

# An Algebraic Process for Visualization Design

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# Why theory?

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Theory is the foundation.

Fundamental properties of what visualization is, and how it works.

Many Vis Theory papers are about taxonomies.

# Our basic idea

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Rethink theory of vis design. Not about:

“Dataset is X, so vis should be Y”

but rather:

“We can X the data; can we Y the image?”

Our design approach studies **changes**

in data to be visualized

in images produced by visualization

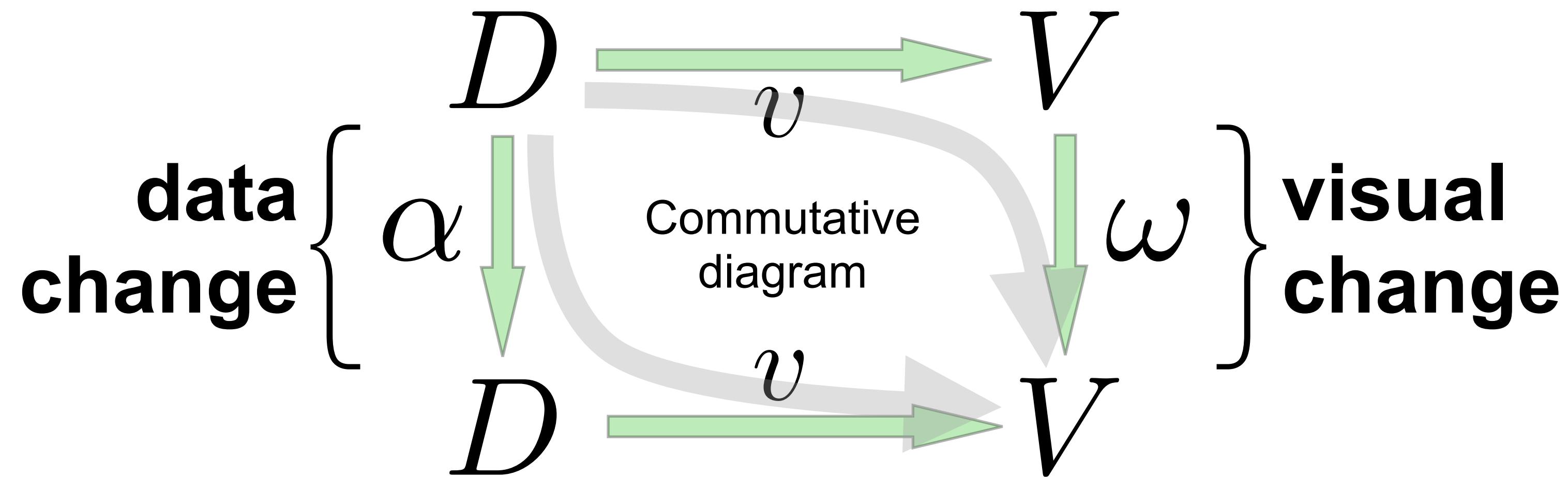
# The basic design question

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“Are important data changes  
well-matched  
with obvious visual changes?

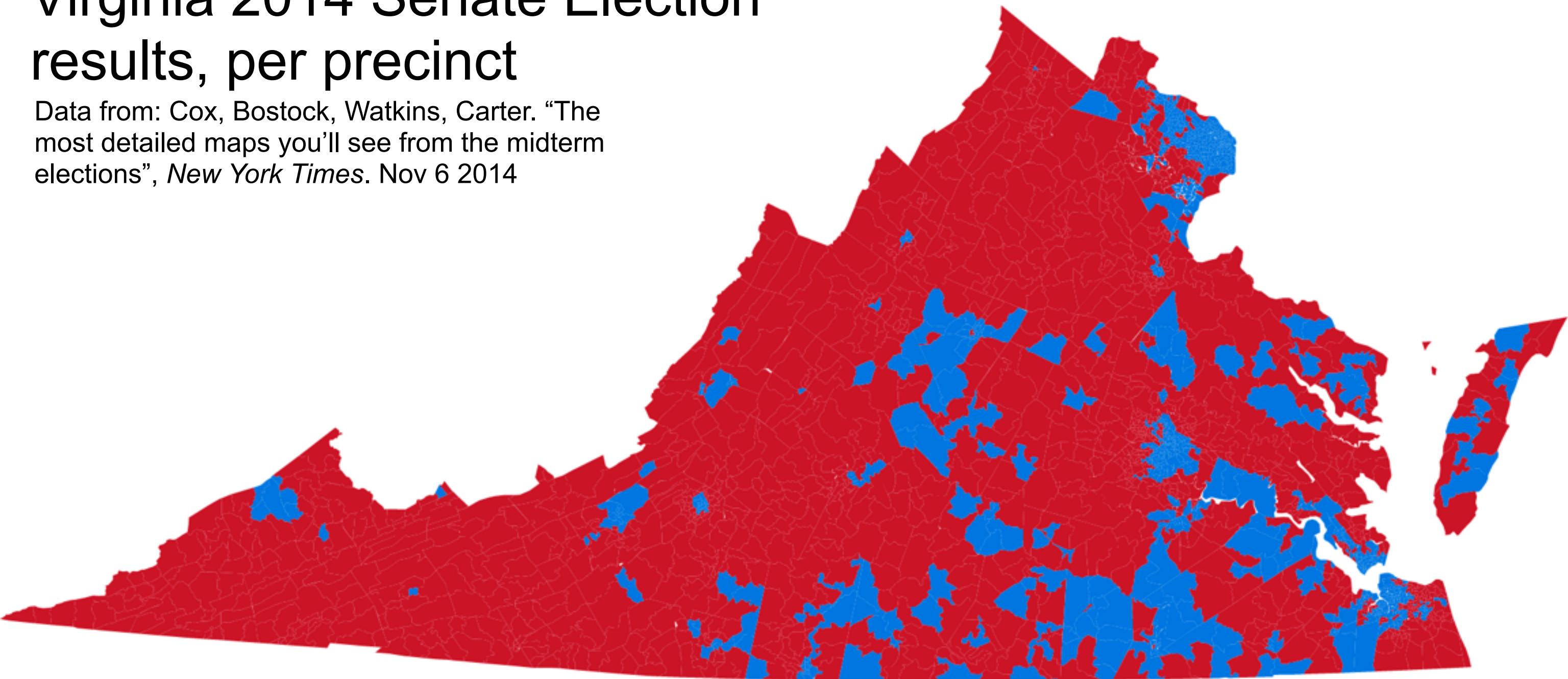
# Algebraic visualization design

Contribution: a math vocabulary for targeted questions about a vis method, and for understanding why one method might be better, for a certain task



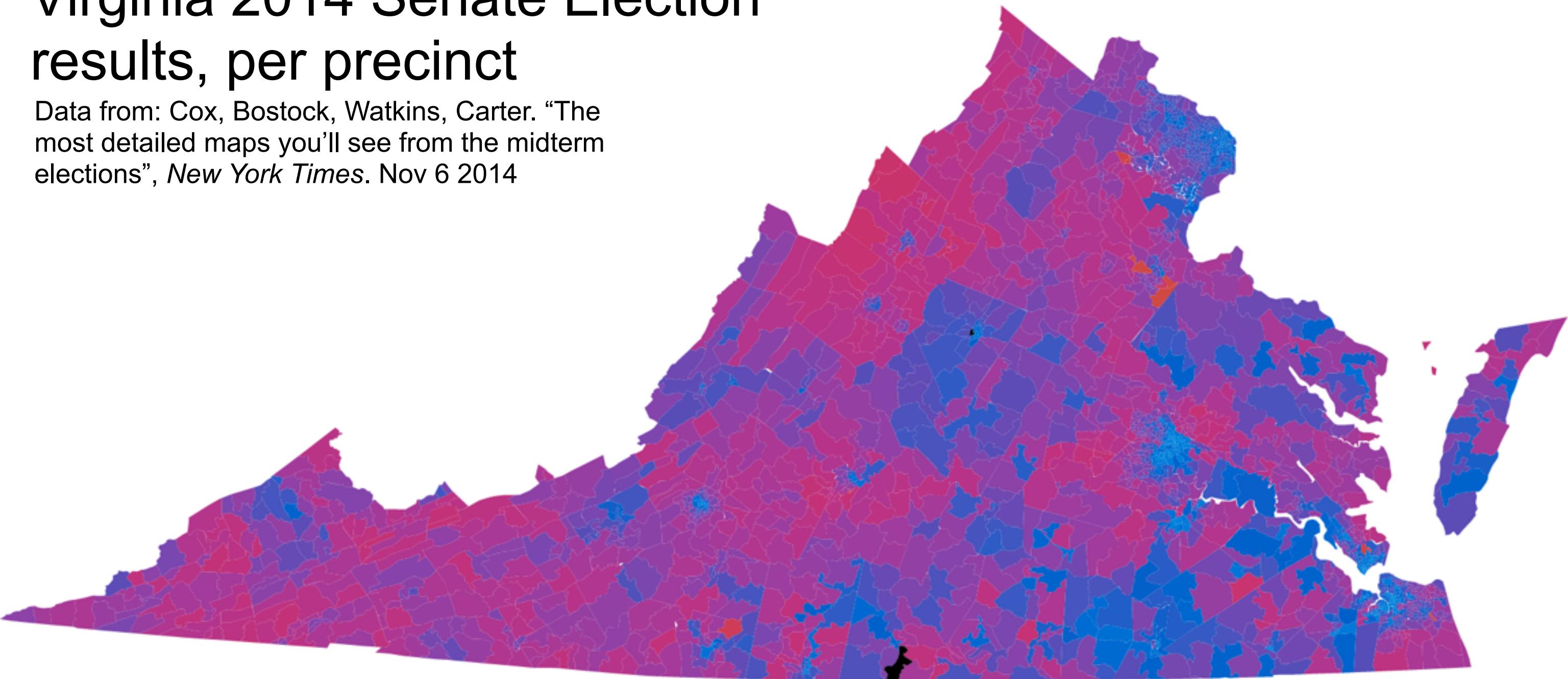
# Virginia 2014 Senate Election results, per precinct

Data from: Cox, Bostock, Watkins, Carter. "The most detailed maps you'll see from the midterm elections", *New York Times*. Nov 6 2014



# Virginia 2014 Senate Election results, per precinct

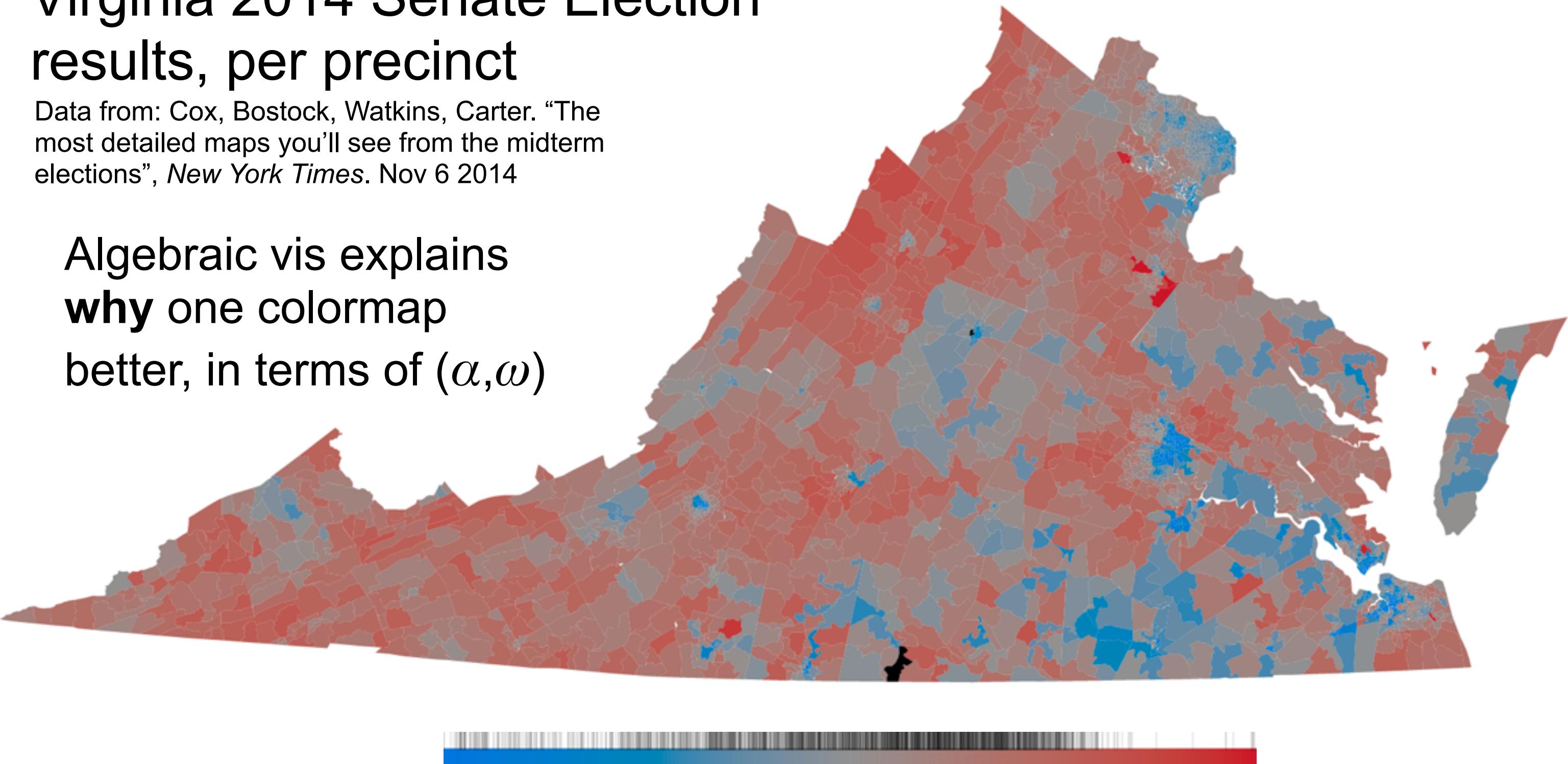
Data from: Cox, Bostock, Watkins, Carter. "The most detailed maps you'll see from the midterm elections", *New York Times*. Nov 6 2014



