

LIVESTOCK DISEASE RISK FOREWARNING BULLETIN

Powered by Artificial Intelligence

PUBLISHED BY: DIRECTOR ICAR-NIVEDI

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JUNE

NOW LINKED WITH FARMER REGISTRATION AND UNIFIED BENEFICIARY INFORMATION SYSTEM (FRUITS)

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Citation: Suresh K P, Hemadri D, Patil S S, Krishnamoorthy P, Siju S J and Shome B R. Livestock Disease Risk Forewarning Bulletin – June 2022, ICAR-NIVEDI, Bengaluru, 10(04): 1-122.

Month & Year: April, 2022.

Published by: Director, ICAR- National Institute of Veterinary Epidemiology and Disease Informatics (NIVEDI), Yelahanka, Bengaluru-560064.

PME NUMBER: F.No.13/NIVEDI/PMEC/Forewarning Bulletin/2022-23/01

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Printed by Naveen Printers No 155 CHS, 4th Phase, Yelahanka New Town, Bengaluru, Karnataka 560064 E-Mail Id: naveenprinters04@gmail.com

Front Page Design by: Ms. Apoorva Hemadri

Disclaimer

The forewarnings are based on the retrospective disease data available in the NADRES database. Hence, for those states wherein data is limited/less, the forewarning may not be realistic. Further the forewarning will not take into consideration the control measures that are *in situ*.

Acknowledgement

I would like to acknowledge the constant support and inspiration from Hon'ble Secretary, DARE and Director General, ICAR, Government of India, New Delhi.

I would like to express sincere everlasting gratitude to Hon'ble Deputy Director General (Animal Science) for his constant encouragement, support and guidance.

I would also like to express sincere gratitude to the Department of Animal Husbandry and Dairying (DAHD), Ministry of Agriculture and Farmers Welfare, Government of India for providing the livestock population data for the preparation of this bulletin.

Animal Husbandry Departments of state governments and also AICRP on ADMAS centres are gratefully acknowledged for the timely submission of reports of livestock disease outbreak data. I am thankful to all the scientific and technical staff of ICAR-NIVEDI for their feedback and support. I sincerely acknowledge the Statistical Division of DAHD for providing the data on livestock census.

Furthermore, I would also like to acknowledge with much appreciation the crucial role of Dr. K. P. Suresh, Principal Scientist and support received from the scientists, Dr. D. Hemadri, Dr. S.S. Patil, Dr. P. Krishnamoorthy, Dr. S.J. Siju and SRF's, JRF's, Young Professionals and other contractual staff working in Disease Informatics Lab/Spatial Epidemiology Lab in preparing this monthly bulletin.

Director (Acting) ICAR- NIVEDI

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1. About the Bulletin...

Livestock sector plays a crucial role in the rural economy of India as around 20.5 million people depend upon livestock for their livelihood. Even though the investment in the livestock sector is meagre, tremendous achievements have been observed in the sector during the last decade. As it is an important component in poverty alleviation programmes, continuous emphasis is being laid on this sector for enhancing the quality of the primary and secondary products in the international market, which in turn demands improved animal health. Therefore, livestock development programmes cannot succeed unless a well-organized animal health service is built up and in place for safeguarding the livestock against economically important diseases.

India has made a noteworthy success in the eradication of Rinderpest (RP), CBPP, AHS and Dourine. However, there are several other infectious and non-infectious diseases prevailing in the country causing huge annual economic loss. Prevention, control and eradication of the animal diseases need a thorough understanding of the epidemiology as well their economic impact.

National Institute of Veterinary Epidemiology and Disease Informatics (NIVEDI) has the mandate to carry out research activities in the area of veterinary epidemiology and disease informatics. With the eradication of RP successfully, India has not only proved its ability to face the challenges but also to succeed, despite various limitations. Similar efforts are needed to control and eradicate diseases like FMD, PPR, Brucellosis, CSF, HS etc., which cause huge economic loss annually to the livestock industry. To this end, ICAR-NIVEDI has identified 13 priority diseases, based on the past incidence patterns and has built a strong database of these diseases. The database, which is the backbone of the National Animal Disease Referral Expert System (NADRESv2), is used for providing monthly livestock disease forewarning, which is compiled in this monthly bulletin to alert the animal husbandry departments, both at the National/state level, to take appropriate control measures. We hope users/stakeholders find this bulletin useful in their quest to control livestock diseases. This forewarning bulletin will assist the field Veterinarians in adopting appropriate preventive and control measures, thereby reducing the occurrence of livestock disease outbreaks.

2. SUMMARY OF THE FOREWARNING BULLETIN....

The association between infectious diseases and the climate was known from ancient times. Hippocrates observed in the 5th century that epidemics were associated with natural phenomena rather than divinities or demons. In modern times, our increasing capabilities to detect and predict climate variations joined with growing evidence for global climate change, have powered interest in understanding the impacts of climate on animal health, particularly the emergence and transmission of infectious disease agents. Simple reasoning suggests that climate can affect infectious disease patterns because disease agents (viruses, bacteria, and parasites) and their vectors are sensitive to temperature, moisture, and other ambient environmental conditions.

India being an agriculture-based country, the livestock sector plays a vital role in contributing to the economy. A robust reporting and forewarning system enable the concerned authorities in disease preparedness and awareness of the risk associated with livestock disease. Therefore, the economic loss due to morbidity and mortality of the animals is reduced thereby helps to increase the productivity in terms of egg, meat, and dairy products. National Animal Disease Referral Expert System database is a weatherbased forewarning system enabled with an artificial intelligence system developed by ICAR- National Institute of Veterinary Epidemiology & Disease Informatics Bengaluru, Karnataka state, India that forecast potential threats from pathogens two months in advance to provide the stakeholders with sufficient timeline for awareness and preparedness to act. Artificial Intelligence (AI) and Machine Learning (ML) models use the programmed algorithms that receive and analyse input data to predict output (Infectious risk prediction) values within an acceptable range. As new data fed into these algorithms, they learn and optimize their operations to improve performance, developing intelligence over time.

Summary

The livestock disease forecasting for June 2022 revealed Jharkhand, Assam, Karnataka, Kerala and West Bengal as the top five states with high predicted livestock disease outbreaks.

Among the predicted diseases, control programmes are in full swing for FMD and PPR in the country and due attention is demanded by the predicted disease outbreaks of these diseases. Among the expected disease outbreaks, the predicted FMD outbreaks are high in Kerala (14), Karnataka (13) and Jharkhand (11) and whereas predicted PPR outbreaks are high in Jharkhand (14). Further the co-occurrence of FMD and HS can be expected in Himachal Pradesh, Jharkhand, Karnataka, Kerala, Meghalaya, Odisha, and West Bengal. Among the different diseases in livestock, the predicted outbreaks are expected to be high for Black Quarter (67), Fascioliasis (58) and FMD (55).

The major challenges for the effective disease control programme being the lack of thorough understanding about the complexity of disease dynamics, wide host range of pathogens, widening of niche of pathogens due to climate change etc. The effective control programme for major livestock diseases in the country can be efficiently addressed by planning and execution of available control measures in the high risk areas and routine surveillance and monitoring of diseases.

		1			1					a	S&G				Total number of
State Name Anthrax Babesiosis I	Babesiosis		-	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	Pox	SF	Theileriosis	Trypanosomosis	diseases predicted
Andaman and Nicobar 0 0 0	0		0		0	0	3	1	0	0	0	0	0	0	04
Andhra Pradesh402	0		2		0	0	0	0	0	0	2	0	0	0	08
Arunachal Pradesh 0 1 0	1		0		0	0	4	2	0	0	0	0	0	0	07
Assam 1 1 19	1		19		0	2	14	0	11	2	5	6	3	0	67
Bihar 0 0 0	0		0		0	0	1	0	0	0	0	0	0	1	02
Chandigarh 0 0 0	0		0		0	0	0	0	0	0	0	0	0	0	00
Chhattisgarh 0 0 0	0		0		0	0	0	0	0	0	0	0	0	0	00
Dadra and Nagar Haveli 0 0 0	0		0		0	0	0	0	0	0	0	0	0	0	00
Daman and Diu 0 0 0	0		0		0	0	0	0	0	0	0	0	0	0	00
0 0 0	0 0	0			0	0	0	0	0	0	0	0	0	0	00
Gujarat 0 0 0	0 0	0		0		0	0	0	0	1	0	0	1	0	02
0 0	0 0	0		0		1	0	0	0	1	0	2	1	0	05
Himachal Pradesh 0 0 0	0 0	0		0		0	0	2	1	3	1	0	0	0	07
Jammu and Kashmir 0 0 0	0 0	0		0		0	0	0	0	0	6	0	0	0	06
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0 0	0 3	3		0		0	9	0	1	0	0	4	0	0	14
Meghalaya 2 0 3 0	0 3	3		0		0	1	6	4	0	2	5	0	0	23
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0 0 0	0 0	0			0	0	0	0	0	0	0	5	0	0	05
Delhi 0 0 0	0 0	0	_		0	0	0	0	0	0	0	0	0	0	00
0	0		3		0	0	0	1	1	4	1	0	2	-1	17
erry 0 3	3		0		0	0	2	0	0	0	0	0	0	0	05
0 1 0	1 0	0	2		0	0	0	0	0	0	0	0	1	0	02
an 0 1 0	1 0	0			0	2	0	2	0	1	1	0	0	0	07
Sikkim 0 0 0 0	0 0	0		0	0	0	0	0	0	0	0	0	0	0	00
Tamil Nadu 3 0 3 0	0 3	3		0		0	0	0	0	1	2	0	0	0	09
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Uttarakhand 0 0 0	0		0		0	0	0	0	0	2	0	0	0	0	02
West Bengal 2 1 11	1	1 11	11		0	0	2	2	8	7	2	0	9	2	43
Total number of districts likely to 28 35 67 report.	35	201	67		0	18	58	55	46	<u>42</u>	3 <mark>9</mark>	30	51	44	513
	-		_		1										

Table S1. Summary of State wise Livestock Disease forewarning for June, 2022

3. INTRODUCTION TO NADRES v2

The geographic and seasonal distribution of many infectious diseases are associated with climate and therefore the possibility of using seasonal climate forecasts as predictive indicators in disease early warning system (EWS) became imminent. In this context, ICAR-NIVEDI, in its quest for achieving better livestock health, had developed an interactive web portal named "National Animal Disease Referral Expert System (NADRES)" during early part of the first decade of the millennium. The web portal, which was developed from the financial support of National Agricultural Technology Project, was launched in the year 2005. The portal which is interactive, allows the user/stakeholder to access livestock disease forewarning (n=13) at the district level for entire country two months in advance. The portal which was initially built on oracle platform was later changed to MySQL platform to store the administrator provided disease information and other relevant meteorological and risk factor information. However, with the availability of remote sensed satellite images and the advancement in information technology and statistical algorithms, the upgradation of NADRES became inevitable. To this end, a newer version of NADRES (NADRES *V2*) has been developed and is ready for release.

How it is different from previous version?

In brief, it can be said that NADRES v_2 underwent a sea change not only in its internal structure but also in its physical design. As a result, now the central menu bar consists of Home, About us, Risk factors, Analysis, Livestock disease, post prediction validation and contact details. Risk factors menu comprises of details on resolution, time interval, units and source of 11 meteorological and 5 remote sensing parameters. Analytics menu has various analysis options. The newly created livestock disease menu has the details regarding species affected, clinical signs and preventive measures to be adopted for the 13 economically important diseases. Post prediction validation menu contains the outbreak reports vs prediction. The menu bar on the RHS tabs include online GIS, state wise Livestock disease forecast, district wise Livestock disease forecast, Epi-calculator, download links for mobile app, etc. The website now hosts disease maps in the form of choropleth maps for 13 diseases in two time periods (1990-2000 and 2000-2018). Similarly, disease trends plots exhibit periodic regression plots providing future trend for the disease. On the LHS, Login menu is provided for authorized persons to login and enter disease details and other related parameters. Disease maps provide choropleth maps for 13 diseases in two time periods (1990-2000 and 2000-2018) is presented. Disease trends- Periodic regression plots are exhibited for prediction of the diseases. Auto-messaging option has been created to send the reminders in the form of text messages to concerned PI's and Co-PI's of AICRP centers for submission of outbreak reports. This message is sent weekly to all the concerned officials. Additionally, a message is sent to the concerned veterinary officers in Karnataka for initiation of preventive measures for the forewarned diseases at the block level. Plans are in place to incorporate farmers' and local vets' mobile numbers in to the list so that they may be asked to initiate preventive measures for the forewarned diseases.



Fig 3.1. NADRES V2 Home page

The forewarning methodology used is unique and has not been used earlier for livestock disease forewarning in India. Following few paragraphs describe about the forewarning methodology used. It is a well-known fact that weather plays an important role in the precipitation of many diseases and therefore, the climatic parameters such as land surface temperature (LST), precipitation, wind velocity, humidity etc are considered as risk parameters. These parameters along with other non-climatic parameters such as livestock population, density, Normalized Differential Vegetation Index (NDVI), soil moisture constitute the overall risk parameters. A total of 24 such parameters are collected/generated at village level and then aggregated to district level before these are used for analysis.

In addition to the output provided at interactive web portal, the NADRES output are also published in the form of monthly livestock disease forewarning bulletins. The prediction results come with a disclaimer that forewarnings do not take into account of the control measures that already in situ and also may not be realistic for those regions where the data is either unavailable or limited. This bulletin provides the likely occurrence of the 13 shortlisted diseases two months in advance at the district level, disease forewarning maps, prediction accuracy, details on diseases, species affected, clinical signs and its preventive measures.

In summary, it can be said that NADRES $_{V2}$ has underwent substantial changes not only in its internal structure but also in its physical design and can be a useful tool for visitors of the website, farmers, vets, policy makers etc.

4. Forewarning Methodology

I. Materials

Livestock disease data

Previous 10 years' livestock disease outbreak data retrieved from the NADRES database linked with Risk factors data.

Livestock population data

The population data at village level for five major livestock species viz., cattle, buffalo, sheep, goat and pigs were obtained from 20th Livestock census (2019) from Department of statistics, DAHD, GOI.

	Species-wise &	Category-wis	e Livestock Pop	ulation (in thou	isands)
CIN				D	0.4 C1
Sl No	Species	Category	Population in 2012	Population in 2019	% Change
1	Cattle	Exotic	39732	51356	29.3
		Indigenous	151172	142106	-6
		Total	190904	193462	1.3
2	Buffalo	Total	108702	109852	1.1
3	Sheep	Exotic	3781	4088	8.1
		Indigenous	61288	70172	14.5
		Total	65069	74260	14.1
4	Goat	Total	135173	148885	10.1
5	Pig	Exotic	2456	1897	-22.8
		Indigenous	7837	7159	-8.7
		Total	10293	9056	-12
6	Yaks	Total	77	58	-24.7
7	Mithuns	Total	298	386	29.5
8	Horses & Ponies	Total	625	342	-45.3
9	Mules	Total	196	84	-57.1
10	Donkeys	Total	319	124	-61.1
11	Camels	Total	400	252	-37
Total L	ivestock		512056	536761	4.8

Meteorological and Remotely Sensed Data:

The parameters such as air temperature (⁰C), perceptible water (mm), pressure (millibar), relative humidity (%) and sea level pressure (millibar) were extracted from National Centre for environmental prediction (NCEP). The parameters such as potential evapotranspiration (PET), Enhanced Vegetation Index (EVI), Leaf Area Index (LAI), Land Surface Temperature (LST), Normalised Difference Vegetation Index (NDVI) were extracted from remote sensed images from MODIS website (<u>https://modis.gsfc.nasa.gov/</u>). In brief, the MODIS products from NASA-TERRA satellite was downloaded for the Indian locations by specifying the tiles (H24V5, H25V6, H24V6, H24V7, H25V7, H25V8, H26V7, H26V6) from 2001 to till date.

The details are given below;

PRODUCT	Science Data Sets (HDF Layers)
MOD15A2H	Lai_500m(Leaf area index) 8 days average
MOD16A2	PET_500m (Total Potential Evapotranspiration) 8 days average
MOD11A2	LST_Day_1km (Daytime Land Surface Temperature) 8 days average
MOD13A1	i. 500m 16 days NDVI (Normalized Difference Vegetation Index)ii. Enhanced Vegetation Index (EVI) 16 days average

The downloaded HDF files (Datasets, which are multidimensional arrays (layers) of a homogeneous type) were converted to GeoTIFF files (single layer data) using R packages, which were later used to extract the parameters by linking it with the sinusoidal values of the Indian villages. The scale factors were multiplied for the extracted values as specified by the MODIS data products to get the values of the parameters. As shown above, the atmospherically corrected NDVI was collected on 16-day interval at 250-meter resolution using MODIS product MOD13A1 and LST was collected on 8-day interval using MOD11A2 at 1 KM resolution.

