

# LIVESTOCK DISEASE RISK FOREWARNING REPORT

Powered by Artificial Intelligence

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# 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*.

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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 (NADRES<sub>v2</sub>), 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 5<sup>th</sup> century that epidemics were associated with natural phenomenon 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 weather-based 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.

The livestock disease forecasting for December 2021 revealed Jharkhand (180), Uttar Pradesh (79), Karnataka (71), Kerala (48) and West Bengal (44) 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 more in Jharkhand (20) followed by Kerala (13) and Karnataka (12) whereas predicted PPR outbreaks are more in Jharkhand (20). Further the co-occurrence of FMD and HS can be expected in Andhra Pradesh, Gujarat, Jharkhand, Karnataka, Kerala, Madhya Pradesh, Odisha, Rajasthan, Tamil Nadu, Tripura and West Bengal. Among the different diseases in livestock, the predicted outbreaks are expected to be high for FMD (96) and Trypanosomosis (76).

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. Further, India has witnessed emergence and re-emergence of various infectious pathogens during the last decade, of which most of the diseases are of zoonotic in nature which urge for the necessity of strengthening of monitoring and surveillance system in the country. 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.

#### Table S1. Summary of State wise Livestock Disease forewarning for December 2021

Sl. No	State Name	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G	SF	Theileriosis	Trypanosomosis	Total number of
				-							Pox			•••	diseases
1	Andaman and Nicobar	0	2	0	0	0	3	1	0	0	0	0	1	0	07
2	Andhra Pradesh	2	0	0	1	0	1	1	3	1	2	0	0	0	11
3	Arunachal Pradesh	0	0	0	0	0	0	1	0	0	0	0	0	0	01
4	Assam	0	0	6	0	2	11	0	1	0	1	12	1	0	34
5	Bihar	0	2	0	0	0	0	6	0	1	0	1	0	3	13
6	Chandigarh	0	0	0	0	0	0	0	0	0	0	0	0	0	00
7	Chhattisgarh	1	0	0	0	0	0	0	0	0	0	0	0	0	01
8	Dadra and Nagar Haveli	0	0	0	0	0	0	0	0	0	0	0	0	0	00
9	Daman and Diu	0	0	0	0	0	0	0	0	0	0	0	0	0	00
10	Goa	0	1	0	0	0	0	0	0	1	0	0	0	0	02
11	Gujarat	0	0	0	0	0	0	2	3	3	2	0	0	1	11
12	Haryana	0	0	0	0	1	0	1	0	2	0	0	0	0	04
13	Himachal Pradesh	0	0	0	0	0	0	1	0	2	1	0	0	0	04
14	Jammu and Kashmir	0	0	0	0	0	0	0	0	0	8	0	0	0	08
15	Jharkhand	3	24	14	0	10	23	20	11	20	2	6	23	24	180
16	Karnataka	7	0	10	9	8	0	12	7	5	10	3	0	0	71
17	Kerala	1	2	0	0	1	0	13	4	7	0	1	10	9	48
18	Lakshadweep	0	0	0	0	0	0	0	0	0	0	0	0	0	00
19	Madhya Pradesh	0	0	1	0	0	0	2	4	0	0	1	0	2	10
20	Maharashtra	0	0	1	0	0	0	0	1	6	1	1	1	0	11
21	Manipur	0	0	3	0	0	2	3	0	0	0	1	0	0	09
22	Meghalaya	0	0	0	0	0	0	5	0	0	0	2	0	0	07
23	Mizoram	0	0	1	0	0	0	0	0	0	0	2	0	0	03
24	Nagaland	0	0	0	0	0	0	0	0	0	0	2	0	0	02
25	NCT of Delhi	0	0	0	0	0	0	0	0	0	0	0	0	0	00
26	Odisha	1	0	4	0	2	0	9	1	2	4	0	0	0	23
27	Puducherry	0	3	0	0	0	2	1	0	0	1	0	0	0	07
28	Punjab	0	0	0	0	0	0	0	0	1	0	1	0	0	02
29	Rajasthan	0	0	2	0	0	1	6	4	3	0	0	0	0	16
30	Sikkim	0	0	0	0	0	0	0	0	0	0	0	1	0	01
31	Tamil Nadu	5	0	0	0	0	0	2	1	3	3	0	0	0	14
32	Telangana	1	0	0	0	1	0	0	0	0	0	0	0	0	02
33	Tripura	0	3	1	0	0	2	2	2	0	3	3	0	0	16
34	Uttar Pradesh	0	17	0	0	1	4	1	0	8	0	0	11	37	79
35	Uttarakhand	0	0	0	0	0	0	0	0	0	0	0	0	0	00
36	West Bengal	2	4	4	0	0	0	7	2	10	4	0	11	0	44
Total nu to repor	mber of districts likely t	23	58	47	10	26	49	96	44	75	42	36	59	76	641

\*Number of predicted disease incidence was summarised considering only High risk and Very high risk

#### 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 v2 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 20<sup>th</sup> Livestock census (2019) fromDepartment of statistics, DAHD, GOI.

Species-wise & Category-wise Livestock Population (in thousands)													
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	livestock		512056	536761	4.8								

#### Meteorological and Remotely sensed data:

The parameters such as air temperature (<sup>0</sup>C), 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 (https://modis.gsfc.nasa.gov/). 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 (https://disc.gsfc.nasa.gov). 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 highriskcategory. 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.



Fig 4.3. 2020 Top Ten diseases

#### 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 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,

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

Where,  $Y_t$  is the differenced time series value,  $\phi$  and  $\theta$  are unknown parameters and  $\epsilon$  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) reduces 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	98.30
3.	Black Quarter	98.30
4.	Blue Tongue	98.77
5.	Enterotoxaemia	99.85
6.	Fasciolosis	99.85
7.	Foot and mouth disease	94.91
8.	Haemorrhagic septicaemia	95.06
9.	Peste des Petits Ruminants	99.38
10.	Sheep & Goat pox	99.23
11.	Swine fever	99.38
12.	Theileriosis	98.30
13.	Trypanosomosis	99.38

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 December 2021

#### i). District wise Livestock Disease forewarning:

		Livestock Diseases													
Districts of Andaman and Nicobar	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis		
Nicobars	NR	VHR	NR	VLR	NR	VHR	NR	NR	NR	NR	NR	NR	NR		
North & Middle Andaman	NR	HR	NR	VLR	NR	VHR	NR	NR	NR	NR	NR	NR	NR		
South Andaman	NR	MR	NR	VLR	NR	VHR	HR	NR	NR	NR	NR	VHR	NR		

District wise Livestock Disease Risk Forewarning for December 2021: Andaman and Nicobar

If vaccination is already been done, please ignore the disease forecast for that disease.

Districts of Andhra		Livestock Diseases													
Pradesh	Anthrax	Babesiosis	BQ	ВТ	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis		
Anantapur	NR	NR	NR	HR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Chittoor	MR	NR	NR	VLR	NR	NR	HR	VLR	NR	NR	NR	NR	NR		
East Godavari	NR	VLR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR		
Guntur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Krishna	NR	NR	NR	VLR	NR	NR	VLR	NR	VHR	VHR	NR	NR	NR		
Kurnool	VHR	NR	NR	VLR	NR	NR	NR	HR	NR	NR	NR	NR	NR		
Prakasam	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR		
Sri Potti Sriramulu Nellore	VHR	NR	NR	VLR	NR	VHR	VLR	VLR	NR	NR	NR	NR	NR		
Srikakulam	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Visakhapatnam	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Vizianagaram	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
West Godavari	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Y.S.R.	NR	NR	NR	VLR	NR	NR	VLR	VHR	NR	VHR	NR	NR	NR		

#### District wise Livestock Disease Risk Forewarning for December 2021: Andhra Pradesh

If vaccination is already has been done, please ignore the disease forecast for that disease.

Districts of Arunachal	Livestock Diseases												
Pradesh	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Anjaw	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Changlang	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Dibang Valley	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
East Kameng	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
East Siang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kurung Kumey	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Lohit	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Lower Dibang Valley	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lower Subansiri	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Papum Pare	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Tawang	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Tirap	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Upper Siang	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
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

#### District wise Livestock Disease Risk Forewarning for December 2021: Arunachal Pradesh

If vaccination has already been done please ignore the disease forecast for that disease.

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

# District wise Livestock Disease Risk Forewarning for December 2021: Assam

#### Continued

		Livestock Diseases												
Districts of Assam	Anthrax	Babesiosis	BQ	ВТ	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis	
Nagaon	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR	NR	
Nalbari	NR	NR	VLR	NR	VHR	NR	NR	NR	NR	NR	VHR	NR	NR	
Sivasagar	NR	NR	NR	NR	NR	VHR	NR	VLR	NR	NR	VHR	NR	NR	
Sonitpur	NR	VLR	VHR	NR	NR	VHR	VLR	HR	NR	NR	MR	NR	NR	
Tinsukia	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	VHR	NR	
Udalguri	NR	NR	VLR	NR	VHR	NR	VLR	NR	NR	NR	NR	NR	NR	

If vaccination has already been done, please ignore the disease forecast for that disease.

District wise Livestock Disease Risk Forewarning for December 2021: Bihar	
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						Liv	estock l	Diseases					
Districts of Bihar	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Araria	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Arwal	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Aurangabad	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Banka	NR	VLR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Begusarai	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bhagalpur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bhojpur	NR	VHR	NR	NR	NR	MR	VLR	VLR	NR	NR	NR	NR	VHR
Buxar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Darbhanga	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Gaya	NR	VLR	NR	NR	NR	NR	VHR	VLR	VHR	NR	NR	NR	NR
Gopalganj	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jamui	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Jehanabad	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Kaimur (Bhabua)	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Katihar	NR	VLR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Khagaria	NR	HR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Kishanganj	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Lakhisarai	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Madhepura	NR	VLR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Madhubani	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Munger	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Muzaffarpur	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR

	Livestock Diseases													
Districts of Bihar	Anthrax	Babesiosis	BQ	ВТ	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis	
Nalanda	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Nawada	NR	NR	NR	NR	NR	NR	VLR	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	HR	NR	NR	NR	VHR	NR	NR	
Purba Champaran	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR	
Purnia	NR	VLR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR	
Rohtas	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Saharsa	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Samastipur	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	
Saran	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Sheikhpura	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Sheohar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Sitamarhi	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Siwan	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	
Supaul	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	
Vaishali	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR	

#### Continued

If vaccination has already been done please ignore the disease forecast for that disease.

# District wise Livestock Disease Risk Forewarning for December 2021: Chandigarh

Districts of Chandigarh						Liv	estock l	Diseases					
	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Chandigarh	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

If vaccination hasalready been done please ignore the disease forecast for that disease.

						Liv	vestock l	Diseases					
Districts of Chhattisgarh	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Bastar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Bijapur	NR	NR	NR	NR	NR	NR	VLR	VLR	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	VLR	VLR	NR	NR	NR	NR	NR
Dhamtari	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Durg	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Janjgir-champa	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Jashpur	NR	NR	VLR	NR	NR	NR	VLR	VLR	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	VHR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Mahasamund	NR	VLR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Narayanpur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Raigarhh	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Raipur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Rajnandgaon	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Surguja	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Uttar Bastar Kanker	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

## District wise Livestock Disease Risk Forewarning for December 2021: Chhattisgarh

If vaccination hasalready been done, please ignore the disease forecast for that disease.

