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| |  | | --- | | **Harvard University** | |  | |  | |  | |
| CS171 Process Book  Weather and the MBTA |
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# Overview and Motivation

## Overview

### Public Transportation and the Data Wave

Over the past several years, there has been a huge increase in the collection and use of data by all sorts of organizations – there has been a huge Data Wave. Public transit agencies have been collecting enormous amounts of data about their riders, but are faced with the challenge similar to so many others riding the Data Wave that transforming this data into useful, actionable information can be hard! Our goal is to develop visualizations that offer insight into how weather affects ridership which can be used by the Massachusetts Bay Transportation Authority (MBTA) to inform staffing.

### Boston Transportation (MBTA)

Using data from the MBTA’s fare collection system, we’ve been developing an initial predictive model of ridership patterns. We found that historical ridership patterns, the number of entries on the same day last week, two weeks prior, etc., to be an adequate baseline predictive model, but that we could improve upon this baseline by including weather information. This past winter, the city of Boston found out just how inconvenienced its commute could be by cataclysmic amounts of snow. Since the city of Boston is not usually hit by such historical amounts of snow, we looked at more typical amounts of snowfall in previous years and their effect on ridership.

## Motivation

Linear predictors like snowfall and temperature are easy enough to interpret in our forecast model, they are represented by a coefficient. However this is not engaging and does not communicate the relationship effectively outside of a narrow technical audience. Our goal with this project is to create an interface that allows for a more intuitive exploration of how varying amounts of snowfall affect ridership for any MBTA rail station.

# Related Work

* **MBTA Viz**
* An interactive exploration of Boston's subway system.
* <http://mbtaviz.github.io>
* **Making Data Matter**
* The role of Information Design and Process in applying automated data to improve transit service.
* <http://dspace.mit.edu/bitstream/handle/1721.1/81640/859408312.pdf?sequence=1>
* **Intermodal Passenger Flows on London’s Public Transport Network**
* Automated inference of full passenger journeys using fare-transaction and vehicle-location data.
* <http://dspace.mit.edu/bitstream/handle/1721.1/78242/830539087.pdf?sequence=1>

# Questions

* Initial Question: How does weather affect ridership on the MBTA?

After Exploratory Data Analysis (section below) the questions changed:

Many weather variables do not have much correlation with ridership but a large amounts of snowfall does. We’ve created dozens of static plots to find that weather patterns other than snow make little difference on ridership, but our visualization will empower the user to explore the different weather conditions and reach the same conclusion (or prove us wrong). Some questions to motivate the user’s exploration of the MBTA system are…

* At what level of snowfall can we start to see differences in number of riders entering the MBTA system?
* How do various amounts of snowfall affect the ridership of a particular station?
* Is the morning rush hour more resilient to snowfall than the evening rush hour?
* Are stations near downtown Boston and in the Boston suburbs similarly affected by snow?
* Holding weather conditions fixed, do we see a different changes in ridership between the weekend and weekdays?
* Does rainfall have any impact on ridership?

# Data and Processing

## Data

MBTA Ridership Data

Through a partnership with the MBTA we have aggregated entries at 15-minute intervals for all MBTA railway stations going back to 2013. This data does not include light rail above ground Green Line trains on the B, C, D, and E branches. The data is relatively clean, containing a timestamp, station ID, number of entries, and number of exits. Our contact at the MBTA was not very strict about the release of the data since it does not contain any personally identifying information so we do not foresee issues with a publicly available project.

NOAA Weather Data

We have pulled Boston’s weather data from the [NOAA API](https://www.ncdc.noaa.gov/cdo-web/webservices/v2) going back to 2013. The readings are from Boston Logan airport so we believe the data to be fairly reliable for the local Boston area.

