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**SURVEY REPORT** 

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# A SURVEY ON PREDICTION OF AUTISM SPECTRUM DISORDER USING DATA SCIENCE TECHNIQUES

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*Abstract:* Autism Spectrum Disorder is a lifelong brain developmental disorder. Diagnosing the level of Autism and predicting the severity of the same are too complex, and it requires a depth analysis of the historical data on the autism patient. Nowadays, Data science techniques play a vital role in diagnosing autism. Decision Tree, Random Forest, Logistic Regression, Adaboost, Naïve Bayse, K-Nearest Neighbour, Support Vector Machine and etc., are the few techniques labeled under the roof of data science are used to predict such disorders. The paper aims to present a survey on the various models proposed by various researchers to predict the severity of autism using data science techniques.

Keywords: Autism Spectrum Disorder (ASD), Accuracy Parameters, ASD Dataset, Data Science Techniques and Data science Platform.

# I. INTRODUCTION

Autism is an abnormality in the brain structure or function. Autism appears before a child ages three because the brain tends to grow too fast during the first three years of life [1]. The following are the Autism patient-related symptoms:

- Complicated with verbal and non-verbal communication.
- Inability to participate in a conversation.
- Difficult with Social interaction.
- Difficult to develop friends with others, and prefer to play alone.
- Difficult to adjust to changes in the common surroundings environment.
- Tedious body movements such as spinning, head banging and flapping.
- Preoccupation with familiar objects.
- Make a little or inconsistent eye contact.

The cause of Autism is uncertain, but possible factors include hereditary data and difference in the development of certain brain functions, leading to impairment in cognitive and social aspects [2].

A child with Autism has three different levels such as lowfunctioning or Serve Autism, Moderate Autism and High-Functioning or Mild Autism [3]. Figure 1 represents the Autism types and their subtypes.



#### **Figure 1: Types of Autism**

Diagnosis of Autism is complicated in the early stages because of the more common early symptoms. There is no particular medical test, and diagnosis is made by observing patient symptoms and behavior through a historical dataset [4]. Hence, it is essential to have such data containing the details of Autism patients, including their symptoms, history, lab test data, vaccination, etc. Nowadays, there are too many automated techniques to diagnose Autism Spectrum Disorder like data science, machine learning, Artificial Intelligence, deep learning, etc. So, in this paper, we will briefly introduce Autism diagnoses using various data science techniques such as Decision Tree, Random Forest, Support Vector Machine, Adaboost, Glmboost, Convolutional Neural Network (CNN), Rule-Machine Learning (RML), Logistic regression, K-Nearest Neighbors.

This article is structured as, under the section literature review, discusses the various proposed models based on data science techniques linked with Autism. The methodology presents an overview of various data science techniques and a comparative study of different existing models based on features. The result and discussion section analyze the efficacy of the various current models used for predicting Autism. Finally, the conclusion section highlights the features and limitations of the enlisted techniques used for this study.

# **II.LITERATURE REVIEW**

Autism is a wrong connection between human brain cells. The efficacy of data science techniques is quite commendable in predicting the different types of autism disorders based on the syndrome. This section briefly presents the works related to the data science techniques used to predict the level of Autism.

C.S.Kanimozhiselvi et al. developed "Grading Autism children using machine learning techniques", They categorized Autism Spectrum Disorder patients into three different levels such as, high, moderate and low, based on Autism patientrelated symptoms. The implementation work was carried out in different stages: the first step of implementation is collecting Childhood Autism Rating Scale (CARS) based on case histories for preparing a real-world dataset. The second step is predicting the grade of Autism using some classification models consistent with CARS diagnostic criteria. Finally, evaluating the performance and predicting the different levels of Autism [5].

Kazi Shahrukh Omar et al. proposed a paper titled "A Machine Learning Approach to Predict Autism Spectrum Disorder", proposed an effective prediction model based on data science technique, and developed a mobile application for predicting Autism for the people of different age groups. For this, the researcher has merged the Random Forest-CART and Random Forest-ID3 techniques used to analyze the Autism dataset [6].

Tania Akter et al. proposed a "Machine Learning Based Models for Early Stage Detection of Autism Spectrum Disorders" model. It gathered the early detection of ASD dataset in different stages of life (toddler, child, adolescent and adult) and analyzed the results using a range of different classifiers to explore the significant features of Autism [7].

