What attributes influence the selection of a romantic partner?

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Introduction

Objective

We want to know what factors will affect we select romantic partner. We use the survey from Columbia University about speed dating. And we try to use regression form to interpret the data set.

What is the problem

We want to find out what attributes influence the selection of a romantic partner. We want to use the data set[1] to find the facts as following:

- 1. What are the least desirable attributes in a male partner? Does this differ for female partners?
- 2. How important do people think attractiveness is in potential mate selection vs. its real impact?
- 3. Are shared interests more important than a shared racial background?
- 4. In terms of getting a second date, is it better to be someone's first speed date of the night or their last?

Why this is a project related to this class

Data mining include the following technologies[2]:

- 1. Clustering is the task of discovering groups and structures in the data that are in some way or another "similar", without using known structures in the data.
- Classification is the task of generalizing known structure to apply to new data. For example, an e-mail program might attempt to classify an e-mail as "legitimate" or as "spam".
- 3. Regression attempts to find a function which models the data with the least error.
- 4. Summarization providing a more compact representation of the data set, including visualization and report generation.

We want to utilize the technique of regression to find functions which models the data with the least error. And we can find out some interesting facts from these function's coefficient.

why other approach is no good

Other approach just concern about positive attribute but not negative attributes.

why you think your approach is better

Our approach concern about both positive and negative attributes.

statement of the problem

what attributes influence the selection of a romantic partner and how important that attribute for the selection.

Area or scope of investigation

Romantic selection, speed dating, racial preference, gender difference

Theoretical bases and literature review

definition of the problem

Analyzing if particular factors related to the decision of selection and how they influence the selection

theoretical background of the problem

Clustering, Classification, Regression, Summarization

related research to solve the problem

Gender Differences in Mate Selection: Evidence From a Speed Dating Experiment. Racial Preferences in Dating

advantage/disadvantage of those research

Advantage: shows the fact that gender and racial is dummy variables to do the linear regression Disadvantage: do not show enough comparison between important factors

Our solution to solve this problem

1. What are the least desirable attributes in a male partner? Does this differ for female partners?

$$Decision_{ij} = \alpha_i + \sum_{c \in C} \beta_c Rating_{ijc} + \varepsilon_{ij}$$

2. How important do people think attractiveness is in potential mate selection vs. its real impact?

$$\begin{aligned} Decision_{ij} &= \alpha_i + \beta_0 (Attra_{ij} - SelfExpectAttra_i) \\ &+ \beta_1 (Attra_{ij} - SelfExpectAttra_i) (Attra_{ij} < SelfExpectAttra_i) + \epsilon_{ij} \end{aligned}$$
 Observe β_1 is positive or negative and it's value to determine its real impact

Are shared interests more important than a shared racial background?

$$Decision_{ii} = \alpha_i + \beta_0 SameRace_{ii} + \beta_1 SameInterest_{ii} + \varepsilon_{ii}$$

4. In terms of getting a second date, is it better to be someone's first speed date of the night or their last?

$$Decision_{ij} = \alpha_i + \beta_0 FirstDate_{ij} + \beta_1 LastDate_{ij} + \varepsilon_{ij}$$

Where your solution different from others

Our solution take attributes and their different level of importance into account. And the solution not only focus on the levels of various attributes influence the probability of speed dating, but also focus on the difference between the male and the female.

Why your solution is better

Our solution combines subjective attributes and objective attributes together to find which is the most important factor influence success rate of speed dating.

Hypothesis

- 1. Male and female have different negative attributes.
- 2. The expectation is related the decision.
- 3. Shared racial background is more important than shared interest.
- 4. The order of speed dating really influence, the last partner of speed dating has better feedback than the first.

Methodology

How to collect/generate input data

The data we used is based on an experiment conducted by Columbia University [1]. The researcher hold multiple Speed Dating, in which each participant will fill out a survey before and after the event. Our data is collect from these survey. To make sure each session of speed dating is consistent, the researcher control the light and music to be identical.

How to solve the problem

Algorithm design

We want to propose several regression form which can shows us the result we want.

Language used

We will use python to calculate the multiple regression

Tool used

We will like to use the package of statsmodels, matplotlib and numpy.