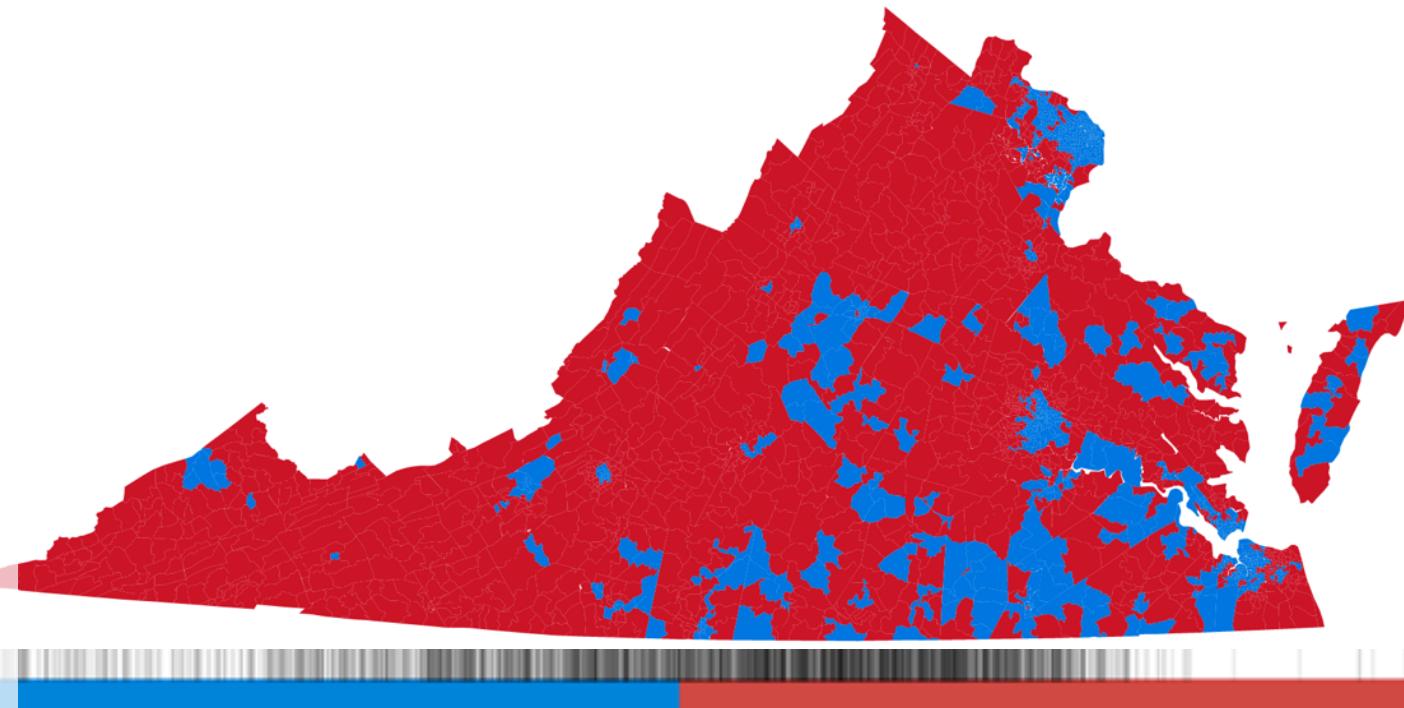
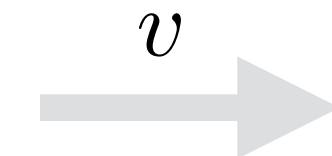
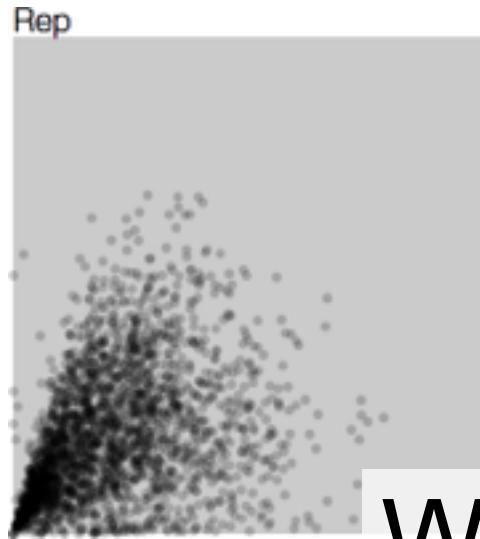
# Virginia 2014 Senate Election results, per precinct

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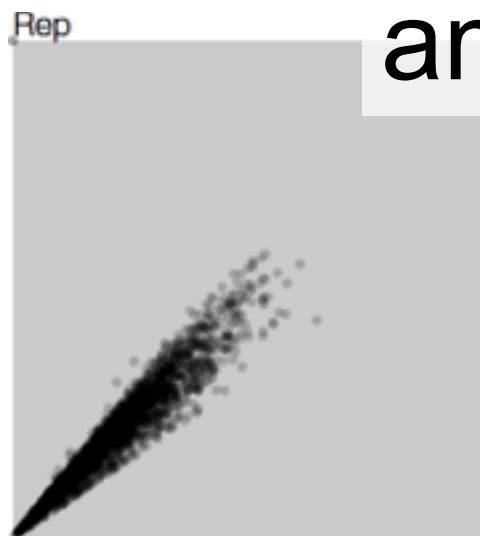
Algebraic vis explains  
why one colormap  
better, in terms of  $(\alpha, \omega)$



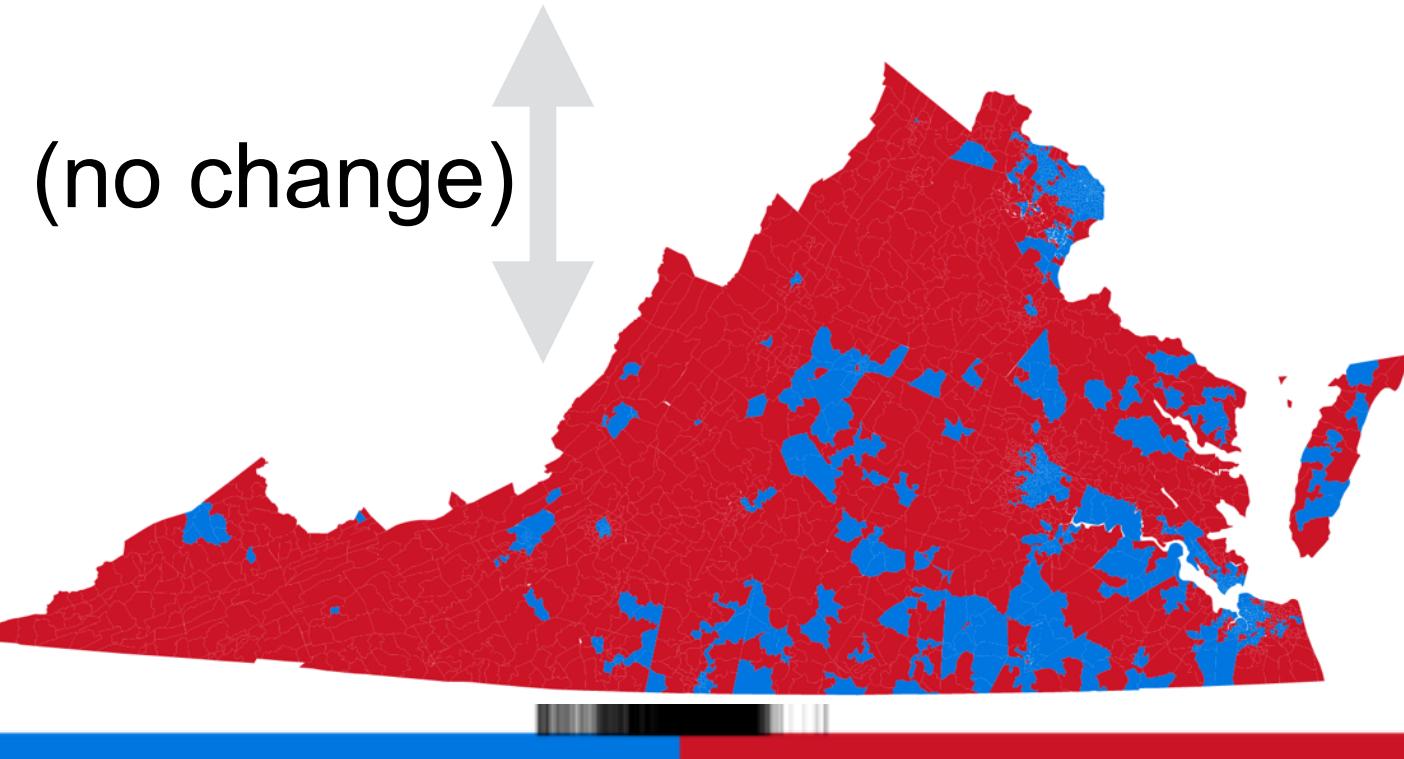
# Let's change the data



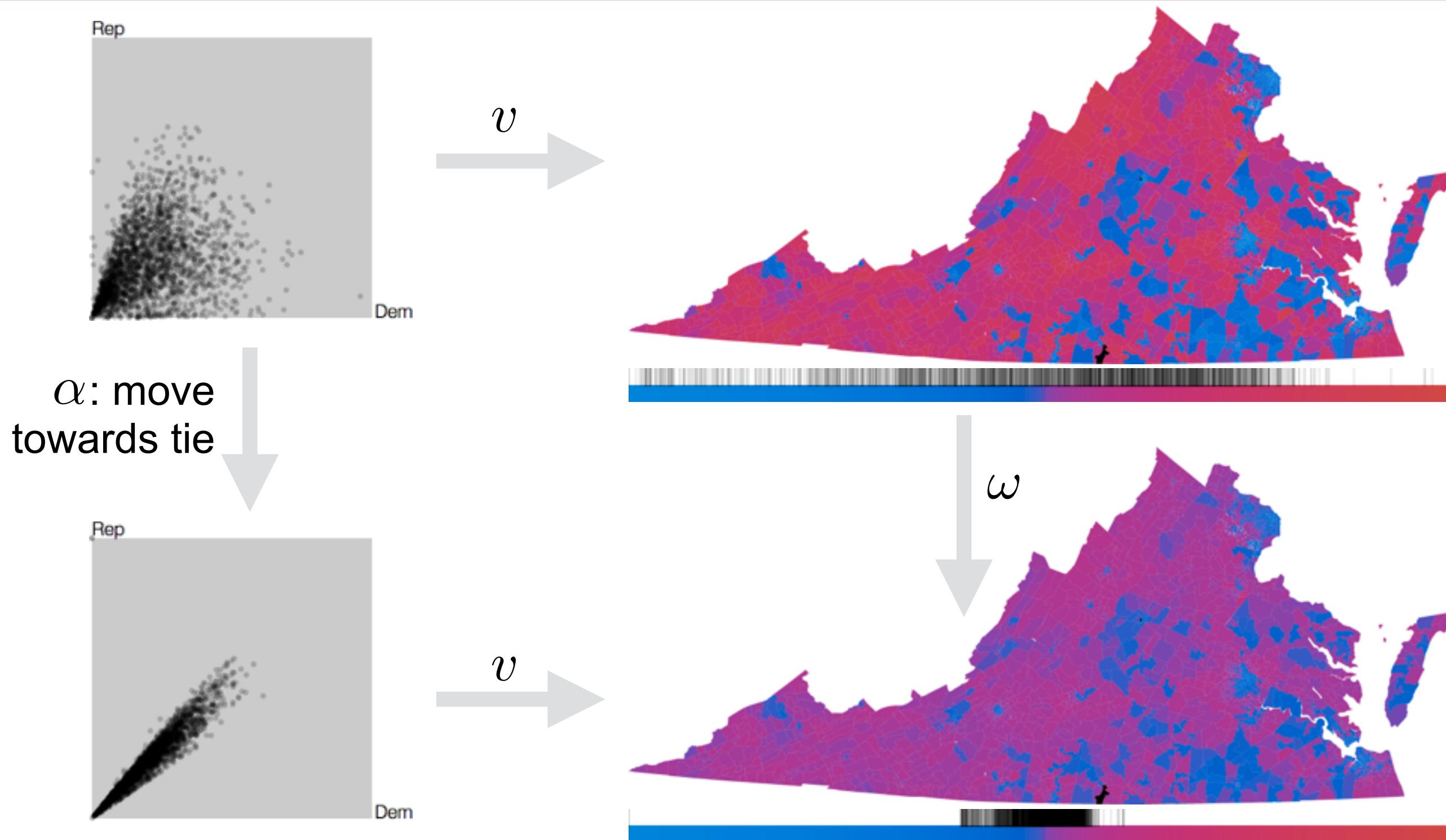
$\alpha$ : move  
towards tie



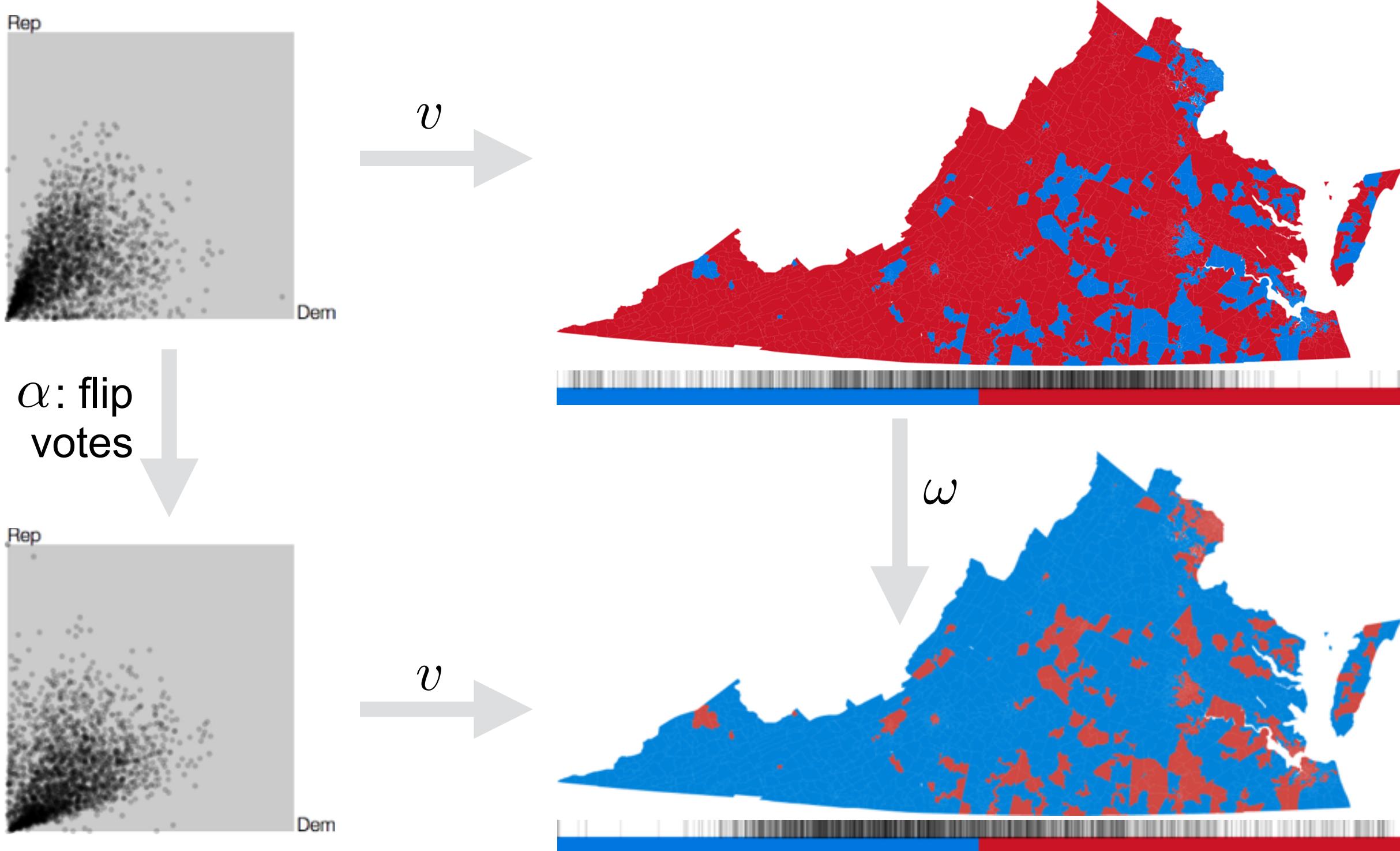
We explicitly  
represent  
source of  
ambiguity



