

Missing value and error in data

Missing Value Handling

- Missing data reduces the representativeness of the sample.
- Makes it difficult to process the data for many analysis models / algorithms.
- Three main approaches to deal with missing values-
 1. Imputation
 2. Omission
 3. Analysis

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Missing Value Handling

- **Imputation**
 - Values are filled in the place of missing data
 - Works well for situation where analysis tools are not robust to missing values
 - Dataset sizes are not reduced but noise gets imposed with the imputation

Estimation methods: regression, maximum likelihood estimation and approximate Bayesian bootstrap

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Missing Value Handling

- **Omission**
 - Samples with invalid data are discarded from further analysis
 - Creates a subset of dataset with no missing values
 - Works well for models that are not robust against data missingness

Example techniques: list-wise deletion, pair-wise deletion

Missing value and error in data

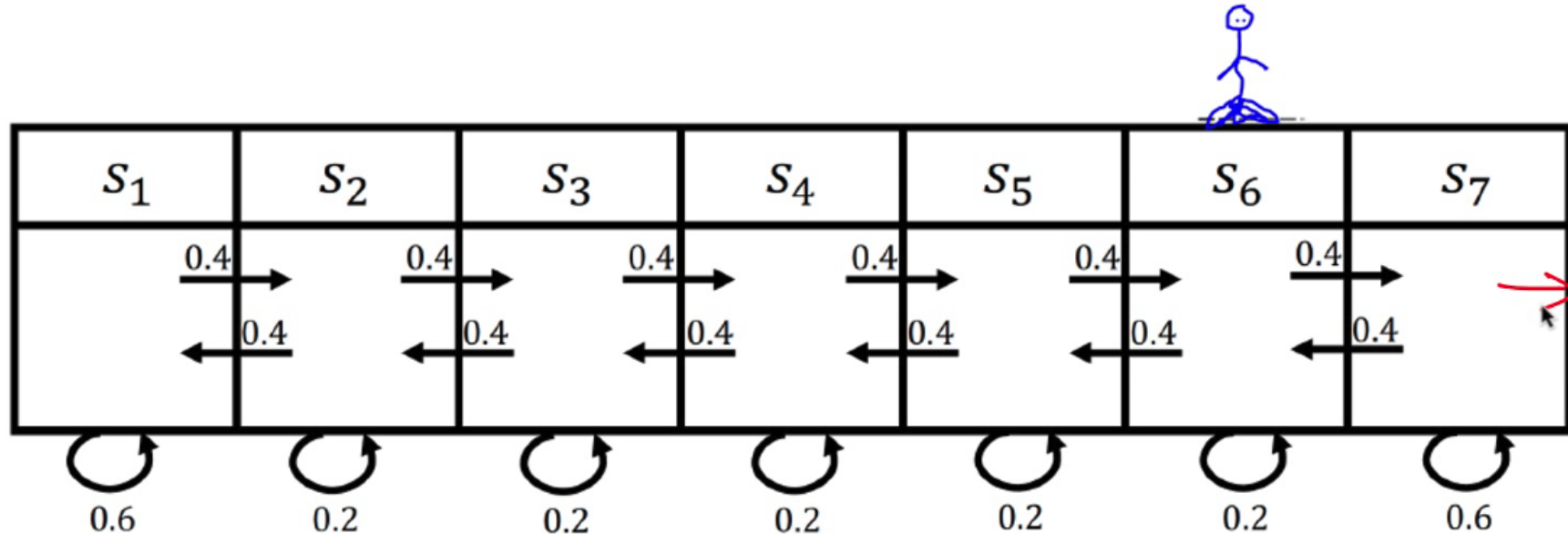
Missing Value Handling

- **Analysis**
 - Samples with invalid data are discarded from further analysis
 - Model-based techniques used to determine missing values
 - Various non-stationary Markov chain models can be used for time series data



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Analysis: Markov chain models can be used for time series data



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Error in Data

- The difference between the recorded data and true value
- Higher error rates in data makes it less representative

Types

- The major types of data error includes-
 - Sampling error
 - Non-sampling error

