PG Department of Computer Science Sacred Heart College (Tirupattur)

In Plant Training

Boscosoft, Yelagiri - April 29th to May 28th Friendzion Technologies- May 4th to May 30th

S.NO	REGNO	NAME	COMPANY
1	BP180501	PRASANTH RAO S	Boscosoft, Yelagiri Hills
2	BP180502	PRIYA S	Boscosoft, Yelagiri Hills
3	BP180503	ARIVAZHAGAN M	Friendzion Technologies
4	BP180504	NISHA M	Boscosoft, Yelagiri Hills
5	BP180505	MANGALA MARY A	Boscosoft, Yelagiri Hills
6	BP180506	VINODHINI V	Boscosoft, Yelagiri Hills
7	BP180507	PUSHPALATHA P	Boscosoft, Yelagiri Hills
8	BP180508	PRABHU S	Boscosoft, Yelagiri Hills
9	BP180509	KOWSALYA M	Boscosoft, Yelagiri Hills
10	BP180510	ANTONY XAVIER S	Friendzion Technologies
11	BP180511	POOJA E	Boscosoft, Yelagiri Hills
12	BP180512	RENIRATHINAM R	Lesoftek, Salem
13	BP180513	IVISALLWYN A	Friendzion Technologies
14	BP180514	WILLIAM A	Friendzion Technologies
15	BP180515	SHALINI K	Boscosoft, Yelagiri Hills
16	BP180516	JOAN GEORGE T	Friendzion Technologies
17	BP180517	JOHNSON JERO S	Friendzion Technologies
18	BP180518	DHANALAKSHMI A	Boscosoft, Yelagiri Hills

	19	BP180519	MASEEHA TASNEEM K	Boscosoft, Yelagiri Hills
	20	BP180520	SOWMIYA A	Boscosoft, Yelagiri Hills
	21	BP180521	JABEER S	Boscosoft, Yelagiri Hills
	22	BP180522	GEETHAPRIYA R	Boscosoft, Yelagiri Hills
	23	BP180523	BHARATHKUMAR S	Boscosoft, Yelagiri Hills
	24	BP180524	KOTHANDARAMAN S	Friendzion Technologies
	25	BP180525	DHANALAKSHMI H	Boscosoft, Yelagiri Hills
ĺ	26	BP180526	ASHWINIPRIYA G	Boscosoft, Yelagiri Hills

sever 20 Head PG Department of Computer Science Sacred Heart College (Autonomous) Tirspattur, Tellere Dt. - 635 601.

CERTIFICATE OF COMPLETION

This certificate is proudly presented to S KOTHANDARAMAN

For successfully completing his industrial inplant training in Adobe Photoshop and Android from 04-05-2019 to 30-05-2019

During the training period the performance of trainee was found to be Good.



PG Department of Computer Science Sacred Heart College (Tirupattur)

Software project- Name List :: April 2019

S.No	REGNO	NAME	GENDER
1	BP170501	VASANTHA KUMARI G	Female
2	BP170502	SUGASHINI V	Female
3	BP170503	DEEPIKA J	Female
4	BP170504	MALINI A	Female
5	BP170505	LOKESH C	Male
6	BP170506	MURALIDHARAN D	Male
7	BP170507	CARMEL RAJ A	Male
8	BP170508	PAVITHRA S	Female
9	BP170509	CHINNAPPA RAJ J	Male
10	BP170510	PAVITHRA K	Female
11	BP170511	STEPHY A	Female
12	BP170512	SHINY MOULISHA R	Female
13	BP170513	JASMINE SHILPA G	Female
14	BP170514	SWETHA C	Female
15	BP170515	SHARADHA S	Female
16	BP170516	VIDYA E	Female
17	BP170517	GOKULASELVAN T	Male
18	BP170518	AMMU S	Female
19	BP170520	SUGANTHI S	Female
20	BP170521	MURUGAN R	Male
21	BP170522	PRABAVATHI P	Female
22	BP170523	PAVITHRA V	Female
23	BP170524	AISWARYA S	Female
24	BP170525	SRIDHAR R	Male
25	BP170526	IMMACULATE MARY J	Female

epel P6 Department of Computer Science Sacred Heart College (Antenamons) Tirspattur, Tellere Dt. - 635 601.

ANALYSIS OF STUDENT'S MARKS AND PREDICTING THEM BY LOGISTIC REGRESSION

A Report

Submitted in partial fulfillment for the award of the degree of

MASTER OF COMPUTER SCIENCE

MURALIDHARAN D

(REG.NO:BP170506)

Under the Guidance of

Miss. D.Gajalakshmi M.C.A.,



PG DEPARTMENT OF COMPUTER SCIENCE

SACRED HEART COLLEGE (AUTONOMOUS)

TIRUPATTUR, VELLORE DT – 635 601

April - 2019

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INTRODUCTION

1.1 Abstract

"Analysis of students marks and predicting them by logistic regression "this is an predicting applications of the students mark details, by using the python language. In this the analysis are made of every student of the particular class and predicting the percentage of passing students in each subject and overall passing percentage of the students in the class. We can also represent the predicted results of the students in graphical representation by plotting in bar diagram, pie chart and different kind of plotting method. The main purpose of the project is to identity whether the students result is pass or fail in the class by fixing the threshold value. We use logistic regression for this operation because the algorithms perfectly suits for our project by which it can be predicted by "pass" or "fail" /"0" or "1" /" yes"/"no".

CHAPTER – I INRODUCTION

1.2 Project Overview 1.2.1 Introduction

"Analysis of students marks and predicting them by logistic regression "this is an predicting applications of the students mark details, by using the python language. In this the analysis are made of every student of the particular class and predicting the percentage of passing students in

each subject and overall passing percentage of the students in the class. We can also represent the predicted results of the students in graphical representation by plotting in bar diagram, pie chart and different kind of plotting method. The main purpose of the project is to identity whether the students result is pass or fail in the class by fixing the threshold value. We use logistic regression for this operation because the algorithms perfectly suits for our project by which it can be predicted by "pass" or "fail" /"0" or "1" /" yes"/"no".

1.2.2 Scope

The scope of the project is used to bring the outcome of the students details regarding the academic of the candidates. We can classify them into different norms and category by classifying the dataset, we can predict different type of possibilities of in various subjects. We predict outcome like passing percentage of the students in each subject and overall subject of each students. Finally, we can also plot the outcome predicted values in histogram and we can perform them in the graphical representations.

1.2.3 Problem Statement

In this project the major problem was all the data are manually for each students and employees for the database due to is the time duration is to long huge man power is involved so that the new system in this project evaluate out comes of search problem in this project

1.2.4 Product position statement

For	For COE of College
Who	Staff members.
That	It is more flexible.
Our Product	Provide up-to-date information and easy to maintain all the details. An classify them and make prediction and represent them in graphical representation manner

SYSTEM ANALYSIS

CHAPTER – II SYSTEM ANALYSIS

2.1 System Study 2.1.1. Introduction

The system study of "Analysis of students marks and predicting them by logistic regression "is that since the data of the students were stored in excel data format due to which the staff remembers of the class are facing problems in getting the proper results .so in this I have proposed the new way representing the data into graphical representations by which the staff members are able to identify the results easily.

2.1.2. Over view

Analysis of students marks and predicting them by logistic regression, this is an predicting applications of the students mark details, by using the python language. In this the analysis are made of every student of the particular class and predicting the percentage of passing students in each subject and overall passing percentage of the students in the class. the data of the students were stored in excel data format due to which the staff remembers of the class are facing problems in getting the proper results .so in this I have proposed the new way representing the data into graphical representations by which the staff members are able to identify the results easily. We can also represent the predicted results of the students in graphical representation by plotting in bar diagram, pie chart and different kind of plotting method. The main purpose of the project is to identify whether the students result is pass or fail in the class by fixing the threshold value.

2.1.3 Proposed System

• Representing the new way of present the data into the graphical reorientations.

2.1.5. 1 ABOUT CLASSIFICATION DETAILS OF THE PROJECTS

Purpose

To produce the complete details about the classifications and prediction.

Overview

The user, who wants to know about the prediction of any case, can easily get the details in graphical representations.

Entry Criteria

The users can directly import the data set, train and test them to get predictions.

Input

Here the input is test data from the data set.

Steps involved

Import .csv file into the python and perform the code form train and test using classification algorithms and make predict the desired output.

Output

The required detail is shown to the knowledge seekers.

2.2 Feasibility Study 2.2.1 Introduction

The scope of the document is to define the system requirements "**Analysis of students marks** and predicting them by logistic regression " software. The impact can be either positive or negative. When the positives nominate the negatives, then the system is considered feasible. Here the feasibility study can be performed in two ways such as technical feasibility and Economical Feasibility.

2.2.1.1 Technical Feasibility:

We can strongly says that it is technically feasible, since there will not be much difficulty in getting required resources for the development and maintaining the system as well. All the resources needed for the development of the software as well as the maintenance of the same is available in the organization here we are utilizing the resources which are available already.

2.2.1.2 Economical Feasibility

Development of this application is highly economically feasible .The organization needed not spend much money for the development of the system already available. The only thing is to be done is making an environment for the development with an effective supervision. If we are doing so, we can attain the maximum usability of the corresponding resources .Even after the development, the organization will not be in condition to invest more in the organization. Therefore, the system is economically feasible.

SYSTEM CONFIGURATION

CHAPTER – III System Configuration

3.1 Requirement Specification

3.1.1 Hardware Requirements

Processor	: Intel i3 processor
RAM	: 4 MB
Hard Disk	: 500 GB
Input/Output	: Keyboard, mouse, monitor

3.1.2 Software Requirements

Front End	: python 3.5 (juypter)
Back End	: Excel file (CSV FORMAT)
Documentation	: Microsoft Word 2013

3.3.2 Reliability

3.3.2.1 SUPP18 Availability

This system will be configured as an online website, hence it will be accessible at24*7 days.

3.3.2.2 Robustness

For every invalid input from the user, the system should display a meaningful error message explaining in what format the input is to be fed into the system.

Ex: Error message for invalid Username or Password in login screen

"Invalid Username", "Incorrect Password"

Session should not expire in between the operations.

3.3.2.3 SUPP20 Accuracy

- ✤ The result should be accurate.
- ✤ The reports should give accurate data.
- ✤ Transactions should not be collapsed in between operation.

3.3.2.4 Safety

Information maintained in the system should be kept confidential especially the user details.

3.3.3 Performance

3.3.3.1 Response Time

Average system response time should be less than five seconds to open the forms, to do the events and to display the result and to generate the report etc.

3.3.3.2 Capacity

✤ It should allow maximum number of users to access the system at the same time.

3.3.3 Utilization of resources

♦ CPU utilization should not exceed 25% when the application is functioning.

3.3.4 Supportability

3.3.4.1 Adaptability

- Deployment time for new version should not take much time.
- System should support all web browsers.
- Application should run in windows family such as Windows XP, Windows 7.

3.3.4.2 Maintainability

Code should be developed in such a way that it should be maintainable, so define the coding standard and follow it strictly.

3.3.4.3 Reusability

Create some components as reusable components, so that it can be integrated with other system.

CONCLUSION AND SAMPLE SCREEN SHOTS

CHAPTER – IV SCREENSHOT

4.7 User Interact Page

The anaconda IDE



The main window of juypter IDE

Home	× +		-	٥	×
$\leftarrow \rightarrow $ 0	localhost.8888/tree	□ ☆ =	1	٩	
	📁 jupyter	Quit Logout			
	Files Running Clusters				^
	Select items to perform actions on them.	Upload New 🗸 📿			
		Name Last Modified File size			
	Ca Anaconda3	a month ago			
	Contacts	2 months ago			
	C Desktop	3 minutes ago			
	Conduction	22 days ago			
	Downloads	11 hours ago			
	Co Favorites	2 months ago			
	Co Links	2 months ago			
	Co Music	2 months ago			
	Co OneDrive	20 days ago			
	C Pictures	a month ago			
	Co Saved Games	2 months ago			
	Ca Searches	2 months ago			
	C Videos	a month ago			~
• • •		~ ■ */	× 17:00 × 16/10/2	018	-

The editing window in juypter

In []: 1 2 3 4 5 6 7

Importing the packages and basic file to perform the prediction.



