

Developing scientific data visualizations for non-specialist audiences

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data, n. (Oxford English Dictionary)

1. As a count noun: an item of information; a datum; a set of data.
2. As a mass noun:
 - a. **Related items of (chiefly numerical) information considered collectively, typically obtained by scientific work and used for reference, analysis, or calculation.**
 - b. Computing. Quantities, characters, or symbols on which operations are performed by a computer, considered collectively.
Also (in non-technical contexts): information in digital form.

visualization, n. (Oxford English Dictionary)

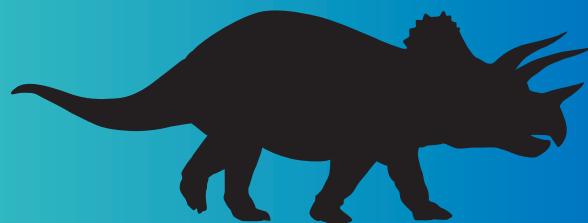
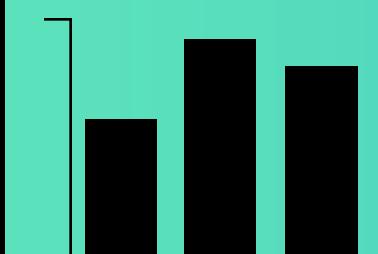
1. The action or fact of visualizing; the power or process of forming a mental picture or vision of something not actually present to the sight; a picture thus formed.
2. **The action or process of rendering visible.**

data visualization

Information (chiefly numerical),
made visible.

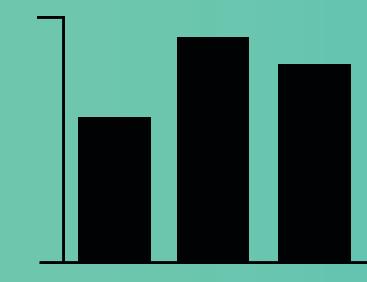
data visualization

Information (chiefly numerical),
made visible.

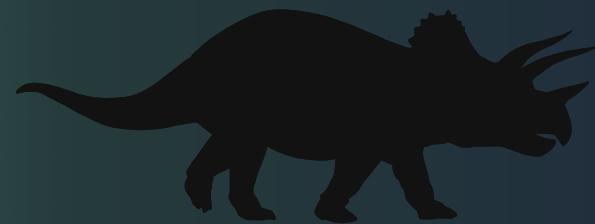


data visualization

Information (chiefly numerical),
made visible.



abstract representations



figurative illustrations

[content]

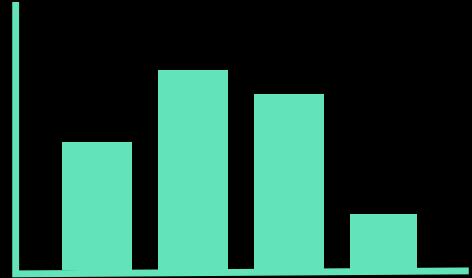
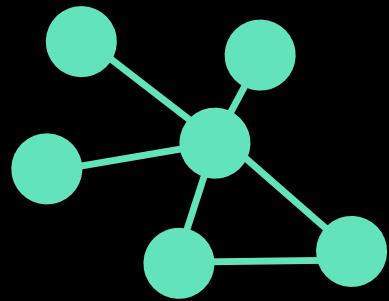
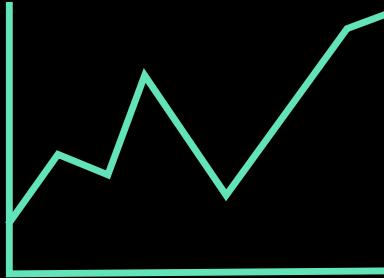
Information (chiefly numerical),
made visible.

[context]

For whom and why?

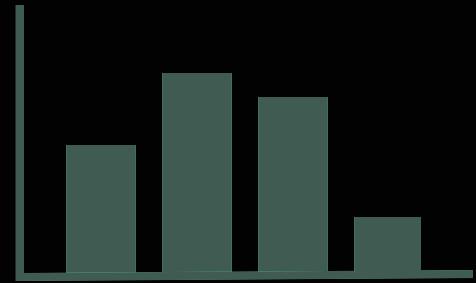
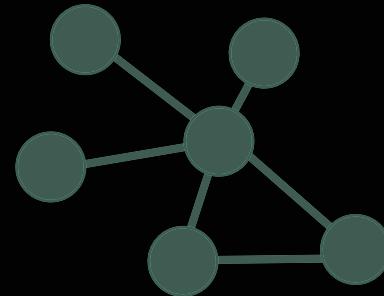
[content]

data / information



[content]

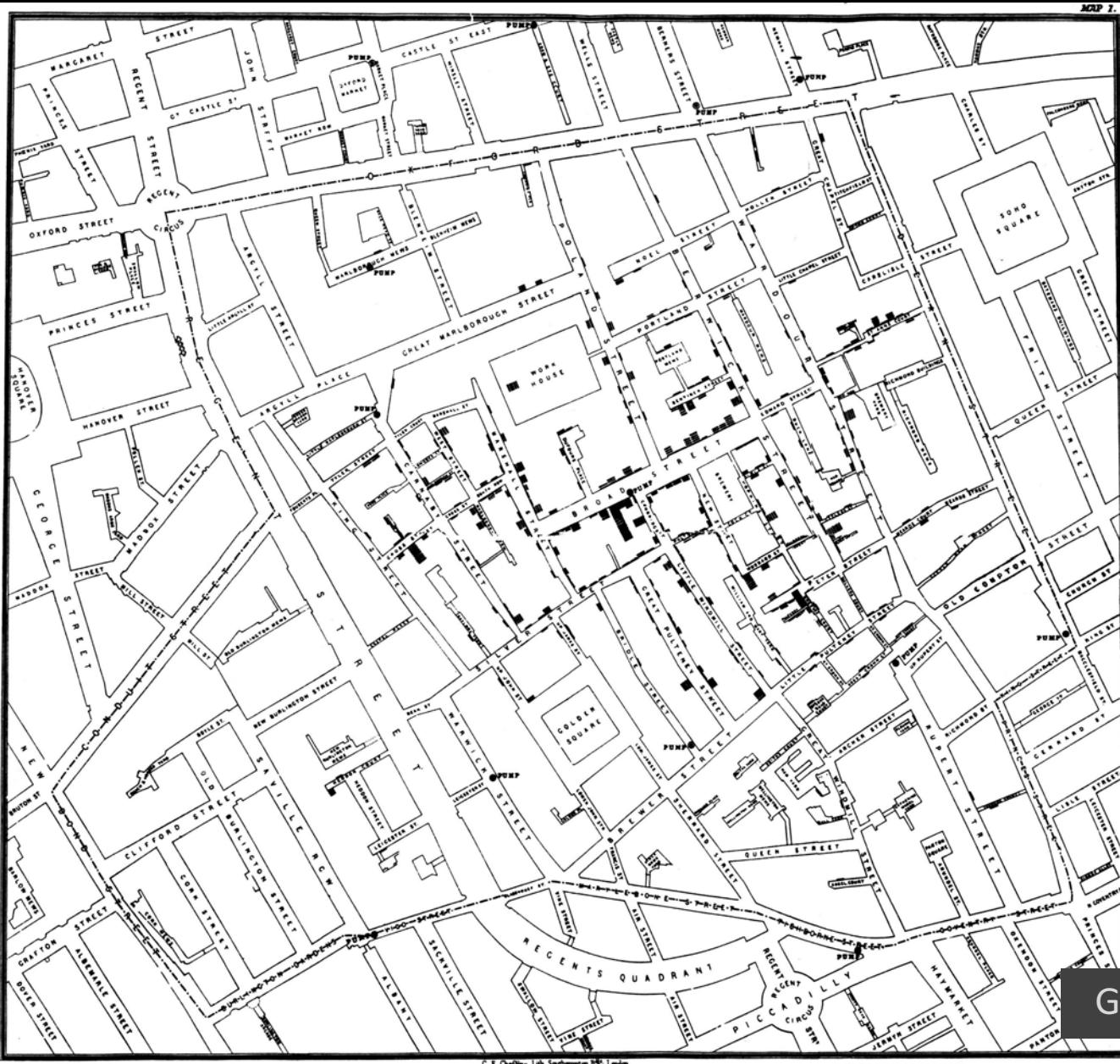
data / information



[context]

audience and purpose

investigative tool



Graphic by John Snow (~1854)

communicate results to peers

Evaluation of Artery Visualizations for Heart Disease Diagnosis

Michelle A. Borkin, Student Member, IEEE, Krzysztof Z. Gajos, Amanda Peters, Dimitrios Mitsouras, Simone Melchionna, Frank J. Rybicki, Charles L. Feldman, & Hanspeter Pfister, Senior Member, IEEE

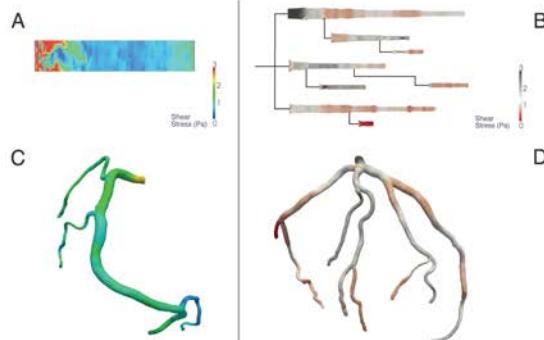


Fig. 1. Left: Traditional 2D projection (A) of a single artery, and 3D representation (C) of a right coronary artery tree with a rainbow color map. Right: 2D tree diagram representation (B) and equivalent 3D representation (D) of a left coronary artery tree with a diverging color map.

Abstract— Heart disease is the number one killer in the United States, and finding indicators of the disease at an early stage is critical for treatment and prevention. In this paper we evaluate visualization techniques that enable the diagnosis of coronary artery disease. A key physical quantity of medical interest is endothelial shear stress (ESS). Low ESS has been associated with sites of lesion formation and rapid progression of disease in the coronary arteries. Having effective visualizations of a patient's ESS data is vital for the quick and thorough non-invasive evaluation by a cardiologist. We present a task taxonomy for hemodynamics based on a formative user study with domain experts. Based on the results of this study we developed HemoVis, an interactive visualization application for heart disease diagnosis that uses a novel 2D tree diagram representation of coronary artery trees. We present the results of a formal quantitative user study with domain experts that evaluates the effect of 2D versus 3D artery representations and of color maps on identifying regions of low ESS. We show statistically significant results demonstrating that our 2D visualizations are more accurate and efficient than 3D representations, and that a perceptually appropriate color map leads to fewer diagnostic mistakes than a rainbow color map.

Index Terms— Quantitative evaluation, qualitative evaluation, biomedical and medical visualization.

1 INTRODUCTION

In the United States, the leading cause of death is heart disease resulting in over 600,000 deaths per year [24]. Early prevention and treatment is vital for saving lives, and visualization plays an essential role

for patient diagnosis in cardiovascular imaging. A new non-invasive diagnostic technique under development uses Computed Tomography Angiography (CTA) data from patients combined with blood flow simulations to calculate hemodynamic risk factors, in particular Endothelial Shear Stress (ESS), in coronary arteries [40]. Visualization methods of this data are of great value for this emerging research and have the potential to lead to faster, more accurate heart disease diagnoses.

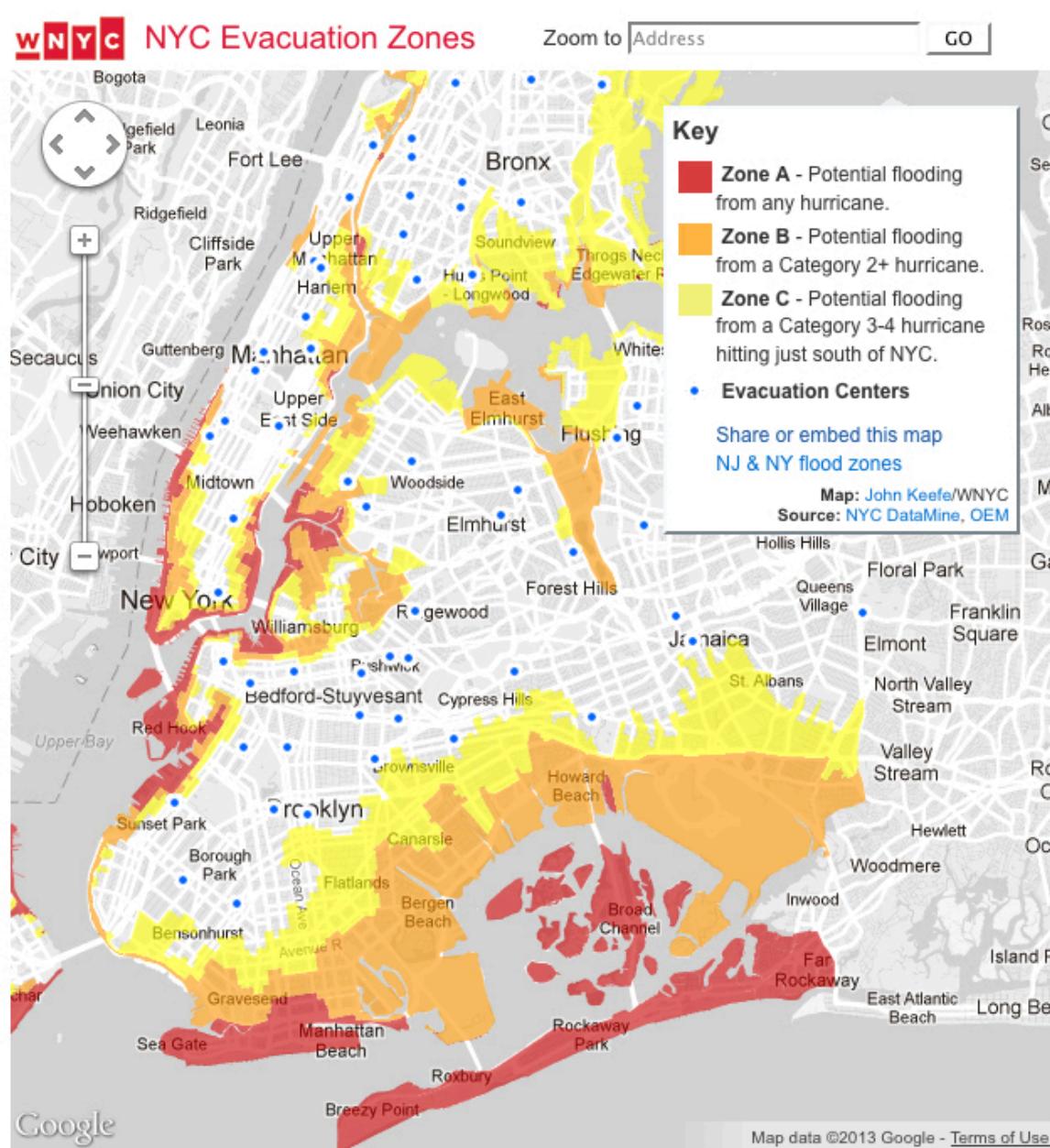
Current visualization techniques, as shown in Fig. 1 (left), use either a 2D side-by-side projection of a single artery or a 3D representation of the structure. These approaches are limited by the lack of advanced information about individual vessels, such as their branching structure, action segments, and specific locations.

From Michelle Anne Borkin et al. 2011. “Evaluation of artery visualizations for heart disease diagnosis.” *IEEE Transactions on Visualization and Computer Graphics* 17(12): 2479-2488

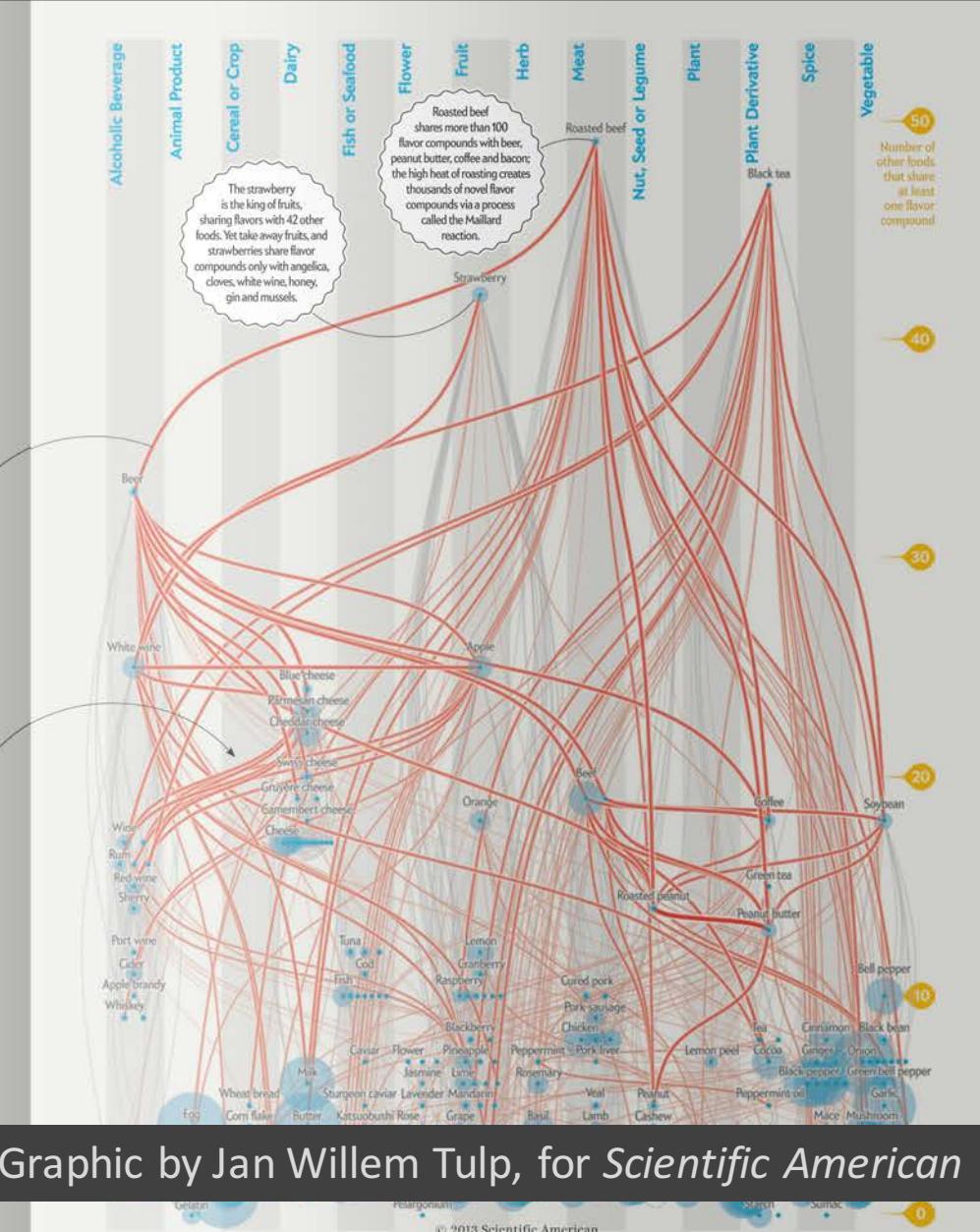
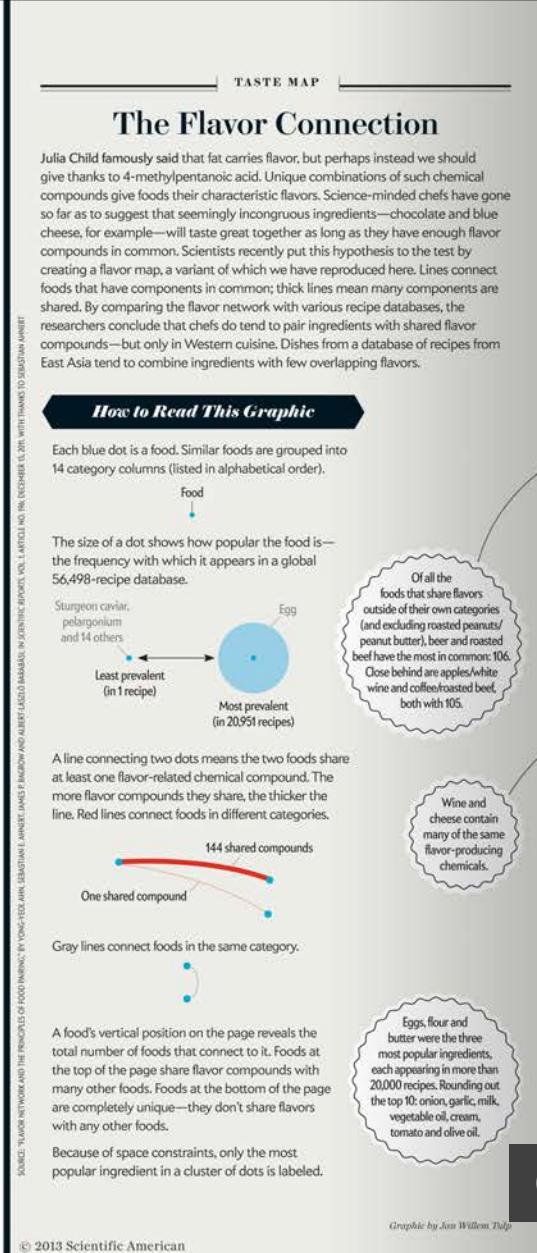
Manuscript received 31 March 2011; accepted 1 August 2011; posted online 23 October 2011; mailed on 14 October 2011.

For information on obtaining reprints of this article, please send email to: tvcg@computer.org.

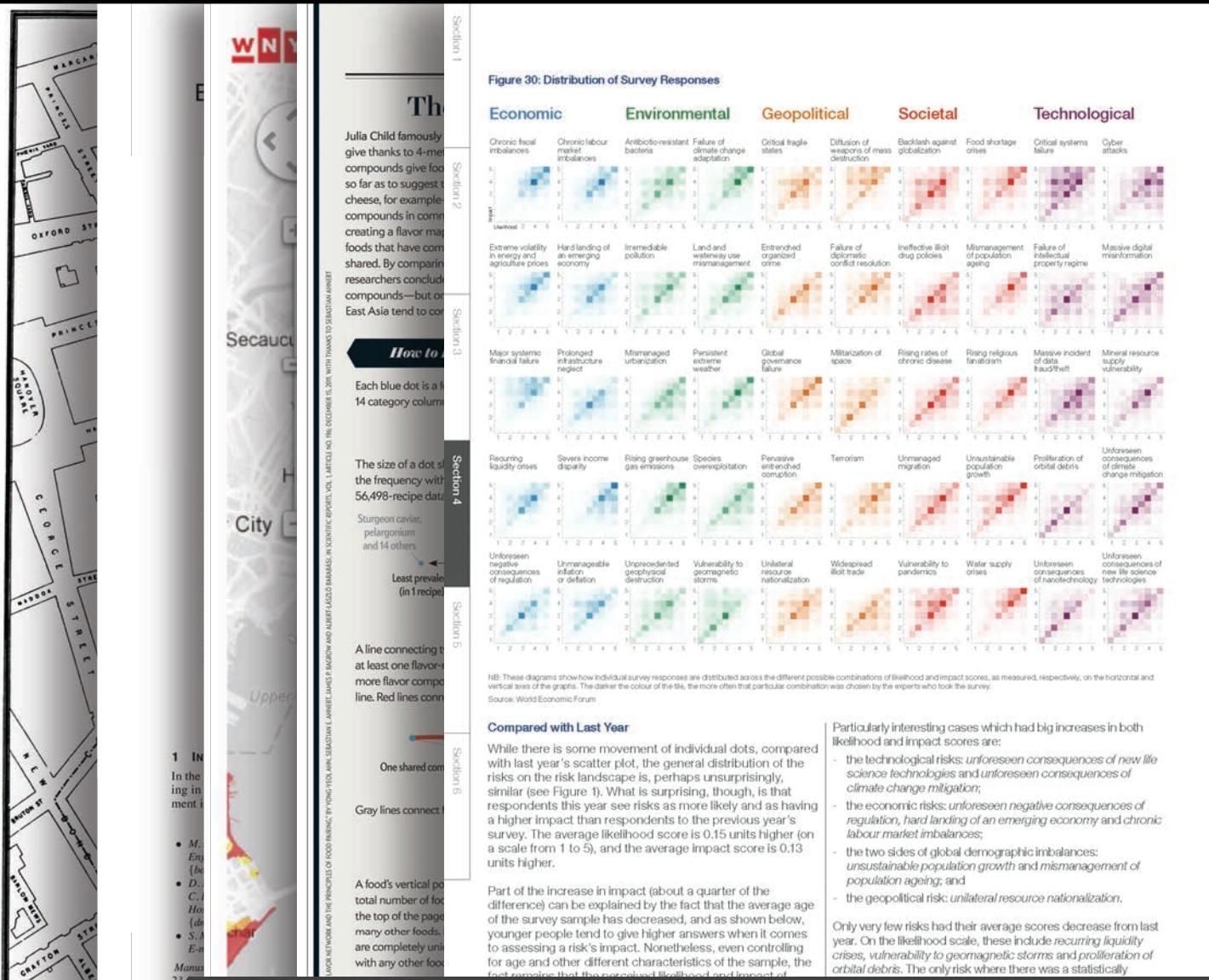
communicate breaking news to public



educate and inspire popular audiences

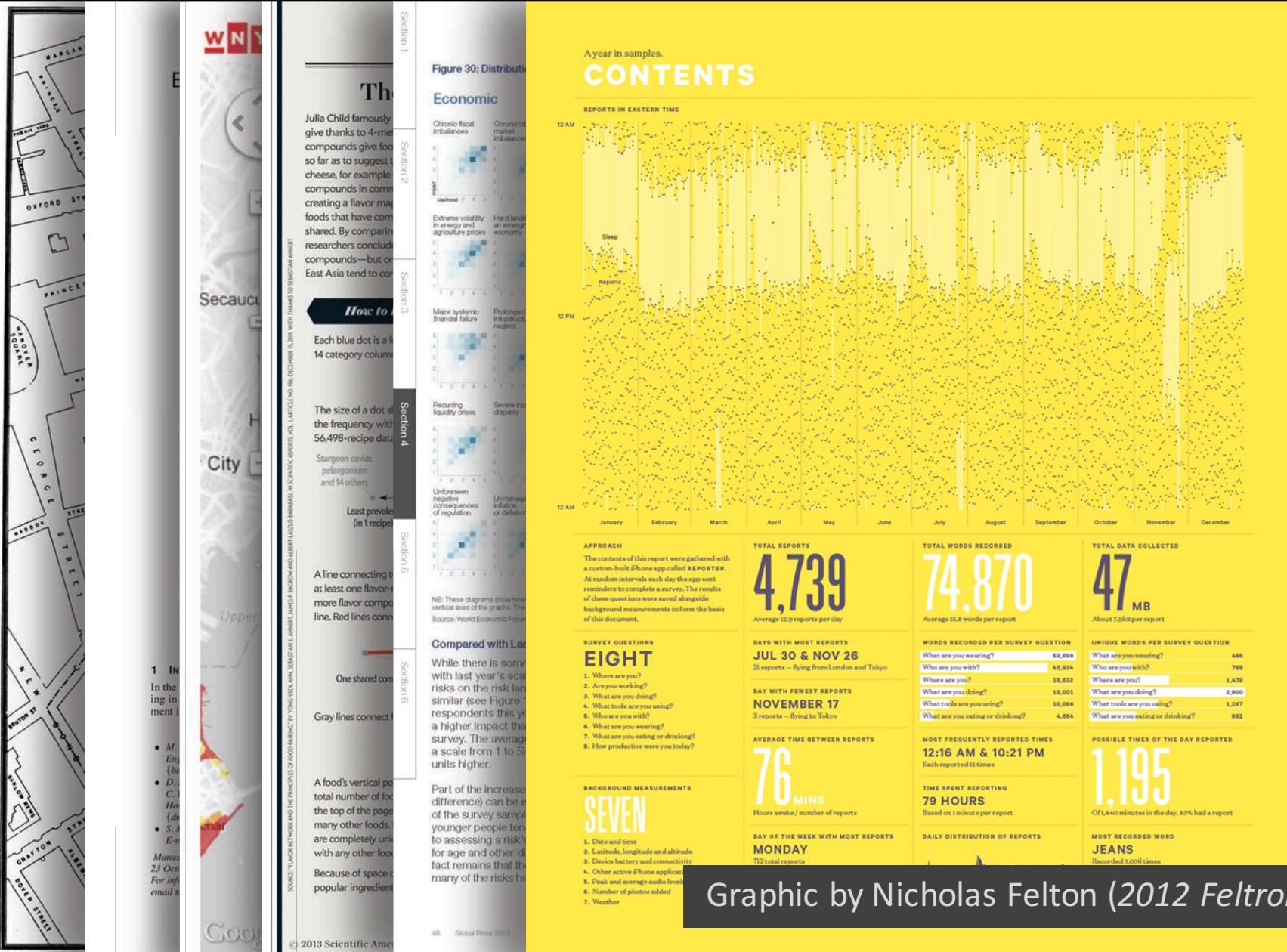


communicate results to stakeholders



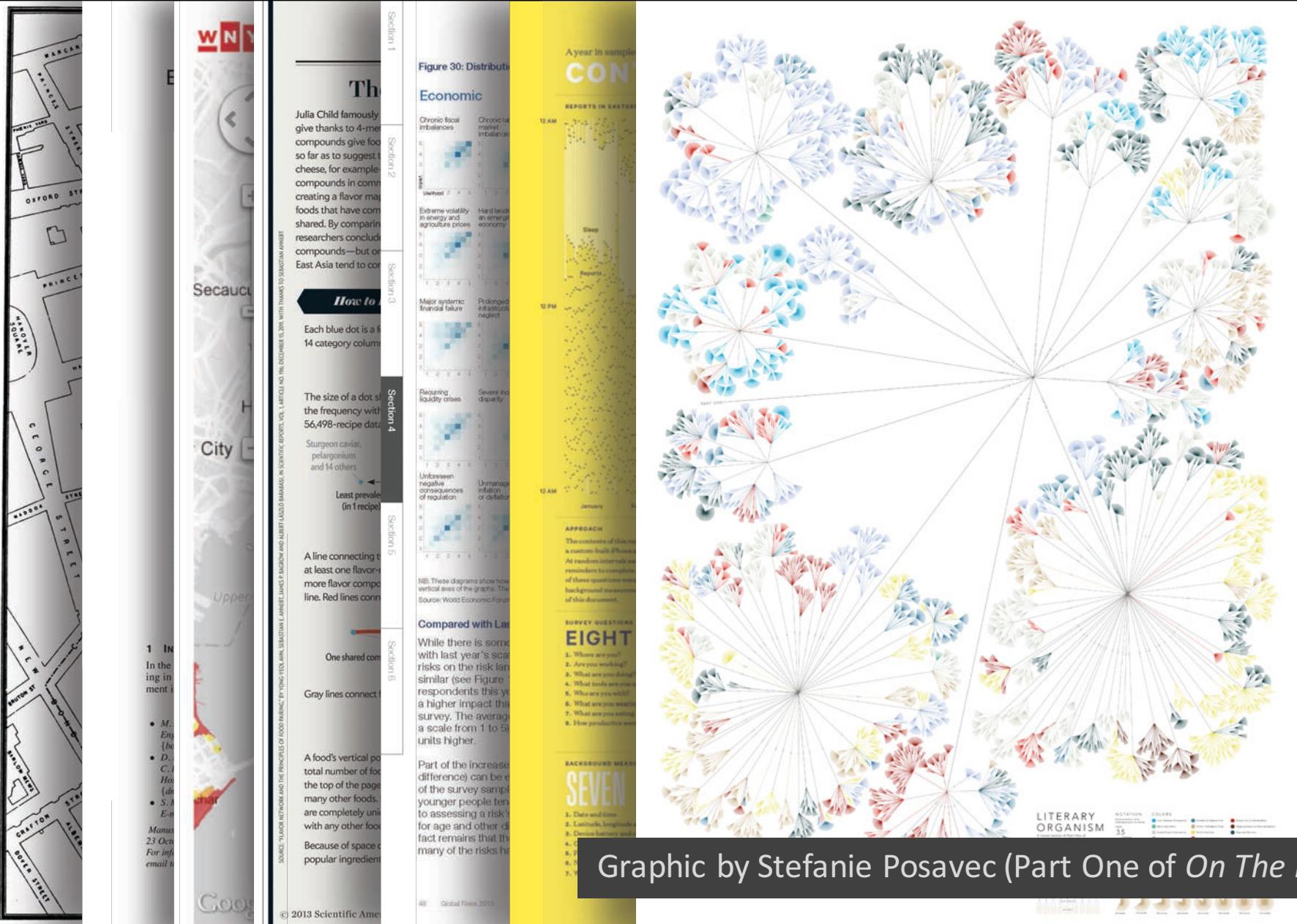
Graphic by Moritz Stefaner, From *World Economic Forum Global Risks Report, 2013*

