



# International Journal of Advance Engineering and Research Development

Special Issue for ICPCDECT 2016, Volume 3 Issue 1

## Resume Extractor and Candidate Recruitment System

Bhoite Mayuri<sup>1</sup>, Bhosale Monika<sup>2</sup>, Chaudhar Kajal<sup>3</sup>, Totre Neha<sup>4</sup>

Department of computer engg, SVPM's COE  
Malegaon (bk), Baramati, Pune

---

**Abstract** — Resume Extractor and candidate Recruitment system is a system which can be very useful for any organization for their recruitment process. The system will be robust enough which will automatically extract the resume content and store it in structure form within the database. On web we may get resumes of different format like .doc, .pdf, .docx. Firstly those are converted into single format of text file. Classification algorithm (Naive bayes) will run on the candidates information from database to identify the profiles of candidates categories or its classes. Also the employer i.e resource manager can specify his particular criteria for required post and also decide the importance level of candidate. Duplicated resumes are also removed and most updated and relevant resume is selected.

---

**Keywords-** Classification, Data mining, Recruitment, Resumes

### I. INTRODUCTION

Now a days all major industries driven by technology. According to current statistics, information available on the internet is about 60 per of what we need.[3] This figure is expected to rise exponentially in the near future. Companies are publishing more and more information on the internet about every aspect of their business and their growth.[4] Resume Extractor and Recruitment candidate system basically extract all the resumes about the candidate only through his/her resumes, without forcing candidate to fill any other information about them. After extraction it stores the information in to the centralized database, allowing the HR managers to search in the data base for their criteria satisfy candidates.

Today many job portals are available but the basic problem in available system is, it requires manual efforts for both employers [2] and candidate. Candidate has to require complete information in given text field and employer also need to apply many filters to select the candidate.[6] Even though employer has applied many filters, he would get thousands of resume. And going through all the resumes and selecting the candidates is very inefficient and time consuming task. This refereed as "problem of resume selection". Some costly extraction systems are available in the market that also do the search on the keyword basis and has many extraction limitations like forcing candidate to fill template and updating the template as per job profiles. Large enterprises and head-hunters receive several thousands of resume from job applicant every day. HRs and managers go through hundred of resumes manually.

Resumes or profile are unstructured documents and have typically number of different formats(eg: pdf, txt, doc).As a result manually reviewing multiple profiles is a very time consuming process. How to ensure you have appropriate candidate in the right job at the right time. This is significant problem faced by large companies today in the market.

### II. EXISTING SYSTEM

As we know, this is an era of an Internet. Everyone want everything in easy way. So for finding a job or for searching a job there is lots of job portal websites. Such as naukri.com,monster.com...etc. These job portals or websites are useful for candidate or user. Researchers from [7] state some of existing online recruiting platforms and also listed their technical advantages and disadvantages. Every job portal has their own application form. But what happen exactly on that websites that every user/candidate has to fill whole application form which includes four to five pages. And after completing this process user has to attach his/her resume also, which is repeated work.

That means which information is present in that application form and the candidate/user's resumes from that some fields are same .This is very lengthy process and also time consuming. And on another side that means on HR side, according to his companies criteria he takes candidates resumes. There is n numbers of resumes are collected. For that HR has to check all these resumes sequentially, which is very hectic work for HR. Resume is a digital document which is structured document. Each resume may be different as according to persons thinking. And these resumes can be in any format like .doc, .pdf, .html, ect. Extracting data from such documents is challenging work to get the "relevant" resume.[8]

### III. PROPOSED SYSTEM

### 3.1. Information Extraction

Information Extraction (IE) is a type of information retrieval whose goal is to automatic extract structured information from unstructured and/or semi-structured machine readable document. Resume or candidate profile is typically unstructured data or electronic documents.

