## Storytelling with Data

Module 10: Infographics — Demonstration and discussion

#### Agenda

Upcoming deliverable

Today's objectives

More on uncertainty

Information graphics

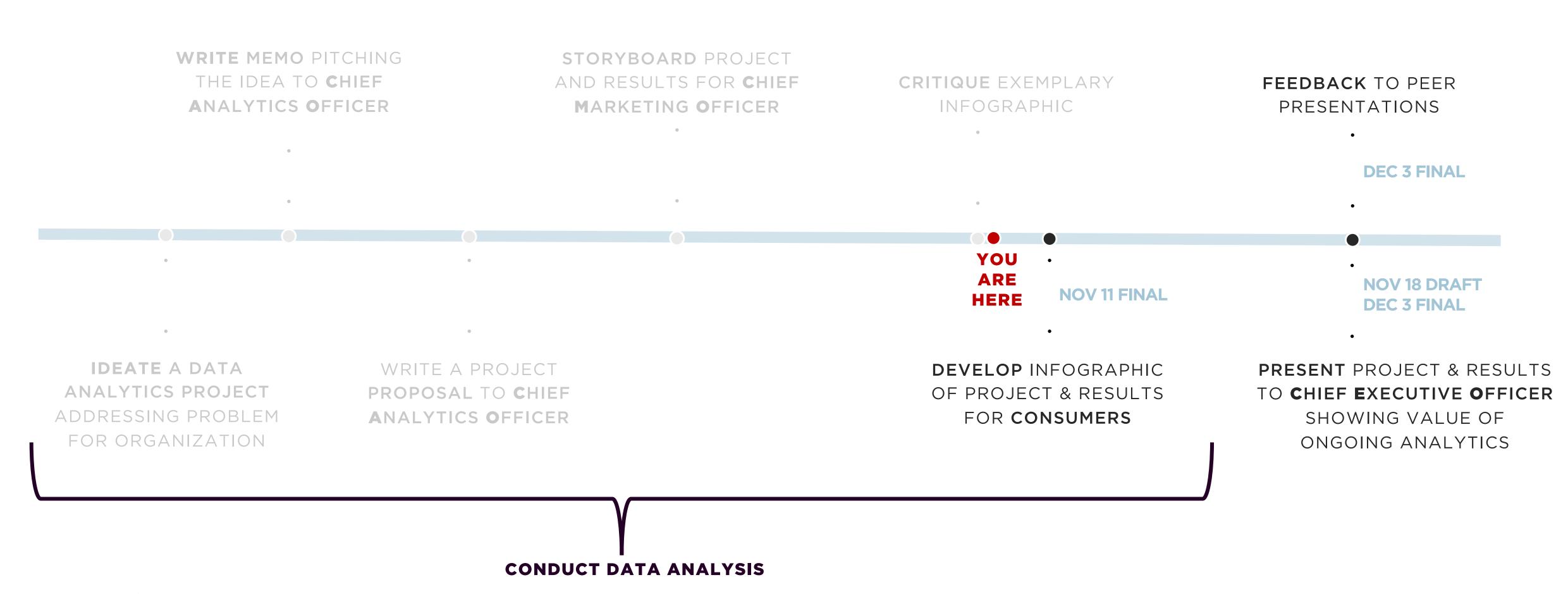
Storytelling with Data - Lecture 10

#### Questions or suggestions?

# Upcoming deliverables

#### Upcoming deliverables

**Information graphic** — reframe your story, this time building off the messages you built for the marketing team in order to craft an infographic that displays the results of the analytic work in a way that is accessible, engaging, and exciting for a **general or consumer audience**.



