

Investigate_a_Dataset

April 12, 2019

Tip: Welcome to the Investigate a Dataset project! You will find tips in quoted sections like this to help organize your approach to your investigation. Before submitting your project, it will be a good idea to go back through your report and remove these sections to make the presentation of your work as tidy as possible. First things first, you might want to double-click this Markdown cell and change the title so that it reflects your dataset and investigation.

1 Project: Investigate a Dataset (No Show Appointments)

1.1 Table of Contents

Introduction

Data Wrangling

Exploratory Data Analysis

Conclusions

Introduction

I have choosed to use the no show appointments, as it relevant to my field of work, and I have been alwasly wondering how can we predict the no show, so here is my analysis for this data set.

Main Questions I tried to answer:

1. Does the time delta (Wait Time) between the Scd_day and app_day is a reason for no show ?
2. what is the impact of the patient age on the show and no show?

Some other question I thought about:

3. Does the patient area has any relevance to the no show ?

```
In [2]: # Use this cell to set up import statements for all of the packages that you
        # plan to use.
```

```
# Remember to include a 'magic word' so that your visualizations are plotted
# inline with the notebook. See this page for more:
# http://ipython.readthedocs.io/en/stable/interactive/magics.html
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
import re
import time as t
import datetime as dt
from datetime import date

df = pd.read_csv('noshowappointments-kaggle2-may-2016.csv')
#df = pd.read_csv('modified_no_show_dataset.csv')
```

Data Wrangling & Cleaning

Tip: In this section of the report, you will load in the data, check for cleanliness, and then trim and clean your dataset for analysis. Make sure that you document your steps carefully and justify your cleaning decisions.

1.1.1 General Properties

```
In [30]: # Load your data and print out a few lines. Perform operations to inspect data
#        types and look for instances of missing or possibly errant data.
df_pat = df.PatientId.astype(int,inplace=True)
df['PatientId'] = df_pat
df.head()
```

```
Out[30]:
```

	PatientId	AppointmentID	Gender	ScheduledDay \
0	29872499824296	5642903	F	2016-04-29T18:38:08Z
1	558997776694438	5642503	M	2016-04-29T16:08:27Z
2	4262962299951	5642549	F	2016-04-29T16:19:04Z
3	867951213174	5642828	F	2016-04-29T17:29:31Z
4	8841186448183	5642494	F	2016-04-29T16:07:23Z

	AppointmentDay	Age	Neighbourhood	Scholarship	Hipertension \
0	2016-04-29T00:00:00Z	62	JARDIM DA PENHA	0	1
1	2016-04-29T00:00:00Z	56	JARDIM DA PENHA	0	0
2	2016-04-29T00:00:00Z	62	MATA DA PRAIA	0	0
3	2016-04-29T00:00:00Z	8	PONTAL DE CAMBURI	0	0
4	2016-04-29T00:00:00Z	56	JARDIM DA PENHA	0	1

	Diabetes	Alcoholism	Handcap	SMS_received	No-show
0	0	0	0	0	No
1	0	0	0	0	No
2	0	0	0	0	No
3	0	0	0	0	No
4	1	0	0	0	No

I converted the patient_id to int to get rid of the decimal format

```
In [31]: cols = df.columns
new_cols = ['patient_id', 'appointment_id', 'gender', 'scheduled_day', 'appointment_day',
            'hipertension', 'diabetes', 'alcoholism', 'handcap', 'sms_received', 'no_sh
df.columns = new_cols
df.head()
```

```

Out[31]:
      patient_id  appointment_id  gender  scheduled_day \
0    29872499824296      5642903      F  2016-04-29T18:38:08Z
1    558997776694438      5642503      M  2016-04-29T16:08:27Z
2     4262962299951      5642549      F  2016-04-29T16:19:04Z
3     867951213174      5642828      F  2016-04-29T17:29:31Z
4     8841186448183      5642494      F  2016-04-29T16:07:23Z

      appointment_day  age  neighbourhood  scholarship  hypertension \
0  2016-04-29T00:00:00Z   62    JARDIM DA PENHA           0           1
1  2016-04-29T00:00:00Z   56    JARDIM DA PENHA           0           0
2  2016-04-29T00:00:00Z   62      MATA DA PRAIA           0           0
3  2016-04-29T00:00:00Z    8  PONTAL DE CAMBURI           0           0
4  2016-04-29T00:00:00Z   56    JARDIM DA PENHA           0           1

      diabetes  alcoholism  handicap  sms_received  no_show
0           0           0           0           0       No
1           0           0           0           0       No
2           0           0           0           0       No
3           0           0           0           0       No
4           1           0           0           0       No

```

Changed the column names to make sure its all following the same pattern and causes no issues in futher processing

```

In [32]: #changing cols [6:] to boolean
df.scholarship = df.scholarship.astype(bool,inplace =True)
df.hipertension = df.hipertension.astype(bool,inplace = True)
df.diabetes = df.diabetes.astype(bool, inplace = True)
df.alcoholism = df.alcoholism.astype(bool, inplace = True)
df.sms_received = df.sms_received.astype(bool,inplace = True)
df.head()

```

```

Out[32]:
      patient_id  appointment_id  gender  scheduled_day \
0    29872499824296      5642903      F  2016-04-29T18:38:08Z
1    558997776694438      5642503      M  2016-04-29T16:08:27Z
2     4262962299951      5642549      F  2016-04-29T16:19:04Z
3     867951213174      5642828      F  2016-04-29T17:29:31Z
4     8841186448183      5642494      F  2016-04-29T16:07:23Z

      appointment_day  age  neighbourhood  scholarship  hypertension \
0  2016-04-29T00:00:00Z   62    JARDIM DA PENHA      False        True
1  2016-04-29T00:00:00Z   56    JARDIM DA PENHA      False        False
2  2016-04-29T00:00:00Z   62      MATA DA PRAIA      False        False
3  2016-04-29T00:00:00Z    8  PONTAL DE CAMBURI      False        False
4  2016-04-29T00:00:00Z   56    JARDIM DA PENHA      False        True

      diabetes  alcoholism  handicap  sms_received  no_show
0      False      False           0          False       No

```

1	False	False	0	False	No
2	False	False	0	False	No
3	False	False	0	False	No
4	True	False	0	False	No

Changed the columns with '1' and '0' to boolean for more consistency in the DataSet

```
In [33]: df.no_show.replace('Yes', True, inplace= True)

df.no_show.replace('No', False, inplace= True)
```

Changed the no_show column also to Boolean

```
In [37]: df.shape
```

```
Out[37]: (110527, 14)
```

Checking some characteristics of the DataSet

```
In [3]: df.isnull().sum()
```

```
Out[3]: Unnamed: 0      0
patient_id      0
appointment_id  0
gender          0
scheduled_day   0
appointment_day  0
age            0
neighbourhood   0
scholarship     0
hypertension    0
diabetes        0
alcoholism      0
handcap         0
sms_received    0
no_show         0
dtype: int64
```

Checking some characteristics of the DataSet

```
In [4]: df.nunique()
```

```
Out[4]: Unnamed: 0      110527
patient_id      62299
appointment_id  110527
gender          2
scheduled_day   103549
appointment_day  27
age            104
neighbourhood   81
```

```

scholarship          2
hypertension         2
diabetes             2
alcoholism           2
handicap             5
sms_received         2
no_show              2
dtype: int64

```

Checking some characteristics of the DataSet

```

In [38]: new_schd_day = []
        for day in df.scheduled_day:
            d = dt.datetime.strptime(day, '%Y-%m-%dT%H:%M:%SZ')
            new_schd_day.append(d.strftime('%Y-%m-%d %H:%M:%S'))
        new_schd_day[:5]
        df.scheduled_day = new_schd_day