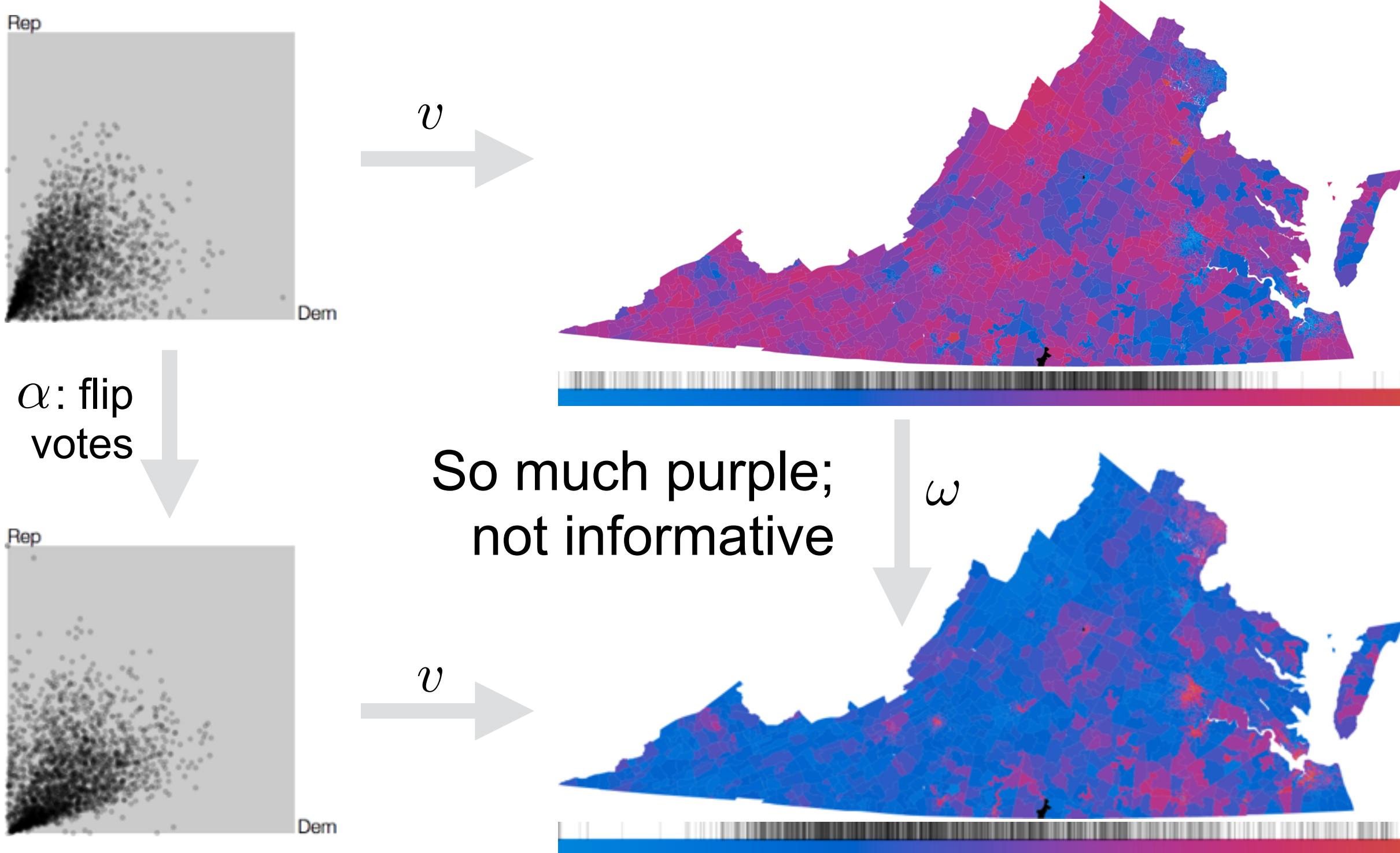
# ... now with a different visualization



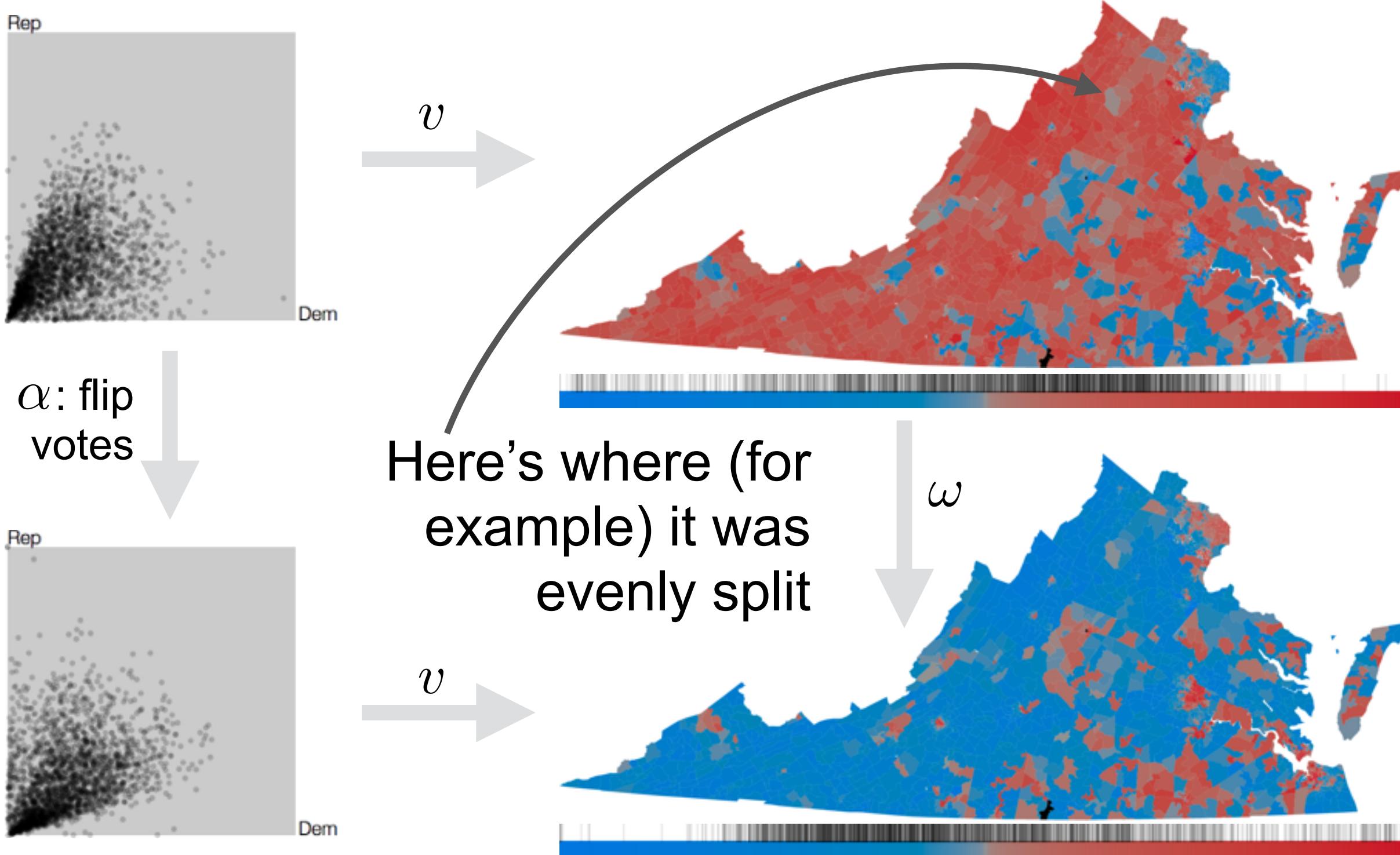
# Where was vote evenly split?



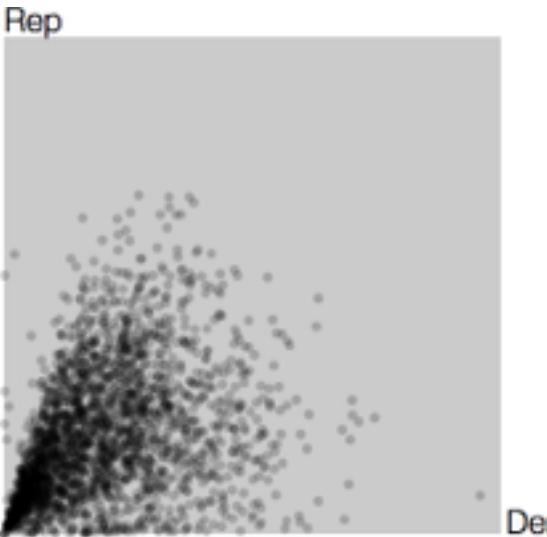
# Where was vote evenly split??



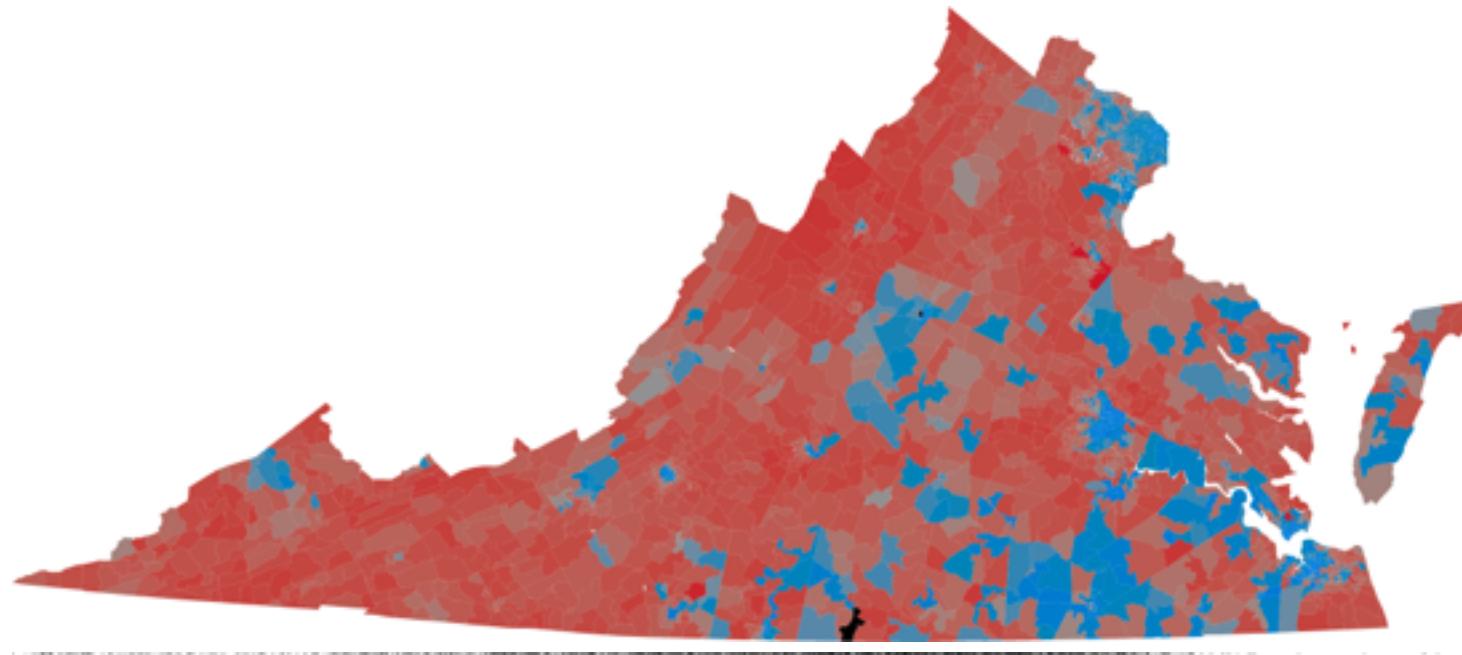
# ... now with a different visualization



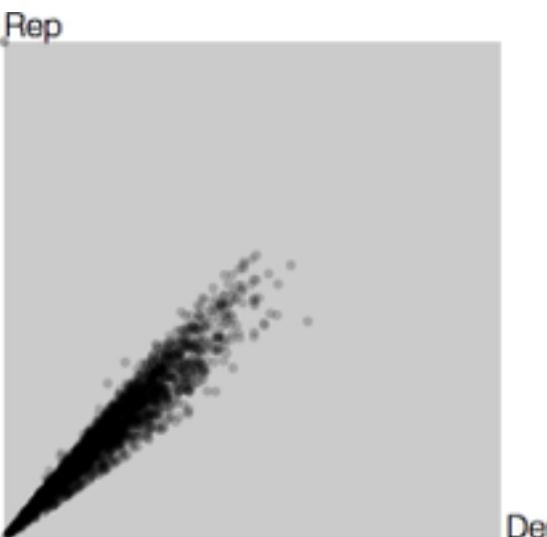
# How about with the first $\alpha$ ?



