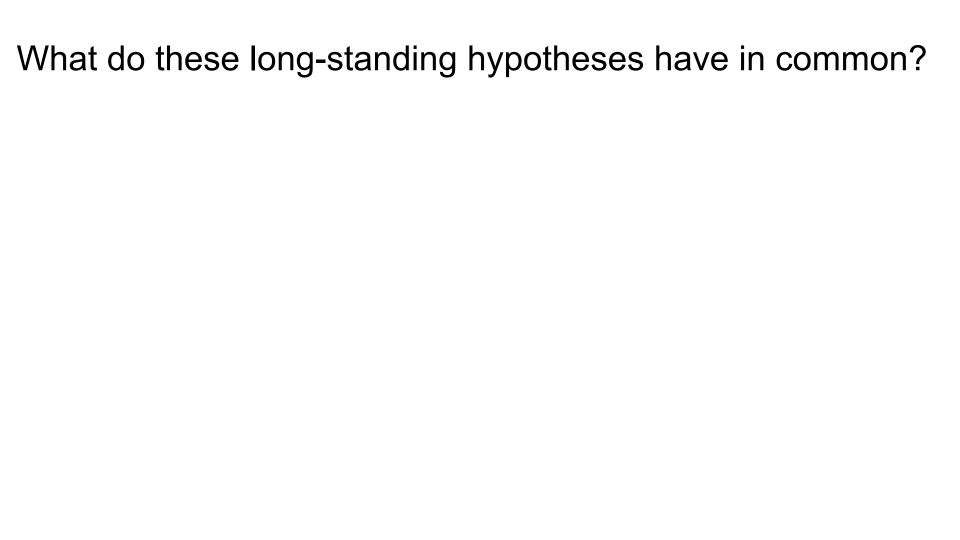


Predicting Social Media Behavior Based on Weather Patterns



The Earth is Flat

Eating sugar causes a "sugar-rush"

Bulls get angry when they see red



Eating sugar causes a "sugar-rush"

Bulls get angry when they see red







Better weather makes people happier (on twitter)



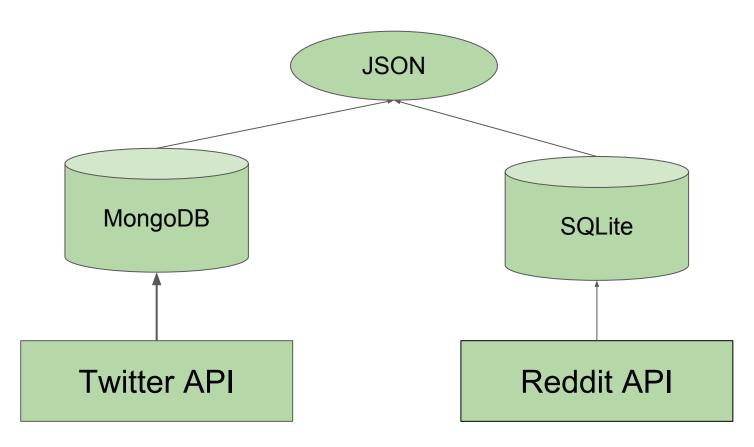
Better weather makes people happier (on twitter)

Problem Overview

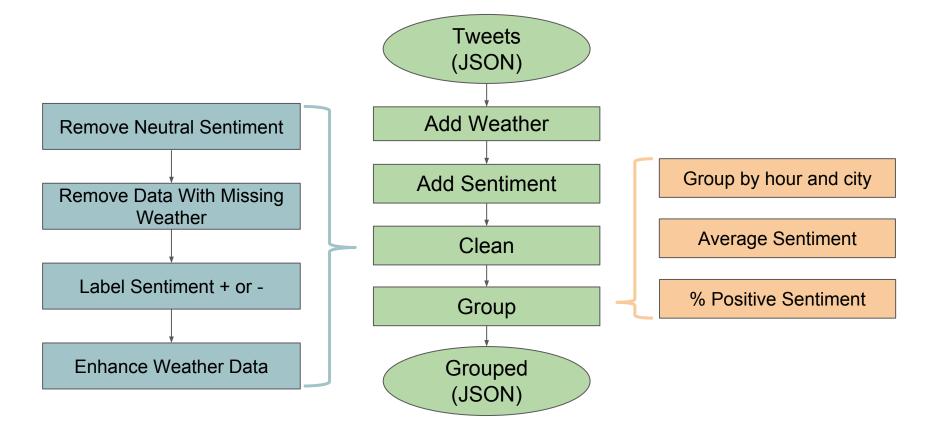
- Is social media sentiment influenced by weather patterns?
- Can we measure the impact of weather on the mood of a city by observing sentiment on social media

