

Project Title: System Verification and
Validation Plan for Natural Language
Processing for Mental Health Risk Prediction

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Revision History

Date	Version	Notes
11/03/2023	1.0	Revision 0
1/08/2024	1.1	Verbiage Changes & adjustments to AI Guidelines
3/03/2024	1.2	Updated Tests to reflect updates to functional requirements
3/04/2024	1.3	Added more evaluation metrics and updated Tests to reflect updates to functional requirements
3/04/2024	1.4	Updated traceability table
3/06/2024	1.5	Updated task 3 information

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1 Symbols, Abbreviations, and Acronyms

symbol	description
UQAM	Université du Québec à Montréal
RMSE	Root Mean Square Error
MZOE	Mean Zero-One Error
MAE	Mean Absolute Error
RS	Restraint Subscale
ECS	Eating Concern Subscale
SCS	Shape Concern Subscale
WCS	Weight Concern Subscale
RP	Reciprocal Rank
DCG	Discounted Cumulative Gain
IDCG	Ideal Discounted Cumulative Gain
NDCG	Normalized Discounted Cumulative Gain

2 General Information

2.1 Summary

The CLEF eRisk competition, better known as the Early Risk Prediction on the internet competition, is an organization that hosts a yearly event where numerous research teams come together to explore new strategies and applications of early detection technologies, particularly in the field of health and safety. The team at McMaster has partnered alongside the University of Quebec in Montreal to help leverage their existing research and work done in prior years of the eRisk competition to help design a new system entry for this year. The team will not be in charge of the system operations portion of the project, but instead will be focused on designing new strategies and implementations for the Natural Language Processing portion of the project. The NLP will be responsible for analyzing textual input from real user data, and will use its provided training data and algorithms to determine the probability of the user showing signs of a selected mental health issue. The competition normally consists of three different tasks, all of which require different Natural Language processing techniques, and to help select strategies, the team will be using data from the prior year's competition, which was used to find symptoms for depression, pathological gambling and eating disorders.

2.2 Objectives

The primary objectives of the current process are to build confidence regarding the various techniques and libraries required to efficiently analyze the datasets provided by the eRisk Competition. The testing outlined in this document is to help demonstrate the team's familiarity and competency to adapt to the various challenges presented by eRisk, whether that be through sentence ranking, early detection through post history, or estimation of severity level through user submissions. This is meant to build confidence for the project's stakeholders on the code efficiency, accuracy and reliability. The exact number of challenges that the team will participate in is variable, as the scope ensures competing in at least one challenge, as even though the project consists of a larger team than prior years, past teams have not always been able to take on all three challenges, extending ourselves to take on all tasks would be going beyond SRS expectations. An additional objective is

to increase efficiency and accuracy of the existing work done by the team in prior years, this is a goal for both the team and the stakeholders at UQAM, but it is not a formalized metric in the SRS, as there is no defined expectation of how much better the project is supposed to perform.

2.3 Relevant Documentation

- [Problem Statement](#)

The Problem statement document abstractly identifies the problem to be solved, characterizing it in terms of its inputs and outputs, which gives context to the related environment and stakeholders.

- [Development Plan](#)

The Development Plan contains the project development overview including team roles, team meeting and communication plans, git workflow and project schedule.

- [SRS](#)

The Software Requirements Specification identifies the various functional and non-functional requirements related to the project, while providing context for the benefits and usage scenarios of the project.

- [Hazard Analysis](#)

The Hazard Analysis identifies various hazards and obstacles to the project's development and operation, along with mitigation strategies to help deal with those potential hazards. It carries a focus on ensuring the project can be delivered to the standards of the eRisk competition while maintaining safety and privacy of user data.

- [Module Guide](#)

The Module Guide decomposes the eRisk NLP project into numerous modules based on the principle of information hiding and separating data structures, in order to help increase understandability of various project components.

- [Module Interface Specification](#)

The Module Interface Specification identifies modules found in the Module Guide and decomposes them into more primitive data types and functions, in order to better observe the module functionality and ensure it's behaving as intended.

- [McMaster AI Guidelines](#)

McMaster's Provisional Guidelines on the use of AI provides useful rulesets that helps the team ensure the safe and responsible use of AI practises.

- [eRisk 2022 Paper](#)

The prior efforts done by the UQAM team in the eRisk competition were instrumental in the team's foundation of understanding and served as a jumping off point for any efforts.

3 Plan

This section will outline the plan for verification and validation of a number of different components within this project (??). This section will first give an overview of the members that are considered part of the verification and validation team and what their roles and involvement will consist of. Next there will be subsections regarding verification of the project's [SRS](#) (??) along with the design plan (??) and the verification and validation plan itself (??). This section will then outline the verification plan that has been created for the implementation of the project and model (??). Automated Testing and verification tools that will be used throughout this project is then the next subsection that is outlined in ?? followed lastly by the validation plan for the software (??).

