



**American International University-
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Course Title: INTRODUCTION TO DATA SCIENCE

Project Title: Data Preparation and Exploration
On Titanic Dataset.

Course: INTRODUCTION TO DATA SCIENCE

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Project overview:

Titanic dataset, which is widely used in the field of data science and machine learning. This dataset consists of information about the passengers who were on board the titanic during its first voyage in 1912, which tragically ended in the ship sinking. The dataset includes different characteristics of the passengers, such as their age, gender, class, fare, and whether they survived or not. The purpose of this project is to analyze the titanic dataset, extract valuable insights from the data, and potentially create predictive models to understand the factors that influenced passenger survival during the titanic disaster.

Dataset description:

The titanic dataset consists of multiple variables (columns) that describe different aspects of the passengers. Here is a brief description of the key variables in the dataset:

Survived: indicates whether a passenger survived or not (0 = no, 1 = yes)

Pclass: passenger class (1 = first class, 2 = second class, 3 = third class)

Name: passenger's name

Sex: passenger's gender (male or female)

Age: passenger's age in years

Sibsp: number of siblings/spouses aboard the titanic

Parch: number of parents/children aboard the titanic

Ticket: ticket number

Fare: fare paid for the ticket

Cabin: cabin number

Embarked: port of embarkation (c = cherbourg, q = queenstown, s = southampton)

Project solution design:

loading and preprocessing the enormous dataset, doing data analysis and visualization, engineering pertinent features, training predictive models, assessing their performance, interpreting the findings, and documenting the findings are the initial steps. We examine the dataset, deal with the missing values, and modify the data types. Techniques for analysis and visualization are used, as well as feature engineering. The outcomes of the training and testing of machine learning models are analyzed and summarized in a project report. This methodical technique guarantees a thorough review of the data and offers useful insights.

Data pre-processing:

Importing the dataset: Initially, the `read.csv` function in R is used to read a dataset into the computer from the "dataset.csv" csv file. The generated data frame is then associated with the dataset variable. The dataset variable's contents are then shown in the console by using the `print` function. This code's output shows every row and column in the dataset as well as all of its data.

```
1 dataset<-read.csv("Dataset_midterm_Section(B).csv")
2 print(dataset)
```

The output provides a way to verify that the data has been read in correctly and allows for a quick visual inspection of the dataset. The output also provides information about the structure of the data, such as the number of columns and rows and the type of data in each column. Overall, this code and its output are a necessary first step in any data analysis project in R.

```
> dataset<-read.csv("Dataset_midterm_Section(B).csv")
> print(dataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0	24.0	0	0	7.7958	S	Third	mannn	TRUE	0
2	0	17.0	0	0	8.6625	S	Third	man	TRUE	0
3	1	21.0	0	0	7.7500	Q	Third	womann	TRUE	0
4	1	NA	0	0	7.6292	Q	Third	woman	TRUE	0
5	1	37.0	0	0	9.5875	S	Third	womannn	TRUE	0
6	NA	16.0	0	0	86.5000	S	First	woman	TRUE	1
7	0	18.0	1	0	108.9000	C	First	mannn	FALSE	0
8	1	33.0	0	2	26.0000	S	Second	woman	FALSE	1
9	0	NA	0	0	26.5500	S	First	man	TRUE	1
10	0	28.0	0	0	22.5250	S	Third	man	TRUE	0
11	0	26.0	0	0	56.4958	S	Third	man	TRUE	1
12	0	29.0	0	0	7.7500	Q	Third	man	TRUE	1
13	0	NA	0	0	8.0500	S	Third	man	TRUE	0
14	0	36.0	0	0	26.2875	S		man	TRUE	1
15	1	54.0	1	0	59.4000	C	First	woman	FALSE	1
16	0	24.0	0	0	7.4958	S	Third	man	TRUE	0
17	0	47.0	0	0	34.0208	S	First	man	TRUE	0
18	1	34.0	0	0	10.5000	S	Second	woman	TRUE	1
19	0	NA	0	0	24.1500	Q	Third	man	TRUE	0
20	1	36.0	1	0	26.0000	S	Second	woman	FALSE	1
21	0	32.0	0	0	7.8958	S	Third	man	TRUE	0
22	1	30.0	0	0	93.5000	S	First	woman	TRUE	1
23	NA	22.0	0	0	7.8958	S	Third	man	TRUE	0
24	0	NA	0	0	7.2250	C	Third	man	TRUE	0
25	1	44.0	0	1	57.9792	C	First	woman	FALSE	1
26	0	NA	0	0	7.2292	C	Third	man	TRUE	0
27	0	40.5	0	0	7.7500	Q	Third	man	TRUE	0
28	1	50.0	0	0	10.5000	S	Second	woman	TRUE	1
29	0	NA	0	0	221.7792	S	First	man	TRUE	0
30	0	39.0	0	0	7.9250	S	Third	man	TRUE	0
31	0	23.0	2	1	11.5000	S	Second	man	FALSE	0
32	NA	2.0	1	1	26.0000	S	Second	child	FALSE	1
33	0	NA	0	0	7.2292	C	Third	man	TRUE	0
34	0	17.0	1	1	7.2292	C	Third	man	FALSE	0
35	1	NA	0	2	22.3583	C		woman	FALSE	1
36	1	30.0	0	0	8.6625	S	Third	woman	TRUE	0
37	1	7.0	0	2	26.2500	S	Second	child	FALSE	1
38	0	45.0	0	0	26.5500	S	First	man	TRUE	0
39	1	30.0	0	0	106.4250	C	First	woman	TRUE	1
40	0	NA	0	0	14.5000	S	Third	man	TRUE	0

