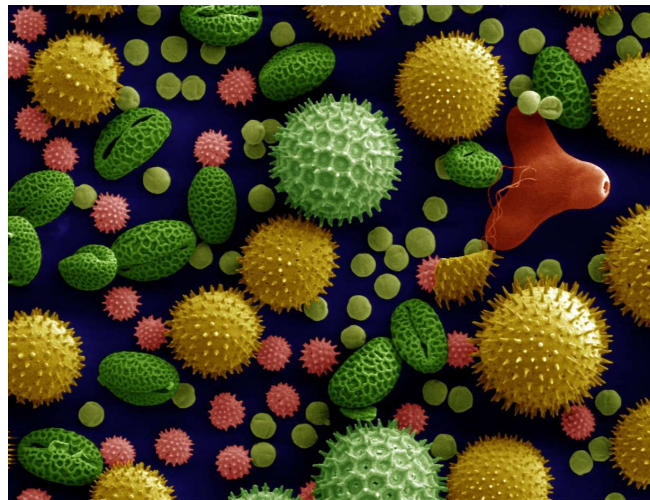


Houston, We Have an *Allergy* Problem



Predicting Pollen Counts in Texas

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Agenda:

1. Introduction & Business Understanding

2. Business Problem

3. Data Understanding

4. Initial Analysis

5. Modeling: *Classification & Time-Series*

6. Results

7. Species-Specific Diagnostic Tool Proposal

8. Further Considerations

9. Conclusion

10. References & Appendix

Introduction:

Allergies & Public Health After Covid-19

1. Heightened Public Awareness of Allergies
2. Seasonal Strain on Healthcare Providers



Allergies: *A Business Problem*



Preparing Medical Providers & the General Public
for Allergy Season Through Public Health Initiatives

The Data: *Pollen*

- All Pollen Data From Houston Health Dept.
- Monthly Tallies of Species-Specific Pollen Counts per m³
- Business Days from January 2013 to April 2022



The Data: *Climate*

- Daily Climate Data from
- 2 Testing Centers in Houston, TX, 1 from Shreveport, LA
- Average Daily Temp (F),
Average Wind Speed (mph),
Precipitation (inches)



Pollen Counts & Allergies

TREE POLLEN:

- 90-1499 /m³: *Heavy*
- 1500+ /m³: *Extremely Heavy*

GRASS POLLEN:

- 20-199/m³: *Heavy*
- 200+ /m³: *Extremely Heavy*

WEED POLLEN:

- 50-499 /m³ : *Heavy*
- 500+ /m³: *Extremely Heavy*



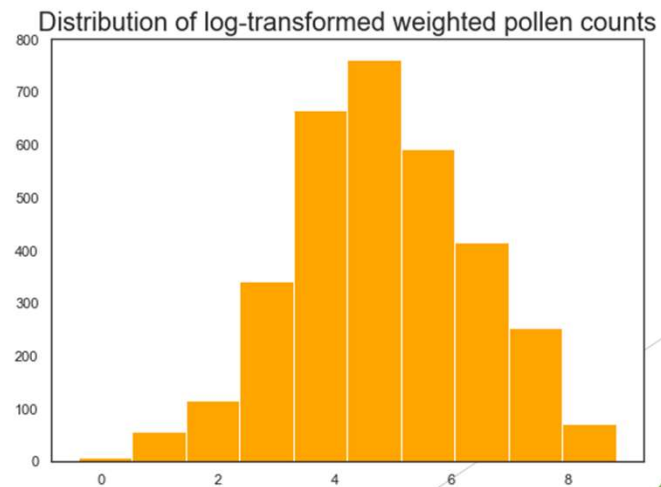
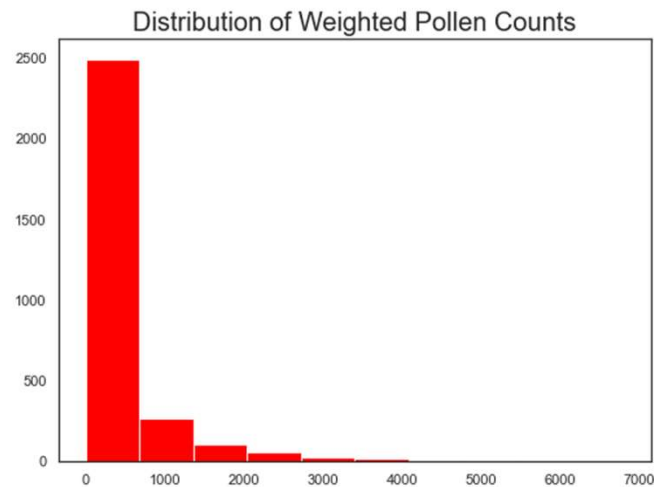
Weighted Pollen Counts