3.1 Verification and Validation Team

The verification and validation team will consist of the core team members (Matthew, Jessica, Ben, Yaruo and Michael) along with the Capstone professor, the team's TA, Marie-Jean and Diego from the Montreal team, and Professor Mosser. The core team will be responsible for creating test suites

that ensure correctness in the solution along with catching possible bugs and issues that may arise. The team will be responsible for creating suitable edge cases to evaluate the correctness of the work along with general automated test suites that will be automatically deployed when new code is pulled.

The core team will be responsible for creating all test suites, along with executing them and documenting the results. The team will also be responsible for making any changes that are required after testing the code. All core team members will have a hand in all sections of testing but different team members will have different focused responsibilities. Firstly, all core team members will be responsible for documenting the results of the automated test suites when their code enters the repository through a pull request. More specifically, Yaruo and Michaels main responsibility will be creating a set of tests including edge cases for the NLP model that will ensure that our model functions as expected for a wide variety of input data. They will be required to create test suites along with automated test suites that will be run periodically when pull requests happen. Jessica, Ben and Matthew on the other hand will have the primary responsibility for training the data vs a training data set in order to determine the accuracy of the model. They will also have the responsibility to ensure code structure in the test suites that are created along with organizing the suites while Yaruo and Michaels main role is coming up with and creating the tests.

The team will meet with Professor Mosser and their TA as well periodically throughout each milestone in order to discuss the requirements for the project regarding the current milestone and to help solidify expectations. This is an opportunity for the team to ask any questions they may have as well. The team also will meet with Marie-Jean and Diego from the Montreal team periodically to help guide the team with regard to requirements of the project and the eRisk competition along with what validation means to them. They will help guide the team to find what the important things are to focus on within our project. Team 8 will also provide the team feedback for every milestone. Lastly, the actual competition itself will be the final validation step when testing the model against a new set of data and reporting how the model performs.

3.2 SRS Verification Plan

All of the members of the core team will be taking part in the SRS Verification Plan along with Group 8, Professor Mosser and the TA. The team will verify the contents of the [SRS](#) by comparing the requirements outlined by the eRisk competition along with our team in Montreal with the requirements stated in our SRS document. This will ensure that the team has covered everything from the SRS document and can see if there is anything new that must be added. Most of the SRS verification plan will involve inspections not only from our team, but also from the team in Montreal along with peer reviews and ad hoc feedback from group 8.

Since our project is unique in the fact that the team is submitting to a competition with a rigid output and guideline structure, the team will also have to compare the rubric for the SRS document with what has been done in the original SRS document to ensure that all the required checkpoints have been covered. On top of this, the team will verify the SRS document by going over the feedback given to on the SRS revision 0 from TA along with the feedback received from group 8 regarding the SRS revision 0. Lastly, when verifying the SRS document, the teamwe will talk with Professor Mosser and ask him any questions that arise during the reviewing and verification process within the group.

These are some of the major areas the team would like to cover in the SRS verification plan to ensure a test exists for each of these components:

- Is a functional overview of the system provided in the SRS?
- Are high-level usage scenarios included?
- Has the software environment and all its elements been specified?
- Are the limitations and exclusions accurate and all encompassing of the true limitations and exclusions of the project?
- Are the inputs and outputs of the system specified?
- Is there any unnecessary overlap in our requirements?

- Are there any other components outside of Text Pre-Processing, Vectorization, Prediction and Output? For those 4 sections, are there any additional functional or non-functional requirements that are missing?
- For every function that is outlined, are the inputs specified sufficient enough to perform the required functionality?
- Do any requirements conflict with each other or does each requirement avoid conflicts with other requirements?
- Are all detailed usage scenarios covered in the document and are completed in adequate detail or are there others we could add?
- Does each component and requirement have a priority?
- Are there any changes in the prioritization chart we should add due to changes in the code since the creation of the original SRS (revision 0)?
- Are the requirements clear enough that it would be possible to give to an independent team for implementation and they would be able to understand the requirements?
- Is every requirement testable?
- Is the verification and acceptance criteria specific enough? If not, what quantifiable metrics can be added to each point?
- Has every definition, abbreviation and acronym been defined in our Glossary?
- Are all imposed technical choices listed in the document?
- Are the risks that may be present in this project along with their mitigation analysis included in this document?
- Is the “Requirements Process and Reports” accurate to what is expected by each of the teams listed in the “Requirements Process and Reports” section?
- Does the Context and Overall Objectives section paint an accurate enough description of the project’s context and objectives?

