```
In [369]: from glob import iglob
import pandas as pd
import pandas_profiling
%matplotlib inline
from matplotlib import pyplot as plt
import seaborn as sns
from pylab import rcParams
rcParams['figure.figsize'] = 8, 5
import numpy as np
In [370]: %config InlineBackend.figure_format = 'svg'
```

Задача: 2. Выбрать датасет и провести его анализ

- Выбрать данные
- Составить список вопросов, по которым хотелось бы получить ответ в виде графиков (и расширять этот список вопросов в процессе выполнения задания)
- Построить соответствующие визуализации

Выбрал данные.

Speed Dating Experiment What attributes influence the selection of a romantic partner?

Эксперимент по Быстрому Знакомству Какие атрибуты влияют на выбор романтического партнера?

https://www.kaggle.com/annavictoria/speed-dating-experiment/version/1 (https://www.kaggle.com/annavictoria/speed-dating-experiment/version/1)

```
In [398]: file_name = 'speed_dating_data.csv'
dating_data = pd.read_csv(file_name, index_col=0 ,sep=',', encoding = "]
```

```
In [399]:
            dating data.head(10)
Out[399]:
                  id gender idg condtn wave round position positin1 order partner ... attr3 3 sinc?
             iid
              1
                1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                           7
                                                                 NaN
                                                                          4
                                                                                 1 ...
                                                                                           5.0
                                                           7
              1 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                          3
                                                                                 2 ...
                                                                                           5.0
                                                           7
              1
                 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                         10
                                                                                 3 ...
                                                                                           5.0
              1 1.0
                              1
                                                           7
                                                                 NaN
                          0
                                      1
                                            1
                                                 10
                                                                          5
                                                                                 4 ...
                                                                                           5.0
              1 1.0
                          0
                              1
                                                           7
                                                                 NaN
                                      1
                                            1
                                                 10
                                                                          7
                                                                                 5 ...
                                                                                           5.0
                                                           7
              1 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                                 6 ...
                                                                                           5.0
              1 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                           7
                                                                 NaN
                                                                          1
                                                                                 7 ...
                                                                                           5.0
                                                           7
              1 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                          2
                                                                                 8 ...
                                                                                           5.0
                                                           7
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                                 9 ...
                                                                                           5.0
              1 1.0
                                                                          8
                                                           7
              1 1.0
                          0
                              1
                                      1
                                            1
                                                 10
                                                                 NaN
                                                                          9
                                                                                10 ...
                                                                                           5.0
            10 rows × 194 columns
In [400]:
            print(dating data.columns)
            Index(['id', 'gender', 'idg', 'condtn', 'wave', 'round', 'position',
                     'positin1', 'order', 'partner',
                     'attr3 3', 'sinc3 3', 'intel3 3', 'fun3 3', 'amb3 3', 'attr5
            3',
                     'sinc5 3', 'intel5 3', 'fun5 3', 'amb5 3'],
                   dtype='object', length=194)
```

Выполним профайлинг данных

Вынужден был отключить вывод информации pandas_profiling.ProfileReport так как не генерился PDF файл из notebook для корректного и удобного представления информации.

```
In [1]: # pandas_profiling.ProfileReport(dating_data)
```

После внимательного изучения файла speed_dating_data_Key.doc

и просмотра статистики ProfileReport

Описание значение столбцов которые показались наиболее интересные.

- iid unique subject number, group(wave id gender)
- id subject number within wave
- gender Female=0 Male=1
- round number of people that met in wave
- order the number of date that night when met partner
- partner partner's id number the night of event
- match 1=yes, 0=no
- age_o age of partner
- · dec o decision of partner the night of event
- attr o rating by partner the night of the event, for all 6 attributes
- age age
- field field of study
 - 1= Law
 - 2= Math
 - 3= Social Science, Psychologist
 - 4= Medical Science, Pharmaceuticals, and Bio Tech
 - 5= Engineering
 - 6= English/Creative Writing/ Journalism
 - 7= History/Religion/Philosophy
 - 8= Business/Econ/Finance
 - 9= Education, Academia
 - 10= Biological Sciences/Chemistry/Physics
 - 11= Social Work
 - 12= Undergrad/undecided
 - 13=Political Science/International Affairs
 - 14=Film
 - 15=Fine Arts/Arts Administration
 - 16=Languages
 - 17=Architecture
 - 18=0ther
- goal What is your primary goal in participating in this event?

```
Seemed like a fun night out=1
To meet new people=2
To get a date=3
Looking for a serious relationship=4
To say I did it=5
Other=6
```

