

## Sprint-3

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Project Name	Project –SignswithSmartConnectivityforBetterRoadSafety

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### ApplicationPackages:

package

com.example.myhp.accidentprevention;importandroid.a

pp.Application;

importandroid.test.ApplicationTestCase;

/\*\*

<ahref="http://d.android.com/tools/testing/testing\_android.html">TestingFund>  
\*/

publicclassApplicationTestextendsApplicationTestCase<Application>

{

    publicApplicationTest()

{

    super(Application.class);

}

}

### MAIN:

\#Manipulatingdata{ #data }

Thissectionisanintroductiontomanipulatingdatasetsusingthe`dplyr`package.

Asoutlinedintheprevioussection,`dplyr`and

`ggplot2`arepartofthe`tidyverse`,whichaimstoprovideauser-friendlyframeworkfordatascience[@grolemund\_data\_2016].

Experience of teaching R over the past few years suggests that many people find it easier to get going with data driven research if they learn the 'tidy' workflow presented in this section.

However, if you do not like this style of R code or you are simply curious, we encourage you to try alternative approaches for achieving the similar results using base R [ @rcoreteam\_language\_2020 ] [

Run the command `help.start()` to see a resources introducing base R, and [Chapter 6 on lists and data frames] (<https://cran.r-project.org/doc/manuals/r-release/R-intro.html#Lists-and-data-frames>) in [An Introduction to R] (<https://cran.r-project.org/doc/manuals/r-release/R-intro.pdf>) in particular for an introduction to data manipulation with base R.

]

, the `data.table` R package [ @R-

`data.table`] or other languages such as [Python] (<https://www.python.org/>) or [Julia] (<https://julialang.org/>).

If you just want to get going with processing data, the `tidyverse` is a solid and popular starting point.

<!-- Todo: add new part here? -->

Before diving into the `tidyverse`, it is worth re-capping where we have gotten so far as we have covered a lot of ground.

Section \@ref(basics) introduced R's basics syntax; Section \@ref(rstudio) showed how to use the Source Editor and other features of RStudio to support data science; and Section

\@ref(pkgs) introduced the concept and practicalities of R packages, with reference to `stats19`, `ggplot2` and `dplyr`.

In this section, we will start with a blank slate.

In Section \@ref(basics) we learned that in R having a 'clean desk' means an `*empty global environment*`. This can be achieved by running the following command, which removes the `list()` of all objects returned by the function `ls()`:

```
```{r}
rm(list=ls())
```
```

## tibbles

Although the data processing techniques in R are capable of handling large datasets, such as the `crashes_2019` object that we created in the previous section, representing 100k+ casualties, it makes sense to start small.

Let's start by re-creating the `crashes` dataset from Section \@ref(basics), but this time using the `tidyverse` `tibble()` function. This is the `tidyverse` equivalent of base R's `data.frame`.

`tibble` objects can be created, after loading the `tidyverse`, as follows:

```
```{r, message=FALSE}
library(tidyverse)
crashes = tibble(
```

```
casualty_type=c("pedestrian","cyclist","cat"),cas
ualty_age           =seq(from
=20,to=60,by=20),vehicle_type=c("car","bus",
"tank"),
dark=c(TRUE,FALSE,TRUE)
)
...
```

In the previous code chunk, we passed four vector objects as \*named arguments\* to the `tibble` function, resulting in columns such as `casualty_type`. A `tibble` is just a fancy way of representing `data.frame` objects, preferred by `tidyverse` users and optimised for data science. It has a few sensible defaults and advantages compared with the `data.frame`, one of which can be seen by printing a `tibble`:

```
```{r} class(crashes
)crashes
```
```

Note the `<chr>`, `<dbl>` or `<lgl>` text below each column, providing a quick indication of the class of each variable- this is not provided when using `data.frame`.

## `filter()` and `select()` rows and columns

In the previous section, we briefly introduced the package `dplyr`, which provides an alternative to base R for manipulating objects. `dplyr` provides different, and some would argue simpler, approaches for subsetting rows and columns than base R. `dplyr` operations for subsetting rows (with the function `filter()`) and columns (with the function `select()`) are demonstrated below. Here we can also see the use of the pipe operator `%>%` to take the dataset and apply the function to that dataset.