3.3 Design Verification Plan

The plan for the design verification will be to go through the SRS document and make sure that each of the outputs and inputs that were planned for are accounted for and included in our design planning and document. The team will also look at reviews given by Group 8 on the SRS document to consider any changes they may want to add or things to consider for the design. The team will go through the same process for feedback received from the TA on the SRS report. They will also utilize the guidance of Marie-Jean and Diego in order to verify that they are on the right path with the design. This assistance can help guide the design and verify that it checks all the boxes and functionalities that are required for the competition.

The following section will go over some design verification questions that will be asked regarding the functions of each of the major components in our design.

Text Pre-Processing Component:

- Does the system tokenize text by dividing it into individual units of words and or sub-words?
- Does the system convert generated tokens into lowercase to preserve consistency?
- Does the system remove stop word tokens (common words that do not commonly have an effect on the meaning)?
- Does the system reduce tokens to their root forms (Ex: “moving” to “move”)?
- Is the output from this component usable by the Vectorization Component?

Vectorization Component:

- Does the system convert tokens into numerical vectors in this stage?
- Are these numerical vectors usable by the Prediction Component of the model?

Prediction Component:

- Is there a created training model within the design?
- What percentage of the given data is used for training vs testing?
- Is there a prediction algorithm that is implemented into the solution? What is the level of accuracy and what is acceptable levels of accuracy (this will be determined further down the road with our team in Montreal on what the team deems acceptable)?
- Does the system make predictions using the vectorization tokens, the training model, as well as with the created prediction algorithm?

Output Component:

- Does the system output the predictions to the console?
- Is the output produced understandable by the project team?
- Is the output in a format that a healthcare worker would be able to read and interpret the results?

3.4 Verification and Validation Plan Verification Plan

For the Verification and Validation Plan Verification Plan the team will be making use of the reviews given for the VnV Plan from group 8 as well as from the TA. This will help the team gain some insight on areas that may be missing or need more work within the document. The team will also be going through the [SRS](#) document and ensuring they have created at least 1 test for each functional and nonfunctional requirement. It is important to ensure that they have sufficient tests for the requirements. The team would also like to do some mutation testing in the code to determine what changes will affect the tests created and in what way. It will also show that there is a possibility of missing some important tests if a mutation expected to result in a change or failure, results in no failed tests.

Below is a checklist of the type of questions and topics the team would like to cover as part of the Verification and Validation Plan Verification Plan.

- Is the brief summary of the software in the summary section enough to get an understanding of the software that is being tested and its general functions?
- Are the objectives for the VnV plan clearly outlined and is each objective covered at some point throughout the document?
- Is the planning section detailed with checklists for each section and steps that will be taken for each plan?
- Is there at least one test created for each requirement in the SRS document?
- Are all functions of each component being tested? This includes the Text Pre-Processing, Vectorization, Prediction and Output Components.
- Are there a wide variety of tests including edge cases? Do the non-functional requirements contain tangible quantifiable values?
- Is the traceability between test cases and requirements clearly shown in section 4.3?

3.5 Implementation Verification Plan

The large majority of the implementation for this project will revolve around achieving an acceptable level of accuracy in our system's predictions. This document goes more in depth about how the team will do this in the implementation in section 4.1.1 but the basis of this process to ensure acceptable model accuracy levels is done through train/test splits, training models on the test sets, predicting the values of test data with these models, and then comparing these predicted values with the known expected values. All of the sections just listed are key areas of the implementation and will need to be tested in order to verify our implementation. There are 4 tests that will be run, typically in tandem with each other in one automatic test suite as outlined in 4.1.1. The largest area will be involving training and testing the data. Data will be provided by eRisk and the data will be used 3 times using different splits for each copy regarding what data is used for training vs testing. These predictions on the testing data will then be evaluated on

a relevant set of metrics in order to test the program's accuracy and implementation. This set of metrics has yet to be determined due to the fact that the eRisk tasks have not been released yet for this year so the team is currently unclear what metrics would be appropriate to evaluate our model at the current time. The team will also be running a series of unit tests on the program in order to test the implementation of each part and function of the code. These tests will be recorded in the unit testing plan.

As new code is added to our repository and as the team works on our model, we will be using implementation verification techniques like code walk-throughs and code inspections periodically. Every time a new pull request is made, a fellow teammate will be required to perform a code inspection for that pull request and give feedback. If there is something unclear, the two teammates will meet and the reviewer will receive a code walk-through from the reviewee. We will also perform code walk-throughs for new changes made to the model when we meet as a team. Lastly we plan to have continuous integration that will allow the team to run automated testing for every pull request that gets created as well as utilizing Fake9 as mentioned below to ensure our implementation follows proper coding standards.

