Assignment Submission Report

Data Mining & Warehousing Assignment 1

IT13011130 - Gunathilaka D. D. T. M.(08/22/2016)

The following report describes the procedure that has been followed while implementing data mining model for Assignment 1 of Data Mining & Warehousing module.

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Declaration

I hear by declare that following task is my own creation and it does not violate any constraints of the assignment.

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Introduction

Case Study

this assignment it has selected a data set which summarizes a heterogeneous set of features about articles published by Mashable in a period of two years. (www.mashable.com). The respective data set consisted, 39797 records with 60 columns. Each row is identified by the URL of the article and the dependent variable (shares) is continuously valued. The goal is to predict the number of shares in social networks (popularity). In other words, we need to find what kind of articles are attracting eyes from general population using the other 58 independent attributes. In the implemented solution it trains a model to predict the number of shares. Therefore "Regression" is used as the data mining technique.

Background Study for Regression

Regression is a "Predictive" data-mining technique that falls under "Supervised Learning "category. In Oracle there are 2 algorithms to implement regression based data mining models. Both algorithms are particularly suited for mining data sets that have very high dimensionality (many attributes), including transactional and unstructured data. In the model build (training) process, a regression algorithm estimates the value of the target as a function of the predictors for each case in the build data. These relationships between predictors and target are summarized in a model, which can then be applied to a different data set in which the target values are unknown.

Generalized Linear Models (GLM)

GLM is a popular statistical technique for linear modeling. Oracle Data Mining implements GLM for regression and for binary classification.

GLM provides extensive coefficient statistics and model statistics, as well as row diagnostics. GLM also supports confidence bounds.

Support Vector Machines (SVM)

SVM is a powerful, state-of-the-art algorithm for linear and nonlinear regression. Oracle Data Mining implements SVM for regression and other mining functions.

SVM regression supports two kernels: The Gaussian kernel for nonlinear regression, and the linear kernel for linear regression. SVM also supports active learning.

Furthermore, Root Mean Squared Error and the Mean Absolute Error are commonly used statistics for evaluating the overall quality of a regression model.

Root Mean Squared Error

The Root Mean Squared Error (RMSE) is the square root of the average squared distance of a data point from the fitted line.

This SQL expression calculates the RMSE.

SQRT(AVG((predicted value - actual value) * (predicted value - actual value)))

Mean Absolute Error

The Mean Absolute Error (MAE) is the average of the absolute value of the residuals (error). The MAE is very similar to the RMSE but is less sensitive to large errors.

This SQL expression calculates the MAE.

AVG(ABS(predicted value - actual value))

Procedure

Data Cleaning

Before starting the implementation, data preprocessed by adjusting data type and excluding URL as predictor a column. "URL" is also the only unique field in the data set which can be used as case id. Unnecessary data needs to be removed from the original data set, since they might cause to increase the error-rate of predictions. Therefore, rows with empty data were removed from original data set.

Data Integration

The original data set was in .CSV format. Once it has been cleansed as next step; "ONLINENEWSPOPULARITY" table was created under Oracle user account which has data mining privileges. Since some attribute data had around 12 decimal points "NUMBER (38,20)" used for their column type to preserve data.

Implementation

There are 2 ways to implement the models.

- 1. Using query
- 2. Using SQL developer GUI

1st approach

Settings for the selected miner models are stored in a table as key-value pairs. Only the column names have to be same. (setting name, setting value)

```
CREATE TABLE miner_model_settings(setting_name VARCHAR2(30),setting_value VARCHAR2(4000));
```

As next step for this example, the settings of the best regression algorithm for the dataset(GLM) with feature selection with 1.3345% predictive confidence were stored in above created table.

```
BEGIN
INSERT INTO MINER_MODEL_SETTINGS VALUES(
DBMS_DATA_MINING.ALGO_NAME,DBMS_DATA_MINING.ALGO_GENERALIZED_LINEAR_MODEL
);
INSERT INTO MINER_MODEL_SETTINGS VALUES(
DBMS_DATA_MINING.GLMS_RIDGE_REGRESSION,DBMS_DATA_MINING.GLMS_RIDGE_REG_DISABLE
);
INSERT INTO MINER_MODEL_SETTINGS VALUES(
DBMS_DATA_MINING.GLMS_FTR_SELECTION,DBMS_DATA_MINING.GLMS_FTR_SELECTION_ENABLE
);
INSERT INTO MINER_MODEL_SETTINGS VALUES(
```

