

```
Pandas libraries can be used for Data Cleansing, Data Analysis and Data Transformation. In this project, we are working on hotel-booking-data.

1. Read hotel-booking-data.txt file using pandas read_table method.
2. Use drop_nai function to drop the Nan values. Don't use inplace=True initially as we need the data present in one column.
3. Use df.reset_index(drop=True, inplace=True)
4. Find the number of bookings made by each company?
5. Use reset_index(inplace=True) function in numpy, and fillna() from df to clean the dataset.

In [1]: Import: numpy as np
import pandas as pd
df=pd.read_table(r'C:\Users\vsyal\OneDrive\Documents\Datasets for Pandas\hotel-booking-data.txt',sep=';')
df

Out[1]:
   Date      Company      Person Name  Room number
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0
3  Hotels      NaN      NaN      NaN
4  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0
...      ...      ...      ...      ...
190 Cleartrip      NaN      NaN      NaN
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0
192 1-Jan-2022  Innojam  Leandra Polaczuk      5002.0
193 1-Jan-2022  Twisterworks  Valeria Ledson      1010.0
194 Hotels      NaN      NaN      NaN

195 rows x 4 columns

In [2]: # From the above df we can see that, sample data has few text in between as well as NaN value. Let's get rid of those first.
df.dropna()

# By doing this it actually removed the text data that was present in those rows. Let's say we need to retain those values for
further analysis, in that case this is not a good solution. I have not used inplace here, so our original df is not affected.

Out[2]:
   Date      Company      Person Name  Room number
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0
3  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0
4  1-Jan-2022  Eadef  Broderic Handscombe      3015.0
...      ...      ...      ...
186 1-Jan-2022  Tagrad    Stephen Lufee      1015.0
187 1-Jan-2022  Meevee    Victoria Lavery      7002.0
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0
192 1-Jan-2022  Innojam  Leandra Polaczuk      5002.0
193 1-Jan-2022  Twisterworks  Valeria Ledson      1010.0

134 rows x 4 columns

In [3]: pd.set_option('display.max_rows', 134)

In [4]: # No of bookings made by each company
df['Company'].value_counts()

Out[4]:
Company
Leeso      3
Fivechat   3
Olinio     2
Jabbersphere 2
Twistervill 2
Miboo      2
Rhyzo      2
Realcube   2
Meevee     2
Taglune    2
Realcube   2
Skintix    2
Stipidlog  2
Alamo      2
Jays       2
Eadef      2
Rhyzo      2
Alyce      2
Jettie     1
Teehee    1
Guruvu    1
Buzibean  1
Camilo    1
Buzzerster 1
Fivechat  1
Tadel     1
Hence     1
Yotilo    1
Oline     1
Flashpoint 1
Vipe      1
Leenti    1
Edgclub   1
Tobon     1
Dazsaphere 1
Olion     1
Miboo     1
Awamba    1
Vinder    1
Zomdog    1
Innojam   1
Tagrad    1
Skiba     1
Twisterviv 1
Jaworks   1
Vipe      1
Vomee     1
Yotifred  1
Voollia   1
Thoughtbridge 1
Centidel  1
Eone      1
Demvee    1
Rhyznoodle 1
Devnug    1
Mecppedia 1
Subbtride 1
Blightbean 1
Photobean 1
Tobocman  1
Tado      1
Awame     1
Dizvanoodle 1
Feeddog   1
Skinder   1
Ozipe     1
Feedspan  1
Melin     1
Tagchat   1
Dynabox   1
Muso      1
Centrus   1
Kinia     1
Ophazay   1
Rabbilattom 1
Alamo     1
Mita      1
Rhyznoodle 1
Feedria   1
Topidlab  1
Filipopia 1
Odo       1
Shuffletag 1
Skinta    1
Buzbeanoom 1
Tazir     1
Ruffpasia 1
Jontotype 1
Ooze      1
Rhyzio    1
Skallib   1
Buzidlog  1
Feedclab  1
Jave      1
Wiffpith  1
Lain      1
Fate      1
Tagoppia  1
Hewesee   1
Rhyzero   1
Rlyglo    1
Dababota  1
Eare      1
Odevot    1
Devshave  1
Topicroom  1
Shilox    1
Myrak     1
Rafio     1
Oadlora   1
Youpia    1
Topicounge 1
Shuffelebat 1
Jays      1
Youspan   1
Twisterviv 1
Name: count, dtype: int64

In [5]: df

Out[5]:
   Date      Company      Person Name  Room number
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0
3  Hotels      NaN      NaN      NaN
4  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0
...      ...      ...      ...