```
```{r}
crashes %>% filter(casualty_age > 50) #
filters rows crashes %>% select(casualty_type) # select just one
column
```
```

It should be clear what happened: `filter()` returns only rows that match the criteria in the function call, only 9 observations with a `casualty_age` greater than 50 in this case. Likewise, `select()` returns data objects that include only columns named inside the function call, `casualty_type` in this case.

To gain a greater understanding of the functions, type and run the following commands, which also illustrate how the `%>%` can be used more than once to manipulate data (more on this soon):

```
```{r}
```

```
crashes_darkness=crashes%>%filter(dark)crashes_a=crashes%>%select(contains("a"))crashes_darkness_a = crashes%>%filter(dark)%>%select(contains("a"))
```

```

Can you guess what the dimensions of the resulting objects will be?

Write down your guesses for the number of rows and number of columns that the new objects, `crashes\_darkness` to `crashes\_darkness\_a`, have before running the following commands to find out. This also demonstrates the handy function `dim()`, short for dimension (results not shown):<sup>^</sup>[  
Note that the number of rows is reported before the number of columns.  
This is a feature of R: rows are also specified first when subsetting using the square brackets in commands such as `crashes[1,2:3]`.  
]

```
```{r,eval=FALSE}dim(crashes)dim(crashes_darkness)?contains#get help on contains() to help guess the output of the next line dim(crashes_a)dim(crashes_darkness_a)
```

```

Look at the help pages associated with `filter()`, `select()` and the related function `slice()` as follows and try running the examples that you will find at the bottom of the help pages for each to gain a greater understanding (note you can use the `package::function` notation to get help on functions also):

```
```{r,eval=FALSE}
?dplyr::filter
?dplyr::select
?dplyr::slice
```

```

## ## Ordering and selecting the 'top n'

Other useful pipe-friendly functions are `arrange()` and `top\_n()`. `arrange()` can be used to sort data. Within the `arrange()` function, optional arguments can be used to define the order in which it is sorted. `top\_n()` simply selects the top 'n' number of rows in your data frame. We can use these functions to arrange datasets and take the top most 'n' values, as follows:

```
```{r}crashes
%>%

```

```

    arrange(vehicle_type)crashes%>%
    top_n(n=1,wt=casualty_age)
  }
}

```

<!--##Longandwidedata-->

## ##Summarise

Apowerfultwo-functioncombinationis`group\_by()`and`summarise()`.Usedtogether,theycanprovide\*groupedsummaries\*ofdats. Intheexamplebelow,wefindthemeanageofcasualtiesindarkandlight conditions.

```

{r}crashes%>%
  group_by(dark)%>%
  summarise(mean_age=mean(casualty_age))

```

Theexampleaboveshowsapowerfulfeatureofthesepipelines.Manyoperationscanbe'chained'together,whilstkeepingreadabilitywithsubsequentcommandsstackedbelowearlieroperations.Thecombinationof`group\_by()`and`summarise()`canbeveryusefulinpreparingdataforvisualisationwitha`ggplot2` function.

Anotherusefulfeatureofthe`tidyverse`fromauserperspectiveistheautocompletionofcolumnnames midpipe.

Ifyouhavenotnoticedthisalready, youcantestitbytypingthefollowing,puttingyourcursorjustbeforethe`)`and pressing `Tab`:

```

{r,eval=FALSE}
crashes%>%select(ca)#pressTabwhenyourcursorisjustafterthea

```

You should see`casualty\_age`and`casualty\_type`popupasoptionsthatcanbeselectedby pressing`Up`and`Down`.

