

### ENSURE LEGIBILITY AND FOCUS ON THE MESSAGE

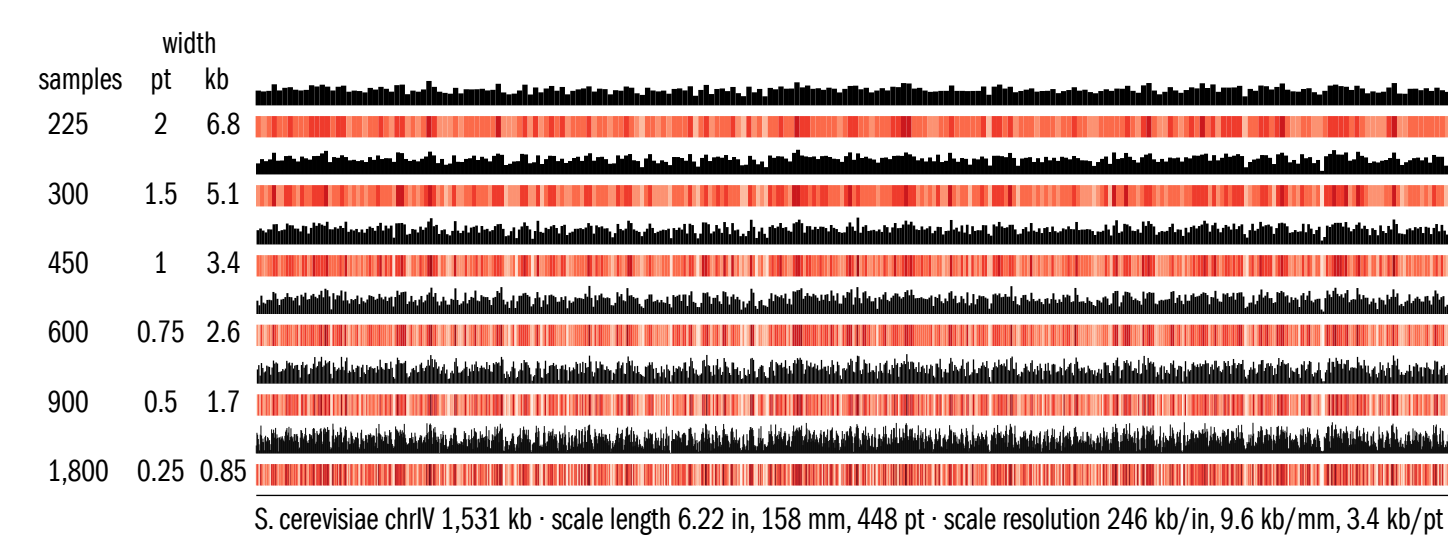
Create legible visualizations with a strong message. Make elements large enough to be resolved comfortably. Bin dense data to avoid sacrificing clarity.

### Distinguish between exploration and communication.

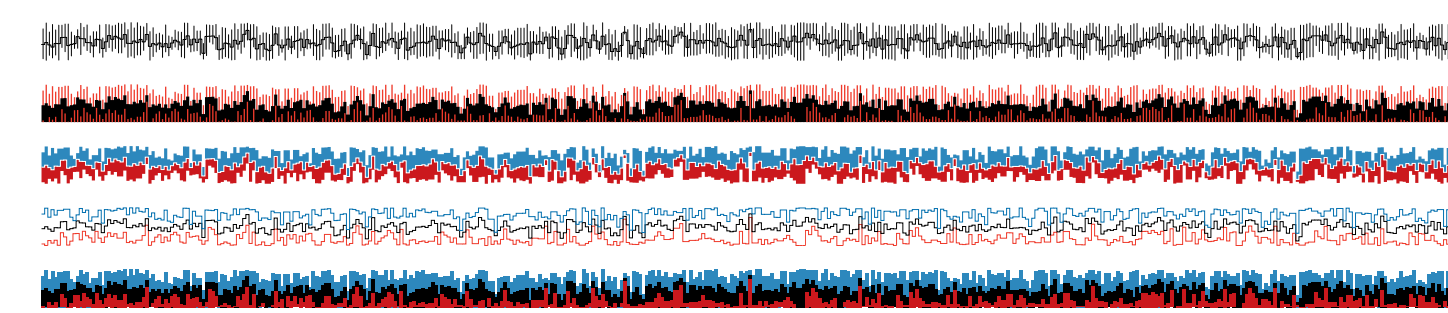
Use exploratory tools (e.g. genome browsers) to discover patterns and validate hypotheses. Avoid using screenshots from these applications for communication – they are typically too complex and cluttered with navigational elements to be an effective static figure.

