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# **Capacity Monitoring Tool**

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# **Problem Statement**

Vodafone has a crucial problem of monitoring the massive capacity of their network nodes in real time. The late response for such an issue might result in complete or partial shutdown of a major server or an active node.



# **Objectives**

Developing a monitoring tool that collects data from vodafone servers to be analyzed, following that, sending alerts to the capacity management team. The analysis of data is intensive and runs through interconnected cycles of clustering, classification and data visualization.



## **Basic Functions**

- The capacity monitoring tool :
  - Collect different files with different formats from different nodes on regular basis (daily, hourly...etc.).
  - Parse the received files from the node and insert the needed data into the database.
  - Notify the capacity management team when the capacity of a certain node is not enough.
  - Figures and analysis of data will be shown in the dashboard website.



#### **Functional requirement**

There are 32 functional requirements, but the most important functional requirements are the following:

Name	Classify
Function Input	selected Algorithm and selected data
Function Output	classified data
Description	classification for the collected data
Expected risks	data error, logical error
Preconditions	calling data from database
Post-conditions	store the classified data into database



## **Functional requirement**

Name	Analyze Data
Function Input	selected Algorithm and selected data
Function Output	analyzed data
Description	data analysis for the selected data
Expected risks	data error, logical error
Preconditions	calling data from database
Post-conditions	store the classified data into database



## **Functional requirement**

Name	Clustering
Function Input	selected Algorithm and selected data
Function Output	clustered data
Description	dividing the selected data for groups using clusters
Expected risks	gathering the data around one cluster and the rest nothing
Preconditions	calling data from database
Post-conditions	store the clusters data into database.



# Non functional requirement

#### • Security

All files will be accessible only by Vodafone server.

#### • Portability

A mobile application will be made for receiving alarms in case of the capacity of a certain node is not enough either via sending an SMS or an e-mail. Also, the mobile application will display charts data, and a website for displaying the dashboards of the nodes viewed for the capacity team on the parsed files.



#### Usecase

**P** 



#### Database





### **Primitive class diagram**





## **Primitive class diagram**





## **Primitive class diagram**





#### Web-based graphical interface tool

+2 http://wodafone.com	] <b>C</b> http://wodafone.com				
Username:	Username: Password: Sign In Sign Up				



#### Web-based graphical interface tool





#### Mobile application

≡	Title		۹	:
	Username:			
	Password:			
		S	ign In	
		Si	gn Up	
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Row No.	Circle	Postpaid - O	Postpaid - In	Postpaid - T	Postpaid - O	Prepaid - Ou	Prepaid - Inc	Prepaid - To	Prepaid - Ou	Blended - O	Blended - In
1	Circle A	472	352	824	16	105	141	246	12	133	<mark>1</mark> 57
2	Circle B	655	398	1053	14	136	177	313	7	162	<mark>188</mark>
3	Circle C	804	395	<mark>119</mark> 9	7	182	243	4 <mark>2</mark> 5	5	2 <mark>0</mark> 6	249
4	?	522	359	881	29	139	168	307	13	181	189

<b>Classification</b>	<u>Tool</u>
<u>Input :</u>	

Blended - To	Blended - O
290	13
350	7
455	5
370	15





#### accuracy: 27.27%

	true Circle A	true Circle B	true Circle C	class precision
pred. Circle A	3	5	1	33.33%
pred. Circle B	0	0	2	0.00%
pred. Circle C	0	0	0	0.00%
class recall	100.00%	0.00%	0.00%	

🔵 Table View 🧕 Plot View



Confusion Matrix (x: true class, y: pred. class, z: counters)



#### **Clustering Tool:**

**<u>Cluto Tool:</u>** 

Input File Sample:





