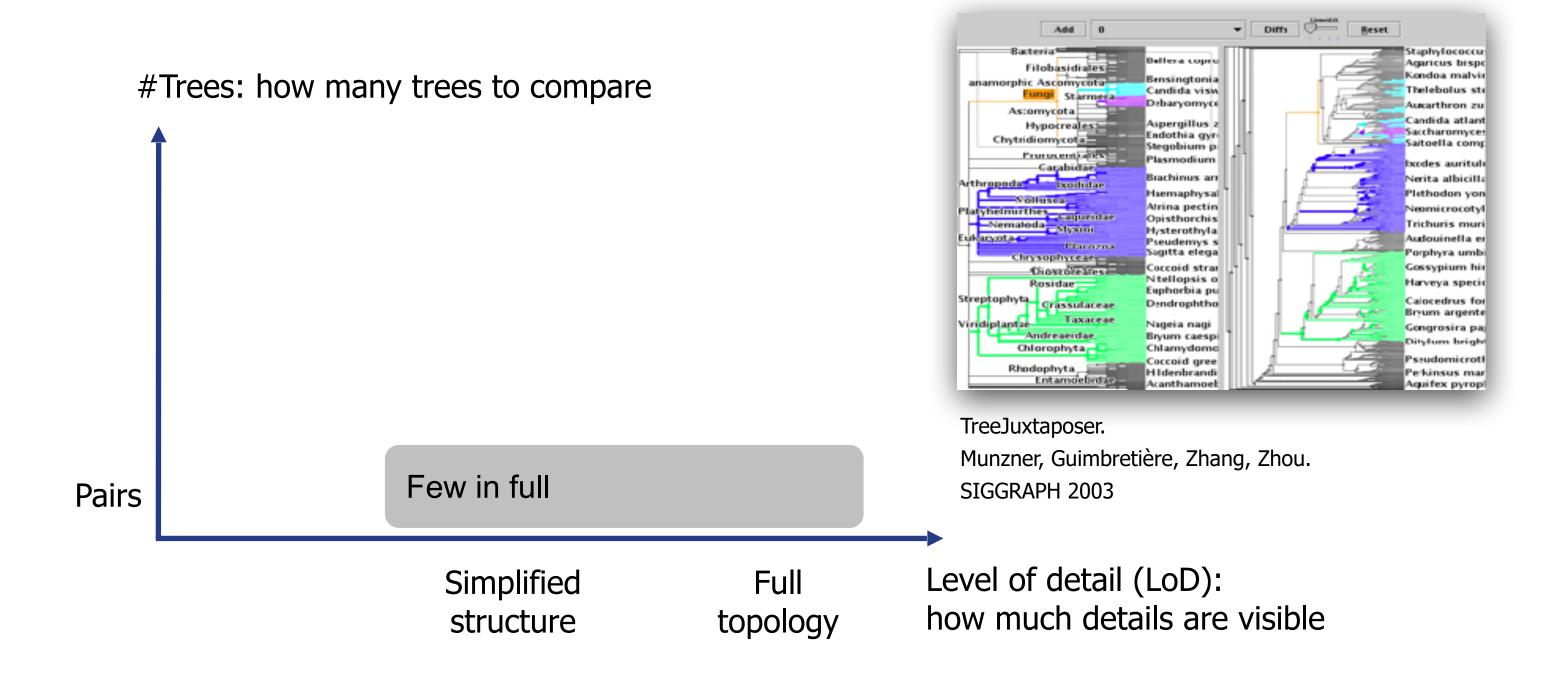
Evolutionary relationships of organisms

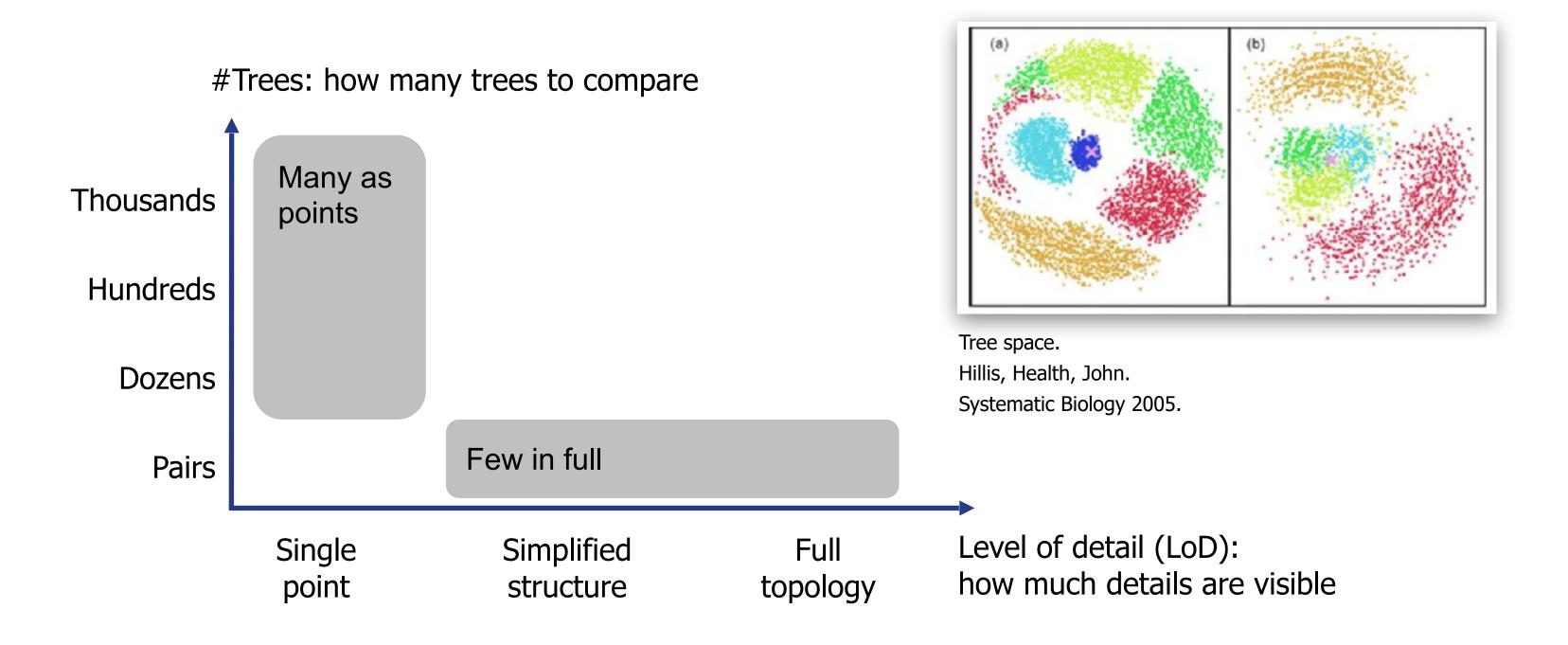


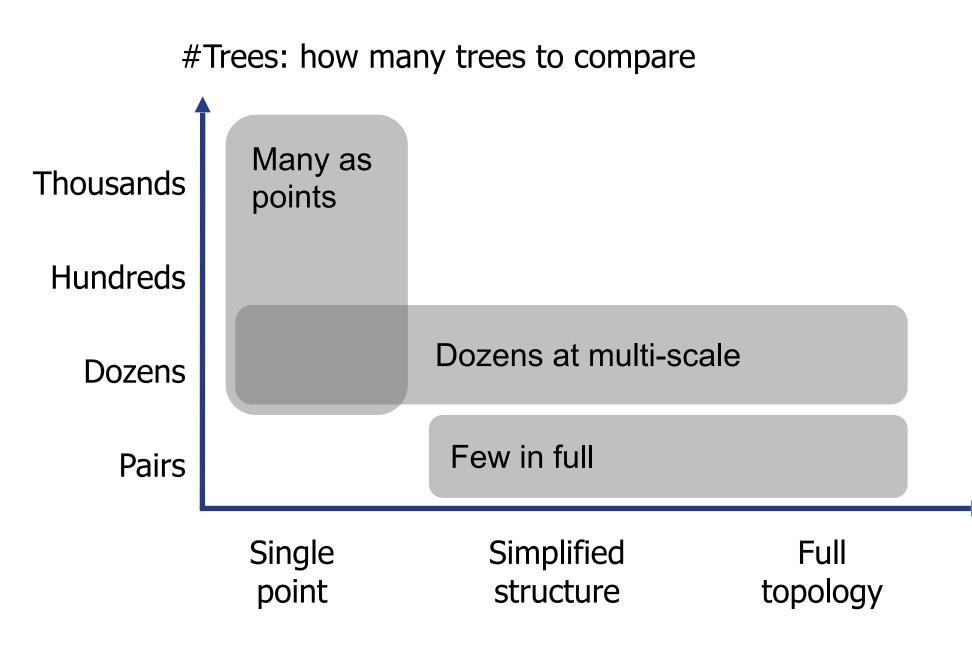
Many phylogenetic trees

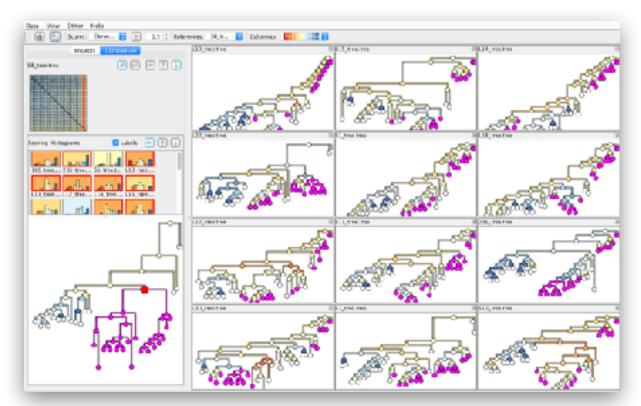












Interactive visual comparison of multiple trees.

Bremm, Landesberger, Heß, Schreck, Weil, Hamacher.

VAST 2011.

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

Comparing many phylogenetic trees

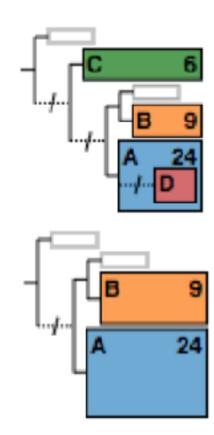
#Trees: how many trees to compare Hundreds / Many as **Thousands** points thousands at multi-scale? Hundreds Dozens at multi-scale Dozens Few in full Pairs Level of detail (LoD): Single **Simplified** Full how much details are visible topology point structure

Contributions include idiom & algorithm levels

Data and task abstractions for comparison of phylogenetic trees

Contributions include idiom & algorithm levels

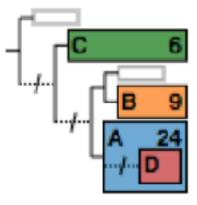
- Data and task abstractions for comparison of phylogenetic trees
- A new visual encoding: Aggregated Dendrogram
 - Compact tree representation that focuses on selected subtrees
 - Adapts to available screen space