### Do not exceed resolution of visual acuity.

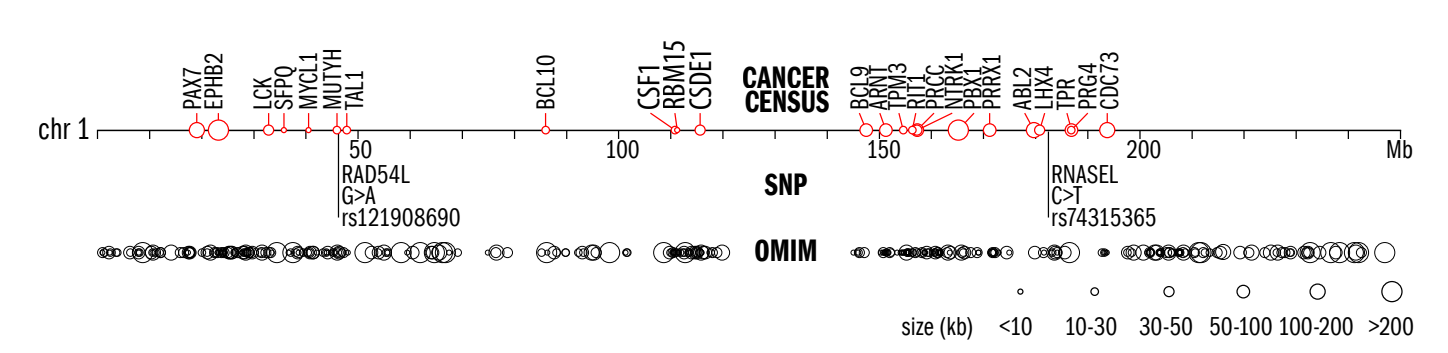
Our acuity is ~50 cycles/degree or about 1/200 (0.3 pt) at 10 inches. Ensure the reader can comfortably see detail by limiting resolution to no more than 50% of acuity. Where possible, elements that require visual separation should be at least 1 pt part.



Rendering human chromosome 1 (249 Mb) with 450 divisions provides a resolution of 550 kb, which is larger than 98% of all human genes. To depict 50% of genes without magnification we would need to use 15 kb bins, which would have to be 1/16,600 of the figure (0.0004 in, 10 microns!)



Approaches to encoding min/avg/max values of downsampled data. In the top hi-low trace, the vertical bars are perceived as a separate layer and effectively show variance without obscuring trends in the average.



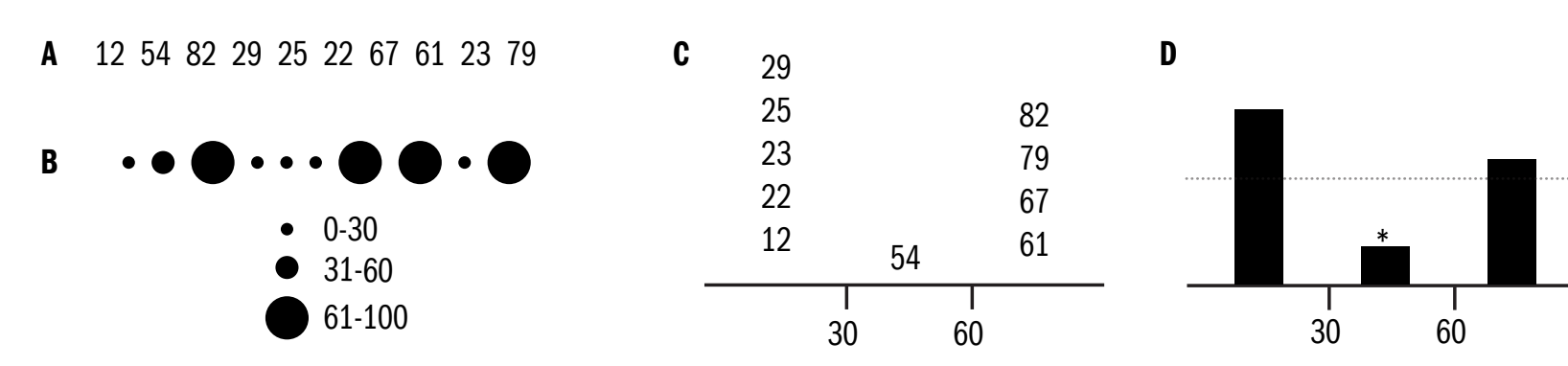
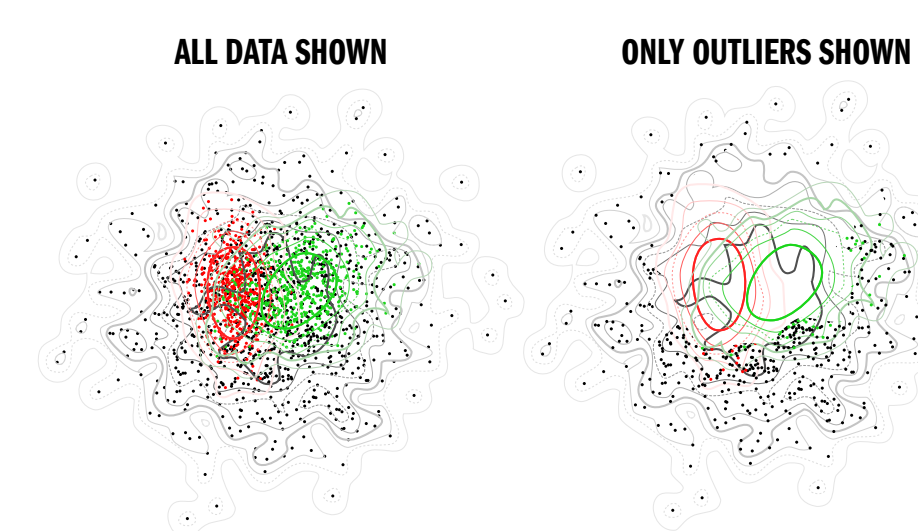
When drawing the position and size of densely packed genes, encode the gene's size using a non-linear mapping. When the number of data values is large, such as in the OMIM gene track, hollow glyphs are effective. For even greater number of points, a density map is preferred.

### Aggregate data for focused theme.

A strong visual message has no uncertainty in its interpretation. Focus on a single theme by aggregating unnecessary detail.

