In [1]:	import pand	_					
In [2]:	df=pd.read_				ndan/Downloads/Sav Data for Assessmen		ask – Prin
In [3]:	df.head()						
Out[3]:	Responsel	D Gender	Age	dAge	Employment_status	Relationship_status	Living_stat
	0	1 0	42	3	1	2	
	1	2 1	51	4	1	2	
	2	3 0	29	2	1	1	
	3	4 1	62	5	1	2	
	4	5 0	36	3	1	1	

5 rows × 165 columns

```
In [4]: df.shape
Out[4]: (73170, 165)
```

## **Brand Columns Analysis**

1: Do not know them|2: I have heard of them but have never used them|3: Previously used - more than 12 months ago|4: Currently use - currently or in the last 12 months

#### 1. Brand\_4

## Method 1: By using pandasql

```
In [21]: val=df['Brand_4'].value_counts().sort_values(ascending=False)
val

Out[21]: 1.0     7297
     2.0     3847
     3.0     966
     4.0     749
     Name: Brand_4, dtype: int64
```

```
In [23]:
          per = (val/sum(val))*100
          per
          1.0
                 56.746248
Out[23]:
          2.0
                 29.916790
          3.0
                  7.512248
          4.0
                  5.824714
          Name: Brand_4, dtype: float64
In [24]:
          tot=sum(val)
In [29]:
          df brand 4 = pd.DataFrame({'Count':val,'Percentage':per,'Total Count':tot
          df brand 4.reset index(inplace=True)
          df brand_4.rename(columns={'index': 'Brand_4'}, inplace=True)
          df_brand_4
Out[29]:
             Brand_4 Count Percentage Total Count
          0
                  1.0
                       7297
                             56.746248
                                            12859
          1
                 2.0
                       3847
                              29.916790
                                            12859
          2
                 3.0
                       966
                               7.512248
                                            12859
          3
                 4.0
                        749
                               5.824714
                                            12859
```

#### Method 2: By using pandasql

```
In [30]:
          from pandasql import sqldf
In [32]: A=sqldf("select * from df",locals())
In [68]:
          Br4= sqldf("SELECT Brand 4,COUNT(*) AS Count Brand 4," \
                    "(COUNT(*) * 100.0 / SUM(COUNT(*)) OVER ()) AS Percentage_4, (SU
              "FROM A where Brand_4 not like 'NA' GROUP BY Brand_4 order by Percent
          Br4
Out[68]:
             Brand_4 Count_Brand_4 Percentage_4 total_Count
                              7297
          0
                 1.0
                                       56.746248
                                                      12859
          1
                 2.0
                              3847
                                       29.916790
                                                      12859
          2
                 3.0
                               966
                                         7.512248
                                                      12859
          3
                 4.0
                               749
                                        5.824714
                                                      12859
```

#### 2. Brand\_13

Out[69]:		Brand_13	Count_Brand_13	Percentage_13	total_Count
	0	2.0	6406	49.805629	12862
	1	3.0	3015	23.441144	12862
	2	4.0	2149	16.708132	12862
	3	1.0	1292	10.045094	12862

#### 3. Brand\_82

Out[70]:		Brand_82	Count_Brand_82	Percentage_82	total_Count
	0	2.0	6763	52.630350	12850
	1	3.0	2471	19.229572	12850
	2	1.0	1912	14.879377	12850
	3	4.0	1704	13.260700	12850

#### 4. Brand\_124

Out[71]:		Brand_124	Count_Brand_124	Percentage_124	total_Count
	0	2.0	6307	48.929403	12890
	1	1.0	4330	33.591932	12890
	2	3.0	1192	9.247479	12890
	3	4.0	1061	8.231187	12890

#### 5. Brand\_128

Out[72]: Brand\_128 Count\_Brand\_128 Percentage\_128 total\_Count 2.0 7177 55.735031 12877 1 1.0 3044 23.639046 12877 11.734100 2 3.0 1511 12877 3 4.0 1145 8.891823 12877

