

Predicting Interview's Attendance

By Implementing Naïve Bayes Technique

Using

Big Data Programming With Hadoop

by

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DESCRIPTION OF THE JOB

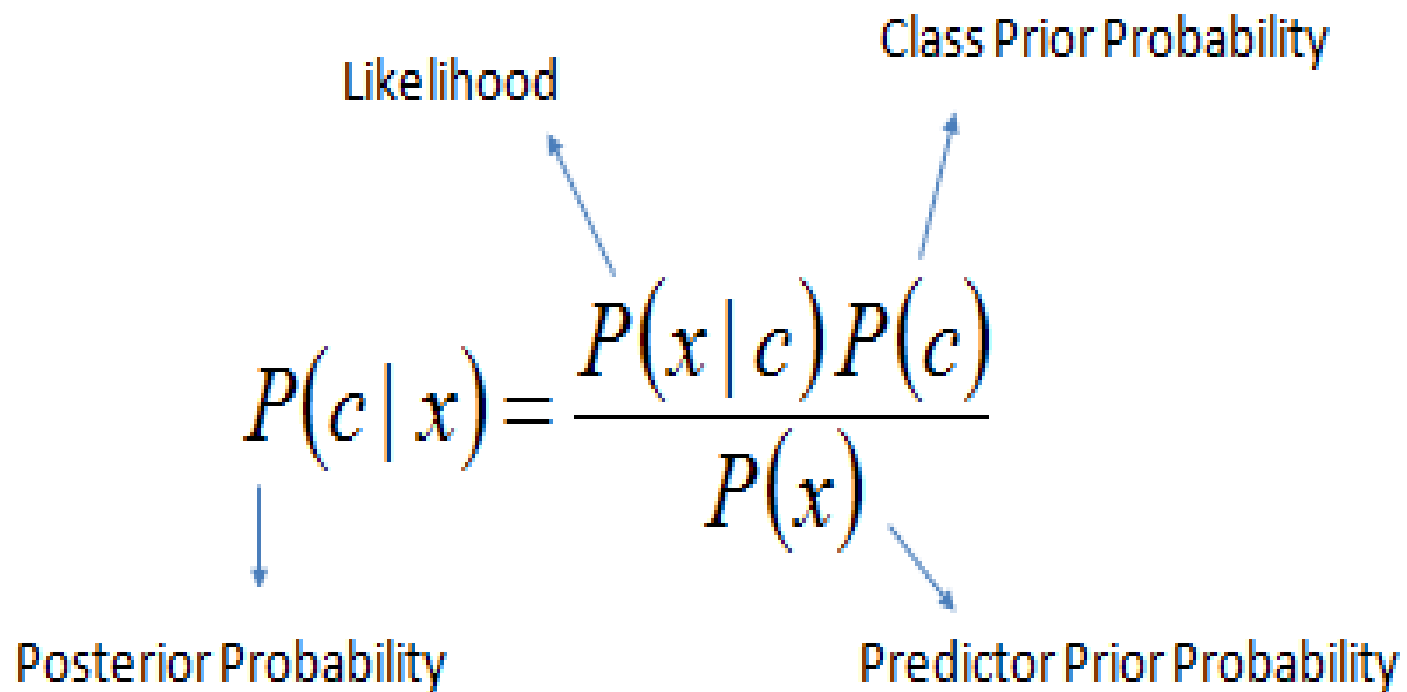
We were given a dataset including various fields like client name, gender, location, answers to various questions asked over phone etc.

Based on this we have to predict the real appearance of candidates in the interview.

TECHNIQUE USED

We have used Naïve Bayes supervised machine learning to predict the appearances in this project. Naive Bayes is a simple but surprisingly powerful algorithm for predictive modeling.

Naive Bayes is called naive because it assumes that each input variable is independent. This is a strong assumption and unrealistic for real data, nevertheless, the technique is very effective on a large range of complex problems.



A diagram illustrating Bayes' Theorem. The equation $P(c | x) = \frac{P(x | c)P(c)}{P(x)}$ is centered. Four blue arrows point from labels to parts of the equation: 'Likelihood' points to $P(x | c)$, 'Class Prior Probability' points to $P(c)$, 'Posterior Probability' points to $P(c | x)$, and 'Predictor Prior Probability' points to $P(x)$.

$$P(c | x) = \frac{P(x | c)P(c)}{P(x)}$$

Labels and arrows:

- Likelihood (points to $P(x | c)$)
- Class Prior Probability (points to $P(c)$)
- Posterior Probability (points to $P(c | x)$)
- Predictor Prior Probability (points to $P(x)$)

$$P(c | X) = P(x_1 | c) \times P(x_2 | c) \times \cdots \times P(x_n | c) \times P(c)$$

