National College of Ireland

Masters in Science in Data Analytics (MSCDAD_A)
Masters in Science in Data Analytics (MSCDAD_B)
Masters in Science in Data Analytics(MSCDAD_C)
Masters in Science in Data Analytics(MSCDA_JANOL23)
Postgraduate Diploma in Science in Data Analytics (PGDDA_JAN23)

Database and Analytics Programming

Team Project (70%)

Semester 2, 2022/23

Submission deadline: 25th April 2023 at 23:59

1 Introduction

This project is designed to evaluate the learning objectives of the Database and Analytics Programming module as outlined below:

- LO1: Analyse, compare, contrast and critically evaluate the characteristics of programming languages, programming environments and database systems commonly utilised for data analytics solution implementation.
- LO2: Critically assess the challenges associated with processing big data datasets and compare and contrast programming for big data vis-à-vis programming for conventional datasets.
- LO3: Evaluate tools and techniques for managing the data pipeline and preparing data for further analysis through data wrangling, cleaning, and validation.
- LO4: Critically assess methods and practices for software development in order to design and implement data programming requirements.
- LO5: Evaluate, design and implement solutions for processing datasets by using key programming patterns and constructs for data analytics, relevant programming languages, and suitable database systems.

2 Objectives

The objective of this project is to identify and carry out a series of analyses on a collection of large datasets that are somehow related or complement each other, utilising appropriate programming languages, programming environments and database systems.

This project is a team project. Teams will have a **maximum of 3** members.

Your project must incorporate the following elements:

- 1. A minimum of two datasets should be used, of which at least one should be semi-structured. For teams of 3, a third dataset (semi-structured) will be required.
- 2. Datasets must be programmatically stored in appropriate database(s) prior to processing.

- 3. Programmatic pre-processing, transformation, analysis and visualisation of the data.
- 4. Programmatically storing the processed output data in appropriate databases.

For example, you could use Python to programmatically retrieve a semi-structured dataset (XML or JSON) or web-scraped or streaming data) and store this data in MongoDB.

You could then use an ETL process (for instance, using Dagster or Luigi Python libraries) to read these data from MongoDB, to process and transform it, finally creating structured datasets that you store in PostgreSQL for later usage.

Following that you could conduct further analysis on these data to find interesting patterns my applying knowledge gained in other modules (e.g., statistical analysis,machine learning), and generate visualisations to better present the results.

Each dataset should contain at least 1,000 records. Some appropriate datasets may be found at:

- https://catalog.data.gov/dataset?res_format=XML
- http://aiweb.cs.washington.edu/research/projects/xmltk/xmldata/
- https://data.gov.ie/dataset?res_format=JSON
- https://catalog.data.gov/dataset?res_format=JSON
- https://data.worldbank.org/

A list of other potential sources will be posted on Moodle.

3 Deliverables

Project Report

The objectives, methodology and results of your analysis should be presented in the form of a project report. This report should discuss the programming and data processing challenges that you encountered and the means and mechanisms you implemented to overcome these challenges.

The report should be around 3,000 words in length (excluding references), should use appropriate academic style and referencing, and be presented in the IEEE conference format. Templates for Microsoft Word and LATEX can be downloaded from the IEEE¹.

The report should contain the following sections:

• Abstract

This should provide a summary of the project objectives, methods and results. Take a look at abstracts from papers in your literature review to get an idea of what constitutes a good/bad

Introduction

Here you should provide a short motivation for the project, describe the relevance of the topic and state the objectives of the project. Note that the proposed analysis should answer a novel question, which should be clearly stated by means of appropriately formed research question(s).

¹https://www.ieee.org/conferences_events/conferences/publishing/templates.html

• Related Work

In this section, you should summarise relevant academic work that addressed similar problems or guided your decisions. Note that this should be a **critical evaluation**. It should be more than a mere summary of the works and should discuss their limitations and implications.

Methodology

This section should contain:

- A detailed description of the underlying dataset(s) and your justification for choosing them.
- Full descriptions and justifications of the data processing activities carried out, such as use of APIs, databases, etc.
- Complete descriptions and justifications of the implemented data processing algorithms.
- Justifications for the choice of technologies used, such as programming languages, libraries and databases.
- Diagrams providing a visual overview of the data gathering, processing and analysis flow.

• Results and Evaluation

Here you should present the results of your work, making appropriate use of figures, tables, etc. You should provide evidence of how the project objectives were met, ensuring that you discuss your research findings, their interpretation(s) and implications.

• Conclusions and Future Work

In this section you should detail what others can/could learn from your work. You should discuss your findings in the context of the research question(s) you elicited earlier. You should present the limitations of your work, i.e. this should be a critical self-evaluation. Lastly, you should suggest potential directions for future work. Typically you would describe what you would do differently or how you would extend your work if you had more time.

Bibliography

Here you should provide a **complete list** of the academic works cited and online materials used in the project. References should be included as in-text citations **using the IEEE citation style**.

