

Development of an employee-performance evaluation scheme using Ensemble of Machine learning techniques for Higher Educational Institution's.

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ABSTRACT

In an era where organizational success hinges on the precise evaluation and management of employee performance, traditional performance evaluation methods have proven insufficient. These conventional approaches, often characterized by periodic reviews and subjective assessments, fail to provide the timely, accurate, and comprehensive insights necessary for effective performance management. This research aims to address these limitations by developing an advanced employee performance evaluation scheme leveraging Machine Learning (ML) and Deep Learning (DL) technologies, specifically tailored for higher educational institutions.

The objectives of this study are fourfold. Firstly, it conducts a comprehensive analysis of existing ML and DL algorithms to compare their strengths, weaknesses, and suitability for ranking employees and identifying Potential performers. This analysis identifies key characteristics that contribute to effective performance classification. Secondly, it implements a hybrid feature selection algorithm that integrates multiple approaches to identify the most relevant performance indicators for higher educational institutions. Thirdly, the research develops an ensemble classification scheme using various DL classifiers to accurately rank and categorize employees based on their performance metrics. Lastly, it fine-tunes the classification model to precisely identify low-performing employees, utilizing the developed AI-based scheme to flag individuals requiring additional support or intervention sets.

The expected outcomes of this research include a detailed comparative analysis of ML and DL algorithms, a validated hybrid feature selection

algorithm, a robust ensemble classification model, and an optimized scheme for accurate low-Potential performers identification. These outcomes will provide higher educational institutions with a powerful tool for enhancing their performance evaluation processes, supporting employee development, and achieving higher overall performance levels.

By integrating cutting-edge ML and DL technologies, this research endeavors to revolutionize employee performance evaluation, addressing the inadequacies of traditional methods and contributing significantly to the field of performance management. The developed evaluation scheme aims to offer practical solutions for higher educational institutions, setting a new standard in employee performance assessment and management.

I. INTRODUCTION

In the contemporary landscape of organizational management, the evaluation of employee performance is a pivotal activity that significantly influences operational efficiency and overall productivity. Traditional methods of performance evaluation, often characterized by subjective assessments and periodic reviews, have proven inadequate in providing timely, accurate, and comprehensive insights into employee performance. This inadequacy is particularly pronounced in higher educational institutions, where diverse roles and responsibilities demand a more nuanced and continuous evaluation approach.

The advent of Machine Learning (ML) and Deep Learning (DL) technologies offers a transformative potential for performance evaluation systems. These technologies enable data-driven,

objective, and real-time assessments, which are crucial for identifying both Nice performers and employees who may require additional support. By leveraging the analytical power of ML and DL algorithms, organizations can move beyond conventional evaluation methods to more sophisticated, accurate, and actionable performance insights.

This research aims to address the limitations of traditional performance evaluation methods by developing an advanced employee performance evaluation scheme using an ensemble of ML and DL classifiers. The primary objectives of this research include conducting a comprehensive analysis of existing ML and DL algorithms, implementing a hybrid feature selection algorithm tailored to higher educational institutions, developing a robust ensemble classification scheme, and fine-tuning the model to accurately identify low-performing employees.

Firstly, the study will conduct a thorough analysis of existing ML and DL algorithms to compare their strengths, weaknesses, and suitability for ranking employees and identifying Potential performers. This analysis will identify key characteristics that contribute to effective performance classification, providing a foundational understanding for the development of the new scheme.

Secondly, a hybrid feature selection algorithm will be implemented to identify the most relevant performance indicators. This algorithm will combine multiple feature selection approaches to ensure that the selected features accurately reflect the multifaceted nature of employee performance in higher educational institutions.

Thirdly, the research will develop an employee performance classification scheme using an ensemble of DL classifiers. This ensemble approach will leverage the strengths of various classifiers to enhance the accuracy and reliability of the performance evaluation, providing a robust tool for ranking and categorizing employees.

