

Online Course
**Data Visualization
for Professionals**



THE UNIVERSITY
of EDINBURGH

Benjamin Bach

May 2022

<http://benjbach.me>

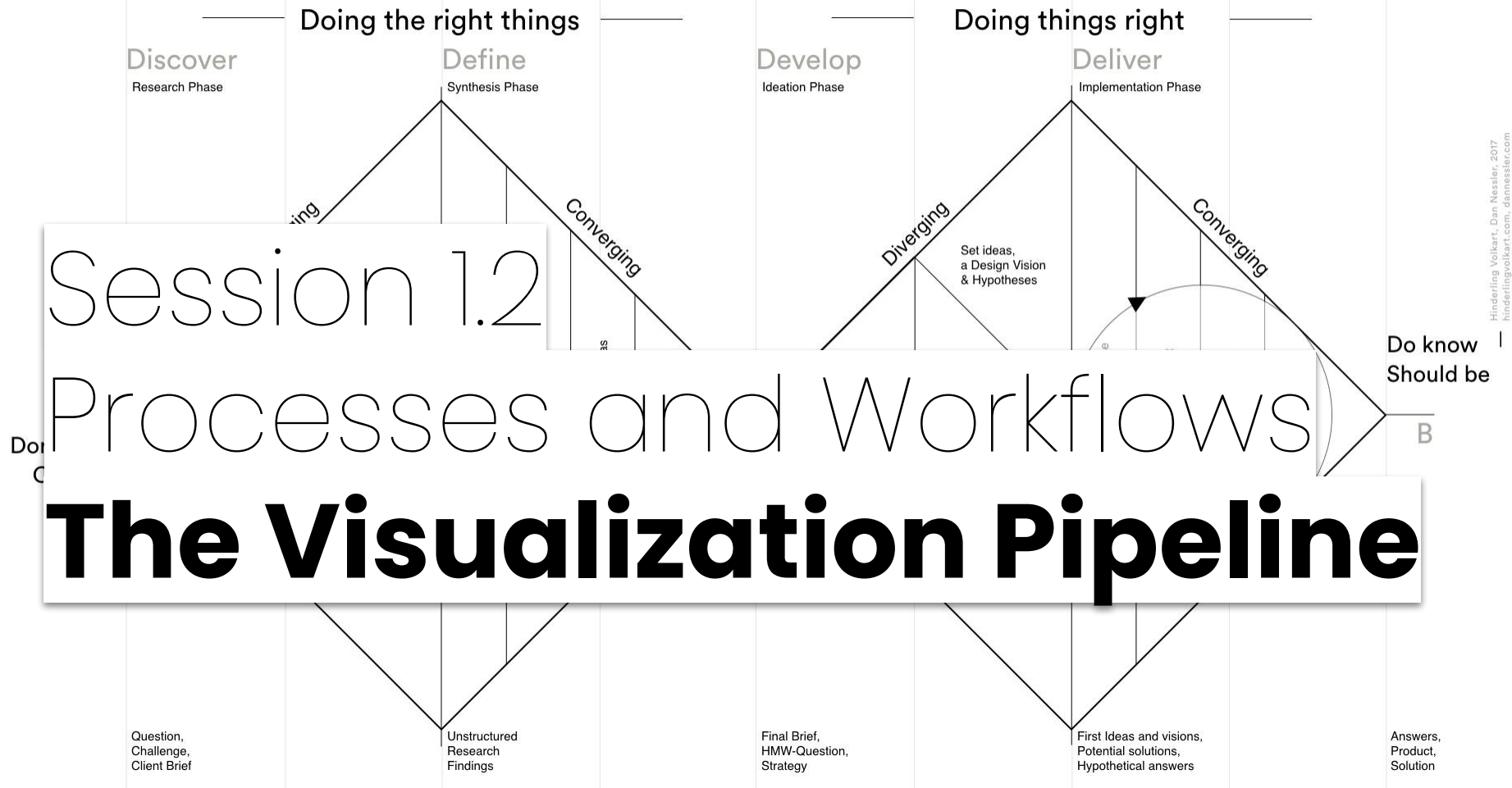
<https://datavis-online.github.io>

-- Not for external use --

Outline

Which processes are involved around understanding and creating visualizations?

1. The visualization pipeline
2. Design Thinking
3. Defining a visualization challenge
4. Exploratory data analysis



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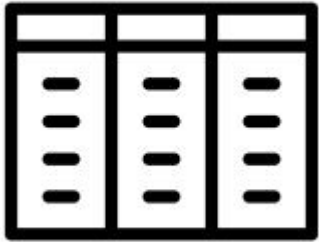
May 2022

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Sense making process



Data

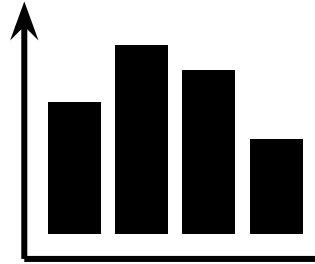
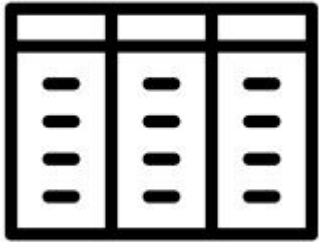
*Numbers,
relations,
records,
text,
analysis, ...*



Action

*Decisions
Emotions,
Knowledge*

Sense making process



Data

*Numbers,
relations,
records,
text,
analysis, ...*



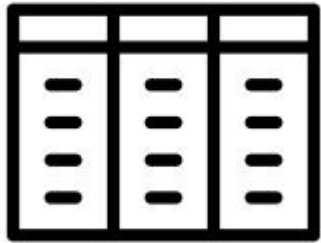
Visualization

*visual
representation*

Action

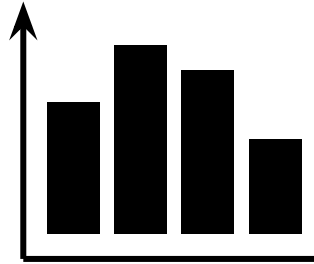
*Decisions
Emotions,
Knowledge*

Sense making process



Data

*Numbers,
relations,
records,
text,
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Visualization

*visual
representation*



Information

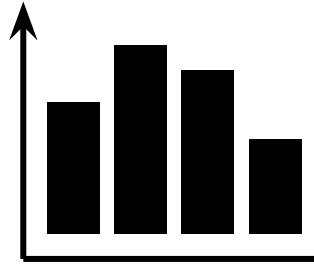
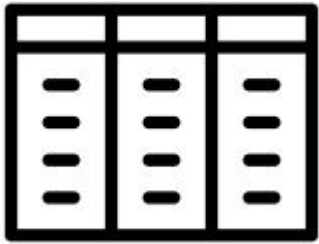
*Insights,
Facts*