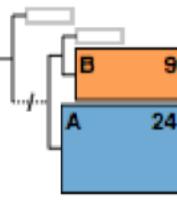


Contributions include idiom & algorithm levels

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





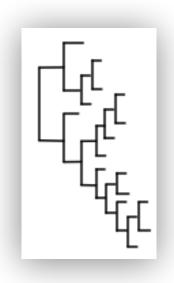


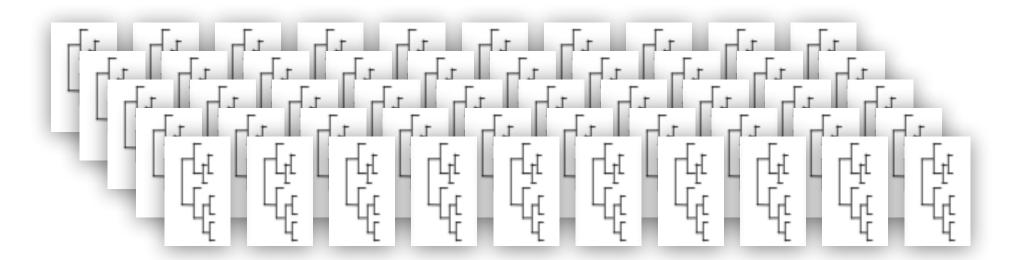
Data & Tasks

- Tree data
- Two crucial tasks

Tree data

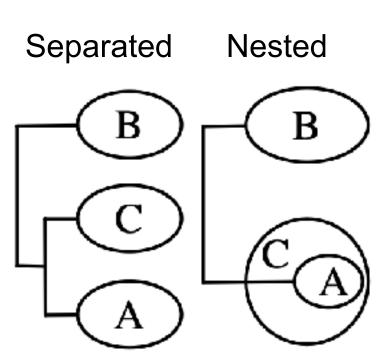
Reference tree vs. Tree collection





Topological relationships between subtrees / leaf nodes

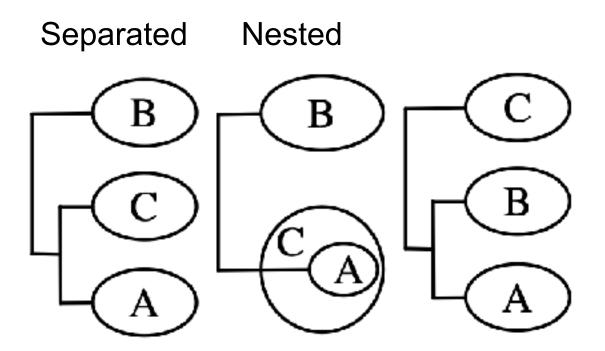
Topological relationships between subtrees / leaf nodes

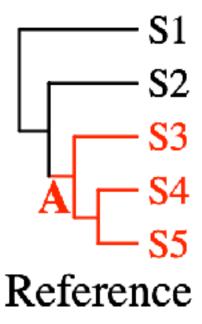


Topological relationships between subtrees / leaf nodes

Topological distance

Leaf node memberships compared to reference tree

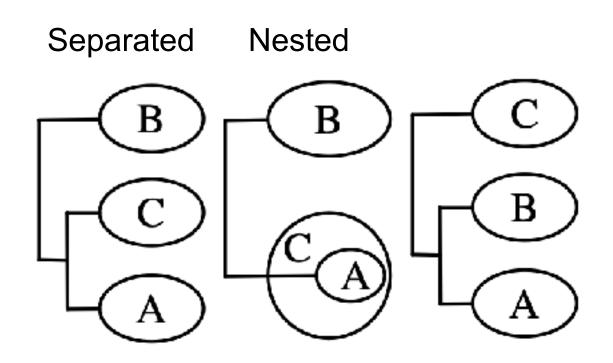


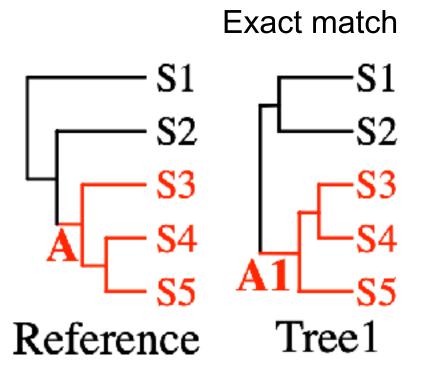


Topological relationships between subtrees / leaf nodes

Topological distance

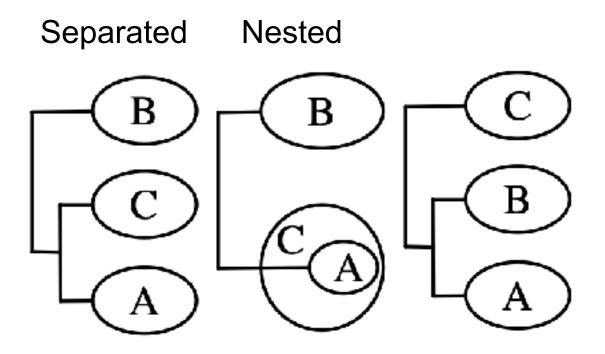
Leaf node memberships compared to reference tree



