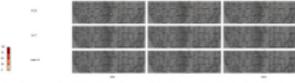
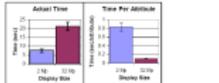
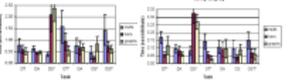
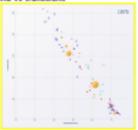
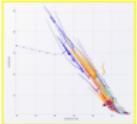
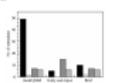
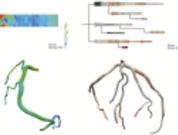
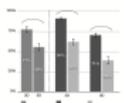


<p>Lecture 14: Evaluation Information Visualization CSPSC 533C, Fall 2011</p> <p>Tamara Munzner UBC Computer Science Wed, 2 November 2011</p>	<p>Readings Covered</p> <p>Evaluating Information Visualizations. Shewhart Carpendale. Chapter in Information Visualization: Human-Centered Issues and Perspectives, Springer LNCS 4950, 2008, p 19-45.</p> <p>The Perceptual Scalability of Visualization. Beth Yost and Chris North. Proc. InfoVis 06, published as IEEE TVCG 12(5), Sep 2006, p 837-844.</p> <p>Turning Pictures into Numbers: Extracting and Generating Information from Complex Visualizations. J Gregory Trullon, Susan S. Kirschbaum, Ted L. Tui, Robert T. Miyamoto, James A. Bates, and Paula D. Raymond. Intl Journ. Human Computer Studies 53(5), 827-850.</p>	<p>Further Readings</p> <p>Task-Centered User Interface Design, Clayton Lewis and John Rieman, Chapters 0-5.</p> <p>The challenge of information visualization evaluation. Catherine Plaisant. Proc. Advanced Visual Interfaces (AVI) 2004</p> <p>Effectiveness of Animation in Trend Visualization. George C. Robertson, Roland Fernandez, Danyel Fisher, Bongshin Lee, and John T. Staelo. IEEE TVCG (Proc. InfoVis 2008), 14(6): 1325-1332 (2008)</p> <p>Artery Visualizations for Heart Disease Diagnosis. Michelle A. Borlin, Krzysztof Z. Gajda, Amanda Peters, Divinosis Measures, Simone Melchione, Frank J. Rybicki, Charles L. Feldman, and Hanspeter Pfister. IEEE TVCG (Proc. InfoVis 2011), 17(12):2479-2488.</p>	<p>Evaluation, Carpendale</p> <ul style="list-style-type: none"> thorough survey/discussion, won't summarize here
<p>Psychophysics</p> <ul style="list-style-type: none"> method of limits <ul style="list-style-type: none"> find limitations of human perceptions error detection methods <ul style="list-style-type: none"> find threshold of performance degradation staircase procedure to find threshold faster method of adjustment <ul style="list-style-type: none"> find optimal level of stimuli by letting subjects control the level 	<p>Cognitive Psychology</p> <ul style="list-style-type: none"> repeating simple, but important tasks, and measure reaction time of error <ul style="list-style-type: none"> Müller's 7:11: 2 short-term memory experiments Fitts' Law (target selection) Hick's Law (decision making given n choices) interference between channels multi-modal studies <ul style="list-style-type: none"> MacLean 2005, Perceiving Ordinal Data Haptically Under Workload using haptic feedback for interruption when the participants were visually (and cognitively) busy 	<p>Structural Analysis</p> <ul style="list-style-type: none"> requirement analysis, task analysis structured interviews <ul style="list-style-type: none"> can be used almost anywhere, for open-ended questions and answers rating/Likert scales <ul style="list-style-type: none"> commonly used to solicit subjective feedback ex: NASA-TLX (Task Load Index) to assess mental workload <ul style="list-style-type: none"> "is frustrating to use the interface" Strongly Disagree Disagree Neutral Agree Strongly Agree 	<p>Comparative User Studies</p> <ul style="list-style-type: none"> hypothesis testing <ul style="list-style-type: none"> hypothesis: a precise problem statement <ul style="list-style-type: none"> ex: Participants will be faster with a coordinated overview+detail display than with an uncoordinated display or a detail-only display with the task requires reading details measurement: factor <ul style="list-style-type: none"> coordinated O-D display uncoordinated O display uncoordinated D display condition of comparison: task requires reading details