nontraditional storytelling

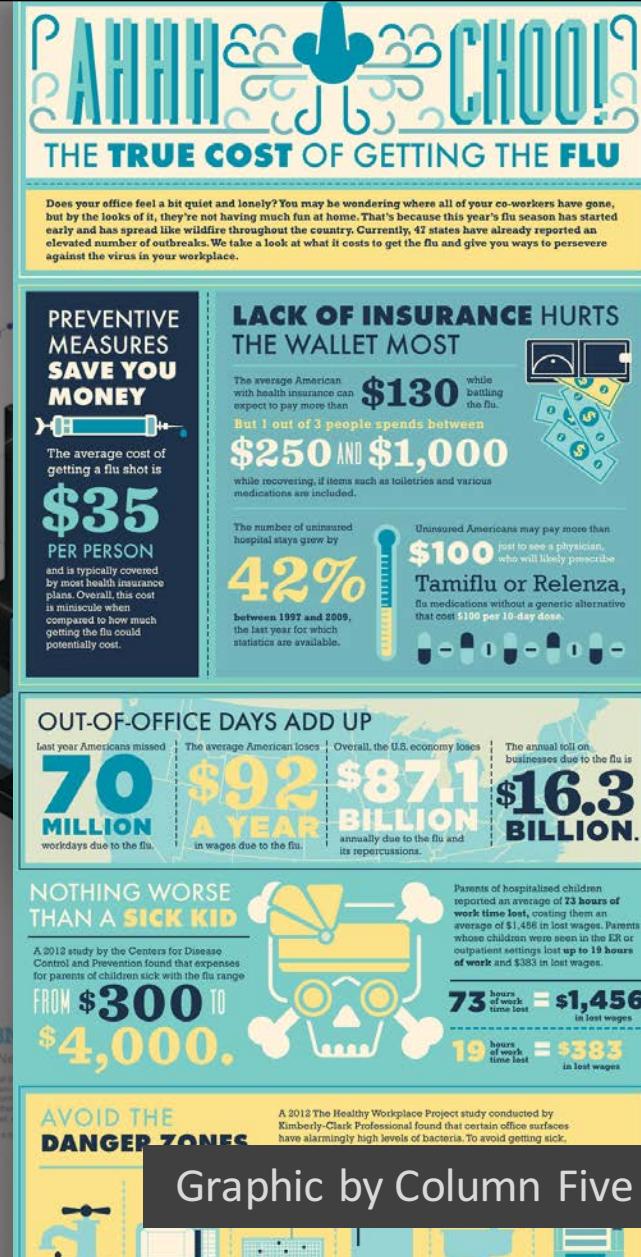
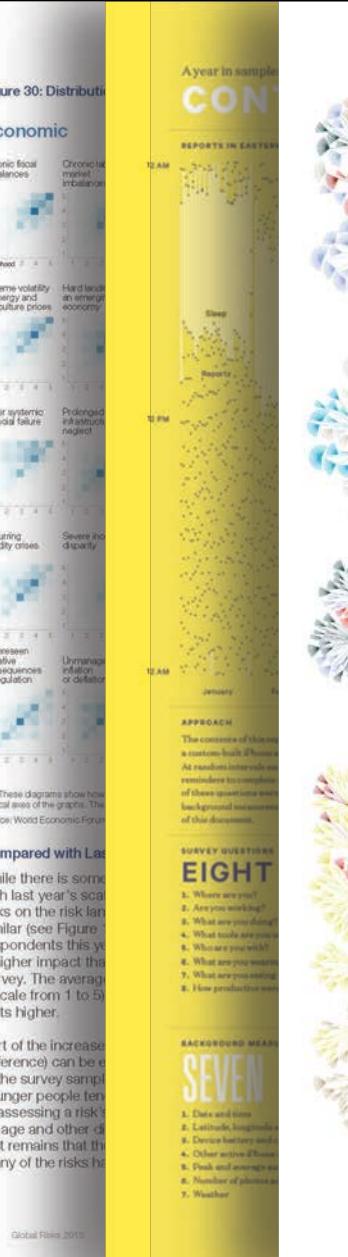
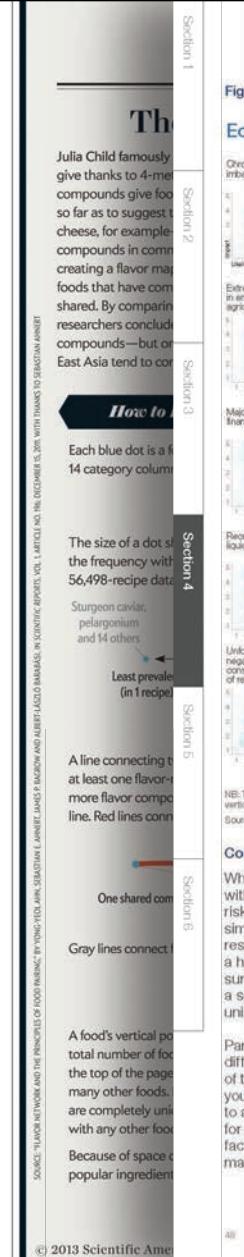


Graphic by Nicholas Felton (2012 Feltron Report)

artistic expression



[infographic]



[content]

Developing scientific data visualizations
for non-specialist audiences

[context]

organizing your [content]

- basic design tips
- choosing the right chart type
- more resources

customizing for a different [context]

- simple edits / reframing
- moderate reworks
- custom solutions



Data Visualization Advice for Scientists

Dataviz engineer Robert Simmon demonstrates how adhering to a few best-practice design principles can make graphics easier to read

By Robert Simmon on February 19, 2016

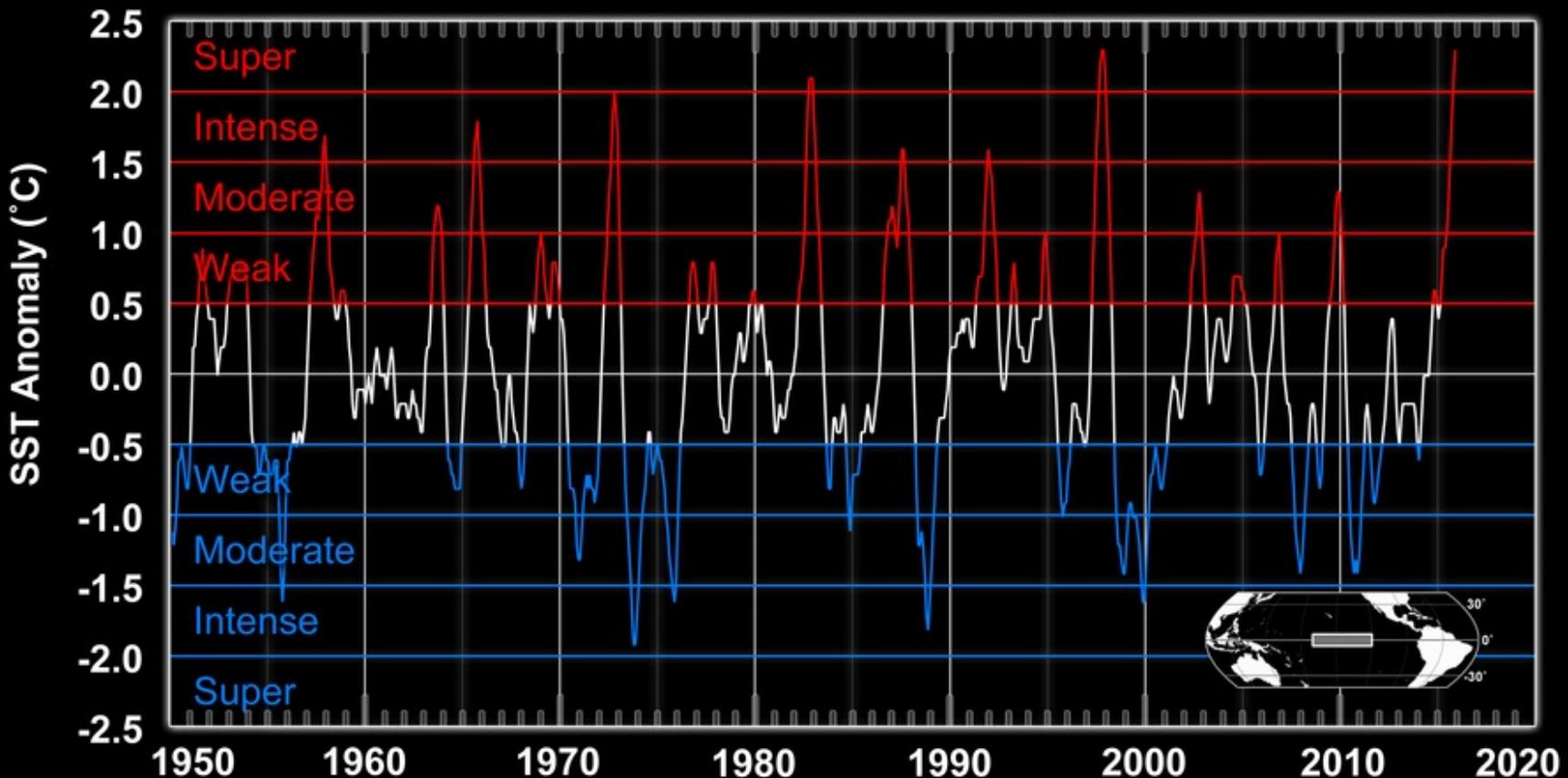
http://bit.ly/LSU_jc1

NOAA Sea Surface Temperature Anomaly (°C)
for Oceanic Niño Region 3.4 (5°S–5°N, 170°W–120°W)

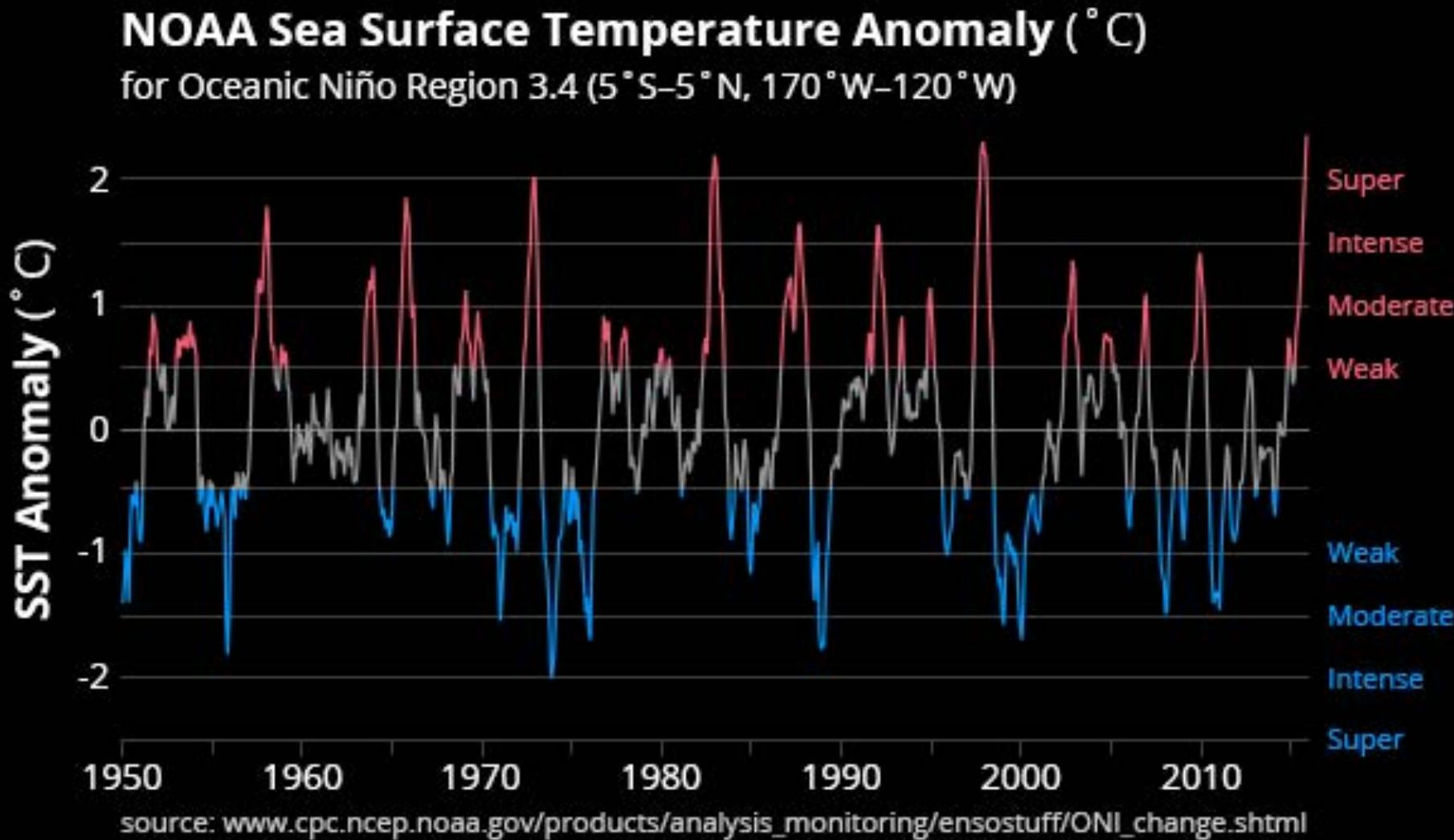
design tips

NOAA Sea Surface Temperature Anomaly ($^{\circ}\text{C}$)

for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



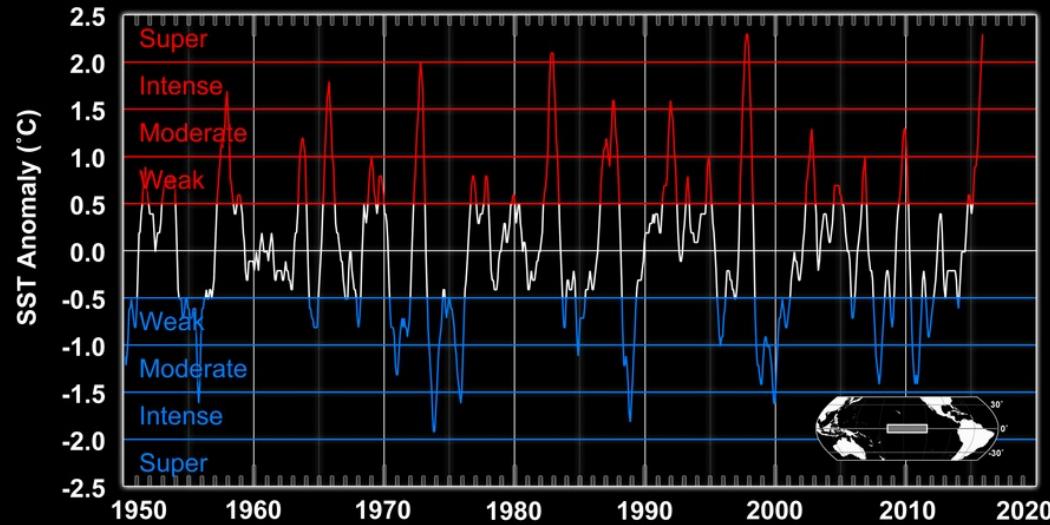
design tips



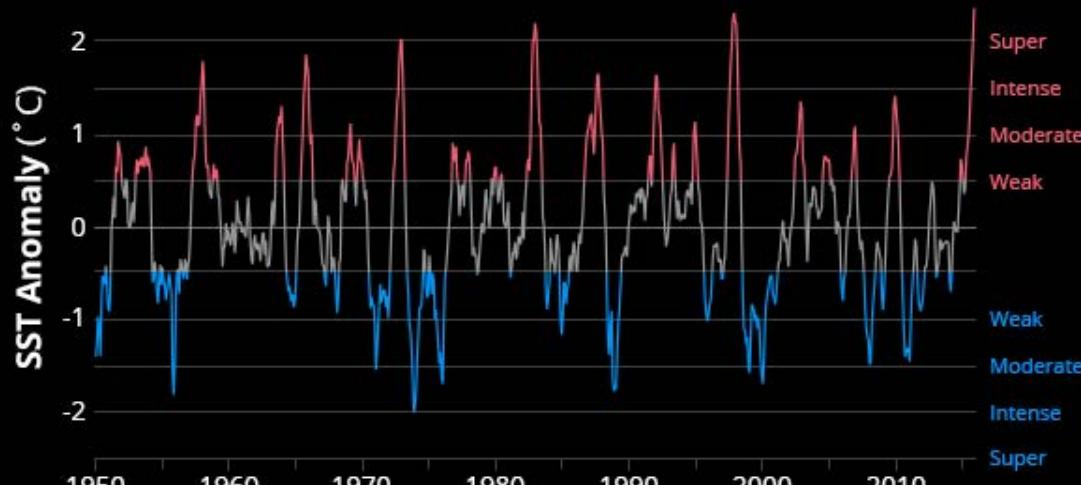
Robert Simmon

design tips

NOAA Sea Surface Temperature Anomaly ($^{\circ}\text{C}$) for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



NOAA Sea Surface Temperature Anomaly ($^{\circ}\text{C}$) for Oceanic Niño Region 3.4 (5°S-5° N, 170°W-120°W)

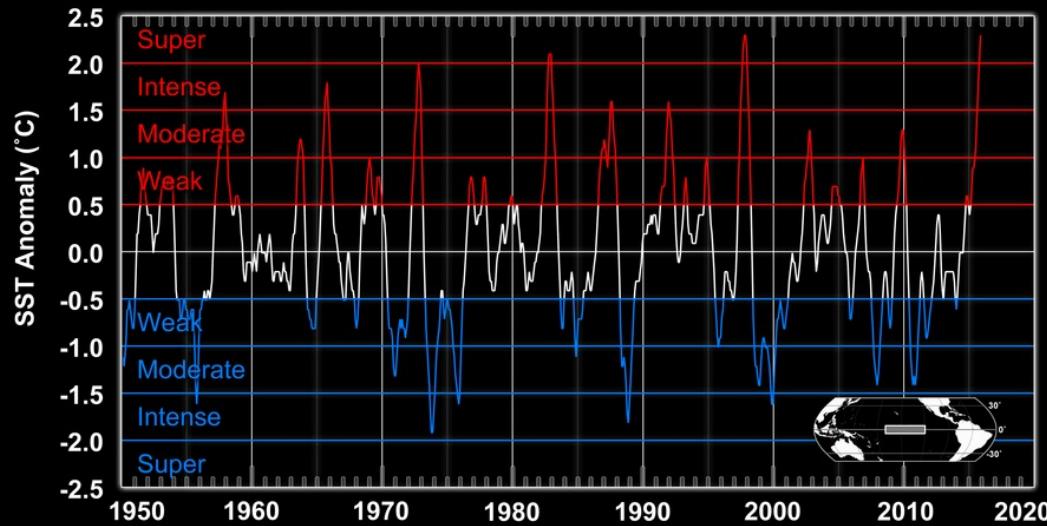


source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

design tips

NOAA Sea Surface Temperature Anomaly (°C)

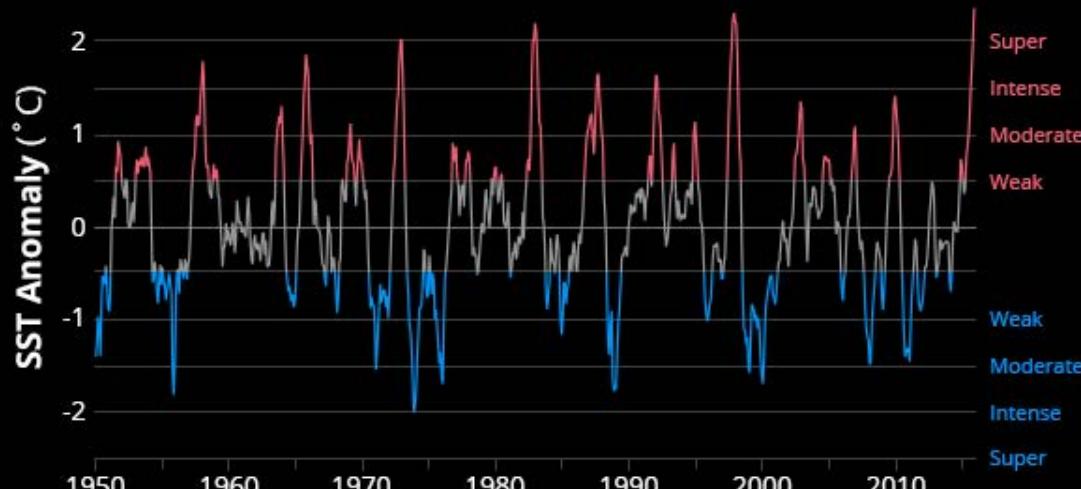
for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



He removed the frame around the graph entirely—it flattens the graphic lessening the difference between foreground and background.

NOAA Sea Surface Temperature Anomaly (°C)

for Oceanic Niño Region 3.4 (5°S-5°N, 170°W-120°W)

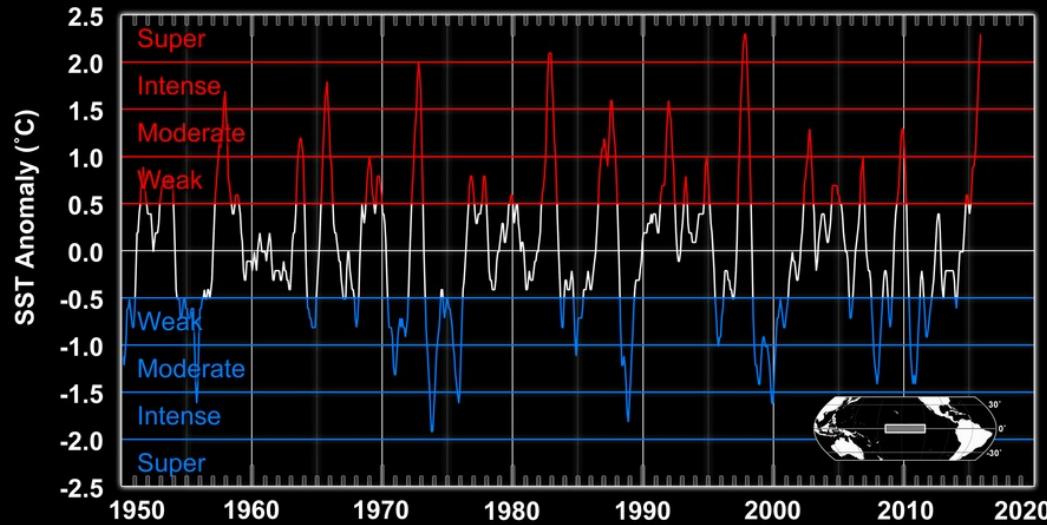


source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

design tips

NOAA Sea Surface Temperature Anomaly (°C)

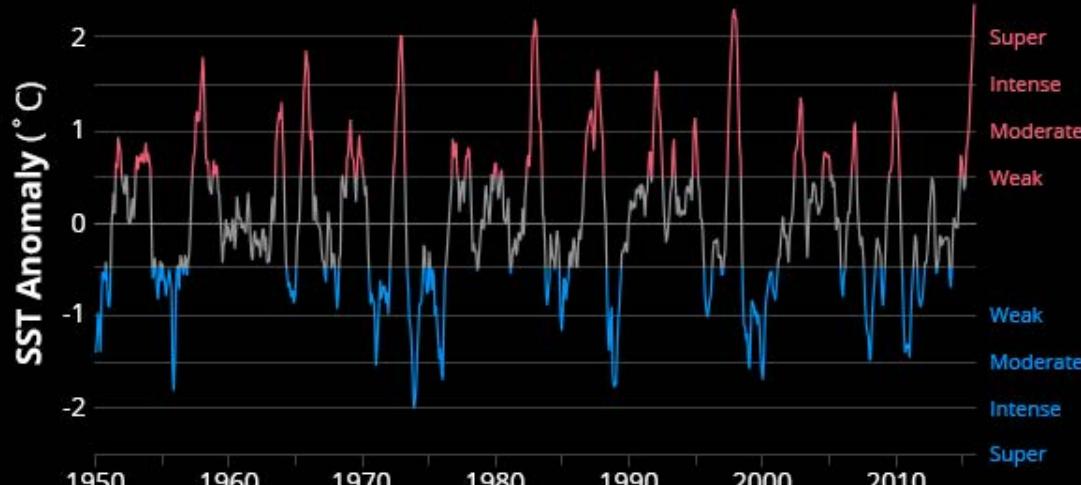
for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



He adjusted the red, neutral, and blue colors so that they'd have equal visual weight.

NOAA Sea Surface Temperature Anomaly (°C)

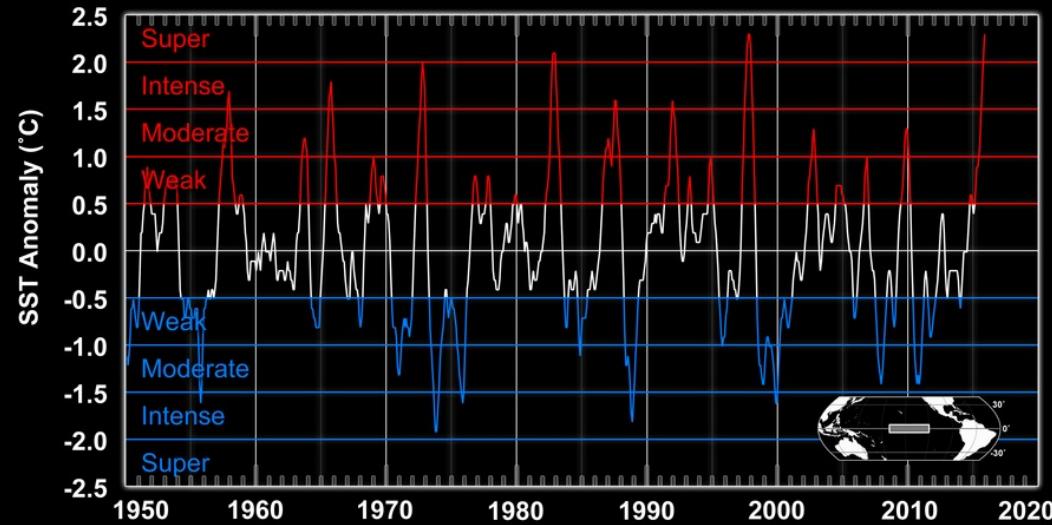
for Oceanic Niño Region 3.4 (5°S-5° N, 170°W-120°W)



source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

design tips

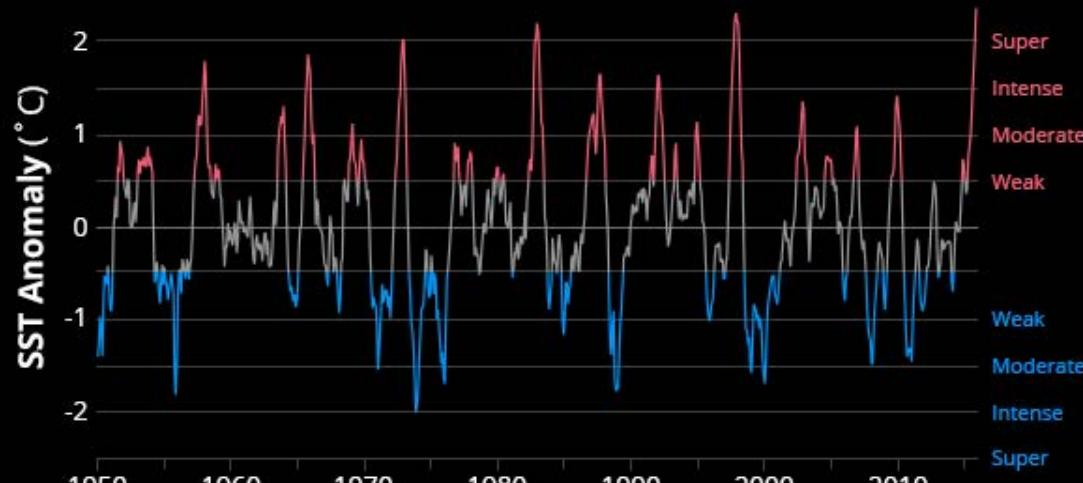
NOAA Sea Surface Temperature Anomaly ($^{\circ}\text{C}$) for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



Moved title to the upper-left corner of the frame: where most western viewers will start reading.