Suman Raja and Sarfaraz Masood proposed "Analysis and Detection of Autism Spectrum Disorder using Machine Learning Techniques". This research work aimed to build a machine learning model that predicts Autism using supervised machine learning algorithms. The steps involved were preprocessing of data, training, testing with specified models, evaluation of results, and predicting Autism [8].

FadiThabtah et al. developed "A New Machine Learning Model based on induction of rules for Autism detection", in which this article proposed a new machine learning method called Rule-Machine Learning (RML) that only detects autistic traits of different cases and controls but also offers users knowledge bases that can be utilized by domain experts in understanding the reasons behind the classification [9].

Jaber Alwidian et al. "Predicting Autism Spectrum Disorder using Machine Learning Technique". This prediction model utilized the Association Classification (AC) of data mining to predict whether an individual has an Autism disorder. They conducted a Comparative performance analysis on the seven Association Classification (CMAR, CBA, FACA, MCAR, FCBA, ECBA, and WCBA) techniques [10].

Roopa B. S and R. Manjunatha Prasad developed" Identification of Best Fit Learning Models Based on Calibration for Better Classification of Autism". Various supervised learning models were first tested on 1101 subjects with 530 ASD subjects and 571 Normal subjects. The performance was calibrated in terms of the brier score, which is a measure to predict autism in a probabilistic way [11].

Charlotte Küpper et al. proposed "Identifying Predictive Features of Autism Spectrum Disorder in a Clinical Sample of Adolescents and Adults using Machine Learning, "focusing on adolescents and adults as assessed with the ASDS module 4. It used the SVM classifier to examine whether ASD detection can be improved by identifying a subset of behavioral features from the ADOS module 4 in a routine clinical sample of N=673 high-functioning adolescents and adults with ASD(n=885) and non-ASD(n=288) [12].

Lakshmi B and Kala A, developed "Prediction of Autistic Spectrum Disorder Based on Behavioural Features Using Machine Learning," in which Neural Networks were chosen for implementing the Autism predictive system. The Neural Networks of the proposed system is a combination of linear and nonlinear functions that take up the vectors comprising various attributes defined in the Autism dataset [13].

Devika Varshini G and Chinnaiyan R, proposed "Optimized Machine Learning Classification Approaches for Prediction of Autism Spectrum Disorder" in which this paper the effectiveness of various machine learning algorithm and pre-processing techniques for the task of classification for medical dataset that are used for predicting the early Autism traits in toddlers and adults is evaluated [14].

The literature review discusses from various authors the techniques which can be utilized to determine Autism risk factors and in the section of methodology, a detailed study about Autism prediction will be carried out.

# III. MATERIALS AND METHODS

Autism is a pervasive developmental disorder. A variety of data science techniques that aid in the classification of different types of autism and their severity levels. The methodology section briefly outlines different classes of data science techniques used in predicting autism.

# A. Decision Tree Algorithm

A decision tree is the most powerful decision support tool that uses a tree-like model of decisions. The decision tree model has two steps in the process: the learning step and the prediction step.

- The first step is the learning step; the model is developed based on Autism training data.
- The second step is the prediction step; the model is used to predict the response for Autism data.

The decision tree algorithm has applied the Real-time Childhood Autism Rating Scale (CARS) dataset, and it gives the greatest accuracy of 1.00 for training data and 0.96 for test data. One of the major limitations of this algorithm is that it is not suitable for a large dataset [5], [15].

# **B.** Random Forest

A Random Forest is an ensemble learning method in which many decision trees are constructed and combined to get a more accurate prediction. The steps involved in the Random Forest algorithm are:

- Random Forest n random numbers of records are taken from the autism data set having k number of records.
- Individual decision trees are constructed for every sample, and every decision tree will produce an output.
- Finally, the result is measured based on Averaging for Classification or Majority Voting and Regression.

A prediction model is proposed to improve the performance that merges the Random Forest- CART model with the Random Forest model - ID3. The outcome of these merged models provides an effective and efficient approach to discover Autism, and it reduces the time and cost of Autism prediction. The drawback of this model is that it is only suitable for small datasets [6].

