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Assessment Guide

Data Intensive Architectures (DIA) : MSc Data Analytics : Dr. H. González-Vélez Issued: 16th October 2020

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Administrative

This module assessment guide has been designed to facilitate the understanding of the scope for the assessment of the Data Intensive Architectures (DIA) module as part of the MSc Data Analytics at the School of Computing of the National College of Ireland.

The learning outcomes of this module are:

- **LO1** Critically compare and contrast multiple distributed system models and their associated enabling technologies.
- **LO2** Demonstrate in-depth knowledge of different types of processing on different data-intensive computational resources.
- LO3 Identify and categorise platforms and software environments for cloud and cognitive computing.
- **LO4** Critically analyse the features of high performance computing platforms and how they enable parallel and distributed programming paradigms.

Assessment of this module is based on **TWO** coursework assignments which represent 20% and 80% of the final grade awarded respectively. As published in the School portal, the deadlines for submission of assignments are:

- 1. Hypothesis (20%): Report–DEADLINE: 29th October 2020 9am.
- 2. Project (80%): Video, Code and Report– DEADLINE: 8^{th} December 2020 9am.

Should any student miss the assessment with a valid reason, (s)he can now apply for an application for coursework Extension/Re-run Form online, via NCI360. PCF forms are no longer in use and will not be accepted by the School of Computing office.

Both assignments have to be electronically submitted using the dropbox provided in the module page in Moodle. Please:

- ensure that your name in full (as per NCI official documents) and student number are clearly visible on the front page of the written reports; and
- name your files starting with the first letter of your given name followed by the first three letters of your surname, your student id, a dash, and the word "report". No spaces or any other alphanumeric characters should be included in the filename. That is to say, when "Mary Murphy" with student id 20123456 submits her coursework report, she should name the file: mmur20123456-report.pdf.

N.B.

All submissions will be electronically screened for evidence of academic misconduct (plagiarism and collusion). The combined mark for the coursework represents 100% of your overall mark for this module.

Hypothesis 20%

The first assignment entails a report which provides the definition of the hypothesis to drive the final project.

(a) Deliverable

1. A short (up to 2 pages) report formatted using the IEEE double-column template. It should include the citations to the key sources using IEEE bibliographic format. Please refer to the following link for the formatting requirements and LATEX/Word templates:

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

(b) Structure

This assignment ought to include the following components:

Title Choose a title that encompasses the specific topic within the Data Intensive Architectures (DIA) area.

Question Phrase your hypothesis addressing its specifics and the measurable side.

Value Explain why this hypothesis is worth investigating and why its answer is non self-evident.

Justification Describe why the hypothesis is feasible to be pursued via a MapReduce approach, clear (measurable), and ethical. Cite the key 2-3 sources (journal articles/conference papers/patents) you are basing your hypothesis on.

References Include the citations to the key 2-3 sources using the IEEE style, as bibliography at the end College of the document.

Project 80%

(a) Datasets

Programmatically analyse and interrogate 2 (or more) curated Open Data [1] datasets, i.e. they should be free to be used, re-used and redistributed by anyone and curated by a recognised European entity. Your data sets should fulfil the following minimum requirements:

- 1. Be related in some way.
- 2. Complement each other such that your study (or something very similar) could not be conducted without one of your datasets.
- 3. Be at least moderately sized for your project to be considered "data-intensive". Whilst there is no upper limit on size, be realistic with respect to the capabilities of your cloud instance(s) and processing times.
- 4. Be ethically employed [2].

