



Visualizing and Predicting Ridership

It can be tough to find a Bikeshare bike when you really need one.



Fortunately, there is plenty of data out there to help us understand the issue.

- Trip data
  - Ten years of rides
- Dock data
  - Three weeks of dock statuses



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- Weather data
    - Ten years of temperatures and rain
  - Neighborhood data
    - Current neighborhood clusters



I set out to determine what drives ridership.  
Here we see strong seasonality.

Number of Trips/Month



Here we see that riders behave differently on weekdays and weekends.



I built a model that predicts CB's hourly ridership.

## In

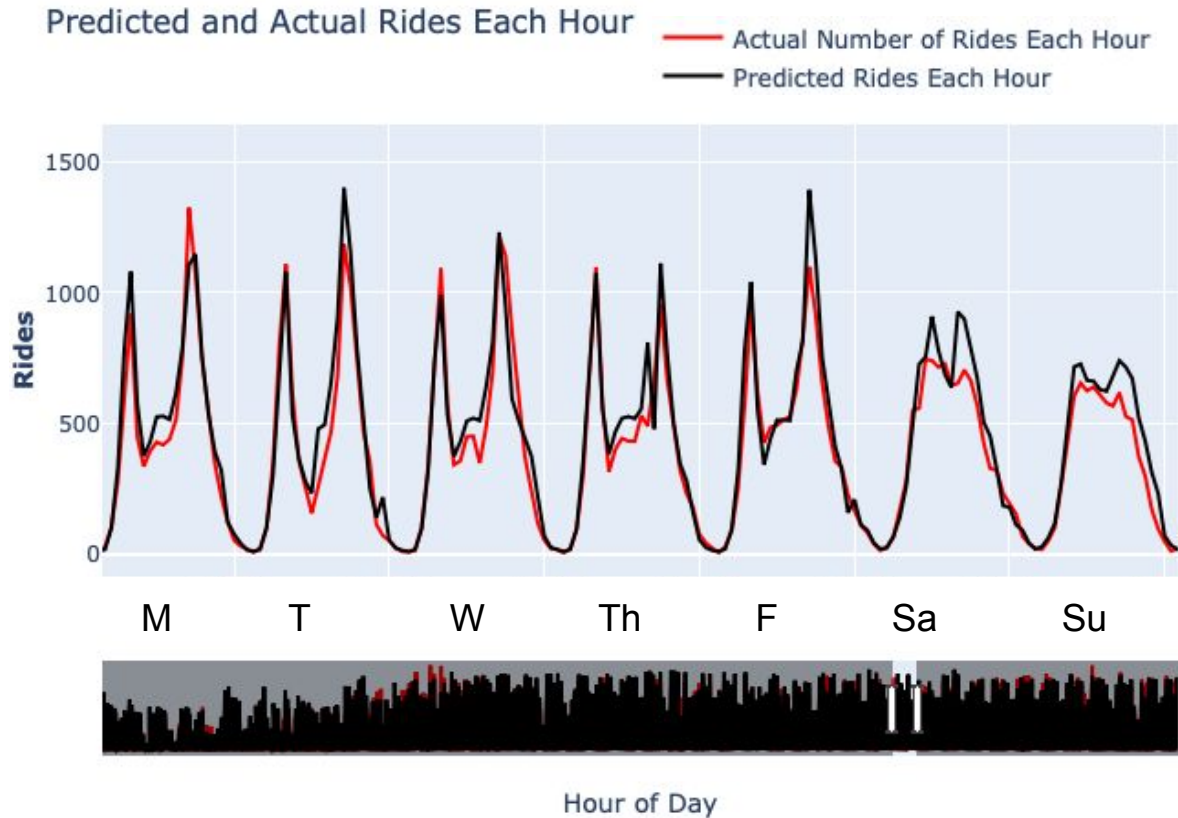
1. Temperature
2. Hour
3. Day of Week
4. Rain
5. Month

## Out

You should expect  
\_\_\_\_\_ CB riders  
during an hour of  
those conditions.