Storytelling with Data – Lecture 10

# Today's Objectives

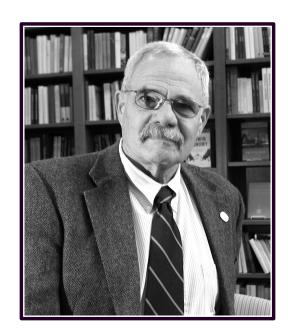
### Objectives

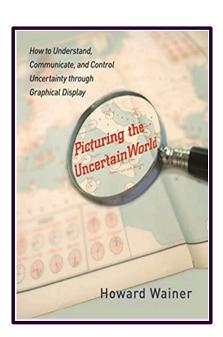
More on communicating uncertainty with data visualization

2 Discussion on infographics

Obtaining feedback to improve your infographic

## More on uncertainty





# Picturing the Uncertain World

#### Wainer

He is is an American statistician, past principal research scientist at the Educational Testing Service, adjunct professor of statistics at the Wharton School of the University of Pennsylvania, and author, known for his contributions in the fields of statistics, psychometrics, and statistical graphics.

#### The most dangerous equation

De Moivre's equation:

$$\sigma_{ar{x}} = rac{\sigma}{\sqrt{n}}$$
 :  $\sigma_{ar{x}} < \sigma$ 

σ the measure of the variability of a population (its standard deviation).

 $\sigma_{\overline{\chi}}$  the variation of averages of subsets of the population.

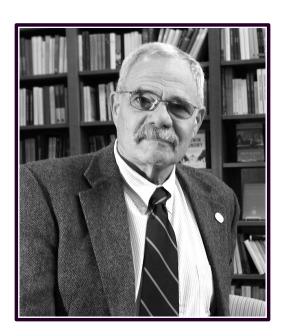
the number of observations in each subset

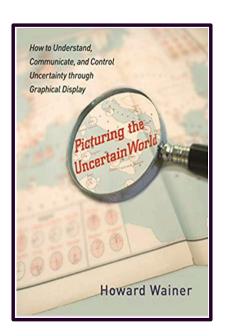
#### Why so dangerous?

Extreme length of time during which ignorance of it has caused confusion

Wide breadth of areas that have been misled

Seriousness of the consequences that ignorance has caused





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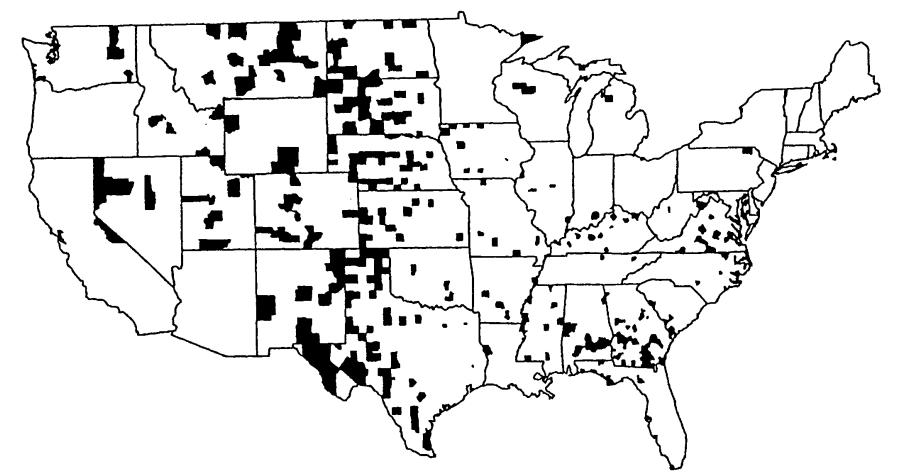


Figure 1.1.

Lowest kidney cancer death rates. The counties of the United States with the lowest 10% age-standardized death rates for cancer of kidney/urethra for U.S. males, 1980–1989 (from Gelman and Nolan, 2002, p. 15, reprinted with permission).

