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Analysis on Indian Education Towards Fully Automated Digitization and Decentralized Education System- A Machine Learning Approach

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Abstract: Education is the major aspect and basis of foundation development in any country. Education directs the right path as well as to build a good human character with all aspect of life. In 2018, India ranked 40th with the score 41.2. In the year 2019, India was on 35th rank on the overall index with a score of 53. This score was based on policy environment, teaching environment and socio-economic environment. National Education Policy NEP-2021 is mainly focused on skill based education system with more flexible learning for the students. Digitization is a way for speed up of the learning process of the students. In this paper, we have analyze the Indian education system in the context of auto digitization and decentralized control through different machine learning data analysis, prediction, feature extraction, classification, visualization tools.

Keywords: Algorithm, Decentralized Education, Digitization, Machine Learning.

1. INTRODUCTION

"Any activity wherever we are able to modify and decompose the worth chain, and move the work around, can get stirred around. Some individuals can say, 'Yes, however you cannot serve Maine a cut of meat.' True, however I will take the reservation for your table sitting anyplace within the world,"

Digital education is not about the videos or lectures on whiteboards by lecturers on the net. It's regarding applicable infrastructure, platforms, technology, tools, interactivity, content etc. we have not yet completely prepared for fully atomization in education. Government universities and colleges don't have the sufficient resource to supply digital education. Private educational sectors are totally different. In India, it will be a very complex task to implement fully digitized educational environment and infrastructures for all educational sectors[1][2].

Globalizations impact on institutional reform methods i.e. competitiveness-driven reforms. The reforms is classified into four categories - Decentralization, Standards, Improved management of institutional resources and Improved teacher enlisting and coaching.

Globalizations impact on academic reform methods i.e. competitiveness-driven reforms Decentralization- The purpose is to extend the management over curriculum and teaching ways of native communities and therefore the academics and principals of the schools themselves. An extension of such reforms is faculty selection and the privatization of academic delivery Standards[3].

higher focus of centralization that for The main reforms are the quest learning standards. Learning customary that an assist learner's academic programme aims to attention. The purpose of providing such standards (established by a central authority) is to administer clear signals of educational expectations to varsities and to folks within the hope, high standards can raise parent demands and school performance, Improved management of academic resources. Introducing new, high-yield resources that may make associate degree particularly giant distinction in student achievement at comparatively low value[4][5]. Better management and allocation of existing resources in colleges. Public education in developing countries ought to focus on increasing and rising basic education because the pay-off -the social rate of return to resources invested with at that level is higher than to resources invested with at the secondary and higher levels.

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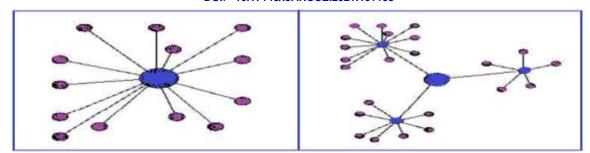


Fig.1 Centralized and Decentralized Education System

Government Universities and Colleges may opt for a completely decentralized implementation process which may not be completely implementable practically but may be possible gradually. Creation of syllabus, course content, practical content, based on current industry standard and market standard which are globally acceptable should be adaptable[6]. So Decentralization means the approach for each school which control by the Government partially only in the aspect of managerial task[7]. The country not only benefits from a wide diversity of theoretical and practical implementation strategies, but it is also able to run these processes simultaneously. Professors, faculties or any academician are afforded maximum opportunity for creativity in their research input output and thus may also have the way to take a new movement towards a new technology for the future[8]. Students benefit from different new technologies and build their mindset for the futuristic development and also have opportunity to be part of the future development process[9].

2. EXPERIMENTS AND OBSERVATIONS

We have taken a dataset name=percentage-of-schools-with-comps-2013-2016.csv.arff (Attribute Relation File Format) as our input and generate different results based on different machine learning algorithms.

2.1 Classifier NAME=weka.classifiers.functions.MultilayerPerceptron

SYNOPSIS=A classifier that uses backpropagation to learn a multi-layer perceptron to classify instances. The network can be built by hand or or set up using a simple heuristic. The network parameters can also be monitored and modified during training time. The nodes in this network are all sigmoid (except for when the class is numeric, in which case the output nodes become unthresholded linear units).