- Weighted using Different Severities of Tree, Grass and Weed
- “high” pollen set at 500+ particles per m³ per HHD
- ~50% of days ‘high’ pollen for Houston, TX

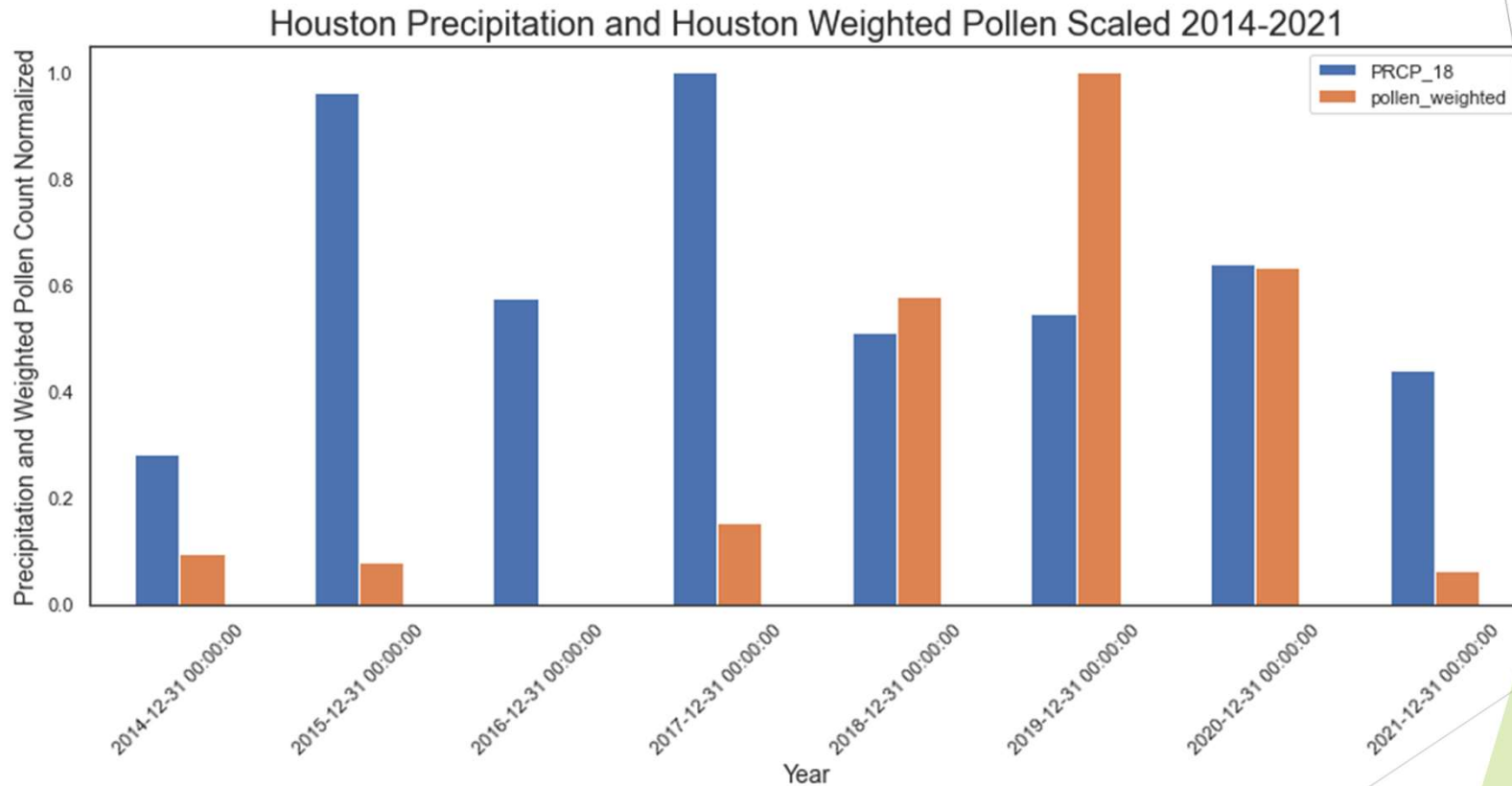


Exploratory Findings

- Many Days had pollen counts of zero especially in summertime
- Max Pollen Count of over 9,000 per m^3 occurred in March, 2019
- Pollen's Logarithmic Distribution



