# Dispatch Radio Mapping

Grant Wilson, J. Hall, Gabriel Perez Prieto

## Agenda

- Problem Statement
- Data Collection and Processing
- Audio Analysis
- Mapping
- Conclusion
- Next Steps

#### **Problem Statement**

- Currently, FEMA identifies areas that require immediate attention (for search and rescue efforts) either by responding to reports and requests put directly by the public or, recently, using social media posts.
- This tool will utilize live police radio reports to:
  - Identify hot spots representing locations of people who need immediate attention
  - Flag neighborhoods or specific streets where the police and first-respondents
    were called to provide assistance related to the event

## **Data Collection and Processing**



DOWNLOAD AUDIO



PARSE AUDIO



SPEECH TO TEXT



EXTRACT STREET NAMES





- BroadCastify San Francisco City Police Dispatch Feed
  - O "The world's largest source of Public Safety, Aircraft, Rail and Marine Radio Live Audio Streams"
  - O BART BroadCastify Archive Tool
    - Selenium Library for Scraping
    - Feed id
    - Dates
  - Save MP3 audio files



### **Audio**

- Converting the audio to text for NLP was a challenge, given the quality of the source audio.
- Police are typically active when radioing dispatch, at varying distance from the microphone, with significant noise from:
  - Movement
  - Environment
  - Static
  - Interference
- Many of the transmissions were difficult for a human to interpret without domain expertise in police coding and terminology.
- Posed a significant challenge for the speech to text

### **Audio**

#### Noise

 Fix audio recorded in noisy environments or with poor equipment to create consistency across your recordings.

#### Sibilance

 This is a characteristic of harsh consonant sounds like "s", "sh", "x", "ch", "t", and "th".

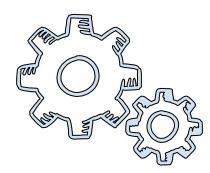
#### Tone

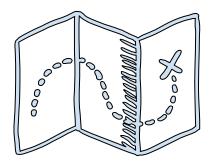
- o Police radio transmits audio on a very thin range of sound, sounding tinny
- Help with equalization to shape the audio from your recorded files to match a listener profile and audio that the text-to-speech model was trained on.

### **Audio**

- Live Audio
  - Disaster mapping in Real Time
- Input can be microphone or direct line in from a radio







## Parse and Segment Saved Audio Files

- Break Audio into Chunks
  - Pydub Library
  - O Detect Audio Average File Loudness
  - O Break if Silence > 4s
  - Convert to WAV
    - 6 s < File > 60 s



## **Speech To Text**

- Google Speech to Text API
  - Transcribe Audio Chunks
  - All streets of San Francisco as a Parameter
    - Give context to Google's API to search for
- Average Confidence on Transcripts
  - O 79%



#### **Extract Street Names**

- Spacy Library
  - Get Street Names from Transcript
  - All streets of San Francisco as a Parameter
    - Give context to search for
- USAdress Library
  - Get Address Numbers
- Google Maps Geocoding API
  - Join Address Number and Street Name
  - O Get Latitude and Longitude



## Example 1

- 'David once again for the night **1910** we **1404 15th Street treat** everyone sits out of can I just want to play for yeah but go back to the temporary place and the Seven Roberts Roberts zebra **211** should come back to black Infinity first like I was like a regular California place it's a paper plate but I have several Robert Roberts it was fabulous party Wright place we try it was he that does go back to a **2004** is that the entity that correct yeah okay'
- Transcript Confidence = 0.8180
- Model Results:
  - 1404 California Ave, 1404 15th Ave, 1404 Treat Ave, 1404 Black Pl

## Example 2

- 'case manager Michelle is **97** 121 Leland a client is experiencing events induced psychosis getting increasingly activated and aggressive and as much customers are impossibly self-induced you doing it for back to the your doesn't get out of here with that license plates some of that about which car are you referring to the **1030** at **1141** of our courts I copy'
- Transcript Confidence = 0.8699
- Model Results:
  - 1141 Lealand Ave

## Example 3

• 'this verse **29** that teacher Elementary is cold report of a **311** occurred last Friday on the Hudson Street side they have pictures not online with these are the secretary okay I'll **98** had a **71** for the **600**'