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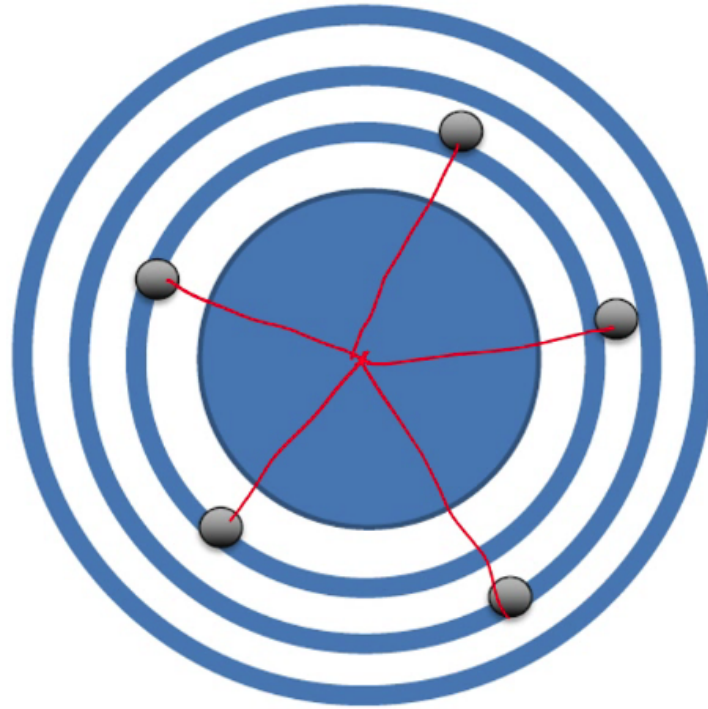
Sampling error

- Error for using data from sample of the population, in place of entire population
- It's the difference between the estimate from the sample and true value for the population
- Could occur for very small sample size
- If the sampling is not random and have some bias

Theory of Measurements

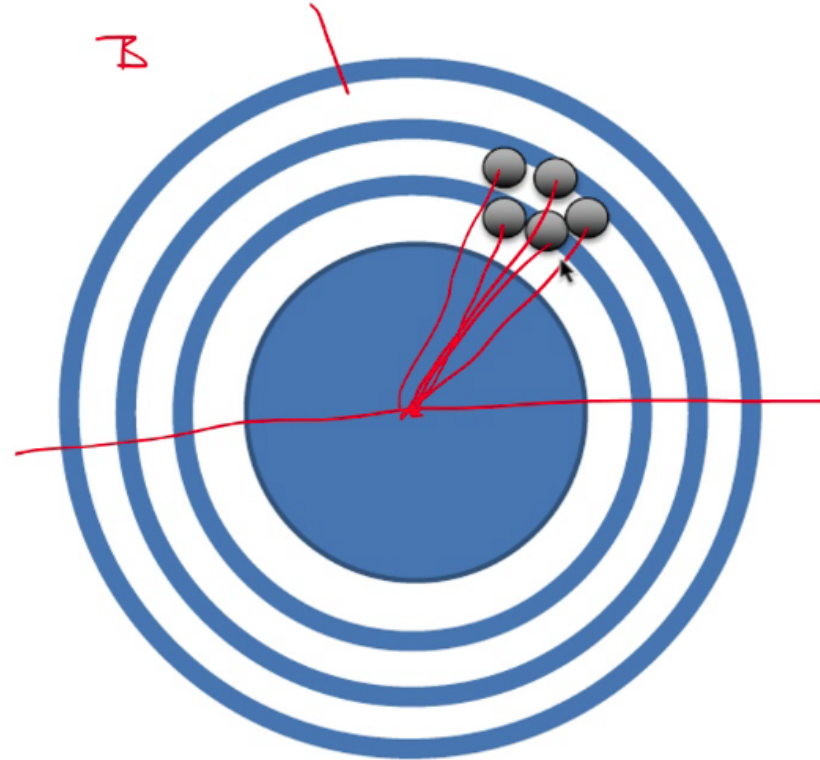
Low Accuracy, Low Precision

A



Low Accuracy, High Precision

B



Accuracy & Precision

Relational Database

Relational Model

- Relational model uses table (called relation) to represent a collection of related data values
- Rows are called records or tuples
- Columns are called attributes
- The number of attributes (i.e., number of columns) is called the degree
- Example database: MySQL, PostgreSQL and SQLite3

Non-relational Database

Non-relational database

- Database that does not use the tabular schema of rows and columns; i.e. it don't use relational model.
- Often refers to NOSQL (not only SQL); Data may be stored as
 - simple key/value pairs
 - JSON documents or
 - a graph consisting of edges and vertices.
- Most NOSQL systems are distributed databases or distributed storage systems
- Example DB: MongoDB, Oracle NoSQL, Apache CouchDB and Redis.

Non-relational Database

NOSQL Systems

- Database NOSQL systems focus on storage of “big data”
- Typical applications that use NOSQL
 - Social media
 - Web links
 - Marketing and sales
 - Posts and tweets
 - Road maps and spatial data
 - Email