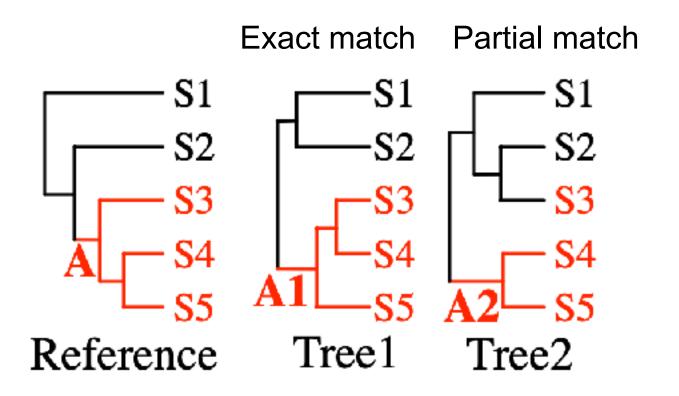


Topological relationships between subtrees / leaf nodes

Topological distance



Leaf node memberships compared to reference tree

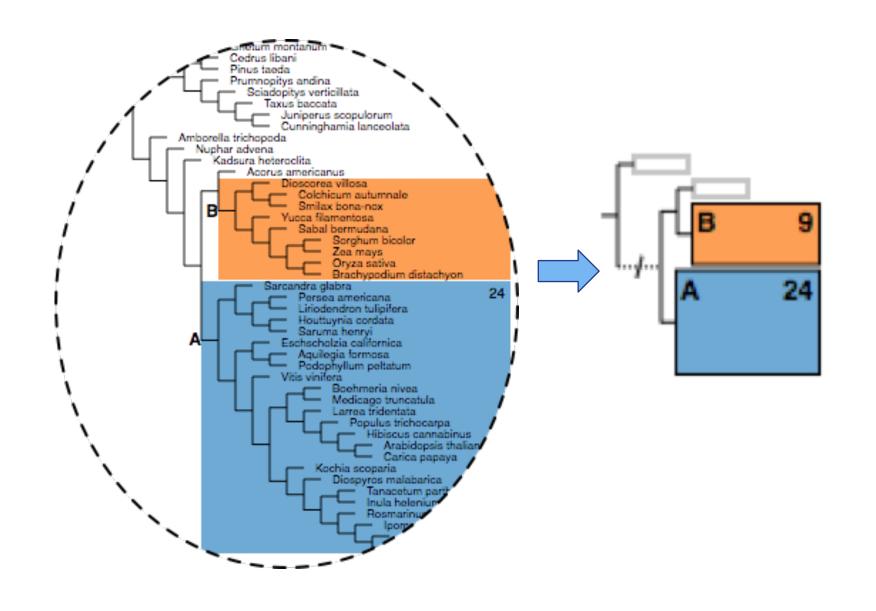


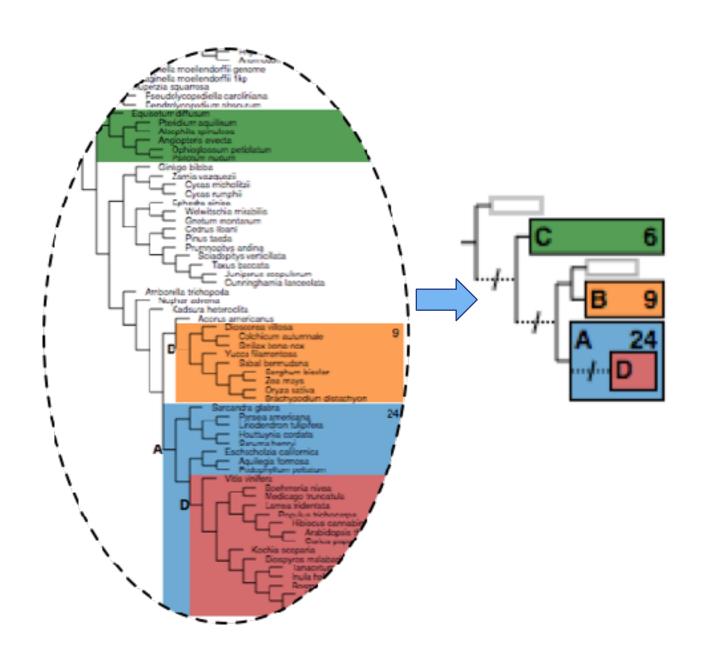
Aggregated Dendrogram (AD)

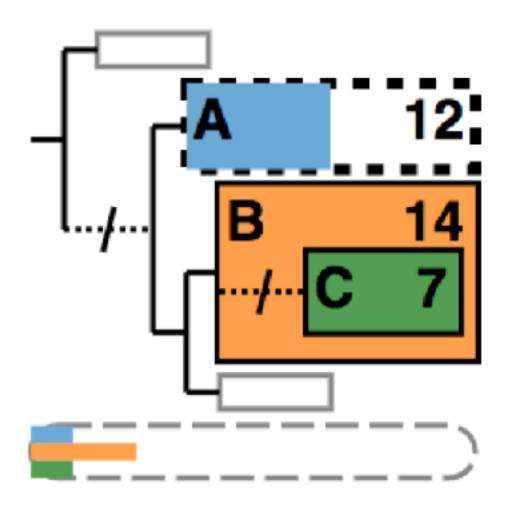
- Intuition
- Visual design

Intuition

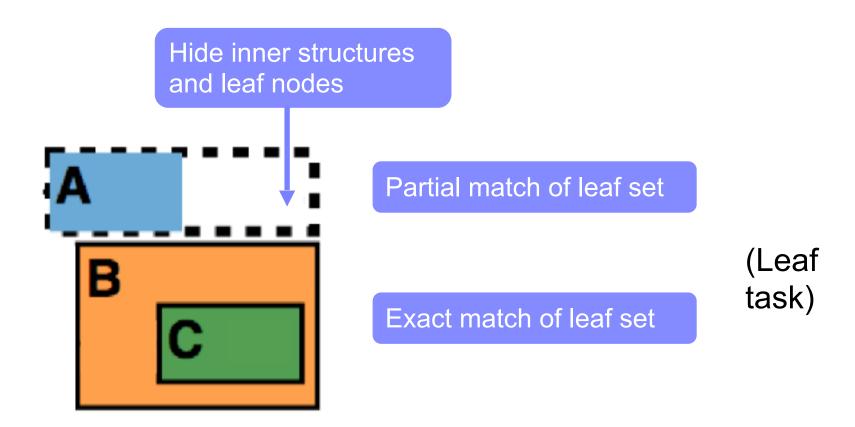
Use glyphs to compress a tree according to user selections