## Processing

MBTA Ridership Data

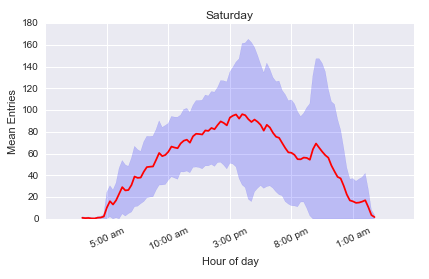
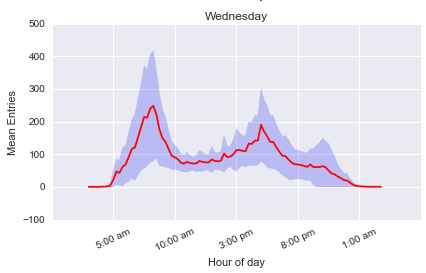
We plan to process this data in an iPython shell using the Pandas module. It provides functions to group our data set by station ID and average the entries over the 15-minute intervals to derive an accurate time series representation of each station. The output will then be made into a json file that also contains information for each station such as the station name, line, geographic coordinates, etc.

NOAA Weather Data

We will format the NOAA data in csv form so it can be loaded into Python via Pandas like above. There are a lot of climate features in the dataset that we will not be using so they can be removed. Using the ‘snowfalli’ feature that gives the amount of snowfall for a particular day in inches, we have added binary variables that correspond to bins of snowfall amount. For example, one variable is ‘2\_to\_4\_snow’ that has a value of 1 if the snowfall for that day is between 2 inches and 4 inches and 0 otherwise. The bins have been hand-picked to have an even distribution of days within each bin since there are fewer days with 12-15 inches of snow than there are days with 2-4 inches of snow. This is expanded on in the exploratory data analysis section. This data can then be merged with the MBTA ridership data using the day portion of the timestamp as the key on which to merge.

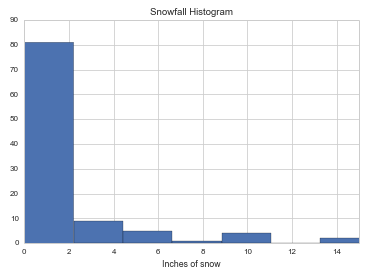
# Exploratory Data Analysis

One of the first things we noticed was that for many stations, simply looking at historical trends gave a pretty accurate estimate of ridership for most weekdays. This made sense since the general population commutes to work and this is a habit with relatively low variation. Most stations had defined peaks in ridership corresponding to morning and afternoon rush hour with a mid-afternoon lull in ridership. Weekends, however, exhibited a completed different trend as ridership peaked around 3 pm with large amounts of variation across the day.

Some initial plots suggested that certain types of weather events, particularly snowfall, result in decreased ridership. Generally, more snow meant less people used the rail system. No matter how much snow fell, however, there were always some patrons that appear to be unperturbed by the snow and continued to use the MBTA. We thought that both patrons and MBTA personnel could benefit from visuals that allow them to explore how resilient particular stations are to snowfall.

Since our data only included two year's worth of ridership, we only had about 100 days where it snowed. About 75% of those days recorded snowfall between trace amounts and two inches with the other 25% ranging from three to twenty inches of snow. We had to make sure our design allowed users to explore various levels of snowfall that were observed in our data. We binned snowfall into amounts that made the most sense in terms of how people perceive amounts of snow and how many days would be included in each bin.