NOAA Sea Surface Temperature Anomaly ($^{\circ}\text{C}$) for Oceanic Niño Region 3.4 (5°S-5° N, 170°W-120°W)

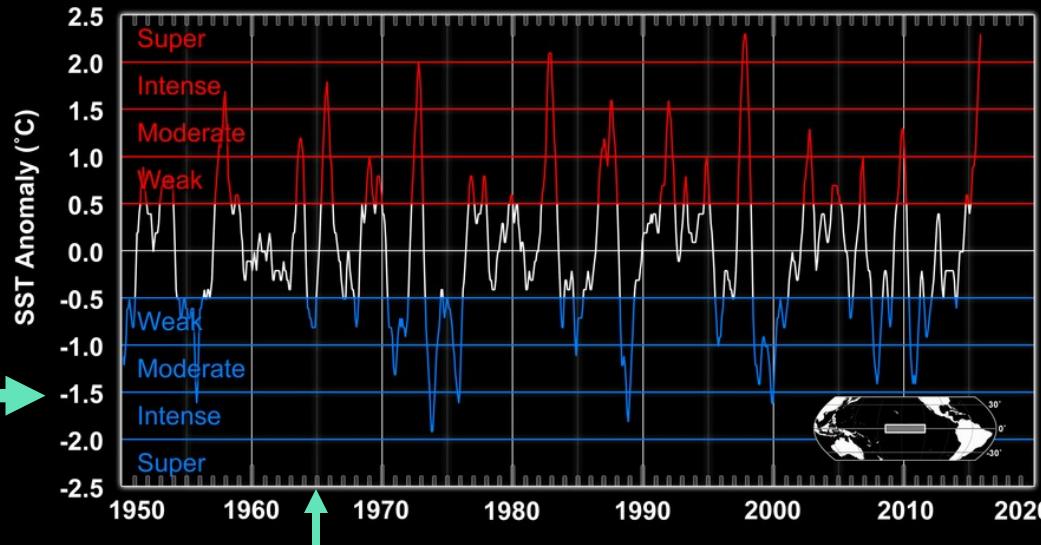


source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

design tips

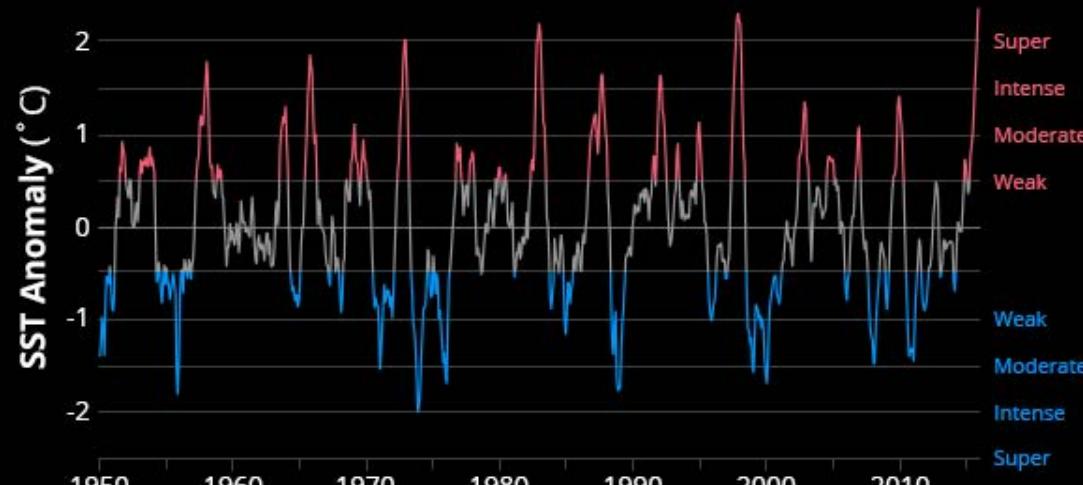
NOAA Sea Surface Temperature Anomaly (°C)

for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



NOAA Sea Surface Temperature Anomaly (°C)

for Oceanic Niño Region 3.4 (5°S-5° N, 170°W-120°W)

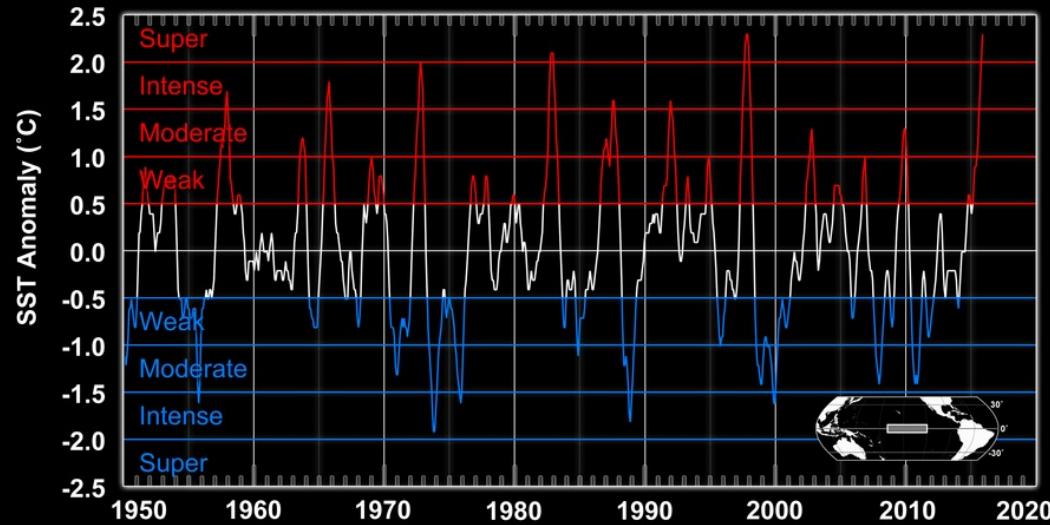


source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

design tips

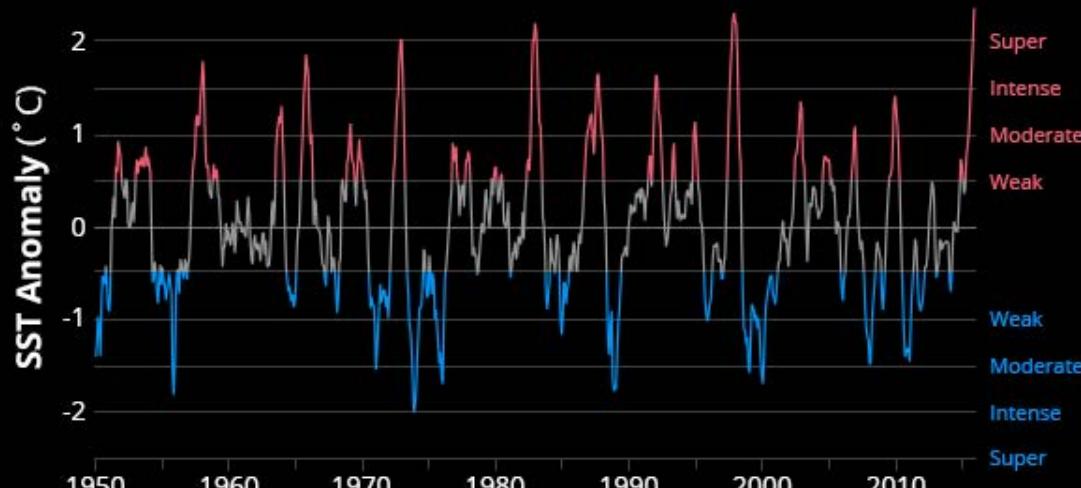
NOAA Sea Surface Temperature Anomaly (°C)

for Oceanic Niño Index Region 3.4 (5°S - 5°N, 170°W - 120°W)



NOAA Sea Surface Temperature Anomaly (°C)

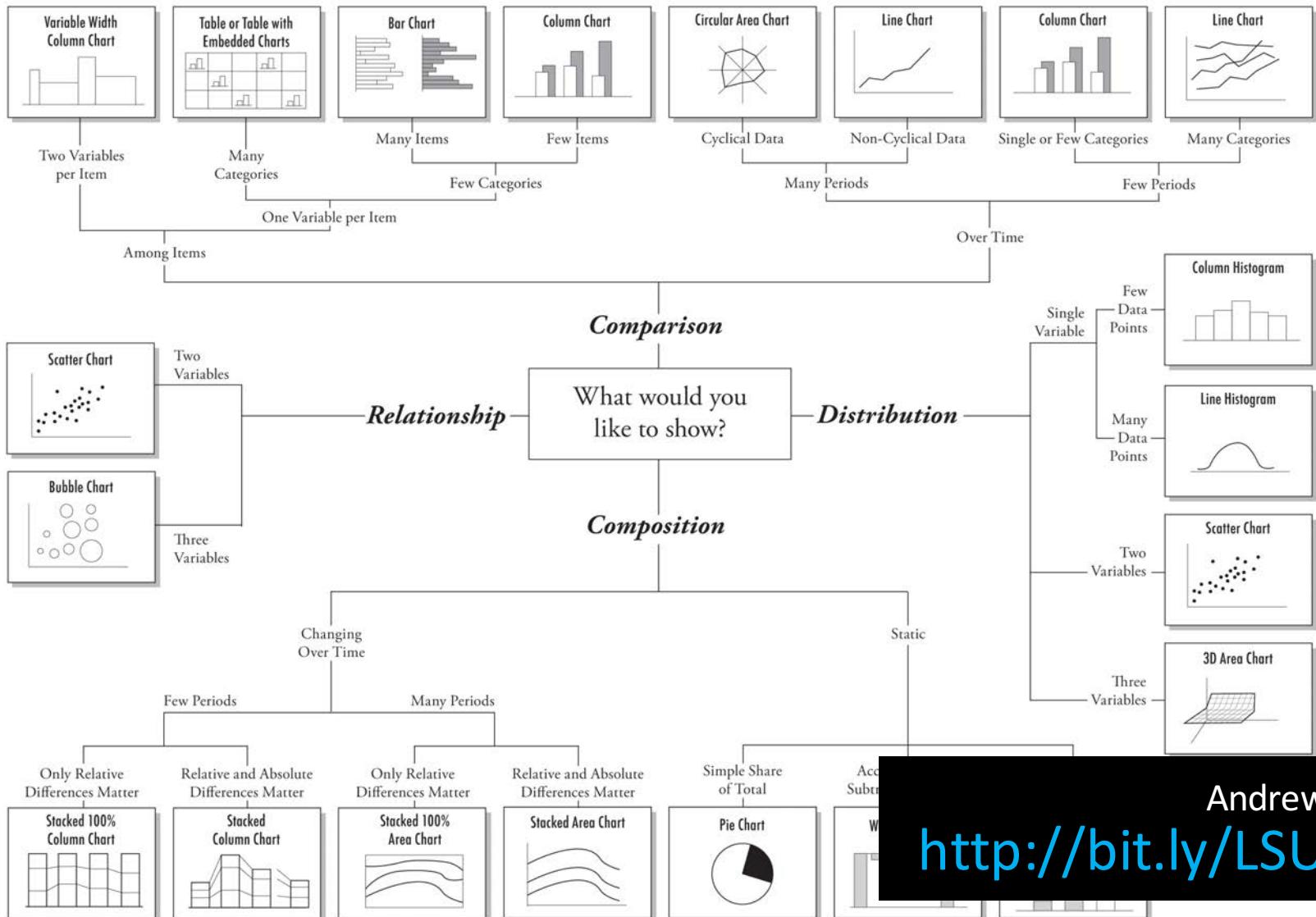
for Oceanic Niño Region 3.4 (5°S-5° N, 170°W-120°W)



source: www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ONI_change.shtml

chart choosers

Chart Suggestions—A Thought-Starter



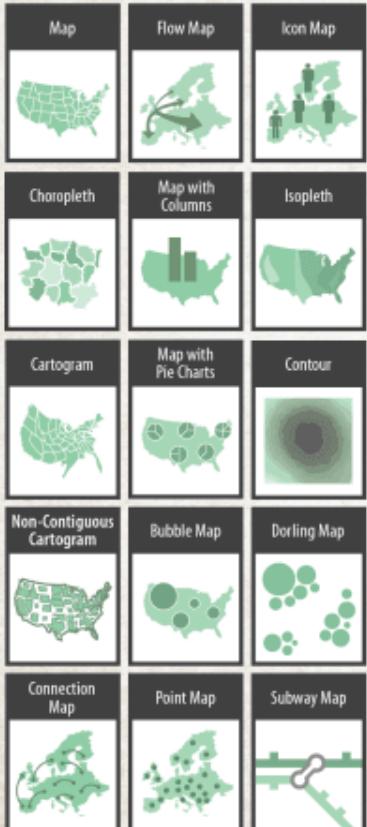
Andrew Abela

http://bit.ly/LSU_jc4

chart choosers

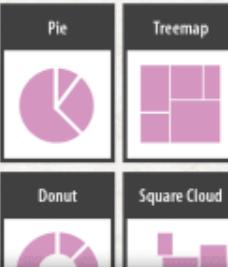
GEOSPATIAL

Relates data to its geography



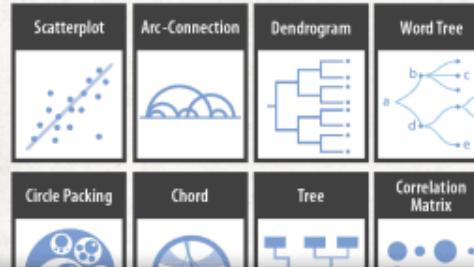
PART-TO-WHOLE

Relates the part of a variable to its total



RELATIONSHIP

Illustrates correlations or relationships between variables



THE GRAPHIC CONTINUUM

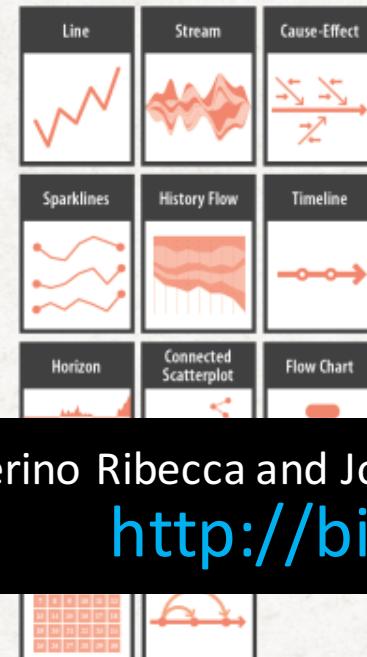
COMPARING CATEGORIES

Compare values across categories



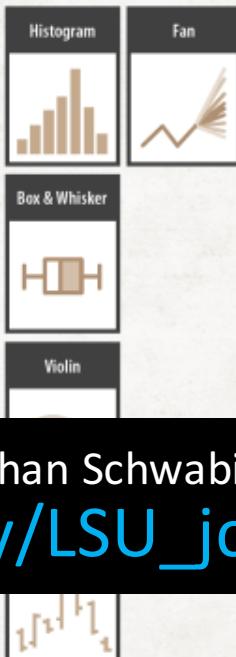
TIME

Track changes over time



DISTRIBUTION

Representation of the distribution of data



Severino Ribecca and Jonathan Schwabish
http://bit.ly/LSU_jc5

chart choosers

Search by Function

View by List



Arc Diagram



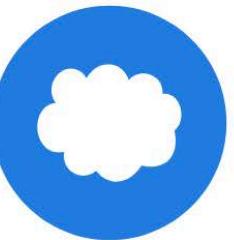
Area Graph



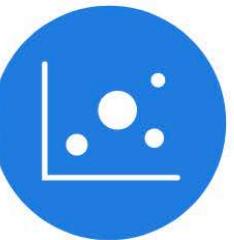
Bar Chart



Box & Whisker Plot



Brainstorm



Bubble Chart



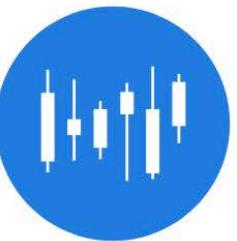
Bubble Map



Bullet Graph



Calendar



Candlestick Chart



Chord Diagram



Choropleth Map



Circle Packing



Connection Map



Density Plot



Dot Plot



Severino Ribeca
http://bit.ly/LSU_jc6

chart choosers

Chart Chooser

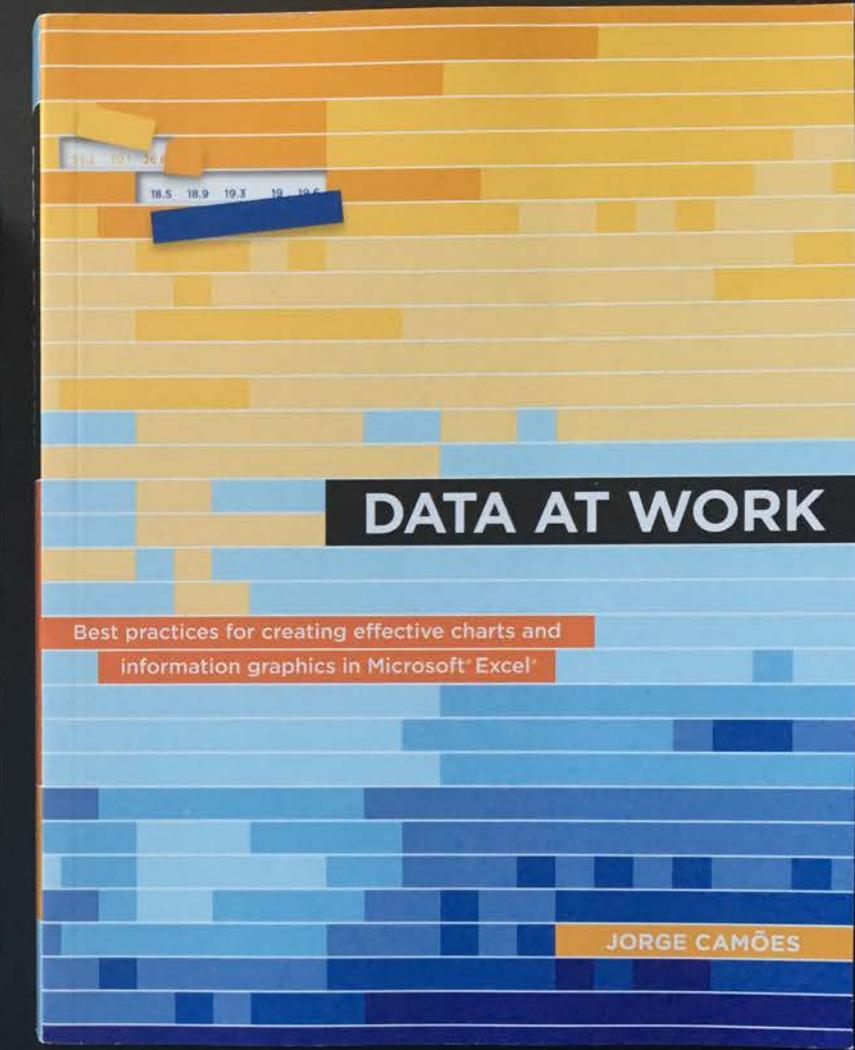
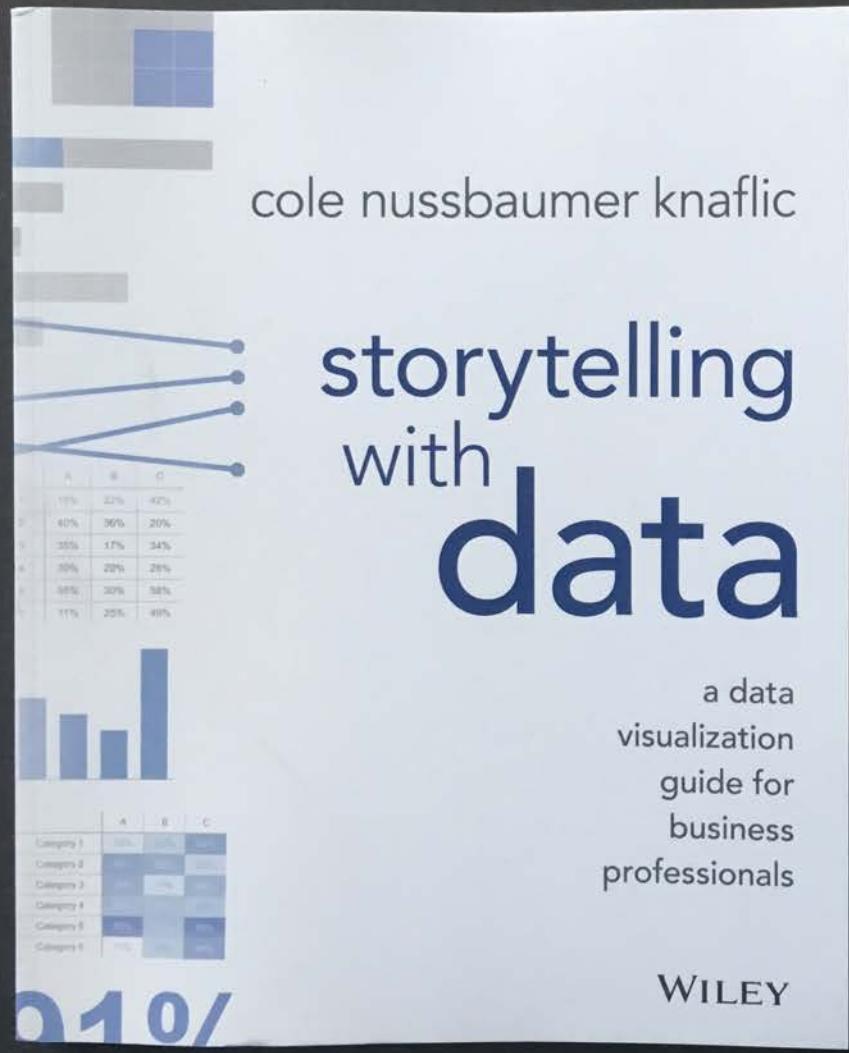


Effective Data Visualization

The Right Chart for the Right Data



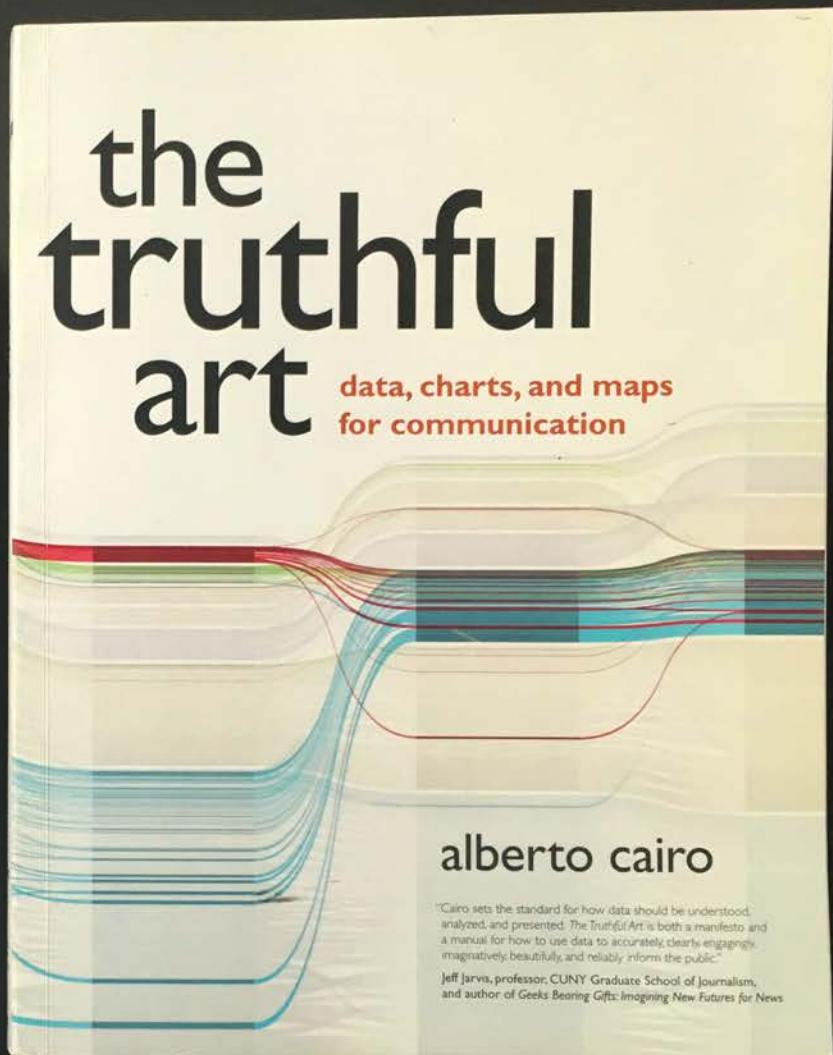
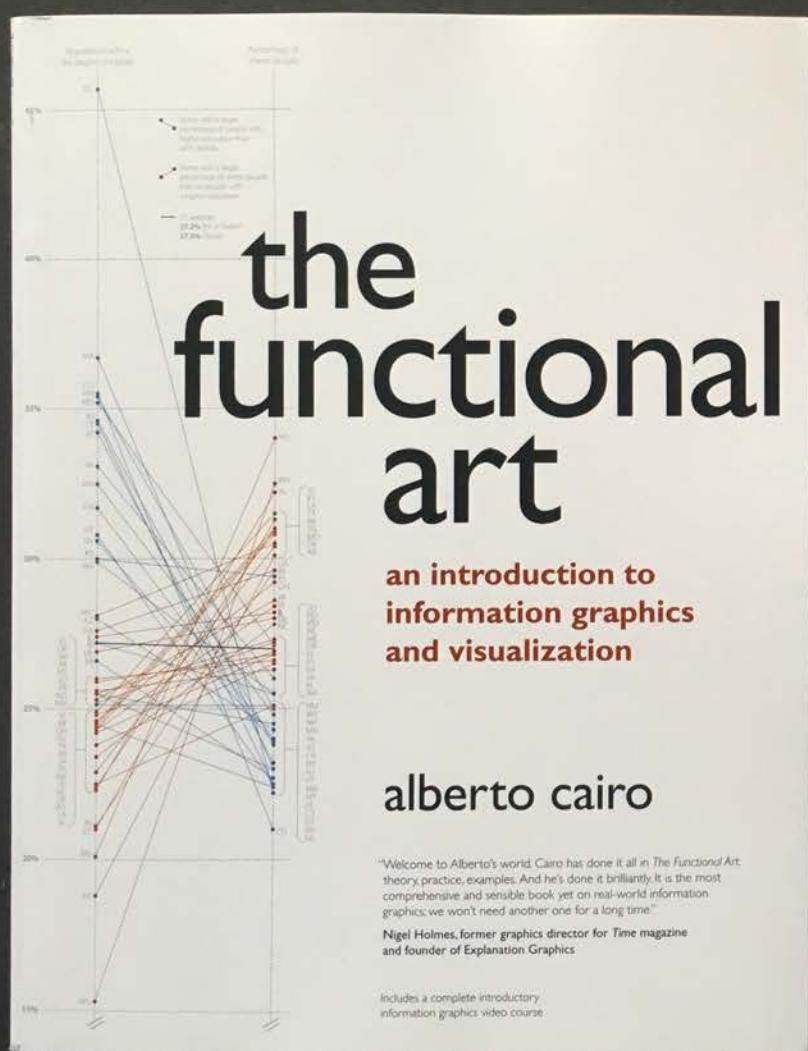
more resources



Jorge Camões http://bit.ly/LSU_jc9

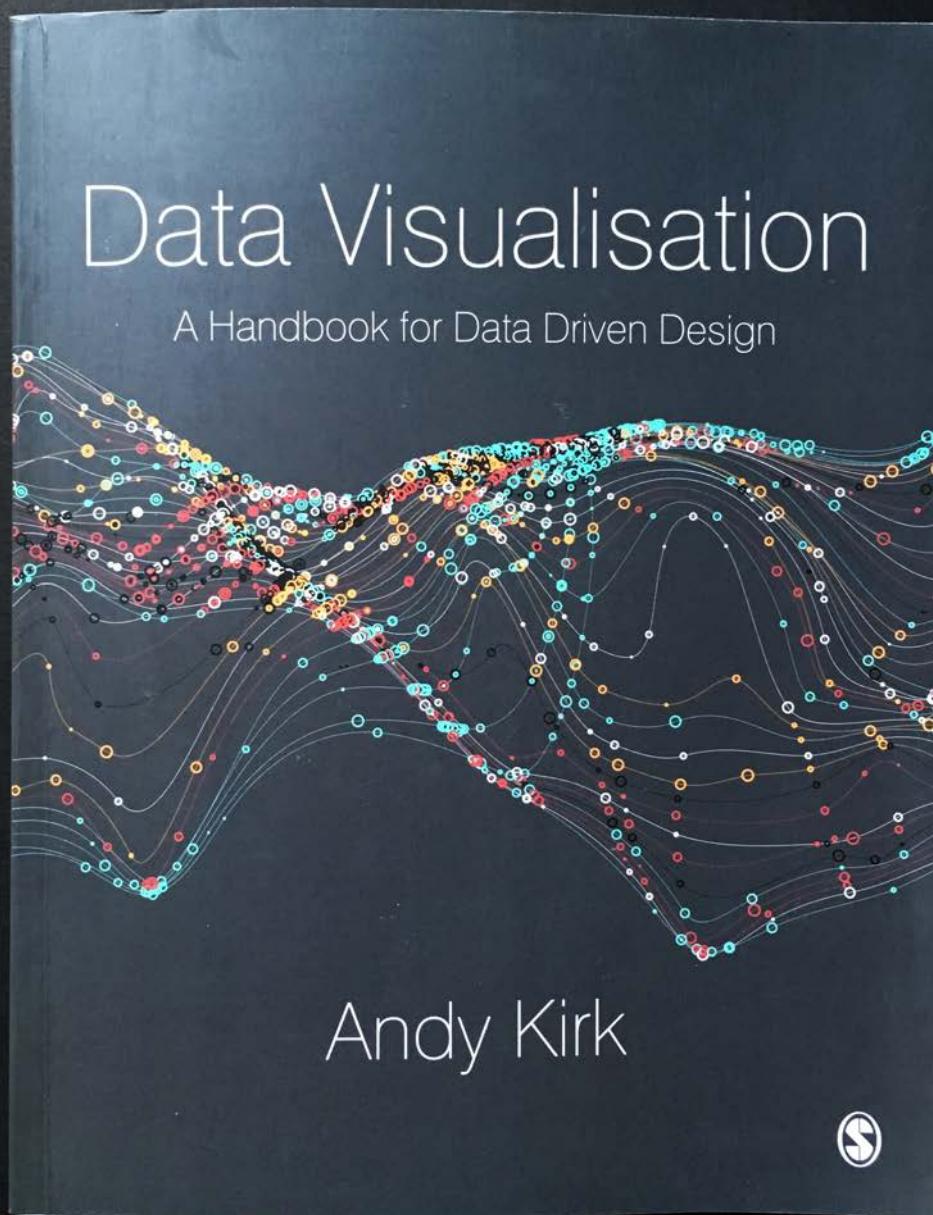
Cole Nussbaumer Knaflic http://bit.ly/LSU_jc8

more resources



Alberto Cairo
http://bit.ly/LSU_jc10

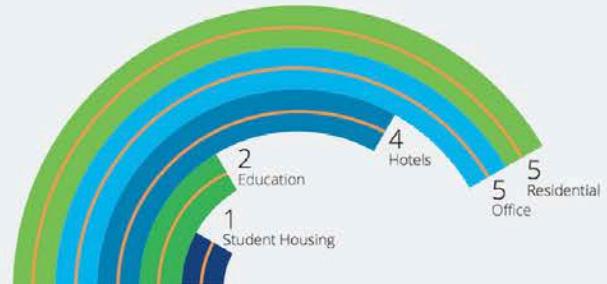
more resources



01 SEP THE PROBLEMS WITH B'ARC CHARTS

ORIGINAL B'ARC CHART

Which sectors are the most active? (Number of schemes under construction)



Almost one month ago I tweeted my contempt for radial bar charts, having come across a particularly egregious case. It seemed to gain a bit of traction on social media but was also met by genuine queries as to my reasoning for this disdain.



23 AUG THE LITTLE OF VISUALISATION DESIGN: PART 42



22 AUG ANOTHER UPDATE ABOUT THE CHARTMAKER DIRECTORY



Andy Kirk
http://bit.ly/LSU_jc11

customizing for a different [context]

- simple edits / reframing
 - shake out label jargon
 - add framing text (focus attention)
 - add annotations
- moderate reworks
 - add audience-specific context
 - combine or divide elements
- custom solutions
 - create a new visualization from scratch

customizing for a different [context]

- simple edits / reframing
 - shake out label jargon
 - add framing text (focus attention)
 - add annotations

reframing

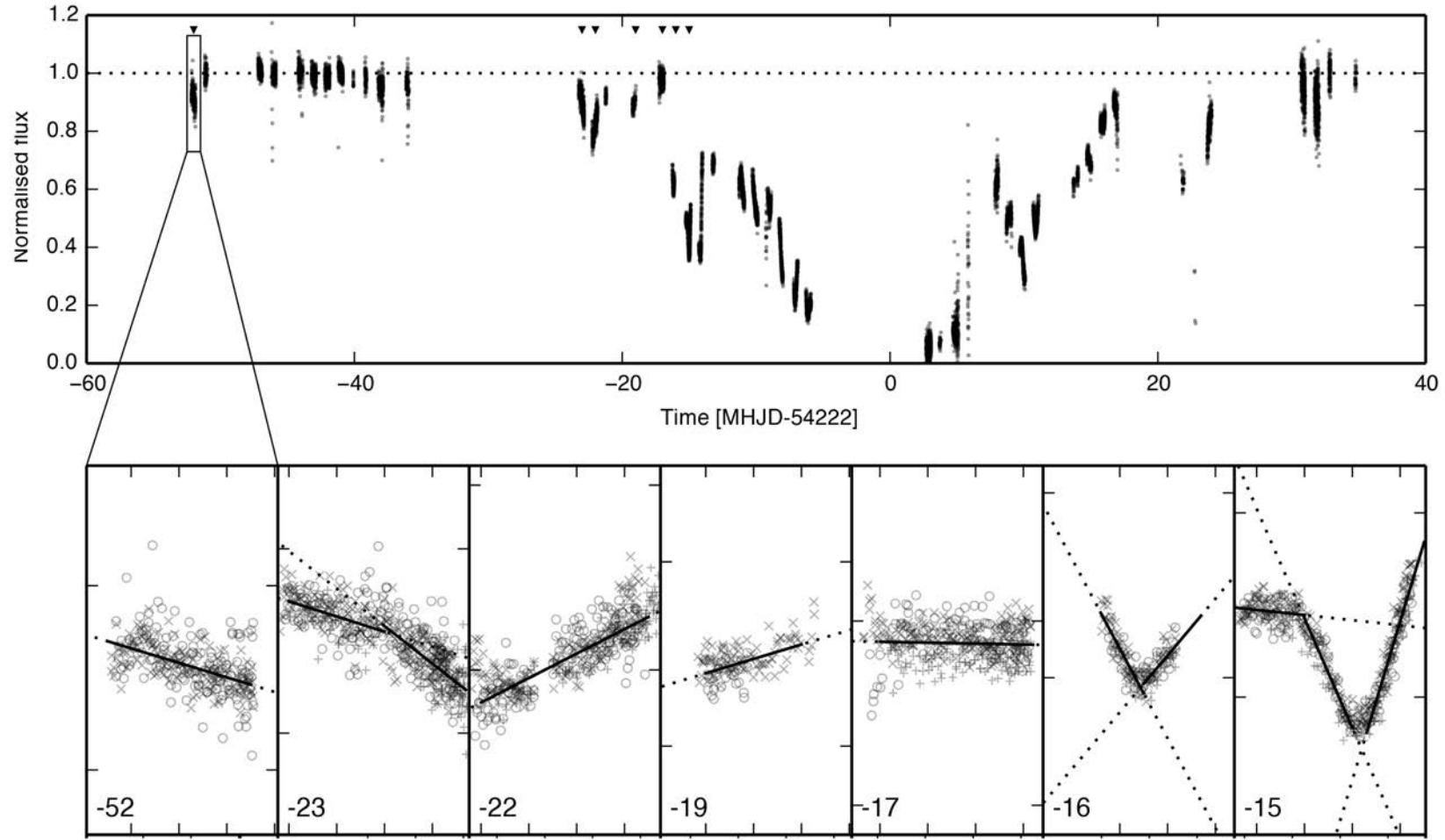


Figure 8. Detailed plot of the J1407 light curve during the eclipse. Top: overview of the light curve. Bottom: nightly light curve for 7 nights during the eclipse, indicated in the upper panel by triangles. The data from the different CCDs are plotted with different symbols, 221: \circ , 227: \times , and 228: $+$. The straight line fits are plotted over the data, solid lines are inside the window used for the fit, dotted line are extrapolated. The y-ticks are spaced at 0.2 au (as in the top panel) and x-ticks at 0.1 d.

