

COMPARATIVE ANALYSIS BETWEEN CLASSIFICATION ALGORITHMS AND DATA SETS (1: N & N: 1) THROUGH WEKA

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Abstract: Data mining is a stage in the knowledge disclosure process comprising of data mining algorithms that used to discovers patterns in the data. Data Mining likewise can be characterize as an analytic procedure proposed to consider a large data in scan for reliable patterns and deliberate connections amongst factors and after that to agree the discoveries by applying on new subsets of data by the distinguished patterns. Classification [14] is the mainly usually attached data mining system, which utilizes an plan of pre-classified cases to construct a model that can order the number of inhabitants in records on the loose. In classification strategies a model is manufactured in view of preparing data and connected to test data. WEKA is an open source data mining apparatus which incorporates usage of data mining algorithms. In this paper explain about two different tasks Compare between different Datasets with single algorithm and Comparison between different classifications.

Keywords: Data mining, classification algorithms, datasets, MAE, RMSE, RAE, RRSE.

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I INTRODUCTION

Data mining [1][20][21][22] is a method of quickly developing interdisciplinary field, which consolidates database management, statistics on datasets, machine learning algorithms and related regions going for removing helpful knowledge from vast accumulations of data. The data mining process comprises of three essential stages: investigation, build model or pattern definition, and validate/confirmation. In a perfect world, if the idea of accessible data permits, it is commonly rehashed iteratively until a "robust/standard" model is distinguished. In any case, in business tradition the alternatives to approve the model at the phase of examination are commonly restricted and, in this way, the underlying outcomes frequently have the status of heuristics that could impact the choice procedure.

Data mining should be possible with vast number of algorithms and strategies which include regression analysis, classification techniques, clustering techniques, and association rules, artificial intelligence, neural networks, and so forth. Basically classification [14] and clustering algorithms also known as supervised and unsupervised classification. Supervised learning means Supervision: The training data (observations, measurements, etc.) are accompanied by labels indicating the class of the observations, that new data is classified based on the training set. Unsupervised learning means the class labels of training data is unknown given a set of measurements, observations, etc. with the aim of establishing the existence of classes or clusters in the data. WEKA[26] incorporates usage of different classification algorithms [14] like "*Bayes Net classifier*", "*NAIVE BAYES*" [25], Meta-Ada Boost- M1, Attribute Selected Classifier, Iterative classifier optimizer, Multiclass Classifier, Randamizable Filtered classifier, Decision Table[24] using single dataset from UCI dataset repository [2].

Naive Bayes and Bayes net algorithms [4] were successfully utilized for apparatus condition checking too [3]. Naive Bayes classification algorithm [6] is a probabilistic classifier and utilizations statistical method for every classification. Bayes Net model represents probabilistic connections among an arrangement of random factors graphically. It shows the quantitative quality of the associations between factors, enabling probabilistic convictions about them to be refreshed consequently as new data that ends up noticeably accessible. This is a "Coordinated Acyclic Graph" (DAG) G that determines a combined likelihood conveyance, where the different nodes of graph stand for random variable and circular segment stand for relationship between variables [7]. A decision table is a straightforward structure to use "divide and conquer method to separate an composite decision making process into an accumulation of more straightforward decisions, subsequently giving an effectively interpretable arrangement [8][9][10][16]. The decision table is simple to understand and

we can get after a tree structure effortlessly to notice how the decision is ended. This is a predictive demonstrating method consumed as a module of classification, clustering and prediction assignments [8], [16].

Yoav Freund and Robert Schapire were introduced a machine learning algorithm, that is known as Ada Boost M1 (or) Adaptive Boosting. This can be developed as a part of concurrence with various dissimilar sorts of learning algorithms to improve their execution. The reschedule of the other learning algorithms ('feeble students') is joined into a weighted aggregate that represents the last yield of the supported classifier. Ada Boost [12] is adaptive as in consequent feeble students are changed for those occurrences misclassified by previous classifiers. Ada Boost is touchy to loud data and anomalies. Attribute selection classifier evaluated by independent and learning algorithm. The independent method evaluated the attributes and subsets of attributes through filters but learning algorithms are evaluated through subset of attributes through Wrappers. In the event that two items are connected, gathering something around one protest can help deductions concerning the others. We label this approach iterative classification [13].

Deductions prepared with lofty trust in beginning iterations are nourished once more interested in the data to fortify deductions about correlated questions in consequent iterations. Multi class classification means each training point fit in to one of n different classes. To predict the new data point through new data function related to which class. Class for running a subjective classifier on top of data that have been gone throughout a discretionary filter. Similar to the classifier, the structure of the filter is constructing only in light of the training data and test occurrences will be prepared by the filter without changing their structure.

