Information Visualization Meets Biology: Models and Methods for Collaboration

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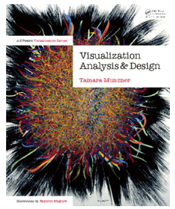
www.cs.ubc.ca/~tmm/talks.html#caulfield17 @tamaramunzner

Visualization (vis) defined & motivated

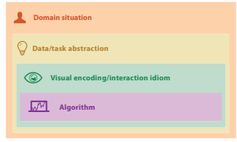
Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

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

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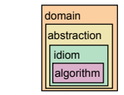
A Nested Model for Visualization Design and Validation.
Munzner. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 09), 15(6):921-928, 2009.

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



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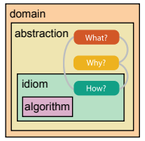
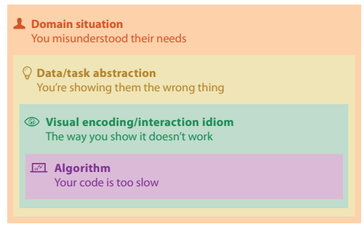


[A Multi-Level Typology of Abstract Visualization Tasks *Brehmer and Munzner. IEEE TVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).*]

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

Why is validation difficult?

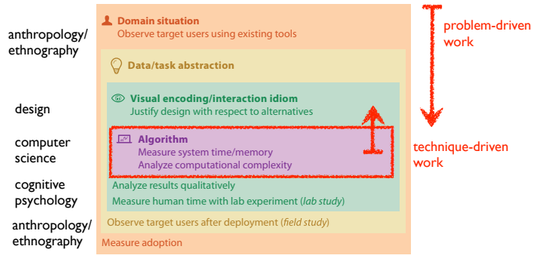
- different ways to get it wrong at each level



[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!



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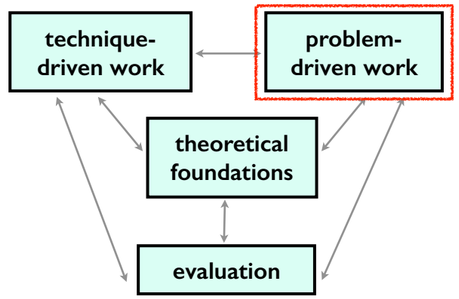
Angles of attack: My own research agenda

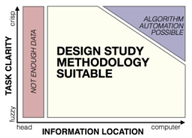


Vis meets bio

- 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
  - often existing infrastructure of computational workflows
    - 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]



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/

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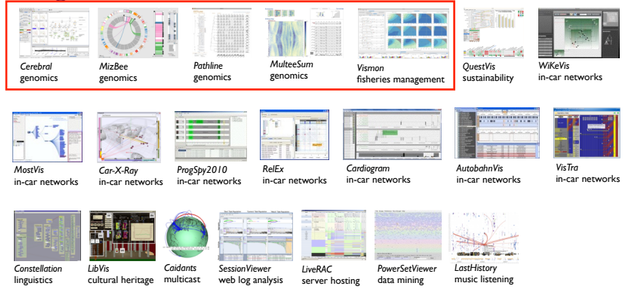
Design Study Methodology: Reflections from the Trenches and from the Stacks.
Sedlmair, Meyer, Munzner. IEEE Trans. Visualization and Computer Graphics 18(12): 2431-2440, 2012 (Proc. InfoVis 2012).

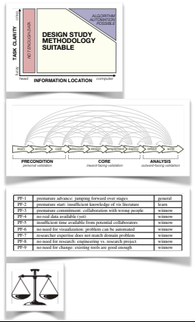
Design Studies: Lessons learned after 21 of them



- commonality of representations cross-cuts domains!

Methodology for problem-driven work

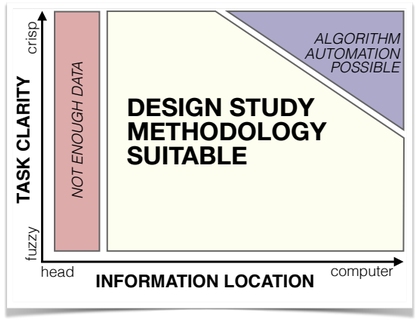
- definitions
- 9-stage framework
- 32 pitfalls & how to avoid them
- comparison to related methodologies