190 Cleartrip      NaN      NaN      NaN
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0
192 1-Jan-2022  Innojam  Leandra Polaczuk      5002.0
193 1-Jan-2022  Twisterworks  Valeria Ledson      1010.0
194 Hotels      NaN      NaN      NaN

195 rows x 4 columns

In [6]: # Let's manipulate the dataframe by moving the text data (hotel, cleartrip etc) to a separate column.
# The error() fn (which gives a true or false value) to check if one of those column value (let's choose Room number) is NaN. isna() will produce a
boolean column (df dataframe)
df['Room number'].isna()

Out[6]:
0      False
1      False
2      False
3      True
4      False
...
190     True
191     False
192     False
193     False
194     True
Name: Room number, Length: 195, dtype: bool

numpy module has a where() function, which acts like an if condition. Based on the condition (maskedvalue), we provide the true and false values. It is kind of an if condition but then imagine as if it goes inside a loop.

df['text_value']=np.where(maskedvalue df['Date'][np.NaN] == 'If masked value is true' then get the value from df['Date'] column , else insert NaN value to the newly created text_value column. df['text_value'] == new column created df['Date'] == the first column in the df

In [7]: maskedvalue=df['Room number'].isna()
df['text_value']=np.where(maskedvalue,df['Date'],np.NaN)
df

Out[7]:
   Date      Company      Person Name  Room number  text_value
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0      NaN
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0      NaN
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0      NaN
3  Hotels      NaN      NaN      NaN      Hotels
4  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0      NaN
...      ...      ...      ...      ...
190 Cleartrip      NaN      NaN      NaN      Cleartrip
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0      NaN
192 1-Jan-2022  Innojam  Leandra Polaczuk      5002.0      NaN
193 1-Jan-2022  Twisterworks  Valeria Ledson      1010.0      NaN
194 Hotels      NaN      NaN      NaN      Hotels

195 rows x 5 columns

In [8]: # Now let's use the fill function in the dataframe to fill it up the values. df.fill stands for backward fill.
# The code df['text_value'].ffill() performs a backward fill on the column text_value in a dataframe df. The backward fill method replaces any NaN values in the column by filling them with the next non-null value
df['text_value'].ffill(inplace=True) # deprecated syntax
df['text_value'].bfill(inplace=True)
df

Out[8]:
   Date      Company      Person Name  Room number  text_value
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0  Hotels
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0  Hotels
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0  Hotels
3  Hotels      NaN      NaN      NaN      Hotels
4  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0  Booking
...      ...      ...      ...      ...