Reading the .csv file(raw data)

In [2]: cd C:\\Users\\madhu\\Downloads\\CSV files

C:\Users\madhu\Downloads\CSV files

Extracting the data in the csv file

```
In [3]: pd.read_csv('student.csv')
Out[3]:
                       Student_Name Gender IOT CC DOS ASP MA Practical 1 Practical 2
               Reg.No
                                                                         89
           0 Bp170501 Vasantha kumari Female
                                            70
                                                69
                                                      87
                                                           68
                                                               56
                                                                                   86
           1 Bp170502
                           Sugashini Female 48 79
                                                      89
                                                           78
                                                               72
                                                                         78
                                                                                   87
           2 Bp170503
                             Deepika Female
                                            50
                                                 98
                                                      78
                                                           86
                                                               89
                                                                         89
                                                                                   90
           3 Bp170504
                              Malini Female
                                            60 86
                                                      98
                                                           94
                                                               92
                                                                         90
                                                                                   90
           4 Bp170505
                             Lokesh
                                       Male
                                            78 89
                                                      74
                                                           74
                                                               87
                                                                         81
                                                                                   87
           5 Bp170506
                         Muralidharan
                                                                         90
                                      Male
                                            40 90
                                                      67
                                                           54
                                                              90
                                                                                   90
           6 Bp170507
                                                                         76
                             Carmel
                                     Male
                                            33 89
                                                      45
                                                           89
                                                               87
                                                                                   87
           7 Bp170508
                           Pavithra 1 Female
                                                                         87
                                            87 78
                                                           85
                                                               78
                                                                                   88
                                                      43
           8 Bp170509
                         Chinnapa raj
                                            20 78
                                                      23
                                                           23
                                                              81
                                                                         77
                                                                                   80
                                       Male
           9 Bp170510
                           Pavithra 2 Female
                                                      87
                                                           68
                                                               56
                                                                         89
                                                                                   86
                                            70 69
          10 Bp170511
                                                                         78
                              Stephy Female
                                             60 79
                                                      89
                                                           78
                                                               72
                                                                                   87
          11 Bp170512
                              Shilpa Female
                                            78 98
                                                      78
                                                           86
                                                              89
                                                                         89
                                                                                   90
```

Creating the object for the csv file

```
In [4]: dd=pd.read_csv('student.csv')
```

Showing the first 5 values

Reg.No Student_Name Gender IOT CC DOS ASP MA Practical 1 0 Bp170501 Vasantha kumari Female 70 69 87 68 56 899 86 1 Bp170502 Sugashini Female 70 69 89 78 72 778 87 2 Bp170503 Deepika Female 50 98 78 89 309 90 3 Bp170504 Malini Female 78 89 74 74 87 81	In [5]:	dd	.head()									
0 Bp170501 Vasantha kumari Female 70 69 87 68 56 89 86 1 Bp170502 Sugashini Female 48 79 89 78 72 78 87 2 Bp170503 Deepika Female 50 98 78 86 89 90 3 Bp170504 Malini Female 60 86 98 94 92 90 90 4 Bp170505 Lokesh Male 78 89 74 74 87 81 87	Out[5]:		Reg.No	Student Name	Gender	ют	сс	DOS	ASP	ма	Practical 1	Practical 2
1 Bp170502 Sugashini Female 48 79 89 78 72 78 87 2 Bp170503 Deepika Female 50 98 78 86 89 90 3 Bp170504 Malini Female 60 86 98 94 92 90 90 4 Bp170505 Lokesh Male 78 89 74 74 87 81 87		0	Bp170501	Vasantha kumari	Female	70	69	87	68	56	89	86
2 Bp170503 Deepika Female 50 98 78 86 89 90 3 Bp170504 Malini Female 60 86 98 94 92 90 90 4 Bp170505 Lokesh Male 78 89 74 74 87 81 87		1	Bp170502	Sugashini	Female	48	79	89	78	72	78	87
3 Bp170504 Malini Female 60 86 98 94 92 90 90 4 Bp170505 Lokesh Male 78 89 74 74 87 81 87		2	Bp170503	Deepika	Female	50	98	78	86	89	89	90
4 Bp170505 Lokesh Male 78 89 74 74 87 81 87		3	Bp170504	Malini	Female	60	86	98	94	92	90	90
		4	Bp170505	Lokesh	Male	78	89	74	74	87	81	87

Showing the last 5 values

In [6]:	dd.tail()												
Out[6]:		Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	ма	Practical 1	Practical 2		
	21	Bp170522	Prabavathi	Female	89	86	98	56	92	90	90		
	22	Bp170523	Swetha	Female	78	89	74	74	87	81	87		
	23	Bp170524	Ishvariya	Female	56	90	34	23	90	90	90		
	24	Bp170525	Sridhar	Male	78	23	67	89	87	76	87		
	25	Bp170526	Mary	Female	89	87	85	89	89	89	90		

Describing the maximum mars in the each subject.

In [9]: dd.describe()

Out[9]:

	ют	cc	DOS	ASP	MA	Practical 1	Practical 2
count	26.000000	26.000000	26.000000	26.000000	26.000000	26.000000	26.000000
mean	65.346154	81.076923	70.769231	74.307692	76.538462	84.461538	87.576923
std	20.834476	14.912875	21.345365	18.561830	19.206730	5.756602	2.715483
min	20.000000	23.000000	23.000000	23.000000	23.000000	76.000000	80.000000
25%	51.500000	78.000000	67.000000	68.000000	72.000000	78.000000	87.000000
50%	70.000000	86.000000	76.000000	78.000000	87.000000	88.000000	87.000000
75%	78.000000	89.000000	87.000000	87.500000	89.000000	89.000000	90.000000
max	89.000000	98.000000	98.000000	94.000000	92.000000	90.000000	90.000000

Showing the attribute in the vertical order.

						-		-	-	
	0	1	2	3	4	5	6	7	8	
Reg.No	Bp170501	Bp170502	Bp170503	Bp170504	Bp170505	Bp170506	Bp170507	Bp170508	Bp170509	Bp17051
Student_Name	Vasantha kumari	Sugashini	Deepika	Malini	Lokesh	Muralidharan	Carmel	Pavithra 1	Chinnapa raj	Pavithra
Gender	Female	Female	Female	Female	Male	Male	Male	Female	Male	Fema
ЮТ	70	48	50	60	78	40	33	87	20	
cc	69	79	98	86	89	90	89	78	78	
DOS	87	89	78	98	74	67	45	43	23	
ASP	68	78	86	94	74	54	89	85	23	
MA	56	72	89	92	87	90	87	78	81	
Practical 1	89	78	89	90	81	90	76	87	77	
Practical 2	86	87	90	90	87	90	87	88	80	

Showing all the register numbers

In [11]:	dd [<mark>' R</mark>	eg.No']					
Out[11]:	0 1	Bp170501 Bp170502					
	2	Bp170503					
	3	Bp170504					
	4	Bp170505					
	5	Bp170506					
	6	Bp170507					
	7	Bp170508					
	8	Bp170509					
	9	Bp170510					
	10	Bp170511					
	11	Bp170512					
	12	Bp170513					
	13	Bp170514					
	14	Bp170515					
	15	Bp170516					
	16	Bp170517					
	17	Bp170518					
	18	Bp170519					
	19	Bp170520					
	20	Bp170521					
	21	Bp170522					

Showing the 5th record of the dataset

In [16]: dd.iloc[5]
Out[16]: Reg.No Bp170506
Student_Name Gender Male
IOT 40
CC 90
DOS 67
ASP 54
MA 90
Practical 1 90
Practical 2 90
Name: 5, dtype: object

Selecting the particular attributes and there corresponding values:

```
In [20]: dd.iloc[[2,3,5,25],[0,1,2]]

Out[20]:

2 Bp170503 Deepika Female

3 Bp170504 Malini Female

5 Bp170506 Muralidharan Male

25 Bp170526 Mary Female
```

Selecting and showing first 5 records:

In [17]:	dd	[0:5]									
Out[17]:		Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	МА	Practical 1	Practical 2
	0	Bp170501	Vasantha kumari	Female	70	69	87	68	56	89	86
	1	Bp170502	Sugashini	Female	48	79	89	78	72	78	87
	2	Bp170503	Deepika	Female	50	98	78	86	89	89	90
	3	Bp170504	Malini	Female	60	86	98	94	92	90	90
	4	Bp170505	Lokesh	Male	78	89	74	74	87	81	87

Selecting the 25th record and showing the 1st values of the record:

```
In [44]: dd.iat[25,1]
Out[44]: 'Mary'
```

```
In [27]: dd.index
Out[27]: RangeIndex(start=0, stop=26, step=1)
```

Selecting and showing the columns of the trained records.

Showing the entire data of the whole record without the attribute:

In [29]:	dd.values	
Out[29]:	array([['Bp170501', 86],	'Vasantha kumari', 'Female', 70, 69, 87, 68, 56, 89,
	['Bp170502',	'Sugashini', 'Female', 48, 79, 89, 78, 72, 78, 87],
	['Bp170503',	'Deepika', 'Female', 50, 98, 78, 86, 89, 89, 90],
	['Bp170504',	'Malini', 'Female', 60, 86, 98, 94, 92, 90, 90],
	['Bp170505',	'Lokesh', 'Male', 78, 89, 74, 74, 87, 81, 87],
	['Bp170506',	'Muralidharan', 'Male', 40, 90, 67, 54, 90, 90, 90],
	['Bp170507',	'Carmel', 'Male', 33, 89, 45, 89, 87, 76, 87],
	['Bp170508',	'Pavithra 1', 'Female', 87, 78, 43, 85, 78, 87, 88],
	['Bp170509',	'Chinnapa raj', 'Male', 20, 78, 23, 23, 81, 77, 80],
	['Bp170510',	'Pavithra 2', 'Female', 70, 69, 87, 68, 56, 89, 86],
	['Bp170511',	'Stephy', 'Female', 60, 79, 89, 78, 72, 78, 87],
	['Bp170512',	'Shilpa', 'Female', 78, 98, 78, 86, 89, 89, 90],
	['Bp170513',	'Jasmin', 'Female', 89, 86, 23, 94, 23, 90, 90],
	['Bp170514',	'Ammu', 'Female', 78, 89, 74, 74, 87, 81, 87],
	['Bp170515',	'Vidhiya', 'Female', 89, 65, 54, 76, 23, 90, 90],
	['Bp170516',	'Sharadha', 'Female', 76, 89, 67, 89, 87, 76, 87],
	['Bp170517',	'Gokul', 'Male', 65, 78, 76, 65, 78, 87, 88],
	['Bp170518',	'Lawrance', 'Male', 43, 78, 76, 88, 81, 77, 80],
	['Bp170519',	'Suganthi', 'Female', 23, 69, 87, 68, 56, 89, 86],
	['Bp170520',	'Pavithra 3', 'Female', 65, 79, 89, 78, 72, 78, 87],
	['Bp170521',	'Murugan', 'Male', 87, 98, 78, 86, 89, 89, 90],
	['Bp1/0522',	Prabavathi, Female, 89, 86, 98, 56, 92, 90, 90],
	['Bp170523',	Swetna, 'Hemale', 78, 89, 74, 74, 87, 81, 87],
	['Bp170524',	'Isnvariya', 'Female', 56, 90, 34, 23, 90, 90, 90],
	['Bp170525',	"Sriunar", "Hate", 70, 23, 67, 89, 87, 76, 87],
	['Bp1/0526',	Mary, remare, 05, 07, 05, 05, 05, 05, 90]],
	atype=object)	

