

# Exploring Refrigerated Truck Dataset

## Import Data:

```
library(ggplot2)
library(tidyverse)

## Attaching packages                                tidyverse 1.3.0
## tibble 2.1.3      dplyr 0.8.3
## tidyr 1.0.0       stringr 1.4.0
## readr 1.3.1       forcats 0.4.0
## purrr 0.3.3

## Conflicts                                tidyverse_conflicts()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

library(rgdal)

## Loading required package: sp
## rgdal: version: 1.4-8, (SVN revision 845)
## Geospatial Data Abstraction Library extensions to R successfully loaded
## Loaded GDAL runtime: GDAL 2.4.2, released 2019/06/28
## Path to GDAL shared files: /Library/Frameworks/R.framework/Versions/3.6/Resources/library/rgdal/gdal
## GDAL binary built with GEOS: FALSE
## Loaded PROJ.4 runtime: Rel. 5.2.0, September 15th, 2018, [PJ_VERSION: 520]
## Path to PROJ.4 shared files: /Library/Frameworks/R.framework/Versions/3.6/Resources/library/rgdal/proj
## Linking to sp version: 1.3-2

setwd("/Users/JacquelineLincroft/Documents/INSH2102 Bostonography/Assignments/FinalProject/")
truckdf <- read.table(file="Ref_Trucks.csv", header=TRUE, sep=",")

# see which destination cities we could choose from
unique(truckdf$Destination)

## [1] ATLANTA      CHICAGO      DALLAS      DENVER      NEW YORK
## [6] LOS ANGELES  MONTREAL, QUE BOSTON      PHILADELPHIA  MIAMI
## [11] SEATTLE      BALTIMORE
## 12 Levels: ATLANTA BALTIMORE BOSTON CHICAGO DALLAS DENVER LOS ANGELES ... SEATTLE

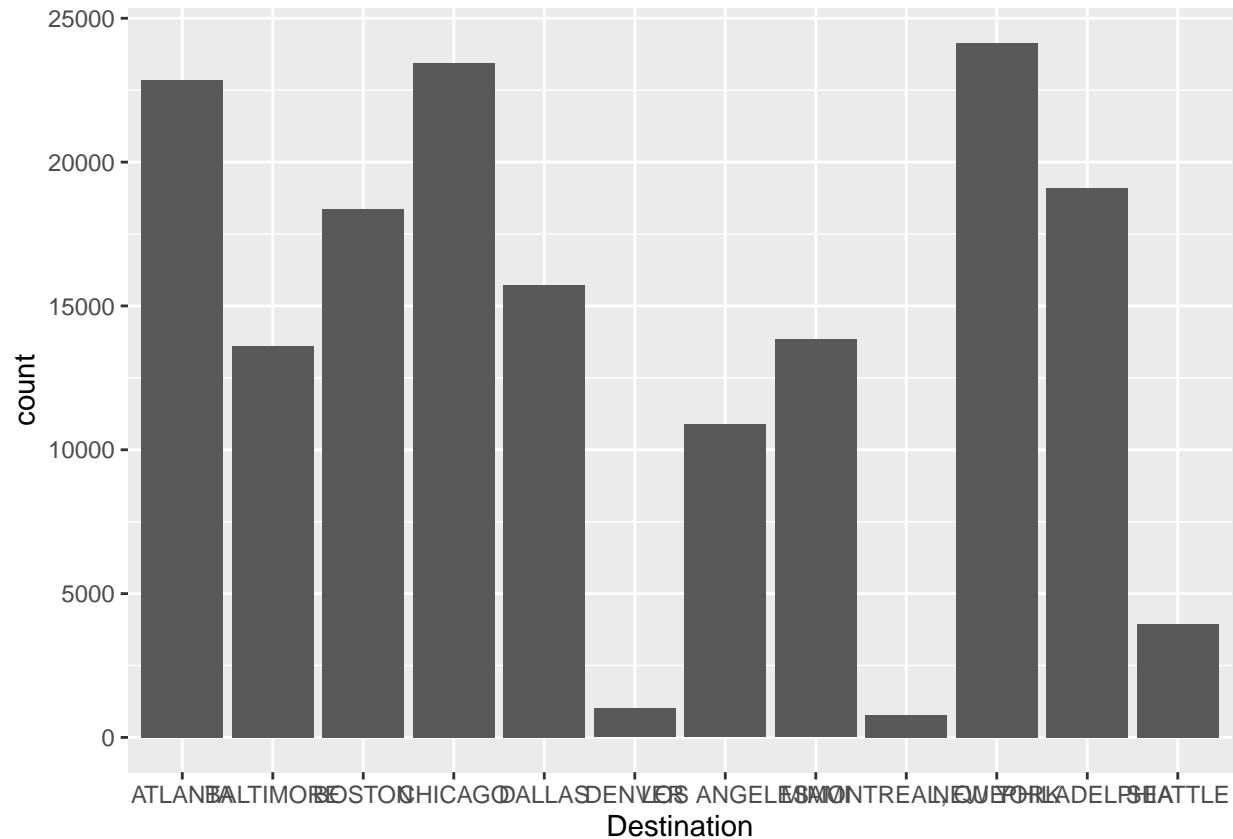
# separate out the data for boston
bostondf <- truckdf[truckdf$Destination == "BOSTON", ]
bostondf <- bostondf[order(bostondf$Origin), ]
boston_mean <- mean(bostondf$Distance, na.rm=TRUE)

# separate out the data for LA
ladf <- truckdf[truckdf$Destination == "LOS ANGELES", ]
la_mean <- mean(ladf$Distance, na.rm=TRUE)
```