#### **Livestock Diseases Districts of Dadra and** Nagar Haveli S&G PPR SF Anthrax **Babesiosis** BQ BT ET Fasciolosis FMD HS Theileriosis Trypanosomosis Pox Dadra and Nagar VLR NR NR NR NR NR NR VLR NR NR NR NR NR Haveli

#### District wise Livestock Disease Risk Forewarning for December 2021: Dadra and Nagar Haveli

If vaccination is already been done please ignore the disease forecast for that disease.

Districts of		Livestock Diseases													
Daman and Diu	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis		
Daman	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR		
Diu	NR	VLR	VLR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR		

#### District wise Livestock Disease Risk Forewarning for December 2021: Daman and Diu

If vaccination is already been done please ignore the disease forecast for that disease.

	Livestock Diseases													
Districts of Goa	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis	
North Goa	NR	VHR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	
South Goa	NR	VLR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	

# District wise Livestock Disease Risk Forewarning for December 2021: Goa

If vaccination is already been done please ignore the disease forecast for that disease.

#### District wise Livestock Disease Risk Forewarning for December 2021: Gujarat

						Liv	estock D	iseases					
Districts of Gujarat	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Ahmadabad	NR	NR	NR	VLR	NR	NR	VLR	MR	VHR	NR	NR	NR	NR
Amreli	NR	NR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Anand	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Banas Kantha	NR	NR	NR	VLR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Bharuch	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Bhavnagar	NR	NR	NR	NR	NR	NR	VLR	VHR	NR	VHR	NR	NR	NR
Dohad	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Gandhinagar	NR	NR	NR	NR	NR	NR	HR	MR	NR	NR	NR	NR	NR
Jamnagar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Junagadh	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Kachchh	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Kheda	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Mahesana	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Narmada	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Navsari	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Panch Mahals	NR	NR	NR	NR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Patan	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	VHR
Porbandar	NR	NR	NR	NR	NR	NR	VLR	NR	VHR	NR	NR	NR	NR
Rajkot	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	VHR	NR	NR	NR
Sabar Kantha	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Surat	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Surendranagar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Тарі	NR	NR	NR	NR	NR	NR	MR	MR	NR	NR	NR	NR	NR
The Dangs	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Vadodara	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Valsad	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

If vaccination has already been done, please ignore the disease forecast for that disease.

		Livestock Diseases													
Districts of Haryana	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis		
Ambala	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Bhiwani	NR	NR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR		
Faridabad	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR		
Fatehabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Gurgaon	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR		
Hisar	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR		
Jhajjar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Jind	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Kaithal	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR		
Karnal	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Kurukshetra	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Mahendragarh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR		
Mewat	NR	NR	NR	NR	NR	NR	VLR	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	NR	NR	NR	NR	NR		
Panipat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Rewari	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR		
Rohtak	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Sirsa	NR	NR	NR	NR	NR	NR	NR	NR	NR	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		

# District wise Livestock Disease Risk Forewarning for December 2021: Haryana

If vaccination hasalready been done please ignore the disease forecast for that disease.

Districts of		Livestock Diseases													
Himachal Pradesh	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis		
Bilaspur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR		
Chamba	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Hamirpur	NR	NR	NR	NR	NR	NR	VLR	NR	HR	NR	NR	NR	NR		
Kangra	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR		
Kinnaur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Kullu	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Lahul & Spiti	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR		
Mandi	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	VHR	NR	NR	NR		
Shimla	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Sirmaur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR		
Solan	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR		
Una	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR		

#### District wise Livestock Disease Risk Forewarning for December 2021: Himachal Pradesh

If vaccination has already been done, please ignore the disease forecast for that disease.

#### **Livestock Diseases Districts of Jammu** and Kashmir S&G BO BT FMD HS **PPR** SF Theileriosis Anthrax **Trypanosomosis Babesiosis** ET **Fasciolosis** Pox NR NR Anantnag NR NR NR NR NR NR NR NR VHR NR Badgam NR NR NR NR VHR Bandipore NR NR NR NR NR NR NR NR NR VHR NR NR NR NR Baramula NR NR NR NR NR NR NR NR NR MR NR NR NR NR NR NR NR NR Doda NR NR NR NR NR NR NR NR Ganderbal NR NR NR NR NR NR NR NR NR VHR NR Jammu NR NR NR Kargil NR Kathua NR Kishtwar NR VHR NR NR NR Kulgam NR NR VLR NR NR NR NR NR NR NR NR VHR NR Kupwara Leh(Ladakh) NR NR NR VLR NR NR NR VLR NR NR NR NR NR VLR Pulwama NR NR NR NR NR NR NR NR VHR NR NR NR NR NR VLR NR NR NR NR NR NR NR NR Punch NR NR Rajouri NR NR NR VLR NR NR NR NR NR NR NR NR NR Ramban NR Reasi NR Samba NR Shupiyan NR VLR NR NR NR NR NR NR NR NR VLR NR VHR NR Srinagar NR Udhampur NR

#### District wise Livestock Disease Risk Forewarning for December 2021: Jammu and Kashmir

If vaccination hasalready been done, please ignore the disease forecast for that disease.

#### District wise Livestock Disease Risk Forewarning for December 2021: Jharkhand

Districts of	Livestock Diseases													
Jharkhand	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis	
Bokaro	NR	VHR	HR	NR	NR	VHR	VHR	NR	VHR	NR	VHR	VHR	VHR	
Chatra	NR	VHR	VLR	NR	NR	MR	VHR	MR	VHR	NR	NR	VHR	VHR	
Deoghar	NR	VHR	HR	NR	VHR	VHR	VHR	VHR	VHR	NR	NR	VHR	VHR	
Dhanbad	NR	VHR	VHR	NR	VHR	VHR	VHR	HR	VHR	NR	VHR	VHR	VHR	
Dumka	NR	VHR	NR	NR	NR	VHR	VHR	VHR	VHR	NR	VHR	VHR	VHR	
Garhwa	NR	VHR	NR	NR	NR	VHR	VLR	VLR	NR	NR	NR	VHR	VHR	
Giridih	NR	VHR	VHR	NR	VHR	VHR	VHR	HR	NR	NR	NR	VHR	VHR	
Godda	NR	HR	NR	NR	NR	VHR	HR	MR	VHR	NR	NR	VHR	VHR	
Gumla	NR	VHR	HR	NR	VHR	VHR	HR	MR	VHR	NR	VHR	VHR	VHR	
Hazaribagh	NR	VHR	VHR	VLR	NR	VHR	VHR	VHR	VHR	NR	NR	VHR	VHR	
Jamtara	NR	VHR	HR	NR	NR	VHR	VHR	VHR	VHR	NR	NR	VHR	VHR	
Khunti	VHR	VHR	VLR	NR	VHR	VHR	VHR	VHR	VHR	VHR	NR	VHR	VHR	
Koderma	NR	HR	VHR	NR	VHR	VHR	VLR	VLR	VHR	NR	NR	MR	VHR	
Latehar	NR	VHR	VHR	NR	NR	VHR	HR	VLR	VHR	NR	NR	VHR	VHR	
Lohardaga	NR	VHR	VHR	NR	NR	VHR	VHR	VLR	VHR	NR	VHR	VHR	VHR	
Pakur	NR	VHR	MR	NR	NR	VHR	VHR	VHR	VHR	NR	MR	VHR	VHR	
Palamu	NR	VHR	NR	VLR	VHR	VHR	VHR	VHR	NR	NR	NR	VHR	VHR	
ashchimiSinghbhum	VHR	VHR	VHR	NR	VHR	VHR	VHR	MR	VHR	NR	NR	VHR	VHR	
Purbi Singhbhum	VHR	VHR	VHR	NR	NR	VHR	VHR	VHR	VHR	NR	NR	VHR	VHR	
Ramgarh	NR	VHR	NR	VLR	NR	VHR	NR	NR	NR	NR	NR	VHR	VHR	
Ranchi	NR	VHR	VHR	VLR	NR	VHR	VHR	VLR	VHR	NR	VHR	VHR	VHR	
Sahibganj	NR	VHR	VHR	VLR	VHR	VHR	VHR	VHR	VHR	VHR	NR	VHR	VHR	
Seraikela - Kharsawan	NR	VHR	VLR	VLR	VHR	VHR	NR	VLR	VHR	NR	NR	HR	VHR	
Simdega	NR	VHR	VLR	NR	NR	VHR	VHR	NR	VHR	NR	NR	VHR	VHR	

If vaccination hasalready been done please ignore the disease forecast for that disease.

						Liv	vestock D	iseases					
Districts of Karnataka	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Bagalkot	NR	NR	VLR	VLR	NR	NR	VLR	LR	NR	NR	NR	NR	NR
Bangalore	NR	VLR	HR	VLR	NR	NR	VHR	VLR	NR	VHR	VHR	NR	NR
Bangalore Rural	VHR	NR	VHR	MR	NR	NR	VHR	VLR	NR	NR	VHR	NR	NR
Belgaum	NR	NR	NR	VLR	NR	NR	LR	LR	NR	NR	NR	NR	NR
Bellary	VHR	NR	VHR	HR	NR	NR	MR	VLR	NR	MR	NR	NR	NR
Bidar	NR	NR	VLR	VLR	VHR	NR	MR	VLR	VHR	MR	VHR	NR	NR
Bijapur	NR	NR	NR	VLR	NR	NR	MR	VLR	NR	NR	NR	NR	NR
Chamarajanagar	VHR	NR	VHR	VLR	NR	NR	HR	HR	NR	NR	NR	NR	NR
Chikkaballapura	MR	NR	NR	MR	VHR	NR	HR	VLR	VHR	MR	NR	NR	NR
Chikmagalur	NR	NR	HR	VLR	NR	NR	HR	HR	NR	VHR	NR	NR	NR
Chitradurga	NR	NR	VLR	HR	NR	NR	VLR	MR	NR	HR	NR	NR	NR
Dakshina Kannada	NR	NR	NR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
Davanagere	HR	NR	VLR	HR	NR	NR	LR	VLR	NR	VHR	NR	NR	NR
Dharwad	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Gadag	NR	NR	VLR	LR	VHR	NR	VLR	MR	NR	HR	NR	NR	NR

# District wise Livestock Disease Risk Forewarning for December 2021: Karnataka
#### Continued

<b>D1</b>						Liv	vestock D	iseases					
Districts of Karnataka	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Gulbarga	NR	NR	VHR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Hassan	NR	NR	VHR	VLR	NR	NR	HR	HR	VHR	NR	NR	NR	NR
Haveri	NR	NR	NR	HR	VHR	NR	VLR	VLR	NR	VHR	NR	NR	NR
Kodagu	NR	VLR	NR	NR	NR	NR	HR	MR	NR	NR	NR	NR	NR
Kolar	NR	NR	NR	HR	VHR	NR	HR	LR	VHR	MR	NR	NR	NR
Koppal	VHR	NR	VLR	HR	VHR	NR	LR	HR	NR	VHR	NR	NR	NR
Mandya	VHR	NR	VLR	MR	NR	NR	HR	HR	NR	NR	NR	NR	NR
Mysore	NR	NR	HR	HR	NR	NR	VLR	MR	NR	NR	NR	NR	NR
Raichur	NR	NR	NR	VLR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Ramanagara	HR	NR	NR	VHR	NR	NR	VHR	MR	NR	MR	NR	NR	NR
Shimoga	NR	NR	VHR	VLR	VHR	NR	LR	VLR	NR	VHR	NR	NR	NR
Tumkur	MR	NR	VHR	HR	VHR	NR	VHR	MR	VHR	VHR	NR	NR	NR
Udupi	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Uttara Kannada	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Yadgir	NR	NR	NR	VLR	NR	NR	VLR	HR	NR	VHR	NR	NR	NR

If vaccination hasalready been done please ignore the disease forecast for that disease.