Selecting the 26 rows and 10 columns:

In [30]: dd.shape
Out[30]: (26, 10)

Counting the number of records:

1. 1	ound method	DataFrame count	of	Per	No	e+,	Ident	Namo	Condor	TOT	<u></u>	DOS	AGD	MD	Practical	1 \	
0	Bp170501	Vasantha kumari	Female	70 xey.	69	87	68	_14ame	Gender	89	00	003	ADE	PIPI	FIACCICAL	± (
1	Bp170502	Sugashini	Female	48	79	89	78	72		78							
2	Bp170503	Deepika	Female	50	98	78	86	89		89							
3	Bp170504	Malini	Female	60	86	98	94	92		90							
4	Bp170505	Lokesh	Male	78	89	74	74	87		81							
5	Bp170506	Muralidharan	Male	40	90	67	54	90		90							
6	Bp170507	Carmel	Male	33	89	45	89	87		76							
7	Bp170508	Pavithra 1	Female	87	78	43	85	78		87							
8	Bp170509	Chinnapa raj	Male	20	78	23	23	81		77							
9	Bp170510	Pavithra 2	Female	70	69	87	68	56		89							
10	Bp170511	Stephy	Female	60	79	89	78	72		78							
11	Bp170512	Shilpa	Female	78	98	78	86	89		89							
12	2 Bp170513	Jasmin	Female	89	86	23	94	23		90							
13	Bp170514	Ammu	Female	78	89	74	74	87		81							
14	Bp170515	Vidhiya	Female	89	65	54	76	23		90							
15	5 Bp170516	Sharadha	Female	76	89	67	89	87		76							
16	5 Bp170517	Gokul	Male	65	78	76	65	78		87							
17	7 Bp170518	Lawrance	Male	43	78	76	88	81		77							
18	Bp170519	Suganthi	Female	23	69	87	68	56		89							
19	Bp170520	Pavithra 3	Female	65	79	89	78	72		78							
20) Bp170521	Murugan	Male	87	98	78	86	89		89							
21	Bp170522	Prabavathi	Female	89	86	98	56	92		90							
22	2 Bp170523	Swetha	Female	78	89	74	74	87		81							
23	Bp170524	Ishvariya	Female	56	90	34	23	90		90							
24	Bp170525	Sridhar	Male	78	23	67	89	87		76							
25	5 Bp170526	Mary	Female	89	87	85	89	89		89							

In [32]: dd.size

Out[32]: 260

Showing the records in descending order:

In [33]: dd.iloc[::-1]

. –												
-		Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	ма	Practical 1	Practical 2	
2	5	Bp170526	Mary	Female	89	87	85	89	89	89	90	l
24	4	Bp170525	Sridhar	Male	78	23	67	89	87	76	87	
23	3	Bp170524	Ishvariya	Female	56	90	34	23	90	90	90	
22	2	Bp170523	Swetha	Female	78	89	74	74	87	81	87	
2	1	Bp170522	Prabavathi	Female	89	86	98	56	92	90	90	
20	0	Bp170521	Murugan	Male	87	98	78	86	89	89	90	
19	9	Bp170520	Pavithra 3	Female	65	79	89	78	72	78	87	
18	8	Bp170519	Suganthi	Female	23	69	87	68	56	89	86	
17	7	Bp170518	Lawrance	Male	43	78	76	88	81	77	80	
10	6	Bp170517	Gokul	Male	65	78	76	65	78	87	88	
1	5	Bp170516	Sharadha	Female	76	89	67	89	87	76	87	
14	4	Bp170515	Vidhiya	Female	89	65	54	76	23	90	90	
13	3	Bp170514	Ammu	Female	78	89	74	74	87	81	87	
12	2	Bp170513	Jasmin	Female	89	86	23	94	23	90	90	
1	1	Bp170512	Shilpa	Female	78	98	78	86	89	89	90	
1	0	Bp170511	Stephy	Female	60	79	89	78	72	78	87	
-	9	Bp170510	Pavithra 2	Female	70	69	87	68	56	89	86	
1	8	Bp170509	Chinnapa raj	Male	20	78	23	23	81	77	80	
7	7	Bp170508	Pavithra 1	Female	87	78	43	85	78	87	88	
-	6	Bp170507	Carmel	Male	33	89	45	89	87	76	87	
4	5	Bp170506	Muralidharan	Male	40	90	67	54	90	90	90	
4	4	Bp170505	Lokesh	Male	78	89	74	74	87	81	87	
:	3	Bp170504	Malini	Female	60	86	98	94	92	90	90	
- 1	2	Bp170503	Deepika	Female	50	98	78	86	89	89	90	
	1	Bp170502	Sugashini	Female	48	79	89	78	72	78	87	

In [34]: dd.iloc[::-1].cummax()

-	
(h + + -	1 2 4 1
_	

	Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	ΜΑ	Practical 1	Practical 2
25	Bp170526	Mary	Female	89	87	85	89	89	89	90
24	Bp170526	Sridhar	Male	89	87	85	89	89	89	90
23	Bp170526	Sridhar	Male	89	90	85	89	90	90	90
22	Bp170526	Swetha	Male	89	90	85	89	90	90	90
21	Bp170526	Swetha	Male	89	90	98	89	92	90	90
20	Bp170526	Swetha	Male	89	98	98	89	92	90	90
19	Bp170526	Swetha	Male	89	98	98	89	92	90	90
18	Bp170526	Swetha	Male	89	98	98	89	92	90	90
17	Bp170526	Swetha	Male	89	98	98	89	92	90	90
16	Bp170526	Swetha	Male	89	98	98	89	92	90	90
15	Bp170526	Swetha	Male	89	98	98	89	92	90	90
14	Bp170526	Vidhiya	Male	89	98	98	89	92	90	90
13	Bp170526	Vidhiya	Male	89	98	98	89	92	90	90
12	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
11	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
10	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
9	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
8	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
7	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
6	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
5	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
4	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
3	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
2	Bp170526	Vidhiya	Male	89	98	98	94	92	90	90
1	Bp170526	Vidhiva	Male	89	88	98	94	92	90	90

In [35]: dd.iloc[::-1].cummin()

```
Out[35]:
```

	Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	ма	Practical 1	Practical 2
25	Bp170526	Mary	Female	89	87	85	89	89	89	90
24	Bp170525	Mary	Female	78	23	67	89	87	76	87
23	Bp170524	Ishvariya	Female	56	23	34	23	87	76	87
22	Bp170523	Ishvariya	Female	56	23	34	23	87	76	87
21	Bp170522	Ishvariya	Female	56	23	34	23	87	76	87
20	Bp170521	Ishvariya	Female	56	23	34	23	87	76	87
19	Bp170520	Ishvariya	Female	56	23	34	23	72	76	87
18	Bp170519	Ishvariya	Female	23	23	34	23	56	76	86
17	Bp170518	Ishvariya	Female	23	23	34	23	56	76	80
16	Bp170517	Gokul	Female	23	23	34	23	56	76	80
15	Bp170516	Gokul	Female	23	23	34	23	56	76	80
14	Bp170515	Gokul	Female	23	23	34	23	23	76	80
13	Bp170514	Ammu	Female	23	23	34	23	23	76	80
12	Bp170513	Ammu	Female	23	23	23	23	23	76	80
11	Bp170512	Ammu	Female	23	23	23	23	23	76	80
10	Bp170511	Ammu	Female	23	23	23	23	23	76	80
9	Bp170510	Ammu	Female	23	23	23	23	23	76	80
8	Bp170509	Ammu	Female	20	23	23	23	23	76	80
7	Bp170508	Ammu	Female	20	23	23	23	23	76	80
6	Bp170507	Ammu	Female	20	23	23	23	23	76	80
5	Bp170506	Ammu	Female	20	23	23	23	23	76	80
4	Bp170505	Ammu	Female	20	23	23	23	23	76	80
3	Bp170504	Ammu	Female	20	23	23	23	23	76	80
2	Bp170503	Ammu	Female	20	23	23	23	23	76	80
1	Bp170502	Ammu	Female	20	23	23	23	23	76	80

```
In [37]: print(dd.shape)
```

(26, 10)

Showing the values of the particular attribute of the dataset:

Showing the 5 fields of the attributes:

In [19]: dd.iloc[:,0:5]

Out[19]:

	Reg.No	Student_Name	Gender	ют	сс
0	Bp170501	Vasantha kumari	Female	70	69
1	Bp170502	Sugashini	Female	48	79
2	Bp170503	Deepika	Female	50	98
3	Bp170504	Malini	Female	60	86
4	Bp170505	Lokesh	Male	78	89
5	Bp170506	Muralidharan	Male	40	90
6	Bp170507	Carmel	Male	33	89
7	Bp170508	Pavithra 1	Female	87	78
8	Bp170509	Chinnapa raj	Male	20	78
9	Bp170510	Pavithra 2	Female	70	69
10	Bp170511	Stephy	Female	60	79
11	Bp170512	Shilpa	Female	78	98
12	Bp170513	Jasmin	Female	89	86
13	Bp170514	Ammu	Female	78	89

Showing the IOT subject students mark details:



Showing the DOS subject students mark details:





Showing the ASP.NET subject students mark details:



Showing the MOBILE APPLICATION subject students mark details:



Predicting the null values:



Making the cross table function:

t[58]:	Reg.No	Bp170501	Bp170502	Bp170503	Bp170504	Bp170505	Bp170506	Bp170507	Bp170508	Bp170509	Bp170510	 Bp170517	Bp170518	Bp170519	Bp1
	Gender			-	-	-	-		-			-			-
	Female	1	1	1	1	0	0	0	1	0	1	 0	0	1	
	Male	0	0	0	0	1	1	1	0	1	0	 1	1	0	
	2 rows x	26 column	s												
	(20 00101111	0												2



Representing the number of male and female in bar chart:



Marks greater than 50 in IOT subject:

In [62]: 1 dd[dd.IOT > 50]

1:		Reg.No	Student_Name	Gender	ют	сс	DOS	ASP	ма	Practical 1	Practical 2	
	0	Bp170501	Vasantha kumari	Female	70	69	87	68	56	89	86	
	3	Bp170504	Malini	Female	60	86	98	94	92	90	90	
	4	Bp170505	Lokesh	Male	78	89	74	74	87	81	87	
	7	Bp170508	Pavithra 1	Female	87	78	43	85	78	87	88	
	9	Bp170510	Pavithra 2	Female	70	69	87	68	56	89	86	
	10	Bp170511	Stephy	Female	60	79	89	78	72	78	87	
	11	Bp170512	Shilpa	Female	78	98	78	86	89	89	90	
	12	Bp170513	Jasmin	Female	89	86	23	94	23	90	90	
	13	Bp170514	Ammu	Female	78	89	74	74	87	81	87	
	14	Bp170515	Vidhiya	Female	89	65	54	76	23	90	90	
	15	Bp170516	Sharadha	Female	76	89	67	89	87	76	87	
	16	Bp170517	Gokul	Male	65	78	76	65	78	87	88	
	19	Bp170520	Pavithra 3	Female	65	79	89	78	72	78	87	
	20	Bp170521	Murugan	Male	87	98	78	86	89	89	90	
	21	Bp170522	Prabavathi	Female	89	86	98	56	92	90	90	
	22	Bp170523	Swetha	Female	78	89	74	74	87	81	87	
	23	Bp170524	Ishvariya	Female	56	90	34	23	90	90	90	
	24	Bp170525	Sridhar	Male	78	23	67	89	87	78	87	
	25	Bp170526	Mary	Female	89	87	85	89	89	89	90	