```

41      1 22.0      0      2 49.5000      C First woman FALSE      1
42      1 36.0      0      2 71.0000      S First woman FALSE      1
43      NA 9.0      4      2 31.2750      S Third child FALSE      0
44      1 11.0      4      2 31.2750      S Third child FALSE      0
45      0 32.0      1      0 26.0000      S Second man FALSE      1
46      0 50.0      1      0 106.4250     C First man FALSE      0
47      0 64.0      0      0 26.0000      S First man TRUE      0
48      1 19.0      1      0 26.0000      S Second woman FALSE      1
49      0 NA      0      0 13.8625     C Second man TRUE      1
50      0 33.0      1      1 20.5250     S Third man FALSE      0
51      0 8.0      1      1 36.7500     S Second child FALSE      1
52      0 17.0      0      2 110.8833    C First man FALSE      1
53      0 27.0      0      0 26.0000      S Second man TRUE      0
54      0 NA      0      0 7.8292      Q Third man TRUE      0
55      0 22.0      0      0 7.2250      C Third man TRUE      1
56      1 22.0      0      0 7.7750      S Third woman TRUE      1
57      0 62.0      0      0 26.5500     S First man TRUE      0
58      1 48.0      1      0 39.6000     C First woman FALSE      1
59      0 NA      0      0 227.5250    C First man TRUE      0
60      1 39.0      1      1 79.6500     S First woman FALSE      1
61      1 36.0      1      0 17.4000     S Third woman FALSE      1
62      0 NA      0      0 7.7500      Q Third man TRUE      0
63      0 40.0      0      0 7.8958      S Third man TRUE      0
64      0 28.0      0      0 13.5000     S man TRUE      0
65      0 NA      0      0 8.0500      S Third man TRUE      0
66      1 NA      0      0 8.0500      S Third woman TRUE      0
67      0 24.0      2      0 24.1500     S Third man FALSE      0
68      0 19.0      0      0 7.8958      S Third man TRUE      0
69      1 29.0      0      4 21.0750     S Third woman FALSE      0
70      0 NA      0      0 7.2292      C Third man TRUE      0
71      0 32.0      0      0 7.8542     S Third man TRUE      1
72      0 62.0      0      0 10.5000     S Second man TRUE      1
73      1 53.0      2      0 51.4792     S First woman FALSE      1
74      0 36.0      0      0 26.3875     S First man TRUE      1
75      1 NA      0      0 7.7500      Q Third woman TRUE      1
76      0 16.0      0      0 8.0500      S Third man TRUE      0
77      0 19.0      0      0 14.5000     S Third man TRUE      0
78      1 34.0      0      0 13.0000     S Second woman TRUE      1
79      1 39.0      1      0 55.9000     S First woman FALSE      1
80      1 NA      1      0 14.4583     C Third woman FALSE      0
81      0 32.0      0      0 7.9250      S Third man TRUE      1
82      1 25.0      1      1 30.0000     S Second woman FALSE      1
83      1 39.0      1      1 110.8833    C First woman FALSE      1
84      0 54.0      0      0 26.0000     S Second man TRUE      0
85      0 36.0      0      0 40.1250     C First man TRUE      0
86      0 NA      0      0 8.7125      C Third man TRUE      0
87      1 18.0      0      2 79.6500     S First woman FALSE      1
88      0 47.0      0      0 15.0000     S Second man TRUE      0
89      0 60.0      1      1 79.2000     C First man FALSE      1
90      0 22.0      0      0 8.0500      S Third man TRUE      0
91      0 NA      0      0 8.0500      S Third man TRUE      0
92      0 35.0      0      0 7.1250      S Third man TRUE      0
93      1 52.0      1      0 78.2667     C First woman FALSE      1
94      0 47.0      0      0 7.2500      S Third man TRUE      0
95      1 NA      0      2 7.7500      Q Third woman FALSE      0
96      0 37.0      1      0 26.0000     S Second man FALSE      0
97      0 36.0      1      1 24.1500     S Third man FALSE      0
98      1 NA      0      0 33.0000     S Second woman TRUE      1
99      0 49.0      0      0 0.0000      S man TRUE      0
100     0 NA      0      0 7.2250      C Third man TRUE      0
[ reached 'max' / getOption("max.print") -- omitted 201 rows ]

```

Then creating a new variable called `xdataset` which is a copy of the dataset variable created in the previous step

```

4 xdataset <- dataset
5 print(xdataset)

```

The creation of `xdataset` allows for a backup copy of the original dataset to be kept while any necessary changes are made to the data. This ensures that the original data is not lost or overwritten accidentally. The output of this code is useful for verifying that the copy was made

correctly and allows for a quick visual comparison of the two datasets to ensure they are identical.

Identifying missing values:

To identify missing value “is.na” function to check for missing values in the xdataset variable. The is.na function returns a logical value of true for each missing value in the data and false for all other values.

```
> is.na(xdataset)
      Gender  age sibsp parch  fare embarked class  who alone survived
[1,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[2,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[3,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[4,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[5,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[6,]  TRUE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[7,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[8,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[9,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[10,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[11,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[12,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[13,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[14,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[15,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[16,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[17,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[18,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[19,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[20,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[21,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[22,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[23,]  TRUE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[24,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[25,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[26,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[27,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[28,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[29,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[30,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[31,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[32,]  TRUE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[33,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[34,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[35,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[36,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[37,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[38,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[39,] FALSE FALSE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
[40,] FALSE  TRUE FALSE FALSE FALSE  FALSE  FALSE FALSE FALSE  FALSE
```

```

[40,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[41,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[42,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[43,] TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[44,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[45,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[46,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[47,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[48,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[49,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[50,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[51,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[52,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[53,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[54,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[55,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[56,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[57,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[58,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[59,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[60,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[61,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[62,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[63,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[64,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[65,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[66,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[67,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[68,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[69,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[70,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[70,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[71,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[72,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[73,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[74,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[75,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[76,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[77,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[78,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[79,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[80,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[81,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[82,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[83,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[84,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[85,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[86,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[87,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[88,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[89,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[90,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[91,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[92,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[93,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[94,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[95,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[96,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[97,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[98,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[99,] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[100,] FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
[ reached getOption("max.print") -- omitted 201 rows ]

```

The output of this code is a matrix that has the same dimensions as the xdataset variable. The matrix contains true values in the cells where the corresponding value in xdataset is missing and false values in all other cells. This output is useful for identifying any missing values in the dataset and determining the extent to which missing values may affect subsequent analysis.


```
> sum(is.na(xdataset))
[1] 65
> |
```

The sum() function is applied to the logical matrix from the previous step to count the total number of missing values in the dataset. Which is 65.

Handling missing values:

performing a procedure to deal with missing values in a dataset. The missing values are found in specific columns, namely **gender**, **age**, **sibsp**, **parch**, **fare**, **alone**, and **survived**. To address these missing values, the code utilizes the mean of each respective column as a replacement. This is achieved through the mean function and logical indexing. The missing values are then replaced with the mean values using the assignment operator, effectively filling in the gaps in the dataset.