Hardware and software requirements

**HADOOP CLUSTER IS USED WHICH IS MADE OF
COMMODITY HARDWARE**

Master's recommendation

4–6 1TB hard disks

**(1 for OS [RAID 1], 2 for the FS image [RAID 5/6],
1 for JournalNode)**

**2 CPUs(8-12 cores per CPU), running at least
2-2.5GHz**

128-512GB of RAM

Bonded Gigabit Ethernet or 10Gigabit Ethernet

SLAVE'S RECOMMENDATION

12-24 1-4TB hard disks in a JBOD (Just a Bunch Of Disks) configuration

2 CPUs(8-12 cores per CPU), running at least 2-2.5GHz

64-128GB of RAM

Bonded Gigabit Ethernet or 10Gigabit Ethernet



SOFTWARE

Hadoop 2.2.0 or above runs well on Linux operating systems like: RedHat Enterprise Linux (RHEL), CentOS, Ubuntu

Hadoop is written in Java. The recommended Java version is Oracle JDK 1.6.31

SOURCE AND DESCRIPTION OF DATA

❑ THE DATA HAS BEEN PROVIDED BY THE INSTITUTE

❑ THE DATA CONSISTS OF 23 DISCRETE SPECIFICATIONS WHICH INCLUDES 1234 NUMBER OF RECORDS

❑ TYPES OF SPECIFICATIONS:

- **DATE OF INTERVIEW**-The date on which the interview is going to take place
 - **CLIENT NAME**-Name of the company
- **INDUSTRY**-The type of industry concerned (Electronics' 'Telecom' 'IT')
 - **LOCATION**-Location of the company ('Delhi' 'chennai')
- **POSITION TO BE CLOSED**-The post offered(Production- Sterile' 'Selenium testing')
 - **NATURE OF SKILL SET**-skill required for the job('Routine' 'Oracle')
- **INTERVIEW TYPE**-Type of interview(Walk in, scheduled, scheduled walk in)
 - **NAME**-Candidate Id
 - **GENDER**
- **CANDIDATE CURRENT LOCATION**-Present address of the candidate
 - **CANDIDATE JOB LOCATION**-Where the candidate will be working
 - **INTERVIEW VENUE**
 - **CANDIDATE NATIVE LOCATION**

- **HAVE YOU OBTAINED THE NECESSARY PERMISSION TO START AT THE REQUIRED TIME-**
 - May be yes , no or not yet
- **HOPE THERE WILL BE NO UNSCHEDULED MEETINGS- TIME-May be yes**
 - , no or not yet
- **CAN I CALL YOU THREE HOURS BEFORE THE INTERVIEW AND FOLLOW UP ON YOUR ATTENDANCE FOR THE INTERVIEW- TIME-**
 - May be yes , no
- **CAN I HAVE AN ALTERNATIVE NUMBER/DESK NUMBER-**
 - Yes or no
- **HAVE YOU TAKEN A PRINT OUT OF YOUR UPDATED RESUME-**
 - yes or no or NA
- **ARE YOU CLEAR WITH THE VENUE DETAILS AND THE LANDMARK-**
 - Yes or no or need to check or not sure
- **HAS THE CALL LETTER BEEN SHARED-**
 - Yes or no or haven't checked
- **EXPECTED ATTENDANCE-**
 - uncertain, yes or no
- **OBSERVED ATTENDANCE-**
 - yes or no
- **MARITAL STATUS-**
 - single or married

DATA CLEANSING

- Rows in which some data were missing or all the 23 data weren't present are ignored
- Some data included incorrect punctuation marks or non word characters and are rectified
- Some values were nearly similar (e.g: liaison and liabilities, liaison & liabilities, liaison liabilities) and were changed to a single value
- Rows where some discrepancy in data was there were rectified

DATA PROCESSING

Concept of Map-Reduce has been used to process the data.

Three jobs have been used namely:-

IA:: MAPPER AND 3 REDUCERS

TEST:: MAPPER AND NO REDUCER

ANALYSIS:: MAPPER AND 1 REDUCER

Work Flow Diagrams

DATASET

data provided by the institute



IA Job(Map-Reduce)

- Cleansing the given dataset and converting nearly similar values to a single value.
- Generates count for each unique values of dependent variables given the appearance, which is used to calculating prior and likelihood probabilities



FIRST OUTPUT

- stores the counts calculated in the IA Job.
- Probabilities can be calculated by simply one division and is done during actual prediction for greater accuracy

```
graph TD; A[TEST Job] --> B[SECOND OUTPUT]; B --> C[ANALYSIS Job (Map-Reduce)]; C --> D[FINAL OUTPUT];
```

TEST Job

predicts the attendance of all the candidates given in the dataset using the learned probabilities and also compares with the observed output.

SECOND OUTPUT

stores for each candidate, the predicted attendance and observed attendance and if they matches

ANALYSIS Job (Map-Reduce)

Calculates the conclusion matrix and also the accuracy of the model.