## 6. Brand\_131

Out[73]: Brand\_131 Count\_Brand\_131 Percentage\_131 total\_Count 0 2.0 6484 50.470927 12847 1 1.0 4010 31.213513 12847 2 3.0 1453 11.310033 12847 4.0 900 7.005527 12847

## 7. Brand\_374

```
Brand_374 Count_Brand_374 Percentage_374 total_Count
Out[74]:
           0
                                      3447
                                                                   6503
                      1.0
                                                  53.006305
                     2.0
                                      2423
                                                  37.259726
                                                                   6503
           1
           2
                     3.0
                                       336
                                                   5.166846
                                                                   6503
           3
                     4.0
                                       297
                                                   4.567123
                                                                   6503
```

```
Out[298]:
               Brand_class Percentage_4 Percentage_13 Percentage_82 Percentage_124 Percent
            0
                        1.0
                                56.746248
                                                10.045094
                                                                14.879377
                                                                                 33.591932
                                                                                                 23
                        2.0
                                29.916790
                                                49.805629
                                                               52.630350
                                                                                48.929403
                                                                                                  5!
             1
            2
                        3.0
                                  7.512248
                                                23.441144
                                                                19.229572
                                                                                  9.247479
                                                                                                  1
                        4.0
                                  5.824714
                                                16.708132
                                                                13.260700
                                                                                  8.231187
```

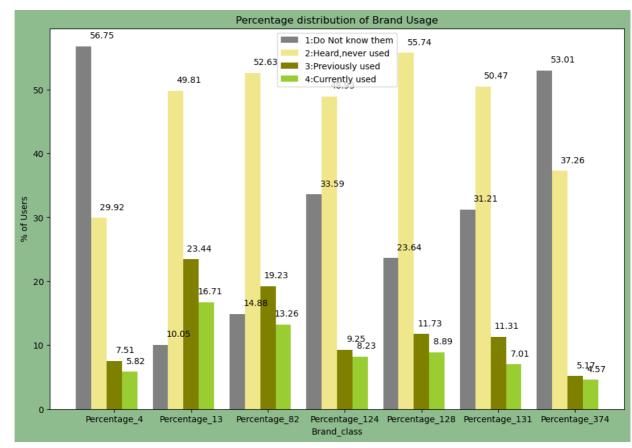
```
In [299... BrandCombined1=BrandCombined.T
    BrandCombined2=BrandCombined1.reset_index()
    BrandCombined2.columns=BrandCombined2.iloc[0]
    BrandCombined3=BrandCombined2.drop(0)
    BrandCombined3.rename(columns={1.0:'1',2.0:'2',3.0:'3',4.0:'4'},inplace=T
    BrandCombined3=BrandCombined3.round(2)
    BrandCombined3
```

```
Out[299]:
                  Brand_class
                                                3
                                                      4
                 Percentage_4 56.75
                                    29.92
                                              7.51
                                                    5.82
            2
                Percentage_13 10.05 49.81 23.44
                                                   16.71
                Percentage_82 14.88 52.63 19.23 13.26
            4 Percentage_124 33.59 48.93
                                             9.25
                                                   8.23
            5 Percentage_128 23.64 55.74
                                             11.73
                                                   8.89
            6 Percentage_131
                              31.21 50.47
                                             11.31
                                                    7.01
            7 Percentage_374 53.01 37.26
                                             5.17
                                                   4.57
```

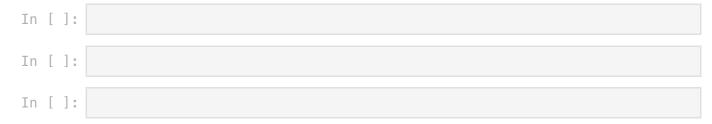
```
In [365... BrandCombined3.to_csv('usage.csv',index=False)
```