DATA COLLECTION AND ANALYSIS

Data Collection Overview



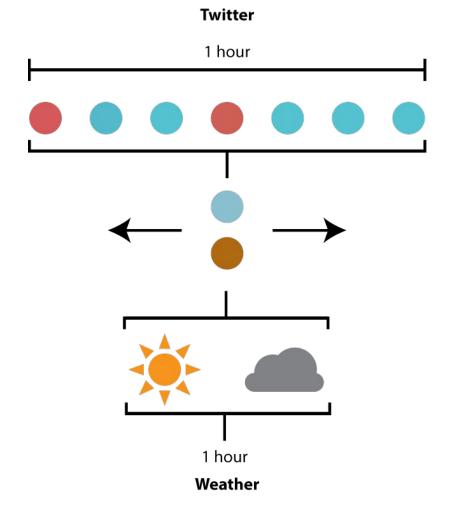
Data Augmentation Overview



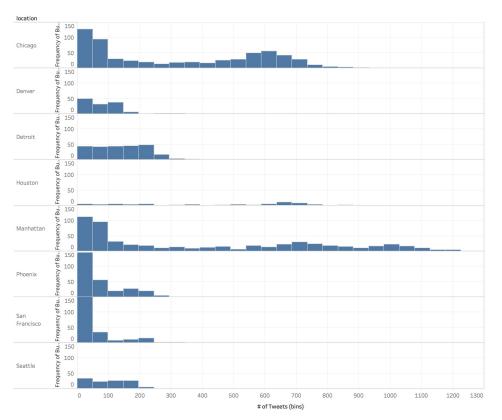
Data Cleaning

Training Data Set Creation:

- 1. Collect 37 days of tweets
- 2. Add weather
- 3. Add sentiment
- 4. Clean
 - a. Non-neutral sentiment
 - b. Weather data available
 - c. Augment weather data
- 5. Group by hour
- 6. Average sentiment by hour
- 7. Enrich tweets with weather data