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```
DBMS_DATA_MINING.PREP_AUTO,DBMS_DATA_MINING.PREP_AUTO_ON
);
INSERT INTO MINER_MODEL_SETTINGS
VALUES(DBMS_DATA_MINING.GLMS_DIAGNOSTICS_TABLE_NAME,'ROW_DIAGNOSTIC_STATISTICS');
COMMIT;
END;
/
```

Following query is executed to build and train the model. In below query 'case_id_column_name' is a unique key which used to identify each record separately while 'target_column_name' is the target that is required to predict by the model.

```
BEGIN

DBMS_DATA_MINING.CREATE_MODEL

(

MODEL_NAME => 'FINAL_GLM_MODEL',

MINING_FUNCTION => DBMS_DATA_MINING.REGRESSION,

DATA_TABLE_NAME => 'ONLINENEWSPOPULARITY',

CASE_ID_COLUMN_NAME => 'URL',

TARGET_COLUMN_NAME => 'SHARES',

SETTINGS_TABLE_NAME => 'MINER_MODEL_SETTINGS'

);

COMMIT;

END;

/
```

When PL/SQL procedure successfully completed, in order to test the created model "ONLINENEWSPOPULARITY" table is used. Therefore, following query is executed to test the implemented model.

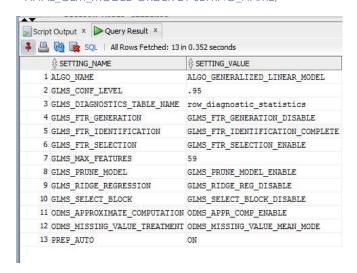
SELECT URL, SHARES, PREDICTION (FINAL_GLM_MODEL USING *) PRED FROM ONLINENEWSPOPULARITY;

Model Settings:

COLUMN SETTING_NAME FORMAT A30

COLUMN SETTING_VALUE FORMAT A30

SELECT SETTING_NAME, SETTING_VALUE FROM USER_MINING_MODEL_SETTINGS WHERE MODEL_NAME = FINAL_GLM_MODEL' ORDER BY SETTING_NAME;



Model Signature (Attributes):

COLUMN ATTRIBUTE_NAME FORMAT A40 COLUMN ATTRIBUTE_TYPE FORMAT A20

SELECT ATTRIBUTE_NAME, ATTRIBUTE_TYPE FROM USER_MINING_MODEL_ATTRIBUTES WHERE MODEL_NAME = 'FINAL_GLM_MODEL'ORDER BY ATTRIBUTE_NAME;

	ATTRIBUTE_NAME	
1	AVERAGE_TOKEN_LENGTH	NUMERICAL
2	AVG_NEGATIVE_POLARITY	NUMERICAL
3	DATA_CHANNEL_IS_ENTERTAINMENT	NUMERICAL
4	KW_AVG_AVG	NUMERICAL
5	KW_MAX_AVG	NUMERICAL
6	KW_MIN_AVG	NUMERICAL
7	LDA_02	NUMERICAL
8	LDA_03	NUMERICAL
9	NUM_HREFS	NUMERICAL
10	N_TOKENS_TITLE	NUMERICAL
11	SELF_REFERENCE_MIN_SHARES	NUMERICAL
12	SHARES	NUMERICAL
13	TIMEDELTA	NUMERICAL

Global statistics:

SELECT * FROM TABLE(DBMS_DATA_MINING.GET_MODEL_DETAILS_GLOBAL('FINAL_GLM_MODEL'))ORDER BY GLOBAL_DETAIL_NAME;

ADJUSTED_R_SQUARE	0.0232352706043227
AIC	632415.191258281
COEFF_VAR	327.051975283298
CORRECTED_TOTAL_DF	33952
CORRECTED_TOT_SS	4267394094335.55
DEPENDENT_MEAN	3387.87129266928
ERROR_DF	33940
ERROR_MEAN_SQUARE	122768615.627308
ERROR_SUM_SQUARES	4166766814390.85
F_VALUE	68.3041559050875
GMSEP	122815640.724438