By T. I. M. van Werkhoven *et al.*, "Analysis of 1SWASP J140747.93–394542.6 eclipse fine-structure: hints of exomoons"

Perplexing Pattern

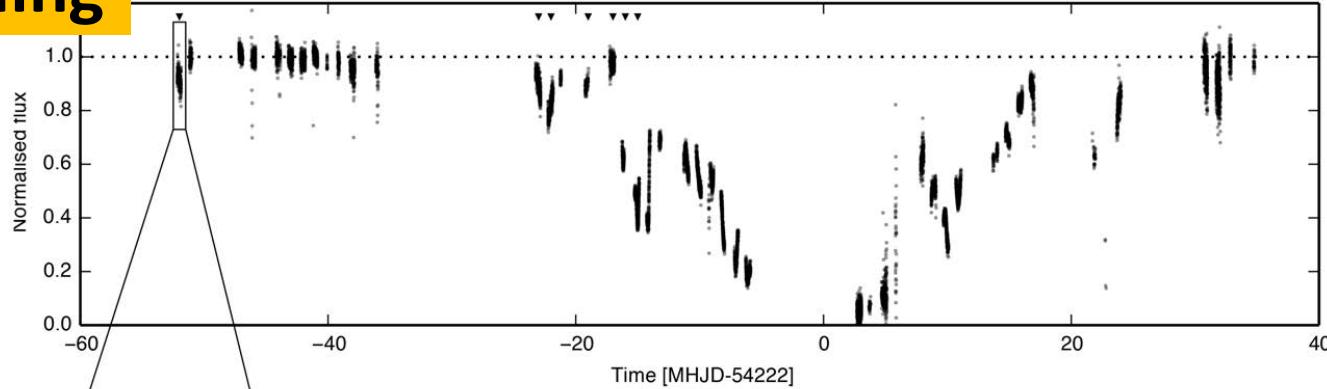
A light curve—an object's variance in brightness plotted over time—is a basic tool for studying stars. A brief boost in brightness can be caused by stellar flares, whereas momentary dips can signal star spots or the shadow of an orbiting planet. But the wildly fluctuating light curve of the star J1407 in 2007 (below) was unlike anything astronomers had ever seen. Something strange was making the star flicker and fade for months at a time.

Between two periods of flickering, J1407's light dimmed for 56 days, suggesting the star was eclipsed by an object 180 million kilometers wide.

J1407'S ODD LIGHT CURVE, DURING A 2007 ECLIPSE



reframing



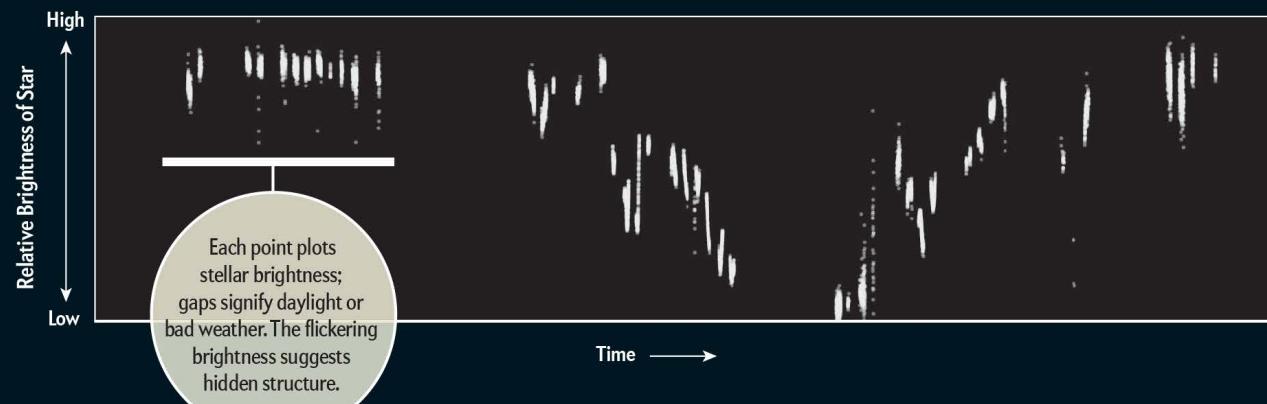
A MYSTERY OF LIGHT AND SHADOW

Perplexing Pattern

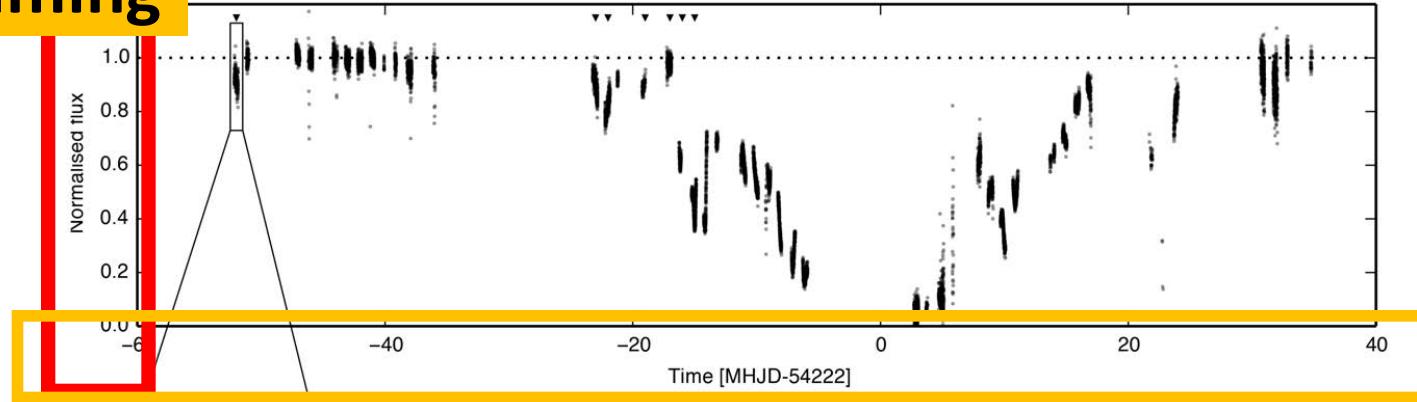
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J1407'S ODD LIGHT CURVE, DURING A 2007 ECLIPSE



reframing



A MYSTERY OF LIGHT AND SHADOW

Perplexing Pattern

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J1407'S ODD LIGHT CURVE, DURING A 2007 ECLIPSE



reframing

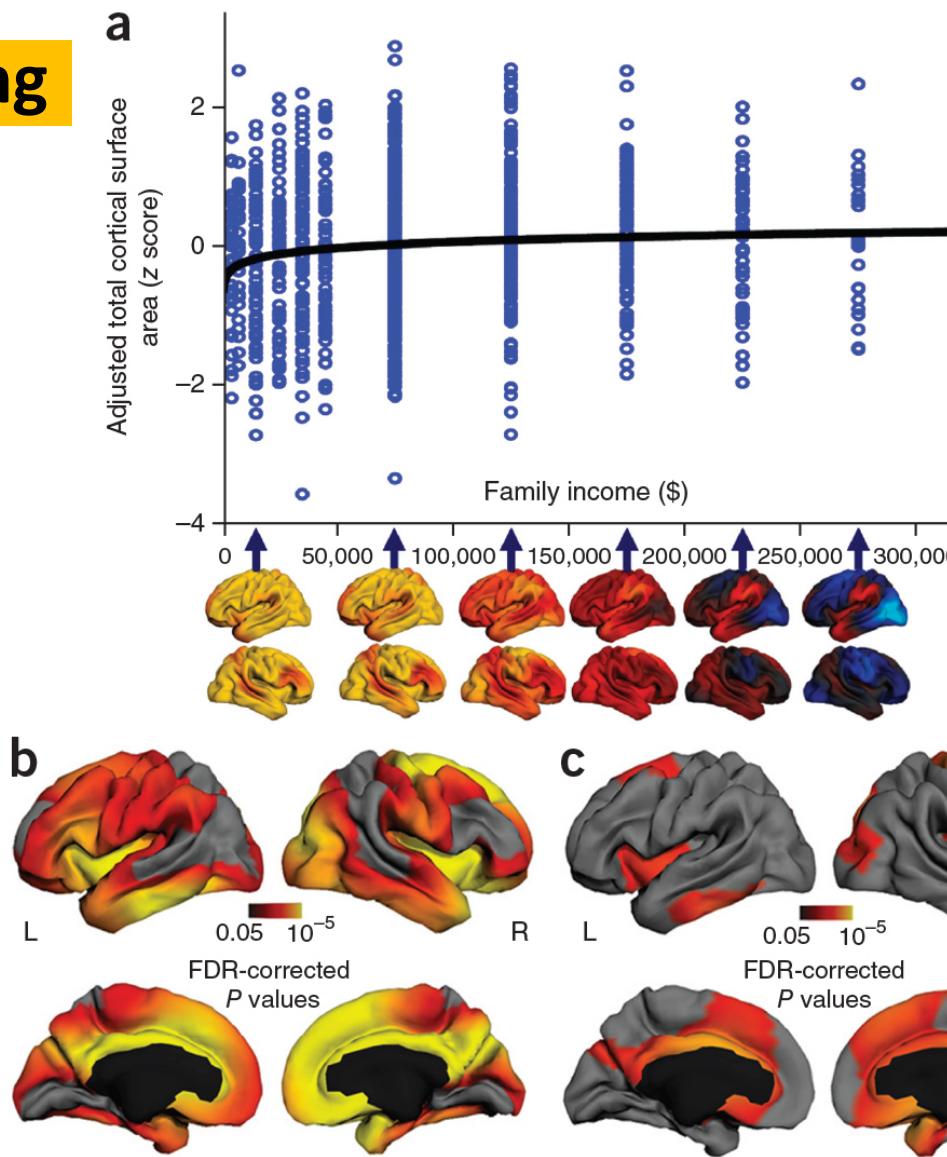
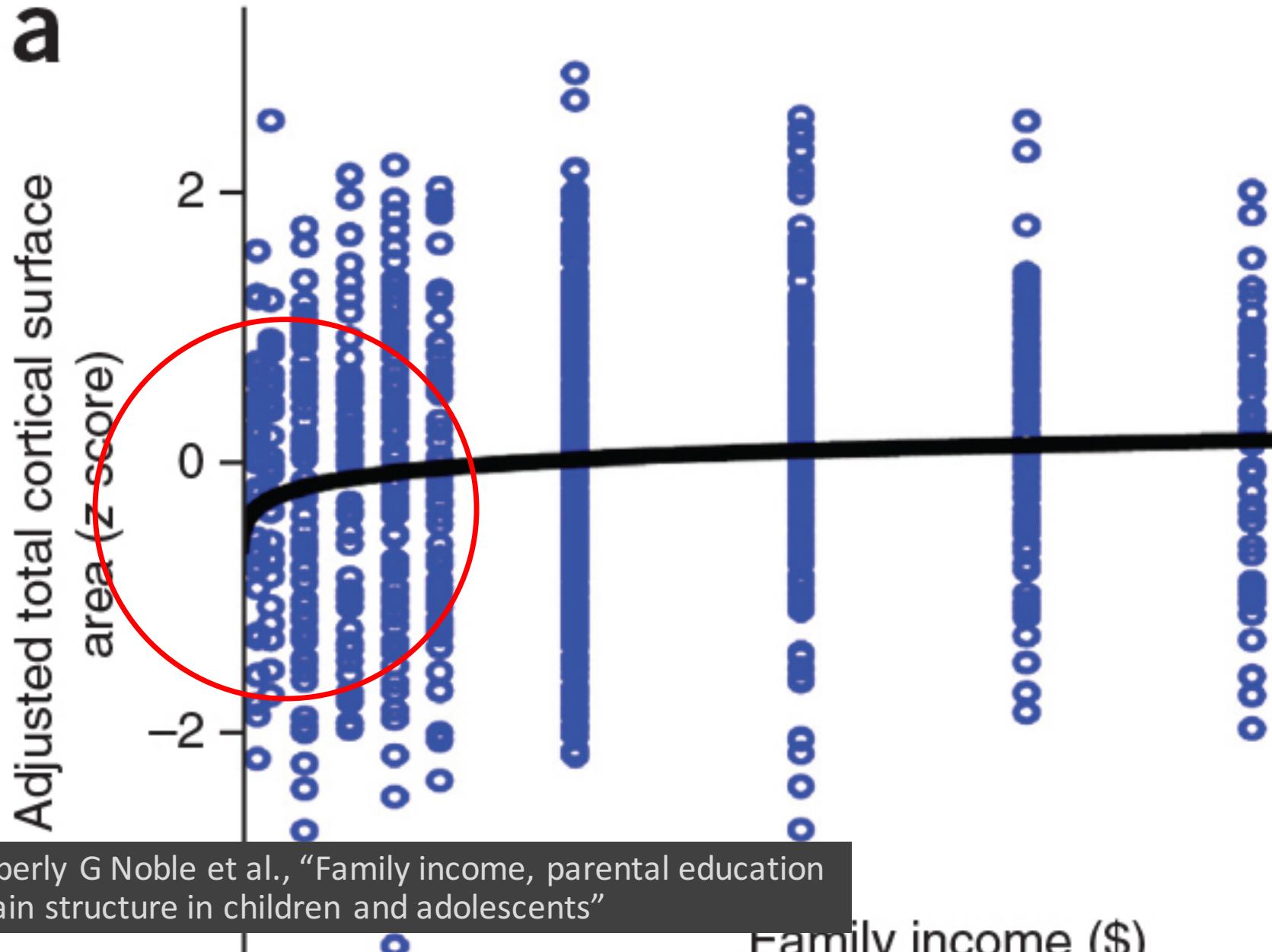


Figure 2 Family income is logarithmically associated with brain structure in children and adolescents ($N = 1,099$). (a) Multiple regression analysis showed that family income was logarithmically associated with brain structure, such that the steepest gradient was at the lowest income levels of the spectrum ($\beta = -0.19$, $P = 0.001$) on the untransformed scale, fitted with a curve. (b) Brain maps of the effect of family income on brain structure, with the brain maps, where the color scale indicates the change in cortical surface area per unit income. (c) When adjusting for age, age 2 , scanner, sex, and genetic variants, In (family income) was significantly associated with widespread regions of children's brains. Relationships were strongest in the left frontal gyrus, and in the right occipital lobe. After adjusting for age, age 2 , scanner, sex, and genetic variants, In (family income) was significantly associated with a smaller number of regions including bilateral regions in the inferior temporal regions, and in the left parahippocampal gyrus. Maps are thresholded at $P < 0.05$ and corrected for multiple comparisons at thresholds of 0.01 and 0.001 and above.

By Kimberly G Noble et al., "Family income, parental education and brain structure in children and adolescents"

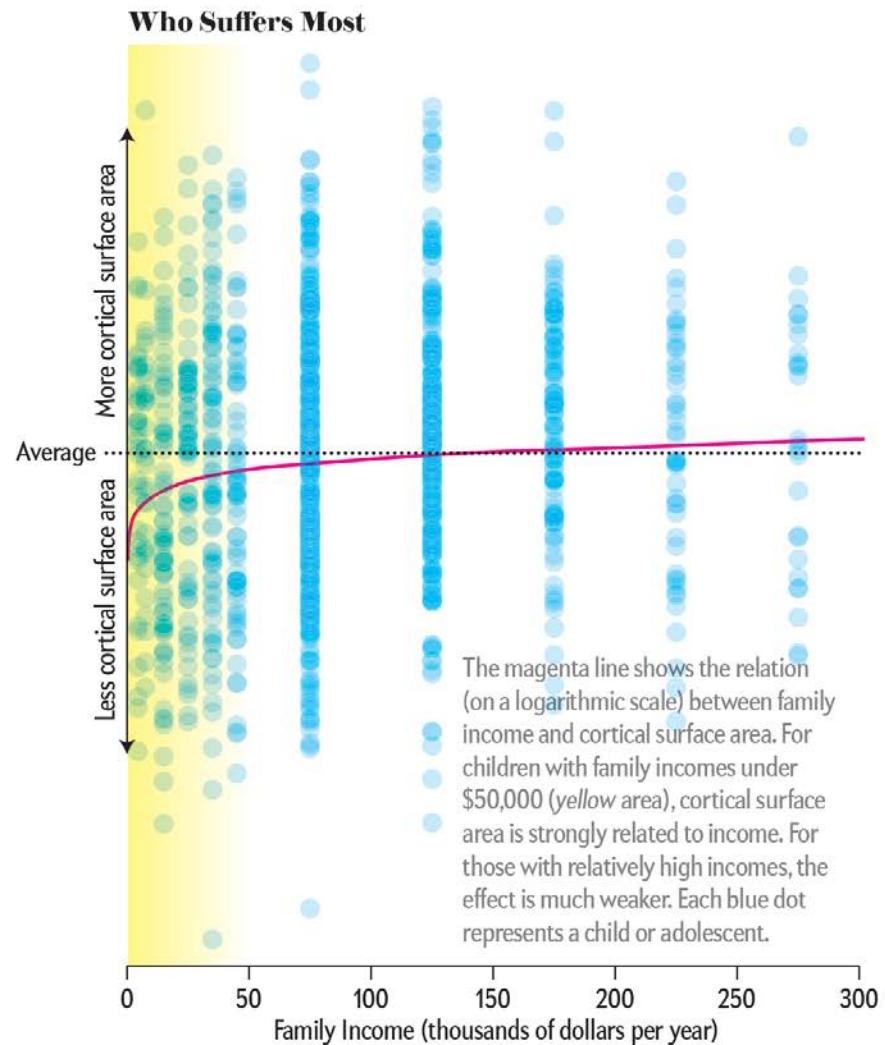
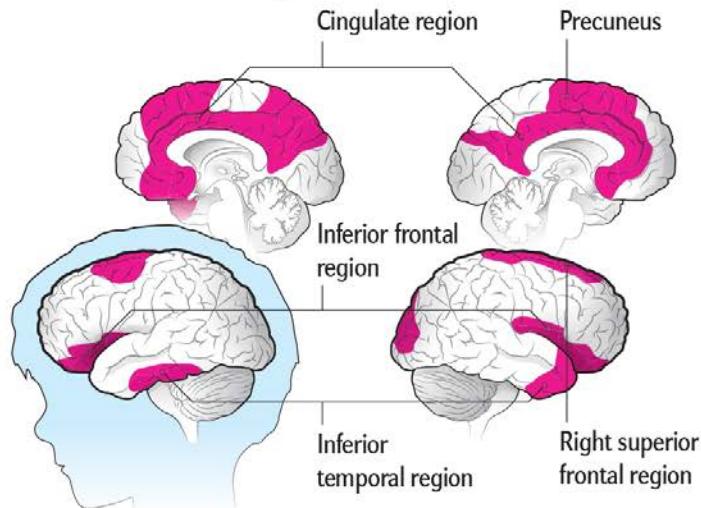
reframing



A Brain on Poverty

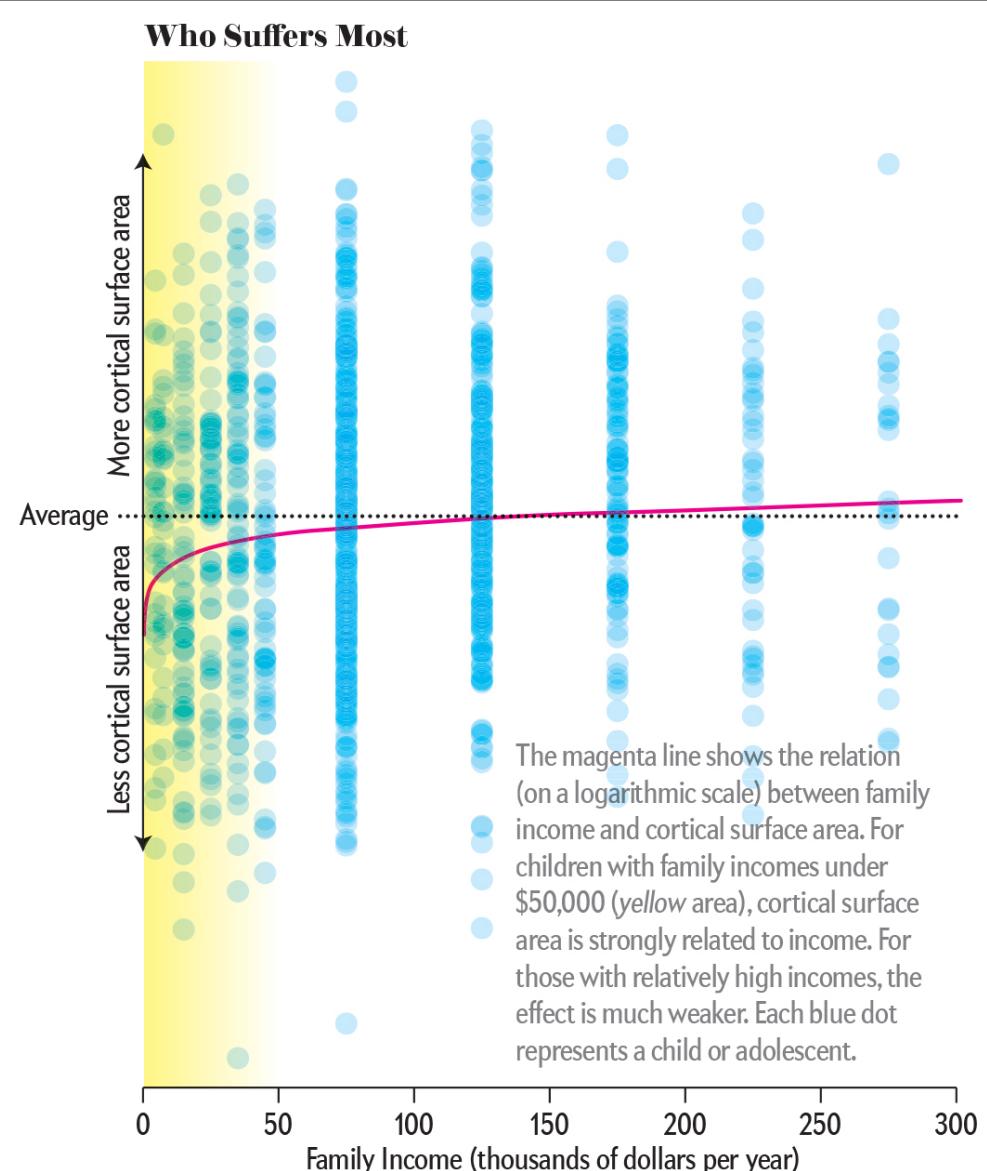
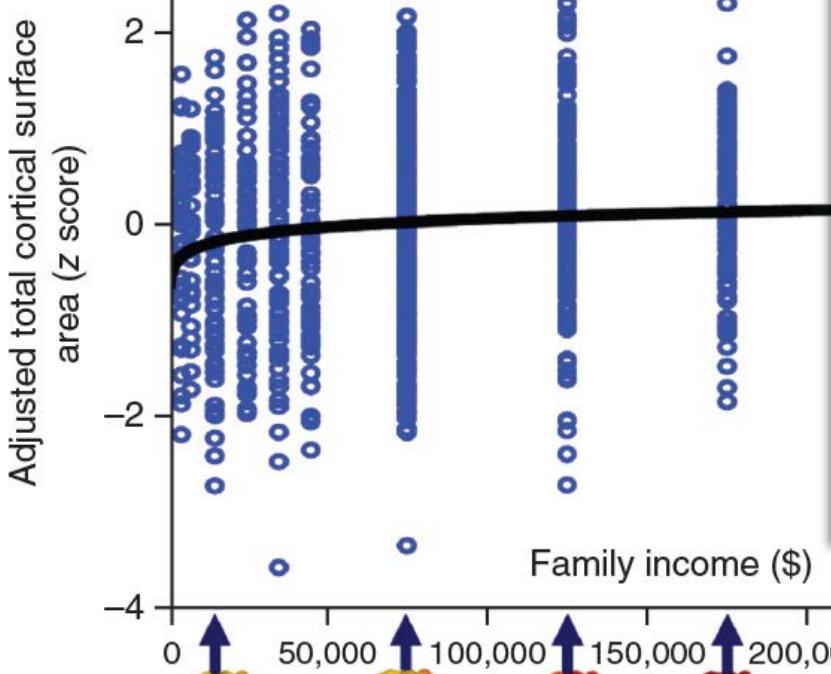
The travails of an impoverished upbringing reduce the surface area of some parts of the cortex more than others. The affected regions (magenta) participate in various forms of mental processing. The researchers demonstrated the connection by plotting collected measures of the affected regions (referred to as the cortical surface area) by socioeconomic status.

Areas of Vulnerability



reframing

a



reframing

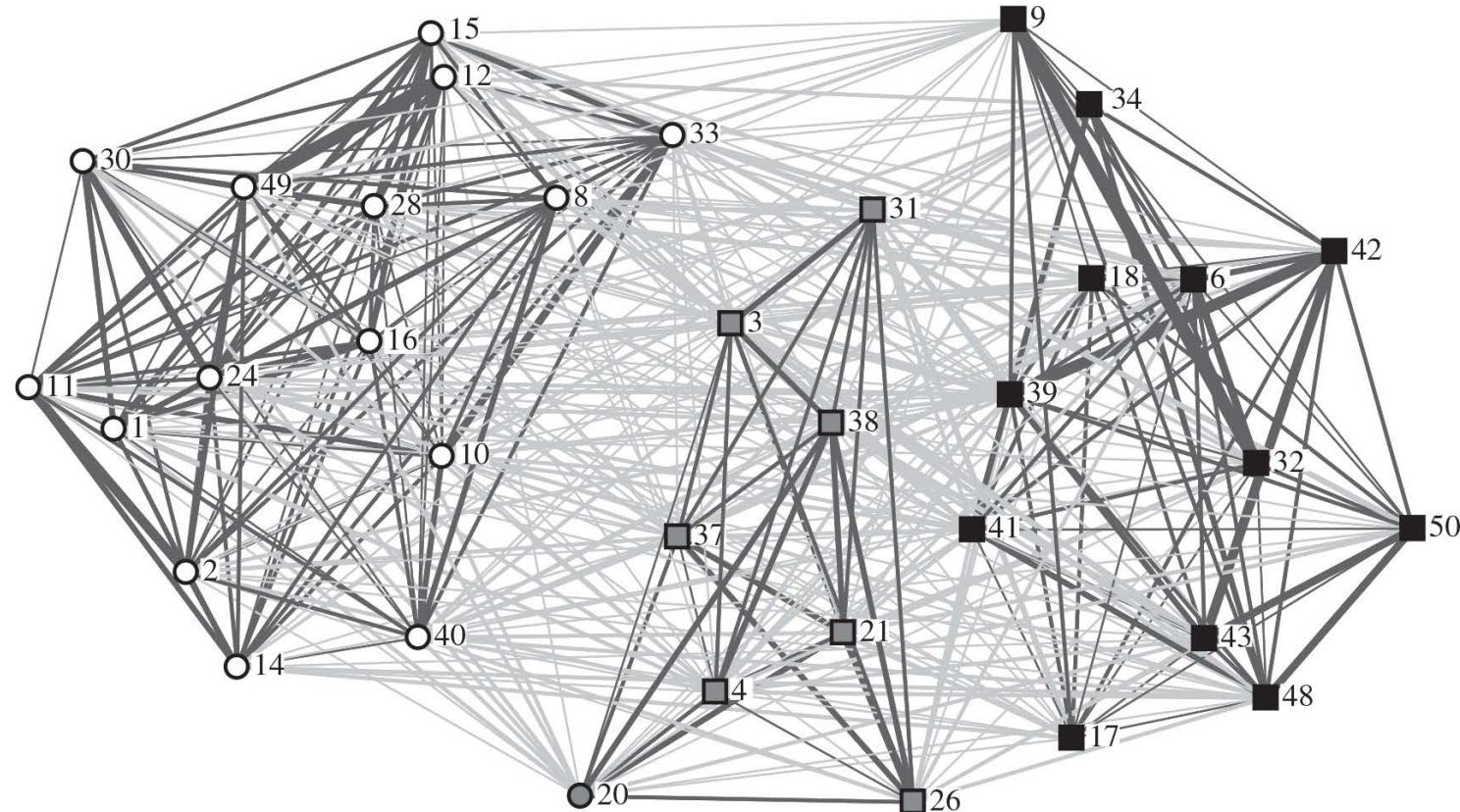


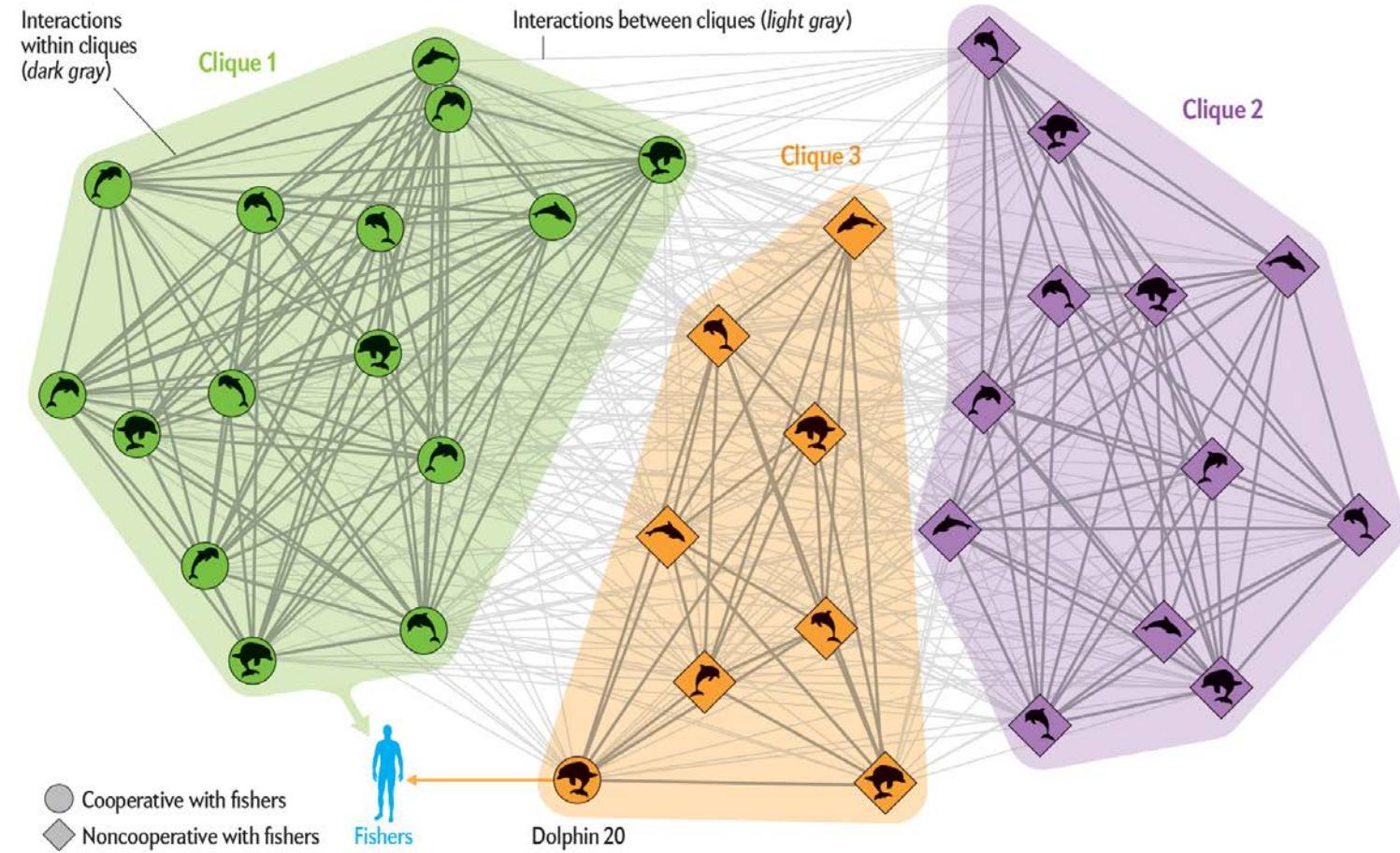
Figure 1. Social network of dolphins from Laguna with three modules defined by the foraging tactic of cooperation with artisanal fishermen. Module 1 (white nodes) comprises cooperative dolphins (circles), module 2 (grey nodes) almost entirely comprises non-cooperative dolphins (squares) (except for dolphin '20') and module 3 (black nodes) comprises only non-cooperative dolphins (squares).