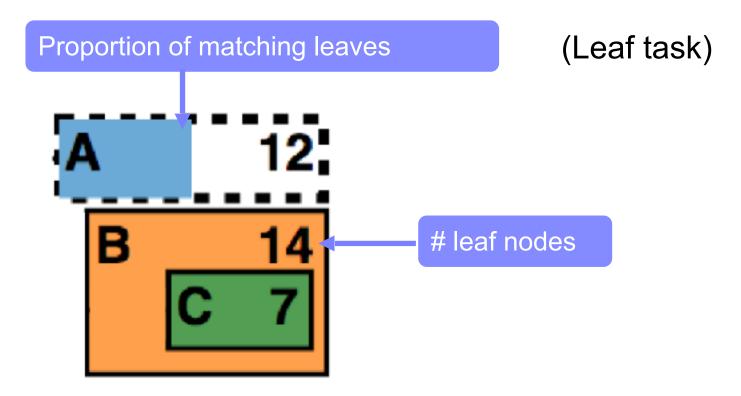




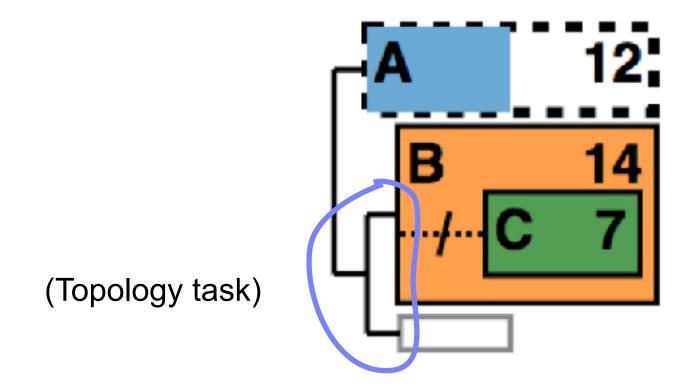
- Focus
 - Selected subtrees



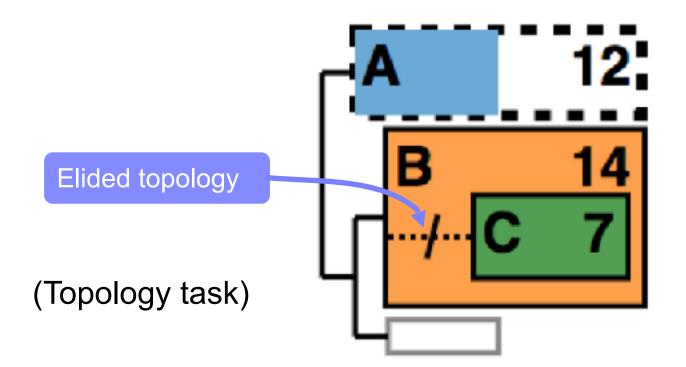
- Focus
 - Selected subtrees



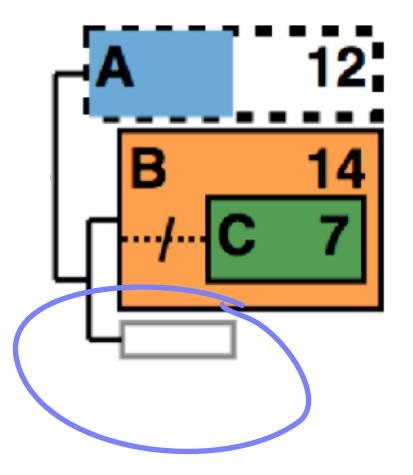
- Focus
 - Selected subtrees
 - Topological relationships between them



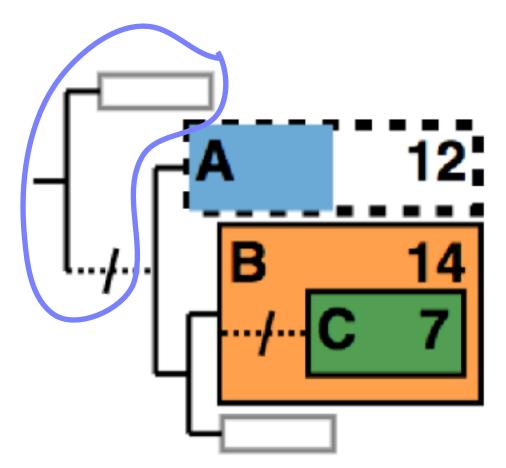
- Focus
 - Selected subtrees
 - Topological relationships between them



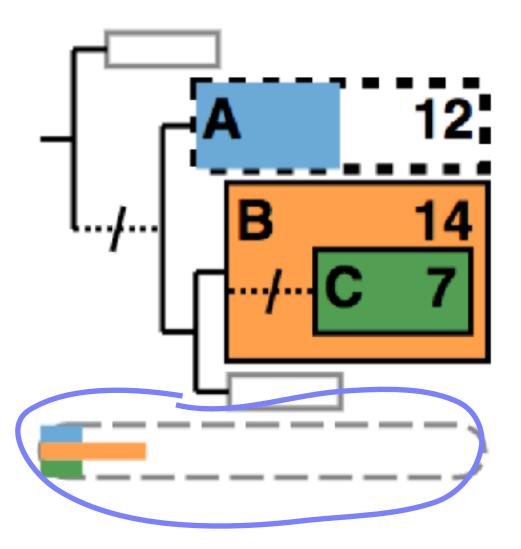
- Focus
 - Selected subtrees
 - Topological relationships between them
- Context
 - Neighboring subtrees