HOCKING_SP	3617.33155447445
J_P	122815621.547349
MODEL_CONVERGED	1
MODEL_DF	12
MODEL_F_P_VALUE	0
MODEL_MEAN_SQUARE	8385606662.0594
MODEL_SUM_SQUARES	100627279944.713
NUM_PARAMS	13
NUM_ROWS	33953
ROOT_MEAN_SQ	11080.0999827307
R_SQ	0.0235804984775775
SBIC	632524.816780708
TERMINATION	0
VALID_COVARIANCE_MATRIX	1

GLM statistics | Coefficient statistics | Features and their p_values:

SELECT * FROM TABLE (DBMS_DATA_MINING.GET_MODEL_DETAILS_GLM('FINAL_GLM_MODEL'));
SELECT * FROM TABLE(DBMS_DATA_MINING.GET_MODEL_DETAILS_GLM(MODEL_NAME =>
'FINAL_GLM_MODEL'));

SET LINE 120

COLUMN FEATURE_EXPRESSION FORMAT A53

SELECT FEATURE_EXPRESSION, COEFFICIENT, STD_ERROR, TEST_STATISTIC,
P_VALUE, STD_COEFFICIENT, LOWER_COEFF_LIMIT, UPPER_COEFF_LIMIT

FROM TABLE(DBMS_DATA_MINING.GET_MODEL_DETAILS_GLM('FINAL_GLM_MODEL'));

SET LIN 80

SET PAGES 20

SELECT FEATURE_EXPRESSION, COEFFICIENT, P_VALUE

 $FROM\ TABLE (DBMS_DATA_MINING.GET_MODEL_DETAILS_GLM ("FINAL_GLM_MODEL"))$

ORDER BY P_VALUE;

() CLASS () ATTRIBUTE_NAME	ATTRIBUTE_SURVAN	ATTRIBUTE_VALUE	() FEATURE_EXPRESSION	() COEFFICIENT	STD_BIRGR	TEST_STATISTIC	P_VALUE	(VSF	@ STD_COEFFICIENT	@ LOWER_COSPF_LIMIT	UPPER_COSFF_LIMIT	EXP_CONTRICIENT	@ EXP_LOWER_COEFF_LIME	DP_UPPER_COST_LINE
(mull) (mull)	(mull)	(mull)	(mull)	-1653.9571709245499	502.106000000172	2.04093893076918	0.00450077011462292			-2795.06302000997	-512.051301030130	(mull)	(mull)	(mull)
(BELL) MM_ANS_ANS	(mull)	(mull)	RY_ANG_ANG	1.7139305338285	0.126240940827702	13.5766723557215	0.4020200000000000000000000000000000000	7.79395137590769	0.203298646271995	1.46648421336783	1.96136685428957	(mull)	(mull)	(mull)
(null) ins_03	(mull)	(mull)	LOA_D3	515.11027974375904	256.763255496202	2.00616920639192	0.044246214903022	1.59541239344579	0.013591460508068	11.0456091808407	1018.37485028660	(mull)	(mull)	(mull)
(mull) ser_max_ard	(mull)	(mull)	en see en o	-0.184554436765581	0.0210009112506000	9.46234067555743	5.000000000000000072422804337987	5.15037577271007	-0.10302930933134	-0.227300640207742	-0.141808232623414	(mull)	(2011)	(mull)
(null) intracrements som image!	(mull)	(mull)	SELF_REFERENCE_NEW_SHARES	0.029338542939396999	0.00314235582588742	9.24101042286843	0.0000000000000000000000000000000000000	1.01742510428119	0.0499957395360447	0.0220704101721606	0.0351976667066045	(mull)	(2011)	(mull)
