

# Data Visualization

A motivational and inspirational talk

Johan Van Horebeek  
October 6th, 2021

## Goals:

- To make a guided walk with some *best practices*
- To present some Tools from Statistics/ML
- To provide some useful and inspirational references



CIMAT

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- To present some Tools from Statistics/ML
- To provide some useful and inspirational references

Not: research talk, technical presentation, to overwhelm



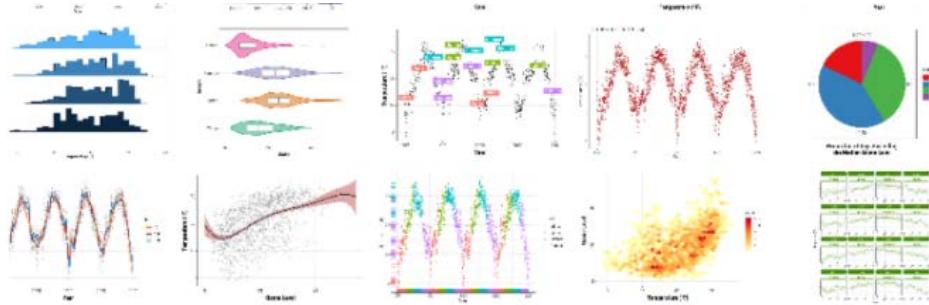
CIMAT

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Centro de Investigación en Matemáticas, A.C.

# 1. Roots and Aims of visualization

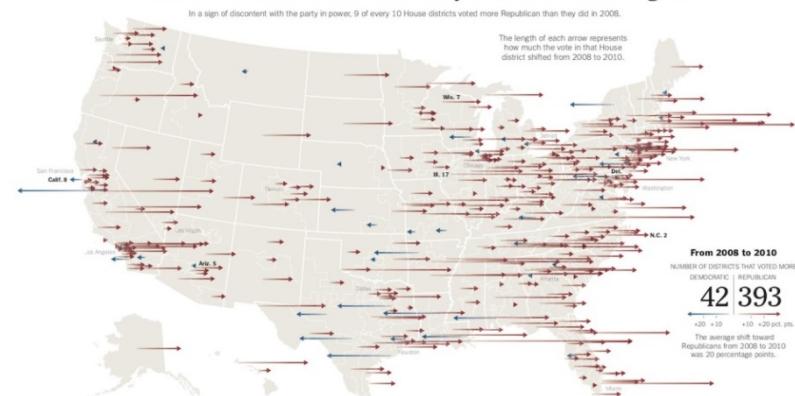
Statistical graphics (roots in statistics community)



Infographics (roots in graphical design community)

The New York Times

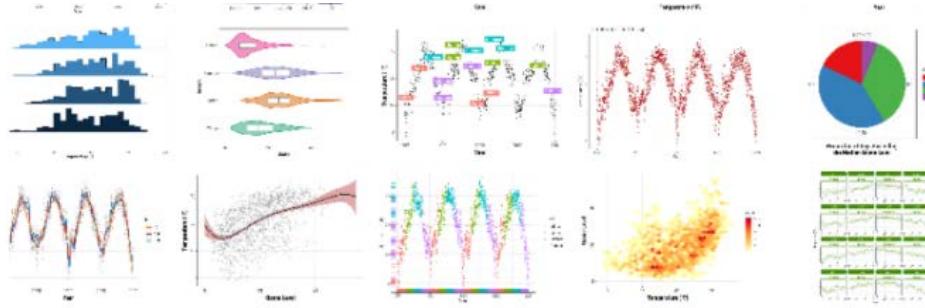
## Districts Across the Country Shift to the Right



Data Story Telling

# 1. Roots and Aims of visualization

## Statistical graphics (roots in statistics community)



Exploration  
Generality  
For you (often)

Tool to evaluate models

## Infographics (roots in graphical design community)

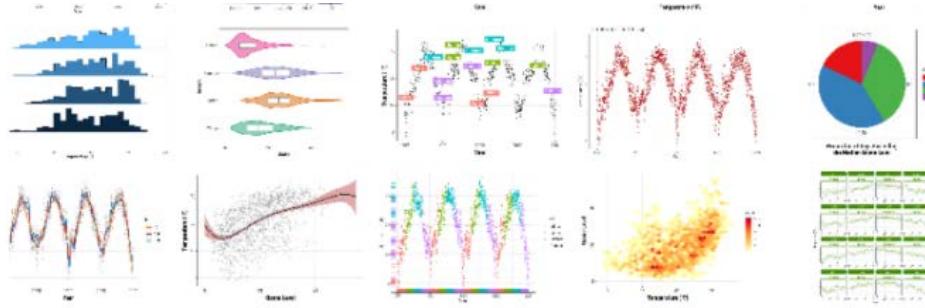


Communication of a message  
To engage  
Unique  
For a particular audience

## Data Story Telling

# 1. Roots and Aims of visualization

## Statistical graphics (roots in statistics community)



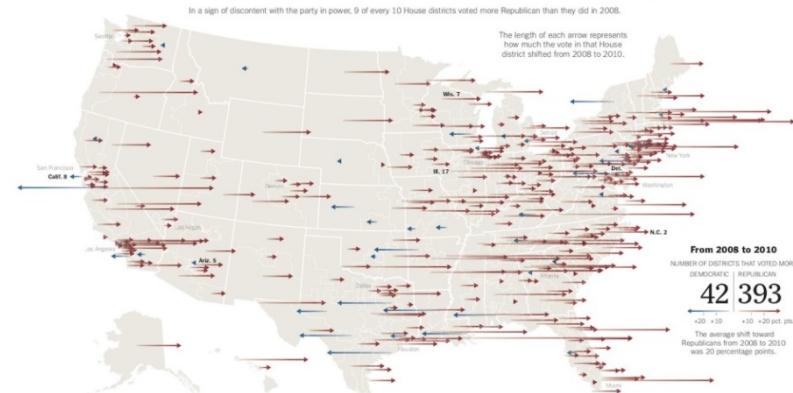
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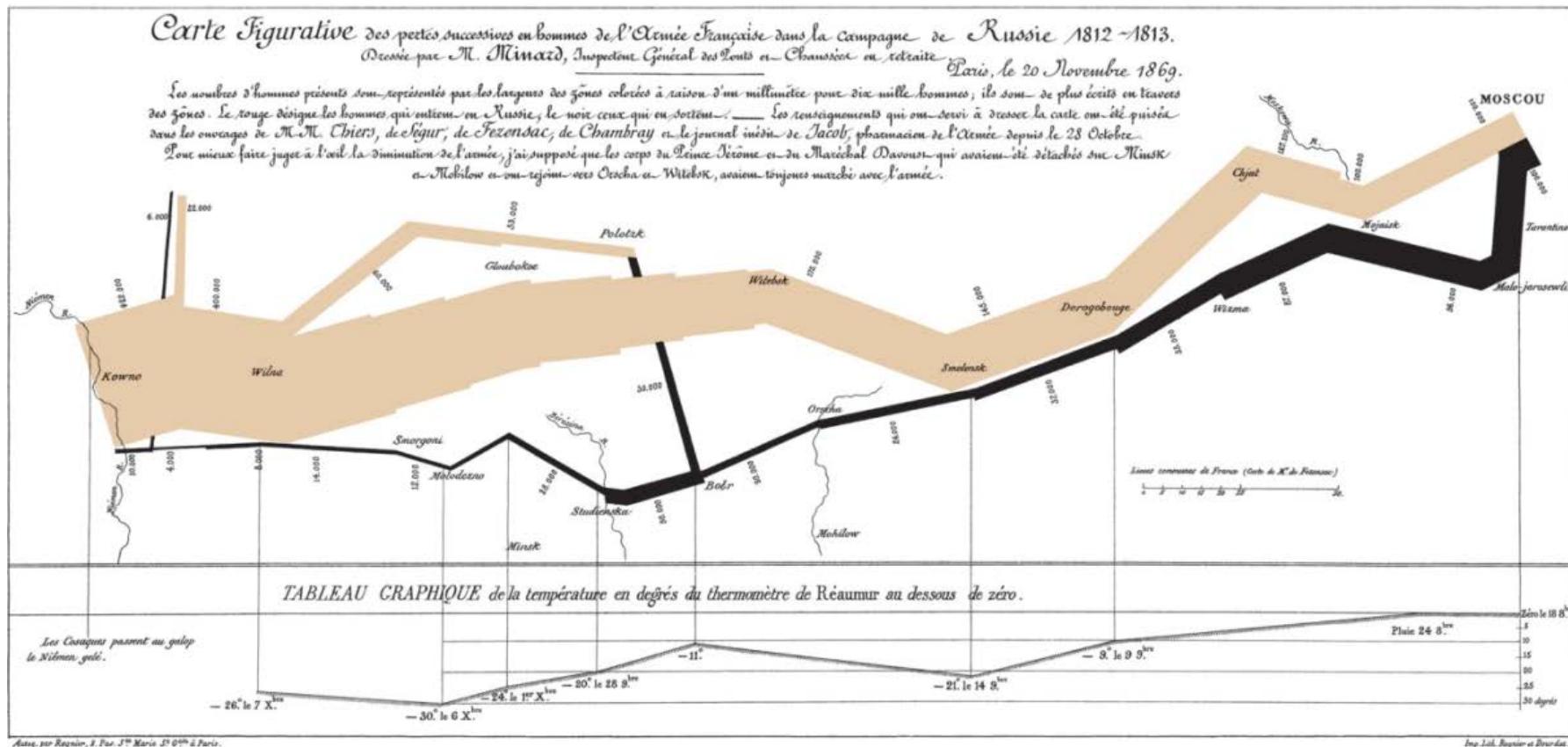


