

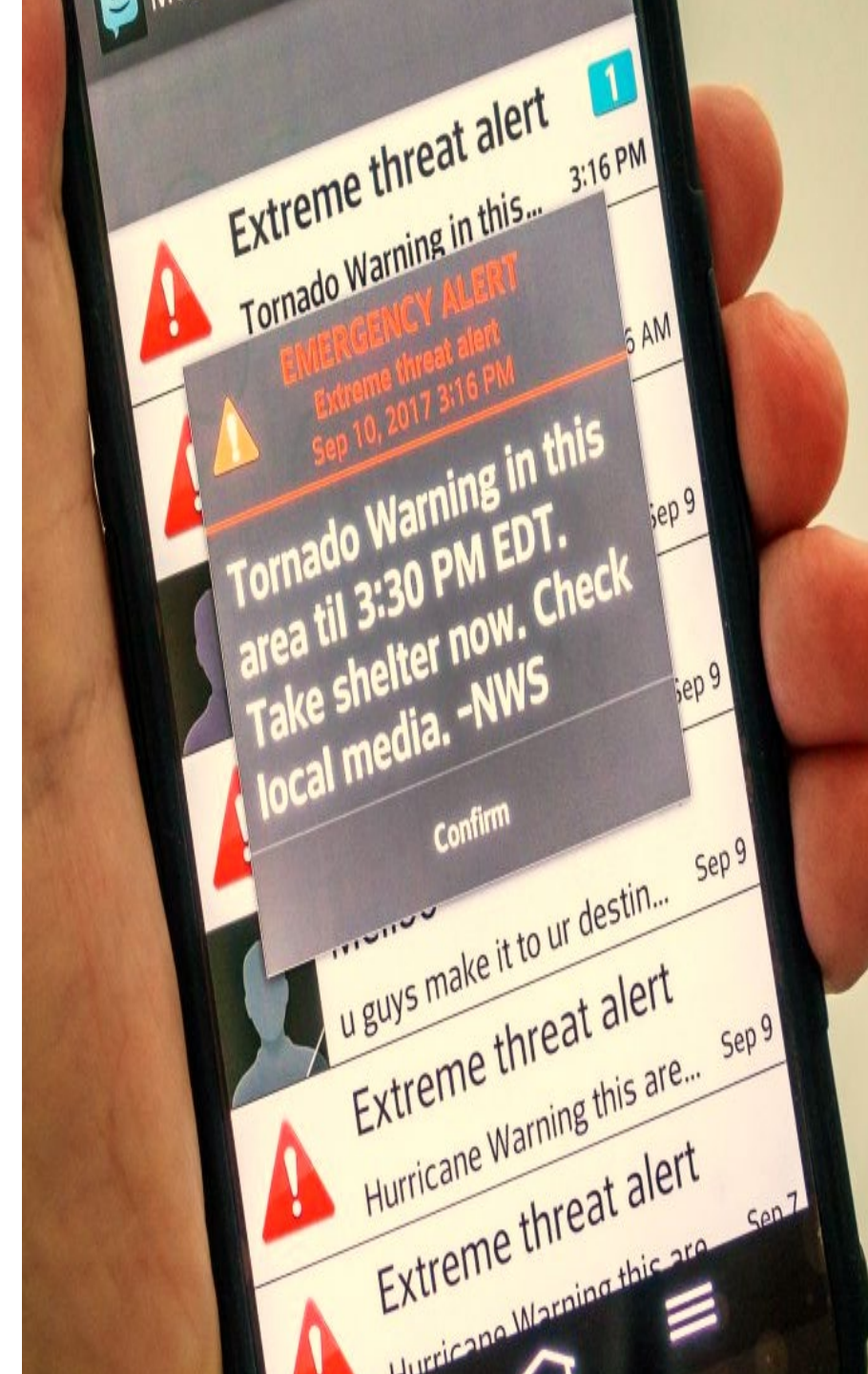


FLIPSWITCH ALERT SYSTEMS

**Presentation to Ran Goldblatt and Associates
at New Light Technologies**

OVERVIEW

- Opportunity
- Methodology
- Analysis
- Conclusions
- Next Steps



OPPORTUNITY

- In the event of an emergency, citizens typically rely on three key channels to alert them:
 - Government Agencies
 - Local First Responders
 - News Media
- Hyperconnected world
 - Everyone's a part of social media
 - No longer be necessary to "wait" for an informed source to issue an alert
- Can we provide an earlier indicator of a natural disaster event based on the content and frequency of social media posts?