In [300... bc3=BrandCombined3

```
In [464... X axis = np.arange(len(bc3['Brand_class']))
          plt.figure(figsize=(12,8),facecolor='darkseagreen')
          plt.bar(X axis - 0.4, bc3['1'], 0.20, label = '1:Do Not know them', color=
          plt.bar(X_axis - 0.2, bc3['2'], 0.20, label = '2:Heard, never used', color=
          plt.bar(X axis + 0.0, bc3['3'], 0.20, label = '3:Previously used',color='
          plt.bar(X_axis + 0.2, bc3['4'], 0.20, label = '4:Currently used',color='y
          # Add label values to the bars
          bar width = 0.135
          for i, value in enumerate(bc3['1']):
              plt.text(i, value + 1, str(value), ha='right', va='bottom')
          for i, value in enumerate(bc3['2']):
              plt.text(i + bar width, value + 1, str(value), ha='right', va='bottom
          for i, value in enumerate(bc3['3']):
              plt.text(i + bar_width+bar_width, value + 1, str(value), ha='right',
          for i, value in enumerate(bc3['4']):
              plt.text(i + bar_width+bar_width+bar_width, value + 1, str(value), ha
          # Set labels and title
          plt.xticks(X axis, bc3['Brand class'])
          plt.xlabel("Brand class")
          plt.ylabel("% of Users")
          plt.title("Percentage distribution of Brand Usage")
          plt.legend()
          plt.show()
          # Saving the figure.
          plt.savefig("output1.jpg")
```



<Figure size 640x480 with 0 Axes>



# **Consider Columns Analysis**

```
In [160... df_c=df
In [161... df_c=df_c.iloc[:,17:24]
df_c
```

Out[161]:		Consider_general_4	Consider_general_13	Consider_general_82	Consider_genera
	0	NaN	1.0	NaN	
	1	NaN	NaN	NaN	
	2	NaN	NaN	NaN	
	3	NaN	NaN	NaN	
	4	NaN	NaN	NaN	
	•••				
	73165	NaN	NaN	NaN	
	73166	NaN	NaN	NaN	
	73167	NaN	NaN	NaN	
	73168	NaN	NaN	NaN	
	73169	NaN	NaN	NaN	

73170 rows × 7 columns

```
In [162... B=sqldf("select * from df_c",locals())
```

## 1. Consider\_general\_4

Out[178]:		Consider_general_4	Count_of_Consider_general_4	Percentage_4	total_count
	0	0.0	4552	81.841064	5562
	1	1.0	1010	18.158936	5562

# 2. Consider\_general\_13

Out[180]:		Consider_general_13	Count_of_Consider_general_13	Percentage_13	total_count
	0	0.0	6767	58.487468	11570
	1	1.0	4803	41.512532	11570

## 3. Consider\_general\_82

Out[181]:		Consider_general_82	Count_of_Consider_general_82	Percentage_82	total_count
	0	0.0	6646	60.760651	10938
	1	1.0	4292	39.239349	10938

### 4. Consider\_general\_124

Out[182]:		Consider_general_124	Count_of_Consider_general_124	Percentage_124	total_count
	0	0.0	6476	75.654206	8560
	1	1.0	2084	24.345794	8560

# 5. Consider\_general\_128

Out[183]:		Consider_general_128	Count_of_Consider_general_128	Percentage_128	total_count
	0	0.0	7339	74.636428	9833
	1	1.0	2494	25.363572	9833

## 6. Consider\_general\_131

Out[184]:		Consider_general_131	Count_of_Consider_general_131	Percentage_131	total_count
	0	0.0	6829	77.277357	8837
	1	1.0	2008	22.722643	8837

### 7. Consider\_general\_374

Out[185]:		Consider_general_374	Count_of_Consider_general_374	Percentage_374	total_count
	0	0.0	2421	79.221204	3056
	1	1.0	635	20.778796	3056

