

Information Visualization Meets Biology: Models and Methods for Collaboration

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VIZBI 2017 Keynote ++ (at Monash Clayton)
 June 19 2017, Clayton Australia

www.cs.ubc.ca/~tmm/talks.html#clayton17

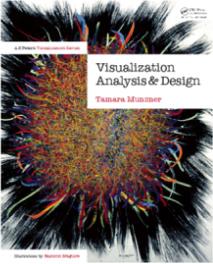
@tamaramunzner

Visualization (vis) defined & motivated

Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

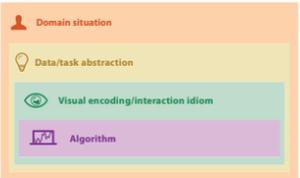
- human in the loop needs the details
 - doesn't know exactly what questions to ask in advance
 - longterm exploratory analysis
 - presentation of known results
 - stepping stone towards automation: refining, trustbuilding
- external representation: perception vs cognition
- intended task, measurable definitions of effectiveness



more at:
 Visualization Analysis and Design, Chapter 1.
 Munzner. AK Peters Visualization Series, CRC Press, 2014.

A Nested Model

for Visualization Design and Validation

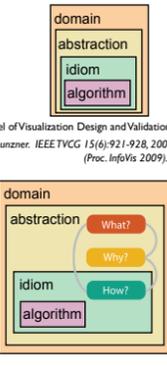


<http://www.cs.ubc.ca/labs/imager/tr/2009/NestedModel>

A Nested Model for Visualization Design and Validation.
 Munzner. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 09), 15(6):921-928, 2009.

Vis analysis framework: Four levels, three questions

- **domain situation**
 - who are the target users? what are their needs & concerns?
- **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



more at:
 Visualization Analysis and Design, Ch 2/3/4. Munzner, CRC Press, 2014.

Why is validation difficult?

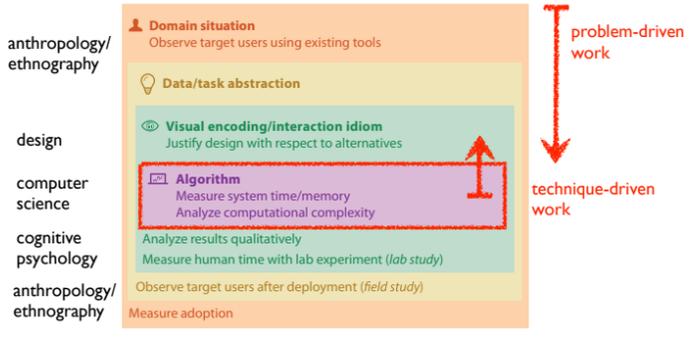
- different ways to get it wrong at each level



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

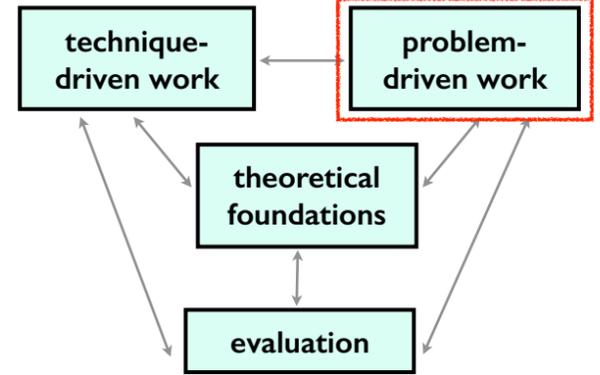
Validation solution: use methods from appropriate fields at each level

- avoid mismatches!



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

Angles of attack: My own research agenda



Vis meets bio

- biology encompasses many rich application domain for vis collaboration
 - challenging multi-level problems that won't be automated away any time soon
 - complex tasks, complex datasets
 - many points where human-in-the-loop decision-making could bear fruit
- landscape of possible tools
 - axis from eureka to speedup
 - sexy use case: eureka moment
 - enable what was impossible before: vis tools for new insights & discoveries
 - workhorse use case: workflow speedup
 - vis tools to accelerate what you're already doing
 - sometimes enables the previously infeasible
 - axis from targeted to address specific pain points, to general purpose for broad use

Collaboration incentives: Bidirectional

- what's in it for bio?
 - bio win: access to more suitable tools, can do better/faster/cheaper science
 - time spent could pay off with earlier access and/or more customized tools
- what's in it for vis?
 - vis win: access to better understanding of your driving problems
 - crucial element in building effective tools to help
 - opportunities to observe how you use them
 - if they're good enough, vis win: research success stories
 - leads us to develop guidelines on how to build better tools in general
 - vis win: research progress in visualization
 - [The Computer Scientist as Toolsmith II, Fred Brooks, CACM 30(3):61-68 1996]

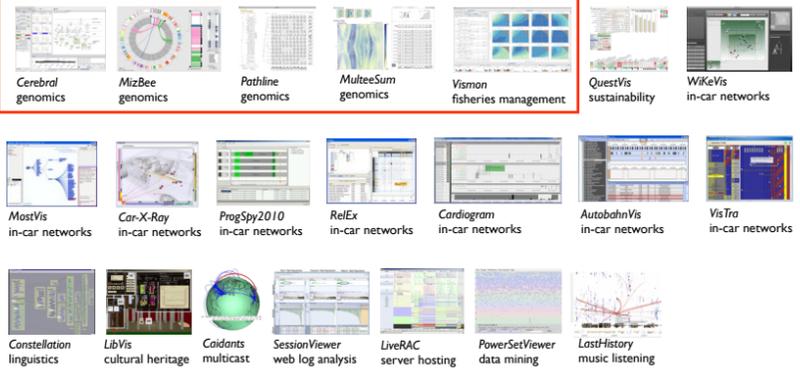
Design Study Methodology
 Reflections from the Trenches and from the Stacks

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

Michael Sedlmair
 Miriah Meyer
 Tamara Munzner

Design Study Methodology: Reflections from the Trenches and from the Stacks.
 Sedlmair, Meyer, Munzner. IEEE Trans. Visualization and Computer Graphics 18(12):2431-2440, 2012 (Proc. InfoVis 2012).

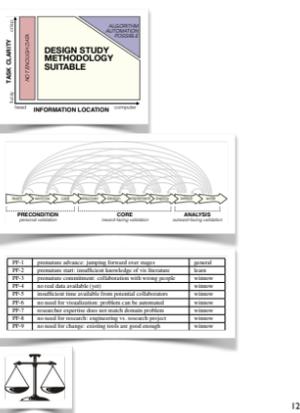
Design Studies: Lessons learned after 21 of them



- commonality of representations cross-cuts domains!

Methodology for problem-driven work

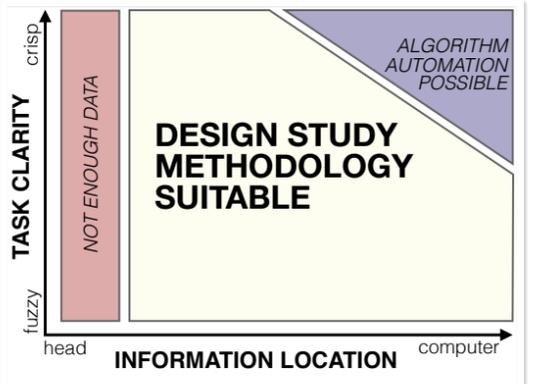
- definitions
- 9-stage framework
- 32 pitfalls & how to avoid them
- comparison to related methodologies



Design study methodology: definitions

- design studies: problem-driven work
 - in collaboration with target users
 - real data, real tasks
 - intensive requirements analysis
 - iterative refinement
 - rapid prototyping
 - deploy tools/systems to target users
 - typical evaluation: field studies
 - case studies provide evidence of utility for target users
 - replicate known results quickly/easily: show workflow speedup
 - examples of new results found using tool