						Liv	vestock Di	iseases					
Districts of Kerala	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Alappuzha	NR	NR	NR	NR	NR	NR	VHR	NR	VHR	NR	NR	NR	VHR
Ernakulum	NR	VHR	NR	NR	NR	NR	VHR	NR	VHR	NR	NR	VHR	VHR
Idukki	NR	NR	NR	NR	NR	NR	VHR	NR	VHR	NR	VHR	NR	NR
Kannur	NR	VLR	NR	NR	NR	NR	HR	NR	VHR	NR	NR	VHR	VHR
Kasaragod	NR	VLR	NR	NR	NR	NR	MR	NR	NR	NR	NR	MR	NR
Kollam	NR	VLR	NR	NR	NR	NR	VHR	HR	NR	NR	NR	VHR	NR
Kottayam	NR	NR	NR	NR	NR	NR	VHR	HR	NR	NR	NR	VHR	NR
Kozhikode	VHR	VLR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	VHR	VHR
Malappuram	NR	VLR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	VHR	VHR
Palakkad	NR	NR	NR	NR	NR	NR	VHR	VLR	VHR	NR	NR	VHR	VHR
Pathanamthitta	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Thiruvananthapuram	NR	VLR	NR	VLR	VHR	NR	VHR	VLR	VHR	NR	NR	VHR	VHR
Thrissur	NR	VHR	NR	NR	NR	NR	VHR	HR	VHR	NR	NR	HR	VHR
Wayanad	NR	VLR	NR	NR	NR	NR	VHR	VHR	NR	NR	NR	VHR	VHR

## District wise Livestock Disease Risk Forewarning for December 2021: Kerala

If vaccination hasalready been done please ignore the disease forecast for that disease.

### District wise Livestock Disease Risk Forewarning for December 2021: Lakshadweep

Districts of						Liv	vestock Di	iseases					
Lakshadweep	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Lakshadweep	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

If vaccination hasalready been done, please ignore the disease forecast for that disease.

Districts of Madhya						L	ivestock	Diseases	5				
Pradesh	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Alirajpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Anuppur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ashoknagar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Balaghat	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Barwani	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Betul	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Bhind	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bhopal	NR	MR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR
Burhanpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Chhatarpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Chhindwara	NR	NR	NR	NR	NR	NR	NR	MR	NR	NR	NR	NR	NR
Damoh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Datia	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Dewas	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Dhar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Dindori	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
East Nimar	NR	NR	VHR	NR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Guna	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Gwalior	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Harda	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Hoshangabad	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Indore	NR	NR	NR	NR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Jabalpur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR
Jhabua	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Katni	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

# District wise Livestock Disease Risk Forewarning for December 2021: Madhya Pradesh

Continued													
Districts of Madhya							Livestock	<b>Diseases</b>					
Pradesh	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Khargone(West Nimar)	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Mandla	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mandsaur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Morena	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Narsimhapur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Neemuch	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	HR
Panna	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Raisen	NR	NR	NR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
Rajgarh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Ratlam	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Rewa	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Sagar	NR	NR	NR	NR	NR	NR	MR	VHR	NR	NR	NR	NR	NR
Satna	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Sehore	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Seoni	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Shahdol	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Shajapur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Sheopur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Shivpuri	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Sidhi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Singrauli	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Tikamgarh	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Ujjain	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Umaria	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Vidisha	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

If vaccination hasalready been done, please ignore the disease forecast for that disease.

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

#### Continued

							Livestoc	k Diseases					
Districts of Maharashtra	Anthrax	Babesiosis	BQ	вт	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Parbhani	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Pune	NR	NR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Raigarh	NR	VLR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ratnagiri	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sangli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Satara	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Sindhudurg	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Solapur	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Thane	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Wardha	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Washim	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Yavatmal	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

If vaccination has already been done please ignore the disease forecast for that disease.

						L	ivestock ]	Diseases					
Districts of Manipur	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theilerio sis	Trypanosomosis
Bishnupur	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Chandel	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Churachandpur	NR	NR	HR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Imphal East	NR	NR	HR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Imphal West	NR	NR	VLR	NR	NR	VHR	HR	VLR	NR	NR	VHR	NR	NR
Senapati	NR	NR	VLR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Tamenglong	NR	NR	HR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Thoubal	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	NR
Ukhrul	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Manipur

If vaccination hasalready been done please ignore the disease forecast for that disease.

						L	ivestock ]	Diseases					
Districts of Meghalaya	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theilerio sis	Trypanosomosis
East Garo Hills	NR	VLR	VLR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
East Jaintia Hills	NR	VLR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
East Khasi Hills	NR	VLR	VLR	VLR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Jaintia Hills	NR	VLR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
North Garo Hills	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ribhoi	NR	NR	NR	NR	NR	NR	MR	NR	NR	NR	NR	NR	NR
South Garo Hills	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Southwest Garo Hills	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	VHR	NR	NR
Southwest Khasi Hills	NR	NR	VLR	NR	NR	NR	MR	VLR	NR	NR	NR	NR	NR
West Garo Hills	NR	VLR	VLR	NR	NR	NR	HR	VLR	NR	NR	VHR	NR	NR
West Khasi Hills	NR	NR	VLR	NR	NR	NR	VHR	NR	NR	NR	MR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Meghalaya

If vaccination hasalready been done please ignore the disease forecast for that disease.

						L	ivestock 2	Diseases					
Districts of Mizoram	Anthrax	Babesiosis	BQ	ВТ	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theilerio sis	Trypanosomosis
Aizawl	NR	VLR	VHR	NR	NR	NR	VLR	NR	NR	NR	VHR	NR	NR
Champhai	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Kolasib	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Lawngtlai	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Lunglei	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mamit	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	VHR	NR	NR
Saiha	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Serchhip	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Mizoram

If vaccination hasalready been done please ignore the disease forecast for that disease.

Districts of						L	ivestock	Diseases					
Nagaland	Anthrax	Babesiosis	BQ	ВТ	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theilerio sis	Trypanosomosis
Dimapur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Kiphire	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Kohima	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Longleng	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mokokchung	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Mon	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR
Peren	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Phek	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	VHR	NR	NR
Tuensang	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Wokha	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Zunheboto	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Nagaland

If vaccination hasalready been done, please ignore the disease forecast for that disease.

Districts of NCT of						L	ivestock ]	Diseases					
Delhi	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theilerio sis	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	NR	NR	NR	NR	NR	NR	NR
North	NR	NR	NR	NR	NR	NR	NR	NR	NR	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	NR	NR	NR	NR	NR
South West	NR	VLR	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

### District wise Livestock Disease Risk Forewarning for December 2021: NCT of Delhi

If vaccination has already been done please ignore the disease forecast for that disease.

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

## District wise Livestock Disease Risk Forewarning for December 2021: Odisha

#### Continued

						]	Livestock I	Diseases					
Districts of Odisha	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Mayurbhanj	VHR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Nabarangapur	NR	NR	NR	VLR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Nayagarh	NR	VLR	HR	VLR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Nuapada	NR	NR	HR	NR	VHR	NR	HR	VLR	NR	NR	NR	NR	NR
Puri	NR	VLR	NR	NR	NR	NR	NR	MR	NR	NR	NR	NR	NR
Rayagada	NR	VLR	VLR	VLR	NR	NR	HR	HR	NR	NR	NR	NR	NR
Sambalpur	NR	NR	VLR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Subarnapur	NR	VLR	NR	VLR	NR	NR	NR	VLR	NR	HR	NR	NR	NR
Sundargarh	NR	VLR	NR	VLR	NR	NR	VHR	NR	NR	VHR	NR	NR	NR

If vaccination hasalready been done, please ignore the disease forecast for that disease.

Districts of						I	Livestock I	Diseases					
Puducherry	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Karaikal	NR	VHR	NR	VLR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Mahe	NR	HR	VLR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Puducherry	NR	VHR	NR	VLR	NR	VHR	NR	NR	NR	VHR	NR	NR	NR
Yanam	NR	VLR	VLR	VLR	NR	VHR	NR	VLR	NR	NR	NR	NR	NR

District wise Livestock Disease Risk Forewarning for December 2021: Puducherry

If vaccination has lready been done please ignore the disease forecast for that disease.

Districts of						]	Livestock 1	Diseases			Ŭ		
Punjab	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	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	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Bathinda	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Faridkot	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Fatehgarh Sahib	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Firozpur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	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	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Ludhiana	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Mansa	NR	VLR	NR	NR	NR	NR	NR	VLR	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	VLR	NR	NR	NR	NR	NR	NR
Patiala	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	VHR	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	VLR	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	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Punjab

If vaccination hasalready been done, please ignore the disease forecast for that disease.

# District wise Livestock Disease Risk Forewarning for December 2021: Rajasthan

Districts of							Livestock	Diseases					
Rajasthan	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Ajmer	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Alwar	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Banswara	NR	NR	NR	VLR	NR	VHR	VLR	NR	NR	NR	NR	NR	NR
Baran	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Barmer	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bharatpur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Bhilwara	NR	NR	NR	VLR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Bikaner	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Bundi	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Chittaurgarh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Churu	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Dausa	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Dhaulpur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Dungarpur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Ganganagar	NR	NR	VHR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
Hanumangarh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Jaipur	NR	NR	NR	NR	NR	NR	VHR	VHR	NR	NR	NR	NR	NR
Jaisalmer	NR	NR	NR	VLR	NR	NR	NR	NR	VHR	NR	NR	NR	NR
Jalor	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Jhalawar	NR	NR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Jhunjhunun	NR	NR	NR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
Jodhpur	NR	NR	i	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Karauli	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR

#### Continued

Districts of							Livestock	Diseases					
Rajasthan	Anthrax	Babesiosis	BQ	вт	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Kota	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Nagaur	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Pali	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Pratapgarh	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Rajsamand	NR	NR	NR	NR	NR	NR	VLR	HR	NR	NR	NR	NR	NR
Sawai Madhopur	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Sikar	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
Sirohi	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Tonk	NR	NR	NR	NR	NR	NR	HR	HR	HR	NR	NR	NR	NR
Udaipur	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR

If vaccination hasalready been done please ignore the disease forecast for that disease.

							Livestock	Diseases	5				
Districts of Sikkim	Anthrax	Babesiosis	BQ	вт	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
East District	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR
North District	NR	NR	NR	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

### District wise Livestock Disease Risk Forewarning for December 2021: Sikkim

If vaccination has already been done please ignore the disease forecast for that disease.