```
> xdataset$Gender[is.na(xdataset$Gender)]<-mean(xdataset$Gender,na.rm=TRUE)
> xdataset$age[is.na(xdataset$age)]<-mean(xdataset$age,na.rm=TRUE)
> xdataset$sibsp[is.na(xdataset$sibsp)]<-mean(xdataset$sibsp,na.rm=TRUE)
> xdataset$parch[is.na(xdataset$parch)]<-mean(xdataset$parch,na.rm=TRUE)
> xdataset$fare[is.na(xdataset$fare)]<-mean(xdataset$fare,na.rm=TRUE)
> xdataset$alone[is.na(xdataset$alone)]<-mean(xdataset$alone,na.rm=TRUE)
> xdataset$survived[is.na(xdataset$survived)]<-mean(xdataset$survived,na.rm=TRUE)
> print(xdataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0.0000000	24.00000	0	0	7.7958	S	Third	mann	1	0
2	0.0000000	17.00000	0	0	8.6625	S	Third	man	1	0
3	1.0000000	21.00000	0	0	7.7500	Q	Third	womann	1	0
4	1.0000000	34.03508	0	0	7.6292	Q	Third	woman	1	0
5	1.0000000	37.00000	0	0	9.5875	S	Third	womannn	1	0
6	0.3198653	16.00000	0	0	86.5000	S	First	woman	1	1
7	0.0000000	18.00000	1	0	108.9000	C	First	mannnn	0	0
8	1.0000000	33.00000	0	2	26.0000	S	Second	woman	0	1
9	0.0000000	34.03508	0	0	26.5500	S	First	man	1	1
10	0.0000000	28.00000	0	0	22.5250	S	Third	man	1	0
11	0.0000000	26.00000	0	0	56.4958	S	Third	man	1	1
12	0.0000000	29.00000	0	0	7.7500	Q	Third	man	1	1
13	0.0000000	34.03508	0	0	8.0500	S	Third	man	1	0
14	0.0000000	36.00000	0	0	26.2875	S		man	1	1
15	1.0000000	54.00000	1	0	59.4000	C	First	woman	0	1
16	0.0000000	24.00000	0	0	7.4958	S	Third	man	1	0
17	0.0000000	47.00000	0	0	34.0208	S	First	man	1	0
18	1.0000000	34.00000	0	0	10.5000	S	Second	woman	1	1
19	0.0000000	34.03508	0	0	24.1500	Q	Third	man	1	0
20	1.0000000	36.00000	1	0	26.0000	S	Second	woman	0	1
21	0.0000000	32.00000	0	0	7.8958	S	Third	man	1	0
22	1.0000000	30.00000	0	0	93.5000	S	First	woman	1	1
23	0.3198653	22.00000	0	0	7.8958	S	Third	man	1	0
24	0.0000000	34.03508	0	0	7.2250	C	Third	man	1	0
25	1.0000000	44.00000	0	1	57.9792	C	First	woman	0	1
26	0.0000000	34.03508	0	0	7.2292	C	Third	man	1	0
27	0.0000000	40.50000	0	0	7.7500	Q	Third	man	1	0
28	1.0000000	50.00000	0	0	10.5000	S	Second	woman	1	1
29	0.0000000	34.03508	0	0	221.7792	S	First	man	1	0
30	0.0000000	39.00000	0	0	7.9250	S	Third	man	1	0

```

70 0.0000000 34.03508 0 0 7.2292 C Third man 1 0
71 0.0000000 32.00000 0 0 7.8542 S Third man 1 1
72 0.0000000 62.00000 0 0 10.5000 S Second man 1 1
73 1.0000000 53.00000 2 0 51.4792 S First woman 0 1
74 0.0000000 36.00000 0 0 26.3875 S First man 1 1
75 1.0000000 34.03508 0 0 7.7500 Q Third woman 1 1
76 0.0000000 16.00000 0 0 8.0500 S Third man 1 0
77 0.0000000 19.00000 0 0 14.5000 S Third man 1 0
78 1.0000000 34.00000 0 0 13.0000 S Second woman 1 1
79 1.0000000 39.00000 1 0 55.9000 S First woman 0 1
80 1.0000000 34.03508 1 0 14.4583 C Third woman 0 0
81 0.0000000 32.00000 0 0 7.9250 S Third man 1 1
82 1.0000000 25.00000 1 1 30.0000 S Second woman 0 1
83 1.0000000 39.00000 1 1 110.8833 C First woman 0 1
84 0.0000000 54.00000 0 0 26.0000 S Second man 1 0
85 0.0000000 36.00000 0 0 40.1250 C First man 1 0
86 0.0000000 34.03508 0 0 8.7125 C Third man 1 0
87 1.0000000 18.00000 0 2 79.6500 S First woman 0 1
88 0.0000000 47.00000 0 0 15.0000 S Second man 1 0
89 0.0000000 60.00000 1 1 79.2000 C First man 0 1
90 0.0000000 22.00000 0 0 8.0500 S Third man 1 0
91 0.0000000 34.03508 0 0 8.0500 S Third man 1 0
92 0.0000000 35.00000 0 0 7.1250 S Third man 1 0
93 1.0000000 52.00000 1 0 78.2667 C First woman 0 1
94 0.0000000 47.00000 0 0 7.2500 S Third man 1 0
95 1.0000000 34.03508 0 2 7.7500 Q Third woman 0 0
96 0.0000000 37.00000 1 0 26.0000 S Second man 0 0
97 0.0000000 36.00000 1 1 24.1500 S Third man 0 0
98 1.0000000 34.03508 0 0 33.0000 S Second woman 1 1
99 0.0000000 49.00000 0 0 0.0000 S man 1 0
100 0.0000000 34.03508 0 0 7.2250 C Third man 1 0
[ reached 'max' / getOption("max.print") -- omitted 201 rows ]
> |
31 0.0000000 23.00000 2 1 11.5000 S Second man 0 0
32 0.3198653 2.00000 1 1 26.0000 S Second child 0 1
33 0.0000000 34.03508 0 0 7.2292 C Third man 1 0
34 0.0000000 17.00000 1 1 7.2292 C Third man 0 0
35 1.0000000 34.03508 0 2 22.3583 C woman 0 1
36 1.0000000 30.00000 0 0 8.6625 S Third woman 1 0
37 1.0000000 7.00000 0 2 26.2500 S Second child 0 1
38 0.0000000 45.00000 0 0 26.5500 S First man 1 0
39 1.0000000 30.00000 0 0 106.4250 C First woman 1 1
40 0.0000000 34.03508 0 0 14.5000 S Third man 1 0
41 1.0000000 22.00000 0 2 49.5000 C First woman 0 1
42 1.0000000 36.00000 0 2 71.0000 S First woman 0 1
43 0.3198653 9.00000 4 2 31.2750 S Third child 0 0
44 1.0000000 11.00000 4 2 31.2750 S Third child 0 0
45 0.0000000 32.00000 1 0 26.0000 S Second man 0 1
46 0.0000000 50.00000 1 0 106.4250 C First man 0 0
47 0.0000000 64.00000 0 0 26.0000 S First man 1 0
48 1.0000000 19.00000 1 0 26.0000 S Second woman 0 1
49 0.0000000 34.03508 0 0 13.8625 C Second man 1 1
50 0.0000000 33.00000 1 1 20.5250 S Third man 0 0
51 0.0000000 8.00000 1 1 36.7500 S Second child 0 1
52 0.0000000 17.00000 0 2 110.8833 C First man 0 1
53 0.0000000 27.00000 0 0 26.0000 S Second man 1 0
54 0.0000000 34.03508 0 0 7.8292 Q Third man 1 0
55 0.0000000 22.00000 0 0 7.2250 C Third man 1 1
56 1.0000000 22.00000 0 0 7.7750 S Third woman 1 1
57 0.0000000 62.00000 0 0 26.5500 S First man 1 0
58 1.0000000 48.00000 1 0 39.6000 C First woman 0 1
59 0.0000000 34.03508 0 0 227.5250 C First man 1 0
60 1.0000000 39.00000 1 1 79.6500 S First woman 0 1
61 1.0000000 36.00000 1 0 17.4000 S Third woman 0 1
62 0.0000000 34.03508 0 0 7.7500 Q Third man 1 0
63 0.0000000 40.00000 0 0 7.8958 S Third man 1 0
64 0.0000000 28.00000 0 0 13.5000 S man 1 0
65 0.0000000 34.03508 0 0 8.0500 S Third man 1 0
66 1.0000000 34.03508 0 0 8.0500 S Third woman 1 0
67 0.0000000 24.00000 2 0 24.1500 S Third man 0 0
68 0.0000000 19.00000 0 0 7.8958 S Third man 1 0
69 1.0000000 29.00000 0 4 21.0750 S Third woman 0 0
70 0.0000000 34.03508 0 0 7.2292 C Third man 1 0