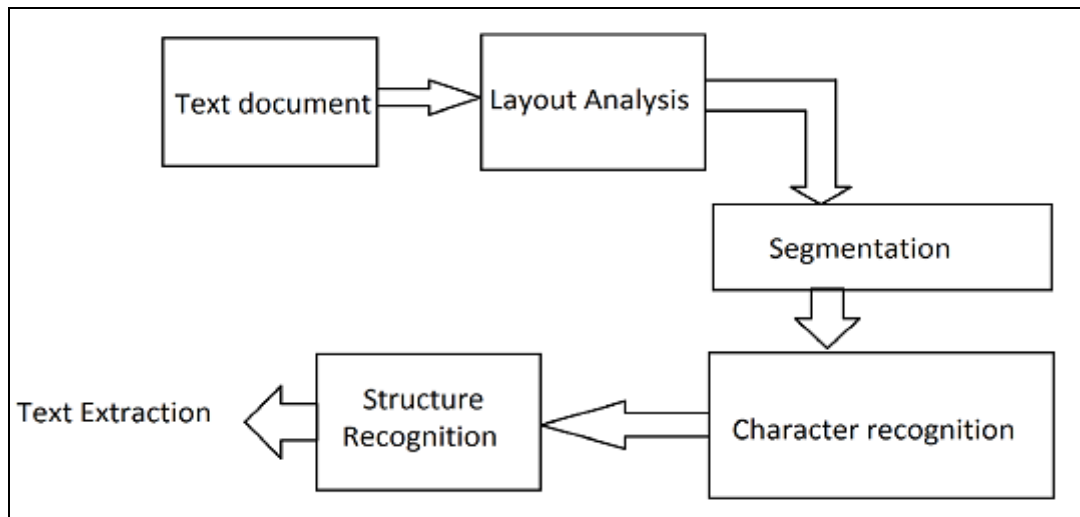


Figure: Content extraction

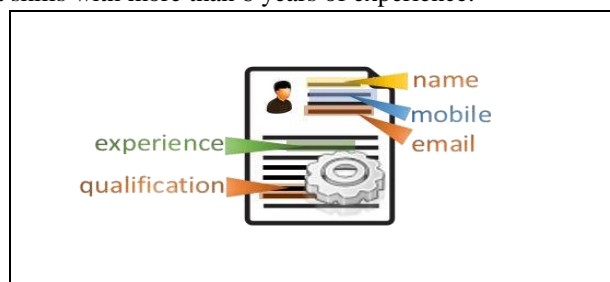
We need to extract information and convert this into structured formats so that we can analyze or query on this data in an effective manner. This information is extracted and provided in a such manner that the features can be extracted and compared. Here in our system we are going to extract the special features of candidate from his/her resume. These features may be experience, skills, education, preferred location.

### 3.2 Document Preprocessing

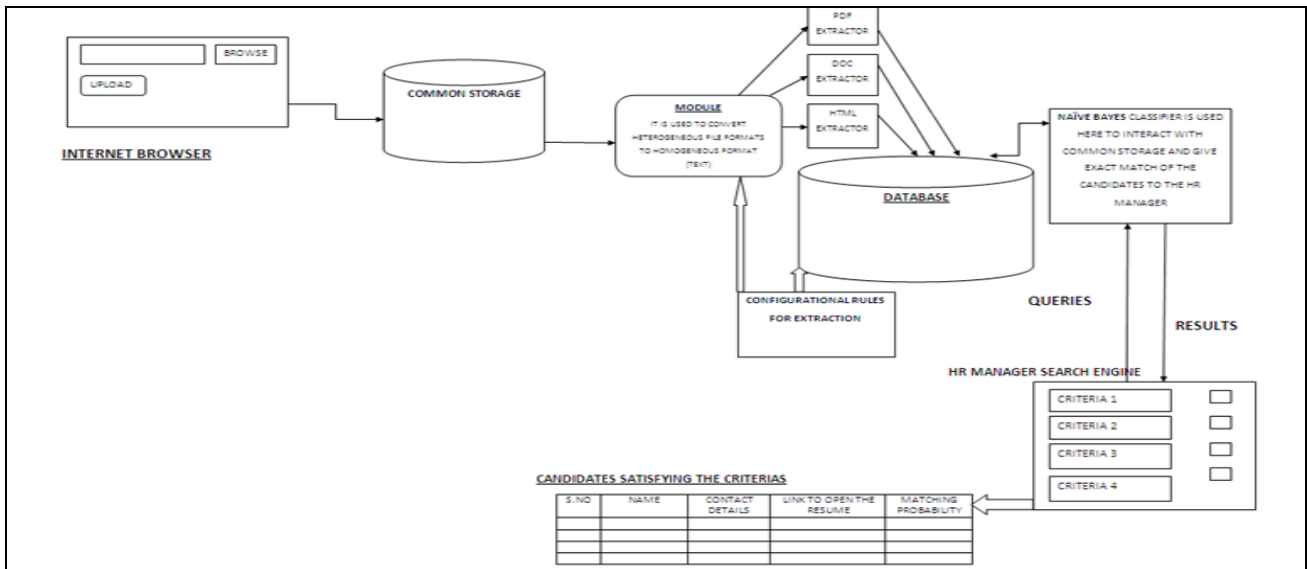
First we convert the input resumes in different file types (eg: pdf, doc) into txt format. We need to maintain one dimension table for storing all the keywords that may appear in the input resume. Then we have to travelling through the text file which is obtained after processing the input resume. So that we can find keywords present in input resume in txt format and store them in a database for that particular resume.

### 3.3 Document Preprocessing

Resume parsing technology means important for both candidates and recruiters. Resume parsing allows you to process online resumes or electronic document of persons profile by extracting data in an intelligent way. It helps recruiters to efficiently manage electronic resume documents sent via the internet. Example: If any candidate recruitment system is only keyword searchable, and you search for someone specifically with six or more years of JAVA experience, they may get everyone with JAVA system. Our system which will consist of resume parser will return only those candidate who are having java skills with more than 6 years of experience.



