Data Mining 1

Introduction to the Student Projects
Outline

1. Requirements for Student Projects
2. Requirements for Project Reports
3. Final Exam
Student Projects

• **Goals**
  
  – Gain practical experience with the complete data mining process
  – Get to know additional problem-specific
    • preprocessing methods
    • data mining methods

• **Expectation**
  
  – Select an interesting data mining problem of your choice
  – Solve the problem using
    • the data mining methods that we have learned so far plus some advanced problem-specific data pre-processing
    • other data mining methods which might be helpful for solving the problem and build on what we learned in class
Procedure

• Teams of **five** students
  – realize a data mining project
  – write a 12 page summary of the project and the methods employed in the project
  – present the project results to the other students (10 minutes presentation + 5 minutes discussion)

• Final mark for the course
  – 20 % written summary about the project
  – 5 % project presentation
  – 75 % written exam
## Schedule

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<th>Week</th>
<th>Wednesday</th>
<th>Thursday</th>
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<tr>
<td>26.10.2020</td>
<td>Introduction to Student Projects</td>
<td>Exercise: Classification 2</td>
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<td>Monday, November 2nd, 2020, 23:59: Submission of Project Proposals</td>
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<tr>
<td>02.11.2020</td>
<td>Lecture: Regression</td>
<td>Project feedback</td>
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<td>09.11.2020</td>
<td>Project feedback</td>
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<td>16.11.2020</td>
<td>Lecture: Text Mining</td>
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<td>23.11.2020</td>
<td>Project feedback</td>
<td>Exercise: Text Mining</td>
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<td>30.11.2020</td>
<td>Lecture: Association Analysis</td>
<td>Presentation of project results</td>
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Wednesday, December 23th, 2020, 23:59: Submission of Project Report
Where to find interesting Data Sets?

- **Public sector data**
  - US government: [https://www.data.gov](https://www.data.gov)
  - UK government: [https://data.gov.uk](https://data.gov.uk)
  - EU: [https://www.europeandataportal.eu](https://www.europeandataportal.eu)
  - Health data (over 125 years): [https://www.healthdata.gov/](https://www.healthdata.gov/)

- **Data registries**
  - Datasets hosted on Amazon AWS [https://registry.opendata.aws](https://registry.opendata.aws)
  - Million Song Dataset, 1000 Genome Project, database of satellite imagery of Earth from NASA, Web Crawl
  - Google’s Dataset Search: [https://datasetsearch.research.google.com/](https://datasetsearch.research.google.com/)
  - Microsoft Datasets: [https://msropendata.com/](https://msropendata.com/)
  - Dataset collection on Github: [https://github.com/awesomedata/awesome-public-datasets](https://github.com/awesomedata/awesome-public-datasets)
  - Data Hub: [http://datahub.io](http://datahub.io)
Where to find interesting Data Sets?

- **Knowledge graphs**
  - Wikidata: [https://www.wikidata.org](https://www.wikidata.org)
  - BabelNet: [https://babelnet.org](https://babelnet.org)
  - DBpedia: [http://wiki.dbpedia.org](http://wiki.dbpedia.org)

- **Language resources**
  - WordNet: [https://wordnet.princeton.edu](https://wordnet.princeton.edu)
  - Project Gutenberg (36,000 ebooks): [http://www.gutenberg.org/](http://www.gutenberg.org/)

- **Competitions**
  - Kaggle: [https://www.kaggle.com/](https://www.kaggle.com/)
  - Data Mining Cup: [http://www.data-mining-cup.de](http://www.data-mining-cup.de)
  - KDD Cup: [https://www.kdd.org/kdd-cup](https://www.kdd.org/kdd-cup)
  - DrivenData: [https://www.drivendata.org](https://www.drivendata.org)
  - CrowdAnalytix: [https://www.crowdanalytix.com](https://www.crowdanalytix.com)
Where to find interesting Data Sets?

- **Covid19**
  - Johns Hopkins University: [https://github.com/CSSEGISandData/COVID-19](https://github.com/CSSEGISandData/COVID-19)
  - Our world in data: [https://github.com/owid/covid-19-data](https://github.com/owid/covid-19-data)
  - Harvard: [https://dataverse.harvard.edu/dataverse/covid19](https://dataverse.harvard.edu/dataverse/covid19)
Where to find Information about additional Methods?

• Pang-Ning Tan, Michael Steinback, Vipin Kumar: Introduction to Data Mining, Pearson / Addison Wesley.


Where to find Information about additional Methods?