```

Working on the Date and Time format to be standrized

```

In [39]: new_appt_day = []
        for day in df.appointment_day:
            d = dt.datetime.strptime(day, '%Y-%m-%dT%H:%M:%SZ')
            new_appt_day.append(d.strftime('%Y-%m-%d'))
        new_appt_day[:5]
        df.appointment_day = new_appt_day

```

Working on the Date and Time format to be standrized

```

In [40]: # df = df.drop(df.age == -1, inplace = True)
        df.drop(df[df.age == -1].index, inplace=True)
        df.reset_index(drop=True,inplace=True)

```

Dropping any record with -1 as age, eventhough it should not affect the statistics, but for age groups

```

In [41]: df_new = pd.read_csv('appt_day_less_schd_day.csv')
        to_be_removed = df_new.appointment_id
        appt_id = df.query('appointment_id in @to_be_removed').appointment_id
        df.drop(appt_id.index,inplace=True)
        df.reset_index(drop=True,inplace=True)
        df

```

```

Out[41]:
   patient_id  appointment_id  gender  scheduled_day \
0    29872499824296         5642903      F  2016-04-29 18:38:08
1    558997776694438         5642503      M  2016-04-29 16:08:27
2    4262962299951         5642549      F  2016-04-29 16:19:04
3     867951213174         5642828      F  2016-04-29 17:29:31
4     8841186448183         5642494      F  2016-04-29 16:07:23

```

5	95985133231274	5626772	F	2016-04-27 08:36:51
6	733688164476661	5630279	F	2016-04-27 15:05:12
7	3449833394123	5630575	F	2016-04-27 15:39:58
8	56394729949972	5638447	F	2016-04-29 08:02:16
9	78124564369297	5629123	F	2016-04-27 12:48:25
10	734536231958495	5630213	F	2016-04-27 14:58:11
11	7542951368435	5620163	M	2016-04-26 08:44:12
12	566654781423437	5634718	F	2016-04-28 11:33:51
13	911394617215919	5636249	M	2016-04-28 14:52:07
14	99884723334928	5633951	F	2016-04-28 10:06:24
15	99948393975	5620206	F	2016-04-26 08:47:27
16	84574392942817	5633121	M	2016-04-28 08:51:47
17	14794966191172	5633460	F	2016-04-28 09:28:57
18	17135378245248	5621836	F	2016-04-26 10:54:18
19	7223289184215	5640433	F	2016-04-29 10:43:14
20	622257462899397	5626083	F	2016-04-27 07:51:14
21	12154843752835	5628338	F	2016-04-27 10:50:45
22	863229818887631	5616091	M	2016-04-25 13:29:16
23	213753979425692	5634142	F	2016-04-28 10:27:05
24	8734857996885	5641780	F	2016-04-29 14:19:19
25	5819369978796	5624020	M	2016-04-26 15:04:17
26	25787851512	5641781	F	2016-04-29 14:19:42
27	12154843752835	5628345	F	2016-04-27 10:51:45
28	5926171692527	5642400	M	2016-04-29 15:48:02
29	1225776163665	5642186	F	2016-04-29 15:16:29
...
110491	793589177751417	5757745	M	2016-06-01 09:46:33
110492	94336536145654	5787655	F	2016-06-08 10:21:14
110493	821969177626116	5757697	F	2016-06-01 09:42:56
110494	443438443334614	5787233	F	2016-06-08 09:35:13
110495	454425189389	5758133	M	2016-06-01 10:19:12
110496	731622885364982	5787937	F	2016-06-08 10:50:42
110497	23621816822757	5759473	F	2016-06-01 13:00:36
110498	9947982555566	5788052	F	2016-06-08 11:06:21
110499	56673438855979	5758455	F	2016-06-01 10:45:50
110500	897388334326	5758779	M	2016-06-01 11:09:20
110501	476946211846992	5786918	F	2016-06-08 09:04:18
110502	94336536145654	5757656	F	2016-06-01 09:41:00
110503	495296829375937	5786750	M	2016-06-08 08:50:51
110504	23621816822757	5757587	F	2016-06-01 09:35:48
110505	823599626588	5786742	F	2016-06-08 08:50:20
110506	98762456447375	5786368	F	2016-06-08 08:20:01
110507	86747784995281	5785964	M	2016-06-08 07:52:55
110508	2695685177138	5786567	F	2016-06-08 08:35:31
110509	645634214296344	5778621	M	2016-06-06 15:58:05
110510	69237724436761	5780205	F	2016-06-07 07:45:16
110511	5574942418928	5780122	F	2016-06-07 07:38:34
110512	72633149253362	5630375	F	2016-04-27 15:15:06

110513	65423877893936	5630447	F	2016-04-27 15:23:14
110514	996997666245785	5650534	F	2016-05-03 07:51:47
110515	36355337746436	5651072	F	2016-05-03 08:23:40
110516	2572134369293	5651768	F	2016-05-03 09:15:35
110517	3596266328735	5650093	F	2016-05-03 07:27:33
110518	15576631729893	5630692	F	2016-04-27 16:03:52
110519	92134931435557	5630323	F	2016-04-27 15:09:23
110520	377511518121127	5629448	F	2016-04-27 13:30:56

	appointment_day	age	neighbourhood	scholarship	hipertension \
0	2016-04-29	62	JARDIM DA PENHA	False	True
1	2016-04-29	56	JARDIM DA PENHA	False	False
2	2016-04-29	62	MATA DA PRAIA	False	False
3	2016-04-29	8	PONTAL DE CAMBURI	False	False
4	2016-04-29	56	JARDIM DA PENHA	False	True
5	2016-04-29	76	REPÚBLICA	False	True
6	2016-04-29	23	GOIABEIRAS	False	False
7	2016-04-29	39	GOIABEIRAS	False	False
8	2016-04-29	21	ANDORINHAS	False	False
9	2016-04-29	19	CONQUISTA	False	False
10	2016-04-29	30	NOVA PALESTINA	False	False
11	2016-04-29	29	NOVA PALESTINA	False	False
12	2016-04-29	22	NOVA PALESTINA	True	False
13	2016-04-29	28	NOVA PALESTINA	False	False
14	2016-04-29	54	NOVA PALESTINA	False	False
15	2016-04-29	15	NOVA PALESTINA	False	False
16	2016-04-29	50	NOVA PALESTINA	False	False
17	2016-04-29	40	CONQUISTA	True	False
18	2016-04-29	30	NOVA PALESTINA	True	False
19	2016-04-29	46	DA PENHA	False	False
20	2016-04-29	30	NOVA PALESTINA	False	False
21	2016-04-29	4	CONQUISTA	False	False
22	2016-04-29	13	CONQUISTA	False	False
23	2016-04-29	46	CONQUISTA	False	False
24	2016-04-29	65	TABUAZEIRO	False	False
25	2016-04-29	46	CONQUISTA	False	True
26	2016-04-29	45	BENTO FERREIRA	False	True
27	2016-04-29	4	CONQUISTA	False	False
28	2016-04-29	51	SÃO PEDRO	False	False
29	2016-04-29	32	SANTA MARTHA	False	False
...
110491	2016-06-01	76	MARIA ORTIZ	False	False
110492	2016-06-08	59	MARIA ORTIZ	False	False
110493	2016-06-01	66	MARIA ORTIZ	False	True
110494	2016-06-08	59	MARIA ORTIZ	False	False
110495	2016-06-01	44	MARIA ORTIZ	False	False
110496	2016-06-08	22	GOIABEIRAS	False	False
110497	2016-06-01	64	SOLON BORGES	False	False

