

STUDENT PLACEMENT PREDICTION USING MACHINE LEARNING

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ABSTRACT

An innovative application that makes use regarding machine learning (ML) algorithms to give job searchers and employers with a smooth and effective platform is the online job portal and placement prediction system. By matching applicant profiles with relevant job advertisements, this method aims to close the gap between job seekers and potential employers and streamline the hiring process. The Online Job Portal and the Placement Prediction module are the two key parts of the system. employment seekers may establish profiles, submit their resumes, and search for suitable employment openings using the Online Job Portal's user-friendly interface Effective job posting, candidate discovery, and recruiting process management are all possible for employers. Utilizing cutting-edge artificial intelligence methods, placement prediction module.

Keywords: Machine Learning Algorithms, MLP, NLP, Deep Learning, Decision Tree.

I. INTRODUCTION

As part of developing a placement administration system at the college level, we want to create a placement calculator that predicts the likelihood that students will be placed. Additionally, it will assist the institution's placement office and professors in giving students the attention they need to succeed during the course. In order to forecast placement, machine learning is being used. The student's academic achievement is the sole element considered by the placement prediction model currently in use, allowing for the prediction of no matter if the student will be placed. We cannot base student placement just on academic performance since some pupils have strong technical, communicative, and aptitude abilities despite having weak academic performance. A creative program called the Online Job Portal and Placement Prediction system makes use of machine learning (ML) algorithms to give job searchers and businesses a seamless and effective platform. By matching applicant profiles with acceptable job advertisements, this method aims to close the gap between job seekers and potential employers and streamline the hiring process. The Online Job Portal and the Placement Prediction module are the two key parts inside the system. The Online Job Portal offers job seekers a user-friendly interface so they can build profiles, submit their resumes, and look Tasks that need a appropriate for them. Effective job posting, candidate discovery, and recruiting process management are all possible for employers.

II. METHODOLOGY

For predicting student placement, a range have been used in machine learning methods used, includes ensemble methods like random forests, decision trees, support vector machines, and neural networks. Using a disciplined technique that includes model selection, feature engineering, and data preparation, training, validation, and assessment , It is conceivable to forecast the best student placements. A common process for predicting student placement is outlined in the stages below:

1. Data gathering and preparation:

- Collect relevant information from several sources, such as academic transcripts, test results, personal statements , and socioeconomic statistics.
- The data should be cleaned and preprocessed, with outliers, inconsistent data, and missing values handled.
- Use methods like label encoding or one-hot encoding to encode categorical information.

2. Engineering Features:

- Identify the essential elements, such as academic achievement, test results, and preferences, that are likely be able to affect student placements.
- If more features are required, add them. For example, compute the cumulative GPA or provide derived attributes.

3. Model choice:

- Depending on the dataset and issue characteristics, choose the best machine learning techniques. Commonly Decision trees, random forests, support vector machines, and neural networks are some of the techniques used.
- To integrate many models for better prediction performance, think about ensemble approaches.

4. Model Education:

- Utilize the practice data to run the chosen models.
- To improve model performance, adjust hyper parameters employing random or grid search techniques search.

5. Model Verification:

- Assess the models' generalizability by validating them using the validation dataset.
- Based on the validation's findings, alter the hyper parameters to prevent overfitting.

6. Model assessment:

Use the relevant assessment measures, such as precision, recall, and accuracy, F1-score, and area under the curve (AUC), to evaluate the model's effectiveness. Compare their results to see which model performs the best.

7. Visualization and Interpretation:

- Particularly for transparent models like decision trees, interpret the model's decision making process.
- To acquire insights, visualize key attributes, model predictions, and boundary conditions.

III. MODELING AND ANALYSIS

To be able to forecast student placements using machine learning, predictive models must be built and their performance must be evaluated. The modeling and analytic procedure for predicting student placement is described in section below:

- **Model selection:** Depending the section issue and the attributes of the data, select the best machine learning methods. Think about methods like gradient boosting, Support vector machines, decision trees, random forests, and logistic regression. The interpretability, scalability, and existence of nonlinear relationships in the data all influence the method of choice.
- **Data preprocessing:** Take care of missing values, encode category variables, and normalize or scale numerical characteristics while preprocessing data. Make sure the data is available in a manner that may be utilized for instruction and evaluation.
- **Identify pertinent:** These factors that help with the forecast of student placements. Metrics of academic achievement, test results, demographic data, and any other elements gleaned from the study may fall under this category.
- **Splitting the data:** Create training, validation, the dataset, as well as test sets. This makes it possible to train models, tune hyper parameters, and assess performance using separate datasets.
- **Model Training:** Utilize the training dataset to educate the chosen machine learning model. To enhance the model's functionality, alter the hyper parameters.

