
Project 4

West Nile Virus Prediction

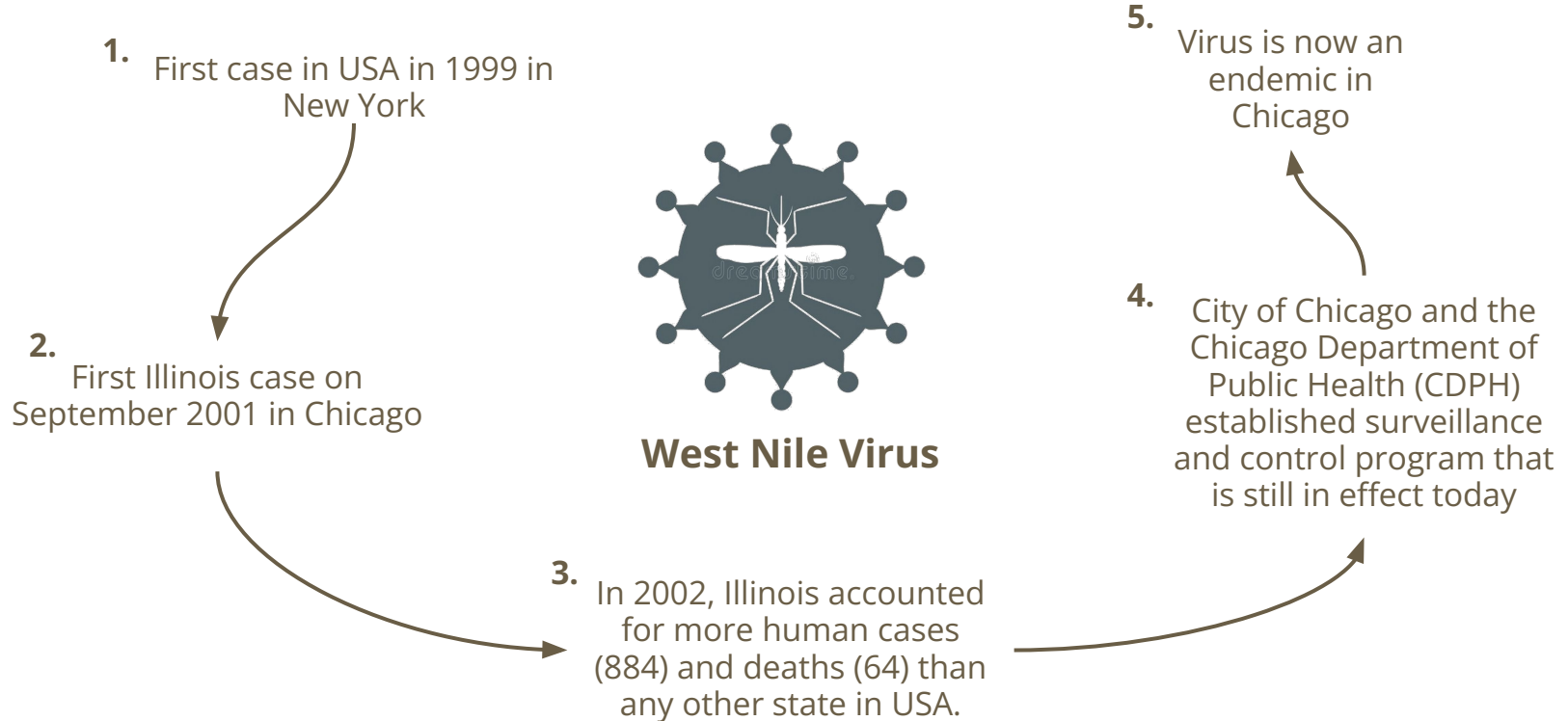
Vincent Chua
May Gee
Tan Cheng Yeow
Tan Cheng Kiat
Low Zhe Wei

Table Of Contents

- Introduction and Problem Statement
- EDA & Feature Engineering
- Modelling
- Cost-Benefit Analysis
- Conclusions and Recommendations



Introduction



Problem Statement

Due to the endemic of West Nile Virus in Chicago, the Department of Public Health has set up a surveillance and control system through which weather, location, testing, and spraying data was collected. CDPH has contacted our team to develop a model to predict the locations where there would be West Nile virus outbreaks.

Using these available datasets, the model will help the City of Chicago and CPHD more efficiently and effectively target spraying of specific neighbourhoods with higher risk of West Nile Virus. This can help the City of Chicago save costs while still keeping the virus at bay. Our model efficacy will be assessed by the Kaggle submission.

Datasets - Kaggle

Spray

- ❖ Total of 14835 observations from 4 features
- ❖ Date, time and location of spray the pesticides

Train

- ❖ Total of 10506 observations from 12 features.
- ❖ Additional **NumMosquitos** and **WnvPresent** features
- ❖ Date, time, location of trap and mosquitos species and numbers