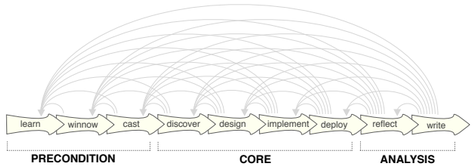
Design study methodology: definitions

- 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

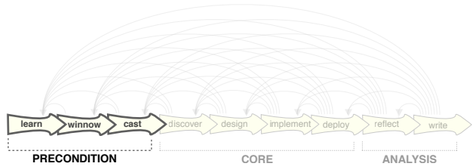


9 stage framework



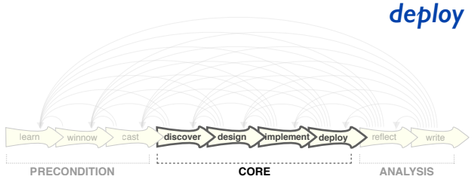
9-stage framework

learn
winnow
cast



9-stage framework

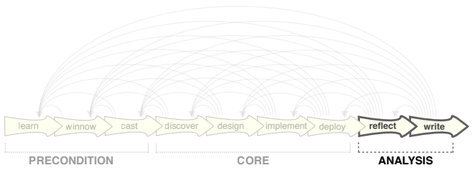
discover
design
implement
deploy



9-stage framework

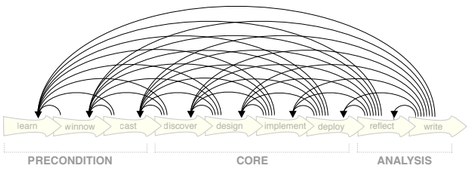
reflect
write

- guidelines: confirm, refine, reject, propose



9-stage framework

iterative



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
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PF-8	no need for research: engineering vs. research project	winnow
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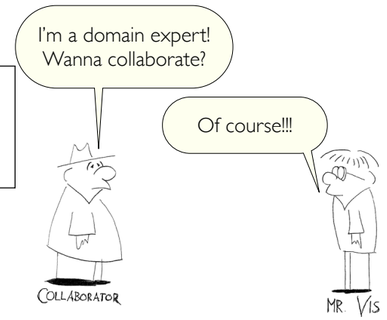
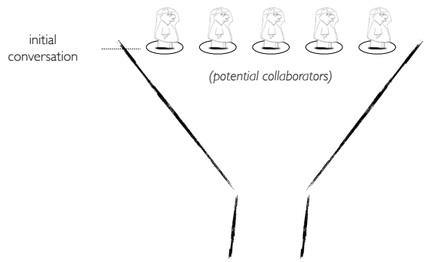
PITFALL

PREMATURE COLLABORATION COMMITMENT



METAPHOR
Winnowing

Collaborator winnowing



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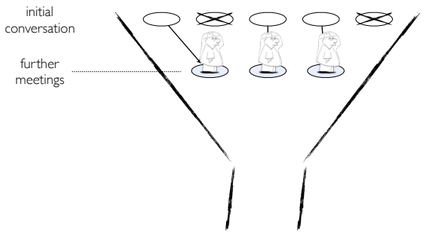

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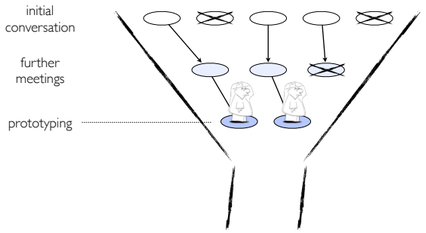

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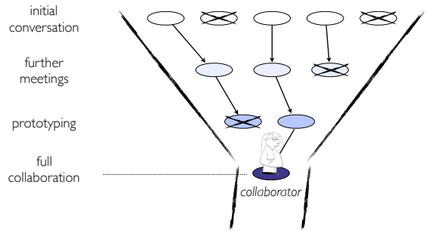

Collaborator winnowing

Talk with many, stay with few!

Design study methodology: 32 pitfalls

- and how to avoid them

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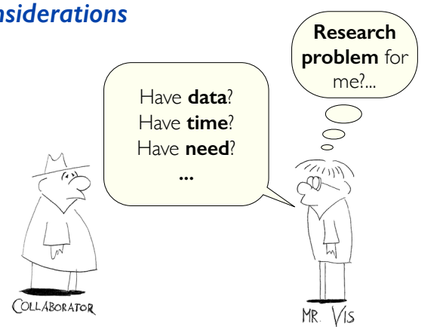
considerations