The parameters such as rainfall, soil moisture and wind speed were obtained from Global Land Data Assimilation System of NASA (<u>https://disc.gsfc.nasa.gov</u>). The remaining parameters were downloaded from climatic research unit (CRU) of University of East Anglia website. It is worth mentioning that the entire process of extraction, assimilation, processing and aligning have been done using R programming language and R environment. After aligning the climatic and non-climatic data with the disease and the livestock population data (aggregated at the district level), the statistical analysis was performed in the R environment.

Initially, two regression models and six machine learning models were applied to test their suitability to fit the data and in all, three models; one regression model (Generalized Linear Model (GLM) and two machine learning models, viz., Gradient Boosting Machine Learning Algorithm (GBM) and Random Forest (RF), which fitted to data well were incorporated for the purpose of disease prediction. The models were trained using the case and control data available at ICAR-NIVEDI. Validation of the models were done by dividing the total observations for a particular disease into marker samples and validation samples and accuracy was tested in terms of discrimination power, which was done using Receiving Operating Characteristics (ROC), Cohen Kappa (Heildke Skill Score) and True Skill statistics (TSS). Once the models produce the probability value, it was used for categorizing the risk. Briefly, when all the models produce the p value of more than 0.5, then the highest p value is used for determining the high-risk category. If all the models or any one model produces the p value of less than 0.5, then the lowest p value was used for categorizing lower risk. This was done to minimize the false alert. Thus, the risk predictions based on the probability values ranging from 0-1 are made as follows; Very High Risk (p=0.81-1.0), High Risk (p=0.61-0.80), Moderate Risk (p=0.41-0.60), Low Risk (p=0.21-0.40), Very Low Risk (p=0.0-0.20) and No Risk (p=0.0) for the occurrence of a said disease. It is believed that categorizing districts in to various risk categories will help the stake holders to effectively utilize the available resources (money and manpower).

II.NADRES v2 Data Flow and Data Processing Diagram

A) Data Flow Diagram:







B) Artificial Intelligence enabled Data Capturing and Forewarning System:

Fig 4.2. Data Capturing and Forewarning system

III. Weighted Outbreak Score

The outbreak data for the month of forecasting is extracted from NADRES database for the period of 10 years from current year. Outbreak data of 13 important livestock diseases are considered. The data is aggregated at district level and the weighted score is defined based on the number of outbreaks for each district in each month considering last 10 years. The weightage score was assigned as 0 for less than three number of outbreaks in the last 10 years for selected month, score 1 for 3-6 number of outbreaks and 2 for more than 6 outbreaks. This weightage score for each district is labelled as risk variable in building the models and risk maps.



Disease Name

Fig 4.3. Top ten livestock diseases (2021)

IV. Forecasting of Weather Parameters

Weather forecasting has been one of the most challenging problems around the world because of both its practical value in meteorology and the popular sphere for scientific research. Weather forecast systems are among the most complex equation systems that computer has to solve. A great quantity of data, coming from satellites, ground stations and sensors located around our planet send daily information that must be used to foresee the weather situation in next hours and days all around. Weather forecasts provide critical information about future weather. There are various techniques involved in weather forecasting, from relatively simple observation of the sky to highly complex computerized mathematical models. Further, forecast products by Indian Metrological department were used for validation of our forecasts (https://mausam.imd.gov.in/imd_latest/contents/extendedrangeforecast.php).

Following are the basic steps of forecasting process:

- 1. Determine the forecast's purpose
- 2. Establish a time horizon
- 3. Select a forecasting technique
- 4. Gather and analyse data
- 5. Perform the forecast
- 6. Monitor the forecast and use it in prediction of disease

Statistical Models used for forecasting of weather and remotely sensed variables

ARIMA stands for Autoregressive Integrated Moving Average. ARIMA is also known as Box-Jenkins approach. Box and Jenkins claimed that non-stationary data can be made stationary by differencing the series, Y_t. The general model for Y_t is written as,

 $\mathbf{Y}_t = \phi_1 \mathbf{Y}_{t-1} + \phi_2 \mathbf{Y}_{t-2} \dots \phi_p \mathbf{Y}_{t-p} + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2} + \dots \theta_q \varepsilon_{t-q}$

Where, Y_t is the differenced time series value, ϕ and θ are unknown parameters and ϵ are independent identically distributed error terms with zero mean. Here, Y_t is expressed in terms of its past values and the current and past values of error terms.

The ARIMA Model combines three basic Methods:

- Auto Regression (AR) In auto-regression the values of a given time series data are regressed on their own lagged values, which is indicated by the "p" value in the model.
- Differencing (I-for Integrated) This involves differencing the time series data to remove the trend and convert a non-stationary time series to a stationary one. This is indicated by the "d" value in the model. If d = 1, it looks at the difference between two-time series entries, if d = 2 it looks at the differences of the differences obtained at d =1, and so forth.
- Moving Average (MA) The moving average nature of the model is represented by the "q" value which is the number of lagged values of the error term.

This model is called Autoregressive Integrated Moving Average or ARIMA (p, d,q) of Y_t . We will follow the steps enumerated below to build our model. ARIMA models were run in 18 combinations of p, d, q. Based on the minimum AIC value, the order of ARIMA model was selected. This order was used for the prediction of all the weather parameters used in developing disease forewarning models.

V. Implementation of Principal Component Analysis

Large datasets are gradually common and are often difficult to interpret. Principal Component Analysis (PCA) is a technique for reducing the dimensionality of such datasets, increasing the interpretability but at the same time, minimizing the information loss. The PCA is employed in NADRES v2 by creating new uncorrelated variables that successively maximize the variance. This means that ` preserving as much variability as possible` translates into finding new variables that are linear functions of those in the original dataset, that successively maximize variance and that are uncorrelated with each other. Determining such new variables, the principal components (PCs) reduce to solve an eigenvalue/eigenvector problem. PCA can be based on either covariance matrix or the correlation matrix and the main use of PCA are descriptive.

In the present study, all the meteorological and remote sensing variables are considering for PCA, with correlation matrix, the final output of principal components which are independent of each were considered for further ML modelling and risk estimation.

VI. Machine Learning Models

Disease outbreak data were aligned with generated risk variables to the respective latitude and longitude, which were subjected to climate-disease modelling. A number of models were fit to aligned data and tested for accuracy in terms of discrimination power. Two regression models, Generalized Linear Models (GLM) and Generalized Additive Models (GAM) and six machine learning algorithms, i.e. Random Forest (RF), Boosted Regression Tree (BRT), Artificial Neural Network (ANN), Multiple Adaptive Regression Spline (MARS), Flexible Discriminant Analysis (FDA) and Classification Tree Analysis (CTA) were employed for disease modelling. Different modelling methods return different types of 'model object' and all these model objects could be used for the predict function to make predictions for any combinations of values of independent variables. Response plots were created to explore and understand model predictions.

The fitted models were assessed for their discriminating power using Receiving Operating Characteristic (ROC) curve, Cohen's Kappa (Heildke Skill Score) and True Skill Statistics (TSS). These measures were used to evaluate the quality of predictions based on presence-absence data. Raster Stack was used to combine the results of individual predictions by different model methods. All the models were assessed for overfitting.

The outcome of best fitted models was in probability of disease occurrence and was categorised into 6 risk levels as No risk (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR) and Very high risk (VHR) for enabling the stakeholders to take appropriate control measures by suitably allocating available resources.

5. Accuracy of Prediction

Serial No.	Diseases	Accuracy (%)
1.	Anthrax	99.69
2.	Babesiosis	98.77
3.	Black quarter	95.22
4.	Bluetongue	99.69
5.	Enterotoxaemia	100.00
6.	Fascioliasis	98.46
7.	Foot and mouth disease	97.22
8.	Haemorrhagic septicaemia	97.38
9.	Peste des petits ruminants	95.99
10.	Sheep & Goat pox	99.23
11.	Swine fever	100.00
12.	Theileriosis	99.54
13.	Trypanosomiasis	97.69

Aggregation and prediction of livestock diseases at district level leading to higher accuracy.

• Formula Used: The Accuracy of disease prediction was calculated using the following formula.

$$\frac{\text{TP} + \text{TN}}{\text{Total}} * 100$$

TP-True Positive Observations, TN-True Negative Observations, Total- Total observations.

- Internal Accuracy was performed using 10 years of data. Accuracy obtained was >90% for all the diseases predicted.
- Despite the power of climate and disease risk models, considerable uncertainties remain, identifying these uncertainties, highlighting importance of improved data may improve the model accuracy, realism, confidence, together with translating uncertainties in model inputs into uncertainties in model outputs, are important benefits of modelling.

6. Moran's I for clustering of Livestock diseases

Moran's I is a tool that measures spatial autocorrelation (feature similarity) based on both feature locations and feature values simultaneously. Given a set of features and an associated attribute, it evaluates whether the pattern expressed is clustered, dispersed, or random. The tool calculates the Moran's I Index value and both a Z score and p-value evaluating the significance of that index. In general, a Moran's Index value near +1.0 indicates clustering while an index value near -1.0 indicates dispersion.

Autocorrelation tool, the null hypothesis states that "there is no spatial clustering of the values associated with the geographic features in the study area". When the p-value is small and the absolute value of the Z score is large enough that it falls outside of the desired confidence level, the null hypothesis can be rejected. If the index value is greater than 0, the set of features exhibits a clustered pattern. If the value is less than 0, the set of features exhibits a dispersed pattern.

7. R Software

R is a programming language and software environment for statistical analysis, graphics representation and reporting. R is a simple and effective programming language, which includes conditionals, loops, user defined recursive functions and input and output facilities. R statistical software version 3.1.3 (version 3.4.3, R Foundation for Statistical Computing, Vienna, Austria. <u>https://www.R-project.org/</u>) was used as an integrated suite for data mining, calculation and graphical display. Several R packages like *openxlx, raster, RMySQL, rgdal, RColorBrewer, sqldf, sp, spdep, xlsx, plyr, randomFores, dismo, SDMTool, dplyr, tmap* and *data table*were used for data extraction, data alignment, annotation, analysis, modelling and risk mapping.

8. Forewarning of Livestock Disease for the Month of June, 2022

i). District wise Livestock Disease forewarning:

Districts of Andaman and NicobarAnthraxBabesiosisBQBTETFascioliasisFMDHSPPRS&GSFTheileriosisTryNicobarNRNRNRVLRNR </th <th></th> <th></th> <th></th> <th></th> <th></th> <th></th> <th>Live</th> <th>Livestock Diseases</th> <th>seases</th> <th></th> <th></th> <th></th> <th></th> <th></th>							Live	Livestock Diseases	seases					
IddleNRNLNLNLNR<	Districts of Andaman and Nicobar	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD		PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
liddleNRNRVLRNRNRNRNRNRNRNRamanNRNRVLRLRNRVHRMRNRNRNRNR	Nicobars	NR	NR	VLR	NR	NR	VHR	HR	VLR	NR	NR	NR		NR
aman NR NR VLR LR NR VHR MR NR NR NR MR NR	North & Middle Andaman	NR	NR	VLR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR
	South Andaman	NR	NR	VLR	LR	NR		MR	NR	NR		NR	NR	NR

District wise Livestock Disease Risk Forewarning for June, 2022: Andaman and Nicobar

If vaccinated, please ignore the disease forecast.

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Anthrax						LITUDEN DISCASES						
ır	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
	NR	NR	NR	NR	NR	NR	VLR	MR	VHR	NR	NR	NR
Chittoor	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
East Godavari NR	NR	HR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Guntur NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Krishna NR	NR	VLR	NR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR
Kurnool VHR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Prakasam NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Sri Potti Sriramulu VHR Nellore	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Srikakulam VHR	NR	HR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Visakhapatnam NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Vizianagaram	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
West Godavari NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Y.S.R. VHR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR

If vaccinated, please ignore the disease forecast.

Districts of Armachal						Liv	Livestock Diseases	liseases					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Anjaw	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Changlang	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dibang Valley	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
East Kameng	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
East Siang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kurung Kumey	NR	VHR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lohit	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lower Dibang Valley	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lower Subansiri	NR	NR	NR	VLR	NR	VHR	NR	NR	NR	NR	NR	NR	NR
Papum Pare	NR	NR	VLR	NR	NR	HR	VLR	NR	NR	NR	NR	NR	NR
Tawang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Tirap	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Upper Siang	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Upper Subansiri	NR	NR	VLR	VLR	NR	VHR	HR	NR	NR	NR	NR	NR	NR
West Kameng	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
West Siang	NR	NR	VLR	NR	NR	VHR	VHR	NR	NR	NR	NR	NR	NR

District wise Livestock Disease Risk Forewarning for June, 2022: Arunachal Pradesh

If vaccinated, please ignore the disease forecast.

District wise Livestock Disease Risk Forewarning for June, 2022: Assam

						Liv	estock	Livestock Diseases					
Districts of Assam	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Baksa	NR	NR	VHR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Barpeta	NR	NR	VHR	NR	NR	VHR	NR	VHR	VLR	NR	NR	NR	NR
Bongaigaon	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Cachar	NR	NR	VHR	VLR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Chirang	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Darrang	VHR	NR	VHR	NR	VHR	VHR	NR	VHR	VLR	NR	NR	NR	NR
Dhemaji	NR	NR	VHR	NR	NR	VHR	NR	VHR	NR	NR	VHR	VHR	NR
Dhubri	NR	NR	VHR	NR	NR	VHR	VLR	VLR	VLR	NR	NR	NR	NR
Dibrugarh	NR	NR	VHR	NR	NR	VHR	VLR	VLR	VLR	NR	NR	NR	NR
Dima Hasao	NR	NR	VLR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Goalpara	NR	NR	VHR	NR	VHR	NR	VLR	VHR	VLR	NR	NR	NR	NR
Golaghat	NR	NR	VHR	NR	NR	VHR	NR	VLR	VLR	NR	NR	NR	NR
Hailakandi	NR	NR	MR	VLR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Jorhat	NR	NR	VHR	NR	NR	VHR	NR	VLR	VLR	NR	VHR	NR	NR
Kamrup	NR	NR	MR	NR	NR	HR	NR	VLR	VLR	HR	VHR	NR	NR
Kamrup Metropolitan	NR	NR	VHR	NR	NR	VHR	NR	VLR	VLR	VHR	VHR	NR	NR
Karbi Anglong	NR	NR	MR	NR	NR	NR	NR	VLR	NR	VHR	VHR	NHR	NR
Karimganj	NR	NR	VHR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Kokrajhar	NR	NR	HR	NR	NR	NR	VLR	VHR	VLR	NR	VHR	NR	NR
Lakhimpur	NR	NR	VHR	NR	NR	VHR	VLR	HR	NR	NR	VHR	NR	NR
Morigaon	NR	NR	VHR	NR	NR	NR	NR	HR	VLR	NR	NR	NR	NR

Continued

						Liv	estock I	Livestock Diseases					
Districts of Assam	Anthrax	Anthrax Babesiosis	Ŋ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Nagaon	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR
Nalbari	NR	NR	VHR	NR	NR	VHR		VHR	VLR	VHR	NR	NR	NR
Sivasagar	NR	NR	VHR	NR	NR	VHR	NR	HR	NR	NR	VHR	NR	NR
Sonitpur	NR	NR	VHR	NR	NR	HR	NR	VHR	VLR	NR	NR	NR	NR
Tinsukia	NR	VHR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	VHR	NR
Udalguri	NR	NR	VHR	NR	NR	VHR	NR	VHR	VLR	NR	VHR	NR	NR
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						Liv	Livestock Diseases	seases					
Districts of Bihar	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Araria	NR	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Arwal	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Aurangabad	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Banka	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Begusarai	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bhagalpur	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Bhojpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Buxar	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Darbhanga	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gaya	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gopalganj	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Jamui	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Jehanabad	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kaimur (Bhabua)	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Katihar	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Khagaria	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kishanganj	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Lakhisarai	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Madhepura	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Madhubani	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Munger	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Muzaffarpur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

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						Live	Livestock Diseases	iseases					
Districts of Bihar	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Nalanda	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nawada	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Pashchim Champaran	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Patna	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Purba Champaran	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Purnia	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Rohtas	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Saharsa	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Samastipur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Saran	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sheikhpura	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sheohar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sitamarhi	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Siwan	NR	NR	NR	NR	NR	NHR	NR	NR	NR	NR	NR	NR	NR
Supaul	NR	NR	NR	NR	NR	NR	NR	NR	MR	NR	NR	NR	NR
Vaishali	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

If vaccinated, please ignore the disease forecast.

Livestock Diseases	
Districts of Chandioarh	0

Trypanosomosis

Theileriosis

SF

S&G Pox

PPR

HS

FMD

Fascioliasis

ET

BT

BQ

Anthrax Babesiosis

NR

Chandigarh

District wise Livestock Disease Risk Forewarning for June, 2022: Chandigarh

If vaccinated, please ignore the disease forecast.