By F. G. Daura-Jorge et al., “The structure of a bottlenose dolphin society is coupled to a unique foraging cooperation with artisanal fishermen.”

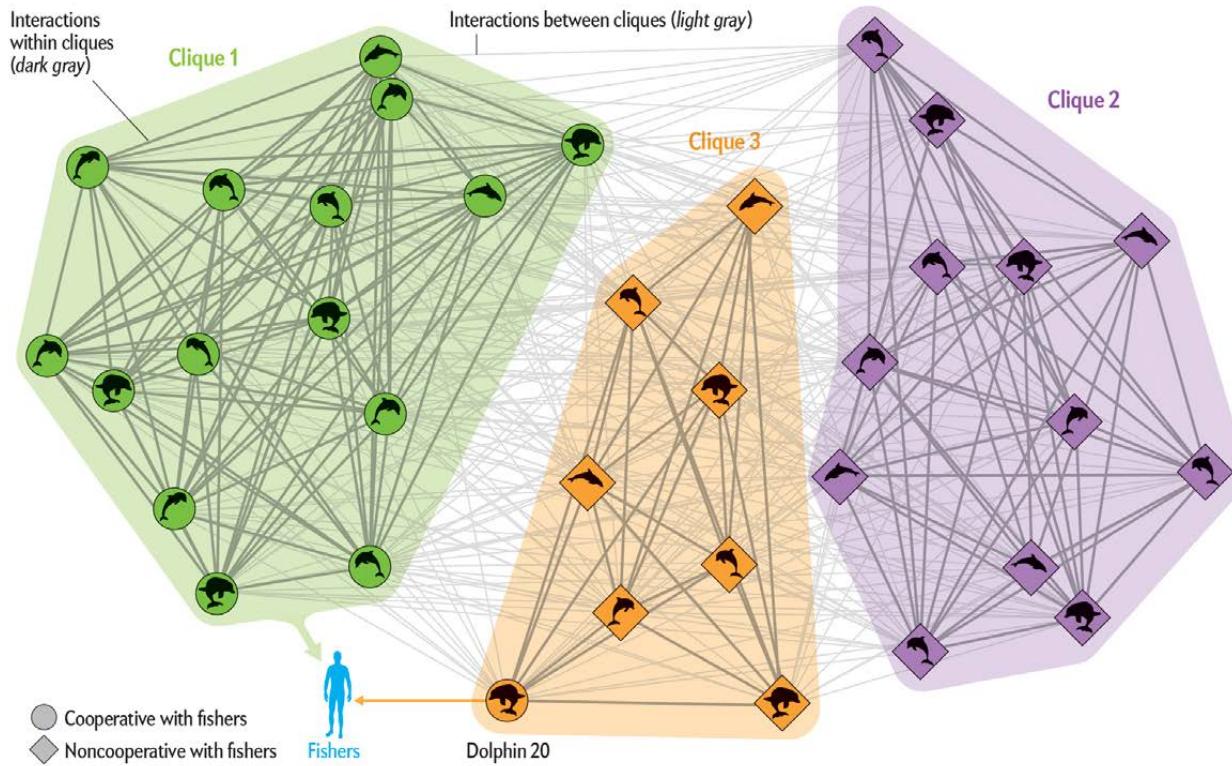
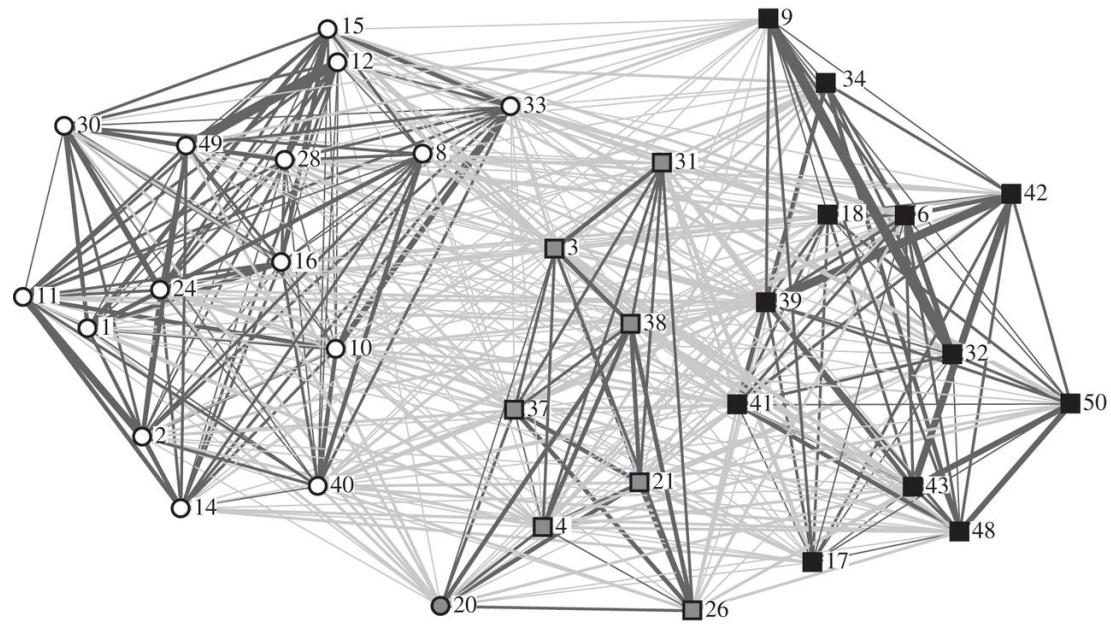
reframing Dolphins and Humans Team Up to Bag Fish

Some—but only some—members of a community of dolphins in the Laguna region of southern Brazil formed a unique alliance with local artisanal fishers attempting to net mullet. The dolphins formed three cliques. All members of clique 1 (green) were cooperative (indicated by circles) with the fishers and highly interactive (lines joining individuals) with one another. Clique 2 (purple) members interacted less with one another than did clique 1 animals and

had no contact with fishers (indicated by squares). Clique 3 (orange) dolphins were also uninterested in the fishers, with the notable exception of one member, known as dolphin “20” (orange circle). This individual aided the humans and liaised between its clique and the cooperative clique 1—and may yet teach clique 3 to work with humans. The collaboration of dolphins with fishers is known to enhance dolphin foraging success and human catch size.



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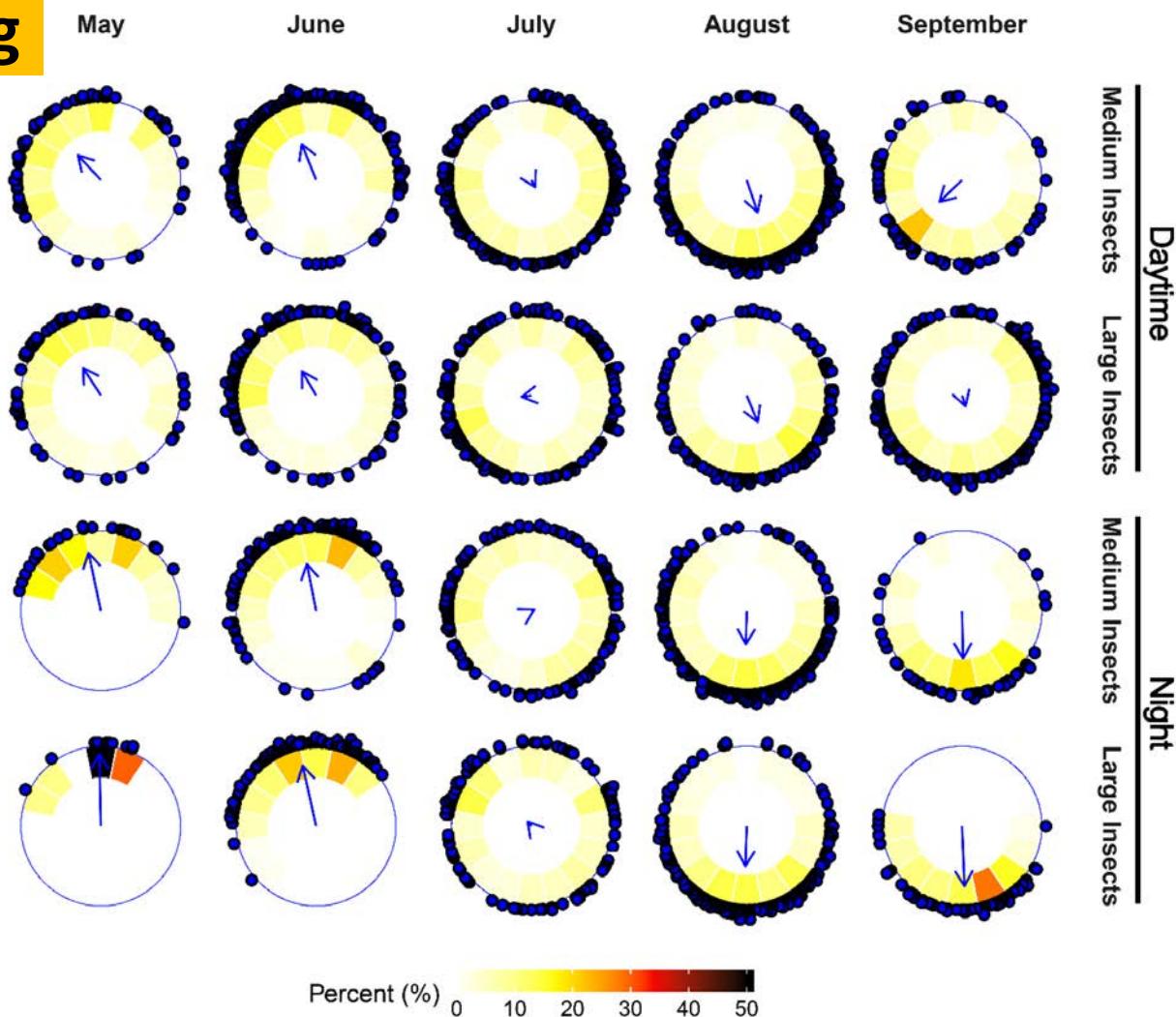
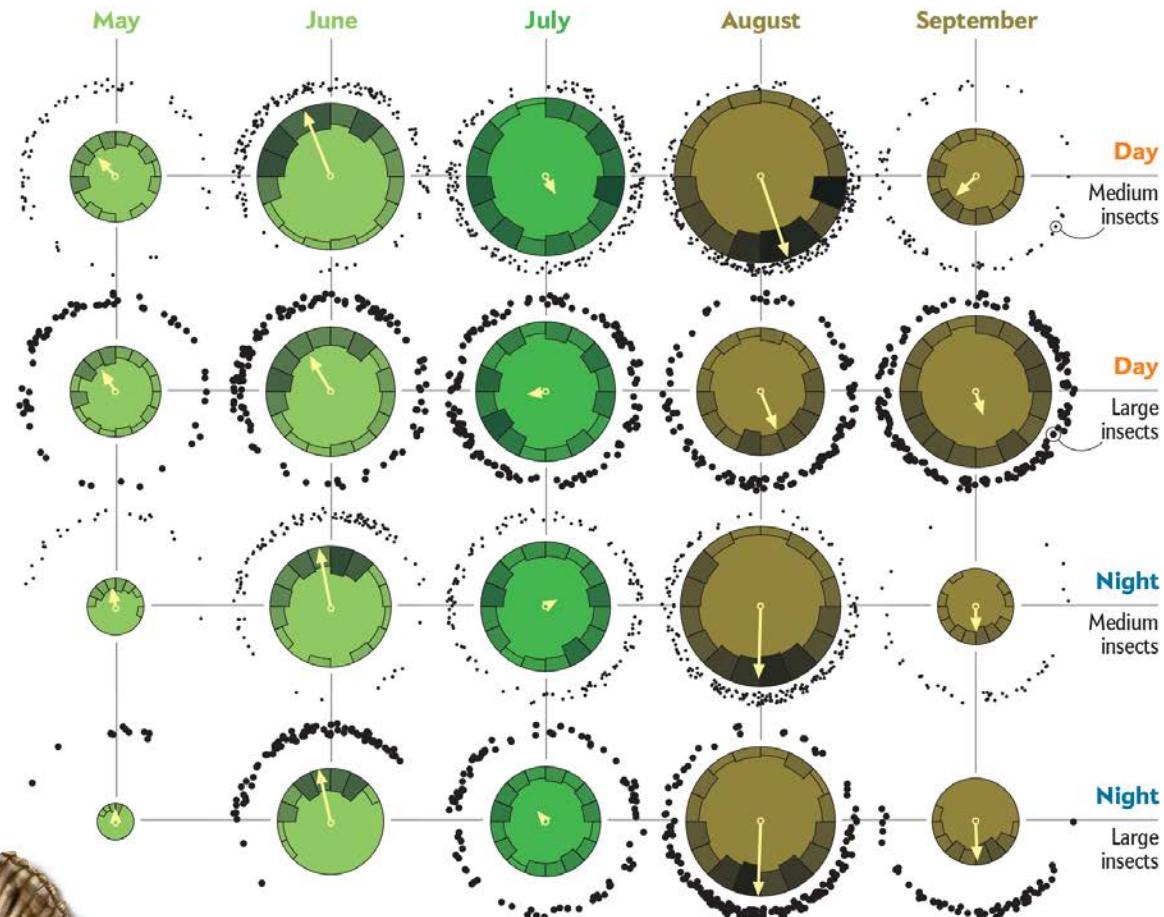


Fig. S2. Monthly track directions of larger insects migrating high above the southern United Kingdom. Migratory tracks of larger insects occurred predominantly in seasonally-beneficial directions in May and June, and in August and September, but were randomly directed in July. Small blue circles indicate the overall mean direction of the dataset, and arrow length is proportional to the number of tracks. By Gao Hu et al., “Mass seasonal bioflows of high-flying insect migrants”

reframing

Flight Patterns

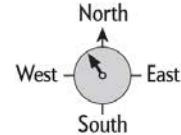
Over southern Britain, the most bug movement occurs in late May and June toward the northwest and in late August and September toward the southeast (arrows). Many insects do not live long and struggle to survive the cold, so it may take up to six generations for one family, flying in slightly different directions from one breeding area to the next, to complete an annual migration.



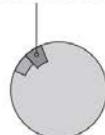
How to Read This Graphic

More than 70 percent of migrations occur during daytime (top two rows); fewer are at night (bottom two rows).

Migration Direction
Arrows depict the overall direction of insect flight.
A longer arrow means more insects stick close to that path.



Darker, larger shapes inside a circle show the more heavily traveled directions.



Relative Number of Insects
Larger circle means more insects are airborne. Black dots indicate episodes when many insects are flying in a given direction.

Trillions of Insects Migrate

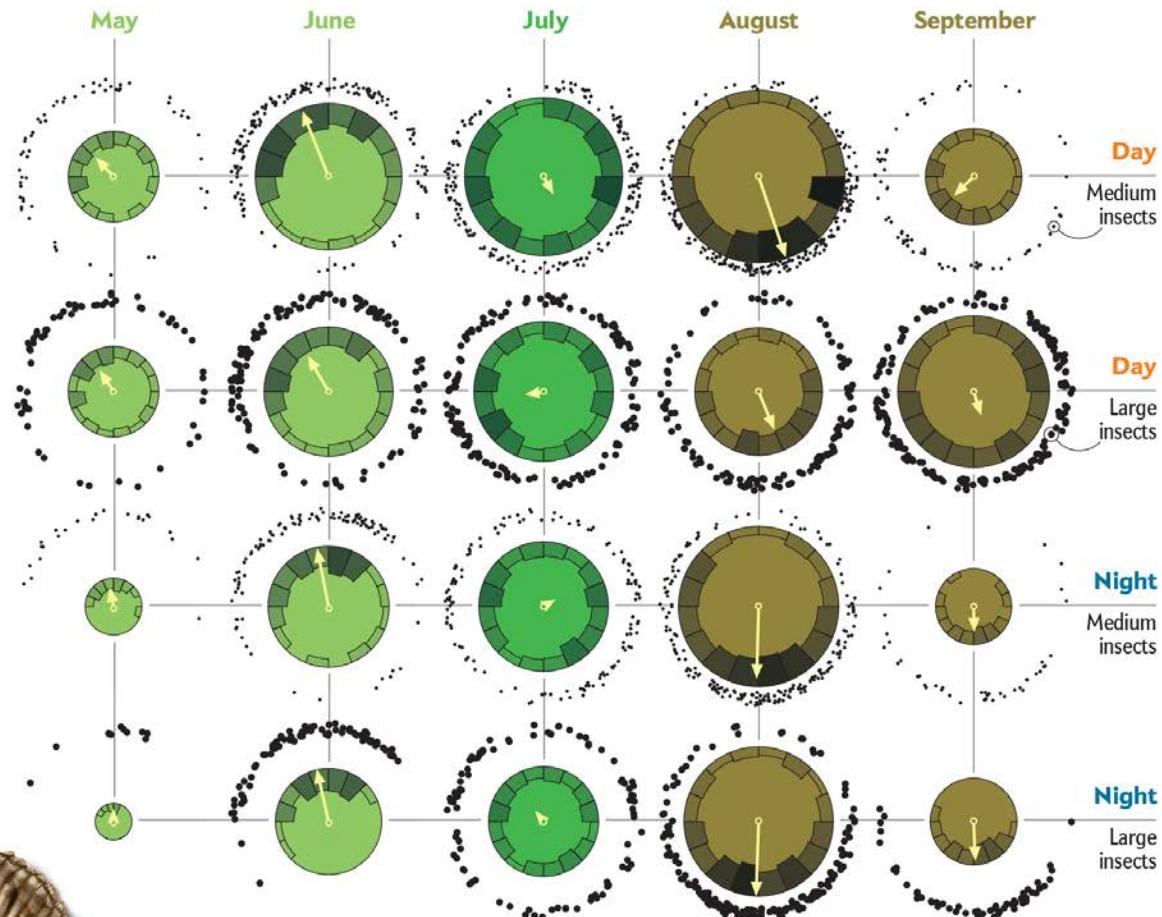
Surprising data show many species make annual treks

Jan Willem
Tulp (charts)
and Jessica
Huppi
(insects)

reframing

Flight Patterns

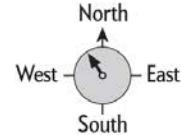
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Surprising data show many species make annual treks

reframing

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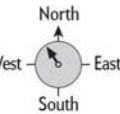


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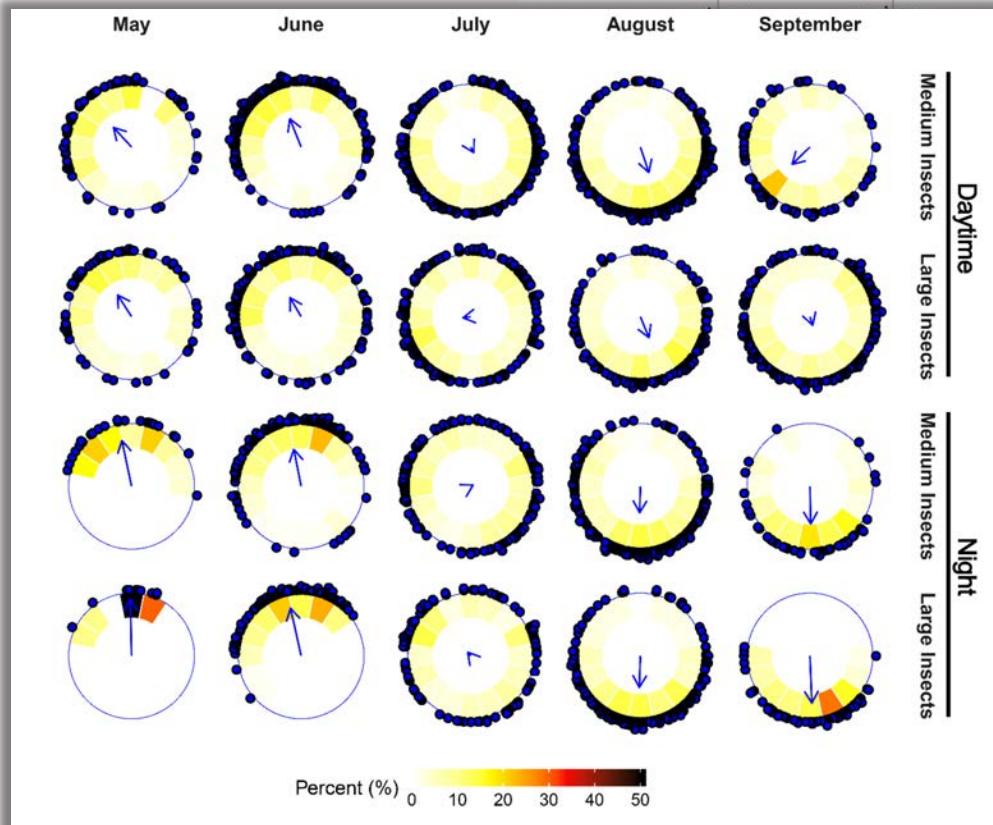
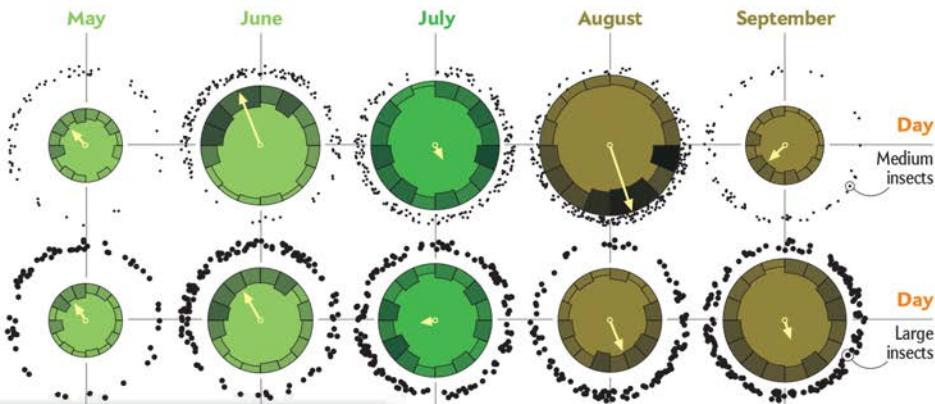
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ns of Insects Migrate

data show many species make annual treks

customizing for a different [context]

- simple edits / reframing
 - shake out label jargon
 - add framing text (focus attention)
 - add annotations
- moderate reworks
 - add audience-specific context
 - combine or divide elements

rework

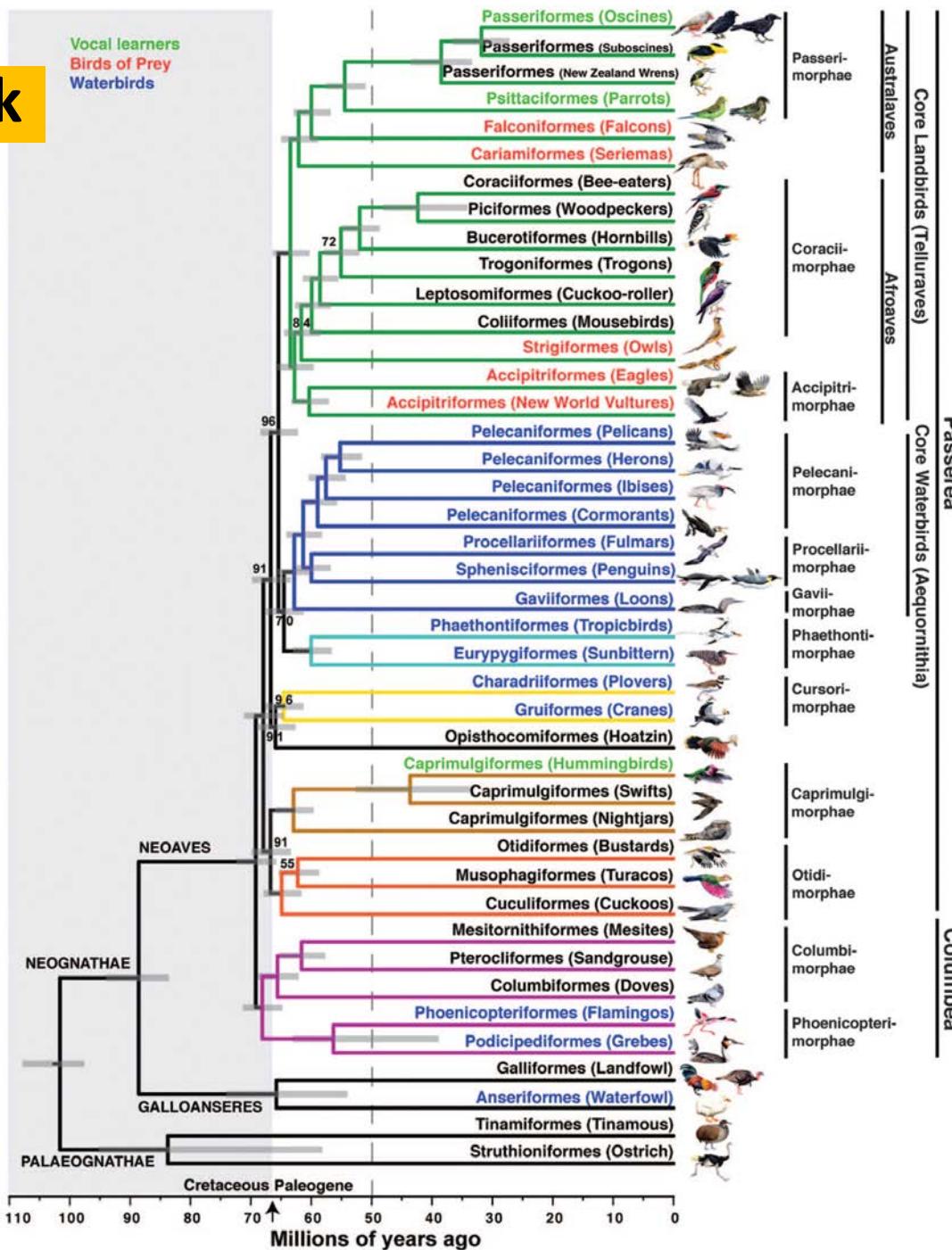


Fig. 1. Genome-scale phylogeny of birds. The dated TENT inferred with ExaML. Branch colors denote well-supported clades in this and other analyses. All BS values are 100% except where noted. Names on branches denote orders (-iformes) and English group terms (in parentheses); drawings are of the specific species sequenced (names in table S1 and fig. S1). Order names are according to (36, 37) (SM6). To the right are superorder (-imorphae) and higher unranked names. In some groups, more than one species was sequenced, and these branches have been collapsed (noncollapsed version in fig. S1). Text color denotes groups of species with broadly shared traits, whether by homology or convergence. The arrow indicates the K-Pg boundary at 66 Ma, with the Cretaceous period shaded at left. The gray dashed line represents the approximate end time (50 Ma) by which nearly all neoavian orders diverged. Horizontal gray bars on each node indicate the 95% credible interval of divergence time in millions of years.

