

About Dataset

RHMCD-20 dataset, we took care to include information from a wide range of sources, including teenagers from Bangladesh, college students, housewives, professionals from businesses and corporations, and other people. This is survey data for Depression and Mental Health Data Analysis. # eracting with

Survey questions:

- Age: Represents the age of the participants.
- Gender: Indicates the gender of the participants.
- Occupation: Represents the participant's occupations.
- Days_Indoors :Indicates the number of days the participant has not been out of the house
- Growing_Stress: Indicates the participant's stress is increasing day by day (Yes/No).
- Quarantine_Frustration: Frustrations in the first two weeks of quarantine (Yes/Maybe/No).
- Changes_Habits: Represents major changes in eating habits and sleeping (Yes/Maybe/No).mo).
- Weight_Change :Highlights changes in body weight during quarantine (Yes/Maybe/No)
- Mood_Swings: Represents extreme mood changes (Low/Medium/High).
- Coping_Struggles: The inability to cope with daily problems or stress (Yes/Maybe/No).
- Work_Interest :Represents whether the participant is losing interest in working (Yes/No).
- Social_Weakness :Conveys feeling mentally weak when interacting with others (Yes/No).th others (Yes/No).

```
In [1]: import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

```
In [2]: df = pd.read_csv('mental_health_finaldata_1.csv')
```

Checking is there's NaN values and duplicates.

```
In [3]: display(df.duplicated().sum())
display(df.isna().sum())
```

0

```

Age          0
Gender       0
Occupation   0
Days_Indoors 0
Growing_Stress 0
Quarantine_Frustrations 0
Changes_Habits 0
Mental_Health_History 0
Weight_Change 0
Mood_Swings 0
Coping_Struggles 0
Work_Interest 0
Social_Weakness 0
dtype: int64

```

So, as we can see there's no dups or NaN values. Now we can see the info about DF

```
In [4]: df.info();
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 824 entries, 0 to 823
Data columns (total 13 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age                                    824 non-null    object
1   Gender                                824 non-null    object
2   Occupation                            824 non-null    object
3   Days_Indoors                          824 non-null    object
4   Growing_Stress                       824 non-null    object
5   Quarantine_Frustrations                824 non-null    object
6   Changes_Habits                        824 non-null    object
7   Mental_Health_History                  824 non-null    object
8   Weight_Change                         824 non-null    object
9   Mood_Swings                          824 non-null    object
10  Coping_Struggles                      824 non-null    object
11  Work_Interest                         824 non-null    object
12  Social_Weakness                       824 non-null    object
dtypes: object(13)
memory usage: 83.8+ KB

```

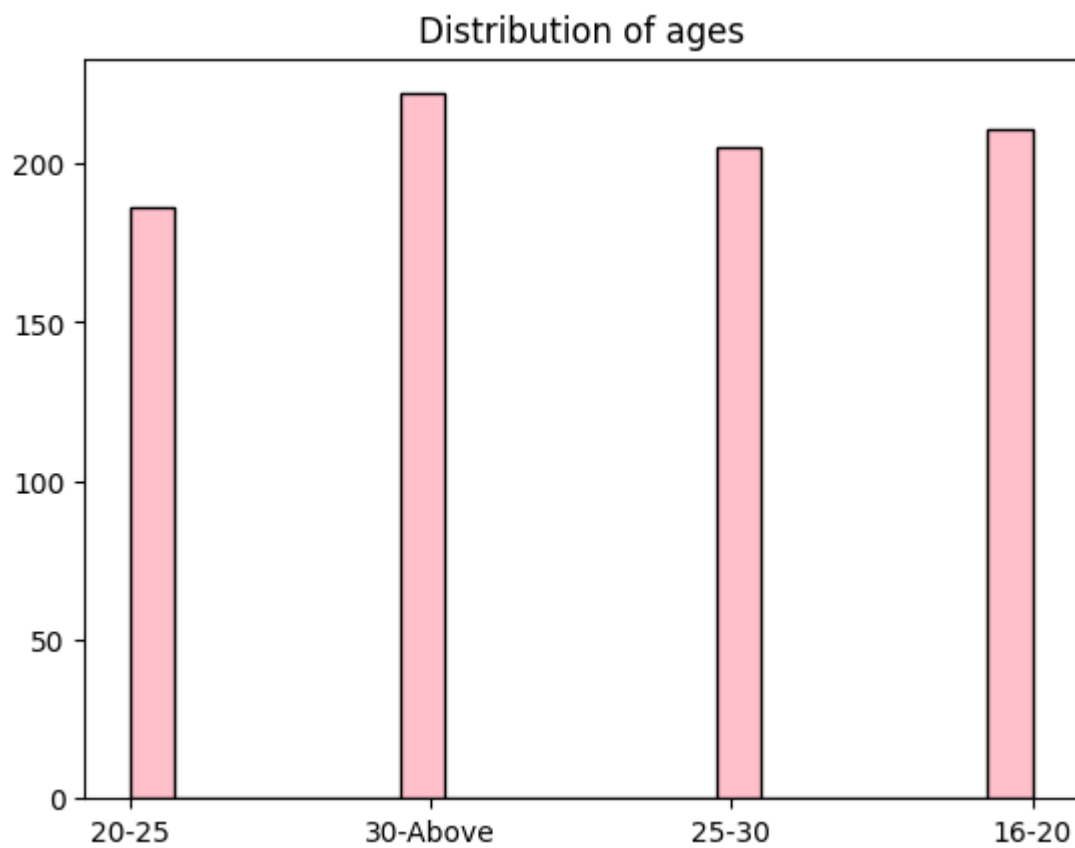
```
In [5]: df.sample(3)
```