	DIS			CK DIS	case r	ISK FOLEW	armig		centre	1 2021.	1 ann	1 1 1 4 4 4	
Districts of						]	Livestock	Disease					
Tamil Nadu	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Ariyalur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Chennai	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Coimbatore	VHR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Cuddalore	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Dharmapuri	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Dindigul	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Erode	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Kancheepuram	NR	NR	MR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Kanniyakumari	NR	VLR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Karur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Krishnagiri	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Madurai	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Nagapattinam	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Namakkal	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Perambalur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Pudukkottai	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Ramanathapuram	VHR	NR	NR	VLR	NR	NR	VLR	NR	NR	VHR	NR	NR	NR
Salem	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	VHR	NR	NR	NR
Sivaganga	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	VHR	NR	NR	NR
Thanjavur	NR	NR	VLR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
The Nilgiris	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Theni	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Thiruvallur	NR	NR	NR	NR	NR	NR	VLR	VHR	VHR	NR	NR	NR	NR
Thiruvarur	NR	NR	NR	VLR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Thoothukkudi	NR	NR	NR	LR	NR	NR	NR	VLR	VHR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Tamil Nadu

### Continued

Districts of Tamil						]	Livestock	Disease	•				
Nadu	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Tiruchirappalli	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Tirunelveli	NR	VLR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Tiruppur	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Tiruvannamalai	VHR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Vellore	VHR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Viluppuram	VHR	NR	NR	NR	NR	NR	VHR	VLR	VHR	NR	NR	NR	NR
Virudhunagar	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR

If vaccination has lready been done please ignore the disease forecast for that disease.

						Ι	Livestock	Diseases	5				
Districts of Telangana	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Adilabad	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Hyderabad	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Karimnagar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Khammam	NR	NR	NR	NR	VHR	NR	VLR	VLR	NR	NR	NR	NR	NR
Mahbubnagar	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Medak	NR	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Nalgonda	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Nizamabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rangareddy	NR	NR	VLR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Warangal	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

## District wise Livestock Disease Risk Forewarning for December 2021: Telangana

If vaccination hasalready been done, please ignore the disease forecast for that disease.

Distaniate of						Ι	Livestock	x Disease	)				
Districts of Tripura	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Dhalai	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	VHR	NR	NR	NR
North Tripura	NR	HR	VHR	NR	NR	NR	VLR	VLR	NR	VHR	VHR	NR	NR
South Tripura	NR	VHR	NR	NR	NR	VHR	HR	VHR	NR	VHR	VHR	NR	NR
West Tripura	NR	VHR	VLR	NR	NR	VHR	VHR	VHR	NR	NR	VHR	NR	NR

### District wise Livestock Disease forewarning for December-2021: Tripura

If vaccination hasalready been done please ignore the disease forecast for that disease.

Districts of Uttar						Li	vestock	Disease					
Pradesh	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Agra	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Aligarh	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Allahabad	NR	HR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR	VHR
Ambedkar Nagar	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Amethi	NR	VLR	NR	NR	NR	NR	VLR	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	VLR	NR	NR	NR	NR	NR	NR
Baghpat	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR
Bahraich	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Ballia	NR	VHR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	VHR	VHR
Balrampur	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Banda	NR	VHR	NR	NR	HR	NR	VLR	VLR	NR	NR	NR	VHR	VHR
Bara Banki	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Bareilly	NR	NR	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	VHR
Basti	NR	NR	NR	VLR	NR	NR	NR	NR	VHR	NR	NR	NR	NR
Bijnor	NR	MR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Budaun	NR	HR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Bulandshahr	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Chandauli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Chitrakoot	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Deoria	NR	NR	NR	NR	NR	NR	VLR	MR	NR	NR	NR	NR	NR
Etah	NR	HR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	HR	VHR
Etawah	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Faizabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	MR
Farrukhabad	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR

## District wise Livestock Disease Risk Forewarning for December 2021: Uttar Pradesh

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

#### Continued

Districts of Uttar						I	Livestock	Disease					
Pradesh	Anthrax	Babesiosis	BQ	BT	ЕТ	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Mahrajganj	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	MR	VHR
Mainpuri	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Mathura	NR	VHR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR
Mau	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Meerut	NR	HR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR	VHR
Mirzapur	NR	HR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	VHR	VHR
Moradabad	NR	MR	NR	NR	NR	NR	VLR	NR	VHR	NR	NR	MR	VHR
Muzaffarnagar	NR	NR	NR	NR	NR	NR	VHR	NR	NR	NR	NR	NR	NR
Pilibhit	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Pratapgarh	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Rae Bareli	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	VHR
Rampur	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Saharanpur	NR	MR	NR	NR	NR	NR	NR	NR	NR	NR	NR	HR	MR
Sambhal	NR	NR	NR	NR	NR	NR	VLR	NR	NR	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	VHR	NR	NR	NR	NR	NR	NR	VHR
Shahjahanpur	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Shamli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Shrawasti	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Siddharthnagar	NR	VLR	NR	VLR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Sitapur	NR	VLR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	VHR	VHR
Sonbhadra	NR	HR	NR	NR	NR	NR	VLR	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	NR	NR	NR	NR	NR	HR
Varanasi	NR	HR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	VHR

If vaccination is already been done please ignore the disease forecast for that disease.

Districts of						]	Livestock	Disease					
Uttarakhand	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Almora	NR	NR	NR	VLR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Bageshwar	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Chamoli	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Champawat	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Dehradun	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Garhwal	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Hardwar	NR	NR	NR	NR	NR	NR	NR	VLR	NR	NR	NR	NR	NR
Nainital	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Pithoragarh	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Rudraprayag	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Tehri Garhwal	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Udham Singh Nagar	NR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Uttarkashi	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: Uttarakhand

If vaccination hasalready been done, please ignore the disease forecast for that disease.

Districts of West Bengal						I	Livestock	Disease					
Deligai	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomosis
Bankura	VHR	VHR	NR	NR	NR	NR	VHR	VLR	VHR	NR	NR	VHR	NR
Bardhhaman	NR	VLR	VHR	NR	NR	NR	VLR	VLR	VHR	VHR	NR	VHR	NR
Birbhum	NR	VHR	NR	VLR	NR	NR	VHR	VLR	VHR	NR	NR	VHR	NR
Dakshin Dinajpur	NR	NR	VHR	NR	NR	NR	VLR	VLR	VHR	NR	NR	NR	NR
Darjiling	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Haora	NR	VHR	VHR	NR	NR	NR	VLR	VLR	VHR	NR	NR	VHR	NR
Hugli	NR	VHR	NR	NR	NR	NR	VLR	VHR	VHR	NR	NR	VHR	NR
Jalpaiguri	NR	NR	VLR	VLR	NR	NR	VHR	VLR	VHR	NR	NR	NR	NR
Koch Bihar	NR	NR	VLR	NR	NR	NR	VLR	NR	NR	NR	NR	NR	NR
Kolkata	NR	VLR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	NR	NR
Maldah	NR	NR	NR	NR	NR	NR	VLR	VLR	VHR	NR	NR	VHR	NR
Murshidabad	VHR	NR	NR	NR	NR	NR	VLR	VLR	NR	NR	NR	VHR	NR
Nadia	NR	VLR	NR	VLR	NR	NR	VHR	VLR	NR	VHR	NR	VHR	NR
North Twenty-Four Parganas	NR	NR	NR	NR	NR	NR	VLR	NR	NR	VHR	NR	HR	NR
Paschim Medinipur	NR	NR	VLR	VLR	NR	NR	VHR	VLR	HR	NR	NR	MR	NR
Purba Medinipur	NR	NR	VHR	VLR	NR	NR	VHR	VLR	NR	VHR	NR	VHR	NR
Puruliya	NR	NR	NR	NR	NR	NR	VLR	HR	VHR	NR	NR	NR	NR
South Twenty Four Parganas	NR	NR	VLR	VLR	NR	NR	VLR	VLR	NR	NR	NR	VHR	NR
Uttar Dinajpur	NR	NR	NR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR

### District wise Livestock Disease Risk Forewarning for December 2021: West Bengal

If vaccination has lready been done please ignore the disease forecast for that disease.

**II)** Glimpse about the risk of predicted diseases:

The Livestock disease risk obtained based on the Machine Learningalgorithm 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 and North & Middle Andaman	Two	Babesiosis
2.	Nicobars, North & Middle Andaman and South Andaman	Three	Fasciolosis
AR 3.	South Andaman	One	Foot and Mouth Disease
4.	South Andaman	One	Theileriosis

# 2. Andhra Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
E.	Kurnool and Sri Potti Sriramulu Nellore	One	Anthrax
2.	Anantapur	One	Bluetongue
3.	Sri Potti Sriramulu Nellore	Two	Fasciolosis
4.	Chittoor	Three	Foot and Mouth Disease
5.	East Godavari, Kurnool and Y.S.R.	Three	Haemorrhagic Septicaemia
6.	Krishna	One	Peste des Petits Ruminants
7.	Krishna and Y.S.R.	One	Sheep & Goat pox

#### 3. Arunachal Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Papum Pare	One	Foot and Mouth Disease

#### 4. Assam

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Chirang, Dhemaji, Dhubri, Dibrugarh, Golaghat and Sonitpur	Six	Black Quarter
2.	Nalbari and Udalguri	Two	Enterotoxaemia
3.	Chirang, Dhemaji, Dhubri, Dibrugarh, Jorhat, Kamrup, Kamrup Metropolitan, Lakhimpur, Nagaon, Sivasagar and Sonitpur	Eleven	Fasciolosis
4.	Sonitpur	One	Haemorrhagic Septicaemia
5.	Darrang	One	Sheep & Goat pox
6.	Baksa, Dhemaji, Dibrugarh, Dima Hasao, Golaghat, Jorhat, Kamrup, Kamrup Metropolitan, Karbi Anglong, Lakhimpur, Nalbari and Sivasagar	Twelve	Swine Fever
7.	Tinsukia	One	Theileriosis

### 5. Bihar

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Bhojpur and Khagaria	Two	Babesiosis
2.	Gaya, Muzaffarpur, Patna, Purba Champaran, Samastipur and Vaishali	Six	Foot and Mouth Disease
3.	Bhojpur and Patna	One	Peste des Petits Ruminants
4.	Patna	One	Swine Fever
5.	Bhojpur, Buxar and Khagaria	Three	Trypanosomosis

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## 6. Chhattisgarh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Koriya	One	Anthrax

### 7. Goa

SI. N	Disease prone districts	Number of districts prone for disease	Disease Name
	North Goa	One	Babesiosis
ICAR	. South Goa	One	Peste des Petits Ruminants

## 8. Gujarat

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Gandhinagar and Junagadh	Two	Foot and Mouth Disease
2.	Banas Kantha, Bhavnagar and Panch Mahals	Three	Haemorrhagic Septicaemia
3.	Ahmadabad, Amreli and Porbandar	Three	Peste des Petits Ruminants
4.	Bhavnagar and Rajkot	Two	Sheep & Goat pox
5.	Patan	One	Trypanosomosis

### 9. Haryana

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Hisar	One	Enterotoxaemia
2.	Faridabad	One	Foot and Mouth Disease
3.	Bhiwani and Kaithal	Two	Peste des Petits Ruminants

#### **10. Himachal Pradesh**

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Kangra	One	Foot and Mouth Disease
2.	Hamirpur and Solan	Two	Peste des Petits Ruminants
3.	Mandi	One	Sheep & Goat pox

