### Tableau License for Students

#### GitHub Student Pack

# **Block Builder**



# Administrative



### What do you like to do outside of school?

-computer science (????) league -cat videos -travel, hiking, exercise

# -gaming: twitch, dota2, witcher,



### What are your goals for this course? - "Making graphs that can sing to human's heart"

data analysis effectively

### - Technical aspects: js, exploratory

### - Practical aspects: presenting data







User interaction/collaboration actions

10.0 8.0 13.( 9.( 11.( 14.0 6.( 4.( 12.( 7.( 5.0

0	8.04
0	6.95
0	7.58
0	8.81
0	8.33
0	9.96
0	7.24
0	4.26
0	10.84
0	4.82
0	5.68

Mean Variance

10.0	8.04
8.0	6.95
13.0	7.58
9.0	8.81
11.0	8.33
14.0	9.96
6.0	7.24
4.0	4.26
12.0	10.84
7.0	4.82
5.0	5.68
9.0	7.5
10.0	3.75

	1		2		3		4	
	Х	Y	Х	Y	Х	Y	Х	Y
	10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
	8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
	13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
	9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
	11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
	14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
	6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
	4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
	12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
	7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
	5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89
Mean	9.0	7.5	9.0	7.5	9.0	7.5	9.0	7.5
Variance	10.0	3.75	10.0	3.75	10.0	3.75	10.0	3.75

	1		2		3		4	
	Х	Y	Х	Y	Х	Y	Х	Y
	10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
	8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
	13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
	9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
	11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
	14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
	6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
	4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
	12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
	7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
	5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89
Mean	9.0	7.5	9.0	7.5	9.0	7.5	9.0	7.5
Variance	10.0	3.75	10.0	3.75	10.0	3.75	10.0	3.75
Correlation	0.816		0.816		0.816		0.816	





X3

X4

### Tableau





# Key question: how to map data to visuals?







#### Bijection (one visual attribute, one data attribute)

#### Surjection (multiple visual attribute to one data attribute)

Set Theory

#### Injection (One to one mapping, but not all data elements are mapped)

# What happens when

### Data Vars > Visual Vars?

# What happens when

### Visual Vars > Data Vars?

## Data Attributes







Level of Information

#### Non-ordered and non-numeric

#### AKA categorical data

#### ['apple', 'pear', 'banana']

# nomal

#### Ordered, not necessarily numeric

#### [1st, 3rd, 5th, 7th]

#### ['G', 'PG', 'PG-13', 'R']

# order for the second second



1st -> 3rd  $PG \rightarrow R$ 

#### Ordered, not necessarily numeric

#### [1st, 3rd, 5th, 7th]

#### ['G', 'PG', 'PG-13', 'R']

#### length is not meaningful

# ordinal

1*st* -> 3*rd PG* -> *R* 

#### Ordered, numeric, not ratio-able

#### ['Jan 12', 'Jan 20']

#### *['17', '44', '23', '30']*

# interval

#### Jan 12/Jan 20 = ???

#### 23°/30° = ???

#### Ordered, numeric, ratio-able (has a "true" 0)

#### [1, 3, 5, 7]

#### [5'8", 6'1", 5'4"]

#### $Q \rightarrow O$ [0-100] -> [A, B, C, D, F]

#### Ratio / Interval (Q)

Ordinal

Nominal



#### $Q \rightarrow O$ $[0-100] \rightarrow [A, B, C, D, F]$ $O \rightarrow N$ $[A, B, C, D, F] \rightarrow [B, C, F, D, A]$

#### Ratio / Interval (Q)

Ordina

#### Nominal



 $Q \rightarrow O$ [0-100] -> [A, B, C, D, F]  $O \rightarrow N$  $[A, B, C, D, F] \rightarrow [B, C, F, D, A]$  $N \rightarrow O$ ["Jack", "Alex"] -> ["Alex", "Jack"]

#### Ratio / Interval (Q)



#### Ordinal

#### Nominal



 $Q \rightarrow O$ [0-100] -> [A, B, C, D, F]  $O \rightarrow N$  $[A, B, C, D, F] \rightarrow [B, C, F, D, A]$  $N \rightarrow O$ ["Jack", "Alex" ] -> ["Alex", "Jack"]  $O \rightarrow O$ "Alex"+"Jack" -> 7 ???

#### Ratio / Interval (Q)



#### Ordinal

#### Nominal



#### Nominal Ordinal ==!= > < <=>=

#### Interval + -





### operations



#### Nominal Ordinal ==!= > < <=>=

#### Interval + -





# consider a distance function...

### operations




### → Ordered

→ Ordinal





→ Quantitative

## structure

### → Tables



→ Multidimensional Table

Key 2



### Multidimensional Table



### Aultidimensional Table





### **Data Types** $( \rightarrow)$ → Items → Attributes Data and Dataset Types Networks & Tables Trees Items (nodes) Items Attributes Links Attributes

Datasets

### → Links → Grids → Positions

Fields	Geometry	Clusters, Sets, Lists
Grids	ltems	Items
Positions	Positions	
Attributes		



### → Tables



### → Networks



 $\rightarrow$  Multidimensional Table



→ Trees



### → Geometry (Spatial)





Position

# data shapes the algorithm space

# data shapes the visual space

# Further Reading

### <u>Stevens, Stanley Smith. "On the theory of scales of</u> <u>measurement." (1946).</u>



# Visual Attributes



### Bertin, Semiologie Graphique, '67

Magnitude Channels: Ordered Attributes

Position on common scale

Position on unaligned scale

Length (1D size)

Tilt/angle

Area (2D size)

Depth (3D position)

Color luminance

Color saturation

Curvature

Volume (3D size)





# (pay attention to your how you judge these differences)

### Position (Common Scale)





### -scatterplots -bar charts -ine charts \_222

### Position (Un-aligned Scale)





### -stacked bars -stacked area -???





A

Unframed Unaligned

### Use design elements to compensate!





В

A В

Framed Unaligned Unframed Aligned







3:00

Angle



### Accurate encoding does not ensure accurate perception!







÷

Volume (3D size)

Effectiveness -

Least

Most 🕨



Most 🕨

Effectiveness

Least

Volume (3D size)

### Luminance and Saturation– really the same?









### Identity Channels: Categorical Attributes



▲

Most







### Spatial Region

### Hue bad for magnitude:





### Hue bad for magnitude:



### Hue is great for identity:











Fig. 2: Palettes of visual stimuli used in our experiments: shape, color, size, shape-color, shape-size, size-color.

### $O \Box + \times * O \Delta \nabla < \Box >$ $\Box + \times \Diamond \Box + \times \Diamond \Box + \times \Diamond$ $\bullet$ + × $\diamond$ $\bullet$ + × $\diamond$ $\bullet$ + × $\diamond$ $\bullet$ + × $\diamond$

Demiralp et al., 2014







Fig. 1: (Left) A crowd-estimated perceptual kernel for a shape palette. The kernel was obtained using ordinal triplet matching. (Right) A two-dimensional projection of the palette shapes obtained via multidimensional scaling of the perceptual kernel.

Demiralp et al., 2014













(huge attention grabber, use with caution)



### Identity Channels: Categorical Attributes



▲

Most



# What happens when

### Data Vars > Visual Vars?

# What happens when

### Visual Vars > Data Vars?

# Data Deconstruction


## Visualization,

## 10 ways