### Show density maps and outliers.

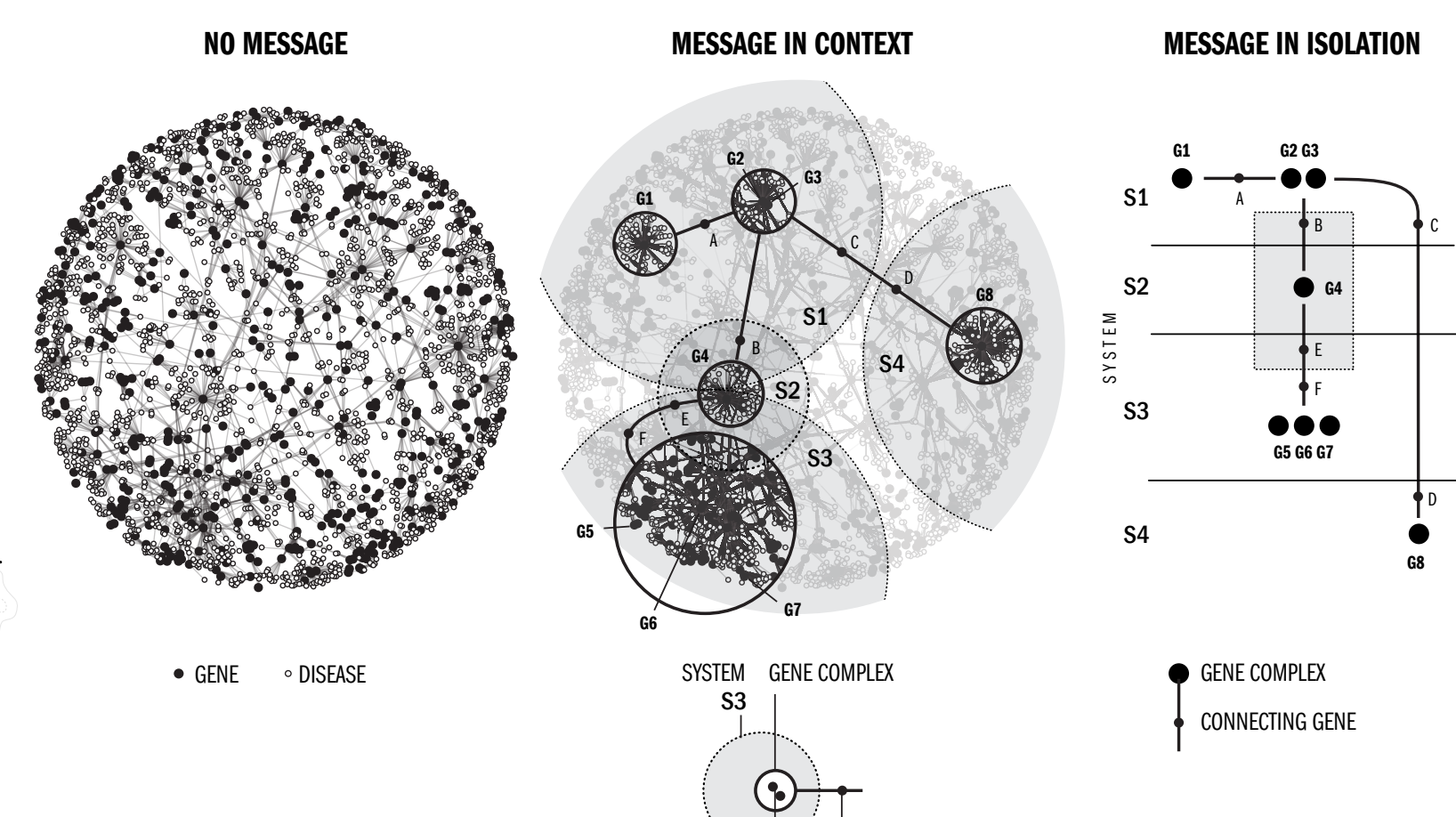
Establishing context is helpful when emergent patterns in the data provide a useful perspective on the message. When data sets are large, it is difficult to maintain detail in the context layer because the density of points can visually overwhelm the area of interest. In this case, consider showing only the outliers in the data set.



What is communicated? (A) The raw data imparts no clear message. (B) Binning indicates ranges, not individual values, are important. (C). Frequency distribution suggests that there is a shortage of medium-sized values. (D) Individual data points can be removed to emphasize trend and significance.

### Consider whether showing the full data set is useful.

The reader's attention can be focused by increasing the salience of interesting patterns. Other complex data sets, such as networks, are shown more effectively when context is carefully edited or even removed.