Marks greater than 50 in DOS subject:

In [63]:	1	dd[dd.I	00S > 50]								
Out[63]:											
		Reg.No	Student_Name	Gender	IOT	CC	DOS	ASP	MA	Practical 1	Practical 2
	0	Bp170501	Vasantha kumari	Female	70	69	87	68	56	89	86
	1	Bp170502	Sugashini	Female	48	79	89	78	72	78	87
	2	Bp170503	Deepika	Female	50	98	78	86	89	89	90
	3	Bp170504	Malini	Female	60	86	98	94	92	90	90
	4	Bp170505	Lokesh	Male	78	89	74	74	87	81	87
	5	Bp170506	Muralidharan	Male	40	90	67	54	90	90	90
	9	Bp170510	Pavithra 2	Female	70	69	87	68	56	89	86
	10	Bp170511	Stephy	Female	60	79	89	78	72	78	87
	11	Bp170512	Shilpa	Female	78	98	78	86	89	89	90
	13	Bp170514	Ammu	Female	78	89	74	74	87	81	87
	14	Bp170515	Vidhiva	Female	80	85	54	78	23	00	90
	15	Bp170518	Sharadha	Female	78	20	87	20	20	78	97
	16	Bp170517	Gakul	Male	85	70	78	85	70	97	0,
	47	D=170510	- Cokar	Male	40	70	70	00	70		00
	17	Bp170518	Lawrance	male	43	/8	/0	80	51	//	08
	18	вр1/0519	Suganthi	remale	23	69	87	68	50	89	86
	19	Bp1/0520	Pavithra 3	Female	65	79	89	78	72	78	87
	20	Bp170521	Murugan	Male	87	98	78	86	89	89	90
	21	Bp170522	Prabavathi	Female	89	86	98	56	92	90	90
	22	Bp170523	Swetha	Female	78	89	74	74	87	81	87
	24	Bp170525	Sridhar	Male	78	23	67	89	87	78	87
	25	Bp170526	Mary	Female	89	87	85	89	89	89	90

CHAPTER – V CONCLUSION

5.1 Conclusion

- This project has been successfully developed and interpreted and system was developed according to the user requirement.
- The system produces accurate result and it also reduces a lot of overheads, which the manual system faced.
- The system was thoroughly tested according to the testing methodologies.
Natural Language Processing

by S Sagayaraj

Submission date: 02-Apr-2019 12:32PM (UTC+0530) Submission ID: 1104361234 File name: BP170522_Document_classification.pdf (308.95K) Word count: 3753 Character count: 20051

DOCUMENT CLASSIFICATION WITH NATURAL LANGUAGE PROCESSING

A Software Project

²⁶Submitted in partial fulfillment for the award of the degree of

MASTER OF COMPUTER SCIENCE

By

PRABAVATHI P

(REG. NO: BP170522)

⁵ Under the Guidance of

Mrs. M. POOVIZHI. M.C.A., M.Phil,



PG DEPARTMENT OF COMPUTER SCIENCE

SACRED HEART COLLEGE (AUTONOMOUS)

TIRUPATTUR, VELLORE DT - 635 601

APRIL-2019

CERTIFICATE

This is to certify that the project work entitled "DOCUMENT CLASSIFICATION WITH NATURAL LANGUAGE PROCESSING" is submitted to Sacred Heart College (Autonomous), Tirupattur-635 601, Vellore District by PRABAVATHI P (REG.NO: 17 BP170522) for the partial fulfillment for the award of the Degree of Master of Science in Computer Science is a bonafied record of the work carried out by him, under my guidance and supervision.

Signature of the Project Guide

Mrs. M.Poovizhi, M.C.A., M.Phil.

Submitted for Viva- Voice examination held on

Internal Examiner

Internal Examiner

1.

2.

ACKNOWLEDGEMENT

I thank God Almighty for his blessings and graces by which I was able to complete project work successfully.

I sincerely thank my parents who have given the gift of life to me to attain many achievements.

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I would like to thank the entire teaching and non-teaching staff members, Department of Computer Science for their help to complete the project successfully.

⁵ Finally, I thank each and every one of my friends, especially who have assisted me in completing the project work.

Document Classification with Natural Language Processing

ABSTRACT:

Text Document classification is the process of assigning the class labels to the text documents in the training model it's used by the large enough collection. It's a more multiplex algorithm to deal with the known attributes and also a limited set of possible values for the each set of attribute. Preprocessing and the representation of the reduced data is the essential for obtaining the effective result of text classification. Text classification is the act of the assigning classes to the text documents. This project helps to analyze and test the dataset with trained dataset model. Natural Language analysis is followed by the statistical analysis process. It's the process to provide the size reduction of the document and also improve the performance of the classification without the compromising accuracy.

INTRODUCTION:

In the world we have the huge amount of data on the web it's randomly increased day by day. All the huge information of the data is valuable and also it contain most of the information is text. So it becomes a very complex or a challenge for the humans to identify the relevant information. In this project document classification is helps to overcome these complex problems.

Text document classification is the dividing the set of inputs for the document into the two /more classes in the each document belong to the one / multiple classes. Text classification is maybe manual / automated. The manual classification helps to consume the time and high accuracy. The automated classification process is the fast and more efficient it's automatically classify the document.

SCOPE:

The scope of this project is used to bring the outcome of the document details regarding the document count and running time. It can be classify them into different norms and category by the classifying the dataset. It can be predict the different type of possibilities of various documents. We can predict the document title and total number of text file will be calculate and generating the graph. Finally, it can also plot the outcome predicted values in histogram and can perform them into the graphical representations.

PROBLEM STATEMENT:

In this project the major problem for all the data are manual document and calculating time for the database due to the time duration is to long huge man power is involved so that the new system in this project evaluates outcomes of search problem in this project.

Product Position Statement:

For	IT Industry
Who	IT Members
That	It is more flexible
Our Product	Provide up-to-date information and easy to maintain all the data. An classify them and make
	prediction and represent them in graphical representation manner.

SYSTEM ANALYSIS

System Study:

Introduction:

Document Classification this is the application for predicting the document details, by using the python language. In this project is to analysis made of every document to analyzing the data. It should be calculating the text files and also calculating the times of seconds. We can also represent the predicted results of the document in graphical representation by plotting in bar diagram, different kind of plotting method. The main purpose of this project is to identify the document and analyzing data of the text file and also calculating the time and also base time of the graph will be generate.

Overview:

Analysis of the text document files is analyzing and predicting by Naïve-Bayes in Natural Language Processing. This is a predicting application of the document file details is using python language. In this project is to analyze the every document and it should be calculating the data and also calculating the number of data files in the particular document and display the graph based on the running times. So that the proposed the new way is representing the data into graphical representations by which the graphs are able to identify the results easily.

It can also represent the predicted result of the document in graphical representation by plotting diagram and also different kind of plotting method. The main purpose of this project is to identify the document data and text file calculation and also generating the graph based on the text file format.

Technical Feasibility:

In this project it can strongly says that it is technically feasible. So there will not be much difficulty in getting required resources for the development and maintaining the system as well. All the resources of the organization needed for the development of the software as well as the maintenance of the same is available in the organization here we are utilizing the resources which are available already.

Economical Feasibility:

Development of this application is highly economically feasible. The resources of the organization needed not spend much money for the development of the system already available. The only thing is to be done is making an environment for the development with the effective supervision. If we are doing so, we can increase the maximum usability of the corresponding resources. Even after the development process, the organization will not be in condition to invest more in the organization. Therefore the system is economically feasible.

Classification details of this project:

Purpose:

To produce the complete details about the classification and prediction

Overview:

The user, who wants to know about the prediction of any case, can easily get the details in graphical representations.

Entry Criteria:

The users can directly import the data set, train and test them to get predictions.

Input:

Here, the input is test data from the data set.

Steps Solved:

Import text file into the python and perform the code form train and test using classification algorithms and make predict the desired output.

Output:

The required detail is shown to the knowledge seekers.

SYSTEM CONFIGURATION

REQUIREMENT SPECIFICATION:

29 Hardware Requirements:

Intel i3 processor
4 GB
500 GB
Keyboard, mouse, monitor

Front End	Python 3.7 (Jupyter Notebook)
Document	Microsoft Word 2007

NATURAL LANGUAGE PROCESSING:

Natural Language Processing is the subfield of the computer science and the artificial intelligence. It's alarmed with the communications between the computer and creature language. It's to process and analyze the large amount of data in Natural Language.

Natural Language Processing is used to apply the machine learning algorithms to text and the speech recognition/categorization. The development of the Natural Language Processing is the challenging process. Because computers are traditionally require the humans to 'speak' to them in to a programming language and it is highly structured or a limited number of voice commands.

NLP is used to interpret the free text and its make analyzable. So there is a large amount of information is stored in the free text files, like a business records and the medical records. NLP allows the analyst to shift the massive troves of free text to find the relevant information of the files.

WORK FLOW OF NATURAL LANGUAGE PROCESSING:

Current approach of natural language processing is based on the Deep Learning. This approach is a type of Artificial Intelligence and its uses the pattern in the data to improve the programs to understanding. Deep Learning model is required a massive amount of labeled data is to train and identify the relevant categorization Earlier NLP is involved a more rules based approaches, where it's a simpler machine learning algorithms.

²MPORTANCE OF NLP:

The benefit of the natural language processing can be considering the following statement that is cloud computing insurance should be the part of every service level agreement. These types of agreements that appear in frequently in the human language and the machine learning an algorithm is historically have bad interpreting. But the improvement of Deep Learning and Artificial Intelligence algorithms is effectively interpreted. Advances of Natural Language processing make it is easy to analyze and learn a greater range of the data sources.

TEXT CLASSIFICATION PROCESS:

Text classification process it contains various sub phases, each phase hast its own importance and needs.

Data collection

Pre-processing phase

Feature Extraction

Feature Selection

Building a classifier

Performance Evaluation



DATA COLLECTION:

The first step of classification process is the corpus building it consist of collect the dissimilar formats of the documents like a HTML, PDF, DOC, WEB substance and so on. so these documents are used to during the training and testing classifier.

PRE-PROCESSING:

Pre-processing is used to represents the text documents into the clear word format. The documents are prepared for the next step in the classification processes are represented by the great amount of features.

FEATURES:

Tokenization:

A document is treated as the string and then it is partitioned into the list of tokens.

Removing stop words:

Stopping words are 'the', 'a', 'and', and etc. these are regularly occur. So that the insignificant words are need to be removed.

Stemming word:

Stemming Algorithm it converts the different words from into the similar canonical forms. So these steps are the process of conflating tokens into their root form. Eg: Connection to connect and computing to compute.

FEATURE EXTRACTION:

The document classification used in the individual of the pre-processing performance that's used to decrease the convolution of the documents and it construct them into easier to switch, the documents have to be distorted into the full text edition to a certificate vector. The unanimously using document exhibition is called the vector space model. So the credentials are representing by the vectors of terms.

Restrictions:

- 1. Towering dimensionality of the document demonstration.
- 2. Failure of association with the flanking of the terms.
- 3. Thrashing of semantic relationships to facilitate is exists along with the expressions into a certificate.

FEATURE SELECTION:

After completing pre-processing and the feature extraction that is significant march of the transcript taxonomy is aspect variety to assemble the vector liberty, which is improve the scalability, competence and precision of the transcript organization. The foremost objective of the aspect variety is to decide on the separation of features from the creative documents. Feature selection is perform to maintenance the terms into the utmost achieve according to the predetermined measurement substance of the words.

Major problem of the content categorization is the towering dimensionality of the aspect liberty. So the several aspect assessment metrics cover be noticed into the information gain, term frequency and the chi-square, expected cross entropy and odds ratio, mutual information, the weight of the evidence of the text and finally Gini index.