						Liv	Livestock Diseases	liseases					
Districts of Chhattisgarh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Bastar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bijapur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bilaspur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dakshin Bastar Dantewada	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Dhamtari	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Durg	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Janjgir-champa	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jashpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kabeerdham	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Korba	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Koriya	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Mahasamund	NR	NR	NR	VLR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Narayanpur	NR	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Raigarhh	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Raipur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rajnandgaon	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Surguja	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Uttar Bastar Kanker	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

District wise Livestock Disease Risk Forewarning for June, 2022: Chhattisgarh

If vaccinated, please ignore the disease forecast.

District wise Livestock Disease Risk Forewarning for June, 2022: Dadra and Nagar Haveli

						Liv	Livestock Diseases	iseases					
Districts of Dadra and Nagar Haveli	Anthrax	Anthrax Babesiosis	BQ	BT	ET	Fascioliasis FMD		SH	РРК	S&G Pox	SF	Theileriosis	Theileriosis Trypanosomosis
Dadra and Nagar Haveli	NR	NR	NR NR	NR	NR	NR	NR	NR	NR	NR NR NR NR NR	NR	NR	NR

If vaccinated, please ignore the disease forecast.

District wise Livestock Disease Risk Forewarning for June, 2022: Daman and Diu

Districts of						Lives	Livestock Diseases	eases						
Daman and Diu	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis FMD	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Theileriosis Trypanosomosis	
Daman	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
Diu	NR	NR	NR	NR	NR	NR	NR	NR	NR VLR NR	NR	NR	NR	NR	
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						Lives	Livestock Diseases	eases					
Districts of Goa	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis FMD	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Theileriosis Trypanosomosis
North Goa	NR	NR	NR NR	NR	NR	NR	NR	VLR	NR	NR VLR NR NR NR	NR	NR	NR
South Goa	NR	NR	NR NR	NR	NR	NR	NR	NR	NR	NR NR NR NR NR	NR	NR	NR
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District wise Livestock Disease Risk Forewarning for June, 2022: Goa

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	Trypanosomosis	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	Theileriosis	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	SF	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	S&G Pox	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	PPR	VLR	HR	NR	VLR	VLR	VLR	NR	VLR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	
ases	SH	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	recast.
Livestock Diseases	FMD	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	lisease fo
Livest	Fascioliasis	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	ed nlease jonore the disease forecast
	ET	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	ted nles
	BT	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	If vaccinat
	BQ	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	If
	Babesiosis	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	•
	Anthrax	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	Districts of Gujarat	Ahmadabad	Amreli	Anand	Banas Kantha	Bharuch	Bhavnagar	Dohad	Gandhinagar	Jamnagar	Junagadh	Kachchh	Kheda	Mahesana	Narmada	Navsari	Panch Mahals	Patan	Porbandar	Rajkot	Sabar Kantha	Surat	Surendranagar	Tapi	The Dangs	Vadodara	Valsad	

If vaccinated, please ignore the disease forecast. *No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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						Livest	Livestock Diseases	eases					
Districts of Haryana	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	Aqq	S&G Pox	SF	Theileriosis	Trypanosomosis
Ambala	NR	NR	NR	LR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bhiwani	NR	NR	NR	NR	NR	NR	NR	NR	MR	NR	NR	NR	NR
Faridabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Fatehabad	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Gurgaon	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Hisar	NR	NR	NR	NR	VHR	NR	NR	NR	VHR	NR	VHR	NR	NR
Jhajjar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR
Jind	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kaithal	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Karnal	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kurukshetra	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	VHR	NR
Mahendragarh	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Mewat	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Palwal	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Panchkula	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Panipat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rewari	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Rohtak	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Sirsa	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Sonipat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Yamunanagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
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If vaccinated, please ignore the disease forecast.

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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						Lives	Livestock Diseases	cases					
Himachal Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	РРК	S&G Pox	SF	Theileriosis	Trypanosomosis
	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
	NR	NR	NR	VLR	NR	NR	NHR	NR	VHR	NR	NR	NR	NR
	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
	NR	NR	VLR	VLR	NR	NR	NR	NR	VHR	NR	NR	NR	NR
1	NR	NR	NR	NR	NR	NR	VHR	NR	HR	VHR	NR	NR	NR
	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
	NR	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
3	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR

If vaccinated, please ignore the disease forecast.

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Districts of Jammu						Liv	Livestock Diseases	iseases					
and Kashmir	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Anantnag	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR
Badgam	NR	NR	VLR	VLR	NR	NR	VLR	NR	VLR	NHR	NR	NR	NR
Bandipore	NR	NR	VLR	VLR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR
Baramula	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Doda	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ganderbal	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jammu	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kargil	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Kathua	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kishtwar	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kulgam	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR
Kupwara	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Leh(Ladakh)	NR	NR	VLR	VLR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Pulwama	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	VHR	NR	NR	NR
Punch	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rajouri	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ramban	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Reasi	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Samba	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shupiyan	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	VHR	NR	NR	NR
Srinagar	NR	NR	VLR	VLR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Udhampur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
				If vaccinated,		please ignore the disease forecast.	ie disease	forecast.					

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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		District wise Livestock Diseas	se Live	stock D	isease	e Risk Forewarning for June, 2022: Jharkhan	varning	for Ju	ie, 202;	: Jhar	khan		
Districts of						Ľ	Livestock Diseases	iseases					
Jharkhand	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Bokaro	NR	VHR	HR	NR	VHR	VHR	VHR	NR	VHR	NR	NR	VHR	VHR
Chatra	NR	VHR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	VHR	VHR
Deoghar	NR	VHR	HR	NR	NR	VHR	VLR	VLR	VHR	NR	NR	VHR	VHR
Dhanbad	NR	VHR	LR	NR	NR	VHR	HR	VLR	VHR	NR	NR	VHR	VHR
Dumka	VHR	VHR	VHR	NR	VHR	VHR	HR	VHR	VHR	NR	NR	VHR	VHR
Garhwa	NR	VHR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	VHR	VHR
Giridih	NR	VHR	HR	NR	NR	VHR	VLR	NR	VLR	NR	NR	VHR	VHR
Godda	NR	NR	VLR	NR	NR	VHR	NR	VLR	VLR	NR	NR	NR	VHR
Gumla	NR	VHR	VLR	NR	NR	VHR	HR	NR	VHR	NR	NR	VHR	VHR
Hazaribagh	NR	VHR	VHR	NR	NR	VHR	VHR	VLR	VHR	NR	NR	VHR	VHR
Jamtara	NR	VHR	MR	NR	NR	VHR	NR	VLR	VHR	NR	NR	VHR	VHR
Khunti	NR	VHR	NR	NR	NR	VHR	VLR	NR	VHR	NHR	VHR	VHR	VHR
Koderma	NR	HR	NR	NR	NR	VHR	NR	NR	VLR	NR	NR	VHR	VHR
Latehar	NR	VHR	VLR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	VHR
Lohardaga	NR	VHR	VLR	NR	NR	VHR	HR	NR	VHR	NR	NR	VHR	VHR
Pakur	NR	VHR	MR	NR	MR	VHR	HR	HR	HR	NR	NR	VHR	HR
Palamu	NR	VHR	VHR	NR	NR	VHR	VLR	VLR	MR	HR	NR	HR	VHR
PashchimiSinghbhum	NR	VHR	VLR	NR	NR	VHR	HR	VHR	VHR	NHR	NR	VHR	VHR
Purbi Singhbhum	NR	VHR	VHR	NR	MR	VHR	HR	VHR	VHR	VHR	NR	VHR	VHR
Ramgarh	NR	VHR	NR	NR	NR	MR	NR	NR	VLR	NR	NR	NR	VHR
Ranchi	NR	VHR	HR	NR	NR	VHR	VHR	NR	VHR	NR	NR	VHR	VHR
Sahibganj	VHR	VHR	LR	NR	VHR	VHR	HR	VHR	MR	VHR	VHR	VHR	VHR
Seraikela - Kharsawan	NR	HR	NR	NR	NR	VHR	NR	NR	VLR	NR	NR	VHR	VHR
Simdega	NR	VHR	NR	NR	NR	VHR	VLR	NR	VHR	NR	NR	VHR	VHR
				If vaccinated,		please ignore the disease forecast.	ne disease	forecast.					

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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District wise Livestock Disease Risk Forewarning for June, 2022: Karnataka

softattaAnthraxBabesiosisBQBTETFascioliasiFMDHSPPRS&GSFatNRN	•						Li	Livestock Diseases	iseases					
otNRNRNRNRNRVLRVLRMRNRNRNR ore NRNRNRNRNRNHRVLRNHRNHRNHRNHRNHRNRNHR <th>Districts of Karnataka</th> <th>Anthrax</th> <th>Babesiosis</th> <th>BQ</th> <th>BT</th> <th>ET</th> <th>Fascioliasi s</th> <th>FMD</th> <th>HS</th> <th>PPR</th> <th>S&G Pox</th> <th>SF</th> <th>Theileriosis</th> <th>Trypanosomosis</th>	Districts of Karnataka	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
ore ore RuralNRVHRVHRVHRVHRVHRVHRNRVHRNR <td>Bagalkot</td> <td>NR</td> <td>NR</td> <td>MR</td> <td>VLR</td> <td>NR</td> <td>MR</td> <td>NR</td> <td>VLR</td> <td>VLR</td> <td>MR</td> <td>NR</td> <td>NR</td> <td>NR</td>	Bagalkot	NR	NR	MR	VLR	NR	MR	NR	VLR	VLR	MR	NR	NR	NR
ore RuralNRNRVLRVLRNR<	Bangalore	NR	NR	VHR	VLR	NR	NR	VHR	VLR	VHR	VHR	NR	NR	NR
mmNRNRVHRNRVLRMRMRMRNR <td>Bangalore Rural</td> <td>NR</td> <td>NR</td> <td>VLR</td> <td>VLR</td> <td>NR</td> <td>NR</td> <td>VHR</td> <td>NR</td> <td>NR</td> <td>NR</td> <td>NR</td> <td>NR</td> <td>NR</td>	Bangalore Rural	NR	NR	VLR	VLR	NR	NR	VHR	NR	NR	NR	NR	NR	NR
HHR NR VLR NR NR NR NR NR VLR VLR HHR NR NR NLR NLR NR NR NR NR NR NLR VLR VLR NR NR NR NR NLR VLR NR	Belgaum	NR	NR	MR	NR	VHR	NR	VLR	MR	MR	NR	NR	NR	NR
rNRNRNRNRVLRVLRNLRNLRNR<	Bellary	VHR	NR	VLR	NR	NR	NR	VLR	VLR	HR	NR	NR	NR	NR
tNRNRNRVLRVHRNRVHRNRVHRNR<	Bidar	NR	NR	HR	NR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
rajanagar <i>VHR</i> NRVLRVLRNRNR <i>HR</i> NR<	Bijapur	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR
ballapuraVHRNRVLRNRNLRNLRNLRNR <td>Chamarajanagar</td> <td>VHR</td> <td>NR</td> <td>VLR</td> <td>VLR</td> <td>NR</td> <td>NR</td> <td>VHR</td> <td>HR</td> <td>NR</td> <td>NR</td> <td>NR</td> <td>NR</td> <td>NR</td>	Chamarajanagar	VHR	NR	VLR	VLR	NR	NR	VHR	HR	NR	NR	NR	NR	NR
agalurNRNRNRNRNLRVLRNR	Chikkaballapura	VHR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
durgaNRNRHRNRVHRNRHRVLRVLRHRNLRHRNLna KannadaNRNRNRVLRNRNRNRNRNRNRNRNRNRagereNRNRNRVLRNRVHRNRVLRNRNRNRNRadNRNRNRVLRNRVLRNRNRNRNRNRNRNRNRNRNRNR	Chikmagalur	NR	NR	HR	NR	NR	NR	MR	VLR	VLR	NR	NR	NR	NR
na KannadaNRNRNRNLRNRNLRNRNLRNR <td>Chitradurga</td> <td>NR</td> <td>NR</td> <td>HR</td> <td>NR</td> <td>VHR</td> <td>NR</td> <td>HR</td> <td>VLR</td> <td>VLR</td> <td>HR</td> <td>NR</td> <td>NR</td> <td>NR</td>	Chitradurga	NR	NR	HR	NR	VHR	NR	HR	VLR	VLR	HR	NR	NR	NR
agere NR NR <u>HR</u> NR <u>VHR</u> NR <u>VHR</u> VLR NL NR <u>VHR</u> NR <u>VHR</u> NR ad NR NR NL NR	Dakshina Kannada	NR	NR	NR	VLR	NR	NR	HR	MR	VLR	NR	NR	NR	NR
ad NR NR MR VLR NR NR NR NR NR NR NR NR VLR VLR VLR NR NR NR	Davanagere	NR	NR	HR	NR	VHR	NR	VHR	VLR	NR	VHR	NR	NR	NR
NR NR VLR NR NR NR VLR VLR VLR VLR VLR NR NR	Dharwad	NR	NR	MR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
	Gadag	NR	NR	VLR	NR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
						Li	Livestock Diseases	iseases						
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Districts of Karnataka	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis	
Gulbarga	NR	NR	MR	NR	VHR	NR	NR	VLR	NR	NR	NR	NR	NR	
Hassan	VHR	NR	VHR	VLR	VHR	NR	VHR	MR	VLR	NR	NR	NR	NR	
Haveri	NR	NR	HR	VLR	NR	NR	VHR	VLR	VLR	MR	NR	NR	NR	
Kodagu	NR	NR	VLR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	
Kolar	NR	NR	VLR	VLR	NR	NR	NR	NR	MR	NR	NR	NR	NR	
Koppal	VHR	NR	MR	NR	VHR	NR	VLR	MR	NR	VHR	NR	NR	NR	
Mandya	NR	NR	VLR	VLR	NR	NR	NR	VLR	MR	NR	NR	NR	NR	
Mysore	NR	NR	VHR	VLR	NR	NR	VHR	MR	VLR	VHR	NR	NR	NR	
Raichur	VHR	NR	VHR	NR	MR	NR	NR	VLR	VLR	NR	NR	NR	NR	
Ramanagara	NR	NR	HR	VLR	NR	NR	VHR	MR	HR	NR	NR	VHR	NR	
Shimoga	NR	NR	VHR	VLR	NR	NR	MR	VLR	MR	NR	NR	NR	NR	
Tumkur	VHR	NR	NHR	VLR	NHR	NR	VHR	VLR	VHR	NHR	NR	NR	NR	
Udupi	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	
Uttara Kannada	NR	NR	HR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR	
Yadgir	NR	NR	VLR	NR	NHR	NR	NR	VLR	VLR	NR	NR	NR	NR	

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

If vaccinated, please ignore the disease forecast.

Continued

	Trypanosomosis	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	Theileriosis	VHR	VHR	MR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	NR	NR	400
	SF	NR	NHR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	100 C
	S&G Pox	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	PPR	VLR	VLR	NR	VLR	VLR	VLR	NR	VHR	VLR	VLR	VLR	NR	VLR	NR	
seases	SH	VLR	NR	NR	VHR	VLR	HR	VHR	NR	VHR	HR	VLR	HR	VHR	NR	
Livestock Diseases	FMD	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	VHR	
Liv	Fascioliasi s	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	ET	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	BT	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
	BQ	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	
	Babesiosis	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	VHR	NR	\$2
	Anthrax	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	
	Districts of Kerala	Alappuzha	Ernakulum	Idukki	Kannur	Kasaragod	Kollam	Kottayam	Kozhikode	Malappuram	Palakkad	Pathanamthitta	Thiruvananthapuram	Thrissur	Wayanad	

District wise Livestock Disease Risk Forewarning for June, 2022: Kerala

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If vaccinated, please ignore the disease forecast.

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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Lakshadweep
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Districts of						Liv	ivestock Disea	seases					
Lakshadweep	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Lakshadweep	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

If vaccinated, please ignore the disease forecast.

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Madhya
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Districts of Madhya						Γ	Livestock Diseases	Diseases					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Alirajpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Anuppur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ashoknagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Balaghat	NR	NR	VLR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Barwani	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Betul	NR	NR	VLR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR
Bhind	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Bhopal	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Burhanpur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Chhatarpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Chhindwara	NR	NR	HR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
Damoh	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Datia	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Dewas	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dhar	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Dindori	NR	NR	VLR	NR	NR	NR	VLR	VHR	NR	NR	NR	NR	NR
East Nimar	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Guna	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Gwalior	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Harda	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Hoshangabad	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Indore	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Jabalpur	NR	NR	VLR	NR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Jhabua	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Katni	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

Districts of Madhya							Livestock	Livestock Diseases					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Khargone (West Nimar)	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mandla	NR	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Mandsaur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Morena	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Narsimhapur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Neemuch	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Panna	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Raisen	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Rajgarh	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Ratlam	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Rewa	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Sagar	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Satna	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sehore	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Seoni	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shahdol	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shajapur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sheopur	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Shivpuri	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Sidhi	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Singrauli	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Tikamgarh	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Ujjain	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Umaria	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Vidisha	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
				If va	ocinate	If vaccinated, please ignore the disease forecast.	e the dise	ease foreca	st.				