(mull) xma_02	(mull)	(mull)	LEA_02	-549.45905184050903	243.451454525786	2.25777495082392	0.0239660171491669	1.30714212930427	-0.0238454693434763	-1026.03253709940	-72.4055657915347	(mu11)	(mull)	(mull)
(null) mus_means	(mull)	(mull)	MUN_MERSE	30.535503451718302	5.54818913374463	5.50340933345795	0.000000374410188838568	1.10728060443344	0.0310432593131223	19.6608673391906	41.410140364240	(mull)	(8011)	(mull)
(mull) MM_MEM_ANG	(mull)	(mull)	NY_NEE AND	-0.457787710538751	0.0716658509387139	6.32720220441402	0.00000000170456373905062	1.04473976626036	-0.0465352025341929	-0.598255203785875	-0.317320217291626	(mull)	(mull)	(mull)
(null) and medative posantry	(mull)	(mull)	AVS_RESATEVE_FOLIANTY	-2433.15193336976	513.107643407279	4.74199101169392	0.00000212496556523904	1.10010914649312	-0.0277237005616956	-3439.06017953147	-1427.44249663635	(mull)	(2011)	(mull)
(SULL) DATA_DRAFFIL_IS_ESTERIAL PRINCEST	(mull)	(mull)	DATA_CHAMPEL_IS_ENTERTAINMENT	-925.21933917301797	160.067036753000	4.02679350357536	0.00000103958369754395	1.14596965100024	-0.0200578061433565	-1156.20344526916	-494.233231076076	(mull)	(2011)	(mul1)
(mull) TEMEDELEA	(mull)	(mull)	TEMEDIATA	1.7310233002511	0.300196392793975	5.61662432426491	0.000000194245911233921	1.2040005298365	0.0330571529190083	1.12694795951506	2.33509964092714	(mu11)	(mull)	(null)
(mall) w_wOmmed_wawas	(mall)	(mull)	N_TORENS_TETLE	81.518304598753085	29.7205294028149	2.74212373393092	0.00410753227705104	1.00778059670912	0.0153398813783985	23.2503024502301	139.789230739266	(mull)	(8411)	(mull)
(null) AVERAGE TOREN LENGTH	(mull)	(mull)	AVERAGE TOKEN LENGTH	-257.74053002559499	80.3379811504002	3.20220273462263	0.00133691281506221	1.20912965902431	-0.0195376537360676	-415.205692802902	-100.275368848288	(mull)	(mull)	(mull)

Validation:

Root Mean Square Error - Sqrt(Mean($(x - x')^2$))

Mean Absolute Error - Mean(|(x - x')|)

COLUMN RMSE FORMAT 9999.99

COLUMN MAE FORMAT 9999.99

SELECT SQRT(AVG((A.PRED - B.SHARES)) * (A.PRED - B.SHARES))) RMSE,

AVG(ABS(A.PRED - B.SHARES)) MAE

FROM (SELECT URL, SHARES, PREDICTION (FINAL_GLM_MODEL USING *) PRED

FROM ONLINENEWSPOPULARITY) A,

ONLINENEWSPOPULARITY B

WHERE A.URL = B.URL:



1 11502.5333274091471849189618877368572007 3051.55541045605563015418474422358995056

2nd approach

In the second method of model implementation a graphical user interface provided by Oracle SQL-Developer tool is used. Once the tool is opened following steps are followed to implement Classification mining model.

