

# Train Accident Findings Report

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## Report of Grand Rapids Train Accident Analysis

### Set Up

Load necessary packages and set max numerical digits displayed in tibble to 8 (necessary for long/lat coordinates):

```
library(tidyverse)
library(sf)
library(osmdata)
library(ggpubr)
library(ggforce)
options(digits = 8)
```

Specify size of all graphs in Knitted Documents:

```
knitr::opts_chunk$set(echo = TRUE, fig.width = 16, fig.height = 8)
```

Upload Grand Rapids Crash Dataset (2008 - 2017):

```
crash_data <- read_csv(here::here("data", "CGR_Crash_Data.csv"))
head(crash_data[1:6])
```

```
## # A tibble: 6 x 6
##       X      Y OBJECTID ROADSOFTID BIKE  CITY
##   <dbl> <dbl>   <dbl>      <dbl> <chr> <chr>
## 1 -85.7  42.9       1    2589528 No    Grand Rapids
## 2 -85.6  42.9       2    2593183 No    Grand Rapids
## 3 -85.7  43.0       3    2582102 No    Grand Rapids
## 4 -85.6  42.9       4    2579820 No    Grand Rapids
## 5 -85.7  43.0       5    2594624 No    Grand Rapids
## 6 -85.7  43.0       6    2599372 No    Grand Rapids
```

Configure features necessary for Grand Rapids map (using openstreetmap api)

```
location_gr <- getbb("Grand Rapids") %>%
  opq()

major_roads_gr <- location_gr %>%
  add_osm_feature(key = "highway", value = c("motorway", "trunk", "primary", "secondary", "tertiary")) %>%
  osmdata_sf()

#minor_roads_gr <- location_gr %>%
#  add_osm_feature(key = "highway", value = c("unclassified", "residential")) %>%
#  osmdata_sf()
```

```

water_gr <- location_gr %>%
  add_osm_feature(key = "waterway", value = c("river")) %>%
  osmdata_sf()

boundary_gr <- location_gr %>%
  add_osm_feature(key = "boundary", value = "administrative") %>%
  add_osm_feature(key = "name", value = "Grand Rapids") %>%
  osmdata_sf()

```

## Initial Exploration of Direct Train Accidents

I am interested in studying the impact gates at a railroad crossing have on the number of accidents associated with trains.

Let's start by visualizing the number of Grand Rapids crashes associated with a train from 2008 - 2017.

```

crash_data_train <- crash_data %>%
  filter(TRAIN == "Yes")

crash_data_train %>%
  select(`Longitude` = X, `Latitude` = Y, CRASHDATE, TRAIN, `Principal Road` = PRNAME, NUMOFINJ)

```

```

## # A tibble: 3 x 6
##   Longitude Latitude CRASHDATE  TRAIN `Principal Road` NUMOFINJ
##   <dbl>    <dbl> <date>    <chr> <chr>          <dbl>
## 1    -85.6     42.9 2008-02-19 Yes   BURTON           1
## 2    -85.7     43.0 2014-05-03 Yes   11TH             1
## 3    -85.7     43.0 2017-12-27 Yes   CENTURY          0

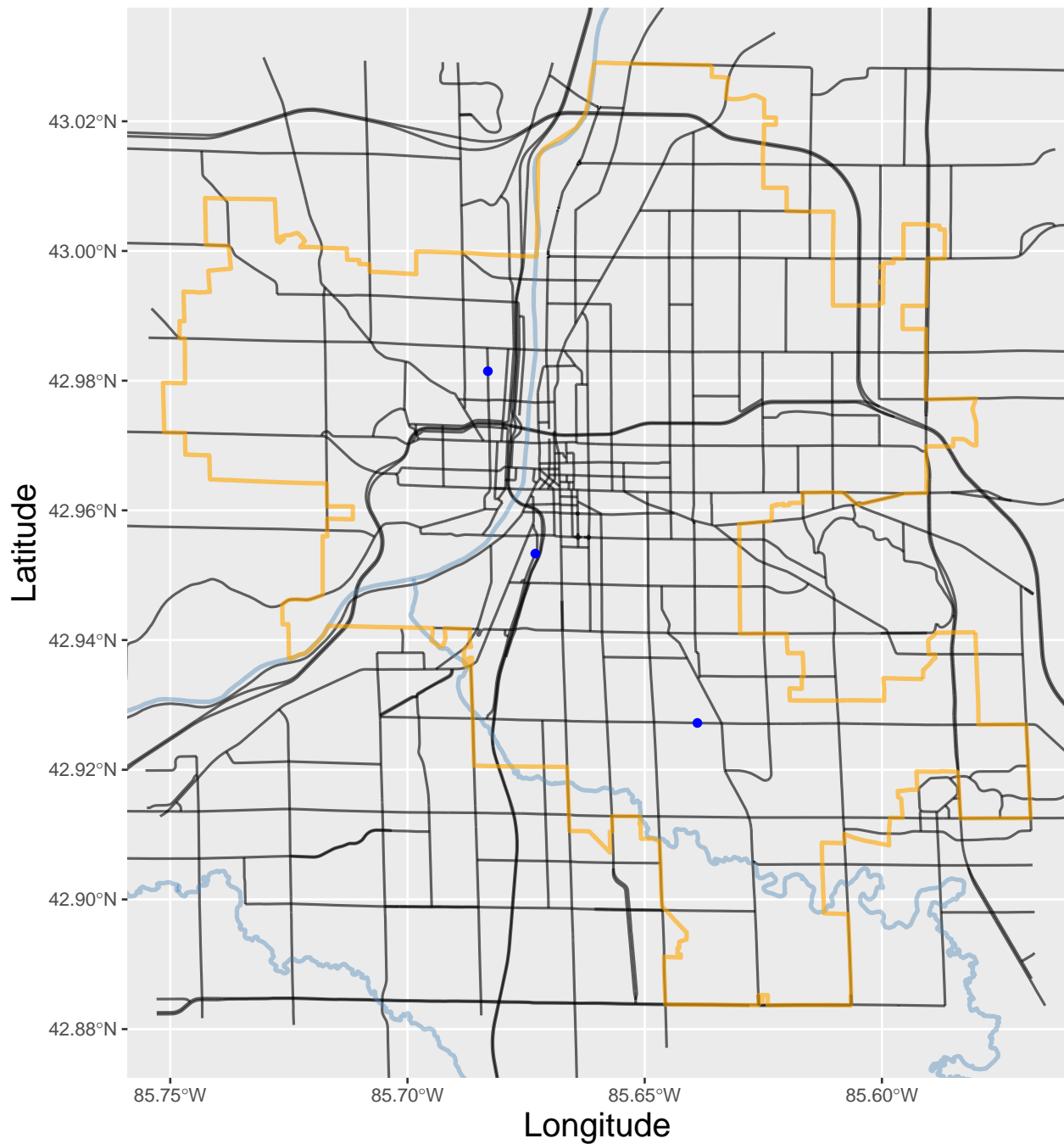
```

```

ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = crash_data_train, mapping = aes(x = X, y = Y), color = "blue") +
  coord_sf(xlim = c(-85.57, -85.75), ylim = c(42.88, 43.03)) +
  labs(title = "Grand Rapids City Limits", x = "Longitude", y = "Latitude") +
  font("title", size = 20, color = "blue", face = "bold") +
  font("x", size = 16) +
  font("y", size = 16)