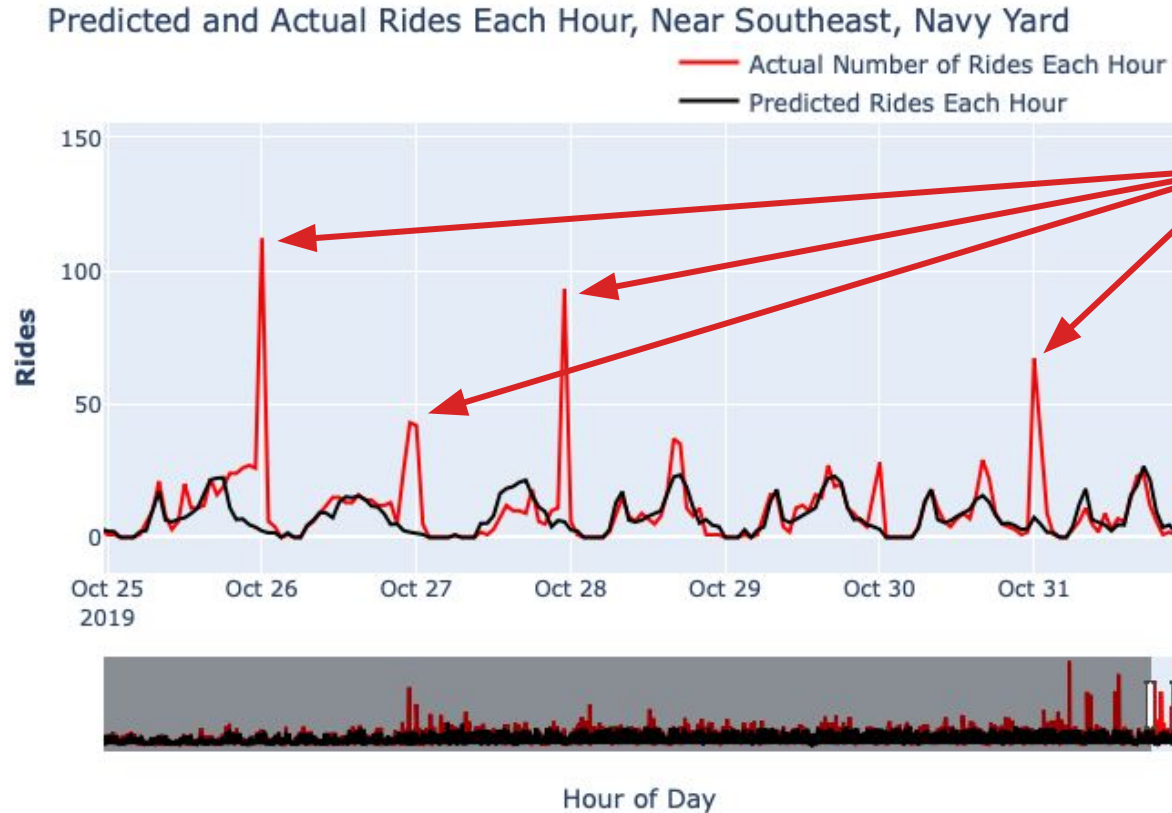


This is how the actual values and the model's predicted values compare in a typical week.





There are some fun anomalies that could be accounted for in a more sophisticated model.

