

Information Visualization

Aggregate & Filter 1

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Lect 17, 10 Mar 2020
<https://www.cs.ubc.ca/~tmm/courses/436V-20>

Upcoming

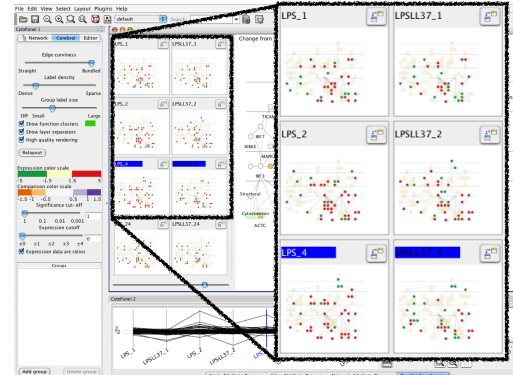
- Foundations 5: out Thu Mar 12, due Wed Mar 18 11:59pm
- Milestone 2: due Wed Mar 25 11:59pm
 - (with update announce last week, schedule status component)

Correction

Idiom: **Small multiples**

- encoding: same
- data: none shared
 - different attributes
- different items (different condition keys, same gene keys), same attributes: expression values for node colors**
 - (same network layout for nodes=genes)
- navigation: shared

System: **Cerebral**



[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008) 14:6 (2008), 1253–1260.]

Reminder

Beyond slides: Textbook for further reading (optional)

- Intro
 - Ch 1. What's Vis, and Why Do It?
- Data Abstraction
 - Ch 2. What: Data Abstraction
 - Ch 4. Analysis: Four Levels for Validation
- Task Abstraction
 - Ch 3. Why: Task Abstraction
- Marks & Channels
 - Ch 5. Marks and Channels
- Multivariate Tables
 - Ch 7. Arrange Tables
- Interactive Views
 - Ch 11. Manipulate View
- Maps
 - Ch 8. Arrange Spatial Data (only 8.1–8.3)
- Color
 - Ch 10. Map Color and Other Channels
- Networks & Trees
 - Ch 9. Arrange Networks and Trees
- Aggregation
 - Ch 13. Reduce Items and Attributes
 - Ch 14. Embed: Focus+Context
- Rules of Thumb (upcoming)
 - Ch 6. Rules of Thumb

Visualization Analysis & Design, free through library: [catalog page](#) [EZProxy direct link](#)


Filter & Aggregate

Exercise: Too much stuff

- Cars dataset: 7 attributes
 - MPG quantitative
 - Cylinders ordinal
 - Horsepower quantitative
 - Weight quantitative
 - Acceleration quantitative
 - Model Year ordinal
 - Origin categorical
- This table has 100 million items
- Pair up, discuss how to have scalable approach, create sketch to illustrate
 - [8 min]
 - Socrative: true when done

How to handle complexity: 1 previous strategy + 3 more

→ Derive



- derive new data to show within view
- change view over time
- facet across multiple views
- reduce items/attributes within single view

Manipulate

- Change
- Select
- Navigate

Facet

- Juxtapose
- Partition
- Superimpose

Reduce

- Filter
- Aggregate
- Embed

How?

Encode

- Arrange
 - Express
 - Order
 - Use
- Map from categorical and ordered attributes
 - Color
 - Hue
 - Saturation
 - Luminance
 - Size, Angle, Curvature, ...
 - Shape
 - Motion
 - Direction, Rate, Frequency, ...

Manipulate

- Change
- Select
- Navigate

Facet

- Juxtapose
- Partition
- Superimpose

Reduce

- Filter
- Aggregate
- Embed

What?

Why?

How?