METHODOLOGY

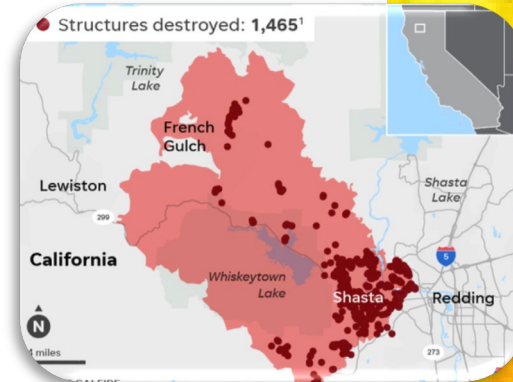
- **Focus on a key event**
 - **Allow us to focus on the mechanics surrounding the disaster event**
 - **Build with an eye to scale up to include other disaster events**
- **Procure Twitter activity around that event**
- **Use NLP to create a model to predict/signal future events**

METHODOLOGY

FOCUS EVENT SELECTION

CARR FIRE

- **7th largest California Wildfire on record**
 - Burned 229,651 acres
 - Destroyed over 1,600 structures
 - Over 3,500 firefighter personnel responded
- **Key dates to analyze:**
 - July 23 – 24, 2018
 - First full week of July 22
- **Well covered throughout social media**



METHODOLOGY

SOCIAL MEDIA SELECTION

TWITTER VIA INTERNET ARCHIVE

- Curated tweet libraries
- Access about 1,000,000 tweets a day
- Available for the time periods we sought



CAL FIRE @CAL_FIRE · 3h

#CarrFire [update] northwest of Anderson (Shasta County) is now 98,724 acres and 20% contained. Evacuations and road closures in place. Unified Command: CAL FIRE and Whiskeytown National Park. Photo credit: CAL FIRE
[fire.ca.gov/current_incide...](https://fire.ca.gov/current_incidents)



