



University of Colorado  
Boulder

# CSCI 4502/5502

# Data Mining

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Fall 2020  
Lecture 16 (Oct 22)

# Homework Assignments

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- ◆ Homework I to 4: graded, check scores in Canvas
- ◆ Homework 5: due at **9:30am, Thursday, Oct 22**
  - ◆ will be graded soon
- ◆ Scores statistics and common mistakes
  - ◆ will be posted in Canvas soon



# Midterm Exam (I)

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- ◆ Availability survey: Tuesday, Oct 20
  - ◆ check for **exam schedule confirmation** from TA
- ◆ Midterm review: Thursday, Oct 22
- ◆ Practice exam review: Tuesday, Oct 27
  - ◆ practice exam posted in Canvas (Proctorio & PDF)
  - ◆ in-person (wait for confirmation) & remote/online



# Midterm Exam (2)

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- ◆ Closed-book exam
- ◆ Formula will be provided
- ◆ No coding, no calculator
  - ◆ write out calculation steps
- ◆ Confirmation from TA
  - ◆ exam time & mode
- ◆ Exam time
  - ◆ 9:35am-10:50am (Oct 29)
  - ◆ 7:00pm-8:15pm (Oct 29)
  - ◆ extra slot (TBD)
- ◆ Exam mode: Proctorio;  
manual proctoring; in-  
person (class time only)



# Office Hours

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- ◆ Instructor

- ◆ Th Oct 22: 7pm-8pm

- ◆ Tu Oct 27: 11am-12pm

- ◆ Tu Oct 27: 7pm-8pm  
(moved from Oct 29)

- ◆ TA:

- ◆ M Oct 26: 8am-9am

- ◆ W Oct 28: 4pm-5pm

- ◆ GSS

- ◆ W Oct 28: 10am-11am



# Midterm Review

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- ◆ Chap 1: Introduction
- ◆ Chap 2: Getting to know your data
- ◆ Chap 3: Data Preprocessing
- ◆ Chap 4 & 5: Data Warehouse, Data Cube
- ◆ Chap 6 & 7: Frequent Pattern Mining
- ◆ Chap 8 & 9: Classification
- ◆ Chap 10 & 11: Cluster Analysis



# Chap I: Introduction

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- ◆ Why data mining?
- ◆ data explosion (generation & sharing)
- ◆ data rich but information poor
- ◆ Data mining (knowledge discovery from data)
  - ◆ automated analysis of massive data
  - ◆ quality vs. efficiency
  - ◆ interesting patterns: valid, novel, potentially useful, ultimately understandable by human



# Data Mining: Various Views

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- ◆ **Data view**

- ◆ kinds of data to be mined

- ◆ **Knowledge view**

- ◆ kinds of knowledge to be discovered

- ◆ **Method view**

- ◆ kinds of techniques utilized

- ◆ **Application view**

- ◆ kinds of applications adapted



# Chap 2: Getting to Know Your Data

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- ◆ **Data objects and attribute types**
- ◆ nominal, binary, ordinal, numeric
- ◆ interval-scaled, ratio-scaled
- ◆ discrete vs. continuous
- ◆ **Measuring data similarity and dissimilarity**
- ◆ data matrix vs. dissimilarity matrix
- ◆ proximity measures for nominal, binary, ordinal, numeric attributes



# Descriptive Summarization

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- ◆ Basics: N, min, max
- ◆ Central tendency
  - ◆ mean, median, mode, midrange
- ◆ Dispersion
  - ◆ quartiles, IQR, five number summary
  - ◆ variance, standard deviation
- ◆ Graphic displays: box plot, histogram, quantile plot, quantile-quantile plot, scatter plot



# Chap 3: Data Preprocessing

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- ◆ Why preprocessing data?
- ◆ Data cleaning
- ◆ Data integration
- ◆ Data reduction
- ◆ Data transformation and discretization