Finally, the model will be fine-tuned to focus on the precise identification of low-performing employees. By optimizing the classification model, the developed AI-based scheme will effectively identify and flag individuals who require additional support or intervention, thereby enabling timely and targeted performance management strategies.

The anticipated outcomes of this research include a detailed comparative analysis of ML and DL algorithms, a validated hybrid feature selection algorithm, a robust ensemble classification model, and a fine-tuned scheme for accurate low-performer identification. These outcomes will provide higher educational institutions with a powerful tool for enhancing their performance evaluation processes, supporting employee development, and achieving higher overall performance levels.

In conclusion, this research endeavors to revolutionize employee performance evaluation through the integration of cutting-edge ML and DL technologies. By addressing the limitations of traditional methods and developing a sophisticated evaluation scheme, this study aims to contribute significantly to the field of performance management and provide practical solutions for higher educational institutions& scenarios.

II. LITERATURE REVIEW

Reference	Method Used	Findings	Results (in Numerical Values if Available)	Limitations
[1]	Blockchain Assisted Decentralized Employee Performance Assessment (BDEPA) System	Proposed a system that uses blockchain for secure, decentralized employee performance assessment	Higher security using ECC, reduced energy costs with Quantum blockchain, verified employee authenticity with AHMFA, secured data access with EDH	Limited discussion on integration with existing systems, scalability issues in large organizations
[2]	Multicriteria evaluation of publication performance using ARAMIS method	Evaluates research organization performance by department	N/A	Does not address the impact on individual employee performance evaluation
[3]	Systematic	Explores the	18 articles on	Limited by scope of

	literature review on digital competence	relationship between digital autonomy and innovative work behavior	relationship topics, 12 on impact topics	articles reviewed, does not provide direct numerical performance metrics
[4]	Age-differentiated leadership (ADL) model	Analyzed impact of ADL on employee health	Improved health outcomes in 947 employees	Specific to age-differentiated leadership, not generalizable to all leadership styles
[5]	Performance evaluation framework using machine learning and big data	Developed a performance evaluation system for intelligent manufacturing enterprises	Uses ridge regression, lasso regression, elastic network regression	Limited to intelligent manufacturing, may not apply to other sectors
[6]	Empirical study on employee-customer familiarity	Investigates the impact of employee-customer familiarity on purchase intentions	Positive correlation found	Focuses on home bedding industry, results may not generalize to other industries
[7]	Conceptual model using PLS-SEM	Explores CSR impact on financial performance of IECES	Positive impact of CSR on financial performance, mediating role of employee loyalty and corporate reputation	Insignificant role of customer satisfaction as mediator, limited to IECES
[8]	Fuzzy neural network model for performance evaluation	Combines financial and non-financial indicators for state-owned enterprises	Shows improved evaluation accuracy	Limited application to state-owned enterprises, may not generalize to private sector
[9]	Analysis of social relationships on intra-organizational social media	Examines the impact on idea generation	Proactive creativity and group identification improve idea generation quality	Focuses on telecommunications sector, results may not generalize to other sectors
[10]	Cross-sectional survey on SE policies and task performance	Evaluates the impact of SE policies on task performance	SE policies improve performance via health and productive capabilities, moderated by UTAUT predictors	Small sample size of 88 employees, specific to cobots
[11]	PLS-SEM analysis on CSA and manufacturing performance	Investigates the role of CSA in enhancing manufacturing performance	Positive impact of CSA on performance	Focuses on small and medium-sized enterprises in Kenya, results may not generalize globally
[12]	Test person study on learning evaluation measures	Compares different learning evaluation measures in manual assembly	Improves learning process design	Limited to manual assembly, may not apply to automated processes
[13]	Hybrid storage blockchain method for business	Proposes a query efficiency enhancement	Reduces storage overhead, improves query performance	Focuses on business environment evaluation, limited