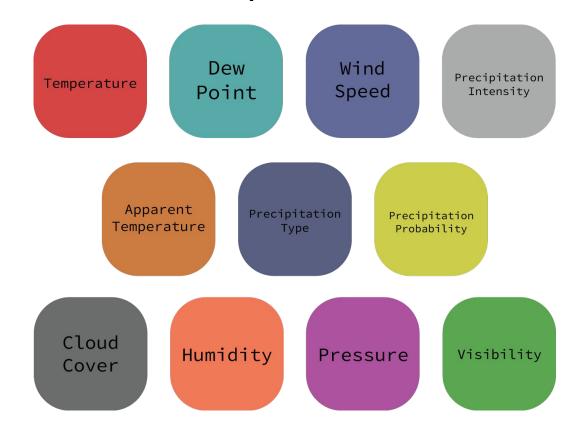


Data Grouping

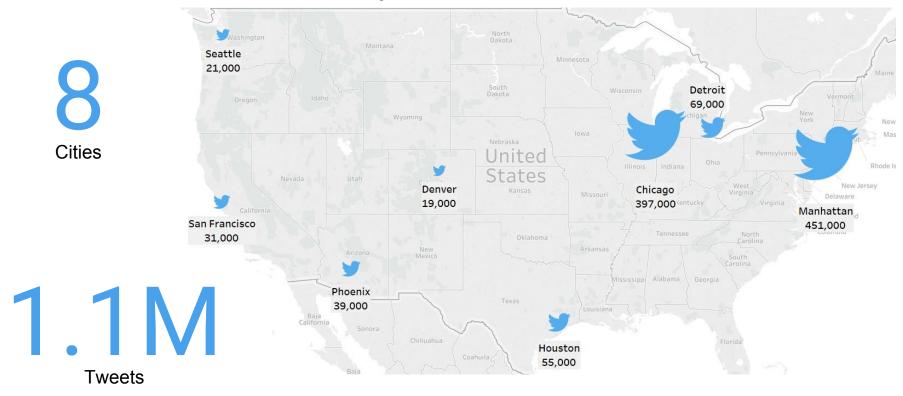


- We aggregated our data into a single point for each hour for each city
- In aggregating the data, we calculated the average sentiment and % positive sentiment
- In order for these statistics to be valid, we need enough tweet in each hour for the average to be close to the actual average
- If we had more data we may have excluded any data points with less than ~20 tweets

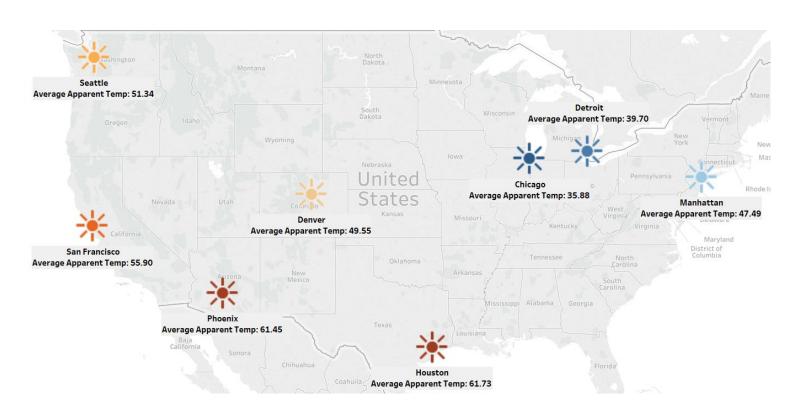
Look at all the weather parameters!