Weather

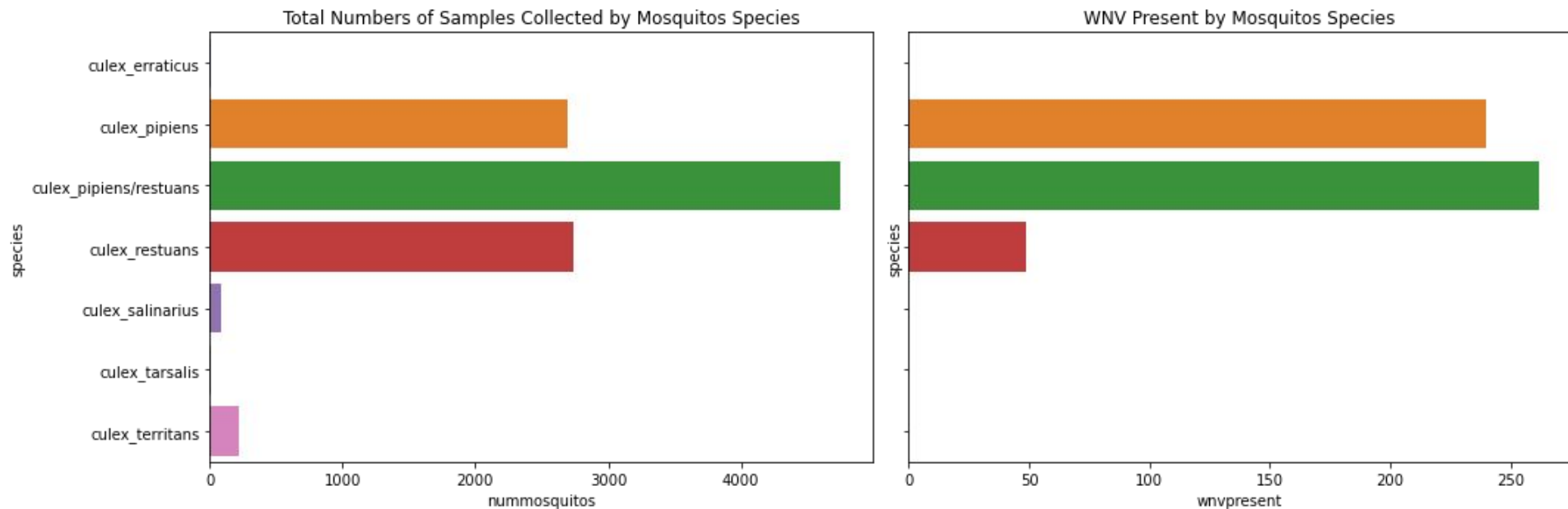
- ❖ Total of 2944 observations from 22 features
- ❖ Date, temperature, windspeed, station pressure etc.

Test

- ❖ Total of 11,6293 from 11 features.
- ❖ Date, time, location of trap and mosquitos species

Exploratory Data Analysis (EDA) - Species (Train)

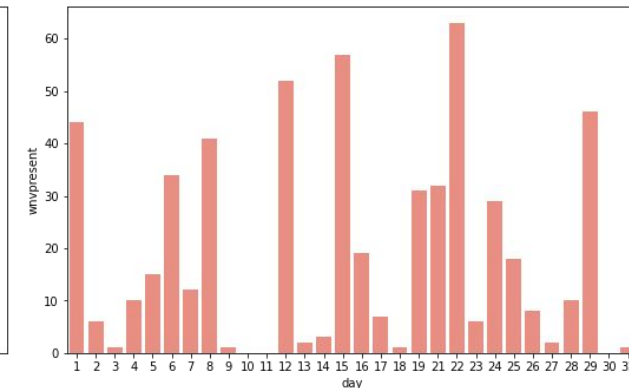
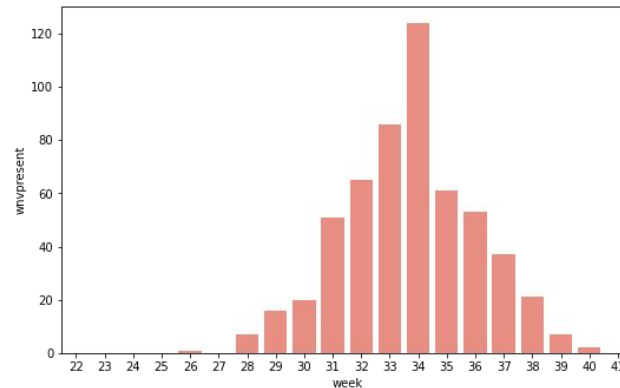
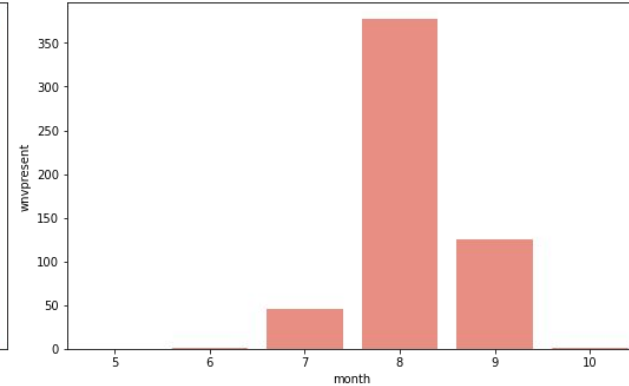
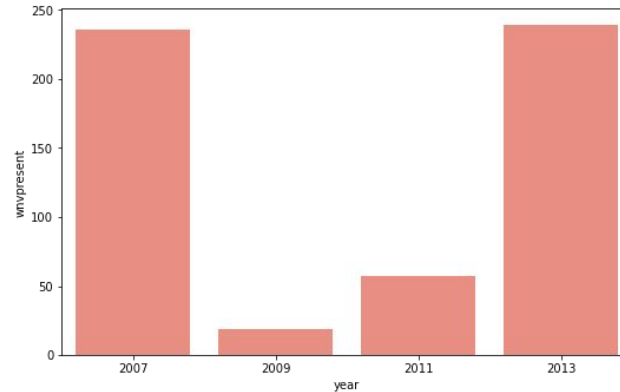
- Imbalanced Class of Datasets with only 5.2% WNV Present cases.
- Majority of mosquitos species are **piens & restuans**



