

# Home-Work(2)

## Homework - 2

### Section A

Only use ggplot2 for plotting This section is for testing your ggplot2, and data exploration skills. Dataset msleep from ggplot2 package will be using through this section. Use ? to check the documentation of msleep.

**Problem 1** We are interested in those animals whose awake time over 12 hours. Create a bar chart as the following figure. Remove the NA values from feeding types: carnivore, omnivore, insectivore and herbivore. hints: You may adjust the angle of x-axis label by using `theme(axis.text.x=element_text())`, and the legend labels by using `scale_fill_discrete()`.

```
install.packages('readr')
```

```
## Installing package into '/home/rstudio-user/R/x86_64-pc-linux-gnu-library/3.6'
## (as 'lib' is unspecified)
```

```
install.packages('tidyverse')
```

```
## Installing package into '/home/rstudio-user/R/x86_64-pc-linux-gnu-library/3.6'
## (as 'lib' is unspecified)
```

```
library('ggplot2')
library('readr')
library('tidyverse')
```

```
## — Attaching packages —————
—— tidyverse 1.3.0 —
```

```
## ✓ tibble 2.1.3      ✓ dplyr  0.8.3
## ✓ tidyr  1.0.0      ✓ stringr 1.4.0
## ✓ purrr  0.3.3      ✓ forcats 0.4.0
```

```
## — Conflicts —————
—— tidyverse_conflicts() —
## ✗ dplyr::filter() masks stats::filter()
## ✗ dplyr::lag()    masks stats::lag()
```

```
data("msleep")
head(msleep)
```

```
## # A tibble: 6 x 11
##   name genus vore order conservation sleep_total sleep_rem sleep_cycle
##   <chr> <chr> <chr> <chr> <chr>          <dbl>      <dbl>      <dbl>
## 1 Chee... Acin... carni Carn... lc          12.1        NA        NA
## 2 Owl ... Aotus omni Prim... <NA>         17          1.8        NA
## 3 Moun... Aplo... herbi Rode... nt         14.4         2.4        NA
## 4 Grea... Blar... omni Sori... lc         14.9         2.3        0.133
## 5 Cow   Bos   herbi Arti... domesticated    4          0.7        0.667
## 6 Thre... Brad... herbi Pilo... <NA>         14.4         2.2        0.767
## # ... with 3 more variables: awake <dbl>, brainwt <dbl>, bodywt <dbl>
```

```
?msleep
```

```
##### Dropping NA values
m<-c('vore','order','awake')
new<-msleep[m]
new<-new[new$awake>12,]
new<-new%>%drop_na(vore)
sum(is.na(new$vore))
```

