

Data Science Project- Dashboard

Title of Dashboard- "The Covid-19"

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Original dataset before preprocessing:

	A	B	C	D	E	F	G	H	I	J	K	L
1	Country/Region	Continent	Population	TotalCases	NewCases	TotalDeaths	NewDeaths	TotalRecovered	NewRecovered	ActiveCases	Serious,Critical	Tot Cases/1M De
2	USA	North America	331198130	5032179		162804		2576668		2292707	18296	15194
3	Brazil	South America	212710692	2917562		98644		2047660		771258	8318	13716
4	India	Asia	1381344997	2025409		41638		1377384		606387	8944	1466
5	Russia	Europe	145940924	871894		14606		676357		180931	2300	5974
6	South Africa	Africa	59381566	538184		9604		387316		141264	539	9063
7	Mexico	North America	129066160	462690	6590	50517	819	308848	4140	103325	3987	3585
8	Peru	South America	33016319	455409		20424		310337		124648	1426	13793
9	Chile	South America	19132514	366671		9889		340168		16614	1358	19165
10	Colombia	South America	50936262	357710		11939		192355		153416	1493	7023
11	Spain	Europe	46756648	354530		28500					617	7582
12	Iran	Asia	84097623	320117		17976		277463		24678	4156	3806
13	UK	Europe	67922029	308134		46413					73	4537
14	Saudi Arabia	Asia	34865919	284226		3055		247089		34082	1915	8152
15	Pakistan	Asia	221295851	281863		6035		256058		19770	809	1274
16	Bangladesh	Asia	164851401	249651		3306		143824		102521		1514

	I	J	K	L	M	N	O	P	Q	R
1	vered	NewRecovered	ActiveCases	Serious,Critical	Tot Cases/1M pop	Deaths/1M pop	TotalTests	Tests/1M pop	WHO Region	
2	6668		2292707	18296	15194	492	63139605	190640	Americas	
3	17660		771258	8318	13716	464	13206188	62085	Americas	
4	7384		606387	8944	1466	30	22149351	16035	South-EastAsia	
5	6357		180931	2300	5974	100	29716907	203623	Europe	
6	7316		141264	539	9063	162	3149807	53044	Africa	
7	8848	4140	103325	3987	3585	391	1056915	8189	Americas	
8	0337		124648	1426	13793	619	2493429	75521	Americas	
9	0168		16614	1358	19165	517	1760615	92022	Americas	
10	12355		153416	1493	7023	234	1801835	35374	Americas	
11				617	7582	610	7064329	151087	Europe	
12	7463		24678	4156	3806	214	2612763	31068	EasternMediterranean	
13				73	4537	683	17515234	257873	Europe	
14	17089		34082	1915	8152	88	3635705	104277	EasternMediterranean	
15	16058		19770	809	1274	27	2058872	9304	EasternMediterranean	
16	13824		102521		1514	20	1225124	7432	South-EastAsia	

R code for Data PreProcessing:

```
dataset<-read.csv(choose.files())
```

```
library(dplyr)
```

```
library(WriteXLS)
```

```
library('writexl')
```

```
install.packages('data.table')
```

```
install.packages('WriteXLS')
```

```
head(dataset)
```

```
ncol(dataset)
```

```
nrow(dataset)
```

```
colnames(dataset)
```

```
data.frame(dataset)
```

```
# we are beginning with data preprocessing
```

```
# 1) checking null values in the dataset
```

```
sum(is.na(dataset))
```

```
sum(is.na(dataset$NewCases))
```

```
sum(is.na(dataset$NewDeaths))
```

```
sum(is.na(dataset$NewRecovered))
```

```
#df2<-data.frame(dataset[, c("Country.Region", "NewCases", "NewDeaths", "NewRecovered")])
```

```
# these are features with high no of na values therefore we will split them into different dataset
```

```
sum(is.na(dataset$Population)) # since there is only one row with na value we can drop the row
```