EDA - WNV Present

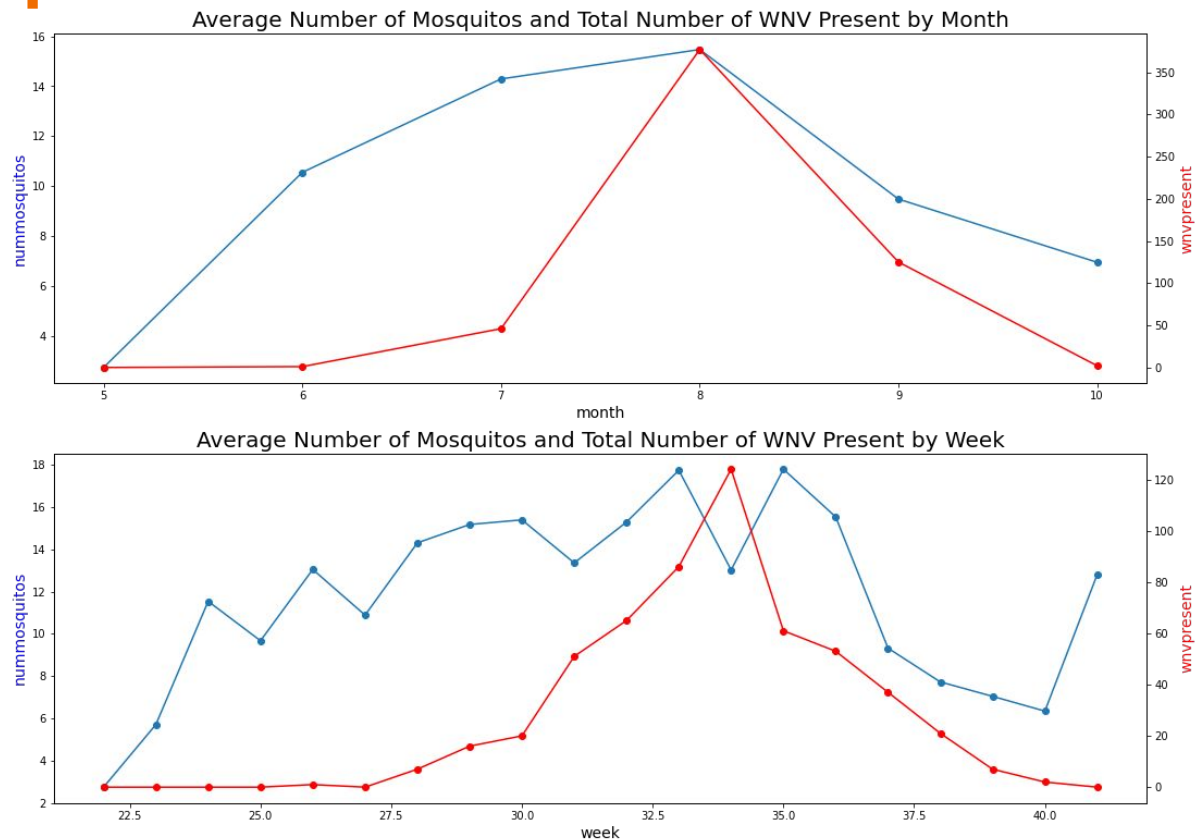
- The months of WNV present cases was observed within July to September and peak in August
- Years and Day does not show any consistency in case present

Total Number of WNV Cases by Year, Month, Week and Day



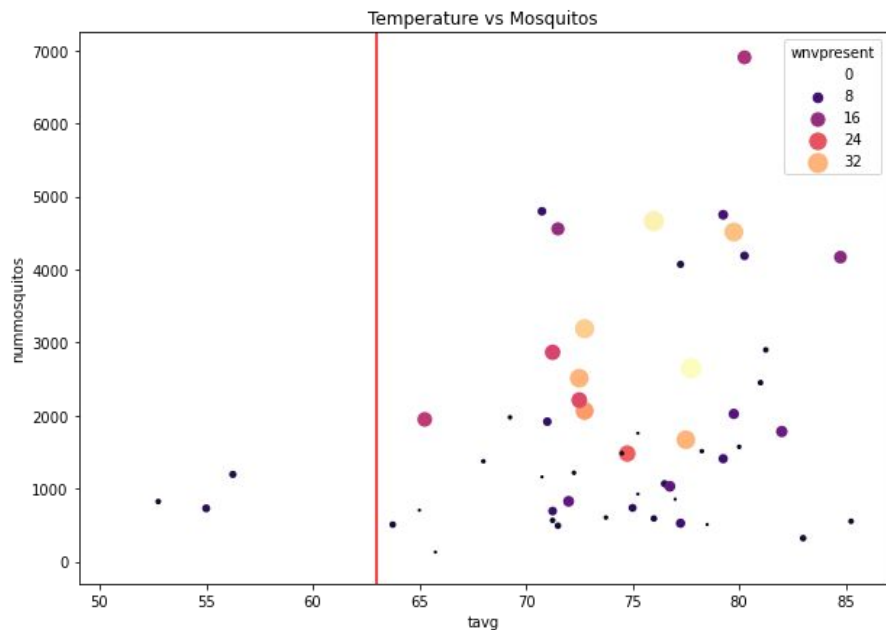
EDA - Number of Mosquitos

- For both **month** and **week**, when the numbers of mosquitos increasing, the WNV cases tend to increase