## 11. Jammu & Kashmir

Sl. No	Disease prone districts		Number of districts prone for disease	Disease Name	रापजास्विस NIVED
1.	Anantnag, Badgam, Ganderbal, Kulgam, Pulwama and Srinagar	Bandipore, Kupwara,	Eight	Sheep & Goat po	ох

## 12. Jharkhand

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Khunti, Pashchimi Singhbhum and Purbi Singhbhum	Three	Anthrax
2.	All twenty-four districts	Twenty-four	Babesiosis
3.	Bokaro,Deoghar,Dhanbad,Giridih,Gumla,Hazaribagh,Jamtara,Koderma,Latehar,Lohardaga,Pashchimi Singhbhum,PurbiSinghbhum,Ranchi andSahibganj.	Fourteen	Black Quarter
4.	Deoghar, Dhanbad, Giridih, Gumla, Khunti, Koderma, Palamu, Purbi Singhbhum, Sahibganj and Seraikela - Kharsawan	Ten	Enterotoxaemia
5.	Except one district (Chatra) remaining all twenty-three districts	Twenty-three	Fasciolosis

6.	Except four districts (Garhwa, Koderma, Ramgarh and Seraikela - Kharsawan) remaining all twenty districts	Twenty	Foot and Mouth Disease
7.	Deoghar, Dhanbad, Dumka, Giridih, Hazaribagh, Jamtara, Khunti, Pakur, Palamu, Purbi Singhbhum and Sahibganj	Eleven	Haemorrhagic Septicaemia
8.	Except four districts (Garhwa,	Twenty	Peste des Petits Ruminants
	Giridih, Palamu and Ramgarh) remaining all twenty districts		
1CAR 9.	Khunti and Sahibganj	Two	Sheep & Goat pox
10.	Bokaro, Dhanbad, Dumka, Gumla, Lohardaga and Ranchi	six	Swine Fever
11.	Except one district (Koderma) remaining all twenty-three districts	Twenty-three	Theileriosis
12.	All twenty-four districts	Twenty-three	Trypanosomosis

# 13. Karnataka

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Bangalore Rural, Bellary, Chamarajanagar, Davanagere, Koppal, Mandya and Ramanagara	Seven	Anthrax
2.	Bangalore-urban, Bellary Bangalore Rural, Chamarajanagar, Mysore, Chikmagalur, Gulbarga, Hassan, Shimoga and Tumkur	Ten	Black Quarter
3.	Bellary, Chitradurga, Davanagere, Haveri, Kolar, Koppal, Mysore, Ramanagara and Tumkur	Nine	Bluetongue
4.	Bidar, Chikkaballapur, Gadag, Haveri, Kolar, Koppal, Shimoga and	Eight	Enterotoxaemia

	Tumkur		
5.	Bangalore Ruler, Bangalore Urban, Chamarajanagar, Chikkaballapura, Chikmagalur, Dakshina Kannada, Hassan, Kodagu, Kolar, Mandya, Ramanagara and Tumkur.	Twelve	Foot and Mouth Disease
6.	Chamarajanagar, Chikmagalur, Hassan, Kolar, Mandya, Raichur and yadgir	Seven	Haemorrhagic Septicaemi
7.	Bidar, Chikkaballapur, Hassan, Kolar and Tumkur	Five	Peste des Petits Ruminant
8.	Bangalore Urban, Chikmagalur, Chitradurga, Davanagere, Gadag, Haveri, Koppal, Shimoga, Tumkur and Yadgir	Ten	Sheep & Goat pox साम जा सू वि NINET
9.	Bangalore Urban, Bangalore Rural and Bidar	Three	Swine Fever

### 14. Kerala

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Kozhikode	One	Anthrax
2.	Ernakulam and Thrissur	Two	Babesiosis
3.	Thiruvananthapuram	One	Enterotoxaemia
4.	Alappuzha, Ernakulam, Idukki, Kannur, Kollam, Kottayam, Kozhikode, Malappuram, Palakkad,	Thirteen	Foot and Mouth Disease
ICAR	Pathanamthitta, Thiruvananthapuram, Thrissur and Wayanad		रापजास् विस NIVEDI
5.	Kollam, Kottayam, Thrissur and Wayanad	Four	Haemorrhagic Septicaemia
6.	Alappuzha, Ernakulam, Idukki, Kannur, Palakkad, Thiruvananthapuram and Thrissur	Seven	Peste des Petits Ruminants
7.	Idukki	One	Swine Fever
8.	Ernakulam, Kannur, Kollam, Kottayam, Kozhikode, Malappuram, Palakkad, Thiruvananthapuram, Thrissur and Wayanad	Ten	Theileriosis
	that and a second secon		

# 15. Madhya Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	East Nimar	One	Black Quarter
2.	Raisen and Shivpuri	Two	Foot and Mouth Disease
3.	Bhopal, East Nimar, Indore and Sagar	Four	Haemorrhagic Septicaemia
4.	Jabalpur	One	Swine Fever
5.	Narsimhapur and Neemuch	Two	Trypanosomosis

#### 16. Maharashtra

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Ahmadnagar	One	Black Quarter
2.	Dhule	One	Haemorrhagic Septicaemia
3.	Ahmadnagar, Aurangabad, Gadchiroli, Jalgaon, Nashik and Pune	Six	Peste desPetitsRuminants
4.	Ahmadnagar	One	Sheep & Goat pox
5.	Jalgaon	One	Swine Fever
ICAP6.	Akola	One	Theileriosis

### 17. Manipur

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Churachandpur, Imphal East and Tamenglong	Three	Black Quarter
2.	Imphal West and Thoubal	Two	Fasciolosis
3.	Bishnupur, Imphal West and Senapati,	Three	Foot and Mouth Disease
4.	Imphal West	One	Swine Fever

### 18. Meghalaya

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	East Garo Hills, East Khasi Hills, Jaintia Hills, West Garo Hills and West Khasi Hills	Five	Foot and Mouth Disease
2.	Southwest Garo Hills and West Garo Hills	Two	Swine Fever
## 19. Mizoram

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Aizawl	One	Black Quarter
2.	Aizawl and Mamit	Two	Swine Fever

## 20. Nagaland

X	Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name	
I	CAR <sup>1</sup> .	Mon and Phek	Two	Swine fever	रापजासू विसं NIVEDI

### 21. Odisha

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Mayurbhanj	One	Anthrax
2.	Ganjam, Khordha, Nayagarh and Nuapada	Four	Black Quarter
3.	Jagatsinghapur and Nuapada	Two	Enterotoxaemia
4.	Cuttack, Dhenkanal, Ganjam, Jagatsinghapur, Nabarangapur, Nayagarh, Nuapada, Rayagada and Sundargarh	Nine	Foot and Mouth Disease
5.	Rayagada	One	Haemorrhagic Septicaemia
6.	Baudh and Khordha	Two	Peste desPetitsRuminants
7.	Ganjam, Khordha, Subarnapur and Sundargarh	Four	Sheep & Goat pox

## 22. Puducherry

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Karaikaland, Mahe and Puducherry	Three	Babesiosis
2.	Puducherry and Yanam	Two	Fasciolosis
3.	Karaikal	One	Foot and Mouth Disease
4.	Puducherry	One	Sheep & Goat pox

## 23. Punjab



Disease prone districts	Number of districts prone for disease	Disease Name
Patiala	One	Peste desPetitsRuminants
Patiala	One	Swine Fever
	Patiala	Patiala One One

# 24. Rajasthan

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Ganganagar and Jodhpur	Two	Black Quarter
2.	Banswara	One	Fasciolosis
3.	Alwar, Ganganagar, Jaipur, Jhunjhunun, Sikar and Tonk	Six	Foot and Mouth Disease
4.	Bhilwara, Jaipur, Rajsamand and Tonk	Four	Haemorrhagic Septicaemia
5.	Jaisalmer, Jhalawar and Tonk	Three	Peste desPetitsRuminants

## 25. Sikkim

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	East District	One	Theileriosis

## 26. Tamil Nadu

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Coimbatore, Ramanathapuram, Tiruvannamalai, Vellore and Viluppuram	Five	Anthrax
2.	Kancheepuram and Viluppuram	Two	Foot and Mouth Disease
3.	Thiruvallur	One	Haemorrhagic Septicaemia
4.	Thiruvallur, Thoothukkudi and Viluppuram	Three	Peste desPetitsRuminants
5.	Ramanathapuram, Salem and Sivaganga	Three	Sheep & Goat pox

### 27. Telangana

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Warangal	One	Anthrax
2.	Khammam	One	Enterotoxaemia
. 1			

## 28. Tripura

Sl. No	Disease prone districts	Number of disease prone for districts	Disease Name
-1.	North Tripura, South Tripura and West Tripura	Three	Babesiosis
2.	North Tripura	One	Black Quarter
3.	South Tripura and West Tripura	Two	Fasciolosis
4.	South Tripura and West Tripura	Two	Foot and Mouth Disease
5.	South Tripura and West Tripura	Two	Haemorrhagic Septicaemia
6.	Dhalai, North Tripura and South Tripura	Three	Sheep & Goat pox
7.	North Tripura, South Tripura and West Tripura	Three	Swine Fever

## 29. Uttar Pradesh

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Allahabad, Ballia, Banda, Budaun, Etah, Fatehpur, Gonda, Gorakhpur, Jalaun, Kanpur Nagar, Kheri, Kushinagar, Mathura, Meerut, Mirzapur, Sonbhadra and Varanasi	Seventeen	Babesiosis
2.	Banda	One	Enterotoxaemia
3. CAR	Ballia, Jaunpur, Meerut and Sant Ravidas Nagar (Bhadohi)	Four	Fasciolosis
4.	Shrawasti and Sonbhadra	Two	Haemorrhagic Septicaemia
5.	Muzaffarnagar	One	Foot and Mouth Disease
6.	Bareilly, Basti, Fatehpur, Gonda, Gorakhpur, Hardoi, Kanpur Nagar and Moradabad.	Eight	Peste desPetitsRuminants
7.	Allahabad, Ballia, Banda, Etah, Fatehpur, Gorakhpur, Jalaun, Kushinagar, Mirzapur, Saharanpur and Sitapur	Eleven	Theileriosis
8.	Aligarh, Allahabad, Ambedkar Nagar, Baghpat, Bahraich, Ballia, Banda, Bara Banki, Bareilly, Bijnor, Budaun, Etah, Farrukhabad, Fatehpur, Gonda, Gorakhpur, Hardoi, Jalaun, Jaunpur, Kanpur Nagar, Kheri, Kushinagar, Lalitpur, Mahrajganj, Mainpuri, Mathura, Meerut, Mirzapur, Moradabad, Pratapgarh, Rae Bareli, Sant Ravidas Nagar (Bhadohi), Sitapur, Sonbhadra, Sultanpur, Unnao and Varanasi	Thirty-seven	Trypanosomosis

## **30. West Bengal**

Contraction of the second s

Sl. No	Disease prone districts	Number of districts prone for disease	Disease Name
1.	Bankura and Murshidabad	Two	Anthrax
2.	Bankura, Birbhum, Haora and Hugli	Four	Babesiosis
3.	Barddhaman, Dakshin Dinajpur, Haora and Purba Medinipur	Four	Black Quarter
4.	Bankura, Birbhum, Jalpaiguri, Nadia, Paschim Medinipur, Purba Medinipur and Uttar Dinajpur	Seven	Foot and Mouth Disease
5.	Hugli and Puruliya	Two	Haemorrhagic Septicaemia
6.	Bankura, Bardhhaman, Birbhum, Dakshin Dinajpur, Haora, Hugli, Jalpaiguri, Maldah, Paschim Medinipur and Puruliya	Ten	Peste desPetitsRuminants
7.	Barddhaman, Nadia, North Twenty Four Parganas and Purba Medinipur	Four	Sheep & Goat pox
8.	Bankura, Bardhhaman, Birbhum, Haora, Hugli, Maldah, Murshidabad, Nadia, North Twenty Four Parganas, Purba Medinipur and South Twenty Four Parganas	Eleven	Trypanosomosis

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

Γ	SI	Disease	<b>Species Affected</b>	Clinical Signs	Preventive Measures
	No.				
	1	Anthrax (AX)	Most of the mammals and ruminants are highly susceptible. Pigs and Horses are moderately	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	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
X	ICAR		susceptible. Carnivores are relatively	shoulder over a period of one week before death.	biosecurity measures may be followed. Carcass may be disposed by deep burying
			resistant.		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.