### USE EFFECTIVE VISUAL ENCODINGS TO ORGANIZE INFORMATION.

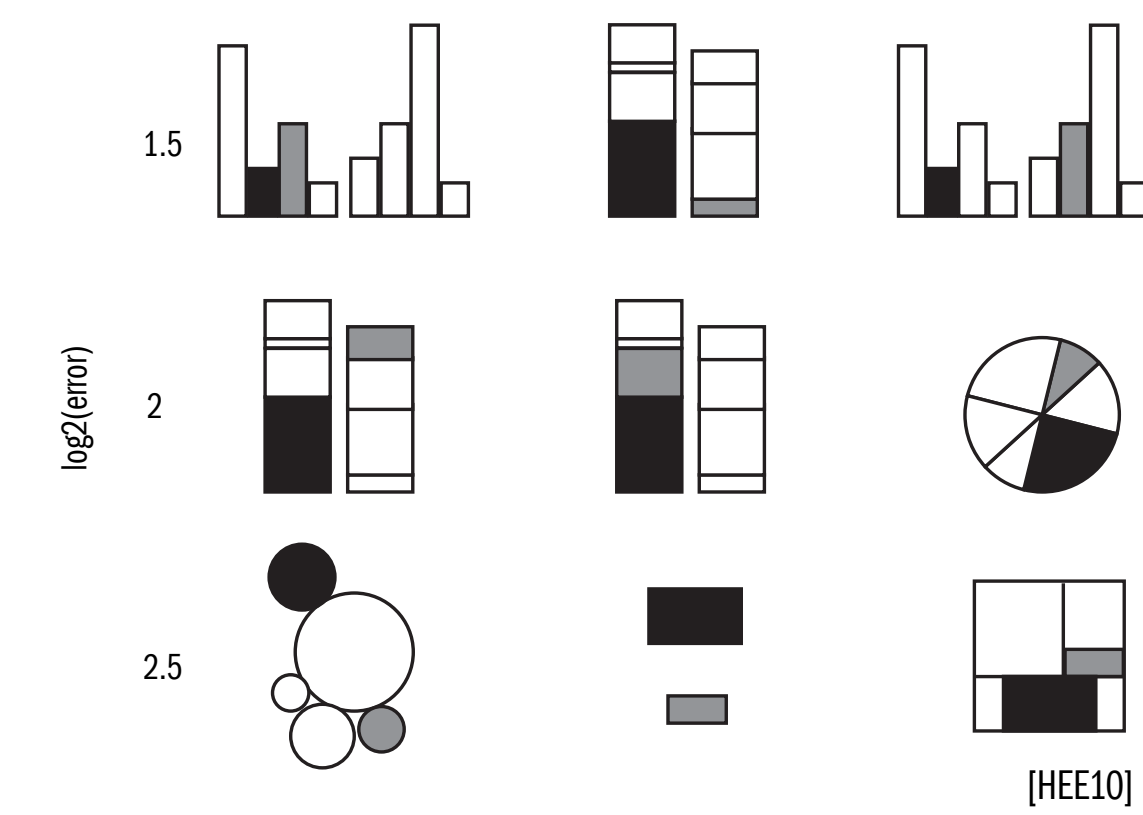
Match the visual encoding to the hypothesis. Use encodings specific and sensitive to important patterns. Dense annotations should be independent of the core data in distinct visual layers.

### Use the simplest encoding.

Choose concise encodings over elaborate ones.

### Help the reader judge accurately.

Accuracy and speed in detecting differences in visual forms depends on how information is presented. We judge relative lengths more accurately than areas, particularly when elements are aligned and adjacent. Our judgment of area is poor because we use length as a proxy, which causes us to systematically underestimate.



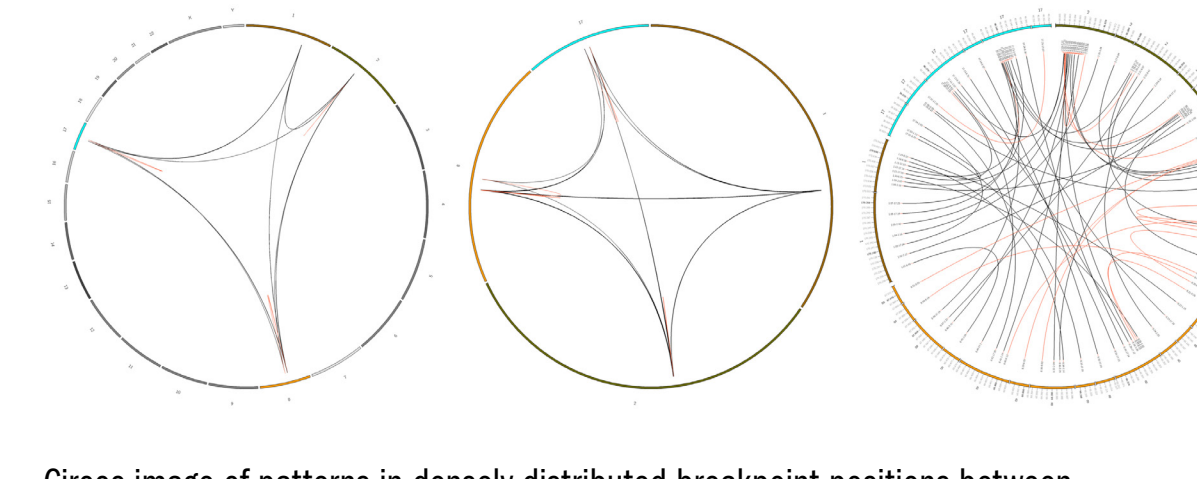
### Use perceptual palettes.

Selecting perceptually favorable colors is difficult because most software does not support the required color spaces. Brewer palettes [BRE11] exist for the full range of colors to help us make useful choices. Qualitative palettes have no perceived order of importance. Sequential palettes are suitable for heat maps because they have a natural order and the perceived difference between adjacent colors is constant. Twin hue diverging palettes, are useful for two-sided quantitative encodings, such as immunofluorescence and copy number.



### Crop scale to reveal fine structure in data.

Biological data sets are typically high-resolution (changes at base pair level can be meaningful), sparse (distances between changes are orders of magnitude greater than the affected areas) and connect distant regions by adjacency relationships (gene fusions and other rearrangements). It is difficult to take these properties into account on a fixed linear scale, the kind used by traditional genome browsers. To mitigate this, crop and order axis segments arbitrarily and apply a scale adjustment to a segment or portion thereof.



Circos image of patterns in densely distributed breakpoint positions between chromosomes 1, 2, 9, and 17 are emphasized when scale is zoomed. (A) full genome (B) chrs 1, 2, 9, 17 (C) 31 regions on chrs 1, 2, 9, 17 totaling 223 kb [KRZ09].