110498	2016-06-08	4	MARIA ORTIZ	False	False
110499	2016-06-01	55	MARIA ORTIZ	False	False
110500	2016-06-01	5	MARIA ORTIZ	False	False
110501	2016-06-08	0	MARIA ORTIZ	False	False
110502	2016-06-01	59	MARIA ORTIZ	False	False
110503	2016-06-08	33	MARIA ORTIZ	False	False
110504	2016-06-01	64	OLON BORGES	False	False
110505	2016-06-08	14	MARIA ORTIZ	False	False
110506	2016-06-08	41	MARIA ORTIZ	False	False
110507	2016-06-08	2	ANTÔNIO HONÓRIO	False	False
110508	2016-06-08	58	MARIA ORTIZ	False	False
110509	2016-06-08	33	MARIA ORTIZ	False	True
110510	2016-06-08	37	MARIA ORTIZ	False	False
110511	2016-06-07	19	MARIA ORTIZ	False	False
110512	2016-06-07	50	MARIA ORTIZ	False	False
110513	2016-06-07	22	MARIA ORTIZ	False	False
110514	2016-06-07	42	MARIA ORTIZ	False	False
110515	2016-06-07	53	MARIA ORTIZ	False	False
110516	2016-06-07	56	MARIA ORTIZ	False	False
110517	2016-06-07	51	MARIA ORTIZ	False	False
110518	2016-06-07	21	MARIA ORTIZ	False	False
110519	2016-06-07	38	MARIA ORTIZ	False	False
110520	2016-06-07	54	MARIA ORTIZ	False	False

	diabetes	alcoholism	handcap	sms_received	no_show
0	False	False	0	False	False
1	False	False	0	False	False
2	False	False	0	False	False
3	False	False	0	False	False
4	True	False	0	False	False
5	False	False	0	False	False
6	False	False	0	False	True
7	False	False	0	False	True
8	False	False	0	False	False
9	False	False	0	False	False
10	False	False	0	False	False
11	False	False	0	True	True
12	False	False	0	False	False
13	False	False	0	False	False
14	False	False	0	False	False
15	False	False	0	True	False
16	False	False	0	False	False
17	False	False	0	False	True
18	False	False	0	True	False
19	False	False	0	False	False
20	False	False	0	False	True
21	False	False	0	False	True
22	False	False	0	True	True

23	False	False	0	False	False
24	False	False	0	False	False
25	False	False	0	True	False
26	False	False	0	False	False
27	False	False	0	False	False
28	False	False	0	False	False
29	False	False	0	False	False
...
110491	False	False	0	False	False
110492	False	False	0	False	False
110493	True	False	0	False	False
110494	False	False	0	False	False
110495	False	False	0	False	False
110496	False	False	0	False	False
110497	False	False	0	False	False
110498	False	False	0	False	False
110499	False	False	0	False	False
110500	False	False	0	False	False
110501	False	False	0	False	False
110502	False	False	0	False	False
110503	False	False	0	False	False
110504	False	False	0	False	False
110505	False	False	0	False	False
110506	False	False	0	False	False
110507	False	False	0	False	False
110508	False	False	0	False	False
110509	False	False	0	False	True
110510	False	False	0	False	True
110511	False	False	0	False	False
110512	False	False	0	True	False
110513	False	False	0	True	False
110514	False	False	0	True	False
110515	False	False	0	True	False
110516	False	False	0	True	False
110517	False	False	0	True	False
110518	False	False	0	True	False
110519	False	False	0	True	False
110520	False	False	0	True	False

[110521 rows x 14 columns]

Cleaning up records with scheduled_day > appointmet_day

In [43]: *#adding time delta 'wait_time' column to the Data Set*

```
appt_day = pd.to_datetime(df.appointment_day)
schd_day = pd.to_datetime(df.scheduled_day)
```

```
df['wait_time'] = appt_day.dt.date - schd_day.dt.date

df.wait_time = df.wait_time.astype('timedelta64[D]', inplace=True)
df.wait_time = df.wait_time.astype(int)
df.head(100)
```

```
Out[43]:
```

	patient_id	appointment_id	gender	scheduled_day \
0	29872499824296	5642903	F	2016-04-29 18:38:08
1	558997776694438	5642503	M	2016-04-29 16:08:27
2	4262962299951	5642549	F	2016-04-29 16:19:04
3	867951213174	5642828	F	2016-04-29 17:29:31
4	8841186448183	5642494	F	2016-04-29 16:07:23
5	95985133231274	5626772	F	2016-04-27 08:36:51
6	733688164476661	5630279	F	2016-04-27 15:05:12
7	3449833394123	5630575	F	2016-04-27 15:39:58
8	56394729949972	5638447	F	2016-04-29 08:02:16
9	78124564369297	5629123	F	2016-04-27 12:48:25
10	734536231958495	5630213	F	2016-04-27 14:58:11
11	7542951368435	5620163	M	2016-04-26 08:44:12
12	566654781423437	5634718	F	2016-04-28 11:33:51
13	911394617215919	5636249	M	2016-04-28 14:52:07
14	99884723334928	5633951	F	2016-04-28 10:06:24
15	99948393975	5620206	F	2016-04-26 08:47:27
16	84574392942817	5633121	M	2016-04-28 08:51:47
17	14794966191172	5633460	F	2016-04-28 09:28:57
18	17135378245248	5621836	F	2016-04-26 10:54:18
19	7223289184215	5640433	F	2016-04-29 10:43:14
20	622257462899397	5626083	F	2016-04-27 07:51:14
21	12154843752835	5628338	F	2016-04-27 10:50:45
22	863229818887631	5616091	M	2016-04-25 13:29:16
23	213753979425692	5634142	F	2016-04-28 10:27:05
24	8734857996885	5641780	F	2016-04-29 14:19:19
25	5819369978796	5624020	M	2016-04-26 15:04:17
26	25787851512	5641781	F	2016-04-29 14:19:42
27	12154843752835	5628345	F	2016-04-27 10:51:45
28	5926171692527	5642400	M	2016-04-29 15:48:02
29	1225776163665	5642186	F	2016-04-29 15:16:29
...
70	67144894855774	5552914	M	2016-04-06 17:59:58
71	1846317738622	5552936	F	2016-04-06 18:12:55
72	7746485718662	5638014	F	2016-04-29 07:37:37
73	45421316129453	5552934	F	2016-04-06 18:12:38
74	9672968175572	5597628	F	2016-04-18 17:29:12
75	148894173528	5597632	F	2016-04-18 17:32:53
76	6549277227425	5597643	M	2016-04-18 17:40:18
77	5753721241256	5642767	F	2016-04-29 17:06:27
78	625926531749	5597672	M	2016-04-18 17:52:49
79	99128824246583	5597673	M	2016-04-18 17:53:25

80	1486714718477	5597685	M	2016-04-18 18:03:12
81	19767951968224	5597689	F	2016-04-18 18:06:35
82	182712485992	5638939	M	2016-04-29 08:36:19
83	227497896765	5637742	M	2016-04-29 07:20:46
84	26879963992389	5637915	F	2016-04-29 07:31:04
85	74727351113223	5623102	F	2016-04-26 13:34:14
86	3376224477447	5595347	M	2016-04-18 12:31:34
87	4143141735632	5595356	M	2016-04-18 12:32:25
88	4448345555999	5595358	M	2016-04-18 12:32:35
89	431493164159576	5640380	M	2016-04-29 10:37:02
90	878252996786747	5595362	M	2016-04-18 12:33:05
91	2294295126913	5598651	F	2016-04-19 07:51:31
92	295467429931514	5638591	M	2016-04-29 08:11:38
93	63225327996426	5639376	F	2016-04-29 09:01:10
94	8192146244379	5640054	M	2016-04-29 10:03:12
95	198624862183842	5640307	M	2016-04-29 10:28:54
96	79376248773989	5623692	M	2016-04-26 14:28:39
97	5253342488842	5565493	F	2016-04-11 09:00:00
98	372596436556933	5571906	F	2016-04-12 09:44:42
99	124621344153	5641893	F	2016-04-29 14:38:28