Continued

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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District wise Livestock Disease Risk Forewarning for June, 2022: Maharashtra

							Livestock Diseases	Diseases					
Districts of Maharashtra	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Ahmadnagar	NR	NR	NR	NR	NR	NR	VLR	NR	VLR	NR	NR	VHR	NR
Akola	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	VHR	NR
Amravati	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	MR	NR	NR	NR
Aurangabad	NR	NR	NR	NR	NR	NR	NR	NR	VHR	VHR	NR	NR	NR
Bhandara	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Bid	NR	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
Buldana	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Chandrapur	NR	NR	VLR	NR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Dhule	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gadchiroli	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Gondiya	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Hingoli	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Jalgaon	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Jalna	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kolhapur	NR	NR	VLR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Latur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Mumbai	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Mumbai Suburban	NR	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Nagpur	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Nanded	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Nandurbar	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Nashik	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Osmanabad	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

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							Livestock	Livestock Diseases					
Districts of Maharashtra	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Parbhani	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Pune	<i>NHR</i>	NR	VLR	NR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR
Raigarh	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ratnagiri	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sangli	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Satara	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sindhudurg	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Solapur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Thane	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Wardha	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Washim	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Yavatmal	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
				If vs	If vaccinated	ed, please ignore the disease forecast.	e the dist	ease foreca	st.				

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						Live	Livestock Diseases	seases					
Districts of Manipur	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosom
Bishnupur	NR	NR	VLR	NR	NR	VHR	VLR	VLR	VLR	NR	VHR	NR	NR
Chandel	NR	NR	VHR	VLR	NR	VHR	VLR	NR	NR	NR	NR	NR	NR
Churachandpur	NR	NR	NHR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Imphal East	NR	NR	HR	VLR	NR	VHR	NR	NR	NR	NR	VHR	NR	NR
Imphal West	NR	NR	VLR	VLR	NR	VHR	NR	VLR	NR	NR	VHR	NR	NR
Senapati	NR	NR	VLR	NR	NR	VHR	VLR	VLR	NR	NR	NR	NR	NR

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District wise Livestock Disease Risk Forewarning for June, 2022: Manipur

If vaccinated, please ignore the disease forecast.

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VLR

						Live	Livestock Diseases	seases					
Districts of Meghalaya	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
East Garo Hills	NR	NR	HR	NR	NR	NR	VLR	HR	NR	NR	MR	NR	NR
East Jaintia Hills	NR	NR	VLR	VLR	NR	NR	VHR	VLR	VLR	NR	NR	NR	NR
East Khasi Hills	VHR	NR	VLR	NR	NR	NR	VHR	NR	NR	VHR	VHR	NR	NR
Jaintia Hills	NR	NR	NR	VLR	NR	NR	VHR	NR	VLR	NR	NR	NR	NR
North Garo Hills	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ribhoi	VHR	NR	VLR	NR	NR	NR	NR	NR	NR	VHR	VHR	NR	NR
South Garo Hills	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Southwest Garo Hills	NR	NR	VLR	NR	NR	NR	VHR	HR	VLR	NR	VHR	NR	NR
Southwest Khasi Hills	NR	NR	VLR	NR	NR	VHR	MR	LR	VLR	NR	NR	NR	NR
West Garo Hills	NR	NR	HR	NR	NR	NR	VHR	HR	NR	NR	VHR	NR	NR

2022: Meghalaya
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						Live	Livestock Diseases	seases					
Districts of Mizoram	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Aizawl	NR	NR	VLR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Champhai	NR	NR	HR	VLR	NR	NR	VLR	HR	NR	NR	VHR	NR	NR
Kolasib	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Lawngtlai	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Lunglei	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Mamit	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Saiha	NR	NR	VLR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Serchhip	NR	NR	VLR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
				If vaccinated	nated, plo	1, please ignore the disease forecast	e disease	forecast.	22				

						Liv	Livestock Diseases	seases					
Districts of Nagaland	Anthrax	Babesiosis	BQ	вт	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Dimapur	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	VHR	NR	NR
Kiphire	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kohima	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Longleng	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	VHR	NR	NR
Mokokchung	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Mon	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Peren	NR	NR	VLR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Phek	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	VHR	NR	NR
Tuensang	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR
Wokha	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Zunheboto	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	VHR	NR	NR
				If vaccin	nated, ple	If vaccinated, please ignore the disease forecast.	e disease	forecast.	1				

District wise Livestock Disease Risk Forewarning for June, 2022: Nagaland

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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Districts of NCT of						Liv	Livestock Diseases	seases					
Delhi	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	РРК	S&G Pox	SF	Theileriosis	Trypanosomosis
Central	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
East	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
New Delhi	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
North	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
North East	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
North West	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
South	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
South West	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
West	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
				If wood	and mla	If wassingted alongs issues the dissact fourset	o diego o	fornact					

District wise Livestock Disease Risk Forewarning for June, 2022: NCT of Delhi

If vaccinated, please ignore the disease forecast.

District wise Livestock Disease Risk Forewarning for June, 2022: Odisha

Districts of						Γ	Livestock Diseases	Diseases					
Odisha	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Anugul	VHR	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
Balangir	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Baleshwar	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Bargarh	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Baudh	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bhadrak	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Cuttack	NR	NR	HR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Debagarh	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dhenkanal	NR	NR	VLR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR	NR
Gajapati	NR	NR	VLR	NR	NR	NR	VLR	MR	VLR	NR	NR	NR	NR
Ganjam	NR	NR	VLR	NR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Jagatsinghapur	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Jajapur	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Jharsuguda	NR	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Kalahandi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kandhamal	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Kendrapara	NR	NR	HR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Kendujhar	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Khordha	NR	NR	HR	NR	NR	NR	VLR	NR	HR	NR	NR	NR	NR
Koraput	VHR	NR	NR	VLR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Malkangiri	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	VHR	NR	NR	NR

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						Ι	Livestock Diseases	Diseases					
Districts of Odisha	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Mayurbhanj	NR	NR	VLR	NR	NR	NR	NR	VLR	VHR	NR	NR	NR	NR
Nabarangapur	VHR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nayagarh	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nuapada	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Puri	NR	NR	VLR	NR	NR	NR	HR	NR	NR	NR	NR	HR	NR
Rayagada	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	VHR	NR
Sambalpur	NR	NR	VLR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR
Subarnapur	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Sundargarh	HR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	VHR
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	Trypanosomosis	NR	NR	NR	NR
	Theileriosis	NR	NR	NR	NR
	SF	NR	NR	NR	NR
	S&G Pox	NR	NR	NR	NR
	APR	NR	VLR	NR	VLR
iseases	SH	NR	NR	NR	VLR
Livestock Diseases	FMD	NR	NR	NR	NR
Ι	Fascioliasis	NR	NR	VHR	VHR
	ET	NR	NR	NR	NR
	BT	NR	VLR	NR	VLR NR
	βQ	NR	NR	NR	VLR
	Anthrax Babesiosis	VHR	VHR	VHR	NR
	Anthrax	NR	NR	NR	NR
Districts of	Puducherry	Karaikal	Mahe	Puducherry	Yanam

District wise Livestock Disease Risk Forewarning for June, 2022: Puducherry

If vaccinated, please ignore the disease forecast.

Punjab
District wise Livestock Disease Risk Forewarning for June, 2022: Punjab
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Districts of							Livestock Diseases	Diseases					
Punjab	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Amritsar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Barnala	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bathinda	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Faridkot	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Fatehgarh Sahib	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Firozpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	<i>VHR</i>	NR
Gurdaspur	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Hoshiarpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jalandhar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kapurthala	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ludhiana	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Mansa	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Moga	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Muktsar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Patiala	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rupnagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sahibzada Ajit Singh Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sangrur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shahid Bhagat Singh Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Tarn Taran	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
				If	vaccinat	If vaccinated, please ignore the disease forecast.	ore the dist	case fore	cast.				

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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District wise Livestock Disease Risk Forewarning for June, 2022: Rajasthan

Districts of							Livestock Diseases	Diseases					
Rajasthan	Anthrax	Babesiosis	Ъд	BT	ET	Fasciolia sis	FMD	SH	PPR	S&G Pox	SF	Theileriosi s	Trypanosomosis
Ajmer	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Alwar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Banswara	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Baran	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Barmer	NR	NR	NR	NR	NR	NR	NR	VLR	MR	NR	NR	NR	NR
Bharatpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bhilwara	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Bikaner	NR	NR	VLR	NR	NR	NR	VLR	NR	VLR	NR	NR	NR	NR
Bundi	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Chittaurgarh	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Churu	NR	NR	VLR	NR	VHR	NR	NR	VLR	VHR	NR	NR	NR	NR
Dausa	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Dhaulpur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dungarpur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Ganganagar	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Hanumangarh	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR	NR	NR	NR
Jaipur	NR	NR	VLR	NR	VHR	NR	HR	VLR	VLR	NR	NR	NR	NR
Jaisalmer	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Jalor	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Jhalawar	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jhunjhunun	NR	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Jodhpur	NR	NR	VLR	NR	NR	NR	NR	VLR	MR	NR	NR	NR	NR
Karauli	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR

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Districts of							Livestock Diseases	Diseases					
Rajasthan	Anthrax	Babesiosis	BQ	BT	ET	Fasciolia sis	FMD	SH	PPR	S&G Pox	SF	Theileriosi s	Trypanosomosis
Kota	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nagaur	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Pali	NR	NR	VLR	VLR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Pratapgarh	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Rajsamand	NR	MR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sawai Madhopur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sikar	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Sirohi	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Tonk	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Udaipur	NR	VHR	VLR	NR	NR	NR	VHR	VLR	VLR	NR	NR	NR	NR
				If	If vaccinated	d, please ignore the disease forecast.	nore the d	isease for	ecast.	3			

							Livestock Diseases	Diseases					
Districts of Sikkim	Anthrax	Anthrax Babesiosis	BQ	BT	ET	Fasciolia sis	FMD	SH	PPR	S&G Pox	SF	Theileriosi s	Trypanosomosis
East District	NR	NR	NR	NR VLR NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
North District	NR	NR	NR	NR VLR NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
South District	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
West District	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
				If	If vaccinate	ed, please ignore the disease forecast.	iore the di	isease for	ecast.				

District wise Livestock Disease Risk Forewarning for June, 2022: Sikkim

Districts of						Γ	Livestock Disease	Disease					
Tamil Nadu	Anthrax	Babesiosis	Ъд	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Ariyalur	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Chennai	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Coimbatore	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Cuddalore	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dharmapuri	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dindigul	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Erode	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kancheepuram	NR	NR	NHR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Kanniyakumari	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Karur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Krishnagiri	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Madurai	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nagapattinam	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Namakkal	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Perambalur	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Pudukkottai	NR	NR	VLR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Ramanathapuram	NR	NR	VLR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Salem	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Sivaganga	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Thanjavur	NR	NR	VLR	NR	NR	NR	NR	VLR	NR	VHR	NR	NR	NR
The Nilgiris	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Theni	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Thiruvallur	VHR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Thiruvarur	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
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District wise Livestock Disease Risk Forewarning for June, 2022: Tamil Nadu

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Districts of Tomil						Γ	Livestock Disease	Disease					
Nadu	Anthrax	Anthrax Babesiosis	ЪЯ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Tiruchirappalli	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Tirunelveli	NR	NR	NR	VLR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Tiruppur	NR	NR	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR
Tiruvannamalai	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Vellore	VHR	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Viluppuram	VHR	NR	HR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Virudhunagar	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
	C.	C		If vac	If vaccinated	I, please ignore the disease forecast	e the dise	ase fore	cast.				

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						L	Livestock Diseases	Diseases					
Districts of Telangana	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Adilabad	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Hyderabad	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Karimnagar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	VHR	NR	NR	NR
Khammam	NR	NR	VLR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR
Mahbubnagar	NR	NR	VLR	NR	VHR	NR	NR	VLR	VLR	NR	NR	NR	NR
Medak	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Nalgonda	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
Nizamabad	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Rangareddy	NR	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Warangal	NR	NR	MR	NR	VHR	NR	VLR	NR	VLR	NR	NR	NR	NR
				If va	If vaccinated	d, please ignore the disease forecast.	e the dise	ase fore	cast.				

						T	ivestock	Livestock Disease					
Districts of Tripura	Anthrax	Ant <i>hr</i> ax Babesiosis	BQ	BT	ET	Fascioliasis FMD	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Dhalai	NR	NR	VLR NR NR	NR	NR	NR	NR	NR VLR NR	NR	NR	NR	NR	NR
North Tripura	NR	NR	VLR NR	NR	NR	NR	NR	VLR V	VLR	NR	NR	NR	NR
South Tripura	NR	NR	VLR	NR	NR	HR	VHR	VHR VLR	MR	VHR	NR	NR	NR
West Tripura	NR	NR	VLR	NR NR	NR	NR	NR	NR VLR	NR	NR	NR	NR	NR
r	č		2	If vac	cinated	If vaccinated, please ignore the disease forecast.	e the dist	ease fore	cast.				

District wise Livestock Disease forewarning for June, 2022: Tripura

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Districts of Uttar						Li	Livestock Disease	Disease					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	SH	PPR	Pox Pox	SF	Theileriosis	Trypanosomosis
Agra	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Aligarh	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Allahabad	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	VHR
Ambedkar Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Amethi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Auraiya	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Azamgarh	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Baghpat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	MR
Bahraich	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR
Ballia	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Balrampur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Banda	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bara Banki	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Bareilly	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Basti	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Bijnor	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Budaun	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Bulandshahr	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	HR
Chandauli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Chitrakoot	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Deoria	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Etah	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Etawah	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Faizabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Farrukhabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

Continued Districts of Uttar						L	Livestock Disease	Disease					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomos is
Fatehpur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	MR
Firozabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gautam Buddha Nagar	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ghaziabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ghazipur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gonda	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Gorakhpur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	MR	VHR
Hamirpur	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Hapur	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Hardoi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR
Jalaun	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jaunpur	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	MR
Jhansi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jyotiba Phule Nagar	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Kannauj	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kanpur Dehat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kanpur Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kanshiram Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kaushambi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kheri	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kushinagar	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Lalitpur	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Lucknow	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
MahaJanuarya Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Mahoba	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR

Districts of Uttar						T	Livestock Disease	Disease					
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomos is
Mahrajganj	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Mainpuri	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Mathura	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mau	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Meerut	NR	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR
Mirzapur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Moradabad	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	HR
Muzaffarnagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Pilibhit	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Pratapgarh	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR
Rae Bareli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	MR
Rampur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Saharanpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR
Sambhal	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Sant Kabir Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sant Ravidas Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shahjahanpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shamli	NR	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR
Shrawasti	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Siddharthnagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sitapur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sonbhadra	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Sultanpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Unnao	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Varanasi	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR
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If vaccinated, please ignore the disease forecast. *No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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^{*}No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

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Districts of West						Γ	Livestock Disease	Disease					
Deligat	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasi s	FMD	SH	PPR	S&G Pox	SF	Theileriosis	Trypanosomos is
Bankura	NR	NR	HR	VLR	NR	NR	HR	VHR	HR	NR	NR	NR	NR
Bardhhaman	VHR	NR	VHR	NR	NR	NR	NR	VHR	HR	NR	NR	VHR	VHR
Birbhum	NR	NR	VHR	NR	NR	NR	NR	VHR	VHR	NR	NR	NR	NR
Dakshin Dinajpur	NR	NR	VHR	NR	NR	NR	NR	VLR	VHR	NR	NR	VHR	NR
Darjiling	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Haora	NR	NR	VHR	NR	NR	NR	NR	VLR	MR	VHR	NR	VHR	NR
Hugli	NR	NR	VHR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NHR	VHR
Jalpaiguri	NR	NR	MR	VLR	NR	NR	NR	HR	NR	NR	NR	NR	NR
Koch Bihar	NR	NR	HR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
Kolkata	NR	NR	NR	NR	NR	VHR	NR	NR	MR	NR	NR	NR	NR
Maldah	NR	NR	VHR	NR	NR	NR	NR	VLR	HR	NR	NR	NR	NR
Murshidabad	VHR	NR	HR	VLR	NR	NR	NR	HR	MR	NR	NR	NR	NR
Nadia	NR	NR	VLR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR
North Twenty-Four Parganas	NR	NR	VHR	VLR	NR	NR	VLR	NR	VLR	VHR	NR	VHR	NR
Paschim Medinipur	NR	NR	VHR	VLR	NR	NR	VHR	VHR	HR	NR	NR	MR	NR
Purba Medinipur	NR	NR	VLR	NR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR
Puruliya	NR	VHR	VLR	NR	NR	VHR	NR	LR	HR	NR	NR	NR	NR
South Twenty Four Parganas	NR	NR	MR	NR	NR	NR	NR	VLR	VLR	NR	NR	VHR	NR
Uttar Dinajpur	NR	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR
				If vacci	nated, 1	If vaccinated, please ignore the disease forecast.	the disea	se forecas	t.				