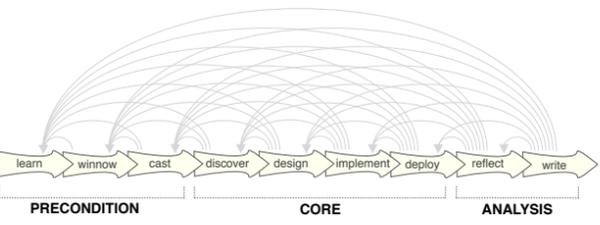
Design study methodology: definitions



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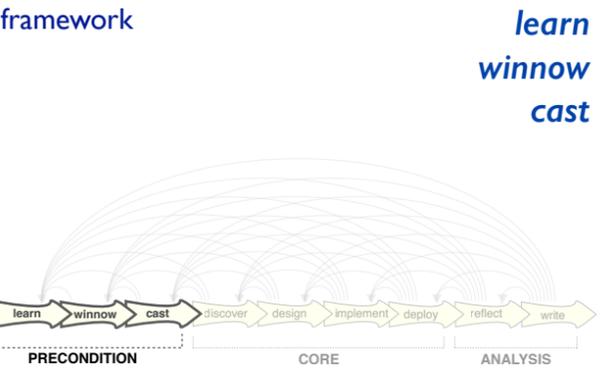
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9 stage framework

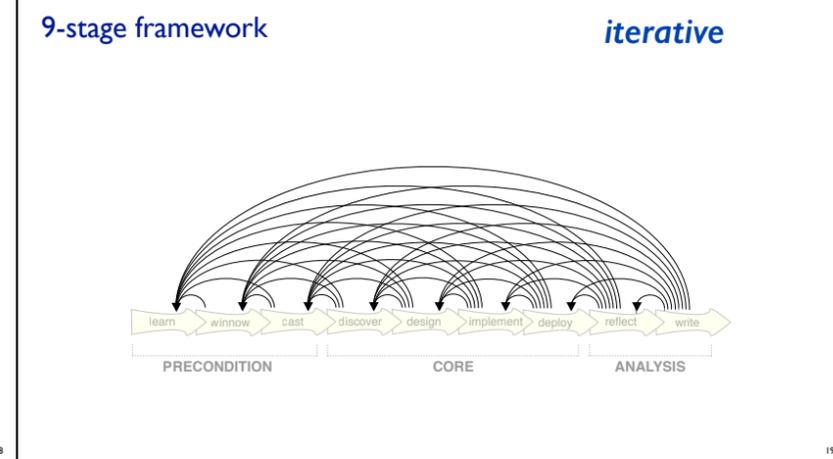
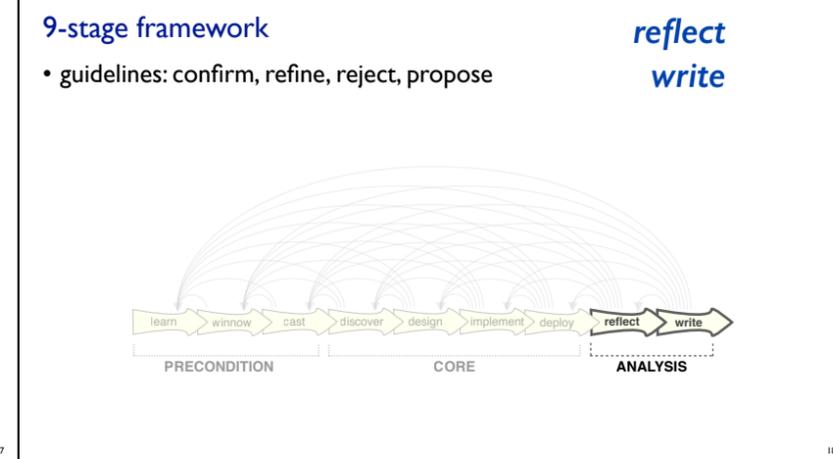
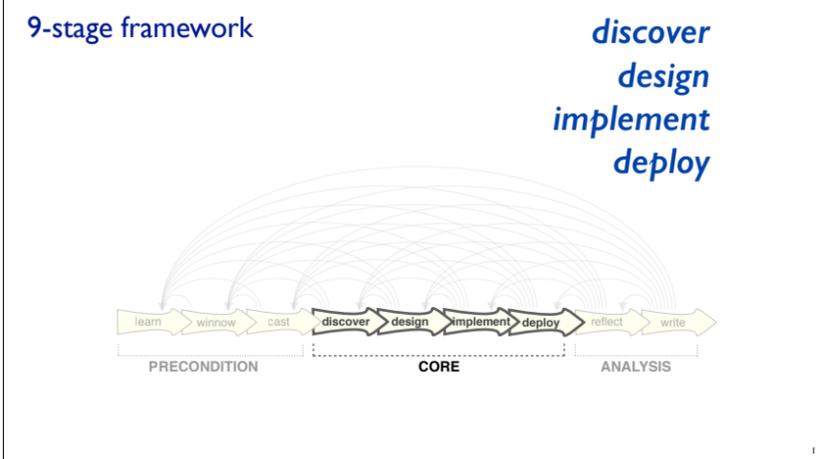


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9-stage framework



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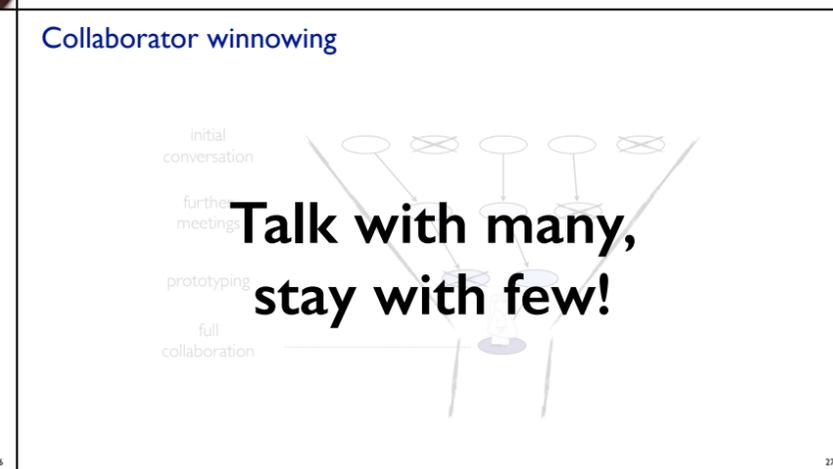
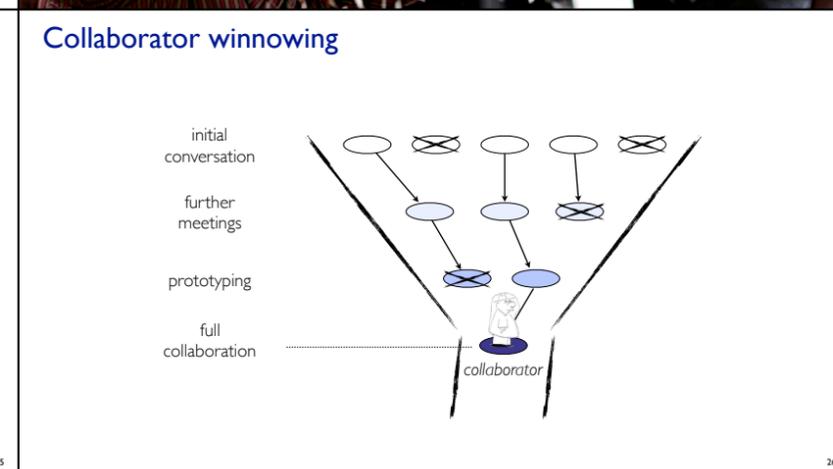
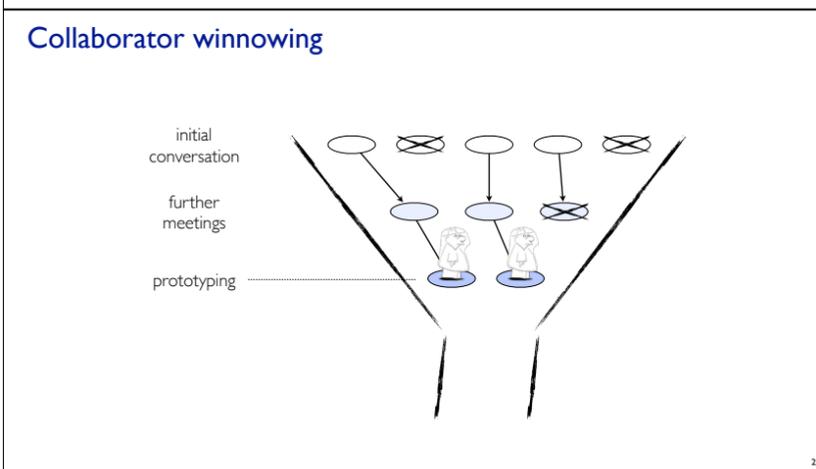
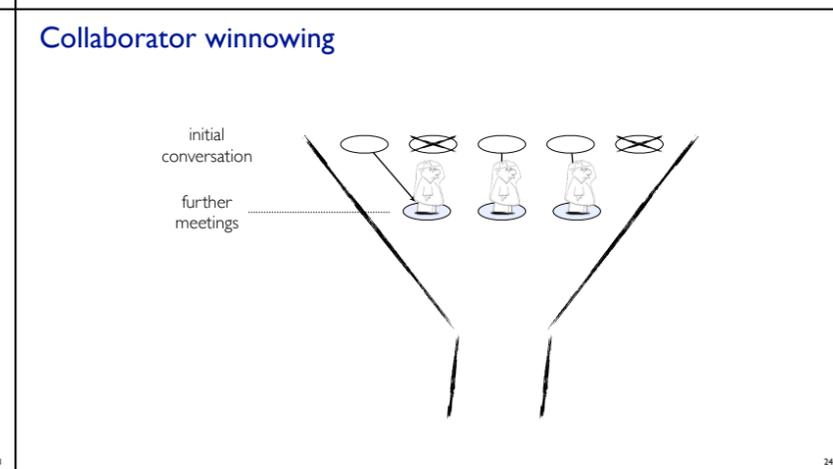
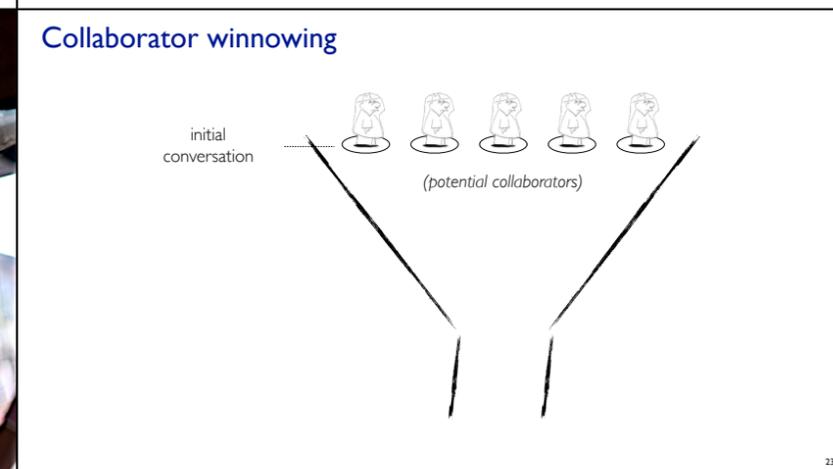
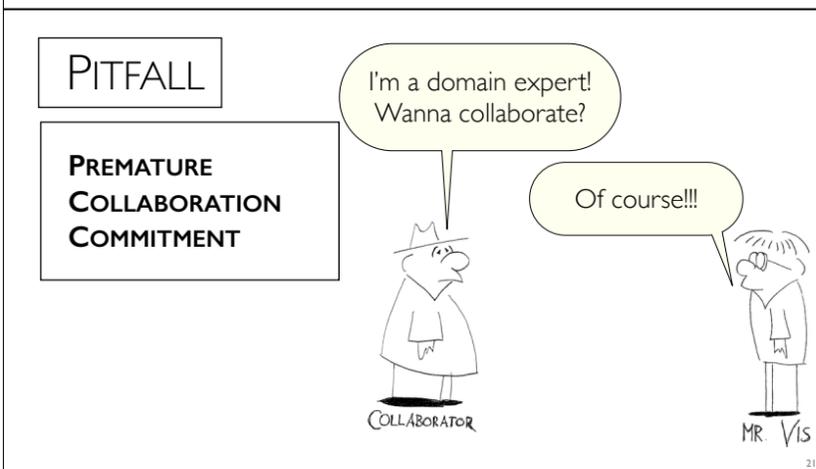