				<b>T</b> T
4.	Bluetongue (BT)	Sheep are more susceptible than goats.	Fever, swelling of face, 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 – 'bluetongue'.	Vector control using 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. Strict biosecurity measures.
5	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.	Fasciolosis (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	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.	Regular vaccination and seromonitoring. Disinfection with sodium carbonate (4%) or 10% washing soda and strict biosecurity measures to be followed and animal
ICAR	B	severe in cattle and pigs.	Concurrent to oral lesions, vesicles also appear in inter digital skin and	followed and animal movement may be controlled.
			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.	
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.	the veterinarians and as

9.	Peste desPetits Ruminants <i>(</i> PPR <i>)</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.	Vaccination of susceptible animals of above 3 months old age. Restriction on animal movement, strict biosecurity measures and proper disposal of carcass.
10. ICA I	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.	The affected animal should be treated with Diaminazine compounds or chloride and sulphate salts of Quinapyramine. Periodical spray of insecticide in and around animal shed to remove the flies.
ICAR	B			रापजास्विस NIVFD

## iv) Risk Prediction - Livestock Disease Forewarning Maps



Risk Prediction of Anthrax for the month of December 2021

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Risk Prediction of Babesiosis for the month of December 2021



### Risk Prediction of Black quarter for the month of December 2021



Risk Prediction of Bluetongue for the month of December 2021

\



#### Risk Prediction of Enterotoxemia for the month of December 2021



#### Risk Prediction of Fascioliasis for the month of December 2021



Risk Prediction of Foot and mouth disease for the month of December 2021



Risk Prediction of Haemorrhagic septicaemia for the month of December 2021



Risk Prediction of Peste des petits ruminants for the month of December 2021



#### Risk Prediction of Sheep and Goat pox for the month of December 2021



Risk Prediction of Swine fever for the month of December 2021



#### Risk Prediction of Theileriosis for the month of December 2021



Risk Prediction of Trypanosomiasis for the month of December 2021



#### V. Forecasting of remote sensing and meteorological parameters till December 2021(*Ex. Tamil Nadu*)



## VI. SIGNIFICANT WEATHER PARAMETERS TABLE

<b>Disease Names</b>	Significant Parameters				
Anthrax	Precipitable Water, Surface Pressure				
Babesiosis	Air Temperature, Precipitation, Surface Pressure Minimum Temperature, Vapour Pressure, Dew Poin Temperature, Wind Speed, NDVI and LST Night				
Black quarter	Precipitable Water, Precipitation, Surface Pressure, Se 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 PETAir Temperature, Precipitation, relative humidity, Temperature, Maximum Temperature, Vapour Pressure, Vwind, Rainfall, Soil Moisture, NDVI and EVIPrecipitable Water, Uwind, Vwind, wet_dry frequency, LST Night and EVI				
Fasciolosis					
Foot and Mouth Disease					
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, NDV and LST Night				
Trypanosomosis	Air Temperature, Precipitation, Surface Pressure, Vapou Pressure, NDVI and LST Night				

A NUMBER OF STREET

### 9. Post prediction Validation

DIMAPUR | Publish Date: 4/14/2019 AH&VS TEAM VISITS AFFECTED AREAS UNDER MEDZIPHEMA, Source: http://www.nagalandpost.com

Following reports of a good number of buffaloes dying in a recent outbreak of suspected Haemorrhagic 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 that affects cattle and water buffaloes with a high mortality rate in infected animals).

AH&VS, deputy director & principal investigator, AICRP-ADMAS, Dr S. Amenla Walling, in a press release reported that the team consisted of the department's director, Dr Temsumeren, along with additional director, Dr. Budhi Lama, and other officials from the department. The press release added that the area is prone to such kind of disease outbreaks and the department officials reminded villagers to cooperate with the department and vaccinate their animals against such outbreaks. The team told the villagers that even an outbreak can be contained more effectively if villagers report the matter on time to the nearest Veterinary Health Centre.

The villagers admitted in the meeting that they had not reported the recent outbreak to the department initially. The director appreciated the CVO Dimapur and his Rapid Response Team for their quick action after receiving information and for remaining stationed in the outbreak area to date. Free medicine was also distributed among the villagers. The department, through the press release also appealed to everyone to report such matters to the nearest Veterinary Health Centre (so that qualified staff may intervene quickly), instead of publicizing it in other ways. It stated that the department is prepared to extend services to any outbreak of diseases in 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. Amenla 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 the circumstances.









FMD POST PREDICTION

District wise Livestock Disease Risk Forewarning for April 2021: Arunachal Pradesh

Published Date: 2021-04-14 14:06:16 Subject: PRO/SOAS> Foot & mouth disease - India: (Arunachal Pradesh) mithun Archive Number: 20210414.8306219

FOOT & MOUTH DISEASE - INDIA: (ARUNACHAL PRADESH) MITHUN

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

Districts of Arunachal													
Pradesh	Anthrax	Babesiosis	BQ	BT	ET	Fascioliasis	FMD	HS	PPR	S&G Pox	SF	Theileriosis	Trypanosomi asis
Anjaw	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
Changlang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
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	NR	NR	NR
East Siang	NR	NR	NR	NR	NR	NR	VHR	NR	VLR	NR	VHR	NR	NR
Kurung Kumey	NR	NR	NR	NR	NR	NR	NR.	NR	NR	NR	NR.	NR	NR
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	NR
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	NR
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	NR
Upper Siang	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR	NR
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
		If vaccination	has alr	eady be	en done	please ignore	the dise	ase fore	cast for	that dis	ease.		

Date: Tue 13 Apr 2021 6:32 AM IST

Source: The Sentinel [edited]

https://www.sentinelassam.com/north-east-india-news/arunachal-news/many-mithuns-infected-with-fmd-foot-and-mouth-disease-533689

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 <promed-SoAs@promedmail.org> Andaman and Nicobar Report June-2020



#### HIMACHAL PRADESH REPORT AUGUST-2020



#### ASSAM REPORTAUGUST-2020



#### KARNATAKA REPORT SEPTEMBER-2021

INTERNATIONAL SOCIETY	and south	Livestock Diseases												
FOR INFECTIOUS DISEASES	Districts of Karnataka	Anthrax	Babesiosis	BQ	BT	ET	Fasciolosis	FMD	HS	PPR	S&G Pox	SF	Theileri osis	Trypanosomosi
Published Date: 2021-09-30 11:40:53	Gulbarga	NR	NR	VLR	VLR	NR	NR	VIP	VLR	NR	VHR	NR	NR	NR
Publisheu Date. 2021-09-30 11.40.33	Hassan	NR	NR	VHR	VLR	NR	NR	VHR	VLR	VLR	VHR	NR	NR	NR
Subject: PRO/SOAS> Foot & mouth disease - India (10): (Karnataka) cattle	Haveri	VHR	NR	VLR	NR	NR	NR	HR	HR	NR	HR	NR	NR	NR
Archive Number: 20210930.8698795	Kodagu	NR	NR	HR	NR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
ICHIVE Multiber. 20210350.0090795	Kolar	NR	NR	NR	VLR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
	Koppal	VHR	NR	VLR	VLR	NR	NR	LR	MR	VLR	NR	NR	NR	NR
FOOT & MOUTH DISEASE - INDIA (10): (KARNATAKA) CATTLE	Mandya	NR	NR	NR	VLR	NR	NR	VHR	VLR	NR	NR	NR	NR	NR
	Mysore	NR	NR	VHR	NR	NR	NR	VHR	VLR	VLR	NR	NR	NR	NR
A ProMED-mail post http://www.promedmail.org	Raichur	VHR	NR	VLR	NR	NR	NR	VLR	LR	VLR	NR	NR	NR	NR
ProMED-mail is a program of the	Ramanagara	NR	NR	NR	VLR	NR	NR	VHR	NR	VLR	NR	NR	NR	NR
International Society for Infectious Diseases	Shimoga	NR	NR	VHR	NR	VHR	NR	VHR	MR	VLR	NR	NR	NR	NR
http://www.isid.org	Tumkur	VHR	NR	HR	VLR	NR	NR	VHR	VHR	VLR	NR	NR	NR	NR
A CONTRACTOR OF	Udupi	NR	NR	NR	NR	NR	NR	HR	NR	NR	NR	NR	NR	NR
Date: Fri 24 Sep 2021 18:33 IST	Uttara Kannada	NR	NR	NR	NR	NR	NR	HR	VLR	NR	NR	NR	NR	NR
Source: The Hindu [edited]	Yadgir	NR	NR	VLR	VLR	NR	NR	VLR	VLR	VLR	NR	NR	NR	NR

#### The outbreak of foot-and-mouth disease (FMD) among cattle in Hassan has left both farmers and veterinarians worried.

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.

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.

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 Arkalgud, told media.

Following reports of the disease, veterinarians have been treating affected animals. "Against 24 sanctioned posts of veterinary doctors in Arkalgud 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.

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.

Considering the cases, the department has launched a ring vaccination programme. Under this, the animals in a 5-kilometre [3.1 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 lakh [658 000] animals," said KR Ramesh, Deputy Director of Veterinary and Animal Husbandry in Hassan.

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

### 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 nonreporting 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.