Twitter Sentiment Analysis



Twitter Cities' Average Temperature

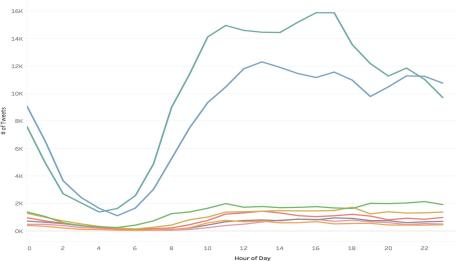


Twitter Data Insights

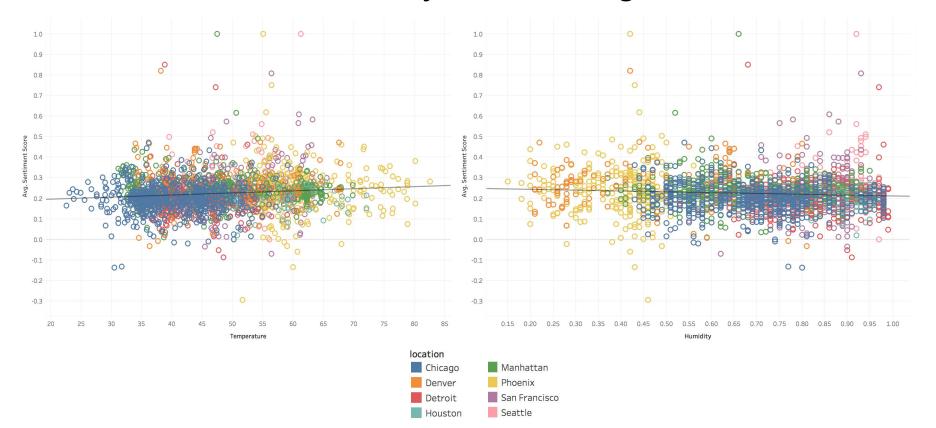
Average Sentiment across America



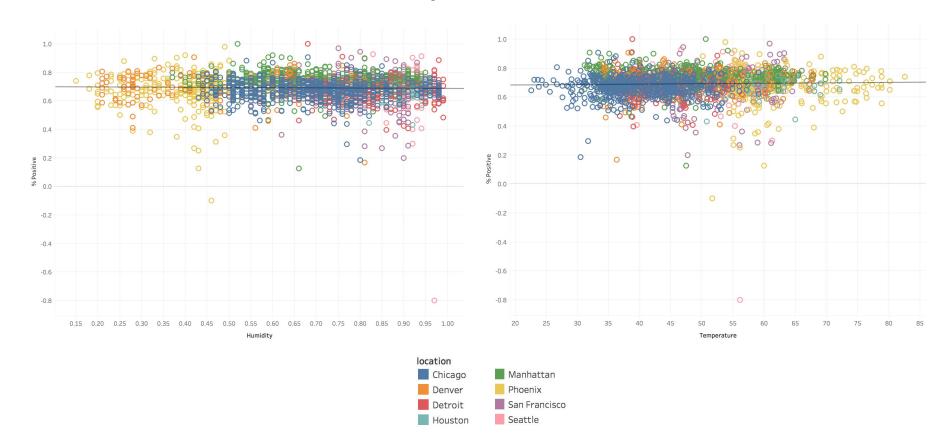
Twitter Usage during an Average American Day