Why to do all this when proper resume parser will derive information on when a skill was last used. The parser will then go through all job descriptions and if the skill is mentioned, the start and end dates for those positions are used to calculate the total number of years of experience the person had in those skills. Even the most recently updated end date of a position where the skill is mentioned become for that skill. As , with tagged data from a parser, our search can be more intelligently defined, and thus, the results much more narrow. Instead of receiving 30 resumes, our system will get three.



Figur: System Architecture

### 3.4 Search Profile

For given search criteria for resume we check in database for the presence of given input criteria search result contains name of resume, matching percent. This will get done using naive bayes algorithm. Candidates probability to satisfy the given criteria is calculated and the resumes with high probability are then sent to resource manager. From the vast data from the electronic documents and www it is not reliable step towards the business success to properly classify such information into our need. Naive bayes classifier works good regard of other classifying techniques due to its simplicity.[9]

### 3.5 Remove Duplication

As we are collecting resumes from 4 different websites it may happen that some candidates have their resumes on more than one website. In this case we are finding the most updated resume. And the result will be sent to the HR. The updated resume will be selected by date and experience field.

## IV. MATHEMATICAL MODEL

Let  $C_i =$  set of candidates,  $| 1 \leq i \leq n$

Let  $R_j =$  Set of resumes of each candidate on k-sites.  $| 1 \leq j \leq m$  and  $m < n$ ,  $k \neq 0$ ,  $k \leq m$

Problem is to find all  $R_j$  from  $C_i$ ,  $\forall k$  sites and reduce them to single  $R'$ .

$R'_i =$  non duplicate current resume of  $C_i$

$R'_i = \int_{j=1}^k R_j * dR_j$

$R'_i =$  Set of distinct current non duplicate resume of each candidate.

Let  $J_c =$  Candidate for job.

$\forall R'_i$  do -> if  $R'_i$  satisfies  $J_c$  select candidate

Let  $L_i =$  List of final sets candidates satisfying  $J_c$

## V. ALGORITHMIC STRATEGY

### 5.1 Psuedo-Algorithm

#### Candidate /Client side

if(login successful for candidate)

```
{  
upload resume  
}  
On succesful resume uplodation  
{  
(extract resume from common data repository )  
FTP  
if(resumes of particular candidate exist more than one)
```

#### HR side

```
if(login successful for hr)  
{  
provide selection criteria  
{  
Based on selection criteria  
execute naive biased algorithm  
Get the probable eligible candidates  
}
```

#### 5.2 Naive Bayesian classifier

Here is NAIVE BAYES algorithm which includes:

- Strong independence (Naive) assumption.
- Bayes theorem.
- Prior and posterior probability.
- How the Naive bayes classifier is used in classification of resumes

##### 5.2.1 Introduction

Bayesian classifier are statistical classifier. They can predict class membership probabilities, such that a given sample belongs to a particular class. Bayesian classifier is based on byes theorem. Naive Bayesian classifier assume that the effect of an attribute value on a given class is independent of the values of the other attribute. This assumption is called class conditional independence. I t is made to simplify the computation involved, and in the sense is considered as "NAIVE".

##### 5.2.2 Explanation of bayes rule

Bayes rule:

$$P(H|E) = P(E|H) * P(H)P(E),$$

The basic idea of Bayesian rule is that the outcome of a hypothesis or an event(H)can be predicted on basic evidences(E) that can be observed from above rule.

(1) **A priori probability** of H or P(H): This is the probability of an event before the evidence is observed.

(2) **A posterior probability** of H or P(H j E): This is the probability of an event after the evidence is observed.

##### 5.2.3 Naive Bayesian classifier

The Naive Bayesian works as follows:

1. Let, T be the training set, each with their class labels. There are k classes C1,C2,...Ck ,is the sample represented by n-dimensional vector, X={x1, x2,...,xn} depicting n measured values of the n attribute A1,A2,...An respectively.
2. Given sample X, the classifier will predict that X belongs to the class having the highest posteriori probability, conditioned on X. That is X is predicted to belong to the class C if and only if  $P(C_i|X) > P(C_j|X)$  for  $j \neq i$ . Thus we find the class.
3. As P(X) is the same for all classes only  $P(X|C_i)P(C_i)$  need be maximized .If the class apriori probabilities, P(Ci), are not known, then it commonly assumed as the classes are equally, likely, i.e  $P(C1)=P(C2)=...=P(Ck)$ . We would therefore maximize  $P(X|C_i)$ . Otherwise we maximize  $P(X|C_i)P(C_i)$ . Here we have to note that the class priori probabilities may be estimated as  $P(C_i)=\text{freq}(C_i, T) | T |$ .
4. Given data sets with many attributes, it would be computationally expensive to compute  $P(X|C_i)$ . In order to reduce computation in evaluating  $P(X|C_i)P(C_i)$ . The naive assumption of class conditional independence is

made. This presumes that the values of the attribute are conditionally independent of one another. Given the class label of sample. Mathematically this means that  $P(X|C_i) = \prod_{k=1}^n P(x_k|C_i)$  where  $k=1$ .