```

## Grand Rapids City Limits



As you can see, from 2008 to 2017 there were only three accidents that occurred in Grand Rapids involving a train (In 2017, 2014, and 2008).

### Pulling In Grand Rapids Train Crossings:

Although we have a very limited number of accidents directly involving a train, this is not a dead-end for our analysis.

Next let's upload a dataset from Transportation.gov (<https://data.transportation.gov/Railroads/Crossing-Inventory-Data-Current/m2f8-22s6>) that provides information about every railroad crossing in the USA:

```
rr_crossing_data <- read_csv(here::here("data", "Crossing_Inventory_Data_-_Current.csv"))

head(rr_crossing_data[1:6])
```

```
## # A tibble: 6 x 6
##   `Revision Date` `Reporting Agency Type I~ `Reporting Agency Typ~ `Reason Code`
##   <date>                <dbl> <chr>                <dbl>
## 1 1970-01-01                1 Railroad                15
## 2 1970-01-01                1 Railroad                15
## 3 1970-01-01                1 Railroad                15
## 4 1970-01-01                1 Railroad                15
## 5 1970-01-01                1 Railroad                15
## 6 1970-01-01                1 Railroad                15
## # ... with 2 more variables: Reason Description <chr>, Crossing ID <chr>
```

Now we filter this data for only railroad crossings within the Grand Rapids city limits (Note that there are three crossing located in the city center where Latitude = 42.96336 that I removed as it seems like the long/lang for these were placeholders):

```
rr_crossing_data_gr <- rr_crossing_data %>%
  filter(`State Name` == "MICHIGAN", `City Name` == "GRAND RAPIDS", `Intersecting Roadway` == "Yes", La

head(rr_crossing_data_gr[1:6])
```

```
## # A tibble: 6 x 6
##   `Revision Date` `Reporting Agency Type I~ `Reporting Agency Typ~ `Reason Code`
##   <date>                <dbl> <chr>                <dbl>
## 1 1991-04-03                1 Railroad                16
## 2 1997-12-31                1 Railroad                16
## 3 2001-08-31                1 Railroad                16
## 4 2010-01-01                2 State                  16
## 5 2010-01-01                2 State                  16
## 6 2010-09-01                1 Railroad                16
## # ... with 2 more variables: Reason Description <chr>, Crossing ID <chr>
```

Let's Make Sure that there are no duplicates in the data:

```
nrow(distinct(rr_crossing_data_gr, Latitude))
```

```
## [1] 74
```