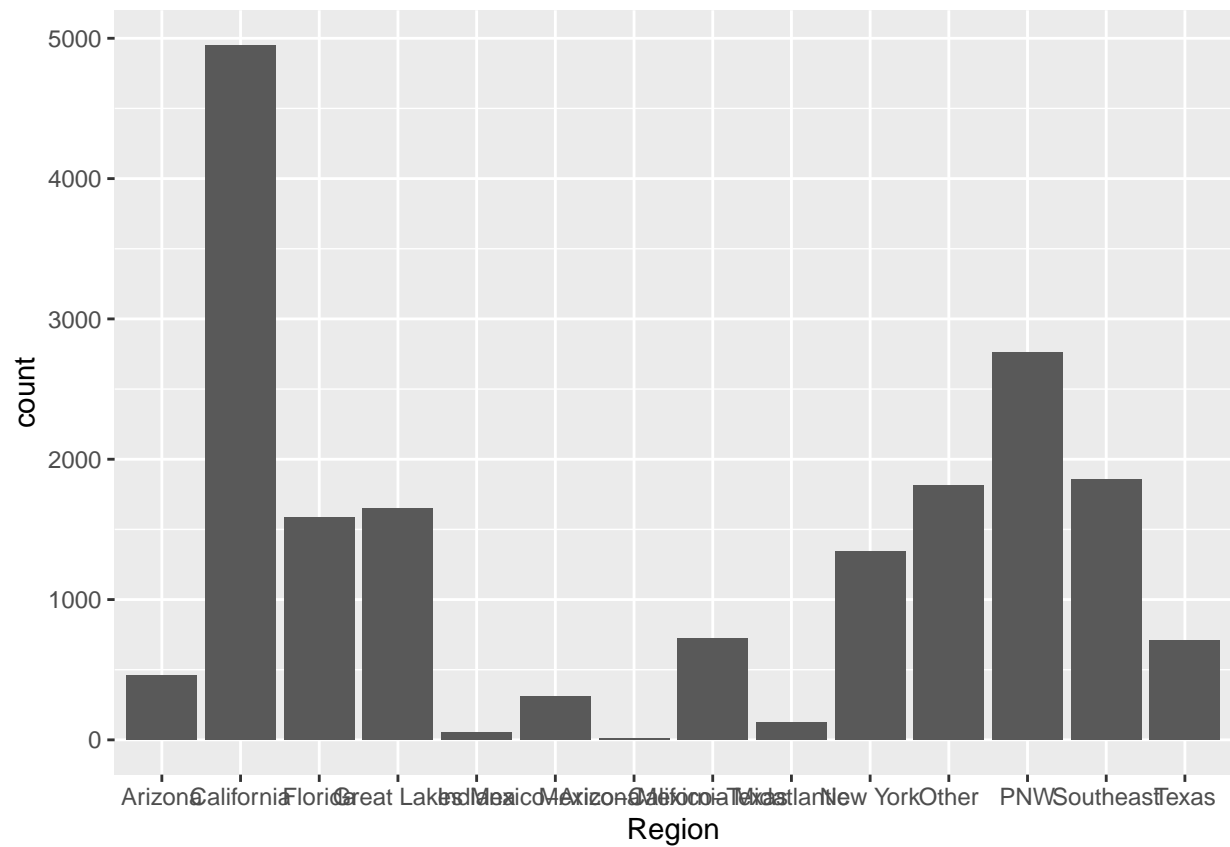
```
# separate out data for seattle
seattledf <- truckdf[truckdf$Destination == "SEATTLE", ]
seattle_mean <- mean(seattledf$Distance, na.rm=TRUE)
```

