Forecasting Street and Sidewalk Cleaning Services in San Francisco

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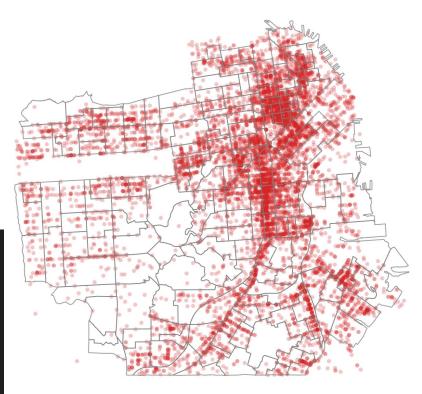
San Francisco 311

"Street and sidewalk cleaning requests are generated internally and through calls received by the City's 311 customer service center. [...] Public Works' Radio Room triages the request to the appropriate crew." (source)

We examine call counts at the hour level between

1/1/2009 and 12/31/2022:

	Descriptive Statistics
count	122712.000000
mean	17.626622
std	22.591759
min	0.000000
25%	2.000000
50%	8.000000
75%	25.000000
max	553.000000
Name:	calls, dtype: float64
Sum:	2162998



Goals and Applications

Our goal: Use machine learning to predict the number of requests that the city will receive in any given hour.

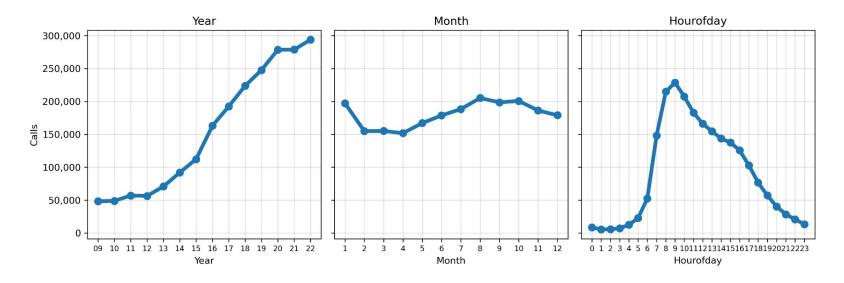
Policy applications: Understand how much staffing will be needed for the 311 call center and public works street cleaning teams at a given time, anticipate sidewalk and street cleaning costs.



SF Public Works Clean Corridors

Approach and Challenges

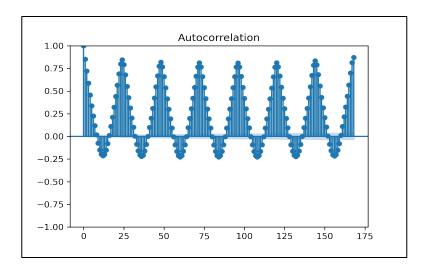
- 311 grew considerably during this time period.
- We were concerned it would be difficult to forecast within year trends with such a strong secular trend across years.
- We focused on capturing hour-level patterns in our modeling

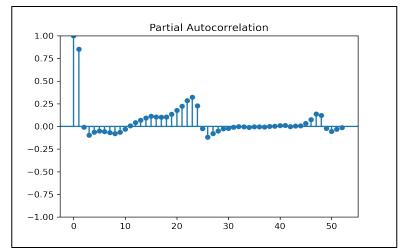


Feature engineering

- Autocorrelation significant hourly at the week level (and beyond)
- Partial autocorrelation significant hourly up to about 2 days

Set up models to use 48 hours of previous call data.



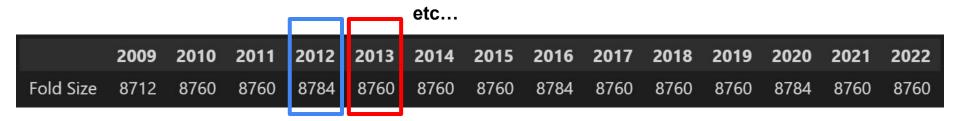


Data and Cross Validation Approach

Lagged Dataset (48 hours)