FLIPSWITCH

CONTENT

REFINED
KEYWORD LIST

TWEET RATE

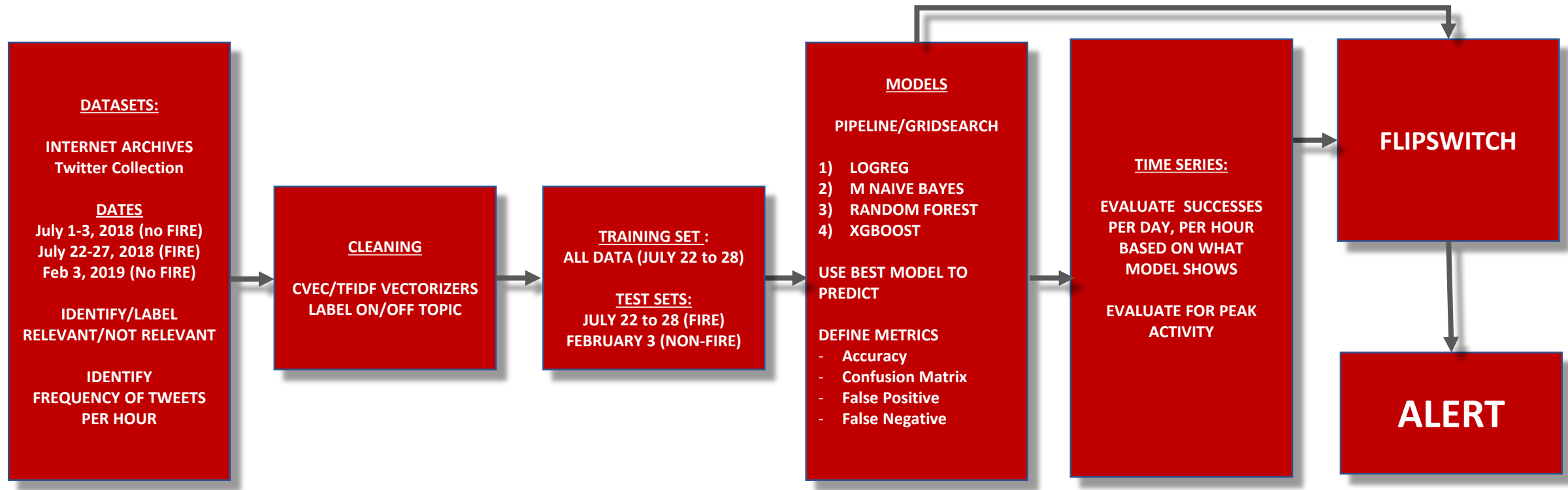
PER HOUR
PER DAY



**CONFIRM
ALERT**

ANALYSIS: OUR PROCESS

THE STRUCTURE TO CREATE THE FLIPSWITCH



ANALYSIS

MODEL PERFORMANCE

- Models performed well overall
- Random Forest and XG Boost were most effective

| MODEL | ACCURACY SCORE |
|---------------------|---------------------------|
| LOGISTIC REGRESSION | CVEC 99.6% TFIDF 98.4% |
| NAÏVE BAYES | CVEC 98.3% TFIDF 98.2% |
| RANDOM FOREST | CVEC 99.6% TFIDF 99.8% |
| XG BOOST | CVEC 99.7% TFIDF 99.7% |

ANALYSIS

TOP MODEL PERFORMANCE

- Compared Confusion Matrices of both models on test set
- XG Boost outperformed Random Forest in minimizing false positives and false negatives

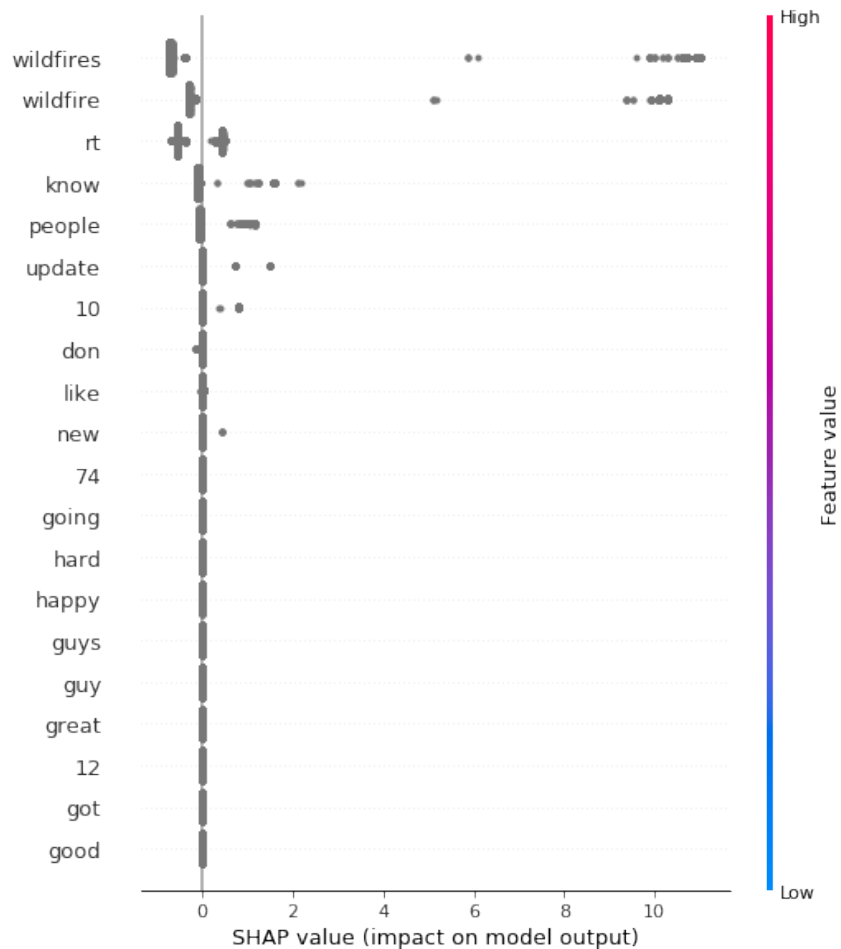
RANDOM FOREST

| | Actual Positive | Actual Negative |
|--------------------|-----------------|-----------------|
| Predicted Positive | 173,875 | 34 |
| Predicted Negative | 0 | 6 |