Reducing Items and Attributes

Filter

- Items
- Attributes

Aggregate

- Items
- Attributes

Reduce items and attributes

Reducing Items and Attributes

- Filter
- Attributes

Aggregate

- Items
- Attributes

Reduce

- Filter
- Aggregate
- Embed

Filter

- eliminate some elements
 - either items or attributes
- according to what?
 - any possible function that partitions dataset into two sets
 - attribute values bigger/smaller than x
 - noise/signal
- filters vs queries
 - query: start with nothing, add in elements
 - filters: start with everything, remove elements
 - best approach depends on dataset size

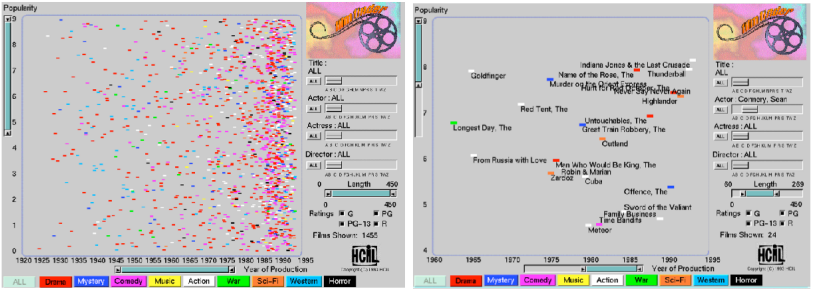
Reducing Items and Attributes

Filter

- Items
- Attributes

Idiom: FilmFinder

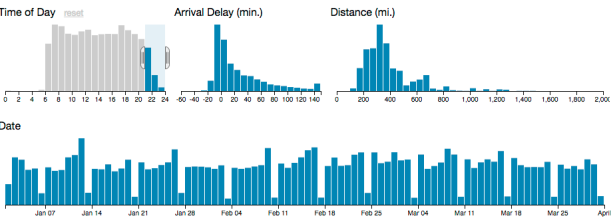
- dynamic queries/filters for items
 - tightly coupled interaction and visual encoding idioms, so user can immediately see results of action



Idiom: cross filtering

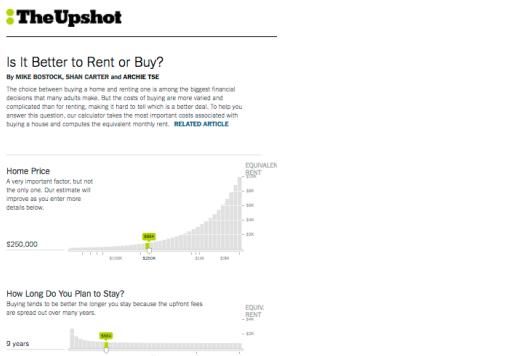
System: Crossfilter

- item filtering
- coordinated views/controls combined
 - all scented histogram bislders update when any ranges change



[<http://square.github.io/crossfilter/>]

Idiom: cross filtering



[https://www.nytimes.com/interactive/2014/upshot/buy-rent-calculator.html?_r=0]

Aggregate

- a group of elements is represented by a smaller number of derived elements

➔ Aggregate

→ Items

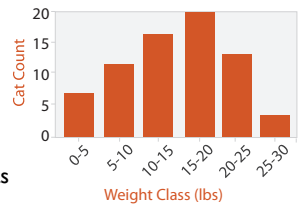


→ Attributes



Idiom: histogram

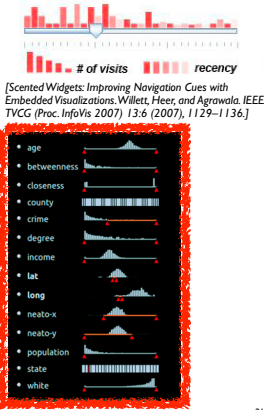
- static item aggregation
- task: find distribution
- data: table
- derived data
 - new table: keys are bins, values are counts
- bin size crucial
 - pattern can change dramatically depending on discretization
 - opportunity for interaction: control bin size on the fly



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Idiom: scented widgets

- augmented widgets show *information scent*
 - better cues for *information foraging*: show whether value in drilling down further vs looking elsewhere
- concise use of space: histogram on slider

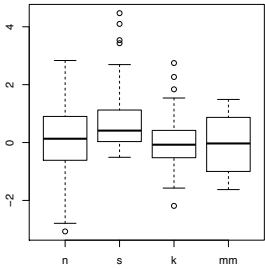


[Multivariate Network Exploration and Presentation: From Detail to Overview via Selections and Aggregations. van den Elzen, van Wijk, IEEE TVCG 20(12):2014 (Proc. InfoVis 2014).]

