2016 CampusLab

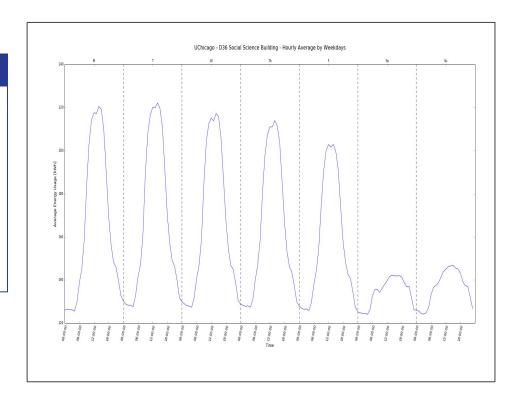
Detecting Unusual Trends in Electricity Usage Data



University of Chicago Energy Data

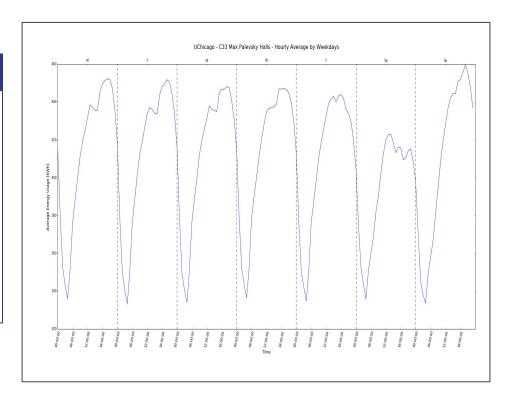
Data

- UChicago is one of the largest energy consumers in Chicago (with O'Hare)
- ~120 buildings
- Over 300 meters
- Meters and buildings do not always have a clear relationships



University of Chicago Energy Data

- Right now data is available at half hour intervals, but moving to more granular readings
- Purchase power hourly
- How would we detect if usage in a building is abnormal?
- Goals: Pattern & Anomaly Detection



Tools and Implementation

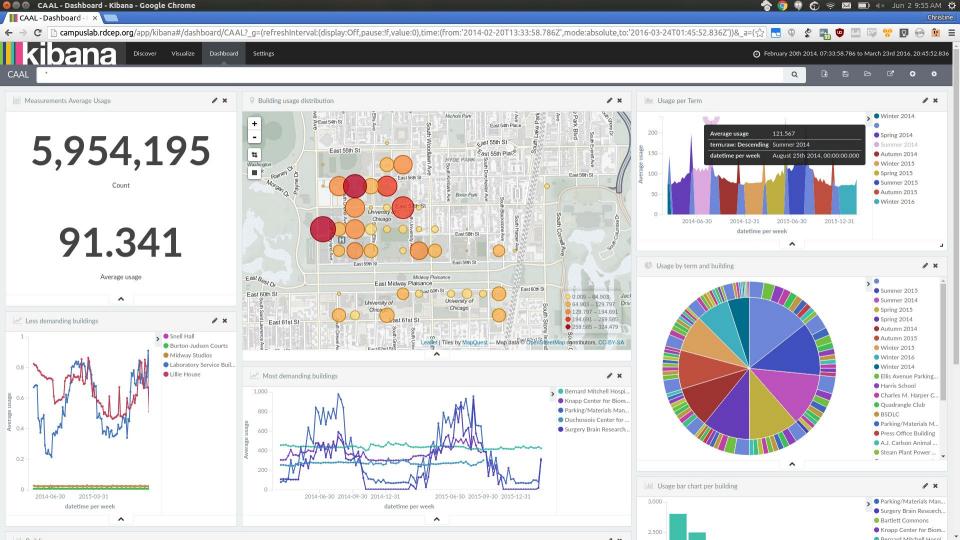
Tools

- PostgreSQL
- Amazon EC2, EMR, and S3
- ElasticSearch & Kibana
- MRJob
- Python Threading and Queue Class

Algorithms

For anomaly detection:

- Tarzan (Suffix Trees, Markov Models)
- K-Nearest Neighbors



Data Flow

Storage

AWS EC2 with Postgres

AWS EMR with ElasticSearch Extraction & Transformation

MRJob

K-NN

Vectorization

Tarzan

Discretization

Analysis

MRJob

Multithreading

Pandas

Algorithms: Tarzan

```
void Tarzan (time_series R, time_series X, int l_1, int a, int l_2, real c)
let x = \text{Discretize\_time\_series}\ (X, l_1, a)
let r = \text{Discretize\_time\_series}\ (R, l_1, a)
let T_x = \text{Preprocess}\ (r, x)
for i = 1, |x| - l_2 + 1
let w = x_{[i,i+l_2-1]}
retrieve z(w) from T_x
if |z(w)| > c then print i, z(w)
```

Table 4: Outline of the Tarzan algoritm: l_1 is the feature window length, a is the alphabet size for the discretization, l_2 is the scanning window length and c is the threshold

Tarzan Algorithm

Finding Surprising Patterns in a Time Series

Database in Linear Time and Space – Keogh et al. SIGKDD 2002

Step 1:

Discretize time series into strings

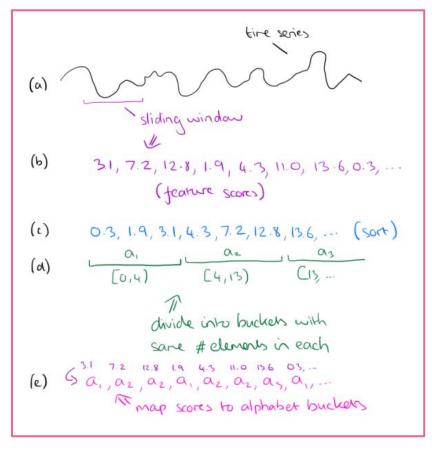
Step 2:

Preprocess strings with Suffix Trees

Step 3:

Score strings with Markov Models

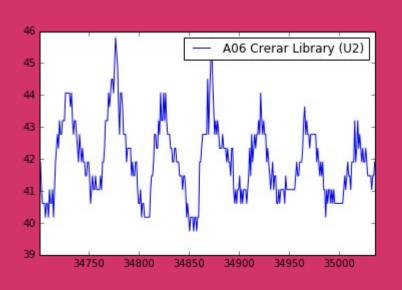
source: http://www.cs.ucr.edu/~eamonn/sigkdd_tarzan.pdf



Discretization

- Slide window over feature scores.
- 2. Generate one real number (slope)
- 3. Bucket slopes by quantiles
- 4. Assign symbol to each bucket
- Replace numbers in original time series with label from bucket

Step 1: time series ----> string





edaafmhkkkjgjklmonkkippooohhafdeelib bgkgccebahphdejneileahplnnnmmoonn oojkllgfggkkdeccdgphcddfhpfedjggfhba hhhommpmmlalnljjhjkgfeaaabaipjabaab aelidedccmpbfmjfmlkloophhkjjjkjggfdob bbapleabfcebkngffcnmoggdnmpafnmlli kjoomplljgebedddcbcnjggfcfmnmphakk kahhhhhpoofhpppmmkgdebfmphhhcd dccbnddeggnmhfcnceahhhhhhpononof bknnljhjljdccceilcbbhaelp

Algorithms: K-Nearest Neighbors

K Nearest Neighbors

- Define a window for comparison and k parameter
- For each building, compare the readings from all other buildings to find the k nearest neighbors
- Find what that building is doing during its 'nearest neighbors'
- Calculate distance to between those readings and the current readings
- Repeat for each building

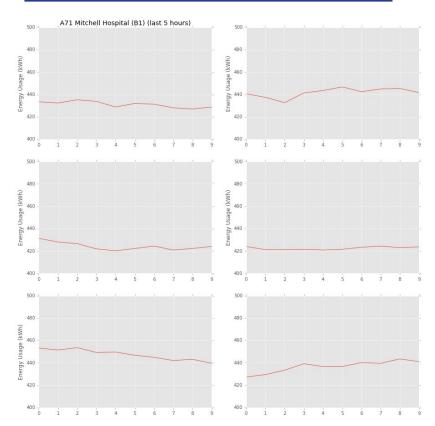
All Other Buildings

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	'						
6.5	2.2	4.5	7.6	3.4	9.1	 	
3.2	2.4	5.7	8.4	2.8	1.0	 	
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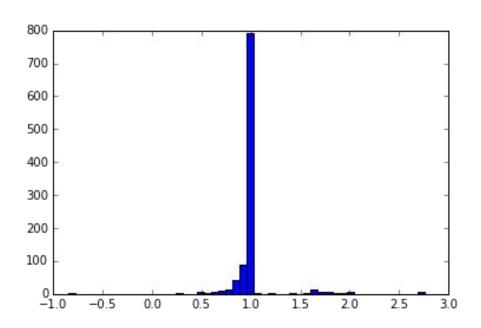
K Nearest Neighbors

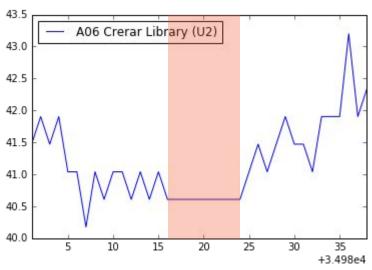


- Nearest Neighbors for Mitchell Hospital meter
- Distance between current readings (upper left) and average of 5 nearest neighbors:
 - 0 14.146

Results

Distribution of "Surprise Scores"





	datetime	A06 Crerar Library (U2)
34995	2016-04-10 03:00:00	41.040
34996	2016-04-10 03:30:00	40.608
34997	2016-04-10 04:00:00	40.608
34998	2016-04-10 04:30:00	40.608
34999	2016-04-10 05:00:00	40.608
35000	2016-04-10 05:30:00	40.608
35001	2016-04-10 06:00:00	40.608
35002	2016-04-10 06:30:00	40.608
35003	2016-04-10 07:00:00	40.608
35004	2016-04-10 07:30:00	40.608

"Surprise Result"

April 10, 2016 | Crerar Library

4 hours of unchanging usage from 3:30am - 7:30am

Run Times

K-NN ~ 31 min

K-NN with MR Job ~ 26 min

Tarzan ~ 17 min

Tarzan w/ Multithreading ~ ∞ min

Next Steps

Sharding with EMR and ElasticSearch

Effectively use MR Job and Multithreading with both Algorithms

Reduce Run Times

Questions?