	environment evaluation	method		application to other fields
[14]	AHP method for online job portal selection	Identifies best job portal for recruitment	Significant results using AHP	Small sample size of 15 candidates, specific to two-wheeler automotive companies
[15]	Case study on capability management approaches	Demonstrates and evaluates KYKLOS and Compass approaches	KYKLOS evaluated positively by modeling experts, Compass by business experts	Initial difficulty in ease of use for business experts, ongoing research required
[16]	Integrated multi-criteria method using Fuzzy AHP and Fuzzy TOPSIS	Evaluates sustainable alternatives in food sector	"Cr1 – Quality" most prioritized, "A1 - correct destination and disposal of burnt cooking oil" most significant	Limited to restaurant sector, may not apply to other industries
[17]	Ontology-based semantic knowledge framework for CGDP	Unveils relationship among CGDP attributes	Correlation analysis validates framework	Limited empirical validation, focuses on financial aspects
[18]	BPR model integrating CSFs and BPs' performance	Reengineers business processes for performance improvement	Improved process time, cycle time, quality, and cost	Focuses on Egyptian tax authority case study, results may not generalize
[19]	Feature selection and meta-heuristic algorithms for employee turnover	Identifies factors affecting employee turnover	Mutual Information algorithm shows better prediction	Limited to one company's HR dataset, may not apply broadly
[20]	Survey-based inquiry using BWM for FMS performance	Evaluates FMS performance in German manufacturing	Quality, flexibility, and productivity prioritized	Focuses on German manufacturing sector, limited generalizability
[21]	Expert interviews on RM applicability in BPS	Surveys RM benefits in BPS	Experts see great benefit but require clearer examples	Limited transfer and clarity for BPS experts
[22]	Hybrid framework for RMS performance metrics	Develops framework for RMS evaluation	Lead time, reconfiguration time, product flexibility prioritized	Case analysis limited to manufacturing sector, may not generalize
[23]	ML and ANN techniques for nurse absenteeism prediction	Predicts nurse absenteeism at KAUH	82% accuracy with ANN	Focuses on healthcare sector, specific to KAUH
[24]	Fuzzy AHP technique for ERP software selection	Evaluates criteria for ERP software selection in SMEs	Success within company most significant	Focuses on SMEs, specific to ERP software
[25]	PLS-SEM and ML analysis on technostress and academic performance	Examines impact of technostress on student burnout	Techno-overload, techno-uncertainty, techno-invasion, techno-complexity, and techno-insecurity	Specific to higher education, limited application to business environments

			predict burnout	
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Research Gaps in Employee-Performance Evaluation Scheme Using Ensemble of Machine Learning

Research Gap	Explanation	Paper(s) Addressing Gap
Integration with Existing Systems	Limited discussion on how proposed models can be integrated with current organizational systems and workflows.	[1], [4]
Scalability in Large Organizations	Issues related to the scalability of proposed models and systems when applied to larger organizational contexts.	[1], [7]
Generalizability Across Sectors	Most studies focus on specific industries or sectors, limiting the generalizability of their findings.	[5], [6], [8], [9], [11], [16], [18], [20], [23]
Real-Time Data Processing	Lack of focus on the ability of models to process and evaluate employee performance in real-time.	[2], [12], [14]
Comprehensive Metric Inclusion	Need for more comprehensive inclusion of both financial and non-financial performance metrics.	[8], [22], [24]
Longitudinal Studies	Absence of long-term studies to evaluate the sustained impact of employee performance evaluation models.	[4], [5], [19]
Cultural and Geographical Diversity	Limited exploration of how cultural and geographical differences impact the effectiveness of performance evaluation models.	[7], [11], [23]
Integration of Employee Feedback	Insufficient integration of direct employee feedback into performance evaluation models.	[6], [12], [21]
Cross-Functional Team Performance	Lack of studies addressing performance evaluation in cross-functional or interdisciplinary teams.	[9], [10], [15]
Adaptability to Technological Changes	Need for models that can adapt to rapid technological advancements and changes in workplace tools.	[10], [19], [20], [25]

This table outlines the key research gaps identified from the reviewed papers, providing an explanation for each gap and listing the specific papers that address these gaps. This format helps to clearly identify areas needing further research and highlights the contributions of existing studies& scenarios.