3.6 Automated Testing and Verification Tools

There will be three main tools the team will be using for automated testing and verification. Firstly, we will be using Flake9 which is a fork of Flake8. This will help the team follow proper coding standards and prevent problems arising regarding topics such as syntax errors, typos, incorrect styling, bad formatting and more. This will also help save time for when we do peer code reviews as a team. We will also be creating test suites using Pytest since it is a free open-sourced, simple and scalable Python-based Test Automation Framework. We will be using GitHub Actions in order to implement continuous integration and continuous delivery into our project. This will allow us to automatically build, test, and deploy our pipeline. This will create a pipeline that allows the team to build and test every pull request when they happen and also when the team deploys and merges a pull request onto another branch.

3.7 Software Validation Plan

There are a multitude of methods and parts to our software validation plan. Firstly, our primary validation method will be the actual process of competing in the eRisk competition. During the competition they will take our model and test it with new data and observe the output. This will be the final and largest validation step for us that will help validate the correctness and precision of our design. The team will also have a set of training data that will be used throughout the design process to train against, the team can use this to test how close the results are to the expected results, validating how accurate the design is.

Another large aspect of the software validation plan will be our review sessions and meeting with Marie-Jean and Diego throughout the project. This will be a time for them to look at our software and run it themselves. A large checkpoint when they will have an opportunity to have our project demoed for them is around the Rev 0 demo. This will give them an opportunity to validate that our project passes all the requirements of the competition or if there is something else we need to add or change. This is an important time for us to take this feedback and use it in order to improve our project. This will also be a time for us to ask questions and for them to confirm the requirements are all being met in our design. These questions and requirements would largely surround the overall output of the design and the overall process (largely focusing on the training of the data and the prediction model created and used). It would also include validation of each of the program's components individually and as a collective (components being the Text Pre-Processing, Vectorization, Prediction and Output components).

Before that however, The team will meet with professor Mosser to discuss if he thinks that all the requirements documented line up with and are accurate to the requirements outlined for the course. The team will do this with Marie-Jean and Diego as well with respect to the requirements from the competition. The team will also go over the SRS document one last time to ensure that every requirement is met.

Regarding the demo with Professor Mosser, Marie-Jean and Diego the team will include the following questions:

- Is the output format what is expected by the competition?

- Are there any requirements not being met within the competition (question for Marie-Jean and Diego)?
- Are there any requirements that are not being met regarding the project rubric from McMaster (question for Professor Mosser)?
- Is our text pre-processing component sufficient in which we use tokenization, lowercasing, stopword removal, and stemming or Lemmatization? If not, what should we add to our text pre-processing component or what methods that we currently have employed are not needed?
- Is our current method of vectorization appropriate for the task we are completing in your opinion? If not, what would you change or what different approaches do you recommend we look into.
- Are there any areas of our code or processes we take that are not allowed in the competition or do not follow the competitions standards?
- What are your thoughts on our topic modeling approach? Do you have any suggestions?
- Are the metrics we use for determining accuracy appropriate in your opinion? If not, what form of metrics should we be using? What level of accuracy would you consider acceptable for this metric?
- In your opinion is the run time of our program to output the results acceptable? If not, what would be an acceptable time?

4 System Test Description

4.1 Tests for Functional Requirements

Our main requirements for the system are built around achieving an acceptable level of accuracy in our system's predictions. This is the primary focus of this section.

4.1.1 Ensuring Model Acceptability

The system is based in machine learning and will therefore need to be tested and evaluated using a typical machine learning evaluation method: creating

train/test splits, training models on the test sets, predicting the values of test data with these models, and then comparing these predicted values with the known expected values. This section first includes information explaining how the team plans to go about this process as well as identifying areas of uncertainty before providing relevant tests.

4.1.1.1 Training and Test Data

Training data given by eRisk will be split 80/20 into train and test sets. Each task has some concept of higher or lower risk, either in the form of positive/negatives, survey answers, a continuous scale, or something else. Depending on the task a metric will be created from expected values representing this risk (an example may be averaging survey answers) it is unclear what these metrics will involve currently as eRisk has not yet released this year's tasks. Train and test splits will be created to have similar distributions of this risk metric (ie. similar numbers of low/mid/high risk individuals). How similar the distributions should be is as of yet unknown and will require further experimentation when eRisk releases their datasets for this year.

When testing the system three different train/test splits will be produced, all splits will be formed on the full training data provided by eRisk, thus giving three copies of said training data with the train and test subsets being different for each copy. These copies will try to maximize the difference between their splits. The model will be trained on a training split and then will give predictions on the test set. These predictions will be evaluated on a relevant set of metrics (outlined below in section ??), and an average of the metrics across all three splits will be used to determine model performance.