XG BOOST

| | Actual Positive | Actual Negative |
|--------------------|-----------------|-----------------|
| Predicted Positive | 173,909 | 0 |
| Predicted Negative | 0 | 6 |

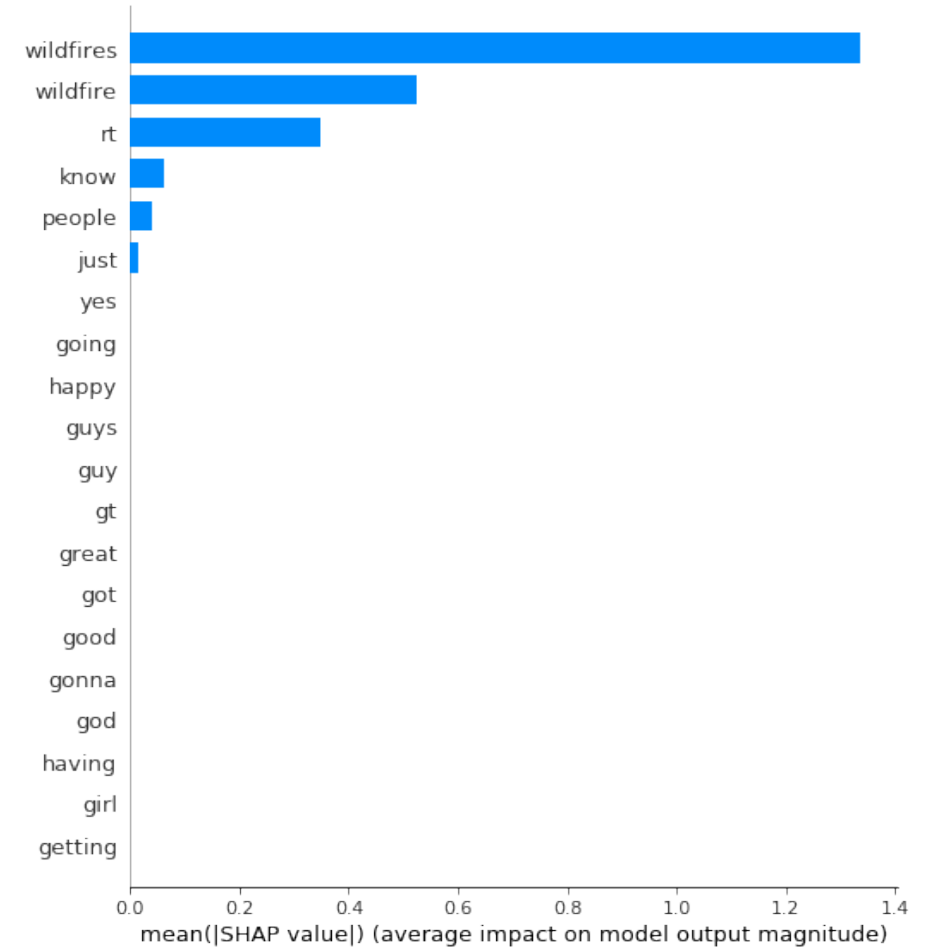
ANALYSIS



SHAP VALUE ASSESSMENT

Measures the impact of variables taking into account the interaction with other variables.