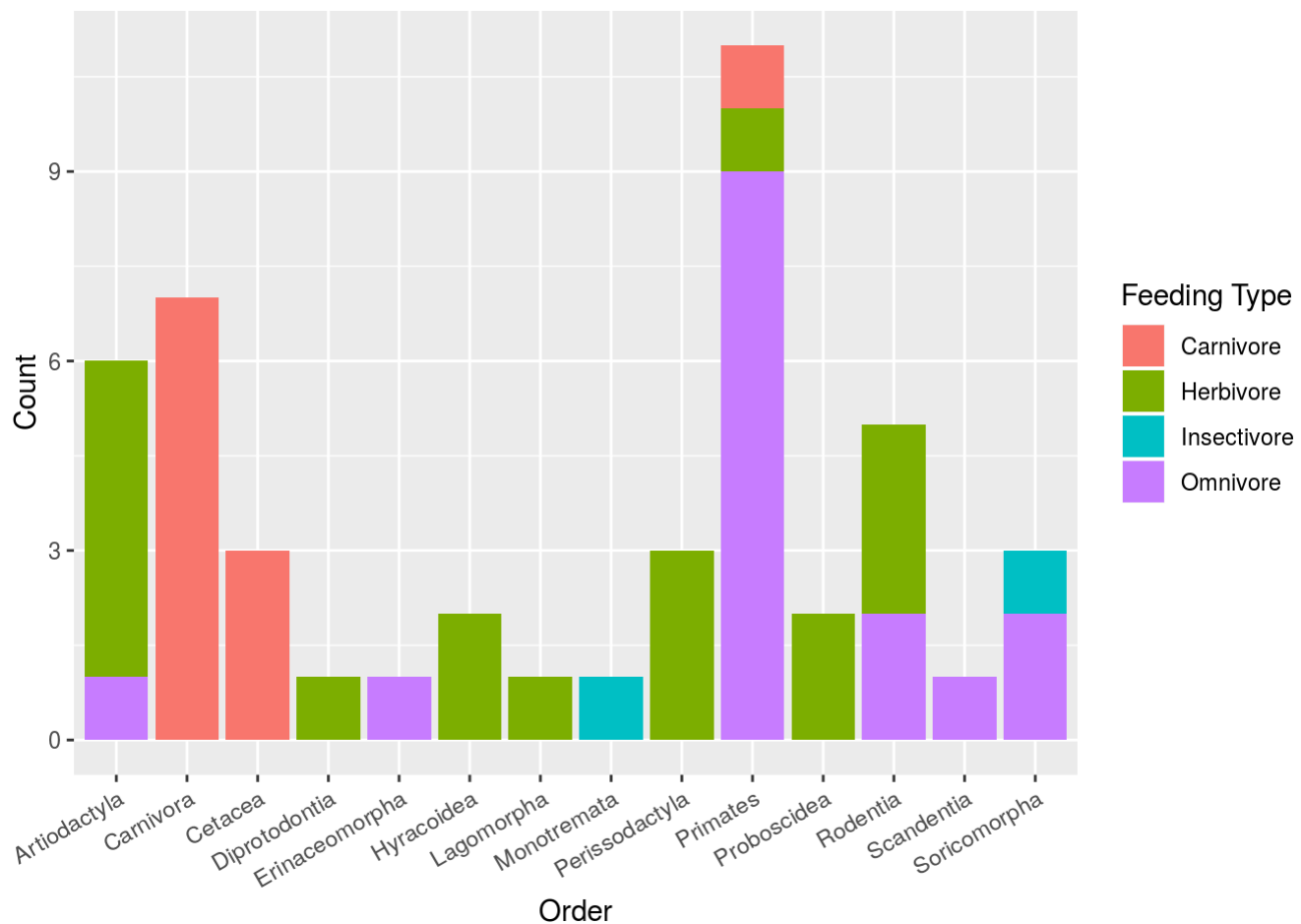
```
## [1] 0
```

```
head(new)
```

```
## # A tibble: 6 x 3
##   vore order      awake
##   <chr> <chr>      <dbl>
## 1 herbi Artiodactyla  20
## 2 carni Carnivora    15.3
## 3 carni Carnivora    13.9
## 4 herbi Artiodactyla  21
## 5 herbi Artiodactyla  18.7
## 6 herbi Rodentia     14.6
```

```
##### Plotting the bar graph
```

```
ggplot(data=new)+geom_bar(mapping=aes(order,fill=vore))+scale_fill_discrete(name='Feeding Type',
labels=c('Carnivore','Herbivore','Insectivore','Omnivore'))+theme(axis.text.x = element_text(ang
le=30,hjust=1))+xlab('Order')+ylab('Count')
```



Problem 2 We would like to investigate how's the relationship between total amount of sleep (hr) and brain weight(kg) among feeding types: carnivore, omnivore, insectivore and herbivore. Plot total amount of sleep (hr) versus brain weight (kg), applying color mapping on the feeding types(vore). Remove the NA group from feeding types. Include a smoothing line on the plot. What do you notice in the plot?

```
m<-c('vore','brainwt','sleep_total')
new<-msleep[m]
new<-new%>%drop_na(vore)
sum(is.na(new$vore))
```