This may not seemlikemuch,but whenanalysinglargedatswithdozensofvariables,it canbeagodsend.

Ratherthanprovidingacomprehensiveintroductiontothe`tidyverse`suiteofpackages,thissectionshouldhaveofferedenoughtogetstarted withusingitforroadsafetydataanalysis.Forfurtherinformation,checkoutup-to-dateonlinecoursesfromrespectedorganisationslike[DataCarpentry](https://datacarpentry.org/R-ecology-lesson/index.html)andthefreeonline[books](https://bookdown.org/)suchas[RforDataScience](https://r4ds.had.co.nz/)[@golemund\_data\_2016].

## ##Tidyverseexercises

1. Use `dplyr` to filter rows in which `casualty_age` is less than 18, and then 28.
2. Use the `arrange` function to sort the `crashes` object in descending order of age (\*\*Hint: \*\*see the `?arrange` help page).
3. Read the help page of `dplyr::mutate()`. What does the function do?
4. Use the `mutate` function to create a new variable, `birth_year`, in the `crashes` data.frame which is defined as the current year minus their age.
5. \*\*Bonus:\*\* Use the `%>%` operator to filter the output from the previous exercises so that only observations with `birth_year` after 1969 are returned.

```
```{rdplyr,eval=FALSE,echo=FALSE}
#answerscrashes
%>%
  arrange(desc(casualty_age))
crashes%>%filter(casualty_age>21)cras
hes%>%
  mutate(birth_year=2019-
  casualty_age)%>%filter(birth_year>1969)
```
```

## Slides:

```
---
title:"RoadSafety(andtransport)ResearchwithR"
#subtitle:"remojiifont::emoji("bike")`<br/>ForEnglandandWales'subtitle:"`r
emojiifont::emoji("rocket")`<br/>RACFoundation,DataDriven'author:"Robi
nLovelace"
date:'2020'outpu
t:
  xaringan::moon_reader:
    #css:["default","its.css"]
    #chakra:libs/remark-
    latest.min.jslib_dir:libs
  nature:
    highlightStyle:githubhighlightLines:
    true
```

```
#bibliography:
#-../vignettes/ref.bib
#-../vignettes/ref_training.bib
---
```

```
```{rsetup,include=FALSE,eval=FALSE}
#getcitations
refs=RefManageR::ReadZotero(group="418217",,params=list(collection="JFR868KJ",limit=
100))
refs_df=as.data.frame(refs)
#View(refs_df)
```

```
#citr::insert_citation(bib_file="vignettes/refs_training.bib")RefManageR::WriteBib(refs,
"refs.bib")
#citr::tidy_bib_file(rmd_file="vignettes/pct_training.Rmd",messy_bibliography="vignett
es/refs_training.bib")
options(htmltools.dir.version=FALSE)knitr::o
pts_chunk$set(message=FALSE)library(RefM
anageR)BibOptions(check.entries=FALSE,
  bib.style="authoryear",c
  ite.style='alphabetic',sty
  le="markdown",first.ini
  ts=FALSE,hyperlink=F
  ALSE,dashed=FALSE)
my_bib=refs
```

```

```
```{r,eval=FALSE,echo=FALSE,engine='bash'}
#publishresultsonline
cp-Rvcode/rrsrr-slides*~/saferactive/site/static/slides/cp-
Rvcode/libs~/saferactive/site/static/slides/
cd~/saferactive/sitegit
add-A
gitstatus
gitcommit-
am'Updateslides'gitpush
cd-
```

```

#Slide/links<https://itsleeds.githubu>

[b.io/rrsrr/https://bookdown.org/](https://bookdown.org/)

<https://www.pct.bike/>

---

background-image:url(<https://media.giphy.com/media/YlQQYUIEAZ76o/giphy.gif>)