```
Out[5]:
```

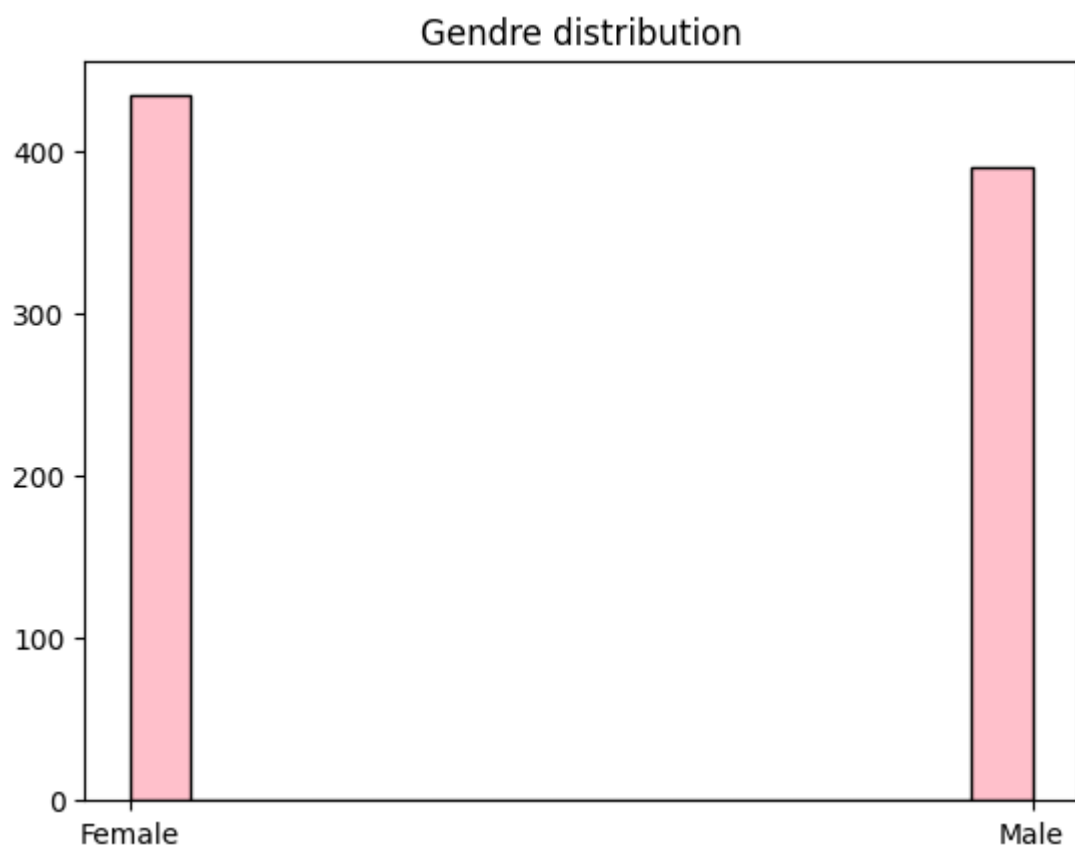
| | Age | Gender | Occupation | Days_Indoors | Growing_Stress | Quarantine_Frustrations | C |
|------------|--------------|--------|------------|--------------|----------------|-------------------------|-------|