Steps

Create new project and work flow in created data miner user account under Oracle Data Miner. Name the project as "Online News Popularity" (any name) and define the workflow as "Popularity by Number of Shares" (any name)



Figure 1 Create Project & workflow

Next drag and drop Data Source from the tool palette to the workflow which is under "Data" section of palette. Once the "Data Source" node is added to workflow it will open a window to select a schema. This schema would be the table which provides training data to the model. Therefore, in our case select "DMUSER. ONLINENEWSPOPULARITY" as the schema. Since we have cleaned irrelevant columns previously all the existing columns of the table are required to build the model. Therefore, click "Finish"

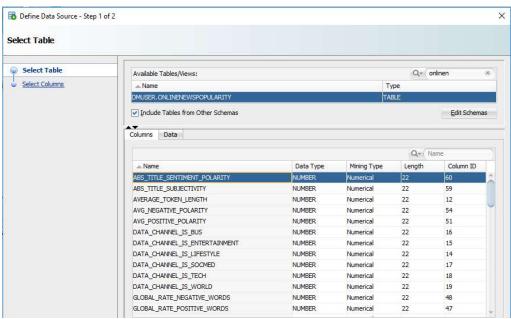


Figure 2 Select Data Source

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As the next step the model node should be created. Therefore drag and drop node from tool palette which is under "Models" section. Then link the two nodes () by using which is under "Linking Nodes" category or right click > connect.

Next double-click or right click → Edit and open edit window of "Regress Build" (i.e. Regression) node. Then use following setting:

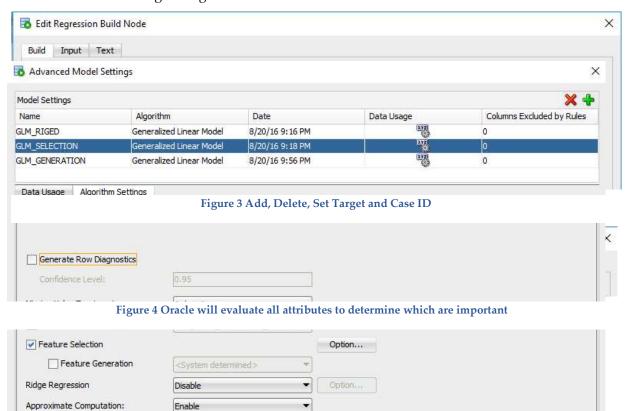


Figure 5 Advance settings for best confidence in dataset

Double Click on the model for advanced settings:

Press Ok twice to go back to work flow.

Now right click on regress build and click on Run button.

Once the model has successfully executed drag and drop "Apply" node and new "Data-Source" node from tool palette.

Next double click on newly added "Data-Source" node and "ONLINENEWSPOPULARITY" table in data source node.

In apply node:

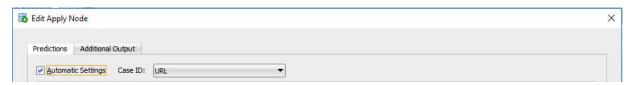


Figure 6 Set case ID for quicker predictions



Figure 7 Set Column order

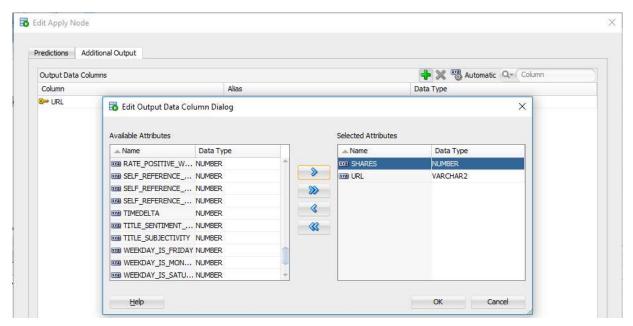


Figure 8 Add Case Id as reference, actual shares to compare with predictions

Next re-run the entire workflow. This will apply implemented all algorithms to the records of "ONLINENEWSPOPULARITY" table. Now models are ready to provide predictions.

In order to display predictions a separate table called "REGRESSION_RESULT" is created. This table keeps the predictions, probability of those predictions and relevant case Ids made by models.