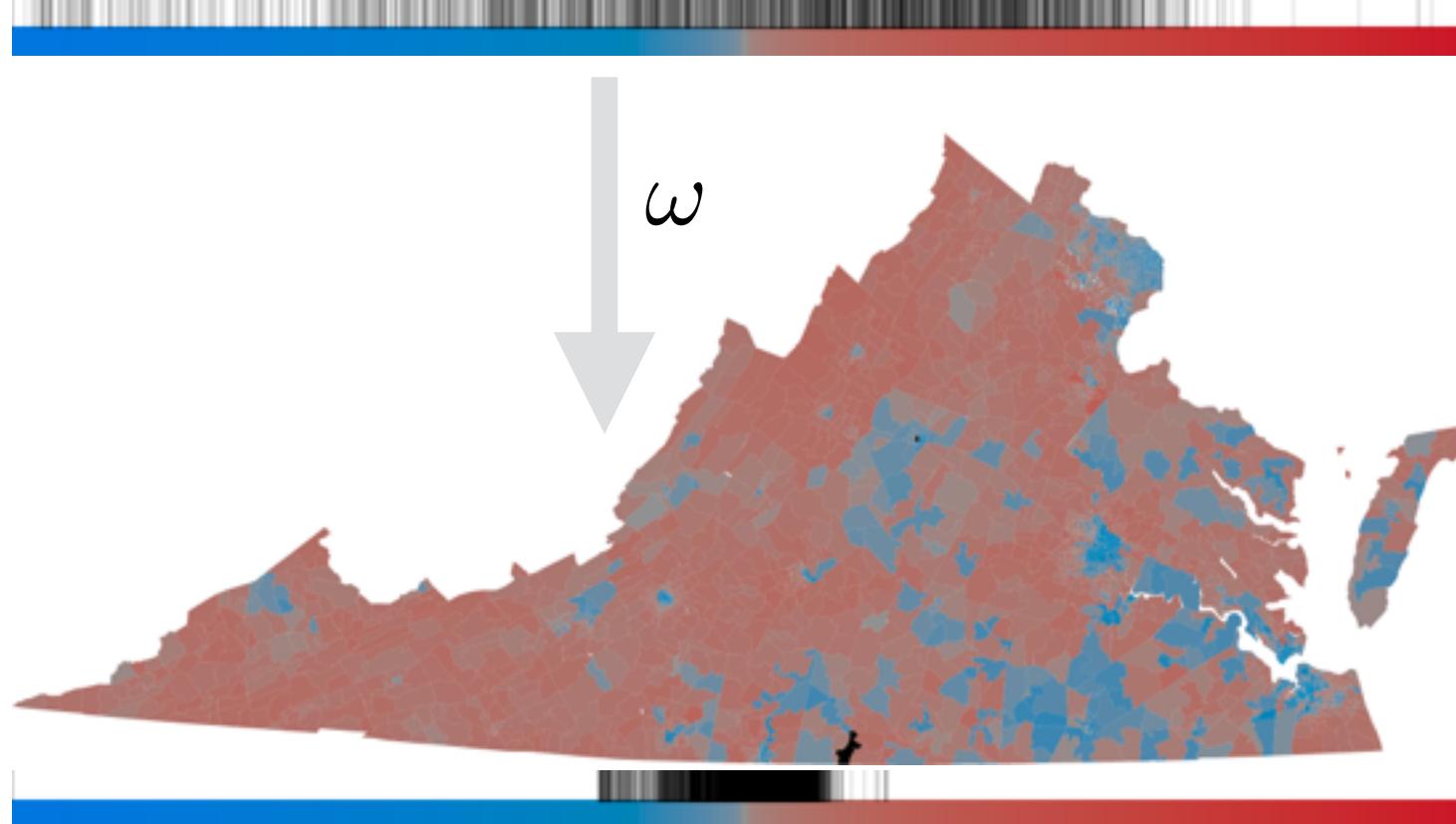
$v$



$\alpha$ : move  
towards tie



$v$



# Design goal: Task $\rightarrow \alpha, \omega \rightarrow$ affordance

## Low-level abstract tasks

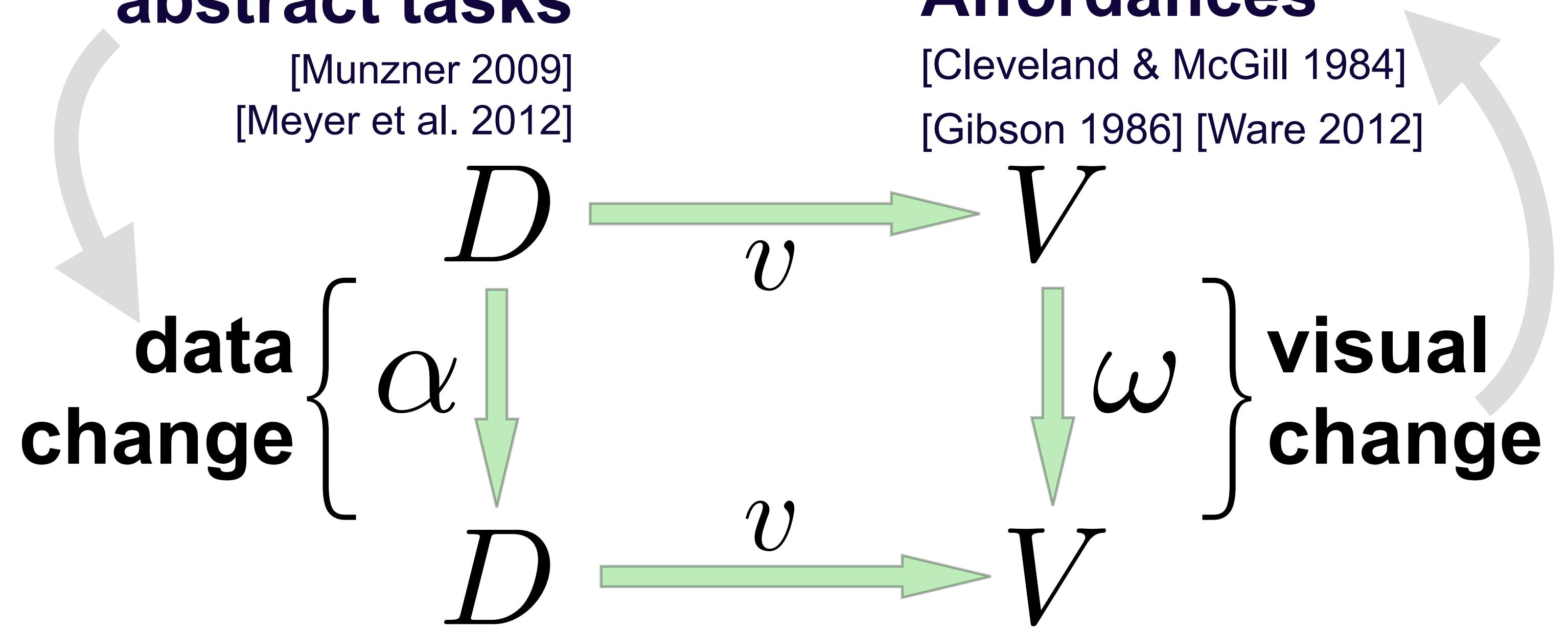
[Munzner 2009]

[Meyer et al. 2012]

## Perception, Affordances

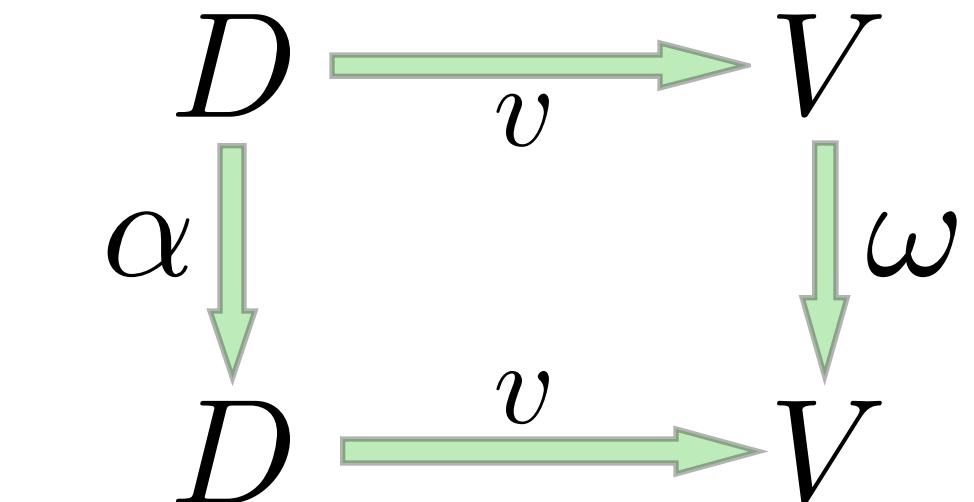
[Cleveland & McGill 1984]

[Gibson 1986] [Ware 2012]



# Three Algebraic Design Principles

Derived from: “Are important  $\alpha$  well-matched with obvious  $\omega$  ?”



Does  $\omega$  make sense, given  $\alpha$  ?

→ 1. Principle of Visual-Data Correspondence

For all important  $\alpha$ , is  $\omega$  obvious?

→ 2. Principle of Unambiguous Data Depiction

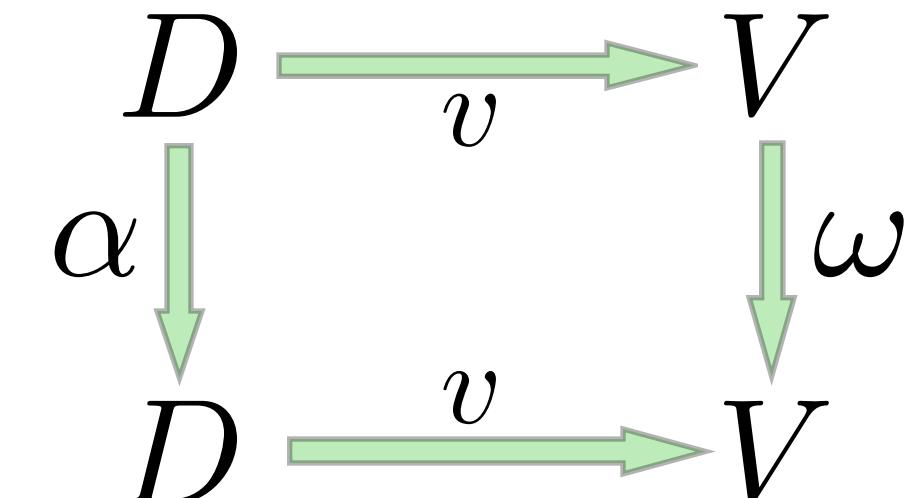
Can obvious  $\omega$  arise without data change ( $\alpha=1$ )?

→ 3. Principle of Representation Invariance

# 1. Principle of Visual-Data Correspondence

Important  $\alpha$  produce obvious and meaningful  $\omega$

- $\alpha$  and  $\omega$  well-matched, “ $\alpha \approx \omega$ ”
- $\omega$  makes sense, given  $\alpha$
- **Congruence:** visual (external) structure  $\cong$  viewer’s mental (internal) structure [Tversky et al. 2002]
- **Effectiveness:** important data attributes mapped to readily perceived visual attributes [Mackinlay 1986]
- **Visual embedding:** visualization preserves distance (in spaces of data, perception) [Demiralp et al. 2014]



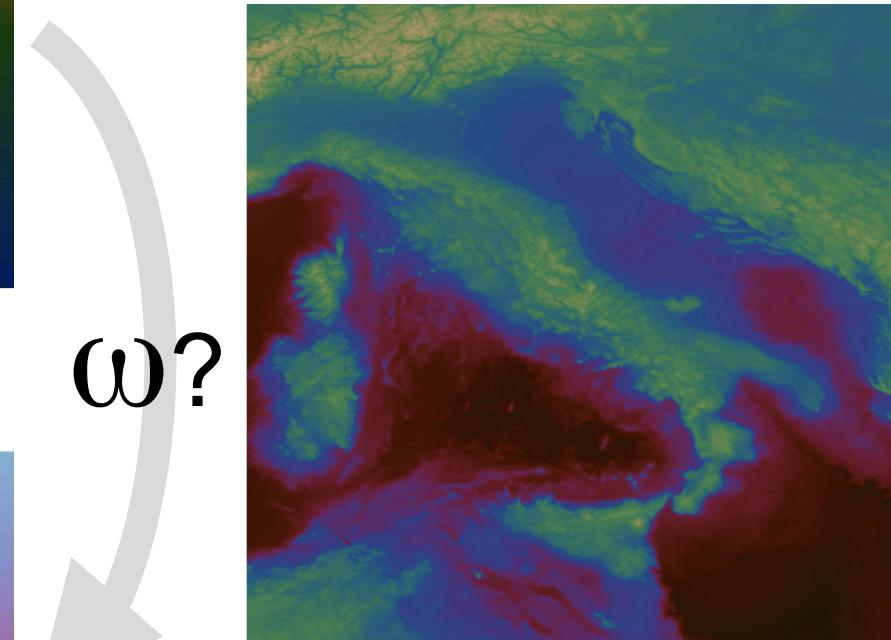
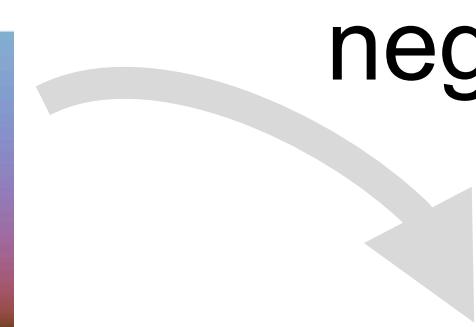
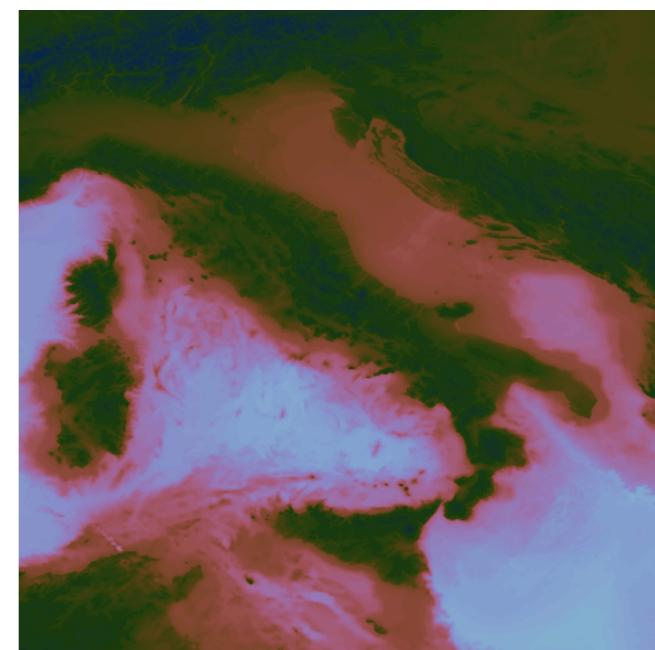
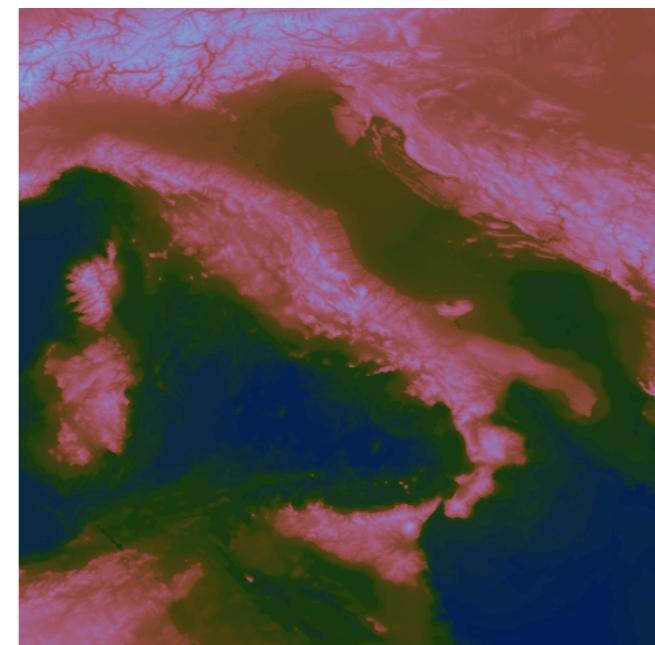
# Correspondence example: elevation colormap

Data: signed elevation  
relative to sea level

$$D \xrightarrow{v}$$

$$\alpha(e) = -e$$

$$D \xrightarrow{v}$$

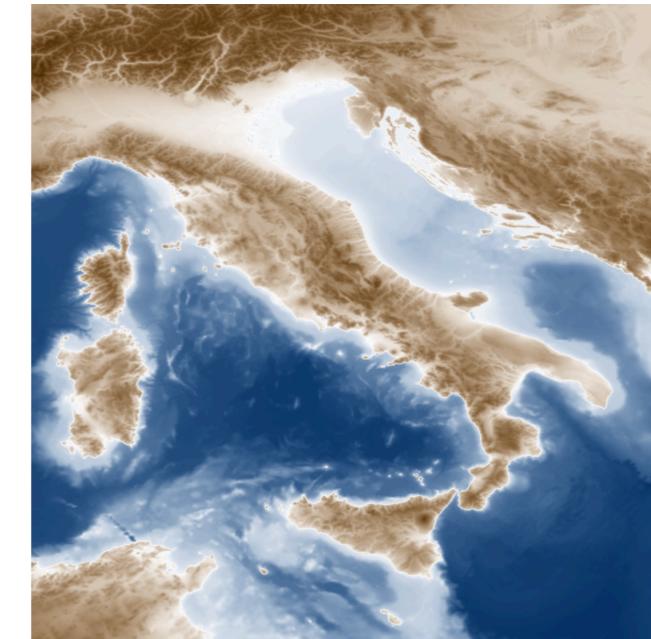


$\omega$ ?  
 $\omega$  not well-matched  
with perception:  
“jumbler”