| 604 | 30- Above | Female | Housewife | 1-14 days | No | | Yes |
| 721 | 20-25 | Male | Others | 1-14 days | Maybe | | No |
| 386 | 25-30 | Male | Student | 1-14 days | Yes | | Maybe |

Let's see age distribution.

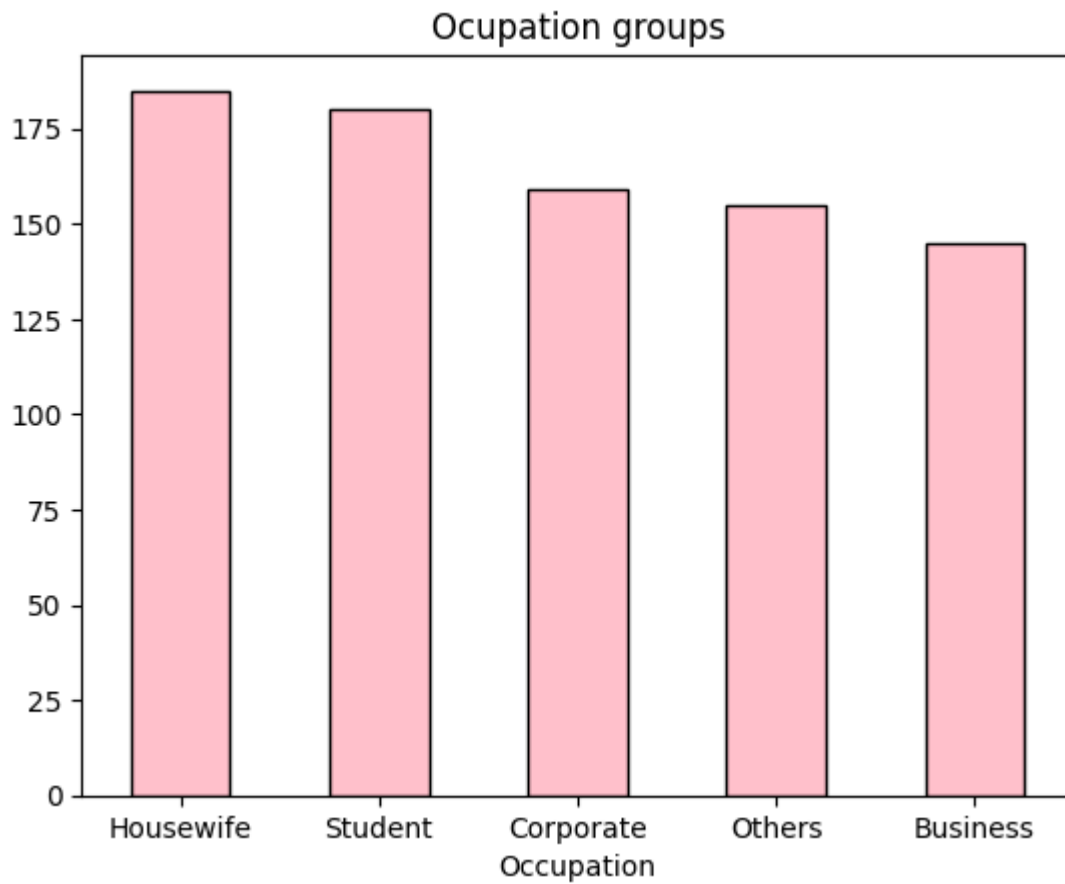
```