### USE EFFECTIVE DESIGN PRINCIPLES TO EMPHASIZE AND COMMUNICATE PATTERNS.

Well-designed figures illustrate complex concepts and patterns that may be difficult to express concisely in words. Figures that are clear, concise and attractive are effective – they form a strong connection with the reader and communicate with immediacy. These qualities can be achieved with methods of graphic design, which are based on theories of how we perceive, interpret and organize visual information.

### Reduce unnecessary variation.

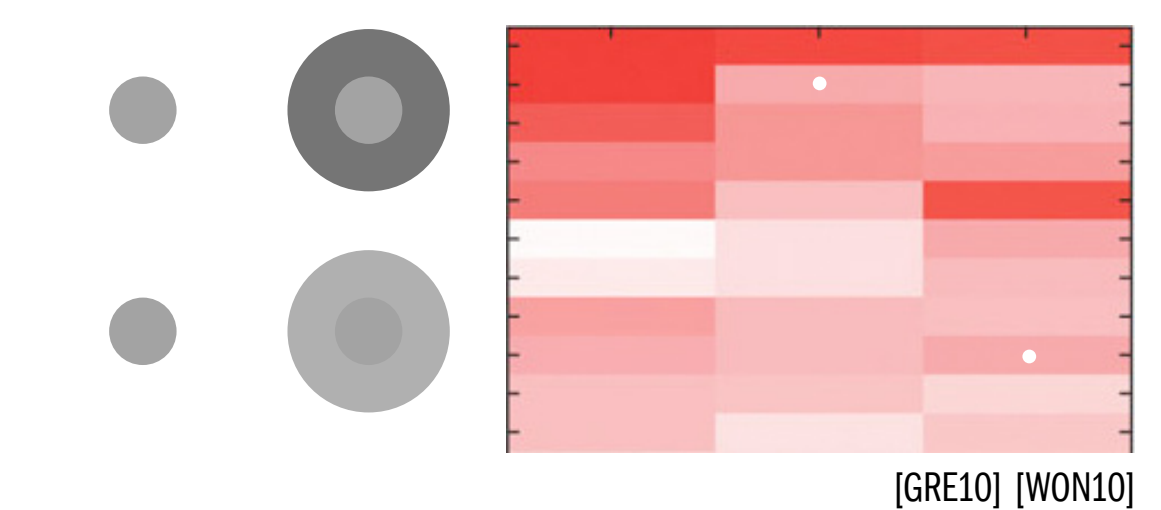
The reader does not know what is important in a figure and will assume that any spatial or color variation is meaningful. The figure's variation should come solely from data or act to organize information.



You can count the objects on the left almost instantly, without conscious effort. This is called pre-attentive cognition. This example uses the concept of proximity to partition the dots into pre-attentive groupings. Elements can be effectively organized by reducing spatial variation, here achieved by symmetrical layout and two levels of spacing.

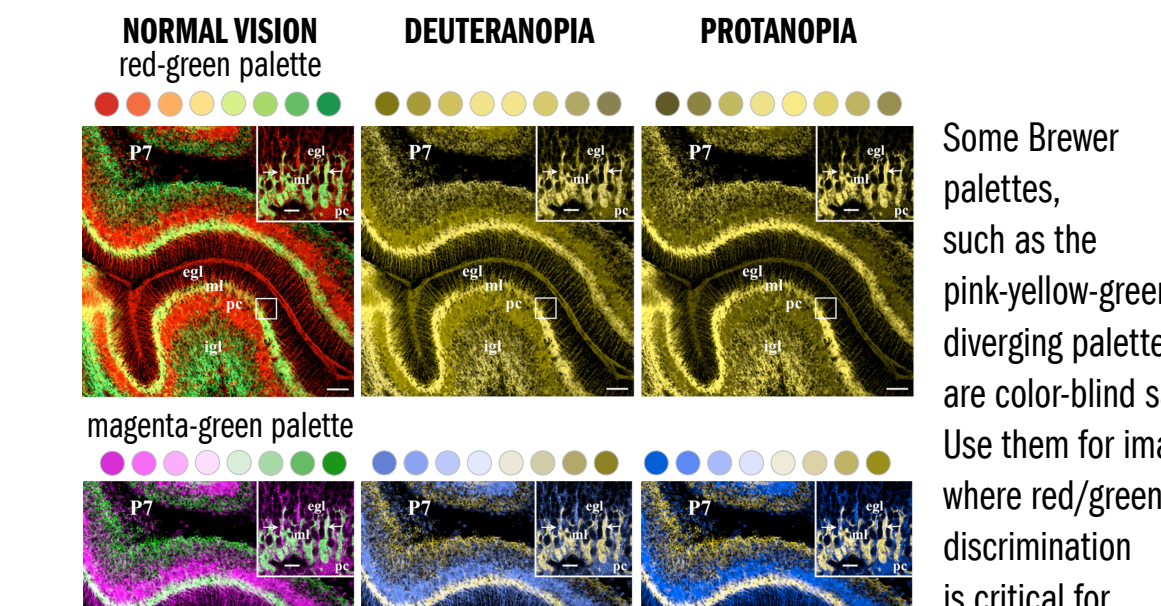
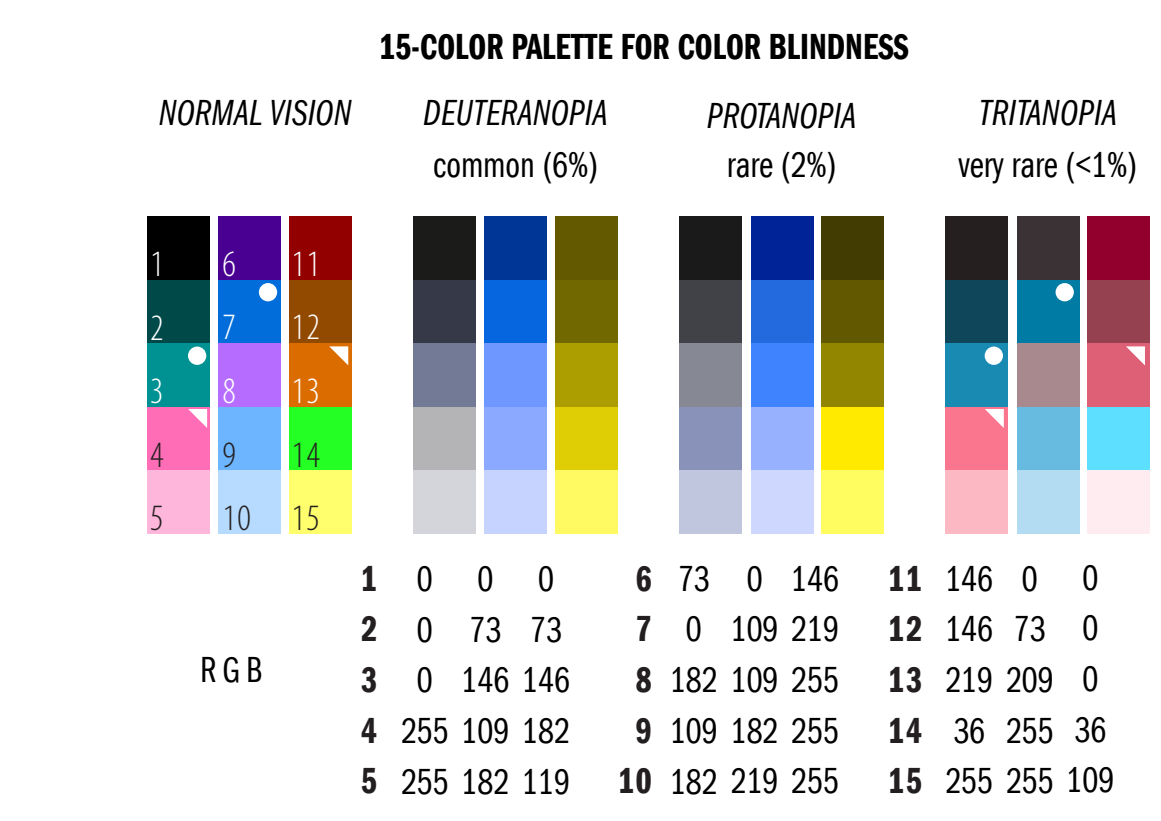
### Be aware of the luminance effect.