*No risk/No data available (NR), Very low risk (VLR), Low risk (LR), Moderate risk (MR), High risk (HR), Very high risk (VHR)

If vaccinated, please ignore the disease forecast.

II) Glimpse about the risk of predicted diseases:

The Livestock disease risk obtained based on the Machine Learning algorithm were further categorized into risk events using High Risk and Very High Risk.

1. Andaman and Nicobar

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Nicobars, North & Middle Andaman and South Andaman	Three	Fascioliasis
2.	Nicobars	One	Foot and Mouth Disease

2. Andhra Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Kurnool, Sri Potti Sriramulu Nellore, Srikakulam and Y.S.R.	Four	Antrax
2.	East Godavari and Srikakulam	Two	Black Quarter
3.	Anantapur and Krishna	Two	Sheep & Goat pox

3. Arunachal Pradesh

SI. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Kurung Kumey	One	Babesiosis
2.	Lower Subansiri, Papum Pare, Upper Subansiri and West Siang	Four	Fascioliasis
3.	Upper Subansiri and West Siang	Two	Foot and Mouth Disease

4. Assam

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Darrang	One	Antrax
2.	Tinsukia	One	Babesiosis
3.	Baksa, Barpeta, Cachar, Darrang, Dhemaji, Dhubri, Dibrugarh, Goalpara, Golaghat, Jorhat, Kamrup Metropolitan, Karimganj, Kokrajhar, Lakhimpur, Morigaon, Nalbari, Sivasagar,	Nineteen	Black Quarter
	Sonitpur and Udalguri		
4.	Darrang and Goalpara	Two	Enterotoxaemia
5.	Barpeta, Darrang, Dhemaji, Dhubri, Dibrugarh, Golaghat, Jorhat, Kamrup, Kamrup Metropolitan, Lakhimpur, Nalbari, Sivasagar, Sonitpur and Udalguri	Fourteen	Fascioliasis
6.	Barpeta, Darrang, Dhemaji, Goalpara, Kokrajhar, Lakhimpur, Morigaon, Nalbari, Sivasagar, Sonitpur and Udalguri	Eleven	Haemorrhagic Septicaemia
7.	Cachar and Karimganj	Two	Peste des Petits Ruminants
8.	Kamrup, Kamrup Metropolitan, Karbi Anglong, Nagaon and Nalbari	Five	Sheep & Goat pox
9.	Dhemaji, Jorhat, Kamrup, Kamrup Metropolitan, Karbi Anglong, Kokrajhar, Lakhimpur, Sivasagar and Udalguri	Nine	Swine Fever
10.	Dhemaji, Karbi Anglong and Tinsukia	Three	Theileriosis

5. Bihar

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Siwan	One	Fascioliasis
2.	Bhojpur	One	Trypanosomiasis

Alla.

6. Gujarat

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Amreli	One	Peste des Petits Ruminants
2.	Kachchh	One	Theileriosis

7. Haryana

Sl.	No	Disease prone districts	Number of districts prone for disease	Disease Name
AR	1.	Hisar	One	Enterotoxaemia
	2.	Hisar	One	Peste des Petits Ruminants
	3.	Hisar and Jhajjar	Two	Swine Fever
	4 .	Kurukshetra	One	Theileriosis

8. Himachal Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Kullu and Shimla	Two	Foot and Mouth Disease
2.	Solan	One	Haemorrhagic Septicaemia
3.	Kullu, Mandi and Shimla	Three	Peste des Petits Ruminants
4.	Shimla	One	Sheep & Goat pox

9. Jammu & Kashmir

Sl. No	Disease prone districts	i.	Number of districts prone for disease	Disease Name
1.	Anantnag, Badgam,	Bandipore,	Six	Sheep & Goat pox
	Kulgam, Pulwama and	Shupiyan		

10. Jharkhand

	Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
	1.	Dumka and Sahibganj	Two	Antrax
	2.	All districts except Godda	Twenty-three	Babesiosis
Ho	3.	Bokaro, Deoghar, Dumka, Giridih, Hazaribagh, Palamu, Purbi Singhbhum and Ranchi	Eight	Black Quarter
IC	4.	Bokaro, Dumka and Sahibganj	Three	Enterotoxaemia
	5.	All districts except Ramgarh	Twenty-three	Fascioliasis
	6.	Bokaro, Dhanbad, Dumka, Gumla, Hazaribagh, Lohardaga, Pakur, Pashchimi Singhbhum, Purbi Singhbhum, Ranchi and Sahibganj	Eleven	Foot and Mouth Disease
	7.	Dumka, Pakur, Purbi Singhbhum, Pashchimi Singhbhum and Sahibganj	Five	Haemorrhagic Septicaemia
A Water	8.	Bokaro, Deoghar, Dhanbad, Dumka, Gumla, Hazaribagh, Jamtara, Khunti, Lohardaga, Pakur, Pashchimi Singhbhum, Purbi Singhbhum, Ranchi and Sahibganj	Fourteen	Peste des Petits Ruminants
	9.	Khunti, Palamu, Pashchimi Singhbhum, Purbi Singhbhum and Sahibganj	Five	Sheep & Goat pox
	10.	Khunti and Sahibganj	Two	Swine Fever
	11.	All twenty-four districts except Godda, Latehar and Ramgarh	Twenty-one	Theileriosis
	12.	All districts	Twenty-four	Trypanosomosis

11. Karnataka

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Bellary, Chamarajanagar, Chikkaballapura, Hassan, Koppal, Raichur and Tumkur	Seven	Anthrax
2.	Bangalore, Bidar, Chikmagalur, Chitradurga, Davanagere, Hassan, Haveri,	Thirteen	Black Quarter
	Mysore, Raichur, Ramanagara, Shimoga, Tumkur and Uttara Kannada		
ICAR 3.	Belgaum, Chitradurga, Davanagere, Gulbarga, Hassan, Koppal, Tumkur and Yadgir	Eight	Enterotoxaemia
4.	Bangalore, Bangalore Rural, Chamarajanagar, Chitradurga, Dakshina Kannada, Davanagere, Hassan, Haveri, Kodagu, Mysore, Ramanagara, Tumkur and Uttara Kannada	Thirteen	Foot and Mouth Disease
5.	Chamarajanagar	One	Haemorrhagic Septicaemia
6.	Bangalore, Bellary, Ramanagara and Tumkur	Four	Peste des Petits Ruminants
7.	Bangalore, Bijapur, Chitradurga, Davanagere, Koppal, Mysore and Tumkur	Seven	Sheep & Goat pox
8.	Ramanagara	One	Theileriosis

12. Kerala

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Idukki and Pathanamthitta	Two	Anthrax
2.	Ernakulam, Pathanamthitta and T1issur	Three	Babesiosis
3.	All districts	Fourteen	Foot and Mouth Disease
4.	Kannur, Kollam, Kottayam, Malappuram, Palakkad, Thiruvananthapuram and Thrissur	Seven	Haemorrhagic Septicaemia
к 5.	Kozhikode	One	Peste des Petits Ruminants
6.	Ernakulam and Idukki	Two	Swine Fever
7.	All districts except Idukki, T1issur and Wayanad	Eleven	Theileriosis

13. Madhya Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Chhindwara	One	Black Quarter
2.	Betul, Chhindwara, Dindori and Jabalpur	Four	Haemorrhagic Septicaemia

14. Maharashtra

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Pune	One	Anthrax
2.	Bid	One	Haemorrhagic Septicaemia
3.	Aurangabad	One	Peste desPetitsRuminants
4.	Aurangabad and Pune	Two	Sheep & Goat pox
5.	Ahmadnagar and Akola	Two	Theileriosis

15. Manipur

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Chandel, Churachandpur and Imphal East	Three	Black Quarter
2.	All nine districts except Churachandpur, Tamenglong and Uk1ul	Six	Fascioliasis
4.	Tamenglong	One	Haemorrhagic Septicaemia
5.	Bishnupur, Imphal East, Imphal West and Tamenglong	Four	Swine Fever

16. Meghalaya

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Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	East Khasi Hills and Ribhoi	Two	Anthrax
2.	East Garo Hills, West Garo Hills and West Khasi Hills	Three	Black Quarter
3.	Southwest Khasi Hills	One	Fascioliasis
4.	East Jaintia Hills, East Khasi Hills, Jaintia Hills, Southwest, Garo Hills, West Garo Hills and West Khasi Hills	Six	Foot and Mouth Disease
5.	East Garo Hills, Southwest Garo Hills, West Garo Hills and West Khasi Hills	Four	Haemorrhagic Septicaemia
6.	East Khasi Hills and Ribhoi	Two	Sheep & Goat pox
7.	East Khasi Hills, Ribhoi, Southwest Garo Hills, West Garo Hills and West Khasi Hills	Five	Swine Fever

17. Mizoram

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Champhai	One	Black Quarter
2.	Champhai	One	Haemorrhagic Septicaemia
3.	Champhai	One	Swine fever

18. Nagaland

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Dimapur, Longleng, Phek Tuensang and Zunheboto	Five	Swine fever

19. Odisha

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
ar 1.	Anugul, Koraput, Nabarangapur and Sundargarh	Four	Anthrax
2.	Cuttack, Kendrapara and Khordha	Three	Black Quarter
3.	Puri	One	Foot and Mouth Disease
4.	Anugul	One	Haemorrhagic Septicaemia
5.	Dhenkanal, Khordha, Mayurbhanj and Sambalpur	Four	Peste desPetitsRuminants
6.	Malkangiri	One	Sheep & Goat pox
7.	Puri and Rayagada	Two	Theileriosis
8.	Sundargarh	One	Trypanosomosis

20. Puducherry

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Karaikal, Mahe and Puducherry	Three	Babesiosis
2.	Puducherry and Yanam	Two	Fascioliasis
21. Punjab

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Faridkot	One	Babesiosis
2.	Firozpur	One	Theileriosis

22. Rajasthan

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Udaipur	One	Babesiosis
3.	Churu and Jaipur	Two	Enterotoxaemia
4.	Jaipur and Udaipur	Two	Foot and Mouth Disease
6.	Churu	One	Peste desPetitsRuminants
8.	Hanumangarh	One	Sheep & Goat pox

23. Tamil Nadu

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Thiruvallur, Vellore and Viluppuram	Three	Anthrax
2.	Cuddalore, Kancheepuram and Viluppuram	Three	Black Quarter
3.	Tiruppur	One	Peste desPetitsRuminants
4.	Thanjavur and Thoothukkudi	Two	Sheep & Goat pox

24. Telangana

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Mahbubnagar and Warangal	Two	Enterotoxaemia
2.	Karimnagar and Khammam	Two	Sheep & Goat pox

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25. Tripura

Sl. No	Disease prone districts	Number of disease prone for districts	Disease Name
1.	South Tripura	One	Fascioliasis
2.	South Tripura	One	Foot and Mouth Disease
3.	South Tripura	One	Sheep & Goat pox

26. Uttar Pradesh



27. Uttarakhand

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Almora and Bageshwar	Two	Peste desPetitsRuminants

28. West Bengal

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1. Barddhaman and Murshidabad		Two	Anthrax
2.	Puruliya	One	Babesiosis
3.	Bankura, Barddhaman, Birbhum, Dakshin Dinajpur, Haora, Hugli, Koch Bihar, Maldah, Murshidabad, North	Eleven	Black Quarter
	TwentyFourParganas,Paschim Medinipur		
4.	Kolkata and Puruliya	Two	Fascioliasis
5.	Bankura and Paschim Medinipur	Two	Foot and Mouth Disease
6.	Bankura, Barddhaman, Birbhum, Jalpaiguri, Koch Bihar, Murshidabad, Nadia and Paschim Medinipur	Eight	Haemorrhagic Septicaemia
7.	Bankura, Barddhaman, Birbhum, Dakshin Dinajpur, Maldah, Paschim Medinipur, Puruliya	Seven	Peste desPetitsRuminants
8.	Haora and North Twenty Four Parganas	Two	Sheep & Goat pox
9.	Barddhaman, Dakshin Dinajpur, Haora, Hugli, North Twenty Four Parganas, South Twenty Four Parganas	Six	Theileriosis
10.	Barddhaman and Hugli	Two	Trypanosomosis

iii) Diseases, Species affected Clinical signs and its preventive measures.

Sl No.	Disease	Species Affected	Clinical Signs	Preventive Measures
1 ICAR	Anthrax (AX)	Most of the mammals and ruminants are highly susceptible. Pigs and Horses are moderately susceptible. Carnivores are relatively resistant.	Convulsion and sudden death with oozing of blood from natural orifices such as rectum and nose prior to death. Occasionally oedema develops in the throat and shoulder over a period of one week before death.	Ring vaccination and reporting of the disease is advised. Vaccination to be done in consultation with the veterinarians and as decided by state animal husbandry authorities. Strict biosecurity measures may be followed. Carcass may be disposed by deep burying covered with lime powder. Contaminated area may be disinfected with 4% formalin or 10% caustic soda. Grazing area may be restricted.
2	Babesiosis (BA)	Cattle. Cross breeds are more susceptible.	High temperature, jaundice like symptoms, yellowish mucosal membrane of eye, rectum and coffee colour urine.	Periodical application of acaricides in and around the animal shed and on the animals. For therapeutic application, Diaminazine or Imidocarb can be useful.
3.	Black Quarter (BQ)	Common disease of cattle and sheep, but occasionally goats and pigs also suffer from the disease.	High fever and lameness followed by swelling in the neck, shoulder, lumbar, gluteal and sacral regions. Skin over the affected area become dark and crepitate on palpation. Loss of feed intake, colic, lateral recumbency, dyspnoea and death.	Affected animals may be treated with suitable antibiotics. Vaccination to be done in consultation with the veterinarians and as decided by state animal husbandry authorities. Strict biosecurity measures may be followed. Grazing area may be restricted. Carcass may be disposed hygienically.

4.	Bluetongue (BT)	Sheep are more	Fever, swelling of face,	Vector control using
		susceptible than goats.	neck, eyelids respiratory distress, nasal discharge, Salivation, necrotic ulcers on tongue, dental pad, gum, lips hyperaemia of muzzle and may bleed at muco-cutaneous junction. Affected tongue may become swollen, cyanotic and purple blue in colour –	insecticides and good water management. Vaccination of susceptible animals preferably in the month of May. Do not shear sheep during winter months. Restriction in animal movement, segregation of affected animals and symptomatic treatment.
			'bluetongue'.	Strict biosecurity measures.
5. CAR	Enterotoxaemia (ET)	Common disease of sheep and goats especially among the young animals.	Dullness, opisthosomas, convulsions, coma and sudden death. Affected adult sheep, which survive for several days May show diarrhoea and staggering.	Affected animals may be treated with suitable antibiotics. Vaccination to be done in consultation with the veterinarians and as decided by State Animal Husbandry Authorities. Strict biosecurity measures may be followed. Carcass may be disposed hygienically. Grazing area to be restricted, stall fed, vitamins and probiotics may be provided.
6.	Fascioliasis (FA)	Cattle, buffalo, sheep and goats.	Progressive anaemia, pale mucous membrane, sub- mandibular oedema (Bottle jaw), loss of appetite, weakness, isolated from flock while grazing, loss in production.	The animal should not be allowed to graze in water stagnant fields or submerged fodder should not be given directly to the animals. The submerged fodder can be processed through hay/silage preparation in order to destroy the metacercariae. The affected animals can be treated with Carbon tetrachloride/ Rafoxanide/Nitroxynil/ Niclofolan /Closantel/Oxyclozanide,

				under the strict supervision of veterinarian.
7.	Foot and Mouth Disease (FMD)	Cattle, buffalo, sheep, goats and pigs are often affected domesticated species, but the disease is more severe in cattle and pigs.	Fever, loss of feed intake, drop in milk production, drooling of saliva like ropey string, vesicles develop on the tongue, lips, gums, and palate and eventually rupture. Concurrent to oral lesions, vesicles also appear in inter digital skin and coronary band of the feet. The animal may open and close its mouth with a characteristic smacking sound. Sheep and goats may show lameness. In pigs, lesions may be seen on snout and also on the feet.	Regular vaccination and seromonitoring. Disinfection with sodium carbonate (4%) or 10% washing soda and strict biosecurity measures to be followed and animal movement may be controlled.
8.	Haemorrhagic septicaemia (HS)	Common disease for cattle and buffaloes, but can also occur among other species such as pigs, sheep, goats and many wild animals.	The disease starts with high fever, respiratory distress and haemorrhages maybe seen on the mucous membranes. There is lacrymation, nasal discharge, drop in milk production and anorexia. As the disease progress ear droops and the animals will be prostrated with cyanosis of mucous membranes. There may be oedema along the head, neck, thorax, vulva and anal areas. Sudden death occurs within few hours of clinical signs.	biosecurity measures may be followed. Carcass may be disposed hygienically and stress factors may be reduced by following good animal husbandry practices.