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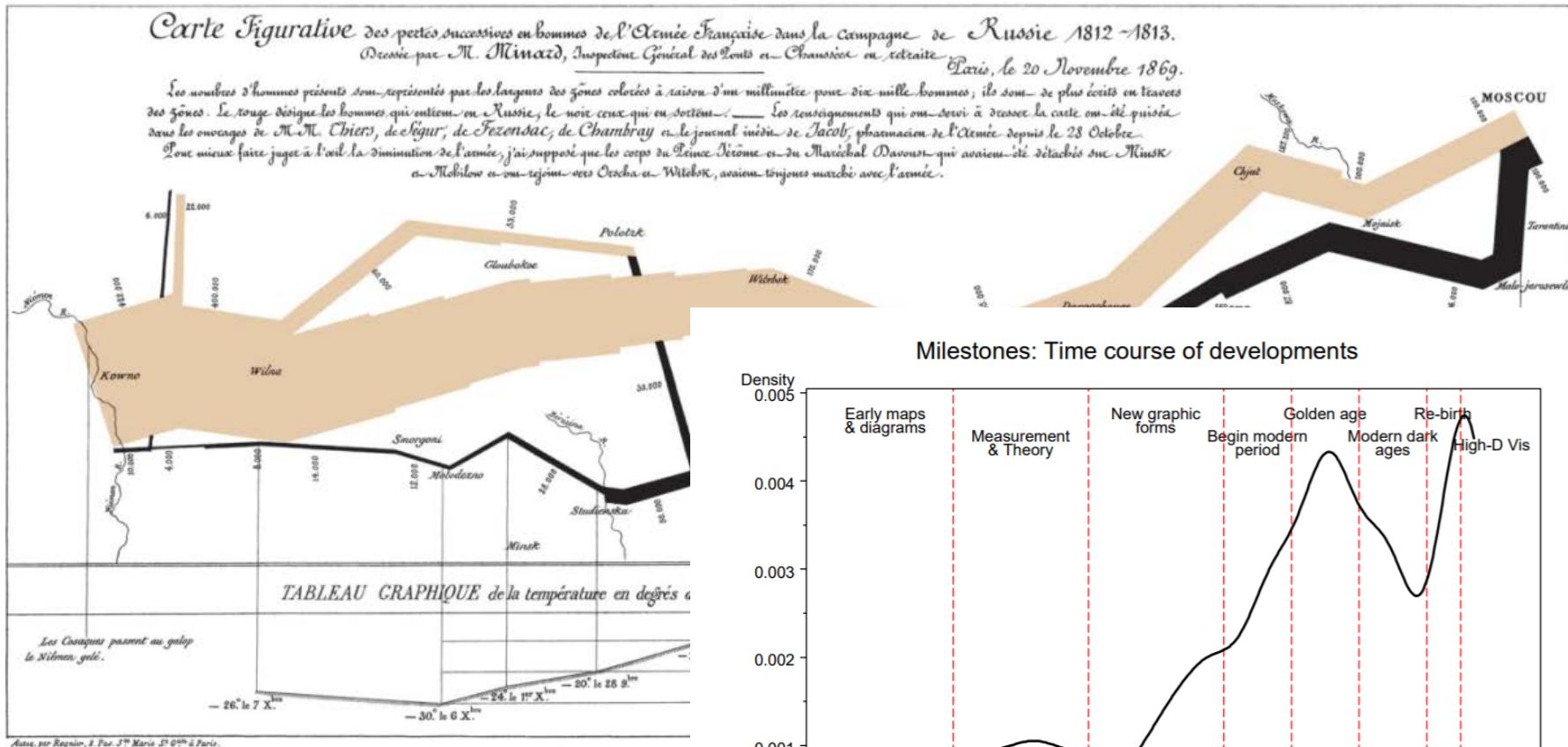
## Visualization is from all times



Battles of Napoleon in Russia

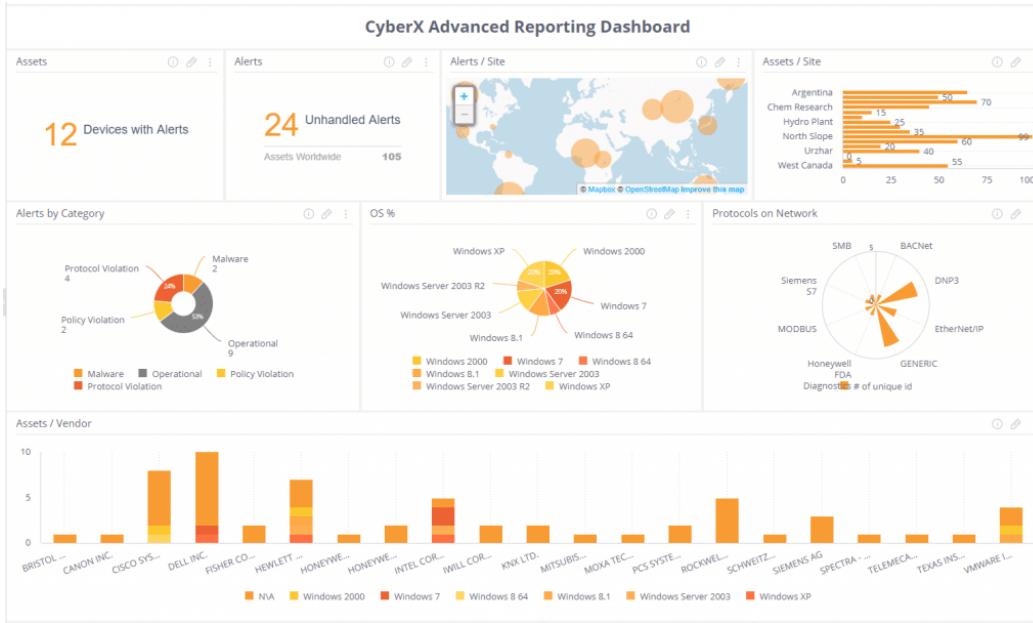
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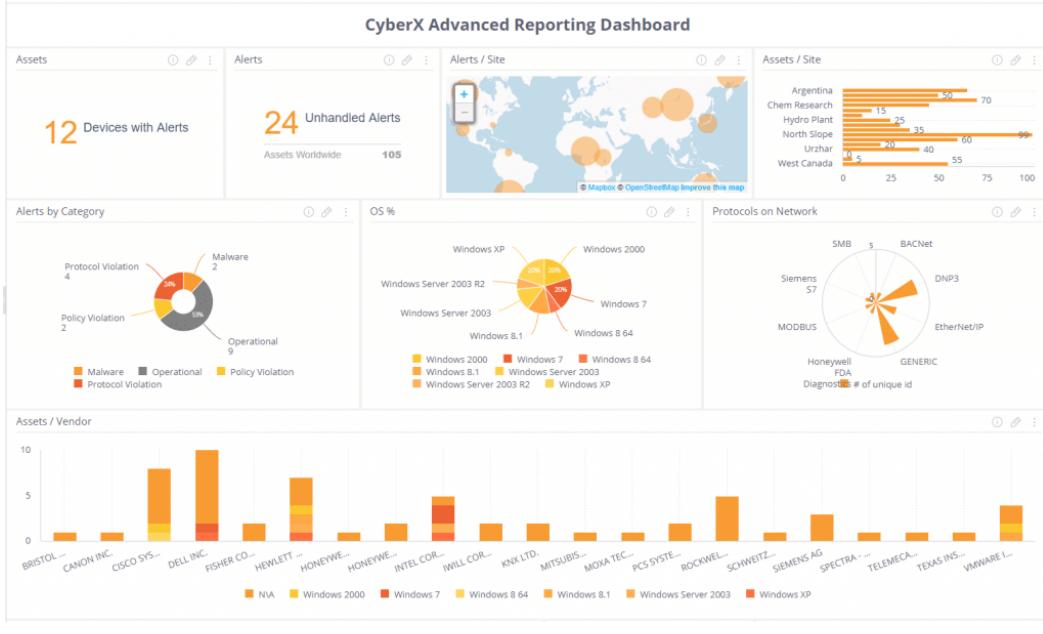


Battles of Napoleon in Russia

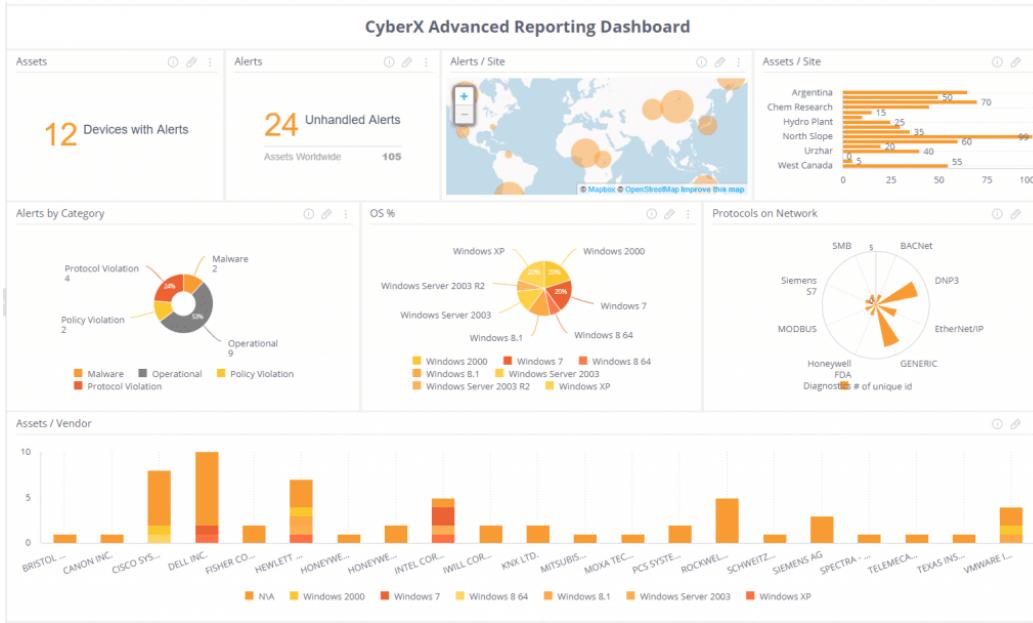
## 2. Visualization: not just engineering



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### Visual perception

### Effective communication

*Perfection is achieved, not when there is nothing more to add, but when there is nothing left to take away*

Antoine de Saint-Exupéry



### Truthful communication

How to avoid artefacts in the visualization that are no intrinsic features?