In [6]: df['Age'].hist(bins=20, grid=False, edgecolor='black', color='pink')
plt.title('Distribution of ages');
```



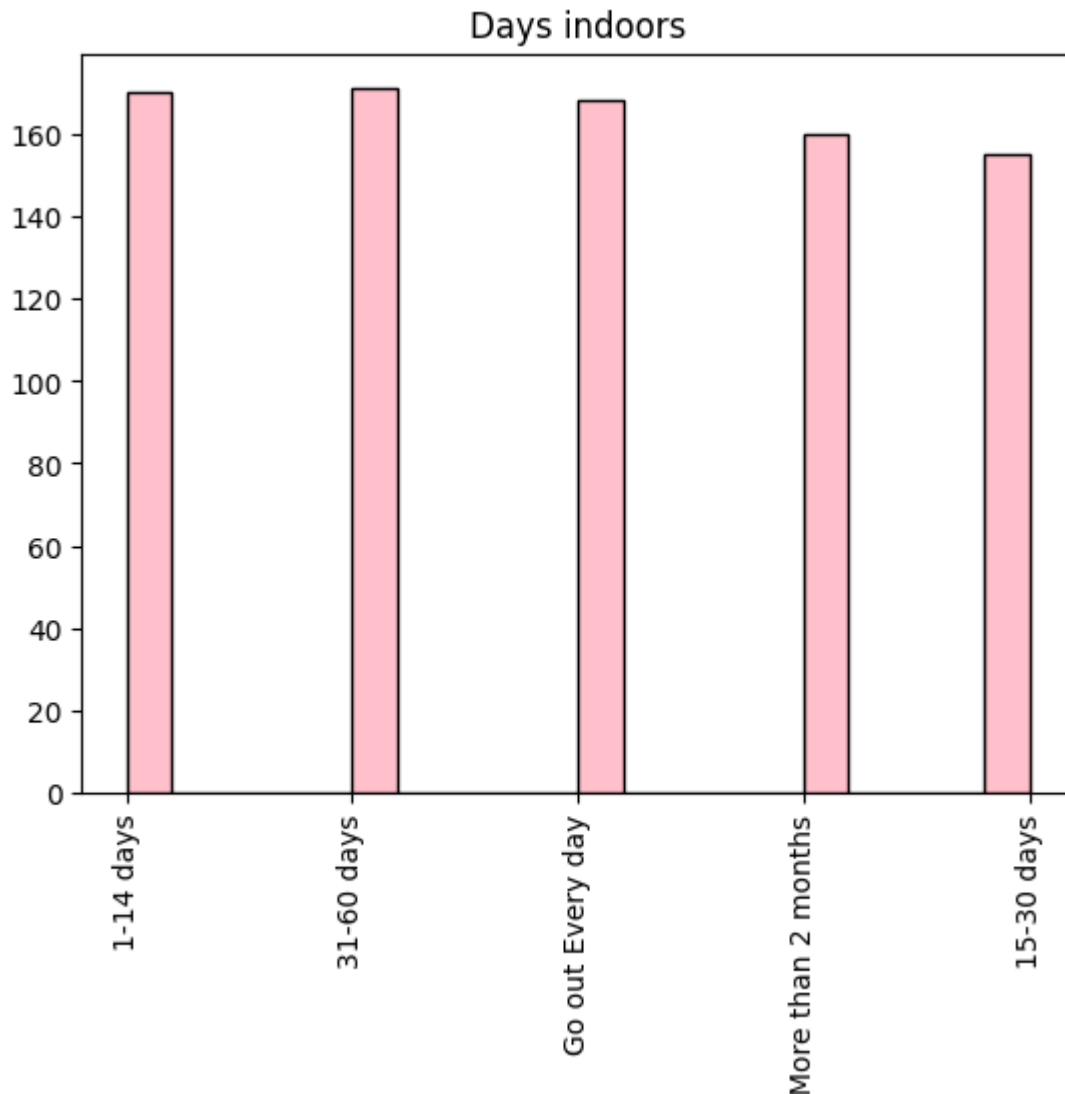
```
In [7]: df['Gender'].hist(bins=15, grid=False, edgecolor='black', color='pink')  
plt.title('Gendre distribution');
```