#### C. Support Vector Machine (SVM)

The Support Vector Machine works by mapping to a high dimensional feature space so that data points can be grouped, even when the data are not otherwise linearly separable. A separator between the categories is found, and then data are transformed in such a way that the separator could be drawn as a hyper plane. This SVM model gives numerous features of data analysis such as better handling of missing values, high accuracy and consumes the minimum time of prediction [7],[12],[13].

#### D. Adaboost

Adaboost is a popular boosting technique. It is a short form of adaptive boosting, which uses the same training set over and thus need not be large, but the classifiers should be simple so that they do not overfit. It measures the importance of features by calculating the increase in the model's prediction error after permuting the feature [7].

#### E. Glmboost

The Boosted General Linear Model by allowing the linear model to be related to the response variable through a link function and by allowing the magnitude of the variance of each dimension to be a function of its predicted value. [7].

# F. Convolutional Neural Network (CNN)

The Convolutional Neural Network is an interconnected network architecture for deep learning. Convolution is a mathematical operation that allocates the integration of two sets of information. Convolution is applied to the Autism dataset to filter the dataset and then produce a feature map. CNN produces highly accurate results for autism prediction and consumes the minimum time for prediction [8].

# G. Rule-Machine Learning (RML)

RML is based on covering classification, employing a search technique for rule discovery. The RML then evaluates the exposed rule and deletes any redundancies. A rule is represented as  $(A1,v1)\wedge(A2,v2)\wedge\ldots\wedge(An,vn)\rightarrow$ Cn where the predecessor is a combination of variable values, and the subsequent is a category value (ASD, No ASD). The merits of the RML algorithm are that it produces accurate results, consumes less time to predict the results, and is better at handling noisy data [9].

# H. Logistic Regression

Logistic regression is statistical software to understand the relationship between the dependent variable and one or more independent variables by estimating probabilities using a logistic regression equation. This type of analysis is then used for predicting the event and the overall performance of this technique shows good efficiency [11], [16].

# I. K-Nearest Neighbours (K-NN)

K-NN models are very useful in handling both regression and classification problems. The KNN model uses the feature of resemblance to predict the value of some new data points. The allocation of values to the new data points is how strongly it resembles the points in the training set. The steps involved in K-NN are:

- 1. Choose the K number of the neighbours.
- 2. Compute the Euclidean distance of K neighbours.
- 3. Take the K-NN as per the designed Euclidean distance.
- 4. Among these neighbours, count the data points in every category.
- 5. Allocate the new data points to that category holding the maximum neighbours.
- 6. Finally construct the K-NN model.

The outcome of this KNN model provides an effective and efficient approach to detect Autism, and it reduces the time and cost of predicting Autism [17].

Table 1 presents an overview of various existing techniques used for identifying the Autism disorder. It tabulates the study based on the parameters, platform, dataset, data science techniques used, and their features and limitations.

#### Table 1: Overview of Various Existing Models

Year	Parameters / Platform	Dataset / Techniques	Features	Limitations
2019	1. Accuracy	Dataset:	1. It	1. Only
	2.Precision	Real time	identifies	useful for
	3.Recall	Childhood	the early	small
		Autism Rating	stages of	datasets.
	Platform:	Scale (CARS)	Autism.	2. It has a
	Artificial	data	2. The	limited set
	Intelligence		decision	of
	-	Techniques:	tree has	classificatio
		1.Naive	the	n models.
		Bayesian,	highest	
		2.Decision	degree of	
		Tree,	accuracy.	