By Erich D. Jarvis et al., “Whole-genome analyses resolve early branches in the tree of life of modern birds”

rework

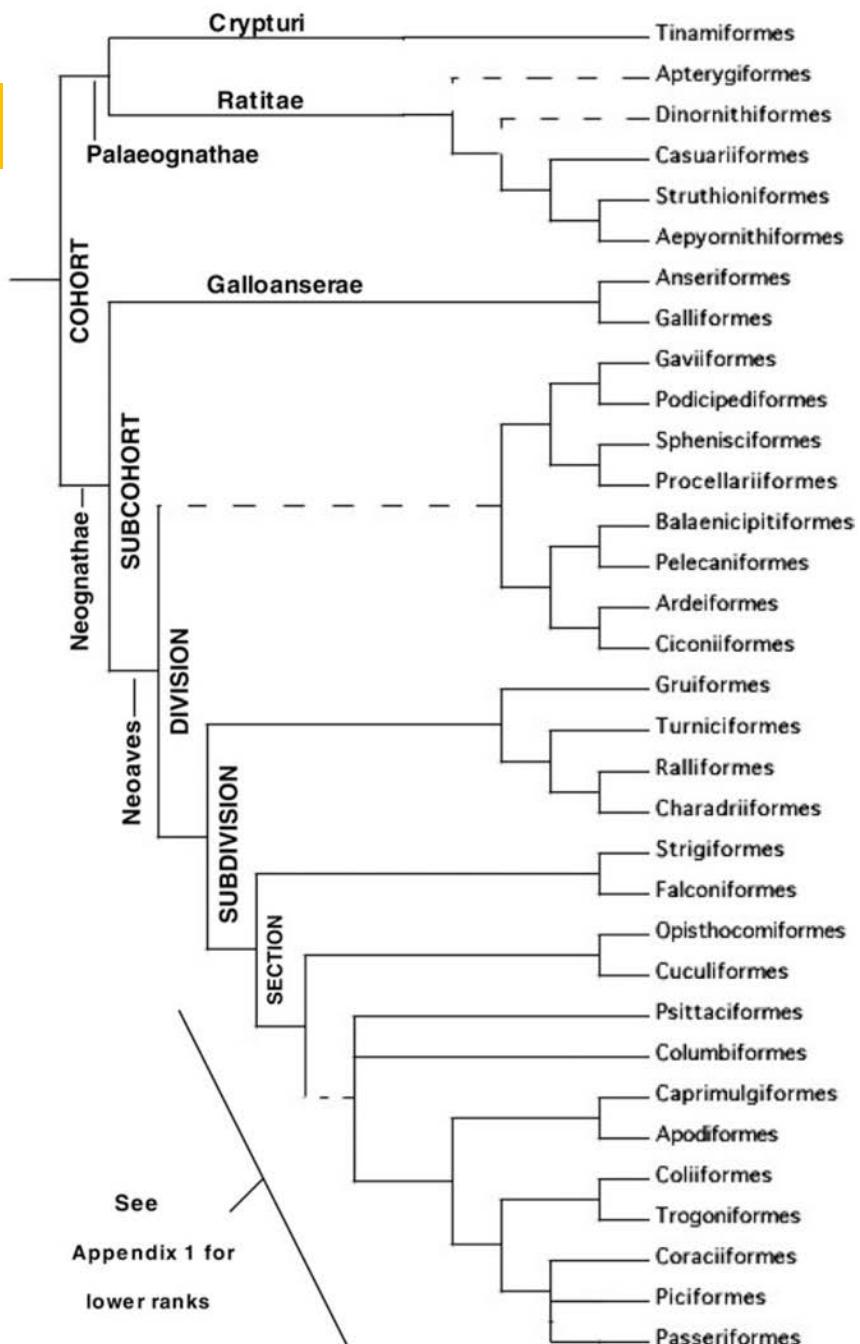
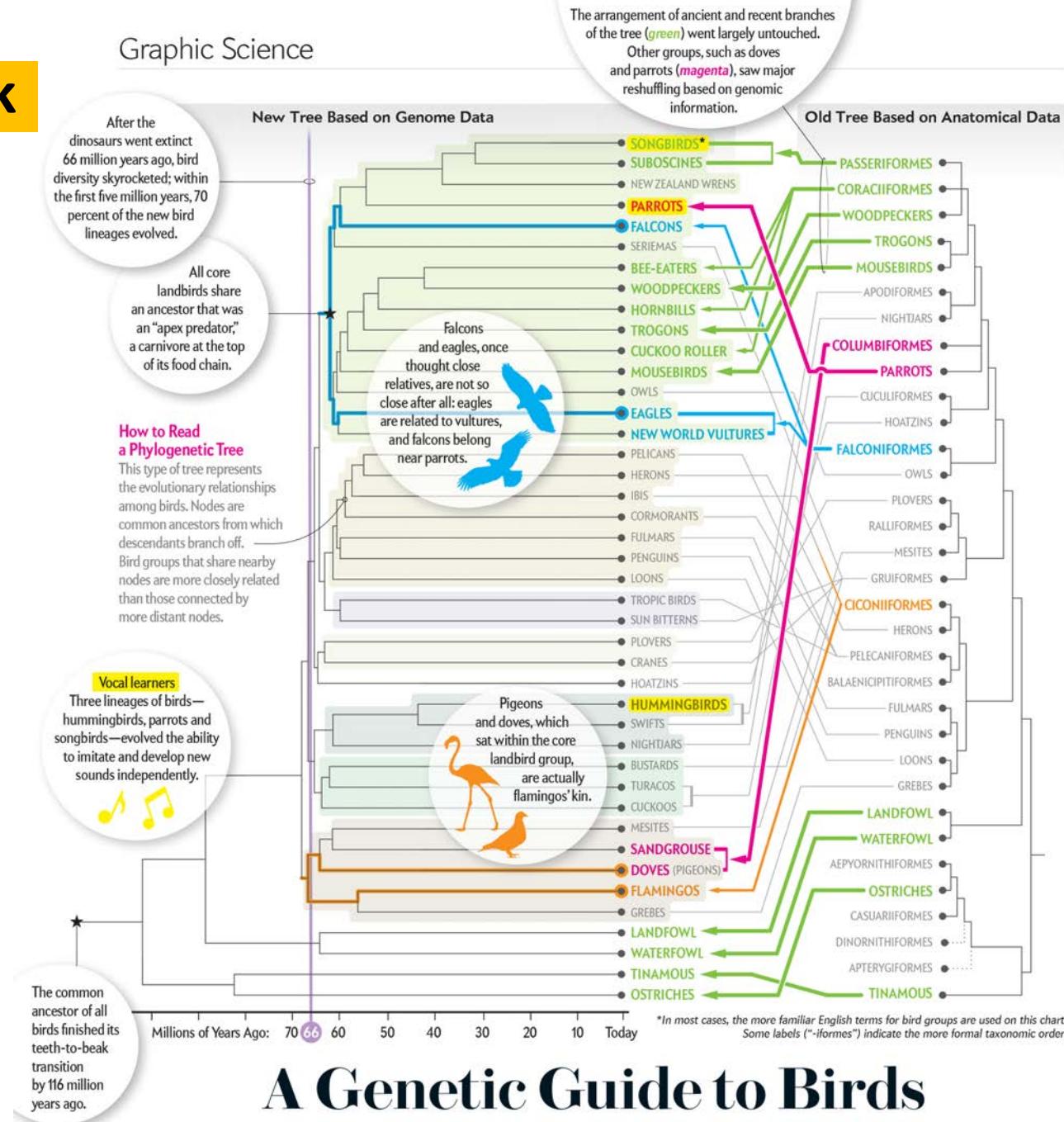


Figure 11. Simplified summary tree for uppermost, supraordinal ranks of avian classification. Dashed internodes correspond to marginally supported clades. For complete classification, see Appendix 1.

Bradley C. Livezey and Richard L. Zusi, "Higher-order phylogeny of modern birds (Theropoda, Aves: Neornithes) based on comparative anatomy. II. Analysis and discussion"

rework

Graphic Science

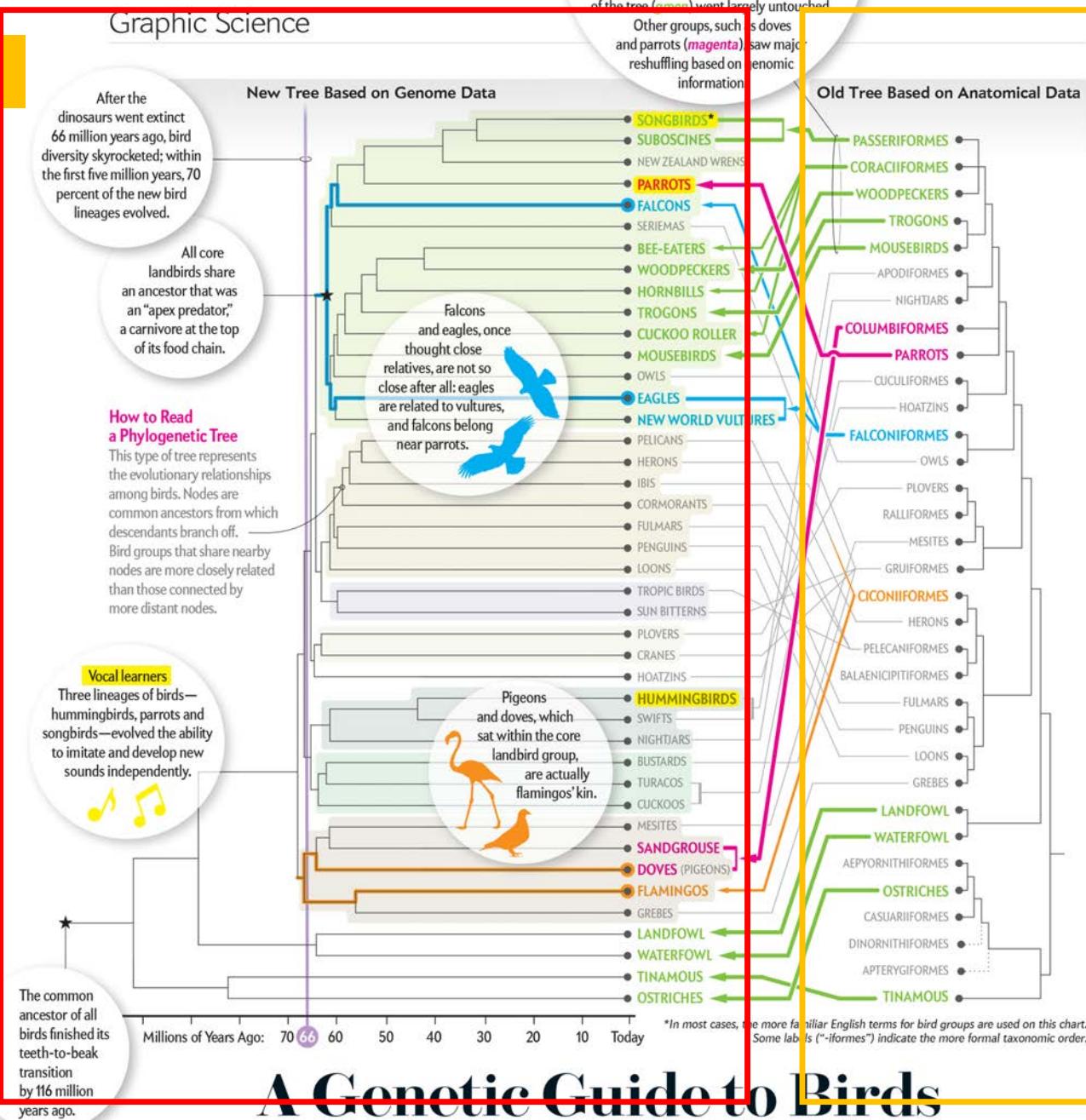


A Genetic Guide to Birds

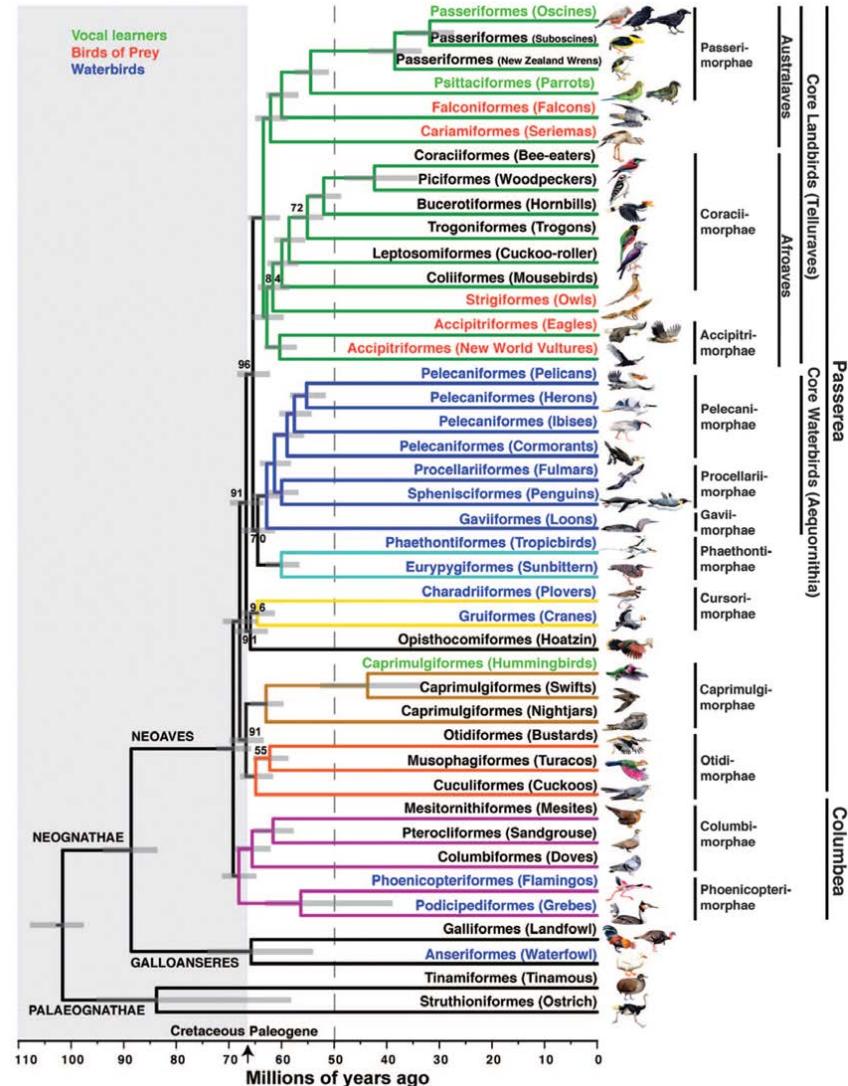
The avian family tree gets a makeover

rework

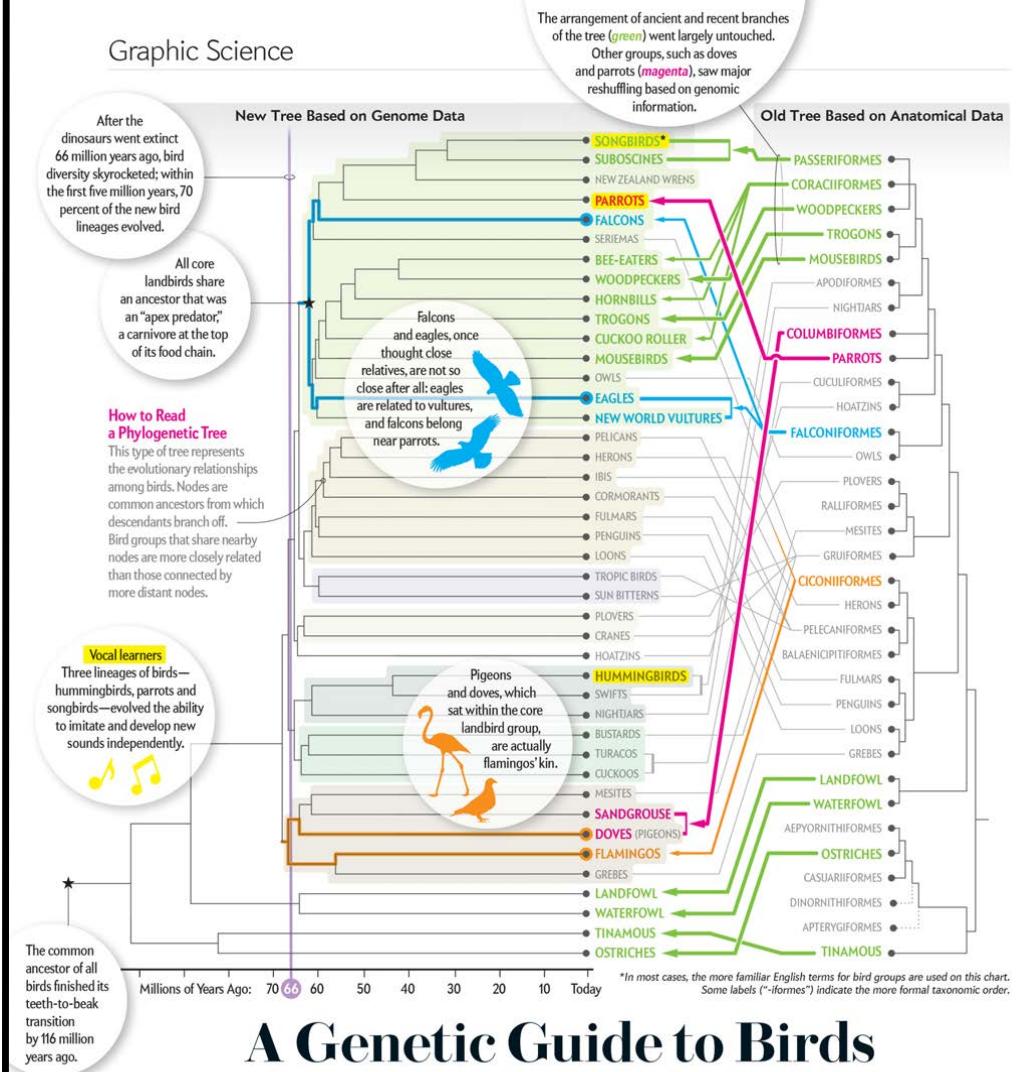
Graphic Science



rework



Graphic Science



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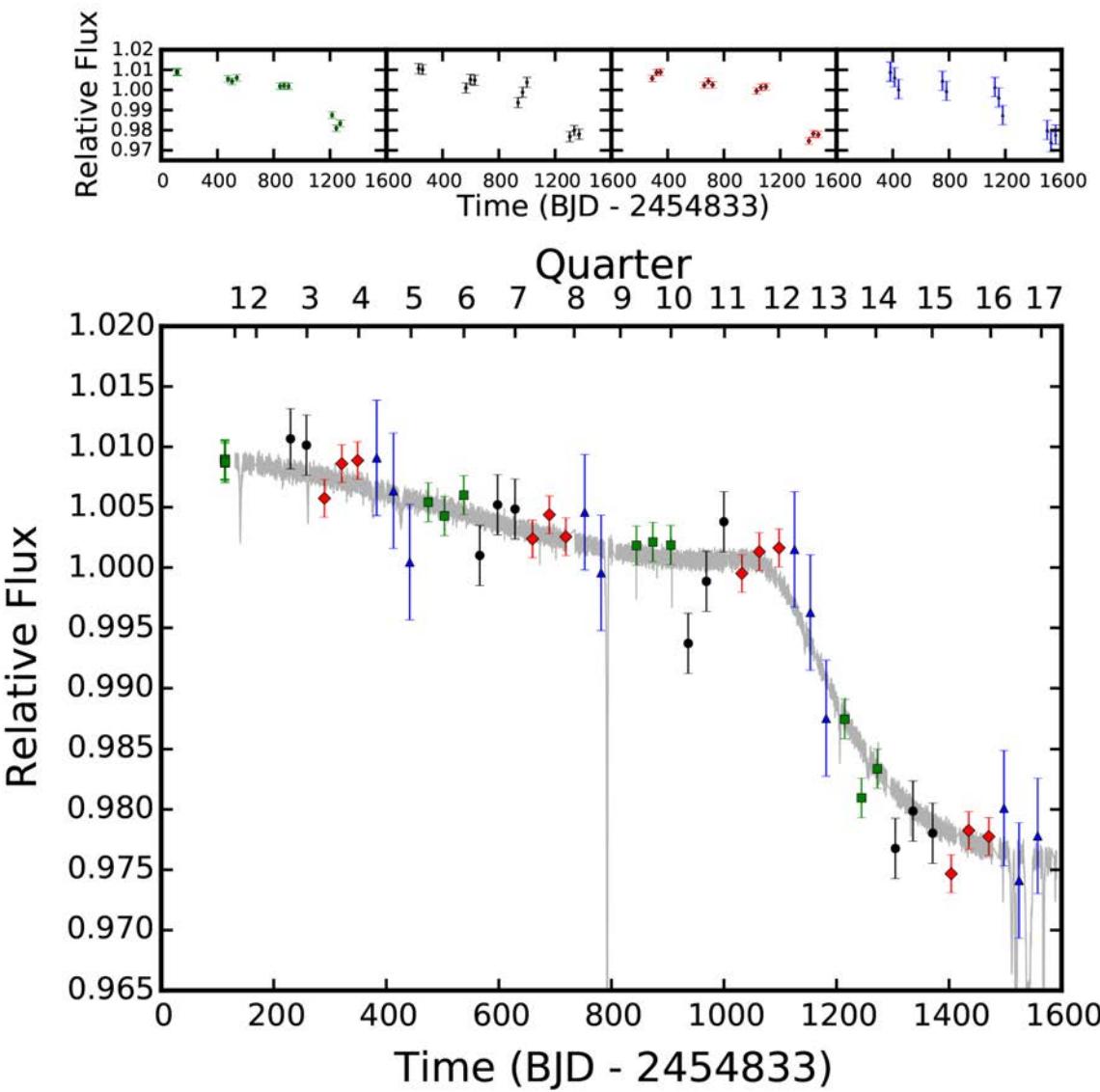


FIG. 3.— Photometry of KIC 8462852 as measured from the FFI data. The four colors and shapes (green squares, black circles, red diamonds, and blue triangles) represent measurements from the four separate channels the starlight reaches as the telescope rolls. The four subpanels show the flux measurements from each particular detector individually. The main figure combines all observations together. In the process of creating the fit, we allow a vertical offset between the data from each individual quarter to account for changes in the flat field with detector orientation. For the purposes of this figure, we plot the maximum likelihood values; the blue, green, and black points have been shifted upward by 0.6%, 0.1%, and 0.2%, respectively, to account for the uncertainty in the fit. The red points have not been shifted. In all four channels, the photometry is consistent with a rapid decrease in flux of $\approx 2.5\%$. The light gray curve represents the photometry created by fitting a spline to the FFI photometry. The narrow vertical error bars are visible but narrow relative to the cadence of FFI observations in the online version of the journal.

By Benjamin T. Montet and Joshua D. Simon, “KIC 8462852 Faded Throughout the Kepler Mission”

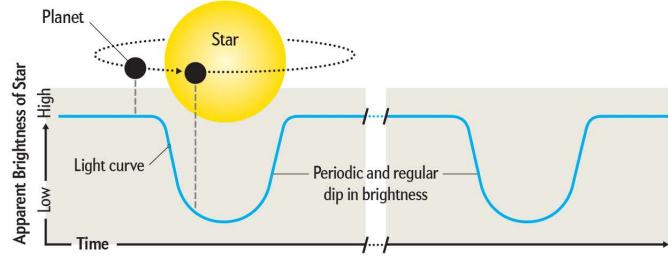
OBSERVATIONS

Enigmatic Light Patterns

To astronomers, there is usually no mystery behind a star fading in the sky. Starspots as well as the shadows of planets or debris disks routinely dim the otherwise steady light from mature stars. But none of these explanations seems to apply for one mercurial middle-aged sun known as KIC 8462852—also called Boyajian's star.

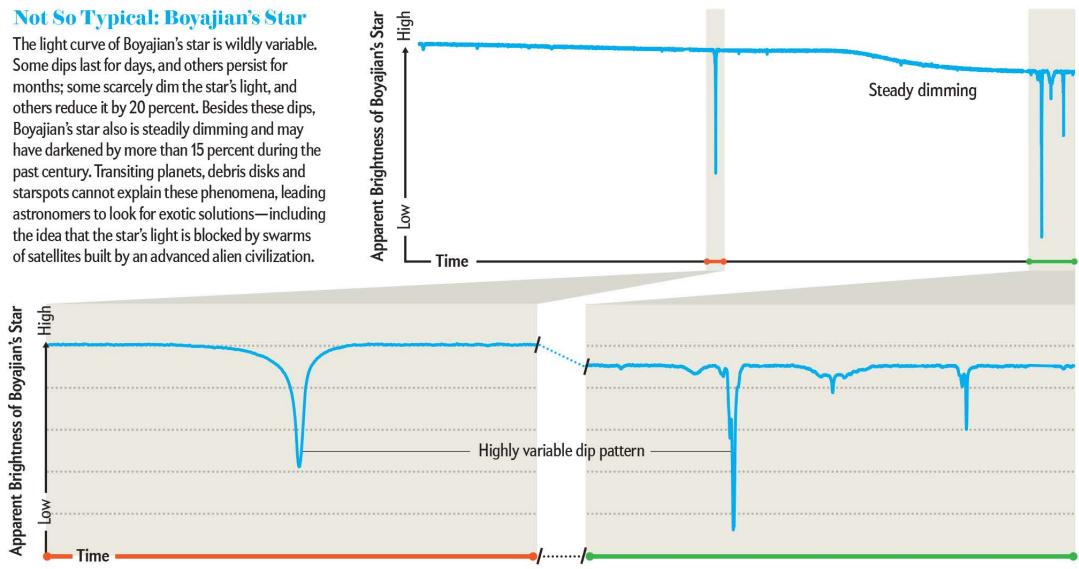
Typical Light Curve

A dimming star can be studied by its light curve—its brightness plotted over time. A planet or disk “transiting” across a star causes a dip in the curve; for planets, this dip recurs every orbital period. Starspots create patterns in light curves based on a star’s rotation rate and activity cycle.



Not So Typical: Boyajian's Star

The light curve of Boyajian's star is wildly variable. Some dips last for days, and others persist for months; some scarcely dim the star's light, and others reduce it by 20 percent. Besides these dips, Boyajian's star also is steadily dimming and may have darkened by more than 15 percent during the past century. Transiting planets, debris disks and starspots cannot explain these phenomena, leading astronomers to look for exotic solutions—including the idea that the star's light is blocked by swarms of satellites built by an advanced alien civilization.

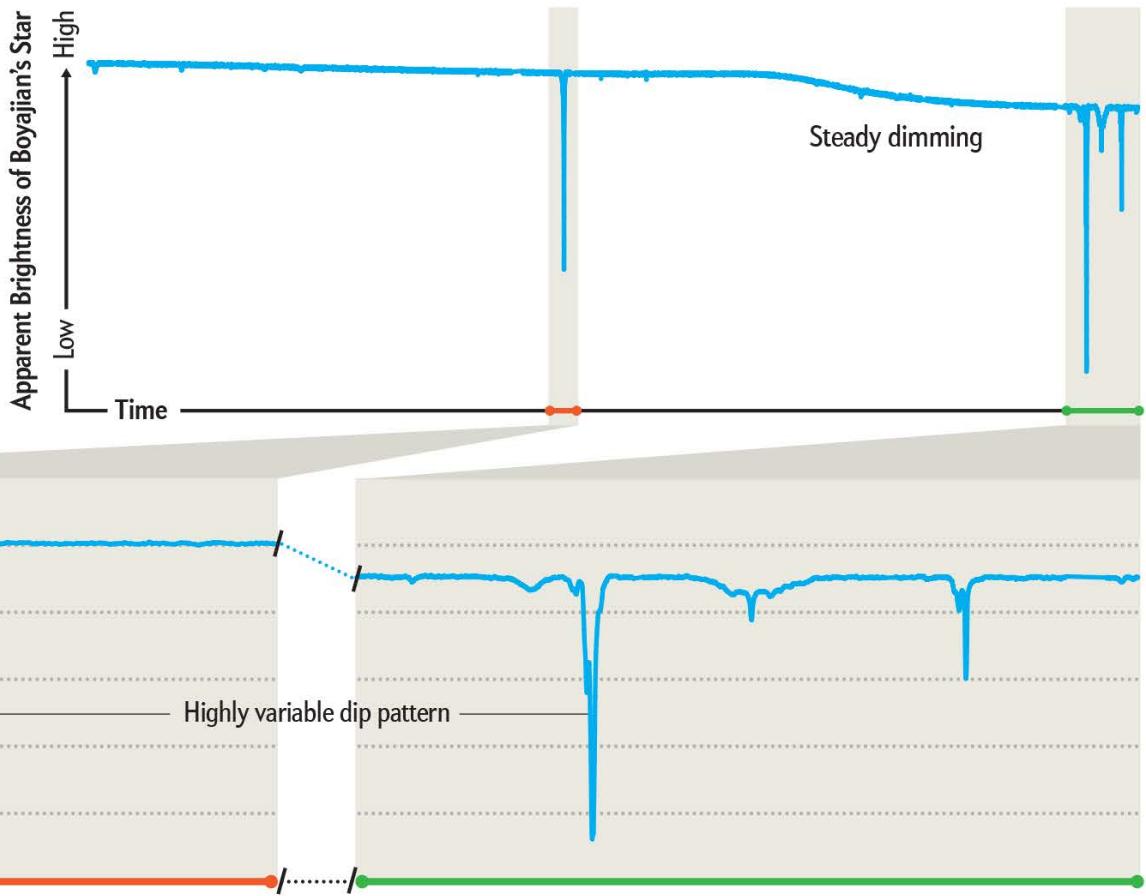


By Tiffany Farrant-Gonzalez, In “Strange News From Another Star,” By Kimberly Cartier and Jason T. Wright

rework

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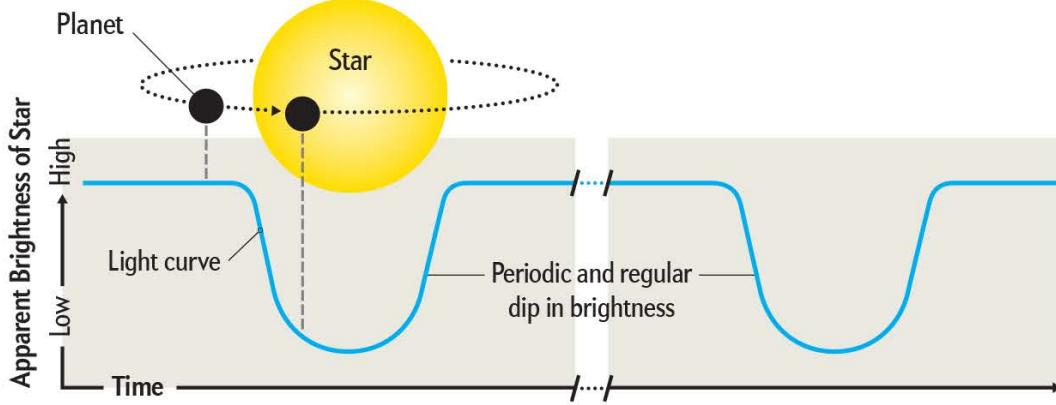
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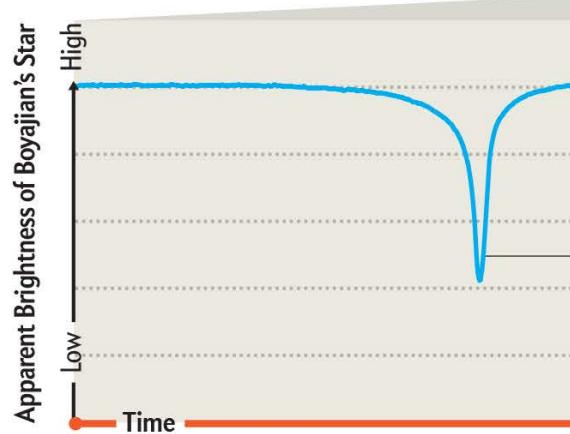
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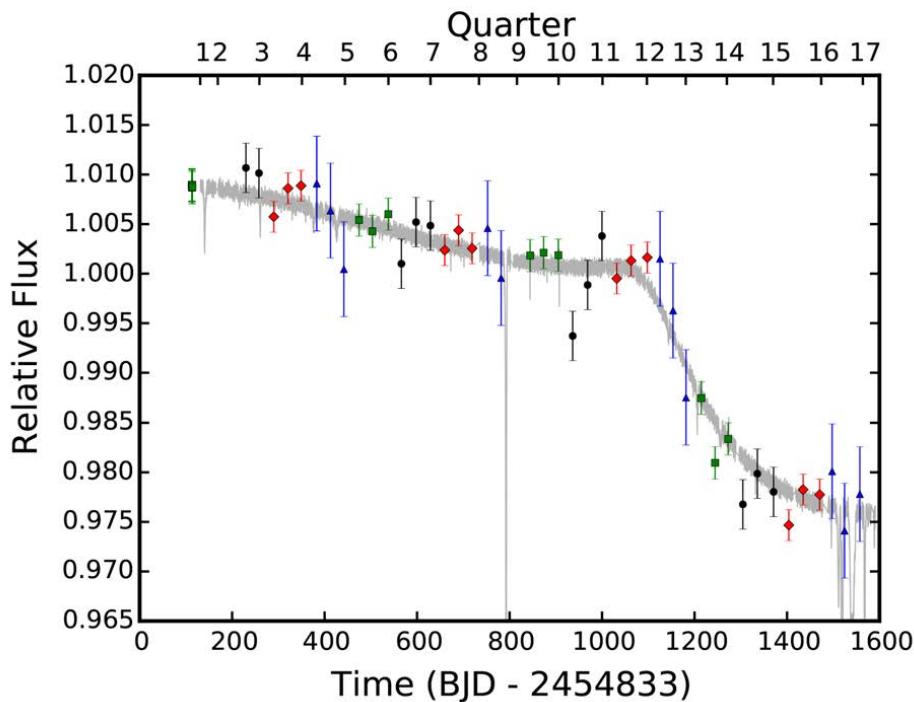
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By Tiffany Farrant-Gonzalez, In “Strange News From Another Star,” By Kimberly Cartier and Jason T. Wright

rework



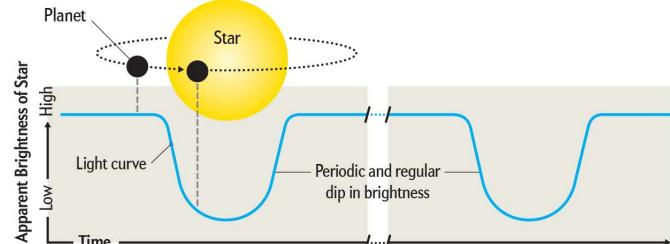
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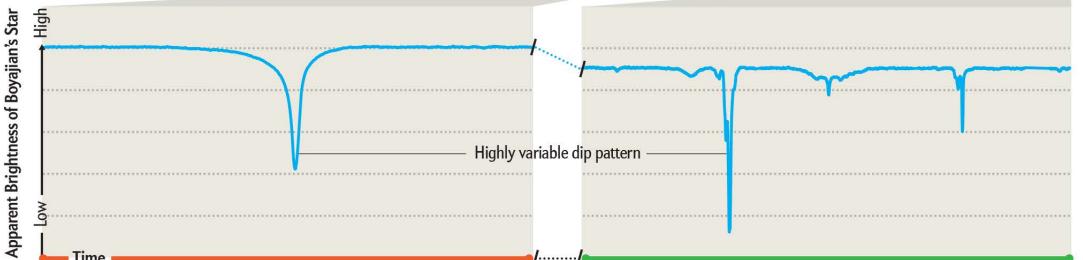
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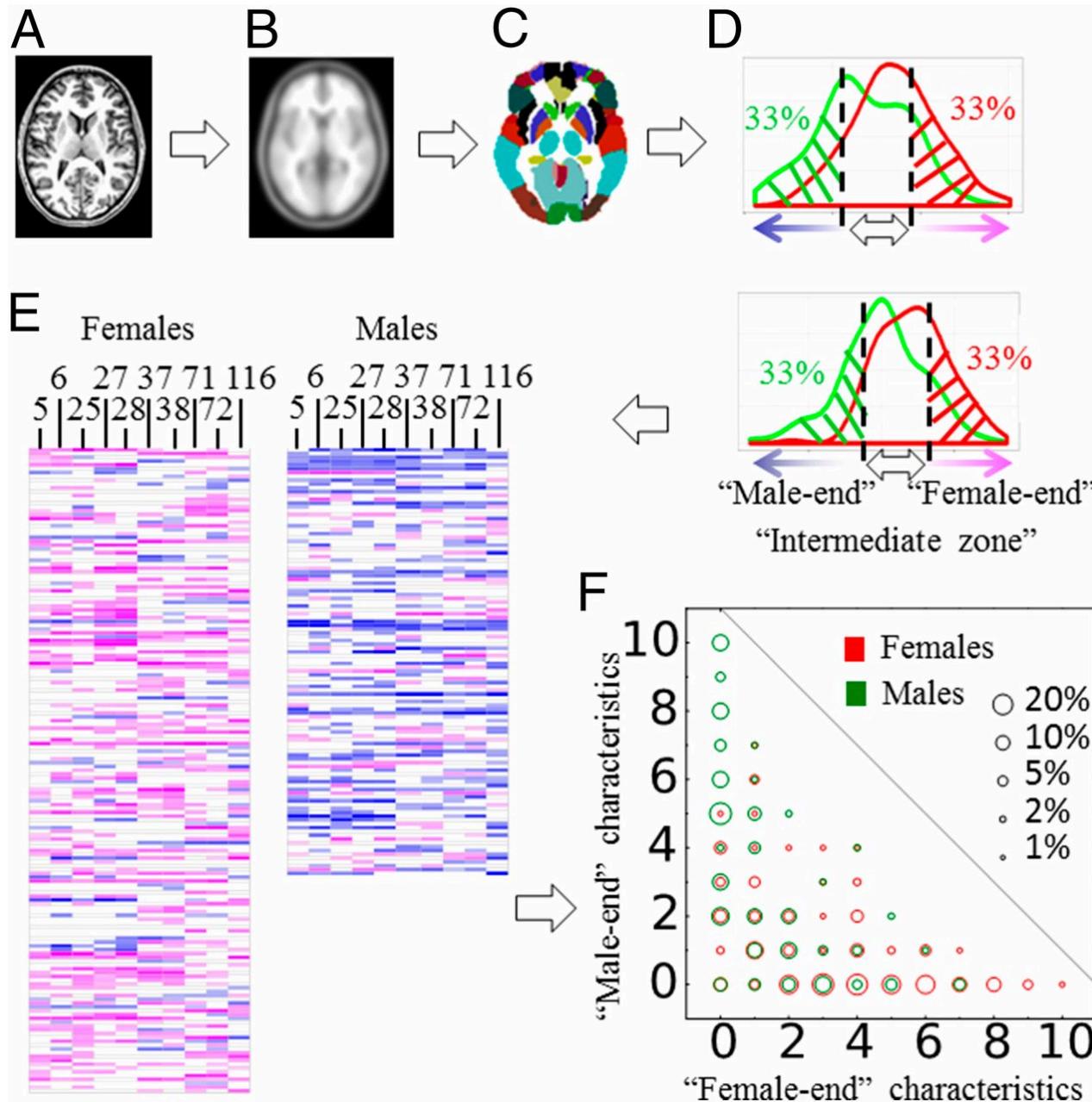