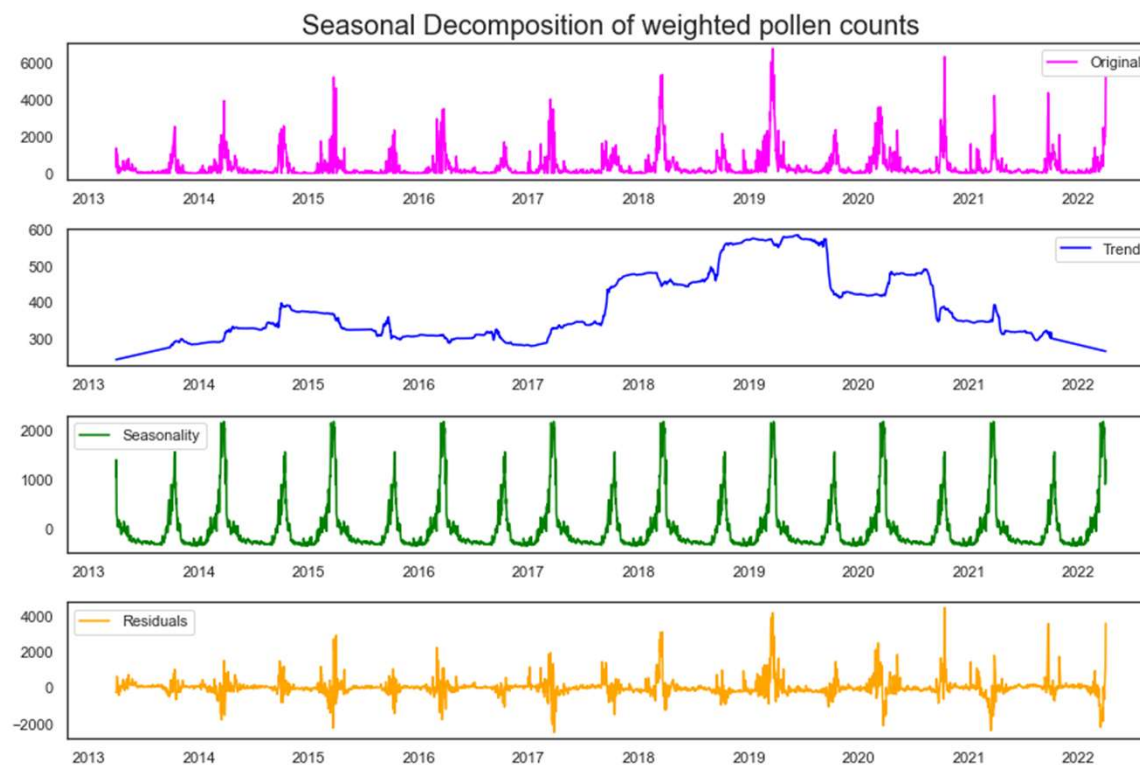
Exploratory Findings



No Clear Correlation Between Local Precipitation* and Pollen Counts for Current or Prior Year