Project Presentation

As a team you should create a video presentation (maximum 10 minutes long) that will act as a discussion point for your work. It should be used to provide a discussion on what you did, how you did it, why you did it and what you discovered.

Note that although individual members will be presenting different parts of the video, each member of the team is expected to be able to present all aspects of the work individually and without assistance from other group members, if required.

Code Artefact

You should create a *zip* or *gz* archive all assets such as program code, data and system configuration details. If working as a team, there is an additional requirement for you to set up a **private** GitHub repository where each team member should commit their own code and any changes they make to code created by other team members.

Note: Having one team member be responsible for making all commits is not acceptable.

4 Submission

The project carries 70% of the total marks for the module, with a mark of 40% or greater being required to pass.

There should be only **one submission per team**, consisting of:

- A **project report** that must include the full name of each team member (as per NCI official documents) and their student number. These must be clearly visible on the front page of the report. The report should be named *teamX.pdf* where X is your team identifier and should be uploaded as a PDF document to the **Project Report** Turnitin link on Moodle.
- A code artefact, which should be uploaded as a zip or gz archive to the Code Artefact link on Moodle. This should be named team X.zip or team X.gz, where X is your team identifier.
- A video presentation that must include the full name of each team member (as per NCI official documents) and their student number. These must be clearly visible at the start of the video. This should be uploaded as a mp4 video to the **Project Presentation** link on Moodle.
- A work breakdown report describing in detail the contribution of each team member. Again, this should be in PDF format and should include the full name of the team members (as per NCI official documents) as well as their student numbers. This should be named teamXworkbreakdown.pdf (where X is your team identifier) and should be uploaded to the Work Breakdown Report Turnitin link on Moodle.

Late submissions will not be accepted unless an extension has been requested through NCI360 and officially approved.

5 Marking

The project will be marked according to the grading rubric provided at the end of this document.

6 Academic Integrity

Any written work created by others must be properly cited and should be paraphrased or summarised where possible, otherwise it should be included in quotes. Figures not created by you should include an acknowledgement detailing the name(s) of the creator(s). Code found on the internet should not be claimed as your own, but instead a comment should be included in the source code indicating where you obtained it.

Students are strongly advised to familiarise themselves with the Guide to Academic Integrity produced by the NCI Library².

Note: All submissions will be electronically screened for evidence of academic misconduct, e.g. plagiarism, collusion and misrepresentation. Any submission showing evidence of such misconduct will be referred to the college's academic misconduct committee for disciplinary action.

²https://libguides.ncirl.ie/academicintegrity

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Grading Rubric - Database and Analytics Programming Project Semester 1 - 2022/23

| Criterion | Solid H1 $\geq 80\%$ | $H1 \ge 70\% < 80\%$ | $H2.1 \ge 60\% < 70\%$ | $\mathbf{H2.2} \ge 50\% < 60\%$ | $Pass \ge 40\% < 50\%$ | Fail < 40% |
|---------------------|--------------------------|---------------------------|---------------------------|---------------------------------|---------------------------|----------------------------|
| Project | Very challenging pro- | Challenging project ob- | Reasonable project ob- | Reasonable project ob- | The objectives are clear, | The objectives of the |
| Objectives | ject objectives are ex- | jectives are well presen- | jectives are well presen- | jectives are clear, are | if unambitious and are | project are unclear, |
| (10%) | ceptionally well presen- | ted, are fully met and | ted, fully met and ad- | mostly met and ad- | at least partially met | have not been discussed. |
| | ted, fully met and thor- | thoroughly discussed. | equately discussed. | equately discussed. | and briefly discussed. | It is not possible to |
| | oughly discussed | | | | | discern if the objectives |
| | | | | | | have been met. |
| Literature | An excellent critical | A very good critical | A good analysis of | An adequate analysis of | A limited analysis of | Little or no relevant lit- |
| Review | analysis of substantive | analysis of substantive | relevant literature. The | mostly relevant literat- | some relevant literature | erature reviewed. Very |
| (10%) | and highly relevant lit- | and relevant literature. | critical analysis aspect | ure. The critical ana- | but it lacks evidence of | limited evidence of un- |
| | erature. | | could be somewhat | lysis aspect could be sig- | understanding. | derstanding. |
| | | | stronger. | nificantly stronger. | | |
| Data Complexity and | The datasets have been | The datasets have been | The datasets have been | The datasets have been | The datasets were | Only one somewhat |
| Handling | well prepared and mean- | well prepared and mean- | well prepared and ex- | appropriately prepared | appropriately handled | trivial dataset was used. |
| (20%) | ingfully explored. All | ingfully explored. All | plored. At least one | for analysis. At least | given the objectives. | No database was used |
| | datasets were stored in | datasets were stored in | dataset was stored in | one dataset was stored | The use of databases | to store the datasets. |
| | appropriate databases | appropriate databases | an appropriate data- | in an appropriate data- | is very basic and some | No obvious develop- |
| | before and after pro- | before and after pro- | base. At least one data- | base. At least one | inappropriate choices | ment was carried out. |
| | cessing. At least two | cessing. At least two | set has a high degree of | of the datasets is non- | may be evident. The | |
| | datasets have a high | datasets have a high de- | complexity. | trivial. | datasets are somewhat | |
| | degree of complexity. | gree of complexity. | | | trivial. | |
| | At least one dataset | | | | | |
| | was programmatically | | | | | |
| | retrieved - through an | | | | | |
| | API or by web scraping. | | | | | |
| Data Processing Im- | The data processing al- | The data processing al- | The use of data pro- | The use of data pro- | Appropriate but basic | Poor or no implement- |
| plementation | gorithms used play a | gorithms used play a | cessing algorithms is | cessing algorithms is | use of data processing | ation. If an implement- |
| (20%) | well conceived and es- | well conceived and es- | well-thought and appro- | meaningful and appro- | algorithms. Basic use of | ation is provided, it |
| | sential role in meet- | sential role in meet- | priate for the project | priate for the project | data programming lan- | demonstrates inappro- |
| | ing the project object- | ing the project object- | objectives. Compre- | objectives. There is | guages and a limited | priate use of data pro- |
| | ives. The implement- | ives. Multiple data pro- | hensive use of at least | evidence of appropriate | number of techniques. | cessing algorithms. |
| | ation significantly ex- | cessing techniques / lan- | one data programming | use of at least one data | | |
| | ceeds the stated min- | guages were employed. | language and multiple | programming language | | |
| | imum requirements. | | techniques. | and a small number of | | |
| | | | | appropriate techniques. | | |