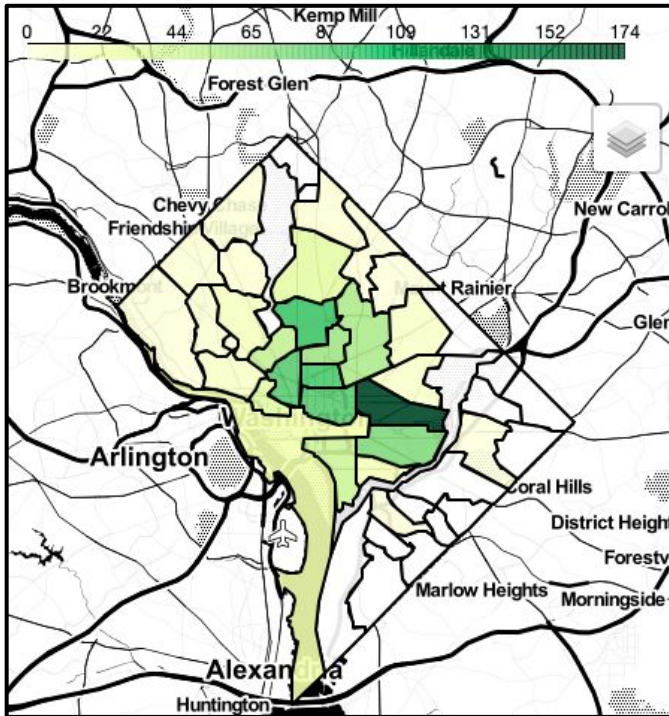




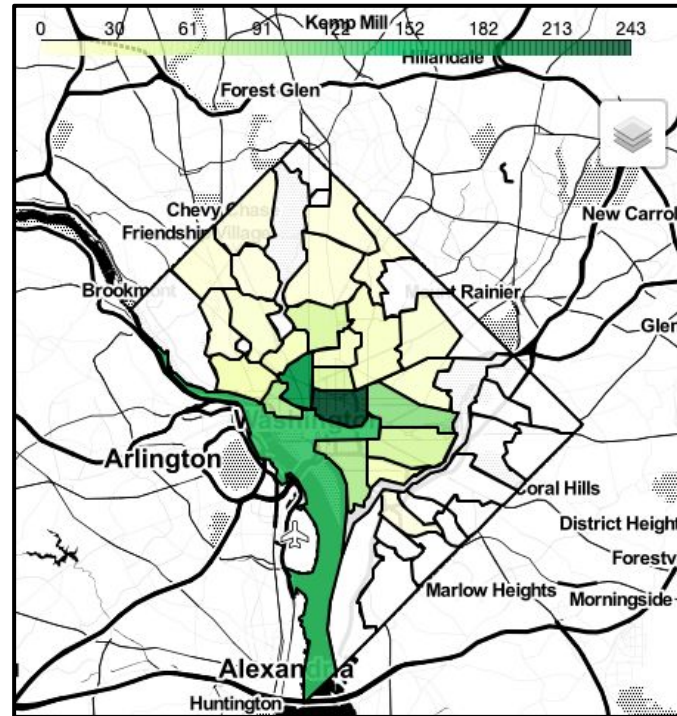
We can map where and when people start and finish rides.

