# **Project: Examining No-show appointments Dataset**

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### Introduction

This dataset collects information from 100,000 plus medical appointments in Brazil and is focused on the question of whether the patients show up for their appointment or not. Here, we will be studying the features which will determine the response to appointments.

For analysing this data set, the very first process we will do is frame questions which we can answer from the information given in the dataset.

## **Asking Questions:**

- 1. Which age group visits the doctor more often?
- 2. Who shows up for the appoinment most of the times among males and females?
- 3.Do the patients show up for the hospitals loacted in specific areas?
- 4. What factors are important for us to know in order to predict if a patient will show up for their scheduled appointment?

```
In [82]: ▶
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

## **Data Wrangling**

After framing our questions, we need to wrangle our data to make sure all the data we need is in great quality. Three steps are involved in Data Wrangling.

### 1st Step: Gathering of Data.

```
In []:

df=pd.read_csv('C:/Users/somya/Desktop/Investigate/noshowappointments-kagglev2-may-2016.csv
df.head()
```

## 2nd Step: Assessing of Data.

We will find out problems in data's quality or structure.

```
In [84]:

df.shape

Out[84]:
(110527, 14)

In [85]:

df.duplicated().sum()

Out[85]:
0
```

In [86]: H df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 110527 entries, 0 to 110526 Data columns (total 14 columns): PatientId 110527 non-null float64 AppointmentID 110527 non-null int64 Gender 110527 non-null object ScheduledDay 110527 non-null object AppointmentDay 110527 non-null object Age 110527 non-null int64 Neighbourhood 110527 non-null object Scholarship 110527 non-null int64 Hipertension 110527 non-null int64 110527 non-null int64 Diabetes Alcoholism 110527 non-null int64 110527 non-null int64 Handcap 110527 non-null int64 SMS\_received No-show 110527 non-null object dtypes: float64(1), int64(8), object(5) memory usage: 11.8+ MB M In [87]: df.dtypes Out[87]: PatientId float64 AppointmentID int64 Gender object ScheduledDay object AppointmentDay object int64 Neighbourhood object Scholarship int64 Hipertension int64 Diabetes int64 Alcoholism int64 Handcap int64 SMS received int64 No-show object dtype: object In [ ]: Н

df.columns

In [88]: ▶

df.describe()

#### Out[88]:

	PatientId	AppointmentID	Age	Scholarship	Hipertension	Diabete
count	1.105270e+05	1.105270e+05	110527.000000	110527.000000	110527.000000	110527.00000
mean	1.474963e+14	5.675305e+06	37.088874	0.098266	0.197246	0.07186
std	2.560949e+14	7.129575e+04	23.110205	0.297675	0.397921	0.25826
min	3.921784e+04	5.030230e+06	-1.000000	0.000000	0.000000	0.00000
25%	4.172614e+12	5.640286e+06	18.000000	0.000000	0.000000	0.00000
50%	3.173184e+13	5.680573e+06	37.000000	0.000000	0.000000	0.00000
75%	9.439172e+13	5.725524e+06	55.000000	0.000000	0.000000	0.00000
max	9.999816e+14	5.790484e+06	115.000000	1.000000	1.000000	1.00000

In [89]: ▶

df.nunique()

#### Out[89]:

PatientId	62299
AppointmentID	110527
Gender	2
ScheduledDay	103549
AppointmentDay	27
Age	104
Neighbourhood	81
Scholarship	2
Hipertension	2
Diabetes	2
Alcoholism	2
Handcap	5
SMS_received	2
No-show	2
dtype: int64	

**Observations after assessment of Data-**

There are no duplicate rows in the dataset.

There are no missing values in the data.

Some of the datatypes are incorrect.

Some of the column names are wrongly spelled.

### 3rd Step: Data Cleaning

First we will drop the columns which we will not use to predict the no-show appointments.

