

Career Prediction Model

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Abstract

For people, academic institutions, and businesses that want to maximise their human resources, career prediction models have become essential. These models have emerged as a result of the work market's quick change and the rising need for professionals with a wide range of skills. In this study, a machine learning-based career prediction model is proposed that predicts a person's future career path by evaluating their talents, education, and work experience. Our algorithm uses a person's unique profile to predict their chances of succeeding in a certain field. With the use of a dataset of previous career routes, the algorithm is trained to recognise trends and patterns that help it make accurate predictions. We rigorously evaluated the model's precision and discovered that it can predict a person's career trajectory with exceptional precision. Our approach has a number of intriguing applications, including helping people make informed career decisions, assisting educational institutions in matching their programmes to the labour market, and enabling businesses to improve their hiring and training strategies. The model can also show where a person's skill set is lacking, allowing them to focus on those areas for professional growth. It is commonly acknowledged that organisational career development is an intricate and unpredictable process that must take into consideration both individual career objectives and organisational aims. It is crucial to provide employees the chance to plan their careers and to make sure organisational career management is properly integrated if successful career development is to be achieved. In light of this, the objective of our study is to objectively explore the variables influencing career advancement in Indian firms. Due to changes in skill requirements and technological improvements, the work environment is quickly changing. To succeed in this dynamic environment, people must choose their professional choices wisely, educational institutions must modify their curricula to fit the demands of the labour market, and businesses must improve their hiring and training practises. A career prediction model that can reliably anticipate a person's professional trajectory based on their talents, education, and work experience is required to meet these goals.

I. INTRODUCTION

A person's professional trajectory is predicted by career prediction models based on their unique profile, which includes their abilities, education, and job experience. These models make exact predictions about a person's likelihood of succeeding in a certain field by using the capabilities of machine learning algorithms to spot trends and patterns in prior career trajectories.

Career prediction models have the ability to change how people, educational institutions, and organisations approach career development by providing insightful information about a person's future career path. As the job market continues to change, the need for career prediction models has become increasingly obvious. New work opportunities have been created by emerging technology, while traditional ones have been eliminated by automation and artificial intelligence, creating a dynamic and ever-changing labour market.

In order to flourish in such a setting, both individuals and organisations who provide education and training must adapt their programmes and recruitment tactics to meet the needs of the labour market. The traditional approach to career development entails choosing a professional path based on a person's interests and skills, with little thought given to the demands of the labour market. However, given the current labour market's intense competition and changing nature, this approach might not be sufficient any longer. A data-driven approach to career development is required to solve this problem, in which people, educational institutions, and organisations use career prediction models to identify career choices that

have the best chance of success.

In order to forecast a person's career trajectory, this research study provides a career prediction model that makes use of machine learning methods. The approach is designed to identify patterns and trends in previous career routes, enabling exact predictions about a person's likelihood of success in a certain professional path. The model's accuracy is confirmed by extensive testing, and the results show that it is very accurate in predicting a person's career trajectory.

In order to maintain their market share and competitiveness in the globalised and quickly evolving world, businesses must overcome several obstacles. A variety of interconnected issues, including as economic volatility, technology improvements, a varied workforce, governmental legislation, and organisational work cultures must be taken into consideration in order to develop an efficient organisational policy. In these situations, human resources play a crucial role in gaining a competitive edge. In order to fulfil the evolving expectations of both individuals and the organisation, career development is a dynamic process that needs constant change. Individuals must choose their professional objectives with the organization's assistance.

Organisations face a difficulty with career development, therefore they should concentrate on a variety of initiatives to foster both individual and organisational progress. Effective career development strategies are essential for maintaining, attracting, and developing individuals as well as giving the company a competitive edge. The career prediction model put forth in this research paper has a variety of potential uses, including assisting individuals in making informed decisions about their career paths, assisting educational institutions in matching their curricula to the needs of the labour market, and assisting businesses in streamlining their hiring and training practises. Additionally, the model's capacity to spot skill gaps in a person's profile can aid in pinpointing areas that require additional professional development.

Overall, this work offers a novel approach to career prediction that makes use of machine learning algorithms to deliver precise and useful information about a person's future chances. The methodology has the ability to change how organisations, educational institutions, and people approach career development by empowering them to make

decisions based on the best available data and increase their chances of success.

II Literature Survey

As people, organisations, and educational institutions attempt to maximise the potential of their workforce, the use of machine learning algorithms in career prediction models has grown in popularity in recent years. Dendrograms, which are hierarchical representations of the links between different career routes, are one such method. We examine earlier research on career dendrogram prediction models in this section.