## Trying Out Some Visualizations:

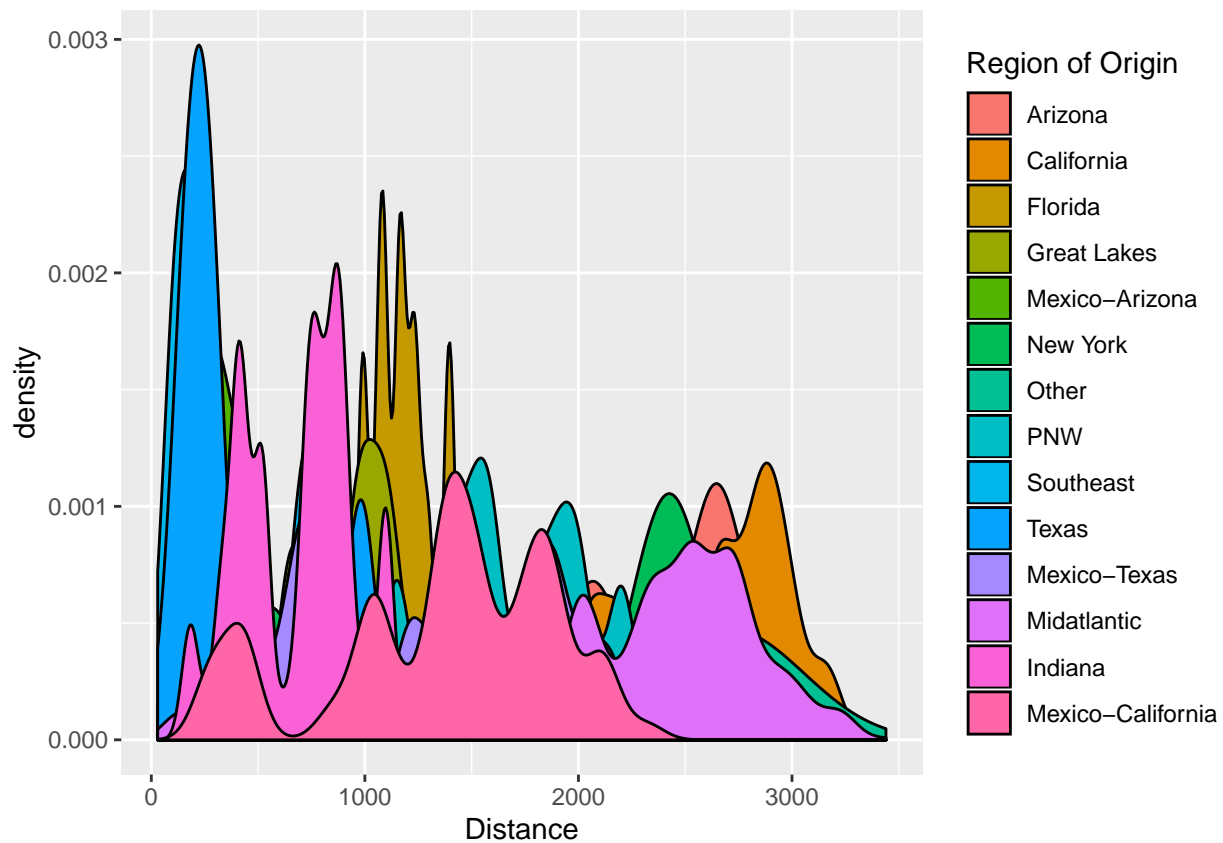
```
# how much data do we have for each of the destination cities?
hist_destination <- ggplot(data=truckdf, aes(x=Destination)) + geom_histogram(stat="count")
hist_destination
```