• Check out the solutions to your problem that other people have tried.
  – for instance by looking at submissions of the KDD Cup or Data Mining Cup as well as Kaggle
  – or search for relevant scientific papers using Google Scholar
Some Project Ideas (not binding)

• Web Log Mining
  – Learn a classifier for the categorizing the visitors of your website.
  – Which features matter? Number of pages visited, time on site, ..
    (Bing Liu Chapter 12.x)
  – Preprocess some web log data outside RapidMiner
  – Learn and evaluate classifier within RapidMiner

• Wikipedia Contributors / Hoax Articles
  – Examine the edit history of Wikipedia contributors
  – Cluster users by different attributes (no of edits, edits/day, topic, ...)
  – Or learn a classifier for the categorizing Wikipedia contributors

• Sentiment Analysis for Discussion Forum / Rating Site / Tweets
  – Are people positive or negative about topic / product? (Bing Liu 11.x)

• SPAM Detection
  – eMail, blog or discussion forum (Bing Liu 6.10, 11.9)
Some Projects realized in previous Semesters

- Mannheim Police Reports
  - Learn classifiers for police reports
  - Identify type of incident, severity of incident, location of incident
- Bundesliga Betting Rules
  - Find rules that help you to predict the outcome of a Bundesliga game
- last.fm Playlist Analysis
  - Cluster last.fm users according to the style of the songs they are listening to
  - Find commons sets of songs for the different clusters
- Analysis of Training Data of a Fitness Center
  - Find different customer groups by clustering exercise data
  - Find frequent combinations of exercises
- Sentiment Analysis of Tweets about Movies
  - Learned classifier from IMDB movie reviews
  - Applied and tested with tweets afterwards
- Classifying a Document‘s Perspective
  - using the example of Israeli – Palestinian Essays
Project Outlines

- maximum 4 pages including title page, using DWS master thesis layout
  - Include a project name and your team number on the first page!
- due Monday, November 2nd, 2020, 23:59
- send by eMail to Heiko, Nico, Ralph & Sven
- answer the following questions:
  1. What is the problem you are solving?
  2. What data will you use?
     - Where will you get it?
     - How will you gather it?
  3. How will you solve the problem?
     - What preprocessing steps will be required?
     - Which algorithms do you plan to use?
     - Be as specific as you can!
  4. How will you measure success? (Evaluation method)
  5. What do you expect your results to look like? (Model/Clusters/Patterns)
Coaching Sessions

• We will give you tips and answer questions concerning your project.
• Registration via email is mandatory!
  – until Monday night!
  – including the questions that you like to discuss
• We will assign you a time slot afterwards and inform you about the slot via email

• Every team has to attend at least one coaching session!
Project Report

- 12 pages (exactly!) including title/toc page and reference page
  - max. 10 pages, no appendix
  - Each extra page and each day of late submission downgrades your mark by 0.3!
- due **Wednesday, December 23th, 2020, 23:59**
- send by email to Heiko, Nico, Ralph & Sven
Project Report

• Outline for project report:
  1. Application area and goals (Business Understanding)
  2. Structure and size of the data set (minimum 1 page) (Data Understanding)
  3. Preprocessing
  4. Data Mining
     – (External Knowledge)
     – ML approaches
     – Evaluation
  5. Results
Project Report

• Requirements

1. You must use the DWS master thesis layout.
2. Please cite sources properly. Preferred citation style [Author, year].
3. Also submit your RapidMiner processes/Python scripts and (a subset) of your data.
4. Include your project name and your team number on the first page!
Checklist

• Business Understanding:
  – What is the actual problem (in the domain)?
  – What is the target variable?
  – Clustering/Classification/Regression?

• Data Understanding:
  – Are examples sorted (time series)?
  – What is the distribution of labels / target variable?
  – Are all attributes and their types listed?
  – Are attributes explained?
  – What is the quality of the data?
Checklist

• Preprocessing
  – Are missing values replaced (in case needed)?
  – Checked for outliers (and handled them)?
  – Validity tests of attributes (Height above sea level < 9000)?
  – Check for inconsistencies (age=42, birthday=03/07/1997)
    • Check for duplicates
  – Data normalization
  – Additional features generated?
  – Has binning been tried out?
  – Correlation analysis implemented?
  – Feature subset selection implemented?

• External Knowledge:
  – Are additional datasets used?
Checklist

• ML approaches
  – How many different ML approaches were tried out?
  – Do you have at least one symbolic and one non symbolic approach?
  – Do you have at least one baseline (majority class / mean value / domain specific ...)?

• Evaluation
  – Is there a train test split or 10 fold cross validation implemented
  – Is eval stratified?
  – Cost matrix or not?
  – Are the hyper parameters tuned (in which range / which attributes) ?
  – Are the tests systematic?
  – Analyse a symbolic model (how does the decision tree / rules /... looks like)
  – What features do have a high impact on the result?
Checklist

• Result
  – Is the result is critically evaluated
  – Is the result analyzed against the baseline
  – What does the result mean given the problem (could you use it)
Deadly Errors to Avoid

• Normalize numeric data before calculating any similarity metrics

• If your data is unbalanced
  – balance your training data
  – do NOT balance your test data
  – report P/R/F1, not accuracy
Final Exam

• Date: Thursday, December 10th, 2020
• Duration: 60 minutes
• Structure: 5 - 6 open questions that
  – check whether you have understood the content of the lecture
  – require you to describe the ideas behind algorithms and methods
  – might require you to do some simple calculations