	appointment_day	age	neighbourhood	scholarship	hipertension \
0	2016-04-29	62	JARDIM DA PENHA	False	True
1	2016-04-29	56	JARDIM DA PENHA	False	False
2	2016-04-29	62	MATA DA PRAIA	False	False
3	2016-04-29	8	PONTAL DE CAMBURI	False	False
4	2016-04-29	56	JARDIM DA PENHA	False	True
5	2016-04-29	76	REPÚBLICA	False	True
6	2016-04-29	23	GOIABEIRAS	False	False
7	2016-04-29	39	GOIABEIRAS	False	False
8	2016-04-29	21	ANDORINHAS	False	False
9	2016-04-29	19	CONQUISTA	False	False
10	2016-04-29	30	NOVA PALESTINA	False	False
11	2016-04-29	29	NOVA PALESTINA	False	False
12	2016-04-29	22	NOVA PALESTINA	True	False
13	2016-04-29	28	NOVA PALESTINA	False	False
14	2016-04-29	54	NOVA PALESTINA	False	False
15	2016-04-29	15	NOVA PALESTINA	False	False
16	2016-04-29	50	NOVA PALESTINA	False	False
17	2016-04-29	40	CONQUISTA	True	False
18	2016-04-29	30	NOVA PALESTINA	True	False
19	2016-04-29	46	DA PENHA	False	False
20	2016-04-29	30	NOVA PALESTINA	False	False
21	2016-04-29	4	CONQUISTA	False	False
22	2016-04-29	13	CONQUISTA	False	False
23	2016-04-29	46	CONQUISTA	False	False
24	2016-04-29	65	TABUAZEIRO	False	False
25	2016-04-29	46	CONQUISTA	False	True

26	2016-04-29	45	BENTO FERREIRA	False	True
27	2016-04-29	4	CONQUISTA	False	False
28	2016-04-29	51	SÃO PEDRO	False	False
29	2016-04-29	32	SANTA MARTHA	False	False
..
70	2016-04-29	62	SOLON BORGES	False	False
71	2016-04-29	30	BONFIM	True	False
72	2016-04-29	61	JARDIM CAMBURI	False	False
73	2016-04-29	68	REPÚBLICA	False	True
74	2016-04-29	64	MARIA ORTIZ	False	False
75	2016-04-29	60	JABOUR	False	False
76	2016-04-29	28	ANTÔNIO HONÓRIO	False	False
77	2016-04-29	27	JABOUR	False	False
78	2016-04-29	21	MARIA ORTIZ	False	False
79	2016-04-29	67	MARIA ORTIZ	False	False
80	2016-04-29	68	JABOUR	False	False
81	2016-04-29	49	JABOUR	False	False
82	2016-04-29	71	JABOUR	False	False
83	2016-04-29	36	RESISTÊNCIA	False	False
84	2016-04-29	29	RESISTÊNCIA	False	False
85	2016-04-29	69	RESISTÊNCIA	False	True
86	2016-04-29	10	ILHA DE SANTA MARIA	False	False
87	2016-04-29	2	ILHA DE SANTA MARIA	False	False
88	2016-04-29	1	JUCUTUQUARA	False	False
89	2016-04-29	0	MONTE BELO	False	False
90	2016-04-29	11	JUCUTUQUARA	False	False
91	2016-04-29	10	BONFIM	False	False
92	2016-04-29	2	BONFIM	False	False
93	2016-04-29	1	BONFIM	False	False
94	2016-04-29	10	BONFIM	False	False
95	2016-04-29	1	BONFIM	False	False
96	2016-04-29	3	BONFIM	False	False
97	2016-04-29	35	BONFIM	False	False
98	2016-04-29	51	BONFIM	False	False
99	2016-04-29	1	BONFIM	False	False

	diabetes	alcoholism	handcap	sms_received	no_show	wait_time
0	False	False	0	False	False	0
1	False	False	0	False	False	0
2	False	False	0	False	False	0
3	False	False	0	False	False	0
4	True	False	0	False	False	0
5	False	False	0	False	False	2
6	False	False	0	False	True	2
7	False	False	0	False	True	2
8	False	False	0	False	False	0
9	False	False	0	False	False	2
10	False	False	0	False	False	2

11	False	False	0	True	True	3
12	False	False	0	False	False	1
13	False	False	0	False	False	1
14	False	False	0	False	False	1
15	False	False	0	True	False	3
16	False	False	0	False	False	1
17	False	False	0	False	True	1
18	False	False	0	True	False	3
19	False	False	0	False	False	0
20	False	False	0	False	True	2
21	False	False	0	False	True	2
22	False	False	0	True	True	4
23	False	False	0	False	False	1
24	False	False	0	False	False	0
25	False	False	0	True	False	3
26	False	False	0	False	False	0
27	False	False	0	False	False	2
28	False	False	0	False	False	0
29	False	False	0	False	False	0
..
70	False	False	0	False	False	23
71	False	False	0	True	False	23
72	False	False	0	False	False	0
73	True	False	0	True	False	23
74	False	False	0	True	False	11
75	False	False	0	False	False	11
76	False	False	0	False	True	11
77	False	False	0	False	False	0
78	False	False	0	True	False	11
79	False	False	0	True	True	11
80	False	False	0	True	False	11
81	False	False	0	True	False	11
82	False	False	0	False	False	0
83	False	False	0	False	False	0
84	False	False	0	False	False	0
85	False	False	0	False	False	3
86	False	False	0	True	False	11
87	False	False	0	False	False	11
88	False	False	0	False	False	11
89	False	False	0	False	False	0
90	False	False	0	True	True	11
91	False	False	0	True	False	10
92	False	False	0	False	False	0
93	False	False	0	False	False	0
94	False	False	0	False	False	0
95	False	False	0	False	False	0
96	False	False	0	True	False	3
97	False	False	0	True	False	18

98	False	False	0	True	False	17
99	False	False	0	False	False	0

[100 rows x 15 columns]

Adding another column to the dataset 'wait_time

```
In [44]: # Adding another column that has the age groups
df['age_group'] = (df.age.apply(lambda x: min(int(x / 10) , 9))
)
df.age_group.value_counts().sort_index()
type(df.age[5])
df.head(100)
```

```
Out[44]:
```

	patient_id	appointment_id	gender	scheduled_day \
0	29872499824296	5642903	F	2016-04-29 18:38:08
1	558997776694438	5642503	M	2016-04-29 16:08:27
2	4262962299951	5642549	F	2016-04-29 16:19:04
3	867951213174	5642828	F	2016-04-29 17:29:31
4	8841186448183	5642494	F	2016-04-29 16:07:23
5	95985133231274	5626772	F	2016-04-27 08:36:51
6	733688164476661	5630279	F	2016-04-27 15:05:12
7	3449833394123	5630575	F	2016-04-27 15:39:58
8	56394729949972	5638447	F	2016-04-29 08:02:16
9	78124564369297	5629123	F	2016-04-27 12:48:25
10	734536231958495	5630213	F	2016-04-27 14:58:11
11	7542951368435	5620163	M	2016-04-26 08:44:12
12	566654781423437	5634718	F	2016-04-28 11:33:51
13	911394617215919	5636249	M	2016-04-28 14:52:07
14	99884723334928	5633951	F	2016-04-28 10:06:24
15	99948393975	5620206	F	2016-04-26 08:47:27
16	84574392942817	5633121	M	2016-04-28 08:51:47
17	14794966191172	5633460	F	2016-04-28 09:28:57
18	17135378245248	5621836	F	2016-04-26 10:54:18
19	7223289184215	5640433	F	2016-04-29 10:43:14
20	622257462899397	5626083	F	2016-04-27 07:51:14
21	12154843752835	5628338	F	2016-04-27 10:50:45
22	863229818887631	5616091	M	2016-04-25 13:29:16
23	213753979425692	5634142	F	2016-04-28 10:27:05
24	8734857996885	5641780	F	2016-04-29 14:19:19
25	5819369978796	5624020	M	2016-04-26 15:04:17
26	25787851512	5641781	F	2016-04-29 14:19:42
27	12154843752835	5628345	F	2016-04-27 10:51:45
28	5926171692527	5642400	M	2016-04-29 15:48:02
29	1225776163665	5642186	F	2016-04-29 15:16:29
..
70	67144894855774	5552914	M	2016-04-06 17:59:58
71	1846317738622	5552936	F	2016-04-06 18:12:55