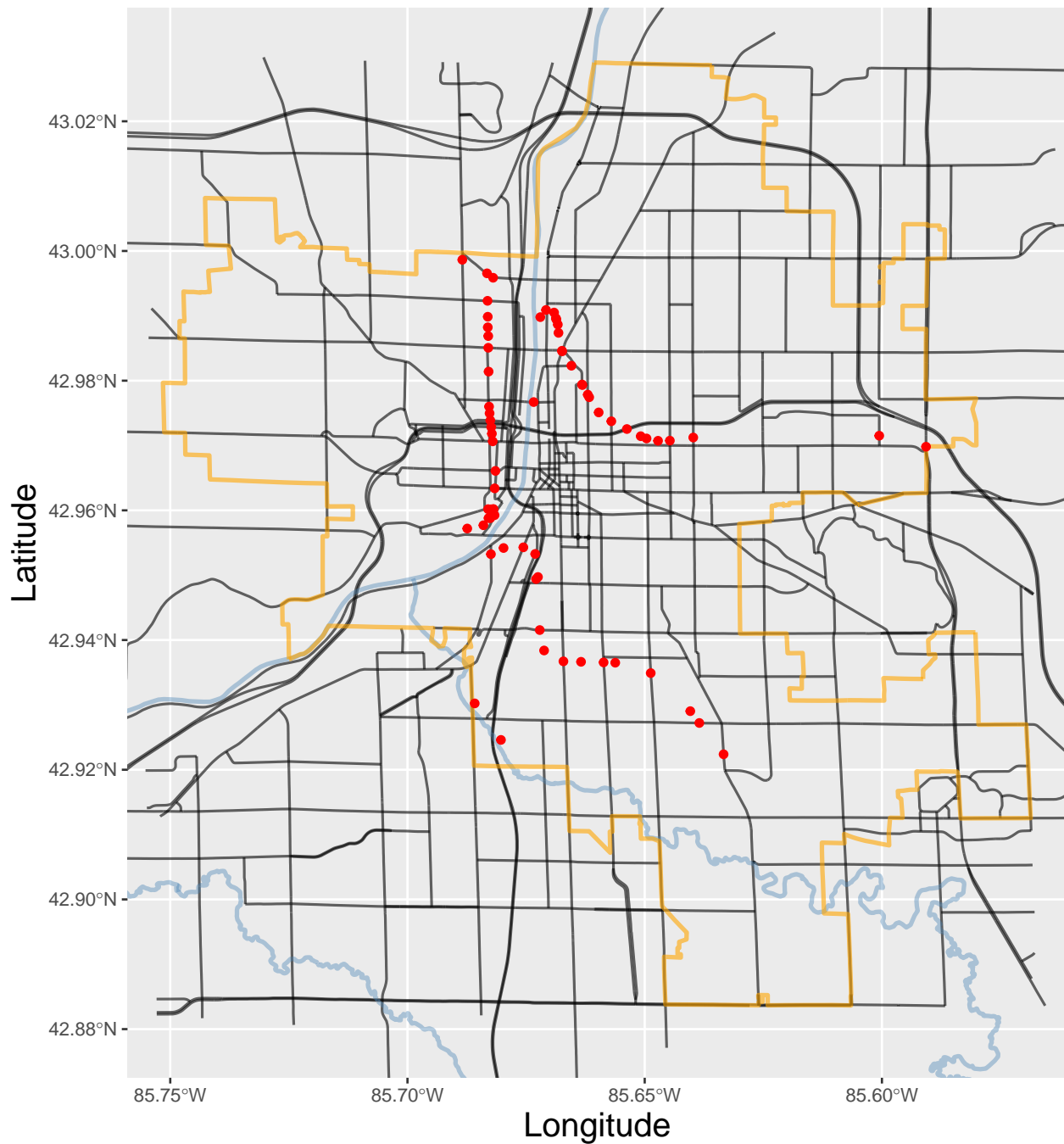
```
nrow(rr_crossing_data_gr)
```

```
## [1] 74
```

We can visualize this data in red on our Grand Rapid's Map:

```
ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = rr_crossing_data_gr, mapping = aes(x = Longitude, y = Latitude), color = "red", size = 100) +
  coord_sf(xlim = c(-85.57, -85.75), ylim = c(42.88, 43.03)) +
  labs(title = "Grand Rapids City Limits", x = "Longitude", y = "Latitude") +
  font("title", size = 20, color = "blue", face = "bold") +
  font("x", size = 16) +
  font("y", size = 16)
```

## Grand Rapids City Limits



### Exploring Accidents in the Vicinity of Railroad Crossings:

We now would like to know how many crashes fall within a .0005 (longitudinal units) radius of each railroad crossing.

To do so, we first write a function called `in_radius()` that detects if a longitude/latitude coordinate is located within a .0005 radius of a given anchor coordinate:

```
in_radius <- function(x1, y1, x2, y2) {  
  if_else(((x1 - x2) ^ 2 + (y1 - y2) ^ 2) <= .0005 ^ 2, 1, 0)
```

```
}
```

