

# GA Project 4:

## West Nile Virus Carrying Mosquitoes in Chicago

*Presented By:*

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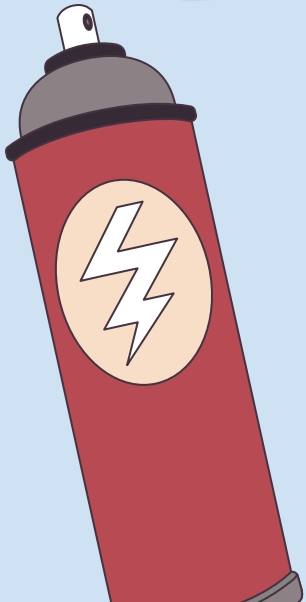
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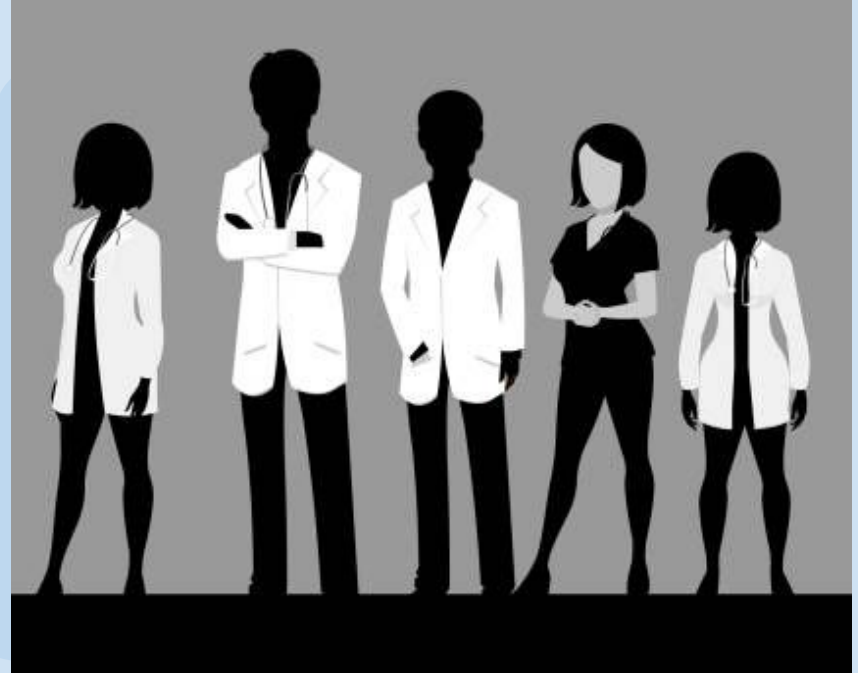
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- Conclusion



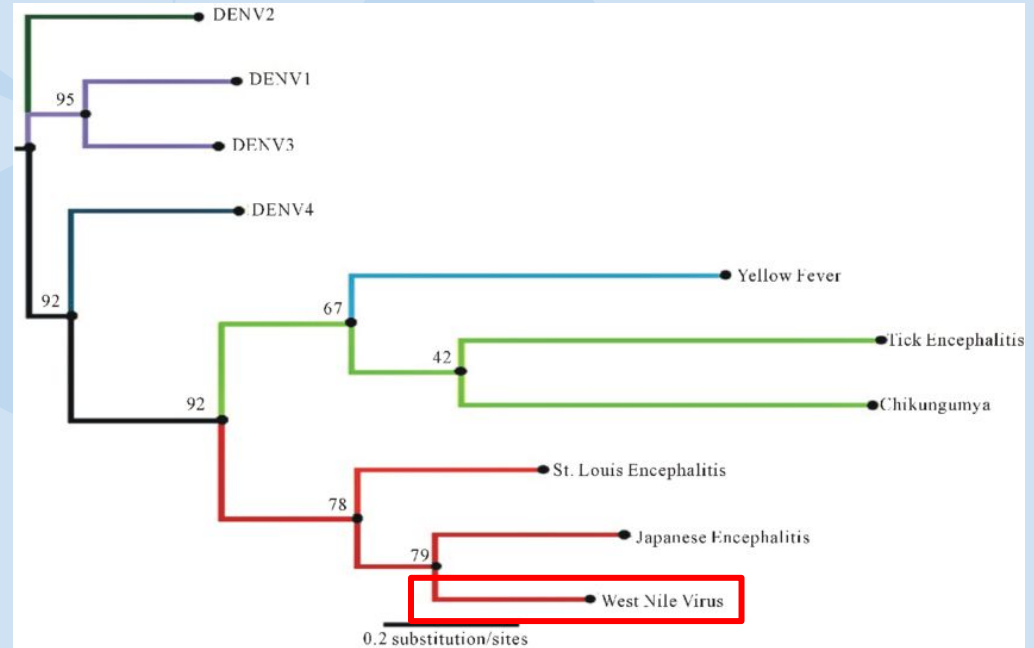
# **BACKGROUND: Who Are We & Problem Statement**

- **Private Consultant to CDC**
  - Environmental Factors affecting WNV transmission
  - Reduce WNV carrying mosquitoes



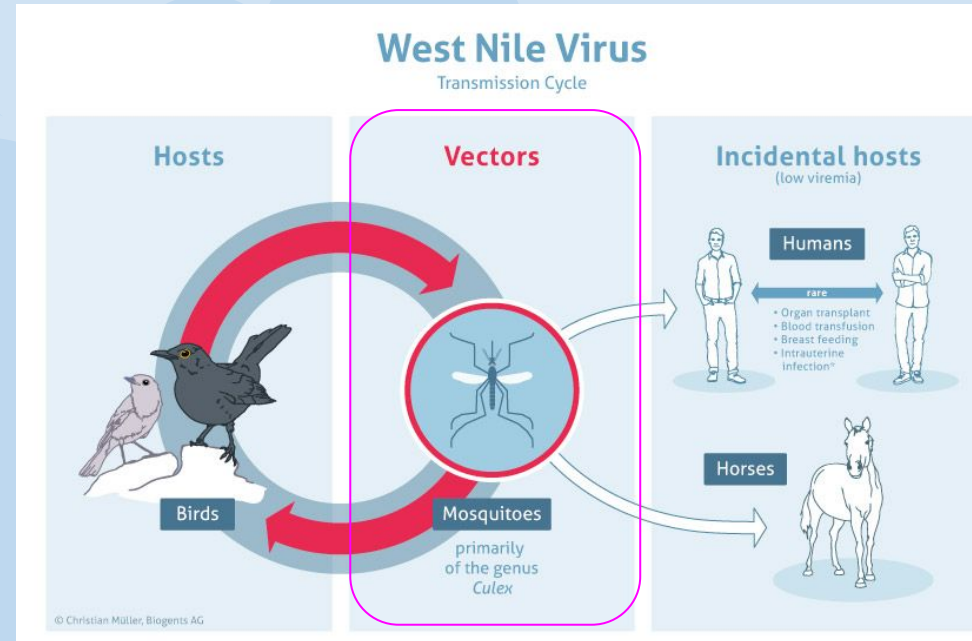
# BACKGROUND: What is West Nile Virus & Resource

- West Nile Virus
  - **Vector**-borne virus
  - Originated in **Africa**
  - **1 in 5** falls **mildly** ill
  - **1 in 150** falls **severely** ill



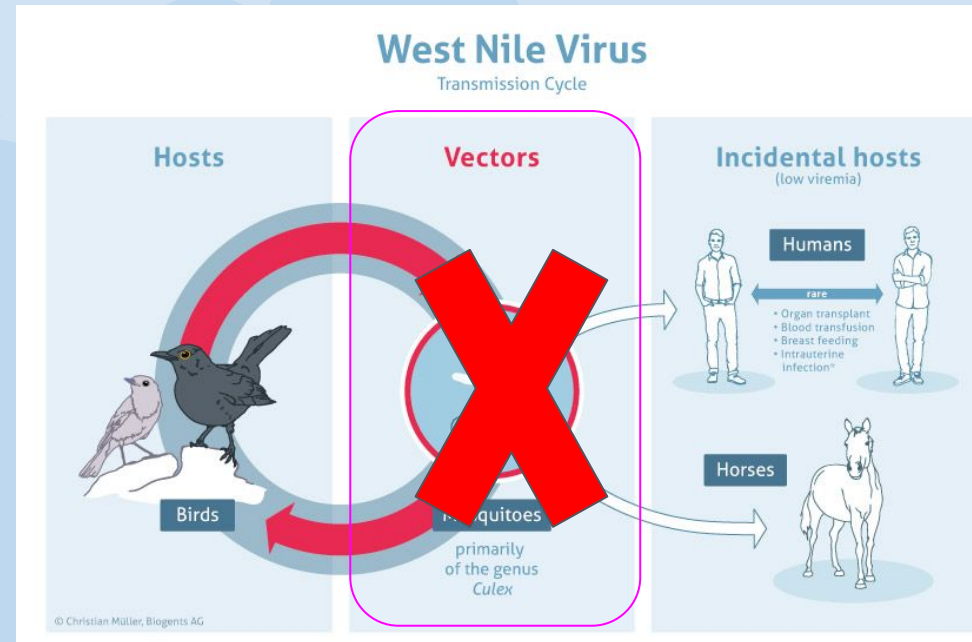
# BACKGROUND: What is West Nile Virus & Resource

- West Nile Virus
  - Vector-borne virus
  - Originated in Africa
  - 1 in 5 falls mildly ill
  - 1 in 150 falls severely ill
- **Prevention** is the best cure