9.	Peste desPetits Ruminants (<i>PPR</i>)	Goats and sheep are most affected domestic animals.	Fever, nasal and ocular discharge, respiratory distress, necrotic lesions in buccal mucosa, gum, dental pad, palate, tongue and diarrhoea. Animals may die because of dehydration and pneumonia.	animals of above 3 months
10. ICAR	Sheep and Goat pox (SGP)	Sheep and Goats	Respiratory distress and pock lesions over the non- hairy parts of body, more common in teat, udder, scortum, head, neck, ear, perineum, inner aspect of thighs and under tail.	Vaccination of susceptible animals of above 3 months old age. Symptomatic treatment of affected animals. Restriction on animal movement, strict biosecurity measures and proper disposal of carcass.
11.	Swine Fever (SF)	Pigs	Fever, Conjunctivitis, purplish discolouration of snout, ears, abdomen, inner side of the legs and staggering gait.	Vaccination of susceptible animals. Restriction on animal movement, strict biosecurity measures and proper disposal of carcass.
12.	Theileriosis (TE)	Large Ruminants. Cross bred cattle are more vulnerable.	High temperature, yellowish eye, sometime eye maybe heavily swollen, icteric mucosal membrane of rectum, dark yellowish urine, sometime may reach to coffee colour. Antibiotic is of no use to check the fever.	Periodical application of acaricides in and around the animal shed and on the animals. Therapeutic treatment with Buparvaquone can be useful in both early and advanced stages of the infection.

13.	Trypanosomosis (TR)	Domestic and wild carnivores and herbivores including cattle, buffalo, horse, donkey, camel, dog and cats. Buffaloes are known as carriers.	Fluctuating high fever which is not responded by antibiotics, swollen lymph gland, chronic emaciation and weakness, loss of appetite, gradual loss of production.	< 15 ·
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iv) Risk Prediction - Livestock Disease Forewarning Maps



Risk Prediction of Anthrax for the month of June 2022



Risk Prediction of Babesiosis for the month of June 2022



Risk Prediction of Black quarter for the month of June 2022



Risk Prediction of Bluetongue for the month of June 2022



Risk Prediction of Enterotoxemia for the month of June 2022



Risk Prediction of Fascioliasis for the month of June 2022



Risk Prediction of Foot and mouth disease for the month of June 2022



Risk Prediction of Haemorrhagic septicaemia for the month of June 2022



Risk Prediction of Peste des petits ruminants for the month of June 2022



Risk Prediction of Sheep and Goat pox for the month of June 2022



Risk Prediction of Swine fever for the month of June 2022



Risk Prediction of Theileriosis for the month of June 2022



Risk Prediction of Trypanosomiasis for the month of June 2022









VI. SIGNIFICANT WEATHER PARAMETERS TABLE

D isease Names	Significant Parameters			
Anthrax	Precipitable Water, Surface Pressure			
Babesiosis	Air Temperature, Precipitation, Surface Pressure, Minimum Temperature, Vapour Pressure, Dew Point Temperature, Wind Speed, NDVI and LST Night			
Black quarter	Precipitable Water, Precipitation, Surface Pressure, Sea Level Pressure and Vapour Pressure			
Bluetongue	Air Temperature, Cloud, Precipitable Water, Precipitation, Surface Pressure, Uwind, Vwind, Vapour Pressure, Elevation, NDVI and PET			
Classical Swine Fever	Cloud, Precipitation, relative humidity Minimum Temperature, Vapour Pressure and Rainfall			
Enterotoxaemia	Surface Pressure, NDVI and PET			
Fascioliasis	Air Temperature, Precipitation, relative humidity Temperature, Maximum Temperature, Vapour Pressure, Vwind, Rainfall, Soil Moisture, NDVI and EVI			
Foot and Mouth Disease	Precipitable Water, Uwind, Vwind, wet_dry frequency, LST Night and EVI			
Haemorrhagic Septicaemia	Cloud, Precipitation and Vwind			
Peste des Petits Ruminants	Cloud, Precipitable Water, Surface Pressure, Sea Level Pressure, Maximum Temperature, Vwind and NDVI			
Sheep & Goat pox	Cloud, Surface Pressure, Maximum Temperature, Vwind, DTP, NDVI and PET			
Theileriosis	Air Temperature, Precipitation, Vapour Pressure, NDVI and LST Night			
Trypanosomosis	Air Temperature, Precipitation, Surface Pressure, Vapour Pressure, NDVI and LST Night			





the villagers. The department, through the press release also appealed to everyone to report such matters to AH&VS, deputy director & principal investigator, AICRP-ADMAS, Dr S. Amenla Walling, in a press release Dr. Budhi Lama, and other officials from the department. The press release added that the area is prone to such vaccinate their animals against such outbreaks. The team told the villagers that even an outbreak can be The director appreciated the CVO Dimapur and his Rapid Response Team for their quick action after receiving the nearest Veterinary Health Centre (so that qualified staff may intervene quickly), instead of publicizing it in septicaemia (HS), a team from Animal Husbandry and Veterinary Services (AH&VS) department visited the affected areas under Medziphema on April 12. (Haemorrhagic septicaemia is a contagious bacterial disease kind of disease outbreaks and the department officials reminded villagers to cooperate with the department and information and for remaining stationed in the outbreak area to date. Free medicine was also distributed among other ways. It stated that the department is prepared to extend services to any outbreak of diseases in animals Following reports of a good number of buffaloes dying in a recent outbreak of suspected Haemorrhagic The villagers admitted in the meeting that they had not reported the recent outbreak to the department initially. reported that the team consisted of the department's director, Dr Temsumeren, along with additional director, contained more effectively if villagers report the matter on time to the nearest Veterinary Health Centre. that affects cattle and water buffaloes with a high mortality rate in infected animals).

to control such things. The press release also pointed out that to control the recent outbreak, the department had to direct its officials to make their own transport arrangements to go to the affected areas because the State Election department did not consider an appeal to exempt the department's emergency duty vehicle from election duty. Meanwhile, when contacted, Dr S. Armenia Walling told Nagaland Post that it is difficult to say if the disease has been fully contained since its free grazing season for the animals, but the department is doing its best under

	NIVEDI PREDICITONS	EDICITO	SZ
Districts of Nagaland	HS prediction for February 2019	HS prediction for March 2019	HS prediction for April 2019
Peren	VLR	VLR	VHR
Dimapur	VLR	NR	НR
Kohima	VLR	VLR	NR
Wokha	VLR	NR	VLR



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FMD POST PREDICTION

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ProMED-mail is a program of the International Society for Infectious Diseases http://www.isid.org

https://www.sentinelassam.com/north-east-india-news/arunachal-news/many-mithuns-infected-with-fmd-foot-and-mouth-disease-533689 Date: Tue 13 Apr 2021 6:32 AM IST Source: The Sentinel [edited]

1 1 1 1 1 1 1 1						Livest	Livestock Diseases	eases					
Districts of Arunachai Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FND	HS	PPR	S&G Por	SF	Theileriosis	Trypanosomi asis
Anjaw	NR	NR	N	NR	NR	NR	NR	NR	N	NR	R	NR	NR
Changlang	NR	NR	R	NR	NR	NR	M	N	R	R	Ŕ	NR	R
Dibang Valley	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
East Kameng	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	R	NR	R
East Siang	NR	NR	NR	NR	NR	NR	NHR	NR	VLR	NR	VHR	NR	R
Kurung Kumey	NR	NR	NR	NR	NR	NR	NR	R	NR	NR	R	NR	Ŕ
Lohit	NR	VLR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lower Dibang Valley	NR	NR	VLR	NR	NR	NR	VLR	VLR	VLR	NR	VHR	NR	R
Lower Subansiri	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	VLR	NR
Papum Pare	NR	NR	NR	NR	NR	NR	VHR	NR	VHR	NR	VHR	NR	R
Tawang	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Tirap	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	Ř
Upper Stang	NR	NR	R	NR	NR	NR	NR	N	NR	R	R	NR	R
Upper Subansiri	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
West Kameng	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
West Siang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

A large number of mithuns [or gayal, a large domestic bovine] have been affected by FMD (foot-and-mouth disease) in various parts of Arunachal Pradesh, and a few have died, said official sources on Monday [13 Apr 2021]. Animal Husbandry & Veterinary Department Deputy director Dr Taba Heli, a top mithun expert in North Eastern region, reported that the disease has taken a severe form in the entire Siang belt, particularly in East Siang, West Siang, and Upper Siang districts.

Though the number of deaths is yet to be known, the disease has spread in scattered areas of Papum Pare district also. The Department has allocated district funds for procurement of medicines to take all possible steps to contain the disease, he said. Mithun deaths have been reported also from Itanagar and Jullang area.

Pointing out that mithuns are the pride of indigenous people in the state, Nyishi Elite Society president Bengia Tolum has urged the department to take all possible steps to save them.

--Communicated by: ProMED-SoAs from HealthMap Alerts cpromed-SoAs@promedmail.org>



Andaman and Nicobar Report June-2020



HIMACHAL PRADESH REPORT AUGUST-2020

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	District wise Livestock Disease Risk Forewarning for September 2021: Karnataka	vise Liv	estock	Disea	se R	isk F	orewa	rning	for	Sept	emb	er 20	21: K	arnataka
FOR INFECTIOUS DISEASES	Districts of Karnataka	Anthrax	Babcsiosis	80	BI	5	Liv	Livestock Diseases sis FMD 1	HIS	844	S&G	SF	Theileri	Trypanesomosis
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FOOT & MOUTH DISEASE - INDIA (10): (KARNATAKA) CATTLE	Mandya	NR	NR	NR	VLR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
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Date: Fri 24 Sep 2021 18:33 IST	Ultara Kannada	NK	NK	M	NK	NK	NK	HK	VLR		NR	NK	NK	NK
Source: The Hindu [edited] https://www.thehindu.com/news/national/karnataka/foot-and-mouth-diseaso-reported-in-hassan/article36652024.ace	Varticle 3665202	24.ece	NK	VLR	VLR	NK	NR	VLR	VLR	VLR	NK	NK	NK	NK
The outbreak of foot-and-mouth disease (FMD) among cattle in Hassan has left both farmers and veterinarians worried.	veterinarians wo	orried.												
Cases have been reported from Arkalgud, Arsikere, Channarayapatna, and Sakleshpur taluks. While the farming community, depending on cattle for regular income, are worried about their livelihood, the veterinarians are struggling hard to provide treatment, amidst a shortage of staff members.	le the farming co	ammunity.	depending	g on catt	le for r	egular	ncome, a	re worri	ed abc	out the	ir livelih	rood, th	Ū.	
The Department of Veterinary and Animal Husbandry has reported over 150 cases in the district so far. As of [Fri 24 Sep 2021], the animals in 16 villages of the district are being treated. So far the death of one animal had been reported. However, farmers claim more animals have died over the last month. The death of cattle has an impact on milk production, affecting milk producers.	o far. As of [Fri 24 re death of cattle	4 Sep 2021 e has an im	l, the anir pact on m	nals in 10 ilk produ	5 villag	es of th affectio	te district ng milk pro	are beir oducers	ug trea	ited. So	far the	e death	of one	
The vaccination for cattle is done once in six months under the National Animal Disease Control Programme. However, the vaccination drive was not done in the last year, owing to the COVID-19 pandemic. "The outbreak of foot and mouth disease is due to the failure of the government in conducting the vaccination. If the vaccination had been done as per the schedule, farmers would not have suffered". H Yoga Ramesh, president of Potato Club in Arkaigud, told media.	rogramme. Howe ination, If the vac	ever, the vi coination h	ocination ad been d	drive wo	as not o	schedu	the last y lie, farme	ear, owi	ng to t I not h	he CO'	VID-19	pander "H Yog	nic. "The	1
Following reports of the disease, veterinarians have been treating affected animals. "Against 24 sanctioned posts of veterinary doctors in Arkaigud taluk, we are only five people. Every doctor is in charge of 2 or more hospitals. We are struggling hard to treat animals", said a veterinarian.	inclianed posts o	of veterina	y doctors	in Arka	gud ta	luk, we	are only f	ive peol	ple. Ev	ery do	tor is i	n charg	e of 2 or	
Unless the vacancies were filled up the department cannot deliver services fully. There was a shortage of staff members of other cadres as well, he added	tage of staff mer	mbers of o	ther cadre	aw se se	II, he ac	Ided.								
Considering the cases, the department has launched a ring vaccination programme. Under this, the animals in a 5-kilometre [31 mi] radius of the village, where the disease breakout was reported, would be vaccinated. "We have sufficient stock for the ring vaccination. We are planning a mass vaccination in October [2021], where we will cover 6.58 takh [658 000] animals," said KR Ramesh, Deputy Director of Veterinary and Animal Husbandry in Hassan.	e animals in a 5-1 i in October (202	kilometre 11, where v	3.1 mi] rad	flus of th er 6.581	ie villaș akh (6	se, whe 58 000	re the dis)] animals.	ease br	eakou' R Ran	t was n tesh, D	eported eputy [1, would	fbe	
Communicated by:														
Promeu-SoAs from HealthMap Alerts														

9.1 Correlational Assessment

The number of outbreaks predicted and outbreaks actually reported were reported in table 9.1. It has been observed from the table that outbreaks predicted two months in advance and timely alerts were issued that helped the stakeholders to take appropriate preventive measures with in time and accordingly the reported outbreaks were very less. Though the use of artificial intelligence system is more beneficial for accurately predicting the livestock disease outbreaks, there are yet number of limitations, namely, there are expected to be under reporting and also non-reporting cases which created the uncertainties in the model predictions while translating model inputs in to model outputs. However, identifying these uncertainties in the prediction using statistical models and highlighting the importance of quality data may improve the model accuracy and confidence while building the model for livestock disease forecasting.

		Januai	ry-2021	Februa	ry-2021
SI No	Livestock diseases	No. of districts predicted for the disease risk events	No. of districts reported the disease outbreaks*	No. of districts predicted for the disease risk events	No. of districts reported the disease outbreaks*
ICAR	Anthrax	20	3	29	राप्4ास्विस ∾ I V E D
2	Babesiosis	47	56	44	42
3	Black quarter	37	0	47	5
4	Bluetongue	0	1	0	2
5	Enterotoxaemia	20	6	21	8
6	Fascioliasis	56	50	46	50
7	Foot and mouth disease	69	13	54	21
8	Haemorrhagic septicaemia	40	9	47	6
9	Peste des petits ruminants	52	21	69	16
10	Sheep & Goat pox	41	9	35	6
11	Swine fever	43	6	54	4
12	Theileriosis	41	72	51	43
13	Trypanosomosis	37	81	45	43

Table 9.1: Number of districts predicted for livestock diseases risk events and reported outbreaks

*Which takes in to account of action taken for prediction and non-reporting of cases

10. Launch of Mobile Android app&link to download

Livestock forewarning application (LDF) can be downloaded following the link provided:<u>http://www.nivedi.res.in/android_nadres/LDF.apk</u>and google play storelinkalso provided <u>https://play.google.com/store/apps/details?id=info.androidhive.ldf</u>

Further launch of LDF application was done, the news provided below.



Livestock Disease Forewarning (LDF Mobile App)

10:35 🌣 🖀 🕨 🔍 🕶 🖬	10:36 💠 🖲 🖻 🕨 🔍 🕶 🖬 🖗	10:36 💠 🌒 🖹 🕨 🔍 🔽 🕯
≡ Home	■ Livestock Disease Forewarning	\equiv Livestock Disease Forewarning
Select a Language	Livestock Disease Forewarning	Livestock Disease Forewarning
-	Select disease name	Select disease name
	Anthrax 👻	Anthrax
माडोमहा ICAR	Select State name	
Livestock Disease Forewarning	KARNATAKA 👻	🕡 Result
About App:	Select district name	Disease name : Anthrax State name : KARNATAKA
National Institute of Veterinary Epidemiology and Disease Informatics (NIVEDI) has the mandate to	Bagalkot 👻	District name : Bagalkot Month: December Result: Very Low Risk
carry out research activities in the area of veterinary epidemiology and disease informatics. With the eradication of RP successfully, India has	Select month name	S OK
not only proved its ability to face the challenges, but also to succeed, despite various limitations. Similar efforts are needed to control and	December -	December
eradicate the diseases like FMD, PPR, Brucellosis, CSF, BT, HS etc., which cause huge economic loss annually to the livestock industry. To this end,	SUBMIT	SUBMIT
ICAR-NIVEDI has identified 13 priority diseases, based on the past incidence patterns and has built a strong database of these diseases. The	RESET	RESET
database, which is backbone of these diseases. The database, which is backbone of the National Animal Disease Referral Expert System (NADRES), is used for providing monthly livestock disease		
	< ● ■	
< • E	< • •	• •

To extend the reach of the NADRES forewarning bulletin among the various stakeholders, a Mobile Application named Livestock Disease forewarning app "LDF-Mobile *App*" was developed. The forewarning methodology adapted in the "mobile app" remains the same as monthly bulletin; it provides user interface to know the predicted forewarning results stored in NADRES MySQL database. A PHP web-based service is developed in Java to extract the results of forewarning two months in advance by keying state name, district name and disease name and display the same in the mobile app. In addition to forewarning, the LDF-Mobile App also provides the details of clinical samples to be collected in case of outbreaks of the listed diseases for laboratory confirmation and immediate preventive measures to be taken up in case of positive prediction/disease confirmation. The LDF mobile app is available at ICAR-NIVEDI website. It is available on Google play store.