#### Combined

```
In [192...
          cb=sqldf("Select Consider general 4 as Consider type, Percentage 4, Perce
                    ",Percentage 124,Percentage 128,Percentage 131,Percentage 374"
                    " from Cr4 inner join Cr13 on Cr4.Consider_general_4=Cr13.Consid
                    " inner join Cr82 on Cr13.Consider_general_13=Cr82.Consider_gene
                   " inner join Cr124 on Cr82.Consider_general_82=Cr124.Consider_ge
                    " inner join Cr128 on Cr124.Consider general 124=Cr128.Consider
                    " inner join Cr131 on Cr128.Consider general 128=Cr131.Consider
                    " inner join Cr374 on Cr131.Consider general 131=Cr374.Consider
                    ,locals())
          cb
Out[192]:
              Consider_type Percentage_4 Percentage_13 Percentage_82 Percentage_124
           0
                       0.0
                              81.841064
                                            58.487468
                                                          60.760651
                                                                        75.654206
                       1.0
                              18.158936
                                            41.512532
                                                          39.239349
                                                                        24.345794
In [193...
          type(cb)
           pandas.core.frame.DataFrame
Out[193]:
In [218...
          cb1=cb.T
          cb1
                                 0
                                            1
Out[218]:
```

79.221204

Percentage\_374

```
In [263... import matplotlib.pyplot as plt import numpy as np
```

20.778796

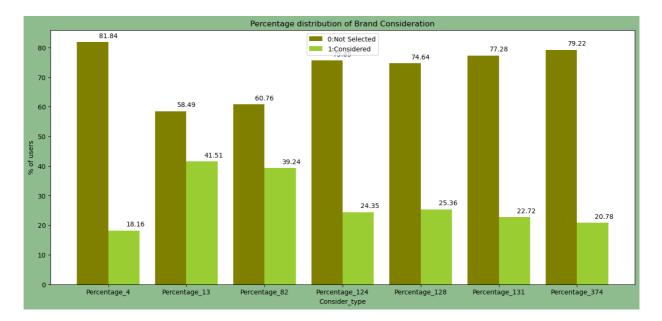
```
In [265... cb2=cb1.reset_index()
    cb2.columns=cb2.iloc[0]
    cb2=cb2.drop(0)
    cb2.rename(columns={0.0:'0:Not Selected',1.0:'1:Considered'},inplace=True
    cb2=cb2.round(2)
    cb2
```

#### Out [265]: Consider\_type O:Not Selected 1:Considered

1	Percentage_4	81.84	18.16
2	Percentage_13	58.49	41.51
3	Percentage_82	60.76	39.24
4	Percentage_124	75.65	24.35
5	Percentage_128	74.64	25.36
6	Percentage_131	77.28	22.72
7	Percentage_374	79.22	20.78

```
In [368... cb2.to_csv('brand_consideration.csv',index=False)
```

```
In [465... X_axis = np.arange(len(cb2['Consider_type']))
         plt.figure(figsize=(16,7),facecolor='darkseagreen')
         plt.bar(X axis - 0.2, cb2['0:Not Selected'], 0.4, label = '0:Not Selected
         plt.bar(X_axis + 0.2, cb2['1:Considered'], 0.4, label = '1:Considered',co
          # Add label values to the bars
         bar_width = 0.35
         for i, value in enumerate(cb2['0:Not Selected']):
             plt.text(i, value + 1, str(value), ha='center', va='bottom')
         for i, value in enumerate(cb2['1:Considered']):
              plt.text(i + bar_width, value + 1, str(value), ha='center', va='botto
         # Set labels and title
         plt.xticks(X axis, cb2['Consider type'])
         plt.xlabel("Consider_type")
         plt.ylabel("% of users")
         plt.title("Percentage distribution of Brand Consideration")
         plt.legend()
         plt.show()
```