Next we create a new accident subset named `near_rr_crash` that only contains crashes that occurred within a .0005 radius of a railroad crossing:

```
near_rr_crash <- crash_data %>%
  rowwise() %>%
  filter(1 %in% in_radius(X, Y, rr_crossing_data_gr$Longitude, rr_crossing_data_gr$Latitude))

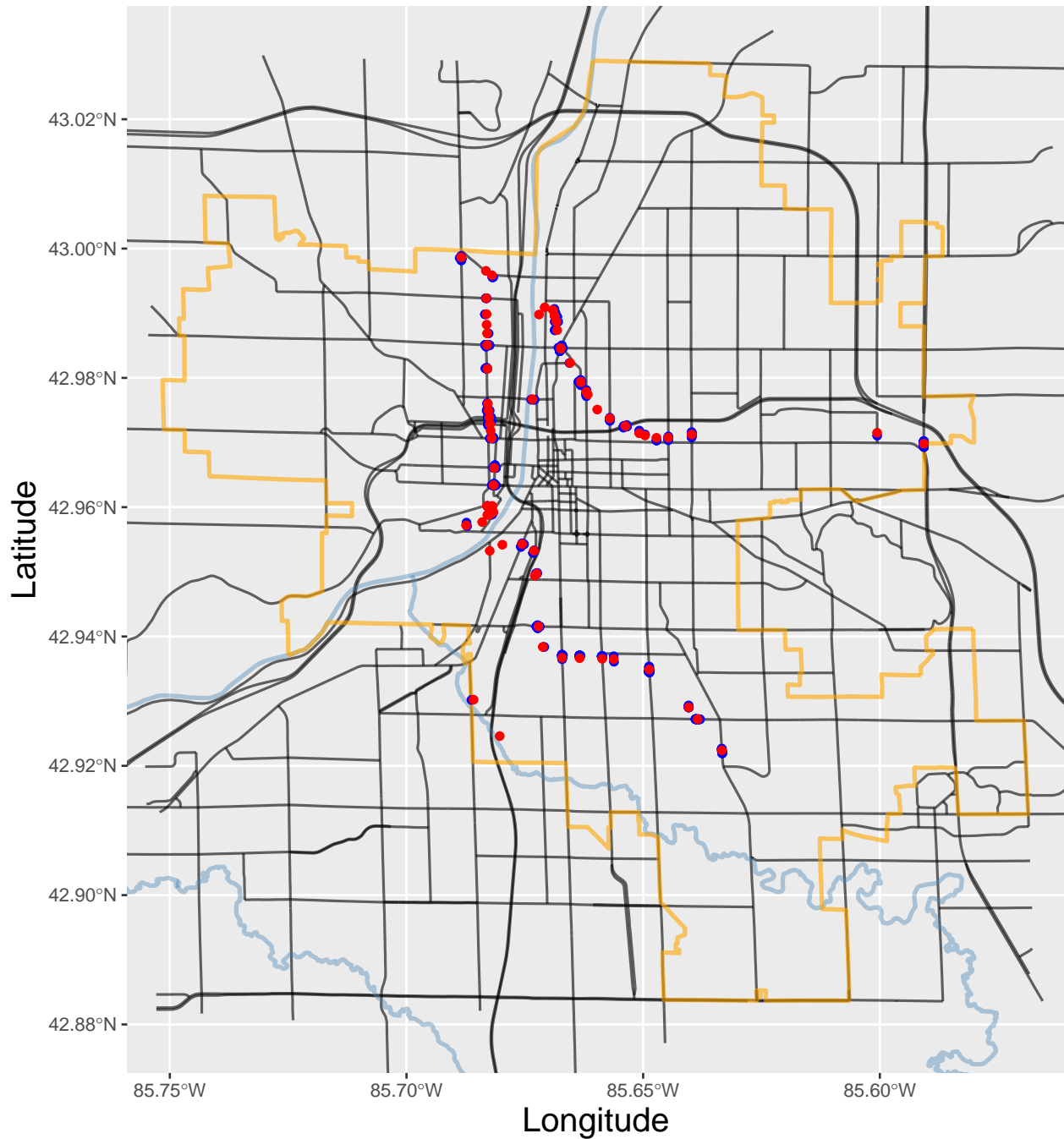
head(near_rr_crash[1:6])
```

```
## # A tibble: 6 x 6
## # Rowwise:
##       X      Y OBJECTID ROADSOFTID BIKE  CITY
##   <dbl> <dbl>   <dbl>    <dbl> <chr> <chr>
## 1 -85.6  42.9     120    2587376 No    Grand Rapids
## 2 -85.7  43.0     343    2579633 No    Grand Rapids
## 3 -85.6  42.9     663    2591863 No    Grand Rapids
## 4 -85.6  42.9     664    2597546 No    Grand Rapids
## 5 -85.6  42.9     665    2583626 No    Grand Rapids
## 6 -85.6  42.9     666    2579464 No    Grand Rapids
```

We plot this data on our grand rapids map to visualize the result:

```
ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = near_rr_crash, mapping = aes(x = X, y = Y), color = "blue") +
  geom_point(data = rr_crossing_data_gr, mapping = aes(x = Longitude, y = Latitude), color = "red") +
  geom_circle(data = rr_crossing_data_gr, mapping = aes(x0 = Longitude, y0 = Latitude, r = .0005)) +
  coord_sf(xlim = c(-85.57, -85.75), ylim = c(42.88, 43.03)) +
  labs(title = "Grand Rapids City Limits", x = "Longitude", y = "Latitude") +
  font("title", size = 20, color = "blue", face = "bold") +
  font("x", size = 16) +
  font("y", size = 16)
```