		3.K Nearest Neighbor		
		4.Support		
		Vector		
		Machine.		
2019	1.Accuracy	Dataset:	1. Offers	1. Less
	2.Specificity	AQ-10 dataset	an	suitable for
	3.Sensitivity		effective	large
	4.Precision	Techniques:	and	datasets.
	5.False	1.Merging	efficient	2. Inspection
	Positive	Random	method	applications
	Rate(FPR)	Forest-CART 2.Merging	for detecting	are not designed for
	Platform:	Random	Autism	the age
	Android	Forest-ID3	traits in	group of
	application		different	below three.
	with the help		age	
	of Amazon		groups.	
	Web Service		2. It	
	(AWS)		reduces	
			time and	
2019	1.Accuracy	Dataset:	cost. 1. Using a	1. Some of
2019	2.Kappa	1.Kaggle	variety of	the
	Statistics	2. UCI ML	feature	classifiers
	3.AUROC	Repository	selection	produced
	4.Sensititvity		and	inconsistent
	5.Specificity	Techniques:	ranking	results.
	6.Logloss	1.Adaboost	methods,	2. The
		2. FDA 3. C5.0	it is highly	currently offered ASD
	Platform:	4.Gimboost	predictive for	data was
	R	5.LDA	Autism.	insufficient
	Programming	6.MDA		to fully
	0 0	7.PDA		resolve this
		8.SVM		ASD
		9.CART		prediction.
2020	1.Specificity	Dataset:	1. It is	1. When
	2.Sensitivity 3.Accuracy	UCI ML Repository	preferable to handle	comparing SVM, ANN,
	5.Accuracy	Repository	missing	and CNN.
	Platform:	Techniques:	values.	SVM and
	Python	1.Logistic	2. It has a	ANN have
	-	Regression,	high	lower
		2.SVM,	degree of	accuracy.
		3.Naive	accuracy.	
		Bayes,	3. It takes	
		4.KNN, 5.ANN,	the least amount of	
		6.CNN	time.	
2020	1. Accuracy	Dataset:	1. Overall,	1. One of the
	2. Sensitivity	Data is	Machine	limitations
	3. Specificity	collected by a	Learning	of this work
	4. Precision	mobile	techniques	is that it
	5.F1-Measure	application	performed	does not
	Platform:	Techniques:	well.	include
	WEKA	1. RIPPER		instances involving
	,, <b>LIN</b> 1	2. RIDOR		toddlers,
		3. Nnge		which are
		4. Bagging		uncommon
		5.CART		and difficult
		6. RML		to obtain.
		7.PRISM		
		8.Adaboost		
2020	1 Accuracy	9. C4.5	1 It	1 It is not
2020	1. Accuracy 2.F-Measure		1. It demonstra	1. It is not appropriate
2020		9. C4.5 <b>Dataset:</b>	demonstra	1. It is not appropriate for other
2020	2.F-Measure	9. C4.5 Dataset: UCI ML		appropriate

			1	
		Techniques:	classificati	Autism
	Platform:	1.CMAR	on	medical data
	Java	2.CBA	accuracy.	sets.
	combination	3.FACA	2. It	
	with the	4.MCAR	shortens	
	WEKA tool	5.FCBA	the	
		6.ECBA	screening	
		7.WCBA	process.	
		7.WCD/1	3.It	
			identifies	
			the fewest	
			number of	
			ASD codes,	
			reducing	
			the	
			problem's	
			complexity.	
2020	1.Brierscore	Dataset:	1. Logistic	1. It
	2.Precision	Autism Brain	and SVM	consumes
	3.Recall	Imaging Data	models	more time.
	4.F-Score	Exchange	have the	
		Repository	lowest	
	Platform:	repository	Brierscore	
	1.Scikit	Techniques:	DIGSCOLE	
	2. Tensorflow	1.Decision	2.	
	2. rensorflow			
		Tree	Obtained	
		2.Random	the	
		Forest	greatest	
		3.SVM	accuracy	
		4.K-NN	(AUC	
		<ol><li>5.Naïve Bayes</li></ol>	Score of	
		6.Logistic	94.52).	
		Regression		
2020	1.Sensitivity	Dataset:	1. It	1. The
	2.Specificity	Data came	demonstra	ADOS was
	promony	from four	ted high	usually
	Platform:	specialized	specificity	considered
	R version	-	and	in clinical
		I	sensitivity	decision-
	Rstudio	clinics in		making, but
	1.1.456	Germany	2. It	it did not
			greatly	always
		Techniques:	aided in	determine
		1.SVM model	the	the
		2.ADOS	complicat	diagnosis.
		Algorithm	ed	
		~	diagnostic	
			process of	
			Autism.	
2020	1.Accuracy	Dataset:	1.	1. It is less
2020	2.Precision	UCI ML	1. Effective	suitable for
	3.F1 score	Repository	in handling	large
	4.Recall	<b>T</b>	handling	datasets.
	<b>DI</b> 10	Techniques:	the noise	2. It
	Platform:	Adam	data.	provided
	Python with			less
	Keras data			accuracy
	processing	neural		when using
	package	network.		a large
				dataset.
2020	1.Precision	Dataset:	1.	1. It
	2.F1 Score	UCI ML	Effective	provided
	3.Accuracy	Repository	for	less
	,	· · · · · · · · · · · · · · · · · · ·	preprocess	precision.
	Platform:	Techniques:	ing	r
	WEKA	1.Logistic	methods.	
	W LIXA	Regression	methous.	
		2. KNN		
		3.Random		
		Forest		