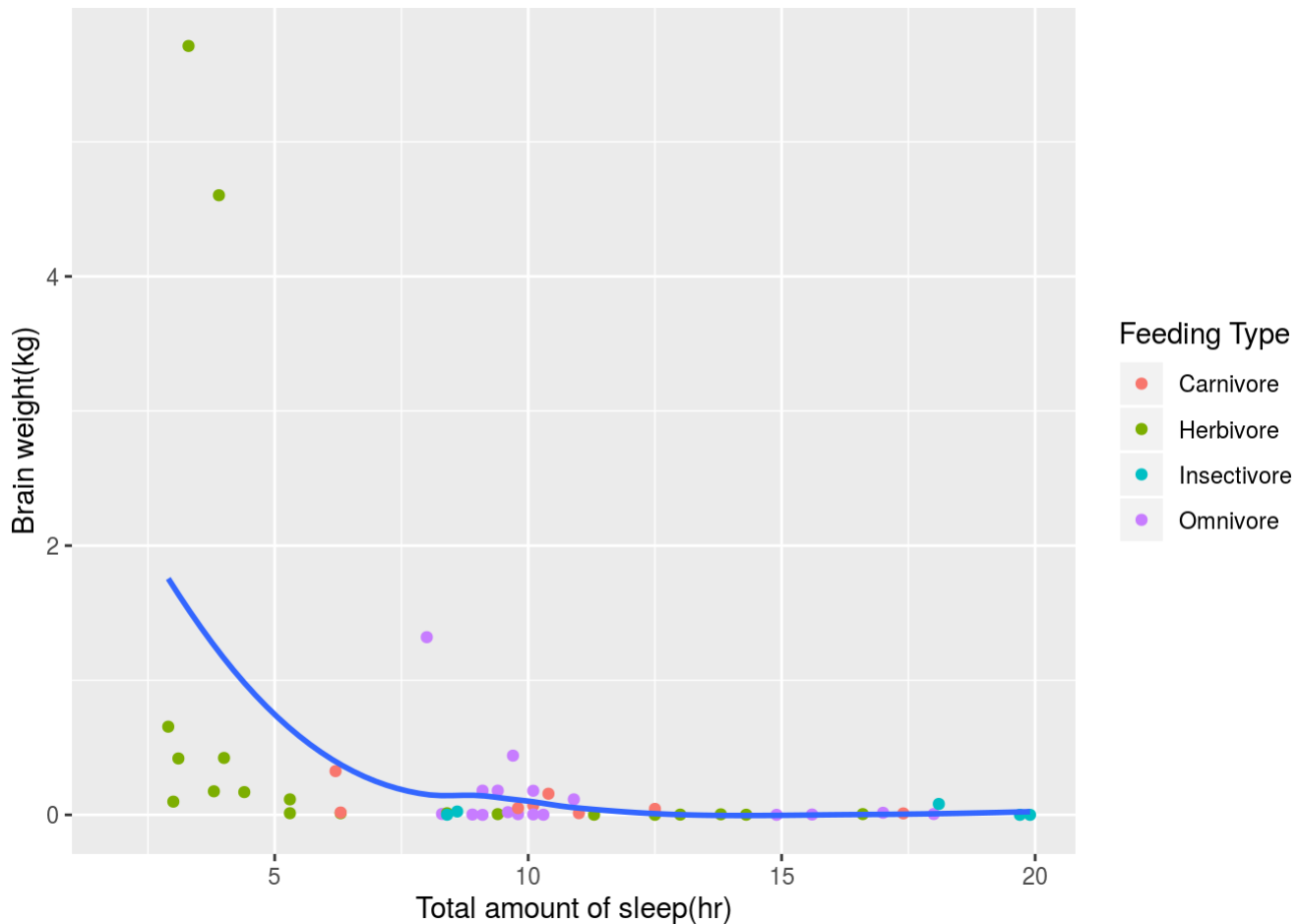
```
## [1] 0
```

```
head(new)
```

```
## # A tibble: 6 x 3
##   vore   brainwt sleep_total
##   <chr>   <dbl>     <dbl>
## 1 carni  NA         12.1
## 2 omni   0.0155      17
## 3 herbi  NA         14.4
## 4 omni   0.00029     14.9
## 5 herbi  0.423        4
## 6 herbi  NA         14.4
```

## ##### Plotting the graph

```
ggplot(data=new,aes(x=sleep_total,y=brainwt))+geom_point(aes(color=vore))+geom_smooth(se=FALSE)+
xlab('Total amount of sleep(hr)')+ylab('Brain weight(kg)')+scale_color_discrete(name='Feeding Type',labels=c('Carnivore','Herbivore','Insectivore','Omnivore'))
```