• date - In general, how frequently do you go on dates?

```
Several times a week=1
Twice a week=2
Once a week=3
Twice a month=4
Once a month=5
Several times a year=6
Almost never=7
```

• go out - How often do you go out (not necessarily on dates)?

```
Several times a week=1
Twice a week=2
Once a week=3
Twice a month=4
Once a month=5
Several times a year=6
Almost never=7
```

- · career -What is your intended career?
- career_c : career coded

```
1= Lawyer
```

2= Academic/Research

3= Psychologist

4= Doctor/Medicine

5=Engineer

6= Creative Arts/Entertainment

7= Banking/Consulting/Finance/Marketing/Business/CEO/Entreprene ur/Admin

8= Real Estate

9= International/Humanitarian Affairs

10= Undecided

11=Social Work

12=Speech Pathology

13=Politics

14=Pro sports/Athletics

15=0ther

16=Journalism

17=Architecture

Out[377]:

	Id	gender	age	career	round	order	partner	match	age_o	dec_o	field	goal	date	g
iid														
1	1.0	0	21.0	lawyer	10	4	1	0	27.0	0	Law	2.0	7.0	
1	1.0	0	21.0	lawyer	10	3	2	0	22.0	0	Law	2.0	7.0	
1	1.0	0	21.0	lawyer	10	10	3	1	22.0	1	Law	2.0	7.0	
1	1.0	0	21.0	lawyer	10	5	4	1	23.0	1	Law	2.0	7.0	
1	1.0	0	21.0	lawyer	10	7	5	1	24.0	1	Law	2.0	7.0	
4														•

2/15/2019

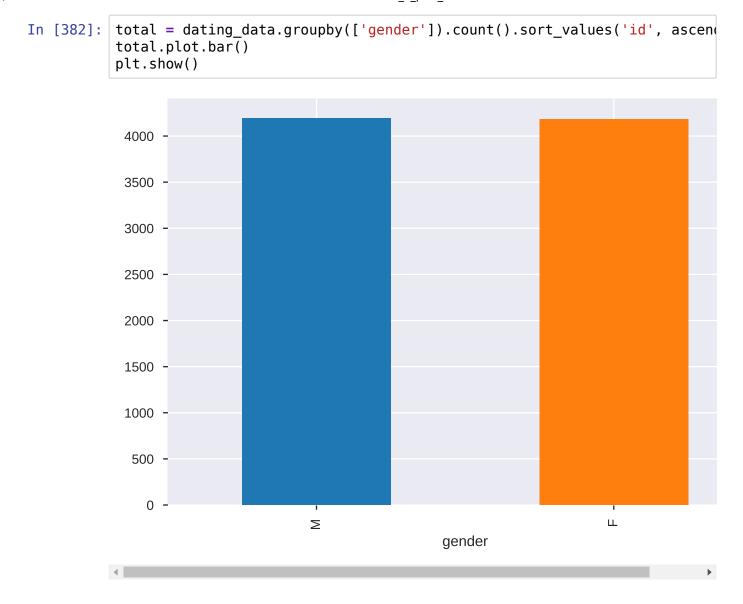
Feature_engineering (Преобразование значений и названий столбцов)

```
In [378]:
          def _feature_engineering_job_by_career_and_field_columns(row):
              Результате работы ``pandas profiling.ProfileReport`` показал основы
              ``career`` -What is your intended career?
              ``field`` - field of study
              1. Lawyer 154 0.0% | Law 121 0.0% | Law 462 0.0% | law 123 0.0%
                               199 0.0% | Professor | Academic |
              professor
                                                                       Scientist
              3. Finance 202 0.0% | Business 521 0.0% | business
                                                                       110 0.0% | 1
              4. Consulting
                             147 0.0% | Social Worker
                                                           136 0.0%
              5. Other values (357) 6887
                                               0.0%
              1. Social Work 378 0.0%
              International Affairs
                                           252 0.0%
              3. Electrical Engineering
                                           164 0.0% | Engineering
              4. Psychology 139 0.0%
              career = str(row['career']).lower()
              field = str(row['field']).lower()
              job = career
              for i in range(2):
                   if job in ['lawyer', 'law']:
                       return 'lawyer'
                  if 'lawyer' in job:
    return 'lawyer'
                  if job in ['professor', 'academic', 'scientist', 'economic']:
                       return 'scientist'
                  if 'academic' in job:
                       return 'scientist'
                  if 'scientist' in job:
                       return 'scientist'
                  if job in ['finance', 'business', 'mba', 'economist']:
                       return 'finance'
                  if 'business' in job:
                       return 'business'
                   if 'bank' in 'job':
                       return 'business'
                   if job in ['social', 'consulting']:
                      return 'social'
                  if 'social' in job:
                       return 'social'
                   if 'engineer' in job:
                       return 'engineer'
                   if 'psychology' in job:
                       return 'psychology'
                   if 'teach' in job:
                       return 'teacher'
                   job = field
              # Default action return small len career or field
              if len(career) < len(field):</pre>
                   return career
```