The implemented work-flow appears as below.

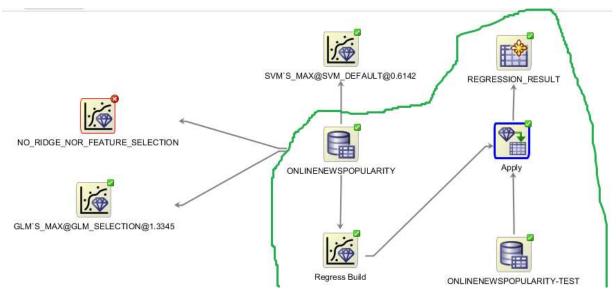


Figure 9 Selected Model for deploy

The accuracy of the models can be visualized by selecting "Compare Test Results" option in menu which appears once right click on "Class Build" node.

Right click on apply node \rightarrow deploy \rightarrow select node, dependent node and children node to generate query for the model.

In model settings:

Missing value treatment set to "Delete Row" to remove rows with null value

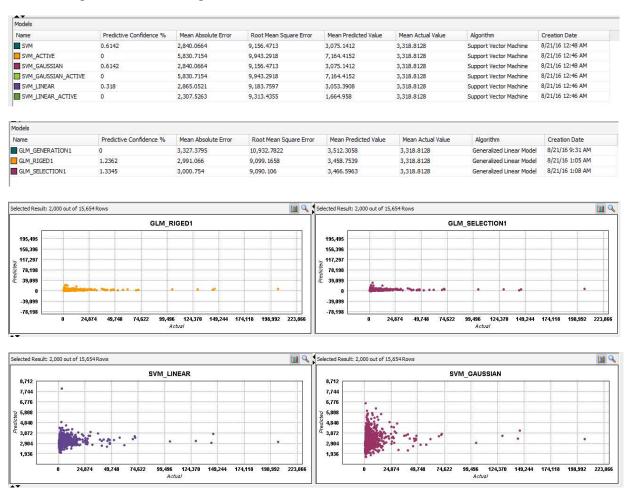
Shares (Independent variable) and URL not used in prediction



Figure 10 URL and Shares attributes are not used

Results & Discussion

The overall performances of implemented models are shown below.



According to the above result SVM with linear kernel has the lowest spread and relatively highest accuracy. But when favoring the predictive confidence, we have to pick GLM with feature selection (confidence 0.318< 1.3345) as the best model.

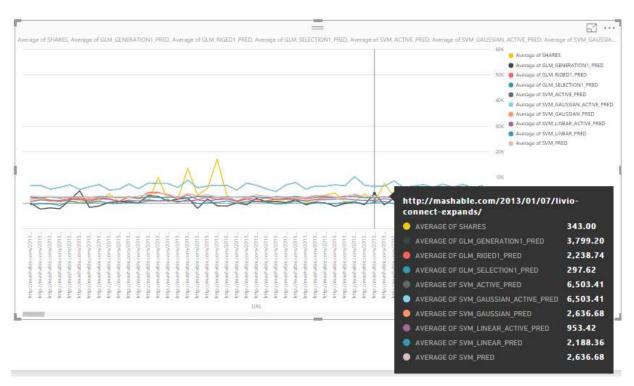


Figure 11 Prediction value distribution per post

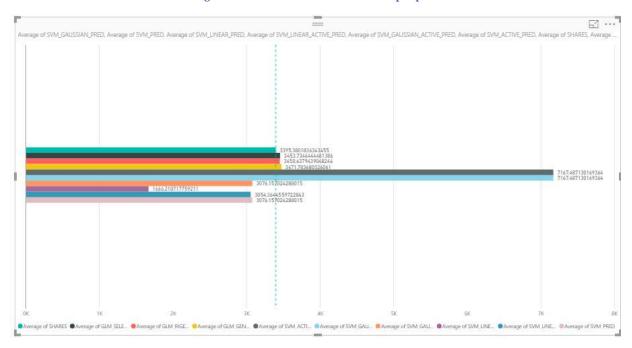


Figure 12 Overall Prediction value Average

Integration

The implemented models can be integrated with software components. Word form with VBA backend implemented for data insertion. Diagrams and visualization are displayed through an Excel instance but all the data manipulation done in the oracle server. Power BI used to visualize spread of training data.

		AVG. KEYWORD (AVG. SHARES)	CLOSENESS TO LOATOPIC 2
Article Features	- to- M to to	1200	0.066657527258
map-y/manas-comyas-ayou		MIN. SHARES OF REFERENCED ARTICLES IN MASHABLE	OLOSENESS TO LOATORICA
DAYS BETWEEN THE ARTIOLS THE DATASET ACQUISITIO	IS DATA CHANNEL ENTERTAINMENT? [YES - 1 NO - I	1200	0.0677605397969
NUMBER OF WORDS IN THE TITLE	0 IS DATA CHANNEL BUSINESS 7 [YES - 1] NO - 0]	MAX. SHARES OF REFERENCED ARTI CLES IN MASHABLE	CLOSENESS TO LOATORIC4
		1200	0.0666579921915
10	0	AUG. SHARES OF REFERENCED ARTI CLES IN MASHABLE	TEXT SUBJECTIVITY
NUMBER OF WORDS IN THE CONTENT 151	IS DATA CHANNEL SOCIAL MEDIA? [YES -1 NO -0]	0	0.297205397205
AUTE OF LINE OUR WORDS IN THE CONTENT	IS DATA CHANNEL TECHT? [YES -1 NO -0]	WAS THE ARTICLE PUBLISHED ON AMONDAY? [1 0]	TEXT SENTIMENT POLARITY
		0	0.0218956228956