- Transcript Confidence = 0.8534
- Model Results:
  - o 98 Hudson Ct

## **Mapping**

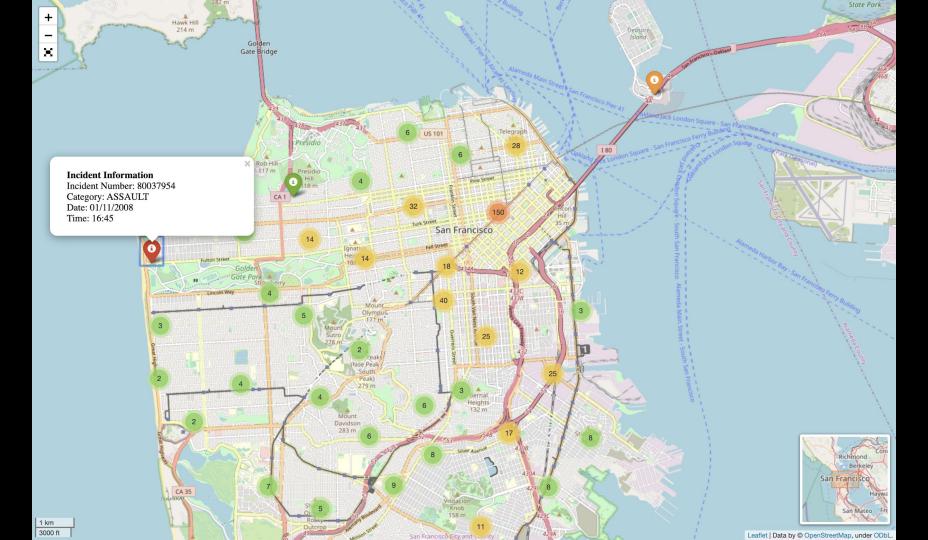
- Folium vs. Google Maps API
  - More creativity in mapping
  - Better represented the data as it relates to key interests of disaster response agencies
    - Surrounding businesses, parks, and bus stops
- Historic San Francisco Crime Data
  - Similar to disaster reporting
    - Time/Date/Location
    - Type of dispatch
      - EMS
      - Police
      - Fire

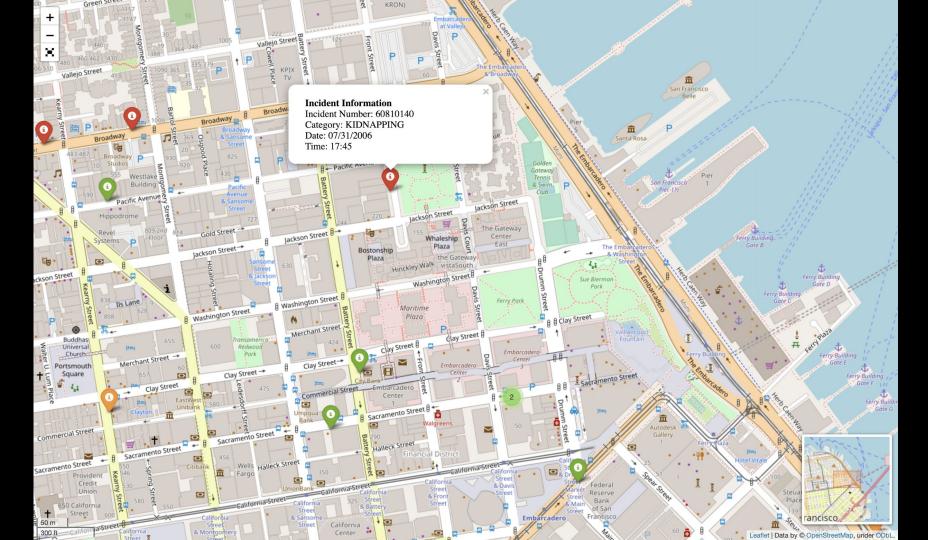
## **Mapping**

- Two maps
  - Individual Detail
    - Point Mapping
  - Neighborhood-area reporting
    - Frequency Mapping