1 /			
Classifier Output === Run information ===			
Scheme: weka.classifiers.functions.MultilayerPerceptron -L 0.3 -M 0.2 -N 500 -V 0 -S 0 -E 20 -H a			
Relation: percentage-of-schools-with-comps-2013-2016-weka.filters.AllFilter-weka.filters.MultiFilter-			
weka.filters.AllFilter			
nstances: 110 Attributes: 13			
tate_UT year Primary_Only Primary_with_U_Primary			
rimary_with_U_Primary_Sec_HrSec U_Primary_Only			
J_Primary_With_Sec_HrSec Primary_with_U_Primary_Sec			
J_Primary_With_Sec Sec_Only			
ec_with_HrSec. HrSec_Only			
Il Schools			
Test mode: 10-fold cross-validation			
=== Cross-validation === === Summary ===			
Correlation coefficient 0.9944			
Iean absolute error2.0938			
Loot mean squared error 2.9674			
elative absolute error 9.222 %			
toot relative squared error 10.7302 %			
Total Number of Instances110			

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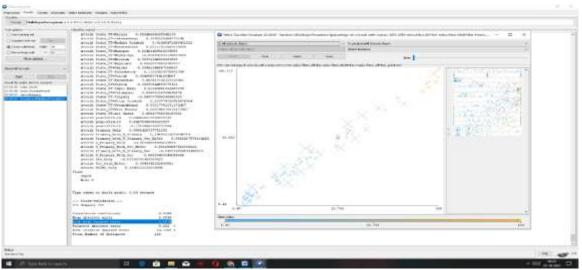


Fig.2 MultilayerPerceptron Classifier with Output Visualization

2.2 Classifier NAME=weka.classifiers.meta.Bagging

SYNOPSIS=Class for bagging a classifier to reduce variance. Can do classification and regression depending on the base learner.

Classifier Output === Run information ===

Scheme: weka.classifiers.meta.Bagging -P 100 -S 1 -num-slots 1 -I 10 -W weka.classifiers.trees.REPTree -- -M 2 -V 0.001 -N 3 -S 1 -L -1 -I 0.0 Relation: percentage-of-schools-with-comps-2013-2016-weka.filters.AllFilter-weka.filters.MultiFilter-

Freiation: percentage-of-schools-with-comps-2013-2010-weka.filters.AllFilter-weka.filters.MultiFilter Fweka.filters.AllFilter Instances: 110 Attributes: 13

mstances. 110 Au	1100003. 15		
State_UT yea	r Primary_Only	y Primary_with_U_Primary	
Primary_with_U_Primary_Sec_HrSec		U_Primary_Only	
U_Primary_With_Sec_HrSec		Primary_with_U_Primary_Sec	
U_Primary_With_Se	c	Sec_Only	
Sec_with_HrSec.		HrSec_Only	
All Schools			
Test mode: 10-fold	cross-validation		
=== Classifier model (full training set) ===			
Bagging with 10 iterations and base learner			
weka.classifiers.trees.REPTree -M 2 -V 0.001 -N 3 -S 1 -L -1 -I 0.0			
Time taken to build model: 0.03 seconds			
=== Cross-validation === === Summary ===			
Correlation coefficient	nt 0.9666		
Mean absolute error	4.6265		
Root mean squared e	rror 7.75	78	
Relative absolute error	or 20.376	9 %	
Root relative squared	l error 28.05	25 %	
Total Number of Inst	ances 110		

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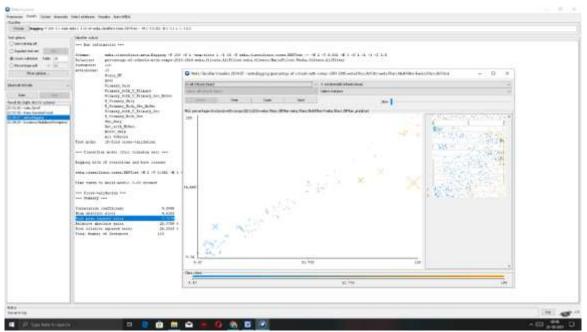