(b) Processing

In terms of what to do with your datasets, please observe the following minimum requirements:

- 1. Programmatically prepare your datasets this includes:
 - (a) Extracting them from well-curated Irish/European **Open Data** repositories and placing them into your own block/blob storage or similar. There is a large number of repositories with relevant data, where some examples include, but are not limited to:
 - Ireland's Open Data Portal https://data.gov.ie/
 - Ordnance Survey Ireland Open Data Portal https://data-osi.opendata.arcgis.com/
 - Dublinked: Open Data for the Dublin Region https://data.smartdublin.ie/
 - Central Statistics Office, Ireland: http://www.cso.ie
 - eHealth Ireland Open Data Portal https://data.ehealthireland.ie/
 - Open Data available from Fáilte Ireland https://failteireland.ie/Research-Insights/ Open-data.aspx
 - Fingal Open Data http://data.fingal.ie
 - UK's open government data repository: http://data.gov.uk
 - European Data Portal https://www.europeandataportal.eu/en and the EU Open Data Portal http://data.europa.eu/
 Additional relevant Open Data Publishers in Ireland are listed in https://wiki.openstreetmap.org/wiki/Ireland/Open_Data. If additional datasets are required, you can, in exceptional circumstances, resort to public cloud providers Open Data repositories such as the Amazon's public dataset repository: https://aws.amazon.com/datasets or Google's Public Data Directory: http://www.google.com/publicdata/directory.
 - (b) Clean them.
 - (c) Conform/transform and combine the datasets.
 - (d) Providing at least a cursory exploratory study to motivate your project focus and formally describe the data.
 - (e) Prepare the data for at least one complete analysis with particular emphasis on how they interrelate.
- 2. Perform analysis using MapReduce [3].
- 3. Interrogate the combined dataset and MapReduce results to provide at least 3 interesting insights into the data you have chosen.

As discussed during lectures, some ideas about possible projects can be found through the case studies reported in [4].

N.B.

Your project MUST explicitly address Data Quality as defined by the ISO/IEC 25012 standard.

(c) Deliverables

- 1. A short video (up to 7 minutes in duration) containing a short individual oral presentation explaining the key findings of the project and a demonstration of the MapReduce code execution.
- 2. A zip archive of all code and datasets used to produce the results also with a (brief) explanation on how they were run in order to facilitate their rerunning if needed.

3. A 6-page report formatted using the IEEE double-column template. The report has to include all figures and any references to existing work. Please refer to the following link for the formatting requirements and LATEX/Word templates:

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

(d) Structure of the Report

All relevant findings should be compiled into an accompanying report, which should be submitted along with any programming code elements and the video demo. Your project report should discuss the challenges that you encountered whilst handling your chosen datasets and the means and mechanisms you implemented to overcome these challenges. It should be structured as follows:

Abstract: a roughly 200-word executive summary of the project and the key results

Introduction: set the scene of the project, i.e., the objectives of the project (for example what are you trying to find out)

Data: describe your datasets including metadata elements, the format the datasets are represented with, where the datasets can be retrieved from and their licensing, how they were generated, when they were generated (ideally have records from 2019 and 2020), and who generated them.

Methodology: essentially, how have you addressed Data Quality in your project?. Conceivable here, you would also discuss how other people have used the datasets you have chosen. Ethical consideration on the datasets and their intended use should also be integrated here.

Implementation and Architecture: how have you built your application workflow, what components and/or forms of analytics have you used and why?

Results: what did you find out about your data sets? e.g.: what was surprising? what was expected? what did you find out with respect to your motivational question that is presented in the introduction? Finally discuss any interesting aspects of your results or key challenges you solved in achieving your results.

Conclusions and future work: what (in general) did you learn and find out? If you were to do the project again, what would you do differently? If you had more time (e.g. in your final project) what would you do next to extend your work?