Commencing to writing only can recognize to facilitate numerous classification algorithms contain and to utilized for characterizing the issues in thrusts and others turning individuals. So as to state firmly that a specific algorithm is well again to contrasted with different algorithms for relative investigation should be finished. Henceforth, this paper predominantly manages the exhibitions of different algorithms in different perspectives at the same different datasets with single algorithm.

II BREAST CANCER DATASET

Breast cancer [5][15][27] is a standout amongst the most well-known cancers among ladies on the planet. Early identification of breast cancer is basic in lessening their existence misfortunes. Data mining is the way toward examining huge data and exactness it into helpful information disclosure along with the part of data mining come nears to developing quickly particularly classification systems are exceptionally viable approach to characterizing the data, which is basic in managerial method for medical experts. This examination at hands the distinctive data mining classifiers going on the database records of breast cancer; next to utilizing classification exactness with and with no include determination systems. Breast cancer dataset contains 287 rows or instances and 10 features or attributes (age, menopause, tumor-size, inv-nodes, node-caps, breast quad, deg-malig, breast, irradiat, class) and 287 instances are classified into 2 categories, the classification is done which algorithm is best based on some factors those are "CCI" (correctly classified instances), "ICI" (incorrectly classified instances), "MAE" (mean absolute error), "RMSE" (Root mean square error), "RAE" (Relative absolute error), RRSE (Root Relative squared error).

 Table 1 Comparison between different classification

 Algorithms with Single dataset through different factors

		Factors								
SNo.	Algorithm	CCI (instances & %)	ICI (instances & %)	Kappa Statistics	MAE	RMSE	RAE (%)	RRSE (%)	Time taken (Sec)	
1	BayesNet classifier	208 (72.47%)	79 (27.52%)	0.313	0.217	0.3716	76.774	99.096	0.03	
2	Naïve Bayes	208 (72.43%)	79 (27.52%)	0.313	0.216	0.3693	76.244	98.461	0.02	
3	Meta-Ada Boost M1	200 (69.68%)	85 (30.27%)	0.246	0.297	0.3557	104.925	100.184	0.04	
4	Attribute Selected classifier	208 (72.47%)	79 (27.52%)	0.313	0.217	0.3716	76.774	99.096	0.03	
5	Iterative classifier optimizer	207 (72.12%)	80 (27.87%)	0.241	0.246	0.3605	87.041	96.128	0.74	
6	Multiclass Classifier	196 (68.29%)	91 (31.7%)	0.177	0.248	0.3802	87.677	101.37	0.16	
7	Randamizable Filtered classifier	191 (66.55%)	96 (33.44%)	0.22	0.227	0.4642	80.043	123.768	0.02	
8	Decision Table	210 (73.17%)	77 (26.82%)	0.269	0.264	0.3551	93.391	94.689	0.12	

The value of Kappa [18] is defined as
$$\kappa = \frac{p_o - p_s}{1 - p_s}$$

 $MAE = \frac{1}{n} \sum_{j=1}^{n} |y_j - \hat{y}_j| \quad \text{RMSE} = \sqrt{\frac{1}{n} \sum_{j=1}^{n} (y_j - \hat{y}_j)^2} \quad \textbf{RAE} = \frac{\sum_{i=1}^{N} |\hat{\theta}_i - \theta_i|}{\sum_{i=1}^{N} |\bar{\theta} - \theta_i|}$ $\text{RRSE} = \sqrt{\frac{\sum_{i=1}^{N} (\hat{\theta}_i - \theta_i)^2}{\sum_{i=1}^{N} (\bar{\theta} - \theta_i)^2}}$

Above table explain about the different algorithms applied on single dataset (breast cancer dataset) identify the dataset is suitable to this breast cancer dataset. After building a number of different regression models, there is an abundance of criteria by which they can be assessed and compared.



Figure 1: Breast cancer dataset 10 attributes measurements



Figure 2: Comparison between classification algorithms through CCI and ICI factors

Above figure 1) shows the number of instances correctly classified. If more number of instances are classified correctly then we identify particular algorithm is suitable to this dataset, this is the one factor for observation.



Figure 3: Comparison between classification algorithms through Kappa statistics

At the point while two binary variables are endeavors by two people to quantify a similar object, you can utilize Cohen's Kappa [17] (regularly basically called Kappa) as a compute of understanding between the two individuals. Kappa measures the values of data esteems in the fundamental transverse of the table and afterward changes these qualities for the measure of assertion that might be required because of chance alone.

Now is one conceivable translation of Kappa [17][18].

- Poor conformity = Less than 0.20
- Fair conformity = 0.20 to 0.40
- Moderate conformity = 0.40 to 0.60
- Good conformity = 0.60 to 0.80
- Very good conformity = 0.80 to 1.00



Figure 4: Comparison between classification algorithms through MAE and RMSE

"Mean Absolute Error" (MAE) [19] and "Root mean squared error (RMSE") [19] are two of the most wellknown metrics used to measure accuracy for ceaseless variables. Not sure in the event that we envisioning it but rather we think there used to be a period when there were significantly more distributed MAE results. It appears that distributions we go over now generally utilize either RMSE or some version of R-squared.