72	7746485718662	5638014	F	2016-04-29 07:37:37
73	45421316129453	5552934	F	2016-04-06 18:12:38
74	9672968175572	5597628	F	2016-04-18 17:29:12
75	148894173528	5597632	F	2016-04-18 17:32:53
76	6549277227425	5597643	M	2016-04-18 17:40:18
77	5753721241256	5642767	F	2016-04-29 17:06:27
78	625926531749	5597672	M	2016-04-18 17:52:49
79	99128824246583	5597673	M	2016-04-18 17:53:25
80	1486714718477	5597685	M	2016-04-18 18:03:12
81	19767951968224	5597689	F	2016-04-18 18:06:35
82	182712485992	5638939	M	2016-04-29 08:36:19
83	227497896765	5637742	M	2016-04-29 07:20:46
84	26879963992389	5637915	F	2016-04-29 07:31:04
85	74727351113223	5623102	F	2016-04-26 13:34:14
86	3376224477447	5595347	M	2016-04-18 12:31:34
87	4143141735632	5595356	M	2016-04-18 12:32:25
88	4448345555999	5595358	M	2016-04-18 12:32:35
89	431493164159576	5640380	M	2016-04-29 10:37:02
90	878252996786747	5595362	M	2016-04-18 12:33:05
91	2294295126913	5598651	F	2016-04-19 07:51:31
92	295467429931514	5638591	M	2016-04-29 08:11:38
93	63225327996426	5639376	F	2016-04-29 09:01:10
94	8192146244379	5640054	M	2016-04-29 10:03:12
95	198624862183842	5640307	M	2016-04-29 10:28:54
96	79376248773989	5623692	M	2016-04-26 14:28:39
97	5253342488842	5565493	F	2016-04-11 09:00:00
98	372596436556933	5571906	F	2016-04-12 09:44:42
99	124621344153	5641893	F	2016-04-29 14:38:28

	appointment_day	age	neighbourhood	scholarship	hipertension \
0	2016-04-29	62	JARDIM DA PENHA	False	True
1	2016-04-29	56	JARDIM DA PENHA	False	False
2	2016-04-29	62	MATA DA PRAIA	False	False
3	2016-04-29	8	PONTAL DE CAMBURI	False	False
4	2016-04-29	56	JARDIM DA PENHA	False	True
5	2016-04-29	76	REPÚBLICA	False	True
6	2016-04-29	23	GOIABEIRAS	False	False
7	2016-04-29	39	GOIABEIRAS	False	False
8	2016-04-29	21	ANDORINHAS	False	False
9	2016-04-29	19	CONQUISTA	False	False
10	2016-04-29	30	NOVA PALESTINA	False	False
11	2016-04-29	29	NOVA PALESTINA	False	False
12	2016-04-29	22	NOVA PALESTINA	True	False
13	2016-04-29	28	NOVA PALESTINA	False	False
14	2016-04-29	54	NOVA PALESTINA	False	False
15	2016-04-29	15	NOVA PALESTINA	False	False
16	2016-04-29	50	NOVA PALESTINA	False	False
17	2016-04-29	40	CONQUISTA	True	False

18	2016-04-29	30	NOVA PALESTINA	True	False
19	2016-04-29	46	DA PENHA	False	False
20	2016-04-29	30	NOVA PALESTINA	False	False
21	2016-04-29	4	CONQUISTA	False	False
22	2016-04-29	13	CONQUISTA	False	False
23	2016-04-29	46	CONQUISTA	False	False
24	2016-04-29	65	TABUAZEIRO	False	False
25	2016-04-29	46	CONQUISTA	False	True
26	2016-04-29	45	BENTO FERREIRA	False	True
27	2016-04-29	4	CONQUISTA	False	False
28	2016-04-29	51	SÃO PEDRO	False	False
29	2016-04-29	32	SANTA MARTHA	False	False
..
70	2016-04-29	62	SOLON BORGES	False	False
71	2016-04-29	30	BONFIM	True	False
72	2016-04-29	61	JARDIM CAMBURI	False	False
73	2016-04-29	68	REPÚBLICA	False	True
74	2016-04-29	64	MARIA ORTIZ	False	False
75	2016-04-29	60	JABOUR	False	False
76	2016-04-29	28	ANTÔNIO HONÓRIO	False	False
77	2016-04-29	27	JABOUR	False	False
78	2016-04-29	21	MARIA ORTIZ	False	False
79	2016-04-29	67	MARIA ORTIZ	False	False
80	2016-04-29	68	JABOUR	False	False
81	2016-04-29	49	JABOUR	False	False
82	2016-04-29	71	JABOUR	False	False
83	2016-04-29	36	RESISTÊNCIA	False	False
84	2016-04-29	29	RESISTÊNCIA	False	False
85	2016-04-29	69	RESISTÊNCIA	False	True
86	2016-04-29	10	ILHA DE SANTA MARIA	False	False
87	2016-04-29	2	ILHA DE SANTA MARIA	False	False
88	2016-04-29	1	JUCUTUQUARA	False	False
89	2016-04-29	0	MONTE BELO	False	False
90	2016-04-29	11	JUCUTUQUARA	False	False
91	2016-04-29	10	BONFIM	False	False
92	2016-04-29	2	BONFIM	False	False
93	2016-04-29	1	BONFIM	False	False
94	2016-04-29	10	BONFIM	False	False
95	2016-04-29	1	BONFIM	False	False
96	2016-04-29	3	BONFIM	False	False
97	2016-04-29	35	BONFIM	False	False
98	2016-04-29	51	BONFIM	False	False
99	2016-04-29	1	BONFIM	False	False

	diabetes	alcoholism	handcap	sms_received	no_show	wait_time	age_group
0	False	False	0	False	False	0	6
1	False	False	0	False	False	0	5
2	False	False	0	False	False	0	6

3	False	False	0	False	False	0	0
4	True	False	0	False	False	0	5
5	False	False	0	False	False	2	7
6	False	False	0	False	True	2	2
7	False	False	0	False	True	2	3
8	False	False	0	False	False	0	2
9	False	False	0	False	False	2	1
10	False	False	0	False	False	2	3
11	False	False	0	True	True	3	2
12	False	False	0	False	False	1	2
13	False	False	0	False	False	1	2
14	False	False	0	False	False	1	5
15	False	False	0	True	False	3	1
16	False	False	0	False	False	1	5
17	False	False	0	False	True	1	4
18	False	False	0	True	False	3	3
19	False	False	0	False	False	0	4
20	False	False	0	False	True	2	3
21	False	False	0	False	True	2	0
22	False	False	0	True	True	4	1
23	False	False	0	False	False	1	4
24	False	False	0	False	False	0	6
25	False	False	0	True	False	3	4
26	False	False	0	False	False	0	4
27	False	False	0	False	False	2	0
28	False	False	0	False	False	0	5
29	False	False	0	False	False	0	3
..
70	False	False	0	False	False	23	6
71	False	False	0	True	False	23	3
72	False	False	0	False	False	0	6
73	True	False	0	True	False	23	6
74	False	False	0	True	False	11	6
75	False	False	0	False	False	11	6
76	False	False	0	False	True	11	2
77	False	False	0	False	False	0	2
78	False	False	0	True	False	11	2
79	False	False	0	True	True	11	6
80	False	False	0	True	False	11	6
81	False	False	0	True	False	11	4
82	False	False	0	False	False	0	7
83	False	False	0	False	False	0	3
84	False	False	0	False	False	0	2
85	False	False	0	False	False	3	6
86	False	False	0	True	False	11	1
87	False	False	0	False	False	11	0
88	False	False	0	False	False	11	0
89	False	False	0	False	False	0	0