<p>Comparative User Studies</p> <ul style="list-style-type: none"> study design: factors and levels factors <ul style="list-style-type: none"> independent variables ex: interface, task, participant demographics levels <ul style="list-style-type: none"> number of variables in each factor limited by length of study and number of participants 	<p>Comparative User Studies</p> <ul style="list-style-type: none"> study design: within, or between? <ul style="list-style-type: none"> within <ul style="list-style-type: none"> everybody does all the conditions can lead to ordering effects Fitts' Law (target selection) can account for individual differences and reduce noise thus can be more powerful and require fewer participants combinatorial explosion seems levels on number of conditions possible workaround is multiple sessions between <ul style="list-style-type: none"> divide participants into groups each group does only some conditions 	<p>Comparative User Studies</p> <ul style="list-style-type: none"> measurements (dependent variables) <ul style="list-style-type: none"> performance indicators: task completion time, error rates, mouse movement subjective participant feedback: satisfaction ratings, closed-ended questions, interview observations: behaviors, signs of frustration number of participants <ul style="list-style-type: none"> depends on effect size and study design: power of experiment possible confounds? <ul style="list-style-type: none"> learning effect: did everybody use interfaces in a certain order? if so, are people faster because they are more practiced, or because of true interface effect? 	<p>Comparative User Studies</p> <ul style="list-style-type: none"> result analysis <ul style="list-style-type: none"> should know how to analyze the main results/hypotheses BEFORE study hypothesis testing analysis (using ANOVA or t-tests) tests how likely observed differences between groups are due to chance alone ex: a p-value of 0.05 means there is a 5% probability the difference occurred by chance usually good enough for HCI studies pilot! <ul style="list-style-type: none"> should have good idea of forthcoming results of the study BEFORE running actual study trials
<p>Evaluation Throughout Design Cycle</p> <ul style="list-style-type: none"> user/task centered design cycle <ul style="list-style-type: none"> initial assessments iterative design process benchmarking deployment identify problems, go back to previous step <p>Task-Centered User Interface Design, Clayton Lewis and John Rieman, Chapters 0-5.</p>	<p>Initial Assessments</p> <ul style="list-style-type: none"> what kind of problems are the system aiming to address? <ul style="list-style-type: none"> analyze a large and complex dataset who are your target users? <ul style="list-style-type: none"> data analysis what are the tasks? what are the goals? <ul style="list-style-type: none"> find trends and patterns in the data via exploratory analysis what are their current practices <ul style="list-style-type: none"> statistical analysis why and how can visualization be useful? <ul style="list-style-type: none"> visual spotting of trends and patterns talk to the users, and observe what they do task analysis 	<p>Iterative Design Process</p> <ul style="list-style-type: none"> does your design address the users' needs? <ul style="list-style-type: none"> can they use it? where are the usability problems? evaluate without users <ul style="list-style-type: none"> cognitive walkthrough action analysis heuristics analysis evaluate with users <ul style="list-style-type: none"> usability evaluations (think-aloud) bottom-line measurements 	<p>Benchmarking</p> <ul style="list-style-type: none"> how does your system compare to existing ones? <ul style="list-style-type: none"> empirical, comparative studies <ul style="list-style-type: none"> ask specific questions compare an aspect of the system with specific tasks Amor/Staelo task taxonomy paper quantitative, but limited The Challenge of Information Visualization Evaluation, Catherine Plaisant, Proc. AVI 2004

Deployment	Compare Systems vs. Characterize Usage	Perceptual Scalability	Perceptual Scalability
