

Prediction of West Nile Virus in Chicago

Group 1
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Executive Summary

We were hired by the Chicago Department of Public Health (CDPH) to use data to gain insights on the occurrence of West Nile Virus in Chicago. Key insights are:

What are the key factors leading to an outbreak?

- 1** Seasonality, as driven by weather conditions
- 2** Presence of particular mosquito species
- 3** Location

Modelling based on these factors was effective at predicting outbreaks

Is it cost-effective to spray the city to prevent outbreaks?

Yes, economic benefits outweigh the costs based on our cost benefit analysis.

Recommendation

Chicago should build a predictive model using lagged weather, trap and location data to determine which locations it should spray, in order to better combat outbreaks

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Outbreak of the West Nile Virus

Leading cause of mosquito-borne disease in the continental United States

Spread to people by the bite of an infected mosquito

About 1 in 150 infected people develop a serious, and sometimes fatal illness

No vaccines or medications to treat the West Nile Virus in people



Prevention & Eradication

Hired by the Chicago Department of Public Health (CDPH) to use data to gain insights on

- 1. What are the key factors leading to an outbreak?**
- 2. Is it cost-effective to spray the city to prevent outbreaks?**

Allow the CDPH to more efficiently and effectively allocate resources to combat outbreaks



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Context

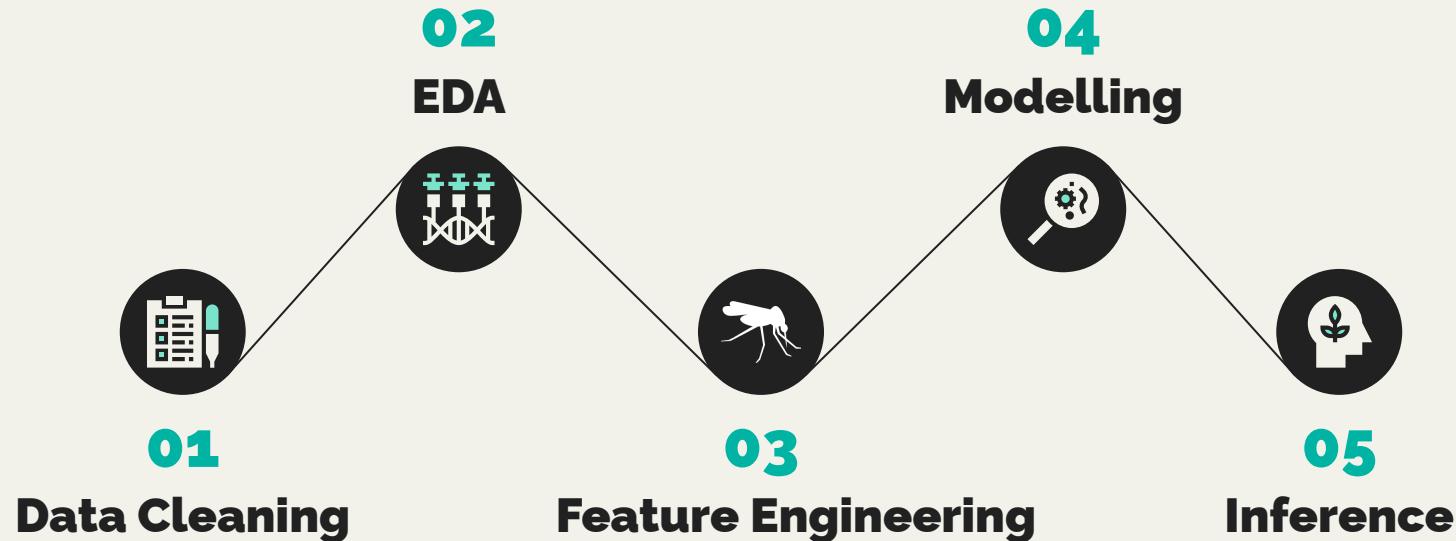
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Methodology



Datasets Used

Train

- Trap data (Location, Species, Date)
- Train data from 2007, 2009, 2011, 2013
- Test data from 2008, 2010, 2012, 2014
- Train data includes number of mosquitoes and presence of WNV

Test

- Weather data from NOAA collected from 2 stations in Chicago
- Weather features matched to trap dataset by distance to nearest station

Weather

Spray

- Limited data on sprays (location, date) from 2011 and 2013
- Excluded from model due to incomplete data

Features Engineered

Totalling 101 explanatory variables

Clusters of Trap Locations

Using KMeans to account for geographical variations e.g. terrain, neighborhood sanitation

Relative Humidity

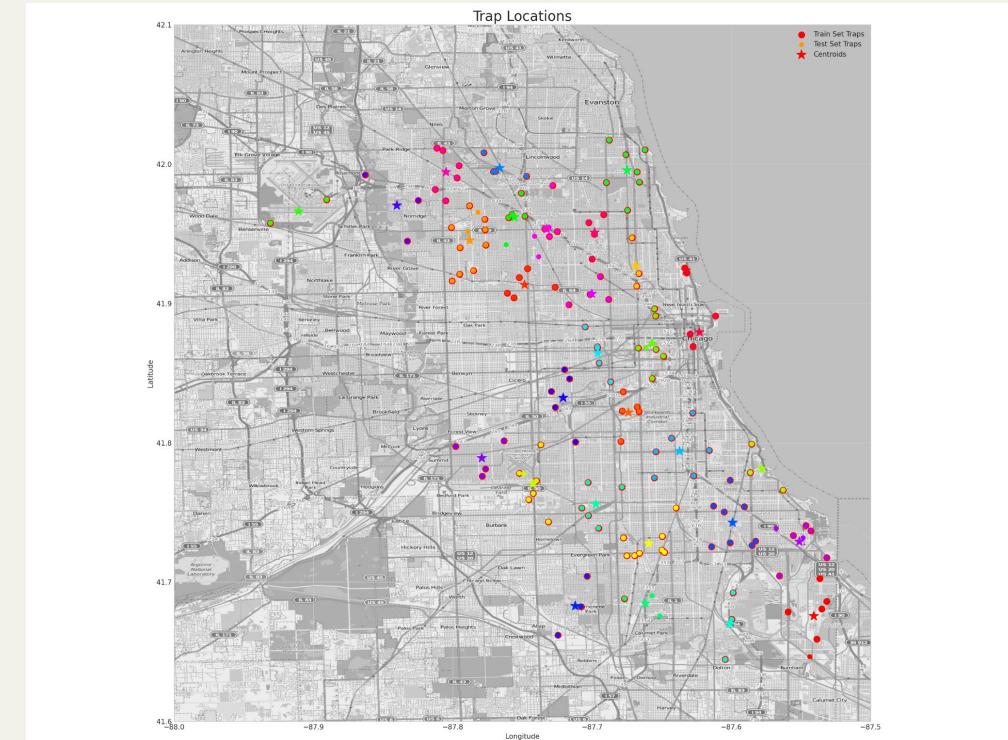
Approximated using dew point and average temperature