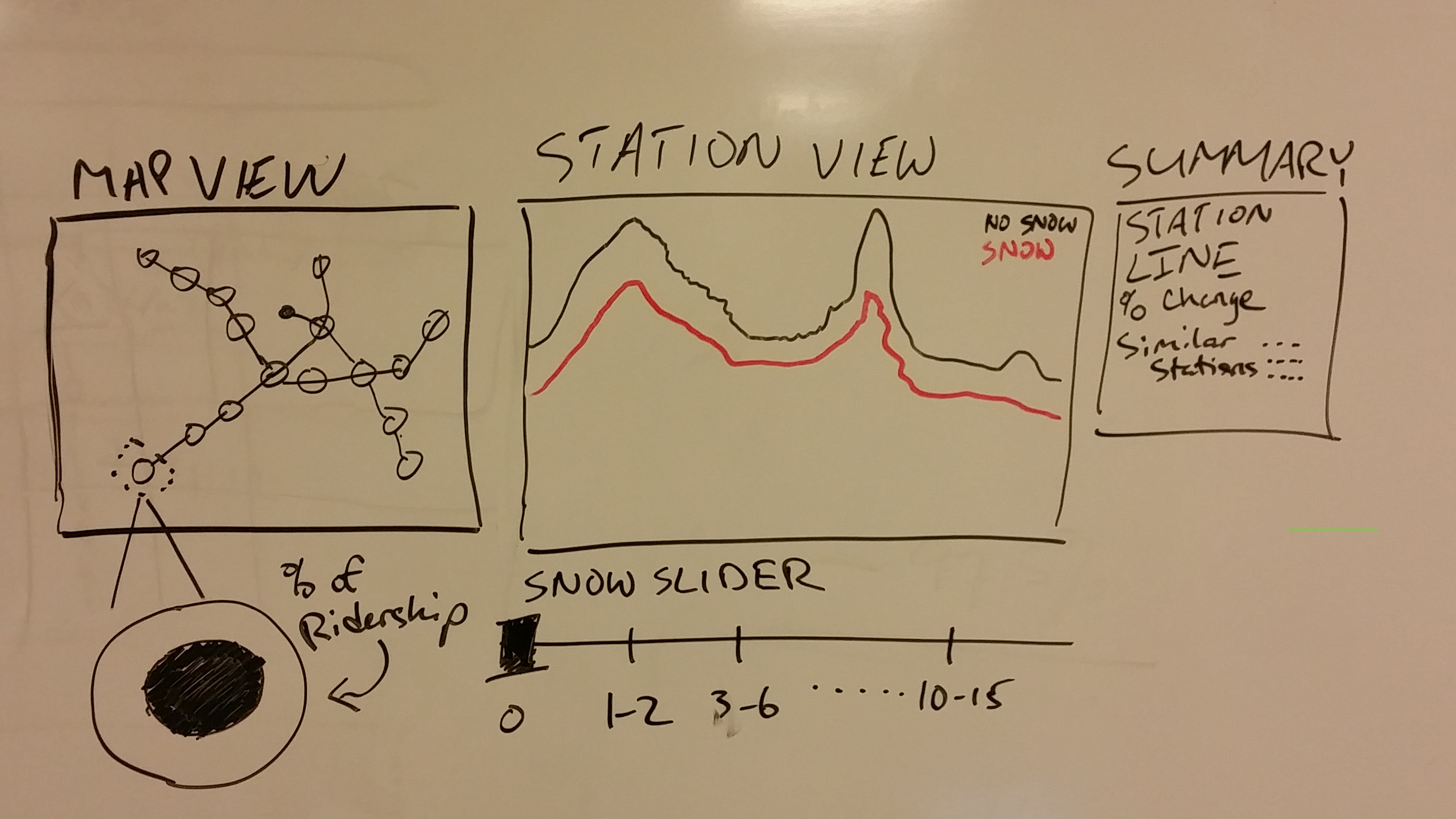
Patterns in other weather-related scenarios such as rainy days or days with variable temperatures weren't so apparent. Nonetheless, we decided to implement our design to allow users to explore various scenarios which they may face when planning their commutes.