How to generate output

We will write a program to read data from csv file and then use statsmodels to generate the output coefficient of each attribute.

How to test against hypothesis

We can observe the the coefficient difference to determine the result. Our goal is to find the special facts from speed dating data set.

How to proof correctness

We may try to reproduce the result of paper [1] to make sure we know and understand how to write the program. And then we will have more confidence of the result from new regression form.

Implementation

code (refer programming requirements)

Python library: statsmodels, numpy, pandas

design document and flowchart

Data acquisition \rightarrow Data transforming \rightarrow Data cleansing \rightarrow Choosing statistic model \rightarrow regression

Data analysis and discussion

Output generation

The attributes includes attractive, sincere, Intelligent, fun and shared interests/hobbies. Initially we ask people to fill out the survey and the score of each attribute should be added up to 100

Hypothesis 1: Male and female have different negative attributes. Our result of this hypothesis is true.

```
Dep. Variable:
                              OLS Adj. R-squared:
Model:
                  Least Squares F-statistic:
Method:
Date:
                  Thu, 16 Mar 2017
                                   Prob (F-statistic):
                        16:42:14 Log-Likelihood:
                              947
lo. Observations:
                                   AIC:
                                                                 4749
                              942
Df Residuals:
                                   BIC:
Df Model:
Covariance Type:
                        nonrobust
 ______
              coef
                                                     [95.0% Conf. Int.]
                      std err
             0.0423
                       0.009
                                 4.697
                                           0.000
                                                       0.025
                                17.284
                                           0.000
                                                       0.060
             0.0679
                        0.004
             0.0585
                        0.011
                                 5.147
                                           0.000
                                                        0.036
             0.1305
                        0.012
                                11.034
                                           0.000
                                                        0.107
                                 7.479
                                           0.000
                                                        0.065
             0.0885
                        0.012
Omnibus:
                          255.942
                                   Durbin-Watson:
                                                                 1.951
Prob(Omnibus):
                            0.000
                                   Jarque-Bera (JB):
                                                               494.621
                           -1.683
                                                             3.93e-108
                                   Prob(JB):
Kurtosis:
                                   Cond. No.
```

male linear regression

		OLS Re	gression	Results		
Dep. Variable	e:		y R-s	quared:		0.823
Model:		(DLS Adj	. R-squared:		0.822
Method:		Least Squar	es F-s	tatistic:		932.2
Date:	Th	nu, 16 Mar 20	17 Pro	b (F-statistic):	0.00
Time:		16:42:	14 Log	-Likelihood:		-2560.2
No. Observat	ions:	16	008 AIC			5130.
Df Residuals:		16	003 BIC	:		5155.
Df Model:			5			
Covariance Ty	/pe:	nonrobu	ıst			
	coef	std err	 t	P> t	95.0% Conf	. Int.]
x1	0.0483	0.011	4.585	0.000	0.028	0.069
x2	0.0737	0.005	15.313		0.064	0.083
x3	0.0792	0.012	6.484	0.000	0.055	0.103
x4	0.0650	0.012	5.623	0.000	0.042	0.088
x5	0.1030	0.010	10.025	0.000	0.083	0.123
Omnibus:		245.9	====== 922 Dur	bin-Watson:		====== 1.768
Prob(Omnibus)):	0.6	000 Jar	que-Bera (JB):		448.753
Skew:		-1.5	552 Pro	b(JB):	3	.58e-98
Kurtosis:		4.6	023 Cor	id. No.		6.32

female linear regression

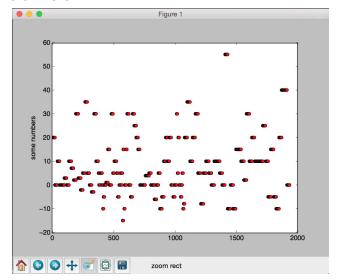
Result:

	Male	Female		
Least desirable	sincere	share interest/hobbit		
	share interest/hobbit	intelligent		
	attractive	attractive		
	intelligent	sincere		
Most desirable	fun	fun		

Hypothesis 2: The expectation is related the decision.

What we found out is that most of the real score assigned to any partner is larger than what he/she expect. The data is biased. Our proposed regression form will fail on this case.