Daylight Hours

Computed from sunrise and sunset timings

Moving Averages

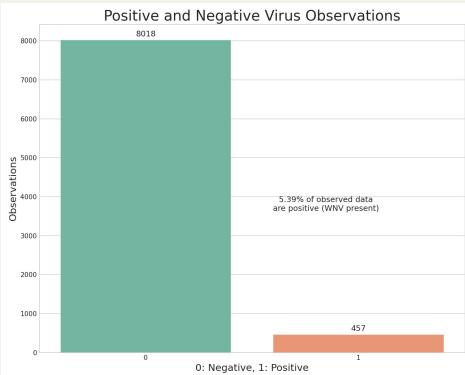
For Temperature, Precipitation, Humidity, Wind Speed, Daylight Hours



Techniques Used in Modelling

SMOTE

Bootstrapping train set to deal with imbalanced classes



StandardScaler

Scales variables for regularisation to prevent overfitting

One-Hot Encoding

Dummified mosquito species and clusters

Feature Selection

Selected moving averages which were correlated with the target

GridSearchCV

Finding and tuning to the best hyperparameters

Logistic Regression, Random Forest, XGB Classifier, Gradient Boost
Classification Models used for Supervised Learning

Best Model

Logistic Regression

Model	Train AUC Score	Test AUC Score	Accuracy	Sensitivity	Specificity	Precision	F1 Score
Dummy (Baseline Model)	0.500	0.500	0.946	0.000	1.000	0.000	0.000
XGB Classifier	0.998	0.843	0.800	0.675	0.807	0.166	0.266
Random Forest	0.983	0.836	0.829	0.570	0.844	0.172	0.264
Gradient Boost	0.961	0.829	0.785	0.711	0.790	0.161	0.263
Logistic Regression	0.901	0.839	0.756	0.754	0.756	0.149	0.249

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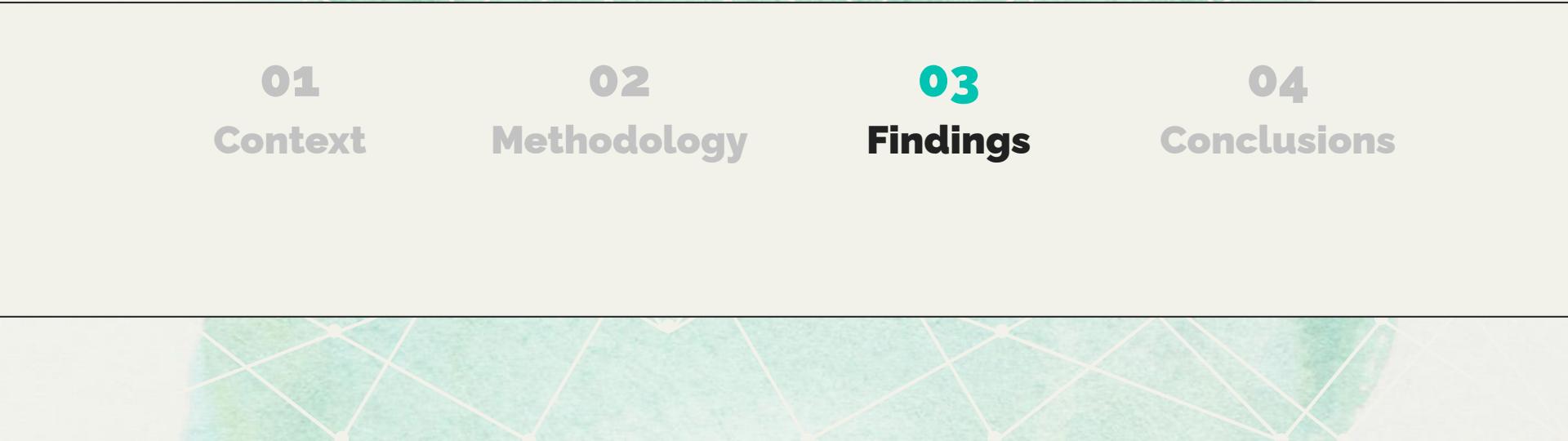


01
Context

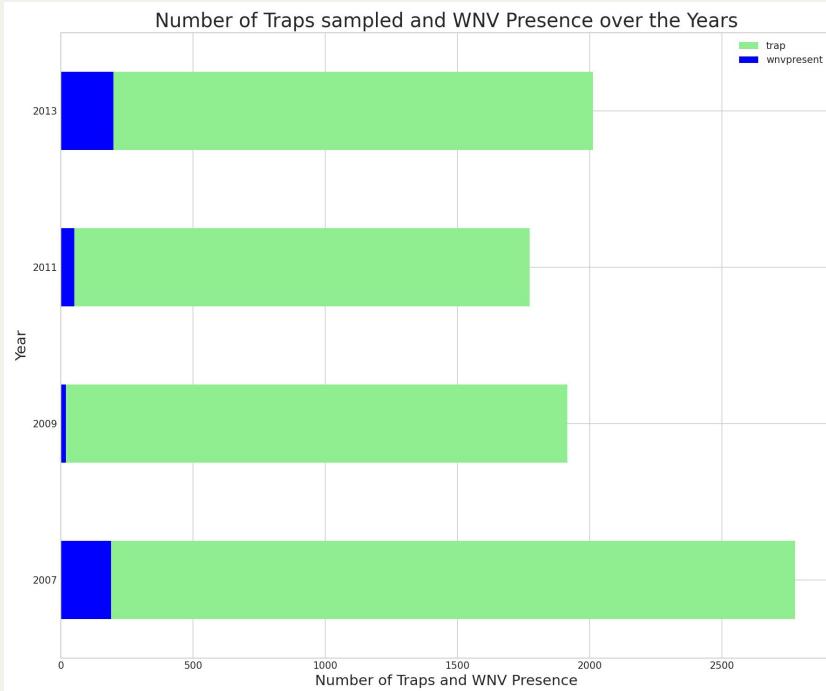
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Outbreaks were worse in 2007 and 2013. Why?



