



# Student placement prediction using machine learning

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**Abstract:** Developing a placement prediction model through machine learning involves abstracting complex patterns from historical data, academic performance, and industry trends. By employing algorithms, the model identifies key features influencing successful placements. This abstraction process allows the model to make accurate predictions, guiding educational institutions and employers in making informed decisions about student placements based on individual aptitudes and market demands.

**Keywords:** historical data, academic performance, industrial trends, machine learning

## I.INTRODUCTION

In the dynamic landscape of workforce recruitment, the integration of machine learning has emerged as a transformative tool for predicting optimal job placements. This technological paradigm shift aims to revolutionize traditional hiring processes by leveraging data-driven insights and predictive analytics.

Placement prediction using machine learning involves the analysis of vast datasets encompassing candidate profiles, skillsets, and historical placement outcomes. Through sophisticated algorithms, machine learning models can identify patterns and correlations that human recruiters might overlook. These models consider diverse factors, such as educational background, work experience, and specialized skills, to generate predictions about the most suitable job placements for individuals.

By harnessing the power of artificial intelligence, organizations can streamline their recruitment processes, minimize human biases, and enhance the accuracy of placement decisions. This approach not only expedites the hiring process but also contributes to increased job satisfaction for candidates and improved retention rates for employers.

In this era of digital transformation, the fusion of machine learning and placement prediction heralds a new era for talent acquisition, where data-driven insights play a pivotal role in shaping the future of workforce dynamics.

## II.NEED OF THE STUDY.

The study on placement prediction using machine learning is essential for optimizing job placement processes. By analyzing historical data and patterns, machine learning models can predict candidates' suitability for specific roles, enhancing recruitment efficiency, reducing hiring biases, and improving overall placement outcomes for both employers and job seekers.

### 2.1 Optimizing Recruitment Processes:

In today's competitive job market, companies receive a large number of applications for job positions. Sorting through these applications manually can be time-consuming and resource-intensive.

Machine learning algorithms can analyze historical placement data, identify patterns, and predict which candidates are more likely to succeed in specific roles. This optimization can significantly streamline the recruitment process by narrowing down the pool of applicants, allowing recruiters to focus their efforts on the most promising candidates.

### 2.2 Enhancing Decision-Making Accuracy:

Traditional methods of candidate assessment may be subjective and prone to biases. Machine learning models can be trained on diverse datasets to make predictions based on objective criteria.

By leveraging predictive analytics, recruiters and employers can make more informed decisions about the suitability of candidates for particular roles.

This not only improves the accuracy of the placement process but also helps in identifying candidates with the right skills and attributes that align with the organization's needs.

### 2.2 Reducing Employee Turnover:

Making a successful placement involves not only finding a candidate with the right skills but also ensuring a good fit with the company culture. Machine learning models can consider a wide range of factors, including the candidate's skills, experience, personality traits, and cultural fit.

Predictive models can help organizations identify candidates who are not only technically qualified but are also likely to stay with the company for the long term. This reduces employee turnover rates, saves costs associated with recruitment and training, and contributes to a more stable and productive work environment.

## III. RESEARCH METHODOLOGY

This research employs a systematic approach to predict placement outcomes using machine learning. It involves data collection from academic and extracurricular records, feature selection, model training, and evaluation. The methodology aims to enhance accuracy in forecasting student placement success, contributing to informed decision-making in career services. The details are as follows;

### 3.1 Data Collection and Preprocessing:

#### a. Data Gathering:

Collect relevant data for the study, including historical placement records, student profiles, academic performance, internship experiences, and any other factors that may influence placements. Ensure data consistency and accuracy by verifying sources and cleaning the dataset.

#### b. Data Exploration and Understanding:

Perform exploratory data analysis (EDA) to understand the distribution, trends, and relationships within the dataset. Identify potential outliers, missing values, and anomalies in the data.

#### c. Feature Selection and Engineering:

Select features that are likely to have a significant impact on placement outcomes.

Create new features or transform existing ones to enhance the predictive power of the model.

Handle missing data and outliers appropriately through imputation or removal.