Design study methodology: 32 pitfalls

PF-10	no real/important/recurring task	winnow
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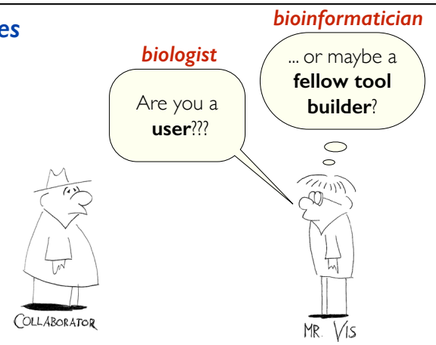
PF-12	not identifying front line analyst and gatekeeper before start	cast
PF-13	assuming every project will have the same role distribution	cast
PF-14	mistaking fellow tool builders for real end users	cast
PF-15	ignoring practices that currently work well	discover
PF-16	expecting *just talking* or *fly 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

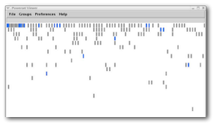
roles



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

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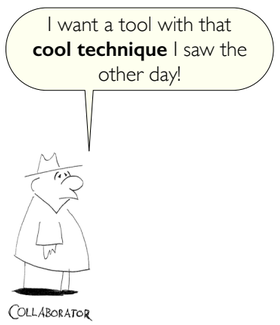

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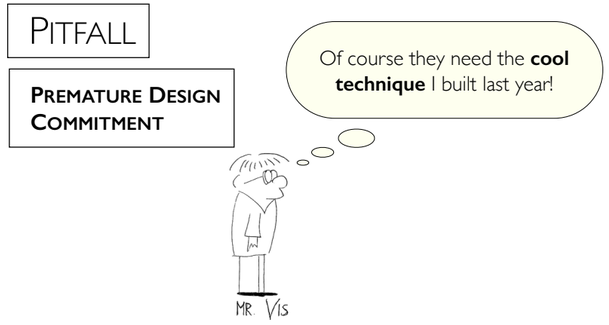
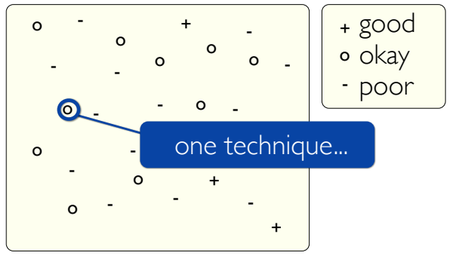


METAPHOR
Design Space



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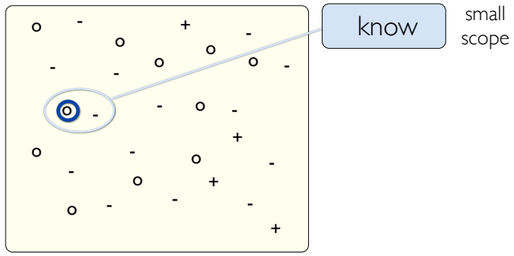

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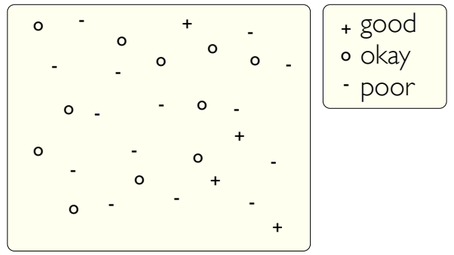

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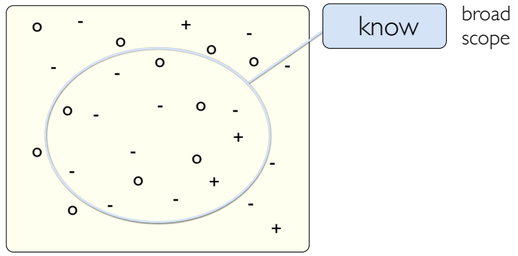

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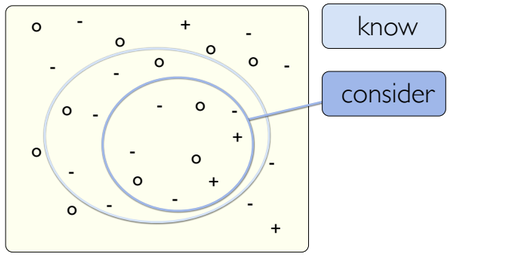

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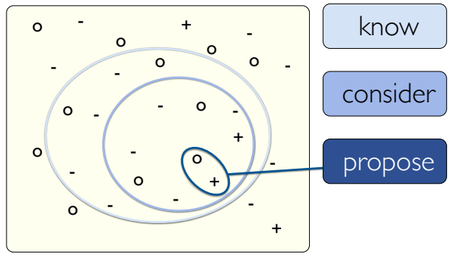