Learning rates, regularization parameters, and the quantity of ensemble trees are examples of hyper parameters.

- Validate the model using the validation dataset to determine how well it generalizes. To avoid overfitting, keep an eye on validation metrics and make appropriate hyper parameter adjustments.
- **Performance Evaluation:** Using the right evaluation criteria, assess the trained model's execution on the test dataset. Accuracy, precision, recall, F1score, and space behind the curve (AUCROC) are examples of common metrics. To determine the ratio false positives, false negatives, real positives, and true negatives., analyze the confusion matrix.
- **Bias Analysis:** Check the predictions of the model for any possible gender, racial, or socioeconomic biases. Reweighting samples, utilizing fairness-aware algorithms, or taking into account feature engineering approaches that balance representation can all help to reduce bias.

- **Fine-Tuning the Model:** Iterate the process by experimenting with other methods, modifying hyper parameters, or adding further features if the model's performance is unsatisfactory.

IV. RESULTS AND DISCUSSION

The collected results open the way for insightful conversations that shed light on the model's performance, insights learned, and prospective improvements after installing the machine learning models for student placement prediction and completing extensive assessments. This section summarizes the findings and debates that followed:

- **Model Performance:** Present the performance metrics acquired by the predictive models on the test dataset. Include metrics such like precision, recall, and accuracy, F1ROC. For example

Accuracy:85.2%

Precision:0.78

Recall:0.87

F1-score :0.87

AUC-ROC:0.89

- **Insights and Patterns:** Talk about any intriguing findings that emerged from the investigation. For example, did certain traits have a bigger effect on placement predictions than others? Were there any particular academic or character traits that strongly connected with successful placements?
- **Comparison with Baseline:** If appropriate, contrast the machine learning model's performance similar to a baseline model or rule-based system. Draw attention to the advancements made possible by utilizing methods of machine learning.
- **Discussion of Study limits:** Identify any study restrictions, such as data availability issues, possible sources of bias, or modeling method limits. Discuss how the outcomes could have been affected by these constraints.
- **Practical Implications:** Talk about how the model's performance in actual situations has practical ramifications. How may the predictive model be implemented in educational settings to help teachers and students decide where to be placed?
- **Feature Importance:** Outline the study of feature importance and highlight the characteristics that were crucial in producing reliable forecasts. Describe how the alignment of these traits with intuition and domain knowledge. Discuss the confusion matrix to understand how False positives, real positives, and true negatives, and false negatives are distributed. Draw attention to any class disparities and the model's response to them.

V. CONCLUSION

A machine learning system with an online job portal and placement prediction has the potential to completely transform the hiring and job seeking processes. This technology attempts to match job searchers with acceptable employment openings and help companies make data-driven recruiting decisions by utilizing cutting-edge machine learning algorithms. The suggested approach is advantageous to both companies and job seekers. The user-friendly design makes it simple for job searchers to create and manage their accounts, and tailored job suggestions increase their chances of landing the ideal position. Employers gain from a successful recruiting process since candidate employer matching and placement prediction offer useful information for selecting employees wisely. The accuracy and effectiveness of the system are improved by the use of machine learning models for candidate-employer matching and placement prediction.

These models develop over time utilizing continuing learning from new data, resulting in better candidate suggestions and placement forecasts. Additionally, the real-time updates and alerts keep both employers and job seekers updated on the status of application submissions and prospective matches, encouraging improved interaction and communication. However, careful consideration of a number of elements is necessary for the system's successful deployment. It is crucial to protect the privacy of the data and mitigate any potential biases in the machine learning models. To maintain system fair and build faith in it, several issues require resolution. As a result, the traditional job search and hiring process may be transformed by the Online Job Portal and Machine Learning-Based Placement Prediction System. The system can successfully match applicants with employment opportunities and help companies find the greatest fit for their openings by utilizing the

effectiveness of data-driven decision-making and sophisticated algorithms. This approach can lead to enhanced job placement results and increased recruiting efficiency through continual development and adaptability to shifting market trends, which will be advantageous to both employers and job seekers.

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