# Correspondence example: elevation colormap

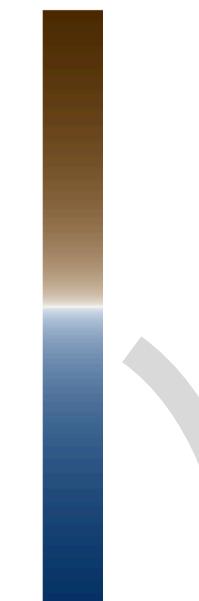
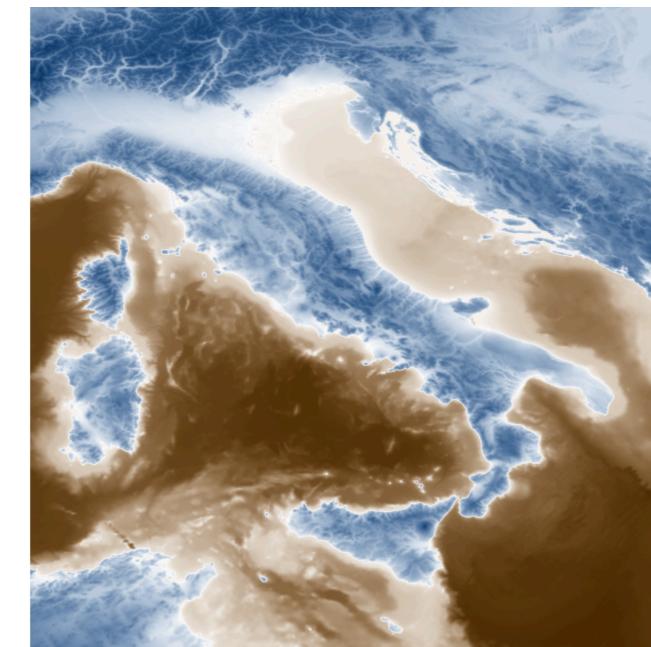
Data: signed elevation  
relative to sea level

$$D \xrightarrow{v}$$

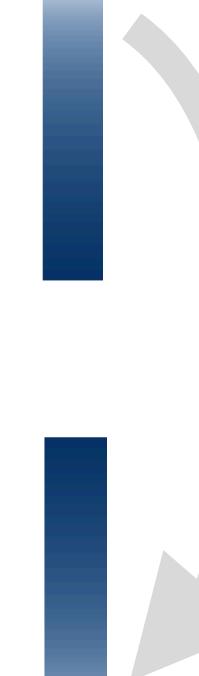


$$\alpha(e) = -e$$

$$D \xrightarrow{v}$$



diverging  
colormap



$\omega$ : negate hue

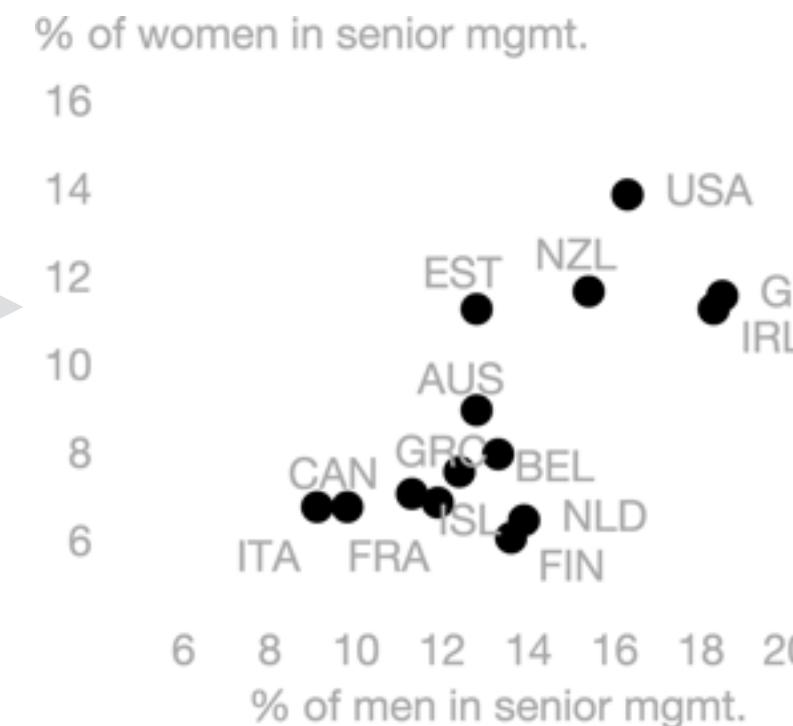
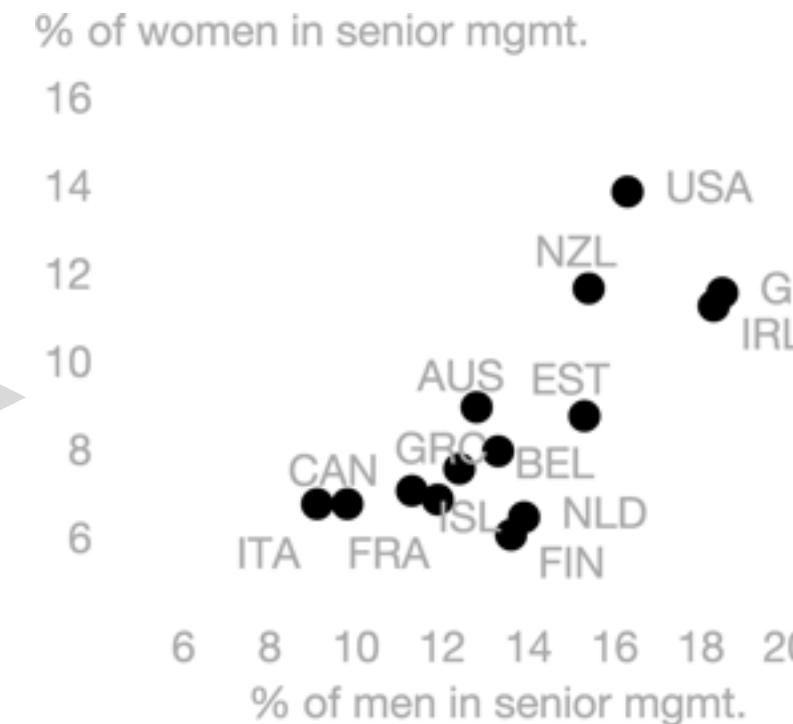
# Correspondence example: scatterplots

Data: % men vs women  
employed as senior  
managers in various  
countries

$$D \xrightarrow{v}$$

$\alpha$ : decrease  
gender gap  
for one  
country: EST

$$D \xrightarrow{v}$$



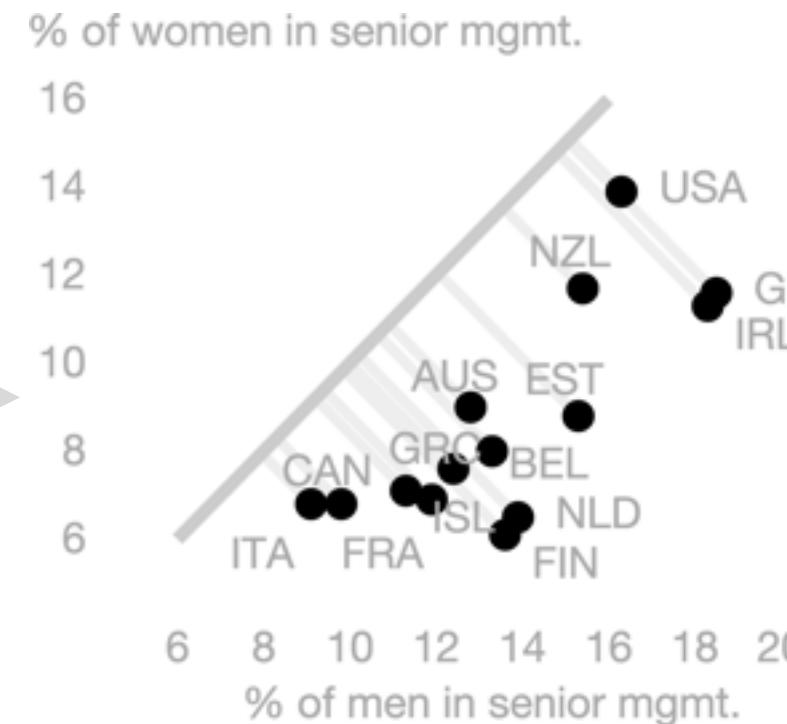
$\omega$ ? Not clear  
how big that  
change was

# Correspondence example: scatterplots

Data: % men vs women employed as senior managers in various countries

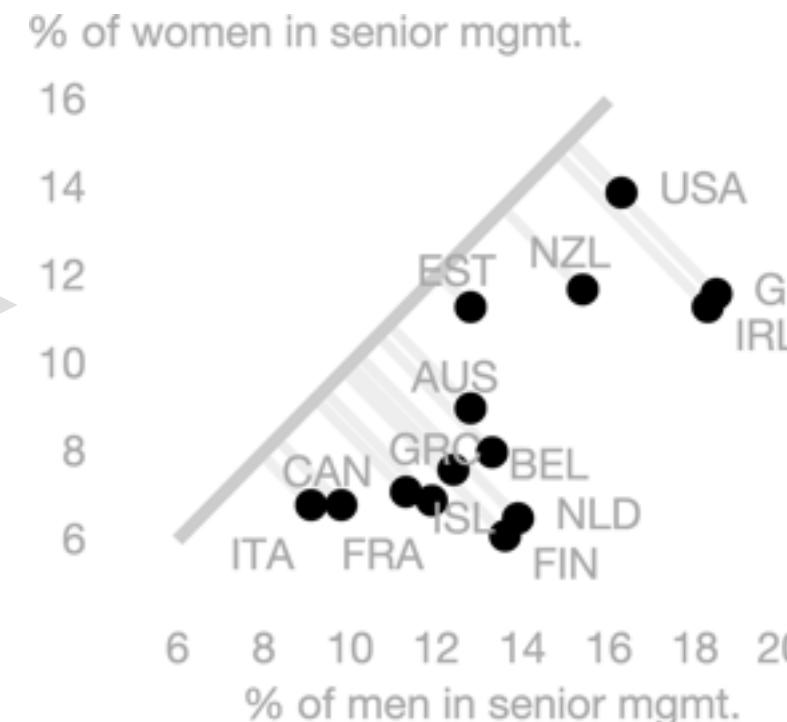


$\alpha$ : decrease gender gap for one country: EST



add diagonal line  
(%men = %women)  
and support lines

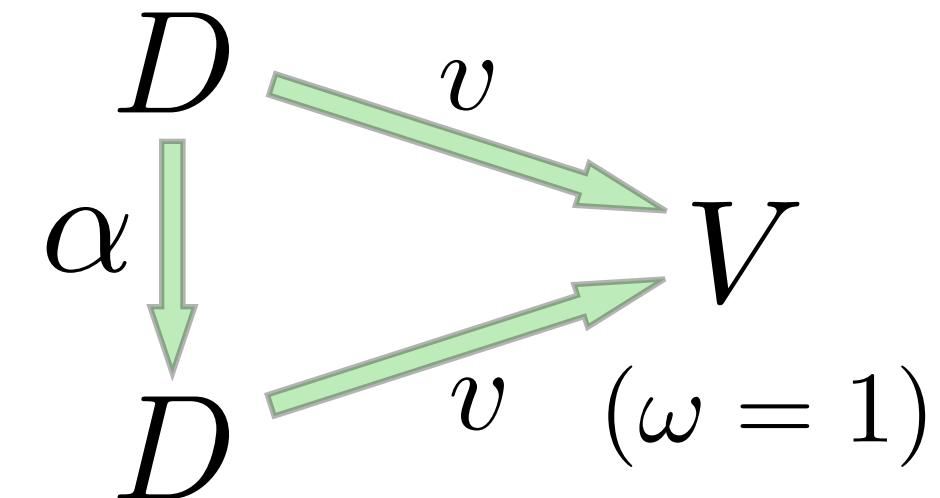
$\omega$ : change in position along a common scale [Cleveland & McGill 1984]



## 2. Principle of Unambiguous Data Depiction

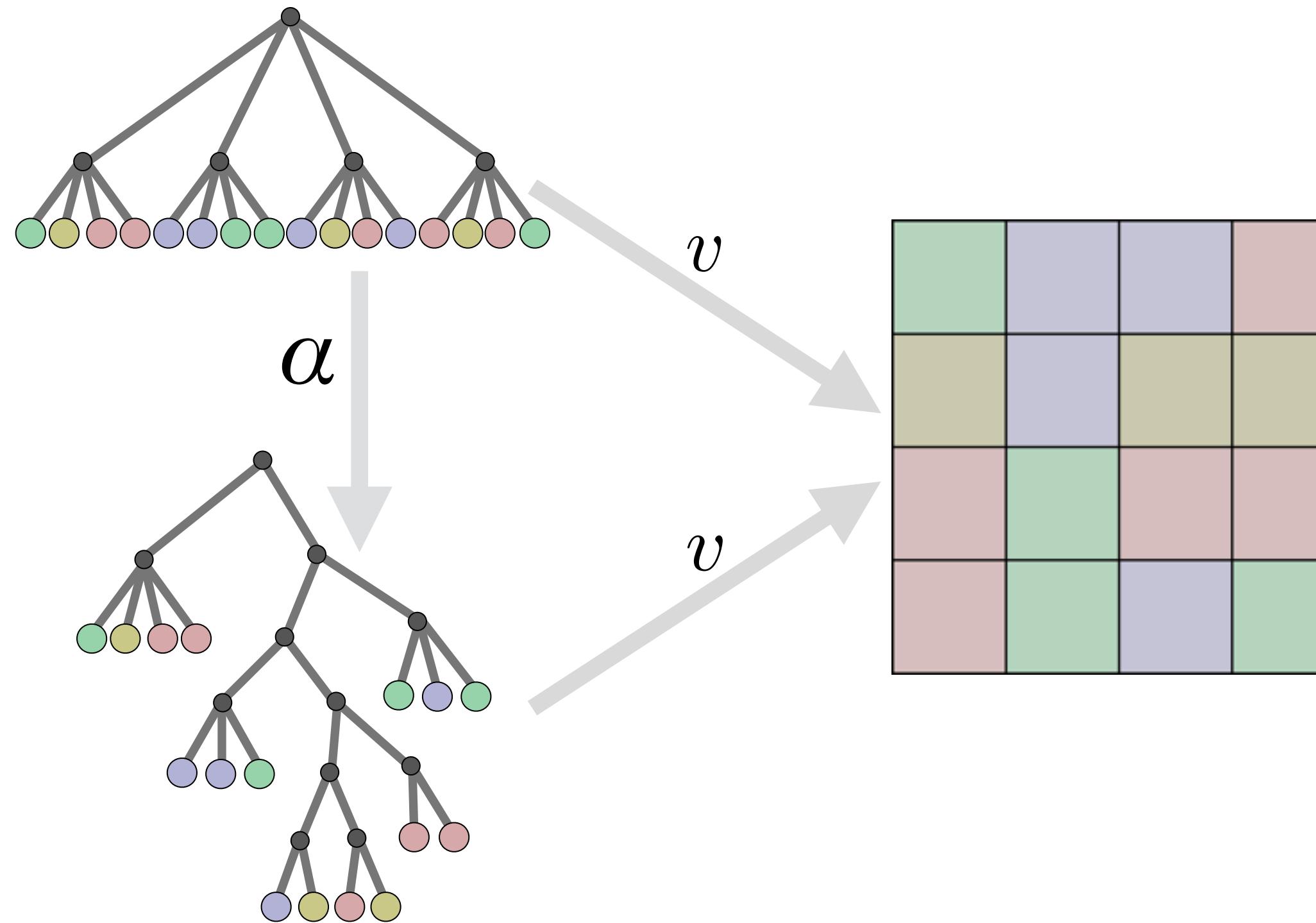
Important  $\alpha$  map to obvious  $\omega$ .

If  $\omega=1$ , then  $\alpha=1$ .