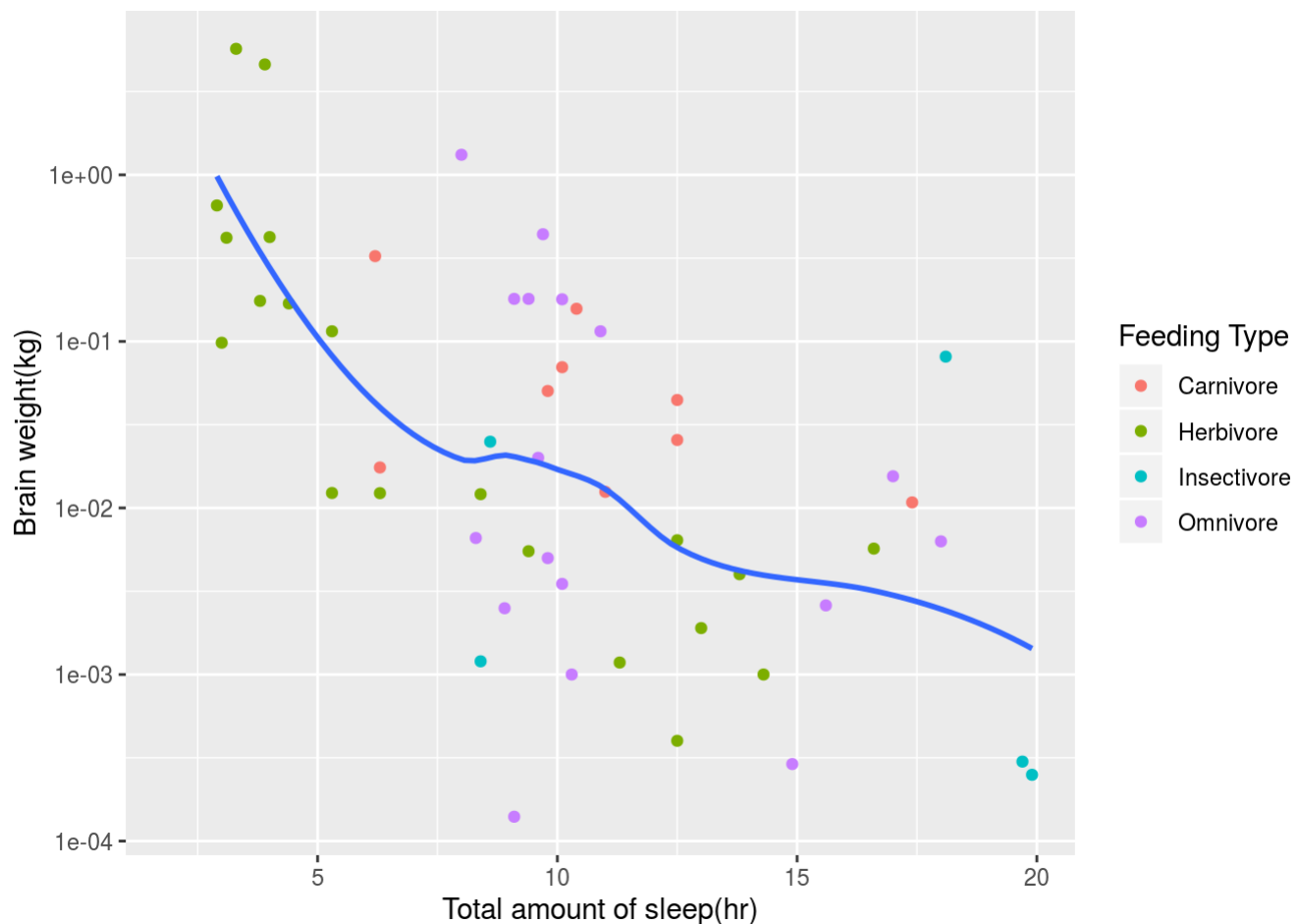


Interpretation: The brain weight of the animals are distributed in the range of 0.001 to 0.4, the smoothing lines helps us to understand the relationship between brain weight and total amount of sleep, the smoothing line linearly decreases and then it reaches to a constant. Most of the animals have similar brain weight with sleeping hours around 10 hours. It is difficult to interpret what is the relationship, because most of the data are skewed around 10 hours. There is also two outliers in the plot.

Problem 3 Still working on the above plot. Apply log transformation on the brain weight Brain Weight (Kg), Log, what do you observe in the plot?

## ##### Plotting using log transformation

```
ggplot(data=new,aes(x=sleep_total,y=brainwt))+geom_point(aes(color=vore))+geom_smooth(se=FALSE)+
scale_y_log10()+xlab('Total amount of sleep(hr)')+ylab('Brain weight(kg)')+scale_color_discrete
(name='Feeding Type',labels=c('Carnivore','Herbivore','Insectivore','Omnivore'))
```



Interpretation: When log transformation is applied to Brain weight the data's are normalised and the smoothing line indicates that the relationship between two variables decreases linearly. After this transformation it is easy to interpret the relationship between total amount of sleep vs brain weight, with Herbivore having maximum brain weight with lesser sleeping hours, whereas the insectivore have higher sleeping hours with lesser brain weight, which can be interpreted easily in comparison with older plot.

## Section B

Only use ggplot2 for plotting Section B uses FY 2019 H-1B Employer Data from U.S. Citizenship and Immigration Services. Download FY2019 H-1B data from: <https://www.uscis.gov/tools/reports-studies/h-1b-employer-data-hub-files> (https://www.uscis.gov/tools/reports-studies/h-1b-employer-data-hub-files) To read the data manual: <https://www.uscis.gov/tools/reports-studies/understanding-our-h-1b-employerdata-hub> (https://www.uscis.gov/tools/reports-studies/understanding-our-h-1b-employerdata-hub) The H-1B is a visa in the United States under the Immigration and Nationality Act, section 101(a)(15)(H) that allows U.S. employers to temporarily employ foreign workers in specialty occupations. A specialty occupation requires the application of specialized knowledge and a bachelor's degree or the equivalent of work experience. Use `read.csv()` to import the dataset to R.

Problem 1 Import the H-1B data. • You may notice the data types of "Initial.Approvals", "Initial.Denials", "Continuing.Approvals", and "Continuing.Denials" are wrong. We need to convert them into numerical columns. • Return a data frame containing the top 5 employers which have the most cases of initial approved H-1B. This data frame should have the columns: employer, initial approvals, initial denials, continuing approvals, and continuing denials. Show the top 5 data frame. • Plot a bar chart of Employer versus Initial approvals, mapping Initial Denials as fill, what do you notice based on the plot?