190 Cleartrip      NaN      NaN      NaN      Cleartrip
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0  Hotels
192 1-Jan-2022  Innojam  Leandra Polaczuk      5002.0  Hotels
193 1-Jan-2022  Twisterworks  Valeria Ledson      1010.0  Hotels
194 Hotels      NaN      NaN      NaN      Hotels

195 rows x 5 columns

In [9]: # Drop the remaining NaN values in the other columns
df.dropna(inplace=True)

In [10]: # Now we have a clean dataset to perform data analysis
df

Out[10]:
   Date      Company      Person Name  Room number  text_value
0  1-Jan-2022  Awamba  Anatole Ridehalgh      4008.0  Hotels
1  1-Jan-2022  Fatz    Aldrich McKevlin      2002.0  Hotels
2  1-Jan-2022  Leeso    Stanley Hadrell      4012.0  Hotels
3  Hotels      NaN      NaN      NaN      Hotels
4  1-Jan-2022  Rhyzo    Lyndell Tice      1006.0  Booking
5  1-Jan-2022  Eadef  Broderic Handscombe      3015.0  Booking
6  1-Jan-2022  Ozez    Dana Harwin      2005.0  Booking
7  1-Jan-2022  Innotype  Benjamin Croatt      2001.0  Booking
10 1-Jan-2022  Jays     Tate Mennethke      3014.0  Booking
13 1-Jan-2022  Riffpedia  Eleanor Vige      3002.0  Cleartrip
14 1-Jan-2022  Taz     Alonso Murdes      4006.0  Cleartrip
16 1-Jan-2022  Brownstoom  Ysabel Lorton      6003.0  Hotels
17 1-Jan-2022  Skinte    Rafi Wecker      3012.0  Hotels
19 1-Jan-2022  Twisterfat  Andrea Hungath      4005.0  Hotels
21 1-Jan-2022  Shuffletag  Cammy Curle      2016.0  Expedia
22 1-Jan-2022  Gaba      Howey Glesner      6001.0  Expedia
25 1-Jan-2022  Filipopia  Prince Caspenhall      2002.0  Booking
26 1-Jan-2022  TopicLab   Noni Tarbett      1012.0  Booking
27 1-Jan-2022  Feedmix    Brandon Turp      3007.0  Booking
28 1-Jan-2022  Skynoodle  Andros Cathcart      1006.0  Booking
30 1-Jan-2022  Mita       Aubert Racher      4004.0  Travel Agent 007
31 1-Jan-2022  Skintix    Curcio Lewis      3006.0  Travel Agent 007
32 1-Jan-2022  Ambu       Weider Brookz      1001.0  Travel Agent 007
33 1-Jan-2022  Babbleslom  Alric Reeder      7001.0  Travel Agent 007
34 1-Jan-2022  Jays      Melany Brimbacombe      3010.0  Travel Agent 007
35 1-Jan-2022  Quinu     Art Giannotti      3013.0  Travel Agent 007
37 1-Jan-2022  Dynatzy    Ericha MacBain      4002.0  Expedia
38 1-Jan-2022  Taghune    Scarlett Berthel      2001.0  Expedia
40 1-Jan-2022  Kinia     Lottie Barnsdal      2014.0  Expedia
41 1-Jan-2022  Centrus   Faydra Hufland      5006.0  Expedia
42 1-Jan-2022  Muso      Felice Kramer      4002.0  Expedia
43 1-Jan-2022  Dynabox   Deane Gerson      5005.0  Expedia
44 1-Jan-2022  Skallith  Willie Norrme      3002.0  Expedia
45 1-Jan-2022  Tagchat   Almira Bartolometti      2004.0  Expedia
47 1-Jan-2022  Buzidlog  Shanna Ferrares      4012.0  Travel Agent 007
48 1-Jan-2022  Zize      Italo Gaynes      1013.0  Travel Agent 007
51 1-Jan-2022  Nlaga     Cory Masell      3002.0  Booking
53 1-Jan-2022  Youspan    Yassin Swager      3012.0  Travel Agent 007