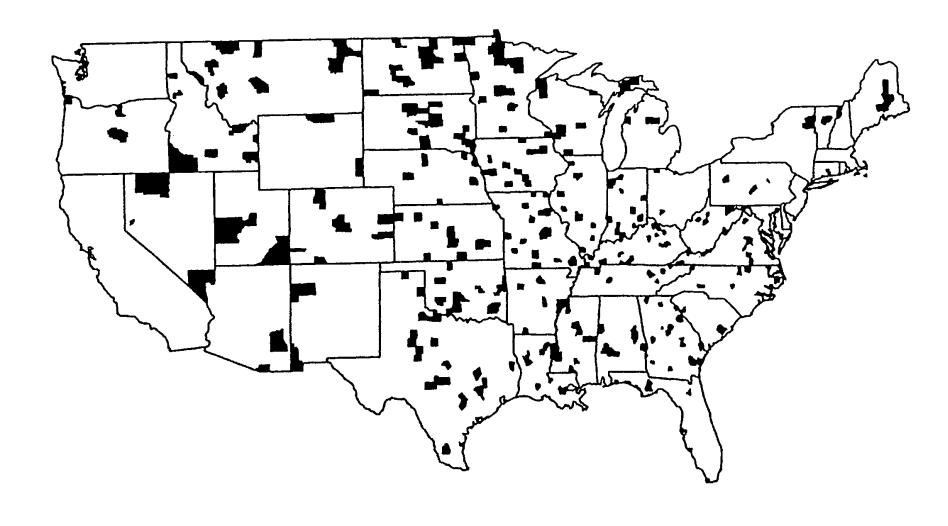
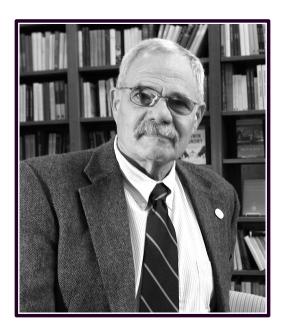
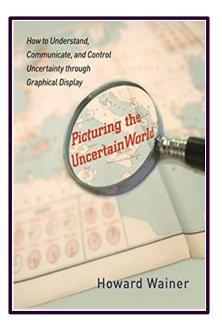


Figure 1.2.
Highest kidney cancer death rates. The counties of the United States with the highest 10% age-standardized death rates for cancer of kidney/urethra for U.S. males, 1980–1989 (from Gelman and Nolan, 2002, p. 14, reprinted with permission).

