

CM4603 – Coursework 1 (Group)

Academic Year	2024-25	
Semester	1	
Module Number	CM4603	
Module Title	Language Processing and Information	
Module Title	Retrieval	
Assessment Method	A group submission of code and report	
Assessment Method	followed by individual viva	
Deadline (time and date)	12 midnight on 29 th November 2024	
Submission	Assessment Dropbox in the Module Study	
345111331011	Area in CampusMoodle	
Word Limit	1500 words	
Use of Generative Artificial Intelligence (AI)	Is authorised	
Module Co-ordinator	Ruvan Weerasinghe	

What knowledge and/or skills will I develop by undertaking the assessment?

Students will be able to collect and collate text data from online sources, explore various natural language pre-processing techniques on the data and perform diverse feature extraction methods in preparing this data for modelling tasks. They will then be able to grasp the effects of these techniques on text processing tasks such as classification and evaluate the resulting models. They will also be able to interpret the results of such modelling including explaining them.

On successful completion of the assessment students will be able to achieve the following Learning Outcomes:

- 1. Describe and critically review natural language processing techniques.
- 2. Select, analyse and apply NLP algorithms to reason with textual content.
- 3. Pre-process and transform textual content for algorithms to satisfy information retrieval needs using a range of similarity metrics.

What knowledge and/or skills will I develop by undertaking the assessment?

Please also refer to the Module Descriptor, available from the module Moodle study area.

What is expected of me in this assessment?

Task(s) - content

Sri Lanka is set on putting behind its past and regaining its position as one of the popular tourist destinations in Asia. You are tasked with achieving this by showcasing some of the best destinations it has to offer. To do this, you are required to access at least 5000 user reviews from at least 100 Sri Lankan hotels on tripadvisor.com and process this information using the following steps.

Task 1: Describe

After performing any cleaning steps on the raw data, fully describe the raw dataset and the cleaned dataset collected from TripAdvisor in terms of the number of hotels, the number of reviews for each hotel, the total number and the number of unique words in the whole dataset. You are also required to plot a histogram of article lengths (in words) of the full dataset. Perform any other exploratory data analysis (EDA) to describe the dataset for bonus points.

Task 2: Establishing Ground Truth

The number of reviews is too large to be able to manually establish a 'gold standard' for the sentiment expressed in each review of this dataset. As an alternative, it is suggested that you use 3 different existing sentiment classifiers with at least one being a sentiment lexicon¹. Use a majority vote scheme to label the reviews of the entire dataset as the way to establish the 'ground truth'.

Task 3: Feature Extraction

Create two (02) sparse vector and two (02) dense vector (non-transformer) representations of the dataset to extract features from the text dataset. Describe the resulting shapes of the data matrices in terms of the number of rows and columns, justifying your choice of vectorization methods used.

Task 4: Text classification

Use each of the feature extraction methods applied in Task 3 to evaluate the performance of three (03) significantly different non-deep learning algorithms for predicting the sentiment of user reviews, justifying

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¹ SentiWordNet, Bing Liu Lexicon and VADER are some examples.

What is expected of me in this assessment?

the rationale for selecting the three algorithms. Interpret the results of the performance of each combination above.

Task 5: Using pre-trained vectors

Compare the performance of the best combinations explored in Task 4 with that of a pre-trained contextual embedding used with the same three (03) algorithms selected in Task 4. Also compare the performance of the resulting models with a deep learning model using this same contextual embedding.

Task 6: Text Clustering

It is desirable to dig deeper to try to understand what the different aspects of the hotel that these sentiments are directed at. For this purpose, you are required to cluster the documents using topic modelling to arrive at a distinct set of aspects from the dataset. Justify your model by randomly taking 50 reviews from the test set and manually labelling the aspects they represent and then checking² how they would be clustered in your topic model.

Task(s) - format

You are required to **formulate solutions for each of tasks 1 through 6** above, clearly explaining your Python code and specifying the outputs produced by the code for the dataset used, **in an iPython Notebook** named **Solution_Group#.ipynb** based on the template given (the # in the filename should be replaced with your group number number – 1 to 10). For each such part, a descriptive summary with an interpretation should be given for the output obtained after each executable cell. The notebook should be compressed as a .zip file. You also need to submit a PDF version of the notebook with the same filename as the notebook, except that the extension should be .pdf.