The probabilities  $P(x_1|C_i), P(x_2|C_i), \dots, P(x_n|C_i)$  can easily estimated from the training set. Recall that here  $X_k$  refers to the value of attribute  $A_k$  for sample  $X$ .

- (a) If  $A_k$  is categorical, then  $P(x_k|C_i)$  is the number of samples of class  $C_i$  in  $T$  having the value  $x_k$  for attribute  $A_k$ , divided by  $\text{freq}(C_i, T)$ , the number of sample of class  $C_i$  in  $T$ . It predicts the class of a previously unseen test case  $T = \{a_1, a_2, \dots, a_n\}$  by selecting the class  $C_i$  that maximizes the following formula:

$$P(c_i | T) = \frac{P(T, c_i)}{P(T)} = \frac{P(c_i) \cdot P(T | c_i)}{P(T)}$$

- (b) Where  $P(T|C_i)$  denotes the conditional probability of the test case  $T$  given class  $C_i$ . Probabilities are estimated from the training set. Since classification focuses on selecting the class that maximizes, rather than assigning explicit probability to each class, The denominator  $P(T)$  in above equation can be omitted as it does not affect the relative class order. Naive bayes classifier is a simple probabilistic classifier based on applying Bayes theorem with naive assumptions.

## VI. ADVANTAGES

- This system provides time efficient and very effective candidate selection process.
- It is easy for user as they just need to upload their resumes on portal .No form filling is require.
- It is highly reliable as employer can specify their criteria along with importance level.
- Automatic E-mail notification to candidate/employer can be possible.

## REFERENCES

- [1] Wenxing Hong, Siting Zheng, Huan Wang\*, "A Job Recommender System Based on User Clustering", JOURNAL OF COMPUTERS, VOL. 8, NO. 8, AUGUST 2013
- [2] Abhishek Sainani, "Extracting Special Information to Improve the Efficiency of Resume Selection Process", 2011
- [3] Jonathan Medema "Reliable Normalization in Rsum Information Extraction" Published in 2008 <http://igitur.archive.library.uu.nl/studenttheses/2009-0211-202406/UUindex.html>
- [4] Roger E. Bohn and James E. Short, "How Much Information? 2009" <http://hmi.ucsd.edu/pdf/HMI-2009-Consumer-Report-Dec-9-2009.pdf>
- [5] D.D. Lewis, "Naive (Bayes) at Forty: The Independence Assumption in Information Retrieval" Proc. European Conf. Machine Learning (ECML) pp.4-15
- [6] Sumit Maheshwari "Mining Special Features to Improve the Performance of Product Selection in E-commerce Environment and Resume Extraction System"(ICEC 2009) Taipei, Taiwan, August 2009, Published by ACM.
- [7] S.T. Al-Otaibi and M. Ykhlef, A survey of job recommender systems, International Journal of the Physical Sciences, vol. 7(29), pp. 5127-5142, July, 2012.
- [8] Charul Saxena "Enhancing Productivity of Recruitment Process Using Data mining Text Mining Tools"
- [9] S. L. Ting, W.H. Ip, Albert H.C. Tsang "Is Nave Bayes a Good Classifier for Document Classification?" International Journal of Software Engineering and Its Applications Vol. 5, No. 3, July, 2011
- [10] M. Gao and Y. Q. Fu, User-Weight Model for Item-based Recommendation Systems, Journal of Software, vol. 7(9), pp.2133-2140, 2012.
- [11] G. Adomavicius and A. Tuzhilin, Toward the next generation of recommender systems: a survey of the state-of-the-art and possible extensions, Knowledge and Data Engineering, IEEE Transactions on, vol. 17(6), pp. 734-749, 2005.
- [13] E. Cesario, F. Folino, G. Manco, and L. Pontieri. An incremental clustering scheme for duplicate detection in large databases. In *Proceedings of the International Database Engineering Application Symposium (IDEAS)*, pages 89–95, Montreal, Canada, 2005.
- [14] **Feature Extraction Using ICA** Nojun Kwak, Chong-Ho Choi, and Jin Young Choi G. Dorffner, H. Bischof, and K. Hornik (Eds.): ICANN 2001, LNCS 2130, pp. 568–573, 2001. \_c Springer-Verlag Berlin Heidelberg 2001