## Grand Rapids City Limits

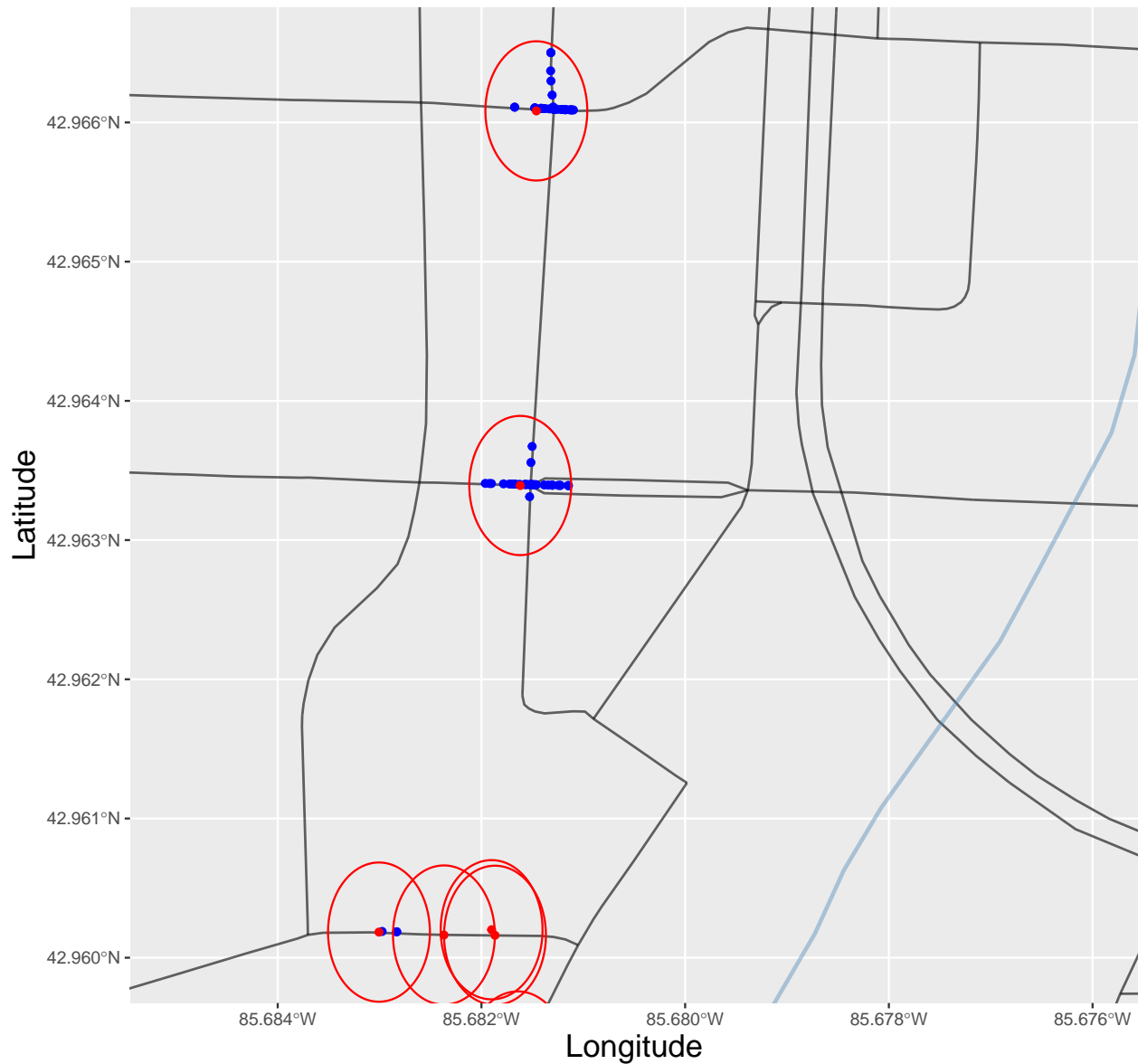


Below is a zoomed-in portion of the map for better detail (located around the GVSU Pew Campus):

```
ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = near_rr_crash, mapping = aes(x = X, y = Y), color = "blue") +
  geom_point(data = rr_crossing_data_gr, mapping = aes(x = Longitude, y = Latitude), color = 'red') +
  geom_circle(data = rr_crossing_data_gr, mapping = aes(x0 = Longitude, y0 = Latitude, r = .001), color = 'red', size = 1000)
```

```
coord_sf(xlim = c(-85.676, -85.685), ylim = c(42.96, 42.9665)) +
labs(title = "Grand Valley Pew Campus", x = "Longitude", y = "Latitude") +
font("title", size = 20, color = "blue", face = "bold") +
font("x", size = 16) +
font("y", size = 16)
```

## Grand Valley Pew Campus



## Summarizing the Results

Now we create a tibble that sums the number of crashes that fall within a .0005 radius of each crossing along with the number of Gate arms at that crossing and the total number of trains that pass through during the day and night:

```
exposure <- rr_crossing_data_gr %>%
  rowwise() %>%
  mutate(Radius_Count = sum(in_radius(Longitude, Latitude, near_rr_crash$X, near_rr_crash$Y))) %>%
```



```

select(Longitude, Latitude, Street, "Gate_Arm_N" = `Count Roadway Gate Arms`, Radius_Count, "Day_Trains_N", "Night_Trains_N")
arrange(desc(Radius_Count)) %>%
ungroup()

head(exposure,10)

```

```

## # A tibble: 10 x 7
##   Longitude Latitude Street Gate_Arm_N Radius_Count Day_Trains_N Night_Trains_N
##   <dbl>    <dbl> <chr>      <dbl>      <dbl>      <dbl>      <dbl>
## 1    -85.7     43.0 PLAIN~         0        111         0         0
## 2    -85.7     43.0 LEONA~         0        108         0         0
## 3    -85.7     43.0 W LEO~         0         84         0         2
## 4    -85.7     42.9 HALL ~         2         79         2         2
## 5    -85.6     43.0 E BEL~         0         74         1         0
## 6    -85.7     43.0 FULTO~         0         45         1         2
## 7    -85.7     43.0 ALPIN~         0         41         1         0
## 8    -85.7     43.0 LAKE ~         2         40         1         2
## 9    -85.7     43.0 11TH ~         0         38         0         2
## 10   -85.7     42.9 JEFFE~         2         32         2         2

```

We can use the number of accidents that occur within .0005 longitudinal units from a crossing (Radius\_Count) to represent traffic exposure. We are deriving latent information from this variable to determine if additional safety protocols should be put in place at any particular train crossing.

The tibble is sorted from largest Radius\_Counts to smallest.

Note that among all 74 railroad crossings in Grand Rapids, 26 have at least one gate arm.

```

exposure %>%
  filter(Gate_Arm_N > 0) %>%
  nrow()

```

```
## [1] 26
```

Of these 26 crossings, 22 have at least one Daylight Train OR one Nighttime train pass through. This is about 85% (22/26):

```

exposure %>%
  filter(Gate_Arm_N > 0, (Day_Trains_N > 0 | Night_Trains_N > 0)) %>%
  nrow()

```

```
## [1] 22
```

Thus, it would make sense that the criteria for adding a gate arm to a crossing or adding additional safety protocols would include a requirement for at least one Daylight Train or Nighttime train at that crossing.