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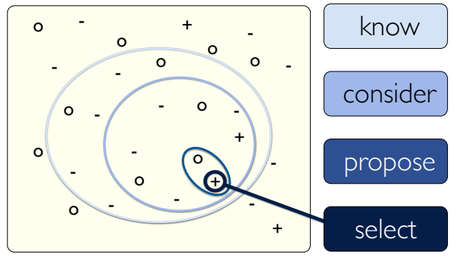

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Design study methodology: 32 pitfalls

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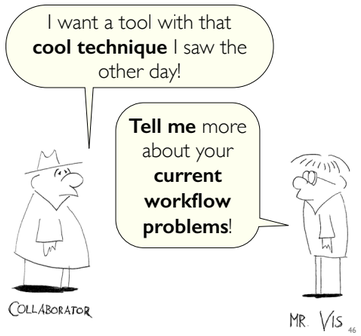
PF-20	premature design commitment: consideration space too small	design

PITFALL

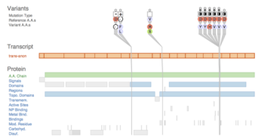
PREMATURE DESIGN COMMITMENT

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



Variant View

Visualizing Sequence Variants in their Gene Context

Joel Ferstay

Cydney Nielsen
@cydneybn

Tamara Munzner
@tamaramunzner

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

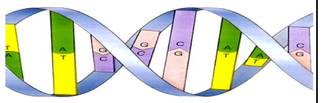
Variant View: Visualizing Sequence Variants in their Gene Context.
Forstay, Nielsen, Munzner. IEEE TVCG 19(12): 2546-2555, 2013 (Proc. InfoVis 2013).

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



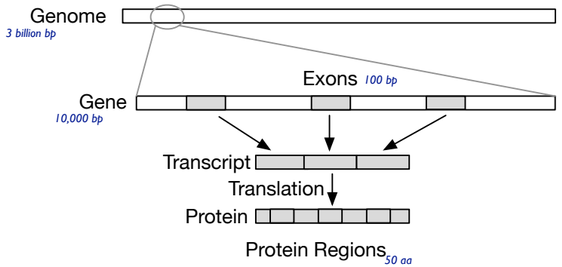
Reference Genome DNA:	ATA TGA TCA ACA CTT	
Sample 1 Genome DNA:	ATA TGG TCA ATA CTT	Harmful?
Sample 2 Genome DNA:	ATA TGA TGA ACA CCT	Harmless?

Cancer Research

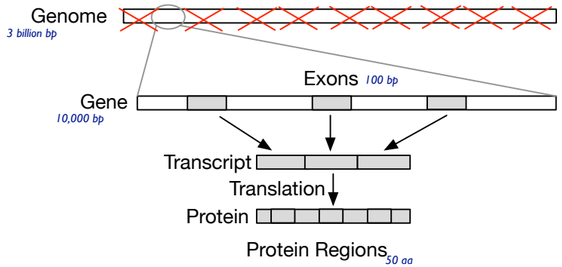
- collaboration with analysts at BC Genome Sciences Center
  - –studying genetic basis of leukemia
- driving task
  - –discover new candidate genes with harmful variants
- two big questions
  - –what to show
    - data abstraction
    - challenge: enormous range of scales in the data
  - –how to show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

Abstractions

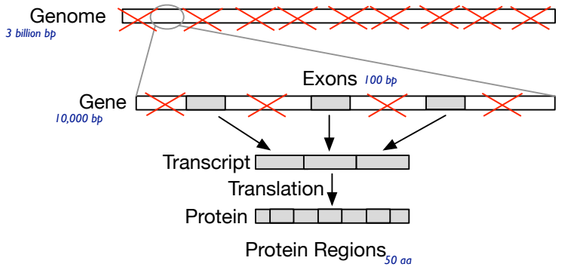
Data: Filtering to relevant biological levels and scales