Design study methodology: 32 pitfalls

- and how to avoid them

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

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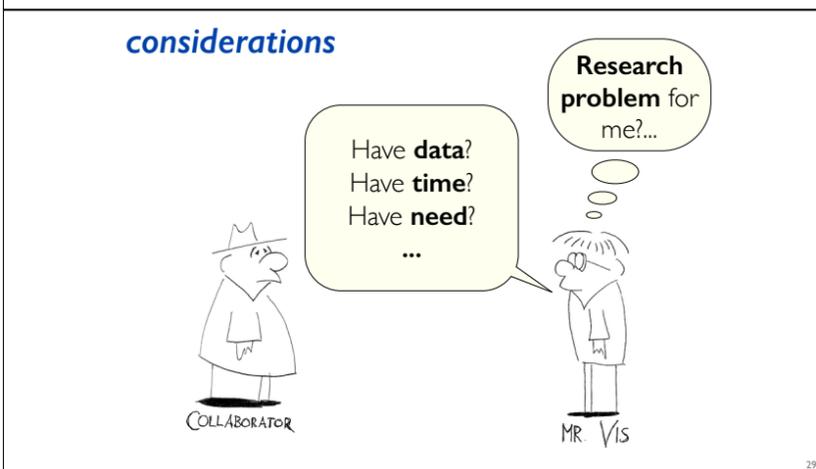


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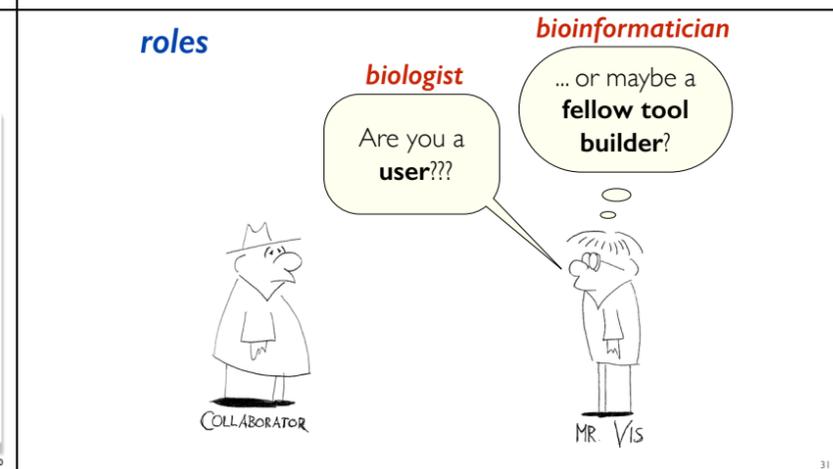
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Design study methodology: 32 pitfalls

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

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Examples from the trenches

- premature collaboration
- fellow tool builders with inaccurate assumptions about user needs
- data unavailable early so didn't diagnose problems

PowerSet Viewer
2 years / 4 researchers

WikeVis
0.5 years / 2 researchers

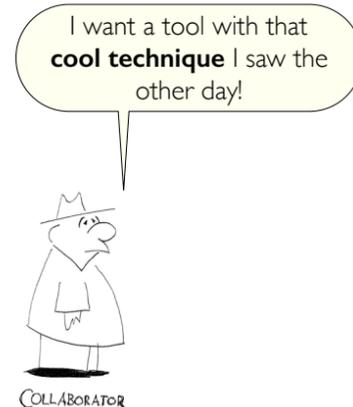
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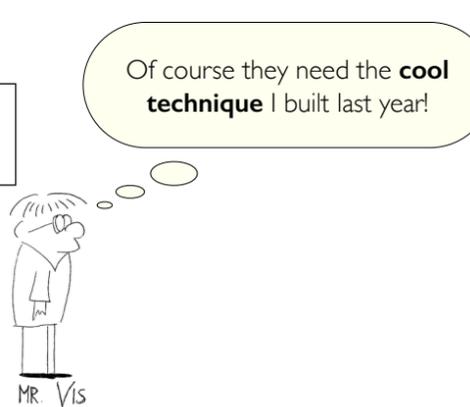
PITFALL