### **Image Analysis**

```
In [468...
          im 4=df
          cols=im 4.columns.tolist()
In [469...
         #image brnd 4
          my_new_dict={}
          my_new_dict['0']='None'
          my new dict['1']='Trustworthy'
          my new dict['2']='Knowledgeble'
          my new dict['3']='Friendly'
          my_new_dict['4']='Expert'
          my_new_dict['5']='Fun/Entertaining'
          my_new_dict['6']='Innovative'
          my new_dict['7']='Helpful'
          my_new_dict['8']='Accurate'
          my new dict['9']='Caring'
          my_new_dict['10']='Gets me a good deal'
          my_new_dict['11']='Upmarket'
          my_new_dict['12']='A market leader'
          my new dict['13']='Straight-forward'
          my new dict['14']='Dependable'
          my_new_dict['15']='For people like me'
          my new dict['16']='Approachable'
          my_new_dict['17']='On my side'
          my_new_dict['18']='Inspires confidence'
          my_new_dict['19']='Socially responsible'
```

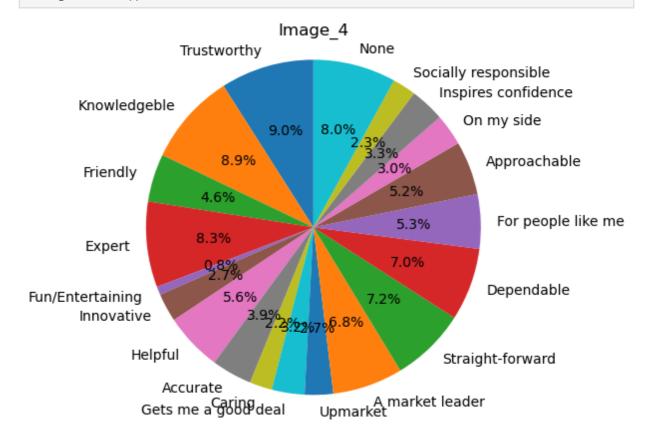
```
In [470...

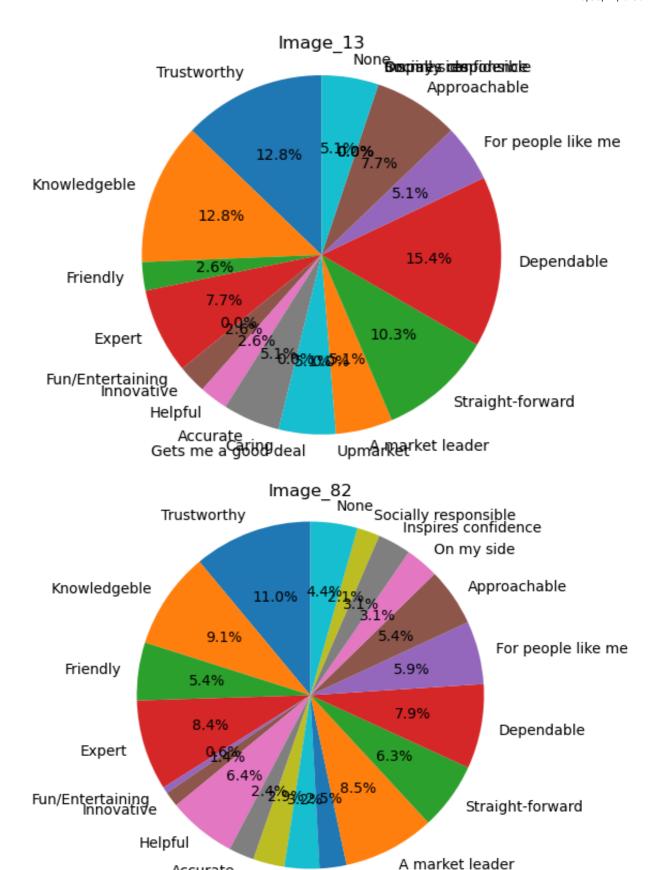
my_str= ['Image_4','Image_13','Image_82','Image_124','Image_128','Image_1
final_dict={}

for i in my_str:
    x=[j for j in cols if i in j]
    image_brnd=df[x]
    image_brnd.dropna(inplace=True)
    image_brnd=image_brnd.astype(int)
    temp_dict={}
    for k in image_brnd.columns.to_list():
        temp=k.rpartition('_')[-1]
        value=image_brnd.apply(lambda x: x.value_counts(normalize=True)).fill
        temp_dict[temp]=value
        final_dict[i]=temp_dict
```

```
/var/folders/4m/mk53z65s6978q3spjfhlw2x80000gn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
  image brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978g3spjfhlw2x80000gn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
 image_brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978g3spjfhlw2x80000gn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
 image_brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978g3spjfhlw2x80000gn/T/ipykernel_19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
  image brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978q3spjfhlw2x80000qn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
 image brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978g3spjfhlw2x80000gn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user guide/indexing.html#returning-a-view-versus-a-copy
  image brnd.dropna(inplace=True)
/var/folders/4m/mk53z65s6978g3spjfhlw2x80000gn/T/ipykernel 19446/52515004
6.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-do
cs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
 image brnd.dropna(inplace=True)
```