Twitter Correlation Analysis - Average Sentiment

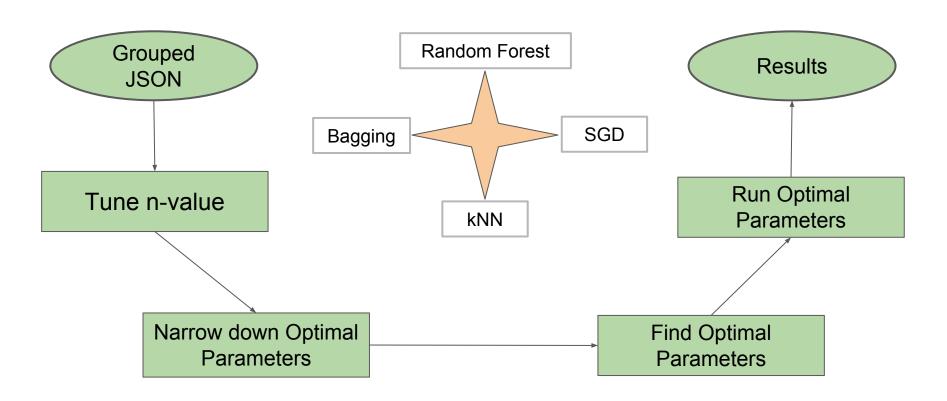


Twitter Correlation Analysis - % Positive

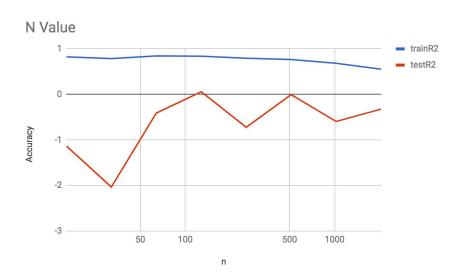


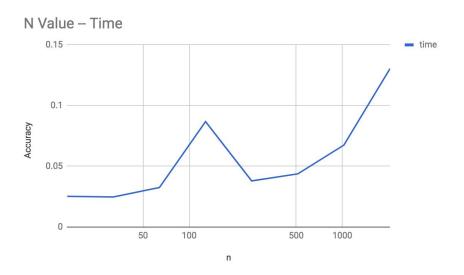
MACHINELEARNING

Machine Learning Overview

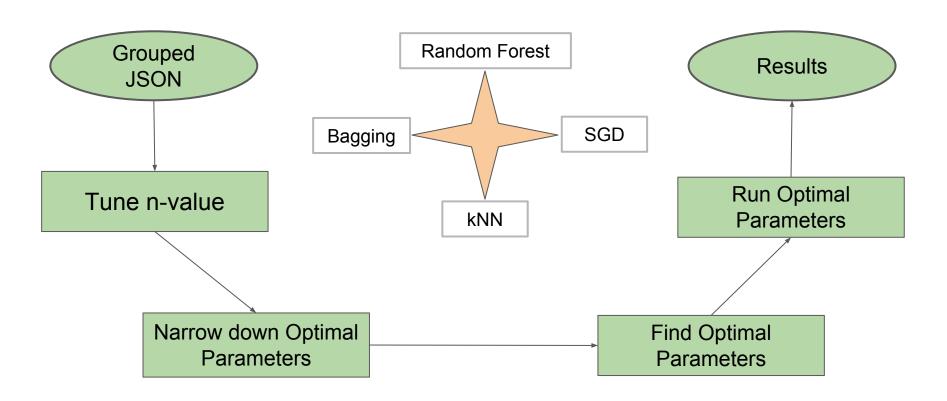


Twitter N-value Optimization

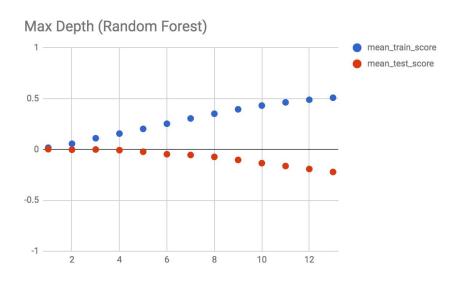


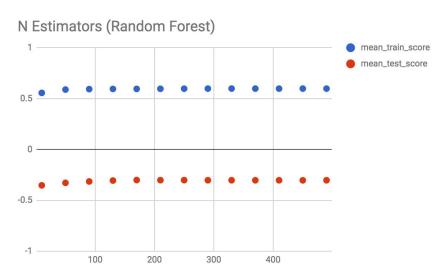


Machine Learning Overview

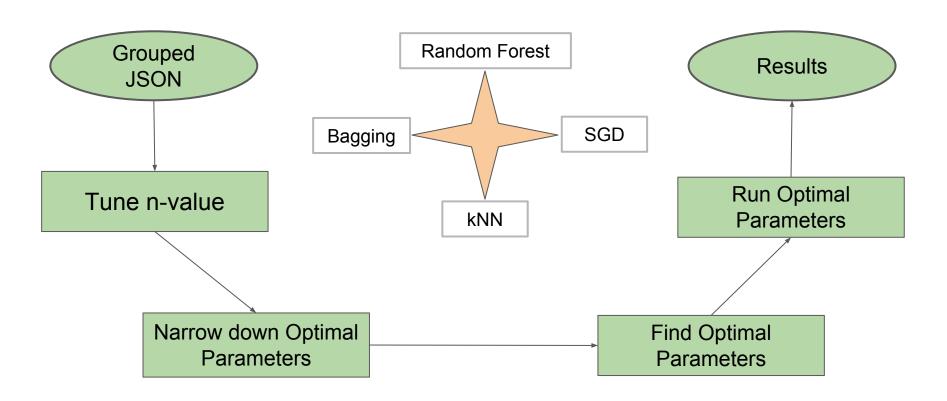


Twitter Parameter Optimization

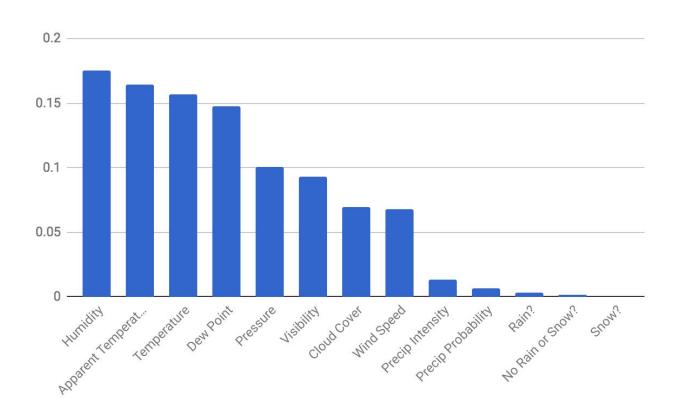




Machine Learning Overview



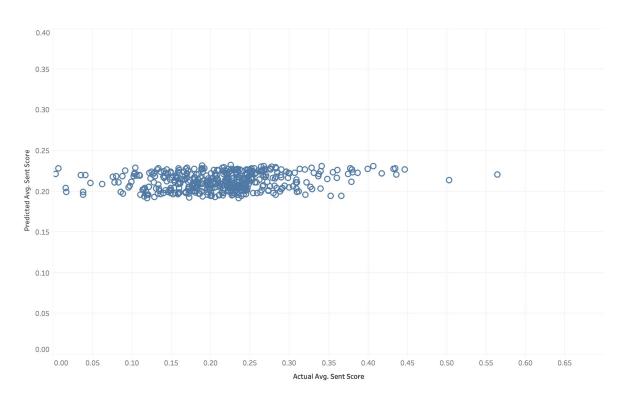
Twitter Feature Importance (Random Forest)