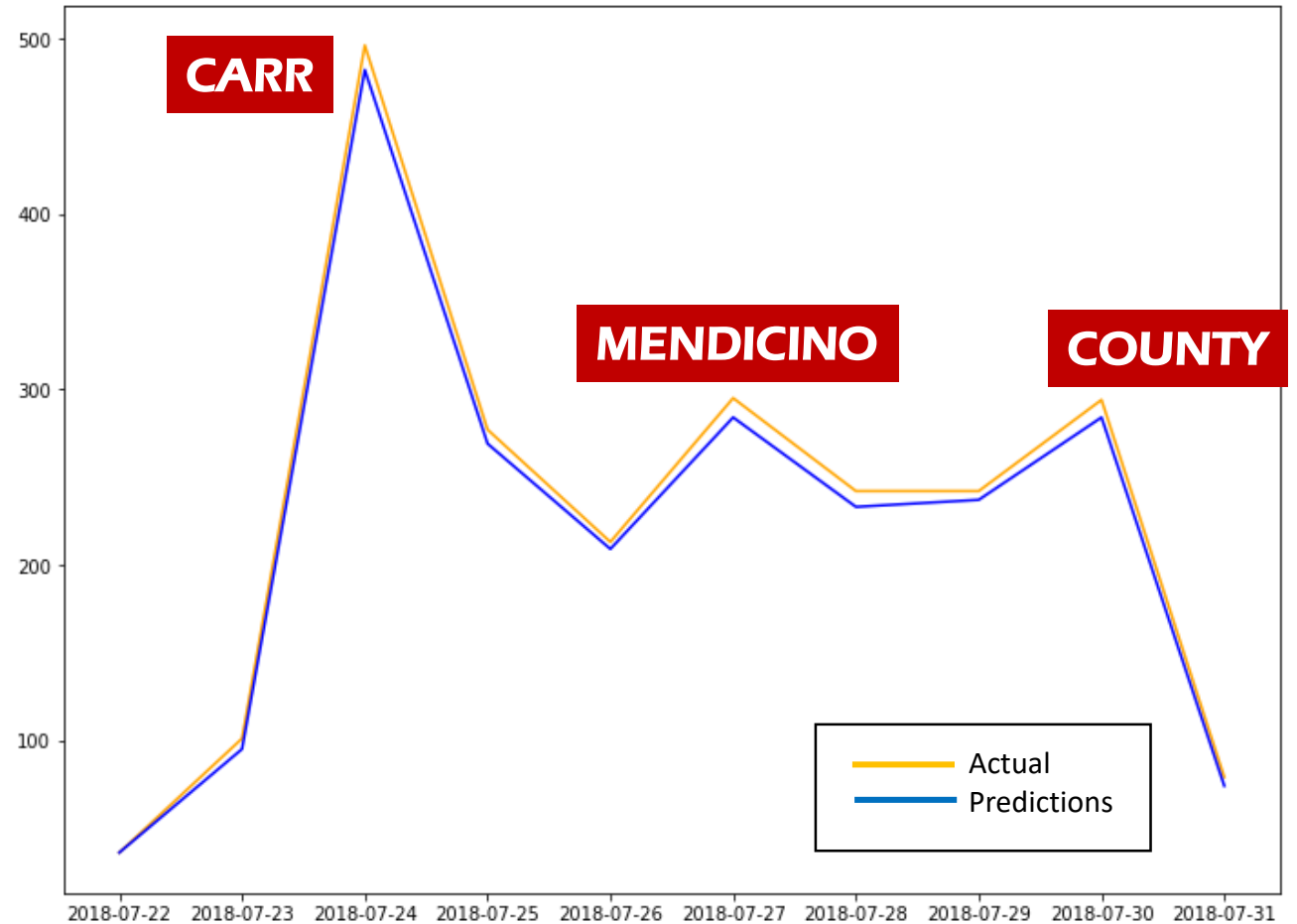
Calculates feature importance by comparing model prediction with and without the feature.



ANALYSIS: TIMESERIES

- Period analyzed: July 22 to 31
- Three clear spikes of twitter activity
- Spikes correlated twitter content and frequency with start of fire activity
- Leveraged largest spikes to train our models