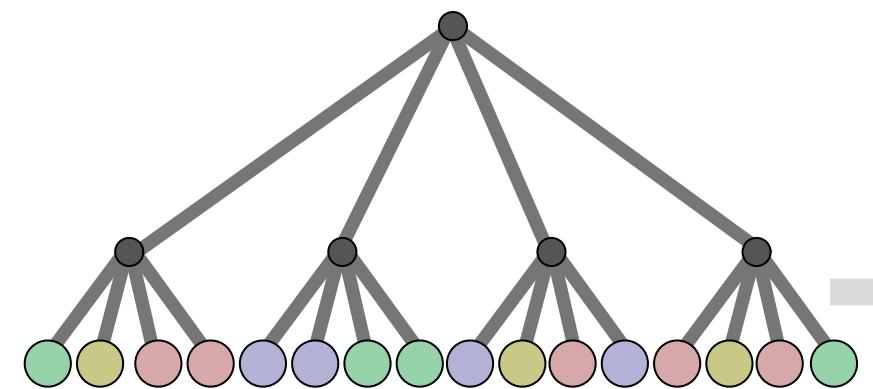
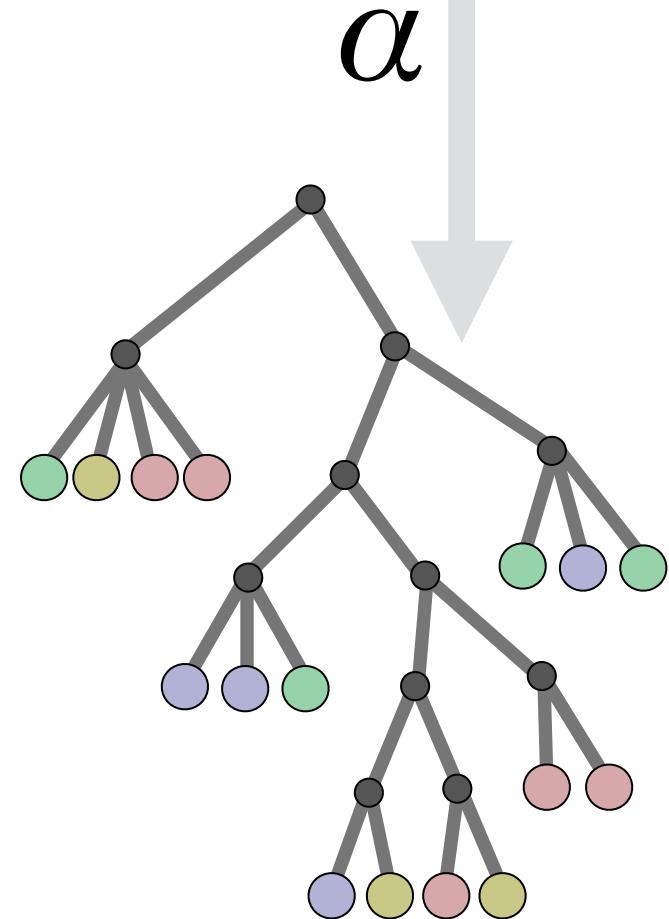
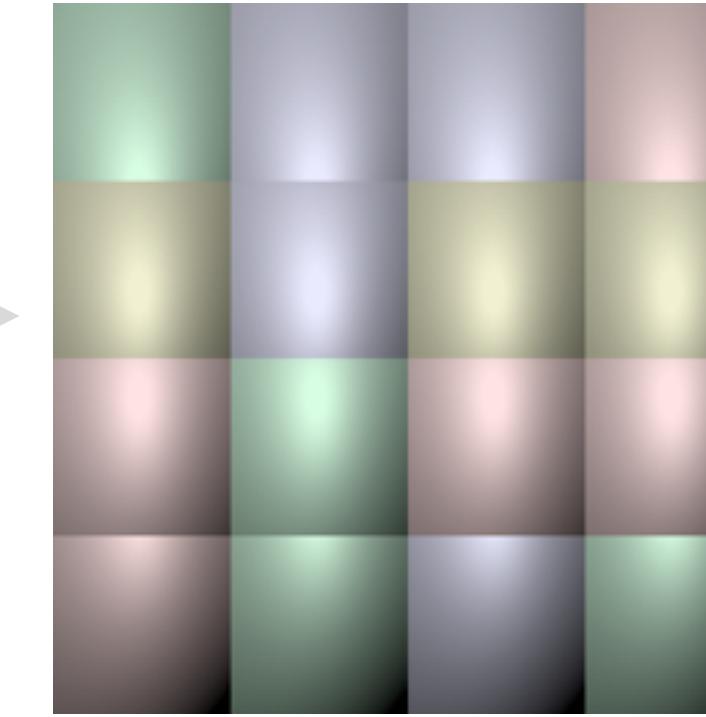
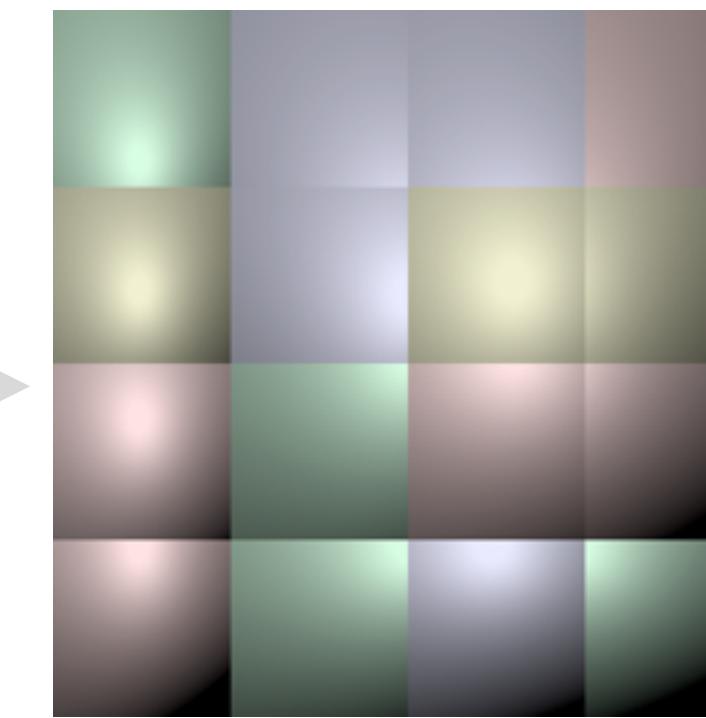
- **Expressiveness:** visualization shows all facts about data (and nothing more) [Mackinlay 1986]
- **Injectivity:** visualization preserves distinctness so viewer can invert it (read it) [Ziemkiewicz & Kosara 2009]
- If not  $v$  injective,  $\alpha$  explicitly indicates the ambiguity:
  - $\alpha$  is the “confuser”

# Unambiguity example: treemaps



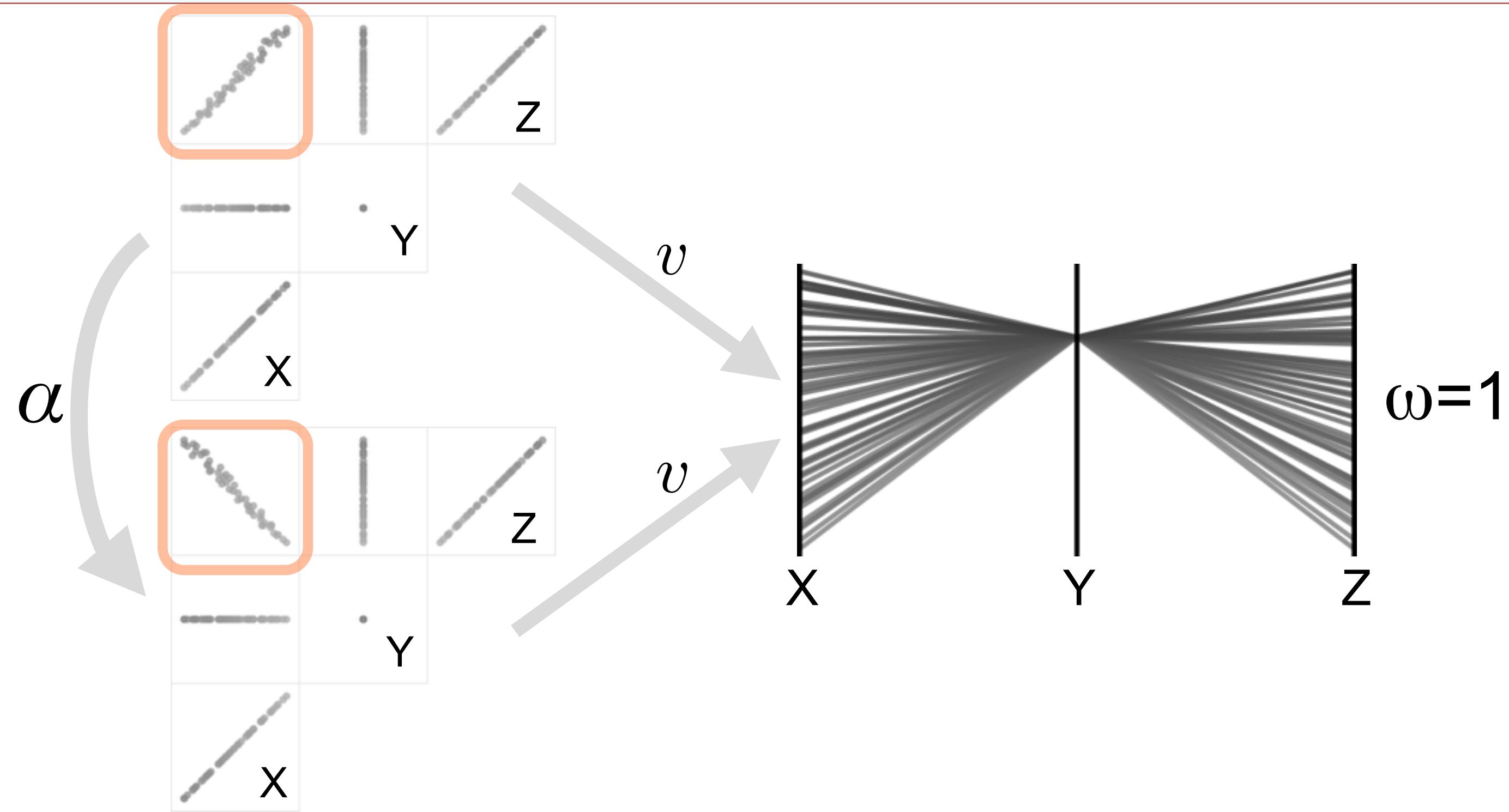
$\omega=1$ :  $\alpha$  is  
“confuser”  
for treemaps

# Unambiguity example: treemaps

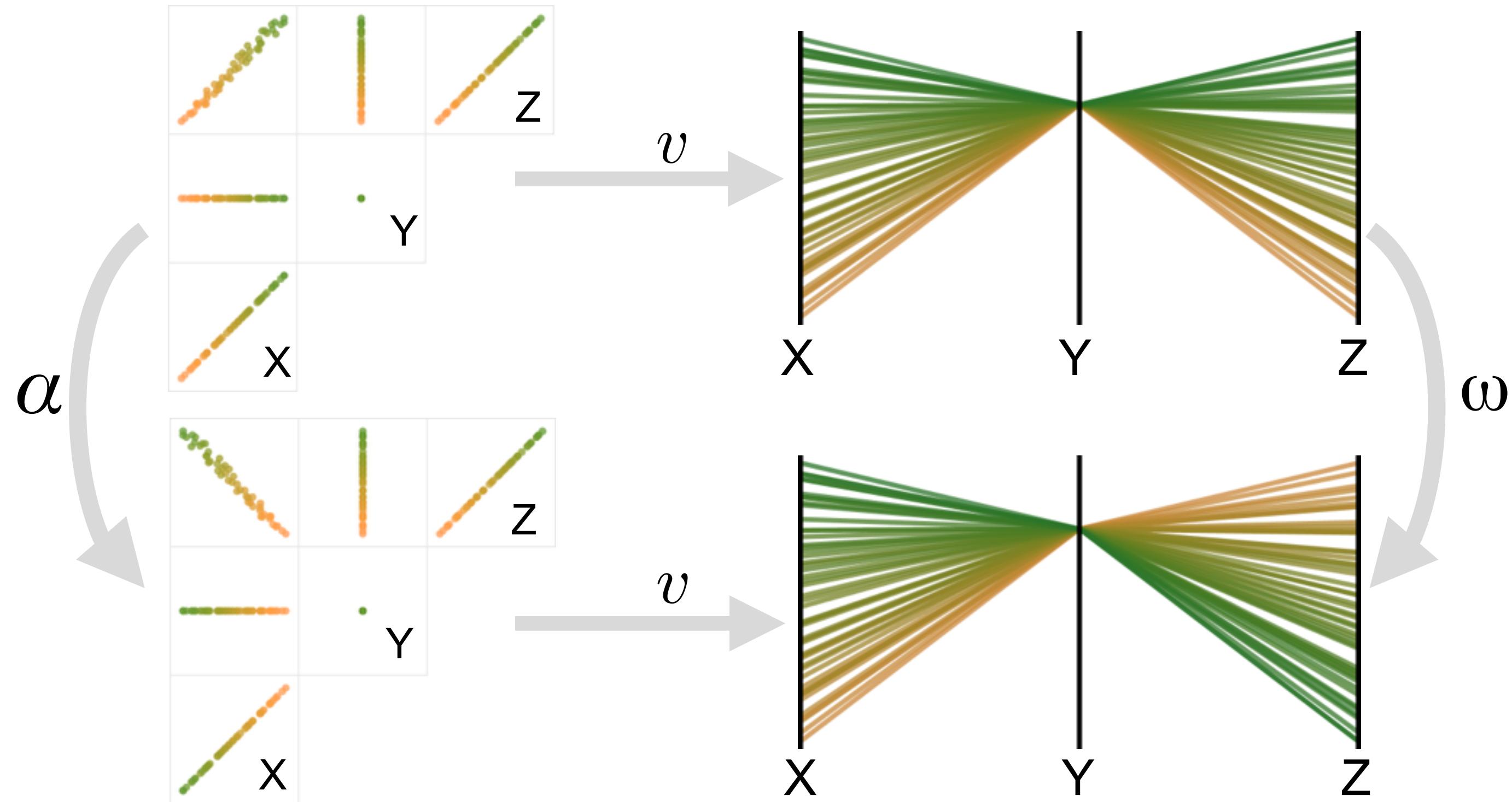
 $v$  $v$ 

$\omega \neq 1$  with  
cushion  
treemaps  
[van Wijk & H.  
van de Wetering  
1999]

# Unambiguity example: parallel coordinates



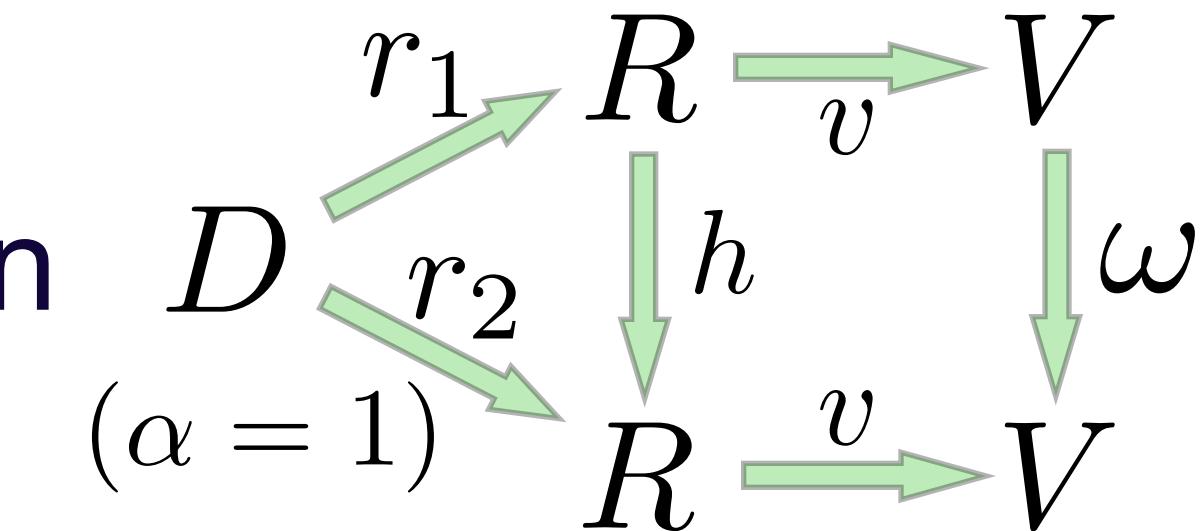
# Unambiguity example: parallel coordinates



