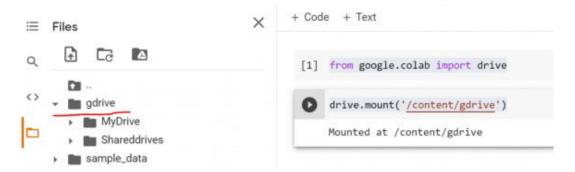
Python

1. How To Get Data From Google Drive Into Colab

The Google Colab is becoming more and more popular in Data Science community. Working with Colab Jupyter notebooks, you are able to mount your Google Drive, so that to be able to get data from it directly. Let's see how we can do it. You should open you notebook and to type the following commands:



It will send you a link to authorize your Google Drive by entering a token and you are done. Then you will see a folder called gdrive under your files:



Now, if you want to get the exact path of the file, you can just right click and choose, <code>Copy path</code> which usually starts with <code>/content/gdrive/MyDrive/</code>

2. How to Impute Missing Values In Pandas

We will provide an example of how you can impute missing values in Pandas following the rules below:

- If the variable is numeric then impute the missing values with the mean
- If the variable is object dtype then impute the missing values with the mode

```
id A
           В
               C
                      D
0
   0
     10.0
            A
               NaN
                    NaN
                BB
                    20.0
1
   1
      9.0
            В
      8.0
                    18.0
2
   2
            A
                CC
3
   3
     7.0
            A
                BB
                   22.0
          NaN
                BB
                    18.0
4
     NaN
     NaN
            В
                CC
                    17.0
5
   5
6
   6
     20.0
            A
                AA
                    19.0
   7 15.0
7
            В
                BB
                    NaN
     12.0
               NaN
                    17.0
8
          NaN
            A AA 23.0
9
   9 NaN
```

```
for i in df.columns:
    if (df[i].dtype=='object'):
        df[i].fillna(df[i].mode()[0], inplace=True)
    elif (df[i].dtype=='float64' or df[i].dtype=='int64'):
        df[i].fillna(df[i].mean(), inplace=True)
```

	id	A	В	C	D
0	0	10.000000	Α	ВВ	19.25
1	1	9.000000	В	ВВ	20.00
2	2	8.000000	Α	CC	18.00
3	3	7.000000	Α	ВВ	22.00
4	4	11.571429	Α	ВВ	18.00
5	5	11.571429	В	CC	17.00
6	6	20.000000	A	AA	19.00
7	7	15.000000	В	ВВ	19.25
8	8	12.000000	Α	ВВ	17.00
9	9	11.571429	Α	AA	23.00

3. How To Read and Write Gzip/Zip Files with Pandas

We can write pandas data frames as gzip and zip files. For example:

```
#lets use this sample dataframe
df=pd.DataFrame({'A':[1,2,3,4]})
```

Save it as gzip or zip

```
# save it as gzip
df.to_csv("dataframe.csv.gz", index=False, compression="gzip")
# save it as gzip
df.to_csv("dataframe.csv.zip", index=False, compression="zip")
```

Read gzip/zip files Pandas

Pandas can also read gzip/zip files that contain a CSV.

```
pd.read_csv('dataframe.csv.gz')

A
0 1
1 2
2 3
```

4. How To Read And Write CSV Files Without Headers With Pandas

Read Without Headers

Assume that you have the following CSV file which is without column names. Let's see how we can read it with pandas. Assume that my file.csv is:

```
George, 25
Maria, 30
John, 18
```

We can read the file as follows:

```
import pandas as pd
df = pd.read_csv("my_file.csv", header=None)
df
```

	0	1
0	George	25
1	Maria	30
2	John	18

In case we want to give specific names to our column names:

```
df1 = pd.read_csv("my_file.csv", header=None, names=['colA', 'colB'])
df1
```

	colA	colB
0	George	25
1	Maria	30
2	John	18

Write Without Headers

Now, let's say that we want to write this dataframe to a csv file but without headers.

```
df.to csv('filename.csv', header=False, index=False)
```

```
!cat filename.csv

George, 25
Maria, 30
John, 18
```

5.Cumulative Count Distinct Values in Python

Sometimes there is a need to do a rolling count of the distinct values of a list/vector. In other words, we want to add up only any **new** element that appears in our list/vector.

```
import pandas as pd

df = pd.DataFrame({'mylist':["e", "a","b","a","b","c", "d", "e"]})

df['CumDistinct'] = (~df.mylist.duplicated()).cumsum()

df

# or by using apply
# df['CumDistinct'] = df.mylist.apply(lambda x: (~pd.Series(x).duplicated()).cumsum())
```

Output:

	mylist	CumDistinct
0	е	1
1	a	2
2	a	2
3	b	3
4	a	3
5	b	3
6	С	4
7	d	5
8	е	5

6.Count The Consecutive Events In Python

We are going to represent a straightforward practical example of how to count consecutive events:

```
import pandas as pd
df = pd.DataFrame({'Score':['win', 'loss', 'loss', 'win', 'win', 'win', 'win', 'win', 'loss', 'loss']})
```

With the following "hack" we are going to get the column of the "Streak":

```
df['Streak'] = df['Score'].groupby((df['Score'] !=
df['Score'].shift()).cumsum()).cumcount() + 1
df
```

Output:

	Score	Streak
0	win	1
1	loss	1
2	loss	2
3	loss	3
4	win	1
5	win	2
6	win	3
7	win	4
8	win	5
9	loss	1
10	win	1
11	loss	1

7. Cumulative Count Distinct Values in R

Sometimes there is a need to do a rolling count of the distinct values of a list/vector. In other words, we want to add up only any **new** element that appears in our list/vector.

```
# assume that this is our vector
x=c("e", "a", "a", "b", "a", "b", "c", "d", "e")
# we apply the "cumsum(!duplicated(x))" command
data.frame(Vector=x,
CumDistinct=cumsum(!duplicated(x)))
```