Action

*Decisions
Emotions,
Knowledge*

Sense making process



Data



Visualization



Information



Action

Collect

Format

Parse

Filter

Analyze

Visualize

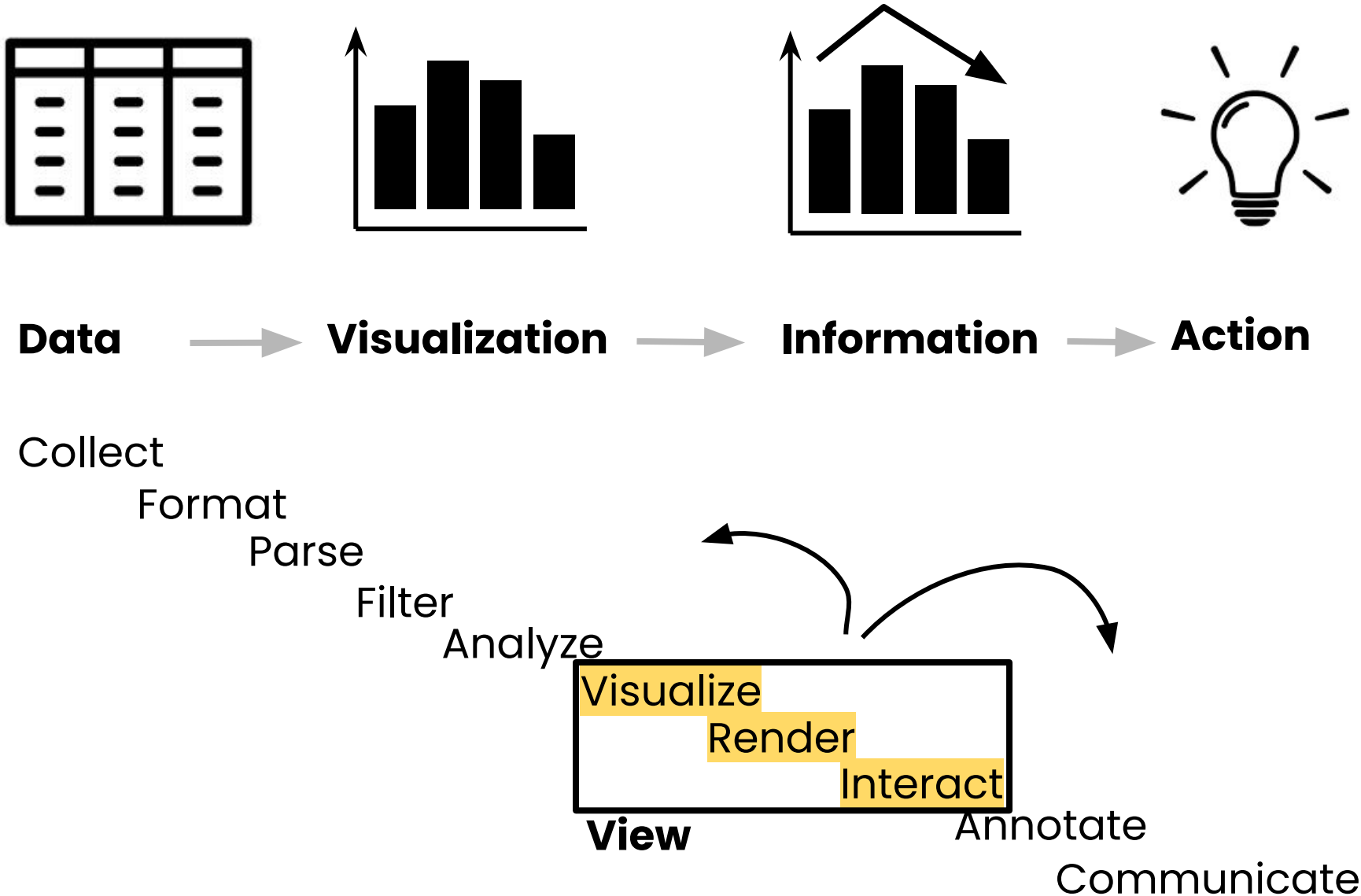
Render

Interact

Annotate

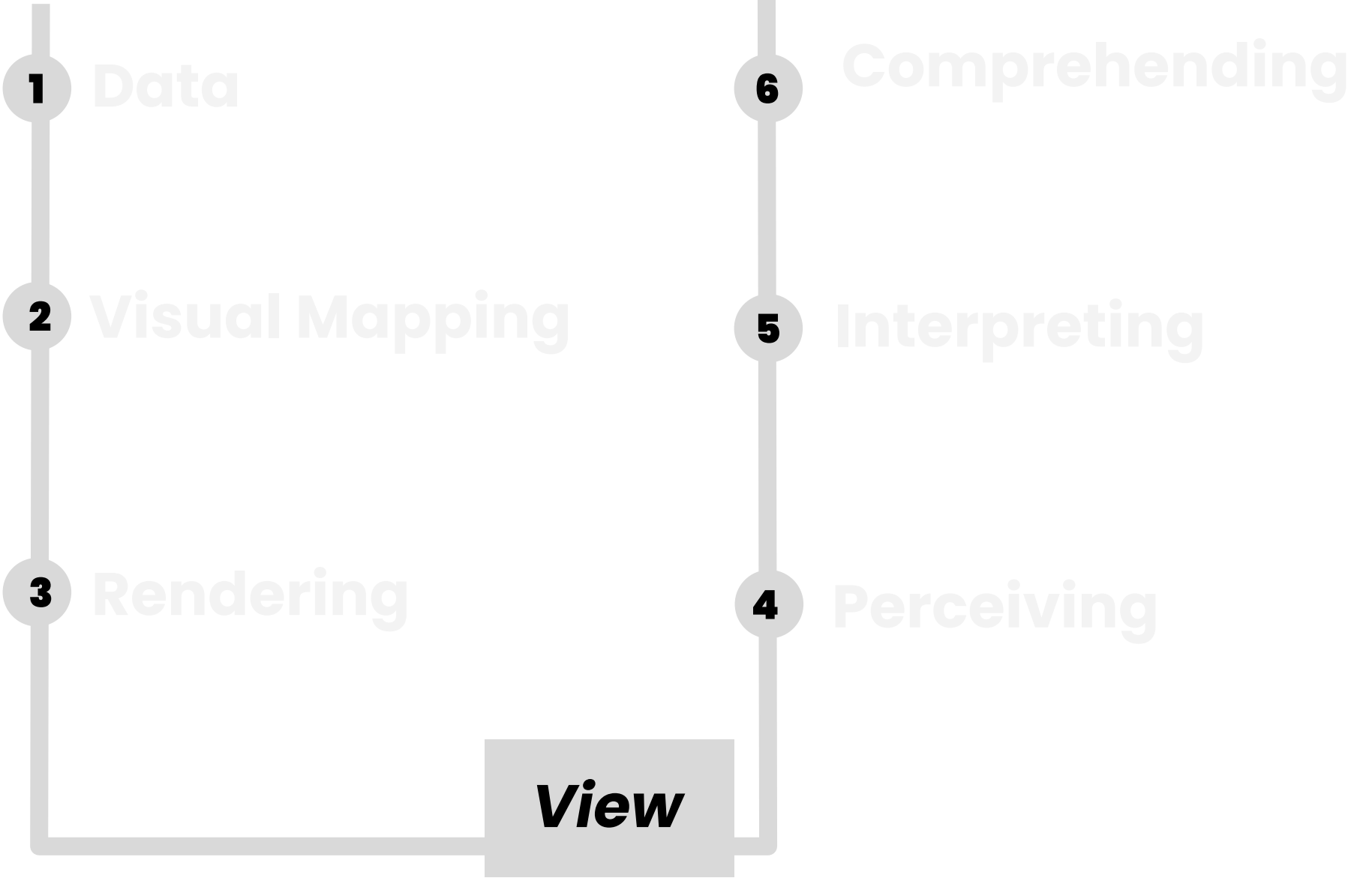
Communicate

Sense making process



Encoding: designer

Decoding: user



Encoding: designer

Decoding: user

1

Data

What is my data?

Which data type?

Ordinal / numerical / categorical?

2

Visual Mapping

3

Rendering

6

Comprehending

5

Interpreting

4

Perceiving

View

Encoding: designer

Decoding: user

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Ordinal / numerical / categorical?

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What is my visual representation?

Which visual variables am I using?

How am I encoding my data?

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Rendering

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View

Encoding: designer

Decoding: user

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Rendering

What is my medium?

monoscopic/stereoscopic?

Tangiblity?

Print / digital?

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Decoding: user

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What does it show?

Where is big, medium, small?

How do things compare?

What relationships exist?

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What does 'up' mean?

What do these patterns show?

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Decoding: user

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What shall I do now?

Is this all true?

What do I learn?

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1. Data

1		mpg	cyl	disp	hp	drat	wt	qsec
2	Mazda RX4	21	6	160	110	3.9	2.62	16.46
3	Mazda RX4 Wag	21	6	160	110	3.9	2.875	17.02
4	Datsun 710	22.8	4	108	93	3.85	2.32	18.61
5	Hornet 4 Drive	21.4	6	258	110	3.08	3.215	19.44
6	Hornet Sportabout	18.7	8	360	175	3.15	3.44	17.02
7	Valiant	18.1	6	225	105	2.76	3.46	20.22
8	Duster 360	14.3	8	360	245	3.21	3.57	15.84
9	Merc 240D	24.4	4	146.7	62	3.69	3.19	20
10	Merc 230	22.8	4	140.8	95	3.92	3.15	22.9
11	Merc 280	19.2	6	167.6	123	3.92	3.44	18.3
12	Merc 280C	17.8	6	167.6	123	3.92	3.44	18.9
13	Merc 450SE	16.4	8	275.8	180	3.07	4.07	17.4
14	Merc 450SL	17.3	8	275.8	180	3.07	3.73	17.6
15	Merc 450SLC	15.2	8	275.8	180	3.07	3.78	18
16	Cadillac Fleetwood	10.4	8	472	205	2.93	5.25	17.98