## Analyzing What Crossings Should Be Scrutinized:

Let's take a look at the top 10 traffic exposure crossings on the Grand Rapids Map:

```

exposure_top_10 <- exposure %>%
  slice_head(n = 10)

```

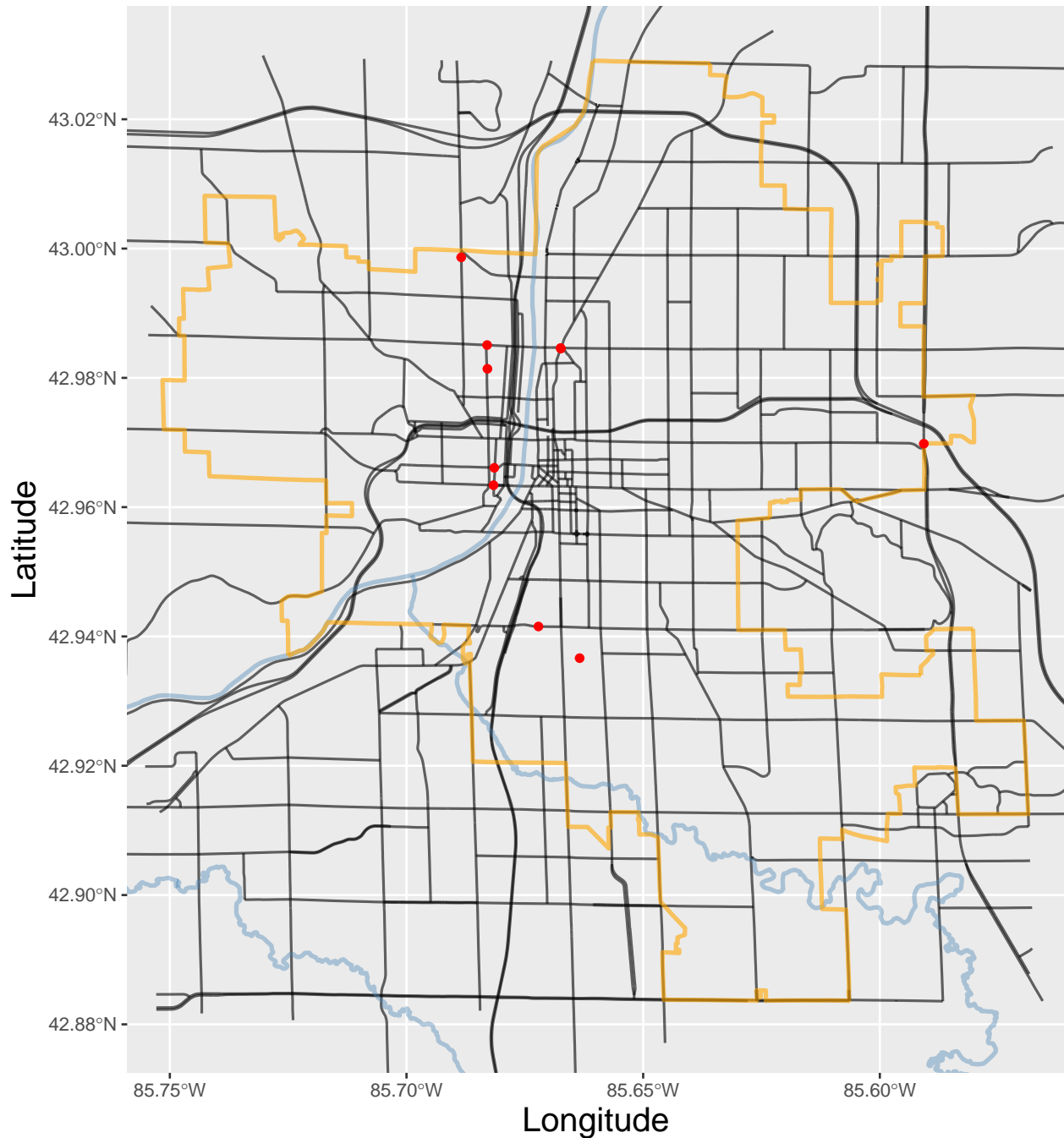
```

ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = exposure_top_10, mapping = aes(x = Longitude, y = Latitude), color = "red")

```

```
coord_sf(xlim = c(-85.57, -85.75), ylim = c(42.88, 43.03)) +
labs(title = "Grand Rapids City Limits", x = "Longitude", y = "Latitude") +
font("title", size = 20, color = "blue", face = "bold") +
font("x", size = 16) +
font("y", size = 16)
```