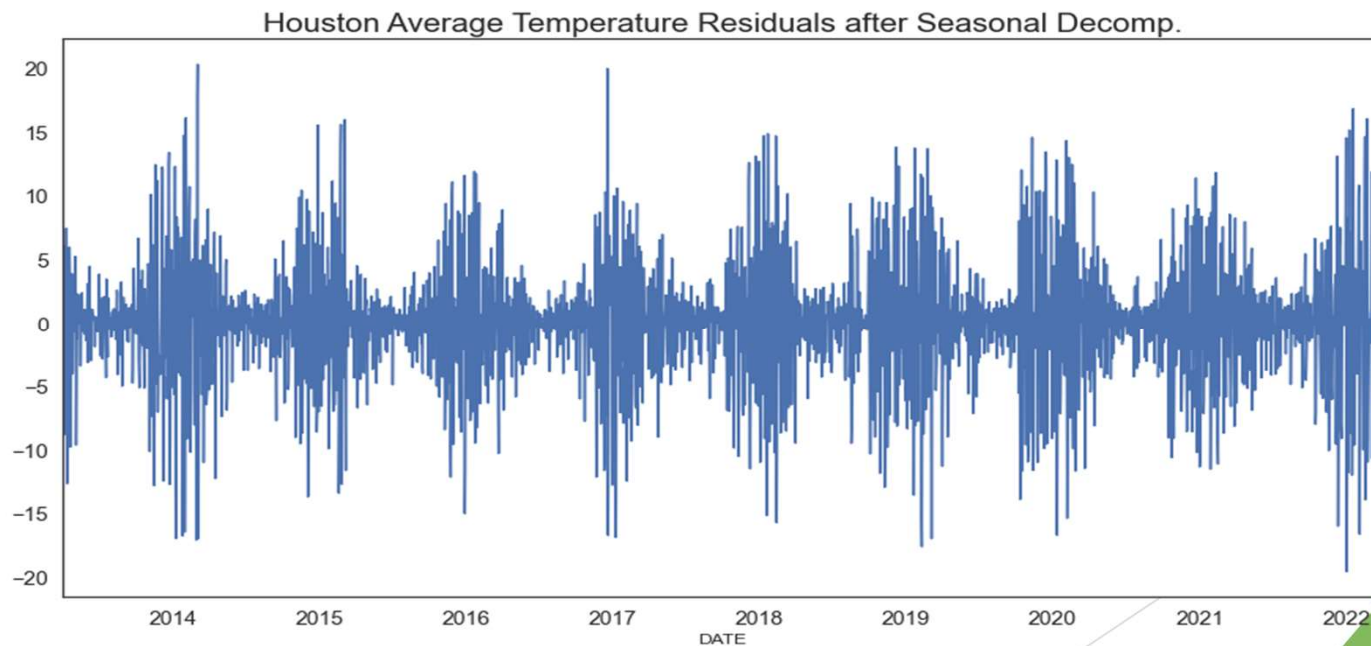
* Also True for Shreveport, & other variables (see Appendix)

Classification Modeling: *The Seasonality Problem*



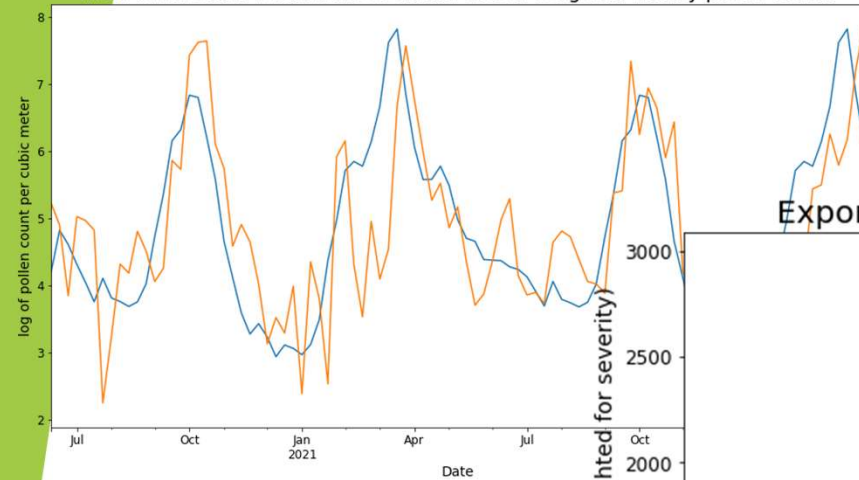
Classification Modeling

- Decision Tree Model: Local Temp most important feature
- Random Forest Classifier Achieved 60% Accuracy
- Seasonality after decomposition?