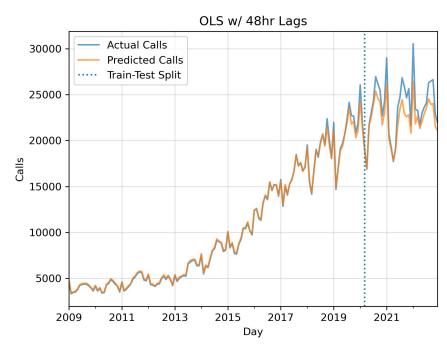
	calls	lag_1	lag_2	lag_3	lag_4	lag_5	lag_6	lag_7	lag_8	lag_9	 lag_39	lag_40	lag_41	lag_42	lag_43	lag_44	lag_45	lag_46	lag_47	lag_48
datetime																				
2009-01-03 00:00:00	0	6.0	0.0	1.0	1.0	3.0	5.0	19.0	20.0	19.0	6.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2009-01-03 01:00:00	0	0.0	6.0	0.0	1.0	1.0	3.0	5.0	19.0	20.0	4.0	6.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2009-01-03 02:00:00	1	0.0	0.0	6.0	0.0	1.0	1.0	3.0	5.0	19.0	10.0	4.0	6.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0
2009-01-03 03:00:00	3	1.0	0.0	0.0	6.0	0.0	1.0	1.0	3.0	5.0	8.0	10.0	4.0	6.0	1.0	0.0	0.0	0.0	0.0	0.0
2009-01-03 04:00:00	0	3.0	1.0	0.0	0.0	6.0	0.0	1.0	1.0	3.0	5.0	8.0	10.0	4.0	6.0	1.0	0.0	0.0	0.0	0.0

Cross Validation: Train, Test, standardize, clip at ±3std, compute metrics



Next-Hour Forecasting

- Given the last 48 hours of data, how many street and sidewalk cleaning calls will come in the next hour?
- This short-term forecasting would be used to prepare dispatchers and cleaning crews



(80-20 train test split pictured for clarity)

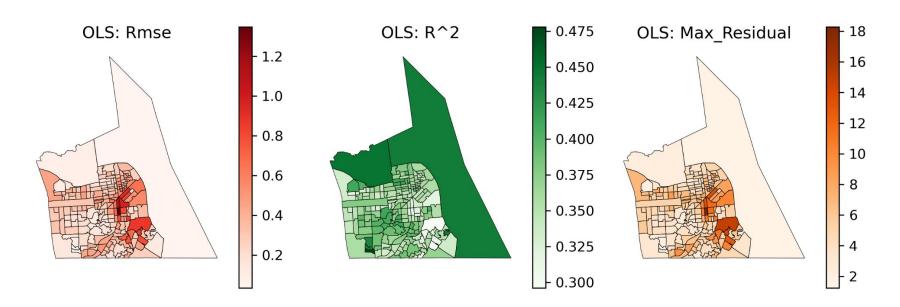
Next-Hour Performance Summary

Model	Avg RMSE	Avg R^2	Avg max residual
Baseline	12.58	0.379	171
Linear regression	8.29	0.595	157.33
Lasso Regression (a=0.0001)	8.28	0.594	157.30
Random Forest (max_depth=None, estimators=200, etc.)	7.95	0.606	153.60
RNN	62.13	0.18	398.465

Average metrics are computed across year-folds. All models have optimized hyperparameters for each fold's predictions

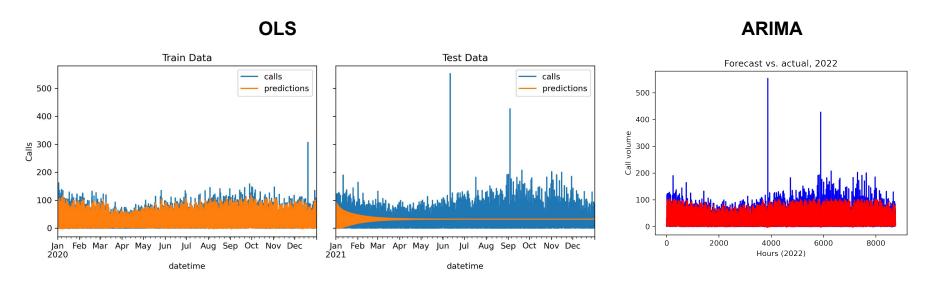
Next-Hour Neighborhood-level Forecasting

- Given the last 48 hours of data, how many street and sidewalk cleaning calls will come in the next hour, and where will they be?
- These forecasts would be used to dispatch cleaning crews preemptively



Long-Run Forecasting

- Given the last 48 hours of data, how many street and sidewalk cleaning calls will come in the next year?
- This is an exploration of how far we can predict trends before quality degrades significantly



Long Run Performance Summary

Model	Avg RMSE	Avg R^2	Avg max residual
Baseline	22.97	-0.75	179.33
Linear regression	17.46	-0.041	166.23
Random Forest (max_depth=None, estimators=200, etc.)	20.277	-0.394	170.54
ARIMA	11.37	0.56	161.376

Average metrics are computed across year-folds. All models have optimized hyperparameters for each fold's predictions