

# CSCI568

## Discussion 4: Data Fundamentals

# Organize Your Mind

Two general purposes

predictive

descriptive

~~Four~~ families of tasks

cluster analysis

classification

exploration

association  
analysis

anomaly  
detection

visualization &  
interaction

A still from the movie 'The Goonies' featuring a young man with dark hair, wearing a grey jacket, holding a wooden bowl. The background is a wall with graffiti. A large, bold, white 'DATA' text is overlaid across the center of the image.

# DATA

**Common “Formats?”**

# Data Mining Starts With Data! (duh)

- type
- quality
- preprocessing
- existing/obvious relationships

# Types

**Categorical**  
(qualitative)

**Numeric**  
(quantitative)

**Nominal**

zip codes  
student ID  
eye color  
gender

**Ordinal**

hardness  
grades  
street numbers

**Interval**

dates  
temperature (in C/F)

**Ratio**

temp in K  
money  
age  
lengths

(DM 26)

# Data Transformations

(changes that don't affect the *meaning* of an attribute)

Categorical  
(qualitative)

Numeric  
(quantitative)

Nominal

preserve 1:1  
mapping

Ordinal

preserve order

Interval

consistent scaling

Ratio

preserve ratio,  
but units can  
change

Bottom line: *statistical* operations should yield same results whether or  
not data is transformed.

(DM 27)

# # of possible values

- finite ... discrete
- infinite.. continuous (limited precision)



# A “Special” Type: Asymmetric Attributes

- We only care about presence (or non-presence)
- Influences the meaning of “similarity”

# Data Sets

## (collections of objects)

- Dimensionality (# of attributes)
- Sparsity (how often a value exists)
- Resolution
  - too fine? pattern buried in noise
  - too coarse? pattern not evident

# Data Sets Have Types Too

DM 29 - 36

# Record-Based

- General (most common)
- Transaction / Market Basket Data
  - One attribute includes a set of values
- Data Matrix
  - sparse data matrices
    - document-term matrices

# Graph-Based

- Relationships among objects
  - data objects are mapped to nodes
  - relationships are represented as links
- Object attributes might be graphs

# Ordered Data

- Temporal data
  - “candy sales peak before Halloween
  - purchase history, predictions
  - subject to temporal autocorrelation
- Sequence data (position, no timestamps)
- Spatial data (eg, weather data by location)
  - subject to spatial autocorrelation

# Unstructured Data

- non-record data can become record data
- but you might not capture certain aspects of the data

# Data Quality Issues

- collected for other purposes
- collected without specified application
- errors
- *often cannot be addressed at the source*



# Data Mining Includes...

- detection and correction of problems
- use of algorithms that tolerate poor quality

# Data Is Rarely Perfect (damn!)

- human error
- limitation of measuring tools
- flaws in collection process
- missing values
- missing records
- duplicate records

# Measurement Error

- recorded value differs from true value
- measure the measurement error (aka “error”) of continuous attributes

# Collection Error

- eg, omitting data objects or attributes
- inappropriately including a data object
- collection process problems
- keyboard errors

# Noise & Artifacts

- NOISE = random component of measurement error
  - distortion or added spurious objects
  - most often with temporal / spatial datasets
  - elimination is often difficult
    - so we use algorithms that are robust against noise
- ARTIFACTS = deterministic errors
  - think: “a repeated streak on a set of photos”

# Precision, Bias, Accuracy (simple, but important!)

- Precision: closeness of repeated measurements
  - std. dev. of a set of values
- Bias: systematic imprecision
  - diff between mean of values and real value
- Accuracy: closeness to true value

# Outliers

- *data objects* that have characteristics that are very different from most of the other objects
- *values* of an attribute that are very different from typical values of that attribute
- outliers are not noise per se
  - are often legitimate data objects/values

# Missing Values (and what to do about it)

- Eliminate data object or attribute
- Estimate missing values
- Ignore missing values during analysis



# Inconsistent Values

- think: “incorrect city for a zip code”
- often easy to detect
  - eg, human height shouldn't be negative
- can use other valid data to resolve inconsistencies

# Duplicate Data

- duplicates or “almost-duplicates”
- eg, John Smith vs John Q. Smith
  - Same object but slightly different attributes
- Values should be resolved/merged
  - aka “deduplication”

DM 42, 43

# What Defines Quality Data?

Data is of “high quality” if it is suitable for its intended use.

Timeliness

Relevance

Documentation / Knowledge



# Exploring Data

# Common Summary Stats

- frequency, mode
- percentiles (esp. ordered data)
- mean, median, trimmed mean
- min, max, range, variance/std-dev
  - absolute average deviation (AAD)
  - median absolute deviation (MAD)
  - interquartile range (IQR)

# Multivariate Summary Statistics

- Covariance
  - “to what degree do the values of two attributes vary together?”
- Correlation
  - “how strongly are two attributes linearly related?”





# Visualization

Simple: the display of information in a graphical format

# Why Visualizations?

- Humans have a great ability for analyzing large amounts of information visually
  - Can detect general patterns/trends
  - Can detect outliers / unusual patterns

# Homework

- Reading 4
- Post one cool data visualization on Piazza