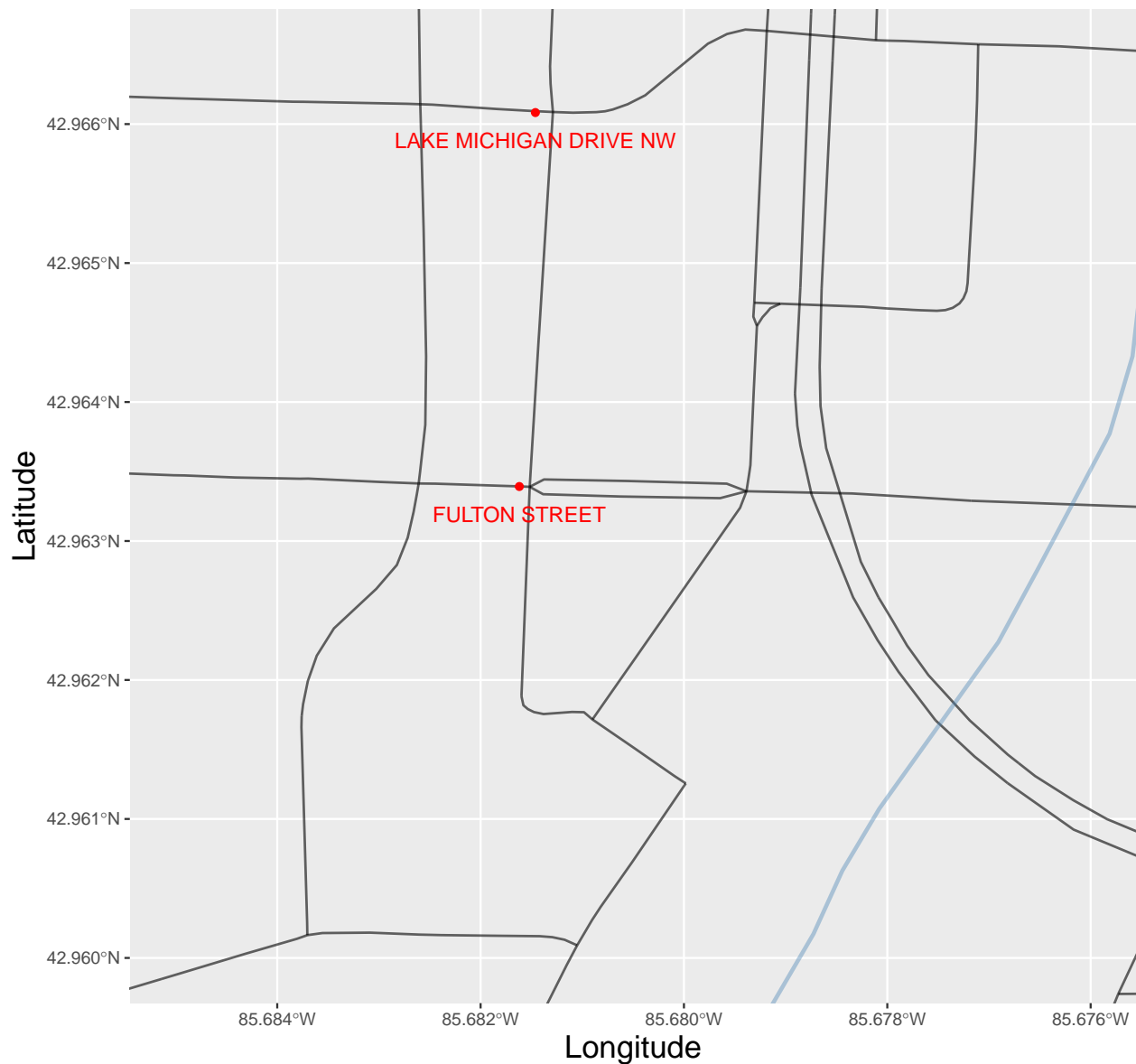
## Grand Rapids City Limits



We can see these crossings are scattered around the city. Notice that two of these crossings are located on the GVSU Pew Campus:

```
ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = exposure_top_10, mapping = aes(x = Longitude, y = Latitude), color = "red", size = 100) +
  #geom_circle(data = exposure_top_10, mapping = aes(x0 = Longitude, y0 = Latitude, r = .0005)) +
  coord_sf(xlim = c(-85.676, -85.685), ylim = c(42.96, 42.9665)) +
  labs(title = "Grand Valley Pew Campus", x = "Longitude", y = "Latitude") +
  geom_text(data = exposure_top_10, mapping = aes(Longitude, Latitude, label = Street), nudge_x = 10, nudge_y = 10,
    font("title", size = 20, color = "blue", face = "bold") +
    font("x", size = 16) +
    font("y", size = 16))
```

## Grand Valley Pew Campus



We see that the crossing at “LAKE MICHIGAN DRIVE NW” (-85.681461,42966083) does have gate arms

while the crossing at “FULTON STREET” (-85.681619,42.963392 ) does not have gate arms. Also note that the total Daylight (1) and Nighttime (2) trains for these crossings are the same while the Radius\_Count for the “FULTON STREET” crossing (45) is greater than the “LAKE MICHIGAN DRIVE NW” crossing (40):

```
exposure_top_10 %>%
  filter(Street %in% c("LAKE MICHIGAN DRIVE NW", "FULTON STREET"))
```

```
## # A tibble: 2 x 7
```

	Longitude	Latitude	Street	Gate_Arm_N	Radius_Count	Day_Trains_N	Night_Trains_N
##	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	-85.7	43.0	FULTON~	0	45	1	2
## 2	-85.7	43.0	LAKE M~	2	40	1	2

The placement of crossing gate arms in Grand Rapids seems to be a bit inconsistent and worthy of a second look.

Note that out of the top 10 crossings with the highest traffic exposure, 7 do not have Gate Arms:

```
exposure_top_10 %>%
  filter(Gate_Arm_N == 0)
```

```
## # A tibble: 7 x 7
```

	Longitude	Latitude	Street	Gate_Arm_N	Radius_Count	Day_Trains_N	Night_Trains_N
##	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	-85.7	43.0	PLAINF~	0	111	0	0
## 2	-85.7	43.0	LEONAR~	0	108	0	0
## 3	-85.7	43.0	W LEON~	0	84	0	2
## 4	-85.6	43.0	E BELT~	0	74	1	0
## 5	-85.7	43.0	FULTON~	0	45	1	2
## 6	-85.7	43.0	ALPINE~	0	41	1	0
## 7	-85.7	43.0	11TH S~	0	38	0	2

Of these 7, 5 have at least one daylight OR nighttime train go through each day (as noted earlier, this seems to be important criteria for adding a gate arm):

```
crossings_of_interest <- exposure_top_10 %>%
  filter(Gate_Arm_N == 0, (Day_Trains_N > 0 | Night_Trains_N > 0))
```

```
crossings_of_interest
```

```
## # A tibble: 5 x 7
```

	Longitude	Latitude	Street	Gate_Arm_N	Radius_Count	Day_Trains_N	Night_Trains_N
##	<dbl>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	-85.7	43.0	W LEON~	0	84	0	2
## 2	-85.6	43.0	E BELT~	0	74	1	0
## 3	-85.7	43.0	FULTON~	0	45	1	2
## 4	-85.7	43.0	ALPINE~	0	41	1	0
## 5	-85.7	43.0	11TH S~	0	38	0	2