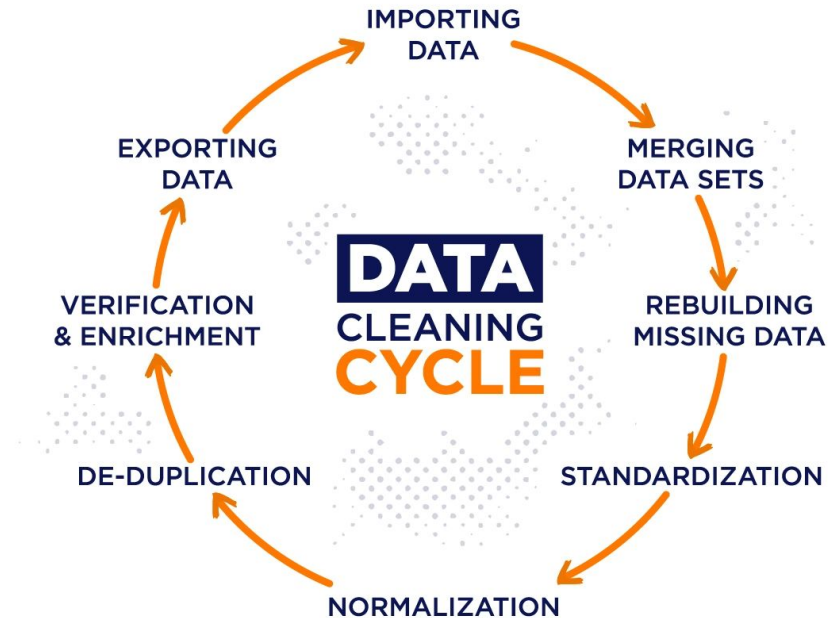
# BACKGROUND: What is West Nile Virus & Resource

- West Nile Virus
  - Vector-borne virus
  - Originated in Africa
  - 1 in 5 falls mildly ill
  - 1 in 150 falls severely ill
- Prevention is the best cure
- Resource and Data from in-house recordings from **2007 up to 2014**



# DATA CLEANING: Filtering / Removing / Replacing / Merge

- Determine Relevant Columns
- Clean nulls, na and drop by:
  - Impute
  - Replace
  - Drop
  - Remove Duplicates
- Merge Train, Spray and Weather



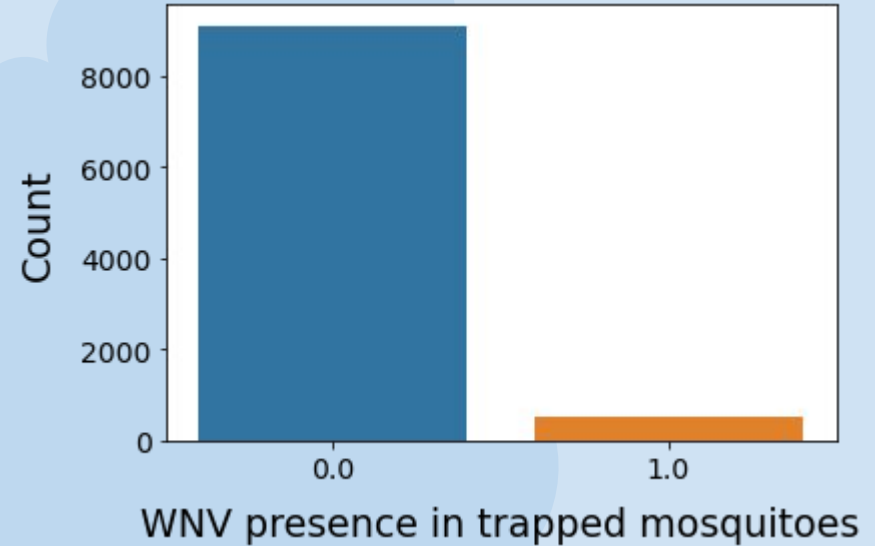


**EDA**



# Unbalanced Class

- Imbalanced Data
- Use **SMOTE** and **ADASYN** at modelling stage
- Minimize False Negatives
- Minimize False Positives



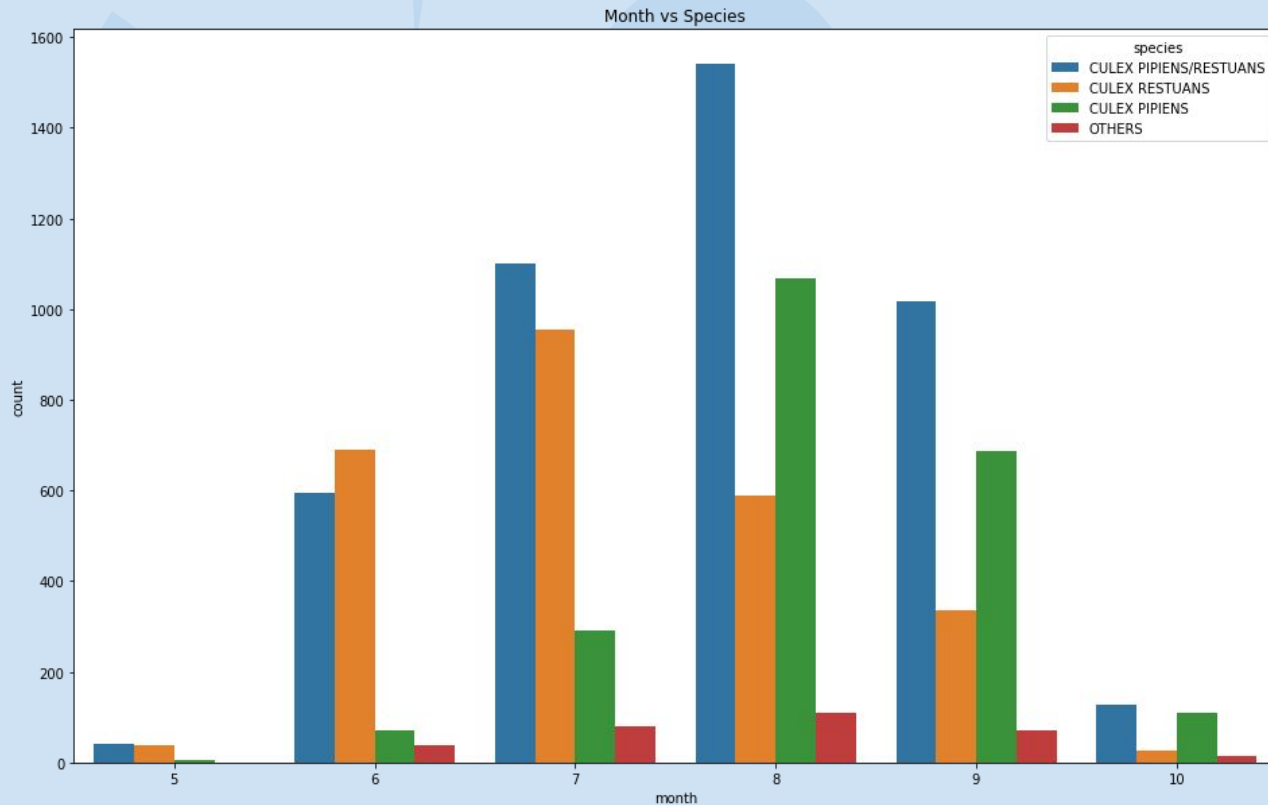
Key:

0 = Not West Nile Virus Carrying Mosquitoes

1 = West Nile Virus Carrying Mosquitoes

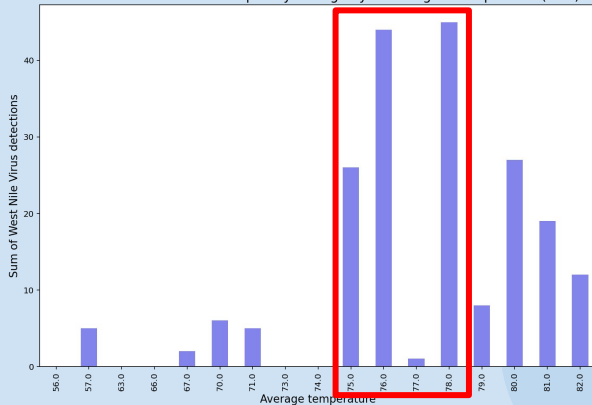
# Mosquito Population vs Months

- Peaks around **August** (hottest month)
- Blue, Orange and Green represents the mosquitoes that **carriers** of **WNV**
- Other species were combined together

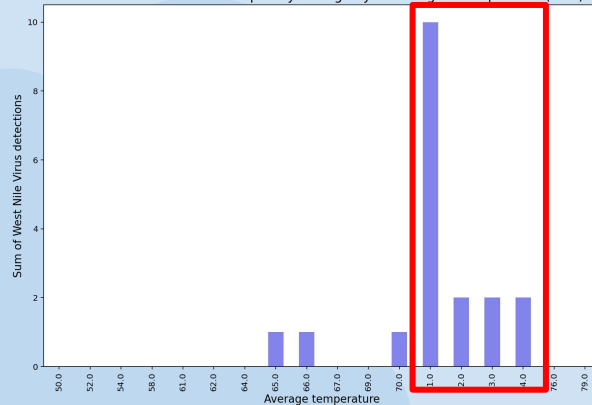


# WNV Mosquito Frequency vs Temperature

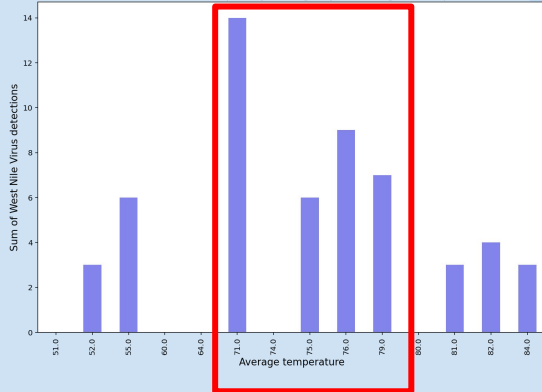
West Nile Virus is detected more frequently during days with higher temperature(>50) in 2007