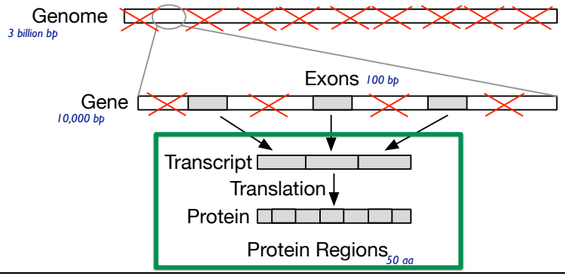
Filter out whole genome; keep genes



Filter out non-exon regions

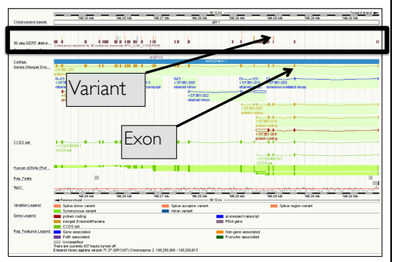


Data abstraction: highly filtered scope of *transcript coordinates*



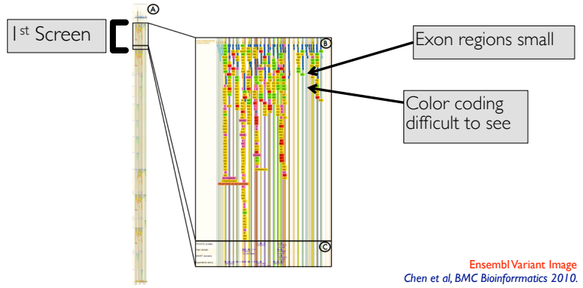
Dominant paradigm: genome browsers

- 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

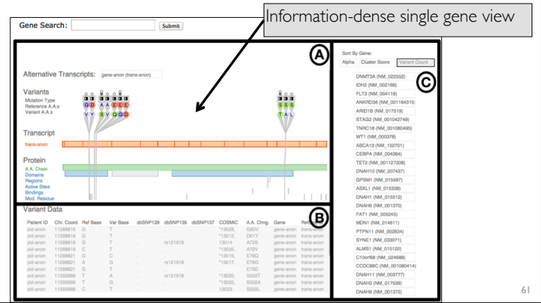


Ensembl Variant Image
Chen et al, BMC Bioinformatics 2010.

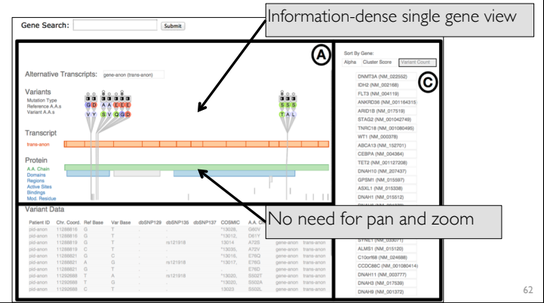
Idioms

Variant View

Variant View

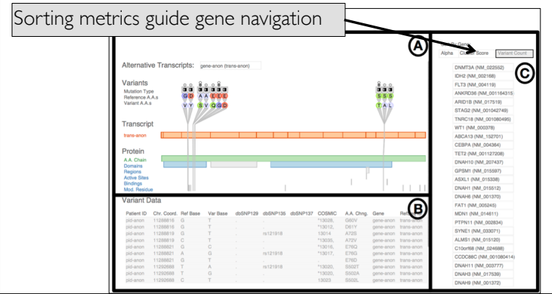


Variant View

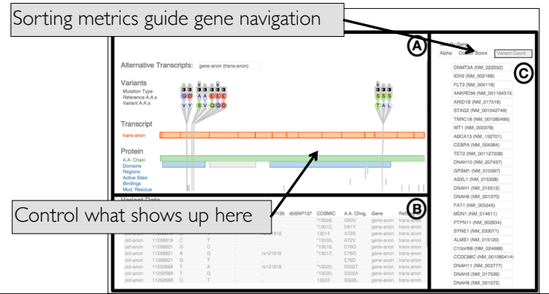


Variant View

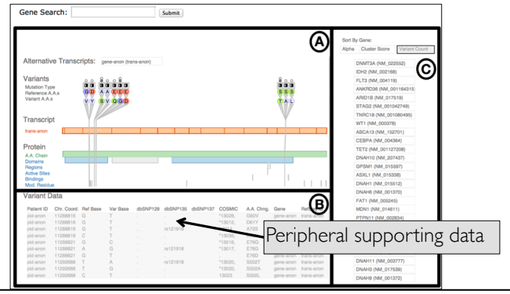
derived data guides human-in-the-loop analytics