## 2. Visualization: not just engineering



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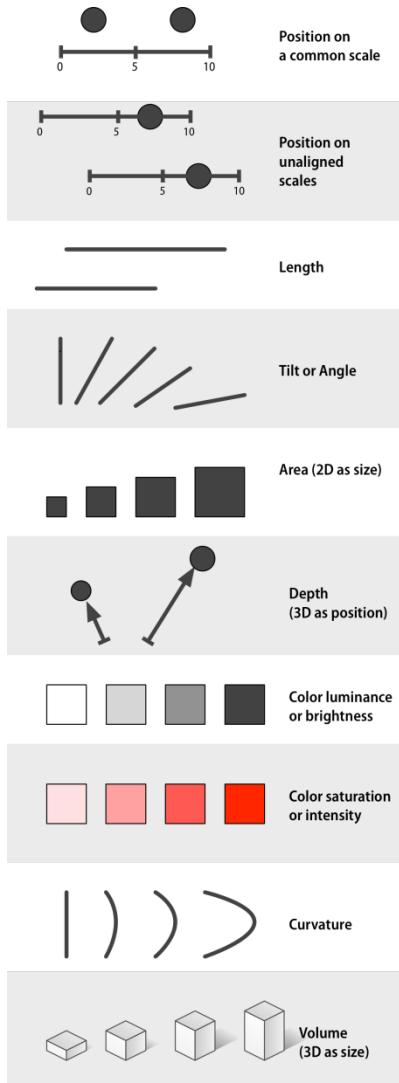
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## 2. Visualization: not just engineering



Design no Decoration

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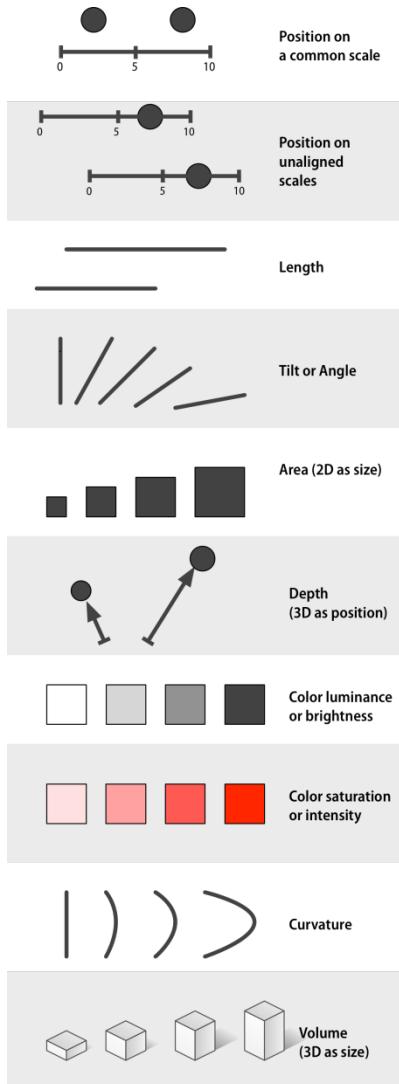
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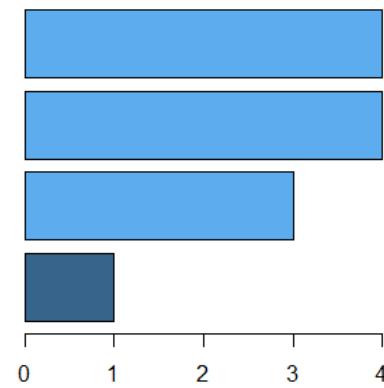
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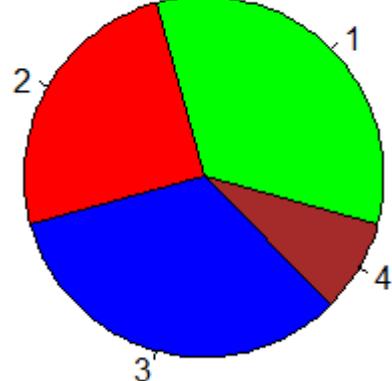
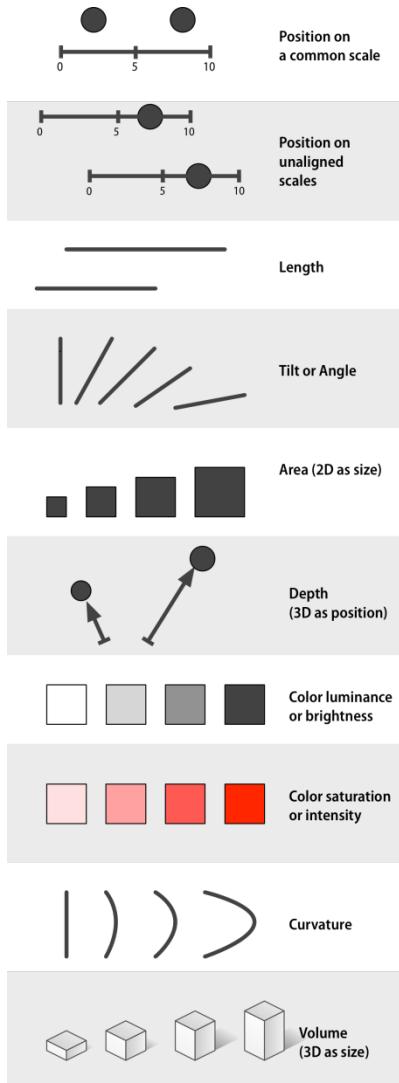
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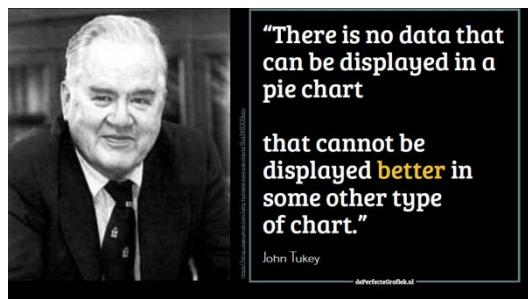
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De perfectografiek

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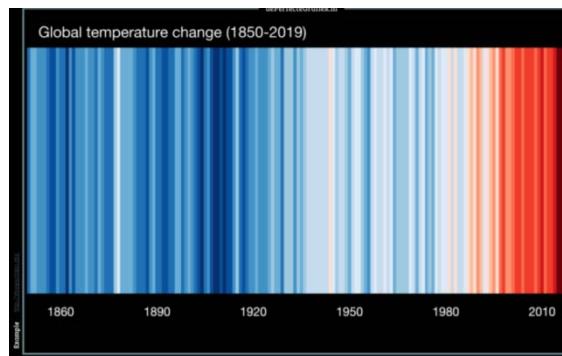
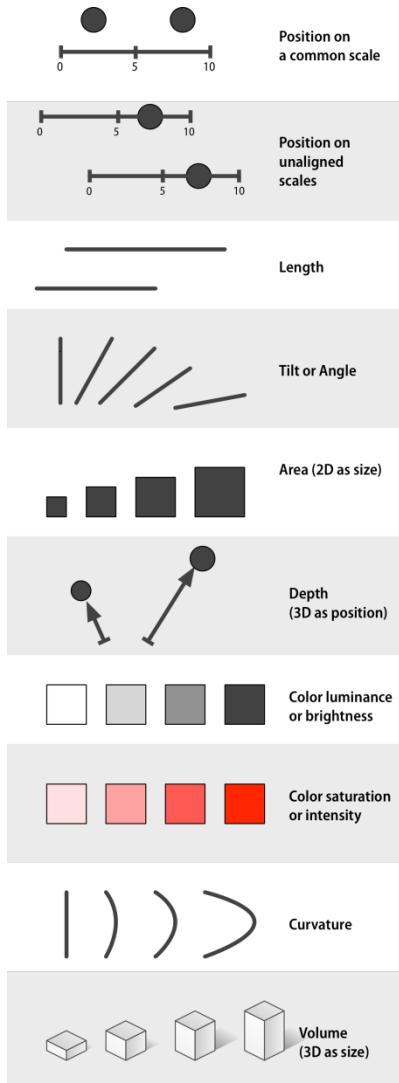
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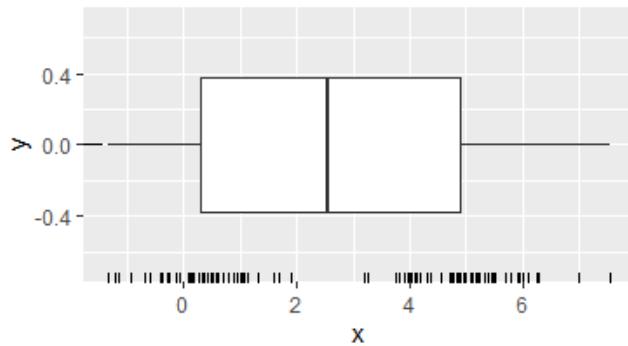
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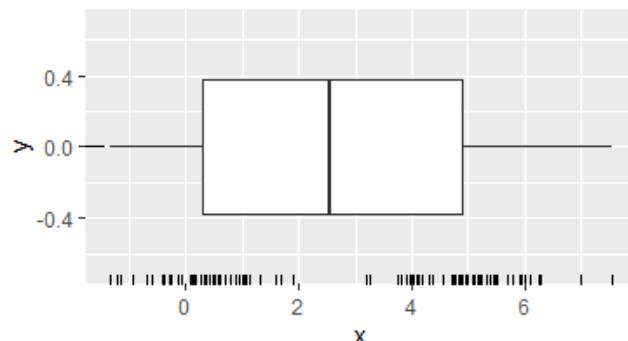