In addition, you should write a comprehensive report of not more than 1500 words in two parts. Part **A** should state any considerations given by your group outside the direct specification of this coursework brief. In addition, a description of what you would do differently to improve the rigour of your exploration if you had to do it again. **Part B** should contain a trace your learning journey as a group through the sources used and the kind of prompts you gave generative AI. Note that you **should NOT** reproduce the steps you carried out step by step in the coursework. The PDF version of this report should be named,

² A common way to compare clustering with ground truth include Rand Index and the Jaccard Coefficient.

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What is expected of me in this assessment?

Report_Group#.pdf where the # in the filename should be replaced with your group number number – 1 to 10.

Your iPython Notebook files (the zipped notebook and the converted pdf) and your study report should be submitted as three separate files to Campus Moodle. Note that the PDF files should NOT be compressed. Submissions which do NOT adhere to these formatting and naming conventions would incur delays in grading and possibly result in penalties.

Participation in the individual physical viva is mandatory for all group members. Non-participation will result in a 'No Show' result which will be interpreted as a Non-Submission for the group member concerned.

How will I be graded?

A number of subgrades will be provided for each criterion on the feedback grid which is specific to the assessment. The overall grade for the assessment will be calculated using the algorithm below.

- At least 50% of the subgrades to be at Grade A, at least 75% of the subgrades to be at Grade B or better, and normally 100% of the subgrades to be at Grade C or better.
- At least 50% of the subgrades to be at Grade B or better, at least 75% of the subgrades to be at Grade C or better, and normally 100% of the subgrades to be at Grade D or better.
- At least 50% of the subgrades to be at Grade C or better, and at least 75% of the subgrades to be at Grade D or better.
- At least 50% of the subgrades to be at Grade D or better, and at least 75% of the subgrades to be at Grade E or better.
- At least 50% of the subgrades to be at Grade E or better.
- Failing to achieve at least 50% of the subgrades to be at Grade E or better.
- NS Non-submission.

NB: Non-participation in the presentation will result in the next lower grade being awarded to the group member concerned.



Feedback grid

	В	C	D	E	F
EXCELLENT	COMMENDABLE/VERY GOOD	GOOD	SATISFACTORY	BORDERLINE FAIL	UNSATISFACTORY
are collected, cleaned appropriately, described using the specification. In	explores at least 5 ways of tokenizing text, commenting on their merits using a		The group describes the dataset used and explores some ways of tokenizing text, and use a metric to compare them.	adequately and explores some ways of tokenizing	The group does not describe the dataset used adequately and uses some standard ways of tokenizing text.
is clearly documented. It is also justified and validated.	dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and justifies them for downstream	vector representations to extract features of the dataset, specifies the shape of the resulting	The group uses two sparse and two dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and interprets the numbers.	representations to extract features of the dataset, specifies the shape of the	The group uses sparse and dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices.
extraction methods are clearly demonstrated. The	modelling and interprets the results obtained by each in a	algorithms to use for modelling and interprets	The group demonstrates the use of 3 significantly different algorithms to use for modelling and interprets the results	algorithms to use for modelling and comments	
	At least 5000 reviews from 100 different hotels are collected, cleaned appropriately, described using the specification. In addition, a meaningful EDA has been included. How the ground truth for the reviews was arrived at is clearly documented. It is also justified and validated. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The	At least 5000 reviews from 100 different hotels are collected, cleaned appropriately, described using the specification. In addition, a meaningful EDA has been included. How the ground truth for the reviews was arrived at is clearly documented. It is also justified and validated. The group describes the dataset used very well and explores at least 5 ways of tokenizing text, commenting on their merits using a justified metric. The group uses two sensible dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and justifies them for downstream tasks. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The	At least 5000 reviews from 100 different hotels are collected, cleaned appropriately, described using the specification. In addition, a meaningful EDA has been included. How the ground truth for the reviews was arrived at is clearly documented. It is also justified and validated. The group describes the dataset used and explores multiple ways of tokenizing text, commenting on their merits using a justified metric. The group uses two sensible sparse and two sensible dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and justifies them for downstream tasks. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The group justifies the choice of the 3 algorithms to use for modelling and interprets the results obtained by each in a	At least 5000 reviews from 100 different hotels are collected, cleaned appropriately, described using the specification. In addition, a meaningful EDA has been included. How the ground truth for the reviews was arrived at is clearly documented. It is also justified and validated. The group describes the dataset used and explores at least 5 ways of tokenizing text, commenting on their merits using a justified metric. The group uses two sensible dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and justifies them for downstream tasks. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The	At least 5000 reviews from 100 different hotels dataset used very well and explores at least 5 ways of appropriately, described using the specification. In addition, a meaningful EDA has been included. How the ground truth for the reviews was arrived at is clearly documented. It is also justified and validated. The group describes the dataset used and explores multiple ways of tokenizing text, commenting on their merits using a justified metric. The group uses two sensible dense vector representations to extract features of the shape of the resulting matrices and justifies them for downstream tasks. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The group justifies the choice of a lagorithms to use for modelling and interprets the clearly demonstrated. The group describes the dataset used and explores some ways of tokenizing text, and use a tokenizing text, and use a metric to compare them. The group uses two sparse and two dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and mentions their suitability for downstream tasks. Two sparse vector and two dense vector feature extraction methods are clearly demonstrated. The group justifies the choice of a lagorithms to use for modelling and interprets the clearly demonstrated. The group describes the dataset used and explores some ways of tokenizing text, and use a metric to compare them. The group uses two sparse and two dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and mentions their suitability for downstream tasks. The group uses two sparse and two dense vector representations to extract features of the dataset, specifies the shape of the resulting matrices and interprets the lasse of the group demonstrates the dataset used and explores multiple ways of tokenizing text, commenting on their merits using a metric. The group uses two sparse and two dense vector representations to extract f