West Nile Virus is detected more frequently during days with higher temperature(>50) in 2009



West Nile Virus is detected more frequently during days with higher temperature(>50) in 2011

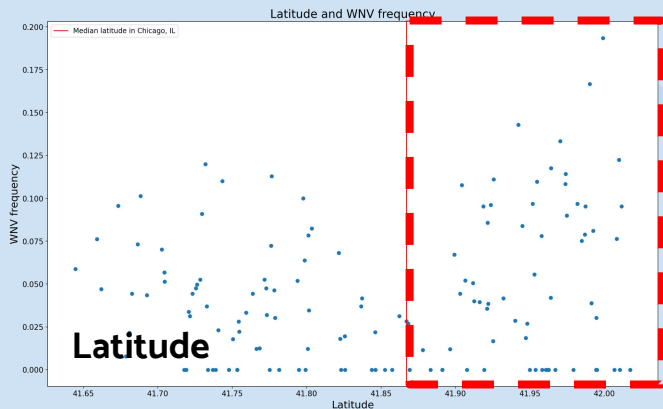


West Nile Virus is detected more frequently during days with higher temperature(>50) in 2013

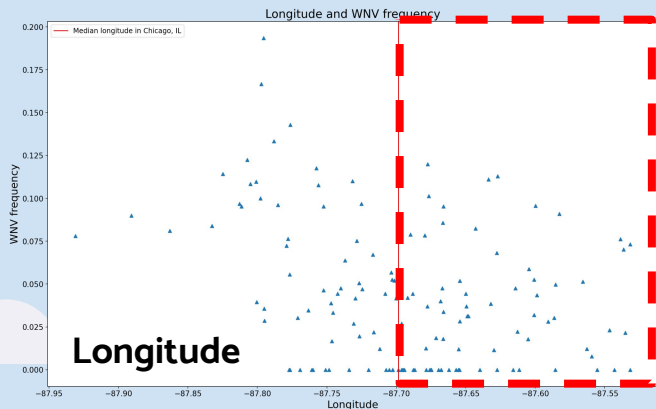


# Majority of the Population

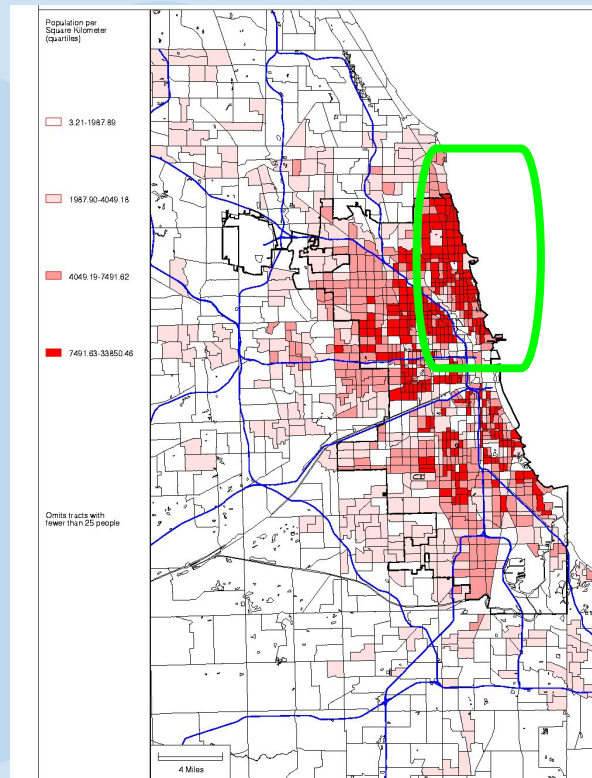
WNV Frequency



Higher density in Northern region



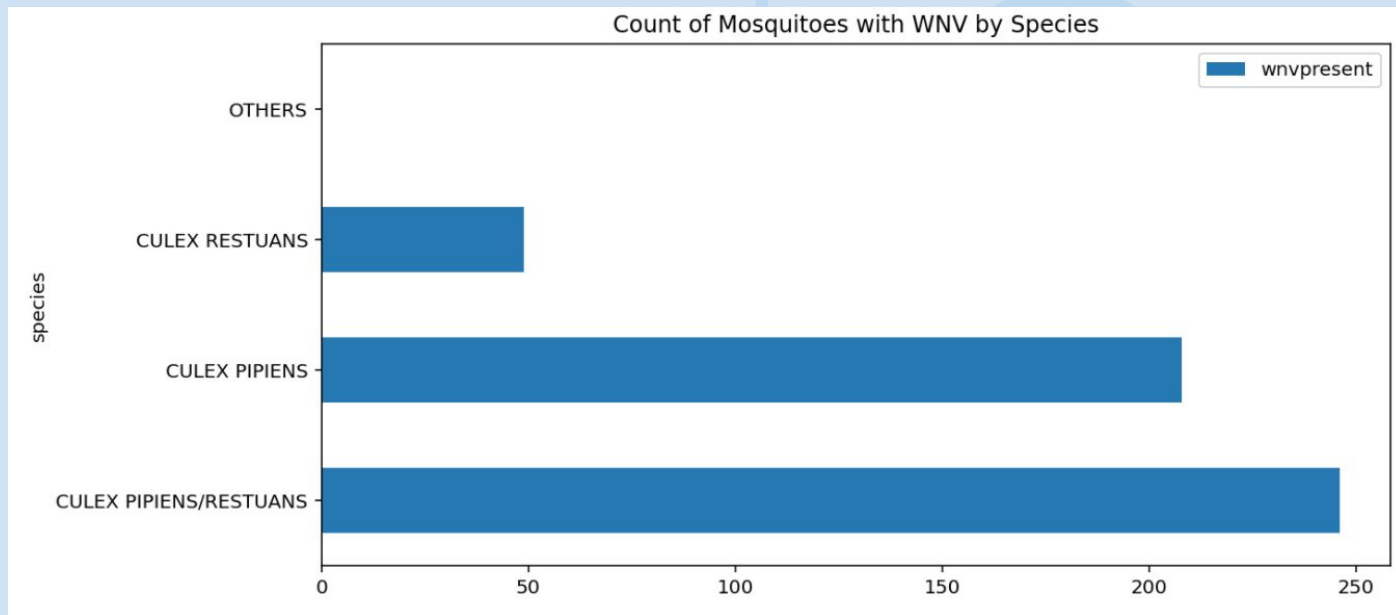
Higher density in Eastern region



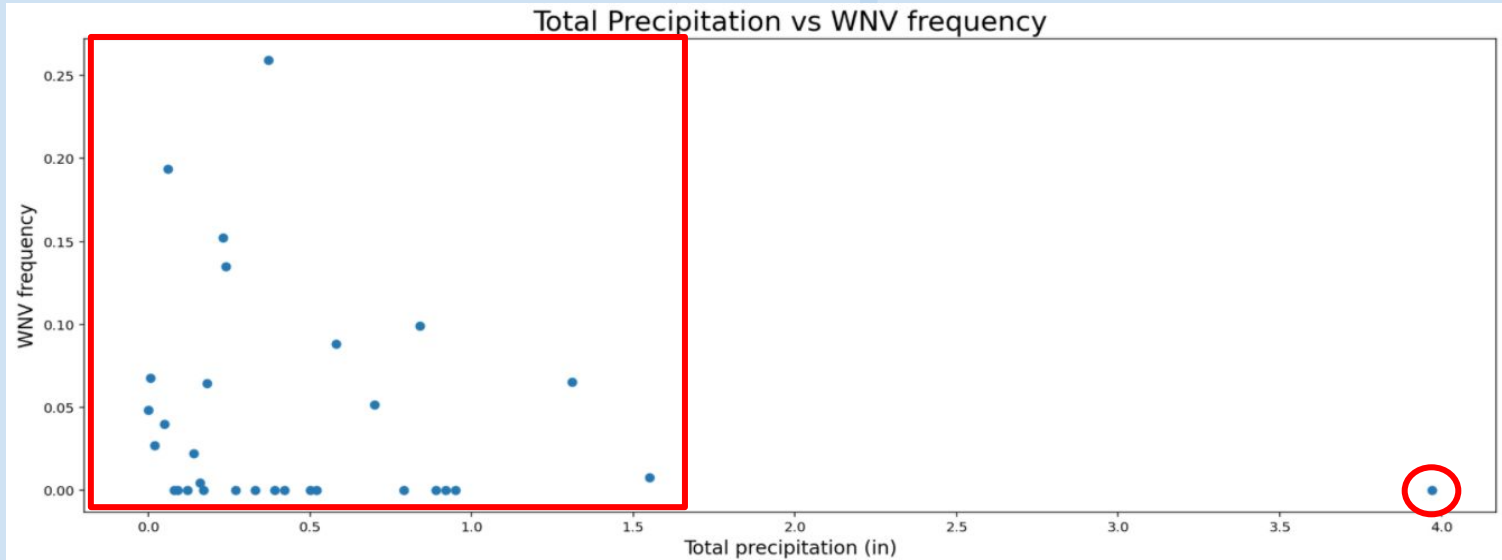
Human Population Density

University of Chicago Map Collection, July 1996

# Mosquito Species with WNV



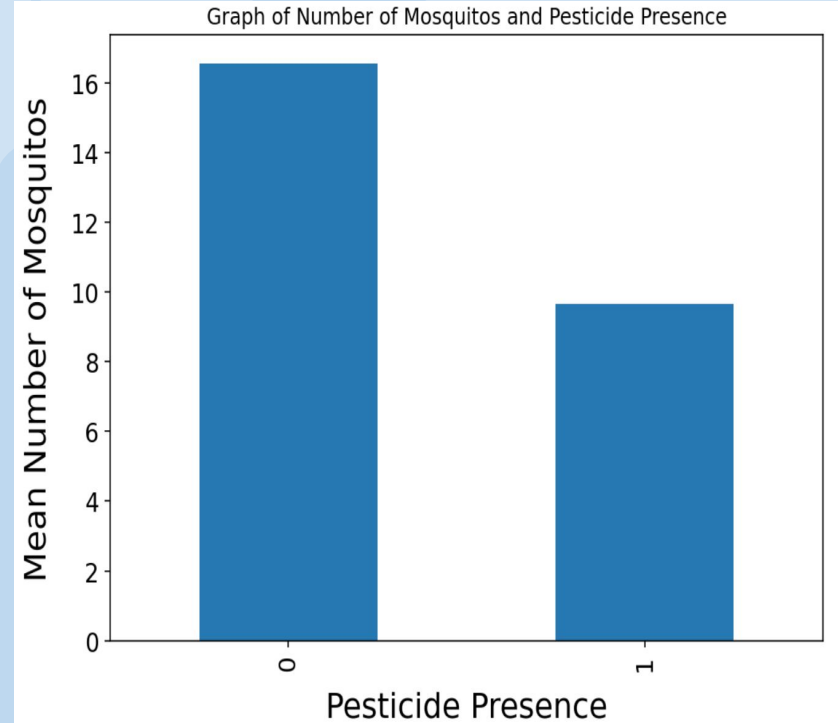