Visual perception

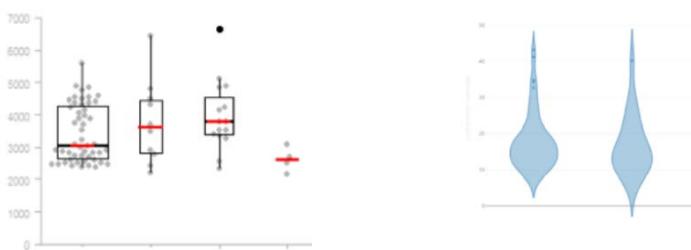
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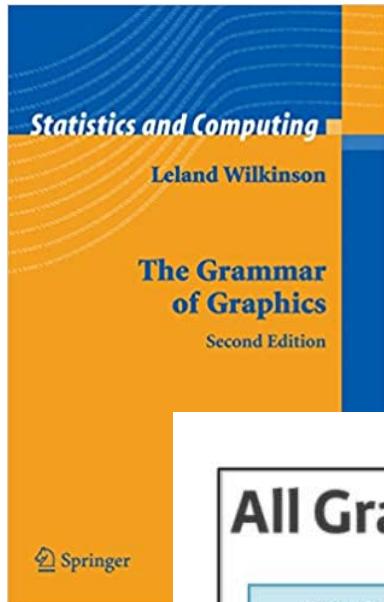


How to avoid artefacts in the visualization that are no intrinsic features?

Do not summarize too fast

### **3. Visualization: a guided walk**

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R: ggplot2 (Hadley Wickham)  
Python: plotnine



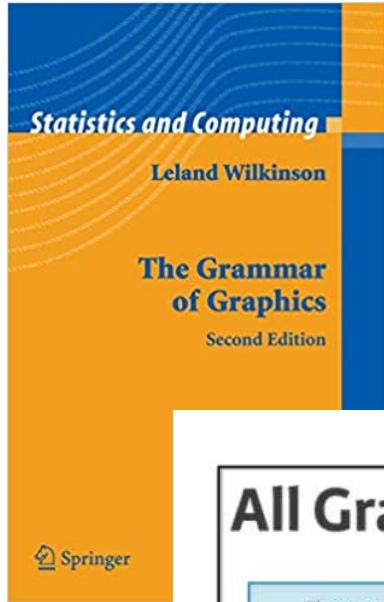
#### All Grammatical Elements

| Element     | Description                                       |
|-------------|---|
| Data        | The dataset being plotted.                        |
| Aesthetics  | The scales onto which we <i>map</i> our data.     |
| Geometries  | The visual elements used for our data.            |
| Facets      | Plotting small multiples.                         |
| Statistics  | Representations of our data to aid understanding. |
| Coordinates | The space on which the data will be plotted.      |
| Themes      | All non-data ink.                                 |

#### ggplot2 Layers - Themes



### 3. Visualization: a guided walk



R: ggplot2 (Hadley Wickham)  
Python: plotnine



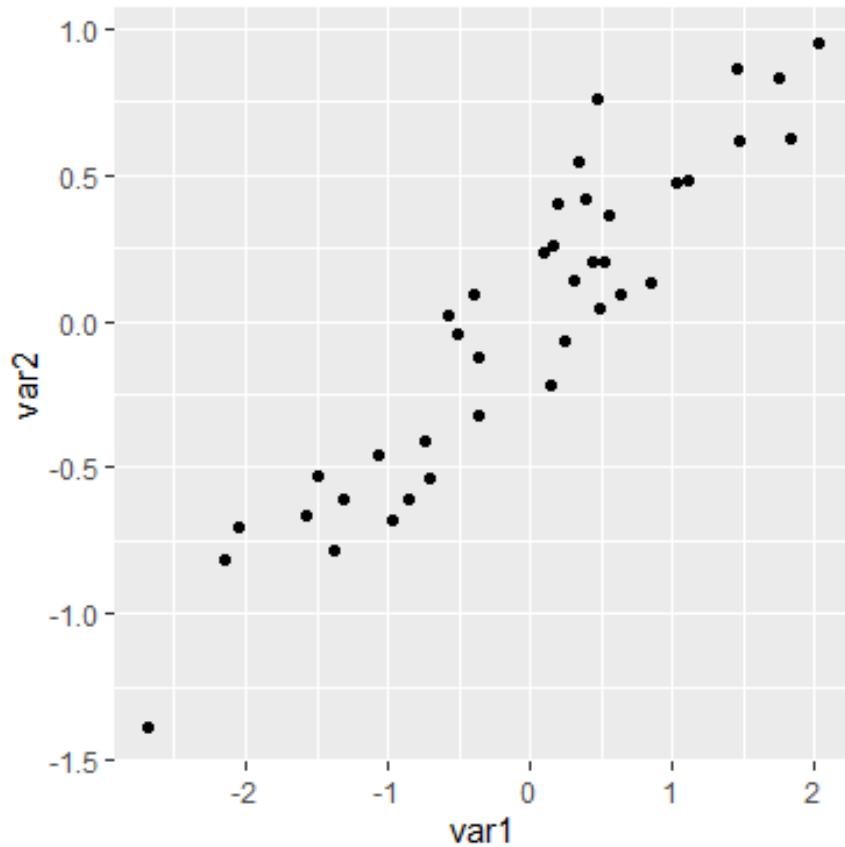
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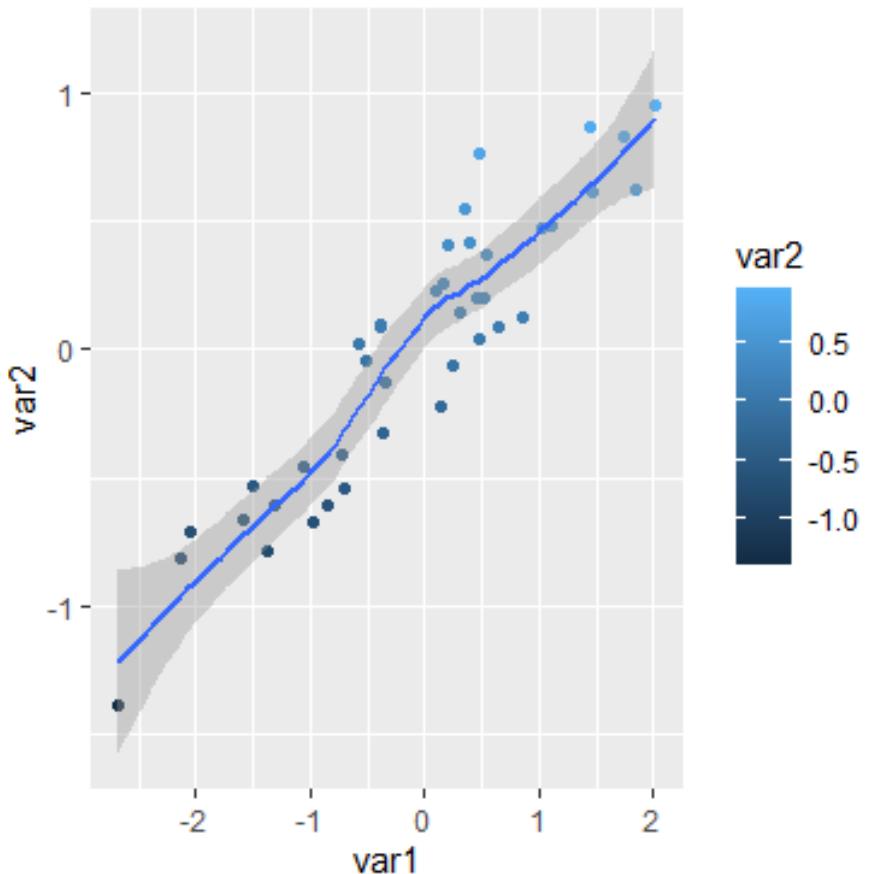
```
d<-data.frame("var1"=x, "var2"=y)
```

```
ggplot(data=d, aes(x=var1, y=var2)) + geom_point()
```