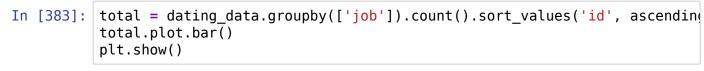
```
return field
```

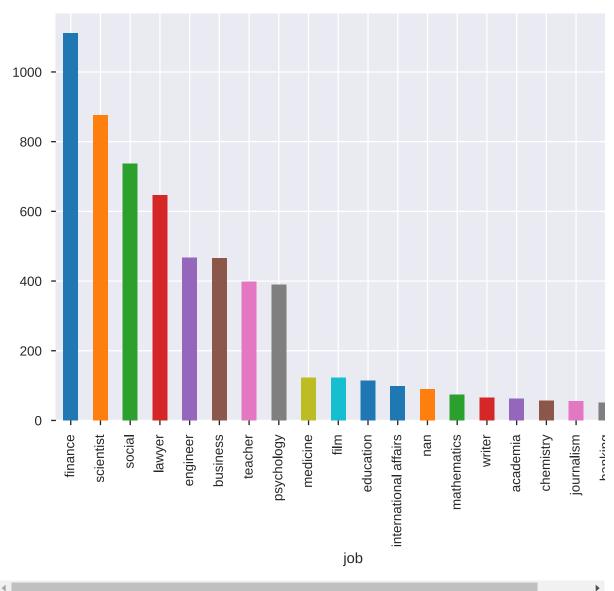
```
In [379]:
           def feature_engineering_clean_data_convert_values(dating_data):
                # gender - Female=0 Male=1
                dating data['gender'] = dating data['gender'].apply(lambda x: 'F' i
                # feature engineering job by career and field columns
                dating data['job'] = dating_data.apply(_feature_engineering_job_by_
                # drop unused column
                dating data = dating data.drop(columns=['career', 'field'])
                return dating_data
In [380]:
           dating_data = feature_engineering_clean_data_convert_values(dating_data
In [381]:
           dating_data.head()
Out[381]:
                           age round order partner match age_o dec_o goal date go_out
                                                                                        job
            iid
                        F 21.0
             1
               1.0
                                  10
                                        4
                                                1
                                                      0
                                                          27.0
                                                                   0
                                                                      2.0
                                                                           7.0
                                                                                  1.0
                                                                                      lawyer
             1 1.0
                        F 21.0
                                  10
                                        3
                                                2
                                                      0
                                                          22.0
                                                                   0
                                                                      2.0
                                                                           7.0
                                                                                      lawyer
             1 1.0
                        F 21.0
                                                          22.0
                                                                      2.0
                                                                           7.0
                                                                                     lawyer
                                  10
                                        10
                                                      1
                                                                   1
                                                                                  1.0
             1 1.0
                        F 21.0
                                  10
                                        5
                                                4
                                                          23.0
                                                                   1
                                                                      2.0
                                                                           7.0
                                                                                      lawyer
                                                      1
                        F 21.0
                                        7
                                                5
                                                                           7.0
             1 1.0
                                  10
                                                      1
                                                          24.0
                                                                   1
                                                                      2.0
                                                                                  1.0 lawyer
```

Кто больше приходи на быстрые свидания, мужчины или женьщины ?