```


The result of the code would be a revised dataset in which the means of the missing columns (gender, age, sibsp, parch, fare, alone, and survived) would be used to fill in the blanks. This indicates that the average value of the non-missing values in that specific column would be used to fill in any missing values in those columns. The resulting dataset would have all the missing values replaced, ensuring that each column is complete with no missing data.

Data type conversion:

Age, gender, and fare are the three mentioned columns. The "as.integer" function is used in the code to change the data type of these columns from their initial format to integers. When the variables in these columns are meant to represent discrete numerical values, like categories or whole integers, this conversion is especially helpful. The code guarantees that the variables in these columns are correctly represented as whole numbers rather than as strings or floating-point values by changing the data type to integers. This conversion aids in consistency maintenance and simplifies further analysis or calculations that could call for integer numbers.

```
> xdataset$Gender <- as.integer(xdataset$Gender)
> xdataset$age <- as.integer(xdataset$age)
> xdataset$fare <- as.integer(xdataset$fare)
>
> print(xdataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0	24	0	0	7	S	Third	mannn	1	0
2	0	17	0	0	8	S	Third	man	1	0
3	1	21	0	0	7	Q	Third	womann	1	0
4	1	34	0	0	7	Q	Third	woman	1	0
5	1	37	0	0	9	S	Third	womannnn	1	0
6	0	16	0	0	86	S	First	woman	1	1
7	0	18	1	0	108	C	First	mannn	0	0
8	1	33	0	2	26	S	Second	woman	0	1
9	0	34	0	0	26	S	First	man	1	1
10	0	28	0	0	22	S	Third	man	1	0
11	0	26	0	0	56	S	Third	man	1	1
12	0	29	0	0	7	Q	Third	man	1	1
13	0	34	0	0	8	S	Third	man	1	0
14	0	36	0	0	26	S		man	1	1
15	1	54	1	0	59	C	First	woman	0	1
16	0	24	0	0	7	S	Third	man	1	0
17	0	47	0	0	34	S	First	man	1	0
18	1	34	0	0	10	S	Second	woman	1	1
19	0	34	0	0	24	Q	Third	man	1	0
20	1	36	1	0	26	S	Second	woman	0	1
21	0	32	0	0	7	S	Third	man	1	0
22	1	30	0	0	93	S	First	woman	1	1
23	0	22	0	0	7	S	Third	man	1	0
24	0	34	0	0	7	C	Third	man	1	0
25	1	44	0	1	57	C	First	woman	0	1
26	0	34	0	0	7	C	Third	man	1	0
27	0	40	0	0	7	Q	Third	man	1	0
28	1	50	0	0	10	S	Second	woman	1	1
29	0	34	0	0	221	S	First	man	1	0
30	0	39	0	0	7	S	Third	man	1	0
31	0	23	2	1	11	S	Second	man	0	0
32	0	2	1	1	26	S	Second	child	0	1
33	0	34	0	0	7	C	Third	man	1	0
34	0	17	1	1	7	C	Third	man	0	0
35	1	34	0	2	22	C		woman	0	1

35	1	34	0	2	22	C	woman	0	1
36	1	30	0	0	8	S Third	woman	1	0
37	1	7	0	2	26	S Second	child	0	1
38	0	45	0	0	26	S First	man	1	0
39	1	30	0	0	106	C First	woman	1	1
40	0	34	0	0	14	S Third	man	1	0
41	1	22	0	2	49	C First	woman	0	1
42	1	36	0	2	71	S First	woman	0	1
43	0	9	4	2	31	S Third	child	0	0
44	1	11	4	2	31	S Third	child	0	0
45	0	32	1	0	26	S Second	man	0	1
46	0	50	1	0	106	C First	man	0	0
47	0	64	0	0	26	S First	man	1	0
48	1	19	1	0	26	S Second	woman	0	1
49	0	34	0	0	13	C Second	man	1	1
50	0	33	1	1	20	S Third	man	0	0
51	0	8	1	1	36	S Second	child	0	1
52	0	17	0	2	110	C First	man	0	1
53	0	27	0	0	26	S Second	man	1	0
54	0	34	0	0	7	Q Third	man	1	0
55	0	22	0	0	7	C Third	man	1	1
56	1	22	0	0	7	S Third	woman	1	1
57	0	62	0	0	26	S First	man	1	0
58	1	48	1	0	39	C First	woman	0	1
59	0	34	0	0	227	C First	man	1	0
60	1	39	1	1	79	S First	woman	0	1
61	1	36	1	0	17	S Third	woman	0	1
62	0	34	0	0	7	Q Third	man	1	0
63	0	40	0	0	7	S Third	man	1	0
64	0	28	0	0	13	S	man	1	0
65	0	34	0	0	8	S Third	man	1	0
66	1	34	0	0	8	S Third	woman	1	0
67	0	24	2	0	24	S Third	man	0	0
68	0	19	0	0	7	S Third	man	1	0
69	1	29	0	4	21	S Third	woman	0	0
70	0	34	0	0	7	C Third	man	1	0
71	0	32	0	0	7	S Third	man	1	1
72	0	62	0	0	10	S Second	man	1	1
73	1	53	2	0	51	S First	woman	0	1
74	0	36	0	0	26	S First	man	1	1
75	1	34	0	0	7	Q Third	woman	1	1
75	1	34	0	0	7	Q Third	woman	1	1
76	0	16	0	0	8	S Third	man	1	0
77	0	19	0	0	14	S Third	man	1	0
78	1	34	0	0	13	S Second	woman	1	1
79	1	39	1	0	55	S First	woman	0	1
80	1	34	1	0	14	C Third	woman	0	0
81	0	32	0	0	7	S Third	man	1	1
82	1	25	1	1	30	S Second	woman	0	1
83	1	39	1	1	110	C First	woman	0	1
84	0	54	0	0	26	S Second	man	1	0
85	0	36	0	0	40	C First	man	1	0
86	0	34	0	0	8	C Third	man	1	0
87	1	18	0	2	79	S First	woman	0	1
88	0	47	0	0	15	S Second	man	1	0
89	0	60	1	1	79	C First	man	0	1
90	0	22	0	0	8	S Third	man	1	0
91	0	34	0	0	8	S Third	man	1	0
92	0	35	0	0	7	S Third	man	1	0
93	1	52	1	0	78	C First	woman	0	1
94	0	47	0	0	7	S Third	man	1	0
95	1	34	0	2	7	Q Third	woman	0	0
96	0	37	1	0	26	S Second	man	0	0
97	0	36	1	1	24	S Third	man	0	0
98	1	34	0	0	33	S Second	woman	1	1
99	0	49	0	0	0	S	man	1	0
100	0	34	0	0	7	C Third	man	1	0

[reached 'max' / getOption("max.print") -- omitted 201 rows]

> |

The output of the code would be an updated dataset where the gender, age, and fare columns have been converted to integers. This means that the original values in these columns, which may have been in a different data type (such as strings or floating-point numbers), will be converted to whole numbers. The resulting dataset will have these specific columns represented as integers, ensuring that they are suitable for analysis or computations that require discrete numerical values.

Transforming passenger class:

A code snippet that performs transformations on the "class" column of a dataset. The original column consists of string labels representing different classes ("first", "second", "third"). The code uses nested ifelse statements to convert these string labels into numerical values (1, 2, 3). The transformed numerical values are then stored back into the "class" column, replacing the original string labels. This conversion allows for easier analysis and computations that require numerical representations of the classes. Additionally, the code handles missing values in the "class" column. If any missing values are present, they are replaced with the mean value of the non-missing values in