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

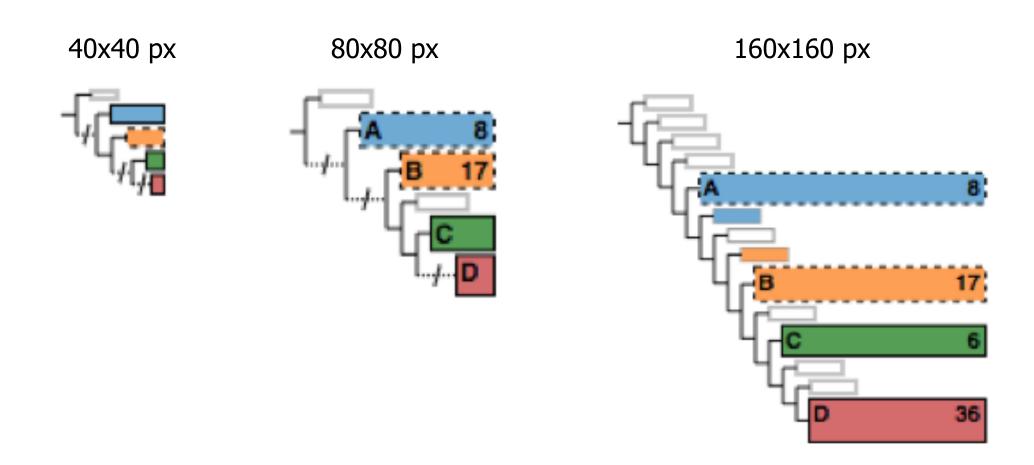


- Focus
 - Selected subtrees
 - Topological relationships between them
- Context
 - Neighboring subtrees
 - Upstream topology and root
 - Missing leaf nodes

