A Visualization Language







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Research Program

Interaction

Human-centered computing; Active collaborations

Theory of visualization

Computational modeling

An interactive environment

Output device Interaction technique

User's tasks

Input device

Visual scene

Problem statements

Ourgent needs to understand how to design visualizations to support understanding of the amount of information from complex systems.

Visualizations

biology m experiments social sciences

physiology medicine neurology s simulations

Information

space

How does visualization support seeing? and what do scientists see from mountains of data?
Theory of visualization

How to enable more effective knowledge discovery process in large information space?

Interactivity



1. A scientific visualization language for diffusion-tensor MRI visualization

Descriptive framework of seeing



2. Experiment: understanding illumination models Experiments



3. Workflow-driven design for timevarying bat flight analysis

Knowledge discovery



1. A scientific visualization language for diffusion-tensor MRI visualization

Collaborators: Computer science: David H. Laidlaw (Brown) Neurology: Alexander P. Auchus (UMMC)

Diffusion-tensor MRI

tractography

Seeds

tensor shapes

More measurement matrices

MRI

Not real-time (data intensive)

"General" graphics theory

Semiotics: the study of sign (Bertin 1967)



Applied to InfoVis by Mackinlay (Stanford, 1986), Fry (MIT/ Harvard, 2006), and Heer (Berkley 2007).

Our 3D semiotics theory



point, line, area, volume



J. Chen, On the semiological analysis of diffusion tensor field visualizations, IEEE TVCG (in progress).



flow direction -> color flow direction -> color flow speed -> texture flow speed -> texture flow speed -> shape size size

How to study these dimensions?

Design space must inform design (visualization technique and problem solving environment)



_ position

– size

- value

- texture

- color

-orientation

-shape

- rendering style

Which dimensions are most important?

Are these the right level of representation in a problem solving environment?

point, line, area, volume

Our approach

Strongly hypothesis-driven experimentation
 End-to-end, breadth-first reciprocal research strategy
 Corpus collection & data-driven research



1. A scientific visualization language for diffusion-tensor MRI visualization Descriptive framework of seeing



2. Experiment: understanding illumination models Experiments

Goal: study the effect of global illumination on task performance in complex visual scenes.

Motivation: Illumination Models

Image courtesy of David Banks (Harvard / U. of Tennessee)

model (GI)



Global illumination

Local illumination model (OpenGL)



Hypotheses

Independent variables:

Illumination model, texture, motion, and scene complexity

Depend variables

Time and error rate

GI > OpenGL

Motion > No motion

Texture > No texture





Task Conditions



Depth Judgment Visual Tracing Contact Judgment

Results: Motion on Performance

Motion reduced error rate but at the cost of longer task execution time.



Results: Illumination Model on Performance

- GI -> higher error rate | global tasks
- GI = GL on error rate | local tasks
- GI -> lower error rate (not significant)



Results: Subjective Responses

More cues = higher score



Beautiful things are useful.



_ position

– size

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-orientation

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-rendering style

Which dimensions are most important?

point, line, area, volume

Are these the right level of representation in a problem solving environment?

Contributions

Significant first step

Inderstanding how illumination model and motion -> time, error rate

Functional value and perceived value are not equivalent

Results could have impact on other types of
 3D vector / tensor field visualizations

Current work: rendering style comparison



Research questions: 1. Are there any differences in accuracy and efficiency when we use artistic rendering? 2. Can artistic rendering replicate the cueing effects in realistic rendering? 3. Does the rendering style influence preferences

and reassuring brain scientists' confidences?

Results:

- abstract tone shading works exceptionally well.
- we did not find significant main effect of halos on task performance
- depth-dependent shadows have a detrimental effect on accuracy and task completion time.

Current work: color encoding for legibility







2D and 3D integration

Boy's surface

goals:

Effects of color to represent selective / associative / quantitative visual dimensions

- Quantify the effectiveness of combined 2D/3D displays

Current work: optimal density











Research question: what is the optimal seeding resolution? Major results: 2x2x2

J. Chen, H. Cai, et al., "Efficacious Graphics Density of Diffusion Tensor MRI Visualizations", (under review).

Current work: ranking encoding for legibility



size



color





transparency

value

Method: - depth->encoding - color > (transparency = value) > size LOAD ''/tmp/allfb_tagged.data'' SELECT ''CC'' SELECT ''FA in [0.2,0.25]'' IN ''IFO'' UPDATE color BY FA IN ''CC'' SELECT ''LA > 0.35'' IN ''CST'' UPDATE shape BY line IN ''CC'' UPDATE shape BY line IN ''IFO'' UPDATE shape BY tube IN ''IFO''

H. Cai, <u>J. Chen</u>, et al., Depth-dependent parallel visualization with 3D stylized dense tubes. (under review). J. Chen, H. Cai, et al,, Gryphon: A scientific visualization language for diffusion MRI tractography visualizations. (under review)



1. A scientific visualization language for diffusion-tensor MRI visualization



2. Experiment: understanding illumination models



3. Workflow-driven design for timevarying bat flight analysis

Knowledge discovery

Goal: invisible visual interfaces for knowledge discovery Collaborators:

Computer science: Andrew Bragdon (Microsoft Research), Andy van Dam, David H. Laidlaw

Biology: Sharon M. Swartz, Rhea von Busse

Problem Domain

Sinematics

- Complex wing bone interaction
- Time-varying wing deformation
- Minetics
 - Unmanned vehicle design

Recording @ 1000 fps Playback @ 30 fps ~ 33x slow down

Video courtesy of Brown University

Conventional problem solving approach

Observations (bio)

Matlab feature extraction (bio, cs, math)

Sisualization (cs)

Hypothesis formation (bio, eng)

Comparison (cs, bio)