54 1-Jan-2022  Skiyu     Sherlock Hyland      6001.0  Travel Agent 007
55 1-Jan-2022  Quinu     Kalind Shaughnessy      6003.0  Travel Agent 007
56 1-Jan-2022  Twisterfat  Lavinella Stoke      4004.0  Travel Agent 007
58 1-Jan-2022  Brightdog  Lucie Jewett      2009.0  Hotels
59 1-Jan-2022  Shuffelebat  Dorita Boulder      6005.0  Hotels
60 1-Jan-2022  Fivechat  Orsola Cowdery      4001.0  Hotels
61 1-Jan-2022  Topicounge  Lilah Attyde      1006.0  Hotels
62 1-Jan-2022  Youopia    Aley Fyer      2002.0  Hotels
64 1-Jan-2022  Dabjam     Pearline Tootan      4010.0  Expedia
65 1-Jan-2022  Kazzo     Alvan Hardwick      4005.0  Expedia
67 1-Jan-2022  Myrak     Gipsy Bellson      5006.0  Expedia
69 1-Jan-2022  Fivechat  Kinia Linley      2001.0  Expedia
70 1-Jan-2022  Meevee    Rodrigre Brookbank      2013.0  Expedia
71 1-Jan-2022  Skibee    Jonathan Ewebank      3007.0  Expedia
72 1-Jan-2022  Topicroom  Ladonna Castagna      3012.0  Expedia
74 1-Jan-2022  Realfree   Jules Evli      4010.0  Hotels
75 1-Jan-2022  Dewshane  Nathaniel Hale      2013.0  Hotels
76 1-Jan-2022  Devocat    Jacques Stiffens      1013.0  Hotels
77 1-Jan-2022  Eare      Felix Donovanson      4010.0  Hotels
78 1-Jan-2022  Dababota  Certe Dominguez      1011.0  Hotels
80 1-Jan-2022  StageKS    Dal McCaughy      1004.0  Booking
83 1-Jan-2022  Rhyzero    Merril Kleinhardt      1014.0  Booking
84 1-Jan-2022  Meevee     Fredrick Hutto      2008.0  Booking
85 1-Jan-2022  Nlaga     Angeline Goodbain      1008.0  Booking
86 1-Jan-2022  Tagopia    Steven Casacchia      2001.0  Booking
87 1-Jan-2022  Feedspan   Abel Do      3016.0  Booking
88 1-Jan-2022  Lipo       Marguerita Cuthren      6001.0  Booking
90 1-Jan-2022  Anyx       Ely Boland      4007.0  Hotels
92 1-Jan-2022  Riffpith   Sera Emlin      2006.0  Expedia
93 1-Jan-2022  Rhyrux     Camella O'Rusdan      1004.0  Expedia
94 1-Jan-2022  Feedfish   Marguerite Blanchard      2016.0  Expedia
95 1-Jan-2022  Kainu     Tabot Martineh      4010.0  Expedia
97 1-Jan-2022  Skinder    Bobbie Dix      1012.0  Expedia
98 1-Jan-2022  Jabbersphere  Merry Lawler      1008.0  Expedia
99 1-Jan-2022  Tambree    Reggie Chalecott      3006.0  Expedia
101 1-Jan-2022  Yadel     Adda Dix      5002.0  Expedia
102 1-Jan-2022  Flashpoint  Virginia Congram      1015.0  Expedia
104 1-Jan-2022  Vipe      Emmetl Schneid      4006.0  Booking
106 1-Jan-2022  Edgclub    Jeffrey Shuffax      1007.0  Expedia
108 1-Jan-2022  Realfree   Oby Redington      1011.0  Hotels
109 1-Jan-2022  Leeso     Leslie Crisp      5003.0  Hotels
111 1-Jan-2022  Jettie    Harmon Humes      4008.0  Hotels
113 1-Jan-2022  Yotilo    Chickie Altman      4001.0  Hotels
117 1-Jan-2022  Olin      Curtis Oney      2005.0  Expedia
118 1-Jan-2022  Eadef     Duley Harrod      7001.0  Expedia
119 1-Jan-2022  Dazsaphere  Dunn Davytsenko      4010.0  Expedia
120 1-Jan-2022  Yabex     Matteo Lovelady      4005.0  Expedia