```
> xdataset$class <- ifelse(xdataset$class == "Third",3,
+                           ifelse(xdataset$class == "Second",2,
+                                   ifelse(xdataset$class == "First",1,NA)))
>
> xdataset$class[is.na(xdataset$class)]<-mean(xdataset$class,na.rm=TRUE)
> xdataset$class <- as.integer(xdataset$class)
> print(xdataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0	24	0	0	7	S	3	mannn	1	0
2	0	17	0	0	8	S	3	man	1	0
3	1	21	0	0	7	Q	3	womann	1	0
4	1	34	0	0	7	Q	3	woman	1	0
5	1	37	0	0	9	S	3	womannnn	1	0
6	0	16	0	0	86	S	1	woman	1	1
7	0	18	1	0	108	C	1	mannnn	0	0
8	1	33	0	2	26	S	2	woman	0	1
9	0	34	0	0	26	S	1	man	1	1
10	0	28	0	0	22	S	3	man	1	0
11	0	26	0	0	56	S	3	man	1	1
12	0	29	0	0	7	Q	3	man	1	1
13	0	34	0	0	8	S	3	man	1	0
14	0	36	0	0	26	S	2	man	1	1
15	1	54	1	0	59	C	1	woman	0	1
16	0	24	0	0	7	S	3	man	1	0
17	0	47	0	0	34	S	1	man	1	0
18	1	34	0	0	10	S	2	woman	1	1
19	0	34	0	0	24	Q	3	man	1	0
20	1	36	1	0	26	S	2	woman	0	1
21	0	32	0	0	7	S	3	man	1	0
22	1	30	0	0	93	S	1	woman	1	1
23	0	22	0	0	7	S	3	man	1	0
24	0	34	0	0	7	C	3	man	1	0
25	1	44	0	1	57	C	1	woman	0	1
26	0	34	0	0	7	C	3	man	1	0
27	0	40	0	0	7	Q	3	man	1	0
28	1	50	0	0	10	S	2	woman	1	1
29	0	34	0	0	221	S	1	man	1	0
30	0	39	0	0	7	S	3	man	1	0
31	0	23	2	1	11	S	2	man	0	0
32	0	2	1	1	26	S	2	child	0	1
33	0	34	0	0	7	C	3	man	1	0
34	0	17	1	1	7	C	3	man	0	0
35	1	34	0	2	22	C	2	woman	0	1

the column. This imputation method ensures that the missing values are filled with a representative value, enabling a complete dataset for further analysis.