Work in multiple environments Extremely complex and dynamic process



Downstroke

Upstroke

Barriers to knowledge discovery

Observations (bio)

Matlab feature extraction (bio, cs, math)

Visualization (cs)

Hypothesis formation (bio, eng)

Comparison (cs, bio)

Education

Inefficient collaborative social dynamics

Error-prone computing

Difficulties in visualization

Our solution: VisBubbles

In the nutshell, it is

- A multiple-view UI with
 bubbles (Bragdon et.al 2010)
- A programming environment for data handling crosslinked to visualization
- A rapid visualization prototyping (2D/3D rendering)
- An asynchronous
 collaborative environment
- Interactivity



Error-prone computing

Difficulties in visualization

Inefficient collaborative social dynamics / education





[f] = wingbeat_frequency(30);
[span, chord] = spanonewing_chord(30);
[span, chord] =spanonewing_chord_l(30);

[area_vector] = area_onewing_vector(30);

function [dr_wrist , dr_wingtip] = downstroke_ratio(run_num) %DOWNSTROKE RATIO Duration of the downstroke divided by the total wingbeat period

temp=csvnead("./data/one wingbeat frame numbers. .

function [f] = wingbeat_frequency(run_num) %WINGBEAT_FREQUENCY Summary of this function goes here % Detailed explanation goes here

-

a=load_from_spreadsheet(run_num,1,1); a=a(:,1);

function [span, chord] = spanonewing chord(run_num) %SPAN_CHORD Summary of this function goes here % Detailed explanation goes here

Sj03=load_from_spreadsheet(run_num,3,3); % bacht referenced function [area_vector] = area_onewing_vector(run num) %WING_AREA_VECTOR produces a vector of nFrames x 1 that tells you area of %one wing over the course of the wingbeat cycle

function [span, chord] = spanonewing_chord_l(run_num) %SPAN_CHORD Summary of this function goes here

% Detailed explanation goes here

Sj03=load_from_spreadsheet(run_num,3.,3); % body referenced

function [xyz] = load_from_spreadsheet(run_num, Sj, whichone)

% make Sj 'all' if you want all the points to come out

% whichone: % 1 is rotated

function [xyz] = load_from_spreadsheet_2(run_num, Sj, whichone)

. 19 % make Sj 'all' if you want all the points to come out

% whichone: % 1 is rotated

> -[dr_wrist , dr_wingtip] = downstroke_ratio(30);

body referenced

run_num)

Sj03=load_from_spreadsheet(run_num,3,3); %

function [area_vector] = area_onewing_vector(

%WING_AREA_VECTOR produces a vector of

%one wing over the course of the wingbeat cycle

nFrames x 1 that tells you area of

[f] = wingbeat_frequency(30);

[span, chord] =spanonewing_chord(30);

[area_vector] = area_onewing_vector(30);

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function [dr_wrist , dr_wingtip] = downstroke_ratio(run_mm) %DOWNSTROKE_RATIO Duration of the downstroke divided by the total wingbeat period	1	aspect_ratio ID chord(m) dorsoventral_wingtip_excu downstroke_ratio_wrist downstroke_ratio_wingtip estimated_peak_lift_force estimated_peak_drag_force frequency(Hz)
temp=cswnood("/data/one_wingbeat_frame_numbers.		(T)camber_m1 1D1
function [f] = wingbeat_frequency(run_num) %WINGBEAT_FREQUENCY Summary of this function goes here % Detailed explanation goes here	1	(T)camber_m2 (T)camber_m3 (T)speed_m1 (T)speed_m2
a=load_from_spreadsheet(run_sum,1,1); a=a(:,1);		(T)speed_m3
function [span, chord] = <i>spanonewing_chord</i> (run_num) %SPAN_CHORD Summary of this function goes here % Detailed explanation goes here	- (span_db32d) - 0.2 0.16 0.12	=spanonewing_chord(30)

0.04 0 0.05 0.10.15 0.2 0.25 0.3 span

Design principles

How to make people more creative?

memory sequencing (spatial locations, predicting next step)

Reduce interruption

e.g., put socks on before the shoes

forming schema (Barlett 32) mental structure representing knowledge)

Consistency

e.g., put shirt on before my jacket

Forming schema

Bubbles interface (Bragdon et al. 2010)

Subser behavior -> interface action

grouping -> linking

New schema

Asynchronous collaboration





Support memory sequencing

Reducing cognitive distances between programming and visualization

Right representation level for visual analysis



Contributions

Memory-driven design for enhancing knowledge discovery

Integrated problem solving environment

Current work: interaction discourse analysis

- Is there an accessible structure in space usage pattern within the knowledge discovery discourse?
- How might one exploit this?
- Answering these questions?
 - Is inherently multidisciplinary
 - Requires expansive effort and vision
 - Promising great rewards



A swimming bat @ Brown (Video courtesy of the Swartz lab)

A key component is mental imagery in discourse.

Current work: Pathway and physiology data analysis

New applications



Conclusions



Global illumination resulted in similar task performance as local illumination



Ranking encoding

ES VARIABLES DE L'IMAGI

2 DIMENSIONS DU PLAN

TAILLE

VALEUR

LES VARIA

COULEUR

ORIENTATIO

FORME



The just-noticeable difference for dense tube visualizations



Color encoding



Legible dimensions: color worked the best.



Workflow driven interface design

Trend: rapid advances in interactive technologies













Trend: increased importance of design process (tools, practice and teaching)

Storytelling, creativity, integrating infoVis + sciVis



Trend: understand uncertainty

Serror bars on Measurement errors?



Trend: into the cloud

 Visualization will make use of the resources in the cloud.

Physiological sciences

Health care and Med student training in the cloud?



Robert Hester



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Thank you!

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