1. Data

Item

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1. Data

Attribute

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1. Data

Value Attribute

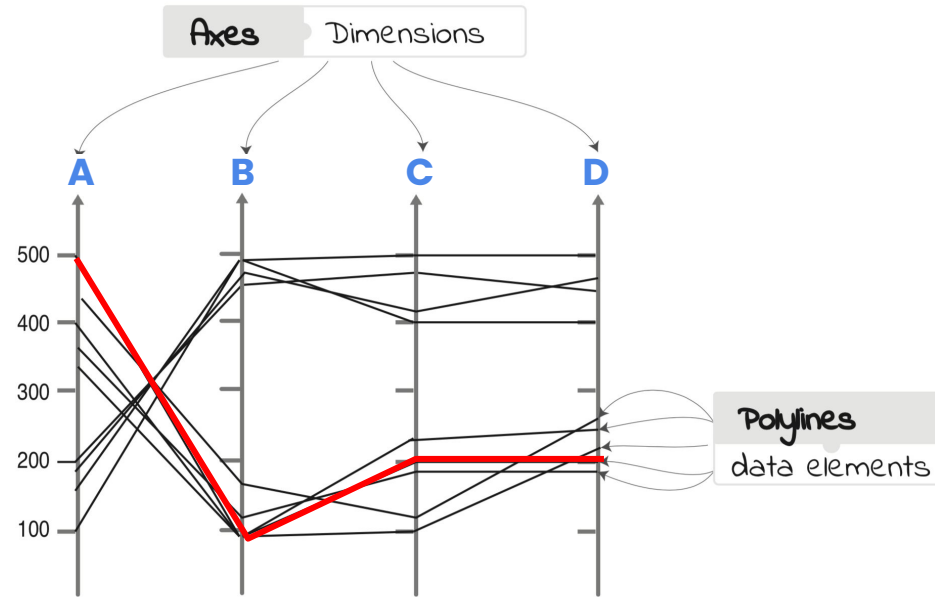
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Data

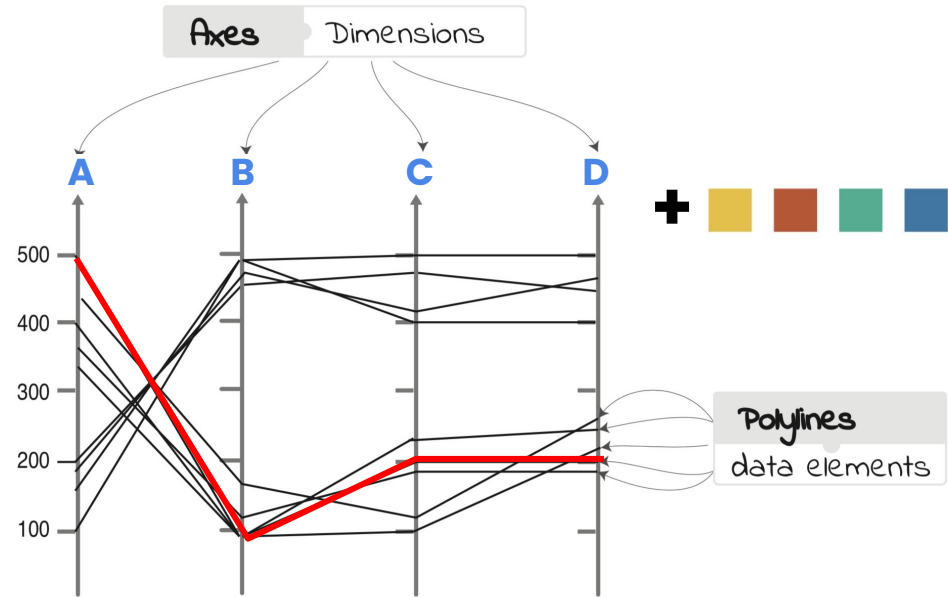


**Visual
Representation**

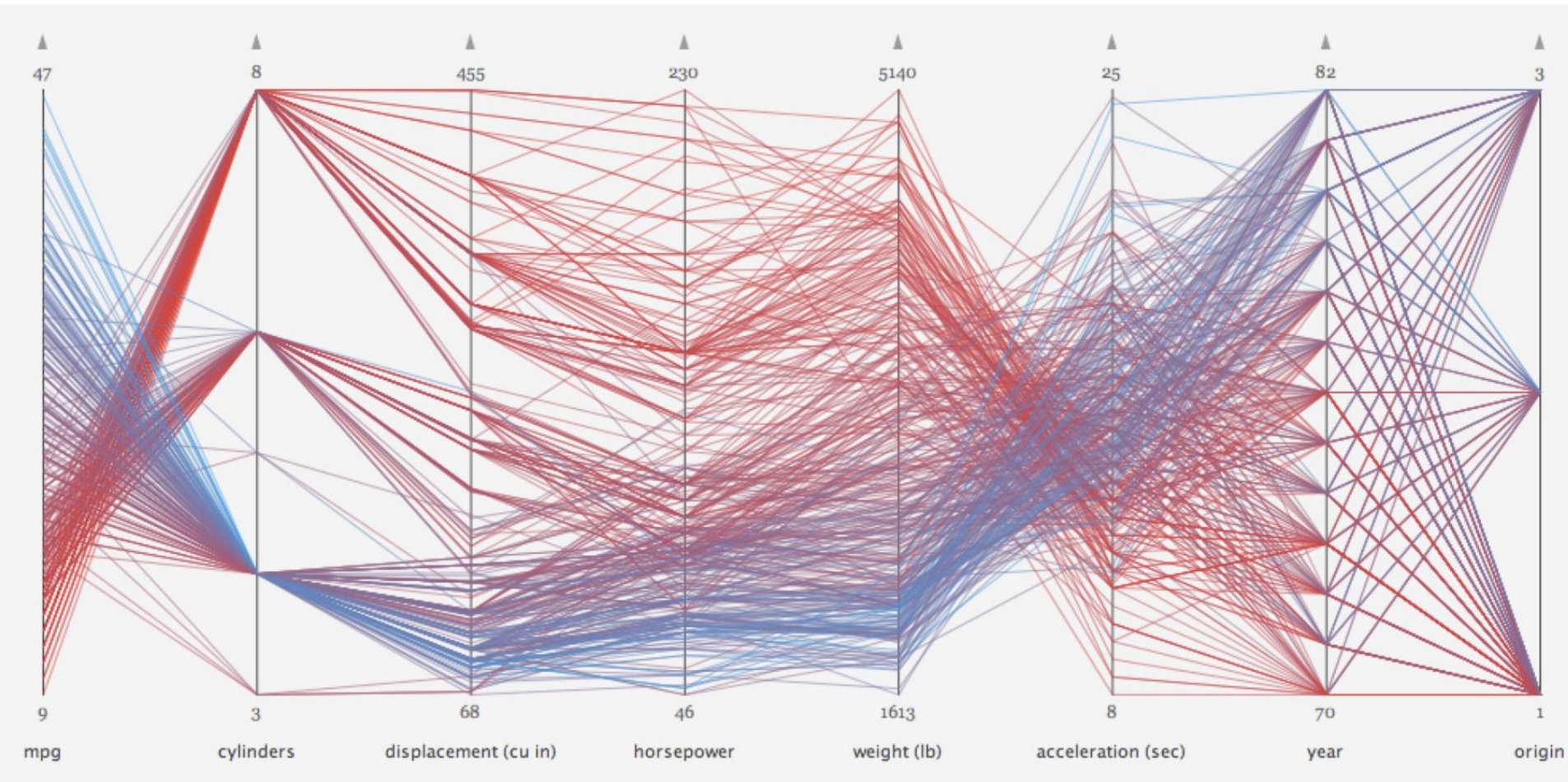
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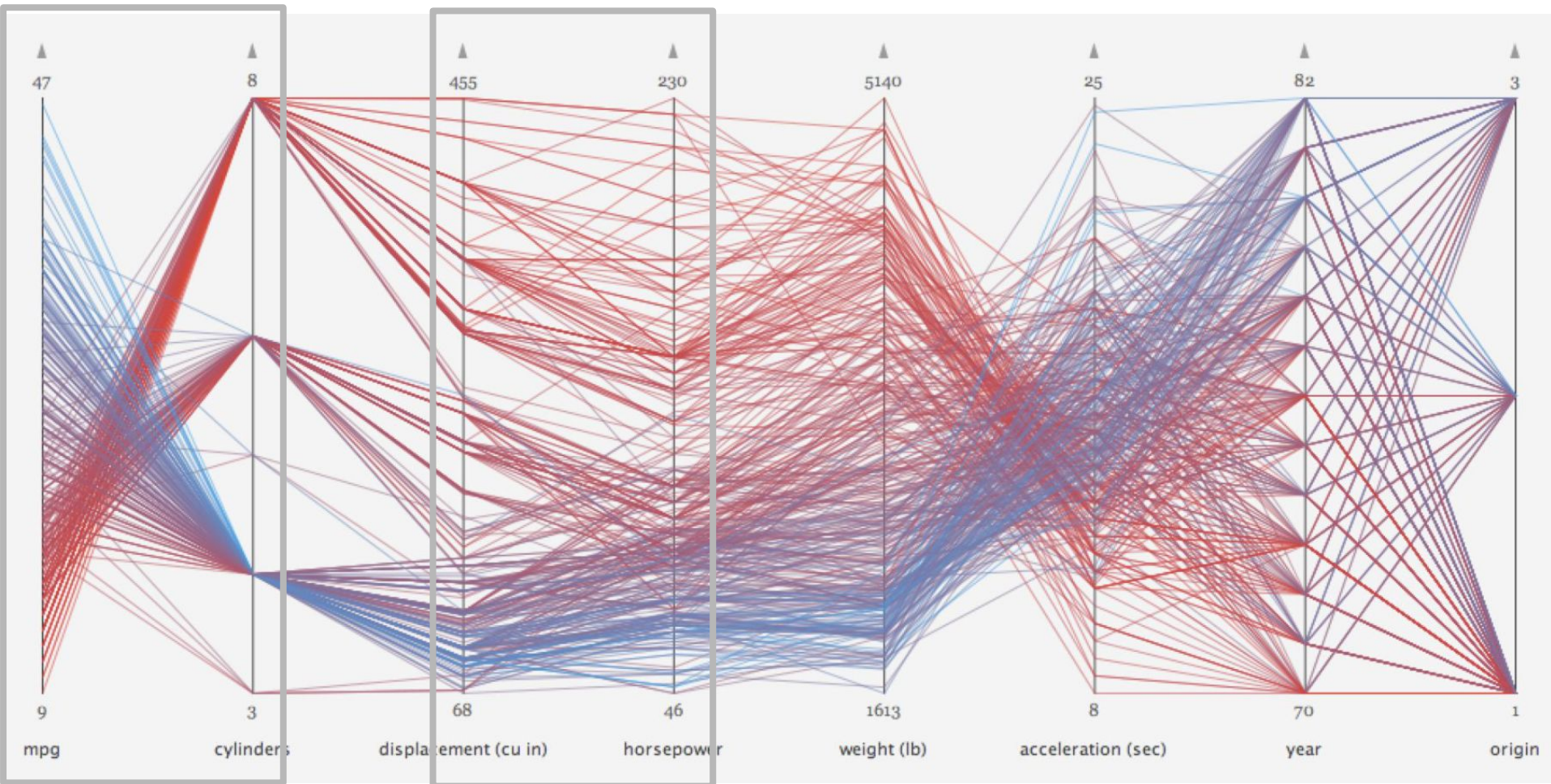
Data