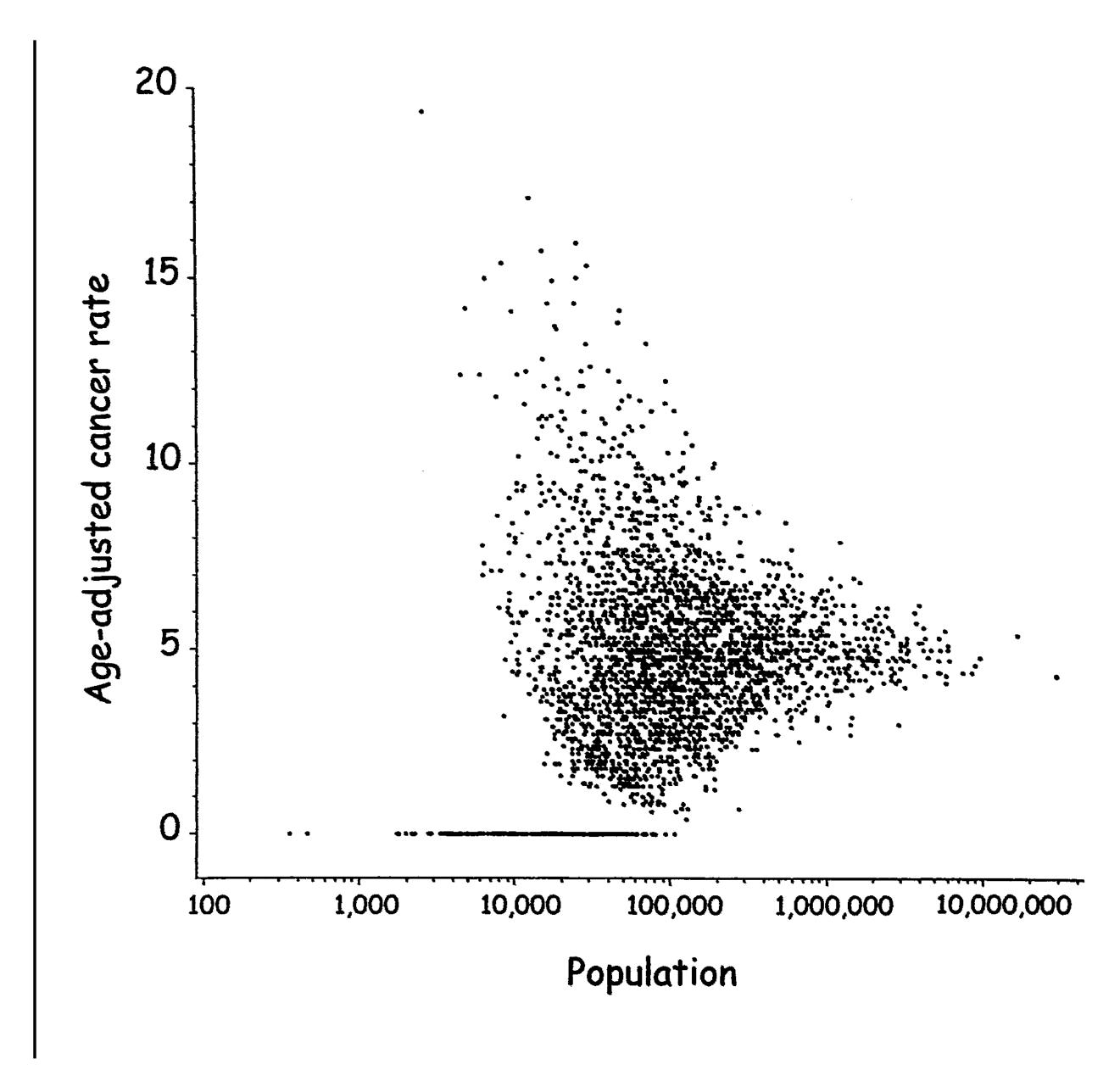




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# Communicating Uncertainty: Fulfilling the Duty to Inform

#### Fischhoff

He is a Professor in Department Engineering and Public Policy and Institute for Politics and Strategy at Carnegie Mellon University. A graduate of the Detroit Public Schools, he holds a BS in mathematics and psychology from Wayne State University and an MA and PhD in psychology from the Hebrew University of Jerusalem.

## Good decisions rely on knowledge of uncertainty

Scientists are often hesitant to share their uncertainty with decisionmakers who need to know it. With an understanding of the reasons for their reluctance, decisionmakers can create the conditions needed to facilitate better communication.

## Failure to express uncertainty has negative value

Communicating knowledge can worsen results if it induces unwarranted confidence or is so hesitant that other, overstated claims push it aside.

### Quantifying uncertainties aid verbal expression

Concern: people will misinterpret quantities of uncertainty, inferring more precision than intended. Response: Most people like getting quantitative information on uncertainty, from them can get the main message, and without them are more likely to misinterpret verbal expressions of uncertainty.

## Posing clear questions guide understanding

Concern: people cannot use probabilities. Response: laypeople can provide high-quality probability judgments, if they are asked clear questions and given the chance to reflect on them.

# Communicating uncertainty protects credibility

Concern: credible intervals may be used unfairly in performance evaluations. Response: probability judgments give us more accuracy about the information; i.e., won't be too confident or lack enough confidence.



When (ish) is My Bus? User-centered Visualizations of			
Uncertainty	in Everyday,	Mobile Predicti	ve Systems
Uncertainty Matthew Ky  CSE I dab  Daivenzipy of Washington registrations of CSE I dab  Daivenzipy of Washington registrations of the CSE I dab  ASSTRACT Uses often eye on scaline p the control of the control of the control to contro	Tara Kola  Tura Kola  Tura Unitaria Caracteria Caracter	Janeke R. Hallman (1988) and School (18th Sc	Sees A. Almanos.  ICIDE (dath)  University of Washington  summonifying of washington  commonifying the taining, are re- commonifying the taining, are re-  trees into).  In operating for taining, are re-  trees into).  In operating the taining, are re-  trees into).  In operating the re-  trees into a back of predicted of arrivers in the control of the control  in the control of the control into a the cortice, and any are re-  trees are r
Amount of the property of the		pex visual representations of procountry unstreamers, rec example, error burs and probability densities require prior experience with statistical models to correctly interpret [2,6]. People can better understand probabilistic infor- mation when it is framed in terms of discrete events. For internal Medican & Giorgan II-16 forms that promoted	

#### When (ish) Is My Bus?

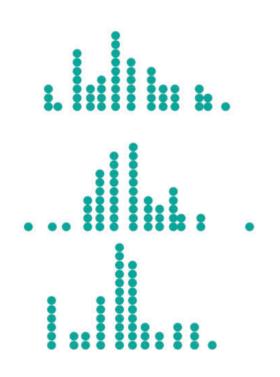
#### Kay

He is an Assistant Professor of Information at UMSI, and works in human-computer interaction and information visualization.