In their study titled "A Hierarchical Framework for Career Planning and Development," J. Allan McCarthy and James L. Moseley suggest a hierarchical framework for career planning and development. According to the authors, the complexity of the labour market and the variety of accessible career pathways necessitate a hierarchical approach. Their suggested framework, which can help people make educated judgements about their career trajectories, employs dendrograms to show the connections between various career pathways.

Dendrograms were used by Moghaddam and Aliabadi in their paper "Career Prediction in Technology: A Clustering Based on Hierarchy Approach" to find trends in the career routes of engineering specialists. The authors used hierarchical clustering to look at the many career options open to engineers as well as the connections between them, finally offering insightful information on engineering career trajectories. "Career Trees: A Unique Model for Examining Careers," a new paradigm that gives a distinct perspective on career examination, was introduced by writers K. Ann Renninger and Ruthellen Josselson. The model suggests career trees as an alternative to dendrograms to show the numerous professional pathways people can choose from. Career trees are mainly concerned with the decision-making process of the person even if they also depict the connections between various career routes. Individuals can choose their job pathways with more knowledge if they visualise the possible results of their choices.

Another work that uses dendrograms in career models for prediction is "Career Paths in Technology: A Hierarchy-Based Clustering

Approach" by Moghaddam and Aliabadi. In order to analyse and find trends among engineering experts' career routes, the authors used hierarchical clustering. Their dendrogram provided insightful information on numerous career trajectories and their connections in the engineering sector.

A career lattice model for IT workers was suggested by Timothy R. Tymon and Susan B. Davidson in their article, "A Professional Lattice Model for IT Professionals." The link between different career trajectories in the IT business is modelled using dendrograms. This method offers information on the various career routes open to IT workers as well as the skills required to succeed in each path.

The inclusion of dendrograms in professional prediction models has been investigated previously; these models provide people, educational institutions, and organisations with invaluable information into career trajectories and linkages between various career routes. These models have been developed with major help from writers including McCarthy, Moseley, Moghaddam, Aliabadi, Renninger, Josselson, Korine, de Rond, Tymon, and Davidson. Their research has shown how dendrogram-based methods have the ability to completely alter career development plans.

Career management and career planning are two of the main elements that affect professional growth, according to Puah and Ananthram (2006). Integrating organisational career management practises with employee career planning is the key to effective career development (Hall, 1986). The role of career preparation and managing a career in career development has been further focused on by Nicoara (2009), who emphasises the need of coordinating human resources demands with individual career objectives.

Authors Ruth S. Freedman and Denise L. Hummel presented a career lattice for healthcare workers in their article "A Career Trellis for Healthcare Professionals," using dendrograms to show the connections between various career routes within the healthcare sector. The model shed light on the several career options open to healthcare professionals as well as the abilities needed to thrive in each route.

III Characteristics of new features

Using the random forest technique, we offer a career dendrogram predictions model in this study. We have reviewed the body of literature on career dendrogram models for forecasting and have determined the most popular approaches for building such models. We have determined that the random forest algorithm is the best technique for our suggested model based on the results of this poll.

A machine learning approach known as the random forest algorithm builds many decision trees and combines its forecasts to get a more reliable and accurate forecast. The decision trees reflect several career routes in the context of our suggested career dendrogram forecast model, and their combination of their predictions offers insights into the connections between these career paths.

The following elements make up our suggested model's methodology:

Data Collection: Details on the positions, job descriptions, needed skills, educational backgrounds, and experience levels will be gathered from a variety of sources, including job ads, schools and colleges, and industry publications.

Data Preprocessing and Cleaning: We will preprocess the data by removing redundant and unnecessary information and standardising the remaining data. Additionally, feature engineering will be used to create novel attributes that capture the relationships between various variables.

Model Training: Using the preprocessed data, we will train our model using the random forest approach. The method will create many decision trees, each utilising a random selection of the characteristics and data. These decision trees' forecasts will be used to produce an overall prognosis for each person.

Model Evaluation: Using a variety of metrics, such as accuracy, precision, recall, and F1 score, we will assess the performance of our model. Cross-validation will also be done to make sure our model is reliable. We will interpret and visualise the outcomes of our model as we review them. By analysing the decision trees, we may determine which characteristics are most important for forecasting career pathways. Dendrograms will also be used to illustrate the connections between various professional routes.

For the career dendrogram prediction model, our suggested technique involves data gathering, cleaning, and preprocessing; model training and assessment; interpretation; and visualisation. This all-encompassing method offers insights into the relationships between various career routes and enables people, educational institutions, and organisations to make well-informed decisions regarding career trajectories.