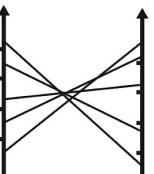
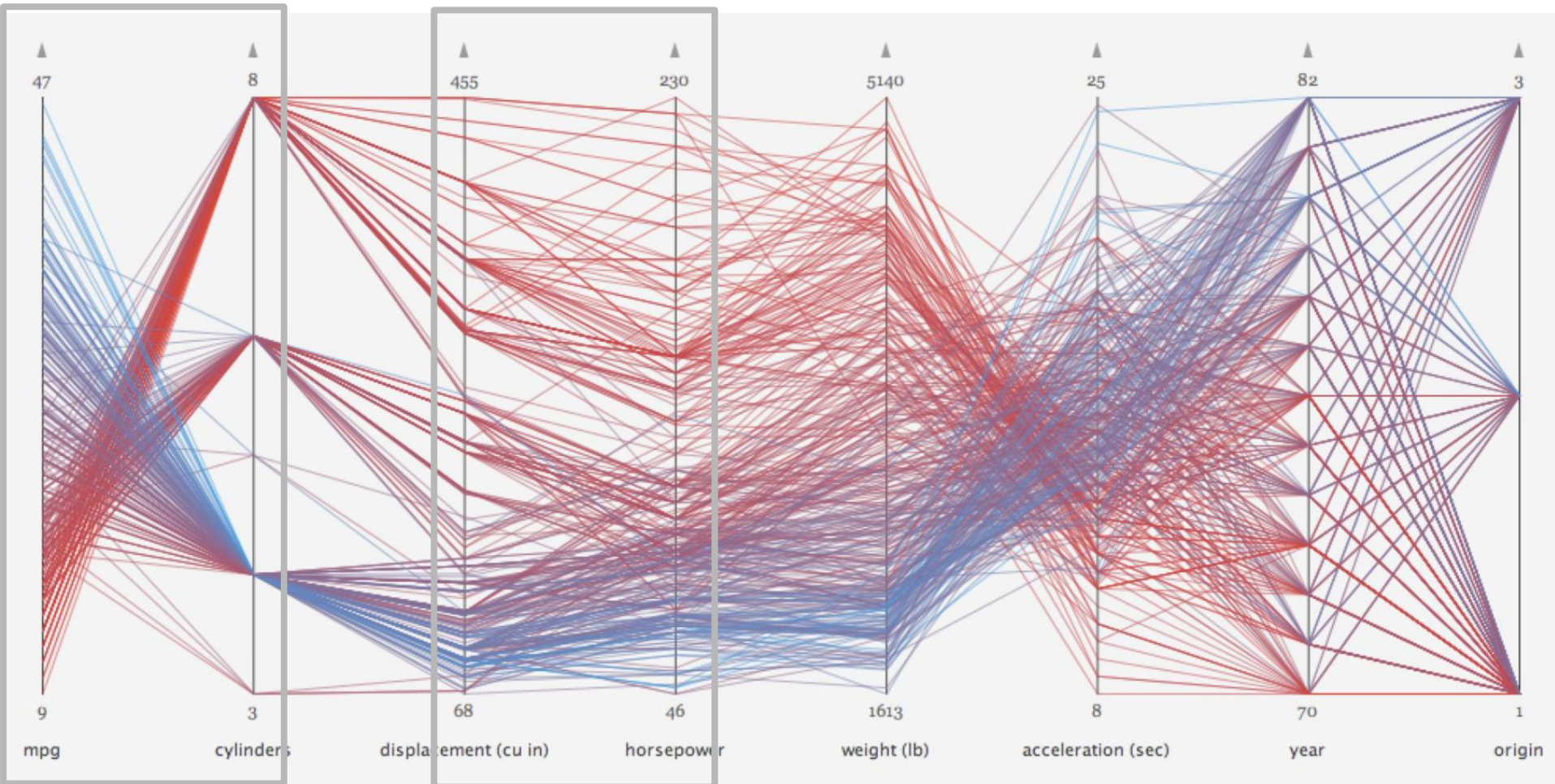
3. Rendering



4. Perceiving



5. Interpretation

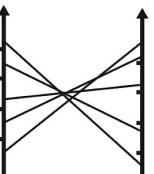
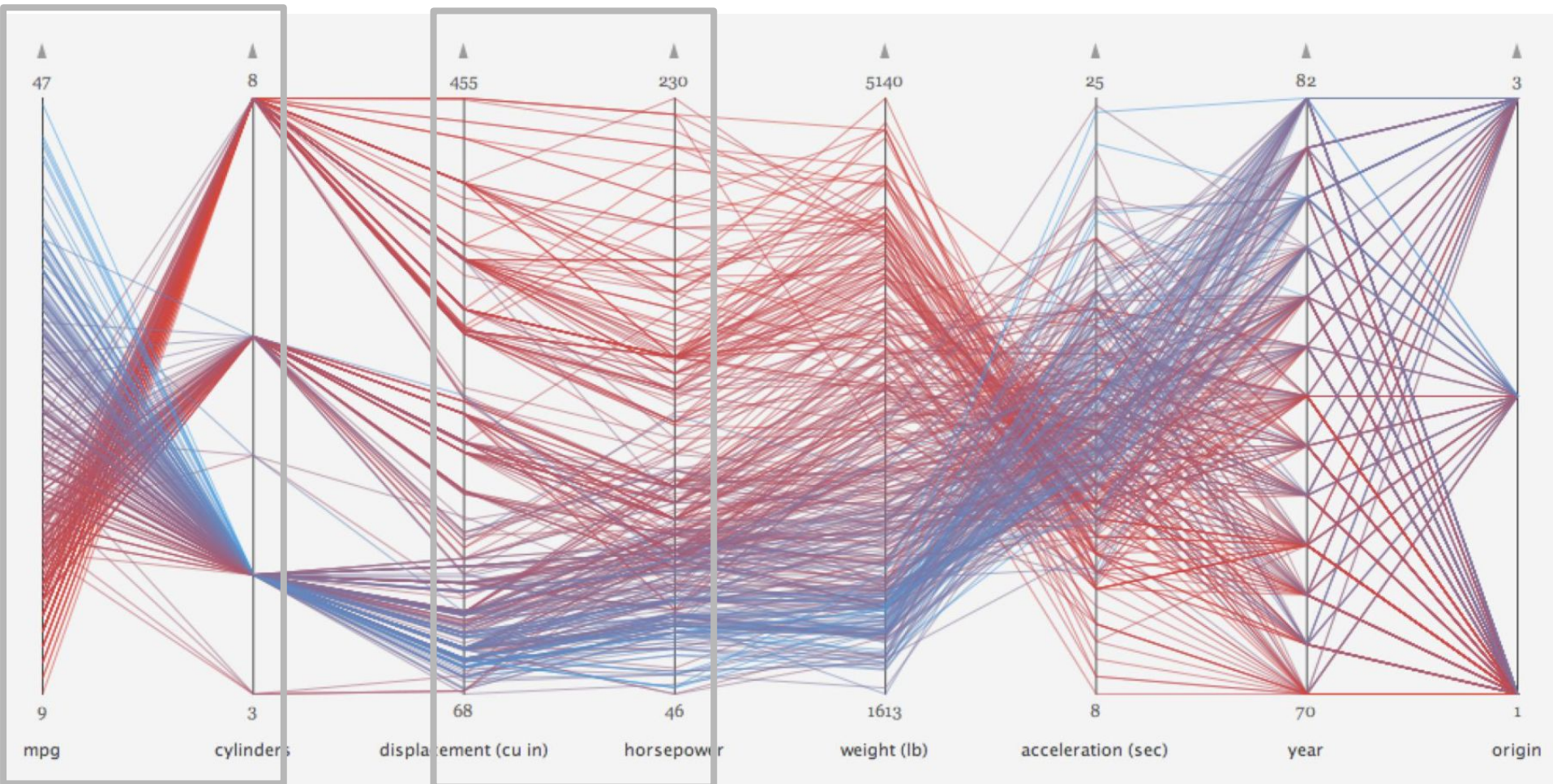