```
In [8]: df['Occupation'].value_counts().plot(kind='bar', edgecolor='black', color='pink')
plt.xticks(rotation=0)
plt.title('Occupation groups');
```



```
In [9]: plt.xticks(rotation=90)
plt.title('Days indoors')
df['Days_Indoors'].hist(bins=20, grid=False, edgecolor='black', color='pink');
```



```
In [10]: colors = ['#ffa7b4', '#ffcbbd', '#ffb972']
```

Let's see the columns

```
In [11]: display(df.columns)
```

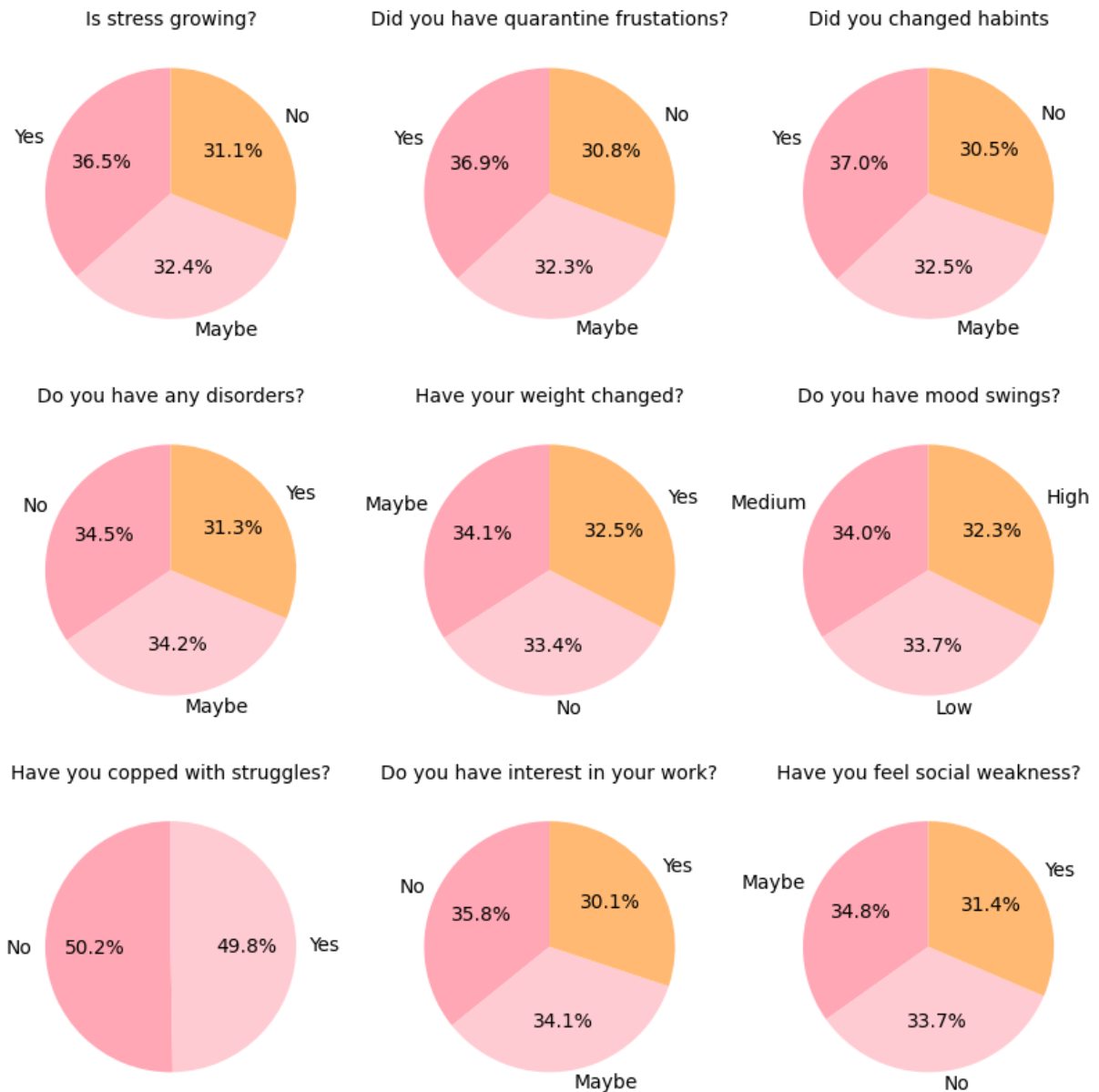
```
Index(['Age', 'Gender', 'Occupation', 'Days_Indoors', 'Growing_Stress',
      'Quarantine_Frustrations', 'Changes_Habits', 'Mental_Health_History',
      'Weight_Change', 'Mood_Swings', 'Coping_Struggles', 'Work_Interest',
      'Social_Weakness'],
      dtype='object')
```