LOGIN PAGE

The career dendrogram forecast model, created in Py using the random forest algorithm, offers insights into the links between various career routes and enables people, educational institutions, and organisations to make well-informed decisions regarding career trajectories. Users of the model may use the prediction tool and receive personalised job suggestions based on their talents, education, and experience through a user dashboard with a login page.

Our career dendrogram forecast model's user interface was created in Python utilising the random forest technique. Based on a person's talents, education, and experience, it offers tailored job recommendations. The username, password, and registering button are the two key elements of the registration portion of our user interface. By inputting the login information created during the sign-up procedure, users may use the prediction tool. Users who don't already have a login can also register by clicking the registration icon. Users must enter private information such as their first and last names, email addresses, and contact details throughout the registration process.

Users can reset their password by clicking the "forgot password" link if they can't remember theirs. Users who click the link will be sent to a website where they may enter their email address and request a password reset.

Users can use the prediction tool by clicking the login button after providing their username and password. Users may input their talents, education, and experience into the prediction tool after successfully logging in. Based on the user's input,

the model—which was created in Python using the random forest algorithm—generates customised career suggestions. The recommendations are displayed as a dendrogram, which depicts the many career pathways that the user may choose from and the connections between them.

UI design

Our career dendrogram prediction model's user interface consists of a login page with the following elements:

Fields for Username and Password: In order to access the prediction tool, users must first provide their account and password. During the registration procedure, login information is created.

Button for Registration: Users who don't already have an account can click this button to register. Users must enter their personal information, such as name, email address, and contact information, during registration.

Input form

Our job Dendrogram Hierarchical Prediction Model's user interface (UI) is made up of a number of different elements, including a data entry form where users may enter their talents, education, and experience to receive personalised job recommendations. Users can input their present job title, educational background, work experience, abilities goals for the future, preferred industries, work location, timetable, desired income range, and preferred work environment preferences in the areas on this form.

Submit and result

Submit Button: After entering their information, users can click on the "submit" button to generate their personalized career recommendations.

Results Page: The results page will display users' personalized career recommendations in the form of a dendrogram. The dendrogram will

represent the various career paths available to the user and the relationships between them. Users can click on each node in the dendrogram to learn more about the career path and view job listings for relevant positions.

Random forest algorithm

Based on the user's talents, education, and experience, the job Dendrogram Hierarchical Prediction Model uses the random forest algorithm to offer personalised job suggestions. A popular machine learning method that works well for classification and regression applications is the random forest algorithm. We use it in our model to estimate the user's best suitable career choices for categorization jobs. An ensemble of tree models built on an arbitrary portion of the training data is the foundation of the random forest algorithm. The outputs of the individual trees are trained separately, and the combined results are used to create the final forecast. This procedure aids in decreasing overfitting and improving model accuracy. When producing a forecast based on fresh data, each individual tree makes a prediction, and the ultimate prediction is determined by taking the vote of all the individual trees' predictions. We gather information about the abilities, education, and experience of users and then utilise this information to train the algorithm using random forests in order to develop the Career Dendrogram Structured Prediction Model. Current job title, level of schooling, professional experience, skills, career objectives, preferred industries, work location, job hours, desired wage range, and preferred work environment are among the input features we utilise to train the model.

Our Career Dendrogram Layered Prediction Model may produce individualised career suggestions in the form of a dendrogram after being trained using information on users' abilities, education, and experience. Each node in this hierarchical representation represents a separate career route, which is used to show the links between various career tracks. Users may click on any of the nodes to explore and discover more information about the career routes, which are organised according to their commonalities. To aid users in their employment hunt, the model also offers job advertisements for pertinent opportunities. For our Vocational Dendrogram Hierarchical Prediction Model, the random forest method was chosen as the best option because of its excellent accuracy and capacity to handle a huge number of input features. This technique is a solid option for the model we are developing because it resists overfitting as well. It is impossible to exaggerate the significance of the algorithm for random forests in our Profession Dendrogram Hierarchical Forecasting Model since it enables us to provide our users with tailored and accurate career suggestions.