### 3. Principle of Representation Invariance

Visualization is invariant w.r.t  
changes in data representation

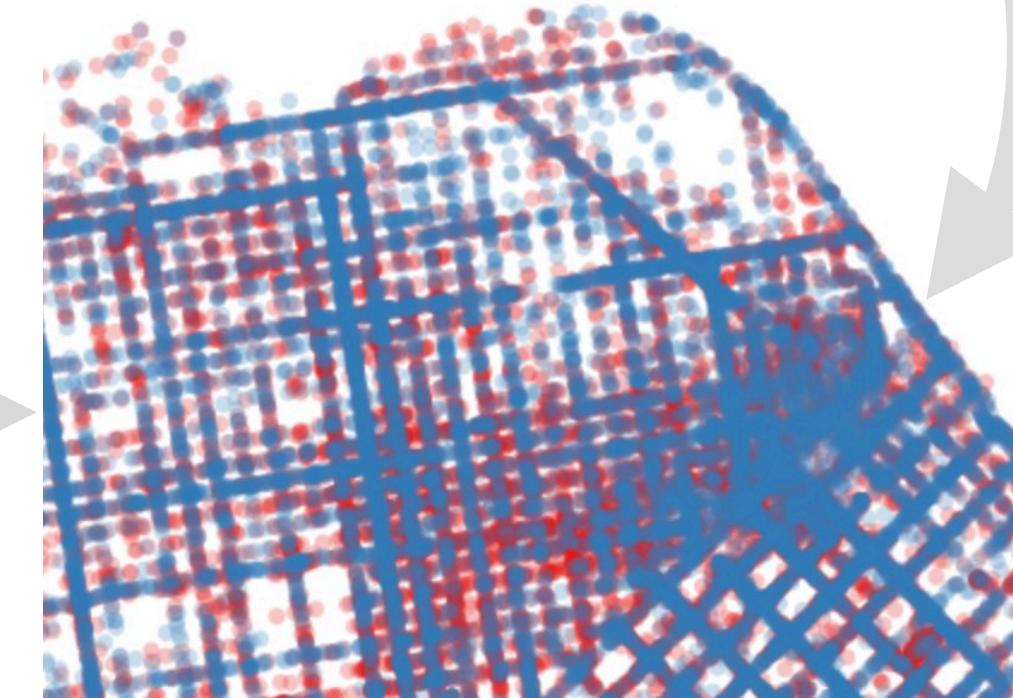
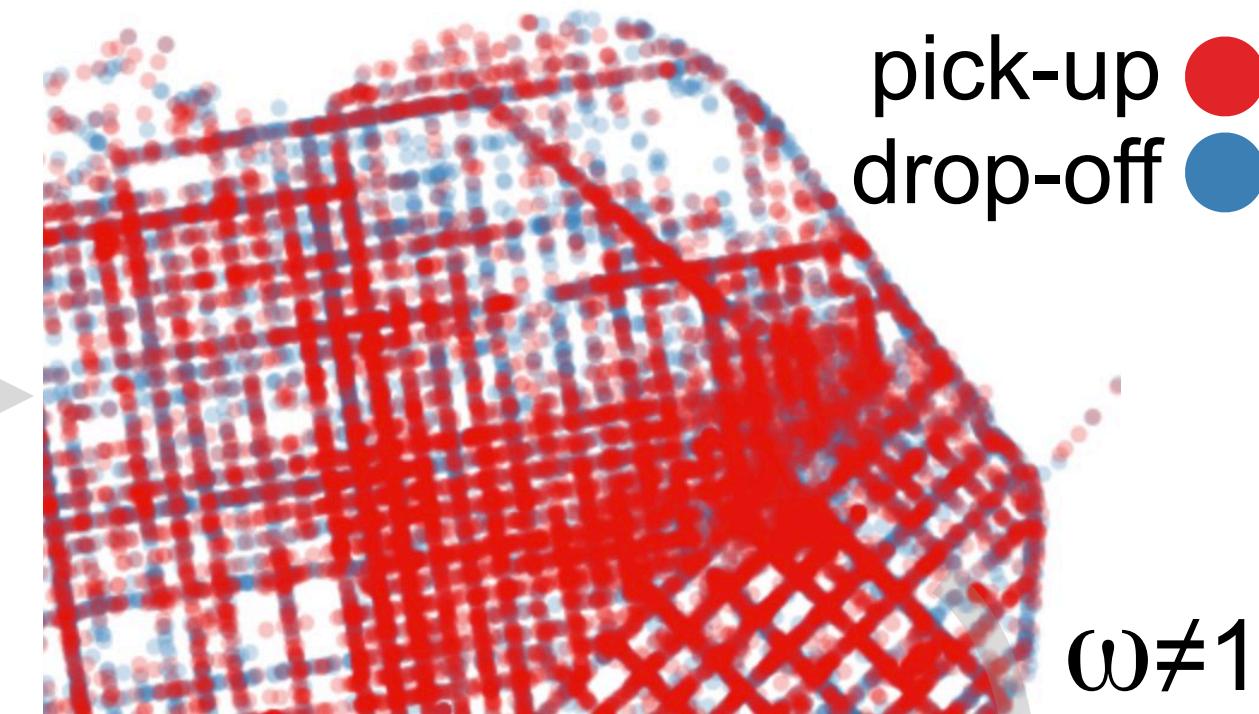
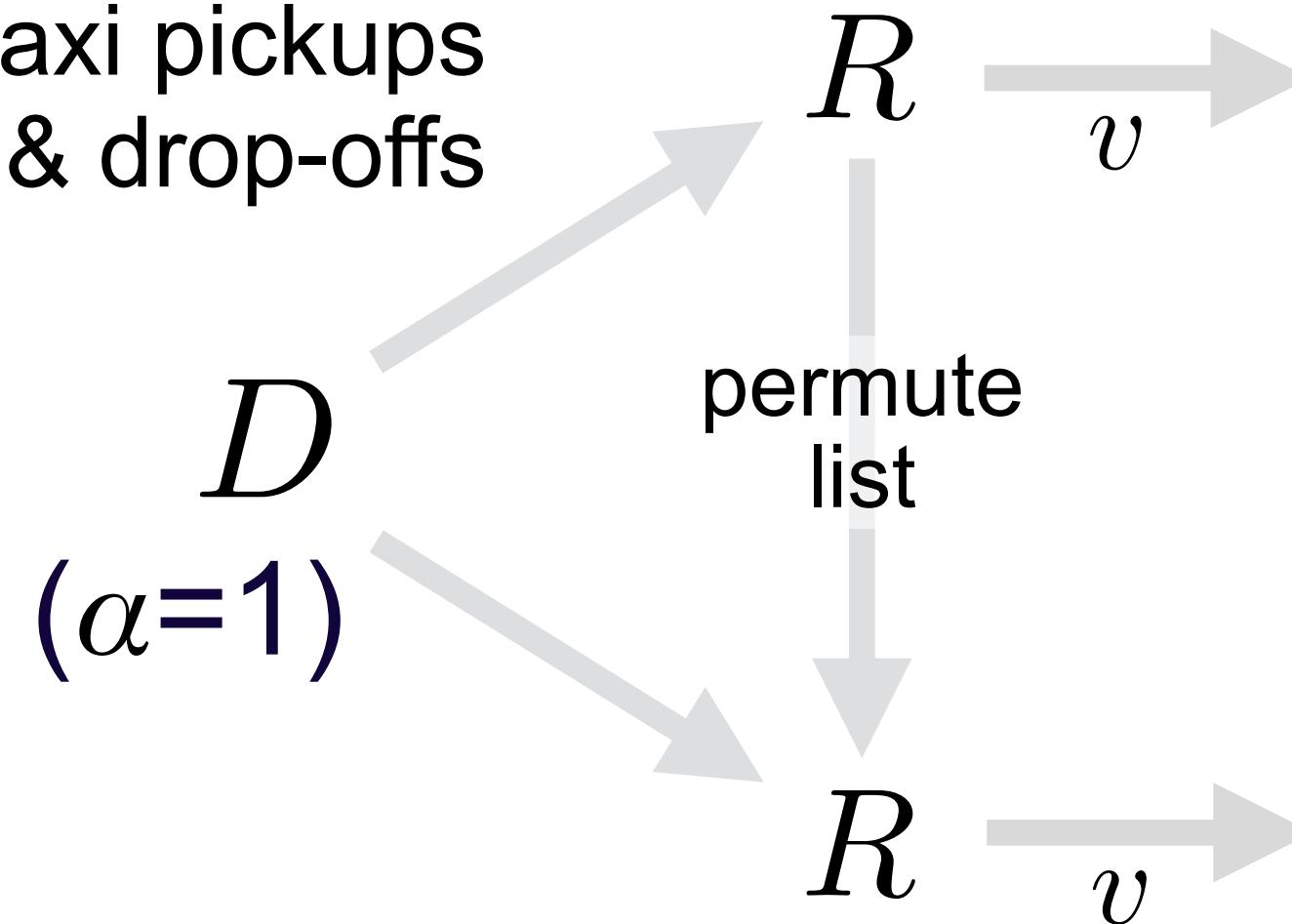
If  $\alpha=1$ , then  $\omega=1$ .



- Underlying **data  $D \neq$  representation  $R$**  of data
  - sets as lists, eigenvectors as vectors
  - **Invariantive:** Scale of measurement (nominal, ordinal, interval, ratio) limits permissible statistics [Stevens 1946]
  - If change  $h$  in representation is visible ( $\omega \neq 1$ ):
    - $h$  is the “hallucinator”

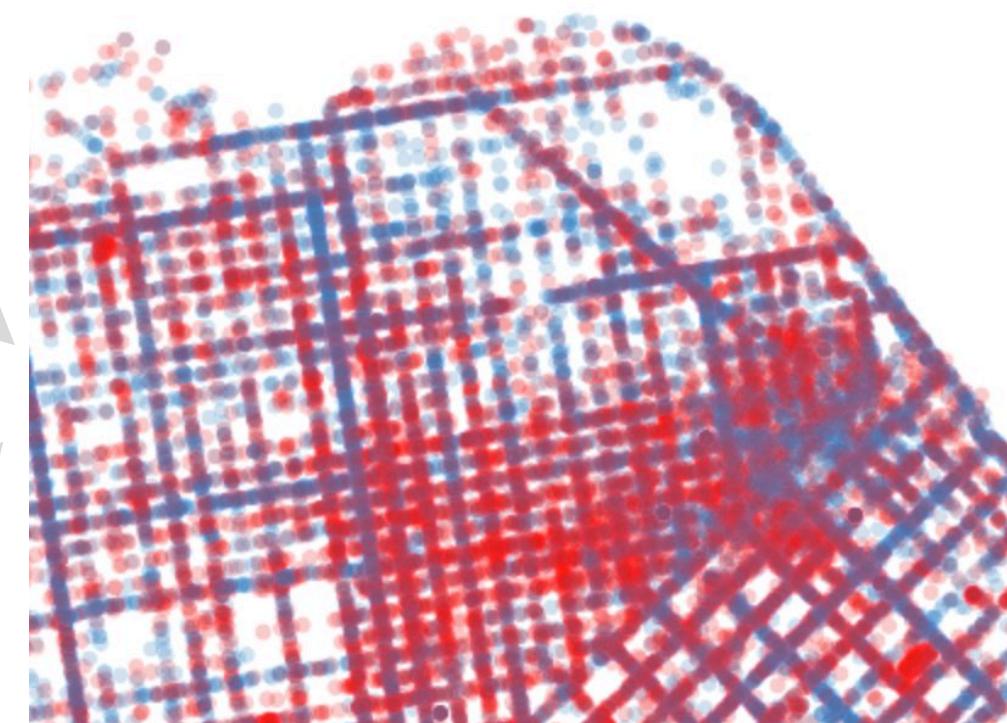
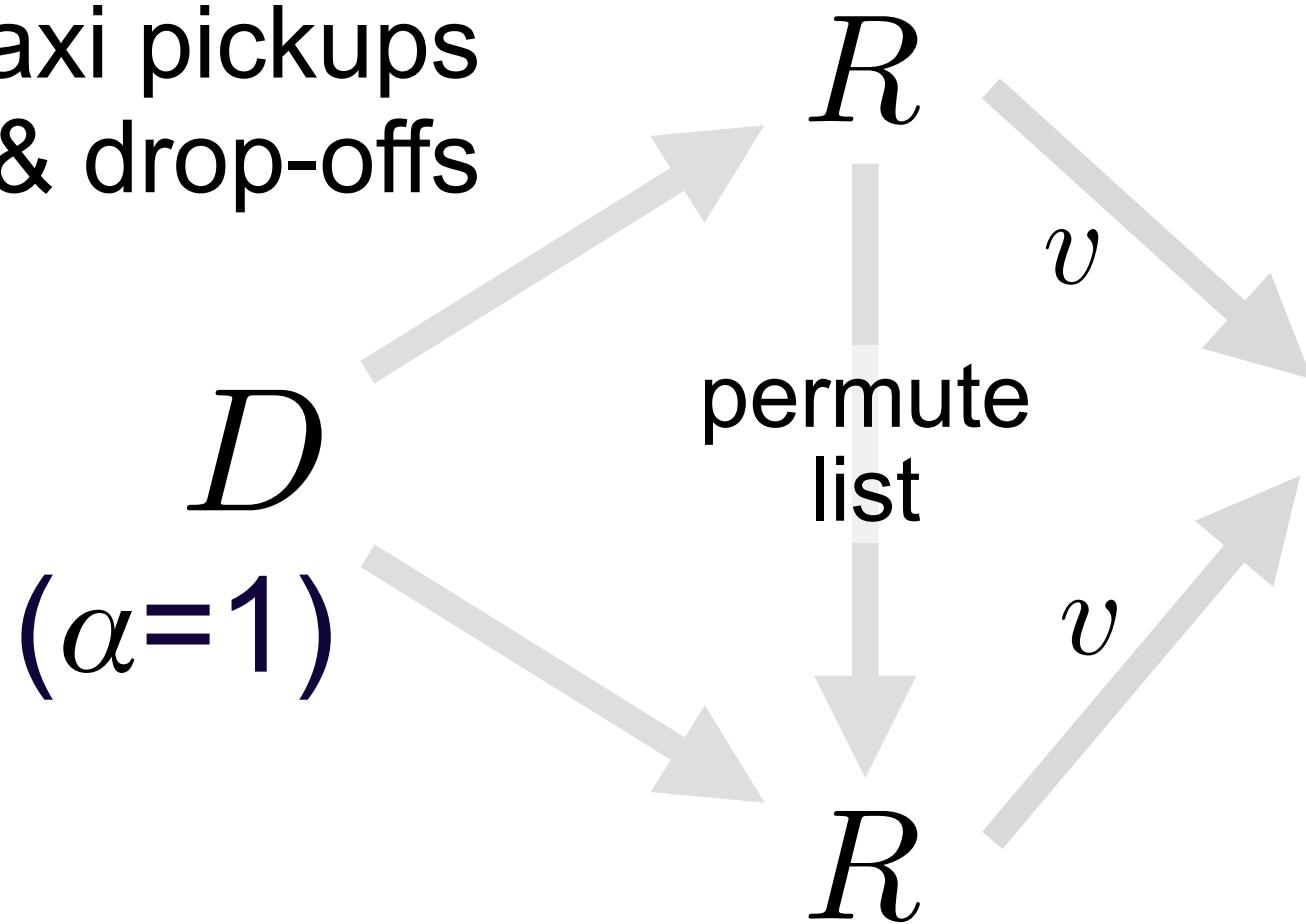
# Invariance example: alpha-blended marks