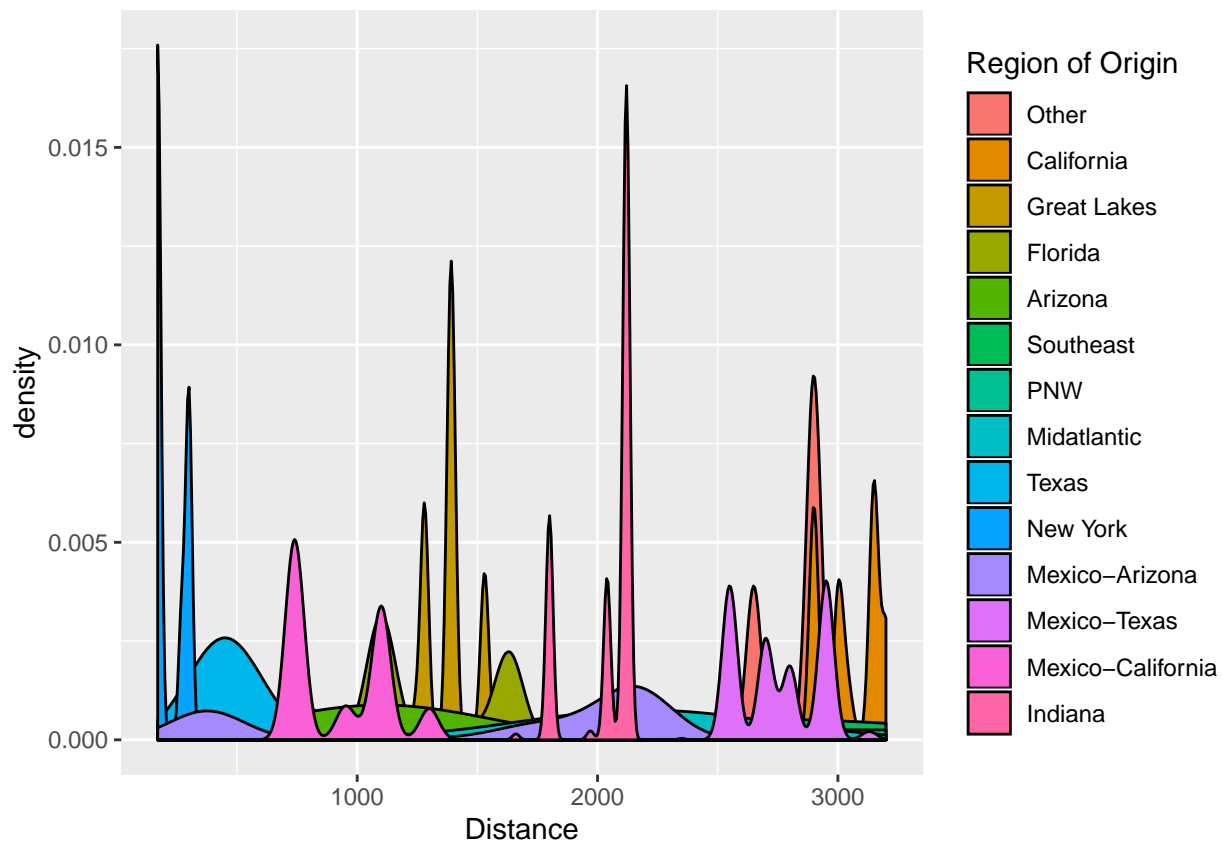
```
# which regions does boston's food come from
hist_boston_region <- ggplot(data=bostondf, aes(x=Region)) + geom_histogram(stat="count")
hist_boston_region
```



```
# which regions send food the furthest distances?
density_region<-ggplot(data=truckdf, aes(x=Distance, fill=Region)) + scale_fill_hue(name="Region of Origin")
density_region
```



```
# which regions send food the furthest distances (to boston)?
density_boston_region<-ggplot(data=bostondf, aes(x=Distance, fill=Region)) + scale_fill_hue(name="Region")
density_boston_region
```