## Where Rides Are Starting

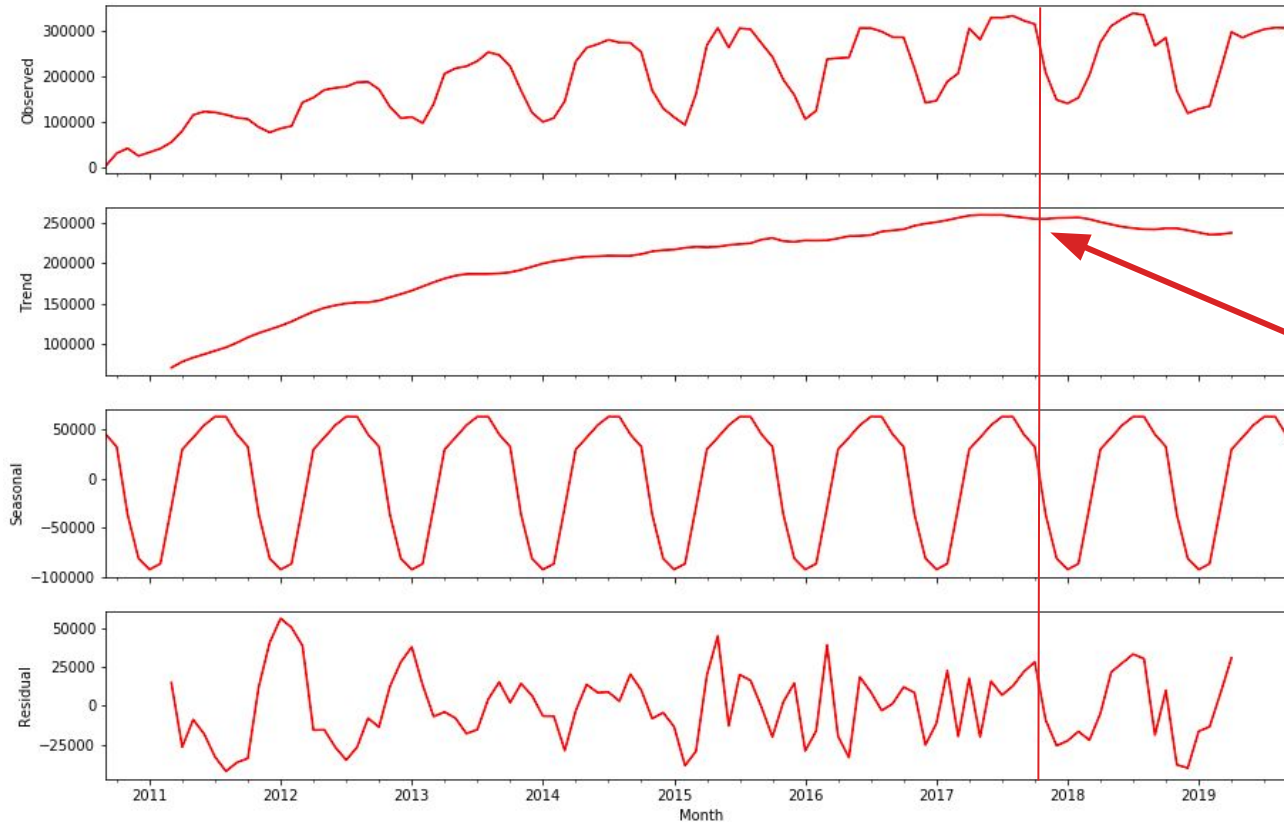
8am  
Hour



5pm  
Hour



There are also questions of long-term trends.  
New travel options impact demand for CB bikes.