90	False	False	0	True	True	11	1
91	False	False	0	True	False	10	1
92	False	False	0	False	False	0	0
93	False	False	0	False	False	0	0
94	False	False	0	False	False	0	1
95	False	False	0	False	False	0	0
96	False	False	0	True	False	3	0
97	False	False	0	True	False	18	3
98	False	False	0	True	False	17	5
99	False	False	0	False	False	0	0

[100 rows x 16 columns]

Adding another column for age groups

```
In [45]: df.to_csv('modified_no_show_dataset.csv')
```

Saving all of these chnages to a new CS file to avoid editing everytime i run the sheet

1.2 Data Visulization:

```
In [46]: figure = plt.figure(figsize=(20,10))
fig_dim = (2,4)

plt.subplot2grid(fig_dim, (0,0), title='Scholarship')
df.scholarship.value_counts().plot(kind='bar')

plt.subplot2grid(fig_dim, (0,1))
df.hipertension.value_counts().plot(kind='bar', title='Hipertension')

plt.subplot2grid(fig_dim, (0,2))
df.diabetes.value_counts().plot(kind='bar', title='Diabetes')

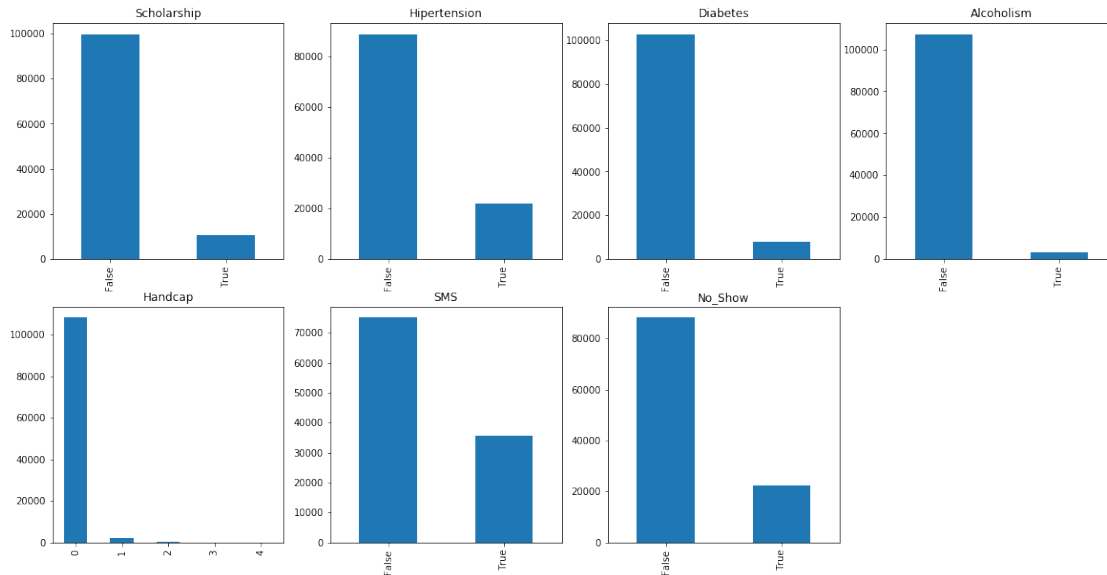
plt.subplot2grid(fig_dim, (0,3))
df.alcoholism.value_counts().plot(kind='bar', title='Alcoholism')

plt.subplot2grid(fig_dim, (1,0))
df.handcap.value_counts().plot(kind='bar', title='Handcap')

plt.subplot2grid(fig_dim, (1,1))
df.sms_received.value_counts().plot(kind='bar', title='SMS')

plt.subplot2grid(fig_dim, (1,2))
df.no_show.value_counts().plot(kind='bar', title='No_Show')

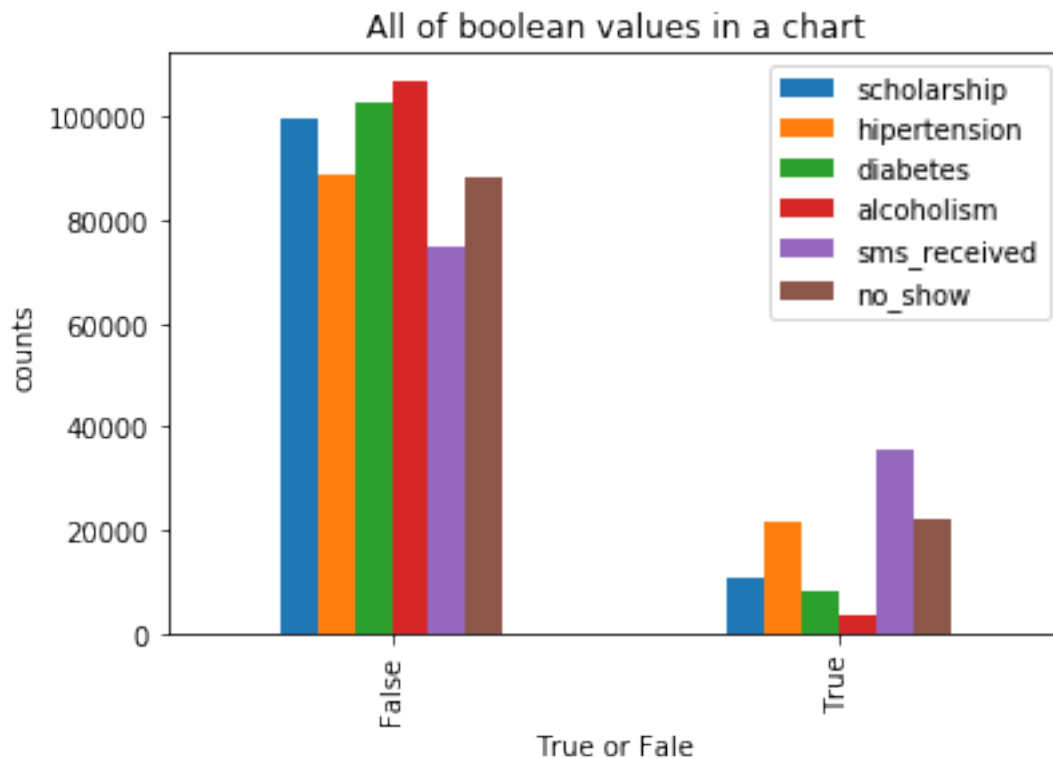
Out[46]: <matplotlib.axes._subplots.AxesSubplot at 0x7fe0171d3c50>
```



Value Counts in bar chart for all of the parameters in the dataset

```
In [15]: df[['scholarship', 'hipertension', 'diabetes', 'alcoholism', 'sms_received', 'no_show']]
plt.title('All of boolean values in a chart')
plt.ylabel('counts')
plt.xlabel('True or False')
```

```
Out[15]: Text(0.5,0,'True or False')
```