**Negative
Correlation**



**Positive
Correlation**

6. Comprehending



Negative Correlation



Positive Correlation

Encoding: designer

1

Data

What is my data?

Which data type?

Ordinal / numerical / categorical?

2

Visual Mapping

What is my visual representation?

Which visual variables am I using?

How am I encoding my data?

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What is my medium?

monoscopic/stereoscopic?

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Decoding: user

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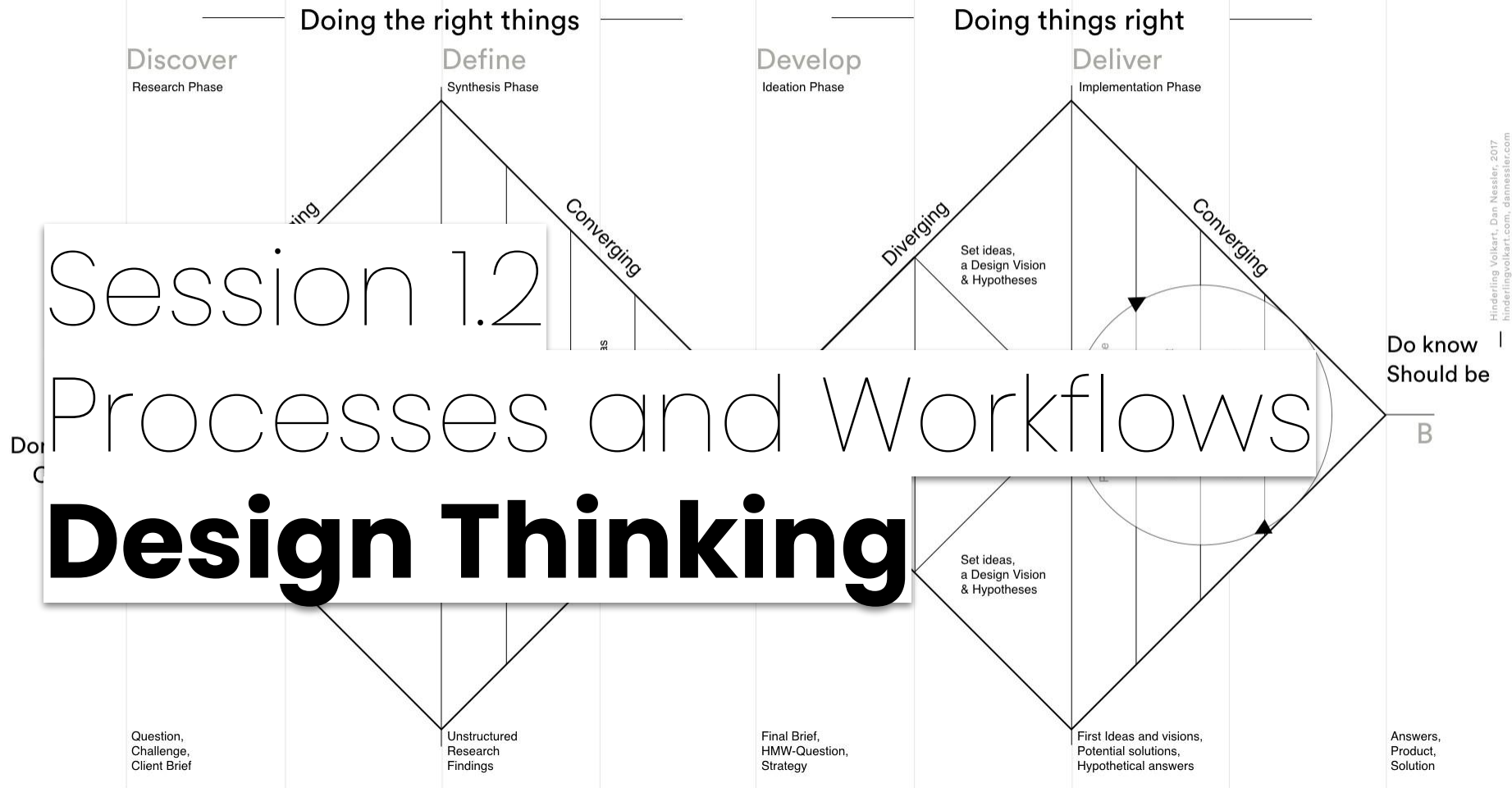
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Visualization Design

What:

- Create effective visualizations

How:

Visualization Design

What:

- Create effective visualizations
- Create efficient visualizations

How:

Visualization Design

What:

- Create effective visualizations
- Create efficient visualizations
- Solve a problem

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- Design a solution

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How:

- Visualization can have **many forms**

Visualization Design

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- It's **not rocket science**

Visualization Design

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- Everyone can **design** visualizations

Visualization Design

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- Everyone can **learn creating** visualizations

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- **Solve your own problems**

Visualization Design

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- **Solve your own problems**
- There are **many rules**

Visualization Design

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- It's **not rocket science**
- Everyone can **design** visualizations
- Everyone can **learn creating** visualizations
- **Solve your own problems**
- There are **many rules**
- There are **many exceptions**

Design Thinking

Design thinking is a human-centered approach to creative problem solving.

- is about **people**
 - empathy, problems, context, problem

<http://www.theagileelephant.com/what-is-design-thinking/>

https://www.ted.com/talks/david_kelley_human_centered_design

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- **hands-on**
 - develop, prototype, test, try, ...
- **Show, don't tell**
- **iterative**
 - failure, progress, iterate, feedback,...