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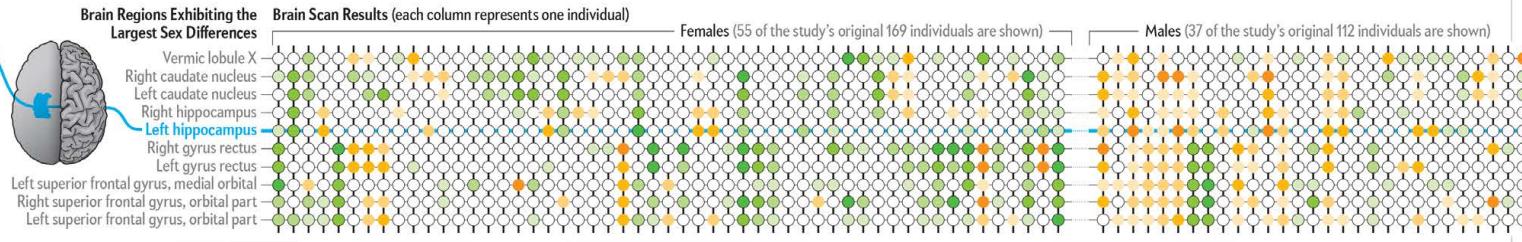
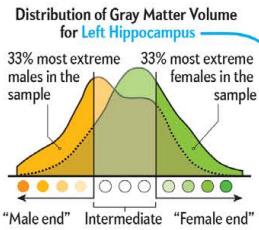
rework



By Daphna Joel et al., “Sex beyond the genitalia: The human brain mosaic”

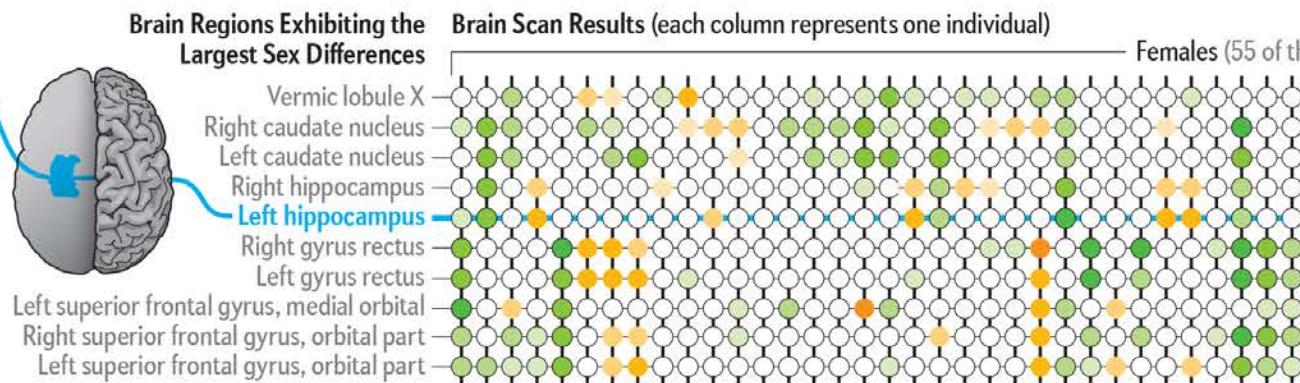
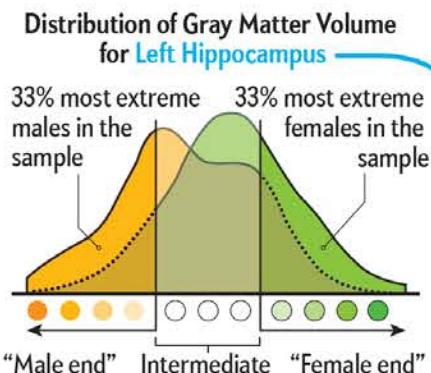
The Mosaic Brain

Sex differences found in the human brain have led to the perception that brains are either male or female. A study by Daphna Joel of Tel Aviv University and her colleagues tells a different story. Joel's research found that the typical brain is a "mosaic," combining some features more common in males and some that appear more frequently in females, pointing to the conclusion that human brains do not belong to two distinct types categorized by sex.



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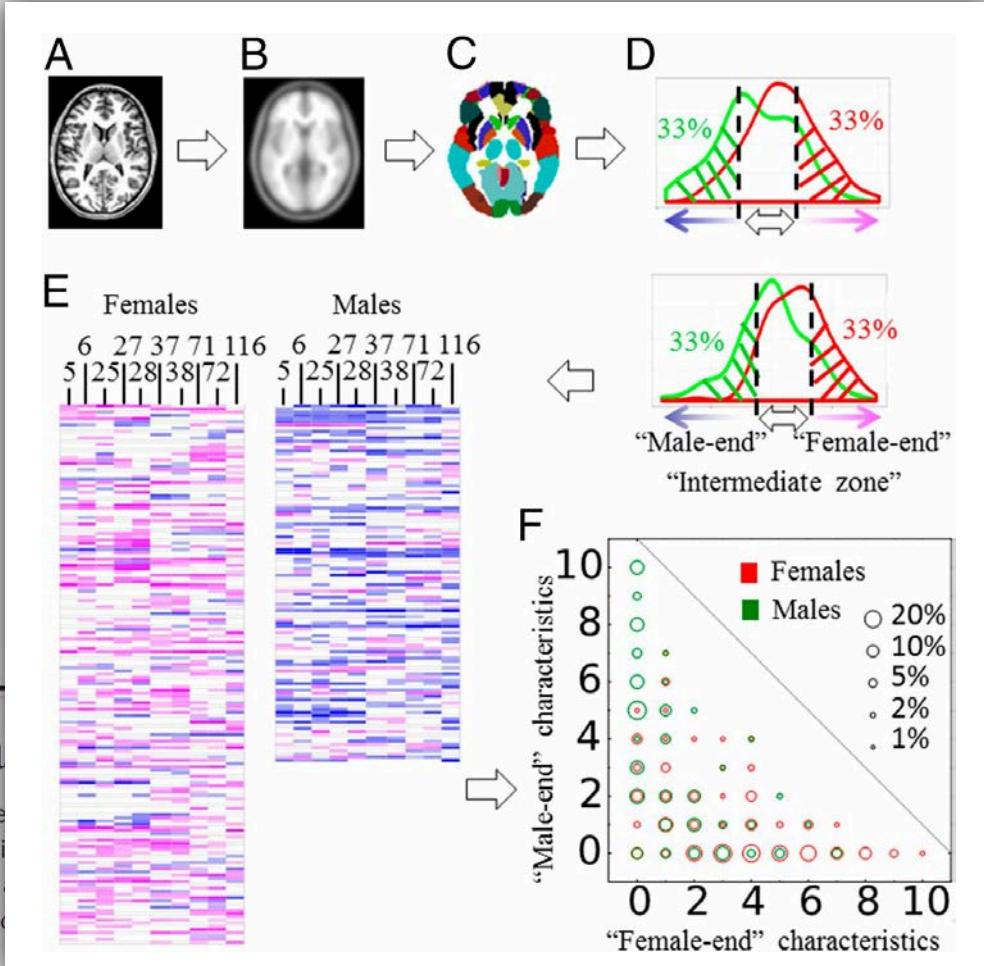
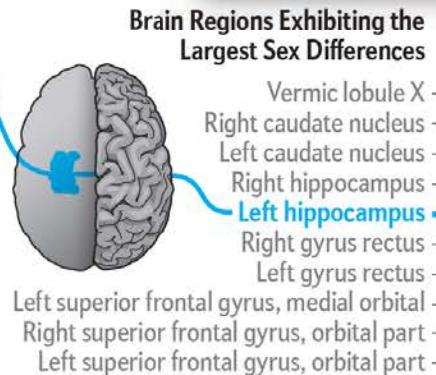
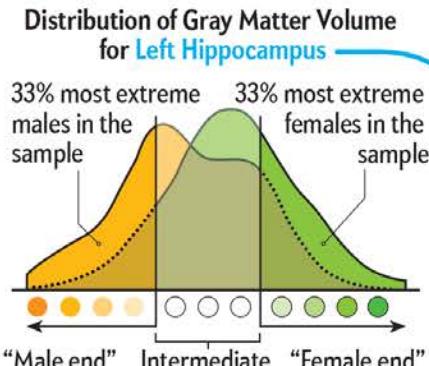
NEURAL SIGNATURES OF

In her 2015 study, Joel examined *Mus musculus* and found significant overlap among the areas showing largest differences between males and females in hippocampus, most females and most males fell in the middle on a continuum of “maleness.”

rework

The Mosaic

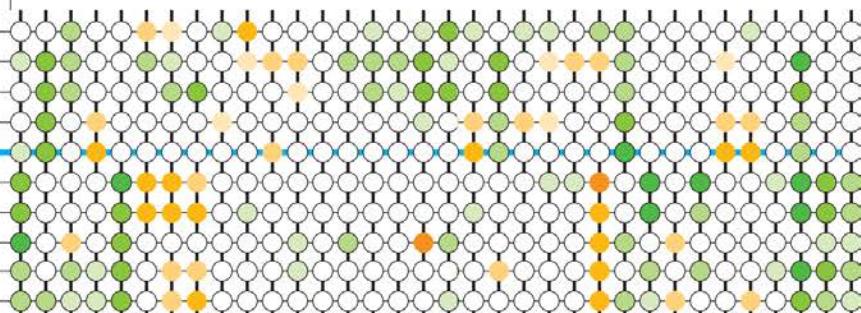
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the study data

Brain Scan Results (each column represents one individual)

Females (55 of the)



customizing for a different [context]

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Table 2 | Author position on rosiglitazone safety and financial conflicts of interest

	Risk of myocardial infarction with rosiglitazone				Use of rosiglitazone			
	Favourable (n=31)	Neutral (n=84)	Unfavourable (n=65)	Rate ratio (95% CI)*	Favourable (n=26)	Neutral (n=116)	Unfavourable (n=38)	Rate ratio (95% CI)*
Any manufacturer	29 (94)	32 (38)	18 (28)	3.38 (2.26 to 5.06)	23 (88)	46 (40)	10 (26)	3.36 (1.94 to 5.83)
Rosiglitazone manufacturer†	27 (87)	25 (24)	13 (20)	4.29 (2.63 to 7.02)	21 (81)	30 (26)	9 (24)	3.60 (2.00 to 6.48)
Pioglitazone manufacturer†	20 (65)	31 (30)	14 (22)	3.96 (2.45 to 6.39)	19 (73)	30 (26)	10 (26)	3.28 (1.88 to 5.73)
None	2 (6)	52 (62)	47 (72)	—	3 (12)	70 (60)	28 (74)	—

Values are numbers (percentages) unless otherwise indicated.

*Comparing favourable versus unfavourable views.

†Categories not mutually exclusive.

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Any manufacturer	29 (94)	32 (38)	18 (28)	3.38 (2.26 to 5.06)	23 (88)	46 (40)	10 (26)	3.36 (1.94 to 5.83)
Rosiglitazone manufacturer†	27 (87)	25 (24)	13 (20)	4.29 (2.63 to 7.02)	21 (81)	30 (26)	9 (24)	3.60 (2.00 to 6.48)
Pioglitazone manufacturer†	20 (65)	31 (30)	14 (22)	3.96 (2.45 to 6.39)	19 (73)	30 (26)	10 (26)	3.28 (1.88 to 5.73)
None	2 (6)	52 (62)	47 (72)	—	3 (12)	70 (60)	28 (74)	—

Values are numbers (percentages) unless otherwise indicated.

*Comparing favourable versus unfavourable views.

†Categories not mutually exclusive.

By Amy T Wang et al., “Association between industry affiliation and position on cardiovascular risk with rosiglitazone: cross sectional systematic review”

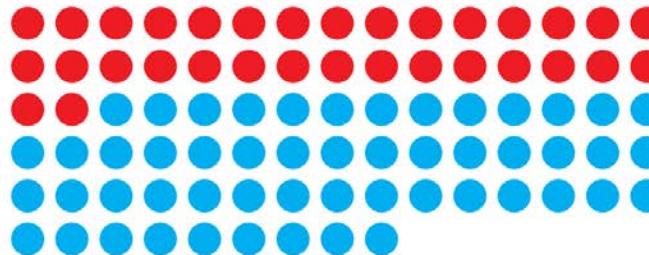
Case Study: Conflicting Interests

To what degree do financial entanglements influence the judgment of scientists? To find out, researchers at the Mayo Clinic in Rochester, Minn., focused on the diabetes drug rosiglitazone, which a meta-analysis had linked to increased risk of heart attacks. They examined articles that cited the meta-analysis or a subsequently released report on a large trial of rosiglitazone and found that scientists with a conflict of interest were more likely to view the drug favorably. "There was a clear and strong link between the orientation of authors' expressed views on the rosiglitazone controversy and their financial conflicts of interest with pharmaceutical companies," the report determined.

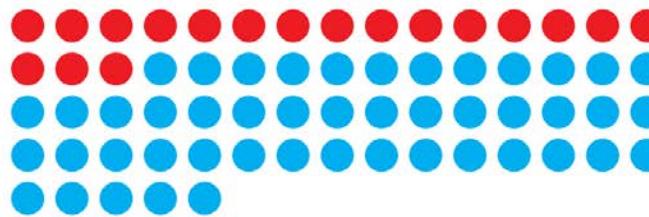
31 authors were classified as
favorable to the drug
(rosiglitazone does *not* increase
the risk of heart attack)



84 authors were
classified as **neutral**



65 authors were classified as
unfavorable to the drug
(rosiglitazone does *increase*
the risk of heart attack)



- Author identified as having a financial conflict of interest with manufacturers of rosiglitazone and/or other antihyperglycemic agents
- No financial conflict of interest

Case Study: Conflicting Interests

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Table 2 | Author position on rosiglitazone safety and financial conflicts of interest

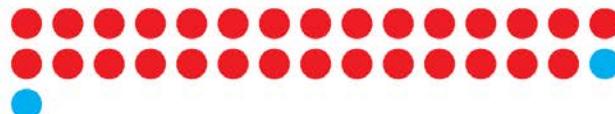
	Risk of myocardial infarction with rosiglitazone			
	Favourable (n=31)	Neutral (n=84)	Unfavourable (n=65)	Rate ratio vs unfavourable
Any manufacturer	29 (94)	32 (38)	18 (28)	3.38 (2.2)
Rosiglitazone manufacturer†	27 (87)	25 (24)	13 (20)	4.29 (2.7)
Pioglitazone manufacturer†	20 (65)	31 (30)	14 (22)	3.96 (2.6)
None	2 (6)	52 (62)	47 (72)	

Values are numbers (percentages) unless otherwise indicated.

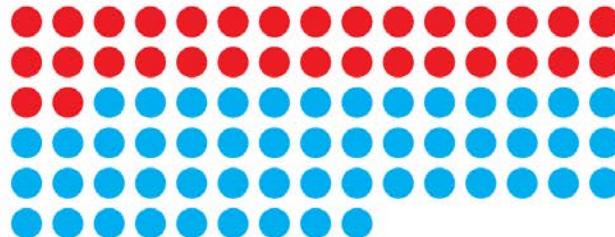
*Comparing favourable versus unfavourable views.

†Categories not mutually exclusive.

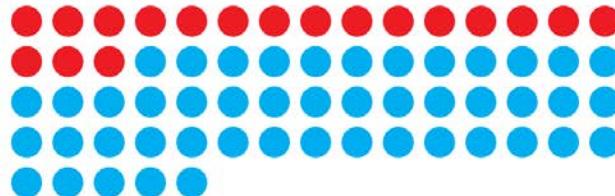
31 authors were classified as **favorable** to the drug (rosiglitazone does not increase the risk of heart attack)



84 authors were classified as **neutral**

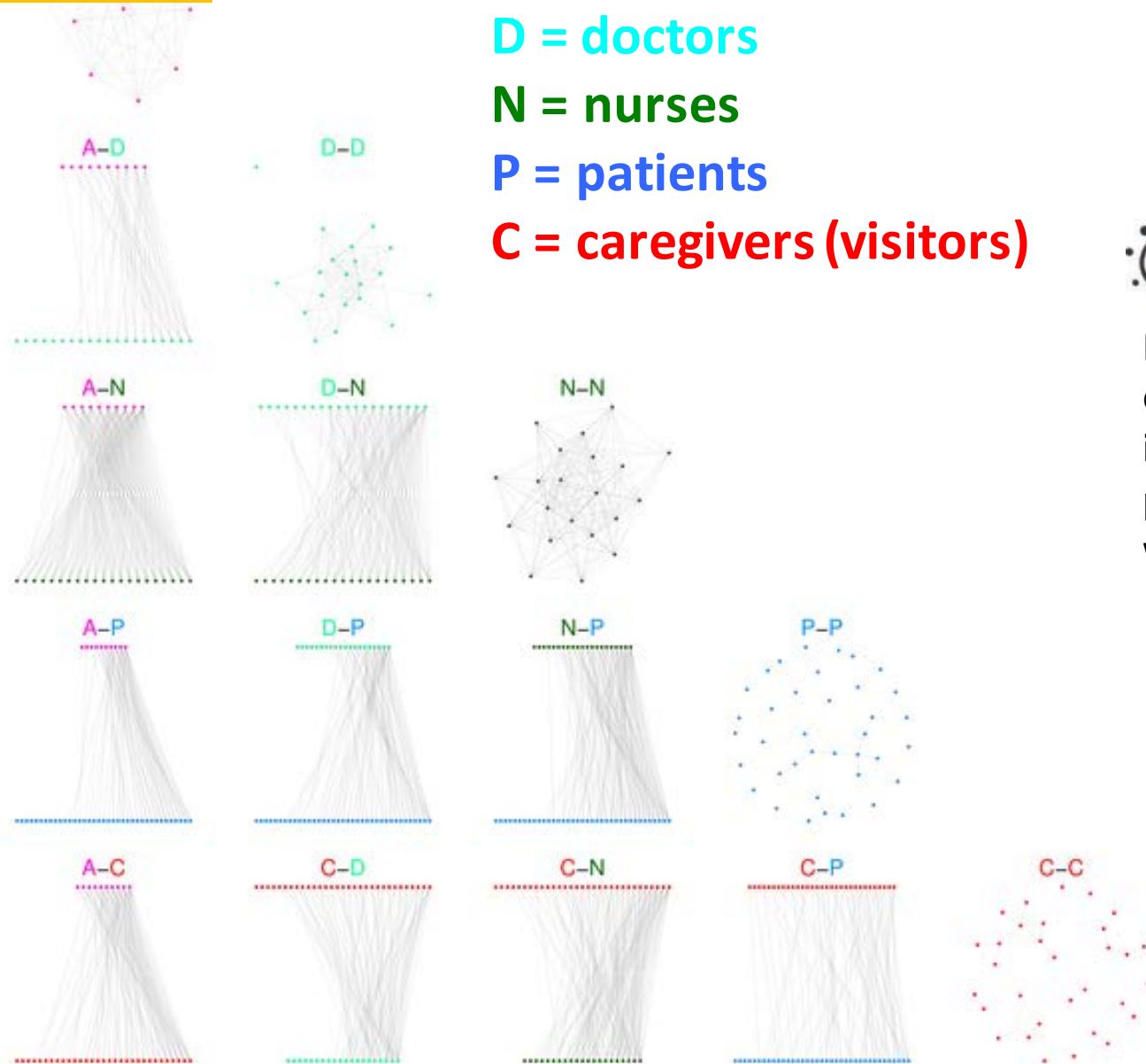


65 authors were classified as **unfavorable** to the drug (rosiglitazone does increase the risk of heart attack)



- Author identified as having a financial conflict of interest with manufacturers of rosiglitazone and/or other antihyperglycemic agents
- No financial conflict of interest

A-A
custom



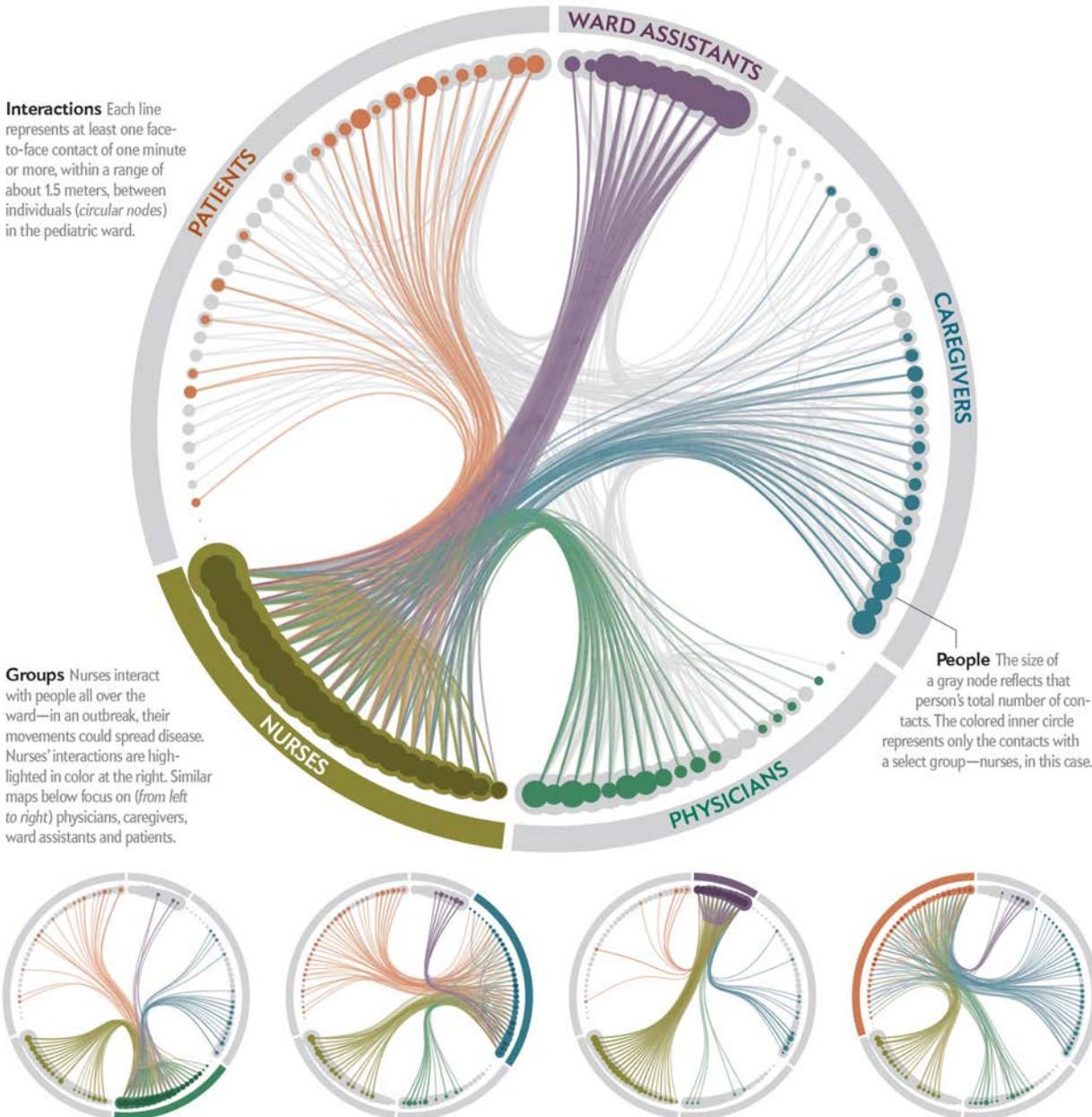
A = ward assistants
D = doctors
N = nurses
P = patients
C = caregivers (visitors)



Figure 6. Cumulative contact networks of individuals, for all pairs of classes and within each class.