# Design Evolution

**Initial Design**

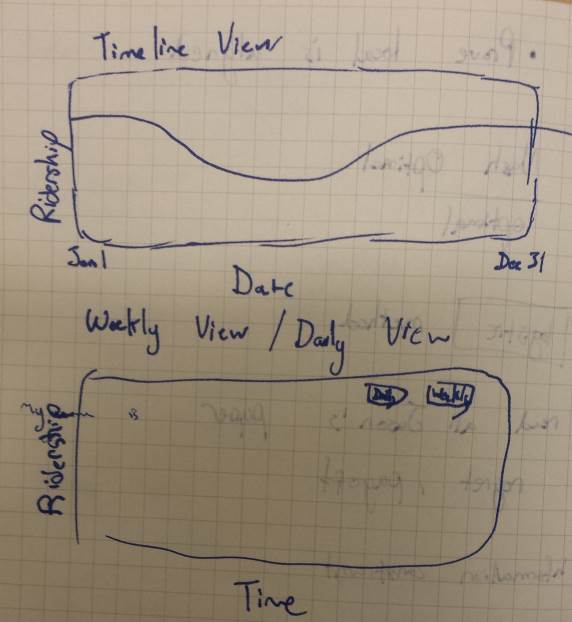
We will implement three views in our design. One view will be primarily dedicated to navigating the data slice. It will allow the user, using a map of the MBTA, to select a station of interest. This view will also enable exploration in the sense that the user will be able to select a particular level of snowfall and this view will display the change in ridership for each station. The high-level view will be one way the user can see which stations across the map are similarly affected. Each icon will be two concentric circles where the ratio of the inner circle area to the outer is proportional to the fraction of ridership for the user-selected snowfall. The second view will display two time series for the selected station; one series is the average entries over the day with no snowfall and the other is the average entries over the day for the user-selected snowfall. Lastly, there will be a summary view which will contain some general information about the station along with the percent change in ridership for the user-selected snowfall and a list of stations with similar changes in ridership.

**Initial Sketch**

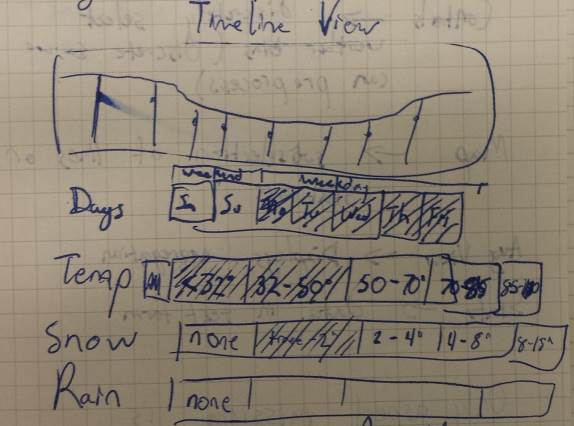


After prototyping our design and examining it critically, including in the peer review design studio, the idea has evolved.

**Add the ability to view the sub selection in time**  
After discussing the design with our TF, Benjy, it became clear that a it would be helpful to have a view showing which days were being used to generate the weather adjusted ridership. In the sketch below this is done in the top graph. We plan to use a timeline that shows the full two years of data and highlight the days that have the selected weather.



**Add day of week controls**We think the user should be able to have more control then in the initial design. We considered allowing the full range of selections (show in the sketch below)



However this prevents effective pre-processing of data and given our audience we should favor clear targeted information over unlimited exploration. Due to the size of the data we want to pre-process the data bins to provide timely interaction. We settled on allowing a weekday / weekend selection and one weather variable at a time

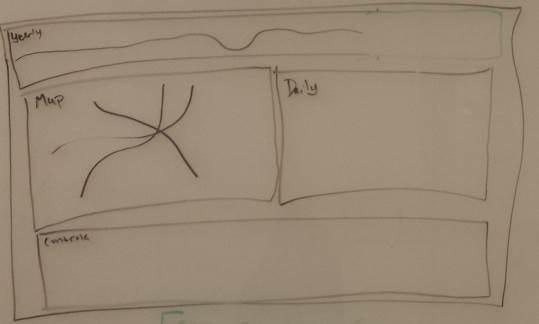
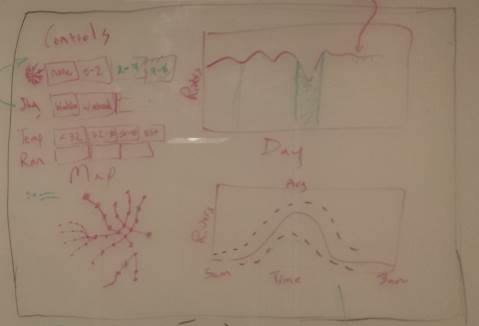
**New Sketch:**

We have removed the text summary panel and now have four divisions. Our layout will fit entirely onto the screen so as to be one cohesive and immersive experience. A timeline panel atop the layout will show the general trend in ridership across all MBTA stations and will be used to highlight the days aggregated in the station ridership panel. For example, if the user has selected the comparison of *0-2 inches of snow*, the timeline panel will render vertical bars at the days in our dataset when there was 0-2 inches of snow. This will help inform our user when our aggregation may contain only a handful of days as in the case of snowfall greater than 15 inches.