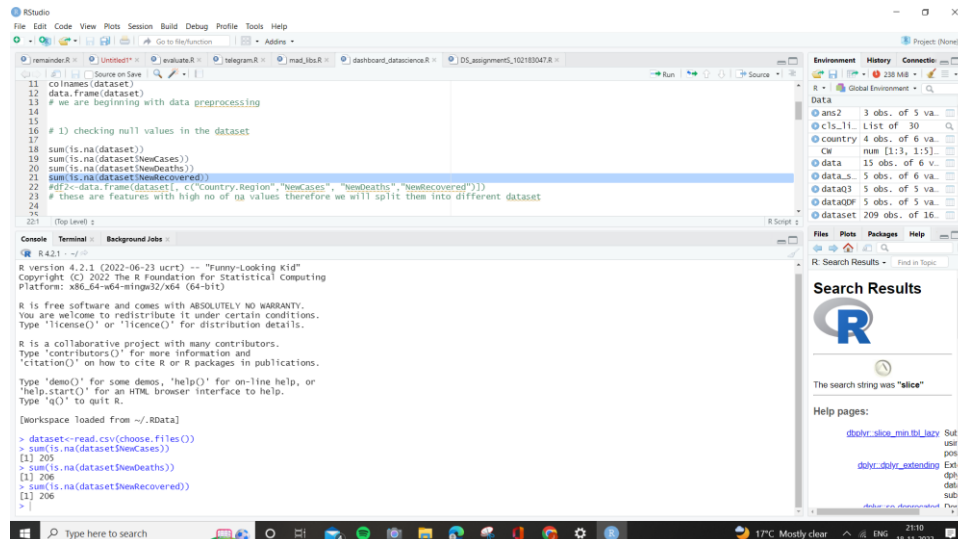
```
dataset[!(is.na(dataset$Population)==TRUE),]
```

```
sum(is.na(dataset$TotalRecovered)) #replacing the nas with mean value
```

```
dataset$TotalRecovered[is.na(dataset$TotalRecovered)] <- mean(dataset$TotalRecovered, na.rm  
=TRUE)
```

```
sum(is.na(dataset$Deaths.1M.pop)) #replacing the nas with median value
```

```
dataset$Deaths.1M.pop[is.na(dataset$Deaths.1M.pop)] <- mean(dataset$Deaths.1M.pop, na.rm =TRUE)
```



2) checking anomalous (inf) values in dataset

```
sum(is.infinite(dataset$Deaths.1M.pop))
```

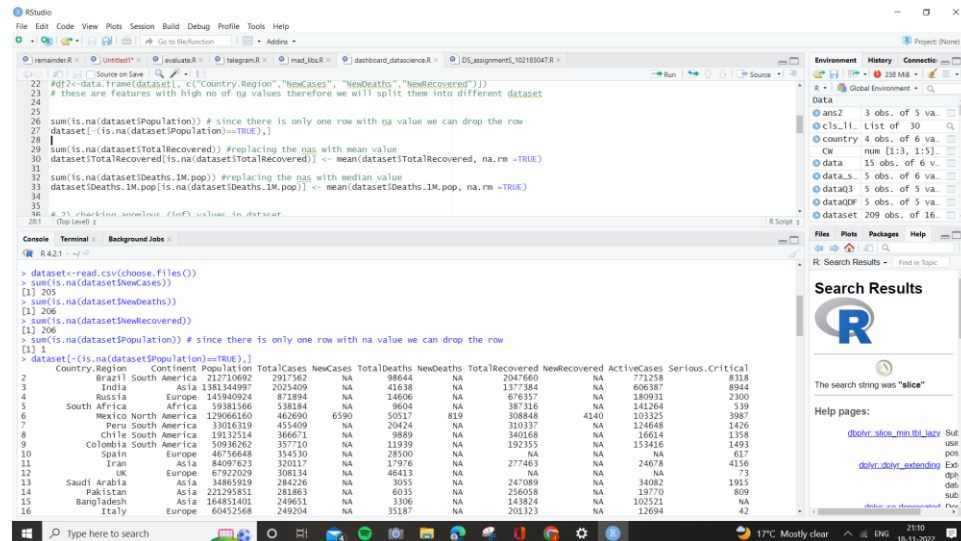
```
#replacing inf values with mean
```

```
dataset$Deaths.1M.pop[which(!is.finite(dataset$Deaths.1M.pop))]<-variable
```

```
#mean(dataset$Deaths...100.Recovered)
```

```
variable=mean(dataset$Deaths...100.Recovered) #saved 39.47385
```