# IV. RESULT AND DISCUSSION

This section intends to list the performance comparison of various models used in this study proposed by various researchers down the years based on their accuracy.

# Table 2: Comparison of Enlisted Techniques Based on Accuracy

Proposers of The	Data Science	Accuracy
Model	Techniques Used	
C.S.Kanimozhiselvi et al.	1. Decision tree algorithm	The training set had the highest accuracy of 100 percent, and the test set had the highest accuracy of 96 percent.
Kazi Shahrukh Omar et al.	1. Merged Random Forest-CART and Decision Tree- CART algorithm	It predicted Autism, in the case of children, adolescents, and adults, with 92.26%, 93.78%, and 97.10% accuracy respectively.
Tania Akter et al.,	1.SVM dataset)(toddler dataset)2.Adaboost (children dataset)3.Glmboost (adolescent dataset)	All of these algorithms provided 100% accuracy.
Suman Raja and Sarfaraz Masood	1.Convolutional neural network (CNN)	It performed best for 99.56 %, 98.30 %, and 96.88 % accuracy in adults, children, and adolescents.
FadiThabtah et al.,	1.Rule-Machine Learning (RML)	It obtained 95% accuracy.
Jaber Alwidian et al.,	1.Weighted Classification Based on Association Rules (WCBA)	The accuracy rate was 86 %.
Roopa B. S and R. Manjunatha Prasad,	1. SVM 2.Logistic Regression	The classification accuracy was 88%.
Charlotte Küpper et al.,	1.SVM model	The rate of the accuracy of the model was 90%
Lakshmi B and Kala A,	1.Neural networks model	It provided 90% of accuracy.

Devika Varshini G	1. KNN	The accuracy of
and Chinnaiyan R,		KNN was
	2.Logistic	69.2%, while the
	regression	accuracy of
		logistic
	3. Random forest	regression and
		random forest
		classifiers was
		68.601%.

According to the table above, the model proposed by Tania Akter et al. provided the highest accuracy when compared to others. The Support Vector Machine, Adaboost, and Glmboost models continue to outperform in terms of autism prediction.

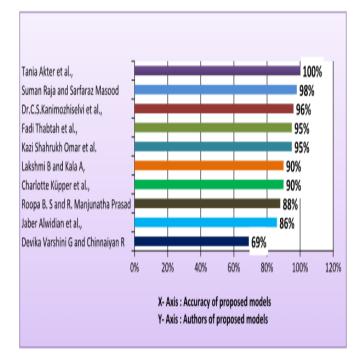


Figure 2: Performance comparison of the various enlisted techniques.

Finally, all of the proposed techniques are compared based on accuracy, which shows that the SVM, Adaboost, and Glmboost outperform with accuracy of 100 %, while the Convolutional neural network (CNN) came in second with accuracy of 98 percent. Finally, the Decision Tree Algorithm and Rule-Machine Learning (RML) came in last with an accuracy of 95%.

# V. CONCLUSION

This paper summarizes various data science techniques for the prediction of Autism Spectrum Disorder. SVM, Adaboost and Glmboost outperformed well than the other data science techniques. It was examined using six commonly used statistical measures, including Accuracy, Kappa Statistics, AUROC, Sensitivity, Specificity, and Logloss. The performance of these techniques was high in all of the proposed models, providing a strong indication of the potential power of data science techniques in serving such a critical domain. According to the findings of the study, one may strongly accept the importance of early detection of Autism Spectrum Disorder with the least amount of time, money, and complexity.

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