One Bar Chart for all value counts for either True or false

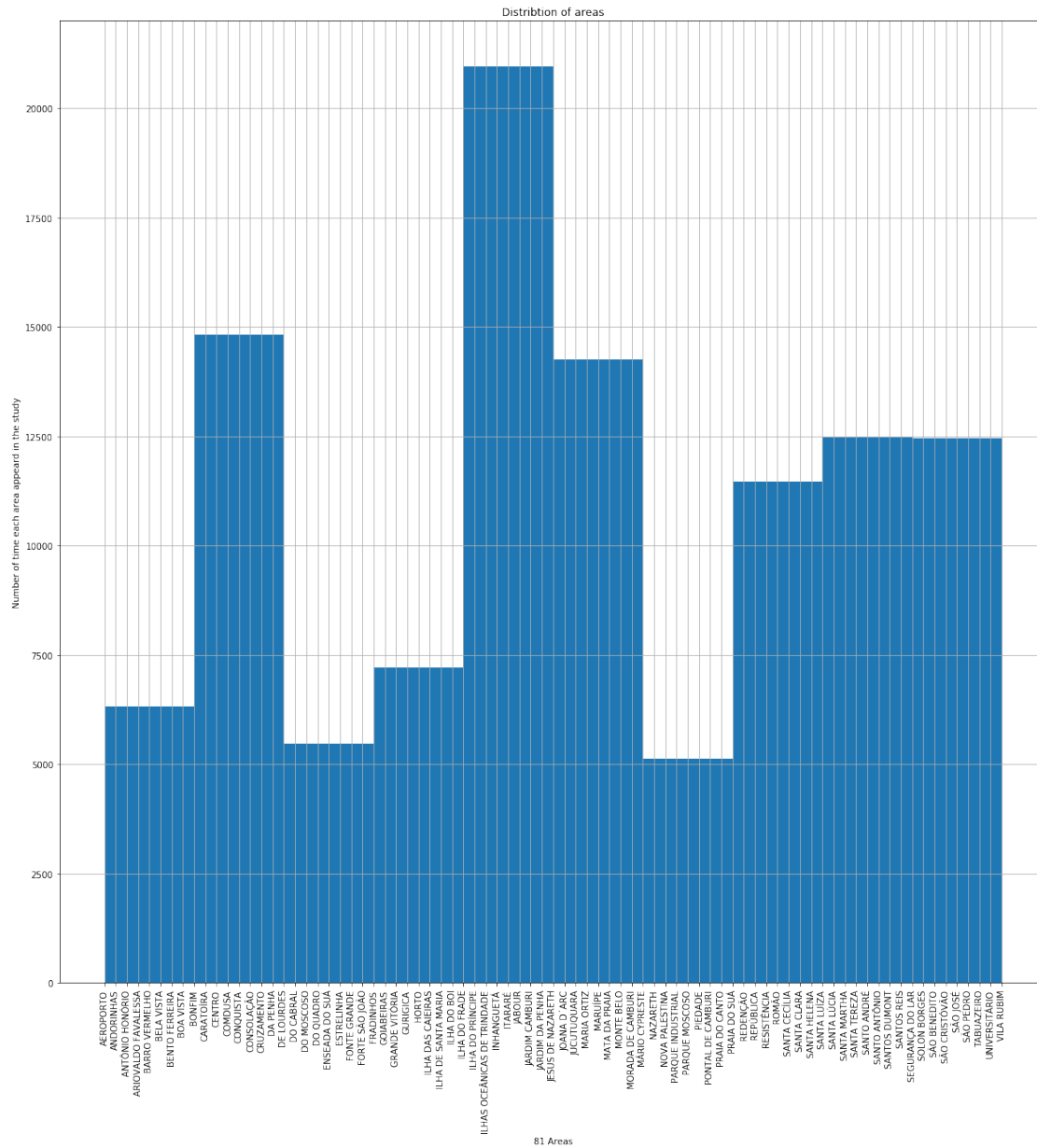
1. So the percentage of no show is around 25%
2. Number of no shows when message received is 9784 (9%)
3. Number of no shows when message wasn't received is 12535 (11.5%)
4. Number of no shows when patient is not hipertensive is 18547 (16.9%)
5. Number of no shows when patient is not on Scholarshop is 19741 (18%)
6. Number of no shows when patient is not Diabetic is 20889 (19%)
7. Number of no shows when patient is not Alcoholic is 21642 (19.7%)
8. Number of no shows when patient is not handicaped is 21912 (19.9%)

Some Statistics to allow me to see which parameter alone impact the no_show

```
In [13]: unique_area = list(df.neighbourhood.unique())
         x_area = unique_area

         df.neighbourhood.hist(figsize=(20,20))
         plt.xticks(rotation='vertical');
         plt.title('Distription of areas')
         plt.xlabel('81 Areas')
         plt.ylabel('Number of time each area appeared in the study')

Out[13]: Text(0,0.5,'Number of time each area appeared in the study')
```



```
In [24]: df.patient_id.value_counts()
```

```
Out[24]: 822145925426128      88
          99637671331          84
          26886125921145       70
          33534783483176       65
          871374938638855       62
          6264198675331         62
          258424392677          62
          75797461494159         62
```

66844879846766	57
872278549442	55
89239687393655	54
8435223536	51
853439686798559	50
14479974122994	46
65433599726992	46
9452745294842	42
81894521843749	42
188232341789524	40
2271579924275	38
9496196639835	38
13364929297498	37
1484143378533	35
986162815579582	34
88834999836575	34
712458866975343	33
6128878448536	30
416755661551767	30
81213966782532	29
8634164126317	24
36994987339512	23
...	
29739554385665	1
98683352133221	1
5394313945329	1
48689197872217	1
9675119787546	1
763619586595	1
983874124283357	1
737858311826761	1
8169988527774	1
2212945531847	1
961392519656997	1
271517596623238	1
8249496395977	1
137479426839	1
6529316371746	1
3212962263947	1
198193457888	1
1425822256863	1
216133833234618	1
37589497678822	1
31638467315	1
89279955685	1
5883897911366	1
56326578686847	1
869587212288428	1

```

735858598529      1
2886912523138      1
68129842443312      1
99264711372        1
57863365759569      1
Name: patient_id, Length: 62298, dtype: int64

```

```

In [25]: df.wait_time.value_counts()
df.query('no_show').wait_time.value_counts()

```

```

Out[25]: 0      38562
2       6725
4       5290
1       5213
7       4906
6       4037
5       3277
14      2913
3       2737
8       2332
21      1861
28      1706
13      1682
9       1605
15      1503
10      1391
20      1187
22      1173
16      1151
12      1115
17      1107
29      1089
19      1044
18      1021
27      1013
11       987
35       963
23       822
34       808
26       731
...
103        5
109        5
111        5
98         5
95         5
112        5
108        5

```

```

96          4
102         4
105         4
119         4
92          3
122         3
107         2
115         2
93          2
94          2
97          2
110         2
126         1
125         1
123         1
117         1
139         1
82          1
101         1
132         1
151         1
146         1
127         1
Name: wait_time, Length: 129, dtype: int64