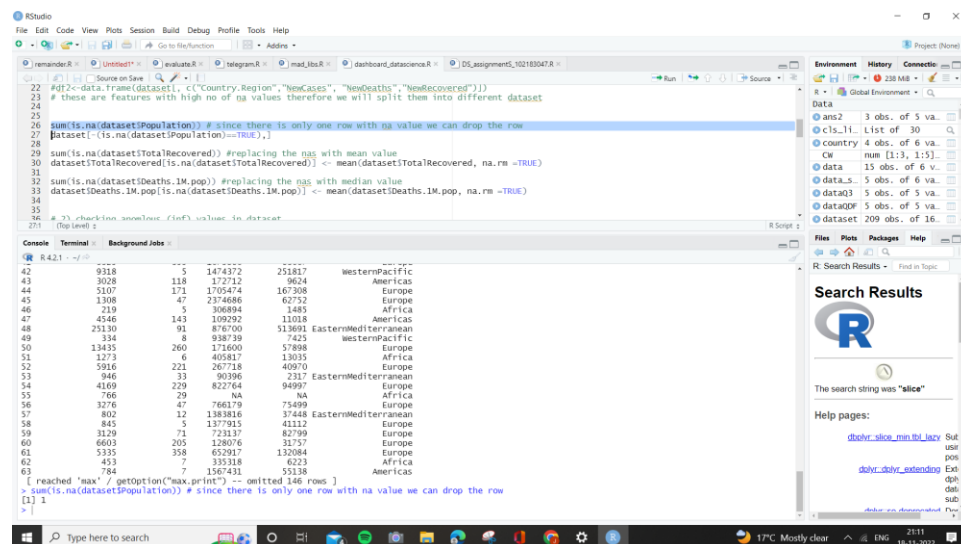
```
#variable=39.47385
```



The screenshot shows the RStudio environment with a script editor containing R code for data cleaning. The code includes comments and functions to handle NA values in a dataset. The console displays the output of the code, showing a dataset with columns: Country, Region, Continent, Population, TotalCases, NewCases, TotalDeaths, NewDeaths, TotalRecovered, NewRecovered, ActiveCases, Serious, and Critical. The data is organized by continent and then by country.

```
# RStudio
File Edit Code View Plots Session Build Debug Profile Tools Help
#> 2<-data.frame(dataset[, c("Country", "Region", "NewCases", "NewDeaths", "NewRecovered")])
# these are features with high no of na values therefore we will split them into different dataset
sum(is.na(dataset$Population)) # since there is only one row with na value we can drop the row
dataset[is.na(dataset$Population)==TRUE,]
sum(is.na(dataset$TotalRecovered)) #replacing the ngs with mean value
dataset$TotalRecovered[is.na(dataset$TotalRecovered)] <- mean(dataset$TotalRecovered, na.rm =TRUE)
sum(is.na(dataset$Deaths.1M.pop)) #replacing the ngs with median value
dataset$Deaths.1M.pop[is.na(dataset$Deaths.1M.pop)] <- median(dataset$Deaths.1M.pop, na.rm =TRUE)
#> 3) check for non/inf values in dataset
#> 4) since the name of continents are containing '/' we have to process them
```

