

SKILL AND JOB RECOMMENDER APPLICATION

INTRODUCTION:

A recent report claims that most college graduates have difficulty in choosing their domain in their job. Many engineers are trying to shift the domain from their field to IT. So, they are doing some courses online and randomly searching for a job. Nowadays, IT fields are the target of many students but they don't know which domain is fit for them. To avoid this situation candidates, need a Job recommendation that analyses the skills to recommend a suitable job for the candidate. The solution is to design a system that reads a resume and their skills. The resumes are going through pre-processing to make the design more efficient. For pre-processing top words and porter stemmer, Porter Stemmer will make every word their root word, and stop words will remove every meaningless word. This makes the system more efficient. Using tf-idf vectorizer for both resume and job description. Then compare the skills in the resume and description.

LITERATURE SURVEY:

Existing works are mainly found for the company to select a candidate who is fit for their vacancy. There are many experiments for calculating the four recommendation algorithm but with a different distance formula namely the Minkowski distance. And some others are tried a different recommender system like collaborative which only helps when there are more data to relate. That won't help for a person who is searching that which job is the correct choice for him/her.

R.J. Mooney and L. Roy used Content-Based Book Recommending where the content-based recommendation helps for a cold start. And some authors also say that a content-based recommender is best when they researched a comparison study of job recommendations .

A recommender system is not only the main part of accurate prediction. There are some other things like vectoring the words and then similarity functions. Authors like Shouning Quad, and Li-Ping Jing aid that for text mining, tf-idf is the best approach for text feature selection. Ravali Boorugu has researched NLP and tried various text summarization techniques. Some papers also say about similarity detection with many languages.

Jeevamol Joy and Renumol V G discussed which similarity is the best one for a content-based recommending system. They finally concluded that cosine similarity is the best similarity for content based recommended system. Cosine similarity is not only used for recommender systems but is used to find the similarity functions between two sentences or two paragraphs.

Mohammad Alobed has tried "A Comparative Analysis of Euclidean, Jaccard, and Cosine Similarity Measure and Arabic Wordnet for Automated Arabic Essay Scoring", and L. Zahrotum also compared Jaccardidean and cosine similarity. They both said that Cosine similarity with all stemming types has the lowest error compared with the Jaccard and Euclidean similarity. There is already a system that worked with both tf-idf and cosine similarity recommendations.

It is used for patient support forums. Tanya V. Yadalam, Vaishnavi, M.Gowda, and Vanditha researched those career recommendations content-based filtering which was mostly like my project but inside it, they mostly discussed security, transparency for the data, and the framework. Most works are just built for the companies or for the purpose of making money from the people by giving some irrelevant choices.

Many were using collaborative recommendation, which recommends the many searched jobs or the jobs which were chosen by some other. It only works if the system deals with more number of resumes which seems it can only be used by the companies. Some systems are asking to log in and some were asked to buy subscriptions.

Logging in makes you redirect some spam mails. In many papers, they have been solved through content recommender which is not enough. A literature paper had done research on content recommender system, vectorizer, and cosine similarity in a row but in that the author doesn't think about the implementation process and only concentrated more on securing the data.

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