

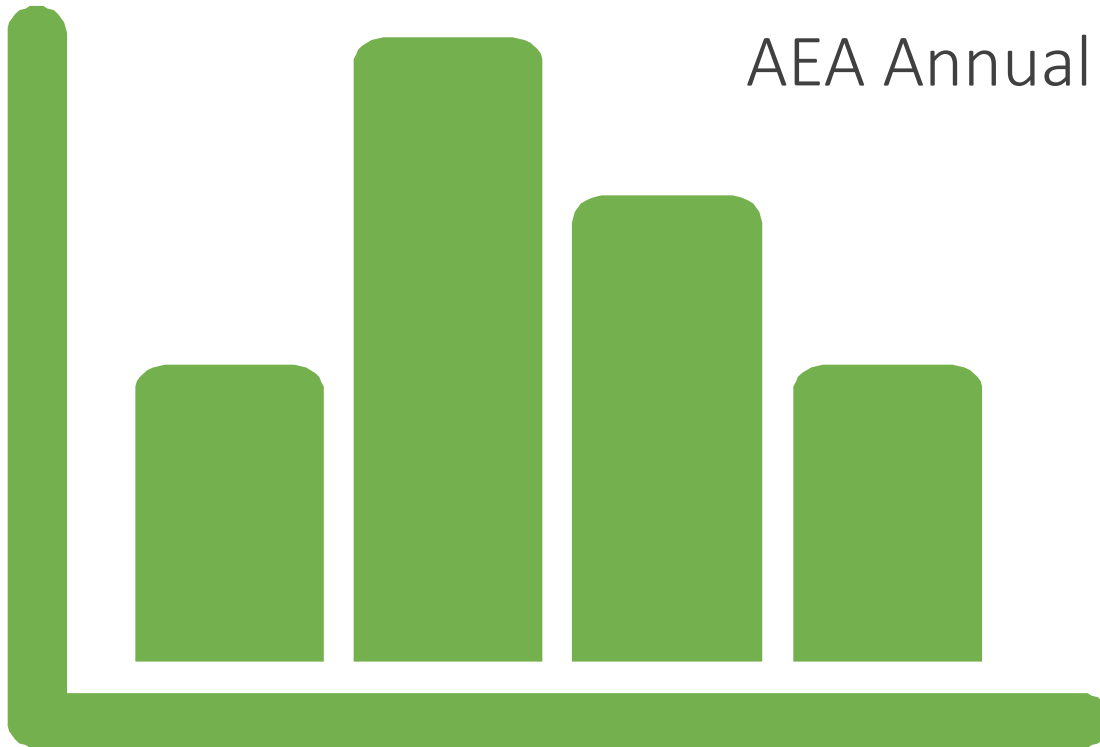
Does it make a difference?

The impact (or not) of data
visualization on evaluation use



Sena Sanjines

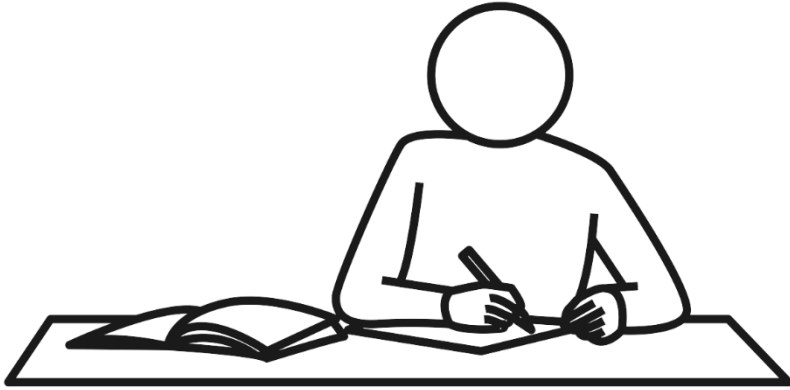
AEA Annual Conference 2018



The problem

We want to avoid this.





Not the people who
read our reports



The people who read
our reports

Does use of data
visualizations increase the
likelihood reports will be
used?



Does the quality of data visualizations increase the likelihood reports will be used?

