

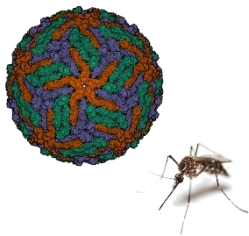
# Machine Learning Algorithms for Predicting the Force of Infection of Japanese Encephalitis

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29 November 2021

## Background- Japanese Encephalitis

- JE is a vaccine preventable **mosquito-borne flavivirus**.
- JE primarily affects **children**. Most adults in endemic countries have natural immunity after childhood infection, but individuals of any age may be affected.
- Most JEV infections are mild (fever and headache), but approximately **1 in 250 infections results in severe clinical illness**.
- Approximately **67,900 JE cases** typically occur annually, of which only about 10% are reported to the World Health Organization.



## Background- Force of Infection and disease burden

- Force of Infection (FOI) is **the rate at which susceptible individuals in a population acquire infection.**
- The FOI is a key parameter used in estimation of disease burden in susceptible countries.
- Current FOI estimates **assume homogeneity** across the entire country- limits assessment of disease burden on a finer scale.
- Due to lack of data- burden estimates are based on **broad assumptions.**
- Hence- there is a need to estimate the FOI on **smaller geographical scales**

## Methodology

We use tree based machine learning models to estimate the FOI for **pixel-level (10s x 10s) data points**.

### A. Data:

- Demographic variables (population density, age groups)
- Geographical data (elevation, Pigs and rice distribution)
- Bioclimatic data

### B. Models used:

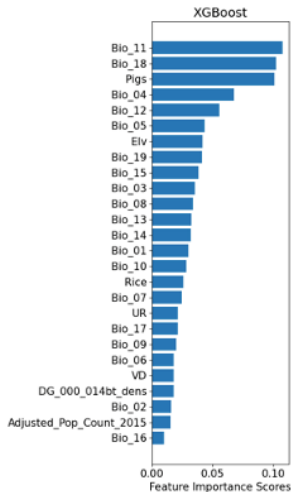
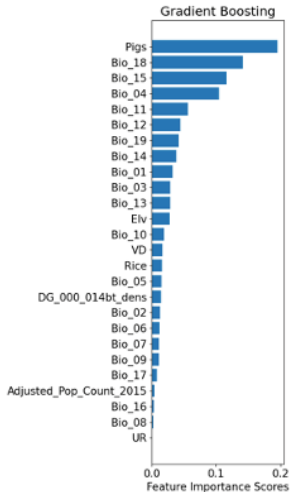
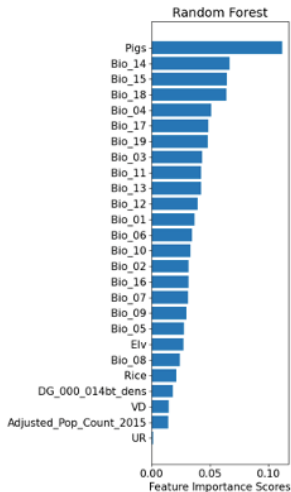
- Principal Component Regression (PCR)
- Random Forest (RF)
- Gradient Boosting (GB)
- XGBoost (XGB)

## Results- MSE

We compared the test Mean Squared Error (MSE) for the four modelling methods:

PCR	0.00496
Random Forest	0.00197
Gradient Boosting	0.00183
XGBoost	0.00172

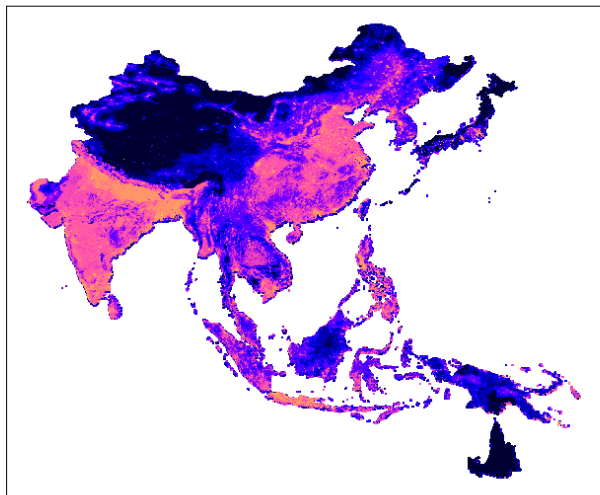
# Results- Feature Importance



## Results

- We also used Individual Conditional Expectation (ICE) plots.
- Our results suggested that **the top 10 features are relatively consistent across methods**. These features included various bioclimatic variables and the pig distribution.
- Features with low importance scores across all three methods, such as adjusted population count and the urban-rural divide (UR), might indicate that they do not contribute much to the prediction of FOI.

## Output- Generating number of JE cases



- [0,0.001]
- (0.001,0.0015]
- (0.0015,0.002]
- (0.002,0.005]
- (0.005,0.0075]
- (0.0075,0.01]
- (0.01,0.015]
- (0.015,0.02]
- (0.02,0.05]
- (0.05,0.07]
- (0.07,0.09]
- (0.09,0.1]
- (0.1,0.15]
- (0.15,0.3]
- (0.3,0.5]
- (0.5,0.7]
- (0.7,1]
- (1,3]
- (3,10]
- (10,20]
- (20,40]
- (40,50]
- (50,70]
- (70,80]



## Conclusion

- The **finer scale estimates of the FOI** allowed for the precise calculation of disease burden, even for countries without surveillance systems in place for JE.
- More accurate burden estimates lead to more **effective understanding of vaccination impact in countries** with significant JE burden.
- The study offers insight on using **tree based algorithms** for understanding public health problems involving heavily correlated predictors.