#CodingIdeal:

```

```{r,eval=FALSE}
od_test$perc_cycle=round(od_test$bicycle/od_test$all)*100l=od_to_sf(od_test,od_data_centroids)
r=stplanr::route(l=l,route_fun=journey)met
=overline(r,"bicycle")
```

```

--

Realit

y

---

##Transportsoftware- whichdoyouuse?

```

```{r,echo=FALSE,message=FALSE,warning=FALSE}
u="https://github.com/ITSLeeds/TDS/raw/master/transport-software.csv"
tms=readr::read_csv(u)[1:5]
tms=dplyr::arrange(tms,dplyr::desc(Citations))
knitr::kable(tms,booktabs=TRUE,caption="Sampleoftransportmodellingsoftwareinusebypractitioners.Note:citationcountsbasedonsearchesforcompany/developername,theproductnameand'transport'.Dataprovider:GoogleScholar searches,October2018.",format="html")
```

```

---

##Datascienceandthetidyverse

-InspiredbyIntroductiontodatasciencewithR(availablefree[online](https://r4ds.had.co.nz/))

```

```{rtds-cover,echo=FALSE,out.width="30%"}knitr::include_graphics("https://d33wubrfki0l68.cloudfront.net/b88ef926a004b0fce72b2526b0b5c4413666a4cb/24a30/cover.png")
```

```

---

##Ageographicperspective

- See<https://github.com/ITSLeeds/TDS/blob/master/catalogue.md>

- Paperonthe\*\*stplanr\*\*paperfortransportplanning(available[online](https://cran.r-



project.org/web/packages/stplanr/vignettes/stplanr-paper.html))  
- Introductory and advanced content on geographic data in R, especially the [transport chapter](https://geocompr.robinlovelace.net/transport.html)(available free [online](https://geocompr.robinlovelace.net/))  
- Paper on analysing OSM data in Python with OSMnx (available [online](https://arxiv.org/pdf/1611.01890))

---

#Gettingsupport

--

With open source software, the world is your support network!

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- Recent example: <https://stackoverflow.com/questions/57235601/>

--

- [gis.stackexchange.com](https://gis.stackexchange.com/questions) has 21,314 questions

- [r-sig-geo](https://r-sig-geo.2731867.n2.nabble.com/) has 1000s of posts

- RStudio's Discourse community has 65,000+ posts already!

--

- Not transport equivalent (e.g. earthscience.stackexchange.com is in beta)

- Potential for a Discourse forum or similar: transport is not (just) GIS

---

## **Textcenter\_analysis:**

---

pagetitle: "Analysis based on test centres" author: "

Amol Nanaware"

date: "06/12/2019" always\_allo

w\_html: true output:

html\_document: default

---

```

```{rsetup,include=FALSE}knitr::opts_chunk$set(echo
=TRUE)
```

```{r,warning=FALSE,echo=FALSE,include=FALSE}packages
<-c("plotly", "tidyverse")
newPackages<-
packages[!(packages%in%installed.packages()[,"Package"])]if(length(newPackages))install.pac
kages(newPackages)

library(tidyverse)library(plotl
y)

```

```{r,echo=FALSE}load("pas
sfail.RData")passfail<-
passfail%>%
  mutate(totalFails=Fail1+ifelse(is.na(Fail2),0,Fail2),Totalpass=Pass1+ifelse(is.na(Pass2),0,
Pass2))
```

```{r,echo=FALSE}

passfailGroup<-summarise(group_by(passfail,
Centre),Pass1=sum(Pass1),Fail1=sum(Fail1),Total1=sum(Total1),Pass2=sum(Pass2,na.rm=T),
Fail2=sum(Fail2,na.rm=T),Total2=sum(Total2,na.rm=T),Totalpass=sum(Totalpass),totalFails=
sum(totalFails))

passfailGroup<-
mutate(passfailGroup,Pass1prop=Pass1/Total1,Pass2prop=Pass2/Total2,totalPassProp=(Totalp
ass/(Total1+Total2)),totalFailsProp=(totalFails/(Total1+Total2)))

```