### ggplot2 Layers - Themes

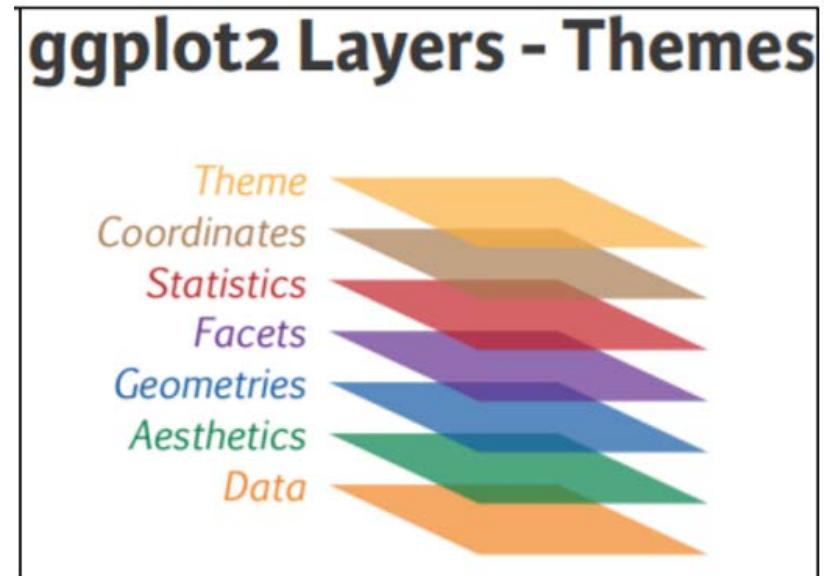


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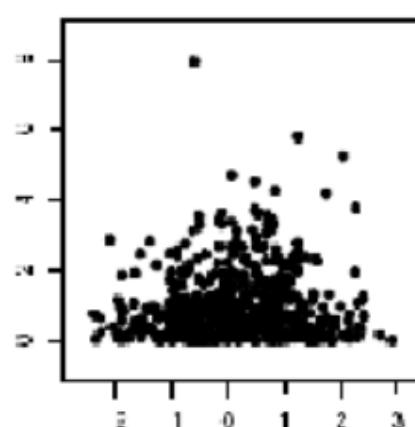
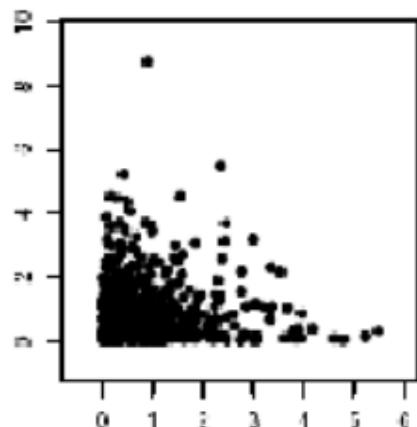
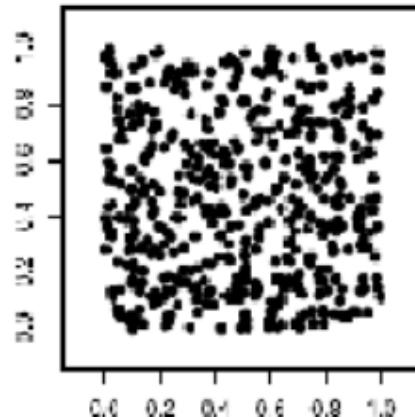
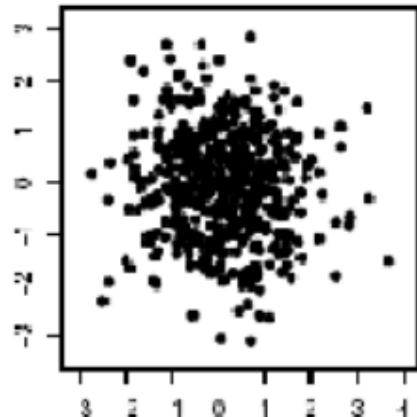


```
d<-data.frame("var1"=x, "var2"=y)
```

```
ggplot(data=d,aes(x=var1, y=var2, colour=var2)) + geom_point() + geom_smooth()
```

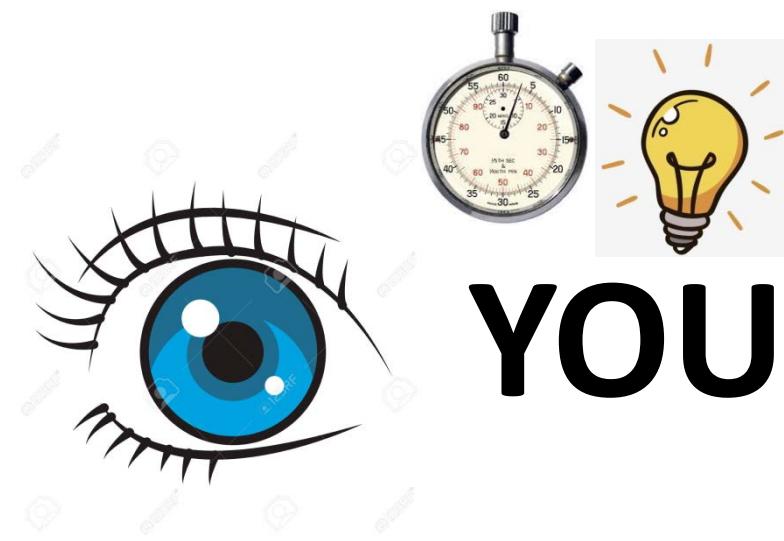
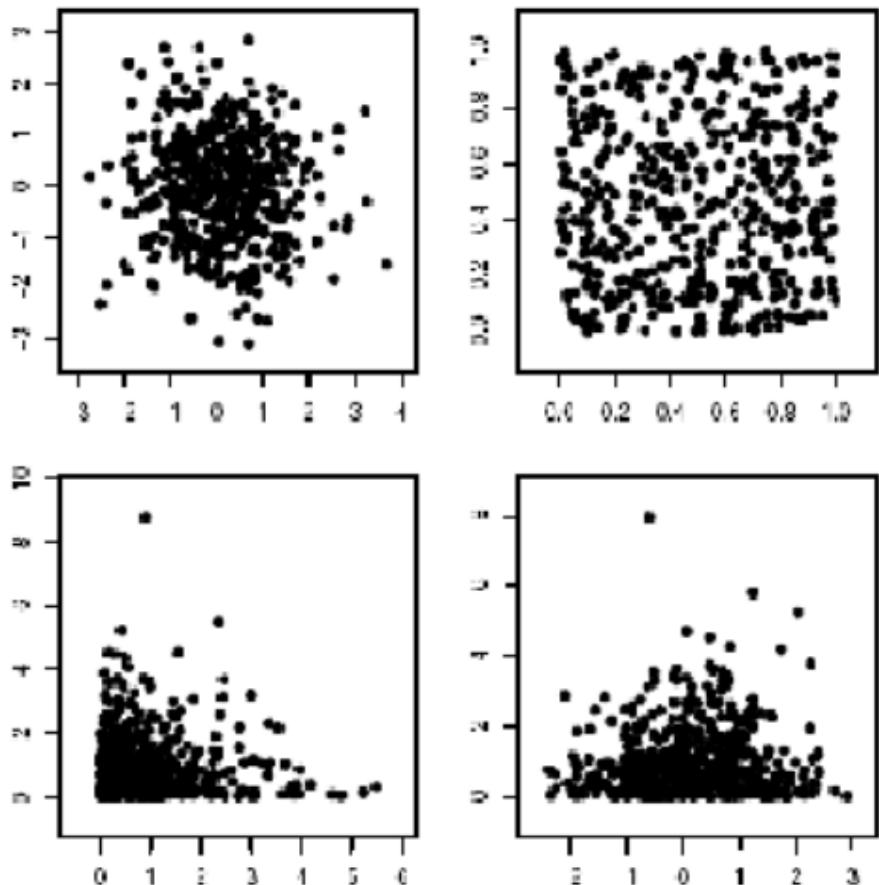


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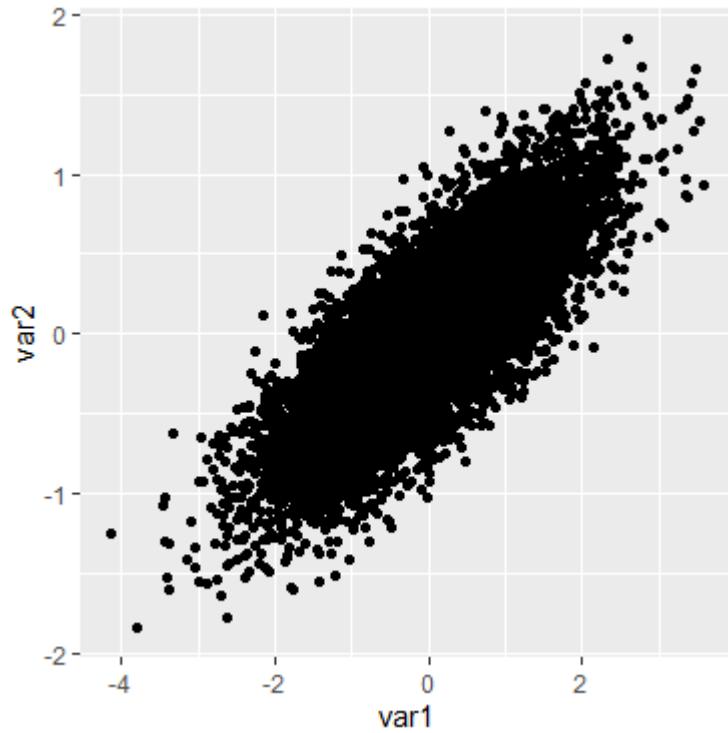
In which case there is a stronger dependency?