We initially intended on displaying the average ridership of each line in the timeline panel, but decided against it for a few reasons. While the questions motivating our implementation focused primarily on weather, we wanted the experience to guide the user from a high-level picture of MBTA ridership down into the idiosyncratic patterns of ridership for days with interesting weather patterns. A single trend showing average ridership across our dataset conveys the importance of seasonality across the whole dataset instead of focusing on the differences between lines. Boston, being a city largely composed of students, sees a large drop in ridership during the summer months before ridership sharply rises in September. Changes like this may not be so obvious on the Blue Line compared to its normal traffic since not many schools like in proximity to it, but this is an important part of the MBTA as a whole.

Below the timeline panel, we chose to divide the screen equally between the MBTA map selection panel and the change in average ridership panel. We debated how to best divide the real estate between the two windows as we wanted to encode information on the map when a click was registered, but the ridership panel was intended to be where the user saw the fruits of exploration so we didn’t want it to be cannibalized by its neighbor. This decision forced us to decide how to intuitively encode information about similarly-affected stations into the map view’s small icons representing each station. Initially, we considered having each icon consist of two concentric circles where the inner circle would fill up the parent circle in an amount proportional to the station’s change in ridership under the user-selected conditions. The icons were just too small given that we only wanted the map to take up half of the screen’s width. We decided a monochrome color scale depicting a station’s change in ridership would be a better way to relate similar stations given the size of the MBTA map panel.

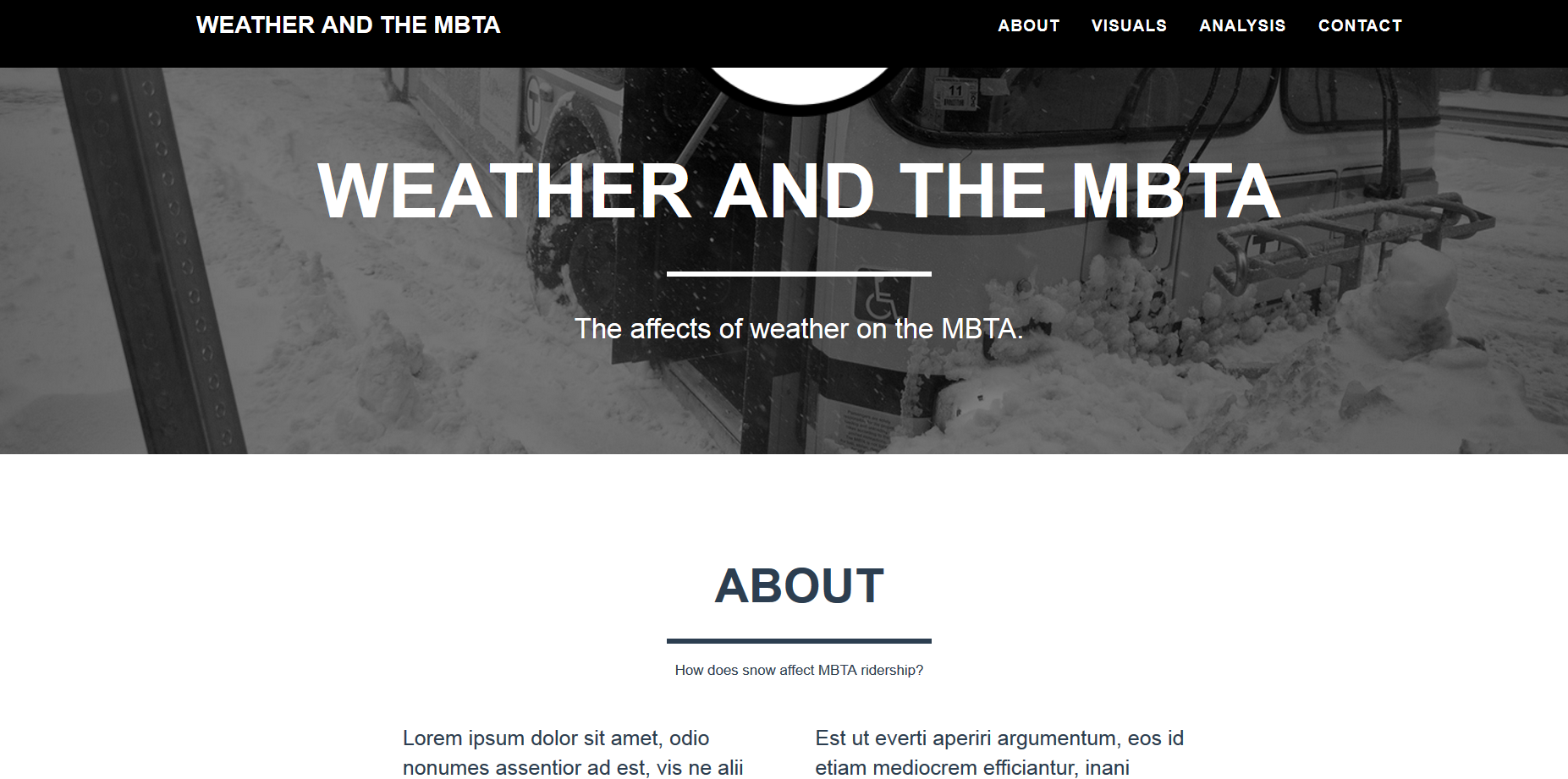
Lastly, we decided to forego the station summary panel from the initial sketches in favor of a user control panel that would hold the buttons for the additional weather conditions we decided to include. Below are two rough sketches of the layout where the left has the controls above the map panel while the other has the controls at the bottom of the screen.