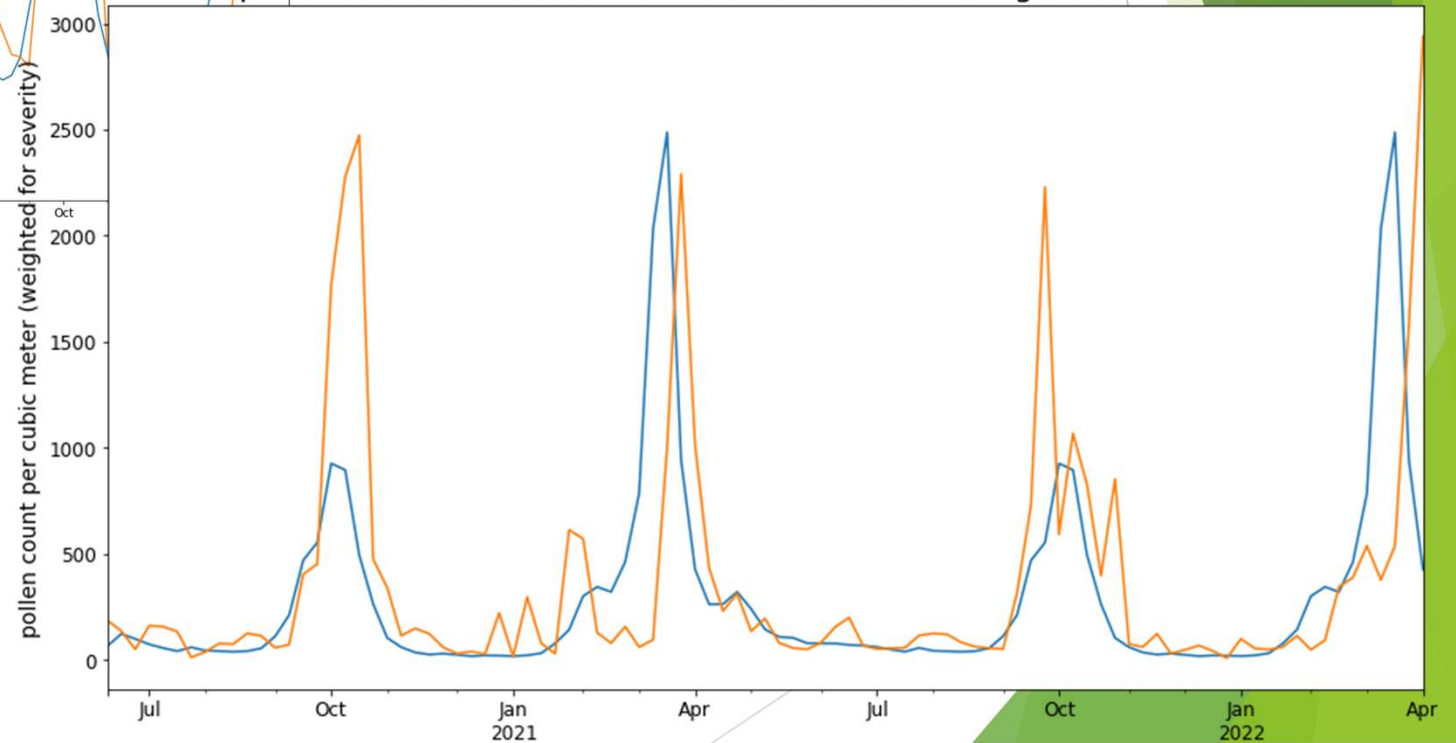


Time-Series Modeling

First Arima Model Predictions vs Actual LOG of weighted weekly pollen counts



Exponentiated First Arima Predictions vs Actual Weighted Pollen Counts



Results

Classification:

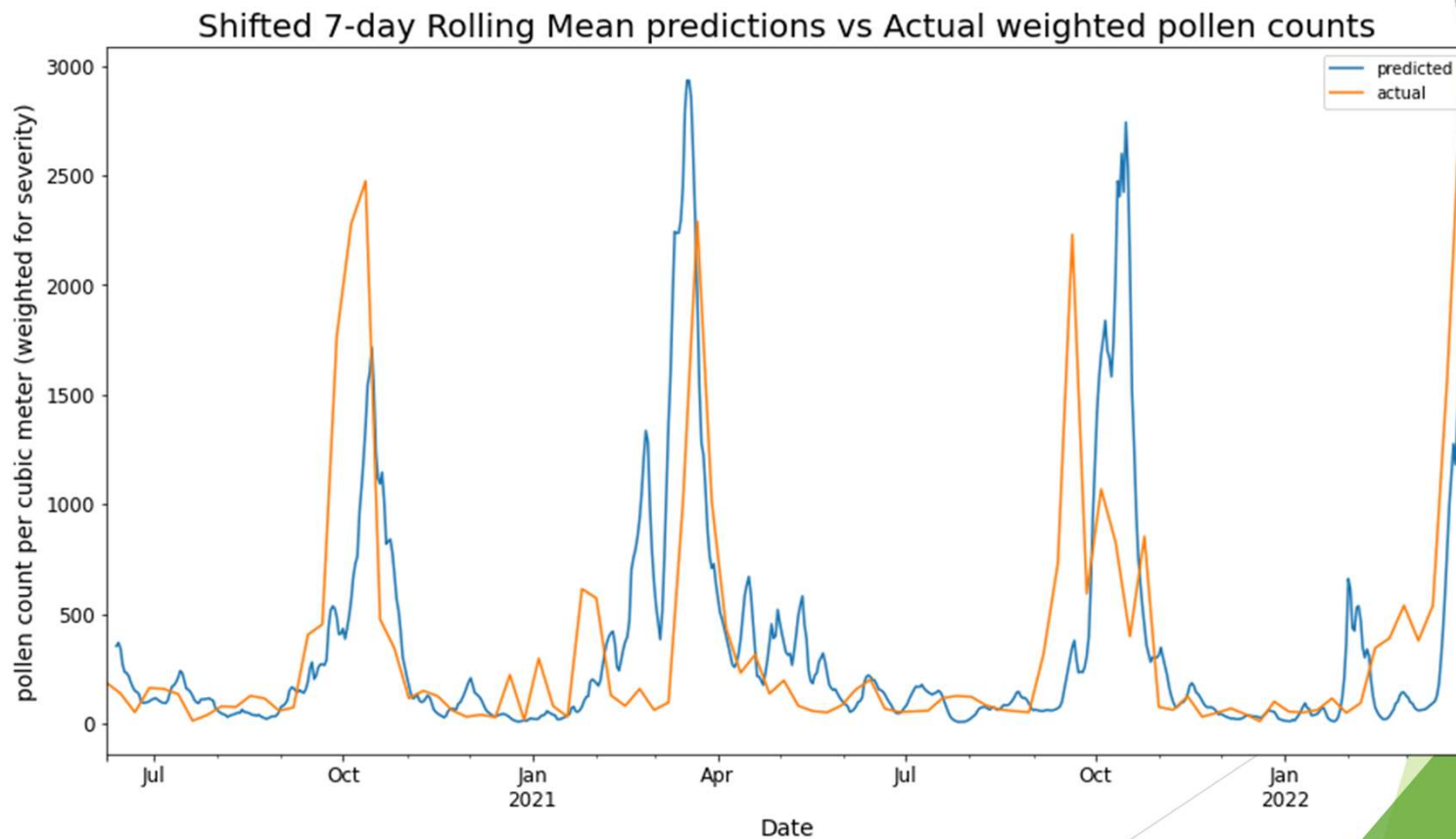
- Best Model was Random Forest 60% Accuracy, 59% Precision

Time Series:

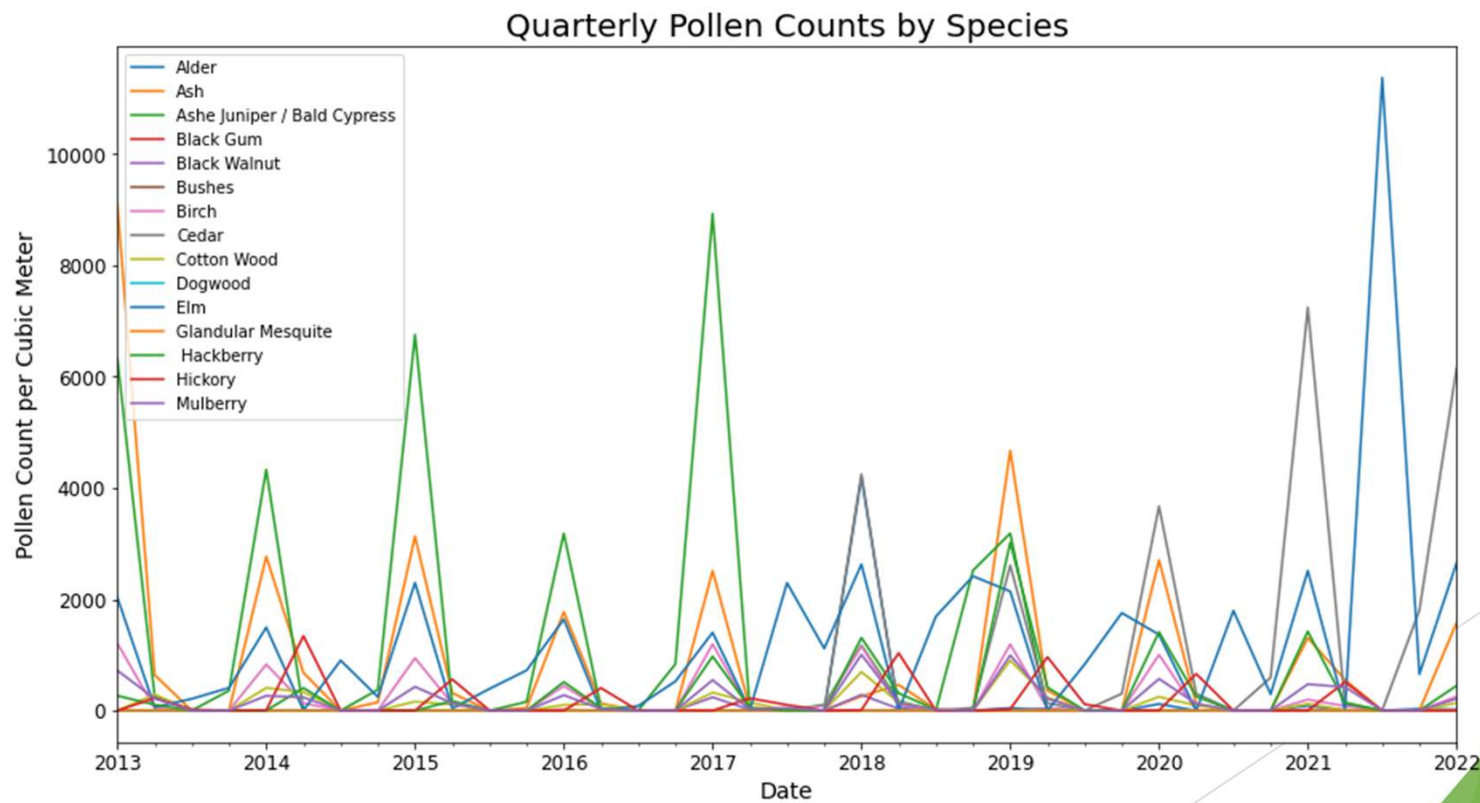
- Best Predictor was Prior Year's Pollen Counts
- Off by ~ 425 on average

Results:

Best Predictions from Last Year

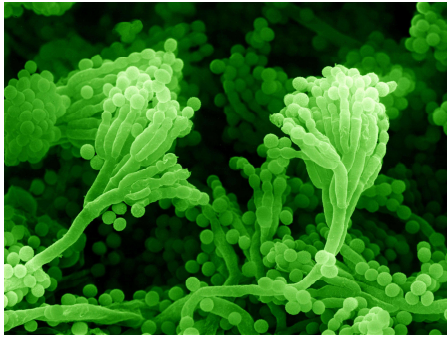


Proposal for Next Steps: *Species-Specific Diagnostic Tool*

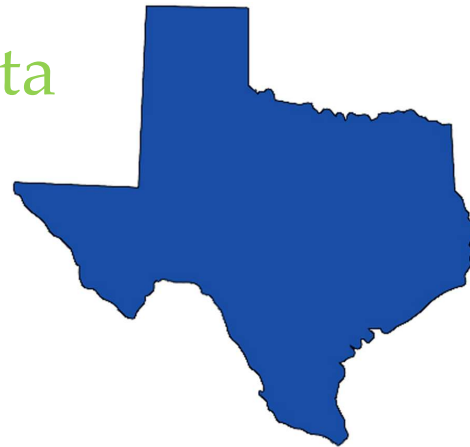


Further Considerations:

1. Mold Spores



2. More Local and Statewide Data



Conclusion

Seasonality is King

- The local seasonality of pollen production is the best predictor of pollen counts.

Thanks!

Questions and Comments



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References

1. Mold Spores

1. More Local and Statewide Data

Appendix