0.694562752795	0	WAS THE ARTI OLE PUBLISHED ON ATUESDAY? [1] 0]	RATE OF POSITIVE WORDS IN THE CONTENT
0. 99999999474	IS DATA CHANNEL WORLD ? [YES - 1 NO - 0]	0	0.0264900652252
	-1	WAS THE ARTICLE PUBLISHED ON A WEDNESD AVY [1]	RATE OF NEGATIVE WORDS IN THE CONTENT
0.910526307259	WORST KEYWORD (MIN. SHARES)	0	0.0122450321126
- 200 - 300 - 100		WAS THE ARTICLE PUBLISHED ON ATHURSDAY? [1] 0	RATE OF POSITIVE WORDS AMONG NON-NEUTRAL
NUMBER OF LINKS	WORST KEYWORD (MAX. SHARES)	0.	0.66666666667
5	21	WAS THE ARTICLE PUBLISHED ON APPLICATE [1] 0]	RATE OF NEGATIVE WORDS AWONG NON-NEUTRAL TO
NUMBER OF LINKS TO OTHER ARTI CLES PUBLISHED BY	WORST KEYWORD (AVG. SHARES)	1/	0.2222222222
NUMBER OF IMAGES	205800 REST KEYWORD (MIN. SHARES)	WAS THE ARTICLE PUBLISHED ON ASATURDAY? [1 0]	AVG. POLARITY OF POSITIVE WORDS
1		0	0.226515151515
NUMBER OF MODOS	942200 925T KEYMORD (MAX. SHARES)	WAS THE ARTICLE PUBLISHED ON A SUNDAY? [1 0]	MIN. POLARITY OF POSITIVE WORDS
NUMBER OF VIDEOS	419732.323233	1	0.032323232323
AVERAGE LENGTH OF THE WORDS IN THE CONTENT	REST KEYWORD (AVG. SHARES)	WAS THE ARTICLE PUBLISHED ON THE WEEKEND? [1]	MAX. POLARITY OF POSITIVE WORDS
4.5761599404	3196.52065181	0.086889121999	0.6
NUMBER OF KEYWORDS IN THE METADATA	AUG. KEYMORD (MIN. SHARES)	CLOSENESS TO LOATOR CO	AUG. POLARITY OF NEGATIVE WORDS
	AND RETHINGS (HERE SHAMES)	222	-0.25
	2001 74022200		
0	2591.74022959	OLOSENESS TO LOATORIC 1	MIN. POLARITY OF NEGATIVE WORDS
0	3951,74032868 AUG. KEYWORD (MAK. SHARES) 2414,40244188	Control Management of the Control of	
	AUG. KEYWORD (MAX. SHARES)	CLOSENESS TO LOATORICS	MIN. POLARITY OF NEGATIVE WORDS
O SOATACHANNEL LIFESTYLET (YES-1) NO-0] 1 MAKK. POLABITY OF NEGATIVE WORDS -0.3 TITLE SUBJECTIVITY Q 454545454545 TITLE SUBJECTIVITY Q 126363666354	AUG. NETWORD (MAN. SHARES) 2454 40244156 ABSOLUTE SUBJECTIVITY LEVEL 0.0454545454545 ABSOLUTE POLIAR TYLEVEL 0.13636363634	CLOSENESS TO LOATORICS	MIN. POLARITY OF NEGATIVE WORDS
O SOATACHANNEL LIFESTYLET (YES-1) NO-0] MAX. POLARITY OF NEGATIVE WORDS -0.2 TITLE SUBJECTIVITY 0.154545454545 TITLE SUBJECTIVITY 0.12626363654 Pritts Submit to evertuate Submit Kandom S	AUG. NETWORD (MAN. SHARES) 2454 40244156 ABSOLUTE SUBJECTIVITY LEVEL 0.0454545454545 ABSOLUTE POLIAR TYLEVEL 0.13636363634	Q. OSSINESS TO LIGATORIC I Q. TEZZZAMSTYSS	MIN. POLARITY OF NEGATIVE WORDS
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