Variant View

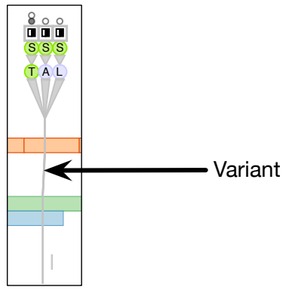


Variant View

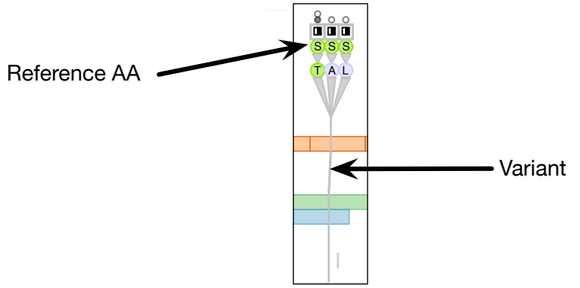


Design information-dense visual encoding

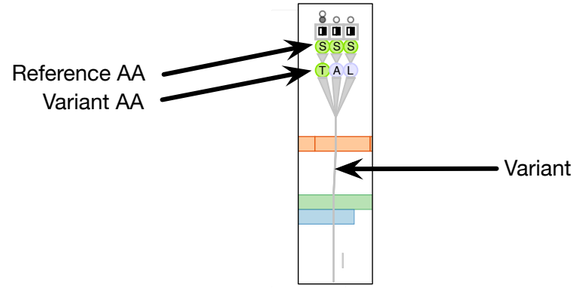
- show all attributes necessary for variant analysis
  - –match salience with importance for analysis task
- variant not just a thin line!
- emphasize with high salience
  - –collocated variants fan out at top
  - –grey variant vertical stroke intersects horizontal colored protein regions



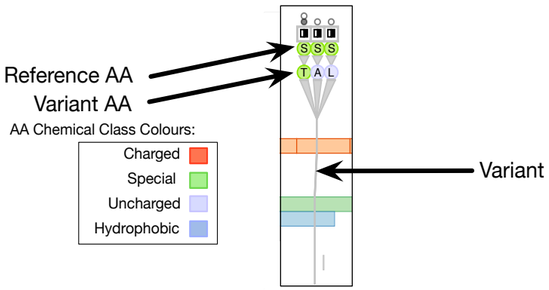
Design information-dense visual encoding



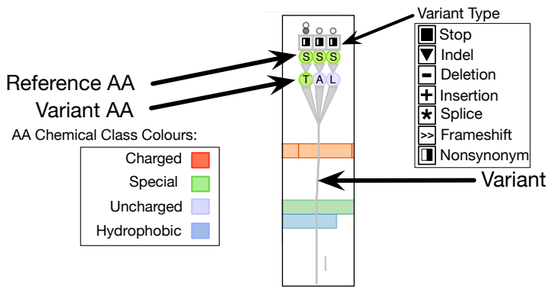
Design information-dense visual encoding



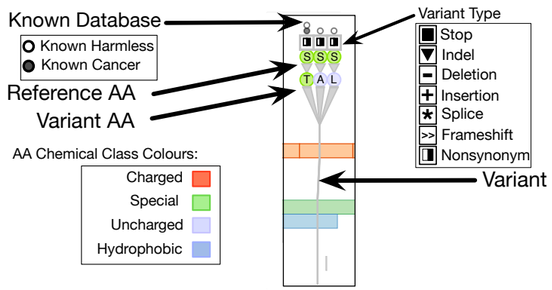
Design information-dense visual encoding



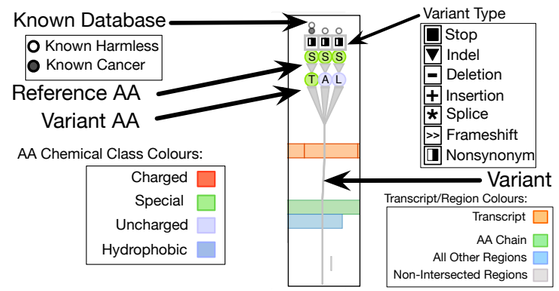
Design information-dense visual encoding