122 1-Jan-2022  Rhyrux     Jurie Parrott      5007.0  Hotels
123 1-Jan-2022  Miboo     Pierce Messo      4007.0  Hotels
124 1-Jan-2022  Leenti    Marven Burbudgie      3010.0  Hotels
125 1-Jan-2022  Quire     Leland Somers      3007.0  Hotels
126 1-Jan-2022  Meetz     Celia Condonin      1016.0  Hotels
127 1-Jan-2022  Realcube   Terrie Berme      2011.0  Hotels
131 1-Jan-2022  Fiveclub   Akim Chang      4004.0  Cleartrip
132 1-Jan-2022  Buzzster   Harwell Vance      2001.0  Cleartrip
134 1-Jan-2022  Camido     Steven Brinbel      1013.0  Cleartrip
135 1-Jan-2022  Buzibean  Moyra Ferris      1008.0  Cleartrip
137 1-Jan-2022  Taglune    Parnmi Powley      3001.0  Expedia
138 1-Jan-2022  Divanu     Cozmo Rosenblum      1009.0  Expedia
139 1-Jan-2022  Vinder    Loretta Vassiere      1004.0  Expedia
140 1-Jan-2022  Quinu     Brandt Groven      1005.0  Expedia
141 1-Jan-2022  Ambo      Perry Roberts      2009.0  Expedia
142 1-Jan-2022  Demivee    Walsh Varnston      1004.0  Expedia
144 1-Jan-2022  Anyx       Georp Jolley      2011.0  Expedia
145 1-Jan-2022  Ozyne     Kim Chalecott      1012.0  Expedia
146 1-Jan-2022  Feedclup   Del Wheeler      2004.0  Expedia
147 1-Jan-2022  Divanoodle  Rael Yeom      3015.0  Expedia
149 1-Jan-2022  Avamm     Katalina Farjan      1009.0  Booking
150 1-Jan-2022  Yotilo    Nanci Symon      2009.0  Booking
151 1-Jan-2022  Photobean  Mary Jerome      6001.0  Booking
153 1-Jan-2022  Realcube   Valeria Hammond      2015.0  Hotels
154 1-Jan-2022  Brightban  Hagan McRobert      1006.0  Hotels
157 1-Jan-2022  Bubbletube  Annabel Tuxell      4010.0  Cleartrip
158 1-Jan-2022  Wordpedia  Estella Jirck      2005.0  Cleartrip
160 1-Jan-2022  Leeso     Zebulen Galliban      5004.0  Expedia
163 1-Jan-2022  Devnug    Benji Quadrio      1016.0  Hotels
164 1-Jan-2022  Jabbersphere  Jarrell Salazar      3011.0  Hotels
165 1-Jan-2022  Rhyznoodle  Delore Cassori      3010.0  Hotels
166 1-Jan-2022  Zomdog    Licha Altyco      7002.0  Hotels
167 1-Jan-2022  Zoocoo    Charlene Pickard      1009.0  Hotels
169 1-Jan-2022  Skintix    Julianne Ghelardi      5004.0  Cleartrip
170 1-Jan-2022  Centidel  Heda Burcombe      2008.0  Cleartrip
171 1-Jan-2022  Voollia    Sander Slagg      2002.0  Booking
173 1-Jan-2022  Yotifred  Marguerite Bodell      3014.0  Booking
176 1-Jan-2022  Voorm     Rudyard Stalbrass      1014.0  Booking
179 1-Jan-2022  Vipe      Pegello Artzol      4009.0  Hotels
180 1-Jan-2022  Ambu      Ashl Coniche      1014.0  Hotels
182 1-Jan-2022  Jaworks    Nomy Sath      3015.0  Expedia
183 1-Jan-2022  Twistervive  Kelen Couthrope      2009.0  Expedia
184 1-Jan-2022  Skiba     George Dwingale      7001.0  Expedia
185 1-Jan-2022  Brightdog  Bernardo Redman      1015.0  Expedia
186 1-Jan-2022  Tagrad     Stephani Laffer      3016.0  Expedia
187 1-Jan-2022  Meevee    Victoria Lavery      7002.0  Expedia
191 1-Jan-2022  Fivechat  Corabella Saye      4008.0  Hotels
```