## ##### Importing the data

```
h1b<-read.csv('h1b_datahubexport-2019.csv',sep=",",stringsAsFactors = FALSE)
head(h1b)
```

```
##   Fiscal.Year                Employer Initial.Approvals
## 1      2019 SOUTHERN CARPET HARDWOOD & TILE IN          1
## 2      2019                UAB HEALTH SYSTEM            0
## 3      2019      BIRMINGHAM VA MEDICAL CENTER            0
## 4      2019                GESTAMP ALABAMA LLC            1
## 5      2019      ARKANSAS HEALTH GROUP                  0
## 6      2019      UNIV OF ARKANSAS AT MONTICELLO            1
##   Initial.Denials Continuing.Approvals Continuing.Denials NAICS Tax.ID
## 1                0                0                0      23      NA
## 2                0                0                1      56      NA
## 3                0                1                0      62      NA
## 4                0                0                0      33      NA
## 5                0                1                0      62      NA
## 6                0                0                0      61      NA
##   State      City      ZIP
## 1    AL BIRMINGHAM 35209
## 2    AL BIRMINGHAM 35233
## 3    AL BIRMINGHAM 35233
## 4    AL    MC CALLA 35111
## 5    AR LITTLE ROCK 72211
## 6    AR MONTICELLO 71656
```

## ##### Transforming the data to numeric

```
h1b<-transform(h1b,Initial.Approvals=as.numeric(gsub(",","",Initial.Approvals)),Initial.Denials=
as.numeric(gsub(",","",Initial.Denials)),Continuing.Approvals=as.numeric(gsub(",","",Continuing.
Approvals)),
                Continuing.Denials=as.numeric(gsub(",","",Continuing.Denials)))

head(h1b)
```

##	Fiscal.Year	Employer	Initial.Approvals
## 1	2019	SOUTHERN CARPET HARDWOOD & TILE IN	1
## 2	2019	UAB HEALTH SYSTEM	0
## 3	2019	BIRMINGHAM VA MEDICAL CENTER	0
## 4	2019	GESTAMP ALABAMA LLC	1
## 5	2019	ARKANSAS HEALTH GROUP	0
## 6	2019	UNIV OF ARKANSAS AT MONTICELLO	1

##	Initial.Denials	Continuing.Approvals	Continuing.Denials	NAICS	Tax.ID
## 1	0	0	0	23	NA
## 2	0	0	1	56	NA
## 3	0	1	0	62	NA
## 4	0	0	0	33	NA
## 5	0	1	0	62	NA
## 6	0	0	0	61	NA

##	State	City	ZIP
## 1	AL	BIRMINGHAM	35209
## 2	AL	BIRMINGHAM	35233
## 3	AL	BIRMINGHAM	35233
## 4	AL	MC CALLA	35111
## 5	AR	LITTLE ROCK	72211
## 6	AR	MONTICELLO	71656

Return a data frame containing the top 5 employers which have the most cases of initial approved H-1B. This data frame should have the columns: employer, initial approvals, initial denials, continuing approvals, and continuing denials. Show the top 5 data frame.

##### Subsetting the data with top approval of H1B based on Initial Approvals

```
m<-(c('Employer','Initial.Approvals','Initial.Denials','Continuing.Approvals','Continuing.Denials'))
h1b_new<-h1b[m]
h1b1<-h1b_new[order(h1b_new$Initial.Approvals,decreasing=T)[1:5],]
h1b1
```

