## **LITERATURE SURVEY**

## FERTILIZERS RECOMMENDATION SYSTEM FOR DISEASE PREDICTION

S.NO	TITLE &	YEAR	TECHNIQUE	PROPOSED SYSTEM
	AUTHOR			
1	Leaf	2015	Artificial Neural	In this paper, a method is proposed for the extraction of shape features from leaf
	Identification		Network,	images. A classifier named as an Artificial Neural Network (ANN) is trained to
	Using Feature		MATLAB	identify the exact leaf class. It is done to attain high efficiency with less
	Extraction and			computational complexity. This work has been tested for the accuracy of network
	Neural Network			with different combination of image features. The results are tested on 80 leaf
	Satnam Singh, M.			images. It is revealed that this method gives 98.8% accuracy with a minimum of
	S. Bhamrah			seven input features. This approach is more promising for leaf identification
				systems that have minimum input and demand less computation time. Neural
				Network also developed and trained to classify and recognize plants from leaf
				images. This work is implemented using the image processing and neural network
				toolboxes in MATLAB.
2	IOT based Crop	2020	Crop	The ML and IoT based suggestions will significantly educate the farmer and help
	Recommendation,		recommendation	them minimize costs and make strategic decisions by replacing intuition and
	Crop Disease		system, crop	passed-down knowledge with
	Prediction, and Its		disease	far more reliable data-driven ML models.
	Solution - Rani		prediction,	
	Holambe, Pooja		Internet of	
	Patil, Padmaja			

	Pawar,		Things, Machine	
	Hrushikesh Joshi,		Learning	
	Saurabh Salunkhe		8	
3	Farmer's	2022	Image Analysis,	A user-friendly web application system based on machine learning and web-
	Assistant:		Deep Learning,	scraping called the 'Farmer's Assistant'. With our system, we are successfully able
	A Machine		Machine Learning	to provide several features - crop recommendation using Random Forest algorithm,
	Learning Based			fertilizer recommendation using a rule-based classification system, and crop
	Application for			disease detection using Efficient Net model on leaf images.
	Agricultural			
	Solutions- Shloka			
	Gupta, Aparna			
	Bhonde, Akshay			
	Chopade, Nishit			
	Jain			
4	Soil Based	2021	Long- or Short-	The proposed system was able to analyse the soil nutrient type efficiently, kind of
	Fertilizer		Term Memory	leaf disease present in the crop and predict the fertilizer in a proficient manner. The
	Recommendation		algorithm.	approach was flexible, and can be extended to the needs of the users in a better
	System for Crop			manner.
	Disease			
	Prediction System			
	- P. Pandi			
	Selvi, P.			
	Poornima			
5	Classification of	2020	Deep Learning	A custom CNN architecture is proposed in this article to classify ladies finger plant
	lady's finger plant			leaf image into three categories namely healthy, disease and leaf burn. The dataset
	leaf using deep			contains 1088 samples of lady's finger plant leaf images, among which 457 images
	learning			are identified as healthy (non-disease) leaves, 509 images as disease and pest
				infected and 122 images as leaf burn due to fertilizer overdose. The images were

	L. Selvam, P.			taken directly from the agriculture farms of different villages of Tiruvannamalai
	Kavitha			district, Tamil Nadu, India. The proposed CNN architecture achieved 96%
				classification accuracy.
6	Plant Leaf	2021	FastAI Image	The motivation of this paper therefore is to solve all these issues by building a
	Disease		Classification	light-weight and cost-efficient deep learning architecture with the proposed
	Recognition			DenseNet-121 model that classifies leaf images from a dataset called 'PlantDoc'
	Using FastAI			across 28 classes with 1874 training images and 468 validation images. A separate
	Image			test dataset is held out only for checking model performance on unknown data.
	Classification			Implementation is done using Fastai framework, because of its faster
	Aditya			computational power, easy workflow, and unique data cleaning functionalities.
	Chakraborty,			Overall, the classification accuracy achieved is 92.5%.
	Debarun Kumer,			
	K. V. Deeba			
7	Corn Leaf	2021	K-means	This paper proposes a method based on K-means clustering and an improved deep
	Diseases		clustering and an	learning model for accurately diagnosing three common diseases of corn leaves:
	Diagnosis Based		improved deep	grey spot, leaf spot, and rust. First, to diagnose three diseases, use the K-means
	on K-Means		learning model	algorithm to cluster sample images and then feed them into the improved deep
	Clustering and			learning model. This paper investigates the impact of various k values (2, 4, 8, 16,
	Deep Learning			32, and 64) and models (VGG-16, ResNet18, Inception v3, VGG-19, and the
	Helong Yu,			improved deep learning model) on corn disease diagnosis. The experiment results
	Jiawen Liu,			indicate that the method has the most significant identification effect on 32-means
	Chengcheng			samples, and the diagnostic recall of leaf spot, rust, and gray spot disease is 89.24
	Chen, Ali Asghar			%, 100 %, and 90.95 %, respectively. Similarly, VGG-16 and ResNet18 also
	Heidari, Qian			achieve the best diagnostic results on 32-means samples, and their average
	Zhang, Huiling			diagnostic accuracy is 84.42% and 83.75%. In addition, Inception v3 (83.05%) and
	Chen, Majdi M.			VGG-19 (82.63%) perform best on the 64-means samples. For the three corn
	Mafarja, H.			diseases, the approach cited in this paper has an average diagnostic accuracy of
	Turabieh			