Design information-dense visual encoding

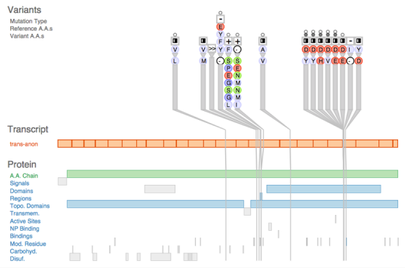


Design information-dense visual encoding

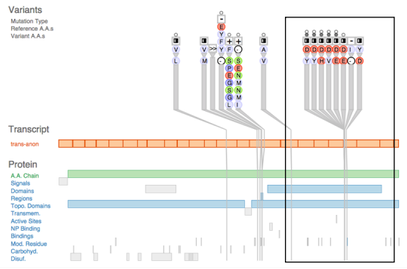


Results

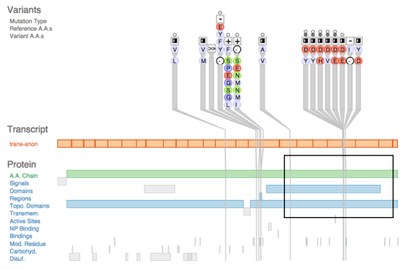
Known leukemia gene: Find fast with sorting metric high score



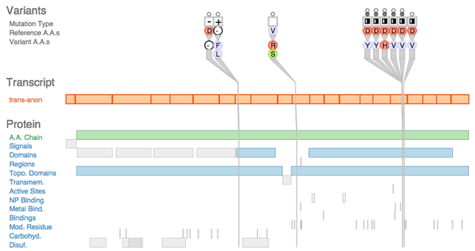
Known leukemia gene: Fanout shows collocation of variants



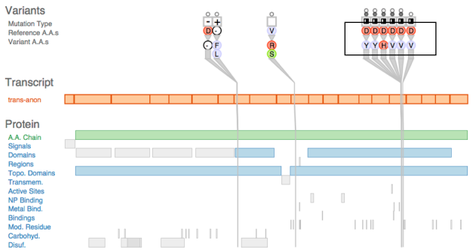
Known leukemia gene: Several functional protein regions affected



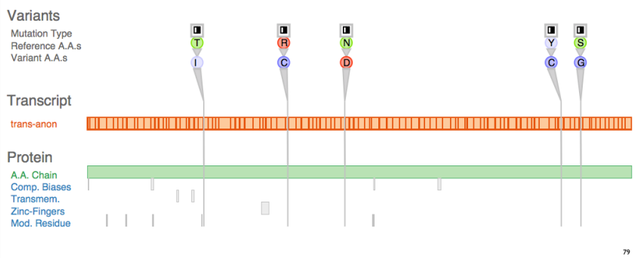
New finding: Good candidate with high metric score



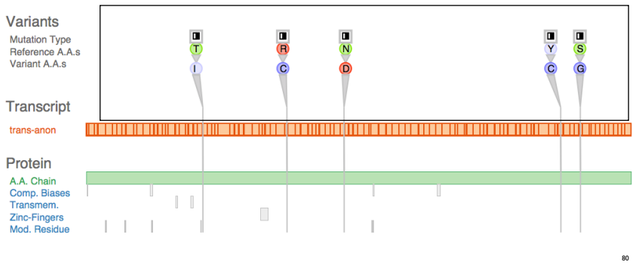
New finding: Protein chemical class change evident



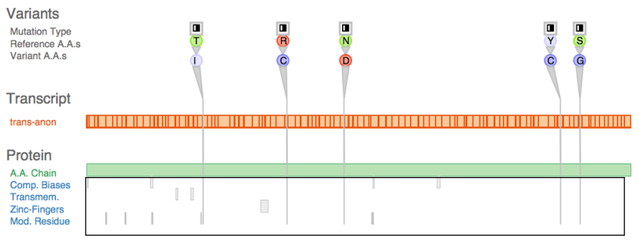
Low scoring gene: in contrast



Low scoring gene: No collocation of variants



Low scoring gene: Mostly unaffected protein regions



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)

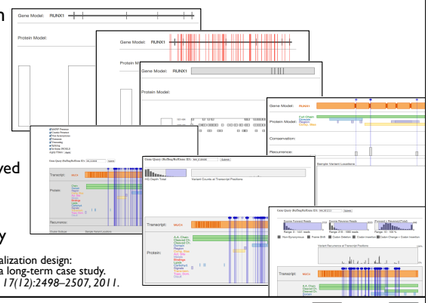
more at:
Design Study Methodology: Reflections from the Trenches and from the Stacks.
Sedlmair, Meyer, Munzner. IEEE TVCG 18(12): 2431-2440, 2012 (Proc. InfoVis 2012).

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



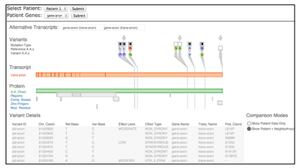
Human-centered approaches in geovisualization design: investigating multiple methods through a long-term case study.
Lloyd and Dykes. IEEE TVCG (Proc. InfoVis), 17(12):2498–2507, 2011.