# Implementation

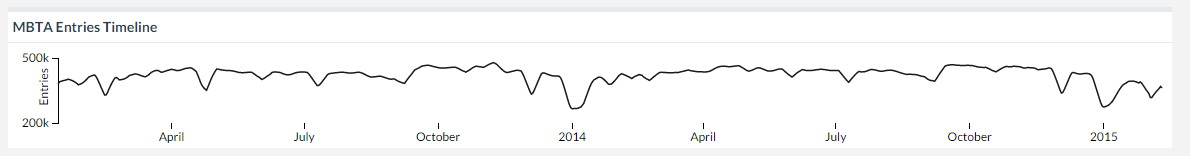
Below we step through the various portions of our implementation and explain the technical choices and evolution of the particular modules.

## Website Mockup **index.html**



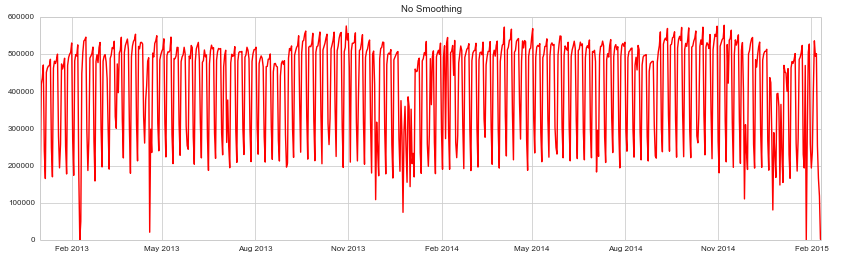
The landing page of our website contains all of our work, allowing the user a free-flowing experience of traversing our site through scrolling, alone. We decided against using separate pages to contain the MBTA timeline panel and the per station ridership plot because we felt that it wouldn’t present motivating interest, ridership on days with poor weather conditions, in the context of normal MBTA travel patterns.