35	1	34	0	2	22	C	2	woman	0	1
36	1	30	0	0	8	S	3	woman	1	0
37	1	7	0	2	26	S	2	child	0	1
38	0	45	0	0	26	S	1	man	1	0
39	1	30	0	0	106	C	1	woman	1	1
40	0	34	0	0	14	S	3	man	1	0
41	1	22	0	2	49	C	1	woman	0	1
42	1	36	0	2	71	S	1	woman	0	1
43	0	9	4	2	31	S	3	child	0	0
44	1	11	4	2	31	S	3	child	0	0
45	0	32	1	0	26	S	2	man	0	1
46	0	50	1	0	106	C	1	man	0	0
47	0	64	0	0	26	S	1	man	1	0
48	1	19	1	0	26	S	2	woman	0	1
49	0	34	0	0	13	C	2	man	1	1
50	0	33	1	1	20	S	3	man	0	0
51	0	8	1	1	36	S	2	child	0	1
52	0	17	0	2	110	C	1	man	0	1
53	0	27	0	0	26	S	2	man	1	0
54	0	34	0	0	7	Q	3	man	1	0
55	0	22	0	0	7	C	3	man	1	1
56	1	22	0	0	7	S	3	woman	1	1
57	0	62	0	0	26	S	1	man	1	0
58	1	48	1	0	39	C	1	woman	0	1
59	0	34	0	0	227	C	1	man	1	0
60	1	39	1	1	79	S	1	woman	0	1
61	1	36	1	0	17	S	3	woman	0	1
62	0	34	0	0	7	Q	3	man	1	0
63	0	40	0	0	7	S	3	man	1	0
64	0	28	0	0	13	S	2	man	1	0
65	0	34	0	0	8	S	3	man	1	0
66	1	34	0	0	8	S	3	woman	1	0
67	0	24	2	0	24	S	3	man	0	0
68	0	19	0	0	7	S	3	man	1	0
69	1	29	0	4	21	S	3	woman	0	0
70	0	34	0	0	7	C	3	man	1	0
71	0	32	0	0	7	S	3	man	1	1
72	0	62	0	0	10	S	2	man	1	1
73	1	53	2	0	51	S	1	woman	0	1
74	0	36	0	0	26	S	1	man	1	1
75	1	34	0	0	7	Q	3	woman	1	1
76	0	16	0	0	8	S	3	man	1	0
77	0	19	0	0	14	S	3	man	1	0
78	1	34	0	0	13	S	2	woman	1	1
79	1	39	1	0	55	S	1	woman	0	1
80	1	34	1	0	14	C	3	woman	0	0
81	0	32	0	0	7	S	3	man	1	1
82	1	25	1	1	30	S	2	woman	0	1
83	1	39	1	1	110	C	1	woman	0	1
84	0	54	0	0	26	S	2	man	1	0
85	0	36	0	0	40	C	1	man	1	0
86	0	34	0	0	8	C	3	man	1	0
87	1	18	0	2	79	S	1	woman	0	1
88	0	47	0	0	15	S	2	man	1	0
89	0	60	1	1	79	C	1	man	0	1
90	0	22	0	0	8	S	3	man	1	0
91	0	34	0	0	8	S	3	man	1	0
92	0	35	0	0	7	S	3	man	1	0
93	1	52	1	0	78	C	1	woman	0	1
94	0	47	0	0	7	S	3	man	1	0
95	1	34	0	2	7	Q	3	woman	0	0
96	0	37	1	0	26	S	2	man	0	0
97	0	36	1	1	24	S	3	man	0	0
98	1	34	0	0	33	S	2	woman	1	1
99	0	49	0	0	0	S	2	man	1	0
100	0	34	0	0	7	C	3	man	1	0

[reached 'max' / getOption("max.print") -- omitted 201 rows]

> |

Reverting passenger class to labels:

The numerical transformations applied to the "class" column back to its original string labels ("first", "second", "third"). To achieve this, the code uses nested ifelse statements to conditionally assign the corresponding string label based on the numerical value. The original numerical values in the "class" column are replaced with their respective string labels. Once the transformation is complete, the resulting "class" column, now containing the original string labels, is printed to the console. This allows for the verification and inspection of the transformed column.

```
> xdataset$class <- ifelse(xdataset$class == 3,"Third",  
+                           ifelse(xdataset$class == 2,"Second",  
+                                   ifelse(xdataset$class == 1,"First",NA)))  
> print(xdataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0	24	0	0	7	S	Third	mannn	1	0
2	0	17	0	0	8	S	Third	man	1	0
3	1	21	0	0	7	Q	Third	womann	1	0
4	1	34	0	0	7	Q	Third	woman	1	0
5	1	37	0	0	9	S	Third	womannn	1	0
6	0	16	0	0	86	S	First	woman	1	1
7	0	18	1	0	108	C	First	mannn	0	0
8	1	33	0	2	26	S	Second	woman	0	1
9	0	34	0	0	26	S	First	man	1	1
10	0	28	0	0	22	S	Third	man	1	0
11	0	26	0	0	56	S	Third	man	1	1
12	0	29	0	0	7	Q	Third	man	1	1
13	0	34	0	0	8	S	Third	man	1	0
14	0	36	0	0	26	S	Second	man	1	1
15	1	54	1	0	59	C	First	woman	0	1
16	0	24	0	0	7	S	Third	man	1	0
17	0	47	0	0	34	S	First	man	1	0
18	1	34	0	0	10	S	Second	woman	1	1
19	0	34	0	0	24	Q	Third	man	1	0
20	1	36	1	0	26	S	Second	woman	0	1
21	0	32	0	0	7	S	Third	man	1	0
22	1	30	0	0	93	S	First	woman	1	1
23	0	22	0	0	7	S	Third	man	1	0
24	0	34	0	0	7	C	Third	man	1	0
25	1	44	0	1	57	C	First	woman	0	1
26	0	34	0	0	7	C	Third	man	1	0
27	0	40	0	0	7	Q	Third	man	1	0
28	1	50	0	0	10	S	Second	woman	1	1
29	0	34	0	0	221	S	First	man	1	0
30	0	39	0	0	7	S	Third	man	1	0

70	0	34	0	0	7	C	Third	man	1	0
71	0	32	0	0	7	S	Third	man	1	1
72	0	62	0	0	10	S	Second	man	1	1
73	1	53	2	0	51	S	First	woman	0	1
74	0	36	0	0	26	S	First	man	1	1
75	1	34	0	0	7	Q	Third	woman	1	1
76	0	16	0	0	8	S	Third	man	1	0
77	0	19	0	0	14	S	Third	man	1	0
78	1	34	0	0	13	S	Second	woman	1	1
79	1	39	1	0	55	S	First	woman	0	1
80	1	34	1	0	14	C	Third	woman	0	0
81	0	32	0	0	7	S	Third	man	1	1
82	1	25	1	1	30	S	Second	woman	0	1
83	1	39	1	1	110	C	First	woman	0	1
84	0	54	0	0	26	S	Second	man	1	0
85	0	36	0	0	40	C	First	man	1	0
86	0	34	0	0	8	C	Third	man	1	0
87	1	18	0	2	79	S	First	woman	0	1
88	0	47	0	0	15	S	Second	man	1	0
89	0	60	1	1	79	C	First	man	0	1
90	0	22	0	0	8	S	Third	man	1	0
91	0	34	0	0	8	S	Third	man	1	0
92	0	35	0	0	7	S	Third	man	1	0
93	1	52	1	0	78	C	First	woman	0	1
94	0	47	0	0	7	S	Third	man	1	0
95	1	34	0	2	7	Q	Third	woman	0	0
96	0	37	1	0	26	S	Second	man	0	0
97	0	36	1	1	24	S	Third	man	0	0
98	1	34	0	0	33	S	Second	woman	1	1
99	0	49	0	0	0	S	Second	man	1	0
100	0	34	0	0	7	C	Third	man	1	0
[reached 'max' / getOption("max.print") -- omitted 201 rows]										
30	0	39	0	0	7	S	Third	man	1	0
31	0	23	2	1	11	S	Second	man	0	0
32	0	2	1	1	26	S	Second	child	0	1
33	0	34	0	0	7	C	Third	man	1	0
34	0	17	1	1	7	C	Third	man	0	0
35	1	34	0	2	22	C	Second	woman	0	1
36	1	30	0	0	8	S	Third	woman	1	0
37	1	7	0	2	26	S	Second	child	0	1
38	0	45	0	0	26	S	First	man	1	0
39	1	30	0	0	106	C	First	woman	1	1
40	0	34	0	0	14	S	Third	man	1	0
41	1	22	0	2	49	C	First	woman	0	1
42	1	36	0	2	71	S	First	woman	0	1
43	0	9	4	2	31	S	Third	child	0	0
44	1	11	4	2	31	S	Third	child	0	0
45	0	32	1	0	26	S	Second	man	0	1
46	0	50	1	0	106	C	First	man	0	0
47	0	64	0	0	26	S	First	man	1	0
48	1	19	1	0	26	S	Second	woman	0	1
49	0	34	0	0	13	C	Second	man	1	1
50	0	33	1	1	20	S	Third	man	0	0
51	0	8	1	1	36	S	Second	child	0	1
52	0	17	0	2	110	C	First	man	0	1
53	0	27	0	0	26	S	Second	man	1	0
54	0	34	0	0	7	Q	Third	man	1	0
55	0	22	0	0	7	C	Third	man	1	1
56	1	22	0	0	7	S	Third	woman	1	1
57	0	62	0	0	26	S	First	man	1	0
58	1	48	1	0	39	C	First	woman	0	1
59	0	34	0	0	227	C	First	man	1	0
60	1	39	1	1	79	S	First	woman	0	1
61	1	36	1	0	17	S	Third	woman	0	1
62	0	34	0	0	7	Q	Third	man	1	0
63	0	40	0	0	7	S	Third	man	1	0
64	0	28	0	0	13	S	Second	man	1	0
65	0	34	0	0	8	S	Third	man	1	0
66	1	34	0	0	8	S	Third	woman	1	0
67	0	24	2	0	24	S	Third	man	0	0
68	0	19	0	0	7	S	Third	man	1	0
69	1	29	0	4	21	S	Third	woman	0	0
70	0	34	0	0	7	C	Third	man	1	0

Converting "alone" column:

Transformation on the "alone" column of a dataset. The original column contains numerical values (0 and 1), which likely represent binary indicators (e.g., 0 for "not alone" and 1 for "alone"). The code utilizes the ifelse statement to conditionally assign string labels ("false" and "true") based on the numerical values in the "alone" column. This transformation converts the numerical values into their corresponding string labels. After the transformation is applied, the resulting "alone" column, now consisting of the string labels ("false" and "true"), is printed to the console. This allows for visual inspection and verification of the transformed column.