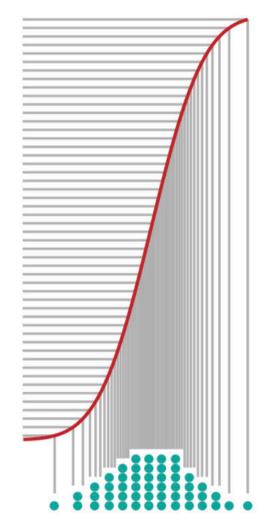
#### **Probability density of Normal distribution**



To generate a discrete plot of this distribution, we could try taking random draws from it. However, this approach is noisy: it may be very different from one instance to the next.



Instead, we use the quantile function (inverse CDF) of the distribution to generate "draws" from evenly-spaced quantiles.



We plot the quantile "draws" using a Wilkinsonian dotplot, yielding what we call a quantile dotplot: a consistent discrete representation of a probability distribution.



By using quantiles we facilitate interval estimation from frequencies: e.g., knowing there are 50 dots here, if we are willing to miss our bus 3/50 times, we can count 3 dots from the left to get a one-sided 94% (1 - 3/50) prediction interval corresponding to that risk tolerance.

# Transparency in statistical communication

# Transparent Statistics: guiding principles

#### Wacharamanotham, et al

The authors are part of a special interest group working to improve transparency in statistical communications.

The special interest group's "guidelines aim to advance a vision of transparent statistical communication for the field. Whatever the methods used, we can at least provide guidance that makes the communication of those methods more transparent, that makes reproduction and replication of work easier, and that makes evaluation of work (e.g., by peer reviewers) easier and more fair."

#### Nine principles

faithfulness to the data and phenomena studied,

robustness to departures from statistical assumptions,

resilience to statistical noise,

full explanations of analysis processes and reporting strategies,

clarity and accessibility of study reports,

preference for simplest analysis procedures,

avoidance of analysis decisions that are contingent on data,

pursuit of statistical power and precision, and

making study material available.

# Group work



#### We want information graphics to ...

Tell a complete story where the purpose is to inform, entertain or persuade the audience. It should:

simple, focused messages new, surprising information credible data sources

visually coherent, integrated use comparisons for context, meaning principles of information design, organized

#### Let's look ahead

# For Next Week, Module 11:

#### Agenda next week

Next deliverables, *final* information graphic **AND** critique of an infographic you select to share.

Combining written and visual communication of analytics with verbal delivery.

#### The minimum

Doumont, Jean-Luc. *Effective Oral Presentations*, in *Trees, Maps, and Theorems*. Principiæ, 2009. Print.

Consider his process of preparing an oral presentation and how oral presentation shares complements written and visual communication.

Tufte, Edward R. *The Cognitive Style of PowerPoint: Pitching Out Corrupts Within*, in *Beautiful Evidence*. Graphics Press, 2006. Print.

Think about the issues he raises of information presented in the form of projected slides.

Duarte, Nancy. *There's always room to improve*, in Chp. 8, *Resonate: Present Visual Stories That Transform Audiences*. Wiley, 2010. Print.

Consider how she organizes and designs the visual parts of a presentation, and its relation to oral delivery.

Schwabish, Jonathan. "Better Presentations: a Guide for Scholars, Researchers, and Wonks." (2016): 1–200. Print.

Skim for presentation best practices.

#### Pay it forward

#### Next steps?

Did you enjoy choosing your own project?

How might you plan to deepen the analysis in a way you'd like to showcase in your portfolio?

# Letter to your future colleagues?

Having had the experience of taking this course, what advice do you have for your future colleagues about to begin this course? In other words, what would you want your younger self to know?



See you next week!