<ul style="list-style-type: none"> how is the system used in the wild? how are people using it? does the system fit into existing work flow? environment? contextual studies, field studies 	<ul style="list-style-type: none"> user/task centered design cycle: <ul style="list-style-type: none"> initial assessments iterative design process benchmarking: head-to-head comparison deployment identify problems, go back to previous step understanding/characterizing techniques <ul style="list-style-type: none"> tease apart factors when and how is technique appropriate line is blurry: intent 	<ul style="list-style-type: none"> what are perceptual/cognitive limits when screen-space constraints lifted? <ul style="list-style-type: none"> 2 vs. 32 lmpical display macro/micro views perceptually scalable <ul style="list-style-type: none"> no increase in task completion times when normalize to amount of data  <p>[The Perceptual Scalability of Visualization. Beth Viet and Chris North. IEEE TVCG 12(5) (Proc. InfoVis 06), Sep 2006, p 837-844.]</p>	<ul style="list-style-type: none"> design <ul style="list-style-type: none"> 2 display sizes, between-subjects <ul style="list-style-type: none"> (data size also increased proportionally) 3 visualization designs, within <ul style="list-style-type: none"> small multiples: bars embedded graphs embedded bars 7 tasks, within <ul style="list-style-type: none"> 42 tasks per participant 3 vis x 7 tasks x 2 trials
<h3>Embedded Visualizations</h3>  <p>[The Perceptual Scalability of Visualization. Beth Viet and Chris North. IEEE TVCG 12(5) (Proc. InfoVis 06), Sep 2006, p 837-844.]</p>	<h3>Small Multiples Visualizations</h3> <ul style="list-style-type: none"> attribute-centric instead of space-centric:  <p>[The Perceptual Scalability of Visualization. Beth Viet and Chris North. IEEE TVCG 12(5) (Proc. InfoVis 06), Sep 2006, p 837-844.]</p>	<h3>Results</h3> <ul style="list-style-type: none"> 20x increase in data, but only 3x increase in absolute task times  <p>[The Perceptual Scalability of Visualization. Beth Viet and Chris North. IEEE TVCG 12(5) (Proc. InfoVis 06), Sep 2006, p 837-844.]</p>	<h3>Results</h3> <ul style="list-style-type: none"> significant 3-way interaction <ul style="list-style-type: none"> between display, size, task  <p>[The Perceptual Scalability of Visualization. Beth Viet and Chris North. IEEE TVCG 12(5) (Proc. InfoVis 06), Sep 2006, p 837-844.]</p>
<h3>Results</h3> <ul style="list-style-type: none"> visual encoding important on small displays <ul style="list-style-type: none"> DS: multis sig slower than graphs on small DS: multis sig slower than embedded on large OS: bars sig faster than graphs for small OS: no sig difference bars/graphs for large spatial grouping important on large displays <ul style="list-style-type: none"> embedded sig faster + preferred over small mult no bar/graph differences 	<h3>Critique</h3>	<h3>Critique</h3> <ul style="list-style-type: none"> first study of macro/micro effects <ul style="list-style-type: none"> breaking new ground many possible followups <ul style="list-style-type: none"> physical navigation vs. virtual navigation <ul style="list-style-type: none"> The Effects of Peripheral Vision and Physical Navigation in Large Scale Visualization. GI 08 Move to improve: Promoting Physical Navigation to Increase user Performance with Large Displays. CHI 07 	<h3>Trends: Animation, Trails, SmallMult</h3> <ul style="list-style-type: none"> Gapminder: animated bubble charts + human <ul style="list-style-type: none"> x/y position, size, color, animation is animation effective? <ul style="list-style-type: none"> presentation vs analysis trend vs transitions  <p>[Effectiveness of Animation in Trend Visualization. Robertson et al. IEEE TVCG (Proc. InfoVis 2008), 14(5): 1325-1332 (2008)]</p>
<h3>Trends</h3> <ul style="list-style-type: none"> many countertrends lost in clutter  <p>[Effectiveness of Animation in Trend Visualization. Robertson et al. IEEE TVCG (Proc. InfoVis 2008), 14(5): 1325-1332 (2008)]</p>	<h3>Small Multiples</h3> <ul style="list-style-type: none"> individual plots get small  <p>[Effectiveness of Animation in Trend Visualization. Robertson et al. IEEE TVCG (Proc. InfoVis 2008), 14(5): 1325-1332 (2008)]</p>	<h3>Design</h3> <ul style="list-style-type: none"> 2 use: presentation vs. analysis (between-subjects) 3 vis encodings: animation vs. traces vs. small multis 2 dataset size: small vs. large <ul style="list-style-type: none"> 3 encoding x 2 size: within-subjects 24 tasks per participant <ul style="list-style-type: none"> 4 tasks x 3 encodings x 2 sizes 	<h3>Results</h3> <ul style="list-style-type: none"> small multiples more accurate than animation animation faster for presentation, slower for analysis <ul style="list-style-type: none"> than small multiples and trends dataset size matters (unsurprisingly)

