# Predicting the Outbreak of West Nile Virus in Chicago

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#### Outline

- Problem
- Data
- Model
- Results
- Conclusion

#### Problem

- First cases of West Nile
  Virus were reported in
  Chicago in 2002
- Predict geographically where there is a high chance that the virus is present



#### Data

- Mosquito traps were set up throughout the city and each year, from spring to late summer, health workers tested the mosquitos in the traps and determined the species as well as whether or not the virus was present (WnvPresent)
- Weather data was also available from NOAA from 2007 to 2014
- Independent Variables (Weather)
  - Tmax
  - o Tmin
  - Tavg
  - PrecipTotal
  - DewPoint
  - WetBulb
  - Heat

- Independent Variables (Main)
  - Address
  - Species
  - o Block
  - Latitude
  - Longitude
  - Trap
  - AddressAccuracy

## Hypothesis

- According to theory, hot and dry conditions favor the presence of the West Nile Virus vs. cold and wet.
- Therefore, temperature and dewpoint will play an important role in terms of feature importance in deciding whether the virus is present.

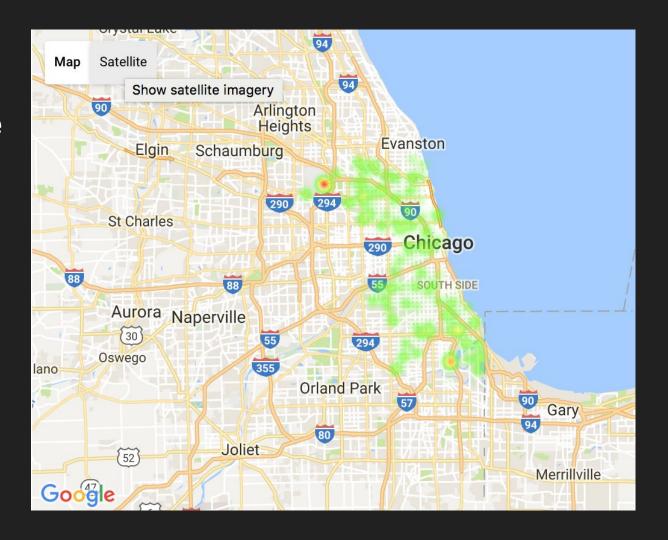
# **Data Processing**

Weather data
 needed to
 be
 combined
 with the
 main
 dataset and
 an inner
 join was
 carried out

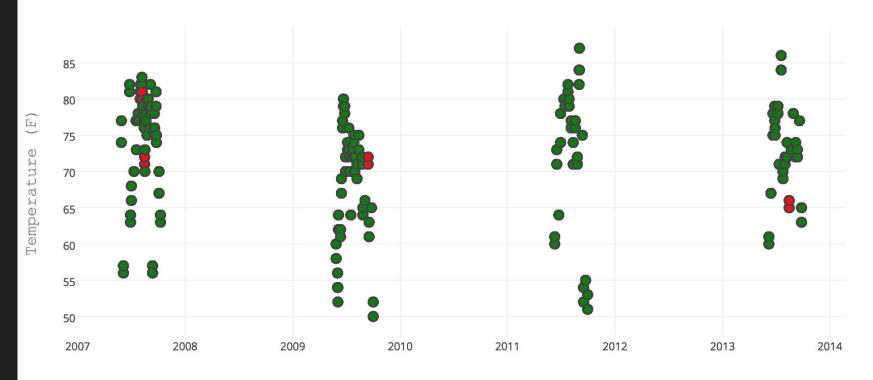
None of the data columns had missing values