# Total Precipitation and WNV



- Very little precipitation required for the mosquitos to breed.
- Excessive rainfall can cause breeding sites to overflow, disrupting mosquito breeding and destroying developing larvae.
- They could also be breeding in other sources of water such as in flower pots.

# Effect of Pesticides on Mosquitos

- **Decrease** in number of mosquitos with of pesticides
- But **not as significant** as we would expect from using pesticides



# Type of Pesticide Use Matters

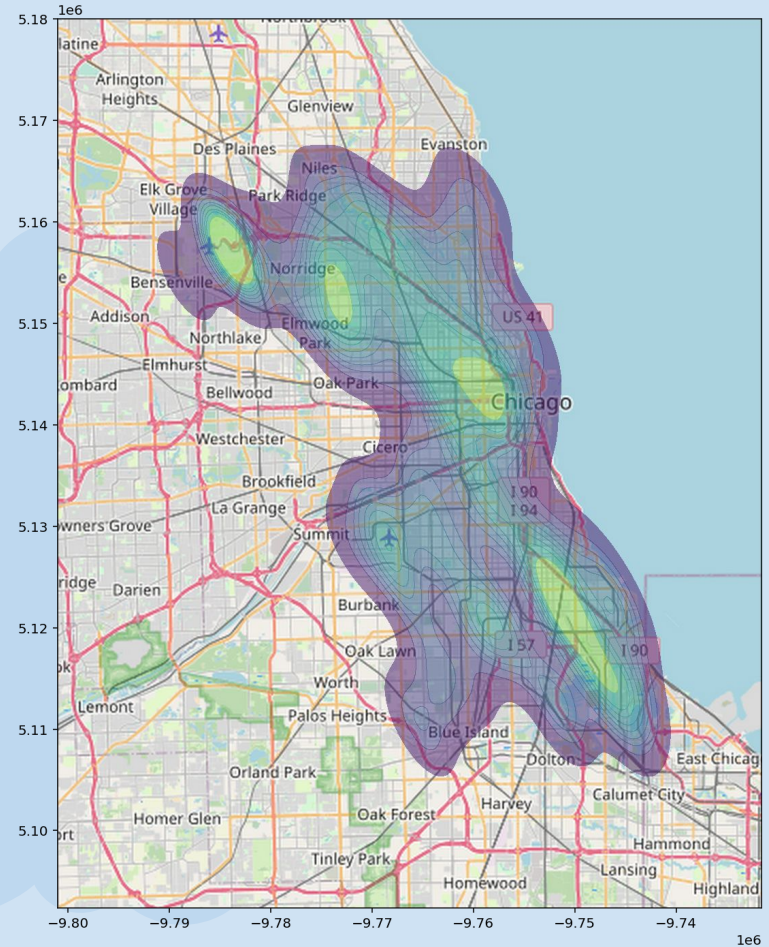
- Zenivex is an **Adulticide**
- **Least effective** mosquito control technique
- Programs spray **indiscriminately**





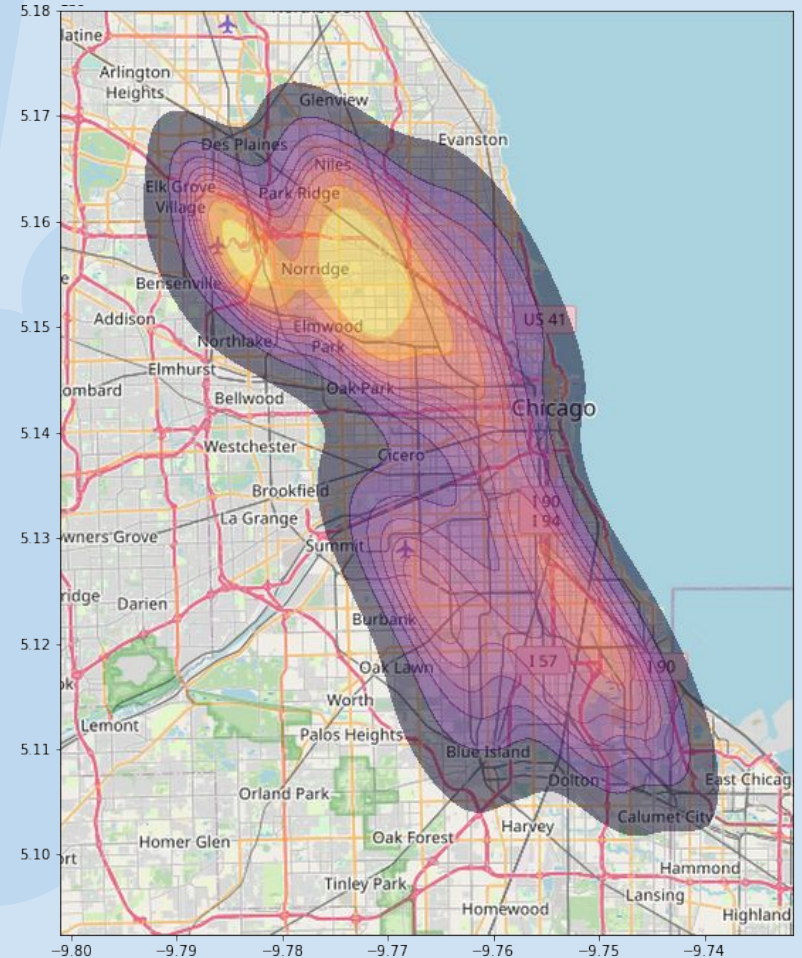
# Mosquito Density

- Light green represents high concentration of Mosquitoes
- Majority located near the coastline



