DECLARATION

We **RAGUL V S** and **SUDHARSAN VISVAK T** hereby declare that the Project Report entitled "IDENTIFYING COVID-19 OR NOT USING SUPERVISED MACHINE LEARNING ALGORITHM" done by me under the guidance of DR J CRUZ ANTONY MCA, PHD., and Sathyabama Institute of Science and Technology is submitted in partial fulfillment of the requirements for the award of Bachelor of Engineering degree in Computer Science and Engineering.

DATE:

PLACE:

SIGNATURE OF THE CANDIDATE

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Identify Covid-19 or not using SMLT

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LIST OF SYSMBOLS

S.N	NOTATION	NOTATION	DESCRIPTION
0	NAME		
1.	Class	Class Name -attribute + public -private # protected +operation +operation	Represents a collection of similar entities grouped together.
2.	Association	Class A NAME Class B Class A Class B	Associations represents static relationships between classes. Roles represents the way the two

				classes see each
				other.
3.	Actor	\frown		It aggregates
		C	\int	several classes into
		-	t	a single classes.
		/		
4.	Aggregation	Class A	Class A	Interaction
		1	1	between the system
		Class B	Class B	and external
				environment
5.	Relation(uses)		uses	Used for additional
	Relation(uses)			process
				communication.
6.	Relation			Extends
	(extends)	ext	ends	relationship is used
				when one use case
				is similar to
				another use case
				but does a bit
				more.
7.	Communication			Communication
				between various
				use cases.
8.	State			State of the
			State	process.

11. Control flow Represents various control flow between the states. 12. Decision box Represents decision making process from a constraint 13. Use case Interaction between the system and external environment. 14. Component Represents physical modules which is a collection of components. 15. Node Represents physical modules which are a collection of components.	9.	Initial State		Initial state of the
11. Control flow Represents various control flow between the states. 12. Decision box Represents decision making process from a constraint 13. Use case Interaction between the system and external environment. 14. Component Represents physical modules which is a collection of components. 15. Node Represents physical modules which are a collection of collecti			$\bigcirc \longrightarrow$	object
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physical modules which are a collection of				components.
which are a collection of	15.	Node		Represents
collection of				physical modules
				which are a
components				collection of
				components

16.	Data	\bigcirc	A circle in DFD
	Process/State		represents a state
			or process which
			has been triggered
			due to some event
			or action.
17.	External entity		Represents external
			entities such as
			keyboard, sensors
			etc.
18.	Transition		Represents
		.	communication
			that occurs
			between processes.
19.	Object Lifeline		Represents the
			vertical dimensions
			that the object
			communications.
20.	Message	Message	Represents the
			message
			exchanged.

1. ABSTRACT:

The Coronavirus disease 2019 (COVID-19) pandemic, which originated in Wuhan China, has had disastrous effects on the global community and has overburdened advanced healthcare systems throughout the world, WHO is continuously monitoring and responding to this pandemic. The current rapid and exponential rise in the number of patients has necessitated efficient and quick prediction of the possible outcome of an infected patient for appropriate treatment using AI techniques. The aim is to predict machine learning based techniques for covid-19 recovery chances possible or not prediction results in best accuracy. The analysis of dataset is done by supervised machine learning technique(SMLT) to capture several information's like, variable identification, uni-variate analysis, bi-variate and multi-variate analysis, missing value treatments and analyze the data validation, data cleaning/preparing and data visualization will be done on the entire given dataset. To propose a machine learning-based method to accurately predict recovery chances by prediction results in the form of whether the covid-19 patient precondition.

2. EXISTING SYSTEM:

In December 2019, a pandemic named COVID-19 broke out in Wuhan, China, and in a few weeks, it spread to more than 200 countries worldwide. Every country infected with the disease started taking necessary measures to stop the spread and provide the best possible medical facilities to infected patients and take precautionary measures to control the spread. As the infection spread was exponential, there arose a need to model infection spread patterns to estimate the patient volume computationally. Such patients' estimation is the key to the necessary actions that local governments may take to counter the spread, control hospital load, and resource allocations. This article has used long short-term memory (LSTM) to predict the volume of COVID-19 patients in Pakistan. LSTM is a particular type of recurrent neural network (RNN) used for classification, prediction, and regression tasks. We have trained the RNN model on Covid-19 data (March 2020 to May 2020) of Pakistan and predict the Covid-19 Percentage of Positive Patients for June 2020. Finally, we have calculated the mean absolute percentage error (MAPE) to find the model's prediction effectiveness on different LSTM units, batch size, and epochs. Predicted patients are also compared with a prediction model for the same duration, and results revealed that the predicted patients' count of the proposed model is much closer to the actual patient count.

2.1 Drawbacks:

The existing method is only patient count and it does not classify whether covid or not the patient is recovered or not.

Accuracy, Recall F1 score metrics are not calculated and machine learning algorithms are not applied.

3. INTRODUCTION

3.1 DATA SCIENCE:

Data science is an interdisciplinary field that uses scientific methods, processes, algorithms and systems to extract knowledge and insights from structured and unstructured data, and apply knowledge and actionable insights from data across a broad range of application domains.