Block	Latitude	Longitude	AddressAccuracy	NumMosquitos	WnvPresent
1.000000	0.091110	-0.090375	0.222134	-0.172388	0.004877
0.091110	1.000000	-0.701795	0.444026	-0.184806	0.028697
-0.090375	-0.701795	1.000000	-0.456775	0.036633	-0.060345
0.222134	0.444026	-0.456775	1.000000	-0.248414	0.008064
-0.172388	-0.184806	0.036633	-0.248414	1.000000	0.196820
0.004877	0.028697	-0.060345	0.008064	0.196820	1.000000
0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
-0.001622	-0.065662	0.080886	-0.074680	0.158969	0.048140
-0.012408	-0.096024	0.099759	-0.105742	0.193559	0.073005
-0.004430	-0.064413	0.086615	-0.081297	0.158800	0.085632
	1.000000 0.091110 -0.090375 0.222134 -0.172388 0.004877 0.000000 -0.001622 -0.012408	1.000000    0.091110      0.091110    1.000000      -0.090375    -0.701795      0.222134    0.444026      -0.172388    -0.184806      0.004877    0.028697      0.000000    -0.005662      -0.012408    -0.096024	1.000000    0.091110    -0.090375      0.091110    1.000000    -0.701795      -0.090375    -0.701795    1.000000      0.222134    0.444026    -0.456775      -0.172388    -0.184806    0.036633      0.004877    0.028697    -0.060345      0.000000    0.000000    0.000000      -0.001622    -0.065662    0.080886      -0.012408    -0.096024    0.099759	1.000000    0.091110    -0.090375    0.222134      0.091110    1.000000    -0.701795    0.444026      -0.090375    -0.701795    1.000000    -0.456775      0.222134    0.444026    -0.456775    1.000000      -0.172388    -0.184806    0.036633    -0.248414      0.004877    0.028697    -0.060345    0.008064      0.000000    0.000000    0.000000    -0.074680      -0.012408    -0.096024    0.099759    -0.105742	1.000000    0.091110    -0.090375    0.222134    -0.172388      0.091110    1.000000    -0.701795    0.444026    -0.184806      -0.090375    -0.701795    1.000000    -0.456775    0.036633      0.222134    0.444026    -0.456775    1.000000    -0.248414      -0.172388    -0.184806    0.036633    -0.248414    1.000000      0.004877    0.028697    -0.060345    0.008064    0.196820      0.000000    0.000000    0.000000    0.000000      -0.001622    -0.065662    0.080886    -0.074680    0.158969      -0.012408    -0.096024    0.099759    -0.105742    0.193559

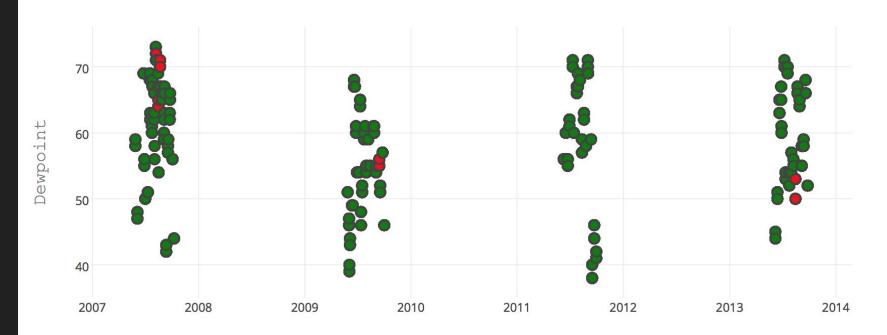
Distribution of Virus Presence From Joined Training Data Set



#### The Impact of Temperature on the Presence of West Nile Virus



#### The Impact of Dewpoint on the Presence of West Nile Virus



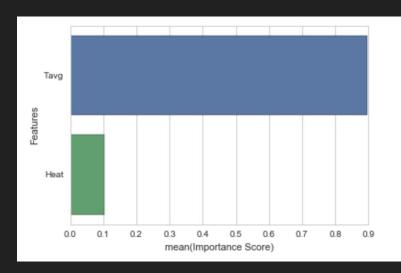
Date

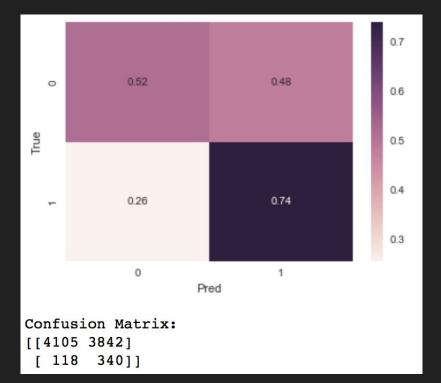
#### Random Forest

- Large Numbers of Features (Temperature, Dewpoint, Heat, Mosquito Species, Address, etc.)
- Classification
- Easy to overfit