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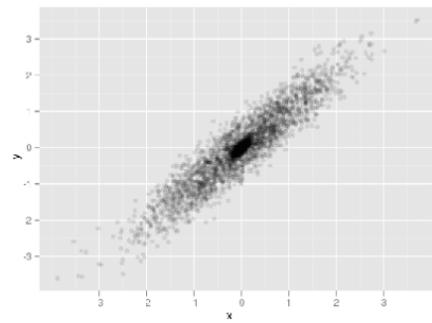
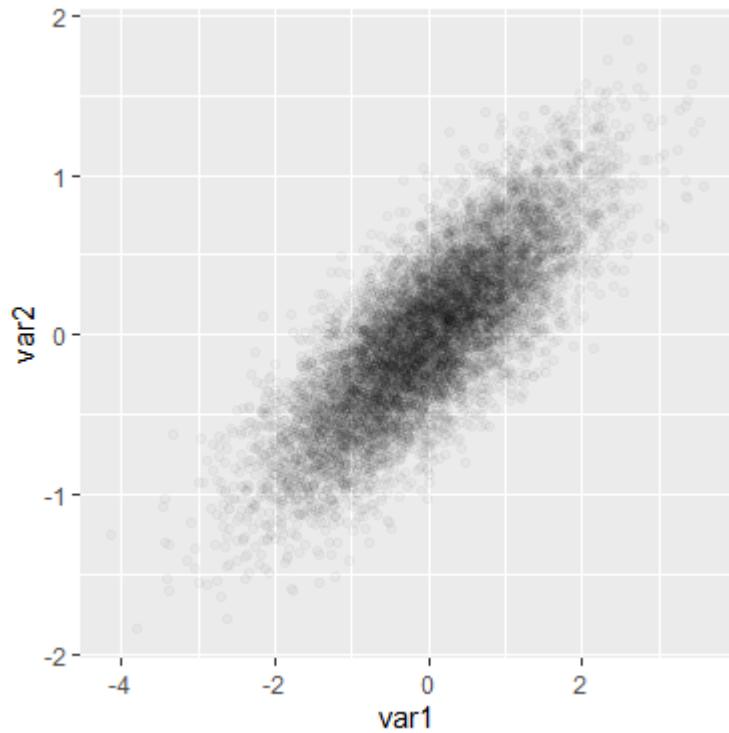
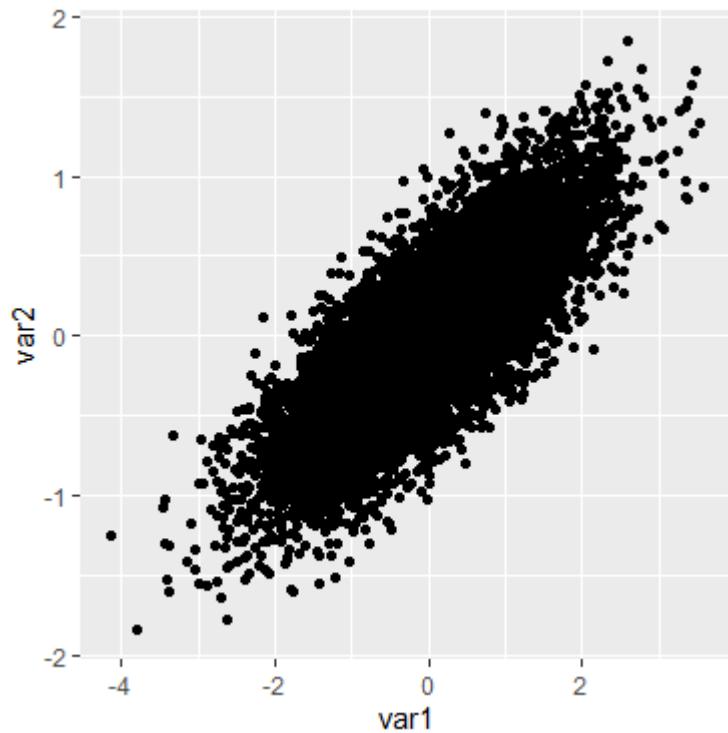


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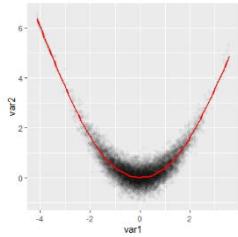
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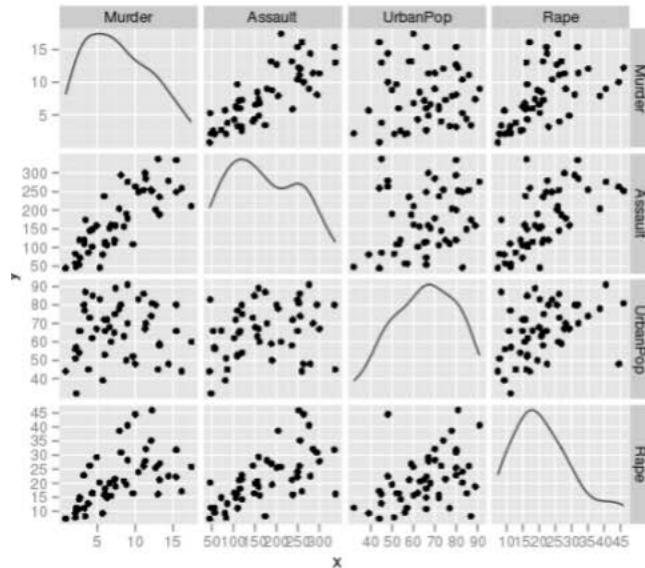
```
ggplot(data=d, aes(x=var1,y=var2)) +  
  geom_point(alpha=0.025)
```



### 3. Visualization: a guided walk

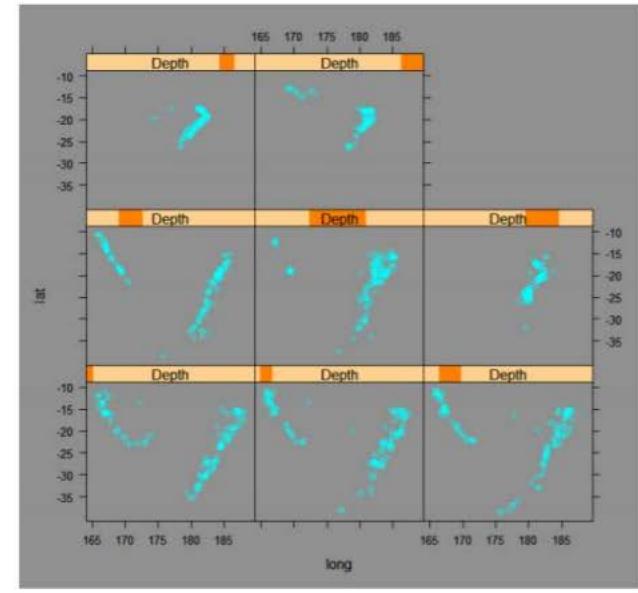
Up to higher dimensions:

abstraction/marginalize



Pairsplots

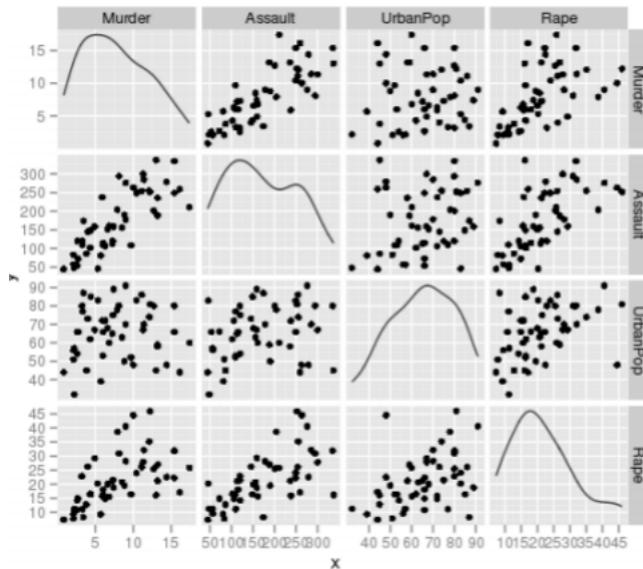
specification/conditionalize



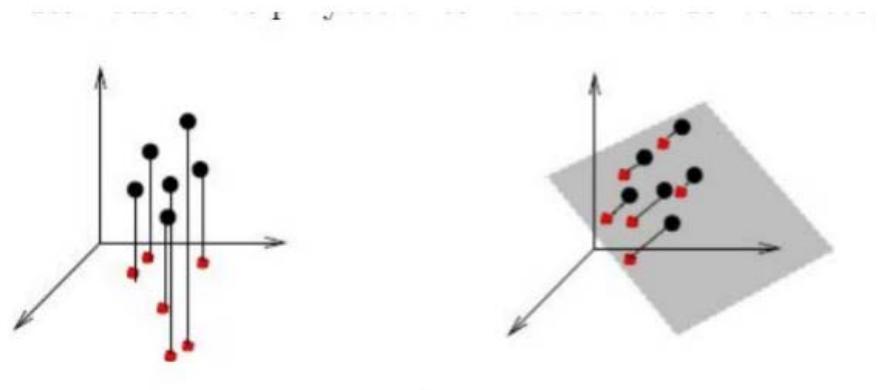
Trellisplots

### 3. Visualization: a guided walk

Up to higher dimensions:



Can we do better?