```{r,echo=FALSE}
passfailGroup$totalPassProp=round((passfailGroup$totalPassProp*100),digits=2)passfailGrou
p$totalFailsProp=round((passfailGroup$totalFailsProp*100),digits=2)

passFailGroup1<-
passfailGroup[c(1,8)]passFailGroup1$Test <-
"Pass"names(passFailGroup1)<-
c("Centre","Count","Test")passFailGroup2<-
passfailGroup[c(1,9)]passFailGroup2$Test<-"Fail"
names(passFailGroup2)<-
c("Centre","Count","Test")passFailcount<-
rbind(passFailGroup1,passFailGroup2)
```

```

#### Analysis based on test centres

<br/> In this section we will analyse data from 2013 till 2018 about each test centre. As shown in the <a href="https://github.com/NanawareAmol/R-project\_Road-safety/blob/master/Result/loc\_spread\_across\_ireland.JPG">map</a>, the test centres are spread across the Ireland and the number of centres is more in highly populated areas such as Dublin, Cork etc.

The bar chart shows the total number of tests that each centre performed and the total pass and fail counts as well as percentages. So, based on the test counts, the top 3 test centres are, \*Fonthill(770685)\*, \*Deansgrange(767484)\*, and \*Northpoint2(729661)\*. The bottom 3 centres which performed the least tests are, \*Donegal Town(16315)\*, \*Cahir Civeen(28806)\* and \*Clifden(38683)\*.

```
```{r, echo=FALSE, fig.width=9, fig.height=4}
p<-
plot_ly(passfailGroup, x=~passfailGroup$Centre, y=~passfailGroup$Totalpass, type='bar', name='
Pass',
text=paste("Total tests=", (passfailGroup$Totalpass+passfailGroup$totalFails), "<br>Passed=", pa
ssfailGroup$totalPassProp, "%", "<br>Failed=", passfailGroup$totalFailsProp, "%"), opacity=0.5, m
arker=list(color=' #3AC3E3', line= list(color='#0D6EB0', width=
1)))%>%add_trace(y=~passfailGroup$totalFails, name='Fails', opacity=0.5,
marker=list(color='#0E84FF', line=list(color='#0D6EB0', width=1)))%>%layout(yaxis=li
st(title='Count'), xaxis=list(title='Test Centres'), barmode='stack')
p
```
```

#### <b>Total test passed for each test centre</b>

The following scatter plot shows the total test pass count for each test centre from the year 2013 till year 2018. The question that can be answered by this graph are, <br/>

1. Which are the top 3 and last 3 centres based on total pass count? <br/>

&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;<b>(Deansgrange, Northpoint2, Fonthill and Cahirciveen, Clifden, Derrybegresp.)</b> <br/>

2. Which year has the highest and lowest total pass count? <br/>

&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;&nbsp;<b>2015 and 2014 respectively</b> <br/> But, in this graph we are not considering the total tests performed by the test centres which shows the actual performance of the tests. For this we will plot another graph.

<br/> <br/>

```
```{r, echo=FALSE, fig.width=9, fig.height=4}
#scatter plot for centre total pass per year
ggplot(data=passfail, aes(x=fct_reorder(Centre, -Totalpass), y=Totalpass, color=Year, size
=Totalpass))+geom_point(alpha=0.5)+
theme(axis.text.x=element_text(size=9, angle=-
90, hjust=0, vjust=0.5), axis.ticks.x=element_blank(), panel.background=element_rect(fill="
white", colour="lightblue"), panel.grid.minor=element_line(size=0.5, linetype='solid', colour
="lightblue"))+labs(x="Test Centres", y="Total pass count")
```
```

```
####<b>Testperformanceforeachtestcentre</b>
```

The graph gives the overall idea of the test performance based on pass rate and the year.