Date Originat	ting	Forwar	rded	Termi	nating	Roami	ng origi	ina <mark>tin</mark> g	Roam	ing forward	ded	Roaming	t
2016-12-22-1005	100943	543	342	29	2	74	0	27998	0	55605	0	0	0
2016-12-22-1010	102685	545	432	13	0	95	0	28849	0	572 <mark>8</mark> 9	0	0	0
2016-12-22-1015	102833	567	315	25	0	83	0	29401	0	57483	0	0	0
2016-12-22-1020	10 <mark>4</mark> 275	584	348	15	0	76	0	295 <mark>4</mark> 0	0	58538	0	0	0
2016-12-22-1025	104724	<mark>541</mark>	319	13	0	84	0	29795	0	58723	0	0	0
2016-12-22-1030	107028	537	216	24	0	56	0	30258	0	60696	0	0	0
2016-12-22-1035	108264	553	148	36	0	76	0	30647	0	61274	0	0	0
2016-12-22-1040	108852	625	193	25	0	56	0	<mark>28495</mark>	0	61879	0	1	0
2016-12-22-1045	109882	671	180	15	0	53	0	27468	0	63053	0	0	0
2016-12-22-1050	112145	720	181	12	0	84	0	29239	0	64497	0	0	0
2016-12-22-1055	112087	675	184	6	0	69	0	27030	0	64681	0	0	0
2016-12-22-1100	11 <mark>453</mark> 1	711	18 <mark>3</mark>	12	0	67	0	32865	0	66477	0	1	0

#### **Results After Clustering:**

- Number of clusters is **5**. \_
- **Similarity Function** used for clustering is **Cosine**. -

#### gCluto - [solution 1]

 Eile
 Project
 Solution
 Window
 Help

 Solution
 solution
 1 - Solution Results

test_tele1	solution	1 - Soluti	on Resul	ts					
	Clustering Options								
	Method: Repeated Bisection								
B Solution 2	CRfun: 12			Simfun: Cosine					
matrix visualization 1	RowModel: None			ColModel: None			Graph Model: Asymetric-D	irect	
mountain visualization 1	ColPrune: 1.000			EdgePrune: 0.000			VertexPrune: 0.000		
e S solution 3	Nearest Nieghbors: 4			MinComponent: 1			CSType: Best		
matrix visualization 1	#Trials: 10			#Iterations: 10					
table36	5-way clustering: [12 of	12]							
matrix visualization 1	Cluster	Size		ISim	ISdev		ESim	ESdev	
Mountain visualization 1	<u>0</u>	2		1.000	0.000		0.999	0.000	
B-S solution 2	1	3		1.000	0.000		0.999	0.000	
matrix visualization 1	2	2		1.000	0.000		1.000	0.000	
mountain visualization 1	3	3		1.000	0.000		1.000	0.000	
Solution 3	4	2		1.000	0.000		1.000	0.000	
B <b>[</b> ] vodafone_system_file	Go to Top								
	Descriptive & Descrimin	nating Features							
	Cluster 0 Size: 2 I	ISim: 1.000 ESim: 0.999							
	Descriptive:	Originating	35.2%	Successful setup	23.7%	Successful released by EXT network	17.8%	Voice	11.7%
	Descriminating:	Unsuccessful normal termination (INT)	29.4%	Content	17.8%	Originating service charging	17.8%	Successful released by EXT network	12.1%
	Cluster 1 Size: 3 I	ISim: 1.000 ESim: 0.999							
	Descriptive:	Originating	34.0%	Successful setup	22.0%	Successful released by EXT network	19.6%	Voice	10.7%
	Descriminating:	Unsuccessful normal termination (INT)	38.9%	Successful released by EXT network	18.4%	Successful setup	11.5%	Originating service charging	8.9%
	Cluster 2 Size: 2 I	ISim: 1.000 ESim: 1.000							
	Descriptive:	Originating	34.4%	Successful setup	23.0%	Successful released by EXT network	18.9%	Voice	11.2%
	Descriminating:	Originating service charging	37.3%	Content	37.3%	Successful setup	11.1%	Unsuccessful normal termination (INT)	5.5%
	Cluster 3 Size: 3 I	ISim: 1.000 ESim: 1.000							
	Descriptive:	Originating	33.9%	Successful setup	22.5%	Successful released by EXT network	19.1%	Voice	11.1%
< >	Descriminating:	Originating service charging	29.5%	Content	29.5%	Originating	24.3%	Unsuccessful normal termination (INT)	9.5%



×

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- 8 :

#### Mountain Visualization for the Result

– 0 × \_ 5 %







# **Thank You!**