	May-	2020	June-2020				
Livestock diseases	No of Districts predicted for the disease	No of districts reported the disease	No of Districts predicted for the disease	No of districts reported the disease			
Anthrax	18	1	31	2			
Babesiosis	57	22	40 21				
Black quarter	77	2	71	1			
Bluetongue	3	NA	NA	NA			
Enterotoxaemia	24		25	1			
Fasciolosis	49	15	55	22			
Foot and mouth disease	64	6	46	5			
Haemorrhagic septicaemia	75	6	57	5			
Peste des petits ruminants	67	6	-46				
Sheep & Goat pox	12	3	43	5			
Swine fever	38	7	28	4			
Theileriosis	67	25	37	19			
Trypanosomosis	70	30	32	46			

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

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 🌣 🖀 ⊳	▼⊿ û 10:36 ¢ ● 🖬 🕨 🛛 🕶 🕶	10:36 💠 🌒 📾 🕨 🛛 💎 🖬 🕼
≡ Home	$\equiv$ Livestock Disease Forewarning	$\equiv$ Livestock Disease Forewarning
Select a Langu	Livestock Disease Forewarning	Livestock Disease Forewarning
	<ul> <li>Select disease name</li> </ul>	Select disease name
	Anthrax 👻	Anthrax 👻
भावग्रेमता KCAR	Select State name	
Livestock Disease Forewarning	KARNATAKA 👻	🕕 Result
About App:	Select district name	Disease name : Anthrax State name : KARNATAKA
National Institute of Veterinary Epidemiology Disease Informatics (NIVEDI) has the manda	e to	District name : Bagalkot Month: December Result: Very Low Risk
carry out research activities in the area veterinary epidemiology and disease informa With the eradication of RP successfully, India	tics.	Content of the second s
not only proved its ability to face the challer but also to succeed, despite various limitat Similar efforts are needed to control	December 🗸	December
eradicate the diseases like FMD, PPR, Brucell CSF, BT, HS etc., which cause huge economic annually to the livestock industry. To this	oss SUBMIT	SUBMIT
ICAR-NIVEDI has identified 13 priority disea based on the past incidence patterns and	ses, has RESET	RESET
built a strong database of these diseases. database, which is backbone of the Nat Animal Disease Referral Expert System (NADI is used for providing monthly livestock dis	onal ES),	
	< • 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 andimmediate 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<sub>2</sub> (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 six livestock diseases were sent through SMS to farmers is presented in Table 11.A. Total of 14,07,823 SMS alerts were sent to farmers in November 2021.

Table 11.A: Number of famers received the SMS alert through FRUITS application on November 2021

Disease Name	District Name	No. of Farmers Received SMS		District Name	No. of Farmers Received SMS
	Bangalore Rural	23513		Bangalore	24939
	Bellary	35622		Bangalore Rural	35625
	Chamarajanagar	25369		Bellary	23552
	Chikmagalur	25369		Chamarajanagar	25417
Anthrax	Chitradurga	13984		Chikkaballapura	36740
	Davanagere	23136		Chikmagalur	14824
	Koppal	13952		Chitradurga	14013
	Mysore	107066		Dakshina Kannada	35990
	Tumkur	86360	FMD	Hassan	77507
Babesiosis	Bangalore	24939	FMD	Kodagu	2688
	Bagalkot	19170		Kolar	23105
	Hassan	77389		Mandya	107117
Black Quarter	Koppal	13952		Mysore	107270
	Shimoga	33083		Ramanagara	56371
	Tumkur	86361		Shimoga	33132
	Bagalkot	19178		Tumkur	86445
	Bellary	23552		Udupi	29532
	Chikkaballapura	36740		Yadgir	219
	Chitradurga	14013		Bangalore	24939
Bluetongue	Davanagere	23221	Theileriosis	Bangalore Rural	35625
	Koppal	14013		Ramanagara	56371
	Tumkur	86445		23 SMS alert has lifferent Animal Dise	

# 12. Appendix

## a) R Code

```
#pars month_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)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),]</pre>

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)</pre>
```