Factors leading to an Outbreak



**Number of
Mosquitos**



**Mosquito
Species**



Location



Seasons



Precipitation



Humidity



Daylight Hours

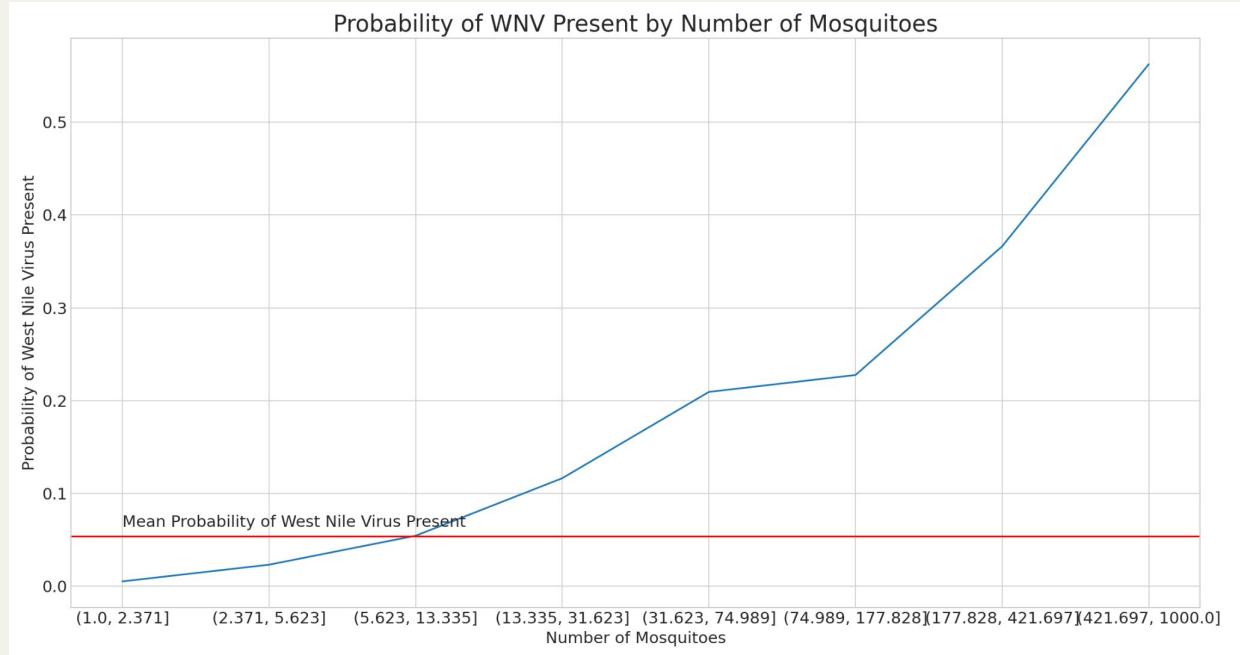
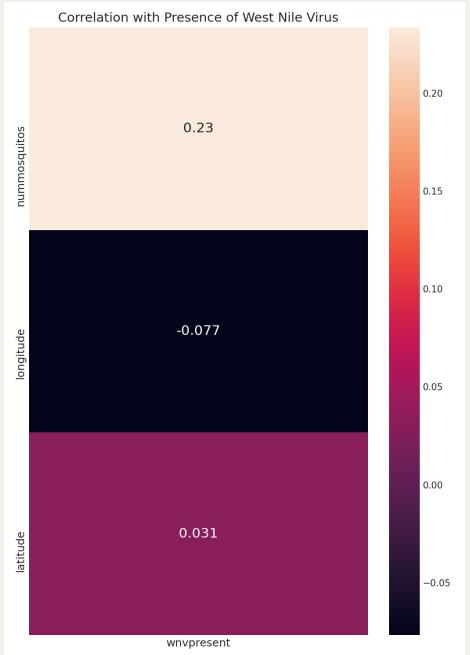


Temperature

Higher number of mosquitoes

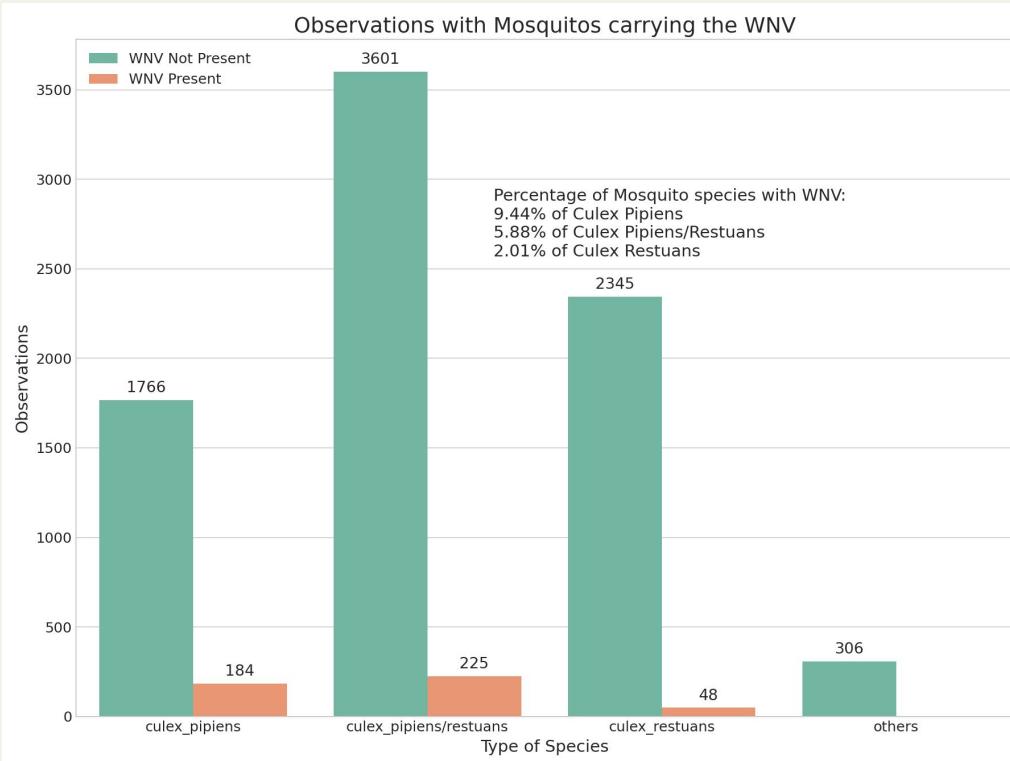
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Higher probability of West Nile Virus



2 out of 6

Species of mosquitoes exhibits the WNV



With WNV Presence:
Culex Pipens (9.44%)
Culex Restuans (2.01%)