By L. Isella et al.,
“Close Encounters in
a Pediatric Ward:
Measuring Face-to-
Face Proximity and
Mixing Patterns with
Wearable Sensors”

custom



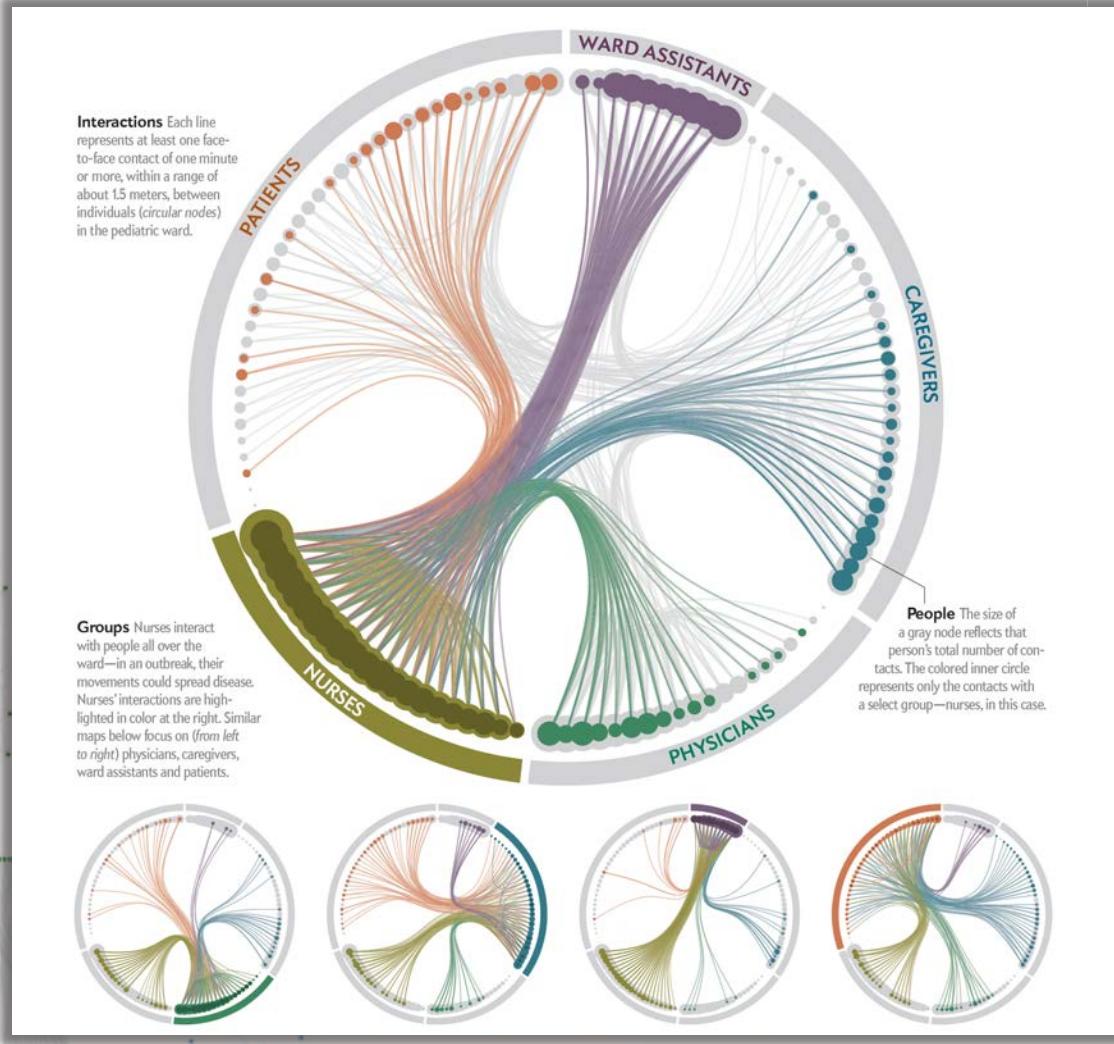
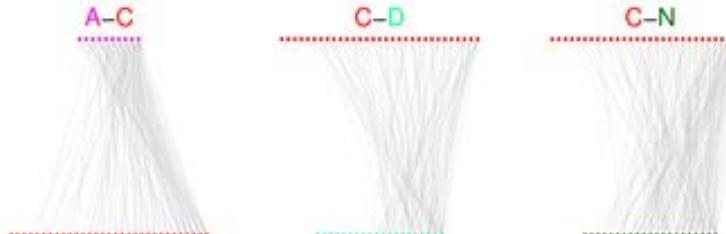
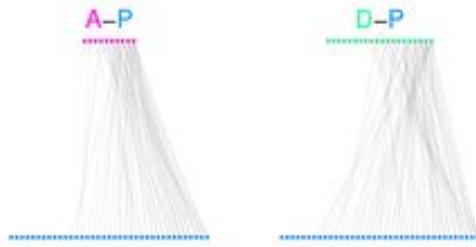
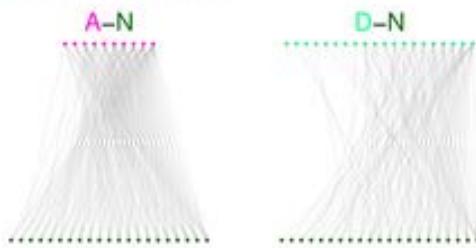
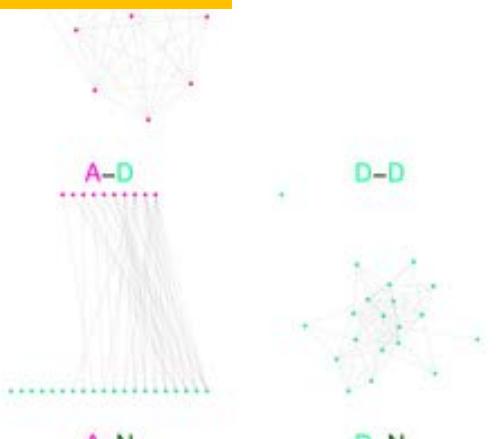
Tag—You're Sick

Patterns of personal contact in a hospital reveal true pathways of transmission

Jan Willem Tulp

A-A

custom



customizing for a different [context]

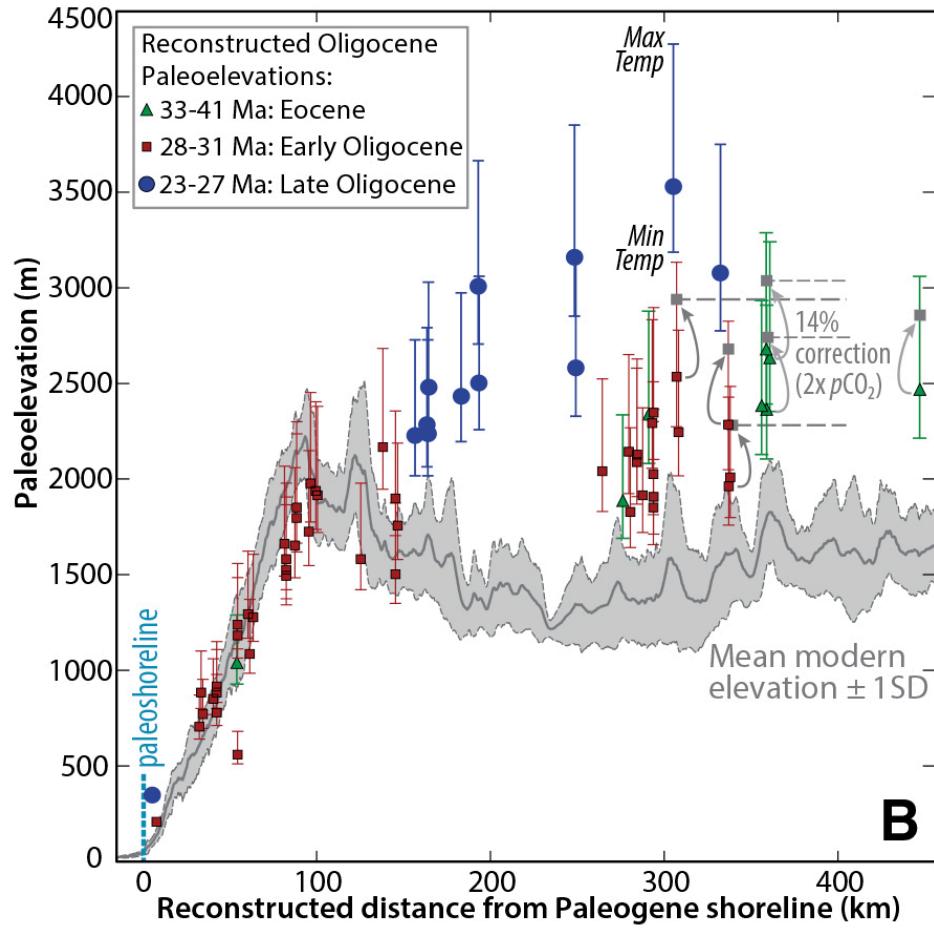
How to proceed?

customizing for a different [context]

How to proceed?

- build in time for many iterations

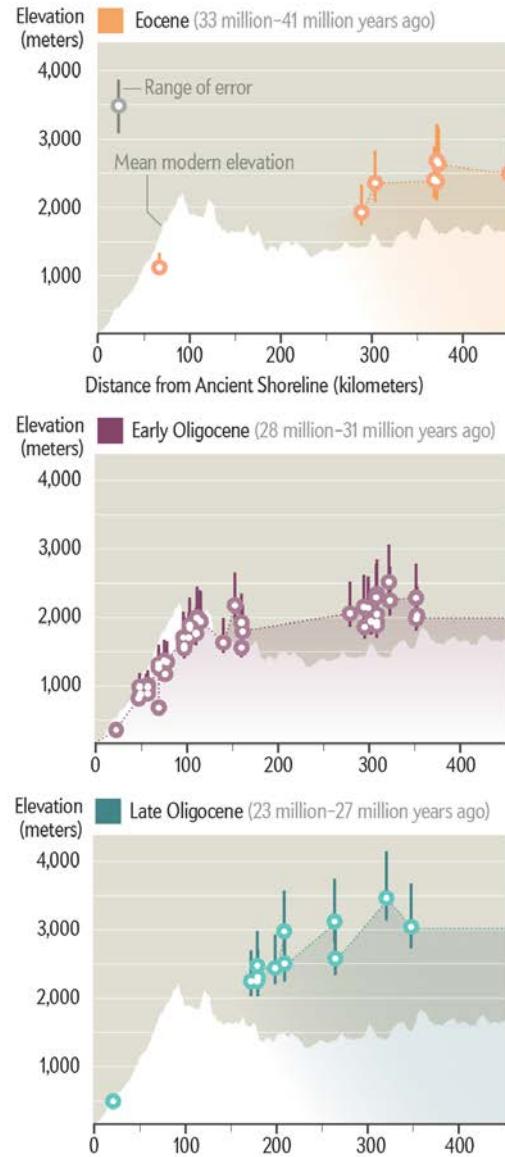
rework



By Elizabeth J. Cassel et al., "Profile of a paleo-orogen: High topography across the present-day Basin and Range from 40 to 23 Ma"

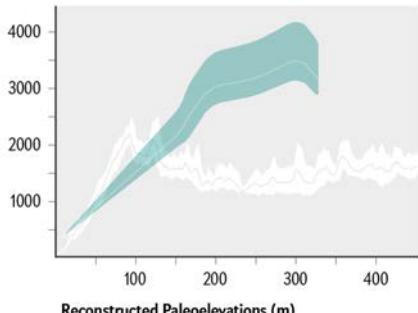
Nevada's Rocky Past

Reconstructed mountain scapes based on rainwater isotopes reveal the state's ups and downs



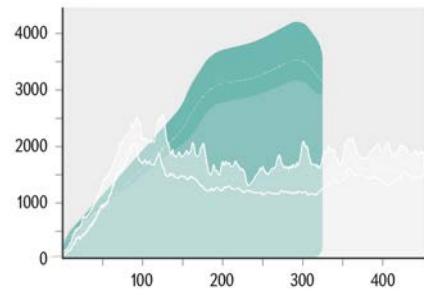
Reconstructed Paleoelevations (m)

■ Late Oligocene (23-27 million years ago)



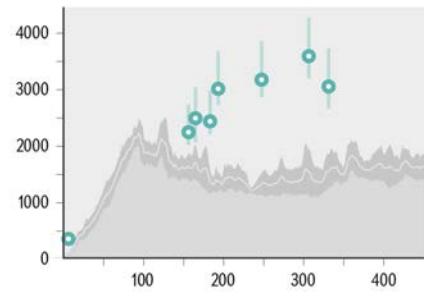
Reconstructed Paleoelevations (m)

■ Late Oligocene (23-27 million years ago)



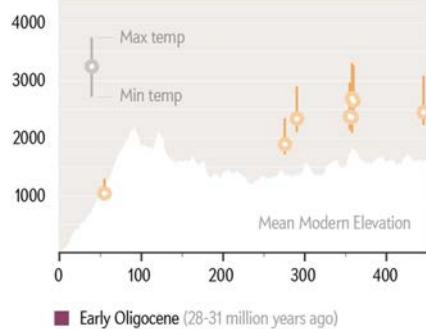
Reconstructed Paleoelevations (m)

■ Late Oligocene (23-27 million years ago)



Reconstructed Paleoelevations (m)

■ Eocene (33-41 million years ago)

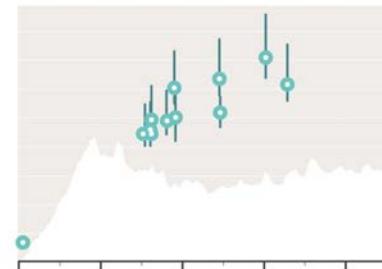


■ Early Oligocene (28-31 million years ago)

Mt. Whitney

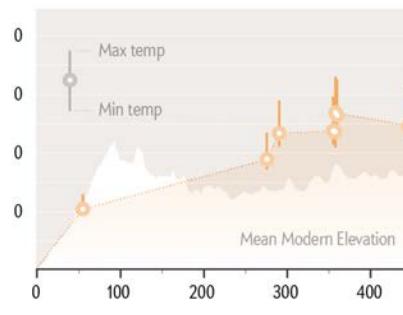
4,421m

■ Late Oligocene (23-27 million years ago)



Reconstructed Paleoelevations (m)

■ Eocene (33-41 million years ago)

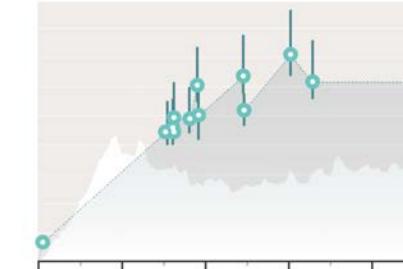


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4,421m

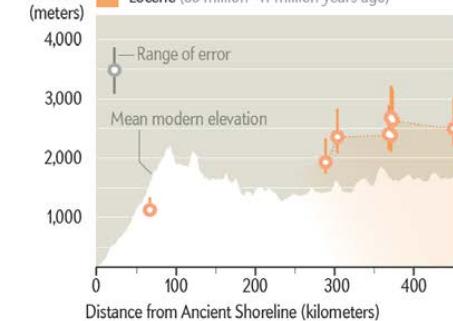
■ Late Oligocene (23-27 million years ago)



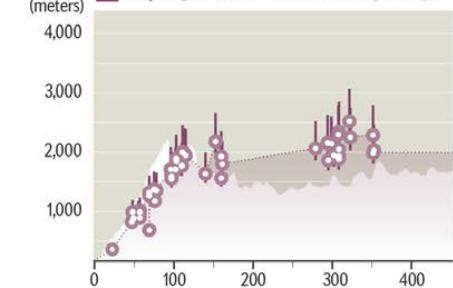
Nevada's Rocky Past

Reconstructed mountain scapes based on rainwater isotopes reveal the state's ups and downs

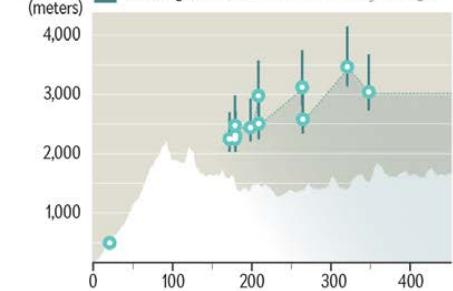
Elevation (meters) ■ Eocene (33 million-41 million years ago)



Elevation (meters) ■ Early Oligocene (28 million-31 million years ago)



Elevation (meters) ■ Late Oligocene (23 million-27 million years ago)



Tiffany Farrant-Gonzalez

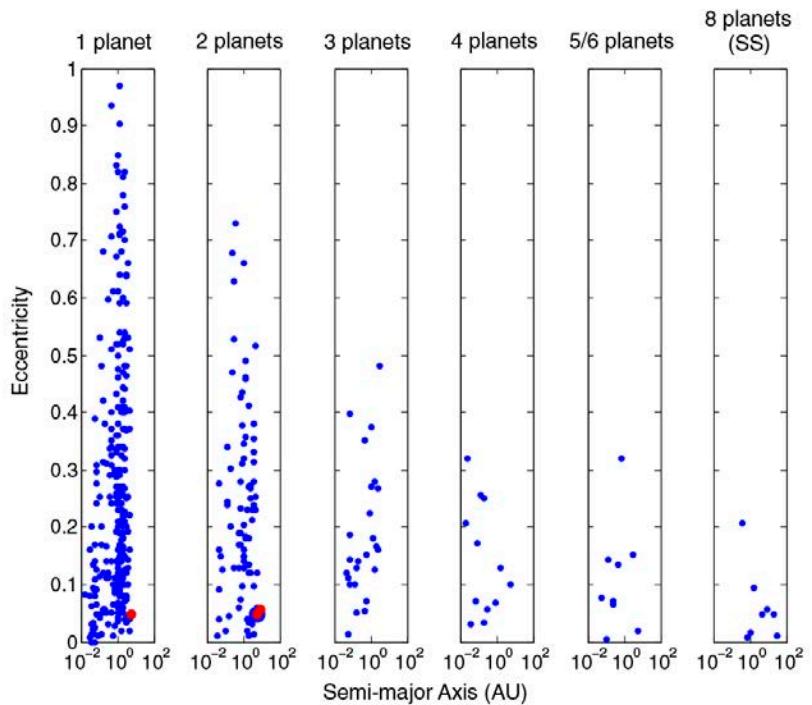
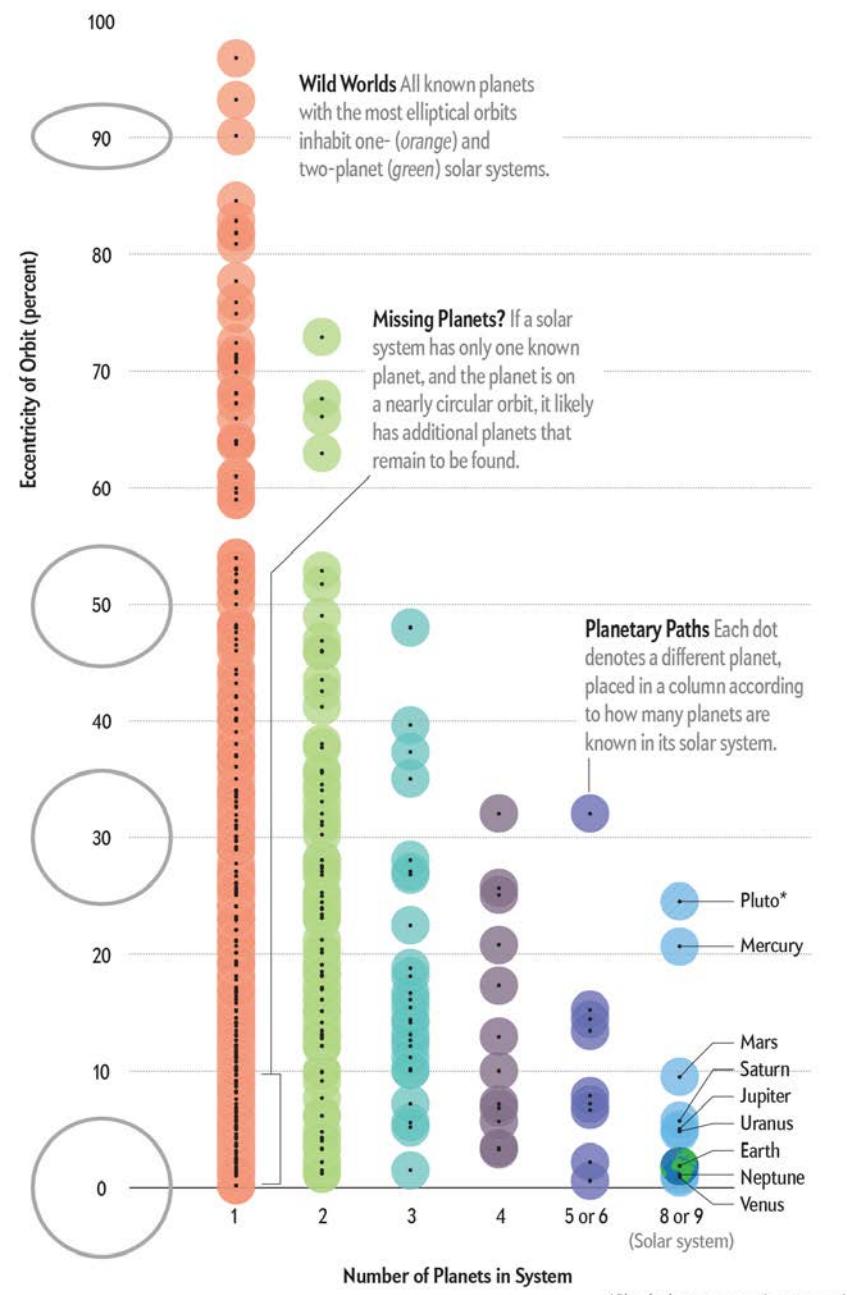


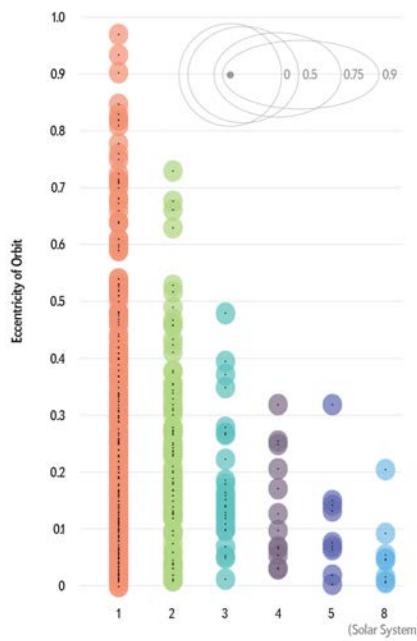
Fig. 2. Eccentricity versus semimajor axis going from low (left) to high multiplicity (right). A red dot is shown for where Jupiter would appear on the one- and two-planet distributions, and for Saturn on the two-planet distribution. This demonstrates that even if the Solar System were detected via RV as a one- or two-planet system, it would still be consistent with the data.

By Mary Anne Limbach and
Edwin L. Turner, "Exoplanet
Orbital Eccentricity: Multiplicity
Relation and the Solar System"

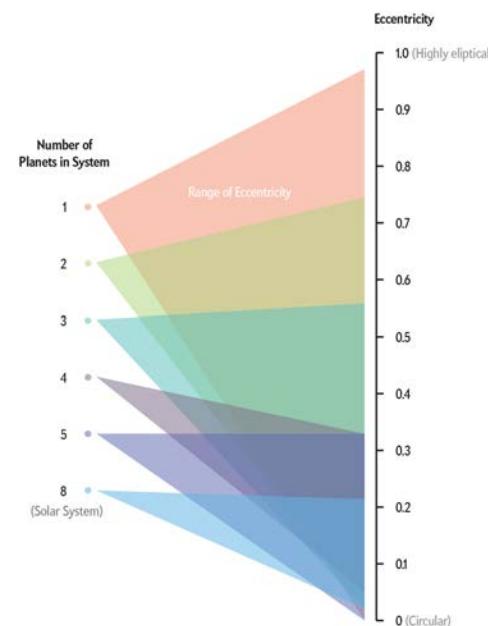


Tiffany Farrant-Gonzalez

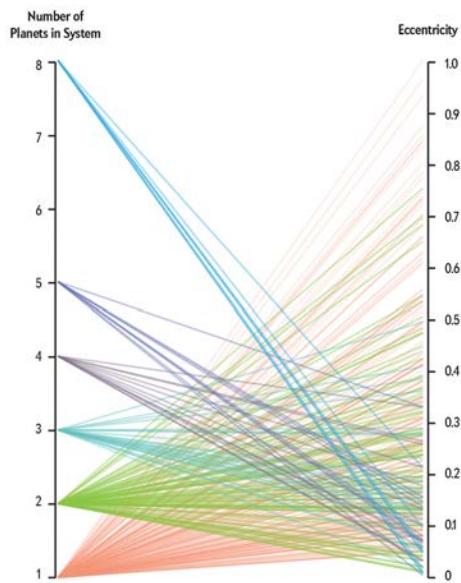
Mockup 1



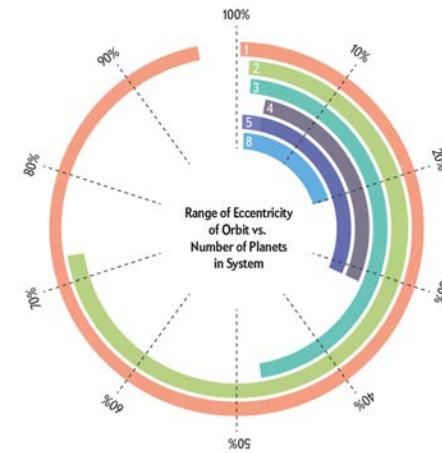
Mockup 2



Mockup 3



Mockup 4



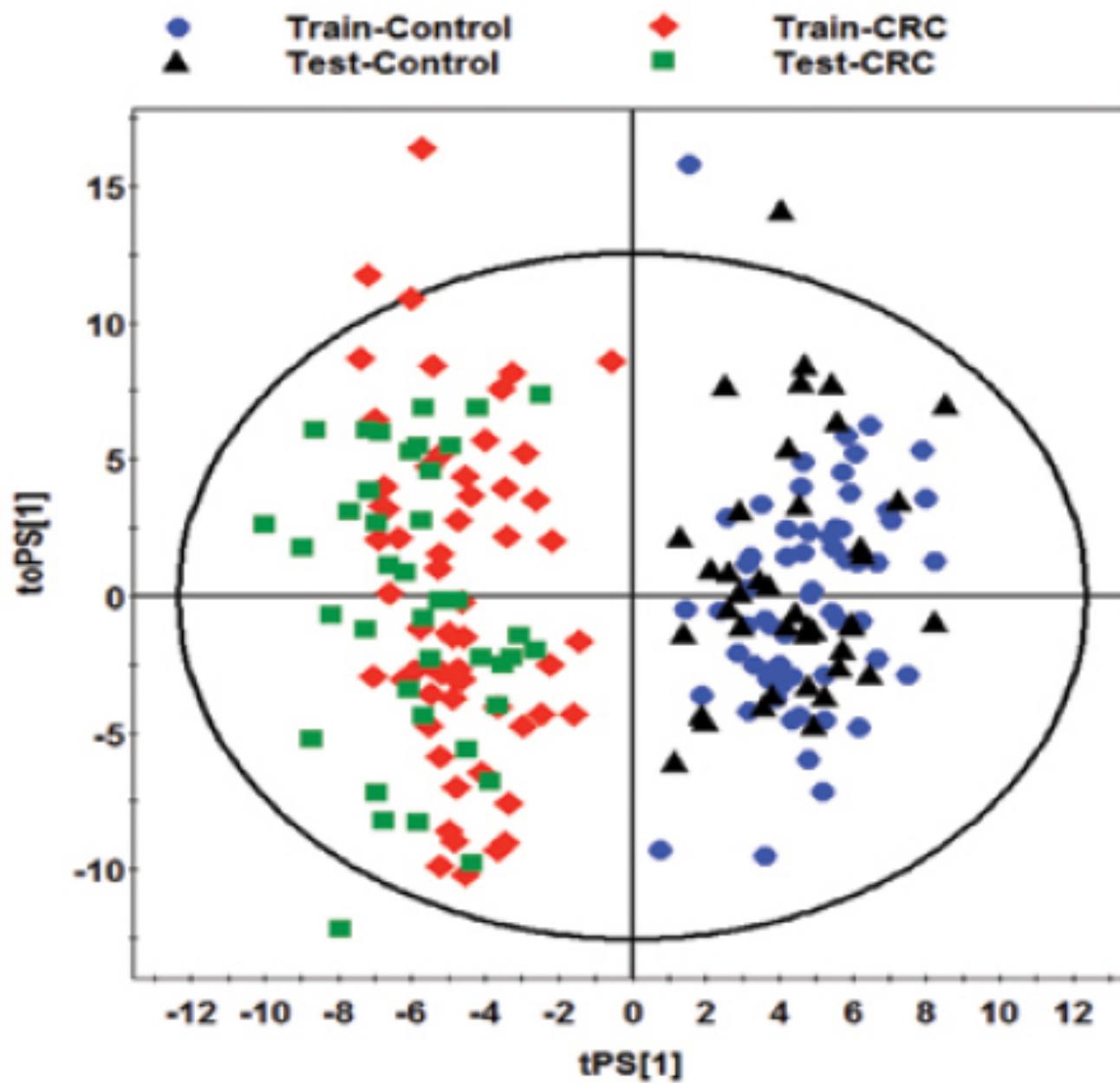
http://bit.ly/LSU_jc2

Tiffany Farrant-Gonzalez

customizing for a different [context]

How to proceed?

- build in time for many iterations
- look at your content with fresh eyes



By Yu Cheng et al., "Distinct Urinary Metabolic Profile of Human Colorectal Cancer"

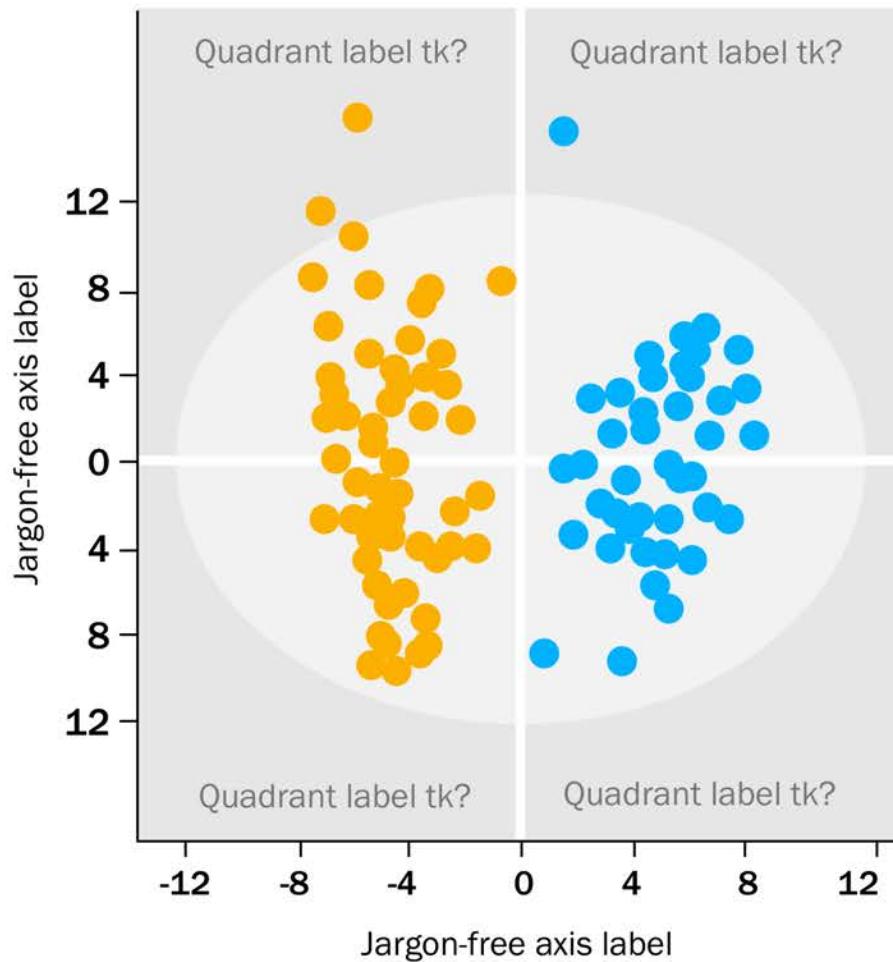
Detecting cancer through urine tests

Lorem ipsum text here.
Lorem ipsum text here. Lorem ipsum text here. Lorem ipsum text here. Lorem ipsum text here. Lorem ipsum text here.

Constructing the model

● Individuals with colorectal cancer

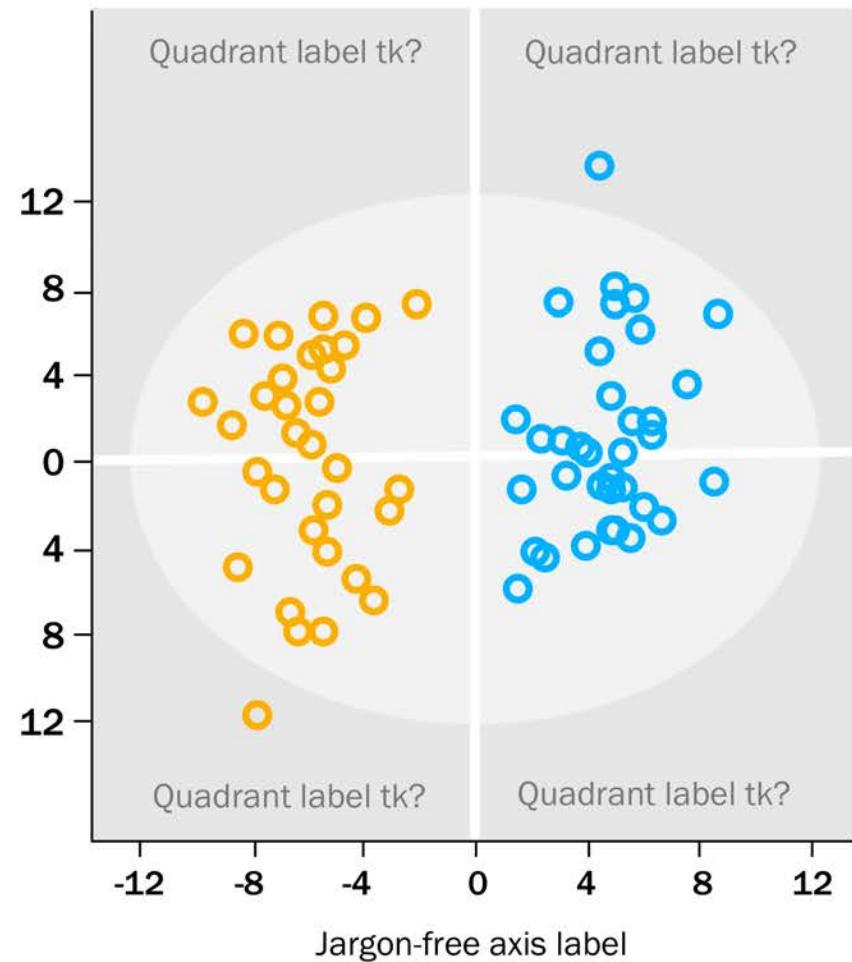
● Healthy individuals



Testing the model

● Individuals with colorectal cancer

● Healthy individuals



customizing for a different [context]

How to proceed?

- build in time for many iterations
- look at your content with fresh eyes
- remember to ask

“for whom and why?”

more resources

design-related SA blog posts

<https://blogs.scientificamerican.com/sa-visual/data-visualization-advice-for-scientists/>

<https://blogs.scientificamerican.com/sa-visual/how-to-choose-the-form-of-an-infographic-it-s-all-about-context/>

[https://blogs.scientificamerican.com/sa-visual/chart choosers](https://blogs.scientificamerican.com/sa-visual/chart-choosers)

<https://extremepresentation.com/design/7-charts/>

<https://policyviz.com/2014/11/11/graphic-continuum-desktop-version/>

<http://www.datavizcatalogue.com/index.html>

<http://stephanieevergreen.com/>

Excel-centric books

<http://www.storytellingwithdata.com/book/>

<https://excelcharts.com/>

other great books/websites

<http://www.thefunctionalart.com/>

<http://www.visualisingdata.com/>

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