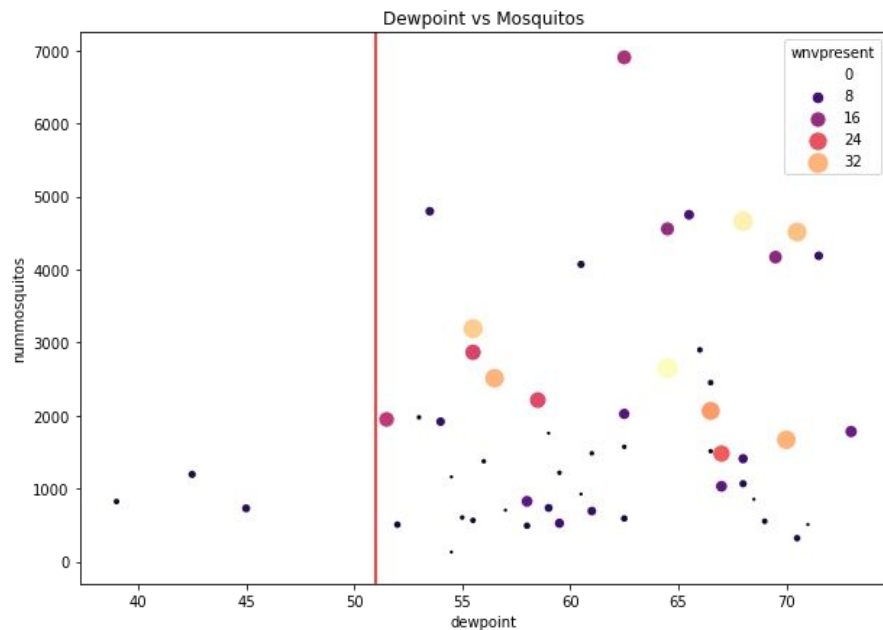


EDA - Temperature & Dewpoint (Weather)

For **average temperatures** above 63°F, we can see that number of mosquitos and west nile clusters are more prevalent.

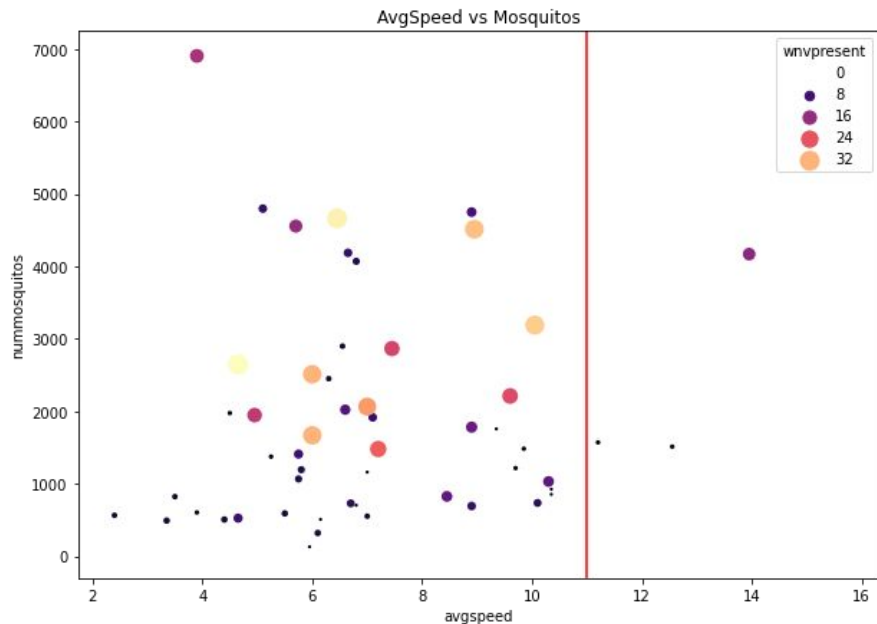


For **DewPoint** above 51°F, we can see that number of mosquitos and west nile clusters are more prevalent.

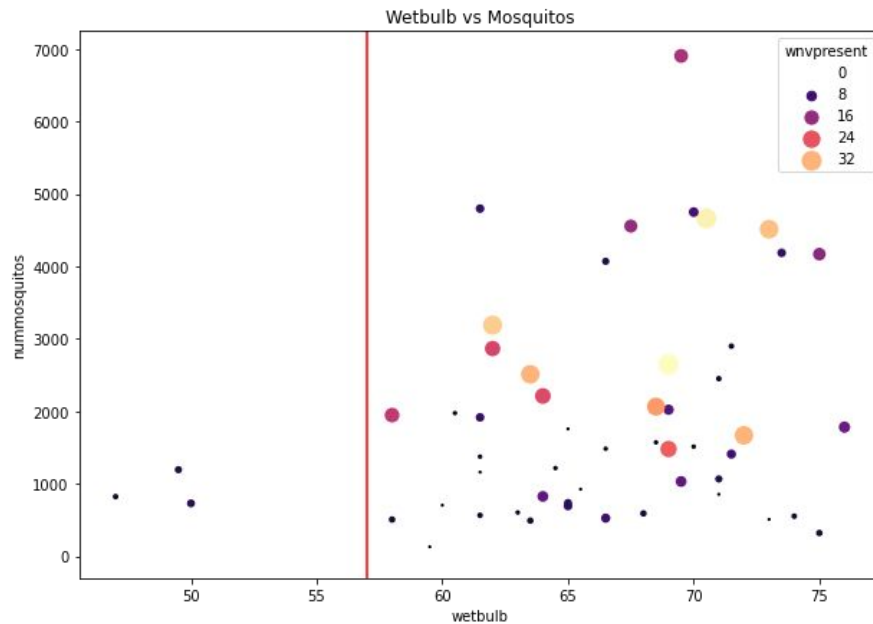


EDA - Wind Speed & Wet bulb (Weather)

From the graph, we can see that number of mosquitos and wnv clusters are more prevalent at **avgspeed** below 11 miles/hour.