Color is a useful encoding – the eye can distinguish about 450 levels of gray, 150 hues, and 10-60 levels of saturation, depending on the color – but our ability to perceive differences varies with context. Adjacent tones with different luminance values can interfere with discrimination, in interaction known as the luminance effect.



### Be aware of color blindness.

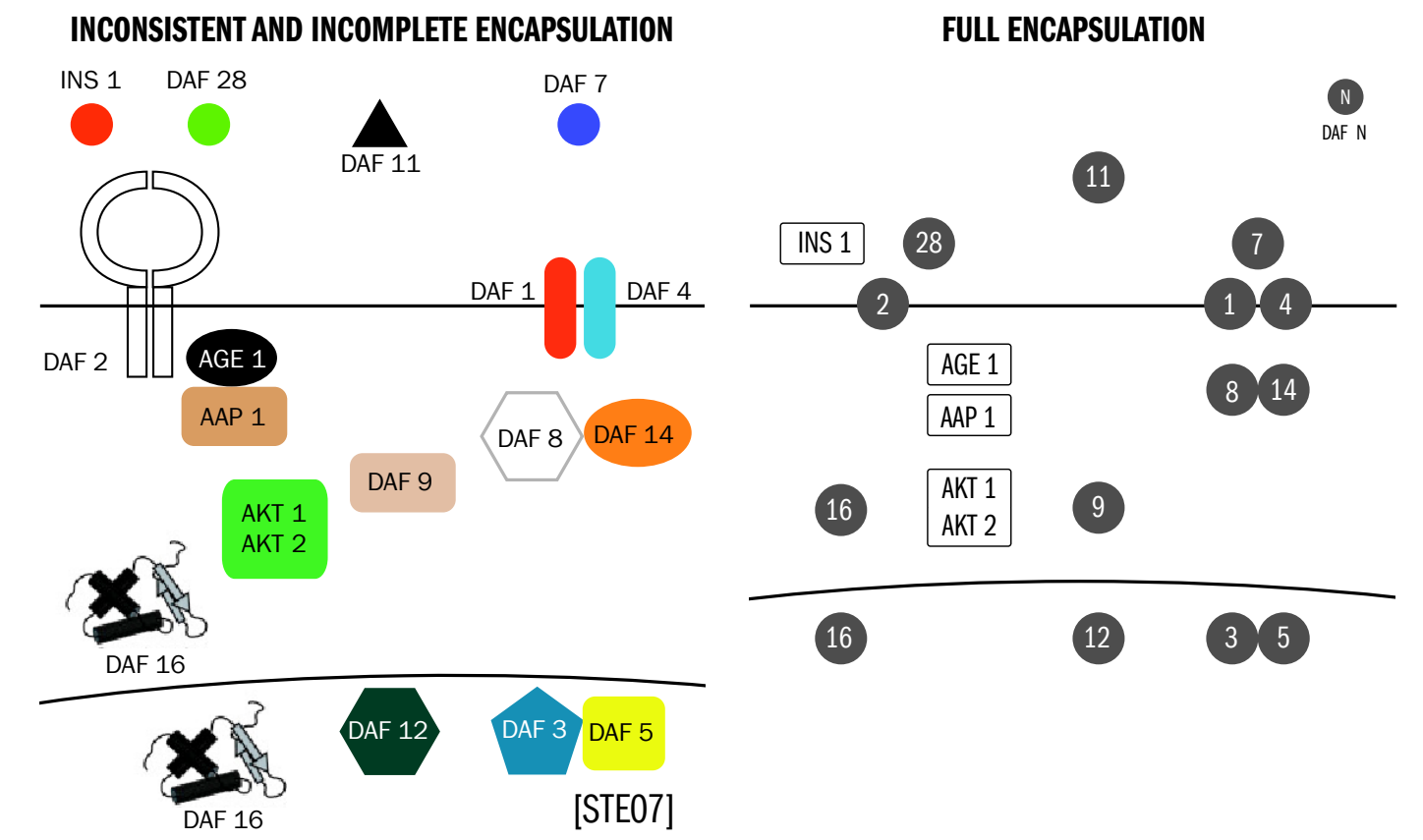
In an audience of 8 men and 8 women, chances are 50% that at least one has some degree of color blindness. Use a palette that is color-blind safe. In the palette below the 15 colors appear as 5-color tone progressions to those with color blindness. Additional encodings can be achieved with symbols or line thickness.