```
In [90]:

df.drop(['PatientId','AppointmentID'],axis=1,inplace=True)

In [91]:

df = df[(df['Age'] > 0) & (df['Age'] < 95)]

df.shape

Out[91]:
(106917, 12)

> Now, we will convert the incorrect datatype of scheduled day and appointment day which is given as String to dateTime.
> Also, we observed that their is no time specified for Appointment day so we don't need time for predicting.
    Hence, we will remove time from the rows.
```

```
In [92]:

df['ScheduledDay']=pd.to_datetime(df['ScheduledDay']).dt.date
```

In [93]: ▶

```
df['AppointmentDay']=pd.to_datetime(df['AppointmentDay']).dt.date
df.head()
```

### Out[93]:

	Gender	ScheduledDay	AppointmentDay	Age	Neighbourhood	Scholarship	Hipertension	D
0	F	2016-04-29	2016-04-29	62	JARDIM DA PENHA	0	1	
1	М	2016-04-29	2016-04-29	56	JARDIM DA PENHA	0	0	
2	F	2016-04-29	2016-04-29	62	MATA DA PRAIA	0	0	
3	F	2016-04-29	2016-04-29	8	PONTAL DE CAMBURI	0	0	
4	F	2016-04-29	2016-04-29	56	JARDIM DA PENHA	0	1	

Some of the columns are spelled wrongly. We will correct them first in order to avoid confusion afterwards.

```
In [94]:

df=df.rename(columns={'Handcap': 'Handicap', 'Hipertension': 'Hypertension','No-show':'NoSh
```

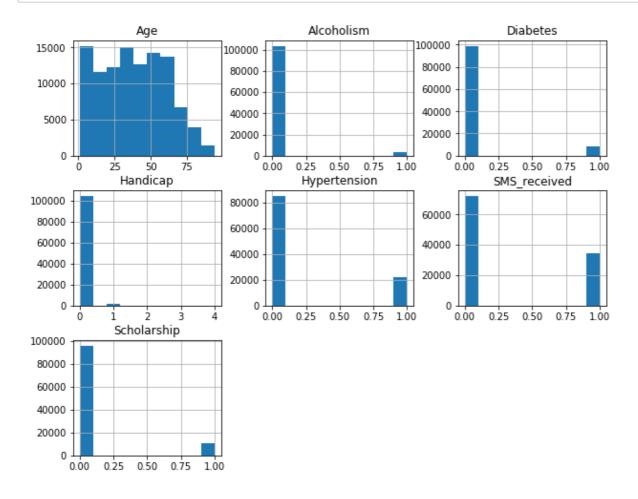
df.head(1)

#### Out[94]:

	Gender	ScheduledDay	AppointmentDay	Age	Neighbourhood	Scholarship	Hypertension	D
0	F	2016-04-29	2016-04-29	62	JARDIM DA PENHA	0	1	_ <del>_</del>

In [95]:

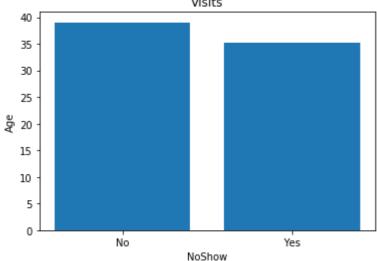
df.hist(figsize=(10,8));



# **Exploratory Data Analysis**

Research Question 1: Which age group visits the doctor more often?

```
H
In [96]:
df.groupby('NoShow').mean().Age
Out[96]:
NoShow
       39.036305
No
Yes
       35.290211
Name: Age, dtype: float64
In [97]:
                                                                                            H
plt.bar(['No', 'Yes'], [39.036, 35.290]);
plt.title('Visits')
plt.xlabel('NoShow')
plt.ylabel('Age');
                          Visits
```



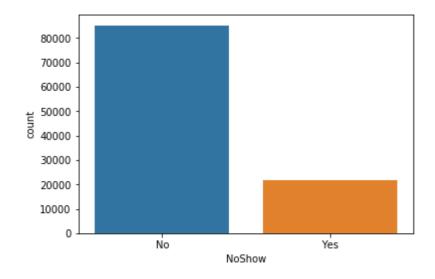
Research Question 2: Who shows up for the appointment most of the times among males and females?

In [98]: ▶