Factors leading to an Outbreak



**Number of
Mosquitos**



**Mosquito
Species**



Location



Seasons



Precipitation



Humidity



Daylight Hours

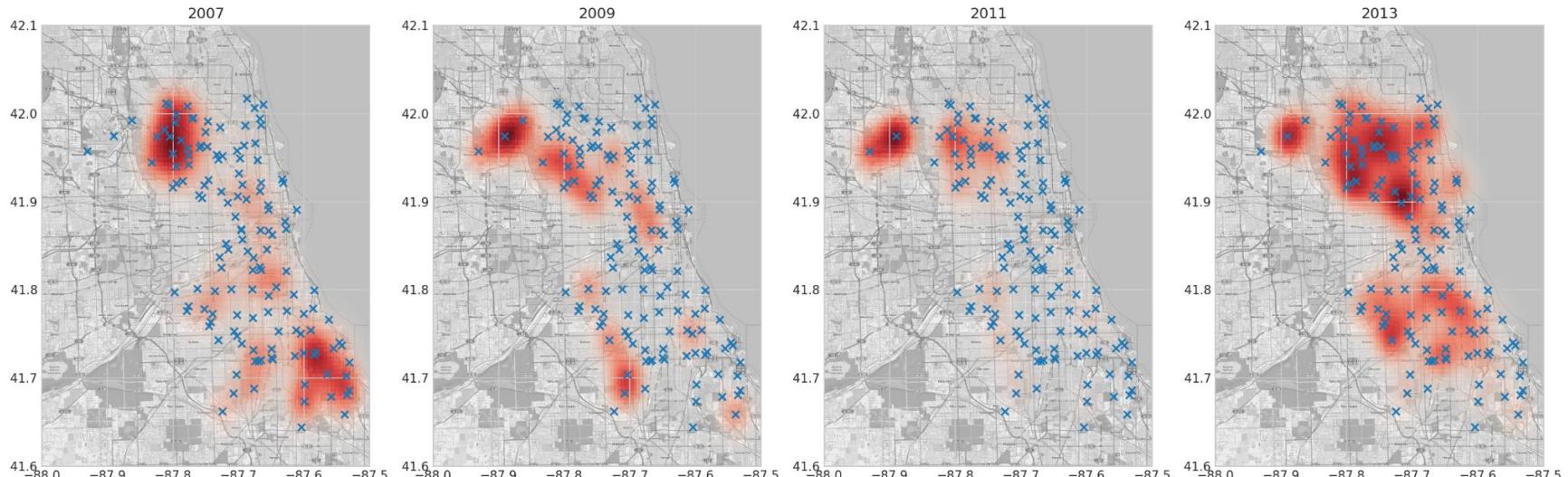


Temperature

Geographical Outbreak Clusters

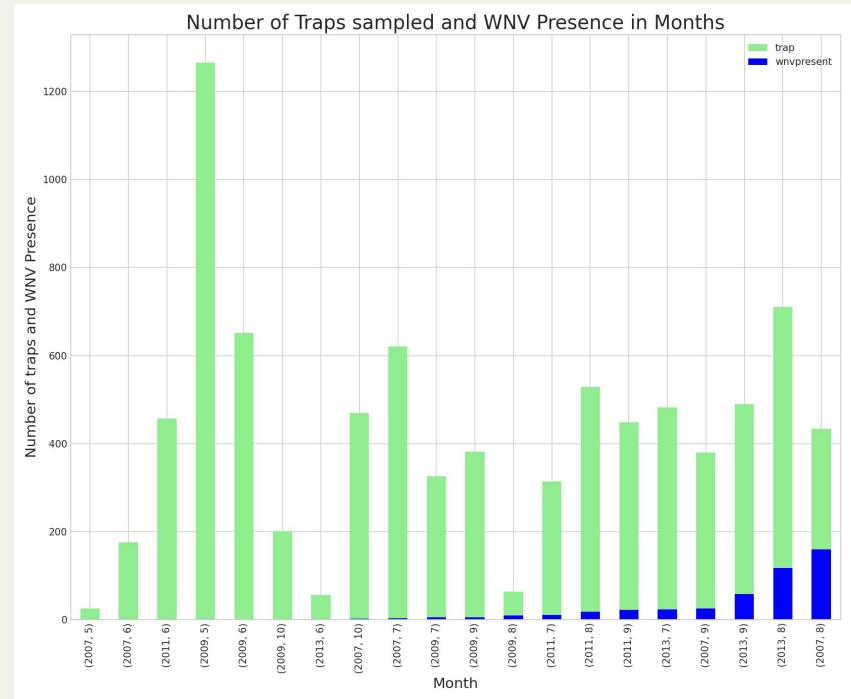
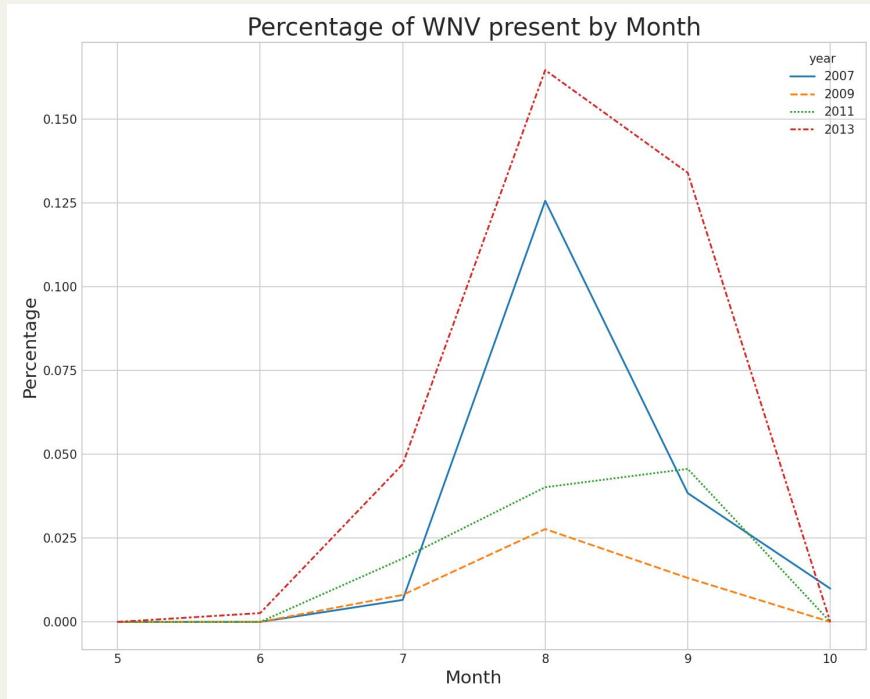
Around specific Neighbourhoods