PREMATURE DESIGN COMMITMENT

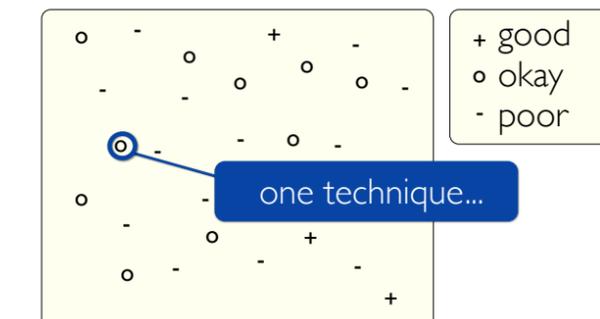


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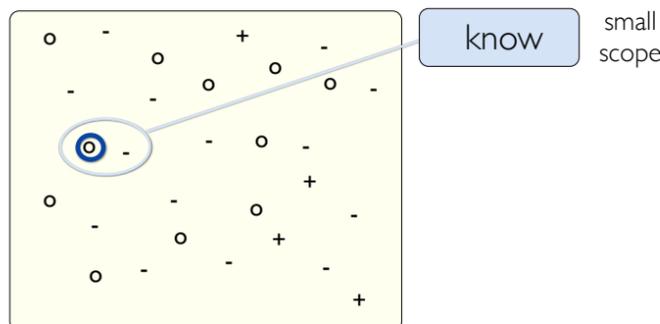
PREMATURE DESIGN COMMITMENT



METAPHOR Design Space



METAPHOR Design Space

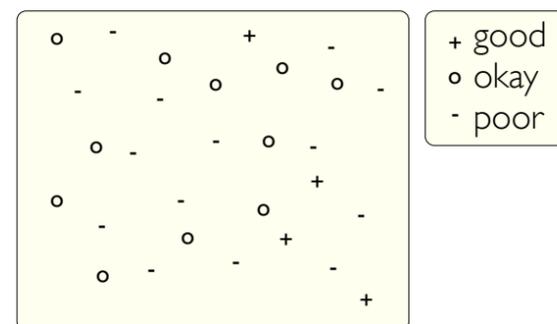


Design study methodology: 32 pitfalls

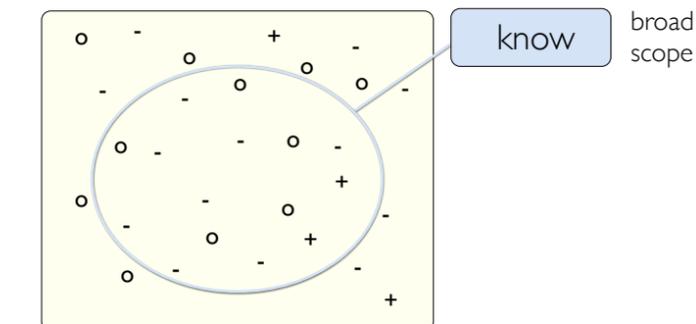
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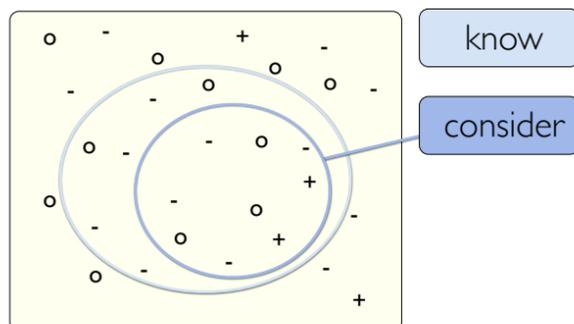
METAPHOR Design Space



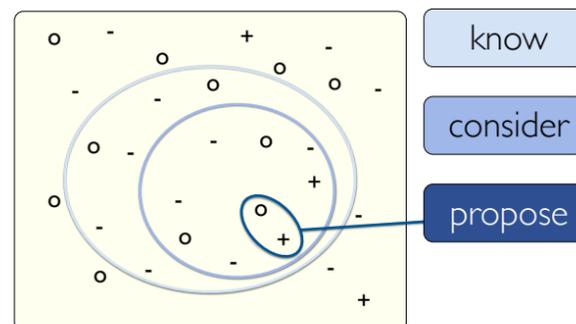
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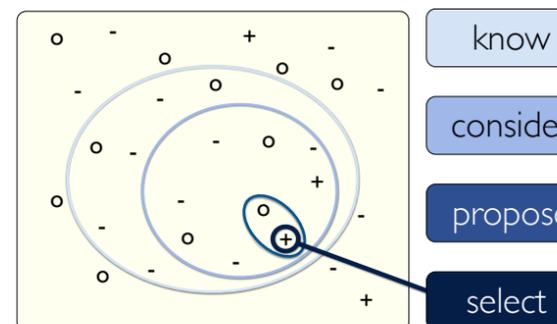
METAPHOR Design Space



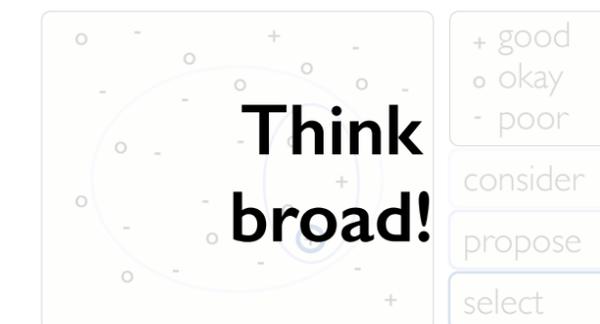
METAPHOR Design Space



METAPHOR Design Space



METAPHOR Design Space



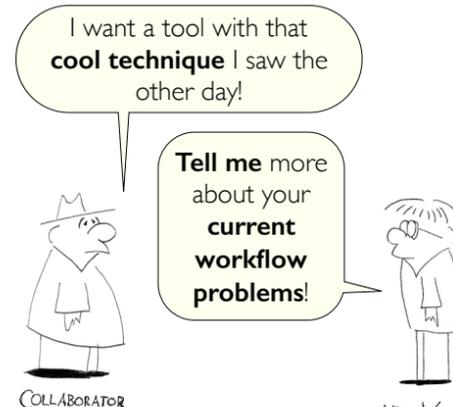
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PITFALL

PREMATURE DESIGN COMMITMENT

DOMAIN EXPERTS FOCUSED ON VIS DESIGN VS DOMAIN PROBLEM



Design study methodology: 32 pitfalls

PF-21	mistaking technique-driven for problem-driven work	design
PF-22	nonrapid prototyping	implement
PF-23	usability: too little / too much	implement
PF-24	premature end: insufficient deploy time built into schedule	deploy
PF-25	usage study not case study: non-real task/data/user	deploy
PF-26	liking necessary but not sufficient for validation	deploy
PF-27	failing to improve guidelines: confirm, refine, reject, propose	reflect
PF-28	insufficient writing time built into schedule	write
PF-29	no technique contribution ≠ good design study	write
PF-30	too much domain background in paper	write
PF-31	story told chronologically vs. focus on final results	write
PF-32	premature end: win race vs. practice music for debut	write

Joel Ferstay

Cydney Nielsen @cydneybn

Tamara Munzner @tamaramunzner

Variant View

Visualizing Sequence Variants in their Gene Context

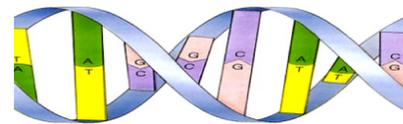
<http://www.cs.ubc.ca/labs/imager/tr/2012/VariantView/>

Variant View: Visualization Design Study

- first after DSM, tried following guidelines explicitly
- a specific **real-world** problem
 - real users and real data,
 - collaboration is (often) fundamental
- **design** a visualization system
 - implications: requirements, multiple ideas
- **validate** the design
 - at appropriate levels: case studies via deployment
- **reflect** about lessons learned
 - transferable research: improve design guidelines for vis in general
 - confirm, refine, reject, propose

Sequence Variant Definition

- Sequence variants
 - Difference between reference and given genome



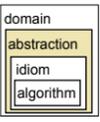
Reference Genome DNA: ATA TGA TCA ACA CTT

Sample 1 Genome DNA: ATA TGG TCA ATA CTT **Harmful?**

Sample 2 Genome DNA: ATA TGA TGA ACA CCT **Harmless?**

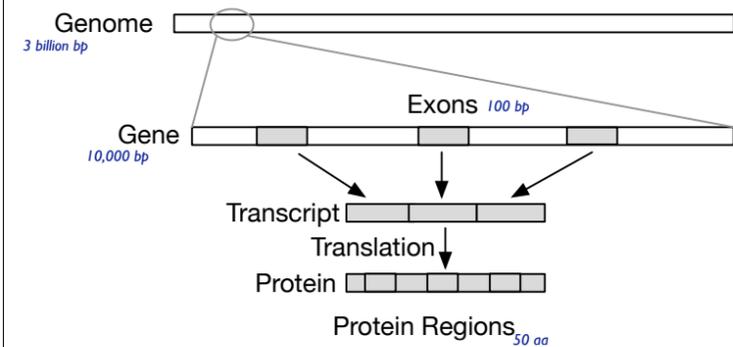
Cancer Research

- collaboration with analysts at BC Genome Sciences Center
 - studying genetic basis of leukemia
- driving task
 - discover new candidate genes with harmful variants
- two big questions
 - what to show
 - data abstraction
 - challenge: enormous range of scales in the data
 - how to show it
 - visual encoding idiom
 - challenge: information density and perceptual considerations

