# Bike share traffic predictions using machine learning 

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## Agenda

- Introduction to bike-sharing
- Motivation and vision
- A short introduction to machine learning
- Overview of software
- Results
- Conclusion



## Bike-sharing

## Why bike-sharing

## The problem

## Back Reviews Write a Review

## "Great fun but.!!"

OOOOO Aug 28, 2013

## Fatima M (1 Review) <br> London, United Kingdom

We hired the bikes for 4 people and had an amazing time! The only catch is that sometimes (specially over weekends) when you want to place the bikes back into their docking stations, you might not be able to get the space. We tried three different stations but couldn't find a single empty space and were thoroughly exhausted after the whole day. Because we were stationed in london for the night at a friends' place, we had to park the bikes in their apartment overnight and only in the morning did we find empty spaces in the nearest docking stations. So great fun and highly recommended but you might want to get back to the docking stations in, shall I say a "lucky" time :)

Above: A customer reviews London's bike-share system on the tripadvisor website
"A good way to get around - but don't expect to find a bike every time."
OOOOO Reviewed July 30,2013
First off: hats off to any city that's doing its bit to promote this kind of transport system. Washington's heart is certainly in the right place. $\$ 7$ for a day of travel around the sights of the National Mall and further afield to Arlington and Georgetown is certainly good value. To clarify: you have 30 minute window to get your bike docked into another station, not the 90 as one review has it, though you do get a 'safety' 15 minutes if you find all of the stations full when you reach your destination. Given that the metro is $\$ 2.70$ a ride (when including the paper card) you can save yourself some money and sore feet with the bike.

The bikes are simply and easy to ride, and you can get a reasonable shunt going if you're so inclined, though I'm not sure why you'd want to unless you're a DC commuter heading into work, and then why are you reading this?

Here's the health warning: there are lots of tourists and bikes can become a premium in certain places (the Lincoln Memorial for example) so you can't expect to rock up and find one. The app is good, but doesn't update strictly in real time: by the time you reach the station the bikes might be gone. Some people are quite canny at dropping a bike then swiping for another 30 minutes, so pairs of bikes are sometimes hard to come by.

Personally, I wouldn't plan a day around using a bike, especially if you have are under time pressure to get back somewhere, but if you can afford to be a bit more relaxed this might be a fun, and useful, bonus attraction.

> Above: A customer reviews Washington's bike-share system on the tripadvisor website

## Users currently have real-time systems



## The vision



## Related work

- Data science for social good (predicting bike-share usage in Chicago's Divvy bike system)
- Jake VanderPlas (modelling the effects of weather on bike usage in Seattle)


## Machine learning



## Training set

## Sunny <br> Downtown <br> Tuesday 8:00 AM <br> 11 bikes

## Sunny

Downtown
Tuesday 11:00 AM
0 bikes
Rainy
Downtown
Tuesday 8:00 AM 2 bikes
Sunny
Downtown
Tuesday 11:00 AM
2 bikes
Sunny
Downtown
Tuesday 1:00 PM
1 bike

## Test set

## Sunny Downtown Tuesday 8:00 AM 11 bikes

## Sunny

Downtown
Tuesday 11:00 AM
1 bike
Sunny
Downtown
Tuesday 8:00 AM 10 bikes
Sunny
Downtown
Tuesday 1:00 PM
Sunny
Downtown
Tuesday 2:00 PM
1 bike

## Software overview



## Libraries used

- Scikit-learn (machine learning algorithms)
- Pybikes (data collection) to collect data from the Washington bike-share system


# Machine learning <br> algorithms 

## Decision Trees



## Random Forests

## Random Forests

- Lots of decision trees
- Output given by the average of the output of all trees in the forest
- Cannot overfit by adding more trees (note: RF can overfit on noisy datasets when there are too few trees!)

Ada Boost

## AdaBoost

- Analogy: student preparing for an exam in physics
- Topics covered: classical physics, thermodynamics, electromagnetism, quantum physics
- They start by doing a practice exam
- They notice they didn't do well on electromagnetism. Ignore all other topics until they grasp electromagnetism.
- Do another practice exam
- Repeat... until it's time for the exam


## Thesis contribution

## Data collection using Pybikes

Feature selection

- Feature importances: Decision tree (RMSE =5,2)


Feature importances: Random forest (RMSE =4,2)


■ Feature importances: Adaboost (RMSE $=3,9$ )


# Why is the "epoch" so important? 

A missing time-related feature that has not been accounted for.

## Genetic algorithms

- Hyperparameters - algorithm configuration
- Can use GA to pick the "optimal" feature set that provides the best prediction performance
- GAs did not improve the accuracy over manually picked hyperparameters

Results

# A customizable machine-learning package for predicting bike-share usage 

Estimator: DecisionTreeRegressor, RMSE $=5.32291481276$




## Improvements on existing solutions?

## Error metric: RMSE

## Improvement



## Further work

## The vision




```
Settings Predictions Alerts (4)
Prediction date(s) 10.12.2014-11.12.2\ldots * Weather Forecast API www.wunderground.com/v
Bike-share system Washington DC * Station to predict 59: Florida Ave & R St. &
```



## Bike-share traffic prediction

Settings Predictions Alerts (4)

Stations predicted to be empty:
73. L'Enfant Plaza / 7th \& C St SW (in 30 mins )
78. Tenleytown / Wisconsin Ave \& Albemarle St Nw (in 15 mins)

Stations predicted to be full
95. Bladensburg Rd \& Bening Rd NE (in 140 mins)
55. Potomac \& Pennsylvania Ave SE (in 30 mins )

## Bike-share traffic prediction

| Settings | Predi |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Prediction date(s) |  | 10.12.2014-11. | $\uparrow$ | Weather Forecast API | www.wunderground.com/T |  |
| Bike-share system |  | Washington DC | $\dagger$ | Station to predict | 59: Florida Ave \& R St. |  |