References: a complete list of academic works and/or online materials used in the project. References should be included as in-text citations according to the IEEE citation style. To find academic works and citation style guidelines, please refer to the NCI Library guide for Data Analytics: http://libguides.ncirl.ie/dataanalytics

(e) Marking Grid

MARKING GRI	D- Data Intensive Arc	hitectures (DIA), Asse	ssment Guide, Dr. Ho	racio González-Vélez	MARKING GRID- Data Intensive Architectures (DIA) , Assessment Guide, Dr. Horacio González-Vélez Due date: 8^{th} December 2020 .
ASSESSMENT	EXCELLENT /	G00D	SATISFACTORY	THRESHOLD	FAIL
CRITERIA	VERY GOOD				
Project Object-	Challenging project	Reasonable project	There are clear ob-	There are some	Cannot discern pro-
ives, Datasets,	objectives are well	objectives are clear,	jectives, which are at	objectives, which	ject objectives, an-
and Ethics:	presented, met,	and at mostly met.	least partially met.	are at least par-	d/or if project ob-
(40% weight)	and thoroughly	Datasets have been	Datasets have been	tially met. Datasets	jectives were met.
	discussed. Datasets	prepared and mean-	somehow prepared	are prepared and	Less than 2 datasets.
	have been well	ingfully explored.	and explored. At	probably somewhat	No obvious devel-
	prepared and ex-	At least one datasets	least one datasets is	trivial. Scant con-	opment conducted.
	plored. At least two	has some degree of	non-trivial. Some	sideration of ethical	No consideration to
	datasets have a high	complexity. Reason-	consideration of eth-	issues	any ethical issues
	degree of complex-	able consideration	ical issues.		
	ity. Comprehensive	of ethical issues.			
	consideration of	<u> </u>			
	ethical issues	0	30		
MapReduce	Excellent/very	Good application of	Adequate applica-	Weak application of	Poor application of
Code design,	good application of	MapReduce design	tion of MapReduce	MapReduce design	MapReduce design
methods, and	MapReduce design	principles in terms	design principles in	principles and lim-	principles and very
analysis.	principles in terms	of appropriate:	terms of appropri-	ited evidence of un-	limited evidence
(30% weight)	of appropriate:	methodology; and	ate: methodology;	derstanding of: ap-	of understanding
	methodology; and	methods for gener-	and methods for	propriate methodo-	of: appropriate
	methods for gener-	ating and analysing	generating and	logy; and methods	methodology; and
	ating and analysing	data.	analysing data.	for generating and	methods for gener-
	data.			analysing data.	ating and analysing
				0	data.
Identified im-	Excellent/very	Good consideration	Adequate consider-	Limited/weak	Very limited and
	good consideration	of potential research	ation of potential re-	consideration of	considera
structure, pro-	of potential research	impact/outcomes.	search impact/out-	sear	of potential research
posal abstract,	impact/outcomes.	Good abstract and	comes. Adequate	impact/outcomes.	impact/outcomes.
and referen-	Excellent/very	structure. Most ref-	, L	Weak abstract and	Poor abstract and
cing.	good abstract and	erencing consistent	ture. Adequate con-	structure. Frequent	structure. Very fre-
(30% weight)	structure. All ref-	and appropriate.	sistent and appro-	inconsistent an-	quent inconsistent
	erencing consistent		priate referencing.	d/or inappropriate	and/or inappropri-
	and appropriate			referencing.	ate referencing.
	70-100	69-09	50-59	40-49	<40
			OF 1710a a 0 /00/ 10	777111177	

THE FINAL MARK MUST BE 40% OR ABOVE TO ACHIEVE A PASS

References

- [1] R. Kitchin, The Data Revolution: Big Data, Open Data, Data Infrastructures & their Consequences. London: Sage, 2014. ISBN: 978-1-4462-8747-7.
- [2] A. Zwitter, "Big data ethics," Big Data & Society, vol. 1, no. 2, p. 2053951714559253, 2014.
- [3] J. Dean and S. Ghemawat, "MapReduce: Simplified data processing on large clusters," Communications of the ACM, vol. 51, pp. 107–113, Jan. 2008.
- [4] J. Kolodziej and H. González-Vélez, eds., High-Performance Modelling and Simulation for Big Data Applications Selected Results of the COST Action IC1406 cHiPSet, vol. 11400 of Lecture Notes in Computer Science. Springer, 2019. (Open Access)–ISBN 978-3-030-16271-9.