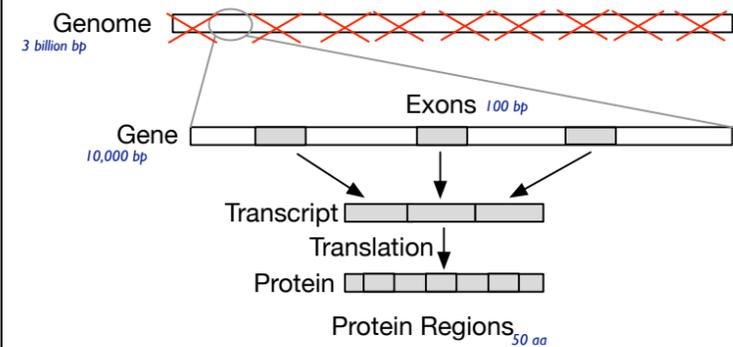


Abstractions

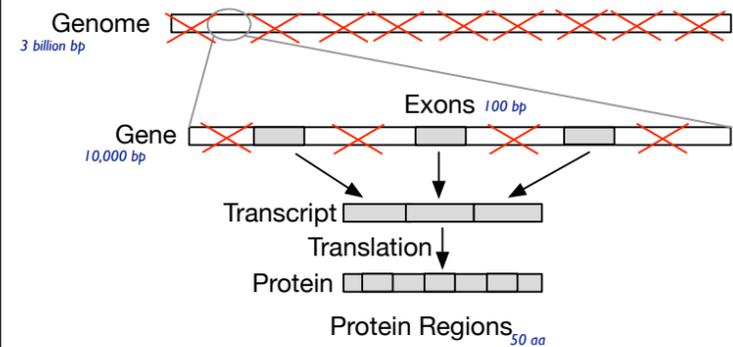
Data: Filtering to relevant biological levels and scales



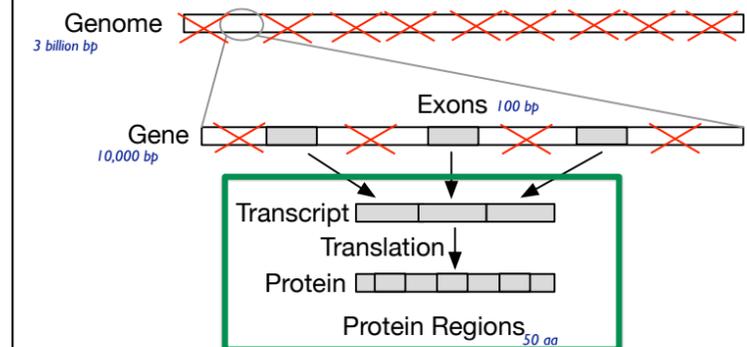
Filter out whole genome; keep genes



Filter out non-exon regions

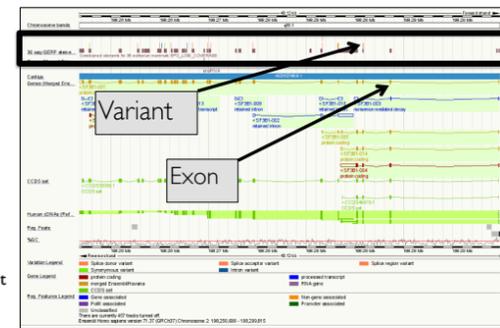


Data abstraction: highly filtered SCOPE of transcript coordinates



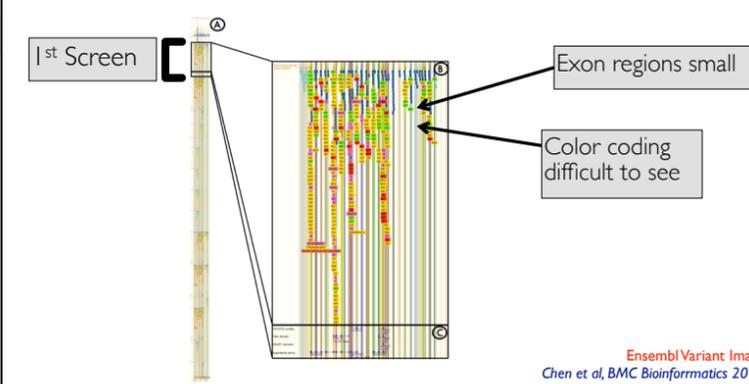
Dominant paradigm: genome browsers

- strengths: flexible and powerful
 - horizontal tracks: user data
 - shared coordinate system: genome coordinates (bp)
- problems
 - tiny features of interest spread out across large extent
 - must zoom far in to inspect known feature, then zoom out and pan to locate next
 - high cognitive load for interaction
 - must already know where to look



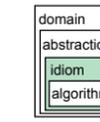
representative example: Ensembl
Chen et al, BMC Bioinformatics 2010.

Features of interest small even in variant-specific view

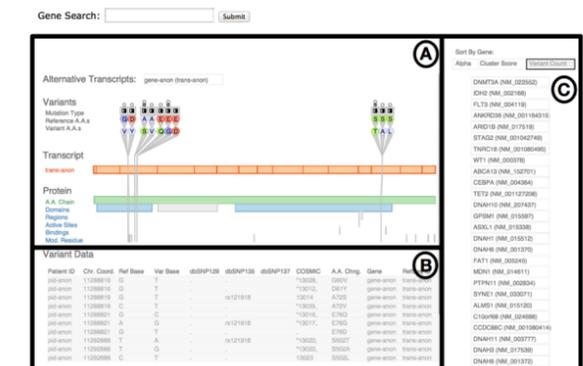


Ensembl Variant Image
Chen et al, BMC Bioinformatics 2010.