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Idiom: boxplot

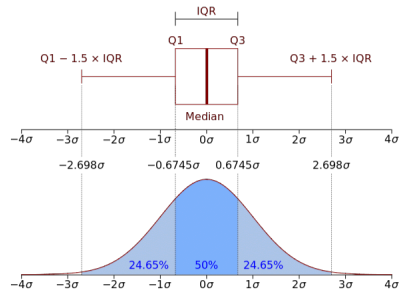
- static item aggregation
- task: find distribution
- data: table
- derived data
 - 5 quant attrbs
 - median: central line
 - lower and upper quartile: boxes
 - lower upper fences: whiskers
 - values beyond which items are outliers
 - outliers beyond fence cutoffs explicitly shown
- scalability
 - unlimited number of items!



[40 years of boxplots. Wickham and Stryjewski. 2012. had.co.nz]

Boxplots

- aka box-and-whisker plots
 - show outliers as points
- bad for non-normal distributions
- really bad for bimodal or multimodal distributions

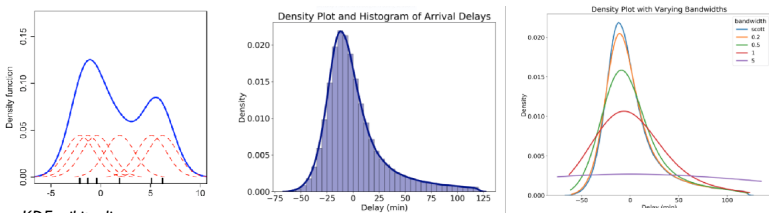


[wikipedia]

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Density plots

- aka kernel density plots, kernel density estimation (KDE)
 - smoothed, continuous version of a histogram estimated from data
 - continuous curve (the kernel, usually Gaussian bell curve) drawn at each data point
 - add curves together for single smooth density estimation
 - bandwidth influences estimate

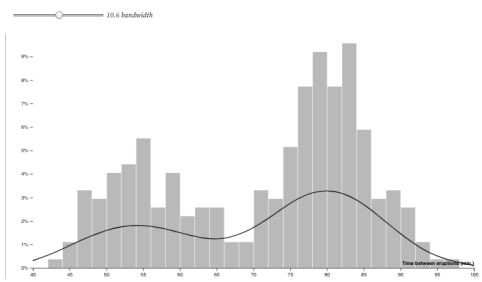


KDE wikipedia

<https://towardsdatascience.com/histograms-and-density-plots-in-python-f6bda88f5ac0>

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KDE in D3: Interactive bandwidth controls

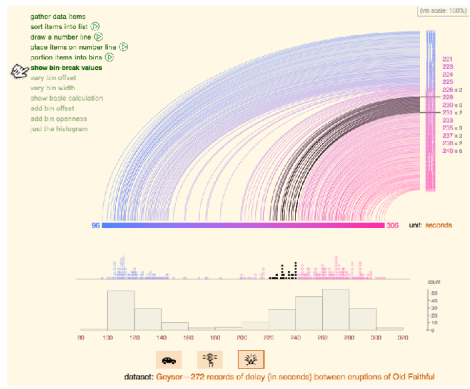


<https://observablehq.com/@d3/kernel-density-estimation>

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Histograms explained

- also great example of scrollytelling!

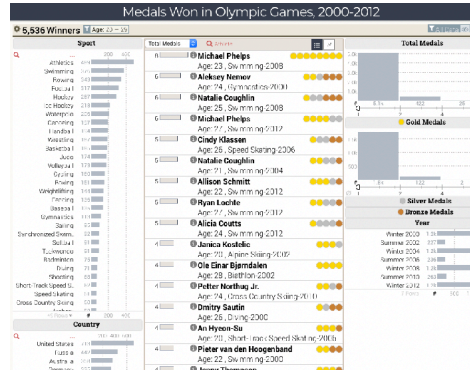


<http://tinlizzie.org/histograms/>

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Example: Keshif

- interactive item filtering with scented widgets
 - also: interaction speed w/ scatterplot vs list view