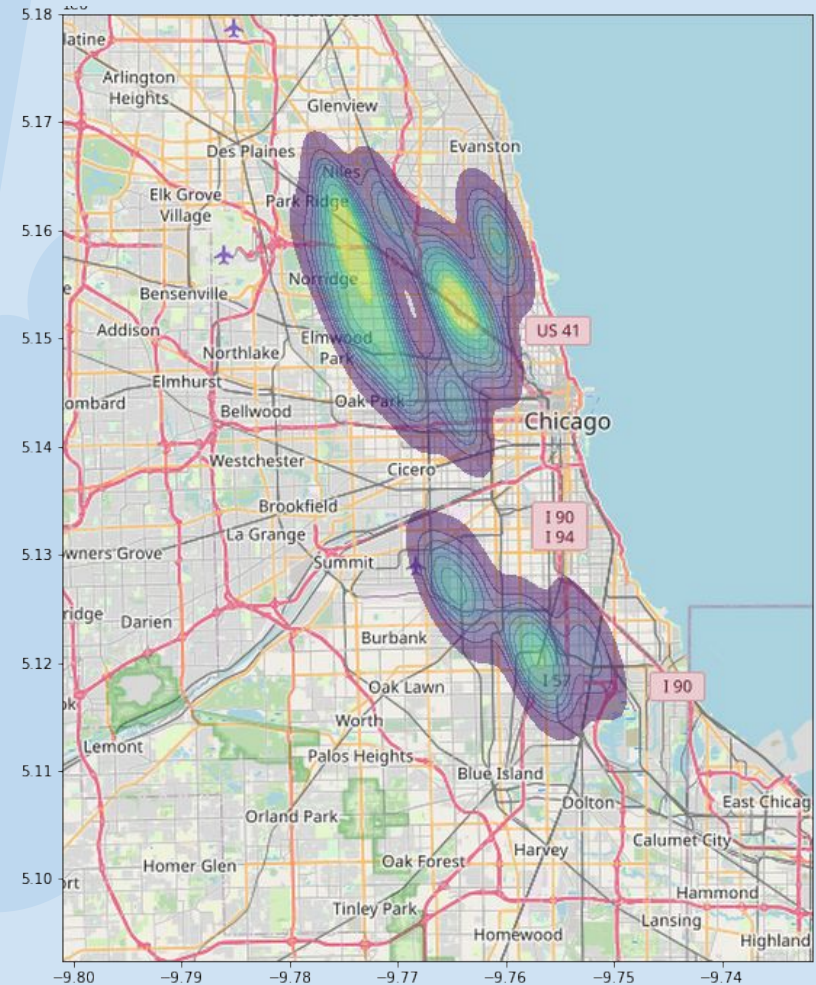
# WNV Mosquito Density

- **Two** main hotspots for WNV mosquitoes



# Spray Location

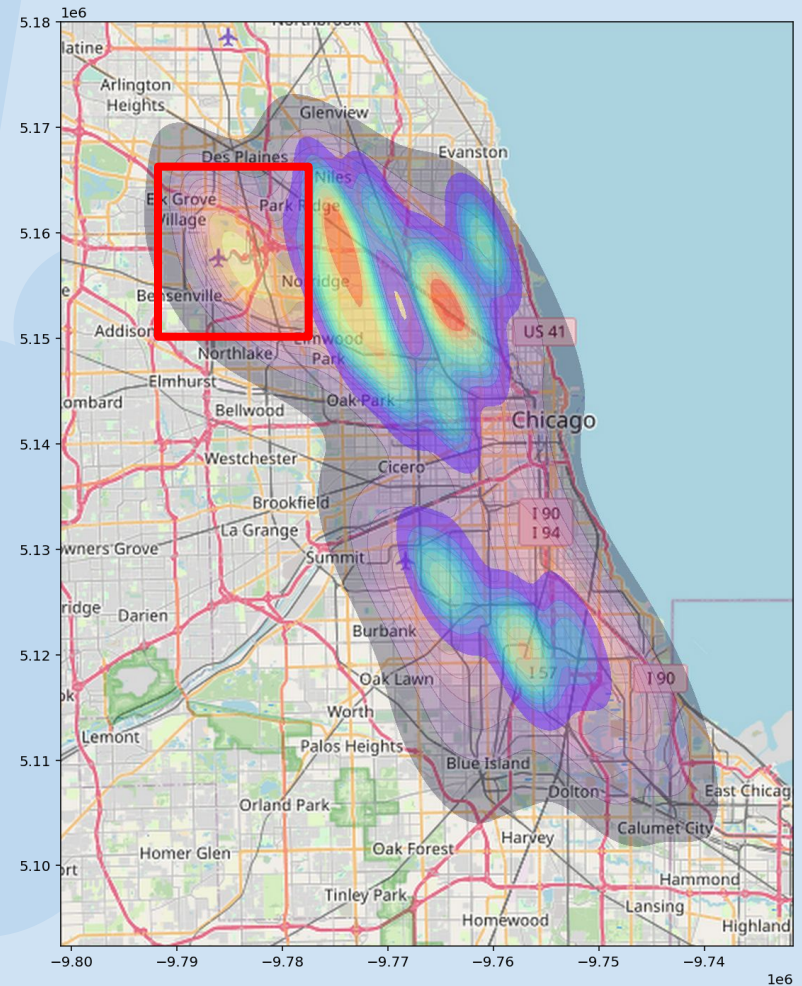
- Concentrated around **2 main areas**
- Density of spray increases in the yellow regions



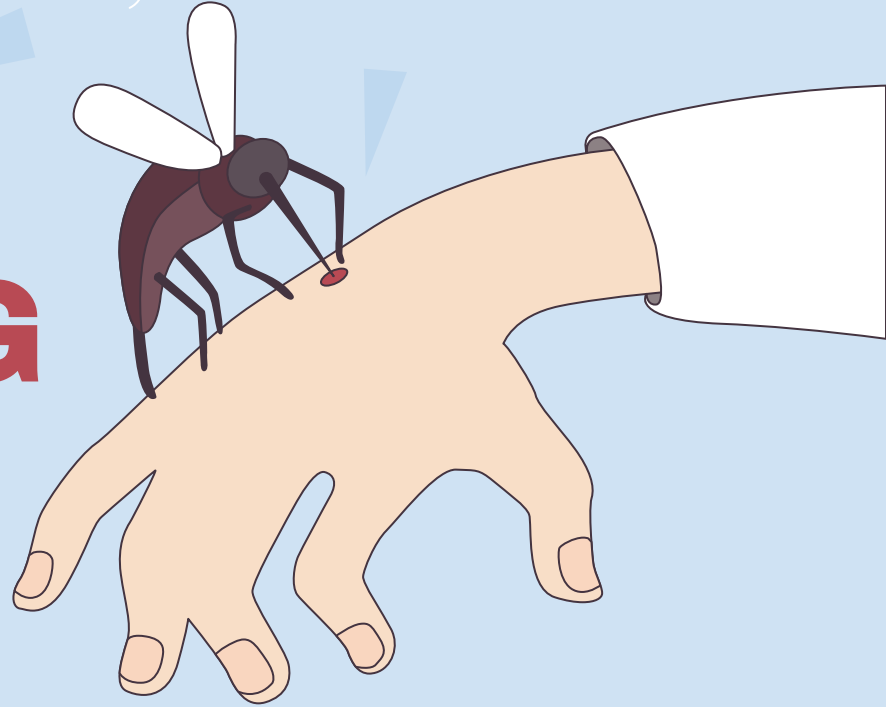


# Overlap of WNV Mosquito and Spray Locations

- Most of the areas currently being sprayed with pesticide are also areas with WNV
- There is **one area** (in red) with high WNV density that is **not covered by pesticide**
- Recommended to include that area for pesticide spraying



# MODELLING



# Models & Resamplers

## Models:

- Naive Bayes
- Logistic Regression
- K-Nearest Neighbor
- Random Forest
- Extra Tree
- Decision Tree
- ADA Boost
- Gradient Boost
- XGB
- Light GB
- SVM

## Resamplers:

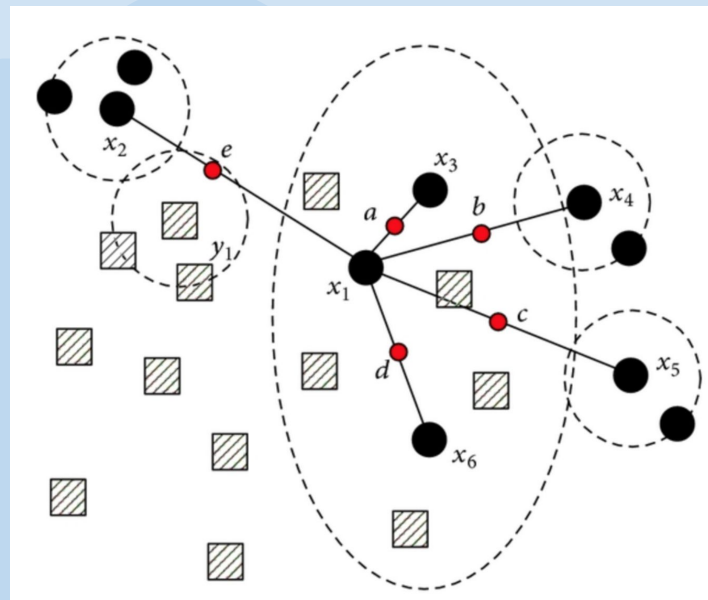
- SMOTE
- ADASYN



# SMOTE

## Synthetic Minority Over sampling Technique

- K-Nearest Neighbours approach
- Draws a line between the neighbours of the minority class
- Generates random points on the lines