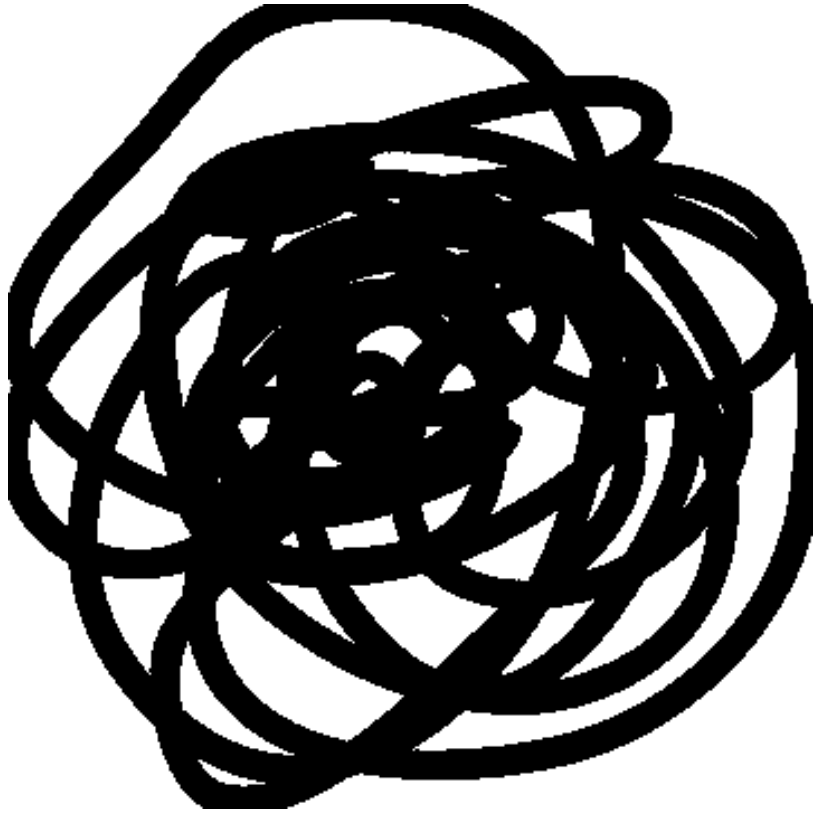


What are data visualizations?

Images based on data that are representative of the data, readable, and support exploration, examination, and communication of the data.
(Azzam, Evergreen, Germuth, & Kristler, 2013, p.9)



The context



Use is messy

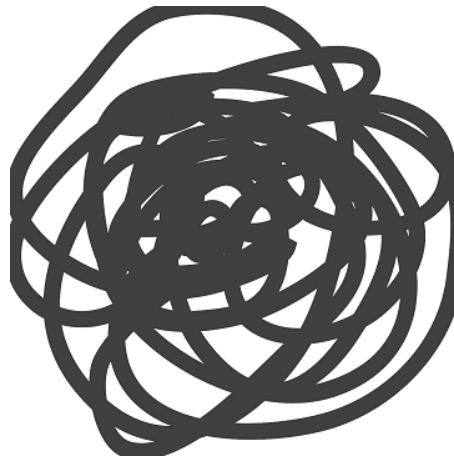


But not in my study!



You were referenced?
Great!
You count as “used”

- ✓ Length of report
- ✓ Type of report (credibility)
- ✓ User affiliation with a university

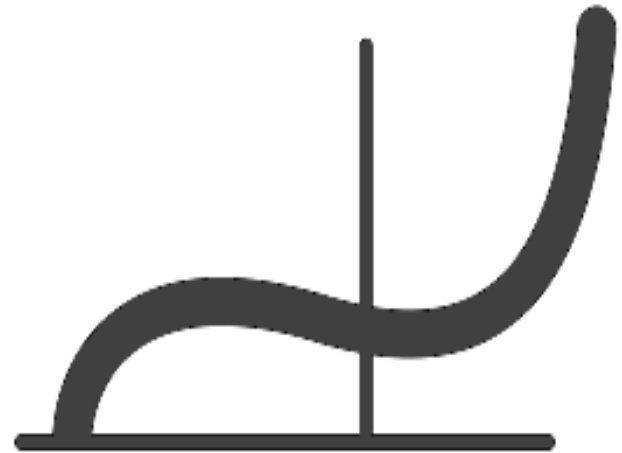


Methodology

Study design



Part 1



Part 2

The Data



Reports submitted as part of legislative testimony between 2000 and 2015 related to teacher quality.*

Report Sample ($N=215$)

Use once	169 (79%)
Used more than once	46 (21%)

*Shared by Reckhow and Tompkins-Stange (2015)

The Variables

Viz:	Percent of data viz to pages
Quality:	Data Visualization Checklist*
Length:	Number of pages
Type:	More like traditional research or advocacy research
Affiliation:	Witness worked for a university

*Evergreen & Emery (2016)

Type of Report

Traditional

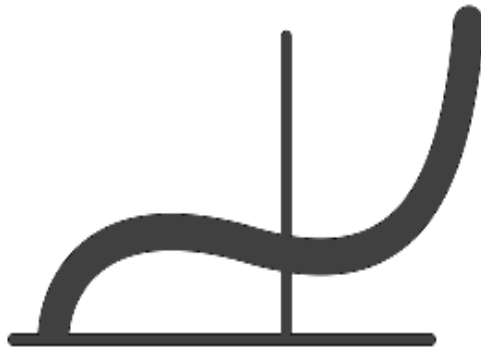
Advocacy



- ✓ Producer
- ✓ Conclusions
- ✓ Policy recommendations
- ✓ Production quality

- ✓ Evidence
- ✓ Citations
- ✓ Tone
- ✓ Methods

Analysis



Analysis

Poisson regression x 2

1. Full sample ($N=215$)
Percent of data viz + length, type, and user affiliation
2. Sub-sample ($N=93$)
DVC score + length, type, and user affiliation

Results

Does use of data
visualizations increase the
likelihood reports will be
used?