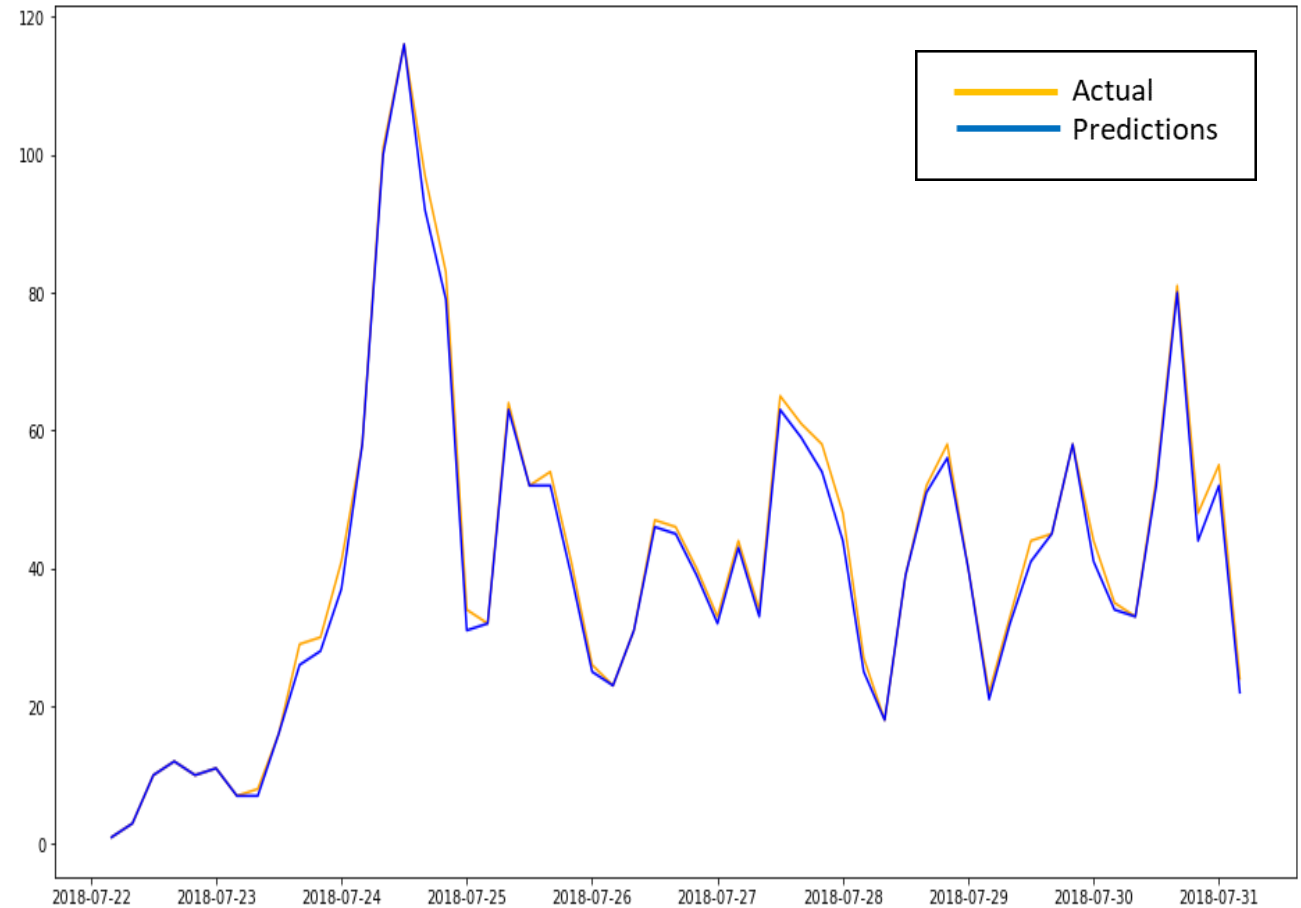
PREDICTED VS. ACTUAL TWEETS BY DAY
JULY 22 to 31



ANALYSIS: TIMESERIES

- **Period analyzed: July 22 to 31**
- **Clear spikes of twitter activity**
- **Spikes correlated twitter content and frequency with start of fire activity**
- **Close alignment of actual and predictions**