GRADE	Α	В	C	D	E	F
DEFINITION /	EXCELLENT	COMMENDABLE/VERY GOOD	GOOD	SATISFACTORY	BORDERLINE FAIL	UNSATISFACTORY
CRITERIA						
(WEIGHTING)						
	Each of the vectorization	The group successfully	The group successfully	The group successfully	The group implements a	The group fails to
	techniques is used for	implements multiple pre-		implements a pre-trained	pre-trained word	successfully implement a
	training 3 significantly	trained word embedding	word embedding scheme	word embedding scheme	embedding scheme for	pre-trained word
Task 4	different non-DL	schemes for feature			feature extraction and	embedding scheme for
	algorithms, checking for	extraction and compares and	·	compares the	reports the performance	feature extraction and
(1 subgrade)			on the performance of		of the algorithms with	reports the performance
	The resulting predictive	1.5	<u> </u>	algorithms with own	own feature extraction	of the algorithms with
	models are evaluated	algorithms with own	feature extraction	feature extraction	scheme.	own feature extraction
	rigorously and	tokenization and feature	schemes.	schemes.		scheme.
	interpreted accurately.	extraction schemes.				
	0 1	A detailed reflection on using	A detailed reflection on	The report comments on	The report comments	The report fails to
	contextual embedding	generative AI in dealing with	the use of generative Al	various tokenization, data		comment on the choices
	and how it can be used	tokenization, class imbalance,	_	pre-processing and model		made with respect to pre-
T1-F	with the algorithms used	overfitting and other	tokenization, data pre-		modelling choices and	processing, modelling or
Task 5	is demonstrated. A clear	important data pre-	processing, modelling and	coursework.	diagnostics.	diagnostics.
(4	and valid comparison of	processing, modelling and	diagnostics steps in the			
(1 subgrade)	these 3 classifiers with a	diagnostics with some	coursework.			
	deep learning classifier is	considerations unspecified in				
	presented, with adequate	the coursework brief.				
	checks for validity.					
	•	A clear application of topic		The reviews are clustered		Little effort is made to
		modelling and its potential	to discover clusters in the	_	made to cluster the data	cluster the data in order
	topic modelling and its	use in aspect-based	reviews. The resulting	within them.		to understand the
Task 6	potential use in aspect-	sentiment analysis is	clusters are validated.		which wasn't valid.	categories of the reviews.
Tusk o	based sentiment analysis	evidenced. The validation of				
(1 subgrade)	is evidenced. The	the clustering achieved with				
(1 Subgrade)	validation of the	the ground truth has been				
	clustering achieved with	properly carried out.				
	the ground truth has					
	been properly carried out.					