Does use of data
visualizations increase the
likelihood report findings will
be used?

No

Parameter Estimates for Data Visualizations ($N=215$)

Parameter	B	SE	Hypothesis Test			
			Wald			
			Chi-Square	df	Sig.	Exp(B)
(Intercept)	-1.043	0.219	22.656	1	0	0.352
Visualizations	0.113	0.100	1.274	1	0.259	1.120
Length	0.000	0.115	0.000	1	0.997	1.000
Type	0.247	0.131	3.567	1	0.059	1.280
Affiliation = 0	-0.167	0.271	0.379	1	0.538	0.846
Affiliation = 1	0 ^a					1

a. Fixed at the displayed value.

Does the quality of data visualizations increase the likelihood reports will be used?



Does the quality of data visualizations increase the likelihood report findings will be used?

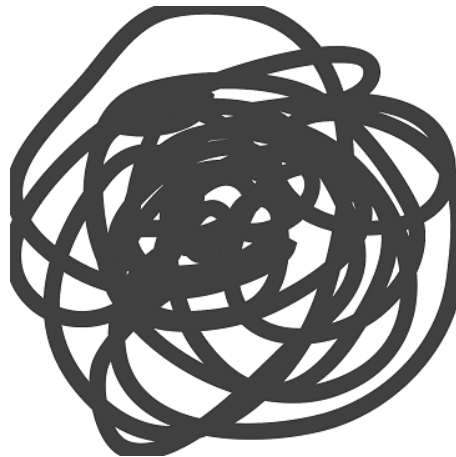
No

Parameter Estimates for Quality of Data Viz (N=93)

Parameter	B	SE	Hypothesis Test			
			Log-L			
			Chi-Square	df	Sig.	Exp(B)
(Intercept)	-0.94	0.34	45.02	1	0.00	0.39
DVC	0.10	0.18	0.31	1	0.58	1.11
Length	-0.51	0.28	4.37	1	0.04	0.60
Type	0.3	0.20	3.10	1	0.05	1.48
Affiliation = 0	-0.38	0.41	0.80	1	0.38	0.67
Affiliation = 1	0 ^a					1

a. Fixed at the displayed value.

- ✓ Length of report
- ✓ Type of report (credibility)
- ✓ User affiliation with a university



- ✓ Length of report
- ✓ Type of report (credibility)

Yes

Parameter Estimates for Data Visualizations ($N=215$)

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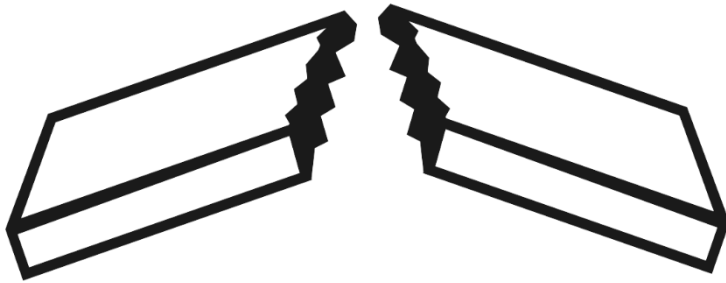
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Considerations and limitations

Considerations



Breaks with prior



Political setting

Limitations



Political context
Frequency of use
Other factors out there

What does it mean?

The takeaway



(www.forhumanpeoples.com)

Mahalo!

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References

Azzam, T., Evergreen, S., Germuth, A. A., & Kistler, S. J. (2013). Data visualization and evaluation. In T. Azzam & S. Evergreen (Eds.), *Data visualization, part 1: New Directions for Evaluation*, 139, 7—32.
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Evergreen, S. & Emery, A. K. (2016). *The data visualization checklist*. Retrieved from http://stephanieevergreen.com/wp-content/uploads/2016/10/DataVizChecklist_May2016.pdf

Reckhow, S., Holden, L., & Tompkins-Stange, M. (2015). *Patron of ideas: How advocacy research influences the education policy debate*. Paper presented at the American Political Science Association Annual Meeting, September 2015. San Francisco, CA