11.FARMERS EMPOWERMENT THROUGH IT: DISEASE RISK COMMUNICATION (COLLABORATION WITH FRUITS, NIC, GOVT. OF KARNATAKA)

In addition to NADRES V_2 (The National Animal Disease Referral Expert System), ICAR-NIVEDI collaborated with NIC, Govt. of Karnataka, Karnataka State for sending the SMS alerts directly to farmers who have registered in FRUITS (Farmers Registration and Unified Beneficiary Information System). The information alerts on risk prediction of three livestock diseases were sent through SMS to farmers is presented in Table 11.A. During 2022, a total of **7,86,945** SMS alerts were sent to farmers.

Table 11.A: Number of famers received the SMS alert through FRUITS application during M	Iarch
2022.	

Disease Name	District Name	No. of farmers received SMS	Disease Name	District Name	No. of farmers received SMS
Anthrax	Bellary	2915		Bangalore Urban	25754
	Davanagere	25375		Bangalore Rural	36869
	Koppal	15380		Chikkaballapur	38695
	Mysore	113070	TH CD	Dakshina Kannada	37590
	Tumkur	91427	FMD	Kodagu	2975
BQ	Hassan	82780		Koppal	15381
	Mysore	113037		Ramanagar	58704
	Tumkur	91427			
1	and a second second		E CO	Grand Total	786945

12. Appendix

a) R Code

#parsmonth_number=8; year_number=2006; current_year=2017;

nadres_func=function (current_year, year_number, month_number)

```
{
args = commandArgs(trailingOnly=TRUE)
if (length(args)<3) {
stop("Correct number of arguments must be supplied", call.=FALSE)
      }
current year=args[1]
year number=args[2]
month number=args[3]
df total<-NULL
month name=data.frame(month=c(1:12),
month names=c("February", "February", "October", "October", "May", "October", "October",
","October","October","January","January")
)
ss<-fread(file="NADRES.csv",header=T,check.names = F)
col pars=names(ss)
vars = paste(col pars[7:ncol(ss)],collapse = "+")
options(verbose = F)
```

for(disease in c(8,10,11,12,24,31,35,37,48,60,62,65,70,72,79))

{

disease=8

```
rs<-dbSendQuery(mydb,"SELECT
index state.state name,index state.state id,index district.district id, index district.district name,
```
year_list.year, outbreak_data_final.month, ls_sp_index.species_name,disease_master.disease_id, disease_master.disease_name, outbreak_data_final.number_of_outbreaks, outbreak_data_final.number_susceptible, outbreak_data_final.number_of_attacks, outbreak_data_final.number_of_deaths

FROM ls_sp_index INNER JOIN (year_list INNER JOIN (disease_master INNER JOIN
(index_district INNER JOIN (index_state INNER JOIN outbreak_data_final ON
index_state.state_id= outbreak_data_final.state_id) ON index_district.district_id =
outbreak_data_final.district_id) ON disease_master.disease_id= outbreak_data_final.disease_id)
ON year_list.year = outbreak_data_final.year) ON ls_sp_index.species_id=
outbreak_data_final.species_id; ")
data = fetch(rs, n=-1)
year change
data<-subset(data,data\$year>=year_number&data\$disease_id==disease)
df<-sqldf("SELECT
state_id,state_name,district_id,district_name,disease_id,disease_name,month,sum(number_of_outbr
eaks)as outbreak FROM data GROUP BY
state_id,district_id,state_name,district_name,month,disease_id,disease_name",drv="SQLite")
ss1<-subset(ss,ss\$disease_id==disease)
attach(ss1,warn.conflicts = F)</pre>

attach(df, warn.conflicts = F)

dd<-merge(ss1, df, by = c("state_id","district_id","disease_id","month"),all.x=TRUE)

attach(dd,warn.conflicts = F)

out<-data.frame(outbreak)

out<-ifelse(outbreak>=1,1,0)

out[is.na(out)]<-0

final<-cbind(dd,out)

final1<-final[which(final\$disease_id==disease),]

cat("For disease: ",as.character(unique(ss1[,"disease_name"])),"\n")

```
ncs = ncol(final1)-5
```

```
temp = data.frame(final1[,8:ncs])
```

```
for(i in 1:ncol(temp)){
```

temp[is.na(temp[,i]), i] <- mean(temp[,i], na.rm = TRUE)

}

final2<-

cbind(final1\$state_id,final1\$state_name.x,final1\$district_id,final1\$district_name.x,final1\$disease_id,final1\$disease_name.x,final1\$out,final1\$month,temp)

setnames(final2,old=c("final1\$state_id","final1\$state_name.x","final1\$district_id","final1\$district_ name.x","final1\$disease_id","final1\$disease_name.x","final1\$out","final1\$month"),new=c("state_id ","state_name","district_id","district_name","disease_id","disease_name","out","month"))

formula=paste("out ~",vars)

as.formula(formula)

model<-glm(formula,data= final2, family = binomial(link="logit"),maxit=20)</pre>

```
new<-data.frame(final2[,8:ncol(final2)])
```

```
prediction<-predict(model,type="response")
```

```
n2=randomForest(as.formula(formula),final2)
prediction_rf<-predict(n2,type="response")
```

```
gbm_model=gbm.step(data=final2, gbm.x = 8:ncol(final2), gbm.y = 7, family = "bernoulli", tree.complexity = 1, learning.rate = 0.01,
```

```
bag.fraction= 0.5, n.trees = 5,keep.fold.fit=T,tolerance.method="fixed", step.size= 5,n.folds = 10)
```

prediction_gbm<-predict(gbm_model,n.trees=gbm_model\$gbm.call\$best.trees,type="response")
prediction=numeric()</pre>

```
for (i in 1:length(prediction_glm)) {
```

```
# if(prediction_glm[i]>prediction_rf[i])
# {
```

```
# if(prediction_glm[i]>prediction_gbm[i])
```

```
# {
```

```
# prediction[i]=prediction_glm[i]
```

```
# }
```

```
if(prediction_glm[i] >= prediction_gbm[i] &&prediction_glm[i] >= prediction_rf[i])
{
```

```
prediction[i]=prediction glm[i];
   }
   if(prediction gbm[i] >= prediction glm[i] &&prediction gbm[i] >= prediction rf[i])
    prediction[i]=prediction gbm[i];
   }
   if(prediction rf[i] >= prediction glm[i] &&prediction rf[i] >= prediction gbm[i]) {
    prediction[i]=prediction rf[i];
   }
summary(prediction)
vv<-round(prediction,2)
 df1<-cbind(final2,vv)
df total<-rbind(df total,df1)
gc()
}
f=function(m){
if(m<=0.0) i=1
else if(m>=0.0 && m<=0.20) i=2
else if(m>=0.21 && m<=0.40) i=3
else if(m>=0.41 && m<=0.60) i=4
else if(m>=0.61 && m<=0.80) i=5
elsei=6
}
df_total$cate=factor(mapply(f,df_total$vv),levels=1:6,labels=c("","","","MR","","HR"))
write.csv(df total, "nadres outbreak.csv")
###### ACCURACY
df total=read.csv("nadres outbreak.csv",header = T)
```

```
dir.create(path = paste(month_name[month_number,2],current_year))
```

df_poa=df_total

```
df_poa$cate=factor(mapply(f,df_poa$vv),levels=1:6,labels=c(0,0,0,0,1,1))
```

df_poa=df_poa[which(df_poa\$month==month_name[month_number,1]),]

df_p=df_poa[,c("disease_name","out","cate")]

df_acc=cbind(data.frame(c(1:ow(df_tot_res))),data.frame(df_tp_tn[,1]),(df_tp_tn[,2]/df_tot_res[,2])*

```
df_acc=setNames(df_acc,c("No","Disease","Accuracy"))
```

print(df acc)

dis_acc=paste(paste(month_name[month_number,2]," ",current_year,"/",sep= ""),"Disease Accuracy ",month_name[month_number,2]," ",current_year,".csv",sep="")

```
write.csv(df_acc,dis_acc,row.names = F)
```

##########PLOT

i=1

```
plot_dir=paste(paste(month_name[month_number,2],"
",current_year,"/",sep=""),month_name[month_number,2]," ",current_year," N",sep="")
dir.create(path = plot_dir)
disease = c(8,10,11,12,31,35,37,48,60,65,70,72,79)
while(i<=length(disease))
{</pre>
```

```
kar=readOGR(dsn = "1shp/2011_Dist.shp",verbose = FALSE)
```

cols=as.character(unique(df_total[df_total\$disease_id==disease[i],"disease_name"]))

```
df_disease=df_total[which(df_total$month==month_name[month_number,1] &df_total$disease_id==disease[i]),]
```

```
df_disease=df_disease[,c(2:5,(ncol(df_disease)-1))]
```

```
df_disease=setNames(df_disease,c("ST_CEN_CD","state_name","DT_CEN_CD","district_name"," vv"))
```

```
kar@data=merge(data.frame(kar@data),data.frame(df_disease),by=c("ST_CEN_CD","DT_CEN_C
D"),all.x=T)
```

kar\$vv[is.na(kar\$vv)]<-0

#View(kar@data)

```
colours<-c("#FFFFFF","#FFFF00","#FFC1C1","#FF7150","#FF8500","#FF0000")
```

kar\$lb=factor(mapply(f,kar\$vv),levels=1:6,labels=c("No Risk / No Data","Very Low Risk","LowRisk","MediumRisk","HighRisk","Very High Risk"))



cols=gsub("&", "and", cols)

disname= gsub("\\."," ",cols)

```
cat("Plot for disease:",disname,"\n")
```

plot_loc=paste(plot_dir,"/",disname,"/",sep="")

dir.create(plot_loc)

file_name=paste(plot_loc,disname,".png",sep="")

plot_title= paste(disname," risk prediction(",month_name[month_number,2],"

",current_year,")",sep="")

png(file_name)

print(spplot(obj=kar,c("lb"),col.regions=colours,main= plot_title,scales=list(draw = TRUE)))

dev.off()

i=i+1

b) Abbreviations

NADRES	:	National Animal Disease Referral Expert System	
R	1	R environment for statistical computing	
BQ	:	Black Quarter	
BT	:	Bluetongue	
ET	:	Enterotoxaemia	राप जा स्विसं
FMD	:	Foot and Mouth disease	
HS	:	Haemorrhagic Septicaemia	
PPR	: 1	Peste des Petits Ruminants	
SGP		Sheep and Goat pox	
SF		Swine Fewer	71
hPa	i i i	Hectopascals	- No-
NR		No risk/No data available	X
VLR	:	Very low risk	
LR	:	Low risk	
MR	:	Moderate risk	
HR	:	High risk	
VHR	:	Very high risk	

13. INFECTIONWITHSARS-COV-2INANIMALS

Aetiology, Epidemiology, Diagnosis, PreventionandControl

Lastupdatedon 3July2020

SARS-Coronavirus-2 (SARS-CoV-2) is the pathogenic agent that causes the disease COVID-19 and was first reported in December 2019. SARS-CoV-2 is thought to have emerged from an animal source and then spilled-over to the human population. Although genetically closely related viruses have been isolated from *Rhinolophus* bats, the exact source of SARS-CoV-2 and route of introduction into the human population has not been established.

The current pandemic of COVID-19 is being sustained through human-to-human transmission. Animal infections with SARS-CoV-2 have been reported by several countries. Several animal species have proven to be susceptible (Table 1) to infection with SARS-CoV-2 either naturally or by experimental infection. Important livestock species (pigs and poultry) have been demonstrated not to be susceptible to infection through experimental studies. Further studies are needed to understand if and how different animals could be affected by SARS-CoV-2.

It is important to monitor infections in animals to better understand their epidemiological significance for animal health, biodiversity, and human health. Evidence from risk assessments, epidemiological investigations, and experimental studies do not suggest that live animals or animal products play a role in SARS-CoV-2 infection of humans.

Infection with SARS-CoV-2 is not included in the OIE List of Diseases. However, consistent with the reporting obligations of Members outlined in Articles 1.1.4. and 1.1.6. of the OIE *Terrestrial Animal Health Code* relating to emerging diseases, the disease should be notified to the OIE through the OIE's World Animal Health Information System or via email.

Aetiology

Classification of the causative agent

Corona viruses (CoVs) are enveloped, positive-sense, single-stranded RNA viruses. SARS-CoV-2 is a *beta coronavirus*, a genus that includes several coronaviruses (SARS-CoV, MERS-CoV, bat SARS-like CoV, andothers) isolated from humans, bats, camels, and other animals. Susceptibility to physical and chemical action.

SARS-CoV-2 is inactivated by 62–71 % ethanol, 0.5% hydrogen peroxide or 0.1% sodium hypochlorite, within 1 minute, or 0.05–0.2 % benzalkonium chloride or 0.02 % chlorhexidine digluconate.

Survival:

In experimental conditions, SARS-CoV-2 remained viable in the environment after aerosolisation for at least 180 minutes. Experiences with other CoVs such as SARS-CoV, MERS-CoV, or endemic human corona viruses show that:

They can persist on surfaces such as metal, glass or plastic for up to 9 days, but can be efficiently inactivated by surface disinfection procedures, as listed above.

SARS-CoV was found to remain infectious for 14 days at 4 °C, but for only 2 days at 20 °C in sewage water.

Epidemiology

Hosts

Although current evidence suggests that SARS-CoV-2 emerged from an animal source, that source has yet to be identified. The pandemic is driven by person-to-person transmission through respiratory droplets from coughing, sneezing, and talking. Genetic sequence data reveal that SARS-CoV-2 is genetically closely related to other corona viruses circulating in *Rhinolophus* bat (horseshoe bat) populations. To date, there is not enough scientific evidence to identify the source of SARS-CoV-2 or to explain the original route of transmission to humans (which may involve an intermediate host).

Several animal species have tested positive for SARS-CoV-2, mostly as a result of close contact with humans infected with SARS-CoV-2. In addition, preliminary findings from experimental infection studies suggest that poultry and pigs are not susceptible to SARS-CoV-2 infection. The list of animal species for which information on natural or experimental infection is available is presented in Table 13.1.



Species	Type of infection	Susceptibility (none/low/high)	Clinical signs	Transmission
-				
Pigs	Experimental	None	No	No
Poultry (chicken, ducks, and turkeys)	Experimental	None	No	No
Dogs	Natural and experimental	Low	No (possible in some cases)	No राषजास
Cats(domestic)	Natural and experimental	High	Yes (none to very mild in some cases)	Yes, between cats
Tigersandlions	Natural	High	Yes	Yes, between animals
Ferrets	Experimental	High	No (very mild in some cases)	Yes, between ferrets
Minks (Americanminks, Neovisonvison)	Natural	High	Yes	Yes, between minks and suggested from mink to humans
Egyptian fruit bats (Rousettus aegyptiacus)	Experimental	High	No	Yes, between Fruitbats
Golden Syrian hamsters	Experimental	High	Yes (none to very mild in some cases)	Yes, between hamsters
Macaques (<i>Macacafascicularis</i> and <i>Macacamulatta</i>)	Experimental	High	Ycs	Ycs

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Table13.1. Summaryof findingsinanimalstodate

Transmission

Information on the routes of transmission of SARS-CoV-2 among animals is limited. However, as for other respiratory viruses, it appears to be transmitted to animals and between animals by direct contact (e.g droplets). SARS-CoV-2 has been found in secretions from the respiratory tract and in faeces.

Viraemia, incubation and infectious period

In laboratory settings, the incubation period in animals appears to be similar to the one seen in humans (i.e., between 2 and 14 days, with a mean duration of 5 days). However, more studies are required to solidly estimate the mean duration of incubation and the infectious periods.