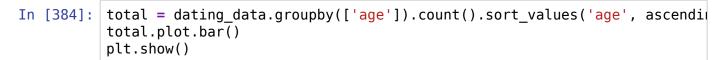


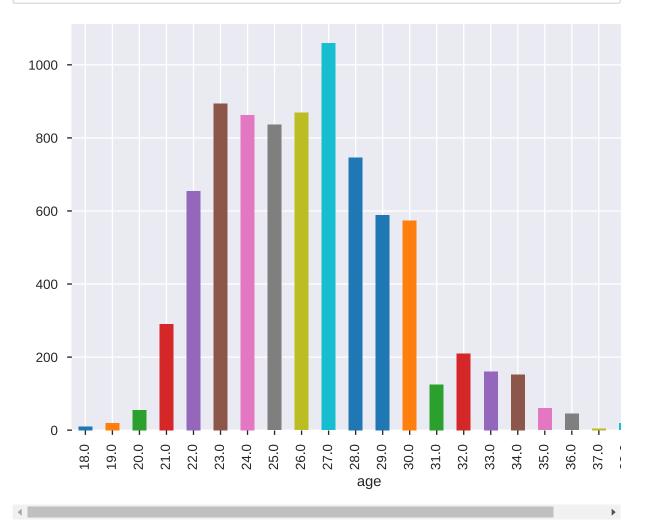
В какой сфере работают люди, участники "быстрых свиданий" Составить топ 20





В каком возрасте мужчины и женщины идуи на быстрые свидания?





1. Сколько раз женщины были выбраны мужчинами которых они не выбирали?

2. Сколько раз мужчины были выбраны женщинами которых они не выбирали?

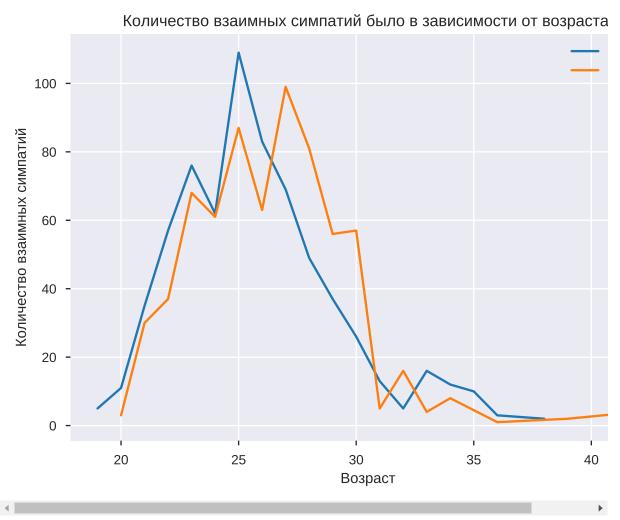
```
In [385]: gender_match_dec_o = dating_data[['gender', 'match', 'dec_o']]
In [386]: female_not_match = gender_match_dec_o[np.logical_and(gender_match_dec_o male_not_match = gender_match_dec_o[np.logical_and(gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_match_dec_o.gender_matc
```

```
In [388]: print("Итого: без взаимности было выбрано женщин:") print(female_not_match_have_been_matched.shape[0]) print("Итого: без взаимности было выбрано мужчин:") print(male_not_match_have_been_matched.shape[0])

Итого: без взаимности было выбрано женщин: 1296
Итого: без взаимности было выбрано мужчин: 839
```

Сколько взаимных симпатий было в зависимости от возраста?

```
In [389]: age_gender_match = dating_data[['age', 'match', 'gender']]
In [390]: age_gender_matched = age_gender_match[age_gender_match.match == 1]
```



1. Какова основная цель участия в этом мероприятии у женщин?

2. Какова основная цель участия в этом мероприятии у мужчин?

• goal - What is your primary goal in participating in this event?

```
Seemed like a fun night out=1
To meet new people=2
To get a date=3
Looking for a serious relationship=4
To say I did it=5
Other=6
```

```
In [392]: goal_id_gender = dating_data[['goal', 'id', 'gender']]
In [393]: dating_goal_female = goal_id_gender[goal_id_gender['gender'] == 'F']
    dating_goal_male = goal_id_gender[goal_id_gender['gender'] == 'M']
```

```
dating goal count female = dating goal female.groupby(['goal']).count()
In [394]:
          dating goal count male = dating goal male.groupby(['goal']).count()[['ic
          plt.plot(
              dating goal count female['goal'],
              dating goal count female.id, label='Female'
          )
          plt.plot(
              dating goal count male['goal'],
              dating goal count male.id, label='Male'
          plt.title('Количество: основная цель участия в этом мероприятии (F & M)
          plt.xlabel('Цель участия в этом мероприятии')
          plt.ylabel('Количество человек выбрали данную цель.')
          plt.text(3, 1200, ' - Seemed like a fun night out=1 \n - To meet new per
          plt.legend()
          plt.show()
```