4.1.1.2 Metrics

The following metrics will be used to evaluate the accuracy of the model for each task. These metrics were selected based on recommendations of the eRisk competition:

Task 1:

1. Reciprocal Rank (RP) is a measure used to evaluate the effectiveness of a ranked list of items. It is often used in the context of question answering or information retrieval systems.

$$RP = \frac{1}{\text{rank}}$$

2. Normalized Discounted Cumulative Gain (NDCG) is a metric used to evaluate the quality of a ranked list of items, considering both the relevance and the position of each item in the list:

$$DCG = \sum_{i=1}^n \frac{2^{rel_i} - 1}{\log_2(i + 1)}$$

$$IDCG = \sum_{i=1}^n \frac{2^{rel_i} - 1}{\log_2(i + 1)}$$

$$NDCG = \frac{DCG}{IDCG}$$

Task 2:

1. F-Score was used in the UQAM team's 2021 submission for the detection of problem gambling task (?). F-Score is a combination of two other measures, precision and recall, and is used to measure the accuracy of positive/negative tests while taking into account false positive and negatives:

$$\text{precision} = \frac{\text{true positives}}{\text{true positives} + \text{false positives}}$$

$$\text{recall} = \frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$$

$$\text{F-score} = 2 \frac{\text{precision} * \text{recall}}{\text{precision} + \text{recall}}$$

Task 3:

The details of the metrics used for task 3 can be found in the SRS: section S2 task 3. In light of this only a brief overview will be provided here.

1. MZOE evaluates what proportion of the predictions are incorrect, smaller is better.

$$MZOE(f, Q) = \frac{|\{q_i \in Q : R(q_i) \neq f(q_i)\}|}{|Q|}$$

2. MAE evaluates the average error in the predictions, smaller is better.

$$MAE(f, Q) = \frac{\sum_{q_i \in Q} |R(q_i) - f(q_i)|}{|Q|}$$

3. The restraint subscale measures specific ED symptoms by looking at specific questions in the questionnaire, smaller is better.

$$RMSE(f, U) = \sqrt{\frac{\sum_{u_i \in U} (R_{RS}(u_i) - f_{RS}(u_i))^2}{|U|}}$$

4. The eating concern subscale measures specific ED symptoms by looking at specific questions in the questionnaire, smaller is better.

$$RMSE(f, U) = \sqrt{\frac{\sum_{u_i \in U} (R_{ECS}(u_i) - f_{ECS}(u_i))^2}{|U|}}$$

5. The shape concern subscale measures specific ED symptoms by looking at specific questions in the questionnaire, smaller is better.

$$RMSE(f, U) = \sqrt{\frac{\sum_{u_i \in U} (R_{SCS}(u_i) - f_{SCS}(u_i))^2}{|U|}}$$

6. The weight concern subscale measures specific ED symptoms by looking at specific questions in the questionnaire, smaller is better.

$$RMSE(f, U) = \sqrt{\frac{\sum_{u_i \in U} (R_{WCS}(u_i) - f_{WCS}(u_i))^2}{|U|}}$$

7. The global ED is a combination of the previous four subscales that attempts to capture the overall ED symptomology in an individual, smaller is better.

$$RMSE(f, U) = \sqrt{\frac{\sum_{u_i \in U} (R_{GED}(u_i) - f_{GED}(u_i))^2}{|U|}}$$

4.1.1.3 Acceptable Error Metric Values

The bounds for which error measures are considered acceptable will be based on a combination of the results other teams achieved in past eRisk competitions and a comparison against certain baselines like guessing an average answer for every prediction. Our model should perform better than previous years and outperform baselines in at least one metric.

4.1.1.4 Tests

1. FRT-INP-1

Control: Automatic

Initial State: No models have been trained by the system yet, input data is available to the system.

Input: Data matching the organizational structure of eRisk competition Data in xml formatting

Output: The model will run to completion and return all required outputs.

Test Case Derivation: The system should run all functions and statements without halting runtime or throwing any errors.

How test will be performed: The appropriate xml data will be fed into the system, the system should complete run without throwing any errors.

2. FRT-INP-2

Control: Automatic

Initial State: No models have been trained by the system yet, input data is available to the system.

Input: Data matching the organizational structure of eRisk competition Data in jsonl formatting

Output: The model will run to completion and return all required outputs.

Test Case Derivation: The system should run all functions and statements without halting runtime or throwing any errors.

How test will be performed: The appropriate jsonl data will be fed into the system, the system should complete run without throwing any errors.

3. FRT-T1-1

Control: Automatic

Initial State: No models have been trained by the task 1 system yet, input data as jsonl files are available

Input: The Task 1 Input data

Output: The system will extract required sentences from the jsonl files and output data that is appropriate to be fed into Task 1.

Test Case Derivation: The system will receive its input data, process then parse according to the Task's input format.

How test will be performed: The tester will run the parser with the given jsonl files as input and compare the parsed data.

4. FRT-T1-2

Control: Automatic

Initial State: No models have been trained by the task 1 system yet, past years input data, training data, and golden truth values are available

Input: The Task 1 Training Data, Input data, and Golden Truth Values from a past year eRisk Competition

Output: The system will output sentence ranking and diagnostic metrics based on the input data.