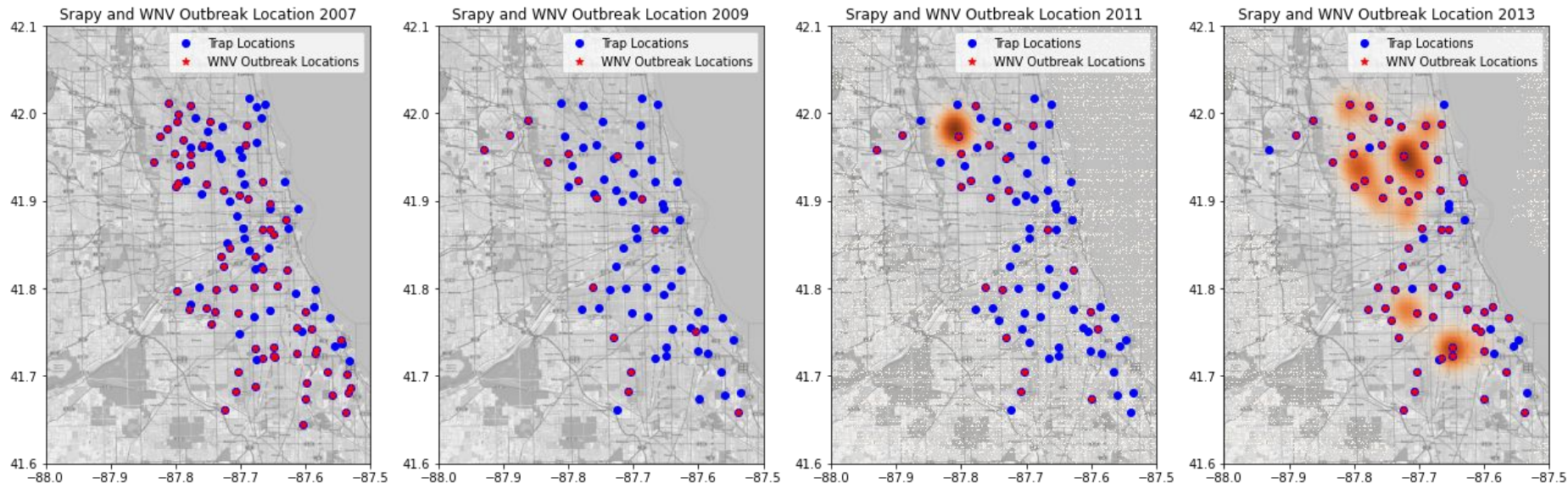


For **WetBulb** above 57°F, we can see that number of mosquitos and west nile clusters are more prevalent.



EDA - Trap / WNV / Spray Location

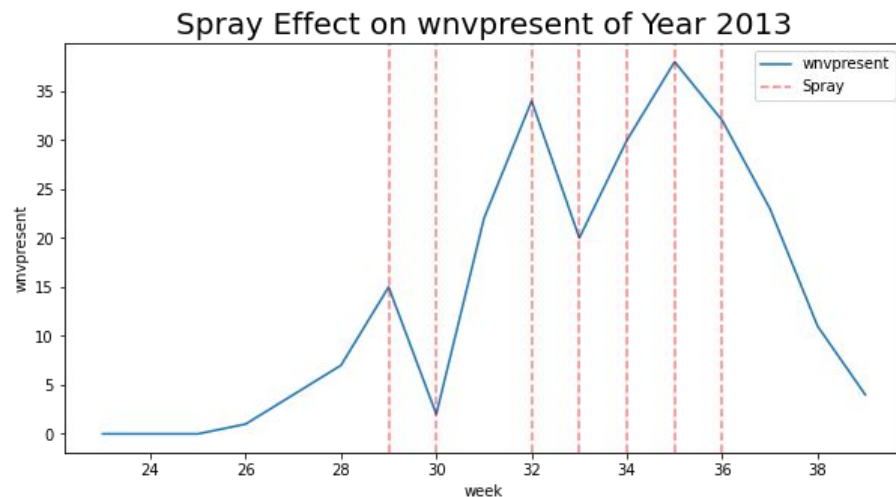
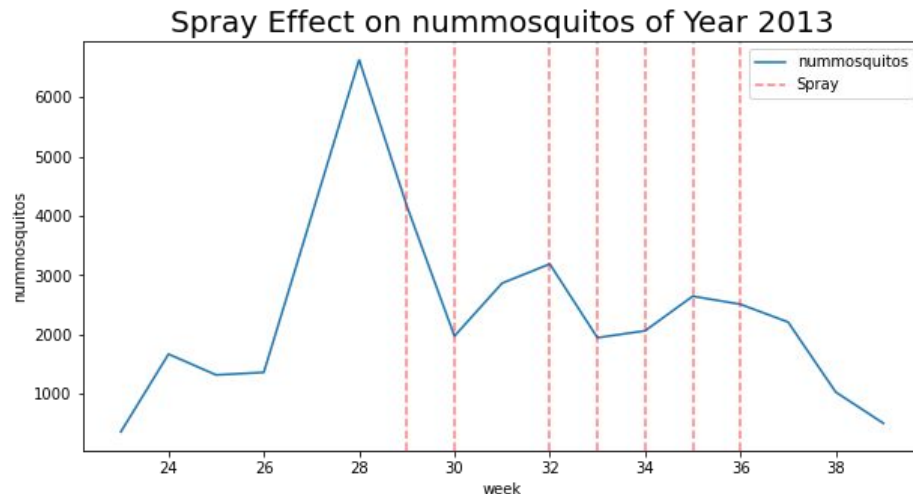
Spray Locations by Year



- In 2007 and 2009, there were no spray data shows in City of Chicago. Spray datasets shows only in 2011, and in 2013, the spraying areas were expanded.