Traps where West Nile Virus is Detected by Year



● Outbreaks of the West Nile Virus
X Location of Traps

High Seasonality Of WNV Outbreaks exhibited



Factors leading to an Outbreak



Number of
Mosquitos



Mosquito
Species



Location



Seasons



Temperature



Humidity



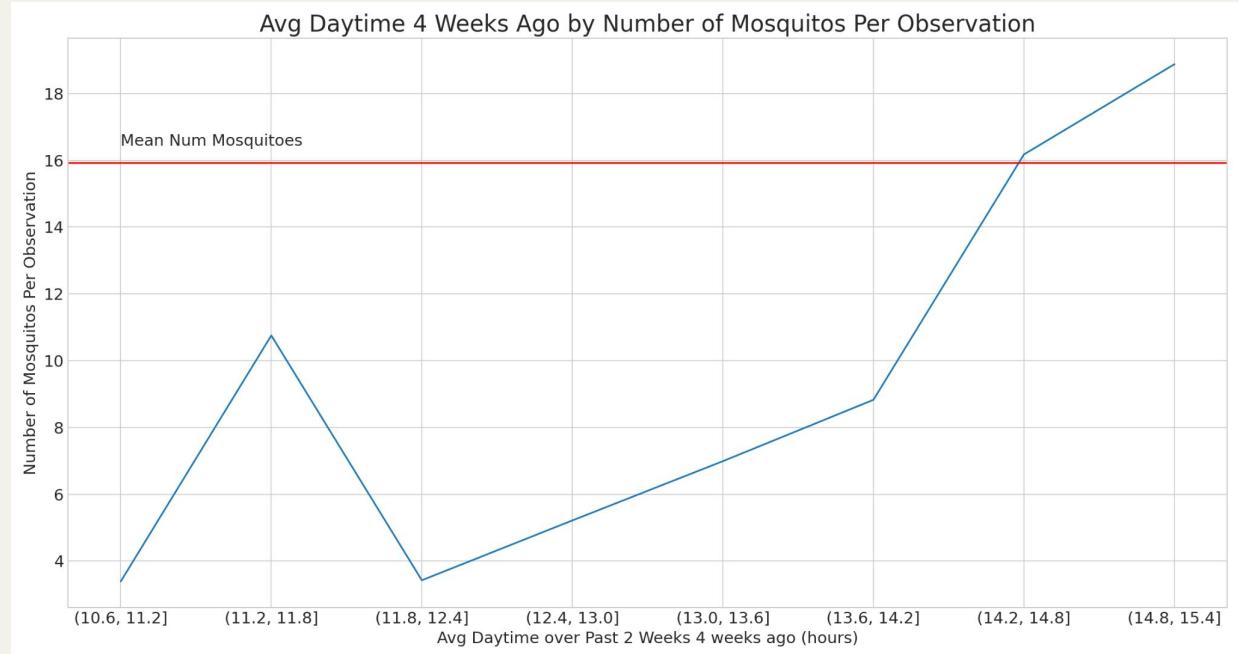
Daylight Hours



Precipitation

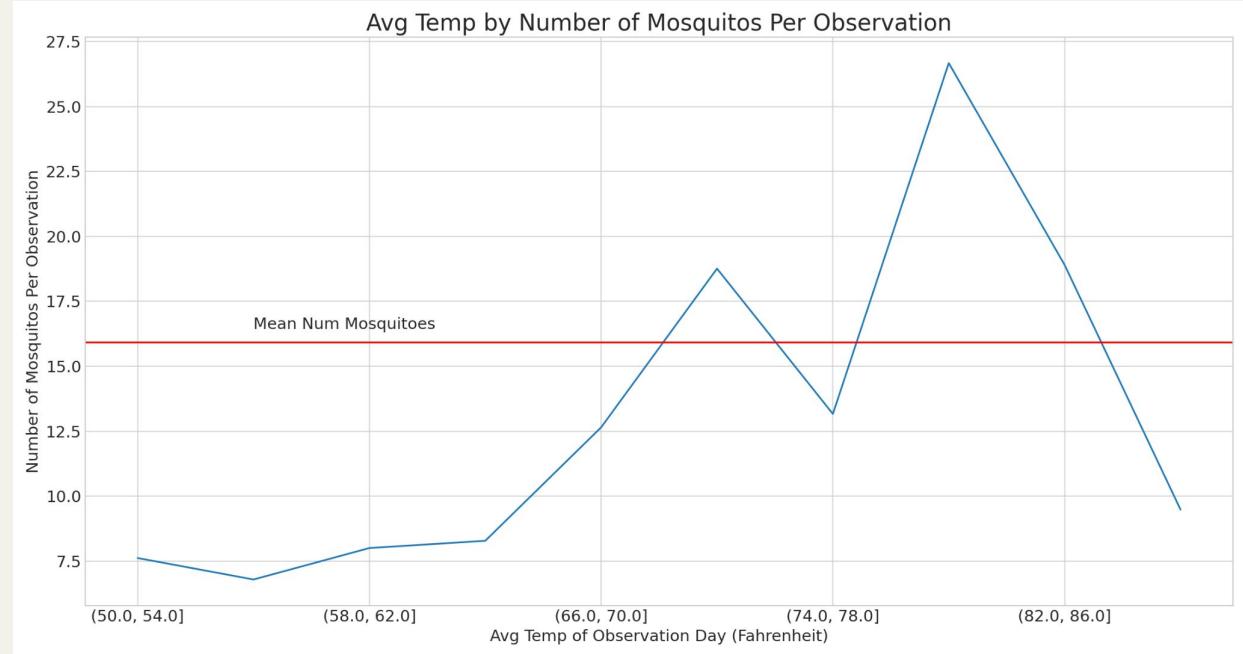
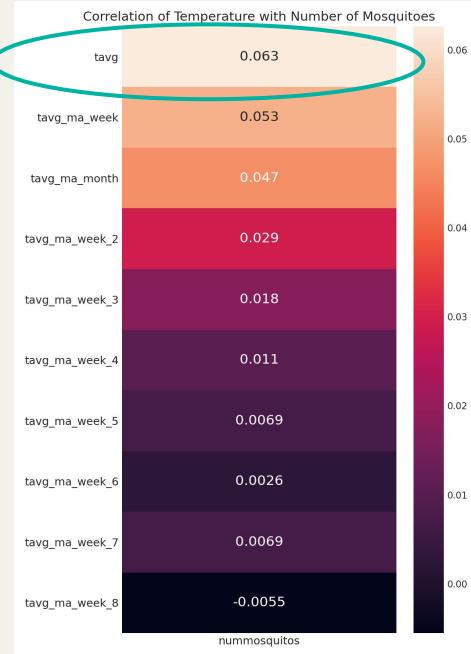
Moving Average of Daylight Hours