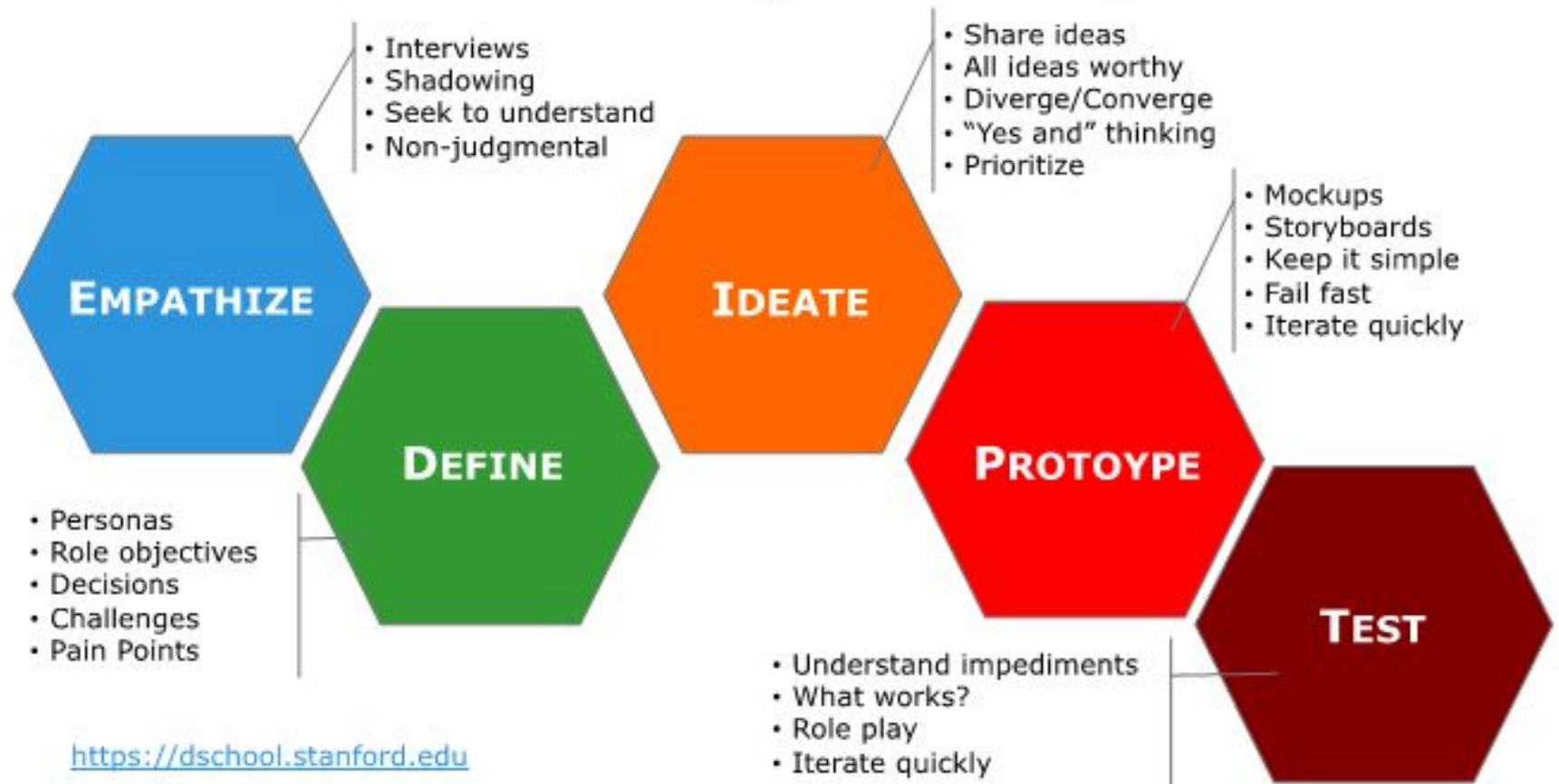
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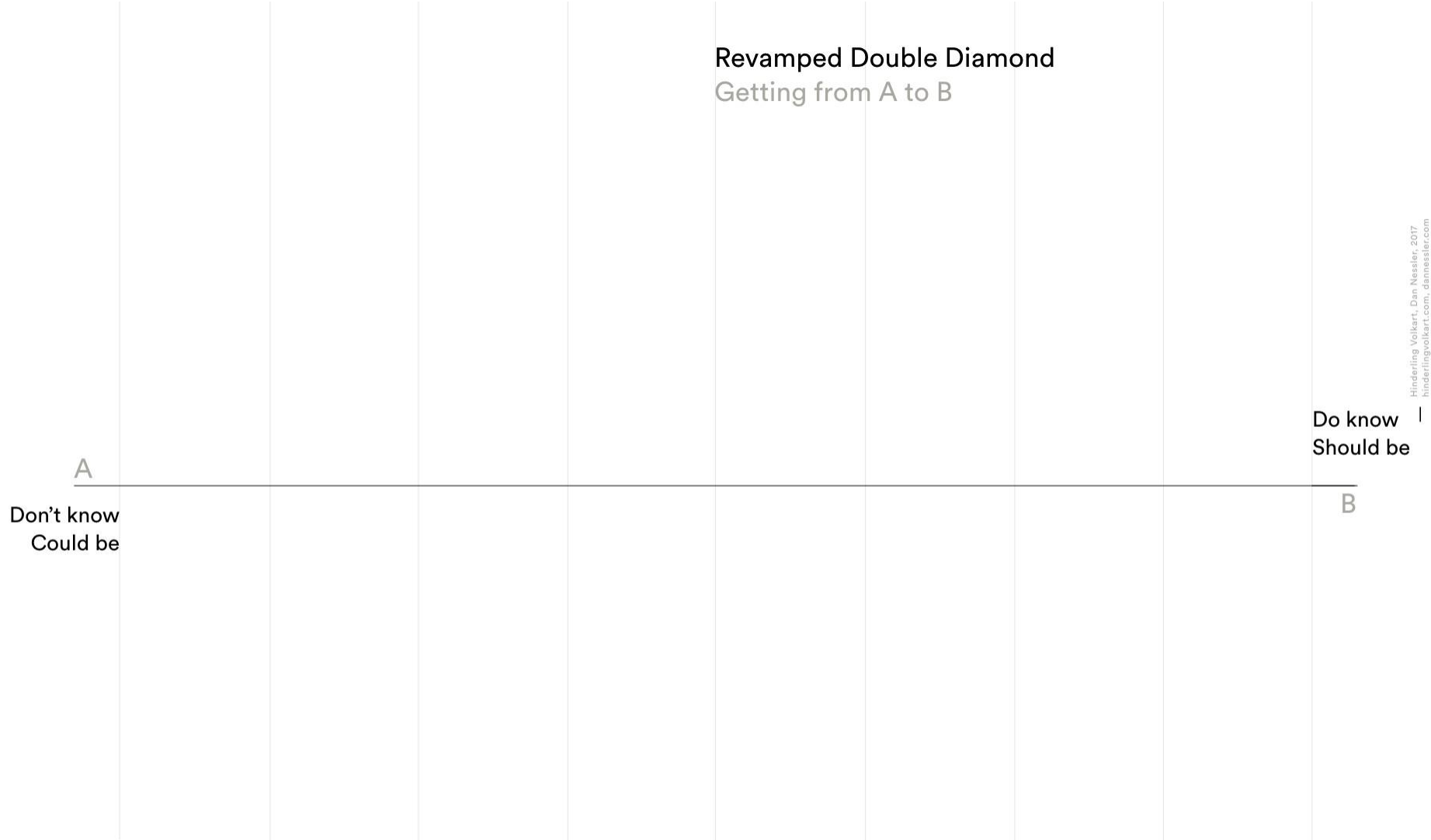
Rowe, Peter G. *Design thinking*. MIT press, 1987.

Design Thinking—5 steps

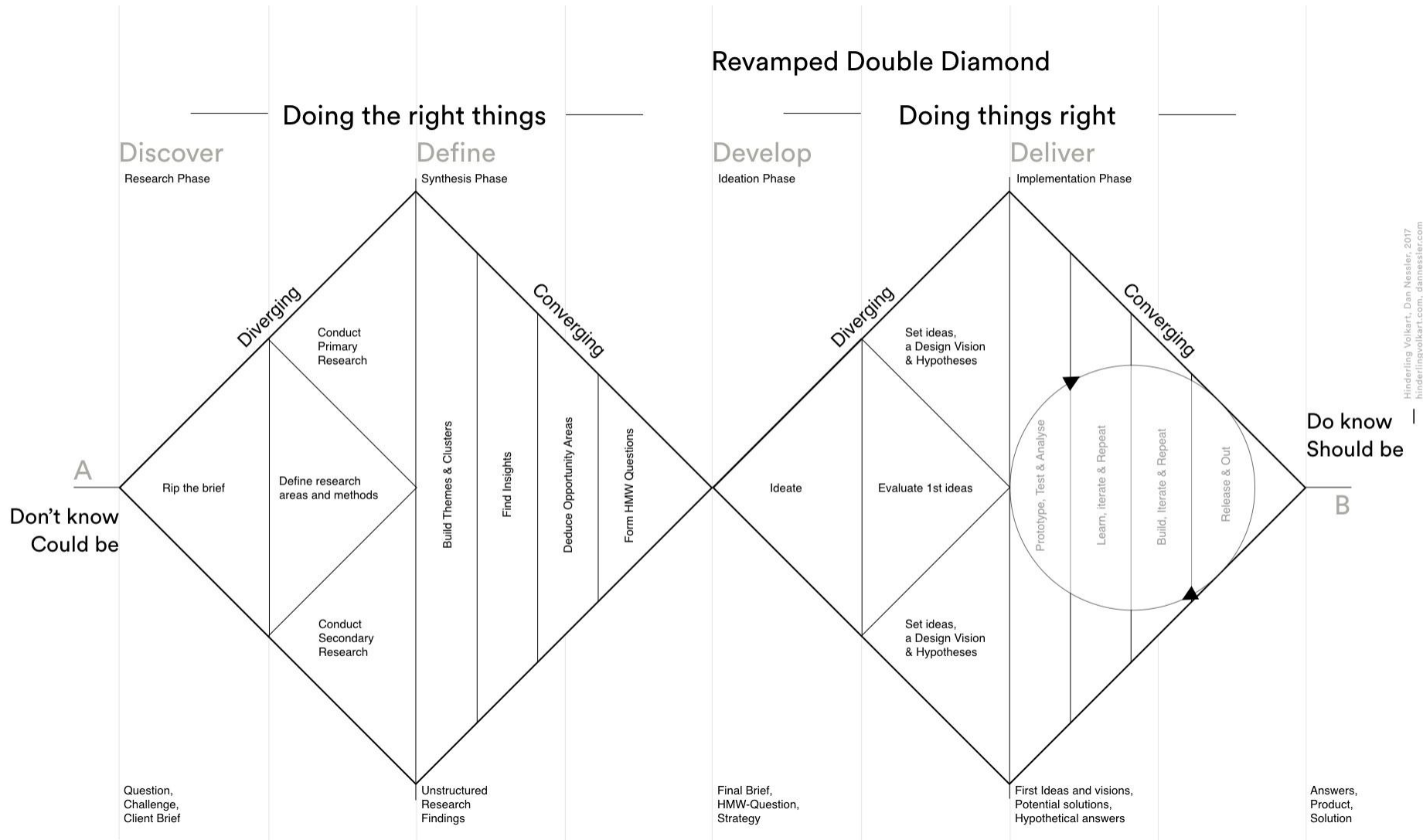
Stanford d.school Design Thinking Process