EDA - Spray Effect

- Spray dataset limited, only have 2 years of data where only 2 sprays carried out in 2011.
- Spray does have the effect on reducing number of mosquitos in both year 2011 and 2013, we cannot conclude that it have strong effect on reducing WMV present



Feature Engineering

1

Extract Year / Month / Week / Day

From Date where WNV test was performed

2

Label Encoding Mosquitos Species

Culex pipiens, culex restuans and the rest of species

3

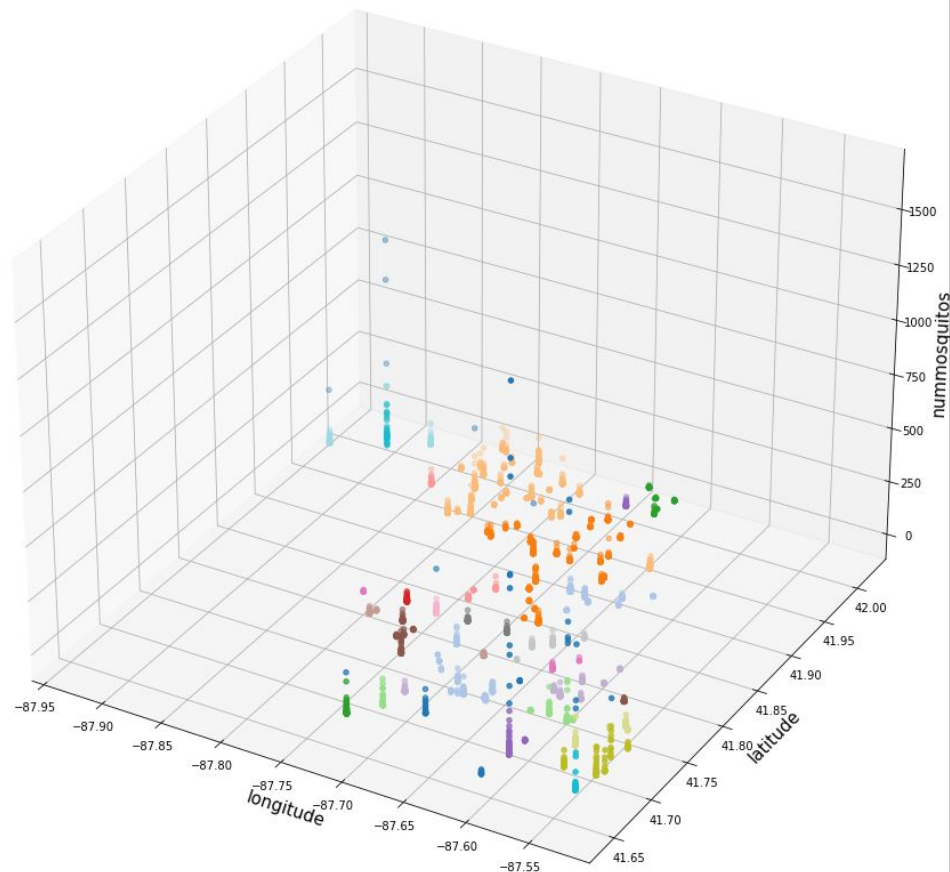
Clustering Trap Locations

Based on Trap Geographic Coordinates and Numbers of Mosquitos, perform DBSCAN clustering

Clustering Trap Location

- ★ There are total of 136 unique trap location in train datasets
- ★ Unsupervised Learning - DBSCAN to cluster **Longitude, Latitude & Numbers of Mosquitos** into different trap clusters
- ★ After our clustering, reduce the trap cluster to 38 for our modeling

3D Scatter Plot for Trap Location and Number of Mosquitos with DBSCAN



Modeling - Using PyCaret

	Data Type
species	Numeric
latitude	Numeric
longitude	Numeric
month	Categorical
week	Categorical
day	Numeric

PyCaret wrongly detects species and day as numeric variables instead of categorical variables so we define our own set of categorical variables below.

```
cat_features = ['species', 'month', 'week', 'day', 'cluster_0', 'cluster_1',  
               'cluster_2', 'cluster_3', 'cluster_4', 'cluster_5', 'cluster_6',  
               'cluster_7', 'cluster_8', 'cluster_9', 'cluster_10', 'cluster_11',  
               'cluster_12', 'cluster_13', 'cluster_14', 'cluster_15', 'cluster_16',  
               'cluster_17', 'cluster_18', 'cluster_19', 'cluster_20', 'cluster_21',  
               'cluster_22', 'cluster_23', 'cluster_24', 'cluster_25', 'cluster_26',  
               'cluster_27', 'cluster_28', 'cluster_29', 'cluster_30', 'cluster_31',  
               'cluster_32', 'cluster_33', 'cluster_34', 'cluster_35', 'cluster_36',  
               'cluster_37']
```


Modeling - PyCaret Results

	Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
lightgbm	Light Gradient Boosting Machine	0.9153	0.8296	0.3442	0.2741	0.3043	0.2601	0.2624	0.7050
xgboost	Extreme Gradient Boosting	0.9162	0.8234	0.3314	0.2728	0.2983	0.2545	0.2562	1.8970
lda	Linear Discriminant Analysis	0.7210	0.8093	0.7547	0.1319	0.2245	0.1471	0.2314	0.3710
gbc	Gradient Boosting Classifier	0.8534	0.8089	0.5201	0.1882	0.2760	0.2144	0.2489	1.9600
lr	Logistic Regression	0.7293	0.8085	0.7318	0.1327	0.2246	0.1476	0.2276	2.6710
ada	Ada Boost Classifier	0.8107	0.7922	0.5605	0.1529	0.2400	0.1706	0.2183	0.5470
rf	Random Forest Classifier	0.9111	0.7743	0.2958	0.2382	0.2630	0.2165	0.2183	0.7930
knn	K Neighbors Classifier	0.7565	0.7304	0.5890	0.1239	0.2047	0.1279	0.1832	0.3070
nb	Naive Bayes	0.4369	0.6845	0.9208	0.0808	0.1485	0.0560	0.1521	0.0820
et	Extra Trees Classifier	0.9107	0.6795	0.2932	0.2349	0.2598	0.2131	0.2151	1.0240
dt	Decision Tree Classifier	0.9102	0.6362	0.2701	0.2253	0.2435	0.1967	0.1986	0.2170
qda	Quadratic Discriminant Analysis	0.3330	0.6332	0.9693	0.0721	0.1342	0.0389	0.1330	0.2310
dummy	Dummy Classifier	0.9467	0.5000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0710
svm	SVM - Linear Kernel	0.6261	0.0000	0.7136	0.1100	0.1766	0.1055	0.1772	0.4670
ridge	Ridge Classifier	0.7214	0.0000	0.7624	0.1330	0.2264	0.1492	0.2350	0.0800

Baseline Model - Dummy Classifier with AUC of 0.5.