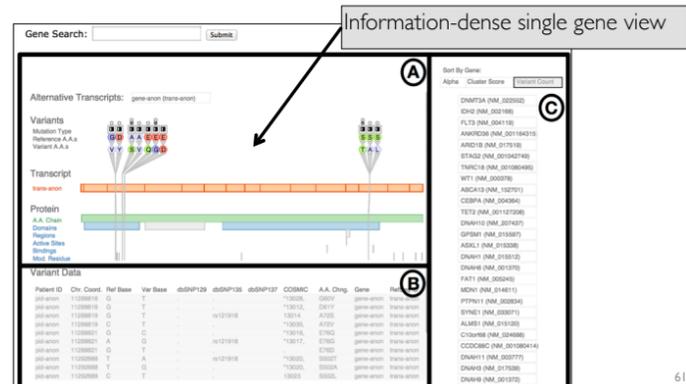
Idioms



Variant View



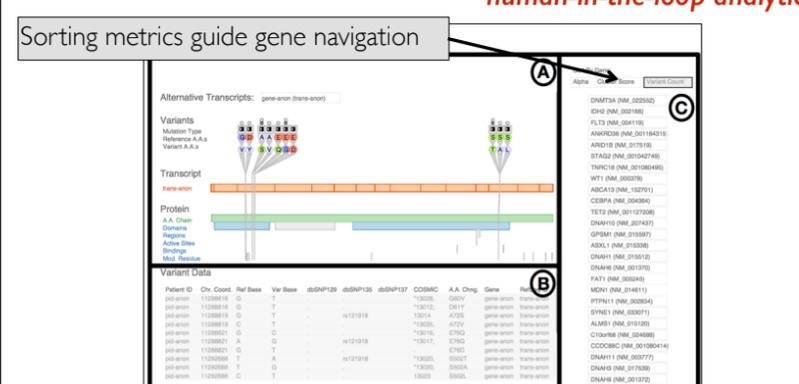
Variant View



Variant View

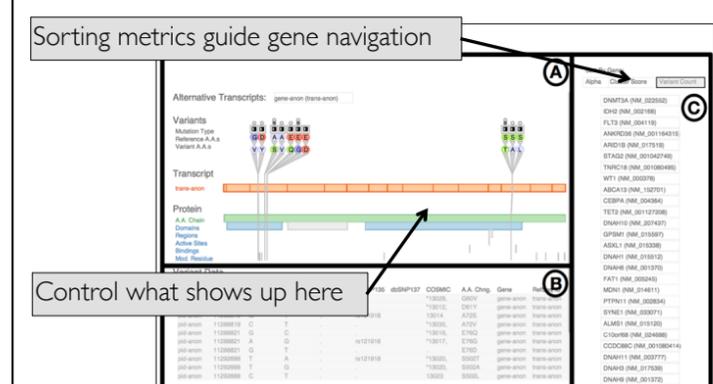


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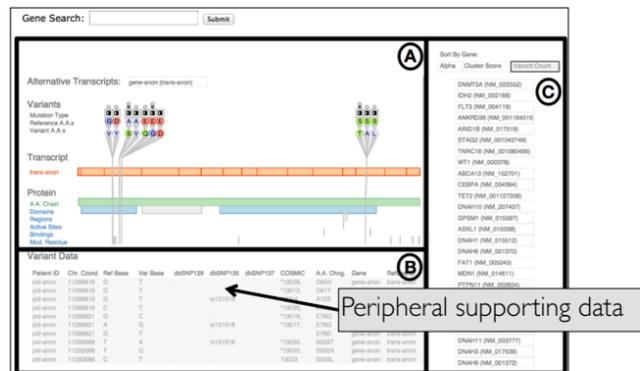


derived data guides
human-in-the-loop analytics

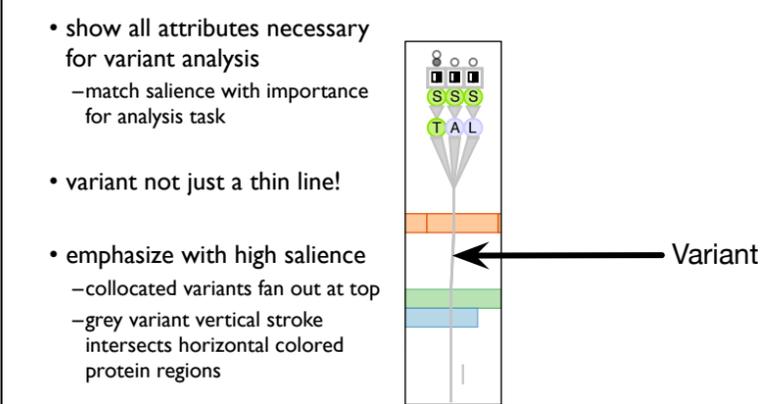
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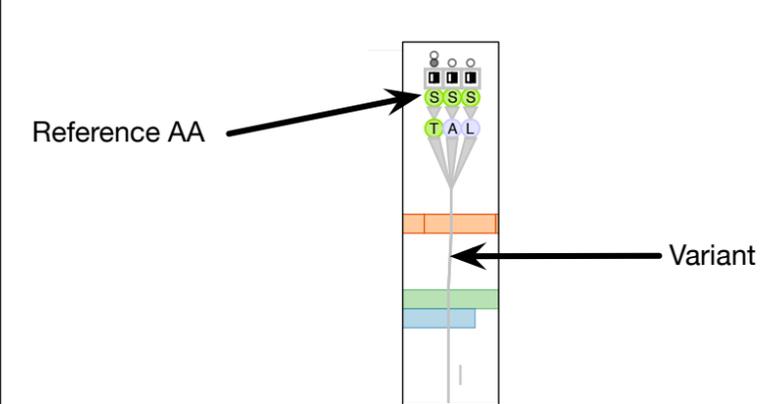
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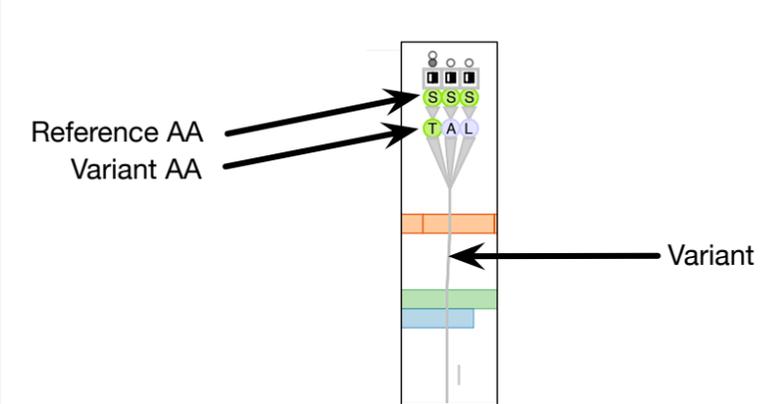
Design information-dense visual encoding



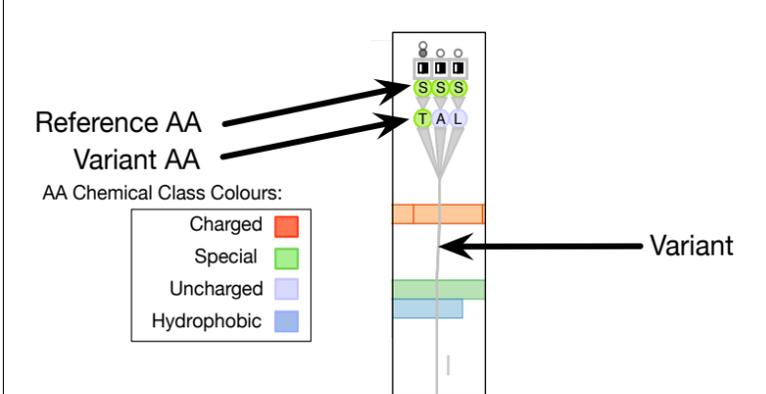
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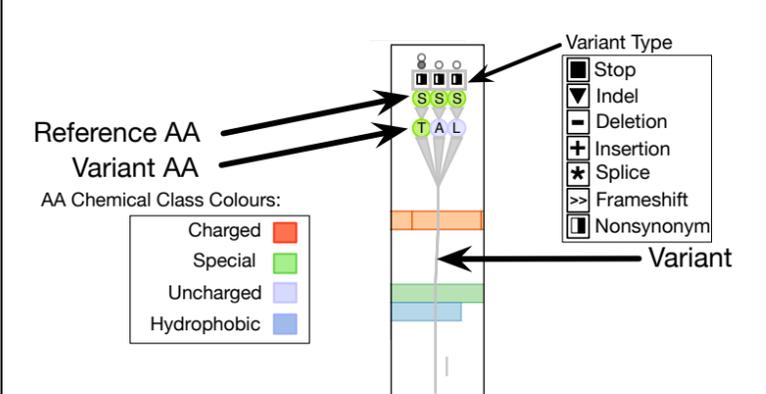
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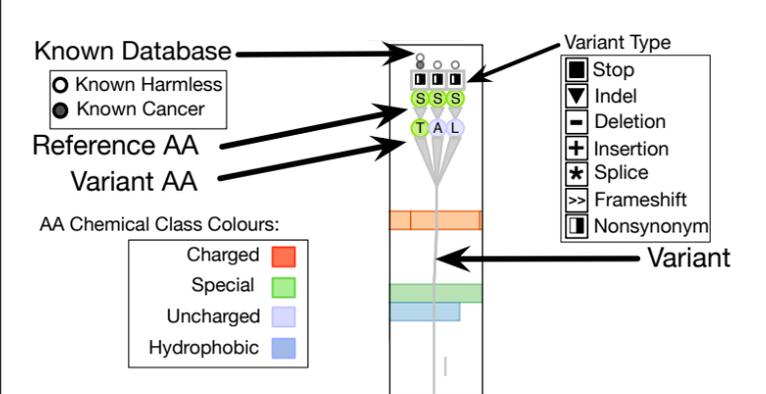
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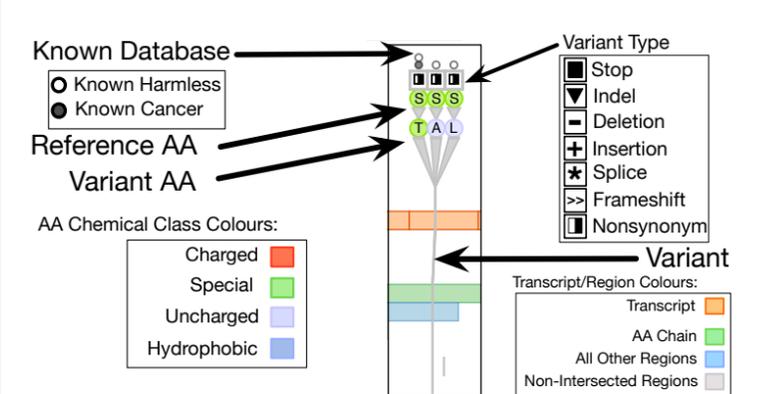
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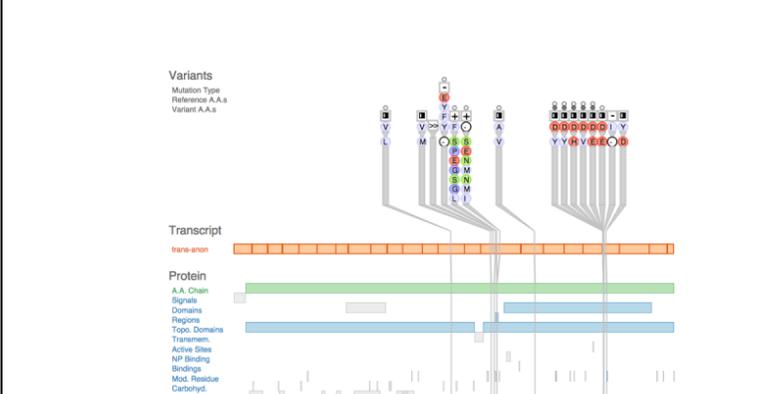