As per the graph we can say that for year 2013, 2015, 2016, 2017 and 2018, the pass rate is higher than 55%. And the highest and lowest performance found in Kilkenny and Monaghan test centres respectively. <br/><br/>

```
```{r,echo=FALSE,fig.width=9,fig.height=4}
```

```
passfail$totPassPercentage<-
```

```
round((passfail$Totalpass/(passfail$Totalpass+passfail$totalFails))*100,digits=2)
```

```
passfail$totFailPercentage<-
```

```
round((passfail$totalFails/(passfail$Totalpass+passfail$totalFails))*100,digits=2)
```

```
#scatterplotforcentrepasspercentageperyear
```

```
ggplot(data=passfail,aes(x=fct_reorder(Centre,-totPassPercentage), y=totPassPercentage,color=Year,size=totPassPercentage)) +geom_point(alpha= 0.5)+
```

```
theme(axis.text.x=element_text(size=9,angle=-90,hjust=0,vjust=0.5),axis.ticks.x=element_blank(),panel.background=element_rect(fill="white",colour="lightblue"),panel.grid.minor=element_line(size=0.5,linetype='solid',colour="lightblue"))+labs(x="TestCentres",y="TotalPass%")#title="Test centrepass%peryear",  
```
```

```
####<b>Totalpasscountlimitsperyear</b>
```

The box plot shows the total pass count against each year. With this we can fetch the details on maximum and minimum pass counts per year, the median pass count and the outstanding pass count values which are shown as outliers (points) per year with the test centre name. <br>

```
```{r,echo=FALSE,fig.width=9,fig.height=4}
```

```
p<-
```

```
plot_ly(passfail,x=passfail$Year,y=passfail$Totalpass,color=~passfail$Year,type="box",text=paste("Centre=",passfail$Centre))%>%
```

```
layout(title="Yearly performance",yaxis=list(title='TotalPassCount'),xaxis=list(title='Year'))
```

```
p
```

```
```
```

Task\_Rmd :

```
---
```

```
title:"Project-Signs with Smart Connectivity for Better Road Safety"
```

```
output:html_document
```

```
---
```

```
```{rsetup,include=FALSE}knitr::opts_chunk$set(echo=TRUE)
```

```
#ggparcoord
```

```
#geom_polygon=>packcircles
```

```
```
```

```
##Libraries
```

```
```{rcars}suppressMessages(library(readxl))
suppressMessages(library(dplyr))suppressM
essages(library(tidyr))suppressMessages(lib
rary(ggplot2))suppressMessages(library(M
ASS))suppressMessages(library(GGally))su
ppressMessages(library(ggExtra))suppress
Messages(library(plotly))suppressMessages(
library(packcircles))
```
```

```
##PreparingTheDataForAnalysis
```

```
```{r}
df<-read_excel("mmAll.xlsx")
#d13<-read_excel("m_m2013.xlsx")
#d14<-read_excel("m_m2014.xlsx")
#d15<-read_excel("m_m2015.xlsx")
#d16<-read.csv("m_m2016.csv",header =T)
#d17<-read_excel("m_m2017.xlsx")
#d18<-read_excel("m_m2018.xlsx")
```

```
names(df)[9]<-
"VehicleandSafetyEquipment"names(df)[10] <-
"VehicleandSafetyEquipment%"names(df)[22]<-
"ChassisandBody%"names(df)[26]<-
"SuspensionTest%"names(df)[36]<-
"IncompleteTests%"
```

```
df$reportYear<-as.factor(df$reportYear)
```

```
```
```

```
##Whichpartfailedthemostperreport year?
```

```
```{r,echo=FALSE}
```

```
#####DATA#####
```

```
cols<-
```

```
c("VehicleMake","VehicleandSafetyEquipment","LightingandElectrical","SteeringandSuspensio
n","BrakingEquipment","WheelsandTyres","Engine,NoiseandExhaust","ChassisandBody",
"SideSlipTest","SuspensionTest","Lighttest","BrakeTest","Emmissions","OTHER")
```