Research Gap: -

While numerous studies have explored the use of machine learning techniques for employee performance evaluation across various sectors, there is a noticeable lack of research focused specifically on higher educational institutions. Most existing studies predominantly concentrate on traditional sectors like finance, healthcare, and retail, where the nature of performance metrics and evaluation criteria significantly differs from those in the educational domain.

Furthermore, the current research that applies machine learning in educational settings primarily addresses student performance,

prediction of dropout rates, or administrative decision-making, rather than faculty and staff performance evaluation section wise. The unique work environment of higher educational institutions, characterized by diverse roles such as teaching, research, and community service, requires a more tailored approach to performance evaluation.

Additionally, while ensemble machine learning techniques have been recognized for their robustness and accuracy in performance prediction, there is limited application of these advanced methods in the context of employee evaluation in educational institutions. Existing studies often rely on single machine learning models, which may not fully capture the multifaceted nature of academic performance metrics.

Therefore, this study aims to develop a comprehensive employee-performance evaluation scheme utilizing an ensemble of machine learning techniques specifically tailored for higher educational institutions. By doing so, this research

intends to address the gap in applying advanced machine learning models with the combined inputs from the reporting officer and the reviewing officer to accurately and fairly evaluate the performance of staff, considering the unique attributes and challenges of the Higher educational Institutions.

Objective of the StudyProcess

1. To Conduct a comprehensive analysis of existing Machine Learning and Deep Learning algorithms for employee performance evaluation, comparing their strengths, weaknesses, and suitability for ranking and identifying Potential performers, Desirable Performers and Nice performers.
2. To develop hybrid feature selection algorithm that combines multiple best approaches to identify the most relevant performance indicators specially tailored for higher educational institutions.
3. To Evaluate Comparative efficacy of newly developed performance indicators versus the best available one.

Aims & Problem Statement

Aims

1. **Comprehensive Analysis of Machine Learning and Deep Learning Algorithms for Employee Performance Evaluation:**
 - Conduct an extensive review of existing Machine Learning (ML) and Deep Learning (DL) algorithms specifically applied to employee performance evaluation.
 - Compare the strengths, weaknesses, and suitability of these algorithms for tasks such as ranking employees and identifying Potential performers.
 - Identify key characteristics that contribute to effective performance classification.
2. **Hybrid Feature Selection for Higher Educational Institutions:**
 - Implement a hybrid feature selection algorithm that integrates multiple approaches to pinpoint the most relevant performance indicators.
 - Focus on higher educational institutions to tailor the algorithm to the unique needs of this sector.
3. **Development of an Employee-Performance Classification Scheme Using Ensemble of Deep Learning Classifiers:**
 - Develop a robust classification scheme utilizing an ensemble of Deep Learning classifiers.

- Ensure the scheme accurately ranks and categorizes employees based on their performance metrics, offering a nuanced understanding of employee performance levels.

4. **Evaluate Comparative efficacy of newly developed performance indicators versus the best available one.:**

- Fine-tune the classification model to enhance its accuracy in identifying Potential performers employees.
- Utilize the developed AI-based scheme to effectively flag individuals who require additional support or intervention due to their performance levels.

Problem Statement

In today's competitive and dynamic work environments, organizations increasingly rely on sophisticated methods to evaluate employee performance. Traditional performance evaluation methods, which often rely on subjective assessments and periodic reviews, fall short in providing a timely and accurate understanding of employee performance. This gap is particularly pronounced in higher educational institutions where diverse roles and multifaceted performance metrics complicate the evaluation process.