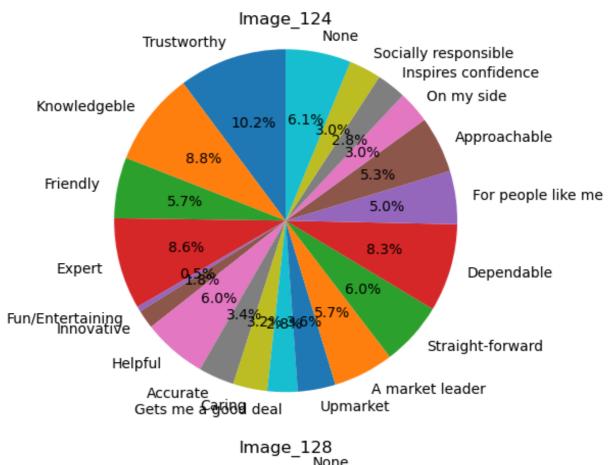
```
In [471...
for i in final_dict.keys():
    #print(i)
    j={my_new_dict[k]: v for k, v in final_dict[i].items()}
    #print(j)
    labels = list(j.keys())
    values = list(j.values())
    fig, ax = plt.subplots()
    ax.pie(values, labels=labels, autopct="%1.1f%%", startangle=90)
    plt.title(i)
    ax.axis('equal')
    plt.show()
```