Country	Region	Continent	Population	TotalCases	NewCases	TotalDeaths	NewDeaths	TotalRecovered	NewRecovered	ActiveCases	Serious	Critical
Brazil	South America	212710692	2917562	NA	98644	NA	2047660	NA	771258	8138		
India	Asia	1381344997	2025409	NA	41638	NA	1377384	NA	608387	8844		
Russia	Europe	145940924	871894	NA	14606	NA	676357	NA	180931	2300		
South Africa	Africa	581344	387316	NA	9604	NA	387316	NA	141264	539		
Mexico	North America	129066160	462690	6590	50517	819	308848	4140	103325	3987		
Peru	South America	33016319	455409	NA	20424	NA	310337	NA	124648	1426		
Chile	South America	19132324	366671	NA	9889	NA	340168	NA	16614	1358		
Colombia	South America	50936262	357710	NA	11939	NA	192355	NA	153416	1493		
Spain	Europe	46756448	354530	NA	28500	NA	NA	NA	NA	607		
Iran	Asia	84097623	320117	NA	17976	NA	277463	NA	24678	4156		
UK	Europe	67922029	308134	NA	46413	NA	NA	NA	NA	73		
Saudi Arabia	Asia	34865918	284226	NA	3033	NA	247089	NA	34082	1915		
Pakistan	Asia	221295851	281863	NA	6035	NA	256058	NA	19770	809		
Bangladesh	Asia	164851401	249051	NA	3306	NA	143824	NA	102321	NA		
Italy	Europe	60452568	249204	NA	35187	NA	201323	NA	12694	42		



The screenshot shows the RStudio environment with a script editor containing R code for data cleaning. The code includes comments and functions to handle NA values in a dataset. The console displays the output of the code, showing a dataset with columns: Country, Region, Continent, Population, TotalCases, NewCases, TotalDeaths, NewDeaths, TotalRecovered, NewRecovered, ActiveCases, Serious, and Critical. The data is organized by continent and then by country.

```
# RStudio
File Edit Code View Plots Session Build Debug Profile Tools Help
#> 2<-data.frame(dataset[, c("Country", "Region", "NewCases", "NewDeaths", "NewRecovered")])
# these are features with high no of na values therefore we will split them into different dataset
sum(is.na(dataset$Population)) # since there is only one row with na value we can drop the row
dataset[is.na(dataset$Population)==TRUE,]
sum(is.na(dataset$TotalRecovered)) #replacing the ngs with mean value
dataset$TotalRecovered[is.na(dataset$TotalRecovered)] <- mean(dataset$TotalRecovered, na.rm =TRUE)
sum(is.na(dataset$Deaths.1M.pop)) #replacing the ngs with median value
dataset$Deaths.1M.pop[is.na(dataset$Deaths.1M.pop)] <- median(dataset$Deaths.1M.pop, na.rm =TRUE)
#> 3) check for non/inf values in dataset
#> 4) since the name of continents are containing '/' we have to process them
```

Country	Region	Continent	Population	TotalCases	NewCases	TotalDeaths	NewDeaths	TotalRecovered	NewRecovered	ActiveCases	Serious	Critical
Brazil	South America	212710692	2917562	NA	98644	NA	2047660	NA	771258	8138		
India	Asia	1381344997	2025409	NA	41638	NA	1377384	NA	608387	8844		
Russia	Europe	145940924	871894	NA	14606	NA	676357	NA	180931	2300		
South Africa	Africa	581344	387316	NA	9604	NA	387316	NA	141264	539		
Mexico	North America	129066160	462690	6590	50517	819	308848	4140	103325	3987		
Peru	South America	33016319	455409	NA	20424	NA	310337	NA	124648	1426		
Chile	South America	19132324	366671	NA	9889	NA	340168	NA	16614	1358		
Colombia	South America	50936262	357710	NA	11939	NA	192355	NA	153416	1493		
Spain	Europe	46756448	354530	NA	28500	NA	NA	NA	NA	607		
Iran	Asia	84097623	320117	NA	17976	NA	277463	NA	24678	4156		
UK	Europe	67922029	308134	NA	46413	NA	NA	NA	NA	73		
Saudi Arabia	Asia	34865918	284226	NA	3033	NA	247089	NA	34082	1915		
Pakistan	Asia	221295851	281863	NA	6035	NA	256058	NA	19770	809		
Bangladesh	Asia	164851401	249051	NA	3306	NA	143824	NA	102321	NA		
Italy	Europe	60452568	249204	NA	35187	NA	201323	NA	12694	42		

```
# 4) since the name of continents are containing '/' we have to process them
```

```
num<-grep("/",dataset$Continent,value=FALSE)
```

```
for (i in num)
```