CLASSIFICATION:

The automatic classification of the documents is the predefined categories has been observed an active attention of the documents can be classified into three ways.

Unsupervised classification

Supervised classification

Semi supervised categorization method.

The task of the automatic classification is extensively studied and the rapid progress in this area, including the approaches of the machine learning.

Naïve Bayes Classifier

Decision Tree

K-Nearest Neighbor 3 Support vector machines

Neural Networks

PERFORMANCE EVALUATION:

The performance of the text classification system is evaluated using four metrics

Accuracy

Precision

Recall

FI measure

10

Precision is measure the exactness of the classifier. The higher precision is less false positives, while the lower precision is more false positives.

Precision = Number of correct extracted text

Total number of extracted texts

Recall is measures the completeness or the sensitivity of the classifier. Higher recall is the lass false negatives and lower recall is more false negatives

Recall= Number of correct extracted texts

Total no of annotated texts

Precision and recall is combined to produce the single metrics that is known as F-measure and which is represents the weighted harmonic of precision and recall. The advantage of the F-

F-measure= 2* precision * recall

Precision + recall

Accuracy is the overall degree which is instances have been correctly classified into the formula, Accuracy= Number of correctly classified instances

Total number of instances

Approaches of Text Classification Task:

In the literature survey is done in the two types of techniques.

- Supervised learning
- Unsupervised learning

Naïve Bayes Classifier:

Naïve Bayes Classifier is the simple and probabilistic classifier is fully based on the applying Bayes theorem with the strong independent assumption. One of the main Naïve Bayes model is work well for the text domain because it is the evidences of the 'Vocabularies' and 'words' are appearing in the text and the size of the words is typically used in the range of thousand.

The huge size of the vocabularies makes the Naïve Bayes Model is works well for the text document classification problem. The Naïve Bayes Classifier model is widely used algorithm for the purpose of the text document classification.

A Basic Naïve Bayes Model:

The most basic models for the text classification is Naïve Bayes Model. The Naïve Bayes classification model will predicts the document topic,

 $Y=\{c1,c2,...,c20\}$

Where Ck is the class otherwise topic based on the document features $x \in \mathbb{N}p$, and p denotes the number of items in bag of words list.

The feature vector contains count or length of xi for the tf-idf value of the i-th term in the bag of words list.

$$\mathbf{x} = \begin{bmatrix} x_1, x_2, \dots, x_p \end{bmatrix}$$

Using Bayes theorem can develop a model to predict the topic class (ck) of a document its feature vector x.

The Naïve Bayes model makes the "Naive" assumption the probability of each term's tf-idfis conditionally independent of every other term. This reduces the conditional probability function to the product

$$P\left(x_1, \dots, x_p \mid C_k\right) = \prod_{i=1}^p P\left(x_i \mid C_k\right)$$

Moreover Bayes theorem for classification of problem becomes,

$$P\left(C_{k} \mid x_{1}, \dots, x_{p}\right) = \frac{P(C_{k}) \prod_{i=1}^{p} P\left(x_{i} \mid C_{k}\right)}{P\left(x_{1}, \dots, x_{p}\right)}$$

Also, Multi-nominal Naïve Bayes classifier using scikit learn commands are:

$$P(x_1, \dots, x_p \mid C_k) = \frac{(\sum_{i=1}^p x_i)!}{\prod_{i=1}^p x_i!} \prod_{i=1}^p p_{k,i}^{x_i}$$

Accuracy: 0.77389803505

Scikit learn Pipelines:

In Scikit-learn pipelines are a sequence of transforms followed by a final estimator. Intermediate steps within the pipeline must be 'transforms', they must implement fit and transform methods. The Count Vectorizer and tfidf Transformer are used as transformers in above example. The final estimator of the pipeline is only needs to implement the fit method. It sees the simplicity of pipelines by using it to re-implement our above analysis using the Naive Bayes model:

Accuracy: 0.77389803505

An Improved Naive Bayes Model:

In this model it seems to improve by removing stop words which are common words in the english language and do not add any information into the text. These includes words such as, "the", "at", "is", etc. Finally can remove them in the CountVectorizer constructor call,

Accuracy: 0.831651619756

A less improvement, but an improvement is none-the-less. In the Naive Bayes classifier can be fast compared to more sophisticated methods due to the decoupling of the class conditional feature distributions, i.e.

 $P\left(x_{1}, \dots, x_{p} \mid C_{k}\right) = \prod_{i=1}^{p} P\left(x_{i} \mid C_{k}\right)$

16

The decoupling is the class conditional distributions it allows for each distribution to be independently estimated as a one dimensional distribution and helps to alleviate problems with the curse of dimensionality.

Implementation of Navies-Bayes to an improved Navies-Bayes Algorithm:

To obtain the training and testing sets directly with the following commands:

from sklearn.datasets import fetch_20newsgroups

twenty_train= fetch_20newsgroups(subset='train', shuffle=True)

twenty_test = fetch_20newsgroups(subset='test', shuffle=True)

To view the total number of articles:

len(twenty_train.data) + len(twenty_test.data)

All the documents contain the data set below the following 20 topics,

twenty_train.target_names

Result:

['alt.atheism','comp.graphics','comp.os.ms,'windows.misc'

'comp.sys.ibm.pc.hardware','comp.sys.mac.hardware','comp.windows.x','misc.forsale','rec.autos',' rec.motorcycles','rec.sport.baseball','rec.sport.hockey','sci.crypt','sci.electronics','sci.med','sci.spac e','soc.religion.christian','talk.politics.guns','talk.politics.mideast','talk.politics.misc','talk.religion. misc']

To particularly view at the actual message within the documents,

9
print("\n".join(twenty_train.data[0].split("\n")))

Result: From: lerxst@wam.umd.edu (where's my thing) Subject: WHAT car is this!? Nntp-Posting-Host: rac3.wam.umd.edu Organization: University of Maryland, College Park Lines: 15

It is wondering if anyone out there could enlighten on this car It seems

the other day. It was a 2-door sports car, 60s/ delayed, early 70s. It was called a Bricklin. The doors were really small. In insertion, the front bumper was separate from the rest of the body.

11 Thanks,

- IL -brought to you by your neighborhood Lerxst -

Using a categorical variable they are encoding the target classes

twenty_train.target[0]

Result: 7

A Basic Naive Bayes Model :

To instantiate a multinomial Naive Bayes classifier using the Scikit-learn library and fit it to our tf-idftf-idf matrix using the commands, 77 from sklearn.naive_bayes import MultinomialNB from sklearn.metrics import accuracy_score mod = MultinomialNB() mod.fit(X_train_tfidf, twenty_train.target)

Result:

MultinomialNB(alpha=1.0, class_prior=None, fit_prior=True)

The word is alpha=1 means it is using Laplace smoothing. Otherwise, now see the accuracy of classifier using Scikit-learns accuracy_score function:

20 X_test_tf = count_vect.transform(twenty_test.data) X_test_tfidf = tfidf_transformer.transform(X_test_tf) predicted = mod.predict(X_test_tfidf)

print("Accuracy:", accuracy_score(twenty_test.target, predicted))

Result:

Accuracy: 0.77389803505

To plot a graph for Navies-Bayes Algorithm:

import matplotlib.pyplot as plt_

x=["alt.atheism", "comp.graphics","comp.os.mswindows.misc","comp.sys.ibm.pc.hardware","comp.sys.mac.hardware","comp.windows.x", "misc.forsale","rec.autos","rec.motorcycles","rec.sport.baseball","rec.sport.hockey","sci.crypt"," sci.electronics","sci.med","sci.space","soc.religion.christian","talk.politics.guns","talk.politics.mi deast","talk.politics.misc","talk.religion.misc"]

y=[0.80, 0.78, 0.79, 0.68, 0.86, 0.87, 0.87, 0.88, 0.93, 0.91, 0.88, 0.75, 0.84, 0.92, 0.82, 0.62, 0.66, 0.95, 0.94, 0.95] y1=[0.69, 0.72, 0.72, 0.81, 0.81, 0.78, 0.80, 0.91, 0.96, 0.92, 0.98, 0.96, 0.65, 0.79, 0.94, 0.96, 0.95, 0.94, 0.95, 0.24] y2=[0.74, 0.75, 0.75, 0.74, 0.84, 0.82, 0.83, 0.90, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.94, 0.96, 0.95, 0.94, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.95, 0.93, 0.94, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.95, 0.93, 0.94, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.95, 0.93, 0.94, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.95, 0.93, 0.94, 0.95, 0.92, 0.93, 0.84, 0.74, 0.85, 0.88, 0.76, 0.78, 0.94, 0.95, 0.93, 0.94, 0.95, 0.92, 0.93, 0.94, 0.95, 0.95, 0.92, 0.93, 0.94, 0.95, 0.95, 0.92, 0.93, 0.95, 0.92, 0.93, 0.95, 0.92, 0.93, 0.94, 0.95, 0.95, 0.92, 0.93, 0.95, 0.92, 0.93, 0.95, 0

plt.subplot(2,2,1) plt.subplots_adjust(hspace=5) plt.subplots_adjust(wspace=10) plt.plot(x,y,'gv-',x,y,'rv-') plt.title("Precision") plt.xlabel("Model") plt.ylabel("Accuracy") plt.subplot(2,2,2)

plt.subplots_adjust(hspace=1) plt.subplots_adjust(wspace=1) plt.plot(x,y,'gv-',x,y1,'rv-') plt.title("Re call") plt.xlabel("Model") plt.ylabel("Accuracy") plt.subplot(2,2,3) plt.subplots_adjust(hspace=1) plt.subplots_adjust(wspace=1) plt.plot(x,y,'g^',x,y2,'r^') plt.title("F1-Score") plt.xlabel("Model") plt.ylabel("Accuracy")

plt.subplot(2,2,4) plt.subplots_adjust(hspace=1) plt.subplots_adjust(wspace=1) plt.plot(x,y,'gv-',x,y3,'rv-') 19 plt.title("Support") plt.xlabel("Model") plt.ylabel("Accuracy") plt.savefig("C:/Users/malini/Desktop/graphnlp1.pdf") plt.show()

Result:



An improved Navies-Bayes model:

To look for an improved model by removing stop words which are common words in the english language and do not add any information into the text. These includes words such as, "the", "at", "is", etc. Also can remove them in the CountVectorizer constructor call,

pipe = Pipeline([('vect', CountVectorizer(stop_words='english')),

('tfidf', TfidfTransformer()),

('model', MultinomialNB()),])

mod = pipe.fit(twenty_train.data, twenty_train.target)

predicted = mod.predict(twenty_test.data)

print(classification_report(twenty_test.target,

predicted, target_names=twenty_test.target_names))

print("Accuracy:", accuracy_score(twenty_test.target, predicted))



Future Scope:

- One thing that did not address was the topic of stemming and lemmatisation, both do with reducing a word down to its base form and also can be used to improve the performance of text classification models.
- Lemmatisation differs from stemming topic because it depends only identifying the intended part of speech and produce meaning of a word in a sentence.
- Natural Language Tool Kit or NLTK and spaCy libraries and both provided by the Stemmers and lemmatizers.