## **Point Mapping**

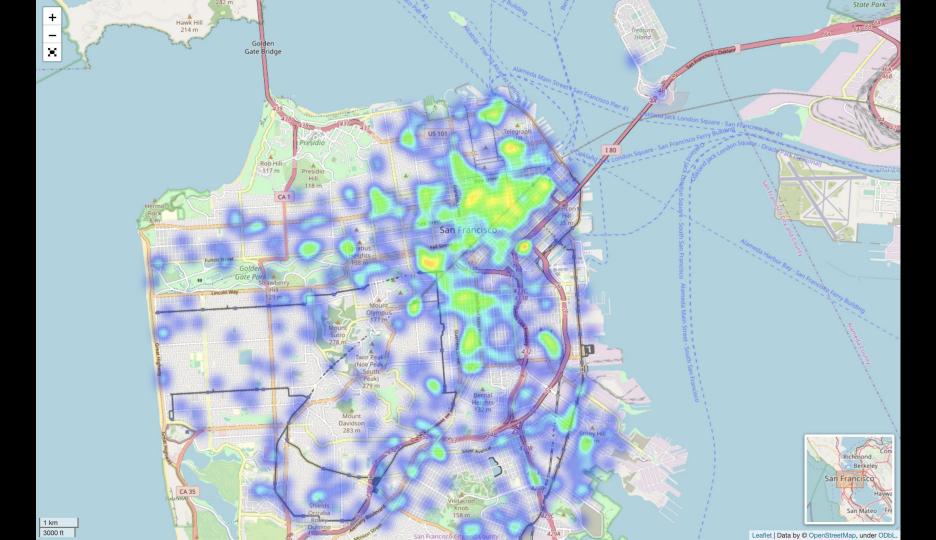
- Drops "pins" for each reported incident
  - Incident Number
  - o Incident Time/Date
  - Response Category
    - Police
    - Fire
    - EMS
  - Incident Category
    - Assault
    - Downed Power Line
    - Fire

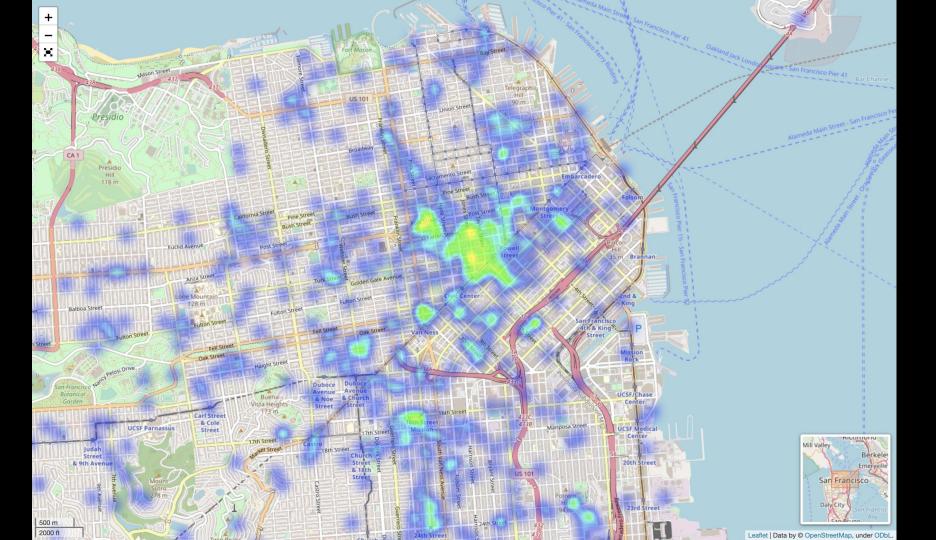




## Frequency Mapping

- Neighborhood-wide incidents
- Quick read on areas requiring assistance
- Balances with detail-oriented Point Mapping
  - Demonstrates call-concentration
  - Neighborhoods needing disaster-assistance





#### **Conclusion**

- Improvements
  - Audio Analysis
  - Live Audio
  - Point Mapping
  - Frequency Mapping
- Utility
  - Identify key neighborhoods
  - Facilitate native integrating with dispatch radio
  - Cleaner address analysis
  - o Isolate Critical Infrastructure weaknesses
    - Neighborhood analysis of incidents

## **Next Steps**

- Clean audio before processing
  - Dolby
- Improve NLP
  - Sentiment/Urgency Analysis
- Address Extraction
- New avenues for speech to text
- Include damage assessment