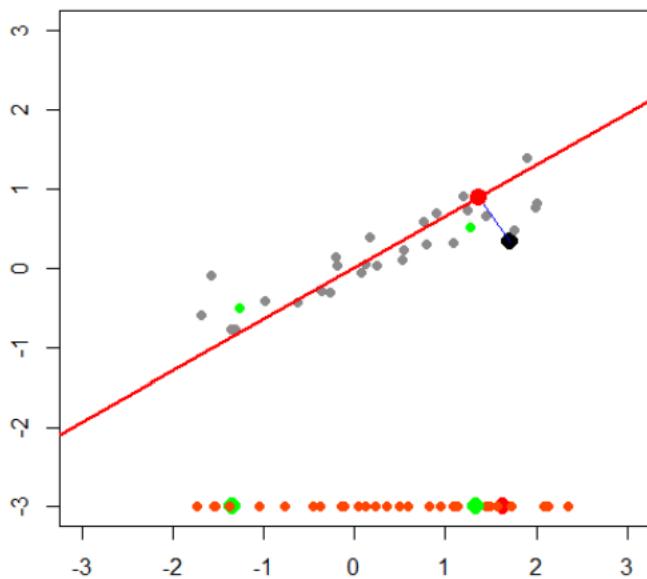


Dimension reduction techniques:

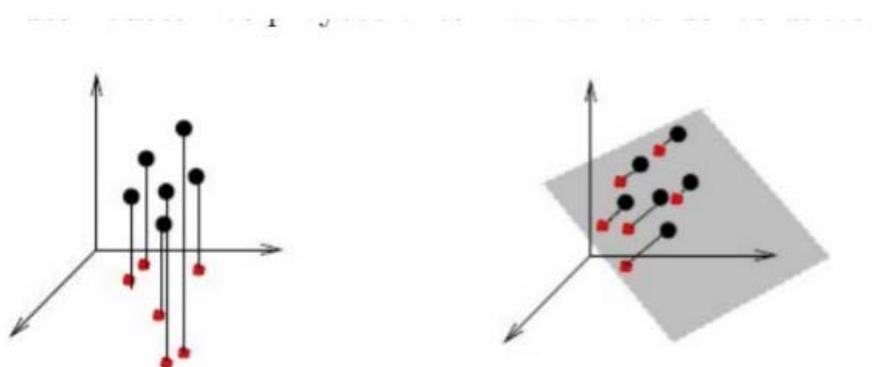
principal component analysis

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Up to higher dimensions:



Can we do better?



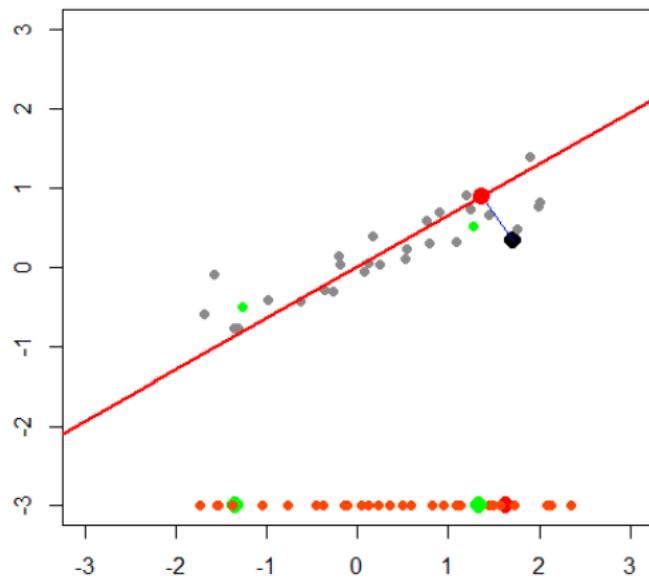
Dimension reduction techniques:

principal component analysis

Informative directions (max variance)  
Least approximation error  
Distance preserving transformation

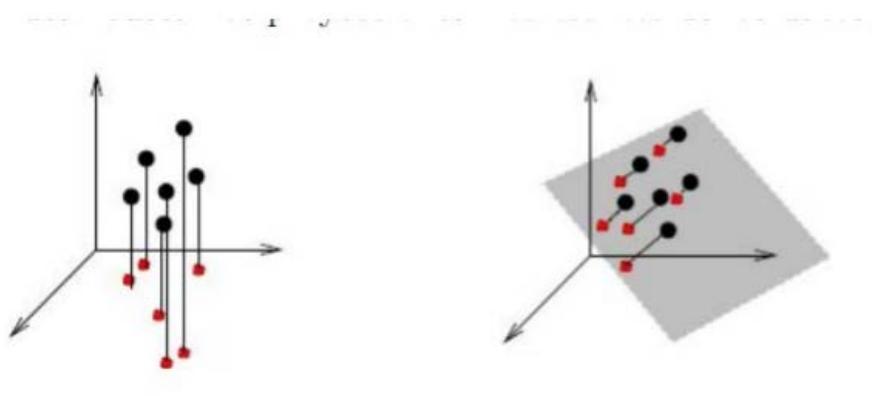
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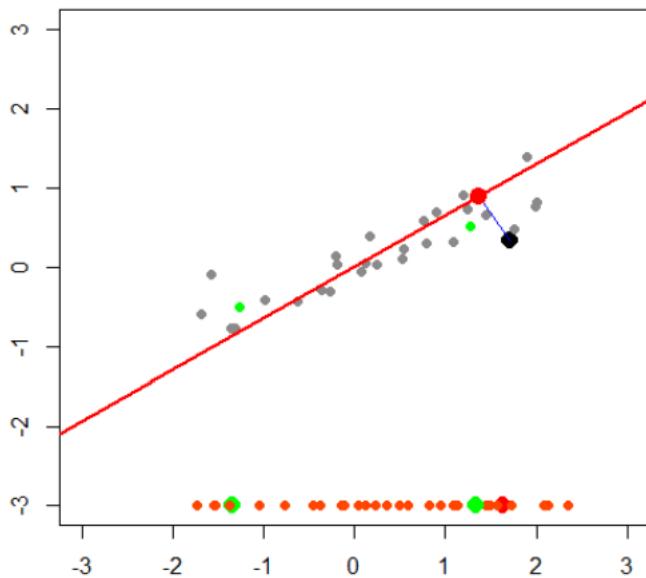
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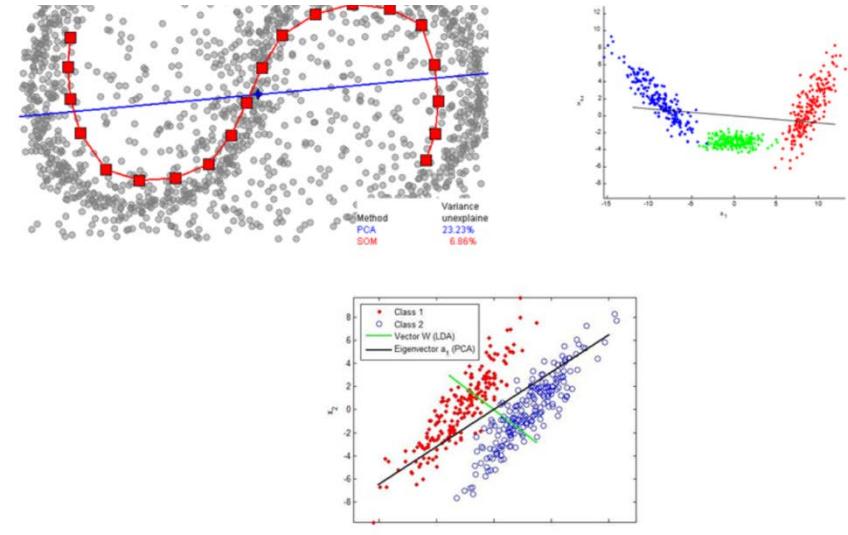
take care : suitable? Optimal?

