

Hundreds of buildings have fires in Atlanta every year. The Atlanta Fire Rescue Department (AFRD) attempts to reduce fire risk by inspecting buildings for potential hazards and fire code violations, but they currently only inspect a subset of the total buildings needing inspection. Our project aims to reduce fire risk in Atlanta by identifying and prioritizing buildings that should be inspected by AFRD.

Fire incidents heat map (2011-present)



Goal 1: Find new properties to inspect

- <u>List of new properties</u>: from external business and property databases
- <u>Prioritized list</u>: using risk scores from the model
- <u>Interactive map</u> to view inspected properties, fire incidents, and potential inspections in Atlanta

Goal 2: Prioritize inspections

- <u>Integrated database of buildings</u> with the most complete property information
- <u>Make a predictive model</u> to generate risk score for properties





SIZE: >2 GB, ~160,000 Records

We used GIS to join and merge various datasets. The merged datasets can help us to identify the type of businesses that require permits and understand the building features that may explain the occurrence of fires.

Machine Learning Algorithms

We are using several machine learning algorithms to predict fire incidents. Linear Regression, Logistic regression, random forest tree, and Support Vector Machine (SVM) models are all tested and SVM is selected as the most appropriate model. More than 250 variables are tested and 58 variables are included in the final results.

D3.js Data Visualization

To make it easier to interpret the model results, we are going to use d3.js library to visualize the model and analysis results.

Final Results

Finding Potential Inspections (Goal 1)



- Joined business licenses and current inspections to obtain the business types of current inspections using:
- Geocoding
- Fuzzy text matching of business names
- Identify business that are similar to the current inspected ones
- Discover other similar businesses via



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Data	Source
Fire Incidents (AFRD)	Atlanta Fire Dep
Fire Permits Data (FSAF)	Atlanta Fire Dep
CoStar Data	Atlanta Fire Dep
Neighborhood Planning Unit	ARC*
Parcel Data	City of Atlanta
Atlanta Business License	City of Atlanta
SCI Data	City of Atlanta
Business Location Data*	Google Places A
Address Based Coordinates	Google Geocode
Demographic Data	U.S. Census Bur
Socio-economic Data	U.S. Census Bur

(AFKD)	Atlanta Fire Department
ata (FSAF)	Atlanta Fire Department
	Atlanta Fire Department
Planning Unit	ARC*
	City of Atlanta
ss License	City of Atlanta
	City of Atlanta
ion Data*	Google Places API
Coordinates	Google Geocode API
Data	U.S. Census Bureau
c Data	U.S. Census Bureau

*ARC: Atlanta Regional Commission

*The type of business (and number) obtained from Google Place API includes: Bars (629), car repairs (686), laundry (280), liquor stores (117), night clubs (163), painter (94), plumber (167), restaurant (1661), roofing contractors (139), schools (617), and hospitals (441).

Predictive Fire Risk Model (Goal 2)

Join Dataset



Final Data for Model Development







Google Places API

• Text-mining of the *Fire Code of Ordinances*

Top 6 Current Inspected Business Types and Potential Inspections



Interactive Inspection Map



Machine Learning Results: Support Vector Machine Model

Model 1 (10 bootstrap result)

Data from 2011-2014 and applied to 2014-2015 fire Average accuracy: 0.77 | Average AUC: 0.75

Actual	1	false negatives (had fire; predicted no fire) n = 38 0.2669	true positives (had fire; predicted fire) n = 104 0.7331
	0	true negatives (no fire; predicted no fire) n = 1577 0.7678	false positives (no fire; predicted fire) n = 468 0.2322
		0 Pred	1 icted

Predicative Factors

Location	NPU (Neighborhood Planning Unit), zip code, submarket, neighborhood, tax district	
Land/ property use	property/business type, land use codes, zoning	
Financial	tax value, appraisal value	
Time-based	year built, year renovated	
Condition	lot condition, structure condition, sidewalks	
Occupancy	vacancy, units available, percent leased	
Size	land area, building square feet	
Building	ding number of units, style, stories, structure,	
	construction materials, sprinklers, last sale date	
Owner	owner or property management company,	
	owner's distance from the location of property	

Model 2 (10 cross validation result)

- Data from 2011-2015
- Average accuracy: 0.78 | Average AUC: 0.73

Actual	1	false negatives (had fire; predicted no fire)	true positives (had fire; predicted fire)
		n = 16 0.3244	n = 39 0.6756
	0	true negatives (no fire; predicted no fire)	false positives (no fire; predicted fire)
		n = 609 0.7914	n = 158 0.2086
		0 Pred	1 icted

Apply Model to Potential Fire Inspections

Predicative fire risk ranging from 0-1 • had fire and rescaled to 1-10 **0** no fire







