Evaluation Document for Explainable AI Project

1. Introduction

Building on your previous semester projects—which involved a Machine Learning challenge and a Deep Learning challenge—this module focuses on **Explainable AI (XAI)**. While you have already tested multiple algorithms to achieve the best performance on leaderboard tasks, the next step is to **interpret and explain** these models' decisions. This document outlines the requirements, deliverables, and grading criteria for your XAI project.

2. Link to Previous Projects

1. Machine Learning Challenge (First Semester):

- You experimented with a range of machine learning models.
- You aimed to produce a strong leaderboard ranking based on performance metrics.
- You delivered a short report and code illustrating your approach and findings.

2. Deep Learning Challenge (First Semester):

- You explored neural network architectures for complex data (images, text, or tabular).
- You optimized hyperparameters for accuracy and efficiency.
- You documented your process and produced a working prototype.

These projects formed a solid foundation in model-building, hyperparameter tuning, and empirical evaluation. Now, you will **extend** these efforts by applying Explainable AI methods to better understand how your algorithms arrive at predictions.

3. Project Scope: Explainable Al

For this semester's project, you will:

1. Choose one (or both) of your previous projects (Machine Learning or Deep Learning).

2. Integrate Explainable Al methods to provide:

- **Global explanations**: overall feature importance and how each feature influences the model's performance.
- Local explanations: interpretability on a single-instance basis, especially focusing on outliers and wrong predictions.
- 3. Provide clear documentation (report), source code, and a presentation.

4. Deliverables

1. Code

- Clean, well-structured, and commented.
- Must integrate XAI libraries or custom interpretability functions (e.g., using methods such as SHAP, LIME, Grad-CAM for deep learning, or permutation-based importance for tree-based models).
- Include any scripts or notebooks used for pre-processing, model training, evaluation, and interpretability.

2. Report

- **Introduction**: Brief recap of the initial project (dataset, goal, main algorithms used).
- **Methodology**: Explain the XAI techniques you chose and why (SHAP, LIME, feature importance, partial dependence plots, etc.).
- Results: Present both global and local explanations. Highlight the most influential features and analyze examples of outliers or misclassifications.
- Discussion: Reflect on the results, including any insights gained about model behavior, data biases, or potential improvements.
- Conclusion: Summarize key takeaways from employing explainability techniques.

3. Presentation

Duration: 20 minutes per group + 5 minutes of Q&A.

- o Groups: **2 to 4** members.
- Focus:
 - Overview of your initial model (or models).
 - Methods used for XAI and reasons for choosing them.
 - Demonstration of real examples (particularly wrong predictions and how the XAI method explains them).
 - Lessons learned & next steps.

5. Timeline

- Next Practice Sessions (29/03 and 01/04):
 - Use this time to polish your code and solidify your presentation flow.
 - Each group will present for 20 minutes, followed by 5 minutes of Q&A.
- Final Deliverables:
 - Code and Report: Submit by the end of the second session (01/04).
 - **Presentation**: Presented live during the practical sessions.

6. Grading Criteria

Component	Weight	Description
Code Quality	30%	Readability (comments, structure)Implementation correctnessAppropriate use of XAI librariesEfficient, reproducible results
Report	30%	 Clarity of explanation Depth of analysis Proper justification for chosen XAI methods Quality of examples and interpretation Organized and coherent structure

Presentation 20% - Organization and clarity

- Ability to communicate technical aspects to peers

- Quality of visuals/demonstrations

- Handling of questions and audience engagement

Innovation & Insight 20% - Novelty or creativity in approach

- Insightful discussion of results

- Thorough exploration of outliers and misclassifications

- Reflection on limitations and future improvements

Total: 100%

7. Additional Recommendations

• **Emphasize Clarity**: Use plain language whenever possible, especially when explaining complex models and interpretability methods.

- Demonstrate Practical Value: Show real-world or business implications of your explanations (e.g., how understanding predictions can lead to better user trust or model refinement).
- **Compare Multiple Methods**: If possible, illustrate more than one XAI technique and discuss their strengths/weaknesses.
- **Be Consistent**: Use consistent variable naming and data labeling across code, report, and presentation.
- **Rehearse**: Make sure the 20-minute presentation is well-timed and that you anticipate potential questions on methodology and results.

8. Conclusion

This evaluation aims to foster a deeper understanding of your Machine Learning or Deep Learning models by leveraging Explainable AI. By showing **how** and **why** a model makes certain predictions, you will gain insights into data quality, model reliability, and potential areas for improvement. This is a crucial step in deploying AI responsibly and effectively.

We look forward to your final code, insightful reports, and engaging presentations!

Good luck with your projects and presentations!