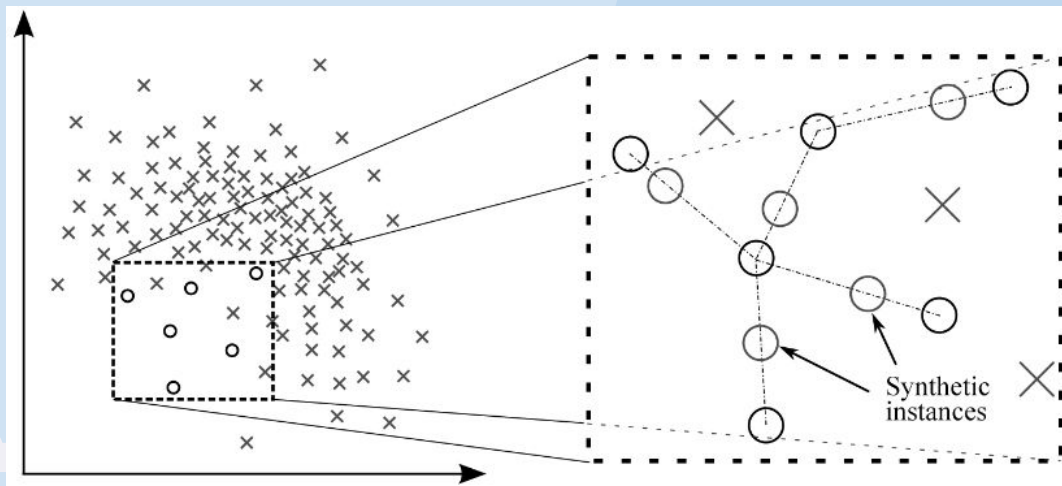


- ▨ Majority class samples
- Minority class samples
- Synthetic samples

# ADASYN

## ADaptive SYNthetic

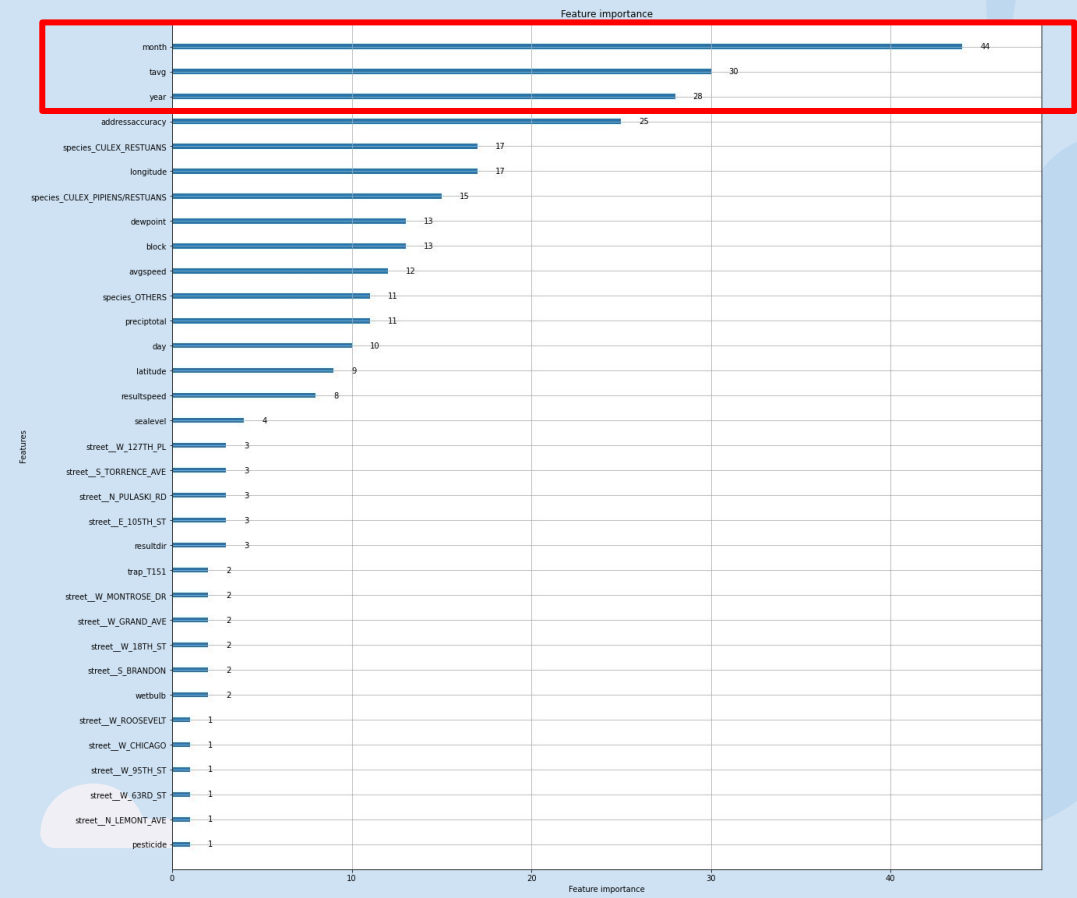
- K-Nearest Neighbours approach
- no assumptions made for the underlying distribution of the data
- Difference: generates the same number of synthetic samples for each original minority sample





Classifier	CV Score	ROC_AUC (train)	ROC_AUC (test)	Accuracy (train)	Accuracy (test)	Sensitivity	Specificity	Precision	Recall	Misclassification	F1 Score
Gaussian Naive Bayes	0.6838	0.77	0.67	0.77	0.67	0.76	0.51	0.08	0.76	0.48	0.1447619
Gaussian Naive Bayes (SMOTE)	0.6753	0.72	0.64	0.72	0.64	0.92	0.23	0.06	0.92	0.73	0.11265306
Gaussian Naive Bayes (ADASYN)	0.6329	0.69	0.6	0.69	0.6	0.86	0.25	0.06	0.86	0.72	0.11217391
Logistic Regression	0.76917	0.83	0.77	0.83	0.77	0	1	0	0	0.05	0
Logistic Regression (SMOTE)	0.7542	0.8	0.75	0.8	0.75	0.49	0.83	0.14	0.49	0.19	0.21777778
Logistic Regression (ADASYN)	0.752	0.79	0.75	0.79	0.75	0.44	0.84	0.13	0.44	0.18	0.20070175
Random Forest	0.8222	0.94	0.83	0.94	0.83	0	1	0	0	0.05	0
Random Forest (SMOTE)	0.8212	0.92	0.84	0.92	0.84	0.61	0.87	0.2	0.61	0.14	0.30123457
Random Forest (ADASYN)	0.8211	0.92	0.83	0.92	0.83	0.62	0.87	0.21	0.62	0.14	0.31373494
Decision Tree	0.6687	0.68	0.65	0.68	0.65	0	1	0	0	0.05	0
Decision Tree (SMOTE)	0.6871	0.64	0.61	0.64	0.61	0.91	0.33	0.07	0.91	0.64	0.13
Decision Tree (ADASYN)	0.7082	0.76	0.72	0.76	0.72	0.75	0.59	0.09	0.75	0.4	0.16071429
Extra Trees	0.8026	0.96	0.81	0.96	0.81	0	1	0	0	0.05	0
Extra Trees (SMOTE)	0.7959	0.92	0.79	0.92	0.79	0.46	0.87	0.17	0.46	0.15	0.24825397
Extra Trees (ADASYN)	0.795	0.91	0.8	0.91	0.8	0.49	0.87	0.17	0.49	0.15	0.25242424
Light Gradient Boost	0.8302	0.87	0.83	0.87	0.83	0	1	0	0	0.05	0
Light Gradient Boost (SMOTE)	0.8195	0.87	0.82	0.87	0.82	0.62	0.86	0.2	0.62	0.15	0.30243902
Light Gradient Boost (ADASYN)	0.8191	0.87	0.83	0.87	0.83	0.56	0.87	0.19	0.56	0.15	0.28373333
K-Nearest Neighbours	0.7165	0.93	0.73	0.93	0.73	0.05	0.99	0.35	0.05	0.05	0.0875
K-Nearest Neighbours (SMOTE)	0.7498	0.93	0.76	0.93	0.76	0.79	0.66	0.11	0.79	0.34	0.19311111
K-Nearest Neighbours (ADASYN)	0.7443	0.93	0.73	0.93	0.76	0.78	0.66	0.11	0.78	0.33	0.19280899
Gradient Boosting	0.8371	0.92	0.85	0.92	0.85	0.02	1	0.25	0.02	0.05	0.03703704
Gradient Boosting (SMOTE)	0.8239	0.91	0.84	0.91	0.84	0.38	0.93	0.22	0.38	0.1	0.27866667
Gradient Boosting (ADASYN)	0.8233	0.91	0.84	0.91	0.84	0.41	0.92	0.22	0.41	0.11	0.28634921
XG Boost	0.8354	0.9	0.85	0.9	0.85	0.04	1	0.46	0.04	0.05	0.0736
XG Boost (SMOTE)	0.8209	0.87	0.83	0.87	0.83	0.62	0.86	0.2	0.62	0.15	0.30243902
XG Boost (ADASYN)	0.8211	0.88	0.83	0.88	0.83	0.53	0.89	0.21	0.53	0.13	0.30081081
SVM	0.7806	0.97	0.8	0.97	0.8	0	1	0	0	0.05	0
SVM (SMOTE)	0.8136	0.88	0.82	0.88	0.82	0.69	0.84	0.19	0.69	0.17	0.29795455
SVM (ADASYN)	0.8143	0.88	0.82	0.88	0.82	0.67	0.84	0.19	0.67	0.17	0.29604651
ADABOOST	0.7765	0.91	0.79	0.91	0.79	0.05	0.99	0.3	0.05	0.06	0.08571429
ADABOOST (SMOTE)	0.7813	0.89	0.8	0.89	0.8	0.46	0.88	0.17	0.46	0.15	0.24825397
ADABOOST (ADASYN)	0.7831	0.89	0.79	0.89	0.79	0.51	0.88	0.19	0.51	0.14	0.27685714

# Best Features



Top 3 features:

Month

Tavg

Year



# PARAMETERS: Confusion Matrix

		(+)	(-)
Real Label	(+)	TP	FN
	(-)	FP	TN
		Predicted Label	

# PARAMETERS: Confusion Matrix

Predicted: ✓ WNV  
Actual: ✓ WNV

(+)

TP

FN

Predicted: ✗ WNV  
Actual: ✓ WNV

Predicted: ✓ WNV  
Actual: ✗ WNV

(-)

FP

TN

Predicted: ✗ WNV  
Actual: ✗ WNV



# PARAMETERS: Confusion Matrix

Predicted: ✓ WNV  
Actual: ✓ WNV

(+)

2234

496

Predicted: ✗ WNV  
Actual: ✓ WNV

Predicted: ✓ WNV  
Actual: ✗ WNV

(-)

48

103

Predicted: ✗ WNV  
Actual: ✗ WNV



# Classifier Evaluation Metric

## Consequences

Precision	$TP/(TP+TN)$
Recall	$TP/(TP+FN)$
Sensitivity	$TP/(TP+FP)$
Specificity	$TN/(TN+FN)$

FP



FN



# Best Model

Light GB with SMOTE

- AUC (Train): 0.87
- AUC (Test): 0.82
- Accuracy: 0.82
- Sensitivity: 0.62
- Specificity: 0.86
- Precision: 0.20
- Recall: 0.62
- F1 Score:

**Real  
Label**

	(+)	(-)
(+)	2234	496
(-)	48	103

**Predicted  
Label**



# Conversion for Cost-Benefit Analysis

Confusion Matrix (Number)



Confusion Matrix (Proportion)

		(+)	(-)
Real	(+)	2234	496
	(-)	48	103
		Predicted	

		(+)	(-)
Real	(+)	0.775	0.172
	(-)	0.017	0.036
		Predicted	



### Confusion Matrix (Costs)

		(+)	(-)
Real	(+)	$\$78.85 + (21.7 * 0.281 * 1000)$	$\$21.7 * 7.31 * 1000$
	(-)	\$ 78.85	0
		Predicted	