The advent of Machine Learning and Deep Learning algorithms has shown promise in revolutionizing performance evaluation by offering data-driven, objective, and continuous assessment capabilities. However, several challenges remain:

1. **Algorithm Suitability and Effectiveness:** Existing ML and DL algorithms exhibit varying strengths and weaknesses when applied to employee performance evaluation. A comprehensive analysis is needed to identify which algorithms are most effective for ranking employees and detecting Potential performers.
2. **Relevant Performance Indicators:** Identifying the most pertinent performance indicators is critical, especially in the context of higher educational institutions where roles and responsibilities are diverse. A hybrid feature selection approach can enhance the accuracy and relevance of the evaluation process by combining multiple feature selection techniques.
3. **Accurate Classification and Ranking:** There is a need to develop an employee-performance classification scheme that leverages the strengths of multiple Deep Learning

classifiers. Such an ensemble approach can improve the accuracy of performance ranking and categorization.

4. **Identification of Potential Performers:** Precisely identifying low-performing employees is essential for timely intervention and support. Fine-tuning the classification model to ensure it accurately flags individuals requiring additional assistance can help organizations maintain high performance standards and support employee development.

This research aims to address these challenges by conducting a thorough analysis of ML and DL algorithms, developing a hybrid feature selection algorithm tailored to higher educational institutions, and creating an ensemble classification scheme for accurate performance evaluation. The ultimate goal is to provide organizations with a powerful tool to identify and support low-performing employees, thereby enhancing overall organizational performance and employee development.

Research Question: -

1. Whether the incorporation of Hybrid feature selection would be more efficacious in ensemble of deep learning classifiers?
2. What combination of feature selection approaches yield the most effective results for employee performance evaluation?
3. What is the efficacy of the developed AI Based classification scheme in identifying employee who require additional support or intervention?

Proposed Methodology

The proposed methodology aims to systematically address the objectives outlined for developing an effective employee performance evaluation scheme using Machine Learning (ML) and Deep Learning (DL) techniques. The methodology is structured into four main phases: Comprehensive Algorithm Analysis, Hybrid Feature Selection Implementation, Ensemble Classification Scheme Development, and Model Fine-tuning for Potential-Performer Identification.

Phase 1: Comprehensive Algorithm Analysis

1. **Literature Review:**
 - Conduct an extensive review of existing literature on ML and DL algorithms applied to employee performance evaluation.
 - Identify and categorize algorithms based on their application domains, performance metrics, and evaluation criteria.

2. **Algorithm Comparison:**

- Select a representative set of ML and DL algorithms including, but not limited to, decision trees, random forests, support vector machines, neural networks, and deep learning architectures like CNNs and RNNs.
- Compare the algorithms based on their strengths, weaknesses, and suitability for ranking employees and identifying Potential performers.
- Evaluate each algorithm's performance using standard metrics such as accuracy, precision, recall, F1-score, and computational efficiency.

3. **Key Characteristic Identification:**

- Identify key characteristics that contribute to effective performance classification, such as feature importance, model interpretability, and adaptability to different datasets.

Phase 2: Hybrid Feature Selection Implementation

1. **Feature Collection:**

- Collect a comprehensive set of performance indicators relevant to higher educational institutions. These indicators may include teaching effectiveness, research output, student feedback, peer reviews, and administrative contributions.

2. **Hybrid Feature Selection Algorithm Development:**

- Develop a hybrid feature selection algorithm that combines multiple feature selection techniques such as filter methods (e.g., Chi-Square, ANOVA), wrapper methods (e.g., Recursive Feature Elimination), and embedded methods (e.g., LASSO, Ridge Regression).
- Integrate these techniques to identify the most relevant performance indicators, ensuring that the selected features capture the multifaceted nature of employee performance in higher educational institutions.

3. **Validation:**

- Validate the selected features through cross-validation techniques and expert consultation to ensure their relevance and effectiveness in performance evaluation.

Phase 3: Ensemble Classification Scheme Development

1. **Ensemble Model Design:**

- Design an ensemble classification scheme that leverages the strengths of multiple DL

- classifiers, such as combining CNNs for structured data and RNNs for sequential data.
- Utilize techniques like bagging, boosting, and stacking to enhance the performance and robustness of the ensemble model.
2. **Model Training:**
- Train the ensemble model on a diverse and comprehensive dataset representing various performance metrics.
 - Implement cross-validation to ensure the model's generalizability and prevent overfitting.
3. **Performance Evaluation:**
- Evaluate the ensemble model using a variety of performance metrics, including accuracy, precision, recall, F1-score, and ROC-AUC.
 - Compare the ensemble model's performance against individual classifiers to demonstrate its superiority in ranking and categorizing employees.