1		1	
			93%. It has a more significant diagnostic effect than the other four approaches and
			can be applied to the agricultural field to protect crops.
Detecting	2022	Machine Learning	The sample images are collected and dataset of the turmeric leaf is prepared. It is
Turmeric		Algorithm	an image processing technique. The conversion of image is done by analysing the
Taphrina			image. Plant illness ID is the main area in farming. The ID of plant sicknesses
Maculans Disease			require close observing, and subsequently this paper embraces advances to oversee
using Machine			turmeric plant infections brought about by organisms to empower creation of top-
Learning			notch crop yields. Different picture handling and ML methods are utilized to
Algorithms			recognize and arrange the illnesses in turmeric leaf. The leaf pictures of various
C. Vanitha, S.			classes are pre-handled and sectioned to advance productive component extraction.
Malathy, P.			ML calculations like k-implies calculation are utilized.
Shenbagavalli, S.			
Krishna, K. Kavin			
Plant Disease	2021	Deep Learning	In this paper, we applied "Convnets" for plant disease detection and classification.
Prediction and		ConvNets	We collected a Plant Village dataset from Kaggle. It contains images of 15
classification			different classes of plant leaves of three different plants potato, pepper, tomato. We
using Deep			divided the dataset into three datasets and applied Convnets on three datasets. We
Learning			achieved an accuracy of 98.3%,98.5%,95% for potato plant disease detection,
ConvNets			pepper plant disease detection, tomato plant disease detection. Experimental results
A. Lakshmanarao,			have shown that our model achieved a good accuracy rate for plant leaf disease
M. Babu, T.			detection and classification.
Kiran			
Multi-Level Deep	2021	Deep Learning	In this research, a multi-level deep learning model for potato leaf disease
Learning Model			recognition has developed. At the first level, it extracts the potato leaves from the
for Potato Leaf			potato plant image using the YOLOv5 image segmentation technique. At the
Disease			second level, a novel deep learning technique has been developed using a
Recognition			convolutional neural network to detect the early blight and late blight potato
			diseases from potato leaf images. The proposed potato leaf disease detection model
	Turmeric Taphrina Maculans Disease using Machine Learning Algorithms C. Vanitha, S. Malathy, P. Shenbagavalli, S. Krishna, K. Kavin Plant Disease Prediction and classification using Deep Learning ConvNets A. Lakshmanarao, M. Babu, T. Kiran Multi-Level Deep Learning Model for Potato Leaf Disease	Turmeric Taphrina Maculans Disease using Machine Learning Algorithms C. Vanitha, S. Malathy, P. Shenbagavalli, S. Krishna, K. Kavin  Plant Disease Prediction and classification using Deep Learning ConvNets A. Lakshmanarao, M. Babu, T. Kiran  Multi-Level Deep Learning Model for Potato Leaf Disease	Turmeric Taphrina Maculans Disease using Machine Learning Algorithms C. Vanitha, S. Malathy, P. Shenbagavalli, S. Krishna, K. Kavin  Plant Disease Prediction and classification using Deep Learning ConvNets A. Lakshmanarao, M. Babu, T. Kiran  Multi-Level Deep Learning Model for Potato Leaf Disease  Algorithm  Algorithm  Algorithm  Algorithm  Algorithm  Algorithm  Deep Learning  CovVnets  Deep Learning  ConvNets  Deep Learning  Deep Learning  Deep Learning

Javed Rashid, I.	was trained and tested on a potato leaf disease dataset. The potato leaf disease
Khan, Ghulam	dataset contains 4062 images collected from the Central Punjab region of Pakistan.
Ali, Sultan H.	The proposed deep learning technique achieved 99.75% accuracy on the potato
Almotiri,	leaf disease dataset. The performance of the proposed techniques was also
Mohammed A.	evaluated on the Plant Village dataset. The proposed technique is also compared
Alghamdi, Khalid	with the state-of-the-art models and achieved significantly concerning the accuracy
Masood	and computational cost.

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