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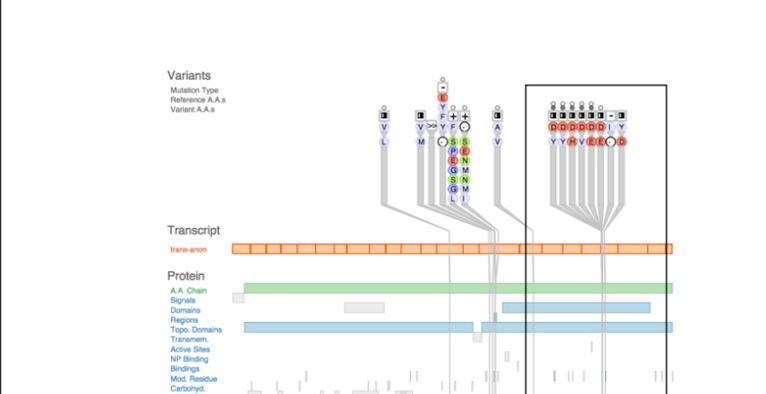


Results

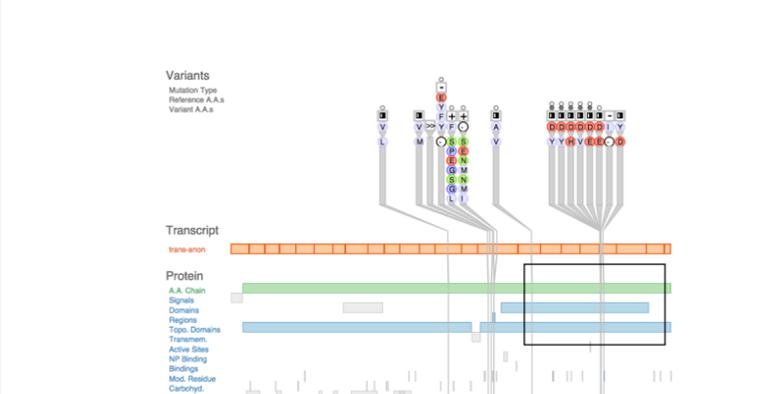
Known leukemia gene: Find fast with sorting metric high score



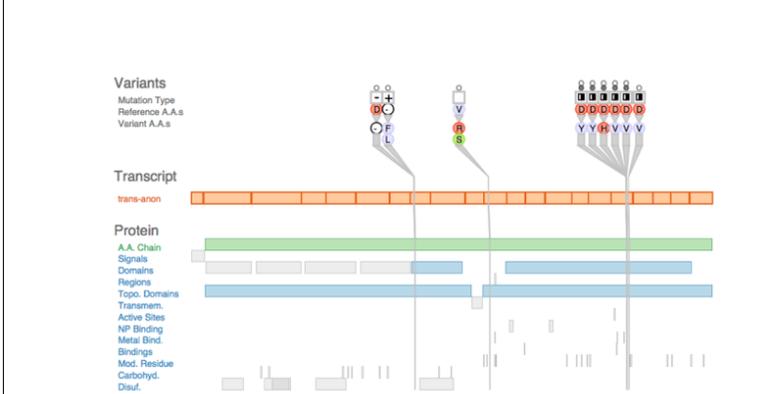
Known leukemia gene: Fanout shows collocation of variants



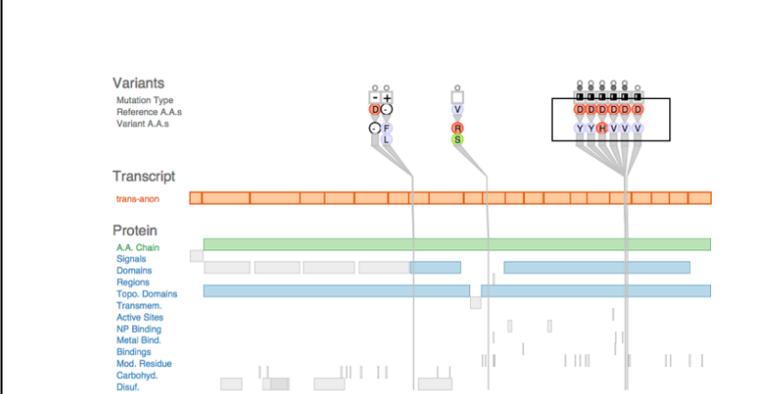
Known leukemia gene: Several functional protein regions affected



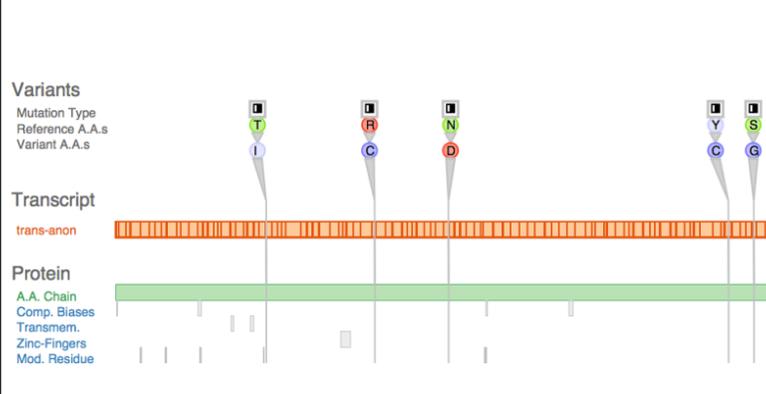
New finding: Good candidate with high metric score



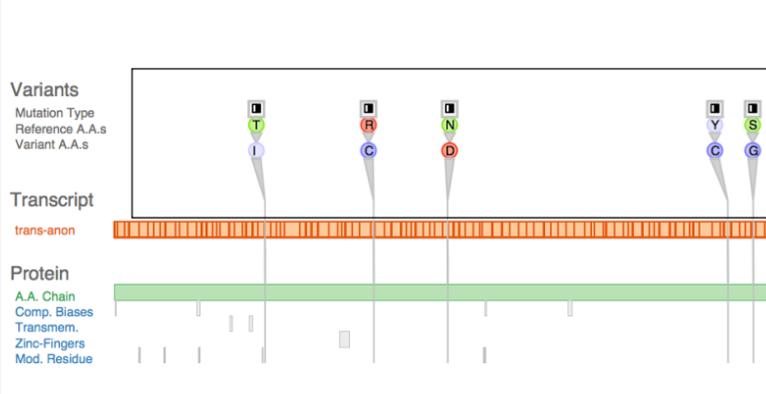
New finding: Protein chemical class change evident