Conclusion:

In this paper, it covers document classification using Scikit-learn and the 20 News Groups dataset. It went over the basics of term frequency-inverse document frequency, pipelines and the Naive Bayes classifier. While possibilities that it able to improve the accuracy of the basic Naive Bayes classifier from 77% to 83%, there are much more sophisticated models like a support vector machines it also achieve higher performance. One important thing that , it did not address the topic of stemming and lemmatisation, where both reducing a word down to its base form and it can be used to improve the performance of text classification models. In this model of Lemmatisation and stemming ,the stemming differs from it. Because it

Only depends on identifying the intended part of speech and meaning of a word in a sentence. Natural Language Tool Kit or NLTK and spaCy libraries and both provided by the Stemmers and lemmatizers



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K-Nearest classification

by S Sagayaraj

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TEXT CATEGORIZATION USING K-NEAREST NEIGHBOR

CLASSIFICATION

A Software Project

16 Submitted in partial fulfillment for the award of the degree of

MASTER OF COMPUTER SCIENCE

by

PAVITHRA.V (REG.NO:BP170523)

9 Under the Guidance of

Mr. R.DENIS. M.C.A., M.Phil,



PG DEPARTMENT OF COMPUTER SCIENCE

SACRED HEART COLLEGE (AUTONOMOUS)

TIRUPATTUR, VELLORE DT – 635 601

April- 2019

21 DECLARATION BY THE STUDENT

I hereby declare that the report entitled "TEXT CATEGORIZATION USING

K-NEAREST NEIGHBOR CLASSIFICATION " submitted by me to PG Department of Computer Science, Sacred Heart College, Tirupattur in partial fulfillment of the requirement for the award of the degree of M.Sc. in Computer Science is an 'Industrial Plant Training' undertaken by me under the Supervision of Department faculty. I further declare that the work in this report has not been submitted and will not be submitted, either in part or in full, for the award of any other degree or diploma in this institute or any other institute or university.

CERTIFICATE

This is to certify that the project work 22 ntitled "Text Categorization Using k-Nearest Neighbor Classification" is submitted to Sacred Heart College (Autonomous), Tirupattur-435601, Vellore District by PAVITHRA.V Reg. No: BP170523 for the partial fulfillment for the award of the Degree of Master of Science in Computer Science is a bona fide record of the work carried out by him, under my guidance and supervision.

Signature of the Project Guide

Mr. R. Denis, M.C.A., M.Phil.,

Submitted for Viva – Voice examination held on

Internal Examiner

Internal Examiner

1.

2.

1.1 Abstract

K-Nearest Neighbor is a popular algorithm for text categorization. In this project represents a document categorization using KNN. This method is to categorize the business, entertainment, politics, sports, and technology and training documents and it finds all the closest neighbors of the sample documents. Its focus the speed and quality of the document classification.

In this project shows the k-nearest neighbors enables to find the document categorization. In this project involves a solution of the similarity functions and previously finds the document categorization.

CHAPTER – I INRODUCTION

1.2 Project Overview

1.2.1 Introduction

Text Categorization can be classification of document different data set as the original documents dataset process a categories. As the volume of data information available, document classification on the intranets Text data and Text Classification is an different text document using.

- 1. K-Nearest Neighbor classification text categiruzation in an algorithm model. Document class approach level categorized Text.
- Text Classification to business document. five Types of data are using this of consumer files.
- 3. Different Text Categorization many documents five categories are worked.
- 4. Text used in K-NN algorithm work below text classified.
- K-NN Text Categorization is mainly concept of machine learning classifier document in java page.
1.2.2 Scope

The scope of the project is used to bring the outcome of the document details regarding the document count and running time. We can classify them into different norms and category by categories the dataset, we can predict different type of possibilities of various document. We predict document title total number of text file will be calculate and generating the document. Finally, we can also plot the outcome predicted values in histogram and we can perform them in the K-NN representations.

1.2.3 Problem Statement

A document classification, matrix formula are using in this projects Cosine Similarity. Non zero vectors assign to same document need to classified those data of this an inner product space that measures the matrix formula are using in **cosine** of the angle between them. The **cosine similarity** as a math function are calculated the seconds minimum time is particularly used in algorithm based to positive space calculate every second assign distance problem , where the document outcome is neatly bounded in this project.

This method is to categorize the business, entertainment, politics, sports, and technology and training documents. Using in K-Nearest Neighbor algorithm some examples can be classified documents in sports and politics same the test which are same in the content easily match in Text data, in our problem be eliminated for the same of efficiency. While categorizing documents, these terms in document categorization, Cosine Similarity both text document are KNN classify Test Document Thread in class using machine learning concepts. java Implementation of K-Nearest Neighbor Algorithm using in File cache.

Text Categorization data for BBC Dataset is under "data" folder. Here dataset import the data level in set of data in similarity between these documents.

1.2.4 Product position statement

For	IT Industry
Who	IT members.
That	It is more flexible.
Our Product	Provide up-to-date information and easy to maintain all the data. An categories them and make prediction and represent them in K-NN representation manner

SYSTEM ANALYSIS

CHAPTER - II SYSTEM ANALYSIS

2.1 System Study 2.1.1. Introduction

Text Categorization used in K-NN algorithm text document project is an k-Nearest Neighbor predicting algorithm of the document text details, by using the java language. In this the analysis are made of every document to analyzing the data it should be calculating the text documents. And also calculating times of the seconds We can also represent the predicted results of the document in K-NN representation by Text in bar diagram, different kind of Text method. The main purpose of the project is to identity the document and categories data of the text file and also calculating time .base of the time the text will be generate.

2.1.2 STATEMENT OF SOLUTION:-

Text categorization, also known as Text classification, is the performance of automatically classifying a dataset of text documents into five categories from a dataset, in this project, we first categorize the algorithms using K-NN Based machine learning approach and then return's the most relevant documents. We will explain the solution method give a K-NN algorithm to classification and analysis.

2.1.3. Overview

Analysis of document text documents and predicting them by K-Nearest Neighbor machine learning models, this is an predicting applications of the document file details, by using java language. In this the analysis of every document should be analyzed and calculating the data and calculating number of data files in the particular document and displayed Test based on the running times so that proposed the new way representing the data into Text representations by which the Text are able to identify the results easily. It can also represent the predicted results of the document in Text representation by K-NN diagram, and different kind of plotting method.

The main purpose of the project is to identity document data and text file calculation and running times and also generating Text based on the data file format.

2.1.3 Proposed System

• Representing the new way of present the data into the categorizing.

2.1.5. 1 ABOUT CLASSIFICATION DETAILS OF THE PROJECTS

Purpose

To produce the complete details about the classifications and Text categorizing.

Overview

The user, who wants to know about the prediction of any case, can easily get the details in classifications and Text categorizing.

Entry Criteria

The users can directly import the data set, train and test them to get predictions.

Input

Here the input is test data from the data set.

Steps involved

Import text data into the java and perform the code form train and test using classification algorithms and make predict the desired output.

Output

The required detail is shown to the knowledge seekers.

2.2 Feasibility Study 2.2.1 Introduction

Text Categorization can be classification of document As the volume of data information available, document classification on the intranets Import text data into the java and perform the code form train and test using classification algorithms and make predict the desired output.

2.2.1.1 Technical Feasibility:

In this, one can strongly says that it is technically feasible, solution based review documents and assign those documents verify quickly access catches few documents development software hardware specify info text documents calculated feasibility solution, since there will not be much difficulty in getting required resources for the development and maintaining the system as well. All the resources needed for the development of the software as well as the maintenance of the same is available in the organization here we are utilizing the resources which are available already.

2.2.1.2 Economical Feasibility

Development of this application is highly economically feasible .The organization needed not spend much money for the development of the system already available. The only thing is to be done is making an environment for the project application using available, and intranets volume of data default text development with an effective supervision. If we are doing so, we can attain the maximum usability of the corresponding resources .Even after the development, the organization will not be in condition to invest more in the organization. Therefore, the system is economically feasible.

SYSTEM CONFIGURATION

CHAPTER - III System Configuration

3.1 Requirement Specification

3.1.1 Hardware Requirements

Processor	: Intel i3 processor
RAM	: 2 GB
Hard Disk	: 500 GB
Input/Output	: Keyboard, mouse, monitor

3.1.2 Software Requirements

Front End	: Java (NetBeans 8.2)
Back End	: PDF Format
Documentation	: Microsoft Word 2013

3.3.2.1ALGOTITHM WORK'S

The algorithm classified in text document where identifying the details in K-NN algorithm. Text representation by K-NN diagram, and different kind of plotting method. The main purpose of the project is to identity document data and text file calculation and running times and also generating Text based on the data file format. "K-Nearest Neighbor classification" was introduced. Model used both document cache file representation. Comparing text document business, sports and text will be same finally machine learning methods are using manual system not using only machine language using. The system produces accurate result and it also reduces a lot of overheads, which the manual system faced. The system was thoroughly tested

according to the testing methodologies. Dataset is including the project Similarity (Document test, Document train Document).

K-NN algorithm The text sample may be classified either to document classified Either to the File Document Factory of assigned or to the second class of file Comparators of tanning data if k=150 it is assigned to the second document because there are more set of documents are available in this project of K-NN algorithm. The second file document class to compare the file cache document class are used. This is the inner circle. If k=150 is assigned the first document and second machine leaning of classifiers in project.

Nearest Neighbor classifier:-

The popular algorithm in nearest neighbor classifier is those training dataset is available one of algorithm that assigns a point of data to the document class of its neighbor in the feature space, that is size of training data set Documents completely in this algorithm is infinity, the one nearest neighbor algorithms classifier sports document and training text below guarantees an error rate of no worse than twice the algorithm minimum achievable error rate given the distribution of the data.

3.3.2. TEXT CATEGORIZATION :-

The machine leaning concept including this project until and file cache page are including and three type of machine learning files are using be the Document similarity methods the usage in search engines. This methods, have been using manual categorization but k-Nearest Neighbor methods like a Text java main page in cluing the string variables and calculus the letters text categorizations router and check the words of data levels. Text categorizations, approaches in KNN Classification Comparator approach categorizes the documents. text categorization meaning of training dataset including the BBC dataset added adds this project. Text classification probabilistic on work to text categorization

Since them, it has been clarify to dataset used for a many numbers of different applications, one. which here briefly review of the most important one. Note that boarders between more different class of applications listed here fully and some artificial, these may be considered special cases of others. Applications we do not explicitly discuss are text categorization by mean of a combination of text classifier and text document categorization through the analysis of textual captions in text, author identifications for literary texts of unknown or disputed authorship language identification for text identify for texts of unknown language automated identification of text genre of a project.

Automatic indexing for Text Categorization systems:-

The Text categorization system applications that has K-NN most of the early research in the field is that algorithm automatic document indexing classifier systems relying on a controlled dictionary, the most prominent example of which is that of document systems. In these latter each document is assigned one or more keywords or key phrases describing of content, where these document keywords and key phrases belong to a finite set called controlled dictionary, after consisting a cosine matrix hierarchical and text document belong to a finite set called controlled dictionary. Usually, this project assignment is done by trained Text documents and is thus a costly activity. If the entries in the controlled vocabulary are viewed as categories, text document is an instance of multi document may thus be Business, sports, polities by the automatic techniques described in this algorithm. Recalling note the text document note that this application may typically require that $k_1 < x > k_2$ value are assigned to each document, for given k1, k2. Document classifiers are assigned to each document-pivoted is probably the best option, so that new documents may be classified as they become available. Various text classifiers and explicitly to conceived for document indexing have been some text described in the literature five kinds of documents Automatic text categories with controlled and dictionaries is matrix functions related to automated data generation. In digital libraries one is usually interested in tagging documents by data this describe then under a variety of aspects data model, some of the data are in matrix functions their role is to describe that semantics of the document by means of K-Nearest Neighbor algorithm, keywords of key phrases. The generation of these data may thus be viewed as a problem identify of document indexing with controlled dictionary, and thus take in k-NN algorithm techniques.

3.3.2.3 DATASET:-

K-Nearest Neighbor is a popular algorithm for text categorization. In this project represents a document categorization using KNN. This method is to categorize the business, entertainment, politics, sports, and technology and training documents and it finds all the closest neighbors of the sample documents. Its focus the speed and quality of the document classification.