From these visualizations, it looks like food from Texas and New York generally travel the shortest distances, and food from Mexico/Arizona/Southeast have a lot of variability in the distance travelled, although this may be due to the specificity/generalality of the way these regions are categorized.

## Trying Out Some Geospatial Visualizations:

```
mapdf <- read.csv(file="map.csv", sep=",")
uscitiesdf <- read.csv(file="uscities.csv", sep=",")

# List the regions of origin in the original dataset
unique(truckdf$Region)

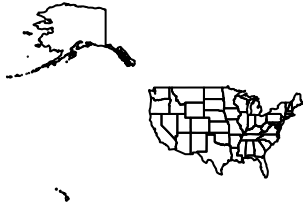
## [1] Arizona      California    Florida      Great Lakes
## [5] Mexico-Arizona New York      Other        PNW
## [9] Southeast    Texas        Mexico-Texas Midatlantic
## [13] Indiana      Mexico-California
## 14 Levels: Arizona California Florida Great Lakes Indiana ... Texas

# We need to map these 14 regions onto more specific states so that we can use geospatial data
# For now, I'm picking a specific state to represent a general area when necessary (ex. Michigan=Great Lakes)

mapping = data.frame(Region=c("Arizona","California","Florida", "Great Lakes", "Mexico-Arizona", "New York",
                              "Southeast", "Texas", "Mexico-Texas", "Midatlantic", "Indiana", "Mexico-California",
                              "Other", "PNW"))
truckdf <- inner_join(truckdf, mapping, by=c("Region"))

library(geojsonio)
usa_spdf <- geojson_read("us-states.json", what = "sp")
mexico_spdf <- geojson_read("mexico.json", what = "sp")
```

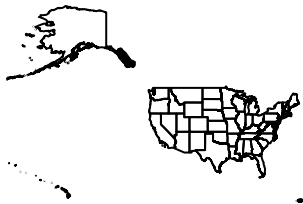
```
usa_github_spdf <- geojson_read("us_states_github.json", what = "sp")
plot(usa_spdf)
```



```
plot(mexico_spdf)
```



```
plot(usa_github_spdf)
```



```
## this is still in progress, but we're working on getting a map of US+Mexico set up so we can visualiz
setwd("/Users/JacquelineLincroft/Documents/INSH2102 Bostonography/Assignments/FinalProject/PoliticalBou
#pb_spdf <- readOGR(
# dsn=setwd("/Users/JacquelineLincroft/Documents/INSH2102 Bostonography/Assignments/FinalProject/Polit
# layer="boundary_l_v2",
# verbose=FALSE)
```