Test Case Derivation: The system will receive its training data and input data corresponding to a prior year. It's analysis will output RP and NDCG accuracy metrics.

How test will be performed: The system should derive its training data and feature sets based on correlation to 21 depression symptoms of beck

depression index. The outputted metrics will be compared to golden truth data to ensure the accuracy is within the desired bounds and is improving on prior year implementation.

5. FRT-T1-3

Control: Automatic

Initial State: No models have been trained by the task 1 system yet, input data is available

Input: The Task 1 Input data

Output: The system will output sentence ranking and diagnostic metrics based on the input data to a txt file.

Test Case Derivation: The system will make an analysis based off the provided input data and return each prediction to a test file in a specified formatting.

How test will be performed: The system's output txt file will be verified that each entry is on its own line in the format "{symptom_number}, Q0, {sentence-id}, {position_in_ranking}, {score}, {system_name}".

6. FRT-T2-1

Control: Automatic

Initial State: No models have been trained by the task 2 system yet, input data is available

Input: The Task 2 Input data consisting of user posts

Output: The system will output sentence ranking and diagnostic metrics based on the input data to a txt file.

Test Case Derivation: The system will make an analysis based off the provided input data and return each prediction to a test file in a specified formatting.

How test will be performed: The system's output txt file will be verified that each entry is on its own line in the format "{Username} {prediction}".

7. FRT-T2-2

Control: Automatic

Initial State: The task 2 system has received input data on a given individual and made a corresponding prediction.

Input: More data corresponding to the same user analyzed in the input state.

Output: The system will output an updated prediction taking into account all data provided.

Test Case Derivation: The system should create a new prediction for the chosen user taking into account all data provided.

How test will be performed: The system should provide a prediction based off the original data input, after receiving new data, the system should combine posts made by the same user to create a new prediction.

8. FRT-T3-1

Control: Manual

Initial State: The task 3 model has been trained on a section of the training data provided by eRisk

Input: A section of the training data that the model has not yet seen

Output: The system will output predictions to a txt file where each line is of the format "{Username} {prediction for Q1} ... {prediction for Q28}"

Test Case Derivation: The system should output predictions in the format required for the eRisk competition

How test will be performed: A tester will run the system on the input data and verify that an output file is created and that it is of the correct format

9. FRT-T3-2

Control: Manual

Initial State: The task 3 model has been trained on a section of the training data provided by eRisk

Input: A section of the training data that the model has not yet seen

Output: The system will output predictions to a txt file where each prediction should be between 0 and 6

Test Case Derivation: The system should output predictions that conform to the questions on the EDE-Q questionnaire

How test will be performed: A tester will run the system on the input data and verify that none of the predictions in the output file are less than 0 or greater than 6

10. FRT-T3-3

Control: Manual

Initial State: The task 3 model has been trained on a section of the training data provided by eRisk and accuracy measures have been calculated for a baseline strategy where a score of 0 is guessed for every question

Input: A section of the training data that the model has not yet seen

Output: The system will print accuracy metrics for the model to the console where the model scores higher than the baseline strategy on at least one metric

Test Case Derivation: If the system fails to outperform a simple guess 0 strategy on at least one metric it is highly likely it does not have any actual prediction power

How test will be performed: A tester will run the system on the input data and verify that the model outperforms the baseline on at least one metric

11. FRT-T3-4

Control: Manual

Initial State: The task 3 model has been trained on a section of the training data provided by eRisk and accuracy measures have been cal-

culated for a baseline strategy where a score of 6 is guessed for every question

Input: A section of the training data that the model has not yet seen

Output: The system will print accuracy metrics for the model to the console where the model scores higher than the baseline strategy on at least one metric

Test Case Derivation: If the system fails to outperform a simple guess 6 strategy on at least one metric it is highly likely it does not have any actual prediction power

How test will be performed: A tester will run the system on the input data and verify that the model outperforms the baseline on at least one metric

12. FRT-T3-5

Control: Manual

Initial State: The task 3 model has been trained on a section of the training data provided by eRisk and accuracy measures have been calculated for a baseline strategy where the average score for a question based on the training data is guessed

Input: A section of the training data that the model has not yet seen

Output: The system will print accuracy metrics for the model to the console where the model scores higher than the baseline strategy on at least one metric

Test Case Derivation: If the system fails to outperform a simple guess average strategy on at least one metric it is highly likely it does not have any actual prediction power

How test will be performed: A tester will run the system on the input data and verify that the model outperforms the baseline on at least one metric

4.2 Tests for Nonfunctional Requirements

4.2.1 Usability Requirements

1. NFRT-U-1

Type: Dynamic, Manual

Initial State: System is completed and trained

Input/Condition: The system will be ran on a Windows desktop/laptop and macOS desktop/laptop device with the correct environment setup

Output/Result: The system should generate an expected result

How test will be performed: After setting up the environment on to both macOS and Windows machines, the system will be ran on both environment with the same command. Test will pass as long as an expected result is generated from both machines.