## MBTA Timeline View



In our initial sketches, we didn’t include the MBTA timeline view. After the first meeting with our TF Benjy, he suggested that we implement a way to show the users how many days we used to compute time series for varying amounts of snow. The entire dataset contains about 100 days on which it snowed, but 75% of those days only had 0-4 inches of snow. Upon allowing the user to select between weekdays and weekends, some scenarios like *weekday ridership for trace amounts of snow* had dozens of day’s worth of data while *weekend ridership for 8-15 inches of snow* only included one day’s worth of data. Upon selected a weather condition, the timeline renders vertical bands at the days which were used to construct the ridership plot. Even though we could do little to remedy these gaps in the data, we felt it was important for the user to realize the inherent variability of data presented on only a handful of data points.

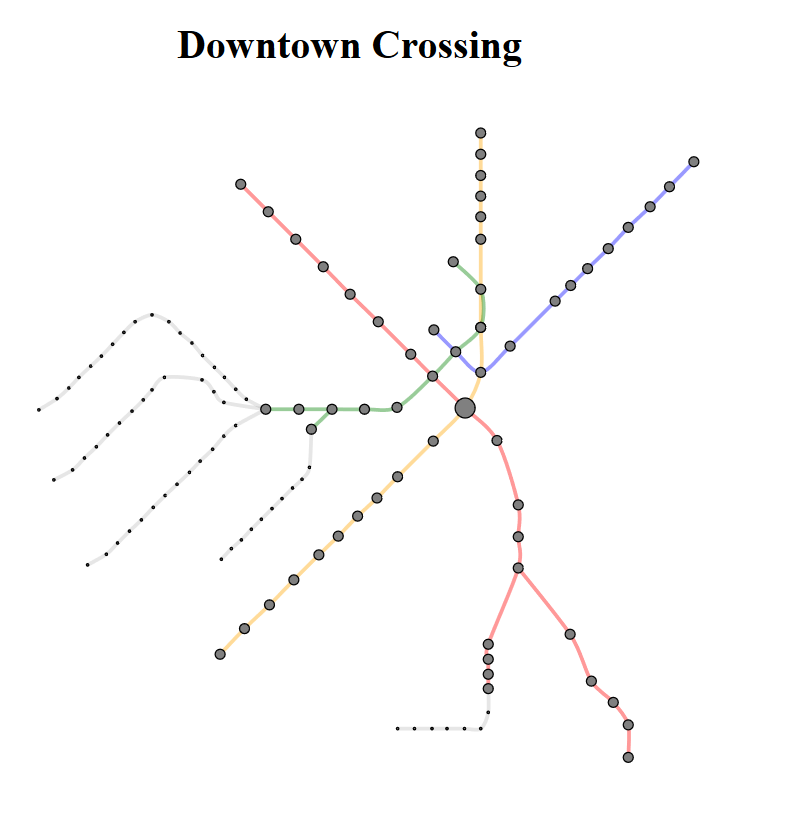
In addition to clarifying our other plots, this timeline is an important storytelling element to our implementation. The line represents the aggregated entries across all MBTA lines after smoothing via moving average. Below is a picture of the timeline without any smoothing. The large difference between weekday and weekend ridership causes the extreme periodicity. We first used a moving average window of 7 days to smooth the series then performed a second iteration of smoothing to achieve a plot to our liking.



The smoothed timeline gently exposes the user to the idea of visualizing the MBTA. Several valleys correspond to familiar time periods when fewer people are using the MBTA; the large dip surrounding the New Year corresponds to the holiday season when many people are on vacation and there is a lull between June and September corresponding to colleges being out of session. While this plot doesn’t directly look at answer any of our questions regarding ridership under weather conditions, its place atop our visualization helps paint a cohesive picture of ridership before looking at the subset of ridership on days with certain weather conditions.

[test\_subwayMap\_simple.html](file:///C:/Users/SonyUser/Documents/GitHub/mbta-weather/test_subwayMap_simple.html)

Initial map with rollover selection, rolled over downtown crossing. After feedback we will add rollover context with click selection



# Evaluation

Section will be updated when we reach this stage of the process