Double Diamond



Double Diamond



Design Decisions

Context:

- Audience knowledge
- Data complexity
- Tasks
- Display medium
- ...

Visual Design:

Design Decisions

Context:

- Audience knowledge
- Data complexity
- Tasks
- Display medium
- ...

Visual Design:

- Familiarity vs. unfamiliarity
- Clarity vs. Memorability
- Novelty vs. Tradition
- Facts vs. Uncertainty
- Reader-driven vs. Author-driven
- ...

Visualization Design Thinking

Empathize - Understand your audience, interviews, observations, reading, conversation

Define

Ideate

Prototype

Test

Visualization Design Thinking

Empathize

- Understand your audience, interviews, observations, reading, conversation

Define

- Create a **Data Challenge**. Set context and constraints.

Ideate

Prototype

Test

Visualization Design Thinking

Empathize

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- **Sketch** design ideas
- Develop **visual mapping**
- **Exploratory data analysis**

Prototype

Test

Visualization Design Thinking

Empathize

- Understand your audience, interviews, observations, reading, conversation

Define

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- Use visualizations **tools**
- High-fidelity paper prototypes
- *"Memento data"*

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- Prototype**
 - Use visualizations **tools**
 - High-fidelity paper prototypes
 - *"Memento data"*
- Test**
 - User-centered **evaluation**



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Benjamin Bach

May 2022

<http://benjbach.me>

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Encoding: designer

1

Data

What is my data?

Which data type?

Ordinal / numerical / categorical?

2

Visual Mapping

What is my visual representation?

Which visual variables am I using?

How am I encoding my data?

3

Rendering

What is my medium?

monoscopic/stereoscopic?

Tangiblity?

Print / digital?

View

Decoding: user

6

Comprehending

What does it mean for me?

What shall I do now?

Is this all true?

What do I learn?

5

Interpreting

What does it mean?

What does color mean?

Wha does 'up 'mean?

What do these patterns show?

4

Perceiving

What does it show?

Where is big, medium, small?

How do things compare?

What relationships exist?

Data Visualization Challenge

Data	

Data Visualization Challenge

Data	Message & Insights

Data Visualization Challenge

Data	Message & Insights
Audience	

Data Visualization Challenge

Data	Message & Insights
Audience	Context

Challenge **Data**

- What is my data?
- Where is it from?
- How is it characterized?
- How complicated is my data?
- How many dimensions?
- How large?
- What data format?
- What is *not* part of my data?
- ...

Challenge **Messages / Insights**

- What am I going to find?
- What am I interested in finding?
- Which questions do I have?
- Which *tasks* do I want to support?
- What am I going to tell with the visualization?
- ...

Challenge **Audience**

- Who is my audience?
- How are they characterized?
- What do they know about the data / topic?
- Why are they interested in my data?
- Why should they care?
- What do they know about visualization?
- What questions might they have?
- ...

Challenge **Context**

- How will people see my visualization?
- Where will this be?
- How will they be able to engage?
- Will they be able to interact?
- ...

Data Visualization Challenge

Data

- What is my data?
- Where is it from?
- How is it characterized?

Audience

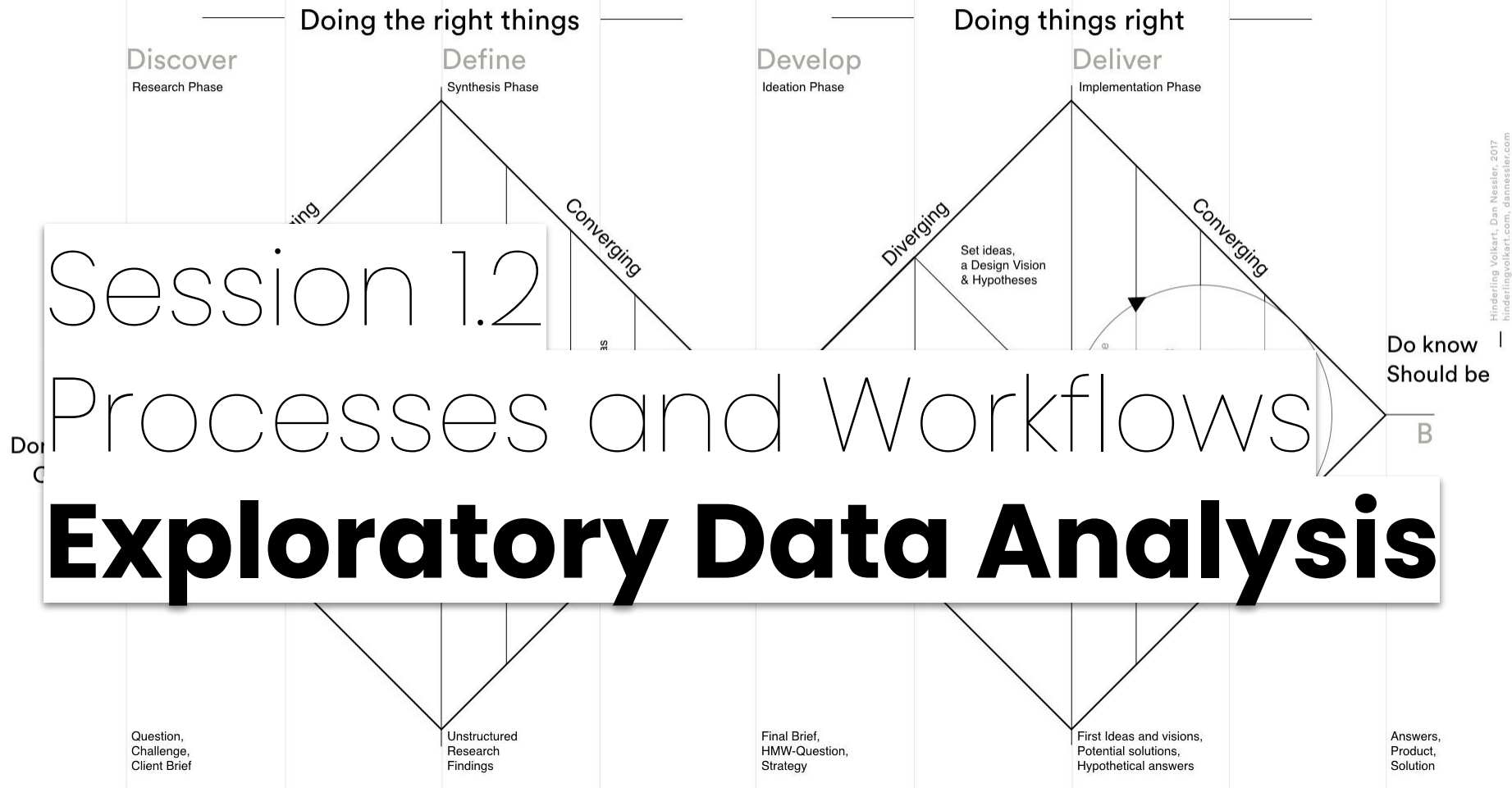
- Who is my audience?
- Why do they care?
- What do they know?

Messages / Insights

- What am I looking for?
- What am I telling?

Context

- Where will visualization be seen?
- How do people engage?