In relation to the no. of mosquitoes



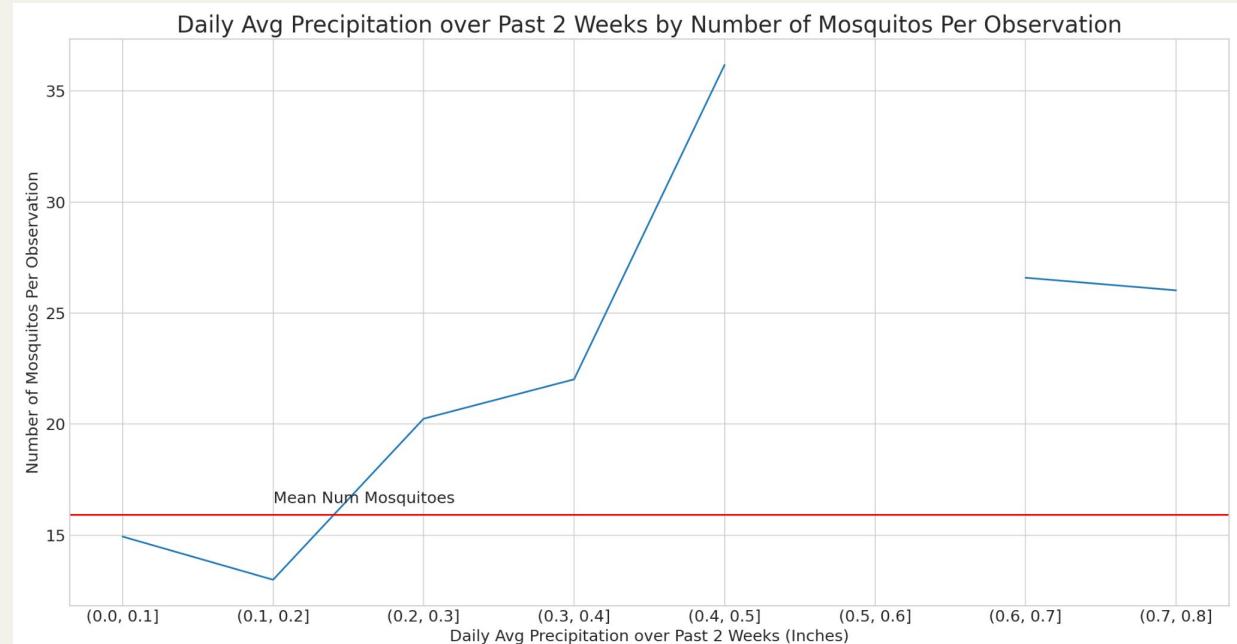
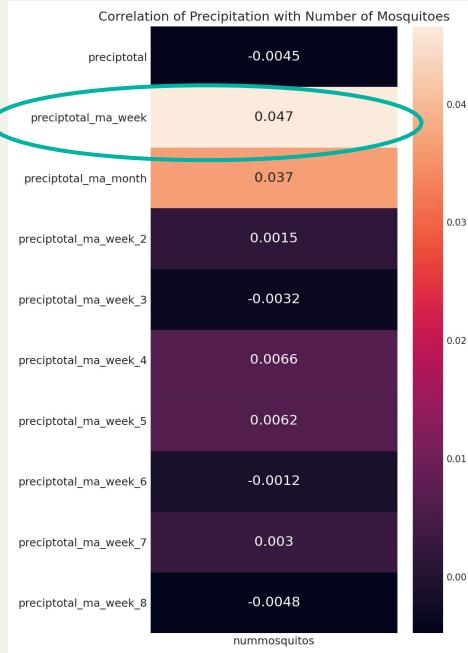
Moving Average of Temperature