Output:

	Vector	CumDistinct
1	е	1
2	а	2
3	а	2
4	b	3
5	а	3
6	b	3
7	С	4
8	d	5
9	е	5

A tibble: 1,000,000 x 4

Rolls Outcome Streak EndOfStreak

8.Count The Consecutive Events In R

Assume that there is a Roulette Wheel which returns Red (50%) and Black (50%). We are going to simulate N=1,000,000 Rolls and keep track of the streaks of Red and Black respectively. The R function which makes our life easier is the **rle** but if we want to track the running streak, then we need also to use the **seq** function. We add also another column, called **EndOfStreak** which indicates if the Streak has ended or not.

```
library(tidyverse)

# number of simulations
n<-1000000

# set a random seed for reproducibility
set.seed(5)

# create the data frame
df<-tibble(Rolls=seq(1:n), Outcome=sample(c("Red", "Black"),n,replace = TRUE,
prob = c(0.5,0.5)))%>%
    mutate(Streak=sequence(rle(Outcome)$lengths), EndOfStreak=ifelse(lead(Outcome)==Outcome, "No", "Yes"))

df%>%print(n=20)

Output:
```

```
1
     1 Black
                 1 Yes
2
     2 Red
                 1 No
3
     3 Red
                 2 Yes
4
     4 Black
                 1 No
5
     5 Black
                 2 Yes
                 1 No
6
     6 Red
7
     7 Red
                 2 No
                 3 No
8
    8 Red
9
    9 Red
                 4 Yes
10
    10 Black
                 1 No
11
    11 Black
                 2 No
12
    12 Black
                 3 No
    13 Black
                 4 Yes
13
14
    14 Red
                 1 Yes
                 1 No
1.5
    15 Black
16
    16 Black
                 2 No
17
   17 Black
                 3 Yes
18
    18 Red
                  1 No
19
    19 Red
                 2 No
20
    20 Red
                  3 No
```

df

9. Replace Missing Values With The Mean Value

Assume that you are dealing with a Data Frame with missing values, and you want to replace them with the mean value of the corresponding column. Let's consider the following Data Frame:

	id	ColumnA	ColumnB	ColumnC	ColumnD
1	1	10	Α		NA
2	2	9	В	BB	20
3	3	8	Α	CC	18
4	4	7	Α	BB	22
5	5	NA		BB	18
6	6	NA	В	CC	17
7	7	20	Α	AA	19
8	8	15	В	ВВ	NA
9	9	12			17
10	10	NA	Α	AA	23

Clearly, we want to consider only the numeric variables. Let's see how we can do it in one-line code.

```
 \begin{split} &\text{df[sapply(df, is.numeric)]} <- \text{lapply(df[sapply(df, is.numeric)], function(x)} \\ &\text{ifelse(is.na(x), mean(x, na.rm = TRUE), x))} \\ &\text{df} \end{split}
```

	id	ColumnA	ColumnB	ColumnC	ColumnD
1	1	10.00000	Α		19.25
2	2	9.00000	В	BB	20.00
3	3	8.00000	Α	CC	18.00
4	4	7.00000	Α	BB	22.00
5	5	11.57143		BB	18.00
6	6	11.57143	В	CC	17.00
7	7	20.00000	А	AA	19.00
8	8	15.00000	В	BB	19.25
9	9	12.00000			17.00
10	10	11.57143	Α	AA	23.00

10.Replace Categorical Variables With Mode

In Data Science project, it is common to replace the missing values of the categorical variables with the mode. Let's see the following example:

	\$ 5120				
	id	ColumnA	ColumnB	ColumnC	ColumnD
1	1	10	Α		NA
2	2	9	В	BB	20
3	3	8	Α	CC	18
4	4	7	Α	BB	22
5	5	NA		BB	18
6	6	NA	В	CC	17
7	7	20	Α	AA	19
8	8	15	В	BB	NA
9	9	12			17
10	10	NA	Α	AA	23

Note that the ColumnB and ColumnC are Character columns. Note also that there is no mode function in R. So let's build it:

```
getmode <- function(v) {
   v=v[nchar(as.character(v))>0]
   uniqv <- unique(v)
   uniqv[which.max(tabulate(match(v, uniqv)))]
}</pre>
```

Now let's replace all the empty strings of the Character variables with their corresponding column mode. Finally, we should convert the character variables to factors.

```
df[sapply(df, is.character)] <- lapply(df[sapply(df, is.character)],
function(x) ifelse(x=="", getmode(x), x))
df[sapply(df, is.character)] <- lapply(df[sapply(df, is.character)],
as.factor)
df</pre>
```

	id	ColumnA	ColumnB	ColumnC	ColumnD
1	1	10	Α	BB	NA
2	2	9	В	BB	20
3	3	8	Α	CC	18
4	4	7	Α	BB	22
5	5	NA	Α	BB	18
6	6	NA	В	CC	17
7	7	20	Α	AA	19
8	8	15	В	BB	NA
9	9	12	Α	BB	17
10	10	NA	Α	AA	23

Data		
o df		10 obs. of 5 variables
\$	id :	int 1 2 3 4 5 6 7 8 9 10
\$	ColumnA:	num 10 9 8 7 NA NA 20 15 12 NA
\$	ColumnB:	Factor w/ 2 levels "A", "B": 1 2 1 1 1 2 1 2 1 1
\$	columnc:	Factor w/ 3 levels "AA", "BB", "CC": 2 2 3 2 2 3 1 2 2 1
\$	ColumnD:	num NA 20 18 22 18 17 19 NA 17 23

As we can see, we replaced the empty strings with the corresponding mode.