## Project Update 2

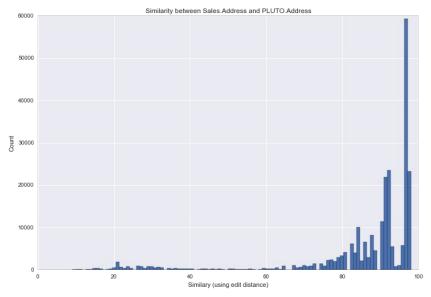
Ben Jakubowski Haonan Zhou

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- Previously work:
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    - scraped data for 12626 current listings.
  - Attempted to merge public datasets
     (NYC PLUTO and Annualized Sales data) yielding approximately 360k sale records from 2011-2015.
  - This merge failed for condos and still yielded a feature poor dataset.



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Needed to develop scraping strategy to avoid being blocked.

#### Solution



# Scrapy: Run Using TOR and Multiple Agents

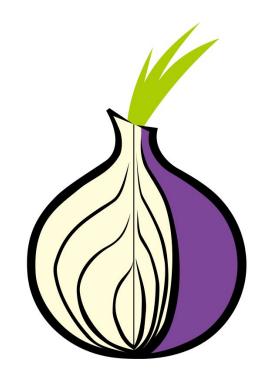
Mar 18th, 2013

Scrapy is a brilliant and well documented crawler written in python. Though it is not as scalable as Apache Nutch but it can easily handle thousands of sites easily. You can get up and running very quickly using the official documentation. Tor gives you power to keep your privacy and security. Tor can hide you so that website can not track your identity. You may read more about TOR in official site. However Tor only works for TCP streams and can be used by any application with SOCKS support.

#### Solution



An open source and collaborative framework for extracting the data you need from websites. In a fast, simple, yet extensible way.

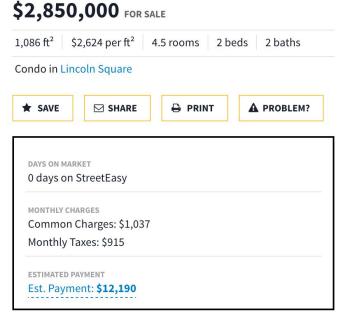


#### StreetEasy Data

- Saved HTML source of all scraped StreetEasy sales pages as text files
  - We chose to save the entire source since we anticipate scraping being a bottleneck, and didn't want to have to re-scrape if we later decided to try additional features from the sale pages.
- Wrote python script to parse the HTML files using BeautifulSoup
- Filtered out all records for properties outside of New York City
- Yielded 372699 records from the current set of 491581 scraped pages
- Current implementation of feature extraction script parses 50000 html pages per hour using a single CPU core
- Trivial to run script in a parallel computing framework to reduce runtime

#### StreetEasy Data - Features

#### 45 West 67th Street #26B



#### **DESCRIPTION**

Location and PARK Views!!! Stunning views from this Renovated 2 Bedroom 2 Bath Condo that has Eastern and Southern Exposures. The Windowed Chef's Kitchen, has Stainless Steel Appliances. There are beautiful hardwood floors throughout. The 2nd bedroom has been turned into a den with lots of built ins, 45 West 67th Street is a boutique smoke free condo in the heart of Lincoln Center and steps to Central Park. Great shopping and restaurants await you!

#### **HIGHLIGHTS**



#### StreetEasy Data - Features

#### **AMENITIES** BUILDING AMENITIES Bike Room Valet Concierge Laundry in Building Live-in Super Storage Available BUILDING 45 West 67th Street New York, NY 10023 Condo in Lincoln Square Built in 1983 173 units SALES LISTINGS: 2 active and 74 previous RENTALS LISTINGS: 5 active DOCUMENTS AND PERMITS: 266 documents MORE ABOUT THE BUILDING

**NEARBY** 

TRANSPORTATION

#### **Subways**

- 1 at 66th St under 500 feet
- B c at 72nd St 0.19 miles
- 1 2 3 at 72nd St 0.3 miles
- A C B D 1 at 59th St-Columbus Circle 0.56 miles
- B c at 81st St **0.57 miles**

View subway lines on Google Maps▶

#### StreetEasy Data - Features

#### List of features extracted from HTML pages:

- saleid
- address
- price
- neighborhood
- borough
- status
- date
- num beds
- num\_baths
- num\_sqft
- type

- url
- monthly\_cost
- amenities list
- transit list
- gps coordinates
- school\_district
- building\_name
- built date
- building\_num\_units
- building\_url
- description

#### Terminal state of HTML pages:

•	sold	187975
•	no_longer_available	151486
•	delisted	23249
•	temporarily_off_market	5461
•	current	1724
•	expired_owner_listing	1472
•	in contract	1332

#### Number of listings by borough:

	manhattan	205510
•	brooklyn	100874
•	queens	36325
•	staten_island	17528
	hronx	12462

#### Most popular neighborhoods:

•	upper_west_side	19253
•	upper_east_side	14007
•	lincoln_square	13441
•	lenox_hill	12648
•	yorkville	11577

#### Most popular types of listings:

•	co_op	121996
•	condo	96294
•	resale	40517
•	sponsor_unit	33640
•	house	25117
•	multi_family	23104
•	townhouse	17007
•	apartment	4863
•	condop	4014

#### Listings bucketed by square footage:

•	NaN	126549
•	1000	102990
•	2000	92580
•	3000	27836
•	4000	12493
•	5000	4882
•	6000	1917
•	7000	1179
•	8000	604
•	9000	453
•	10000	257
•	11000	204

124

12000

Listings bucketed by price per square foot:

•	250	20398
•	500	45083
•	750	45768
•	1000	43576
•	1250	34316
•	1500	21391
•	1750	12489
•	2000	7566
•	2250	4687
•	2500	2975
•	2750	2116
•	3000	1365
•	3250	993

712

3500

Listings in Manhattan bucketed by price per square foot:

•	250	741
•	500	5580
•	750	20308
•	1000	32229
•	1250	29847
•	1500	19958
•	1750	12045
•	2000	7416
•	2250	4630
•	2500	2953
•	2750	2088
•	3000	1355
•	3250	989

707

3500

• Featurization: Examples include:

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#### **HIGHLIGHTS**



Cats Only - No Dogs 📀



Elevator 📀



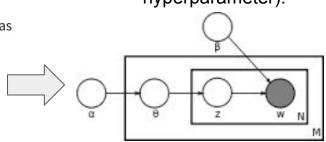
Full-time Doorman

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Use LDA to preprocess description, optimizing K (number of topics) as hyperparameter).



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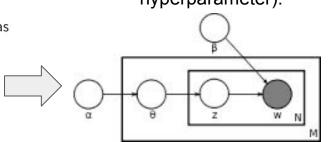


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One-hot encodings of highlights/other categorical features.

- Modeling:
  - Model 1:
    - Model sale prices citywide, using geographic features as predictors to allow for geographic effects in model.
    - Augment obvious features (i.e. number of bedrooms, square feet)
       with topics from LDA
    - Use nonlinear models (random forests, boosted regression trees) to allow for complex interactions.

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  - Model 2:
    - Model each neighborhood separately using simple linear regression model.
    - If competitive, this approach will allow for comparisons between neighborhoods' sale price sensitivity to different predictors.