# Why Preprocessing Data?

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- ◆ Imperfect data
- ◆ incomplete: missing attributes, values
- ◆ noisy: errors, outliers
- ◆ inconsistent: discrepancies
- ◆ Measure of data quality
  - ◆ accuracy, completeness, consistency, timeliness, believability, value added, interpretability, accessibility



# Data Cleaning

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## ◆ Missing data

- ◆ ignore, manual
- ◆ automatic: constant, attribute mean, class attribute mean, most probable

## ◆ Noisy data

- ◆ binning: bin mean, median, boundary
- ◆ regression
- ◆ clustering



# Correlation Analysis

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- ◆ Correlation coefficient (numerical data)

$$r_{A,B} = \frac{\sum_{i=1}^N (a_i - \bar{A})(b_i - \bar{B})}{N\sigma_A\sigma_B} = \frac{\sum_{i=1}^N (a_i b_i) - N\bar{A}\bar{B}}{N\sigma_A\sigma_B}$$

- ◆  $\chi^2$  (chi-square) test (categorical data)

$$\chi^2 = \sum_{i=1}^c \sum_{j=1}^r \frac{(o_{ij} - e_{ij})^2}{e_{ij}}$$

$$e_{ij} = \frac{\text{count}(A = a_i) \times \text{count}(B = b_j)}{N}$$

- ◆ Correlation does not imply causality!



# Data Reduction

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- ◆ Data reduction: Why?  
Goal?
- ◆ Dimensionality reduction
- ◆ attribute subset selection;  
Wavelet transform;  
principle component analysis (PCA)
- ◆ Numerosity reduction
  - ◆ regression, log-linear models
  - ◆ data cube aggregation
  - ◆ histograms, clustering, sampling



# Data Transformation

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- ◆ **Smoothing:** remove noise from data
- ◆ **Aggregation:** summarization
  - ◆ e.g., daily sales => monthly, annual sales
- ◆ **Generalization:** concept hierarchy climbing
  - ◆ e.g., street => city => state
- ◆ **Normalization:** scale to fall within a range
  - ◆ min-max, z-score, decimal scaling



# Chap 4: Data Warehouse

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- ◆ Separate data store for information processing
  - ◆ nonvolatile
- ◆ Characteristics
  - ◆ subject-oriented
  - ◆ integrated
  - ◆ time-variant
  - ◆ day to day operation
  - ◆ vs. decision support
- ◆ OLTP vs. OLAP



# Chap 4 & 5: Data Cube

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- ◆ Multi-dimensional data model
- ◆ Dimensions, facts
- ◆ Schema: star, snowflake, fact constellation
- ◆ Typical operations
- ◆ roll-up, drill-down, slice and dice, pivot, drill-across, drill-through
- ◆ Cuboid cells, materialization of data cube
- ◆ full, partial, or no materialization



# Chap 6 & 7: Frequent Pattern Mining

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- ◆ Basic concepts, roadmap
- ◆ Mining frequent itemsets
- ◆ Mining association rules
- ◆ Correlation analysis
- ◆ Constraint-based association mining



# Mining Frequent Itemsets

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- ◆ **Apriori algorithm**
- ◆ k-itemsets, candidate (k+1)-itemsets
- ◆ Challenges
- ◆ #scans, #candidates, support counting
- ◆ Partition, sampling, dynamic itemset counting, hash-tree
- ◆ **FP-growth**: FP-tree, conditional pattern base, conditional FP-tree
- ◆ Vertical data format



# Associations & Correlations

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- ◆ Association: **Support & confidence**
- ◆ Correlation rule
  - ◆  $A \Rightarrow B$  [support, confidence, **correlation**]

$$\text{lift}(A, B) = \frac{P(A \cup B)}{P(A)P(B)}$$
$$\chi^2 = \sum_{i=1}^c \sum_{j=1}^r \frac{(o_{ij} - e_{ij})^2}{e_{ij}} \quad e_{ij} = \frac{\text{count}(A = a_i) \times \text{count}(B = b_j)}{N}$$