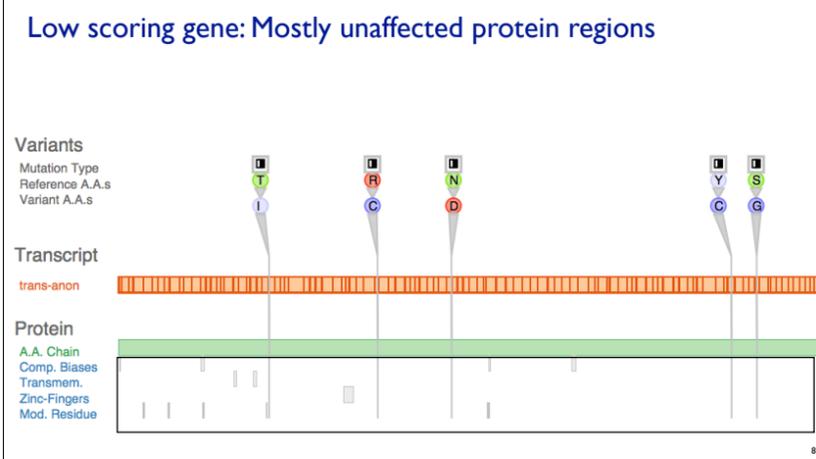


Low scoring gene: in contrast



Low scoring gene: No collocation of variants





Methods

Phase 1: Winnow and Cast

5 months

learn → winnow → cast → discover → design → implement → deploy → reflect → write

- embedded within GSC for all stages
- winnow stage
 - considered and ruled out many potential collaborators
- cast stage
 - gatekeeper (PI)
 - two front-line analysts (postdocs)

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

Phase 2: Core Design

5 months

learn → winnow → cast → discover → design → implement → deploy → reflect → write

- main task abstraction
 - discover gene
- semi-structured interviews
 - every week for 1 hr
- iterative refinement
 - 8 data sketches deployed
 - rapid prototyping to show real data ASAP
 - refine utility & usability

Human-centered approaches in geovisualization design:
 investigating multiple methods through a long-term case study.
 Lloyd and Dykes. IEEE TVCG (Proc. InfoVis), 17(12):2498-2507, 2011.

Phase 3: Two More Tasks

1 month

learn → winnow → cast → discover → design → implement → deploy → reflect → write

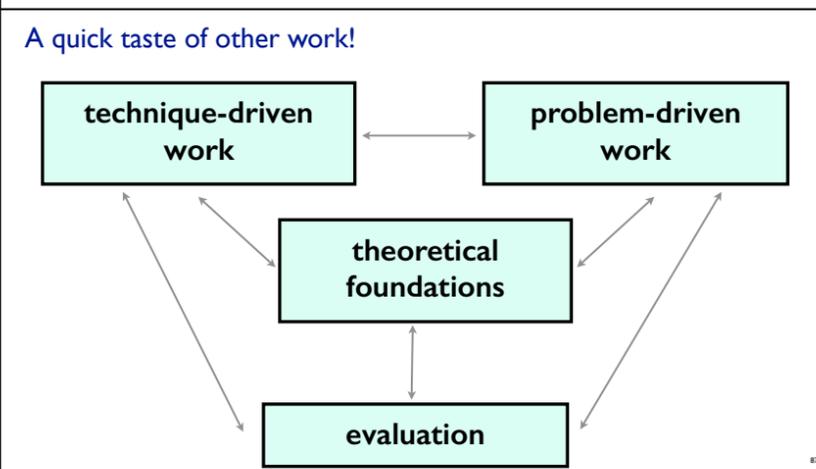
- two new analysts
 - connected by enthusiastic gatekeeper
- new task abstractions
 - compare patients
 - debug pipeline
- transferrable with minimal changes

Phase 4: Reflect and write

3 months

learn → winnow → cast → discover → design → implement → deploy → reflect → write

- abstraction innovation
 - data abstraction: highly filtered transcript coordinates (vs genome coordinates)
- guidelines
 - specialize first, generalize later
 - good for domains with complex data
 - high-level considerations
 - identifying scales of interest
 - what to visually encode directly vs what to support through interaction
 - when (and how) to eliminate navigation



Problem-driven: Genomics

Aaron Barsky
 Jenn Gardy (Microbio)
 Robert Kincaid (Agilent)
 Miriah Meyer
 Hanspeter Pfister (Harvard)
 MizBee

Cerebral
 MulteeSum, Pathline

Problem-driven: Genomics, fisheries

Joel Ferstay
 Cydney Nielsen (BC Cancer)
 Maryam Booshehrian
 Torsten Moeller (SFU)

Variant View
 Vismon

Problem-driven: Tech industry

Heidi Lam
 Diane Tang (Google)
 Peter McLachlan
 Stephen North (AT&T Research)

SessionViewer: web log analysis
 LiveRAC: systems time-series

Problem-driven: Journalism

Matt Brehmer
 Stephen Ingram
 Jonathan Stray (Assoc Press)

Overview

Technique-driven: Graph drawing

James Slack
 Kristian Hildebrand
 Daniel Archambault
 David Auber (Bordeaux)

TreeJuxtaposer
 TopoLayout
 SPF
 Grouse
 GrouseFlocks
 TugGraph

Evaluation: Graph drawing

Dmitry Nekrasovski
 Adam Bodnar
 Joanna McGrenere (UBC)
 Jessica Dawson

Stretch and squish navigation
 Search set model of path tracing

Technique-driven: Dimensionality reduction

Stephen Ingram
 Michael Sedlmair
 Melanie Tory (UVic)

Glimmer
 DimStiller
 Glint
 QSNE

Evaluation: Dimensionality reduction

Melanie Tory
 Michael Sedlmair

Points vs landscapes for dimensionally reduced data
 Guidance on DR & scatterplot choices
 Taxonomy of cluster separation factors

Evaluation: Focus+Context

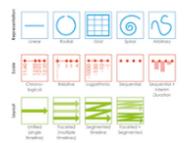
Heidi Lam
 Ron Rensink (UBC)
 Heidi Lam
 Robert Kincaid (Agilent)

Distortion impact on search/memory
 Separate vs integrated views

Curation & Presentation: Timelines



TimeLineCurator
<https://vimeo.com/123246662>



Timelines Revisited
[timelinesrevisited.github.io/](https://github.com/timelinesrevisited)

Johanna Fulda
 (Sud. Zeitung)



Matt Brehmer



Bongshin Lee
 (Microsoft)



Benjamin Bach
 (Microsoft)



Nathalie Henry-Riche
 (Microsoft)

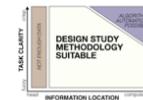


T F E P

Theoretical foundations

- Visual Encoding Pitfalls
 - Unjustified Visual Encoding
 - Hammer In Search Of Nail
 - 2D Good, 3D Better
 - Color Cacophony
 - Rainbows Just Like In The Sky
- Strategy Pitfalls
 - What I Did Over My Summer
 - Least Publishable Unit
 - Dense As Plutonium
 - Bad Slice and Dice

Papers Process & Pitfalls



Design Study Methodology



Abstract Tasks

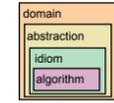
Michael Sedlmair



Miriah Meyer



Matt Brehmer



Nested Model



Visualization Analysis & Design



Visualization Analysis & Design

T F E P

Geometry Center 1990-1995



Geomview

Charlie Gunn



Stuart Levy

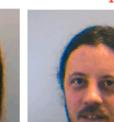


The Shape of Space



Outside In

Mark Phillips



Delle Maxwell



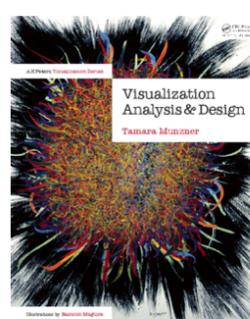
Wrap-up

- models and methods for design and validation
 - collaboration incentives for vis and bio
- example biovis project
 - Variant View
- methodological dream:
 - user-centered design spreading from vis to biovis to bioinformatics
 - task/requirements analysis for *all* tools, not just visual ones
 - focus on both utility and usability

More information

- this talk
<http://www.cs.ubc.ca/~tmm/talks.html#clayton17>
- papers, videos, software, talks, courses
<http://www.cs.ubc.ca/group/infovis>
<http://www.cs.ubc.ca/~tmm>
- theoretical foundations: book
 (+ free tutorial/course lecture slides)
<http://www.cs.ubc.ca/~tmm/vadbook>
 – 20% promo code for book+ebook combo: HVN17
 – <http://www.crcpress.com/product/isbn/9781466508910>

@tamaramunzner



Visualization Analysis and Design.
 Munzner, A K Peters Visualization Series, CRC Press, Visualization Series, 2014.