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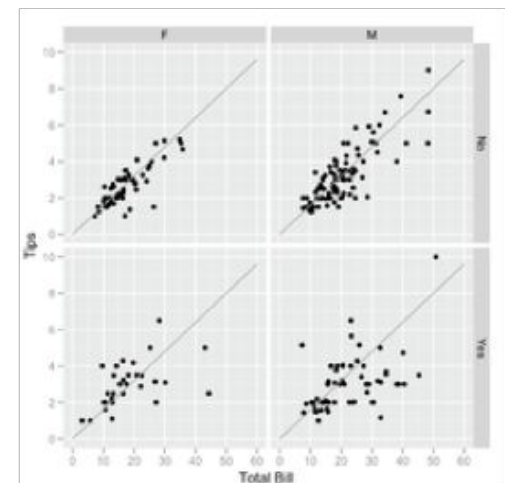
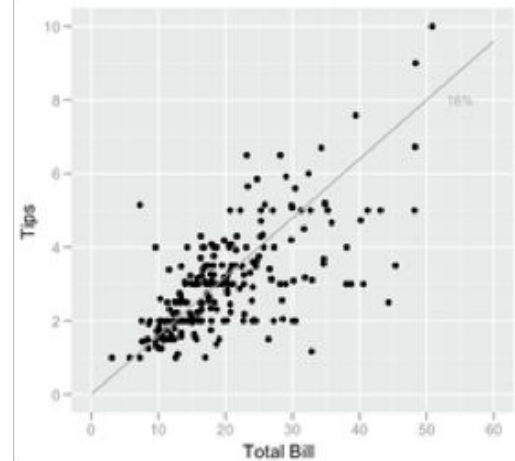
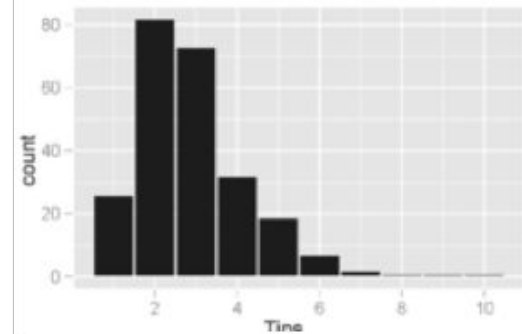
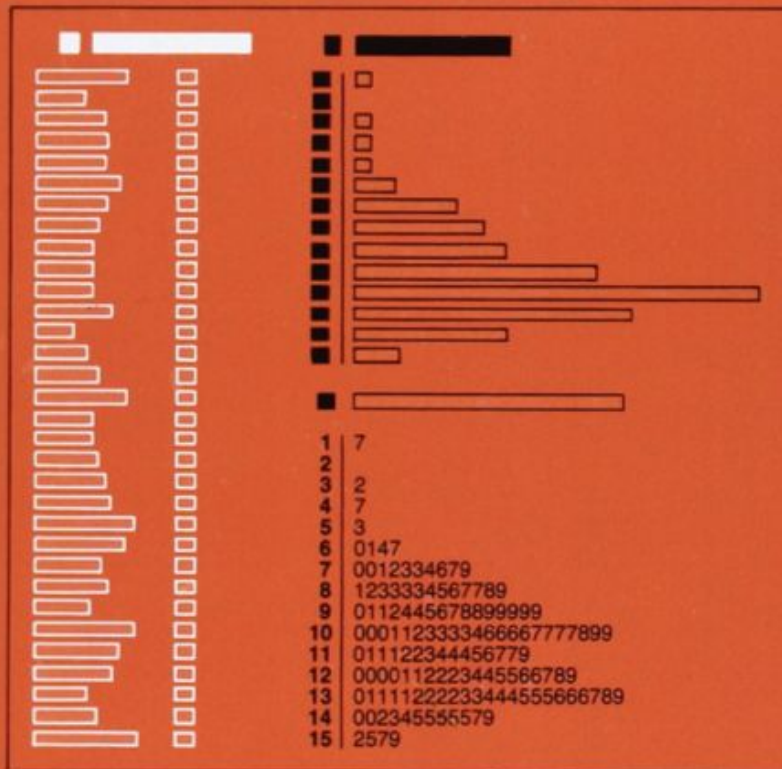
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John W. Tukey

EXPLORATORY DATA ANALYSIS



Exploratory Data Analysis

How –to:

- Grand tour: create as many views as possible
- Obtain as many different perspectives as possible
- Use multiple views
- Generate hypotheses
- Play with data and visualization
- Use simple visualizations first, then become complex

Implications:

- Understand your data
- Generate insights
- Inform your visualization design

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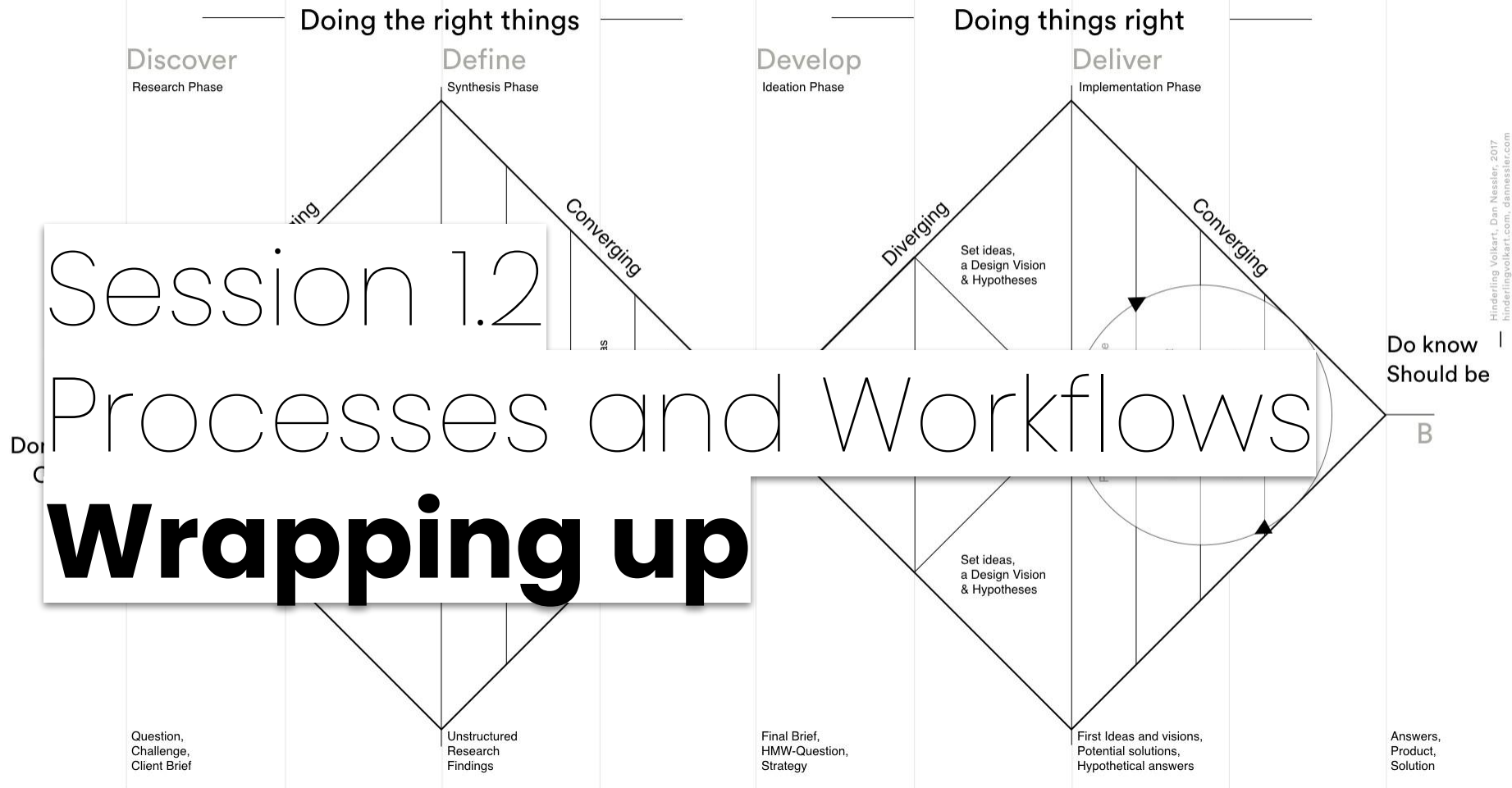
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Wrap-up

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2. Many **processes** you will use visualization:
 - *Creator, user, reader, analyst, ...*
3. **Design Thinking** is essential for creating effective and efficient visualizations
4. Formulating a **visualization challenge** helps you focusing and start your design thinking process
5. **Exploratory data analysis** helps you knowing your data and informing a design process.

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