Best Model is lightgbm with AUC score of 0.8296

AUC score selected as evaluation as AUC measures the performance of the model at distinguishing between the positive and negative classes.

For this problem, we want to clearly identify the true positive and the true negatives so we optimize AUC.

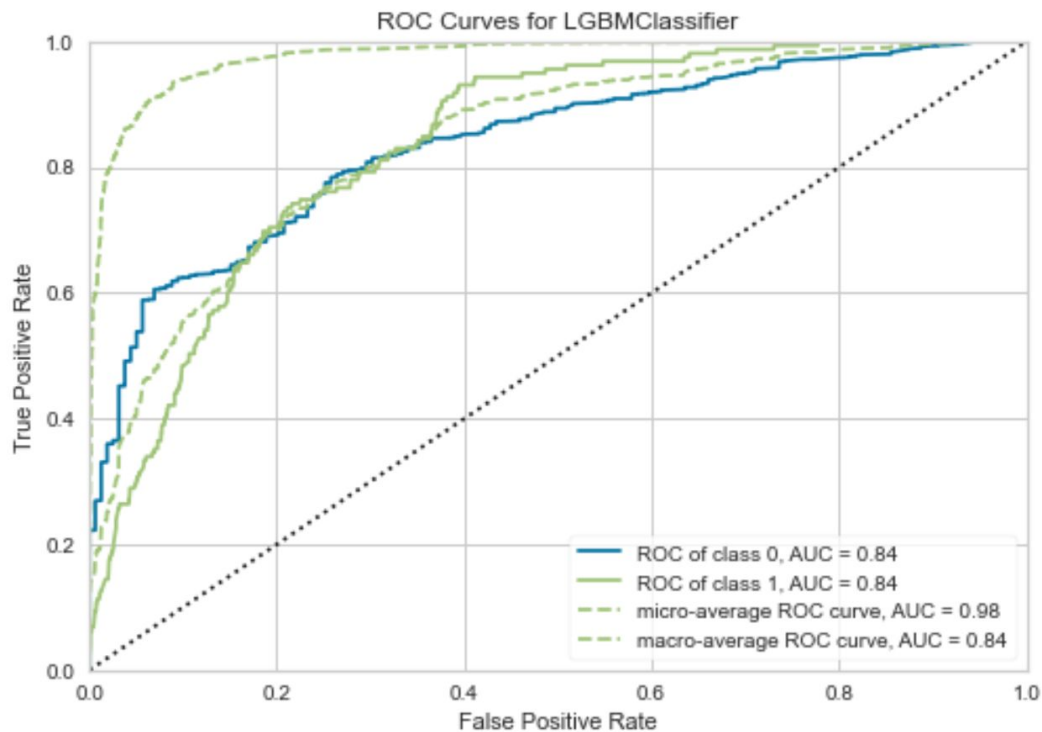
Modeling - Hyperparameter Tuning

```
best = automl(optimize = 'AUC')  
best
```

```
LGBMClassifier(boosting_type='gbdt', class_weight=None, colsample_bytree=1.0,  
                importance_type='split', learning_rate=0.1, max_depth=-1,  
                min_child_samples=20, min_child_weight=0.001, min_split_gain=0.0,  
                n_estimators=100, n_jobs=-1, num_leaves=31, objective=None,  
                random_state=1, reg_alpha=0.0, reg_lambda=0.0, silent='warn',  
                subsample=1.0, subsample_for_bin=200000, subsample_freq=0)
```

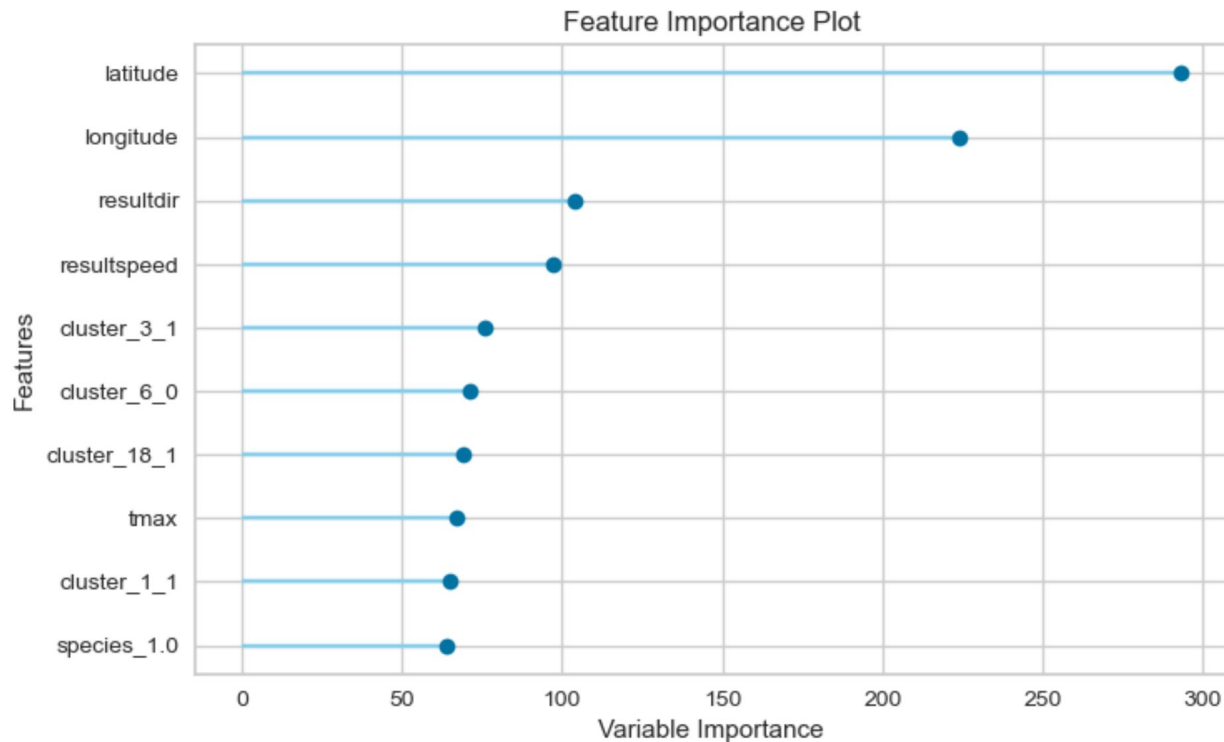
Using automl to perform hyperparameter tuning