}

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")</pre>
```

```
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])
```

```
# {
# nucliation[i]=nucliation_alm
```

```
# 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])\* 100)

```
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	:	R environment for statistical computing
BQ	:	Black Quarter
BT	:	Bluetongue
ET	:	Enterotoxaemia
FMD	:	Foot and Mouth disease
HS	:	Haemorrhagic Septicaemia
PPR	: 1	Peste des Petits Ruminants
SGP		Sheep and Goat pox
SF		Swine Fewer
hPa	Acres 6	Hectopascals
NR		No risk/No data available
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 ofhumans.

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 toemerging 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 upto 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.



a •	Type ofinfection		Clinicalsigns	
Species		(none/low/high)		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 No
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 Macacamulatta)	Experimental	High	Yes	Yes

Contraction of the state of the

Table13.1. Summary of finding sin animals to date

#### 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. Asinhumans, 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 peribronchitis 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 listof 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 tobe 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.



#### References

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- 6. CDC, Corona virus Disease 2019 (COVID-19)–pets & other animals (consulted on 29/05/2020) <u>https://www.cdc.gov/coronavirus/2019-ncov/daily-life-coping/positive-pet.htm</u>



# EPIDEMIOLOGY OF COVID-19

		H	Ep	idemio	logy of C	OVID-19 A	LL INDIA				
Number of Infections (5 laks Increment)	No of days taken to reach since 22 Jan-2020	date reached since 22 Jan- 2020	Cumulative Number of deaths	CFR	avg. daily deaths	R <sub>0</sub> for confirmed cases	Required herd Immunity (Threshold) Ro	Total Vaccine Administered (cum)	% of Immunity by Infection	% of immunity by vaccination	total % of Immunity gained
1st 5 Lakh Cases	156 days	26-06-20	15685	3.06	116.0	1.872	46.58		0.04		0.04
Cum 10 Lakh Cases	20 days	16-07-20	25602	2.49	481.1	1.802	44.51		0.07		0.07
Cum 15 Lakh Cases	12 days	28-07-20	34193	2.23	711.5	1.762	43.25		0.11		0.11
cum 20 Lakh Cases	9 days	06-08-20	41585	2.04	809.4	1.732	42.26		0.14		0.14
cum 25 Lakh Cases	8 days	14-08-20	49036	1.97	932.2	1.732	42.26		0.18		0.18
cum 30 Lakh Cases	8 days	22-08-20	56706	1.90	964.3	1.722	41.93		0.22		0.22
cum 35 Lakh Cases	7 days	29-08-20	63498	1.82	973.0	1.702	41.25		0.25		0.25
cum 40 Lakh Cases	6 days	04-09-20	69561	1.74	996.3	1.692	40.9		0.29		0.29
cum 45 Lakh Cases	6 days	10-09-20	76271	1.70	1111.0	1.692	40.9		0.32		0.32
cum 50 Lakh Cases	5 days	15-09-20	82066	1.64	1157.6	1.692	40.9		0.36		0.36
cum 55 Lakh Cases	6 days	21-09-20	88935	1.62	1146.0	1.692	40.9		0.39		0.39
cum 60 Lakh Cases	6 days	27-09-20	95542	1.59	1101.5	1.692	40.9		0.43		0.43
cum 65 Lakh Cases	6 days	03-10-20	101782	1.57	1039.3	1.692	40.9		0.47		0.47

cum 70 Lakh Cases	7 days	10-10-20	108334	1.55	937.0	1.702	41.25		0.50		0.50
cum 75 Lakh Cases	8 days	18-10-20	114610	1.53	784.5	1.722	41.93		0.54		0.54
cum 80 Lakh Cases	10 days	28-10-20	120527	1.51	591.7	1.752	42.92		0.57		0.57
cum 85 Lakh Cases	10 days	07-11-20	126121	1.48	559.3	1.772	43.57		0.61		0.61
cum 90 Lakh Cases	12 days	19-11-20	132162	1.47	503.2	1.972	49.29		0.65		0.65
cum 95 Lakh Cases	13 days	02-12-20	138648	1.46	498.9	2.332	57.12		0.68		0.68
cum 100 Lakh Cases	16 days	18-12-20	145136	1.45	405.3	3.083	67.56		0.72		0.72
cum 105 Lakh Cases	26 days	13-01-21	151727	1.45	253.6	2.673	62.59		0.75		0.75
cum 110 Lakh Cases	39 days	21-02-21	156385	1.42	119.4	2.052	51.27	10651012	0.79	0.76	1.55
cum 115 Lakh Cases	25 days	18-03-21	159370	1.39	119.4	2.092	52.2	35923500	0.83	2.58	3.40
cum 120 Lakh Cases	10 days	28-03-21	161843	1.35	247.5	2.202	54.59	55180875	0.86	3.96	4.82
cum 125 Lakh Cases	7 days	04-04-21	165101	1.32	464.4	2.202	54.59	76405697	0.90	5.48	6.38
cum 130 Lakh Cases	4 days	08-04-21	167642	1.29	640.5	2.112	52.65	91881530	0.93	6.59	7.53
cum 135 Lakh Cases	3 days	11-04-21	170179	1.26	838.0	2.012	50.3	102000401	0.97	7.32	8.29
cum 140 Lakh Cases	3 days	14-04-21	173123	1.24	981.0	1.902	47.42	111913288	1.00	8.03	9.04
cum 145 Lakh Cases	2 days	16-04-21	175649	1.21	1260.0	1.822	45.12	117305344	1.04	8.42	9.46
cum 150 Lakh Cases	2 days	18-04-21	178769	1.19	1560.0	1.742	42.59	121207098	1.08	8.70	9.78
cum 155 Lakh Cases	2 days	20-04-21	182533	1.18	1882.0	1.701	41.21	127428887	1.11	9.15	10.26
cum 160 Lakh Cases	2 days	22-04-21	186920	1.17	2193.0	1.641	39.06	132754608	1.15	9.53	10.68
cum 165 Lakh Cases	1 days	23-04-21	189544	1.18	2624.0	1.611	37.93	135658324	1.18	9.74	10.92

cum 170 Lakh Cases	2 days	25-04-21	195123	1.15	2789.0	1.581	36.75	139185173	1.22	9.99	11.21
cum 175 Lakh Cases	1 days	26-04-21	197894	1.13	2771.0	1.561	35.94	142524947	1.26	10.23	11.48
cum 180 Lakh Cases	2 days	28-04-21	204832	1.14	3469.0	1.531	34.68	147053392	1.29	10.55	11.85
cum 185 Lakh Cases	1 day	29-04-21	208330	1.13	3498.0	1.521	34.25	149268772	1.33	10.71	12.04
cum 190 Lakh Cases	1 day	30-04-21	211853	1.12	3523.0	1.511	33.82	151998107	~ 1.36	10.91	12.27
cum 195 Lakh Cases	1 day	01-05-21	215542	1.11	3689.0	1.510	33.77	153626325	1.40	11.03	12.43
cum 200 Lakh Cases	2 days	03-05-21	222408	1.11	3433.0	1.501	33.38	156082136	1.44	11.20	12.64
cum 205 Lakh Cases	1 days	04-05-21	226188	1.10	3780.0	1.491	32.93	157750752	1.47	11.32	12.79
cum 210 Lakh Cases	1 days	05-05-21	230168	1.10	3980.0	1.491	32.93	159931238	1.51	11.48	12.98
cum 215 Lakh Cases	2 days	07-05-21	238270	1.10	4051.0	1.481	32.48	165190000	1.54	11.86	13.40
cum 220 Lakh Cases	1 day	08-05-21	242347	1.10	4077.0	1.481	32.48	167493857	1.58	12.02	13.60
cum 225 Lakh Cases	1 day	09-05-21	246116	1.09	3769.0	1.471	32.02	168304868	1.61	12.08	13.69
cum 230 Lakh Cases	2 days	11-05-21	254197	1.11	4040.0	1.471	32.04	173862643	1.65	12.48	14.13
cum 235 Lakh Cases	1 day	12-05-21	258317	1.10	4120.0	1.461	31.58	176045577	1.69	12.63	14.32
cum 240 Lakh Cases	1 day	13-05-21	262317	1.09	4000.0	1.461	31.58	178361846	1.72	12.80	14.52
cum 245 Lakh Cases	2 days	15-05-21	270284	1.10	3983.0	1.461	31.58	181544536	1.76	13.03	14.79
cum 250 Lakh Cases	2 days	17-05-21	278719	1.11	4217.0	1.461	31.58	183817204	1.79	13.19	14.98
cum 255 Lakh Cases	2 days	19-05-21	287122	1.13	4201.0	1.451	31.10	186410600	1.83	13.38	15.21
cum 260 Lakh Cases	2 days	21-05-21	295525	1.14	4202.0	1.451	31.10	189344105	1.87	13.59	15.45
cum 265 Lakh Cases	2 days	23-05-21	303720	1.15	4098.0	1.451	31.10	191877460	1.90	13.77	15.67

cum 270 Lakh Cases	2 days	25-05-21	311388	1.15	3834.0	1.451	31.10	196463495	1.94	14.10	16.04
cum 275 Lakh Cases	2 days	27-05-21	318895	1.16	3754.0	1.451	31.10	201438120	1.97	14.46	16.43
cum 280 Lakh Cases	3 days	30-05-21	329100	1.18	3402.0	1.451	31.10	208907723	2.01	14.99	17.00
cum 285 Lakh Cases	4 days	03-06-21	340702	1.20	2901.0	1.451	31.10	219831571	2.05	15.78	17.83
cum 290 Lakh Cases	5 days	08-06-21	344082	1.22	2565.0	1.451	31.10	223642281	2.08	16.86	18.94
cum 295 Lakh Cases	5 days	13-06-21	374305	1.27	4155.0	1.451	31.10	250656362	2.12	17.99	20.10
cum 300 Lakh Cases	9 days	22-06-21	390660	1.30	1817.0	1.461	31.58	289961746	2.15	20.81	22.96
cum 305 Lakh Cases	10 days	02-07-21	401050	1.31	1039.0	1.471	32.04	339447068	2.19	24.36	26.55
cum 310 Lakh Cases	13 days	15-07-21	412531	1.33	884.0	1.481	32.50	390286930	2.22	28.01	30.23
cum 315 Lakh Cases	13 days	28-07-21	422662	1.34	780.0	1.481	32.50	445696580	2.26	31.99	34.25
cum 320 Lakh Cases	13 days	10-08-21	429179	1.34	501.0	1.502	33.40	517799432	2.30	37.16	39.46
cum 325 Lakh Cases	14 days	24-08-21	434756	1.34	398.0	1.511	33.82	594552135	2.33	42.67	45.00
cum 330 Lakh Cases	12 days	05-09-21	440752	1.34	499.0	1.521	34.25	686693339	2.37	49.28	51.65
cum 335 Lakh Cases	15 days	20-09-21	445385	1.33	309.0	1.511	33.82	817737078	2.40	58.69	61.09

#### DATASOURCE

 1.WORLD WIDE DATA SET:
 https://data.humdata.org/dataset/novel-coronavirus-2019-ncov-cases

 2.VACCINATION DATA SET:
 https://api.covid19india.org/

In state

CFR: Case Fatality rate , number of deaths for every 100 cases Prepared by Spatial Epidemiology Lab , ICAR-NIVEDI, Bengaluru.

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								7.7	जा स वि सं		
Number of Infections (1 laks Increment)	No of days taken to reach since 14 March- 2020	date reached since 14 March -2020	Cumulative Number of deaths	CFR	avg. daily deaths	R <sub>0</sub> for confirmed cases	Required herd Immunity (Threshold) R <sub>0</sub>	Total Vaccine Administered (cum)	% of Immunity by Infection	% of immunity by vaccination	total % of Immunity gained
1st 1 lakh Cases	182 days	09-11-2020	411	0.41	2.0	1.732	42.25		0.14		0.14
Cum 2 Lakh Cases	20 days	10-01-2020	772	0.39	18.0	1.692	40.89		0.28		0.28
Cum 3 Lakh Cases	12 days	10-13-2020	1047	0.35	23.0	1.672	40.18		0.43		0.43
Cum 4 Lakh Cases	14 days	10-27-2020	1377	0.34	24.0	1.642	39.09	r I	0.57		0.57
Cum 5 Lakh Cases	15 days	11-11-2020	1772	0.35	26.0	1.652	39.45		0.71		0.71
Cum 6 Lakh Cases	19 days	11-30-2020	2245	0.37	25.0	1.702	41.24	and the second second	0.85		0.85
Cum 7 Lakh Cases	19 days	12-19-2020	2787	0.40	28.0	1.802	44.5		0.99		0.99
Cum 8 Lakh Cases	20 days	01-08-2021	3258	0.41	24.0	1.862	46.29		1.13		1.13
Cum 9 Lakh Cases	19 days	01-27-2021	3664	0.41	21.0	2.052	51.27	58323	1.28	0.08	1.36
Cum 10 Lakh Cases	18 days	02-14-2021	3986	0.40	18.0	2.983	66.48	340452	1.42	0.48	1.90
Cum 11 Lakh Cases	33 days	03-19-2021	4468	0.41	15.0	3.363	70.27	2298455	1.56	3.26	4.82
Cum 12 Lakh Cases	31 days	04-16-2021	4878	0.41	13.0	3.433	70.87	5680114	1.70	8.06	9.76

Cum 13 Lakh Cases	5 days	04-21-2021	5001	0.38	25.0	2.492	59.88	6353600	1.84	9.01	10.86
Cum 14 Lakh Cases	4 days	04-25-2021	5111	0.37	27.0	2.122	52.88	6847075	1.99	9.71	11.70
Cum 15 Lakh Cases	4 days	04-29-2021	5260	0.35	37.0	1.842	45.71	7229177	2.13	10.25	12.38
Cum 16 Lakh Cases	2 days	05-01-2021	5357	0.33	48.0	1.762	43.24	7425416	2.27	10.53	12.80
Cum 17 Lakh Cases	3 days	05-04-2021	5508	0.32	50.0	1.702	41.24	7577304	v E ₽ 2.41	10.75	13.16
Cum 18 Lakh Cases	3 days	05-07-2021	5683	0.32	58.0	1.642	39.09	7869269	2.55	11.16	13.71
Cum 19 Lakh Cases	2 days	05-09-2021	5815	0.31	66.0	1.612	37.95	7952556	2.69	11.28	13.97
Cum 20 Lakh Cases	3 days	05-12-2021	6054	0.30	80.0	1.592	37.17	8185506	2.84	11.61	14.45
Cum 21 Lakh Cases	3 days	05-15-2021	6340	0.30	95.0	1.572	36.37	8417840	2.98	11.94	14.92
Cum 22 Lakh Cases	3 days	05-18-2021	6613	0.30	91.0	1.562	35.96	8573567	3.12	12.16	15.28
Cum 23 Lakh Cases	4 days	05-22-2021	7171	0.31	139.0	1.552	35.55	8644281	3.26	12.26	15.52
Cum 24 Lakh Cases	4 days	05-26-2021	7883	0.33	178.0	1.542	35.13	8789981	3.40	12.47	15.87
Cum 25 Lakh Cases	4 days	05-30-2021	8642	0.35	190.0	1.542	35.13	9241456	3.55	13.11	16.65
Cum 26 Lakh Cases	5 days	06-04-2021	9511	0.37	174.0	1.542	35.13	10026627	3.69	14.22	17.91
Cum 27 Lakh Cases	7 days	06-11-2021	10805	0.40	185.0	1.551	35.53	11102819	3.83	15.75	19.58
Cum 28 Lakh Cases	9 days	06-20-2021	12061	0.43	140.0	1.561	35.94	12161549	3.97	17.25	21.22
Cum 29 Lakh Cases	9 days	06-29-2021	13094	0.45	115.0	1.571	36.35	13948645	4.12	19.78	23.90
Cum 30 Lakh Cases	8 days	07-07-2021	14108	0.47	127.0	1.581	36.75	15120585	4.26	21.45	25.70
Cum 31 Lakh Cases	7 days	07-14-2021	14938	0.48	119.0	1.601	37.54	16206261	4.40	22.99	27.38
Cum 32 Lakh Cases	7 days	07-21-2021	15618	0.49	97.0	1.621	38.31	17427952	4.54	24.72	29.26

Cum 33 Lakh Cases	6 days	07-27-2021	16327	0.49	118.0	1.641	39.06	18942910	4.68	26.87	31.55
Cum 34 Lakh Cases	5 days	08-01-2021	16838	0.50	102.0	1.641	39.06	20369496	4.82	28.89	33.71
Cum 35 Lakh Cases	5 days	08-06-2021	17516	0.50	136.0	1.661	39.80	21560396	4.96	30.58	35.55
Cum 36 Lakh Cases	5 days	08-11-2021	18121	0.50	121.0	1.671	40.16	22443788	5.11	31.83	36.94
Cum 37 Lakh Cases	5 days	08-16-2021	18744	0.51	125.0	1.671	40.16	24620672	<b>VED</b> 5.25	34.92	40.17
Cum 38 Lakh Cases	5 days	08-21-2021	19429	0.51	137.0	1.671	40.16	25732732	5.39	36.50	41.89
Cum 39 Lakh Cases	5 days	08-26-2021	20135	0.52	141.0	1.671	40.16	27264922	5.53	38.67	44.20
Cum 40 Lakh Cases	3 days	08-29-2021	20542	0.52	136.0	1.651	39.43	28152388	5.67	39.93	45.60
Cum 41 Lakh Cases	4 days	02-09-2021	21150	0.52	152	1.641	39.06	29423239	5.82	41.73	47.55
Cum 42 Lakh Cases	3 days	09-05-2021	21497	0.52	116.0	1.641	39.06	29636101	5.96	42.04	47.99
Cum 43 Lakh Cases	4 days	09-09-2021	22127	0.51	157.0	1.651	39.43	30767695	6.10	43.64	49.74
Cum 44 Lakh Cases	5 days	14-09-2021	22780	0.52	131.0	1.651	39.43	31874863	6.24	45.21	51.45
Cum 45 Lakh Cases	5 days	19-09-2022	23592	0.52	162.0	1.651	39.43	33641007	6.38	47.72	54.10
Cum 46 Lakh Cases	6 days	25-09-2023	24439	0.53	141.0	1.651	39.43	35064476	6.52	49.74	56.26
Cum 47 Lakh Cases						1			X		

CFR: Case Fatality rate , number of deaths for every 100 cases Prepared by Spatial Epidemiology Lab , ICAR-NIVEDI, Bengaluru.





ICAR - National Institute of Veterinary Epidemiology and Disease Informatics

# **Customer/Client Feedback Form**

### Feedback for the Livestock Diseases forewarning bulletin of October-2021, Volume 9 and

#### Issue 10

(Please return this duly fill in after receiving the outbreak report of December-2021)

Sl. No	Diseases Name	No of districts outbreak occurred but not alerted**	Measure taken in case of disease forecasted:	सपज Any other
			Yes or No**	
1.	Anthrax	a l'elerer		[
2.	Babesiosis			
3.	Black Quarter			
4.	Bluetongue			
5.	Enterotoxaemia			
6.	Fasciolosis			
7.	Foot and mouth disease		1	
8.	Haemorrhagic septicaemia			
9.	Peste des Petits Ruminants			
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 outbreak predicted?



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

Very satisfied	Satisfied	Unsatisfied	Not sure
-1			
			1
vement of	report.		
	satisfied	Salistien	satisfied Satisfied Chsatisfied

Sign and Signature with Designation

AICRP centre:

Dated:





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