### Confusion Matrix (Costs)

		(+)	(-)
Real	(+)	<b>\$6176.51</b>	<b>0.172</b>
	(-)	<b>\$6176.51</b>	<b>0.036</b>
		Predicted	

Overall Total Cost = \$ 32,085.55 per spray

Expected total cost per year = \$US 802,138.75

# **COST BENEFIT ANALYSIS**



# **COST**

## **Indirect Cost**

- Higher Tax Rates

## **Direct Cost**

- Cost of Pesticides
- Other Miscellaneous Cost

## **Intangible Cost**

- Lower Quality of Life



# DIRECT COST

## Direct Cost

- Cost of Pesticides
- Other Miscellaneous Cost

**69km<sup>2</sup>**

amount spent on aerial  
spraying in Chicago in 2020



**USD797k**

amount spent on spraying in  
Chicago in 2020

# INDIRECT COST

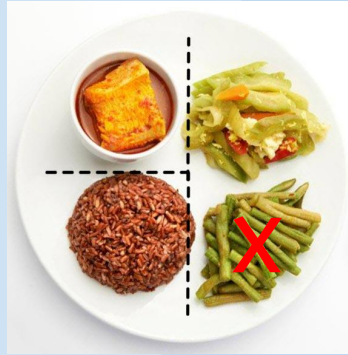
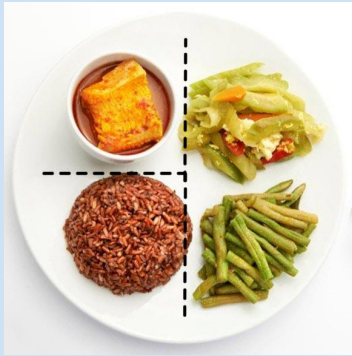
## Indirect Cost

- Higher Tax Rates

- Shared cost of pesticides borne by both the state Illinois, Chicago city, and the residents of Chicago
- Eventual increase in higher taxes may be a huge **financial burden**
  - Especially those of the **lower-income group**.

# INTANGIBLE COST

- Lower Income Group



## Intangible Cost

- Lower Quality of Life



# BENEFITS

**Direct  
Benefits**

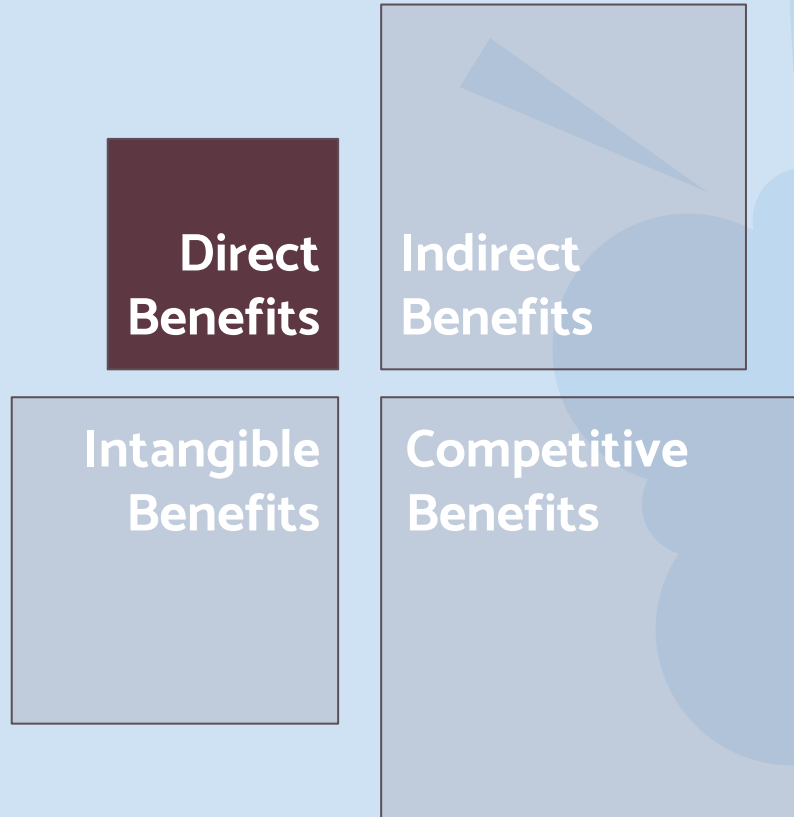
**Indirect  
Benefits**

**Intangible  
Benefits**

**Competitive  
Benefits**



# BENEFITS



Decrease in probability:

- People contracting the West Nile Virus
- Unemployed residents in Chicago
  - Due to contracting the West Nile Virus or,
  - Being caregivers to the patients





# BENEFITS

Direct  
Benefits

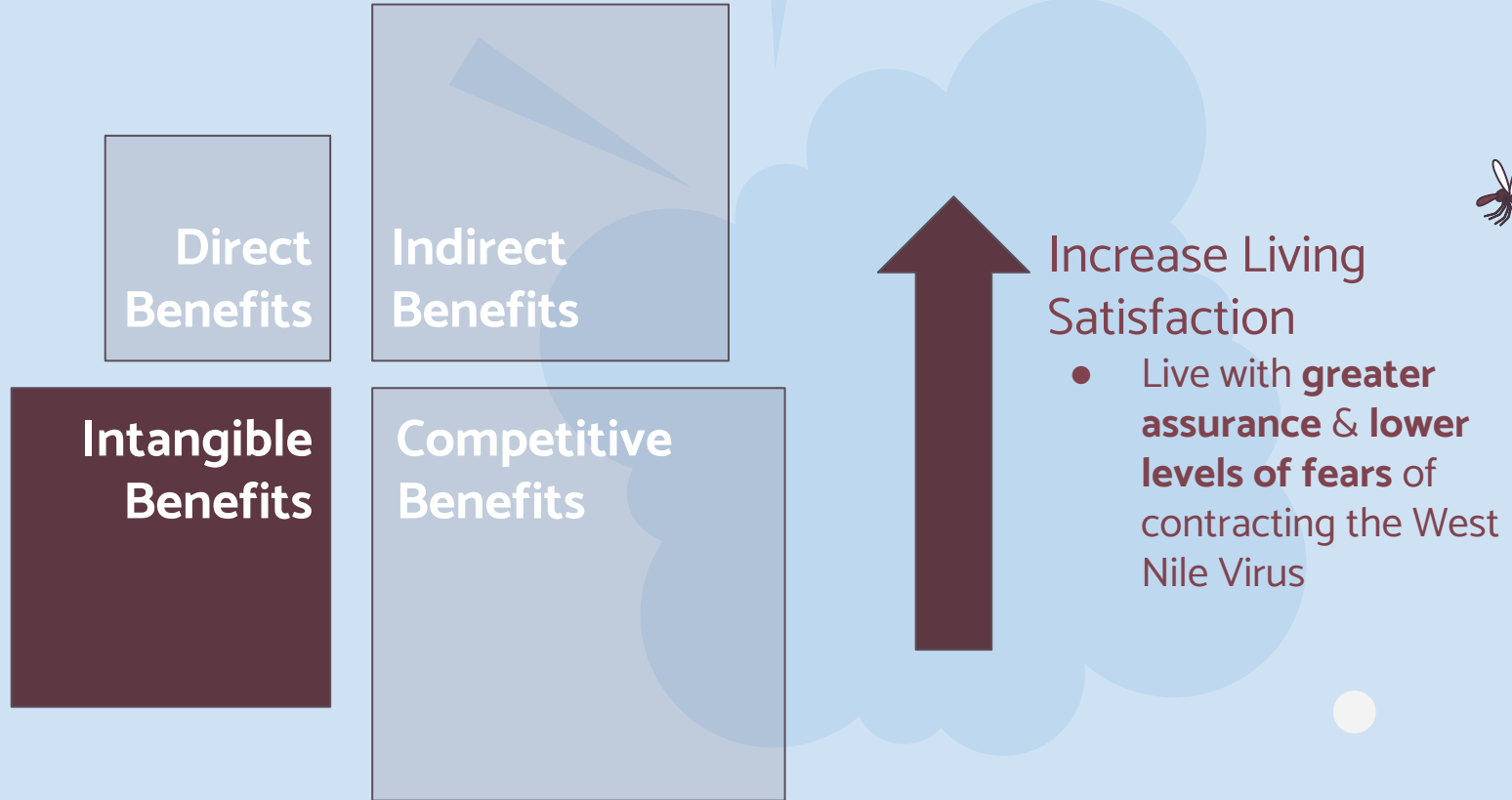
Indirect  
Benefits

Intangible  
Benefits

Competitive  
Benefits

- Decrease in Medical Fees
- Reduce stimulus cheque for patients who cannot work

# BENEFITS



# BENEFITS

Direct  
Benefits

Indirect  
Benefits

Intangible  
Benefits

Competitive  
Benefits

- Stand out from her neighbouring cities
- Promote as a **choice destination** for a safe summer vacation.