Some Brewer palettes, such as the pink-yellow-green diverging palette, are color-blind safe. Use them for images where red/green discrimination is critical for comprehension.

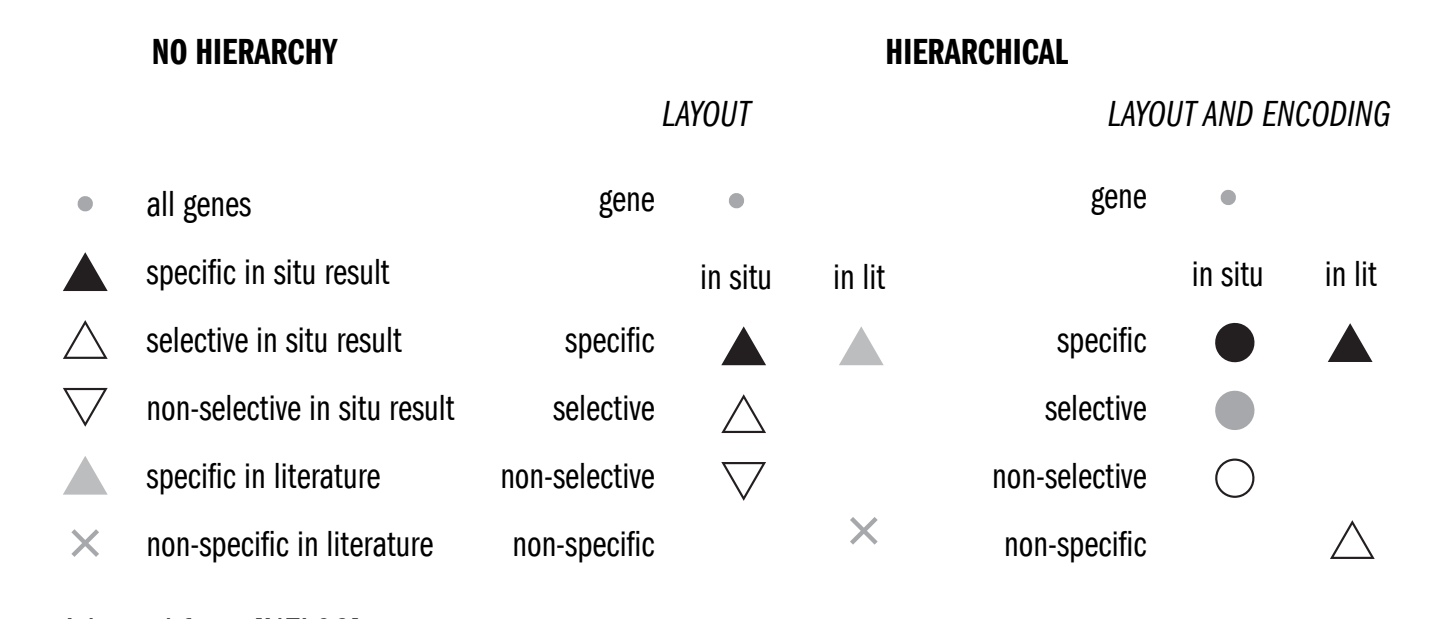
### Encapsulate details.

Including details not relevant to the core message of the figure can create confusion. Encapsulation should be done to the same level of detail and to the simplest visual form. Duplication in labels should be avoided unless required to preserve semantic forms.