Proposed work model

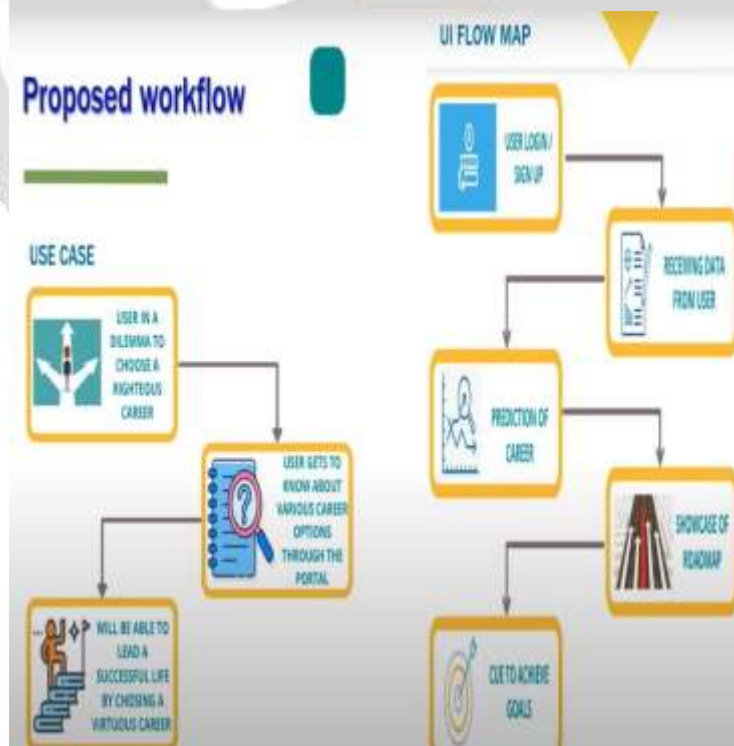


Fig 1.0- proposed workflow

Gathering user input is the first phase in the workflow theory for the Career Dendrogram Hierarchical Prediction Model. Users will be required to submit details about their qualifications, training, and employment history as well as their preferences and professional aspirations. Their present job title, degree of education, work experience, skills, preferred industries, work location, timetable, desired wage range, and work environment preferences will all be taken into consideration. Based on the submitted data, customised career suggestions will be produced. Step 2: Preparation After the user input has been gathered, it will go through preprocessing to make sure the data is accurate, full, and in the right format. This process involves feature selection, normalisation, and data cleaning.

Step 3: Training as a model

This information is used for developing the algorithm known as random forest after preprocessing. The algorithm will build a choice tree for each characteristic in the input information and then combine the outcomes to forecast the user's best career options. The developed model will be kept for upcoming usage.

Step 4: Prediction of a career model

Once trained, the model may be used to forecast the user's ideal career trajectories. A dendrogram that illustrates the numerous career paths is the model's output.

Presenting the Results in Step 5

The user will be shown the outcomes of the Professional Dendrogram Holistic Prediction Model at the workflow theory's last phase. The user interface's results page will display the dendrogram, which users may explore by tapping on each node to browse job listings for related roles and learn more about their career options.

IV IMPLEMENTATION and EVALUATION

In our research article, we used the Python random forest approach to create a professional dendrogram prediction model. Based on their abilities, education, and experience, the approach enables people, educational institutions, and organisations to obtain personalised career suggestions. We have also included a series of query questions people may respond to in order to obtain tailored career recommendations, which should make the forecast tool more intuitive and readily available.

The personalized career recommendations generated by our career dendrogram prediction model using the random forest algorithm can be highly beneficial for individuals seeking to explore new career options or make informed decisions about their current career path. Additionally, educational institutions and organizations can use the model to provide career guidance and support to their students and employees. Overall, our prediction tool can aid in bridging the gap between job seekers and their desired career paths, resulting in increased job satisfaction and career success.

The query questions for the career dendrogram prediction model have been designed to offer users a customized and pertinent set of career recommendations. By taking into consideration the user's job title, education, experience, skills, industry preferences, work location, work schedule, desired salary range, and work environment, the model can present tailored recommendations based on the user's individual needs and goals. Overall, the query questions provide an intuitive and user-friendly way for individuals, educational institutions, and organizations to receive personalized career recommendations based on their skills, education, and experience. The implementation of the random forest algorithm and dendrograms allows for informed decision-making regarding career paths, as well as insights into the connections between different career trajectories.