Modeling - AUC Graph



AUC Score of 0.84

Modeling - Feature Importance



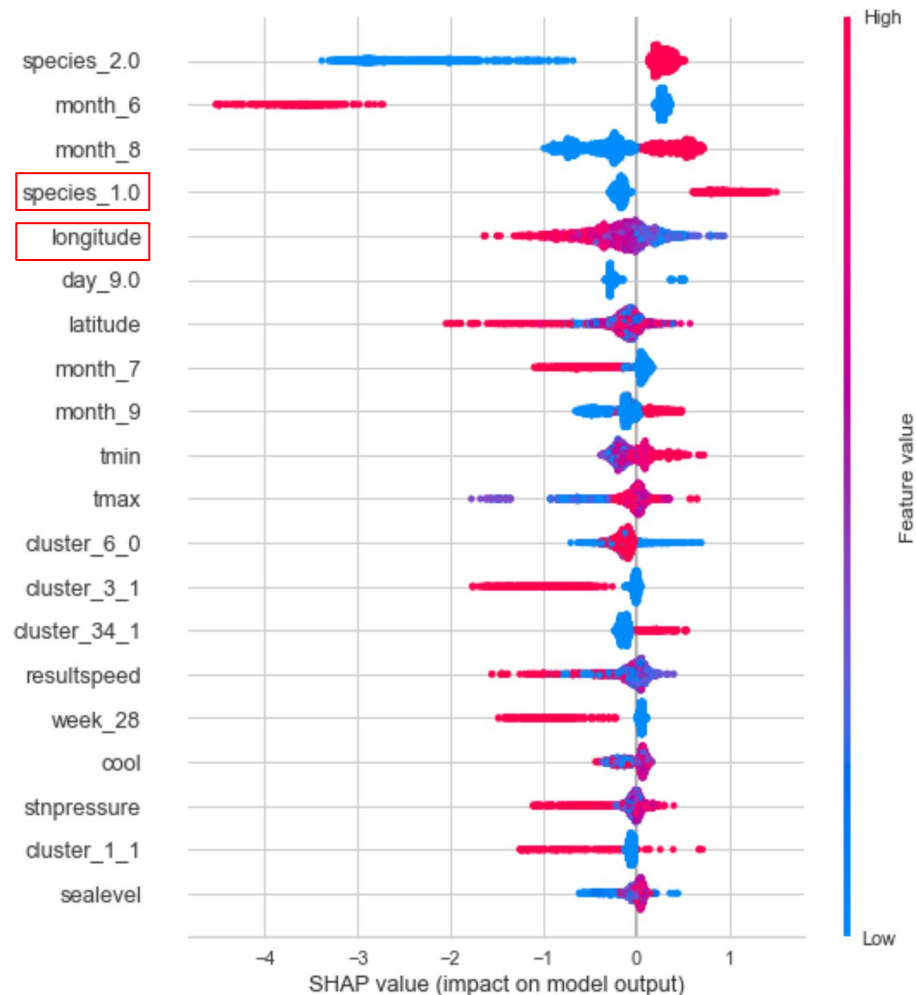
The most important features are latitude and longitude which is the location of the virus.

Other important features include resultdir, resultspeed, tmax, speices_1 and a few clusters.

Modeling - SHAP Values

The most important features are latitude and longitude which is the location of the virus.

Other important features include resultdir, resultspeed, tmax, speices_1 and a few clusters.



The Cost of West Nile Virus

\$800 million



That's how much in hospitalisation and lost productivity the West Nile Virus has cost the USA from 1999 - 2013.

1 in 150



patients with the West Nile Virus that will develop severe symptoms.

\$7,500



The mean hospitalisation and lost productivity cost for mild cases.

\$80,000



The mean hospitalisation and lost productivity cost for severe cases.

Cost Benefit Analysis

Assumptions

Spray Cost

- Zenivex costs \$0.92/acre
- Pest control worker earns \$20/hour
- 8pm - 1am (5 hour spray window)
- 149 traps - all will have spray operations
- 1 worker per trap
- 1km radius spray per trap
- Spray will be 7 times a year

\$840,000

Cost of not Spraying

- All cases are non-severe
- Mean cost for non-severe cases - \$7,500
- 200 additional cases if no spray conducted

\$1,500,000

**\$660,000 savings
annually**

Conclusion and Recommendations

High accuracy of

Conclusion and Recommendations



Use of drones



Expand study to other States