Twitter Machine Learning Results

Algorithm	R2 Train	R2 Test
RandomForest	0.0424	0.0203
K-Neighbors	0.0419	0.0322
Stochastic Gradient	-0.1220	-0.1619
Bagging	0.1000	0.0042

Twitter Machine Learning Results

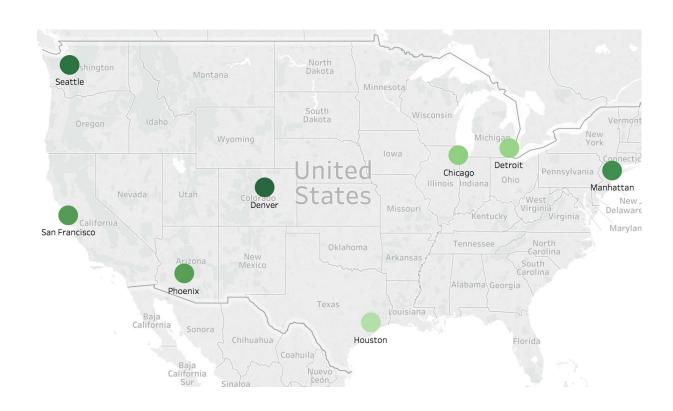


OTHER CONSIDERATIONS

Reddit Sentiment Analysis Extension

8 Cities

3.4M
Comments



Reddit Comments



 $r^2 = .002$

Twitter % Positive Sentiment

 $r^2 = .03$



Conclusion

$$H_0: r^2 > 0$$

$$H_A: r^2 \leq 0$$

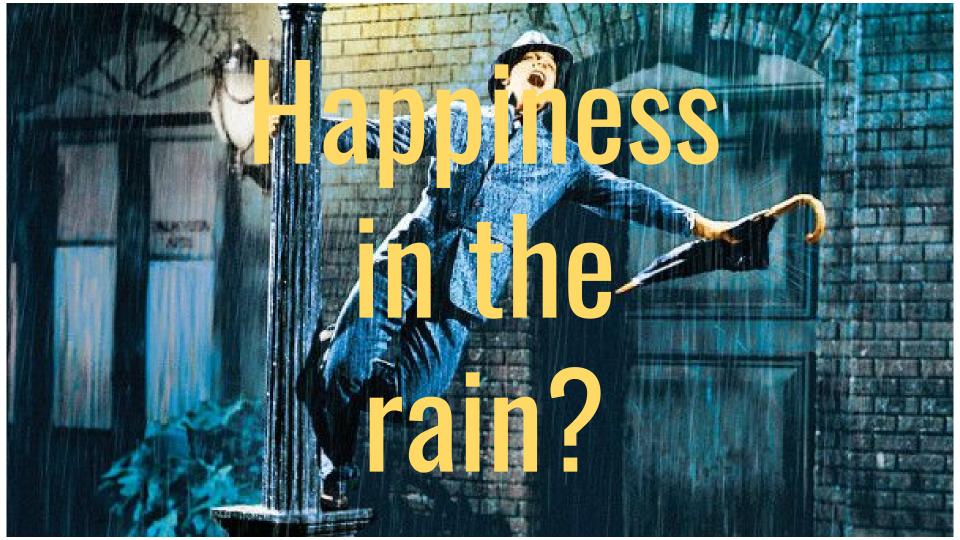
We **reject** the null hypothesis that we can predict sentiment using weather forecasts.

Conclusion

$$H_0: r^2 > 0$$

 $H_A: r^2 \leq 0$

that we can predict entiment using leader forecase.





Reflection and Next Steps

Reconsiderations

- More data (Year-round Twitter data)
- Explore "change in weather" rather than absolute weather
- Filter tweets for specific topics (i.e. tweets about weather)
- Other than weather, what else could impact sentiment?

Controlled Experiment

- Have a group of people tweet 5 times a day at pre-set times
- Clean data!