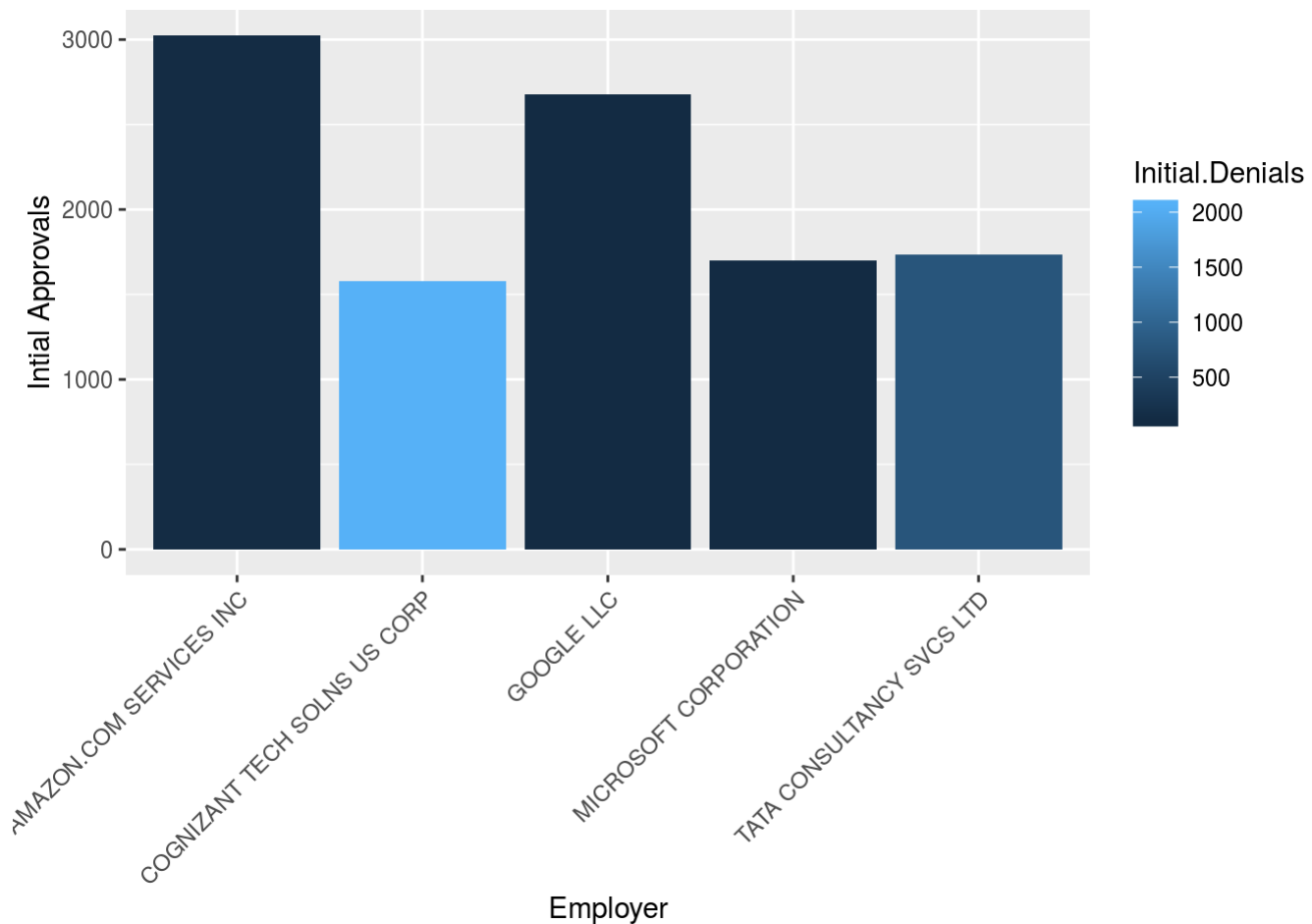
##	Employer	Initial.Approvals	Initial.Denials
## 51300	AMAZON.COM SERVICES INC	3026	122
## 47554	GOOGLE LLC	2678	104
## 59158	TATA CONSULTANCY SVCS LTD	1733	763
## 55790	MICROSOFT CORPORATION	1701	109
## 5060	COGNIZANT TECH SOLNS US CORP	1580	2060

##	Continuing.Approvals	Continuing.Denials
## 51300	4186	133
## 47554	3333	53
## 59158	5859	1376
## 55790	3560	66
## 5060	11783	3910

##### Plotting a bar plot according to approvals:

```
ggplot(data=h1b1,aes(x=Employer,y=Initial.Approvals))+geom_bar(stat='identity',mapping=aes(fill=Initial.Denials))+theme(axis.text.x=element_text(angle=45,hjust=1))+xlab('Employer')+ylab('Initial Approvals')
```



Problem 2 Download geocode data <https://public.opendatasoft.com/explore/dataset/us-zip-code-latitude-and-longitude/> (<https://public.opendatasoft.com/explore/dataset/us-zip-code-latitude-and-longitude/>) export/?location=3,43.25174,-106.27166&basemap=jawg.streets. • Join H-1B data table with geocode data table by State and Zip columns. • This new data frame should include columns: zip, employer, initial approvals, initial denials, continuing approvals, continuing denials, state, city, longitude, and latitude. • Insert a new column prop into this new data frame by the formula: initial denial/initial approval

##### Importing the data

```
geocode<-read.csv('us-zip-code-latitude-and-longitude.csv',sep=';')
head(geocode)
```



```
##      Zip      City State Latitude Longitude Timezone
## 1 71937      Cove   AR 34.39848 -94.39398      -6
## 2 72044 Edgemont   AR 35.62435 -92.16056      -6
## 3 56171 Sherburn   MN 43.66085 -94.74357      -6
## 4 49430  Lamont    MI 43.01034 -85.89754      -5
## 5 52585 Richland   IA 41.19413 -91.98027      -6
## 6 47520 Cannelton  IN 37.93431 -86.67821      -5
## Daylight.savings.time.flag      geoint
## 1      1 34.398483,-94.39398
## 2      1 35.624351,-92.16056
## 3      1 43.660847,-94.74357
## 4      1 43.010337,-85.89754
## 5      1 41.194129,-91.98027
## 6      0 37.934311,-86.67821
```