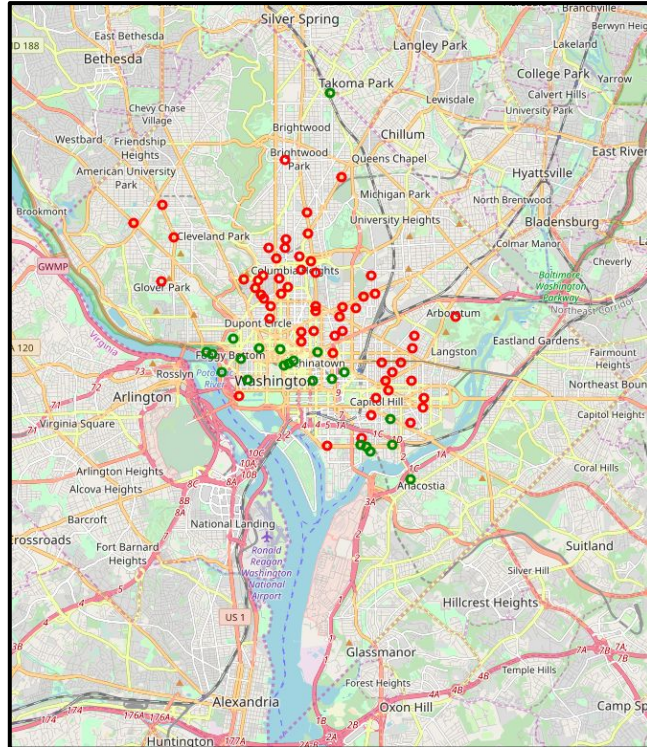


September 2017 -  
DC launches pilot  
program for  
dockless transport

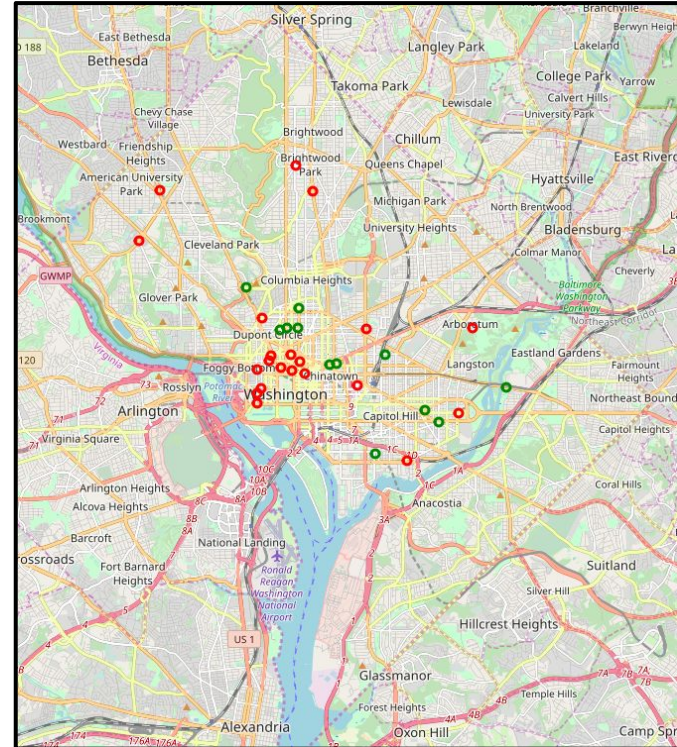


# Capital Bikeshare wants to minimize empty stations and full stations.

9:30am

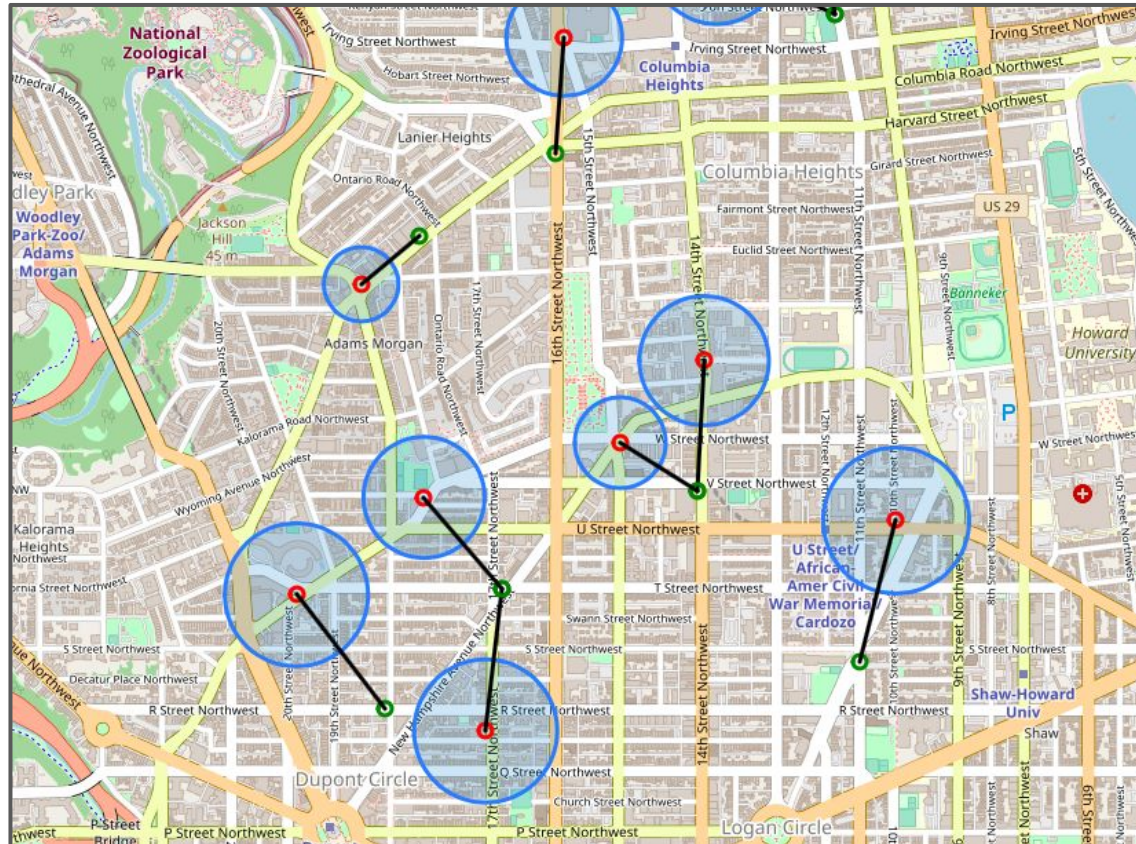


7pm





One possibility: incentivize riders to pick up and dock at underutilized stations



Take a look at my web applications for further exploration!

<https://obscure-garden-58632.herokuapp.com>

# Thanks!