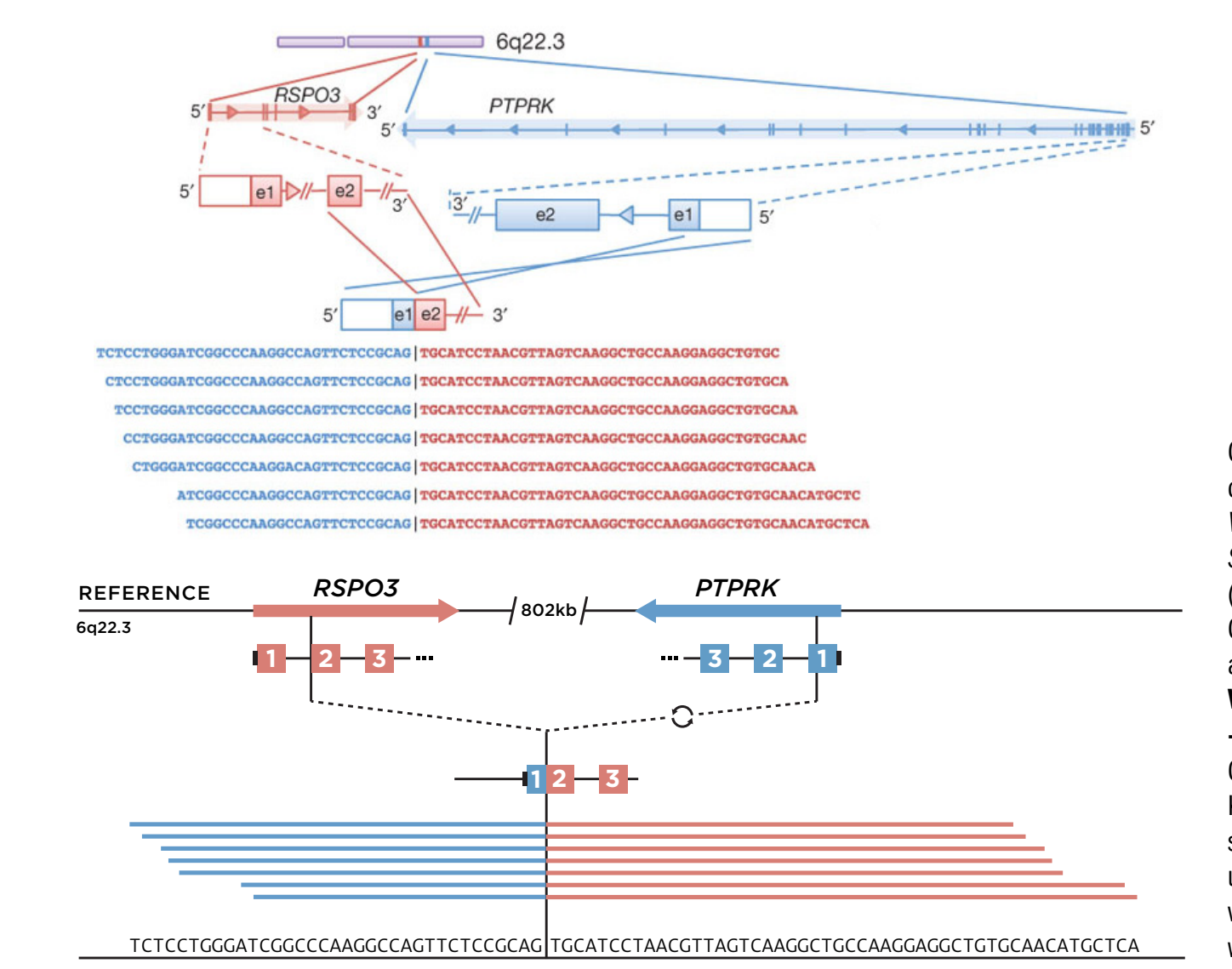
### Respect natural hierarchies.

When the data set embodies a natural hierarchy, use an encoding that emphasizes it clearly and memorably. The use hierarchy in layout (e.g. tabular form) and encoding can significantly improve a muddled figure.



### Use consistent alignment. Center on theme.

Establish equivalence using consistent alignment. Awkward callouts can be avoided if elements are logically placed.



Complex information can be organized by consistent alignment. Notice how placing the gene fusion product in the center emphasizes the subject of the figure. Explaining the concept of gene fusion does not require the internal structure of the two genes, whose size and exon count/distribution should be removed to improve clarity. Original figure from [SES12].

[BRE11] C Brewer (2011) Color Brewer: <http://www.colorbrewer.org>  
 [GRE10] HE Genzco et al. (2010) In situ analysis of tyrosine phosphorylation networks by FLIM on cell arrays. Nat Methods 7: 467-472.  
 [HEE10] J Heer et al. (2010) Crowdsourcing graphical perception: using mechanical turk to assess visualization design. Proceedings of the 28th international conference on human factors in computing systems. Atlanta, Georgia, USA, ACM, pp. 203-212.  
 [KRZ09] M Krzywinski et al. (2009) Circos: an information aesthetic for comparative genomics. Genome Res. vol. 19, pp. 1639-45  
 [KRZ11] M Krzywinski et al. (2011) Hive plots – rational approach to visualizing networks. Brief Bioinform. Dec 9 2011.  
 [NEL03] S Nelander et al. (2003) Prediction of cell type-specific gene modules: identification and initial characterization of a core set of smooth muscle-specific genes. Genome Res 13: 1838-1854.  
 [SES12] S Seshagiri et al. (2012) Recurrent R-spondin fusions in colon cancer. Nature 488: 660-664.  
 [STE07] SE Van Steen et al. (2007) Cell-specific microarray profiling experiments reveal a comprehensive picture of gene expression in the C. elegans nervous system. Genome Biol 8: R135.  
 [WON10] B Wong (2010) Points of view: Color coding. Nat Methods 7: 573.

Content of this poster is drawn from my book chapter Visualization Principles for Scientific Communication (Martin Krzywinski & Jonathan Corum) in the upcoming open access Cambridge Press book Visualizing Biological Data - a practical guide (Seán L. O'Donoghue, James B. Procter, Kate Patterson, eds.), a survey of best practices and unsolved problems in biological visualization. This book project was conceptualized and initiated at the Vizi 2011 conference.

If you are interested in guidelines for data encoding and visualization in biology, see the Points of View column by Bang Wong in Nature Methods.



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