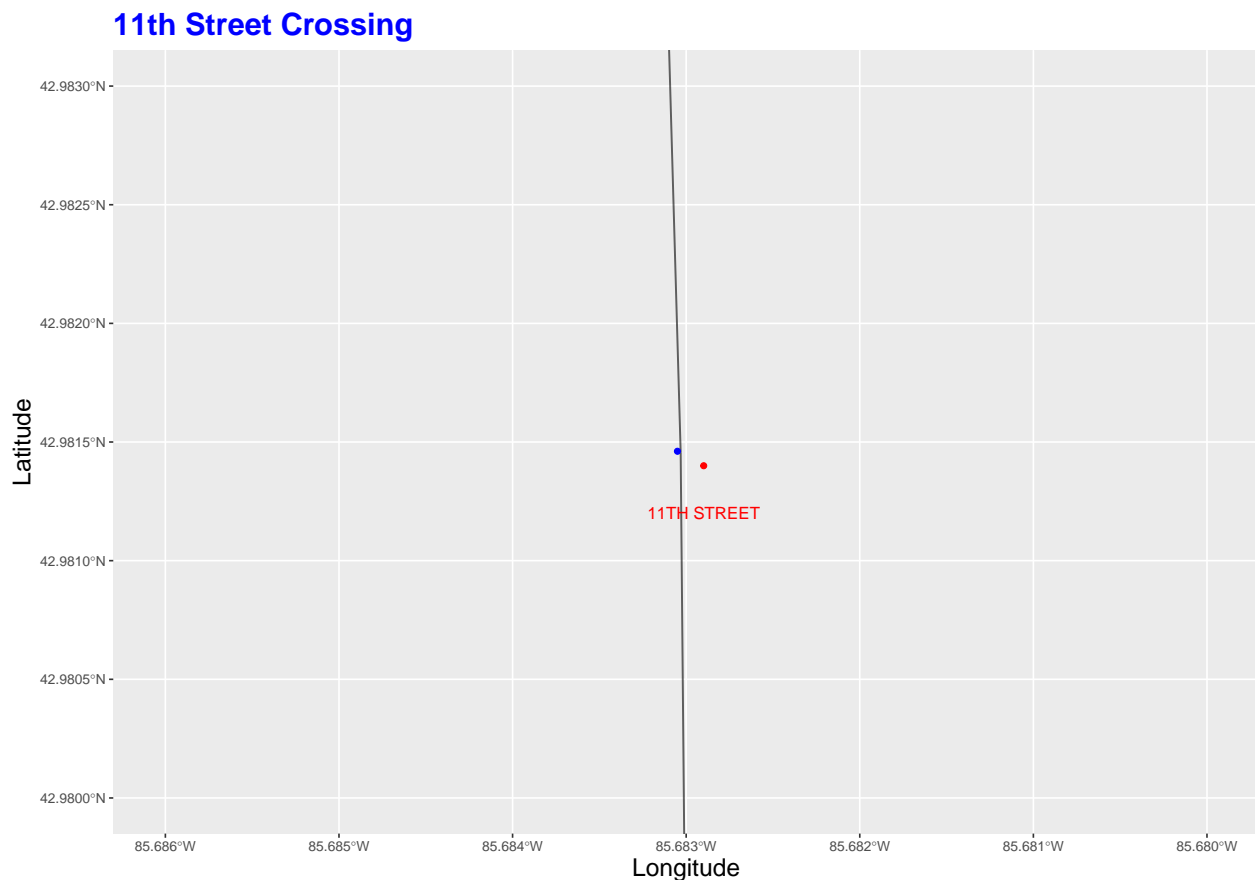
I would recommend that the city of Grand Rapids takes a second look at these crossings to determine if a gate arm or additional traffic precautions are necessary.

It is also worth noting that one of the three accidents we found that directly involved a train was located at the 11TH STREET crossing included in the crossings of interest above:

```
crash_data_train %>%
  rowwise() %>%
  filter(1 %in% in_radius(X, Y, crossings_of_interest$Longitude, crossings_of_interest$Latitude)) %>%
  select(`Longitude` = X, `Latitude` = Y, CRASHDATE, TRAIN, `Principal Road` = PRNAME, NUMOFINJ)
```

```
## # A tibble: 1 x 6
## # Rowwise:
##   Longitude Latitude CRASHDATE  TRAIN `Principal Road` NUMOFINJ
##   <dbl>    <dbl> <date>    <chr> <chr>          <dbl>
## 1    -85.7      43.0 2014-05-03 Yes   11TH              1

ggplot()+
  geom_sf(data = major_roads_gr$osm_lines, size = .6, alpha = .6, color = 'black') +
  #geom_sf(data = minor_roads_gr$osm_lines, size = .3, alpha = .3, color = 'black') +
  geom_sf(data = water_gr$osm_lines, size = 1, alpha = .4, color = 'steelblue') +
  geom_sf(data = boundary_gr$osm_lines, size = 1, alpha = .6, color = "orange") +
  geom_point(data = crash_data_train, mapping = aes(x = X, y = Y), color = "blue") +
  geom_point(data = crossings_of_interest, mapping = aes(x = Longitude, y = Latitude), color = "red") +
  geom_text(data = exposure_top_10, mapping = aes(Longitude, Latitude, label = Street), nudge_x = 10, color = "red", size = 12) +
  coord_sf(xlim = c(-85.68, -85.686), ylim = c(42.98, 42.983)) +
  labs(title = "11th Street Crossing", x = "Longitude", y = "Latitude") +
  font("title", size = 20, color = "blue", face = "bold") +
  font("x", size = 16) +
  font("y", size = 16)
```



## Exploratory Application

To supplement this analysis, I have put together a Shiny Application for the city of Grand Rapids to explore the crash dataset on their own. This application can be accessed using the address below:

[https://lcadagin.shinyapps.io/interactive\\_crash\\_plotting/](https://lcadagin.shinyapps.io/interactive_crash_plotting/)