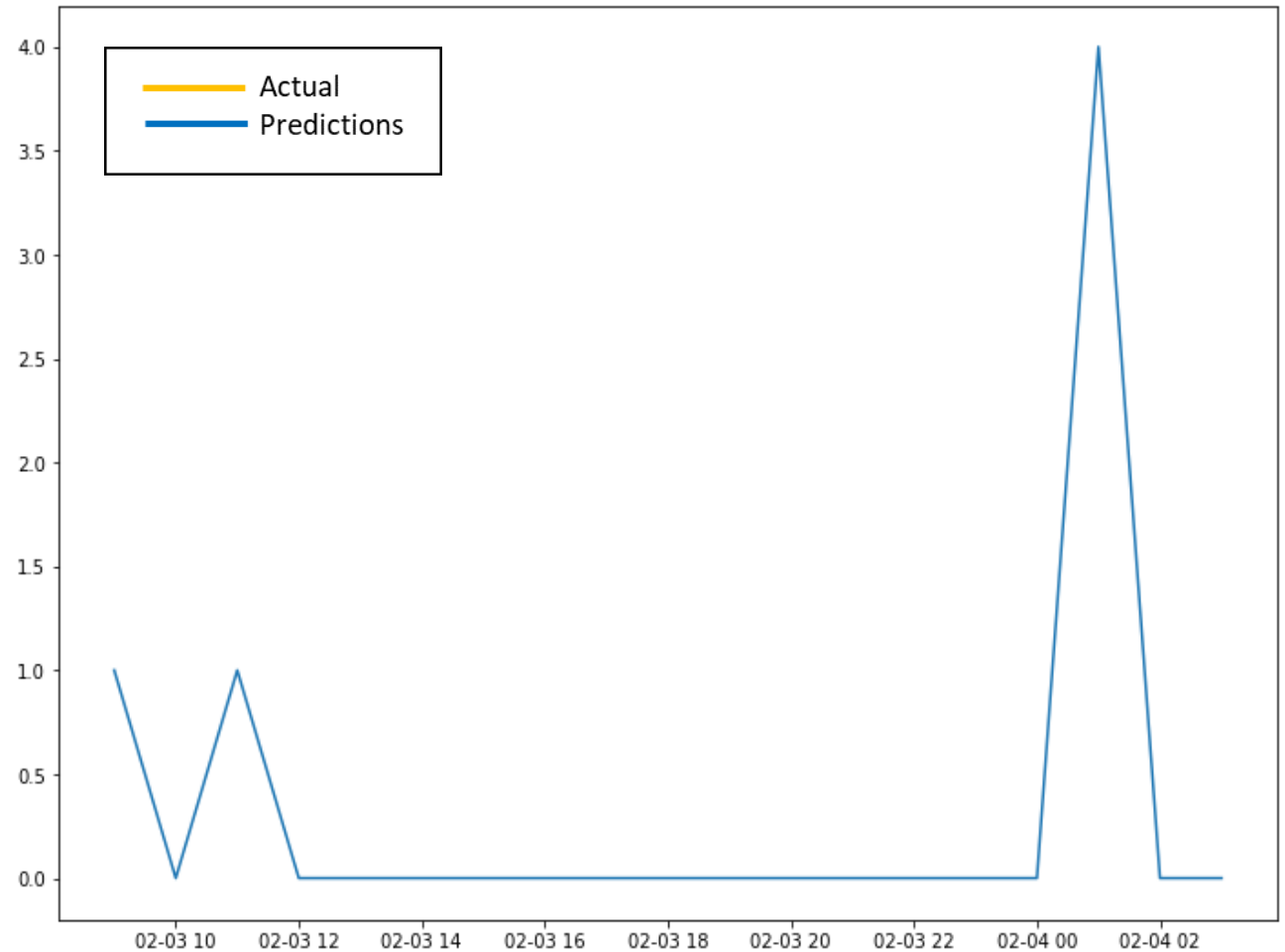
**PREDICTED VS. ACTUAL TWEETS PER DAY
MEASURED EVERY 4 HOURS - JULY 22 to 31**



ANALYSIS: TIMESERIES

- **Period analyzed: February 3, 2019**
- **Confirmed day with no wildfire activity**
- **Chart reflects tweets per hour throughout the day**
- **Predicted and Actual aligned**

PREDICTED VS. ACTUAL TWITTER ACTIVITY February 3, 2019



CONCLUSION/NEXT STEPS

- **FLIPSWITCH = SUCCESS**
 - Identified and demonstrated the two key elements
 - We can train a signal on content and frequency
- **CONNECT MODEL TO DASHBOARD FOR REAL TIME ALERTS**
 - With more time in development, our next phase would connect our model to an app or web-based dashboard
 - Continuously monitor the twitter stream
 - Indicate when confirmed positive posts were reaching crisis levels

CONCLUSION/NEXT STEPS

- **SEEK TO ADD DETAIL OF LOCATION/GEOCODES**
 - Challenge: Not all twitter users share their location
 - Not universally available through Internet Archive curated collection.
 - Our team believes in keeping data pure
 - Uncomfortable simulating a location
 - May/may not be tied to a specific area of response.
 - If universally validated = an asset to the alert.
- **BUILD OUT CATEGORY MODULES FOR OTHER NATURAL DISASTERS**
 - Each set of keywords for an event to be carefully selected and tested.
 - Our terminology was researched, debated, vetted and tested before integration
 - Take same approach with each natural disaster considered to expand this program.

THANK YOU

**WE ARE READY WHEN YOU ARE.
JUST FLIP THE SWITCH...**