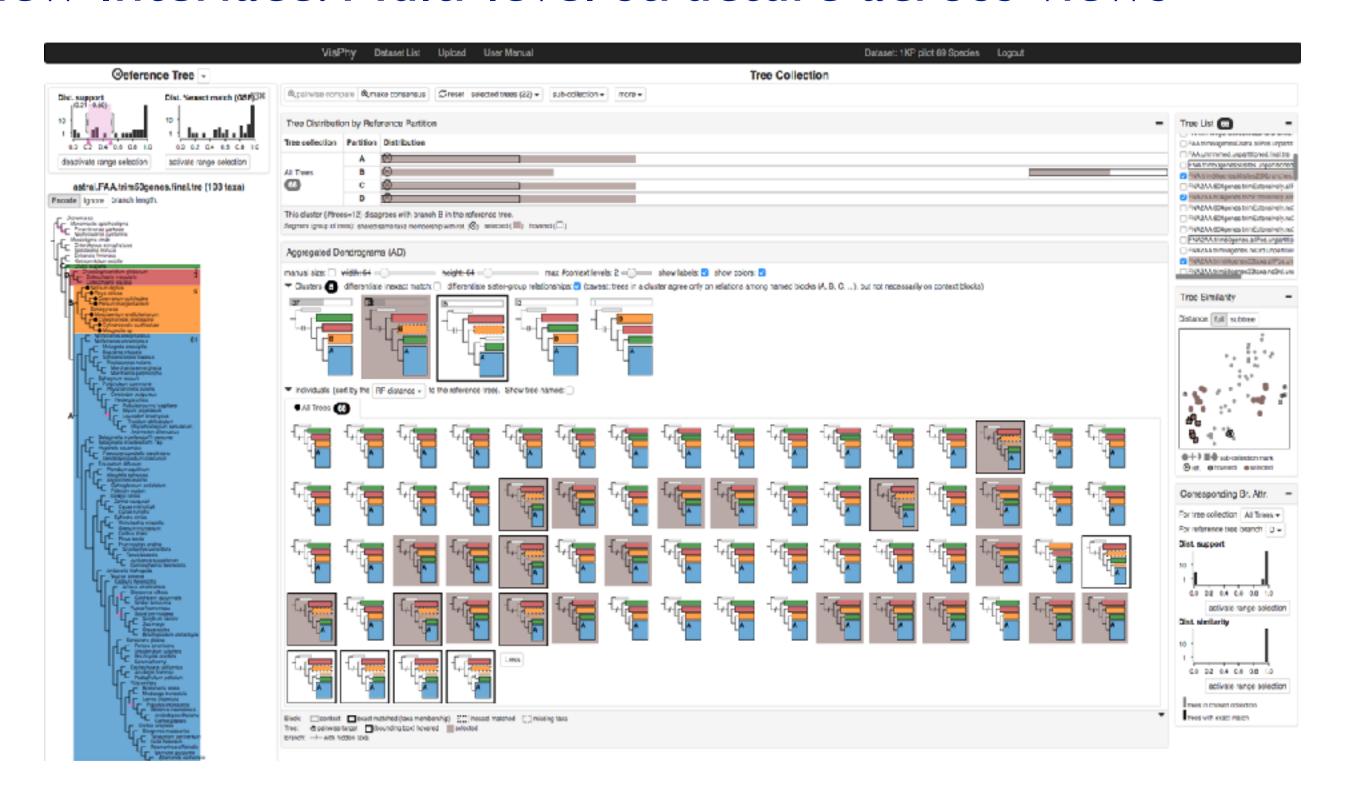


Visual design: algorithm adapts to space

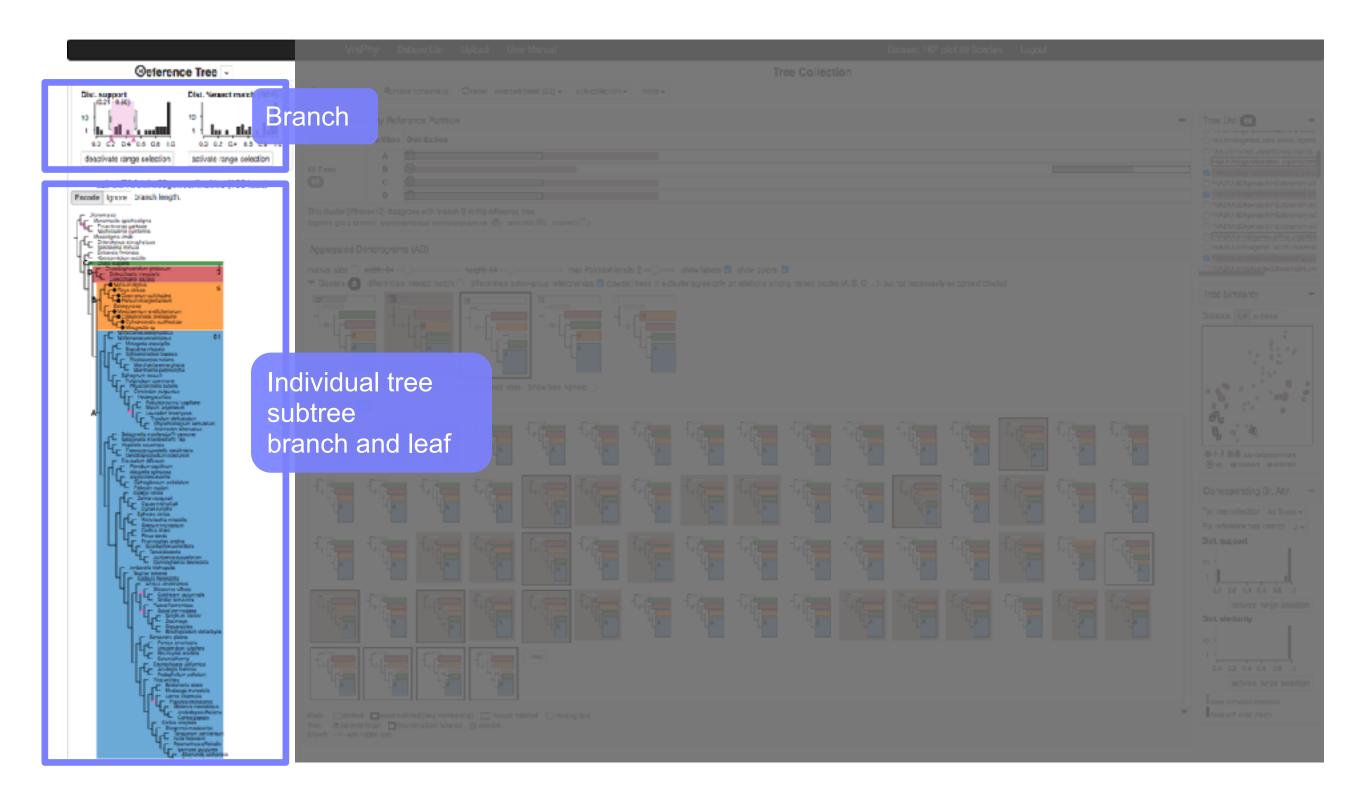
- Show more info when space permitted
 - Labels
 - #leaf nodes
 - Neighboring blocks