```
> xdataset$alone <- ifelse(xdataset$alone == 1,"TRUE",
+                           ifelse(xdataset$alone == 0,"FALSE",NA))
>
> print(xdataset)
```

	Gender	age	sibsp	parch	fare	embarked	class	who	alone	survived
1	0	24	0	0	7	S	Third	mann	TRUE	0
2	0	17	0	0	8	S	Third	man	TRUE	0
3	1	21	0	0	7	Q	Third	womann	TRUE	0
4	1	34	0	0	7	Q	Third	woman	TRUE	0
5	1	37	0	0	9	S	Third	womannn	TRUE	0
6	0	16	0	0	86	S	First	woman	TRUE	1
7	0	18	1	0	108	C	First	mannn	FALSE	0
8	1	33	0	2	26	S	Second	woman	FALSE	1
9	0	34	0	0	26	S	First	man	TRUE	1
10	0	28	0	0	22	S	Third	man	TRUE	0
11	0	26	0	0	56	S	Third	man	TRUE	1
12	0	29	0	0	7	Q	Third	man	TRUE	1
13	0	34	0	0	8	S	Third	man	TRUE	0
14	0	36	0	0	26	S	Second	man	TRUE	1
15	1	54	1	0	59	C	First	woman	FALSE	1
16	0	24	0	0	7	S	Third	man	TRUE	0
17	0	47	0	0	34	S	First	man	TRUE	0
18	1	34	0	0	10	S	Second	woman	TRUE	1
19	0	34	0	0	24	Q	Third	man	TRUE	0
20	1	36	1	0	26	S	Second	woman	FALSE	1
21	0	32	0	0	7	S	Third	man	TRUE	0
22	1	30	0	0	93	S	First	woman	TRUE	1
23	0	22	0	0	7	S	Third	man	TRUE	0
24	0	34	0	0	7	C	Third	man	TRUE	0
25	1	44	0	1	57	C	First	woman	FALSE	1
26	0	34	0	0	7	C	Third	man	TRUE	0
27	0	40	0	0	7	Q	Third	man	TRUE	0
28	1	50	0	0	10	S	Second	woman	TRUE	1
29	0	34	0	0	221	S	First	man	TRUE	0
30	0	39	0	0	7	S	Third	man	TRUE	0

31	0	23	2	1	11	S	Second	man	FALSE	0
32	0	2	1	1	26	S	Second	child	FALSE	1
33	0	34	0	0	7	C	Third	man	TRUE	0
34	0	17	1	1	7	C	Third	man	FALSE	0
35	1	34	0	2	22	C	Second	woman	FALSE	1
36	1	30	0	0	8	S	Third	woman	TRUE	0
37	1	7	0	2	26	S	Second	child	FALSE	1
38	0	45	0	0	26	S	First	man	TRUE	0
39	1	30	0	0	106	C	First	woman	TRUE	1
40	0	34	0	0	14	S	Third	man	TRUE	0
41	1	22	0	2	49	C	First	woman	FALSE	1
42	1	36	0	2	71	S	First	woman	FALSE	1
43	0	9	4	2	31	S	Third	child	FALSE	0
44	1	11	4	2	31	S	Third	child	FALSE	0
45	0	32	1	0	26	S	Second	man	FALSE	1
46	0	50	1	0	106	C	First	man	FALSE	0
47	0	64	0	0	26	S	First	man	TRUE	0
48	1	19	1	0	26	S	Second	woman	FALSE	1
49	0	34	0	0	13	C	Second	man	TRUE	1
50	0	33	1	1	20	S	Third	man	FALSE	0
51	0	8	1	1	36	S	Second	child	FALSE	1
52	0	17	0	2	110	C	First	man	FALSE	1
53	0	27	0	0	26	S	Second	man	TRUE	0
54	0	34	0	0	7	Q	Third	man	TRUE	0
55	0	22	0	0	7	C	Third	man	TRUE	1
56	1	22	0	0	7	S	Third	woman	TRUE	1
57	0	62	0	0	26	S	First	man	TRUE	0
58	1	48	1	0	39	C	First	woman	FALSE	1
59	0	34	0	0	227	C	First	man	TRUE	0
60	1	39	1	1	79	S	First	woman	FALSE	1
61	1	36	1	0	17	S	Third	woman	FALSE	1
62	0	34	0	0	7	Q	Third	man	TRUE	0
63	0	40	0	0	7	S	Third	man	TRUE	0
64	0	28	0	0	13	S	Second	man	TRUE	0
65	0	34	0	0	8	S	Third	man	TRUE	0
66	1	34	0	0	8	S	Third	woman	TRUE	0
67	0	24	2	0	24	S	Third	man	FALSE	0
68	0	19	0	0	7	S	Third	man	TRUE	0
69	1	29	0	4	21	S	Third	woman	FALSE	0
70	0	34	0	0	7	C	Third	man	TRUE	0

