# Lecture 2: Introduction To Data Data Visualization

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#### Outline

- Data Mining Terminology
- Basics of Visualization
  - Graph integrity
  - 2D visualization
  - Basics of higher dimensional visualization

#### Piazza

- Enrollment done for registered students
- Auditing students send me your email id
- "Search for Teammates" enabled

#### Socrative

- Instant Student Feedback
- Accessible via Smart Phone, Tablet, Laptop
- No login needed from Student's end
- Use a consistent name throughout the semester

# In-Class Quizzes

- URL: http://m.socrative.com/
- Room Name: 4f2bb99e

#### Misc Announcements

- Slides for Lecture 1 updated
- Change Office hour timings?
- Installation of Scientific Python

#### Other Relevant Online Classes

- Machine Learning, Stanford: https://www.coursera.org/course/ml
- Mining of Massive Datasets, Stanford: https://www.coursera.org/course/mmds
- Statistical Learning, Stanford: https://class.stanford. edu/courses/HumanitiesandScience/StatLearning/ Winter2015/about

# Data Mining Terminology

# Data Matrix

Table 1.1. Extract from the Iris dataset

(	Sepal length	Sepal width	Petal length	Petal width	Class
	$X_1$	$X_2$	$X_3$	$X_4$	$X_5$
$\mathbf{x}_1$	5.9	3.0	4.2	1.5	Iris-versicolor
$\mathbf{x}_2$	6.9	3.1	4.9	1.5	Iris-versicolor
<b>X</b> <sub>3</sub>	6.6	2.9	4.6	1.3	Iris-versicolor
$\mathbf{x}_4$	4.6	3.2	1.4	0.2	Iris-setosa
<b>X</b> 5	6.0	2.2	4.0	1.0	Iris-versicolor
$\mathbf{x}_6$	4.7	3.2	1.3	0.2	Iris-setosa
<b>X</b> 7	6.5	3.0	5.8	2.2	Iris-virginica
<b>x</b> <sub>8</sub>	5.8	2.7	5.1	1.9	Iris-virginica
:	:	:	:	:	Ė
<b>X</b> 149	7.7	3.8	6.7	2.2	Iris-virginica
$\mathbf{x}_{150}$	5.1	3.4	1.5	0.2	Iris-setosa

#### Data Matrix

$$\mathbf{D} = \begin{pmatrix} & X_1 & X_2 & \cdots & X_d \\ \mathbf{x}_1 & x_{11} & x_{12} & \cdots & x_{1d} \\ \mathbf{x}_2 & x_{21} & x_{22} & \cdots & x_{2d} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \mathbf{x}_n & x_{n1} & x_{n2} & \cdots & x_{nd} \end{pmatrix}$$

#### Data Matrix

- n rows and d columns
- Row ⇒ Tuple/Entities
- Column ⇒ attribute/feature
- Special column called Class
- $x_i$ : i-th row,  $X_j$ : j-th column
- Row ⇒ entities, instances, examples, records, transactions, objects, points, feature-vectors, tuples
- Column ⇒ attributes, properties, features, dimensions, variables, fields
- $n \Rightarrow \text{size}$ ,  $d \Rightarrow \text{dimensionality of data}$

## Geometric View

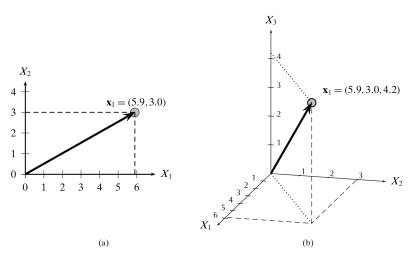
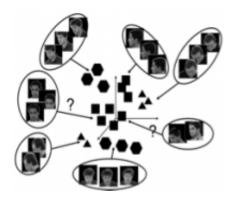


Figure 1.1. Row  $\mathbf{x}_1$  as a point and vector in (a)  $\mathbb{R}^2$  and (b)  $\mathbb{R}^3$ .

# Implications '

- Each photo in the universe is some point in high dimension
- Each book (written or in future) are some point in high dimension



#### Ben Shneiderman, 1996:1

- 1D (sequences)
- Temporal
- 2D (maps)
- 3D (shaped)
- nD (relational)
- Trees (hierarchical)
- Networks (graphs)
- Others (text)

<sup>&</sup>lt;sup>1</sup>The Eyes Have It: A Task by Data Type Taxonomy for Information Visualization [Shneiderman, 96]

# Semantics vs. Types

- Data Semantics: real-world meaning
  - e.g., company name, day of the month, person height, etc.
- Data Type: Interpretation in terms of scales of measurements
  - e.g., quantity or category, sensible mathematical operations etc.