### 3.2 Model Development and Training:

#### a. Model Selection:

Choose suitable machine learning algorithms for placement prediction. Common choices include decision trees, random forests, support vector machines, and gradient boosting.

Consider the nature of the data and the problem, such as whether it's a classification or regression task.

#### b. Training and Validation:

Split the dataset into training and validation sets to train and evaluate the model.

Implement techniques like cross-validation to ensure the model's generalizability.

Tune hyperparameters to optimize model performance.

#### **c. Evaluation Metrics:**

Define appropriate evaluation metrics based on the problem, such as accuracy, precision, recall, F1 score, or mean absolute error. Evaluate the model using the validation set and fine-tune as needed.

### **3.3 Theoretical framework**

The theoretical framework for placement prediction using machine learning encompasses the integration of various concepts and methodologies from both the fields of machine learning and human resource management. At its core, this framework relies on the premise that historical placement data can be leveraged to build predictive models that aid in optimizing the matching process between candidates and job opportunities. Machine learning algorithms play a pivotal role in this framework, as they enable the extraction of patterns and trends from large datasets, allowing for the identification of key factors influencing successful placements.

The framework begins with the collection and preprocessing of relevant data, including candidate profiles, skill sets, educational backgrounds, and historical placement outcomes. Feature engineering is employed to extract meaningful features that capture the essential characteristics of both candidates and job positions. These features serve as input variables for the machine learning models.

Supervised learning algorithms, such as classification models, are commonly employed in placement prediction. These models are trained on historical data, learning the relationships between input features and placement outcomes. Feature importance analysis helps identify the most influential factors in successful placements. Additionally, ensemble techniques and deep learning models may be explored to enhance the predictive capabilities of the system.

Incorporating domain-specific knowledge from the human resource management field is crucial. Factors such as cultural fit, soft skills, and career aspirations are considered alongside technical competencies. The framework also acknowledges the dynamic nature of the job market and the evolving nature of skills required in various industries.

The deployment of the model involves continuous monitoring and updating to adapt to changing trends and market dynamics. Ethical considerations, such as bias mitigation and transparency in decision-making, are integral components of the framework to ensure fairness and accountability in the placement process.

In conclusion, the theoretical framework for placement prediction using machine learning merges data-driven insights with domain expertise, creating a robust system that enhances the efficiency and effectiveness of the job placement process. The integration of machine learning models in human resource management holds the potential to revolutionize the way organizations match candidates with suitable opportunities, ultimately contributing to a more responsive and adaptive job market ecosystem.

## **VI. Model Used**

### **4.1 Logistic Regression:**

Type: Supervised Learning (Classification)

Use Case: Logistic Regression is suitable for binary classification problems, where the goal is to predict whether a candidate will be placed (1) or not (0).

Advantages: It is computationally efficient, interpretable, and provides probabilities for predictions.

Considerations: Assumes a linear relationship between features and the log odds of the target variable.

### **4.2 Random Forest:**

Type: Supervised Learning (Ensemble Method - Decision Trees)

Use Case: Random Forest is effective for both classification and regression tasks. It can handle complex relationships and interactions between features.

Advantages: Robust against overfitting, handles non-linearity well, and provides feature importance scores.

Considerations: May require tuning of hyperparameters for optimal performance

#### **4.3 Gradient Boosting (e.g., XGBoost):**

Type: Supervised Learning (Ensemble Method - Boosting)

Use Case: Gradient Boosting algorithms, like XGBoost, are powerful for predictive modeling. They sequentially build weak models to correct errors made by previous models.

Advantages: High predictive accuracy, handles complex relationships and automatically handles missing data.

Considerations: Requires careful tuning of hyperparameters, and training time can be longer compared to simpler models.

## **V. Conclusion**

In conclusion, leveraging machine learning for placement prediction enhances the efficiency of matching candidates with suitable roles. Models like Logistic Regression, Random Forest, or Gradient Boosting provide accurate insights into historical data, optimizing the placement process. The fusion of data-driven approaches with human resource expertise creates a dynamic and effective framework for predicting successful job placements.

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