The system project and accurate result and it also reduces less of time saving and a lot of overheads, which the manual system faced. The system was high-level of performance need this project model thoroughly tested according to the testing methodologies.

In this project shows the k-nearest neighbors enables to find the document categorization. In this project involves a solution of the similarity functions and previously finds the document categorization. Filtering this document according to

3.3.2.4 Document organization:-

This organization of documents is critical to each space handle with area in business documents severally however is a lot of sequences but efface more project particularly important for corporations, firms and also the classifiers. It makes search on the categorised documents simple easy and it permits individuals to pay time. Some magazines and newspapers are exemplar for the firms in since they received several advertisements and most need and automatic system those to arrange these advertisements to classes in order that thousands of document similarity advertisements concerning sports, business, politics, educations etc. won't need to be divided into documents manually. Newspapers aren't solely printed this paper however they some document exits conjointly within the document electronic surroundings and create use of text categorization throughout some deciding the places of file caches places of article consistent with to the topic cover. This project to find out details see.

3.3.2.5 Finding disambiguos words:-

Applications of text categorization square measure doable to use fine text documents with in an exceedingly text, that have over one document which means and confirm the means to which means of its that category's finding time calculate disambiguos words for instance the document base classifiers base in an categorization article it's presumably accustomed offer the same which means, used to give the some meaning. Text categorization finds the meant thatmeans. That is often referred to as File caches documents application and is usually used in K-Nearest Neighbor algorithm rule.

3.3.2.6 Different Approaches:-

We have mentioned about the decision Tree method, K nearest neighbor classification(k-NN), and lexical chains are generally not used in this text categorization but other research topics, so will not clearly explain them here very briefly in out proposal. But are many other applications used in this text categorization. We will try to explain all approaches including the K-NN classification method we use in our implementation.

3.3.2.7 K -nearest neighbor bayes probabilistic classifiers:-

The K -nearest neighbor probabilistic classifiers are employed in machine learning applications, to shortly make a case for it joint chances of words and text classes to estimate that of follow predefined classes are appropriate for a text. Then. K- nearest neighbor simplifies the computation and will increases to potency with the belief text document. In line with this assumption, text document in one class have freelance of this following additional blessings this document's conditional chances.

According to this assumption, text document in one category have freelance of this following a lot of blessings this document's conditional chances.

Let as justify exploitation mathematics:

P (c/d)=probability for a document text to belong to class File caches document.

 $\overline{P}(c/d) = [P(c)*p(d/c)]/p(d)$ wherever matrix perform

P(c)=probability that any document chosen haphazardly belongs to text class

P(d)=probability that any document chosen haphazardly features a vector illustration.

P(d/c)=probability that text class includes conjointly documents.

There are a lot of formulas to calculate p(d/c) and also the earlier mentioned k nearest neighbor Algorithm assumption permit this calculation of p(d/c) to be done easier.

3.3.2.8 Decision Tree method:-

The K-Nearest Neighbor rule is employed in machine learning, a choice tree classifier uses a tree structure in some internal nodes stands for things and terms and edges holds relevant weights associated with the terms in nodes.

Finally the lead nodes represent text classes. These classifiers reason documents that following a path to the connected lead node. This path is set consistent with the algorithmic partitioning specific rule and a some rule referred to as info gain algorithm rule.

The algorithm program is given below:

- Class within the set of coaching documents are set to 3 or 5 if containing or not.
- ii. The coaching documents are appointed to a root node and a perform be enforced are assigned to a root node and text document a info retrieved from training documents is that the root of the tree.
- iii. The Recursive algorithm partitioning algorithm program is enforced during this a part of document since the perform is recursive and takes nodes as parameters.
- i. if this node has some specific documents stopping conditions, flip this node into a let in order that is appointed to the category most ordinarily used among coaching documents known as text catteries
- ii. if there's chance to make a brand new text class, a word from the coaching documents, this could be word that holds most in an exceedingly document of not. According crate to kid documents current node and call and decision this project.

Document sense Disambiguation:-

Document sense clarification is that the activity of finding text classes, prevalence associate exceedingly in a very text of an ambiguous document, the sense of this explicit word prevalence file document. As an example, bank could have a lot of totally different check files are offered as within the establishment or the document. DSD is extremely vital for several applications, together with tongue process, the text documents by word senses rather of then by words for algorithmic program functions. Once we have tendency to read document prevalence contexts as document and text senses as classes.

3.3.2.9 Hierarchical categorization of web pages:-

Text has recently or used a lot of interest also for its possible applications to belong very much of document automatically classifying web pages, or web sites, under the hierarchical section catalogues hosted on popular internet portals. Where text documents are catalogued in this way, rather than issuing letters in categories an algorithm to general-purpose web search engine model a searcher may find it easier to first navigate in the hierarchy of categories and then restrict her search to a given document of particular category of interest.

Classifying web pages automatically has more advantages, this project and algorithm since the manual categorization of a large size of document enough subset of this web is infeasible. In to more unlike in the previous applications, it is file text modeling typically the case of study that each category must be populated the set of $k_1 < x < k_2$ documents. Should be chosen so as to allow new categories to be added and document to be deleted. With respect to previously discussed text applications, automatic web page categorization

Two essential peculiarities:-

- i. The hypertext nature of this documents like a more links are a rich file structure source of information, as they may be understood as training data stating the relevance of the linked based pages to the linking pages. Techniques as exploiting this intuition in a context have been logics.
- The hierarchical structure of file documents category sets this may be used on decomposing documents the classification in to regular experience problem into most a number of smaller classification problems, each text corresponding to a branching decision at an internal node techniques documents.

This organization of documents is important to each in business documents one by one however is additional project's particularly important for corporations, firms and also the classifiers. It makes search on the categorised documents straightforward and it allows folks to pay time. Some magazines and classifiers documents most are model for the firms in since they received several advertisements and most need and automatic system those to arrange these advertisements to classes in order that thousands of document advertisements concerning sports, business, politics, educations etc. won't must be divided into documents manually. Newspapers aren't solely revealed this paper however they some document exits conjointly within the electronic setting and build use of text categorization throughout some crucial the places of file caches places of article per the subject.

The machine learning approach to text categorization:-

The most popular approach for the document creation of text will be automatic document classifiers consisted in to manually some building, by means of different documents knowledge techniques, an machine learning expert system capable of length text decisions. Such an expert system would like typically consist of a set of manually process defined logical rules, one text document per category of type if the formula then category more types if formula using to identify the category disjunctive normal formula is a disjunction of conjunctive class to added the document is classified under category if u want to more document add satisfies the formula is a disjunction of conjunctive clauses, the document almost classified under category formula in using machine learning in automated text categorization.

A sample rule of the type used in construe is illustrated

The drawback this project approach is the knowledge is question well-known from the expert system identify literature. That is some rules must be manually defined by a knowledge engineer with the text id concepts of a domain expert if the set of categories is updated, then those add two professionals must intervene again, and if the classifier is ported to a completely different domain, expert user needs to intervene and the work has to be repeated both side from scratch. In the others hand, it was originally suggested that this approach can give very good effectiveness to document intervene and the work has to be ready to repeated from scratch. On the other hand, its more specify originally suggested that this approach documents can give very normal effectiveness results, on a subset of the results test collections that output performs even those best classifiers built in this 90 by state-of –art machine learning techniques. However, other classifier has been tested on this same dataset as construe, project not clear whether this was a randomly chosen or a favorable subsets of entire results collection. As project the results above

do not clear randomly chosen or a results again may be obtained in general methods. Since the early machine learning approach to documents gained popularity and has eventually become the dominant one, at least in the research comments a comprehensive introduction to machine learning approach a general inductive process also called the machine learning, automatically builds a classifier for a category files domain expert from these characteristics, the inductive process gleans the characteristics that inductive process characteristics that a new unseen document should have in order to be classified document.

CONCLUSION AND SAMPLE SCREEN SHOTS

CHAPTER - IV SCREENSHOT

4.7 User Interact Page



The main page:-



Importing the Data set:-



Output:-

an Administrator: CAWindows\system32\cmd.exe	- • • -×-	
INFO: Document business-459 classified as business Nar 24, 2019 7:36:15 PM com.machinelearning.doc_classifier.k	mn.core.KNNClassifyT	Q- Search (Chi+i)
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estDocumentThread run	and concernance assuring t	
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Mar 24, 2019 7:36:15 PM com.machinelearning.doc_classi	fier.knn.core.KNNClassifyT	
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estDocumentThread run		
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4.8 FUTURE WORK:-

The k-Nearest Neighbor formula is a simple to modifiable algorithm and is filmable to totally different issues. This formula permits north American country to develop code any to extend potency and accuracy. We tend to assign specifically one classes to at least one article in our implementation for safe of potency. However there exists article containing several terms concerning many classes are offered. Articles don't seem to be perpetually written concerning every topic, however jump from topic to topic in way of life. This means, articles might have multiple classes. This version of our implementation seems to offer less info than it ought to, at that time not be tough to point out different any text documents classes since the k-NN formula already calculates the required info for distribution multiple categories. Keep in mind that we tend to calculates all similarity totals and confidences of every class document and select the best worth to assign our category.

CHAPTER - V CONCLUSION

5.1 Conclusion

This project has been successfully developed and interpreted and system was developed according to the user requirement. The system produces accurate result and it also reduces a lot of overheads, which the manual system faced. The system was thoroughly tested according to the testing methodologies. We several documents are text strategies are exploitation text categorization, thanks to strategies and it's document that text categorization may be a helpful application in standard of living wherever the requirement of automatic system will increase information is automatic calculated the information.

K-Nearest classification

ORIGIN	ALITY REPORT			
3 SIMILA	5% ARITY INDEX	33% INTERNET SOURCES	20% PUBLICATIONS	18% STUDENT PAPERS
PRIMAF	RY SOURCES			
1	www.clip	o <mark>s.ua.ac.be</mark>		13%
2	WWW.Ce	ng.metu.edu.tr		9%
3	de.slides	share.net		2%
4	www.slic	deshare.net		2%
5	elvis.slis	s.indiana.edu		1 %
6	Submitt Student Pap	ed to CSU, San J	Jose State Univ	versity 1%
7	www.lw1	ree.cn		1%
8	Submitt Student Pap	ed to Info Myann ^{er}	nar College	1%
9	Submitt Student Pap	ed to Sacred Hea	art College	1%

10	Submitted to University of Warwick Student Paper	< 1 %
11	Submitted to Selçuk Üniversitesi Student Paper	<1%
12	Célia Valente. "A Tool for Text Mining in Molecular Biology Domains", Repositório Aberto da Universidade do Porto, 2014. Publication	< 1 %
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Geometric Features", International Journal of Pattern Recognition and Artificial Intelligence, 2015.

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Exclude bibliography	Off		

Image Retrieval

by S Sagayaraj

Submission date: 02-Apr-2019 12:37PM (UTC+0530) Submission ID: 1104362585 File name: BP170524_Tag_Based_Image_Retrival.pdf (403.19K) Word count: 2196 Character count: 10971

JOINT HYPERGRAPH LEARNING FOR TAG-BASED IMAGE RETRIEVEL A PROJECT REPORT

Submitted in partial fulfillment for

The award of the degree of

MASTER OF SCIENCE IN COMPUTER SCIENCE

By

AISWARYA.S

Reg. No: BP170524

Under the guidance of

Prof. D. Gajalakshmi, MCA.,



PG DEPARTMENT OF COMPUTER SCIENCE

SACRED HEART COLLEGE (AUTONOMOUS)

Tirupattur, Vellore Dist-635 601

APRIL - 2019

BONAFIDE CERTIFICATE

This is to certify that the project work entitled JOINT HYPERGRAPH LEARNING FOR TAG-BASED IMAGE RETRIEVEL is submitted to Sacred Heart College (Autonomous) Tirupattur by AISWARYA S (Reg.No. BP170524) for the partial fulfillment for the award of the Degree of Master of Science in Computer Science is a bonafide record of the work carried out by her, under my guidance and supervision.