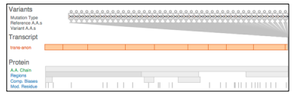
Phase 3: Two More Tasks

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





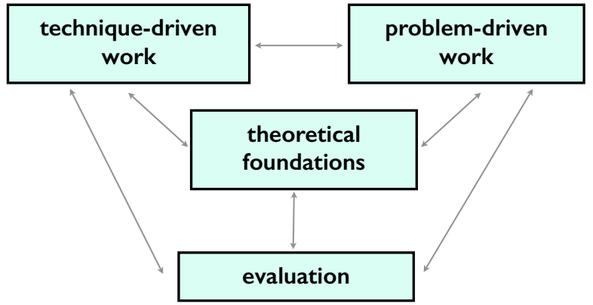
Phase 4: Reflect and write

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

A quick taste of other work!



Problem-driven: Genomics

T F E P

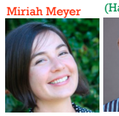
Aaron Barsky

Jenn Gardy (Microbio)

Robert Kincaid (Agilent)

Cerebral

Miriah Meyer



Hanspeter Pfister (Harvard)

MizBee

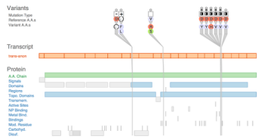
MulteeSum, Pathline

Problem-driven: Genomics, fisheries

T F E P

Joel Ferstay

Cydney Nielsen (BC Cancer)



Variant View

Vismon

Maryam Booshehrian

Torsten Moeller (SFU)

Problem-driven: Tech industry

T F E P

SessionViewer: web log analysis

Heidi Lam

Diane Tang (Google)

Peter McLachlan

Stephen North (AT&T Research)

LiveRAC: systems time-series

Problem-driven: Journalism

T F E P

Matt Brehmer

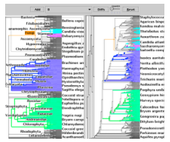
Stephen Ingram

Jonathan Stray (Assoc Press)

Overview

Technique-driven: Graph drawing

T F E P



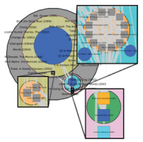
TreeJuxtaposer

James Slack

Kristian Hildebrand

Daniel Archambault

David Auber (Bordeaux)



TopoLayout
SPF
Grouse
GrouseFlocks
TugGraph

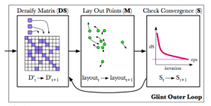
Technique-driven: Dimensionality reduction

T F E P

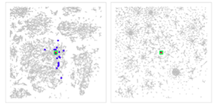
Stephen Ingram

Glimmer

DimStiller



Glint

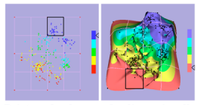


QSNE

Evaluation: Dimensionality reduction

T F E P

Melanie Tory

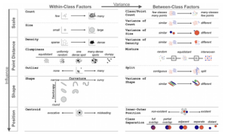


Points vs landscapes for dimensionally reduced data

Guidance on DR & scatterplot choices

Michael Sedlmair

Melanie Tory (UVic)



Taxonomy of cluster separation factors

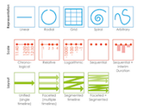
Curation & Presentation: Timelines

T F E P

TimeLineCurator
https://vimeo.com/123246662

Johanna Fulda (Sud. Zeitung)

Matt Brehmer



Timelines Revisited
timelinesrevisited.github.io/

Bongshin Lee (Microsoft)

Benjamin Bach (Microsoft)

Nathalie Henry-Riche (Microsoft)

Theoretical foundations

T F E P

- 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

Nested Model

Design Study Methodology

Michael Sedlmair

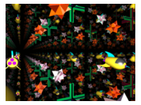
Miriah Meyer

Abstract Tasks

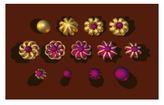
Matt Brehmer

Geometry Center 1990-1995: Expository videos

Geomview



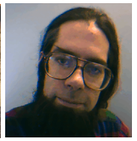
The Shape of Space



Outside In

Charlie Gunn

Stuart Levy



Mark Phillips

Delle Maxwell

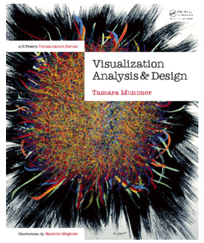
Wrap-up

- 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

@tamaramunzner

- this talk
  http://www.cs.ubc.ca/~tmm/talks.html#caulfield17
- 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



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