[illegible]

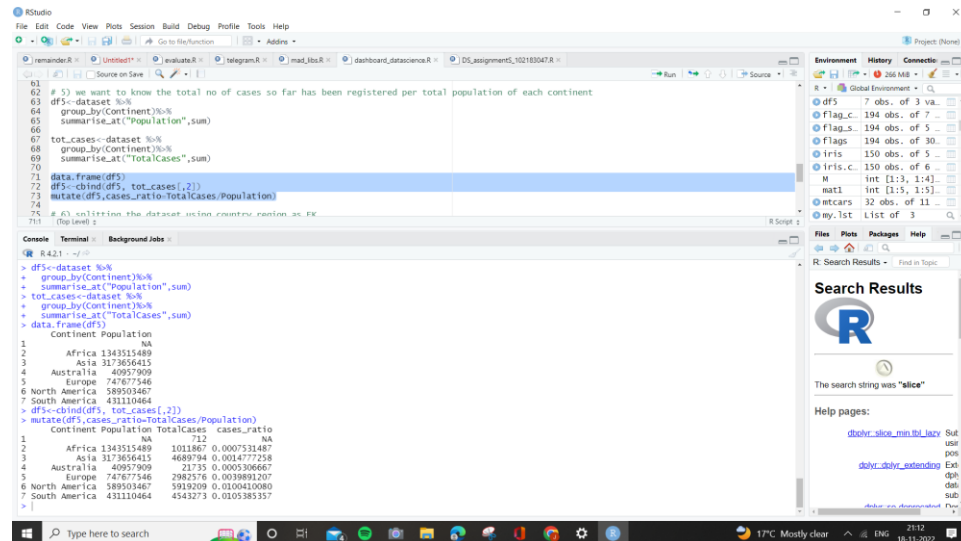
```
df5<-dataset %>%
  group_by(Continent)%>%
  summarise_at("Population",sum)
```

```
tot_cases<-dataset %>%
  group_by(Continent)%>%
  summarise_at("TotalCases",sum)
```

```
data.frame(df5)
```

```
df5<-cbind(df5, tot_cases[,2])
```

```
mutate(df5,cases_ratio=TotalCases/Population)
```



```
# 6) splitting the dataset using country.region as FK
```

```
df1<-data.frame(dataset[,  
c("Country.Region","TotalCases","TotalDeaths","TotalRecovered","ActiveCases")])
```

```
df1
```

```
df2<-data.frame(dataset[, c("Country.Region","NewCases","NewDeaths","NewRecovered")])
```

```
df3<-data.frame(dataset[,c("Country.Region","Tot.Cases.1M.pop","Deaths.1M.pop","Tests.1M.pop")])
```

```
df4<-df5
```

```
#7) preprocessing of df2
```

```
df2
```

```
sum(is.na(df2)) #dropping
```