<p>Critique</p> <ul style="list-style-type: none"> nice idea to investigate the gapminder phenomenon! well done study 	<p>Critique</p> <ul style="list-style-type: none"> field study participants: professional meteorologists <ul style="list-style-type: none"> two people: forecaster, technician interfaces: multiple programs used protocol <ul style="list-style-type: none"> talkaloud videotaped sessions with 3 cameras <p>[Turning Pictures into Numbers: Extracting and Generating Information from Complex Visualizations. Trafletti et al. Intl J. Human Computer Studies 53(5), 827-850.]</p>	<p>Pictures Into Numbers</p>	<p>Cognitive Task Analysis</p> <ul style="list-style-type: none"> initialize understanding of large scale weather build qualitative mental model (QMM) <ul style="list-style-type: none"> verify and adjust QMM write the brief task breakdown part of paper contribution
<p>Coding Methodology</p> <ul style="list-style-type: none"> interface <ul style="list-style-type: none"> which interface used whether picture/chart/graph usage (every utterance!) <ul style="list-style-type: none"> goal <ul style="list-style-type: none"> extract <ul style="list-style-type: none"> quant/qual goal-oriented/opportunistic integrated/unintegrated brief-writing <ul style="list-style-type: none"> quant/qual QMM/via/notes 	<p>Results</p> <ul style="list-style-type: none"> sig difference between vis used at CTA stages <ul style="list-style-type: none"> charts to build QMM images to verify/adjust QMM all kinds during brief-writing many others...  <p>[Turning Pictures into Numbers: Extracting and Generating Information from Complex Visualizations. Trafletti et al. Intl J. Human Computer Studies 53(5), 827-850.]</p> 	<p>Critique</p>	<p>Critique</p> <ul style="list-style-type: none"> video coding is huge amount of work, but very illuminating <ul style="list-style-type: none"> untangling complex story of real tool use methodology of CTA construction not discussed here <ul style="list-style-type: none"> often bottomup/topdown mix
<p>User Study Goals</p> <ul style="list-style-type: none"> compare systems <ul style="list-style-type: none"> (Untangling the Usability of Fishes Menus, Hornbæk 2007) characterize methods <ul style="list-style-type: none"> Slating the Horizon, Heer 2009 formative feedback <ul style="list-style-type: none"> Hotel Visits, Weaver 2007 Genealogical Graphs, McGuffin 2005 summative judgement <ul style="list-style-type: none"> LiveRAC, McLachlan 2008 Genealogical Graphs, McGuffin 2005 Preflux, Heer 2005 convince stakeholders <ul style="list-style-type: none"> (InfoVis Eval in Large Companies, Sudinal 2011) (Evaluation of Artery Visualizations for Heart Disease Diagnosis, Borkin 2011) 	<p>HemoVis: Round 1</p> <ul style="list-style-type: none"> formative qualitative study with experts task taxonomy led to design of HemoVis  <p>[Fig. 1. Borkin et al. Artery Visualizations for Heart Disease Diagnosis. Proc. InfoVis 2011.]</p>	<p>HemoVis: Round 2</p> <ul style="list-style-type: none"> experts balk: demand familiar 3D and rainbows 	<p>HemoVis: Round 2</p> <ul style="list-style-type: none"> experts balk: demand familiar 3D and rainbows quantitative user study <ul style="list-style-type: none"> advanced Harvard Med School students real patient data 91% with 2D/diverging vs. 39% with 3D/rainbow medical experts now convinced to use new system  <p>[Fig. 1. Borkin et al. Artery Visualizations for Heart Disease Diagnosis. Proc. InfoVis 2011.]</p>
<p>Credits</p> <ul style="list-style-type: none"> significant influence from Heidi Lam guest lecture <p>http://www.cs.ubc.ca/~tmcm/courses/cpsc533c-06-fall/#lect10</p>	<p>Readings For Next Time</p> <p>Process and Pitfalls in Writing Information Visualization Research Papers. Tamara Munzner. Chapter from Information Visualization: Human-Centered Issues and Perspectives. Andreas Kerren, John T. Stasko, Jean-Daniel Fekete, Chris North, eds. Springer LNCS Volume 4950, p. 134-153, 2008.</p> <p>Reproducible Research in Signal Processing - What, why, and how. Patrick Vandewalle, Jelena Kovacevic and Martin Vetterli. IEEE Signal Processing Magazine, 26(3):37-47, May 2009.</p>		