4.2.2 Safety and Security Requirements

1. NFRT-SS-1

Type: Dynamic, Manual

Initial State: The system is completed and trained

Input/Condition: The system will be ran on the developer's computer

Output/Result: The generated result should not overuse the processing power of the developer's computer

How test will be performed: CPU, GPU, RAM and live power usage will be monitored while running the software. Usage of all aspects should not increase more than 20 percent.

1. NFRT-SS-2

Type: Automatic, Dynamic

Initial State: The system has been trained and is awaiting data to form predictions on.

Input/Condition: A set of data that the system can predict on.

Output/Result: The resulting predictions, in a form where no sensitive data from the input is present in the output.

Test Case Derivation: Sensitive in this case refers to the post history of the subjects eRisk provides as training and test data. If these posts

can be reconstructed from any part of our system outputs it is possible the identity of the individual could be discovered. This represents an unacceptable breach of privacy and can not happen.

How test will be performed: After the system has produced it's predictions a script will be run that takes all posts in the input data, forms a string out of all consecutive three word triplets (ie. "hello my good friend" forms "hello my good" and "my good friend"), both with and without processing (removal of stopwords, punctuation, etc.), and scans the output to ensure that none of these triples are present.

4.2.3 Legal Requirements

1. NFRT-L-1

Type: Static, Manual

Initial State: The source code and documentation is prepared

Input/Condition: The user reviews the entirety of the source code and related documentation

Output/Result: Copyright licenses, appropriate credits and/or MIT license must attached

How test will be performed: The testers will review the entirety of the project and check to see if appropriate copyright licenses and/or credits are included for resources that require them

Table 2: Traceability Between Functional Test Cases and Functional Requirements, T3FR-4 to T3FR-6

Test IDs	Functional Requirement IDs		
	T3FR-4	T3FR-5	T3FR-6
FRT-INP-1			
FRT-INP-2			
FRT-T1-1			
FRT-T1-2			
FRT-T1-3			
FRT-T2-1			
FRT-T2-2			
FRT-T2-1			
FRT-T3-2			
FRT-T3-3	X		
FRT-T3-4		X	
FRT-T3-5			X

Table 3: Traceability Between Non-Functional Test Cases and Non-Functional Requirements

Test IDs	Non-Functional Requirement IDs
----------	--------------------------------

	GR1	GR2	SR1	SR2
NFRT-U-1	X			
NFRT-SS-1			X	
NFRT-SS-2				X
NFRT-L-1		X		

5 Unit Test Description

[This section should not be filled in until after the MIS (detailed design document) has been completed. —SS]

[Reference your MIS (detailed design document) and explain your overall philosophy for test case selection. —SS]

[To save space and time, it may be an option to provide less detail in this section. For the unit tests you can potentially layout your testing strategy here. That is, you can explain how tests will be selected for each module. For instance, your test building approach could be test cases for each access program, including one test for normal behaviour and as many tests as needed for edge cases. Rather than create the details of the input and output here, you could point to the unit testing code. For this to work, your code needs to be well-documented, with meaningful names for all of the tests. —SS]

5.1 Unit Testing Scope

[What modules are outside of the scope. If there are modules that are developed by someone else, then you would say here if you aren't planning on verifying them. There may also be modules that are part of your software, but have a lower priority for verification than others. If this is the case, explain your rationale for the ranking of module importance. —SS]

5.2 Tests for Functional Requirements

[Most of the verification will be through automated unit testing. If appropriate specific modules can be verified by a non-testing based technique. That can also be documented in this section. —SS]

5.2.1 Module 1

[Include a blurb here to explain why the subsections below cover the module. References to the MIS would be good. You will want tests from a black box perspective and from a white box perspective. Explain to the reader how the tests were selected. —SS]

1. test-id1

Type: [Functional, Dynamic, Manual, Automatic, Static etc. Most will be automatic —SS]

Initial State:

Input:

Output: [The expected result for the given inputs —SS]

Test Case Derivation: [Justify the expected value given in the Output field —SS]

How test will be performed:

2. test-id2

Type: [Functional, Dynamic, Manual, Automatic, Static etc. Most will be automatic —SS]

Initial State:

Input:

Output: [The expected result for the given inputs —SS]

Test Case Derivation: [Justify the expected value given in the Output field —SS]

How test will be performed:

3. ...

5.2.2 Module 2

...