#### Baseline

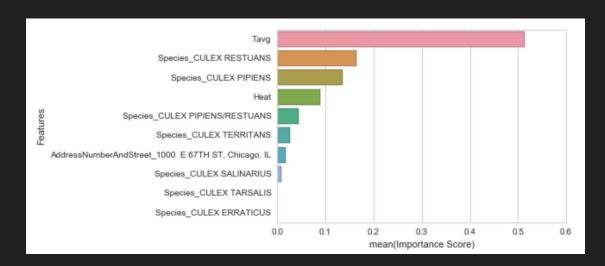
- Two Features: Tavg and Heat
- Accuracy of Training: 0.526
- Accuracy of Testing: 0.529

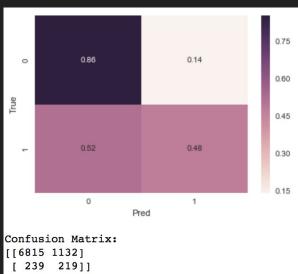




## Feature Engineering

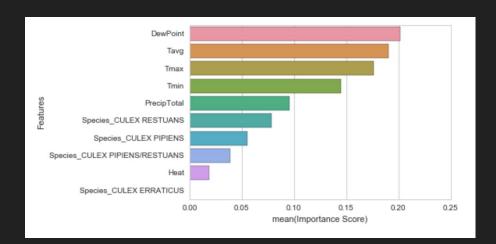
- Features: Tavg, Heat, Address, Species
- Accuracy of Training: 0.872
- Accuracy of Testing: 0.837
- ROC AUC: 0.679

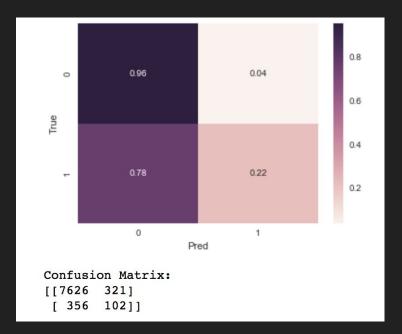




# Feature Engineering Cont.

- Features: Tavg, Heat, Address, Species,
  Dewpoint, PrecipTotal, Tmax, Tmin
- Accuracy of Training: 0.964
- Accuracy of Testing: 0.919
- ROC AUC: 0.784

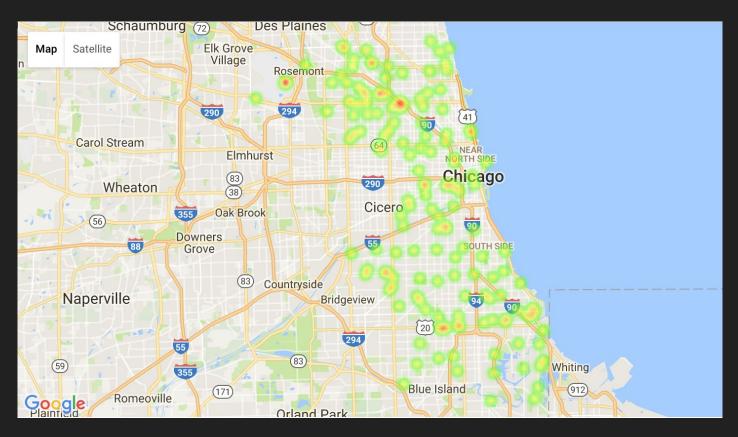




## Tuning Parameters with Grid Search

- N\_estimators 200
- Max\_features 30
- Accuracy of Training: 0.966
- Accuracy of Testing: 0.914

# Prediction (2008, 2010, 2012, 2014)



#### Conclusion

- A lot of features are involved in determining the presence of West Nile Virus
- The features tell us which are the most relevant:
  Dewpoint, Tavg, etc.
- The more information you feed into a model, the greater the accuracy

#### Next Steps

- Since more data is better for the model, we can pull in the mosquito spraying data
- We can look at the number of cases of West Nile Virus reported
- Important for public health planning and allocation of resources to the right place at the right time
- Extend this analysis to studying the Zika Virus

# Acknowledgements

- Jim
- Kate and Austin
- Tim
- Everyone in the class
- GA

#### Citations

- Kaggle Competition: <a href="https://www.kaggle.com/c/predict-west-nile-virus/data">https://www.kaggle.com/c/predict-west-nile-virus/data</a>
- GA DS-DC-13 (Class Notes)

# Any Questions?