```
m<-
```

```
df%>%dplyr::select(c("reportYear",cols))%>%group_by(reportYear)%>%summarise_if(is.num
eric,mean,na.rm=TRUE)
```

```
m<-gather(m,-reportYear,key=Part,value=Failures)
```

```
```
```

```

```{r}
#####PLOT#####
#ggplot(m,aes(x=factor(reportYear),y=,colour=supp,group=supp))+geom_line()

library(MASS)library(GGally
)
#Vectorcolorlibrary(RColorB
rewer)
palette<-brewer.pal(3,"Set1")
my_colors<-palette[as.numeric(m$reportYear)]

#names(x)<-c("2013","2014","2015","2016","2017","2018")
#p<-ggparcoord(m,
columns=2:13,groupColumn="reportYear")+geom_line(size=0.3)+theme_minimal()+geom_point(
)+
#xlab("CarPart")+ylab("Averagefailure rate")

ggplotly(ggplot(data=m,
mapping=aes(x=reportYear,y=Failures,colour=Part,group=1))+geom_point()+
geom_line()+xlab("ReportYear")+ylab("AverageNumberofFailures")

)
...

##EquipmentFailures-OverallStatistics

```{r}library(ggplot2
)
cols<-
c("VehicleandSafetyEquipment","LightingandElectrical","SteeringandSuspension","BrakingEqui
pment","WheelsandTyres","Engine,NoiseandExhaust","ChassisandBody","SideSlipTest","Suspen
sionTest","Lighttest","BrakeTest","Emmissions","OTHER")
a<-
df%>%dplyr::select(cols)b<-
colSums(a)
c<-data.frame(Part=names(b),Percent=unname(b)/sum(df$Total)*100)ggplot(c)+
geom_col(mapping=aes(x=reorder(Part,-
Percent),y=Percent,fill=Percent),col="black")+xlab("")+
ylab("FailurePercentage(%))+scale_fill_gradient(low
="orange",high="tan")+coord_flip()

...

###Thereisabuginthis code.Cananybodyfix it?
```{r,eval=FALSE}
####Thepolygongraphrepresentationoftheabovedata####1<-
data.frame(Part=names(b),Total=unname(b))
packing<-circleProgressiveLayout(l$Total,sizetype='area')

```

```

l$packing<-packingl
dat.gg<-circleLayoutVertices(packing,npoints=50)

p<-ggplot()+geom_polygon(data=dat.gg,aes(x,y,group=id,fill=as.factor(id)),colour="black",
alpha=0.6) +geom_text(data=l, aes(x, y,
size=Total,label=Part))+scale_size_continuous(range=c(1,4))+theme_void()+theme(legend.posit
ion="none")+coord_equal()

ggplotly(p,tooltip=c("Total", "Part"))

```{r}
z<-
df%>%group_by(VehicleMake)%>%summarise(tot=sum(Total),res=sum(PASS)/sum(Total))%
>%arrange(desc(tot))%>%print(Inf())
```

```{r}require(scales
)
q<-z%>%arrange(desc(tot))%>%slice(1:15)

ggplot(q)+
  geom_col(mapping=aes(x=reorder(VehicleMake,-
tot),y=tot,fill="green"))+xlab("VehicleMake")+ylab("NumberofVehicles")+coord_flip()+theme(
legend.position="none")+
  scale_y_continuous(labels=comma)

#ggMarginal(g,type="histogram",fill="transparent")
```

##PassPercentageversusNumberofVehiclesforagivenVehicleMake
```{r}require(scal
es)library(plotly)
p<-ggplot(q,aes(x=tot,
y=res*100))+geom_line(color="red")+
  geom_point(aes(text=VehicleMake))+xlab("NumberofVehicles")+ylab("PassPercentage(%))+
  scale_x_continuous(labels=comma)

ggplotly(p,tooltip="text")
```

```