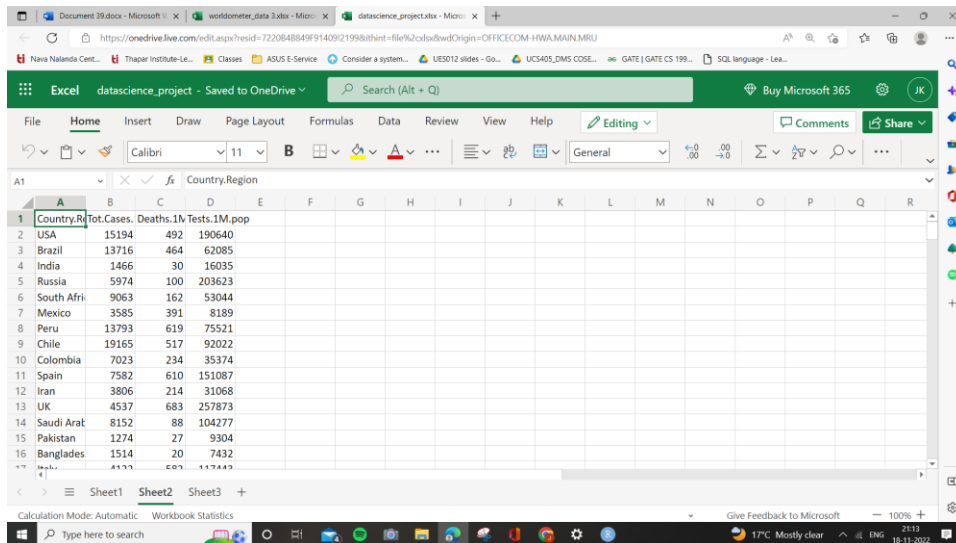
8) now are four datasets are preprocessed for tableau representation, so we will convert them into
xlsx file

```
write.csv(df4, file="C:\\Users\\jaskirat singh\\Downloads\\datascienceproj\\dataset4.csv", row.names =  
FALSE)
```

```
write.csv(df1, file="C:\\Users\\jaskirat singh\\Downloads\\datascienceproj\\dataset1.csv", row.names =  
FALSE)
```

```
write.csv(df3, file="C:\\Users\\jaskirat singh\\Downloads\\datascienceproj\\dataset3.csv", row.names =  
FALSE)
```

The final preprocessed data is:



The screenshot shows a Microsoft Excel spreadsheet titled 'datascience_project - Saved to OneDrive'. The data is organized in a table with the following columns: Country, Region, Tot.Cases, Deaths, 1N Tests, 1M, and pop. The rows list various countries including USA, Brazil, India, Russia, South Africa, Mexico, Peru, Chile, Colombia, Spain, Iran, UK, Saudi Arabia, Pakistan, and Bangladesh, along with their respective statistics.

Country	Region	Tot.Cases	Deaths	1N Tests	1M	pop
USA		15194	492	190640		
Brazil		13716	464	62085		
India		1466	30	16035		
Russia		5974	100	203623		
South Africa		9063	162	53044		
Mexico		3585	391	8189		
Peru		13793	619	75521		
Chile		19165	517	92022		
Colombia		7023	234	35374		
Spain		7582	610	151087		
Iran		3806	214	31068		
UK		4537	683	257873		
Saudi Arabia		8152	88	104277		
Pakistan		1274	27	9304		
Bangladesh		1514	20	7432		

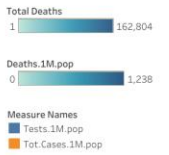
The dashboard:

Covid-19 Data Visualization During Year(2020-22)

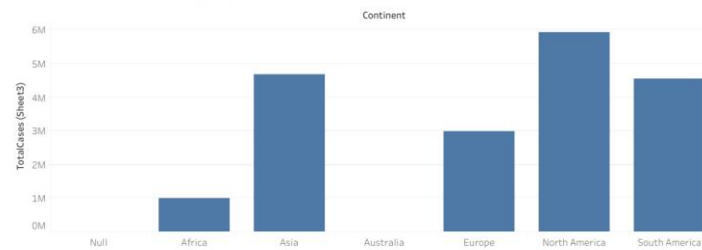
Country-wise total no. of deaths,cases and recoveries



Country wise total deaths, tests conducted per million



Continent-wise covid cases along with population



Country-wise total no of tests conducted through out along with the total cases cases observed