In relation to the no. of mosquitoes



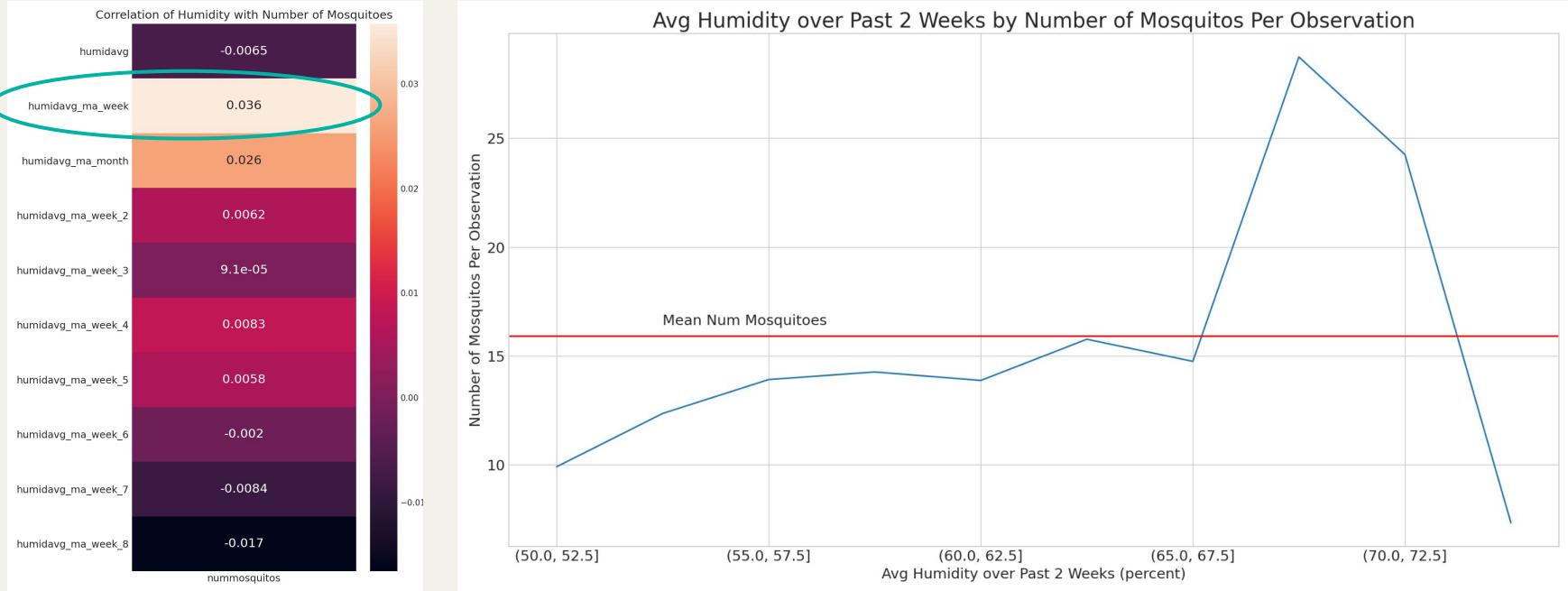
Moving Average of Precipitation

In relation to the no. of mosquitoes



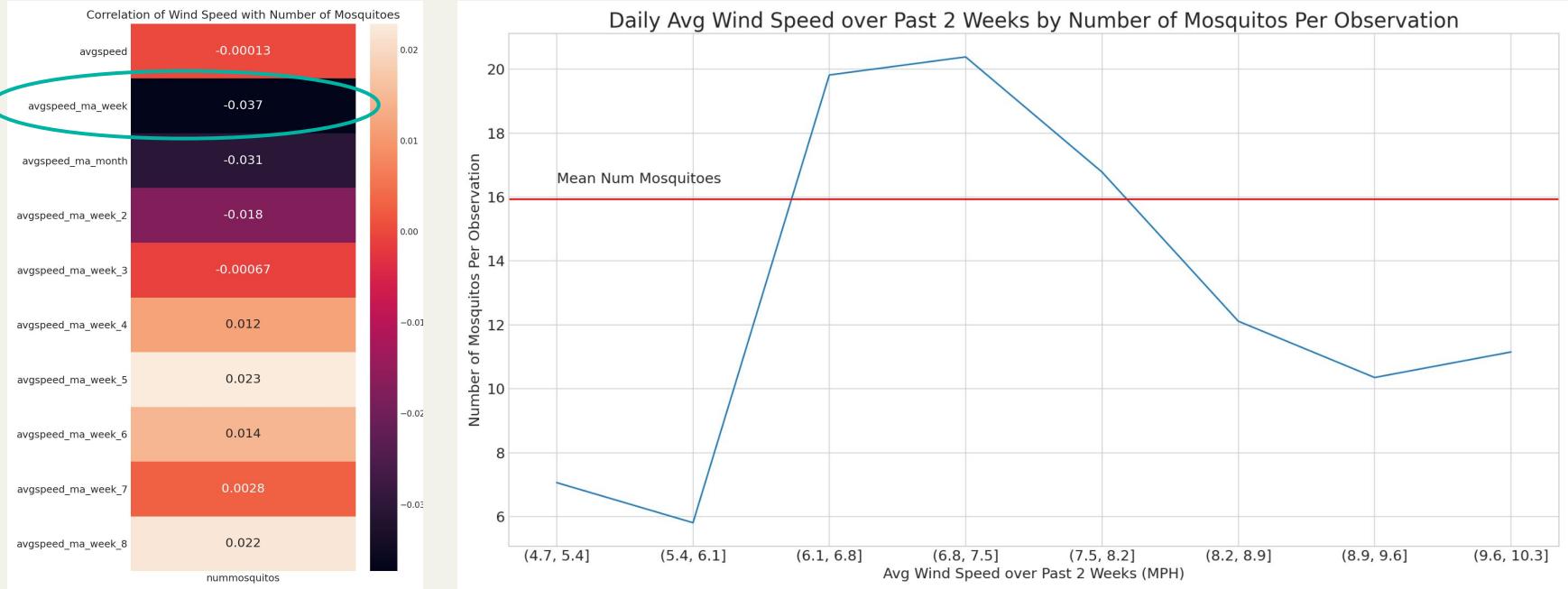
Moving Average of Humidity

In relation to the no. of mosquitoes



Moving Average of Wind Speed

In relation to the no. of mosquitoes

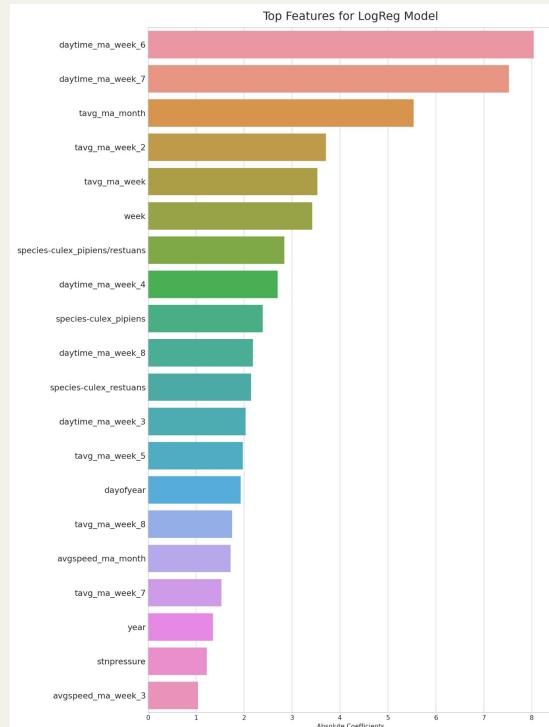


Higher Accuracy of Predictions

Led by Findings and Interpretable Model Coefficients

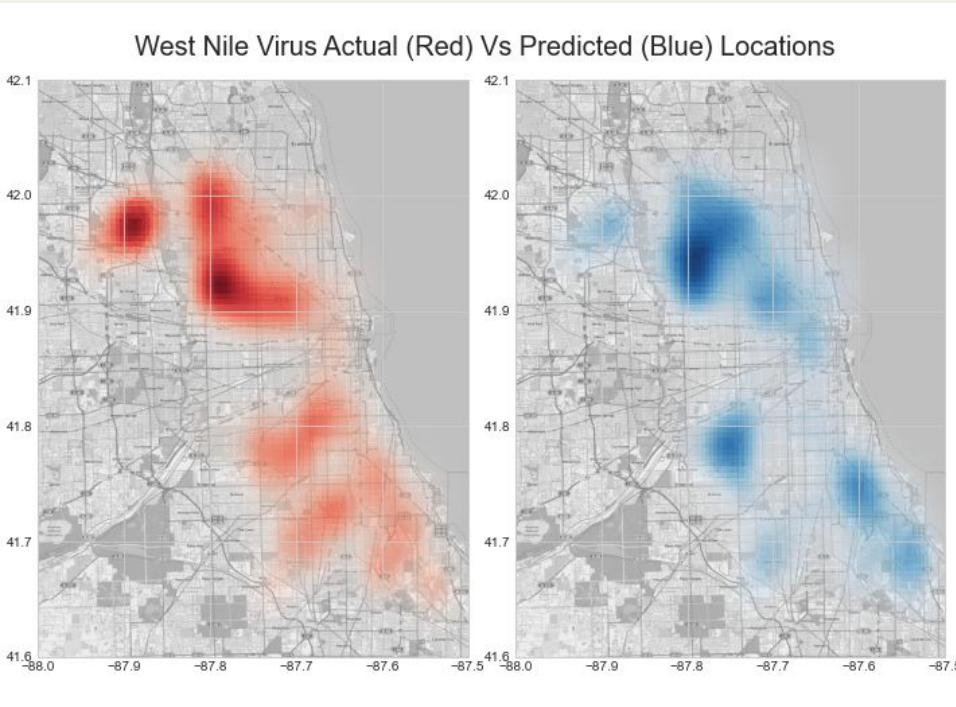
Top Features resulting in higher probabilities of the presence of WNV

- 1) Moving Average for Hours of Daytime
- 2) Moving Average for Average Temperature
- 3) Week of the Year
- 4) Species of Mosquito