Date:

Signature of the guide

Submitted for Viva –Voce examination held on:

Internal Examiners

External Examiner

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I thank God Almighty for his blessings and graces by which I was able to complete project work successfully.

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Finally, I thank each and every one of my friends, especially who have assisted me in completing the project work.

AISWARYA.S

1. INTRODUCTION

1.1 Project Summary

Joint Hypergraph Learning for Tag-based Image Retrieval is a web application. It's helpful for social network users. The social network users can easily create their own account, and use this account. Through this web application, the users can share the images approach and Search for the image in efficient way using tag based image retrieval. While the user sharing the images, the images are dumped into search option and the same can be added into timeline news feed. This application not only limited to image sharing but also the users can able to follow or following other users in the friends list.

1.2 List of Modules

- ➤ Home
- > Admin
 - User Details
 - o View Images
- ≻ User
 - o User Home
 - o Profile
 - o Share
 - TimeLine
 - Search Image
 - \circ Followers
 - \circ Followings
 - Logout
- User Registration

1.3 Conclusion

This chapter says about the main concept of the project and motive of the Application. This chapter concludes with motive of the project.

2. REQUIREMENTS ANALYSIS

2.1 INTRODUCTION

Requirement analysis brings the system requirements of the project. This chapter gives the analysis of the application, E-R diagram, DFD diagram, Product vision, Use case specification of the application. This brings to get an idea about the scope of the project.

2.2 SYSTEM STUDY

2.2.1 Overview

The public networking users anywhere and anytime to share and view the image details through offline.

2.2.3 Proposed System

- This proposed system helpful for the public social networking users. Through this application performance is sharing the images, search and view the images.
- Through this application approach to automatically use different visual features and tags for images.
- This application is offline based approach and can capture more reliable relationships between the images.

Features

- Easily share unlimited images
- Easily view the images and image details

Advantages of proposed system:

This application not only limited to image sharing but also the users can able to follow or following other users in the friends list.

2.2.4 Responsibilities of Admin

- ✓ User Details
- ✓ View Images

2.2.5 Modules and their Description List of modules:

- Home
- Admin
- User
- User Registration

Module No: 01

Module Name: Home

Overview

To view about the website

Module No: 02

Module Name: Admin

Overview:

Securing administrator tools from the users

Activities: 2.1

Activity Name: Admin Home

Purpose

To ensure that the login person is admin

Entry Criteria

Valid admin name and password

Input

Admin name

Password

Output

Now the admin can use his tools

Edit Criteria

Admin exit the login of the moment he/she press the proceed or back button.

Activities: 2.2

Activity Name: User Details

Purpose

To see the how many users are registered

Entry Criteria

Valid admin name and password

Edit Criteria

User can exit the controls

Activities: 2.3

Activity Name: View Images

Purpose

To see how many users are what are the image shared

Entry Criteria

Valid admin name and password

Edit Criteria

User can exit the controls

Activities: 2.4

Activity Name: Logout

Purpose

The purpose is to completing their process to leave in this page.

Module No: 03

Module Name: User

Overview:

We can share, view, following friends and followers are saw

Entry Criteria

To give the valid email id and password.

Input

Email id

Password

Output

Now the user can use this tools

Edit Criteria

User exit the login of the moment he/she press the proceed or back button.

Activities: 3.1

Activity Name: User Home

Purpose

To view the home page of users

Activities: 3.2

Activity Name: Profile

Purpose

To see the user details

Edit Criteria

User can exit the controls

Activities: 3.3

Activity Name: Share

Purpose

To share any images

Entry Criteria

Valid details are given about the image

Input

To see the image details

Output

To see the image

Edit Criteria

User can exit the controls

Activities: 3.4

Activity Name: TimeLine

Purpose

To see the images

Edit Criteria

User can exit the controls

Activities: 3.5

Activity Name: Search Image

Purpose

To search the given images

Entry Criteria

We can search the specific images.

Input

To give the name of the images

Output

To view for name of the images

Edit Criteria

User can exit the controls

Activities: 3.6

Activity Name: Followers

Purpose

To see how many followers are seen in this site

Input

Search the images

Output

To see the images

Edit Criteria

User can exit the controls

Activities: 3.7

Activity Name: Followings

Purpose

To see how many users are followed yourself

Input

Search the images

Output

To see the images

Edit Criteria

User can exit the controls

Activities: 3.8

Activity Name: Logout

Purpose

The purpose is to completing their process to leave in this page.

Exit Criteria:

After completing this task, the user will be redirected at the home page for further process.

Module No: 04

Module Name: User Registration

Purpose:

If the new user means they have to give their details.

Entry Criteria:

User has login into the application.

Input:

To give the

- 6 1. First name
- 2. Last name
- 3. Email
- 4. Password
- 5. Date of birth
- 6. State
- 7. Gender
- Country
- Profile picture

Output:

To add the new user

Edit Criteria

User can exit the controls

2.3 Vision Document

2.3.1 Introduction

Joint Hypergraph Learning for Tag-based Image Retrieval is a web application. It's helpful for social network users. The social network users can easily create their own account, and use this account. Through this web application, the users can share the images approach and Search for the image in efficient way using tag based image retrieval. While the user sharing the images, the images are dumped into search option and the same can be added into timeline news feed. This application not only limited to image sharing but also the users can able to follow or following other users in the friends list.

2.3.2 Positioning

I Problem Statement

The problem of	To maintain the image and user details.
Affects	The users.
The impact of which is	Time consuming to see the images easily.
The successful solution would be	It is a web application for developing this project to full fill the joint hyper learning for tag-based image retrieval Tagging requirements to provide into the successful solution.

II Product Position Statement:

For	The Image retrieval Tagging
Who	Users.
The	Image retrieval Tagging
That	Maintenance of the image, user details.
Unlike	The existing system, this application approach to automatically use different visual features and tags for images.
Product	It is one of the important ways to find images contributed by social networking users.

III STAKEHOLDER AND USER DESCRIPTION

Stakeholder summary:

Name	Represent	Responsibilities
End user	The maintain details about Tag based image retrieval.	To monitor the details of admin to ensure their needs.

User summary:

Name	Represent	Responsibilities
Admin	To maintain the records about users and image	Ensure whether the records have
	details.	been maintained.

V Assumptions and dependence
Hardware:

Processor	: Intel i3
Ram	: 2 GB more
Storage	: 50 GB
Input\Output	: Printer, pen drive, key board, mouse.
Operating system	: Windows XP/7

Software:

Front Tier	: JAVA/J2EE
Data Tier	: MYSQL
Software Tools	: NetBeans IDE 7.2.1

2.4 USE CASE DIAGRAM



1. Use Case Name: User Details

1.1. Brief Description:

This Use Case describes about view the all the user details.



1.1.2. Actors:

Admin

1.1.3. Precondition:

The user has separate login to add new users.

1.2. Flow of Events:

The use Case begins when the user move to see add the new user details.

2. Use Case Name: View Image Details

2.1. Brief Description:

This Use Case describes about view the all the images.



2.1.2. Actors:

Admin

2.1.3. Precondition:

The user has separate login to add their images.

2.2. Flow of Events:

The use Case begins when the user move to see the image details.

2.5 Conclusion

4

This web application has been developed for Joint Hypergraph Learning for Tag-based Image Retrieval. It has been developed by the JAVA and the database is MYSQL server. It can be helpful for the public social users.

3. DESIGN DOCUMENT

3.1 Introduction

This chapter tells the design diagram. It gives brief introduction about activity diagram, architecture design, and database design, interface design and procedural design at last the test case.

3.2 Architecture Design



3.3. Activity design



3.4 DATA FLOW DIAGRAM

Admin



User



3.5 Sequence Diagram



3.7 Test case Design

Login:

TSC_ID	TSC_1	
Objective	To Admin Login	
Test Design	ST1: Go on admin page first login .in that login without giving any data.	
	EO: No error should occur and proper message should be handled	
	ST2: Click on cancel without giving any data	
	EO: No error should occur and window should be closed	
	ST3: Enter user name and password	
	EO: User name should be displayed as it is given and password should be displayed as special character	
	ST4: Enter valid user name and password and click on login	
	EO: Admin page should be opened	
	ST5: Enter valid user name and valid password and click on cancel	
	EO: No error message should be displayed, text boxes should be cleared	
	ST6: Enter valid user name and password and click on login	
	EO: No error should occur and proper message should be displayed to get correct data	
	ST7: Enter valid user name and password and click on login	
	EO: No error should occur and proper message should be displayed to get correct data	
Expected output	The admin is logged in the application.	

New user registration:

TSC_ID	TSC_2
Objective	New User Register
Test design	ST1: Click on register without giving any data.
	EO: No error should occur and proper message should be handled
	ST2: Enter mandatory fields and click on add
	EO: The user details will be add in the database

Conclusion

This web application has been developed for Joint Hypergraph Learning for Tag-based Image Retrieval. It has been developed by the JAVA and the database is MYSQL server. It can be helpful for the public social users.

4. IMPLEMENTATION

4.1 Overview of the Implementation

Joint Hypergraph Learning for Tag-based Image Retrieval application is used to see the image details. The purpose of this document to collect, analyze and define the basic requirements, functional units of public user's. This system mainly focuses on see the image retrieval through offline.

- User Details
- View Images

Steps (In bootstrap web application)

- Create a Application and edit NetBeans IDE
- Create a Java file and run it in local host.

Steps (MYSQL)

- First need to install the software MYSQL server
- Next open the MYSQL Server and connect the MYSQL Server
- The need to create and insert the table fields.

Steps (Microsoft Visio-2003)

Need to install the software Microsoft Visio for the diagram purpose.

It only for the diagrams to draw

DFD, Use case, Sequence, architecture diagram has been drowned by this software.

File \rightarrow New \rightarrow Software \rightarrow UML Diagram.

5. TESTING DOCUMENT

3 5.1UNIT TESTING

Unit testing focuses verification efforts on the smallest unit of software design of the modules. This is also reffered as "Module testing".

ADMIN LOGIN

Login		
Admin ID		
Password		
	Login	

USER LOGIN:

Login		
Email		
Password		
	Login	

USER REGISTRATION

User Registration			
FIRST NAME	LAST NAME PASSWORD STATE		
	PROFILE PICTURE		
INSERT			

SHARE IMAGE:

SHARE IMAGES AT TIMELINE		
IMAG	E NAME IMAGE TAGS	
IMAG	DESCRIPTION	
SELEC	T CATEGORY	
SELEC	T IMAGE	
SHARE		

5.2 Integration Testing

In this combination testing there are two kinds of testing exists that is top-down coordination and base up joining.

6. CONCLUSION

The project **JOINT HYPERGRAPH LEARNING FOR TAG-BASED IMAGE RETRIEVAL** is completed and satisfying this application. It has been developed by the web application and the database is MYSQL server.

Image Retrieval

ORIGINALITY REPORT 5% **6**% 4% INTERNET SOURCES **PUBLICATIONS** STUDENT PAPERS SIMILARITY INDEX **PRIMARY SOURCES** Yaxiong Wang, Li Zhu, Xueming Qian, Junwei 1 0/ Han. "Joint Hypergraph Learning for Tag-Based Image Retrieval", IEEE Transactions on Image Processing, 2018 Publication Submitted to Sacred Heart College 3% 2 Student Paper admissionsinindia.com 1% 3 Internet Source Submitted to University of Greenwich 1% 4 Student Paper www.nitc.ac.in 1% 5 Internet Source www.behavior.net 1% 6 Internet Source Submitted to College of Engineering 1% 7 Trivandrum Student Paper

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