```
sns.countplot(x="NoShow", data=df)
```

#### Out[98]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1868710a9e8>



In [99]:

df['Gender'].value\_counts(normalize=True)

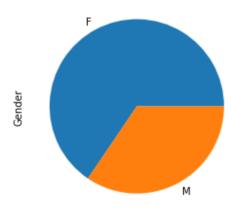
#### Out[99]:

F 0.655303 M 0.344697

Name: Gender, dtype: float64

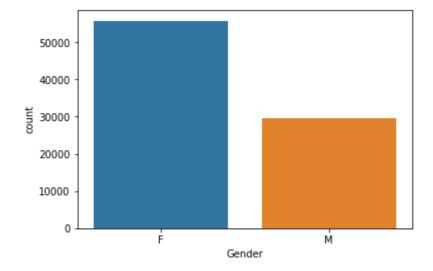
In [100]: ▶

```
df['Gender'].value_counts(normalize=True).plot(kind='pie');
```



```
In [101]:
```

```
shows_up = df[df['NoShow'] == 'No']
sns.countplot(x='Gender', data=shows_up);
```



Research Question 3: Do the patients show up for the hospitals loacted in specific areas?

In [102]:

```
shows = df[df['NoShow'] == 'No']
shows.groupby('Neighbourhood')['NoShow'].count().sort_values(ascending=False).head(10)
```

#### Out[102]:

Neighbourhood JARDIM CAMBURI 6149 MARIA ORTIZ 4367 RESISTÊNCIA 3360 JARDIM DA PENHA 3220 **CENTRO** 2586 SANTA MARTHA 2547 ITARARÉ 2514 **TABUAZEIRO** 2465 SANTO ANTÔNIO 2195 BONFIM 2161 Name: NoShow, dtype: int64

Research Question 4: What factors are important for us to know in order to predict if a patient will show up for their scheduled appointment?

```
In [104]:
```

```
df.groupby('NoShow').mean().Handicap
```

#### Out[104]:

NoShow

No 0.02339 Yes 0.02063

Name: Handicap, dtype: float64

In [105]: ▶

```
df.groupby('NoShow').mean().Scholarship
```

#### Out[105]:

NoShow

No 0.096633 Yes 0.118660

Name: Scholarship, dtype: float64

```
H
In [106]:
df.groupby('NoShow').mean().Hypertension
Out[106]:
NoShow
No
       0.211144
Yes
       0.173674
Name: Hypertension, dtype: float64
In [107]:
                                                                                            H
df.groupby('NoShow').mean().Diabetes
Out[107]:
NoShow
       0.076317
No
Yes
       0.065861
Name: Diabetes, dtype: float64
In [108]:
                                                                                            M
df.groupby('NoShow').mean().SMS_received
Out[108]:
NoShow
No
       0.293361
Yes
       0.441455
Name: SMS_received, dtype: float64
In [109]:
                                                                                            M
df.groupby('NoShow').mean().Alcoholism
Out[109]:
```

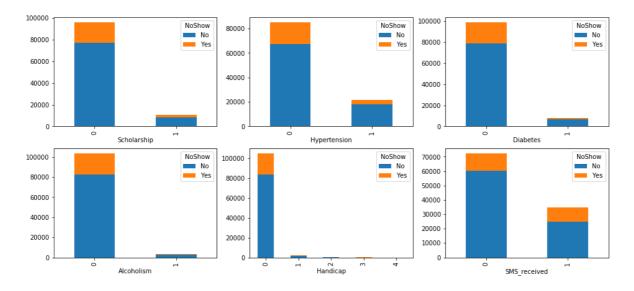
NoShow

No 0.031472 Yes 0.031246

Name: Alcoholism, dtype: float64

In [114]:

```
categories = [ 'Scholarship', 'Hypertension', 'Diabetes', 'Alcoholism', 'Handicap', 'SMS_re
fig = plt.figure(figsize=(16, 11))
for i, v in enumerate(categories):
    ax = fig.add_subplot(3, 3, i+1)
    df.groupby([v, 'NoShow'])[v].count().unstack('NoShow').plot(ax=ax, kind='bar', stacked=
```



## **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, you should save a copy of the report in HTML or PDF form via the **File** > **Download as** submenu. Before exporting your report, check over it to make sure that the flow of the report is complete. You should probably remove all of the "Tip" quotes like this one so that the presentation is as tidy as possible. Congratulations!

Examine\_Dataset

In [ ]:	M