Data: set of locations of taxi pickups & drop-offs



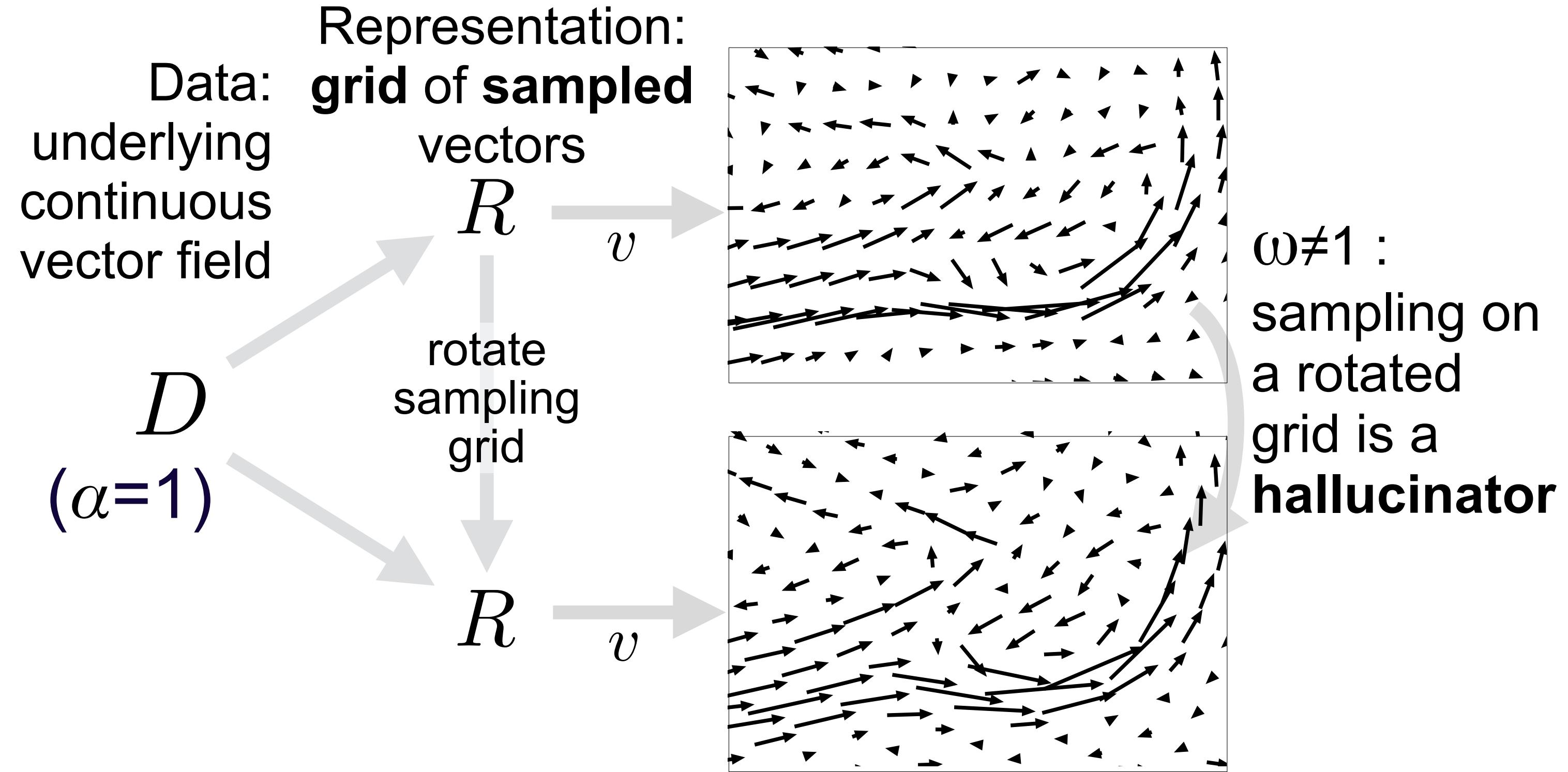
# Invariance example: alpha-blended marks

Data: set of locations of taxi pickups & drop-offs

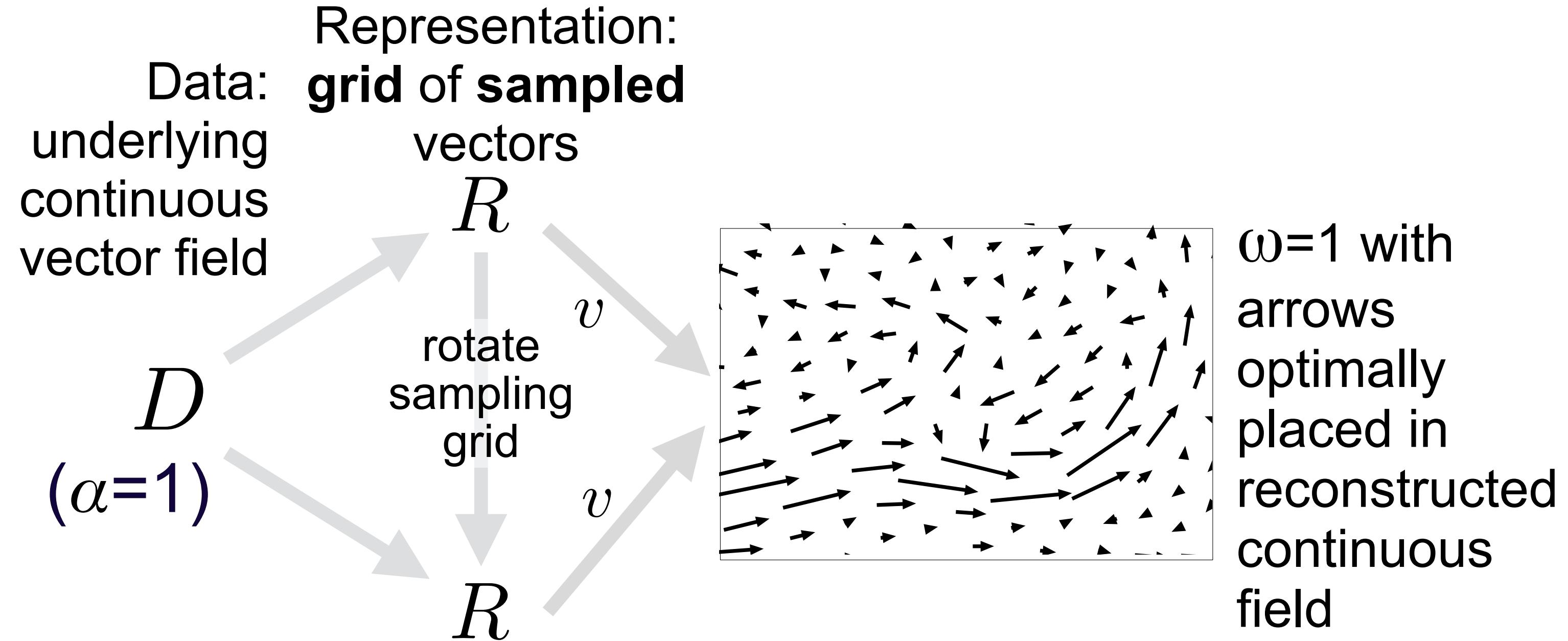


$\omega=1$  with  
**order-invariant**  
(commutative)  
compositing

# Invariance example: quiver plot

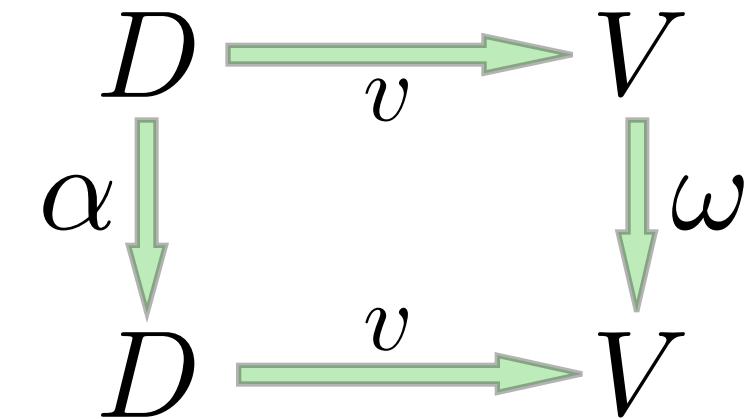


# Invariance example: quiver plot

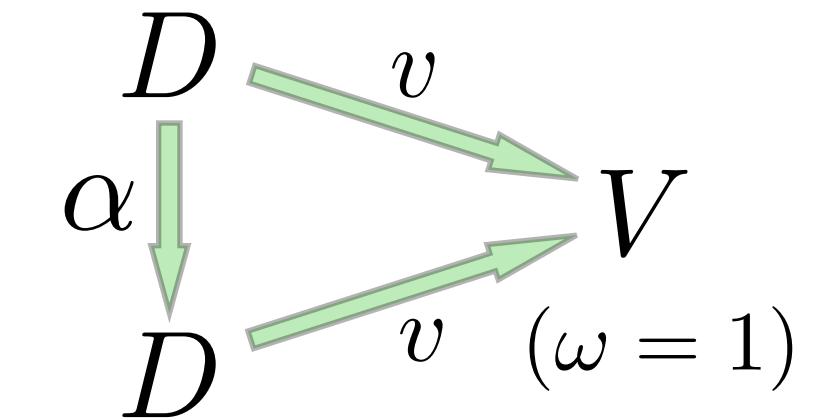


# Summary of 3 Principles

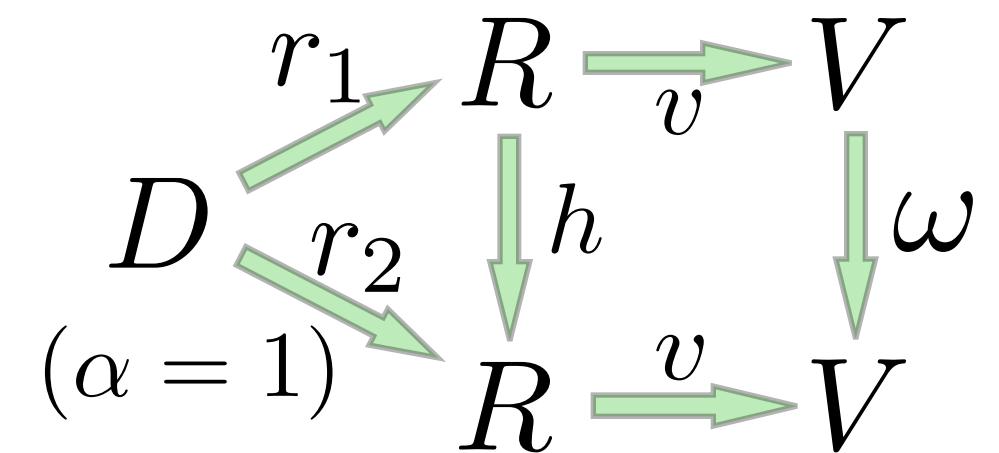
- Visual-Data Correspondence  
or else a **jumbler**  $\alpha$ , or **misleader**  $\omega$



- Unambiguous Data Depiction  
or else a **confuser**  $\alpha$



- Representation Invariance  
or else a **hallucinator**  $h$

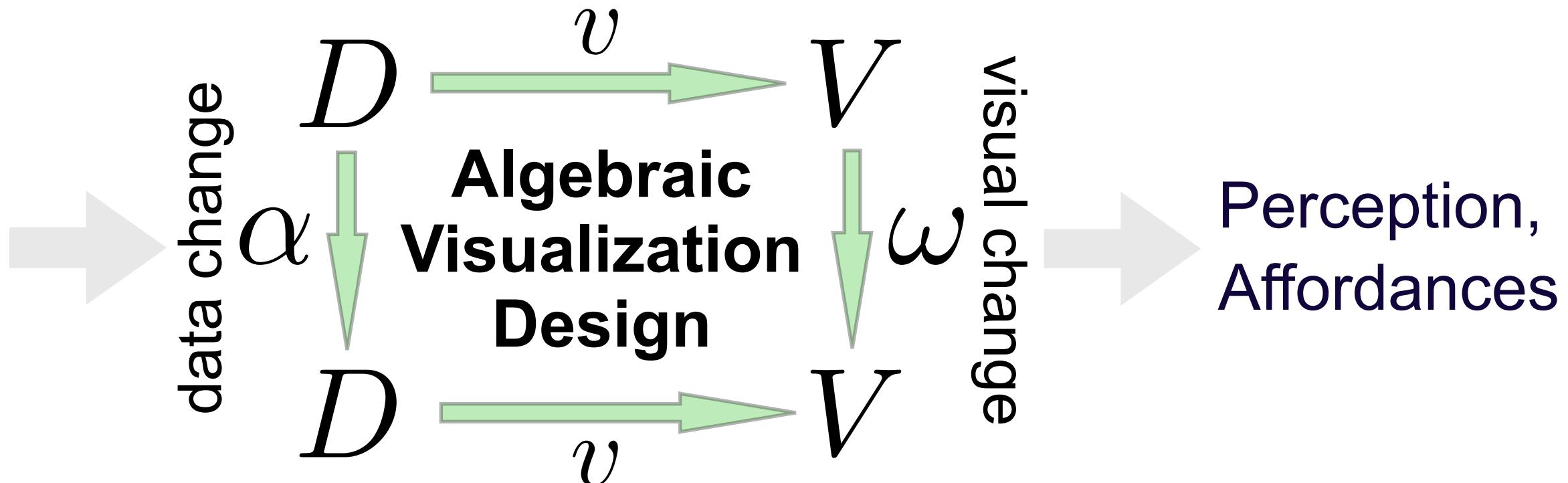


# References

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- [Brehmer & Munzner 2013]: A multi-level typology of abstract visualization tasks. *IEEE TVCG.*, 19(12):2376–2385, 2013
- [Cleveland & McGill 1984]: W. S. Cleveland and R. McGill. Graphical perception: Theory, experimentation, and application to the development of graphical methods. *J. American Statistical Association*, 79(387):531–554, 1984.
- [Demiralp et al. 2014]: Visual embedding: A model for visualization. *IEEE CG&A.*, 34(1):10–15, 2014.
- [Gibson 1986]: J. J. Gibson. The Ecological Approach To Visual Perception, chapter 8: The Theory of Affordances. Lawrence Erlbaum Associates, 1986.
- [Mackinlay 1986]: J. Mackinlay. Automating the design of graphical presentations of relational information. *ACM Trans. Graph.*, 5(2):110–141, 1986.
- [Meyer et al. 2012]: The four-level nested model revisited: blocks and guidelines. In Proc. 2012 BELIV Workshop, pages 11:1–11:6, 2012.
- [Munzner 2009]: T. Munzner. A nested model for visualization design and validation. *IEEE TVCG.*, 15(6):921–928, 2009.
- [Stevens 1946]: On the theory of scales of measurement. *Science*, 103(2684):677–680, 1946.
- [Tversky et al. 2002]: B. Tversky, J. B. Morrison, and M. Betrancourt. Animation: can it facilitate? *Intl. J. Hum.-Comp. Stud.*, 57(4):247–262, 2002.
- [Ware 2012]: C. Ware. Information visualization: perception for design. Elsevier, 2012.
- [van Wijk & H. van de Wetering 1999]: Cushion treemaps: Visualization of hierarchical information. In Proc. Info. Vis., pp 73–78, 1999.
- [Ziemkiewicz & Kosara 2009]: C. Ziemkiewicz and R. Kosara. Embedding information visualization within visual representation. In Z. W. Ras and W. Ribarsky, editors, *Advances in Information and Intelligent Systems*, volume 251 of Studies in Computational Intelligence, pages 307–326. Springer, 2009.

Low-level  
abstract tasks



Thanks to:

- New York Times (Amanda Cox, Mike Bostock, Derek Watkins, and Shan Carter) for Virginia Senate election data <http://www.nytimes.com/interactive/2014/11/04/upshot/senate-maps.html>
- Anonymous reviewers for constructive feedback
- Conversations with: Tamara Munzner, Stephen Ingram, Hadley Wickham, Çağatay Demiralp, Xavier Tricoche, and Thomas Schultz
- 2009 Dagstuhl Scientific Visualization Seminar 09251

Web page for paper, these slides, and eventually more:

- <http://AlgebraicVis.net>

and on Twitter! @algebraicvis