##### *Joining the table*

```
h1b_new<-merge(h1b,geocode,by.x=c('ZIP','State'),by.y=c('Zip','State'))

head(h1b_new)
```

```
##      ZIP State Fiscal.Year      Employer
## 1 10001   NY      2019      HAYMARKET MEDIA INC
## 2 10001   NY      2019      SHINAN BANK AMERICA
## 3 10001   NY      2019  BISLEY INC DBA BISLEY N AMERICA
## 4 10001   NY      2019      TRIALSPARK INC
## 5 10001   NY      2019      33ACROSS INC
## 6 10001   NY      2019 ANIKA PHARMACY CORP DBA LORVEN PHA
## Initial.Approvals Initial.Denials Continuing.Approvals
## 1      0      1      0
## 2      0      0      1
## 3      0      0      1
## 4      1      0      3
## 5      2      0      0
## 6      1      0      0
## Continuing.Denials NAICS Tax.ID City.x City.y Latitude Longitude
## 1      0    54    1585 NEW YORK New York 40.75074 -73.99653
## 2      1    52    1762 NEW YORK New York 40.75074 -73.99653
## 3      0    23    8497 NEW YORK New York 40.75074 -73.99653
## 4      1    54    4239 NEW YORK New York 40.75074 -73.99653
## 5      0    54    3623 NEW YORK New York 40.75074 -73.99653
## 6      0    44    2948 NEW YORK New York 40.75074 -73.99653
## Timezone Daylight.savings.time.flag      geoint
## 1      -5      1 40.750742,-73.99653
## 2      -5      1 40.750742,-73.99653
## 3      -5      1 40.750742,-73.99653
## 4      -5      1 40.750742,-73.99653
## 5      -5      1 40.750742,-73.99653
## 6      -5      1 40.750742,-73.99653
```

##### Subsetting the dataframe

```
m<-c('ZIP','Employer','Initial.Approvals','Initial.Denials','Continuing.Approvals','Continuing.Denials','State','City.x','City.y','Longitude','Latitude')
```

```
h1b_new<-h1b_new[m]
head(h1b_new)
```

```
##      ZIP                      Employer Initial.Approvals
## 1 10001          HAYMARKET MEDIA INC              0
## 2 10001          SHINAN BANK AMERICA              0
## 3 10001    BISLEY INC DBA BISLEY N AMERICA          0
## 4 10001          TRIALSPARK INC                    1
## 5 10001          33ACROSS INC                      2
## 6 10001 ANIKA PHARMACY CORP DBA LORVEN PHA          1
##  Initial.Denials Continuing.Approvals Continuing.Denials State  City.x
## 1              1              0              0    NY NEW YORK
## 2              0              1              1    NY NEW YORK
## 3              0              1              0    NY NEW YORK
## 4              0              3              1    NY NEW YORK
## 5              0              0              0    NY NEW YORK
## 6              0              0              0    NY NEW YORK
##      City.y Longitude Latitude
## 1 New York -73.99653 40.75074
## 2 New York -73.99653 40.75074
## 3 New York -73.99653 40.75074
## 4 New York -73.99653 40.75074
## 5 New York -73.99653 40.75074
## 6 New York -73.99653 40.75074
```

##### Adding proportion as a new column to the dataframe

```
h1b_new$prop<-h1b_new$Initial.Denials/h1b_new$Initial.Approvals
head(h1b_new)
```

```
##      ZIP                      Employer Initial.Approvals
## 1 10001          HAYMARKET MEDIA INC                0
## 2 10001          SHINAN BANK AMERICA                0
## 3 10001    BISLEY INC DBA BISLEY N AMERICA          0
## 4 10001          TRIALSPARK INC                    1
## 5 10001          33ACROSS INC                      2
## 6 10001 ANIKA PHARMACY CORP DBA LORVEN PHA          1
##      Initial.Denials Continuing.Approvals Continuing.Denials State   City.x
## 1          1                0                0    NY NEW YORK
## 2          0                1                1    NY NEW YORK
## 3          0                1                0    NY NEW YORK
## 4          0                3                1    NY NEW YORK
## 5          0                0                0    NY NEW YORK
## 6          0                0                0    NY NEW YORK
##      City.y Longitude Latitude prop
## 1 New York -73.99653 40.75074  Inf
## 2 New York -73.99653 40.75074  NaN
## 3 New York -73.99653 40.75074  NaN
## 4 New York -73.99653 40.75074   0
## 5 New York -73.99653 40.75074   0
## 6 New York -73.99653 40.75074   0
```