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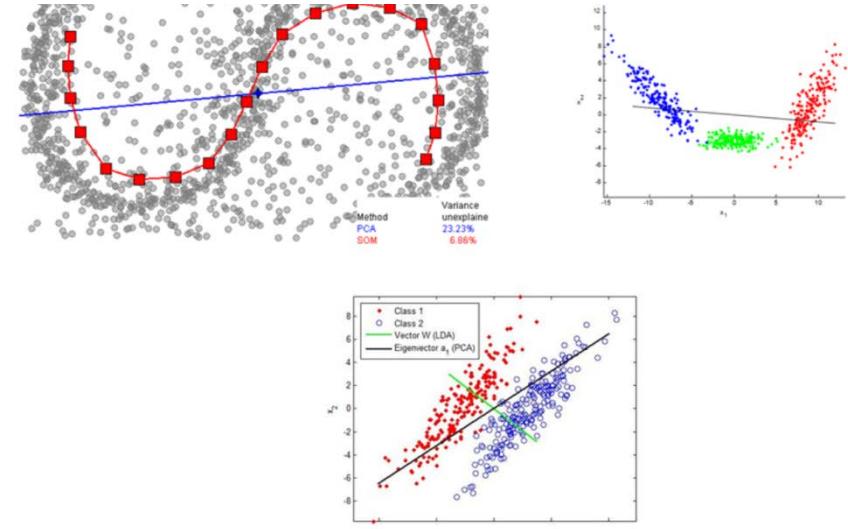
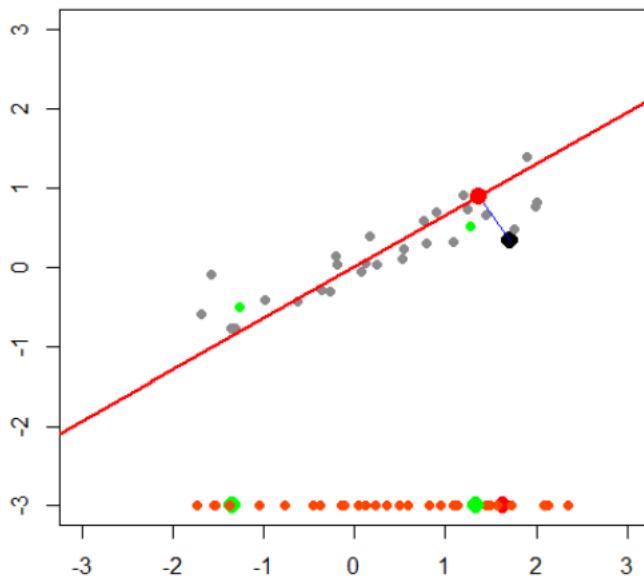
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Up to higher dimensions:



Dimension reduction techniques:

principal component analysis

Informative directions (max variance) (**ICA, PP**)

Least approximation error

Distance preserving transformation

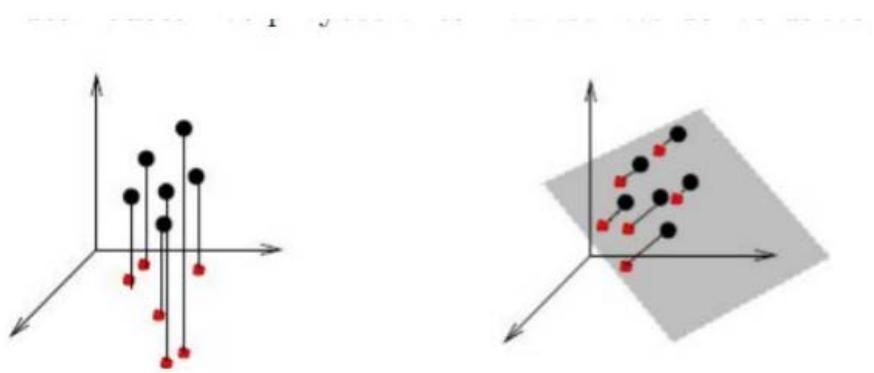
(**CDS, KernelPCA, T-SNE, ISOMAP, SOM, ...**)

take care : suitable? Optimal?

### 3. Visualization: a guided walk

Up to higher dimensions:

Can we do better?



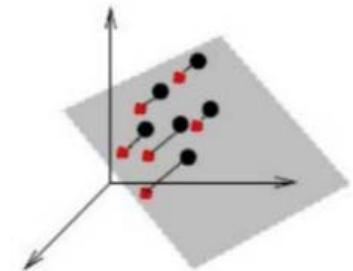
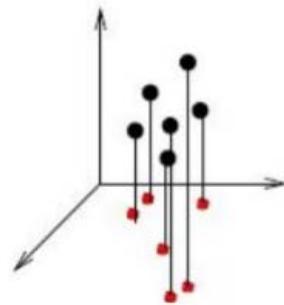
**Dimension reduction techniques:**

think a lot to decide how to proceed

### 3. Visualization: a guided walk

Up to higher dimensions:

Can we do better?



Dimension reduction techniques:

Make a **random** walk



think a lot to decide how to proceed

### 3. Visualization: a guided walk

Up to higher dimensions:

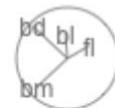
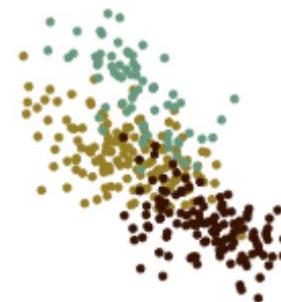


Example (Diana Cook)



Make a **random** walk

challenging rendering problem  
nice mathematical results



[https://www.dicook.org/files/visec2020/slides\\_tourr#1](https://www.dicook.org/files/visec2020/slides_tourr#1)

### **3. Visualization: a guided walk**

Up to higher dimensions:

**Next step: combine**

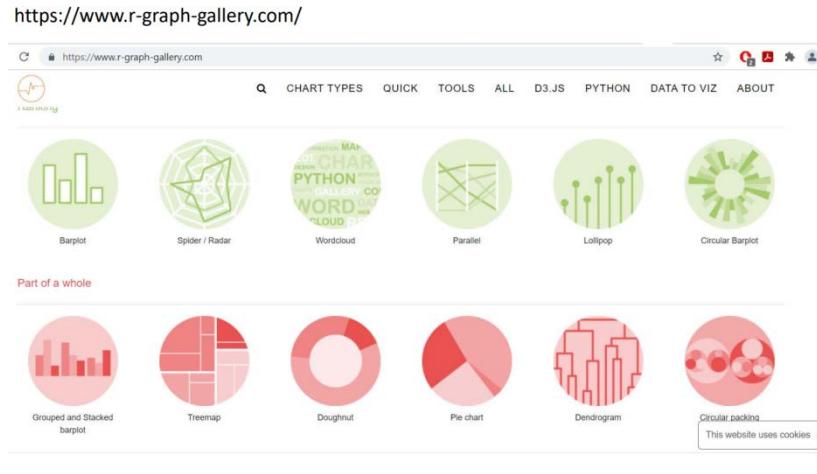
ggobi

# 4. Visualization: in practice

## Increasing popularity of **Galleries**

Python/R

<https://www.r-graph-gallery.com/>



Plotly R Open Source Graphing Library

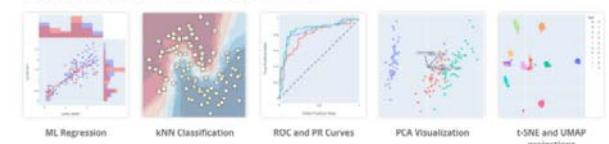
Plotly's R graphing library makes interactive, publication-quality graphs. Examples of how to make line plots, scatter plots, area charts, bar charts, error bars, box plots, histograms, heatmaps, subplots, multiple-axes, and 3D (WebGL based) charts. Plotly R is free and open source and you can [view the source](#), [report issues](#) or [contribute on GitHub](#).

Deploy R AI Dash apps on private Kubernetes clusters: [Pricing](#) | [Demo](#) | [Overview](#) | [AI App Services](#)

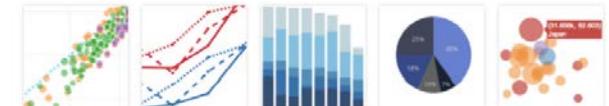
Fundamentals



Artificial Intelligence and Machine Learning



Basic Charts



**Dashboards:** Dash (Python/R),  
Shiny (R)

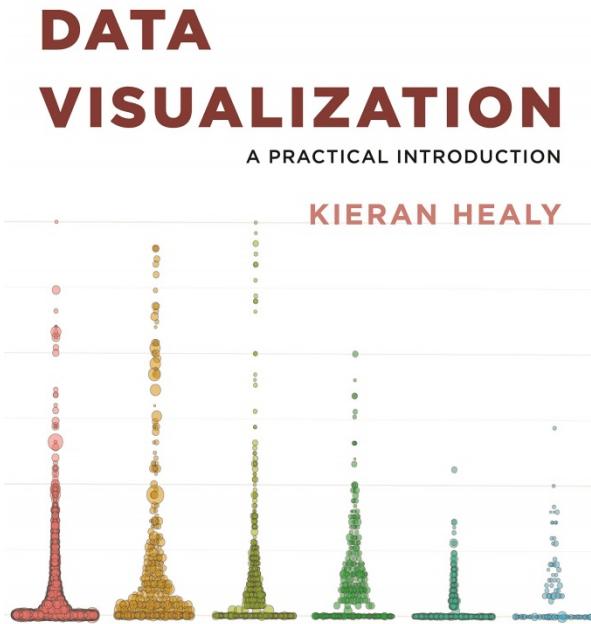
For working with **prototypes**:

<https://flourish.studio/>

Flourish\*



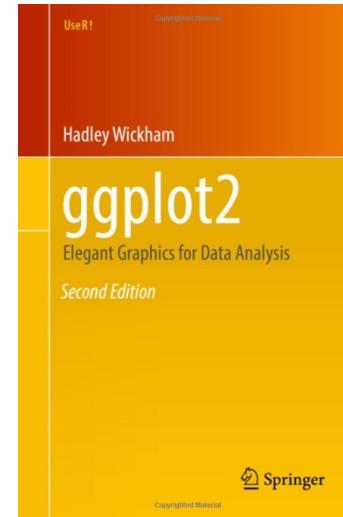
## 4. Visualization: in practice



<https://socviz.co/>

Good practices (Michiel Dullaert)

<https://chart.guide/>



<https://ggplot2-book.org/>