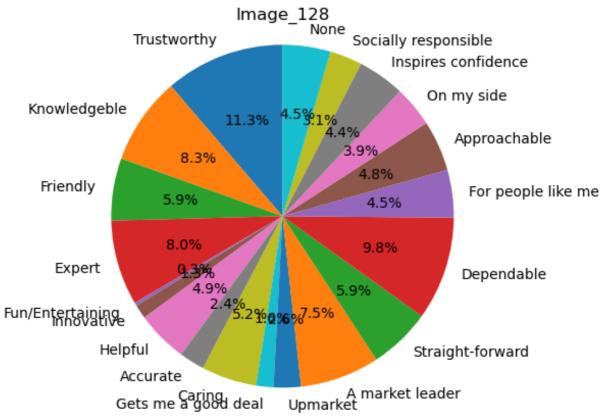


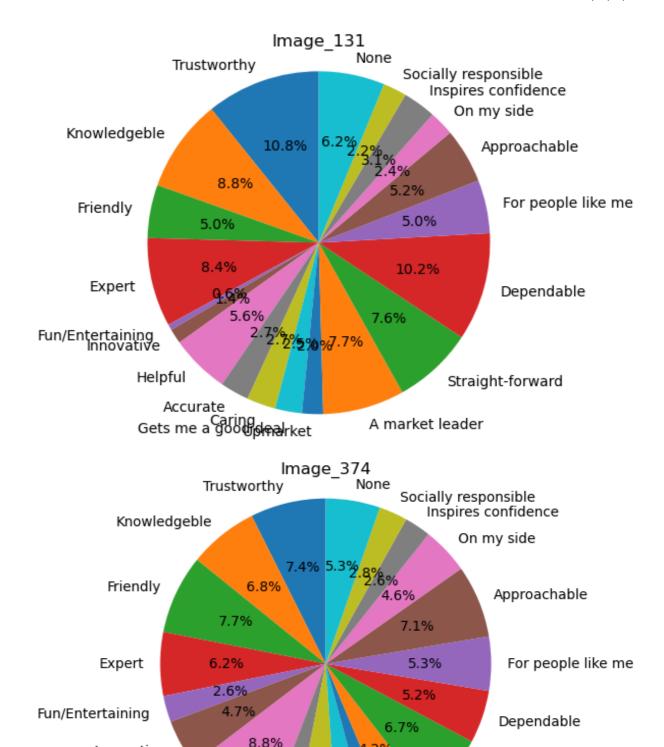


Upmarket

Accurate Gets me a good deal







# Overall Analysis Brand 13 vs Brand 131

2.7%2%1%

In [499... oa1=df

Accurate Caring

Innovative

Helpful

Straight-forward

A market leader

Gets me a good deal

```
In [500... oa2=sqldf("select * from oa1",locals())
```

#### **Brand 13**

```
oa_br13=sqldf("select * from oa2 where Consider_general_13=1",locals())
In [501...
           oa br13['Consider general 13'].value counts().sort values(ascending=False
In [504...
           1.0
                   4803
Out [504]:
           Name: Consider general 13, dtype: int64
In [505...
           oa br13.iloc[ :, 1:10].describe()
Out [505]:
                       Gender
                                                         Employment_status
                                                                            Relationship_statu
                                       Age
                                                   dAge
            count 4803.000000
                               4803.000000
                                            4803.000000
                                                                                  4803.00000
                                                               4803.000000
                      0.476161
                                  47.558401
                                               3.479492
                                                                   2.854674
                                                                                      1.89527
            mean
                     0.499483
                                  14.614186
                                                1.211902
                                                                   2.516607
                                                                                     0.69422
              std
             min
                      0.000000
                                  16.000000
                                                1.000000
                                                                   0.000000
                                                                                     0.00000
             25%
                      0.000000
                                 36.000000
                                               3.000000
                                                                                     2.00000
                                                                   1.000000
            50%
                      0.000000
                                 49.000000
                                               3.000000
                                                                   1.000000
                                                                                     2.00000
             75%
                      1.000000
                                 60.000000
                                               5.000000
                                                                  6.000000
                                                                                     2.00000
                      1.000000
                                 74.000000
                                               5.000000
                                                                  9.000000
                                                                                     5.00000
             max
In [506...
           oa br13['Gender'].value counts().sort values(ascending=False)
                 2516
Out [506]:
                 2287
           Name: Gender, dtype: int64
In [507...
           oa_br13['Age'].describe()
           count
                      4803.000000
Out [507]:
           mean
                        47.558401
           std
                        14.614186
           min
                        16.000000
           25%
                        36.000000
           50%
                        49.000000
           75%
                        60.000000
                        74.000000
           max
           Name: Age, dtype: float64
In [508...
           oa_br13['dAge'].value_counts().sort_values(ascending=False)
```

```
1413
Out[508]:
                1289
           4
                1063
                 738
           2
           1
                 300
           Name: dAge, dtype: int64
In [509...
          oa br13['Employment status'].value counts().sort values(ascending=False)
                2806
           1
Out [509]:
                 829
           4
                 298
           5
                 249
           8
                 226
           2
                 110
           7
                 101
           0
                   71
                   70
           3
                   43
           Name: Employment_status, dtype: int64
In [510...
          oa br13['Relationship status'].value counts().sort values(ascending=False
                3091
           2
Out[510]:
           1
                1180
           3
                 377
           4
                 105
           5
                   38
                   12
           Name: Relationship status, dtype: int64
          oa_br13['Living_status'].value_counts().sort_values(ascending=False)
In [511...
                3027
Out [511]:
           3
                 814
           4
                 611
                 317
           2
                   34
           Name: Living status, dtype: int64
In [512...
          oa br13['dSEG'].value counts().sort values(ascending=False)
                3141
Out[512]:
           2
                1662
           Name: dSEG, dtype: int64
In [513...
          oa_br13['Personal_Income'].describe()
           count
                       4786.000000
Out[513]:
           mean
                      30572.133932
           std
                      34087.896884
                       -999.000000
           min
           25%
                      12000.000000
           50%
                      22958.500000
           75%
                      35000.000000
                     200000.000000
           Name: Personal Income, dtype: float64
```

```
In [514...
          oa br13['Household Income'].describe()
           count
                      4788.000000
Out[514]:
           mean
                     45704.093776
                     40336.372069
           std
           min
                      -999.000000
           25%
                     20000.000000
           50%
                     36000.000000
           75%
                     58518.000000
                    200000.000000
           max
           Name: Household_Income, dtype: float64
          oa_br13.corr()
In [515...
Out[515]:
```