```

71      0 32      0      0      7      S Third      man TRUE      1
72      0 62      0      0     10      S Second     man TRUE      1
73      1 53      2      0     51      S First      woman FALSE     1
74      0 36      0      0     26      S First      man TRUE      1
75      1 34      0      0      7      Q Third      woman TRUE      1
76      0 16      0      0      8      S Third      man TRUE      0
77      0 19      0      0     14      S Third      man TRUE      0
78      1 34      0      0     13      S Second     woman TRUE      1
79      1 39      1      0     55      S First      woman FALSE     1
80      1 34      1      0     14      C Third      woman FALSE     0
81      0 32      0      0      7      S Third      man TRUE      1
82      1 25      1      1     30      S Second     woman FALSE     1
83      1 39      1      1    110      C First      woman FALSE     1
84      0 54      0      0     26      S Second     man TRUE      0
85      0 36      0      0     40      C First      man TRUE      0
86      0 34      0      0      8      C Third      man TRUE      0
87      1 18      0      2     79      S First      woman FALSE     1
88      0 47      0      0     15      S Second     man TRUE      0
89      0 60      1      1     79      C First      man FALSE     1
90      0 22      0      0      8      S Third      man TRUE      0
91      0 34      0      0      8      S Third      man TRUE      0
92      0 35      0      0      7      S Third      man TRUE      0
93      1 52      1      0     78      C First      woman FALSE     1
94      0 47      0      0      7      S Third      man TRUE      0
95      1 34      0      2      7      Q Third      woman FALSE     0
96      0 37      1      0     26      S Second     man FALSE     0
97      0 36      1      1     24      S Third      man FALSE     0
98      1 34      0      0     33      S Second     woman TRUE      1
99      0 49      0      0      0      S Second     man TRUE      0
100     0 34      0      0      7      C Third      man TRUE      0
[ reached 'max' / getOption("max.print") -- omitted 201 rows ]
> |

```

Data visualization:

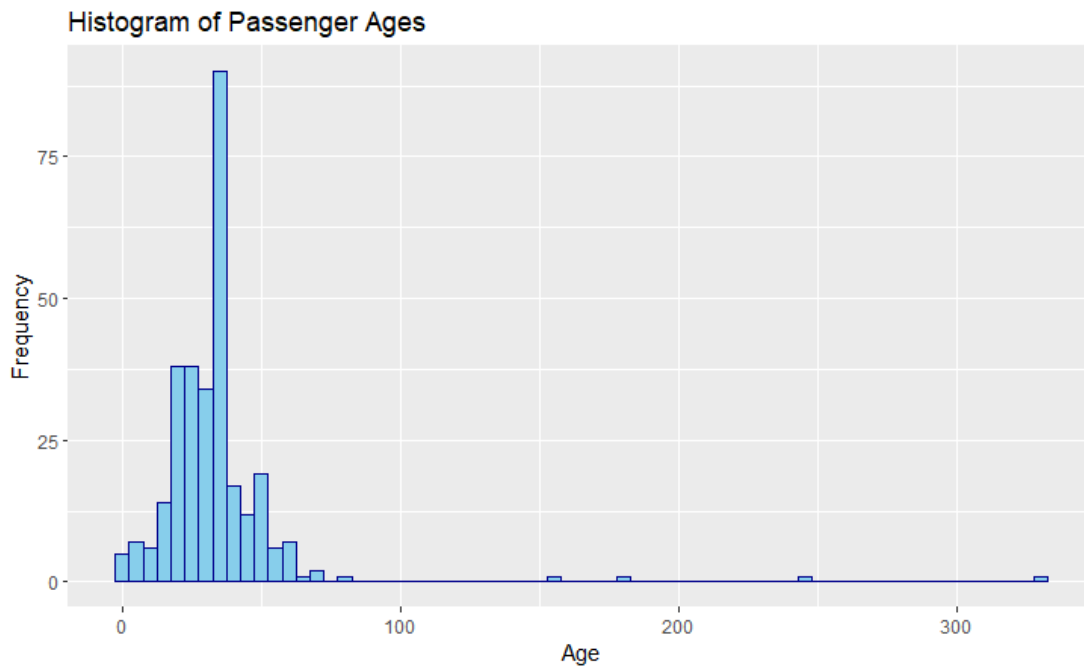
the code checks if the "ggplot2" package is installed. If not, it installs the package using the `install.packages` function. Then, it loads the "ggplot2" library using the `library` function.

```

44 ▾ if (!require("ggplot2")) {
45   install.packages("ggplot2")
46 ▾ }
47
48 library(ggplot2)
49
50 ggplot(data = xdataset, aes(x = age)) +
51   geom_histogram(binwidth = 5, fill = "skyblue", color = "darkblue") +
52   labs(title = "Histogram of Passenger Ages", x = "Age", y = "Frequency")

```

A histogram is created using the `ggplot` function from the "ggplot2" package. The histogram represents the distribution of passenger ages from the `xdataset`. The `geom_histogram` function is used to create the histogram, specifying the binwidth, fill color, and line color. The `labs` function is used to set the title and labels for the x and y axes. The resulting plot is displayed.



Discussion & conclusion:

The analysis of the titanic passenger dataset involved data preprocessing steps and data visualization. The missing values were handled using mean imputation, ensuring that no missing data hindered subsequent analyses. Additionally, variable transformations were performed to enable more accurate analysis and interpretation. The resulting dataset, along with the generated histogram, provides valuable insights and sets the stage for further exploration and modeling tasks. It is important to note that any future analyses or modeling based on this dataset should consider the preprocessing steps performed here to ensure reliable results.