Higher Accuracy of Predictions

Led by Findings and Interpretable Model Coefficients



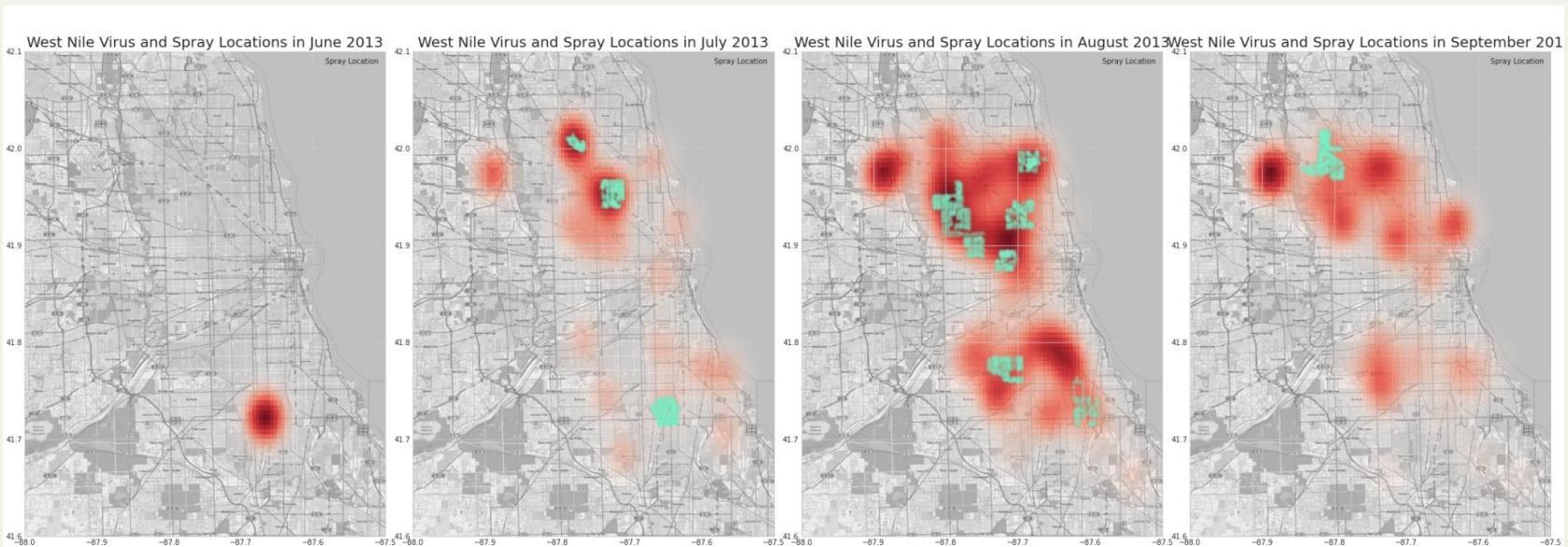
- Model is quite accurate at predicting locations of outbreaks
- Low false negative rate, at the expense of high false positive rate

Confusion Matrix	Predicted Negative	Predicted Positive
Actual Negative	1515 (72.2%)	490 (23.3%)
Actual Positive	28 (1.3%)	68 (3.2%)

Cost-Benefit Analysis

Should the Government spend more on insecticides?

- Spraying seems to be effective at controlling outbreaks within a small radius
- Spraying is reactive rather than proactive, when there is an outbreak in the area



Benefits vs Costs

Economical Cost without Spraying (Sacramento County)

Inpatient Cost per case	\$ 33,143
Outpatient Cost per case	\$ 6,317
Productivity Cost per case	\$ 4,775
Total Economic Cost per case (Inpatient)	\$ 37,918
Total Economic Cost per case (Outpatient)	\$ 11,092

Spraying Cost Estimation

Country	Sacramento County, California	Chicago
Area	2,574 km ²	606 km ²
Sprayed Area	477 km ² (18%)	242 km ² (~40%)
Sprayed Cost per Area	\$ 1,471	\$ 1,471
Total Sprayed Cost	\$ 701,790	\$ 356,045

Chicago Cases in 2018

Total Cases	57
Cases Hospitalised	44
Fatalities	3

Sources: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3322011/> ;
https://www.chicago.gov/content/dam/city/depts/cdph/food-env/general/West_Nile_Virus/WNV_2018databrief_FINALJan102019.pdf

Spraying is economically cost effective and should be carried out

- Economic benefits outweigh the costs from spraying, even if not all cases are prevented

Hospitalisation and Productivity Benefits			
	Benefits	2018 Cases	Total Benefits
Inpatient	\$ 37,918	44	\$1,668,392
Outpatient	\$ 11,092	13	\$144,196
Total			\$1.8 mil



Spraying Cost
\$ 356,045

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Conclusions & Recommendation

What are the key factors leading to an outbreak?

- 1** Seasonality, as driven by weather
- 2** Presence of particular mosquito species
- 3** Location

Backtesting shows predictive models built using these variables do better than baseline

Is it cost-effective to spray the city to prevent outbreaks?

Yes, economic benefits outweigh the costs.

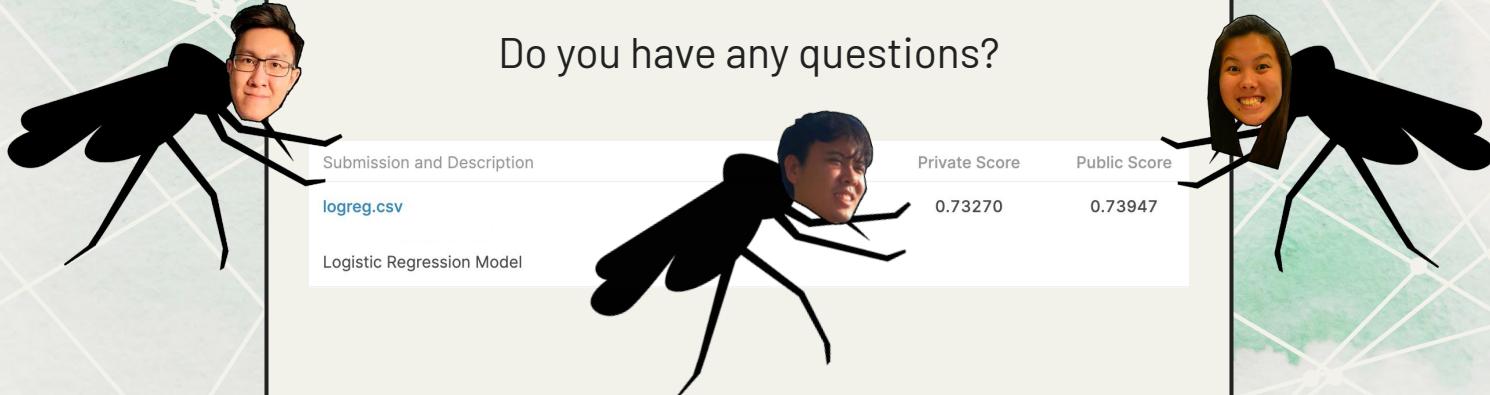
Spraying insecticides in Chicago seems to be reactive currently.

Recommendation

Chicago should build a predictive model using lagged weather, trap and location data to determine which locations it should spray, in order to better combat outbreaks

Thanks!

Do you have any questions?



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