GRADE	Α	В	C	D	E	F
DEFINITION /	EXCELLENT	COMMENDABLE/VERY GOOD	GOOD	SATISFACTORY	BORDERLINE FAIL	UNSATISFACTORY
CRITERIA						
(WEIGHTING)						
	The group demonstrates	The group appears to have	Some group work is	Individuals have	Only some parts of the	Poor execution of the
	understanding beyond	worked well in addressing the	evidenced in addressing	contributed to the	coursework have been	coursework and weak
	the exact specification of	coursework. Individual	the coursework. The	execution of the	executed and commented	reflective report on the
	the coursework in the way	learning process is tracked via	learning process is	coursework. A very similar	on. The learning process	learning process by
	that other considerations	the generative AI prompts	documented by each	learning process is	is not well captured.	members. Poor
Report & Viva	have been addressed. In	that have been employed.	member. Members of the	documented by all	Members of the team	understanding of the
	addition, the learning	Most members of the team	team demonstrate this	members. Members of	don't have a clear idea of	main learning by group
(2 subgrades)	process is well tracked via	demonstrate this	understanding at	the team provide some	what was learnt.	members.
	the generative AI prompts	understanding.	different levels.	evidence of this learning.		
	that have been employed.					
	This is evidenced in each					
	member of the team.					

Coursework received late will be regarded as a non-submission (NS) and one of your assessment opportunities will be lost.



What else is important to my assessment?

What is the Assessment Word Limit Statement?

It is important that you adhere to the Word Limit specified above. The Assessment Word Limit Statement can be found in Appendix 2 of the <u>RGU Assessment Policy</u>. It provides detail on the purpose, setting and implementation of wordage limits; lists what is included and excluded from the word count; and the penalty for exceeding the word count.

What's included in the word count?

The table below lists the constituent parts which are included and excluded from the word limit of a Coursework; more detail can be found in the full Assessment Word Limit Statement. Images will not be allowed as a mechanism to circumvent the word count.

Excluded	Included
Cover or Title Page	Main Text e.g. Introduction, Literature Review, Methodology, Results, Discussion, Analysis, Conclusions, and Recommendations
Executive Summary (Reports) or Abstract	Headings and subheadings
Contents Page	In-text citations
List of Abbreviations and/or List of Acronyms	Footnotes (relating to in-text footnote numbers)
List of Tables and/or List of Figures	Quotes and quotations written within ""
Tables – mainly numeric content	Tables – mainly text content
Figures	
Reference List and/or Bibliography	
Appendices	
Glossary	

What are the penalties?

The grade for the submission will be reduced to the next lowest grade if:

- The word count of submitted work is above the specified word limit by more than 10%.
- The submission contains an excessive use of text within Tables or Footnotes.

What else is important to my assessment?

What is plagiarism?

Plagiarism is "the practice of presenting the thoughts, writings or other output of another or others as original, without acknowledgement of their source(s) at the point of their use in the student's work. All materials including text, data, diagrams or other illustrations used to support a piece of work, whether from a printed publication or from electronic media, should be appropriately identified and referenced and should not normally be copied directly unless as an acknowledged quotation. Text, opinions or ideas translated into the words of the individual student should in all cases acknowledge the original source" (RGU 2022).

What is collusion?

"Collusion is defined as two or more people working together with the intention of deceiving another. Within the academic environment this can occur when students work with others on an assignment, or part of an assignment, that is intended to be completed separately" (RGU 2022).

For further information please see Academic Integrity.

What if I'm unable to submit?

- The University operates a <u>Fit to Sit Policy</u> which means that if you undertake an assessment then you are declaring yourself well enough to do so.
- If you require an extension, you should complete and submit a <u>Coursework Extension Form</u>. This form is available on the RGU <u>Student and Applicant Forms</u> page.
- Further support is available from your Course Leader.

What additional support is available?

- RGU Study Skills provide advice and guidance on academic writing, study skills, maths and statistics and basic IT.
- RGU Library guidance on referencing and citing.
- The Inclusion Centre: Disability & Dyslexia.
- Your Module Coordinator, Course Leader and designated Personal Tutor can also provide support.

What are the University rules on assessment?

The University Regulation 'A4: Assessment and Recommendations of Assessment Boards' sets out important information about assessment and how it is conducted across the University.