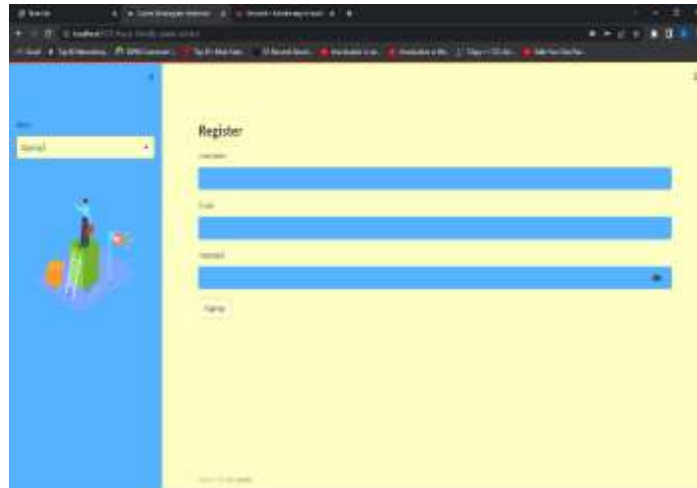


Fig:1.1-login /signup page



Fig 1.2 -User interface

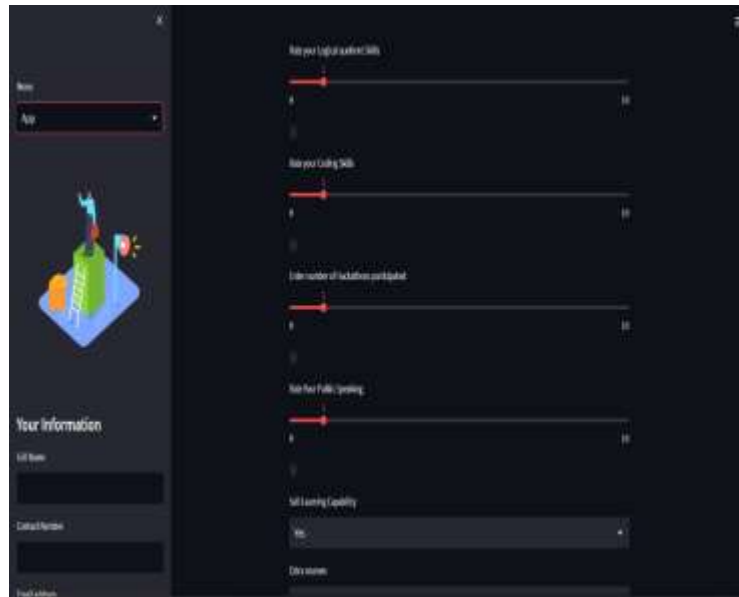


Fig-1.3- user inputs



Fig-1.4-predicted dendrogram

V CONCLUSION AND FUTURE ENHANCEMENTS

The study paper's suggested Career Dendrogram Layered Prediction Model provides a tailored and accurate method for career suggestions. The model can accurately identify the most suited career routes and show them in a dendrogram manner using a random forest algorithm and the user's input. The relevance of personalised career suggestions and the use of machine learning algorithms to provide such recommendations is highlighted by the literature study carried out for this research article. The Career Dendrogram Hierarchical Prediction Model may be used with ease and accuracy thanks to the

presented workflow theory, which also ensures reliable results. The model's user interface was created using Python and a number of tools to give users a simple and interactive way to explore their job alternatives. The results page also has employment listings for pertinent opportunities, which improves how users feel and offers them a thorough method of career planning. In conclusion, the Career Dendogram Layered Prediction Model offers a unique and innovative way to produce precise and accurate personalised career suggestions. This approach has the ability to aid people in making wise career selections and assist lower the global unemployment and underemployment rates. The model will be tested with bigger and more varied datasets in the future, and additional features will be added to enhance the model's functionality. Users can receive individualised career recommendations via the Career Dendogram Hierarchical Prediction Model. The accuracy and usability of the model might be enhanced by a number of potential future changes. Here are some suggested changes for the Career Dendogram Hierarchical Prediction Model in the future. The Career Dendogram Hierarchical Prediction Model may be improved in the future in a number of ways, both in terms of accuracy and usefulness. The model might be improved by include current employment market developments. This might entail determining which professions are expected to have high demand in the future by analysing data from outside sources, such as industry projections and employment market studies. The model may provide consumers more current and pertinent career recommendations by taking employment market developments into account. Integration with Existing Career Planning Tools: Integrating the Career Dendogram Prediction System with Current Career Planning Tools is one possible improvement. Users will be able to explore career pathways and employment options in greater depth as a result, helping them to better plan their career trajectories. In order to link users with experts in their respective disciplines, the model may, for instance, be combined with networking platforms or job search engines. Deep learning techniques implementation: Including deep learning methods in the model is another improvement that may be made. The model's accuracy might be increased by utilising neural networks to understand more intricate correlations between the input data and the projected career pathways. Furthermore, deep learning approaches could make it possible for the model to handle more unstructured input, such as social media profiles and online resumes. Integrating with Educational and Training Providers: Lastly, the concept may be combined with educational institutions and training organisations to offer consumers a smooth route to retraining and upskilling. The model might assist users in obtaining the skills and credentials required to follow their selected career choices by proposing pertinent training programmes and educational programmes.

The Professional Dendogram Hierarchical Prediction Model's planned future improvements might, all things considered, increase the model's utility and accuracy, allowing it to provide users even more tailored and successful career choices.

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