# Rule (Pattern) Constraints

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- ◆ Metarule
- ◆ Anti-monotonicity
- ◆ Monotonicity
- ◆ Succinctness (**skip**)
- ◆ Convertible constraints
  - ◆ with proper ordering of items
- ◆ Strongly convertible (**skip**)
- ◆ Inconvertible (**skip**)



# Chap 8: Classification

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- ◆ Basic concepts
- ◆ Decision tree induction
- ◆ Bayesian classification
- ◆ Rule-based classification (skip)
- ◆ Model evaluation and selection
- ◆ Improve classification accuracy
- ◆ Summary



# Classification vs. Prediction

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- ◆ Classification: categorical class labels
- ◆ Prediction: continuous-valued functions
- ◆ Training vs. testing
- ◆ Supervised vs. unsupervised
- ◆ Evaluation criteria
  - ◆ accuracy, speed, robustness, scalability, interpretability, goodness of rules



# Decision Tree Induction

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- ◆ Top-down, recursive, divide-and-conquer
- ◆ Attribute selection
  - ◆ information gain
  - ◆ gain ratio
  - ◆ gini index
- ◆ Attribute split
  - ◆ discrete, continuous, discrete-binary



# Bayesian Classification

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- ◆ Bayes' Theorem

$$P(H|X) = \frac{P(X|H)P(H)}{P(X)}$$

- ◆ Naive Bayesian classifier

$$P(X|C_i) = \prod_{k=1}^n P(x_k|C_i) = P(x_1|C_i) \times P(x_2|C_i) \times \cdots \times P(x_n|C_i)$$



# Classifier Accuracy Measures

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- ◆ Accuracy, error rate
- ◆ Confusion matrix
- ◆ Costs and benefits of TP, TN, FP, FN
- ◆ Sensitivity, specificity, precision
- ◆ Predictor error measures
  - ◆ error, square error
  - ◆ absolute, mean, relative



# Classifier/Predictor Evaluation

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- ◆ Holdout, random sampling
- ◆ Cross-validation
  - ◆ k-fold stratified cross-validation
- ◆ Bootstrapping
  - ◆ sample with replacement
  - ◆ .632 bootstrap
- ◆ Model selection
  - ◆ t-test, ROC curves



# Ensemble

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- ◆ Ensemble
- ◆ combination of multiple models
- ◆ bagging: majority vote (equal weight)
- ◆ boosting (e.g., Adaboost): weighted



# Chap 9:Advanced Classification

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- ◆ Bayesian belief networks
- ◆ Backpropagation
- ◆ Support vector machines
- ◆ Lazy learning (**skip**)
- ◆ Other classification methods (**skip**)
- ◆ Additional topics regarding classification (**skip**)
- ◆ Summary



# Chapter 10: Cluster Analysis

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- ◆ Basic concepts
- ◆ Partitioning methods
- ◆ Hierarchical methods
- ◆ Density-based methods
- ◆ Grid-based methods  
(skip)
- ◆ Evaluation of clustering  
(skip)
- ◆ Summary



# Cluster Analysis

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- ◆ Unsupervised learning
- ◆ Intra/inter-cluster similarity
- ◆ Requirements
  - ◆ scalability, various data types, arbitrary shape, minimal domain knowledge, noisy data, incremental, high dimensionality, constraint-based, interpretability



# Clustering Methods

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- ◆ Partitioning methods
  - ◆ k-means, k-medoids
- ◆ Hierarchical methods
  - ◆ agglomerative, divisive
  - ◆ BIRCH, CHAMELEON

- ◆ Density-based methods
  - ◆ DBSCAN, DENCLUE
- ◆ Grid-based methods
  - ◆ STING, CLIQUE ([skip](#))



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## ◆ Chapter II: Advanced Cluster Analysis

- ◆ probabilistic model-based clustering
- ◆ clustering high-dimensional data (**skip**)
- ◆ clustering graph and network data (**skip**)
- ◆ clustering with constraints (**skip**)



# Reminder

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- ◆ Practice exam review: Tuesday, Oct 27
- ◆ Midterm exam: Thursday, Oct 29
- ◆ Exam schedule confirmation from TA