```

In [9]: *#calculate the rate of wait time to see how does it affect the no_show*

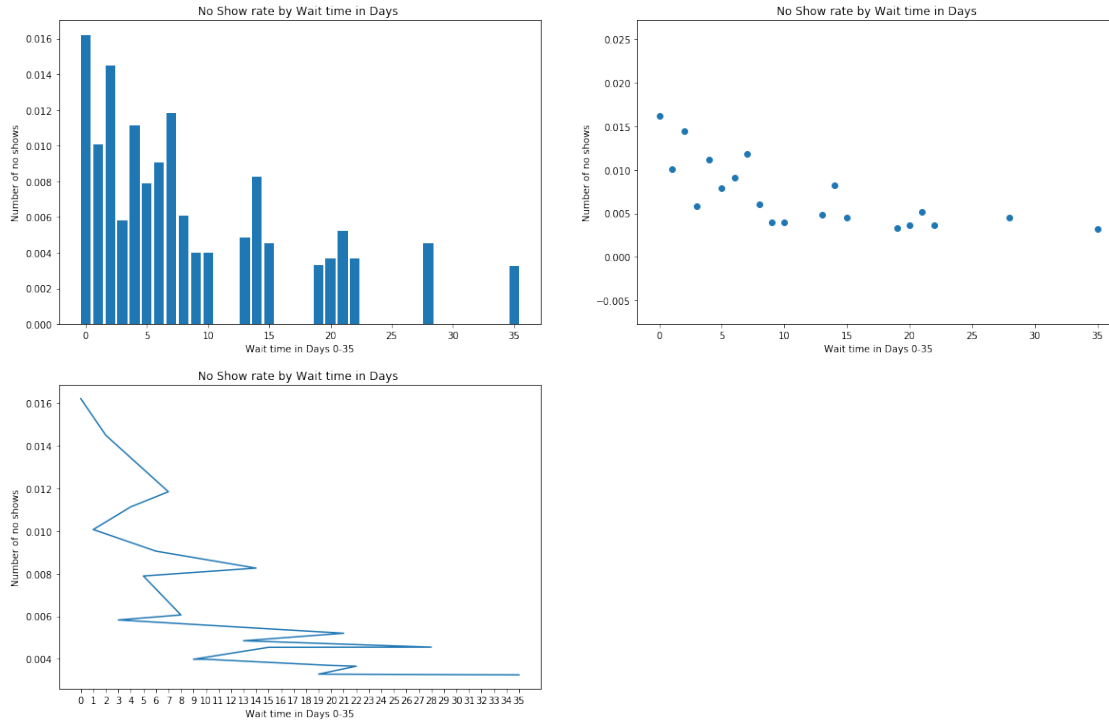
```

rate = df.query('no_show').wait_time.value_counts()[0:20] / df.shape[0]
plt.figure(figsize=(20, 20))
x=df.query('no_show').wait_time.value_counts().index[0:20]
#y=df.query('no_show').wait_time.value_counts().values[0:20]
y=rate
plt.subplot(321)
plt.bar(x,y);
plt.title('No Show rate by Wait time in Days ')
plt.xlabel('Wait time in Days 0-35')
plt.ylabel('Rate of no shows')
plt.subplot(322)
plt.scatter(x,y);
plt.title('No Show rate by Wait time in Days ')
plt.xlabel('Wait time in Days 0-35')
plt.ylabel('Rate of no shows')
plt.subplot(323)
plt.plot(x,y);
plt.xticks(range(0,36));
plt.title('No Show rate by Wait time in Days ')
plt.xlabel('Wait time in Days 0-35')
plt.ylabel('Rate of no shows')

```



```
plt.show()
```

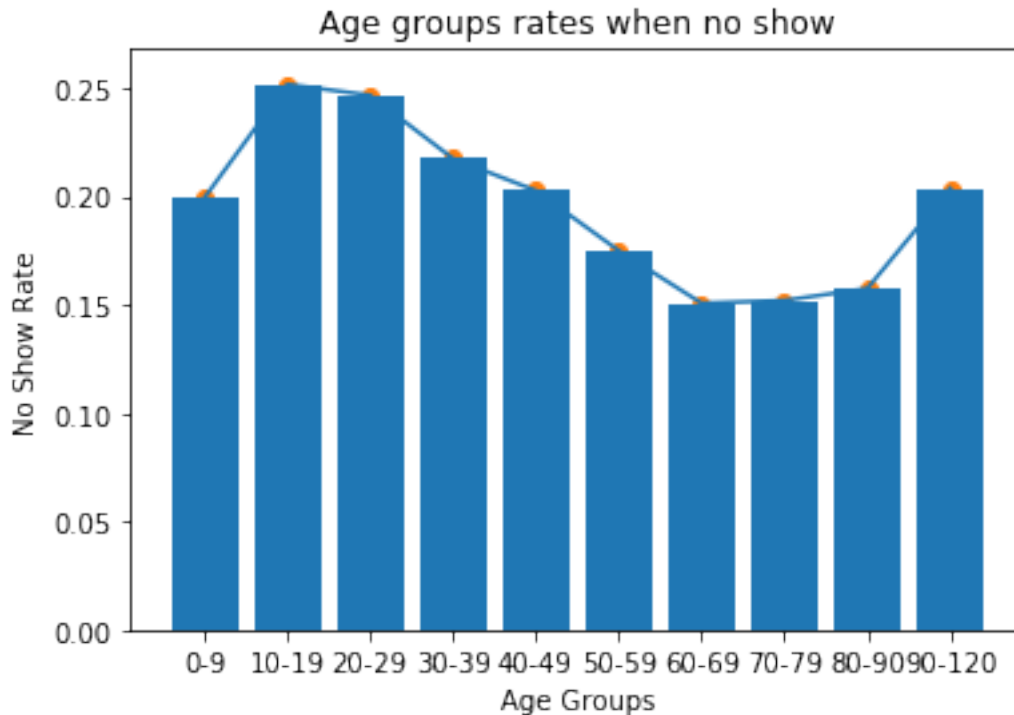


The Above plot show the wait days rate with no show and 0 wait days are the most affecting

```
In [4]: val = []
        for i in range(0,10):
            val.append((df[df.age_group == i].query('no_show').count()/df[df.age_group == i].count()))

        plt.bar([0,1,2,3,4,5,6,7,8,9], val)
        plt.scatter([0,1,2,3,4,5,6,7,8,9], val)
        plt.plot([0,1,2,3,4,5,6,7,8,9], val)
        # plt.xticks([0,1,2,3,4,5,6,7,8,9], rotation='vertical')
        labels = ['0-9', '10-19', '20-29', '30-39', '40-49', '50-59', '60-69', '70-79', '80-90', '90-120']
        plt.xticks(range(0,10), labels);
        plt.title('Age groups rates when no show')
        plt.xlabel('Age Groups')
        plt.ylabel('No Show Rate')
```

```
Out[4]: Text(0,0.5,'No Show Rate')
```



The above bar, scatter and line plot shows Age Groups showing which age group has the most no_show

Exploratory Data Analysis

Tip: Now that you've trimmed and cleaned your data, you're ready to move on to exploration. Compute statistics and create visualizations with the goal of addressing the research questions that you posed in the Introduction section. It is recommended that you be systematic with your approach. Look at one variable at a time, and then follow it up by looking at relationships between variables.

1.3 Question that my analysis answered:

1.3.1 Does the time delta (Wait Time) between the Scd_day and app_day is a reason for no show ?

1.3.2 what is the impact of the patient age on the show and no show?

Conclusions

Tip: Finally, summarize your findings and the results that have been performed. Make sure that you are clear with regards to the limitations of your exploration. If you haven't done any statistical tests, do not imply any statistical conclusions. And make sure you avoid implying causation from correlation!

Tip: Once you are satisfied with your work here, check over your report to make sure that it satisfies all the areas of the rubric (found on the project submission page at

the end of the lesson). You should also probably remove all of the "Tips" like this one so that the presentation is as polished as possible.

As a Conclusion there isn't one parameter that causes a no show, but multiple, for example wait time does have an impact and the highest impact when the wait time is 0

Another factor that I looked into which is the age of the patient, the highest numbers of no show is for ages between 10-29

1.4 Submitting your Project

Before you submit your project, you need to create a .html or .pdf version of this notebook in the workspace here. To do that, run the code cell below. If it worked correctly, you should get a return code of 0, and you should see the generated .html file in the workspace directory (click on the orange Jupyter icon in the upper left).

Alternatively, you can download this report as .html via the **File > Download as** sub-menu, and then manually upload it into the workspace directory by clicking on the orange Jupyter icon in the upper left, then using the Upload button.

Once you've done this, you can submit your project by clicking on the "Submit Project" button in the lower right here. This will create and submit a zip file with this .ipynb doc and the .html or .pdf version you created. Congratulations!

```
In [11]: from subprocess import call
         call(['python', '-m', 'nbconvert', 'Investigate_a_Dataset.ipynb'])
```

```
Out[11]: 0
```