Phase 4: Model Fine-tuning for Potential Performers Identification

1. **Precision Tuning:**
- Fine-tune the ensemble classification model to enhance its precision in identifying low-performing employees.
 - Adjust model parameters and hyperparameters to optimize its performance specifically for low-performer identification.
2. **Threshold Setting:**
- Establish appropriate decision thresholds to balance sensitivity and specificity, ensuring accurate identification of Potential performers without excessive false positives.
3. **Validation and Testing:**
- Validate the fine-tuned model on a separate validation dataset and perform rigorous testing to confirm its reliability and accuracy in real-world scenarios.
 - Utilize techniques like confusion matrices and error analysis to refine the model further.
4. **Implementation and Feedback Loop:**
- Implement the AI-based performance evaluation scheme in a real-world higher educational institution setting.
 - Establish a feedback loop where identified low-performing employees receive targeted interventions and support.

- Continuously monitor and update the model based on feedback and changing performance dynamics.

The proposed methodology offers a structured and comprehensive approach to developing an advanced employee performance evaluation scheme for this process. By combining a thorough analysis of ML and DL algorithms, implementing a robust hybrid feature selection process, developing a powerful ensemble classification scheme, and fine-tuning the model for precise low-performer identification, this methodology aims to provide a reliable and effective tool for enhancing employee performance evaluation in higher educational institutions & scenarios.

Expected Outcome

The implementation of the proposed methodology for employee performance evaluation using an ensemble of Machine Learning (ML) and Deep Learning (DL) techniques is expected to yield several significant outcomes. These outcomes will address the objectives outlined and contribute to the advancement of performance evaluation systems, particularly in higher educational institutions.

1. Comprehensive Understanding of Algorithm Effectiveness

- **Detailed Comparative Analysis:** The comprehensive analysis of existing ML and DL algorithms will provide a clear understanding of their strengths, weaknesses, and suitability for employee performance evaluation. This will enable organizations to make informed decisions about which algorithms to implement based on their specific needs and contexts.
- **Key Characteristics Identification:** The identification of key characteristics that contribute to effective performance classification will highlight the essential features and functionalities that successful algorithms should possess. This insight will guide future research and development in the field.

2. Enhanced Feature Selection Process

- **Hybrid Feature Selection Algorithm:** The development and implementation of a hybrid feature selection algorithm will result in a more accurate and relevant identification of performance indicators. This will ensure that

the most critical factors influencing employee performance are considered, leading to more precise and meaningful evaluations.

- **Relevance to Higher Educational Institutions:** By tailoring the feature selection process to the unique requirements of higher educational institutions, the algorithm will produce performance metrics that are directly applicable and highly relevant to this sector, enhancing the overall evaluation process.
3. **Advanced Employee-Performance Classification Scheme**
- **Robust Ensemble Model:** The creation of an ensemble classification scheme using multiple DL classifiers will lead to a robust and reliable model capable of accurately ranking and categorizing employees based on their performance metrics. This model will leverage the strengths of various classifiers to provide a comprehensive assessment of employee performance.
 - **Improved Accuracy and Reliability:** The ensemble approach is expected to outperform individual classifiers, resulting in higher accuracy and reliability in performance evaluation. This will allow organizations to trust the system's outputs and make better-informed decisions regarding employee performance management.
4. **Precise Identification of Potential Performers Employees**
- **Fine-Tuned Classification Model:** The fine-tuning of the classification model will ensure that low-performing employees are accurately identified. This precision will enable targeted interventions and support for these employees, helping them improve their performance and contributing to overall organizational success.
 - **Effective Intervention Strategies:** By accurately flagging individuals requiring additional support or intervention, the developed AI-based scheme will facilitate the implementation of effective strategies to address performance issues. This proactive approach will help in mitigating performance-related problems before they escalate.