ADView Interface: Multi-level structure across views



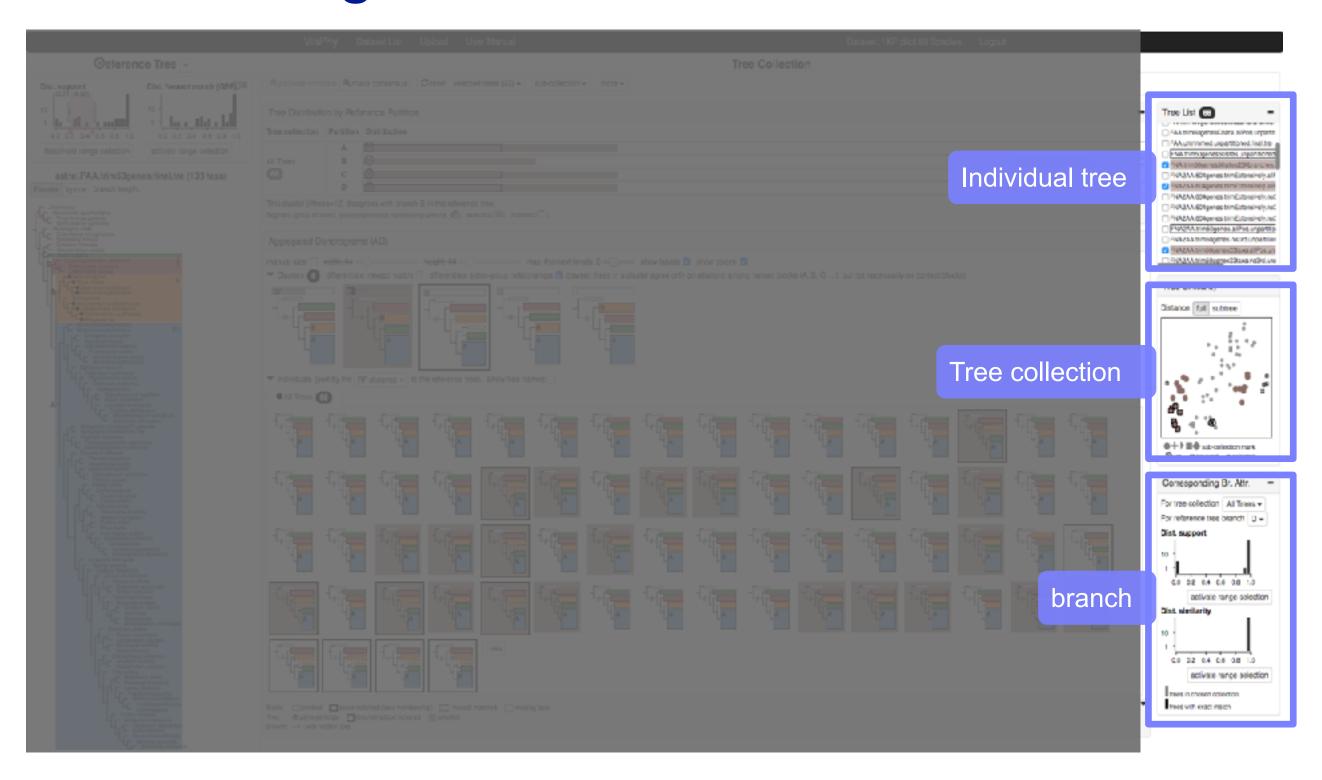
Multi-level structure across views



Interface walkthrough: tree collection main views



Interface walkthrough: tree collection aux. views

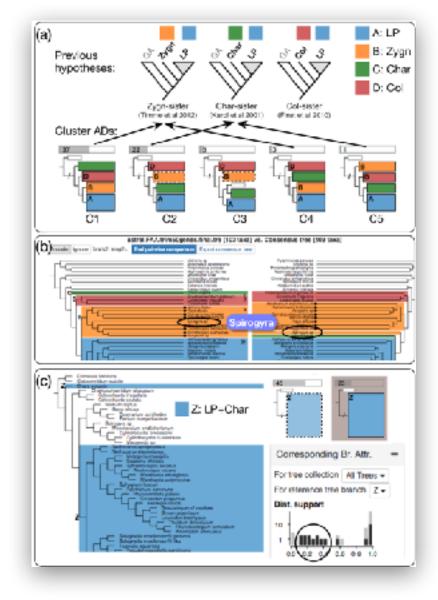


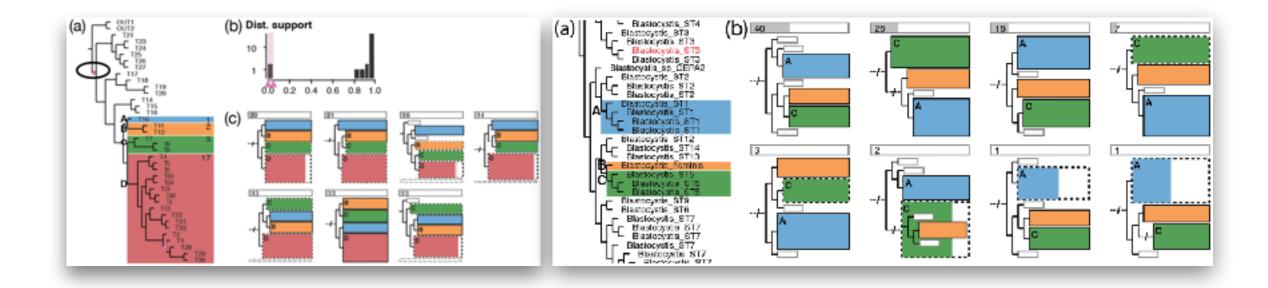
Validation with many biologists

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

Validation with many biologists

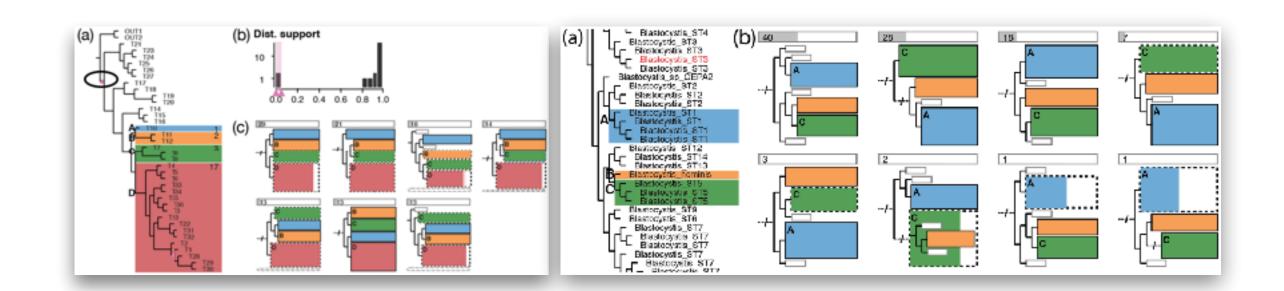
- Work closely with a biology PhD student (second author)
- Demos, interviews and discussions
 - 10 biologists at different times throughout project
- User study sessions
 - 5 biologists
 - Using their own datasets

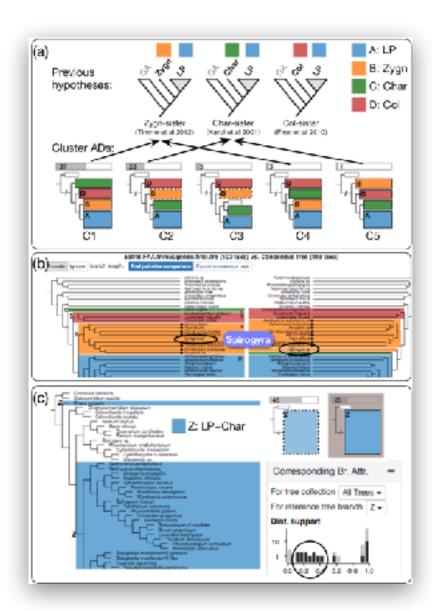




Validation with many biologists

- 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

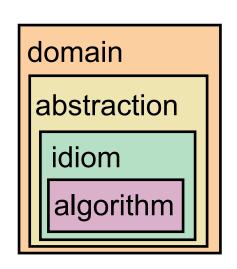




Problem-driven visualization through design studies

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

- three examples
 - different domains
 - -different methods





More information

theoretical foundations: book
 (+ tutorial/course lecture slides)

http://www.cs.ubc.ca/~tmm/vadbook

Visualization Analysis and Design.

Munzner.

AK Peters Visualization Series.

CRC Press, 2014.



• this talk

http://www.cs.ubc.ca/~tmm/talks.html#vinci21