5.3 Tests for Nonfunctional Requirements

[If there is a module that needs to be independently assessed for performance, those test cases can go here. In some projects, planning for nonfunctional tests of units will not be that relevant. —SS]

[These tests may involve collecting performance data from previously mentioned functional tests. —SS]

5.3.1 Module ?

1. test-id1

Type: [Functional, Dynamic, Manual, Automatic, Static etc. Most will be automatic —SS]

Initial State:

Input/Condition:

Output/Result:

How test will be performed:

2. test-id2

Type: Functional, Dynamic, Manual, Static etc.

Initial State:

Input:

Output:

How test will be performed:

Type: Automatic, Dynamic

Initial State: The system has been trained and is awaiting data to form predictions on.

Input/Condition: A set of data that the system can predict on.

Output/Result: The resulting predictions, in a form where no sensitive data from the input is present in the output.

Test Case Derivation: Sensitive in this case refers to the post history of the subjects eRisk provides as training and test data. If these posts can be reconstructed from any part of our system outputs it is possible the identity of the individual could be discovered. This represents an unacceptable breach of privacy and can not happen.

How test will be performed: After the system has produced it's predictions a script will be run that takes all posts in the input data, forms a string out of all consecutive three word triplets (ie. "hello my good friend" forms "hello my good" and "my good friend"), both with

and without processing (removal of stopwords, punctuation, etc.), and scans the output to ensure that none of these triples are present.

5.3.2 Module ?

...

5.4 Traceability Between Test Cases and Modules

[Provide evidence that all of the modules have been considered. —SS]

6 Appendix

6.1 Symbolic Parameters

None.

6.2 Reflection

1. What knowledge and skills will the team collectively need to acquire to successfully complete the verification and validation of your project? Examples of possible knowledge and skills include dynamic testing knowledge, static testing knowledge, specific tool usage etc. You should look to identify at least one item for each team member.

The following knowledge and skills will need to be acquired by the team to successfully complete the verification and validation for the project.

1. Github Actions
 2. Pytest
 3. Mental Health Diagnostic Basics (For creating expected values)
 4. Python script for parsing and checking files
 5. System performance diagnostic skills
2. For each of the knowledge areas and skills identified in the previous question, what are at least two approaches to acquiring the knowledge or mastering the skill? Of the identified approaches, which will each team member pursue, and why did they make this choice?
 1. I will learn more about GitHub actions because using GitHub actions can let us test iteratively every time we push to GitHub repository without interrupting our workflow. One way to acquire this skill is to use the github actions documentation page in order to help with learning the coding skills required to add it to the coding pipeline. Another way to acquire this knowledge is using YouTube videos such as the “GitHub Actions Tutorial” by “TechWorld with Nana” which includes all the necessary information needed for the scope of github actions testing for our project.
 2. Jessica: I will learn more about train/test splits, different distributions of scores in populations and how to control for this in testing.

3. Ben: In order to further enhance my understanding and as a result my skillset of the project's trajectory, I wish to learn more about the specificities of evaluation metrics used for Natural Language Processors. Testing and tuning are essential aspects of creating an artificial intelligence system, but without accurate methods of tracking metrics there is no way to measure the effectiveness of the model. The first strategy presented to learn more would be to look at corresponding scholarly articles endorsed by the creators of each corresponding AI, for example, see how effectiveness is measured by the people who have created that methodology. A secondary approach would be to use more baseline metrics discussed in prior papers from the Montreal Team, as well as studies done by the eRisk competition itself, and use those as a standard to compare other metrics off of.

4. Matthew: In order to learn more and develop my skills regarding python scripts for parsing and checking files I took a couple of key steps. I think these skills are very important because we are given very large files to work with for this project and need to parse these files effectively including splitting them into separate parts and labeling them correctly. I love big data so this was right up my alley so I wanted to take the initiative into looking more into these skills. The largest step I took to deepen my understanding of this skill was first to complete the "Parsing Data in Python: A Tutorial for Beginners" from Pierian Training. This was very useful to refresh my memory and help me get in the right mental state for developing said scripts for parsing files. It also helped refresh my mind of the basics. Once that was done, to move onto slightly more complicated parsing I followed and read "Parsing text with Python" by vipinajayakumar. This was extremely helpful and more focused on more complex parsing then just the basics. I then moved on to practicing parsing the input data we had for testing in different ways. I practiced different manipulation techniques I learned from the two articles above and played with the output to get a deeper feel.

5. Franklin: I will learn more about Pytest, one method is to research the official documentation provided by Pytest itself. This will include installation instructions, usage examples, and various detailed explanations of versatile features and functionalities of Pytest. Another method to learn about Pytest is online courses on platforms like

Udemy and Coursera. Courses such as Python Testing with Pytest by Brian Okken will provide everything I need to know to test systems extensively as they provide video lectures, learning paths, exercise and tests to reinforce learning.