	ResponseID	Gender	Age	dAge	Employment_status
ResponseID	1.000000	-0.011998	-0.035771	-0.031932	-0.006517
Gender	-0.011998	1.000000	0.059309	0.045893	-0.037713
Age	-0.035771	0.059309	1.000000	0.967171	0.325311
dAge	-0.031932	0.045893	0.967171	1.000000	0.301572
Employment_status	-0.006517	-0.037713	0.325311	0.301572	1.000000
Image_374_17	-0.009560	-0.438529	-0.115017	-0.099197	-0.097557
Image_374_18	0.416910	-0.233882	-0.029860	-0.012400	0.073168
Image_374_19	0.143194	-0.126878	0.140958	0.084968	-0.179662
Image_374_0	-0.218422	0.258199	-0.214881	-0.185326	-0.269251
Weighting	0.020745	0.069598	0.001178	0.004648	0.008281

165 rows × 165 columns

#### Brand 131

```
2008.000000
           count
Out[520]:
                       51,982570
           mean
                       13.536133
           std
           min
                       16.000000
           25%
                       42.000000
           50%
                       55.000000
           75%
                       63.000000
                       74.000000
           max
           Name: Age, dtype: float64
          oa_br131['dAge'].value_counts().sort_values(ascending=False)
In [521...
                731
Out [521]:
                539
           3
                 483
           2
                187
           1
                 68
           Name: dAge, dtype: int64
In [523...
          oa_br131['Employment_status'].value_counts().sort_values(ascending=False)
                1076
Out[523]:
                  448
           4
                  147
           5
                   87
           8
                   83
           2
                   58
           0
                   32
           7
                   28
                   28
           3
                   21
           Name: Employment_status, dtype: int64
          oa_br131['Relationship_status'].value_counts().sort_values(ascending=Fals
In [524...
           2
                1313
Out [524]:
                  448
           1
           3
                  172
                   51
           4
           5
                   21
                    3
           Name: Relationship status, dtype: int64
In [525...
          oa_br131['Living_status'].value_counts().sort_values(ascending=False)
                1405
           1
Out[525]:
                 277
                 211
           4
           2
                  100
                   15
           Name: Living_status, dtype: int64
          oa_br131['dSEG'].value_counts().sort_values(ascending=False)
In [526...
                1388
           1
Out[526]:
                  620
           Name: dSEG, dtype: int64
```

```
oa_br131['Personal_Income'].describe()
In [527...
                      2001.000000
           count
Out [527]:
                     31064.037981
           mean
           std
                     34665.946802
                      -999.000000
           min
           25%
                     12000.000000
           50%
                     23000.000000
           75%
                     35000.000000
                    200000.000000
           max
           Name: Personal_Income, dtype: float64
          oa_br131['Household_Income'].describe()
In [528...
                      1998.000000
           count
Out[528]:
           mean
                     45476.774274
           std
                     40375.189182
           min
                      -999.000000
           25%
                     20000.000000
           50%
                     36043.500000
           75%
                     57670.500000
           max
                    200000.000000
           Name: Household_Income, dtype: float64
```