5. Practical Implementation and Continuous Improvement

- **Real-World Application:** The implementation of the AI-based performance evaluation scheme in a real-world higher educational institution setting will demonstrate its practical applicability and effectiveness. This will provide valuable insights into the system's performance in actual use cases.
- **Feedback Loop and Model Refinement:** Establishing a feedback loop will ensure continuous monitoring and updating of the model based on real-world data and feedback. This iterative process will lead to ongoing improvements in the system, making it increasingly effective over time.

6. Contribution to Research and Practice

- **Enhanced Performance Evaluation Practices:** The outcomes of this research will contribute to the advancement of performance evaluation practices by providing a sophisticated tool that integrates cutting-edge ML and DL techniques. This will set a new standard for evaluating employee performance in higher educational institutions and potentially other sectors.
- **Guidance for Future Research:** The findings and insights gained from this research will offer valuable guidance for future studies, helping to address existing gaps and explore new avenues in the field of employee performance evaluation using AI technologies.

The expected outcomes of this research encompass both theoretical advancements and practical benefits. By providing a comprehensive analysis of existing algorithms, developing a hybrid feature selection process, creating an advanced ensemble classification scheme, and fine-tuning the model for precise identification of Potential performers, this research aims to revolutionize employee performance evaluation. The ultimate goal is to equip organizations, particularly higher educational institutions, with a powerful tool to enhance their performance evaluation processes, support employee development, and achieve higher overall performance levels.

Work Plan Schedule

Month	Work to be Done	Work Details
1-6	Literature Review and Algorithm Selection	Conduct an extensive literature review on existing ML and DL algorithms used for employee performance evaluation. Select a representative set of algorithms for detailed analysis.
7-12	Algorithm Comparison and Initial Testing	Compare selected algorithms based on strengths, weaknesses, and suitability for ranking employees and identifying Potential performers. Perform initial testing using standard performance metrics.
13-18	Key Characteristic Identification and Feature Collection	Identify key characteristics of effective performance classification algorithms. Collect a comprehensive set of performance indicators relevant to higher educational institutions.
19-24	Development of Hybrid Feature Selection Algorithm	Develop a hybrid feature selection algorithm that integrates multiple feature selection techniques. Validate the selected features through cross-validation and expert consultation.
25-30	Design and Development of Ensemble Model	Design an ensemble classification scheme using multiple DL classifiers. Begin development of the ensemble model.
31-36	Training and Cross-Validation of Ensemble Model	Train the ensemble model on a diverse dataset. Implement cross-validation to ensure model generalizability and prevent overfitting.
37-42	Performance Evaluation and Fine-Tuning	Evaluate the ensemble model using various performance metrics. Fine-tune the model to enhance precision in identifying low-performing employees.
43-48	Threshold Setting and Model Validation	Establish decision thresholds to balance sensitivity and specificity. Validate the fine-tuned model on a separate validation dataset and perform rigorous testing.
49-54	Implementation in Real-World Setting	Implement the AI-based performance evaluation scheme in a higher educational institution. Monitor performance and gather feedback for continuous improvement.
55-60	Establishment of Feedback Loop and Continuous Monitoring	Establish a feedback loop for continuous monitoring and updating of the model based on real-world data and feedback.
61-66	Comprehensive Analysis and Reporting	Conduct a comprehensive analysis of the implementation results. Prepare detailed reports and documentation of the findings.
67-72	Dissemination and Future Research Scopes	Disseminate findings through publications and conferences. Identify future research directions based on the outcomes and feedback received.

This work plan provides a structured timeline over three years, with clear milestones and specific tasks outlined in six-month intervals. Each phase builds on the previous one to ensure a systematic and comprehensive approach to developing and implementing the employee performance evaluation scheme using an ensemble of Machine Learning and Deep Learning techniques.

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