Fig.3 Bagging Classifier Based Prediction with Output Visualization

2.3 Classifier NAME=weka.classifiers.trees.RandomForest

SYNOPSIS=Class for constructing a forest of random trees. Classifier Output=== Run information === Scheme: weka.classifiers.trees.RandomForest -P 100 -I 100 -num-slots 1 -K 0 -M 1.0 -V 0.001 -S 1 Relation: percentage-of-schools-with-comps-2013-2016-weka.filters.AllFilter-weka.filters.MultiFilter-Fweka.filters.AllFilter Instances: 110 Attributes: 13 Primary_Only year Primary_with_U_Primary State UT Primary_with_U_Primary_Sec_HrSec U_Primary_Only U_Primary_With_Sec_HrSec Primary_with_U_Primary_Sec U_Primary_With_Sec Sec_Only Sec_with_HrSec. HrSec_Only All Schools Test mode: 10-fold cross-validation === Classifier model (full training set) === RandomForest Bagging with 100 iterations and base learner weka.classifiers.trees.RandomTree -K 0 -M 1.0 -V 0.001 -S 1 -do-not-check-capabilities Time taken to build model: 0.06 seconds === Cross-validation ====== Summary === Correlation coefficient 0.9931 Mean absolute error 3.3593 Root mean squared error 4.457 Relative absolute error 14.7953 % Root relative squared error 16.1167 % Total Number of Instances 110

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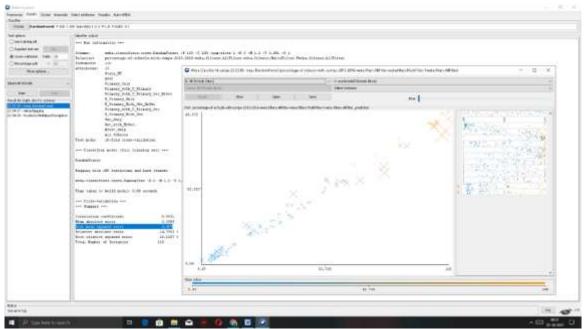


Fig.4 RandomForest Classifier Based Prediction with Output Visualization

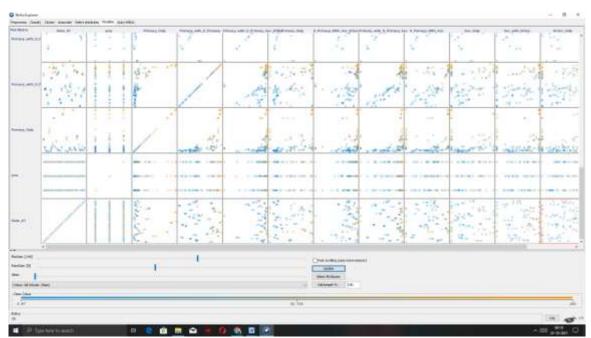


Fig.5 Visualization of % of Schools with Computers for all Students - Primary Schools to H. Secondary Schools

3. DISCUSSION

As we have taken 3 different algorithms and an educational dataset to know the digital and automated cases in different schools in the country. The first classifier that we have taken is the RandomForest in which we got accuracy rate of classified data as 95.543%. The second classifier that we have taken is the metaBagging in which we got accuracy rate of classified data as 92.2422%. The third classifier that we have taken is the MultilayerPerceptron in which we got accuracy rate of accuracy rate of classified data as 97.0326%.

Now from this analysis, it is found that number of computers with proper technological infrastructure should be developed gradually to make the educational system towards fully automated digitized and also there are lots of pros and cons regarding whether to make the education system centralized or decentralized. Now it's a great challenge for the higher educational department to make such policies to implement and provide such things to the universities and autonomous educational bodies for the development of the global education system[10].



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As results from experiments and observation shows that, the Multilayer Perceptron classifier algorithm is an efficient algorithm which gives more accurate results of the correctly classified instances of 97.0326 %.

4. CONCLUSION

No doubt, educational reforms should be needed time to time as per the demand for the global technological advancements. Education should meet the updated needs of a dynamical inconsistent world, society and new generation students. Decentralized educational reforms should be adopt for a global economical and developed country. Policies are an important concern or a challenge for the higher educational departments to implement decentralized and somewhere centralized time to time based on different criteria. Machine Learning will be an excellent tool to analysis, predict, classify and observation for this type of real issues in the educational reforms and for the digitization of the educational system.

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BIOGRAPHY



Mr. Dillip Narayan Sahu received his B.Sc. degree in physics from Sambalpur University, India, Master of Computer Application degree from Sambalpur University, India, M.Tech. degree in Computer Science from Sambalpur University, India, M-Phil degree Computer Science from MATS University, India, and continuing Ph.D. in Computer Science. His research area includes Artificial Intelligence, Machine Learning, Analysis and Design of Algorithms, Data Science, IoT. He has published many papers in National/International Conferences, Seminars, and Journals. Presently He is working as a Lecturer in the Dept. of MCA, Gangadhar Meher University, India.



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