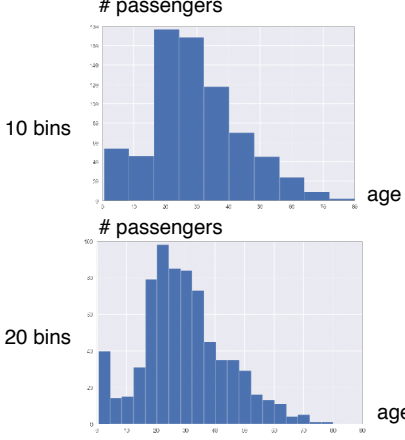


<https://keshif.me/gallery/olympics>

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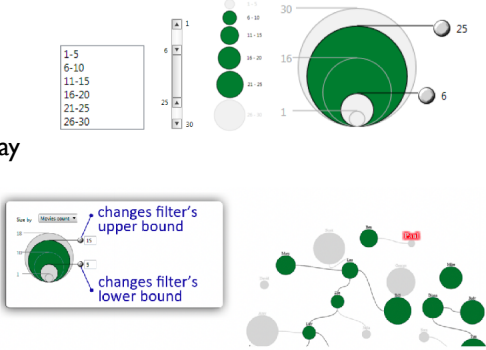
Histogram bins

- good # bins hard to predict
 - make it interactive when possible
- rules of thumb
 - # bins = \sqrt{n}
 - # bins = $\log_2(n)+1$



Interactive legends

- controls combining
 - visual representation of static legends w/
 - interaction mechanisms of widgets
- define & control visual display together

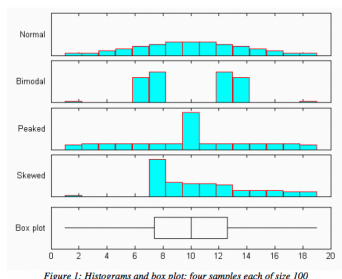


Riche 2010

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Boxplots: Drawbacks

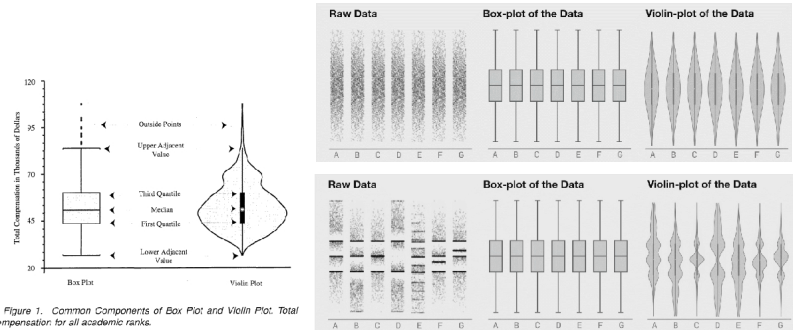
- four distributions with same boxplot



http://stat.mq.edu.au/wp-content/uploads/2014/05/Can_the_Box_Plot_be_Improved.pdf

Violin plots

- boxplot + probability density function

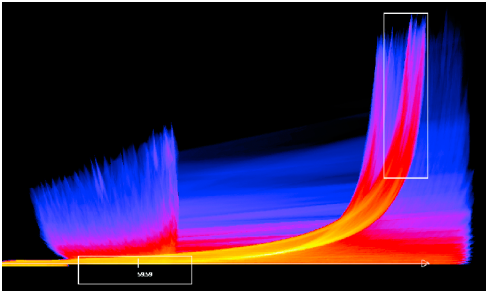


<https://towardsdatascience.com/violin-plots-explained-fb1d115e023d>

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Idiom: Continuous scatterplot

- static item aggregation
- data: table
- derived data: table
 - key attribs x,y for pixels
 - quant attrib: overplot density
- dense space-filling 2D matrix
- color: sequential categorical hue + ordered luminance colormap
- scalability
 - no limits on overplotting: millions of items



[Continuous Scatterplots. Bachthaler and Weiskopf. IEEE TVCG (Proc.Vis 08) 14:6 (2008), 1428–1435. 2008.]

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Credits

- Visualization Analysis and Design (Ch 13, 14)
- Alex Lex & Miriah Meyer, <http://dataviscourse.net/>

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