Grading Rubric (continued)

| Criterion | Solid H1 $\geq 80\%$ | $H1 \ge 70\% < 80\%$ | $H2.1 \ge 60\% < 70\%$ | $H2.2 \ge 50\% < 60\%$ | $Pass \ge 40\% < 50\%$ | Fail < 40% |
|-------------|---|---|---|--|---------------------------------------|---|
| Level of | All parts of the analysis | All parts of the analysis | Most core components | Some components of | Individually all com- | Little or no evidence of |
| Automation | are automated within | are automated within | of the analysis are auto- | the analysis are in- | ponents of the analysis | automation. |
| (10%) | a single process control | a single process control | mated within a single | cluded within a larger | are automated, but not | |
| | flow. Every run of the | flow. | process control flow. | process. However, some | necessarily connected | |
| | process can result in | | | components of the ana- | together as part of a lar- | |
| | different results as new | | | lysis are run as separate | ger process flow. | |
| | data is extracted and | | | processes. | | |
| | subsequently included, | | | | | |
| | such as data obtained | | | | | |
| | through an API. | | | | | |
| Results and | Three or more insight- | Three or more interest- | Three or more interest- | Two or more interesting | Two or more interest- | Little to no non- |
| Conclusions | ful findings are excel- | ing and non-arbitrary | ing non-arbitrary find- | non-arbitrary findings | ing non-arbitrary find- | arbitrary results |
| (20%) | lently presented and | findings are presen- | ings are presented and | are presented and ap- | ings are presented but | and/or findings are |
| | thoroughly discussed in | ted and thoroughly | thoroughly discussed. | propriately discussed. | are poorly discussed. | presented. |
| | the context of the do- | discussed the context | | | | |
| | main using appropri- | of the domain using | | | | |
| | ate references to prior | appropriate references | | | | |
| 0 11 6 | work. | to prior work. | TTT 11 1 . 1 | 4.1 . 1 | A. 3 3 | T 1 |
| Quality of | Exceptionally well writ- | Well written, with no | Well written, but has | Adequately written. | Adequately written, | Poorly written and |
| Writing | ten, with no language | significant language er- | a few significant lan- | but as a few significant | with some significant | littered with typo- |
| (10%) | errors. All figures are | rors. All figures are | guage or style errors. | language and/or style | language and/or style | graphical errors and/or |
| | well conceived, read- | well conceived, readable | Figures are well presen- | errors. Some figures are | errors. Figures may | poor use of English. |
| | able and correctly captioned. The IEEE tem- | and appropriately captioned. The IEEE tem- | ted. The IEEE template and length limit | may be hard to read. The IEEE template | be hard to read or | The IEEE template |
| | | | are adhered to. Refer- | * | presented in a suboptimal manner. The | was not used. Figures |
| | plate is strictly adhered to. The report does not | plate is adhered to. The report does not exceed | ences are complete and | and length limit are mostly adhered to. Ref- | timal manner. The IEEE template may | may be hard to read. References (if any) are |
| | exceed the length lim- | the length limits. Refer- | correctly used. | erences are complete, | not have been followed. | largely incomplete. |
| | its. All references are | ences are appropriately | correctly used. | and correctly used. | References are mostly | rargery incomplete. |
| | appropriately and cor- | and correctly used. | | and correctly used. | complete and correctly | |
| | rectly used. | and correctly used. | | | used. | |
| | rectry used. | | | | useu. | |