FINAL OUTPUT

Stores the conclusion matrix and also accuracy of the model as calculated by the analysis Job.

OUTPUT

Output of IA job

x11_no	6	73		
x11_yes	651	240		
x13_na	103	164		
x14_no	2	14		
x14_yes	682	260		
x15_uncertain	1	1		
x16_na	103	162		
x17_no	1	92		
x17_yes	666	219		
x1_aonhewitt	24	4		
x1_barclays	5	0		
x1_pfizer	51	24		
x1_prodapt	6	11		
x1_ust	10	8		
x2_itproductsandservices	34	11		
x2_pharmaceuticals	96	69		
x3_bangalore	193	99		
x3_gurgaon	22	11		
x4_dotnet	10	8		
x4_productionsterile	0	5		
x4_seleniumtesting	4	1		

Output of Test Job

```

Output Matched      Observed : No      Predicted : no  <--Candidate 1
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1232
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1231
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1230
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1229
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1228
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1227
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1226
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1225
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1224
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1223
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1222
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1221
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1220
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1219
Output Matched      Observed : Yes     Predicted : yes  <--Candidate 1218
      :
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 730
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 729
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 1025
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 728
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 204
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 525
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 279
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 1020
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 724
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 1018
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 1017
Output Not Matched  Observed : Yes     Predicted : no   <--Candidate 523
Output Not Matched  Observed : Yes     Predicted : no   <--Candidate 124
Output Not Matched  Observed : Yes     Predicted : no   <--Candidate 521
Output Not Matched  Observed : Yes     Predicted : no   <--Candidate 379
Output Not Matched  Observed : No      Predicted : yes  <--Candidate 202

```


Output of Analysis Job(Final Output)

Observed yes , Predicted yes : 650

Observed yes , Predicted no : 111

Observed no , Predicted yes : 214

Observed no , Predicted no : 216

Accuracy : 72.712006%

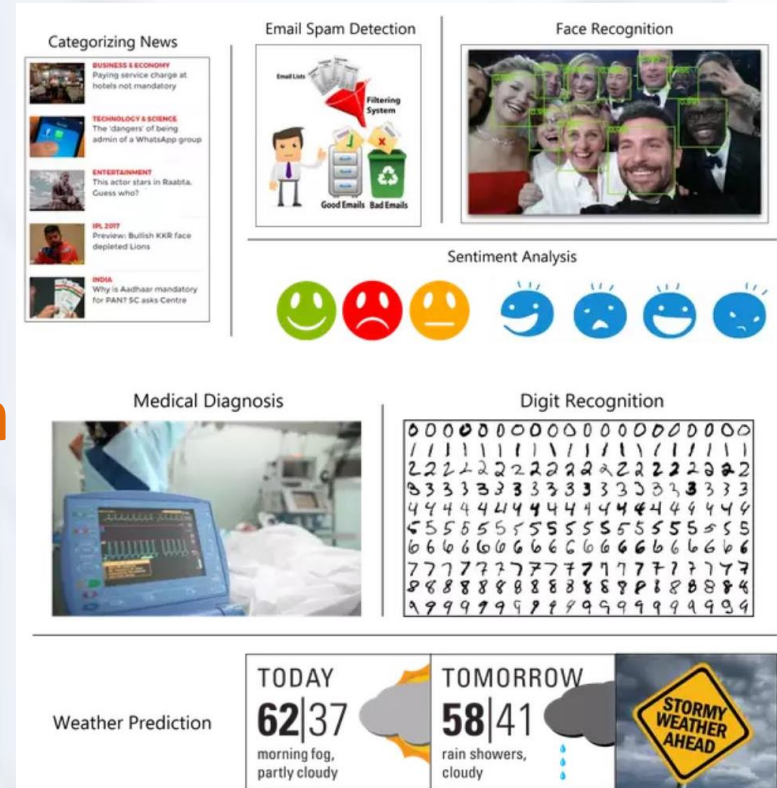
FUTURE IMPROVEMENT

The prediction analysis has been done based on a batch processing system instead of a real time system. For further improvement this has to be transformed into real time system using Apache Spark.

APPLICATION

This Naïve Bayes approach using MapReduce can be used similarly in many other applications:

- ▶ Churn Detection
- ▶ Vote Prediction
- ▶ Email Spam Detection
- ▶ News article categorization
- ▶ Sentiment Analysis
- ▶ Facial recognition
- ▶ Handwriting Recognition
- ▶ Weather Prediction



CONCLUSION

From a given data set, we have calculated the conditional probability for each value of the dependent attributes given the appearance of the candidates using Naïve Bayes supervised machine learning. After that we have predicted the candidate appearance and cross checked with the observed appearance and we have got a satisfactory 73% accuracy. So we can say that this application is working as expected.

Thank
you!!