```
In [12]: columns_to_pie = ['Growing_Stress', 'Quarantine_Frustrations', 'Changes_Habits',
                          'Mental_Health_History', 'Weight_Change', 'Mood_Swings',
                          'Coping_Struggles', 'Work_Interest', 'Social_Weakness'
                          ]

columns_for_title = ['Is stress growing?', 'Did you have quarantine frustations?',
                    'Do you have any disorders?', 'Have your weight changed?', 'Do
                    'Have you copped with struggles?', 'Do you have interest in yo
                    ]
```

```
In [13]: figure, axis = plt.subplots(3, 3, figsize=(10, 10))
axis = axis.flatten()

for ax, column, name_columns in zip(axis, columns_to_pie, columns_for_title):
    ax.set_title(name_columns, fontsize=10)
    df[column].value_counts().plot(kind='pie', autopct='%1.1f%%', ax=ax, colors=col
    ax.set_ylabel('')
plt.show()
```



Additional info about data set

Mood swing distribution by gender

```
In [14]: pivot_table_mood = df.pivot_table(index='Gender', columns='Mood_Swings', aggfunc='sum_row_female' = pivot_table_mood.iloc[0].sum(axis=0)
```

```
In [15]: display(pivot_table_mood)
display(sum_row_female)
```

| Mood_Swings High Low Medium | | | |
|-----------------------------|-----|-----|-----|
| Gender | | | |
| Female | 146 | 144 | 144 |
| Male | 120 | 134 | 136 |

434

```
In [16]: mood_swing_level = ['High', 'Medium', 'Low']
genders = ['Female', 'Male']
```

```
In [17]: sum_row_female
```

Out[17]: 434

```
In [18]: #TODO! additional, but not the main.
row_index = -1
for gender in genders:
    row_index += 1
    for level in mood_swing_level:
        display(f'{gender}, {level}, Persantage , {pivot_table_mood[level][gender]}/
```

```
'Female, High, Persantage , 0.33640552995391704'
'Female, Medium, Persantage , 0.3317972350230415'
'Female, Low, Persantage , 0.3317972350230415'
'Male, High, Persantage , 0.3076923076923077'
'Male, Medium, Persantage , 0.3487179487179487'
'Male, Low, Persantage , 0.3435897435897436'
```

As we can see in percentage points female is likely has high mood swing. For the man this high mood swing is the lower. But the difference for female in those 3 groups almost imperceptibly. Difference less than a percentage point. For the men the difference is higher. Compared for others 2 groups high mood swing has 4 percentage points less than medium and low groups

Conclusion

- Women are more susceptible to mood swings. The high level has 3 percentage points more than the men.
- Stress level is growing, with the most percentage than the others answers by 4 percent.
- Many people have experienced quarantine frustration, still the biggest group, by the same 4 percent.
- What is good is that many people have changed their habits. 37 percent of them.
- Only 31 percent of people certainly have any disorders.
- For 33.4 percent of people their weight has not changed. For 32.5 weight has changed, and for others hard to say.
- 32.5 percent of people have a high mood swing level, 34 percent have a medium mood swing level, and 33.7 percent low mood swing level.
- Half have copped struggles and half have not.
- Most people are not interested in their work - 35.8 percent. 34.1 is not sure and only 30.1 interested in it.
- Most people are not sure if they have a social weakness - 34.8 percent. 33.7 certainly has not. And 31.4 certainly does have.
- Most people stayed indoors for 1-60 days.
- Most of the questioned are older than 30.
- Most of the questioned were female.
- Most of the questioned were the housewives.