Sources of virus

The main source of the virus is respiratory droplets and respiratory secretions, although it is possible to isolate SARS-CoV-2 from faeces of infected animals.

Pathogenesis

In laboratory settings, infected animals showed presence of the virus in the respiratory tract and, in some cases, lesions in the trachea and lungs, associated with dyspnoea and cough.

Occurrence and impact

There have been sporadic reports of companion animals and captive wild animals infected with SARS-CoV-2. With regard to production animals, to date, SARS-CoV-2 has only affected mink farms in the Netherlands, with high morbidity and low mortality.

Diagnosis

Knowledge about the susceptibility of different animal species to SARS-CoV-2 infection and clinical signs is, to date, limited (see Table 1).

Clinical diagnosis

Knowledge about clinical disease manifestations in animals is limited. Current evidence suggests clinical signs may include, but are not limited to, coughing, sneezing, respiratory distress, nasal discharge, ocular discharge, vomiting or diarrhoea, fever and lethargy. As in humans, asymptomatic infections occur.

Lesions

More studies are needed to systematically categorise the lesions resulting from infection with SARS-CoV-2 in animals.

In transgenic mice expressing the human version of the SARS-CoV-2 ACE2 receptor, the typical histopathology outcome was interstitial pneumonia with significant inflammatory cell infiltration around the bronchioles and blood vessels, and viral antigens were detected in bronchial epithelial cells and alveolar epithelial cells. These pathological findings were not seen in wild type mice infected with SARS-CoV-2. In golden Syrian hamsters, histopathological changes were reported in the respiratory tract and spleen. Rhesus macaques infected with SARS-CoV-2 presented lesions similar to those seen in humans. Juvenile cats infected with SARS-CoV-2 presented massive lesions in the

nasal and tracheal mucosa epithelia, and lungs. SARS-CoV-2 can replicate in the upper respiratory tract of ferrets without causing severe disease and only resulting in pathological findings such as severe lymphoplasmacytic perivasculitis and vasculitis, increased numbers of type II pneumocytes, macrophages, and neutrophils in the alveolar septa and alveolar lumen, and mild peri bronchitis in the lungs.

Differential diagnosis

All other causes for respiratory or digestive illness should be excluded before a tentative diagnosis for infection with SARS-CoV-2 is made. Existence of an epidemiological link with a confirmed infection in humans or other animals should be considered when narrowing down the list of differential diagnoses.

Laboratory confirmatory tests are necessary for a final diagnosis.

Laboratory diagnosis

Samples

Depending on the type of test, samples may include single or combinations of oropharyngeal, nasal, and rectal swabs, and blood. Faecal samples may be used in situations where direct sampling is not possible due to risks to the animal or testing staff. Tests should be validated for the purpose, species and matrix to be analysed.

Procedures

Agent identification

- Reverse-transcription polymerase chain reaction (RT-PCR);
- Reverse transcription loop-mediated isothermal amplification (RT-LAMP);
- Other molecular tests developed for use in humans;
- Virus isolation;
- Virus genome sequencing.

Detection of immune response:

- ELISA antibody test;
- Virus neutralisation test (VNT);
- Several other tests for antibody detection.

Prevention and control

Biosecurity and hygiene measures are key to preventing transmission of SARS-CoV-2. People who are suspected or confirmed to be infected with SARS-CoV-2 should restrict contact with mammalian animals, including pets, just like they would with people during their illness.

Animals suspected or confirmed to be infected with SARS-CoV-2 should remain separated from other animals and humans.



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EPIDEMIOLOGY OF COVID-19

0.04 0.14 0.18 0.22 0.25 0.29 0.32 0.36 0.39 0.43 0.47 0.07 0.11 Immunity total % of gained immunity by vaccination % of 0.04 0.11 0.14 0.18 0.22 0.25 0.32 0.36 0.39 0.07 0.29 0.43 0.47 Immunity by Infection % of **Total Vaccine** Administered (uno) 42.26 40.9 46.58 43.25 42.26 41.93 41.25 40.9 40.9 40.9 40.9 40.9 44.51 **Epidemiology of COVID-19 ALL INDIA** (Threshold) Required Immunity herd R₀ 1.762 1.692 1.872 1.802 1.732 1.732 1.722 1.702 1.692 1.692 1.692 1.692 1.692 confirmed R₀ for cases 711.5 116.0 932.2 973.0 996.3 1111.0 1157.6 1101.5 1039.3 809.4 964.3 1146.0 481.1 avg. daily deaths 3.06 2.49 2.23 2.04 1.90 1.82 1.74 1.70 1.64 1.62 1.59 1.57 1.97 CFR Cumulative 34193 41585 49036 56706 63498 88935 15685 82066 95542 101782 25602 Number of 69561 76271 deaths 15-09-20 since 22 Jan-06-08-20 10-09-20 22-08-20 date reached 26-06-20 16-07-20 28-07-20 14-08-20 29-08-20 04-09-20 21-09-20 27-09-20 03-10-20 2020 12 days 8 days 6 days 156 days 9 days 8 days 7 days 6 days 6 days 5 days 6 days 6 days 20 days since 22 aken to Jan-2020 No of days reach Cum 10 Lakh Cases Cum 15 Lakh Cases cum 20 Lakh Cases cum 25 Lakh Cases cum 30 Lakh Cases cum 35 Lakh Cases cum 40 Lakh Cases cum 45 Lakh Cases cum 50 Lakh Cases cum 55 Lakh Cases cum 60 Lakh Cases cum 65 Lakh Cases Infections (5 laks 1st 5 Lakh Cases Increment) Number of

0.50	0.54	0.57	0.61	0.65	0.68	0.72	0.75	1.55	3.40	4.82	6.38	7.53	8.29	9.04	9.46	9.78	10.26	10.68	10.92
								0.76	2.58	3.96	5.48	6.59	7.32	8.03	8.42	8.70	9.15	9.53	9.74
0.50	0.54	0.57	0.61	N 1 20.65	0.68	0.72	0.75	0.79	0.83	0.86	06.0	0.93	0.97	1.00	1.04	1.08	1.11	1.15	1.18
				R Z				10651012	35923500	55180875	76405697	91881530	102000401	111913288	117305344	121207098	127428887	132754608	135658324
41.25	41.93	42.92	43.57	49.29	57.12	67.56	62.59	51.27	52.2	54.59	54.59	52.65	50.3	47.42	45.12	42.59	41.21	39.06	37.93
1.702	1.722	1.752	1.772	1.972	2.332	3.083	2.673	2.052	2.092	2.202	2.202	2.112	2.012	1.902	1.822	1.742	1.701	1.641	1.611
937.0	784.5	591.7	559.3	503.2	498.9	405.3	253.6	119.4	119.4	247.5	464.4	640.5	838.0	981.0	1260.0	1560.0	1882.0	2193.0	2624.0
1.55	1.53	1.51	1.48	1.47	1.46	1.45	1.45	1.42	1.39	1.35	1.32	1.29	1.26	1.24	1 <mark>.</mark> 21	1.19	1.18	1.17	1.18
108334	114610	120527	126121	132162	138648	145136	151727	156385	159370	161843	165101	167642	170179	173123	175649	178769	182533	186920	189544
10-10-20	18-10-20	28-10-20	07-11-20	19-11-20	02-12-20	18-12-20	13-01-21	21-02-21	18-03-21	28-03-21	04-04-21	08-04-21	11-04-21	14-04-21	16-04-21	18-04-21	20-04-21	22-04-21	23-04-21
7 days	8 days	10 days	10 days	12 days	13 days	16 days	26 days	39 days	25 days	10 days	7 days	4 days	3 days	3 days	2 days	2 days	2 days	2 days	1 days
cum 70 Lakh Cases	cum 75 Lakh Cases	cum 80 Lakh Cases	cum 85 Lakh Cases	cum 90 Lakh Cases	cum 95 Lakh Cases	cum 100 Lakh Cases	cum 105 Lakh Cases	cum 110 Lakh Cases	cum 115 Lakh Cases	cum 120 Lakh Cases	cum 125 Lakh Cases	cum 130 Lakh Cases	cum 135 Lakh Cases	cum 140 Lakh Cases	cum 145 Lakh Cases	cum 150 Lakh Cases	cum 155 Lakh Cases	cum 160 Lakh Cases	cum 165 Lakh Cases

11.21	11.48	11.85	12.04	12.27	12.43	12.64	12.79	12.98	13.40	13.60	13.69	14.13	14.32	14.52	14.79	14.98	15.21	15.45	15.67
66.6	10.23	10.55	10.71	10.91	11.03	11.20	11.32	11.48	11.86	12.02	12.08	12.48	12.63	12.80	13.03	13.19	13.38	13.59	13.77
1.22	1.26	1.29	1.33	1987 1 × 1 × 1	1.40	1.44	1.47	1.51	1.54	1.58	1.61	1.65	1.69	1.72	1.76	1.79	1.83	1.87	1.90
139185173	142524947	147053392	149268772	151998107	153626325	156082136	157750752	159931238	165190000	167493857	168304868	173862643	176045577	178361846	181544536	183817204	186410600	189344105	191877460
36.75	35.94	34.68	34.25	33.82	33.77	33.38	32.93	32.93	32.48	32.48	32.02	32.04	31.58	31.58	31.58	31.58	31.10	31.10	31.10
1.581	1.561	1.531	1.521	1.511	1.510	1.501	1.491	1.491	1.481	1.481	1.471	1.471	1.461	1.461	1.461	1.461	1.451	1.451	1.451
2789.0	2771.0	3469.0	3498.0	3523.0	3689.0	3433.0	3780.0	3980.0	4051.0	4077.0	3769.0	4040.0	4120.0	4000.0	3983.0	4217.0	4201.0	4202.0	4098.0
1.15	1.13	1.14	1.13	1.12	1.11	1.11	1.10	1.10	1.10	1.10	1.09	1.11	1.10	1.09	1.10	1.11	1.13	1.14	1.15
195123	197894	204832	208330	211853	215542	222408	226188	230168	238270	242347	246116	254197	258317	262317	270284	278719	287122	295525	303720
25-04-21	26-04-21	28-04-21	29-04-21	30-04-21	01-05-21	03-05-21	04-05-21	05-05-21	07-05-21	08-05-21	09-05-21	11-05-21	12-05-21	13-05-21	15-05-21	17-05-21	19-05-21	21-05-21	23-05-21
2 days	1 days	2 days	1 day	I day	1 day	2 days	1 days	1 days	2 days	1 day	1 day	2 days	1 day	1 day	2 days				
cum 170 Lakh Cases	cum 175 Lakh Cases	cum 180 Lakh Cases	cum 185 Lakh Cases	cum 190 Lakh Cases	cum 195 Lakh Cases	cum 200 Lakh Cases	cum 205 Lakh Cases	cum 210 Lakh Cases	cum 215 Lakh Cases	cum 220 Lakh Cases	cum 225 Lakh Cases	cum 230 Lakh Cases	cum 235 Lakh Cases	cum 240 Lakh Cases	cum 245 Lakh Cases	cum 250 Lakh Cases	cum 255 Lakh Cases	cum 260 Lakh Cases	cum 265 Lakh Cases

16.04	16.43	0 17.00	3 17.83	6 18.94	20.10	22.96	26.55	30.23	34.25	39.46	45.00	\$ 51.65	61.09	71.61					
14.10	14.46	14.99	15.78	16.86	17.99	20.81	24.36	28.01	31.99	37.16	42.67	49.28	58.69	69.17					
1.94	1.97	2.01	2.05	2.08°	2.12	2.15	2.19	2.22	2.26	2.30	2.33	2.37	2.40	2.44					
196463495	201438120	208907723	219831571	223642281	250656362	289961746	339447068	390286930	445696580	517799432	594552135	686693339	817737078	963825324					
31.10	31.10	31.10	31.10	31.10	31.10	31.58	32.04	32.50	32.50	33.40	33.82	34.25	33.82	33.82	,		•		
1.451	1.451	1.451	1.451	1.451	1.451	1.461	1.471	1.481	1.481	1.502	1.511	1.521	1.511	1.511	A.		al france	1	¢
3834.0	3754.0	3402.0	2901.0	2565.0	4155.0	1817.0	1039.0	884.0	780.0	501.0	398.0	499.0	309.0	264.0	383.0	427.0	362.0	250.0	459.0
1.15	1.16	1.18	1.20	1.22	1.27	1.30	1.31	1.33	1.34	1.34	1.34	1.34	1.33	1.33	1.35	1.38	1.36	1.35	1.33
311388	318895	329100	340702	344082	374305	390660	401050	412531	422662	429179	434756	440752	445385	451189	465349	482017	483463	484213	485131
25-05-21	27-05-21	30-05-21	03-06-21	08-06-21	13-06-21	22-06-21	02-07-21	15-07-21	28-07-21	10-08-21	24-08-21	05-09-21	20-09-21	12-10-21	20-11-21	04-01-22	08-01-22	11-01-22	13-01-22
2 days	2 days	3 days	4 days	5 days	5 days	9 days	10 days	13 days	13 days	13 days	14 days	12 days	15 days	22 days	37 days	39 days	4 days	3 days	2 days
cum 270 Lakh Cases	cum 275 Lakh Cases	cum 280 Lakh Cases	cum 285 Lakh Cases	cum 290 Lakh Cases	cum 295 Lakh Cases	cum 300 Lakh Cases	cum 305 Lakh Cases	cum 310 Lakh Cases	cum 315 Lakh Cases	cum 320 Lakh Cases	cum 325 Lakh Cases	cum 330 Lakh Cases	cum 335 Lakh Cases	cum 340 Lakh Cases	cum 345 Lakh Cases	cum 350 Lakh Cases	cum 355 Lakh Cases	cum 360 Lakh Cases	cum 365 Lakh Cases

cum 370 Lakh Cases	2 days	15-01-22	485752	1.31	311.0	ī	1				
cum 375 Lakh Cases	2 days	17-01-22	486451	1.30	350.0	þ	2				
cum 380 Lakh Cases	2 days	19-01-22	487202	1.28	376.0		ı				
cum 385 Lakh Cases	1 day	20-01-22	487693	1.27	491.0	ı.					
cum 390 Lakh Cases	2 days	22-01-22	48884	1.25	596.0		1 at		지 박 SH 및 철 관 N I V F D I		
cum 395 Lakh Cases	1 day	23-01-22	489409	1.24	525.0	ĩ	,				
cum 400Lakh Cases	2 days	25-01-22	490462	1.23	527.0		-				
cum 405 Lakh Cases	2 days	27-01-22	491700	1.21	619.0						
cum 410 Lakh Cases	2 days	29-01-22	493198	1.20	749.0	,		7			
cum 415 Lakh Cases	3 days	01-02-22	496242	1.20	1015.0	'-					100
cum 420 Lakh Cases	2 days	03-02-22	499424	1.19	1591.0	•		8			
Cum 425 Lakh cases											
		DATASOURCE	RCE		The second secon					-]
1.WORLD WIDE DATA SET: 2.VACCINATION DATA SET:		https://data.humdata.org/dataset/novel-coronavirus-2019-ncov-cases https://api.covid19india.org/	<u>rg/dataset/nov</u> ia.org/	el-corona	virus-2019-	ncov-cases			~		
		CF Prepai	R: Case Fata red by Spatia	ality rate al Epide	e , number miology I	c of deaths f ab , ICAR	CFR: Case Fatality rate , number of deaths for every 100 cases Prepared by Spatial Epidemiology Lab , ICAR-NIVEDI, Bengaluru.	cases ngaluru.			





ICAR - National Institute of Veterinary Epidemiology and Disease Informatics

Customer/Client Feedback Form Feedback for the Livestock Diseases Risk Forewarning Bulletin of April -2022, Volume 10 and Issue 04

(Please return this duly fill in after receiving the outbreak report of April -2022)

Sl. No	Disease Name	No. of districts in which outbreaks occurred but not alerted**	Measures taken in case of disease forecasted: Yes or No.**	Any other
1.	Anthrax			
2.	Babesiosis			
3.	Black Quarter			
4.	Bluetongue			-
5.	Enterotoxaemia			(sale
6.	Fascioliasis			
7.	Foot and mouth disease	net -		
8.	Haemorrhagic septicaemia		-1	
9.	Peste des Petits Ruminants	- 11.		
10.	Sheep & Goat pox			
11.	Swine fever			
12.	Theileriosis			
13.	Trypanosomosis			

**Details may be written here.

2. What are the preventive measures taken in case of predicted outbreaks?



3. How would you rate your satisfaction with the following aspects of the services you have received or accessed?

Description	Very satisfied	Satisfied	Unsatisfied	Not sure
Quality of services provided	-			
Timeliness of alerts received				
Benefits from forecasting of livestock diseases				
Your awareness of this				THE P
	ovement of	report.		
service Suggestions for further impro	ovement of	report.		

Sign and Signature with Designation

AICRP centre:

Dated:





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