Problem 3 We are interested in the H-1B cases around Bay Area, California. Create a map of the California, and then adjust the plotting x/y limits to a proper zoom level of Bay Area. Then showing the locations of each employer along with, the prop less than 0.1 (mapped as the color/fill), and the initial approvals (mapped as the size). hints: Install map and mapproj packages, and use the ggplot2::map\_data() to draw "California" region of the US.

```
install.packages('maps')
```

```
## Installing package into '/home/rstudio-user/R/x86_64-pc-linux-gnu-library/3.6'
## (as 'lib' is unspecified)
```

```
install.packages('mapproj')
```

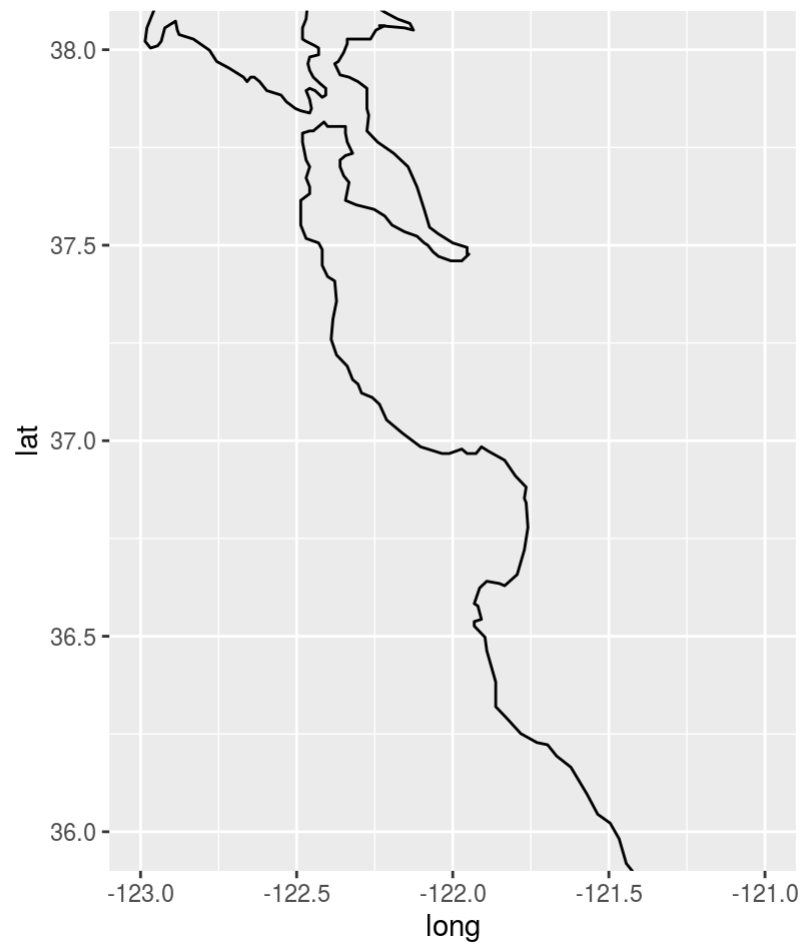
```
## Installing package into '/home/rstudio-user/R/x86_64-pc-linux-gnu-library/3.6'
## (as 'lib' is unspecified)
```

```
##### Reading the california map
```

```
cal<- map_data('state',region=c('California'))
```

```
ca_base<-ggplot(cal)+geom_polygon(mapping=aes(x=long,y=lat,group=group),fill='NA',color='Black')
+coord_quickmap(xlim = c(-123, -121.0), ylim = c(36, 38))
```

```
ca_base
```



```
##### Finding the proportions less than 0.1
```

```
pro<-h1b_new[h1b_new$prop<0.1,]  
head(pro)
```

```
##      ZIP      Employer Initial.Approvals
## NA      NA      <NA>      NA
## NA.1    NA      <NA>      NA
## 4      10001    TRIALSPARK INC      1
## 5      10001    33ACROSS INC      2
## 6      10001    ANIKA PHARMACY CORP DBA LORVEN PHA      1
## 7      10001    SHAREBITE INC      1
##      Initial.Denials Continuing.Approvals Continuing.Denials State
## NA      NA      NA      NA <NA>
## NA.1    NA      NA      NA <NA>
## 4      0      3      1      NY
## 5      0      0      0      NY
## 6      0      0      0      NY
## 7      0      0      0      NY
##      City.x    City.y Longitude Latitude prop
## NA      <NA>    <NA>      NA      NA      NA
## NA.1    <NA>    <NA>      NA      NA      NA
## 4      NEW YORK New York -73.99653 40.75074 0
## 5      NEW YORK New York -73.99653 40.75074 0
## 6      NEW YORK New York -73.99653 40.75074 0
## 7      NEW YORK New York -73.99653 40.75074 0
```

##### Plotting the employers and proportions less than 0.1

```
ca_base+geom_point(data=pro,aes(x=Longitude,y=Latitude,color=prop,size=`Initial.Approvals`))
```