Nominal (Categorical) (N)
Are = or ≠ to other values
Apples, Oranges, Bananas,...

○ □ + △ S U

Ordinal (O)
Obey a < relationship</li>
Small, medium, large

---////// min max

Quantitative (Q)
Can do arithmetic on them
10 inches, 23 inches, etc.

On the theory of scales and measurements [S. Stevens, 46]

Q - Interval (location of zero arbitrary)
Dates: Jan 19; Location: (Lat, Long)

Like a geometric point. Cannot compare directly.

Only differences (i.e., intervals) can be compared

Q - Ratio (zero fixed)

Measurements: Length, Mass, Temp, ...

Origin is meaningful, can measure ratios & proportions Like a geometric vector, origin is meaningful

- N Nominal (labels)
  - Operations:  $=, \neq$
- O Ordinal (ordered)
  - Operations:  $=, \neq, >, <$
- Q Interval (location of zero arbitrary)
  - Operations:  $=, \neq, >, <, +, -$
- Q Ratio (zero fixed)
  - Operations:  $=, \neq, >, <, +, -, \times, \div$

What is the data type of:

• Gender:

What is the data type of:

• Gender: Categorical/Nominal

• Age:

What is the data type of:

• Gender: Categorical/Nominal

• Age: Ordinal

• Height:

#### What is the data type of:

• Gender: Categorical/Nominal

• Age: Ordinal

• Height: Quantitative - Ratio

Date:

#### What is the data type of:

• Gender: Categorical/Nominal

• Age: Ordinal

• Height: Quantitative - Ratio

• Date: Quantitative - Interval

#### Data Dimensions

- Univariate (1D)
- Bivariate (2D)
- Trivariate (3D)
- Multivariate (nD)

# Introduction To Data Visualization

#### Visualization Goals

#### Presentation

- Known facts about data
- Task: Communicate results

#### Exploration

- Data without hypothesis
- Task: Generate hypothesis

#### Confirmation

- Hypothesis is given
- Task: Verify / falsify hypothesis

#### Visualization Goals

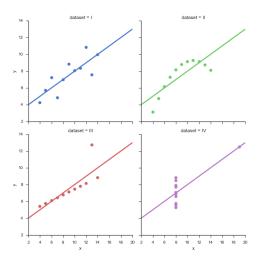
"The greatest value of a picture is when it forces us to notice what we never expected to see."

-John Tukey (1915 - 2000)



# Anscombe's Quartet

Same mean, variance, correlation, and linear regression line

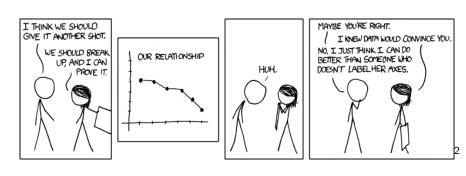


# **Graphical Integrity**

"There are three kinds of lies: lies, damned lies, and statistics."

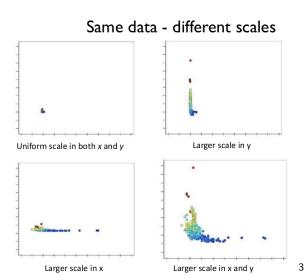
- attributed to Benjamin Disraeli in 19th Century

# Labelling Chart Axes



<sup>&</sup>lt;sup>2</sup>http://xkcd.com/833/

# Lying with Scales

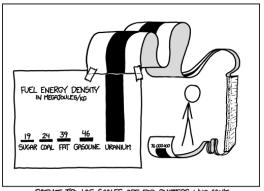


<sup>&</sup>lt;sup>3</sup>Ward, Grinstein, Keim, 2011

#### Scales are Critical!

- What are your bounds upper and lower?
- What scale works? Linear? Log? Clipping? Breaks?
- Relative or absolute values?
- How can you make things comparable?

# Log Scale



SCIENCE TIP: LOG SCALES ARE FOR QUITTERS WHO CAN'T FIND ENOUGH PAPER TO MAKE THEIR POINT PROPERLY.

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<sup>4</sup>http://xkcd.com/1162/

# Graph Types (2D and nD)

# Summary

# Major Concepts:

- Data mining Terminology
- Visualization basics
- Graphical Integrity
- Graph Types for 2D and nD

#### Slide Material References

• Slides from Harvard CS 109 (2013 and 2014)