Figure 5: Comparison between classification algorithms through RAE and RRSE

RAE [19] and RMSE [19] is a popular formula to measure the error rate of a regression display. However, it must be compared between models whose errors are measured in similar units.



Figure 6: Comparison between classification algorithms through Time in Seconds

Naive Bayes and Randamizable Filtered classifier had low time complexity compare to other algorithms but Randamizable Filtered classifier contains less correctly classified instances. Based on kappa statistics and CCI factor Attribute selected classifier is a best algorithm is suitable for breast cancer dataset.

 Table 2 Comparison between different Datasets with single
 algorithm through different factors

	Dataset Name	Factors										
SNo.		Data type	No. Instances	No. Attributes	CCI (instances & %)	ICI (instances & %)	Kappa Statistics	MAE	RMSE	RAE (%)	RRSE (%)	Time taken (Sec)
1	S/W Engineering Team Assessment & prediction in education setting	Sequential, Timeseries	74	116	50 (67.567%)	24 (32.432%)	0.0472	0.3341	0.4928	89.69	114.67	0.16
2	Diabetes	Multivariate	768	9	576 (75%)	192 (25%)	0.4218	0.324	0.424	71.46	89.13	0.14
3	Yeast	Multivariate	1483	1	0 (0%)	1483(100%)	0.0014	0.0014	0.026	100.03	100.037	0.05
4	Iris	Multivariate	149	1	3 (2.01%)	146 (97.98%)	0	0.013	0.083	100.33	10.37	0.001
5	Weather	Multivariate	14	5	6 (42.85%)	8(57.14%)	-0.365	0.45	0.55	95.99	113.27	0.02
6	Supermarket	Multivariate	4627	217	2948 (63.71%)	1679 (36.28%)	0	0.46	0.48	99.99	100	0.46
7	Vehicle	Multivariate	847	19	347 (40.96%)	500 (59.03%)	0.2118	0.2505	0.41	83.42	107.62	0.14
8	Heart-h	Multivariate	295	14	235 (79.66%)	60 (20.33%)	0.544	0.2063	0.3309	66.24	84.06	0.08

Above table explain about the different datasets applying on single algorithm is Attribute selected classifier through different factors. Each dataset contain different instances and attributes and also different data types. We contains different types of datasets are Sequential, Time series, Multivariate and different count of instances and attributes.



Figure 7: Comparison between different datasets through No. of instances and attributes.

Here simply show the number of instances in each dataset. We take number of datasets (S/W Engineering Team Assessment & prediction in education setting, Diabetes, Yeast, Iris, Weather, Supermarket, Vehicle, Heart-h), and each dataset had some instances and attributes, here supermarket dataset had the more instances and attributes compared to other datasets.



Figure 8: Comparison between different datasets through CCI and ICI factors

Heart-h [23] dataset had contain more correctly classified instances (CCI=77.66%) compared to other datasets especially super market dataset contains more instances but CCI is only 63.71%. but heart-h contains 235 instances and 14 attributes only.



Figure 9: Comparison between different datasets through Kappa Statistics

Heart-h dataset contains more kappa statistics compared to other datasets. Some algorithms had –ve kappa statistics. Kappa statistics value is more that means it contains more reliable.



Figure 10: Comparison between different datasets through RAE and RRSE



Figure 11: Comparison between different datasets through MAE and RMSE

We measure "Mean Absolute Error" (MAE) [19] and "Root Mean Squared Error" (RMSE) [19] on different datasets. Observation of above graph RMSE value is more to compare MAE to all the datasets Yeast (CCI=0%) and iris datasets contains very less attributes, so ignore that datasets. Compared to other datasets heart-h dataset contains maximum instances, attributes and less error rates.



Figure 12: Comparison between different datasets through Time in seconds

We observe except iris and yeast datasets, weather dataset build time is less but instances are only 14 and attributes are 5, so we don't consider this dataset. Next one is heart-h got good classification with in less time compared to other datasets.

III CONCLUSION

Super market dataset had more time complexity because it contains more instances and attributes. And another dataset is iris contains less time (0.001sec), it contains 149 instances and only one attribute and also correctly classified instances(CCI) rate is worst percentage (2.01%) so, we did not consider time complexity of iris. Another dataset is heart-h is best to compare other datasets because time complexity is less (0.08 sec) and CCI rate is high compared to ICI. And also kappa statistics also best (0.544). So finally we conclude attribute selected classifier is very suitable to heart-h dataset type is multivariate. So attribute selected classifier is very suitable to heart-h dataset.

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