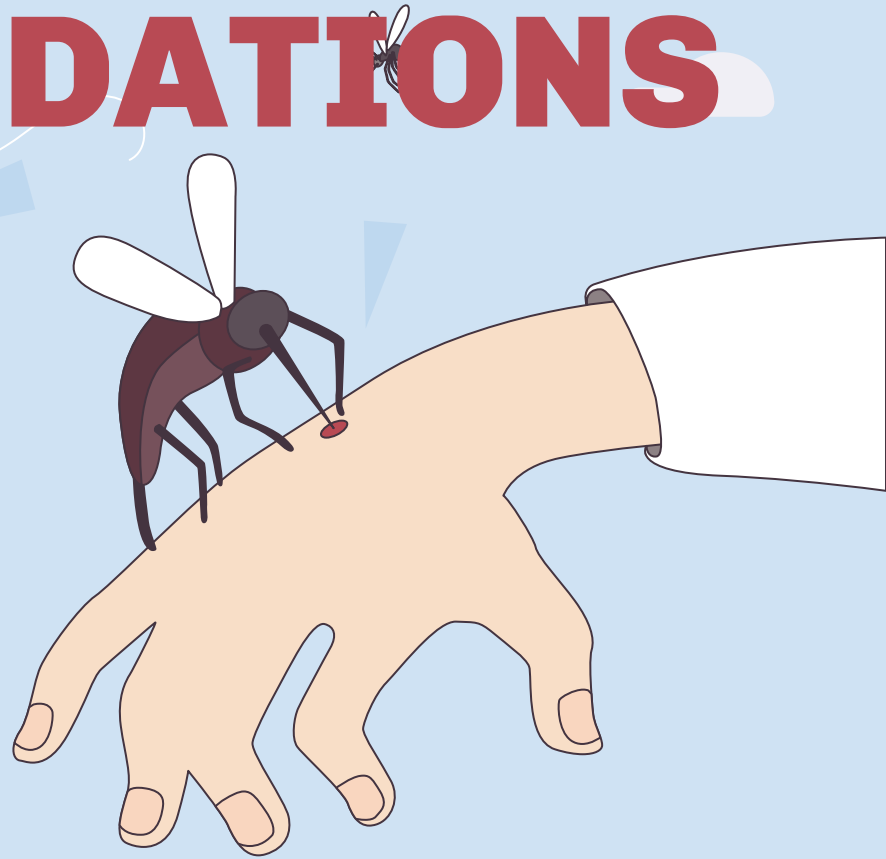
# COST BENEFIT ANALYSIS

Cannot put a price tag on human lives!

- As compared to the COVID-19 pandemic, the benefits of preventing an add-on factor on top of an ongoing pandemic will definitely outweigh the cost.



# RECOMMENDATIONS & FURTHER STEPS



# RECOMMENDATIONS

## Focus on Education

- Increase awareness on Social Media
- Start education early in schools

## Change the Spray Locations

- Blocks 10, 11 & 76 densely populated with the presence of West Nile Virus, but were not sprayed.



# RECOMMENDATIONS

## Consider alternative pesticides

- Zenivex is an adulticide which is less effective than larvicides.

## Utilize effective technologies

- Use drones to spray the pesticides at lower altitude for harder to reach targeted areas.



# FURTHER STEPS

## Research for less invasive solutions

- Research to optimally use wolbachia or other strains

## Conduct decision tree / markov models

- Data on healthcare costs & quality-adjusted life years.
- Economic evaluation to calculate incremental cost-effectiveness ratio.

## Account for environmental factors



- Study relationship between climate change and WNV transmission.
- Resistance to pesticides
- Public involvement in curbing WNV.



The background is a solid light blue. A large, stylized cloud in a slightly darker shade of blue is positioned on the right side. The word "CONCLUSION" is written in a bold, red, sans-serif font across the middle of the image, partially overlapping the cloud. To the left of the cloud, there is a large, detailed illustration of a mosquito with a dark brown body and a long, segmented abdomen featuring white spots. Three smaller, simpler illustrations of mosquitoes are scattered around: one in the upper left, one in the upper right, and one in the lower center. A thin, white, swirling line is located in the upper left quadrant, and a small, solid yellow circle is in the lower left quadrant. There are also some light blue and white geometric shapes, like triangles and semi-circles, scattered around the cloud.

**CONCLUSION**

# **CONCLUSION:**

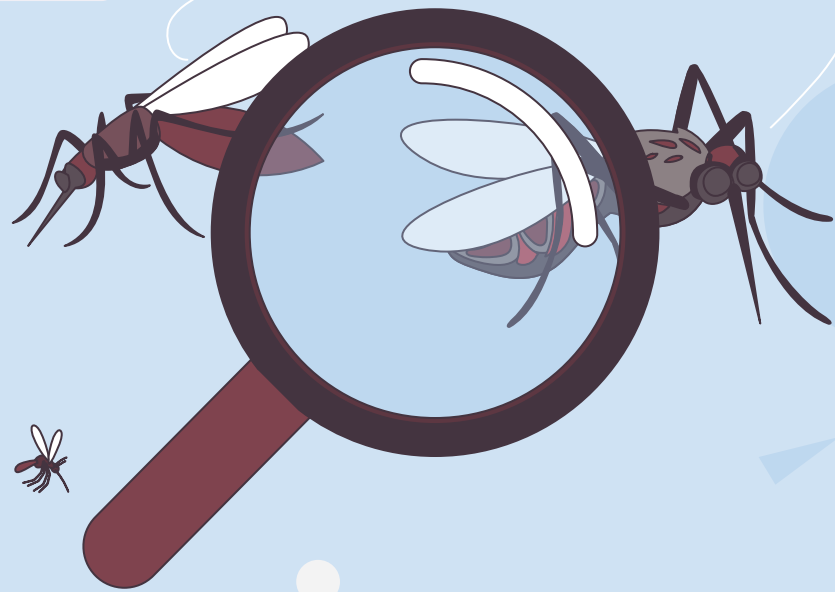
- Forecast of seasonal WNV outbreaks
- Best Model: Light Gradient Boost Model
- Top determinants for the presence of WNV are: Month, Tavg, year
- Conducted a preliminary cost benefit analysis but requires new data to obtain a more detailed analysis
- Future directions with regards to climate change



**Thank you!**

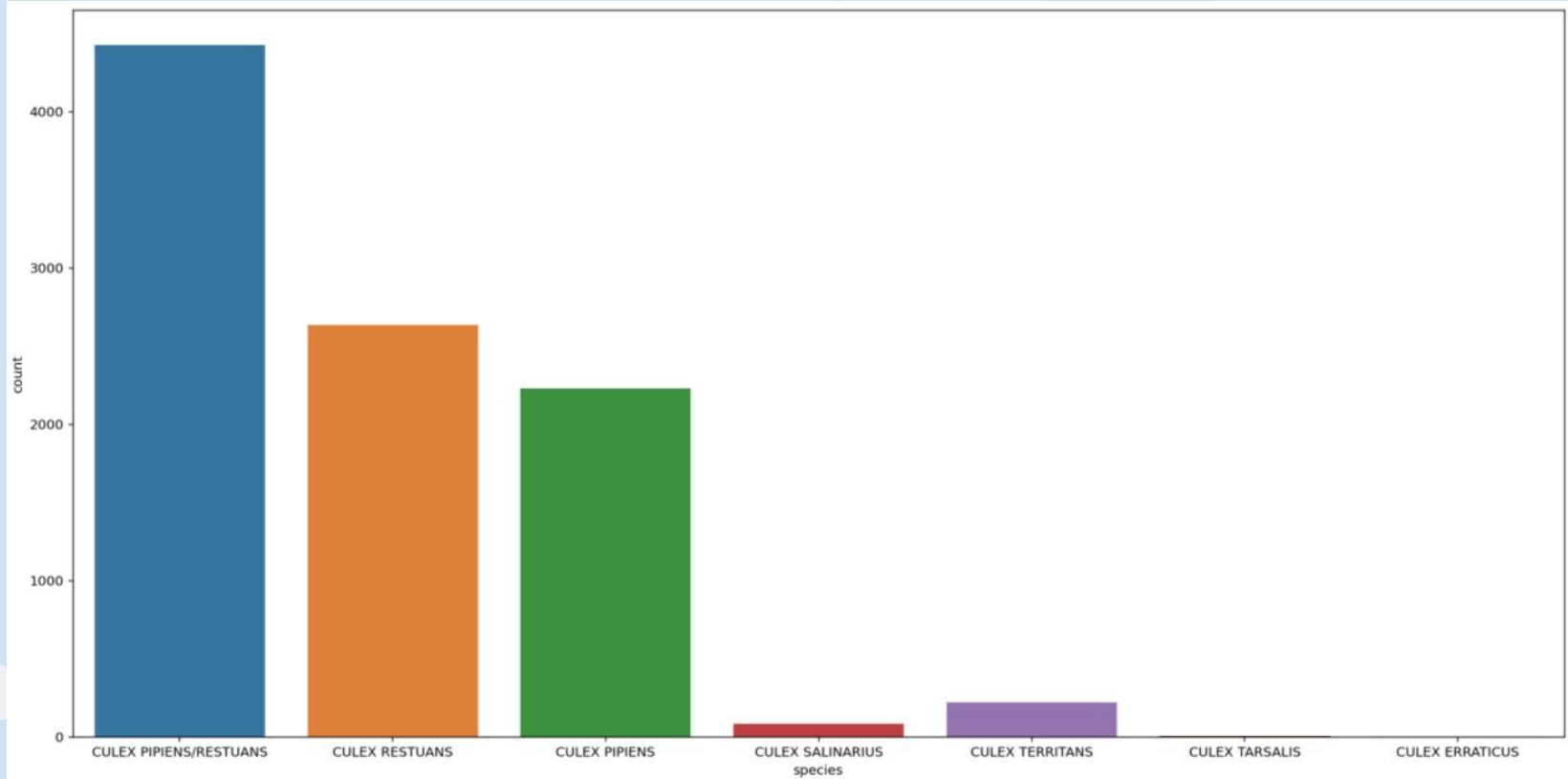
**Any  
Questions?**





# Appendix

# Distribution of all Mosquitos trapped



# Pesticide and WNV occurrence