The below is the graph of real score minus the expectation. As we can see, most of it is larger than zero.



```
Dep. Variable:
                                         R-squared:
                                                                            0.215
                                                                           0.215
265.1
Model:
                                   OLS
                                         Adj. R-squared:
                        Least Squares
Method:
                                         Prob (F-statistic):
Log-Likelihood:
                     Thu, 16 Mar 2017
                                                                        1.87e-102
Date:
                              17:07:59
Time:
                                                                         -7458.4
No. Observations:
Df Residuals:
                                  1933
                                                                        1.492e+04
                                  1931
                                                                        1.493e+04
Df Model:
Covariance Type:
                             nonrobust
std err
                                                   P>|t|
                                                              [95.0% Conf. Int.]
              -1.1685
                            0.091
                                     -12.886
                                                  0.000
                                                                -1.346
                                                                           -0.991
               0.3597
                                      21.160
                                                  0.000
                                                                 0.326
                                                                           0.393
                            0.017
Omnibus:
                              6244.911
                                         Durbin-Watson:
                                                                            1.281
                                         Jarque-Bera (JB):
                                                                         147.485
Prob(Omnibus):
                                 0.000
                                 0.174
                                                                         9.42e-33
Skew:
Kurtosis:
                                 1.692
                                         Cond. No.
```

The linear regression

result

shared race & shared interest

 $\beta_0 = 0.1852101$

 $\beta_1 = 0.02124837$

Race and Intereset [0.1852101	.248371					
		OLS Regre	ssion Re	esults		
Dep. Variable:		у		ared:		0.316
Model: Method:	Loos	OLS		R-squared: tistic:		0.316 1906.
Date:		t Squares Mar 2017		(F-statistic):		0.00
Time:	1114, 10	12:42:50		ikelihood:		-6552.8
No. Observations:		8257	_			1.311e+04
Df Residuals:		8255	BIC:			1.312e+04
Df Model:		2				
Covariance Type:		nonrobust			<u></u>	
co	ef std	err	t	P> t	[0.025	0.975]
x1 0.18	352 Ø	.011	16.587	0.000	0.163	0.207
x2 0.02	12 0	.001	40.465	0.000	0.020	0.022
Omnibus:		102.985	Durbi	n-Watson:		1.458
Prob(Omnibus):		0.000		ie-Bera (JB):		939.290
Skew:		0.278				1.09e-204
Kurtosis:		1.444 	Cond	No.		25.5

First Date and last Date

 $\beta_0 = 0.4880$ $\beta_1 = 0.4535$

Consider first and last

OLS Regression Results

=========	======	========	=====	======	========	=======	========
Dep. Variable: Model: Method: Date: Time: No. Observatio Df Residuals: Df Model: Covariance Typ	ns:	Least Sq Thu, 16 Mar 12: nonr	2017 42:50 1068 1066	Adj. F-sta Prob	ared: R-squared: tistic: (F-statistic) ikelihood:	:	0.001 0.000 1.273 0.260 -772.71 1549. 1559.
	coe	f std err		t	P> t	[0.025	0.975]
x1 x2	0.4886 0.4535		5.5	2.731 0.850	0.000 0.000	0.446 0.411	0.530 0.496
Omnibus: Prob(Omnibus): Skew: Kurtosis:			2.416 0.299 0.116 1.018				1.833 177.162 3.39e-39 1.01

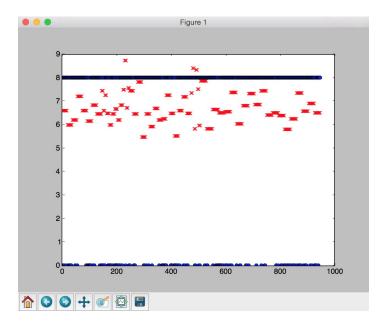
Output analysis

Hypothesis 1: Male and female have different negative attributes. Our result of this hypothesis is true.

Result:

	Male	Female		
Least desirable	sincere	share interest/hobbit		
	share interest/hobbit	intelligent		
	attractive	attractive		
	intelligent	sincere		
Most desirable	fun	fun		

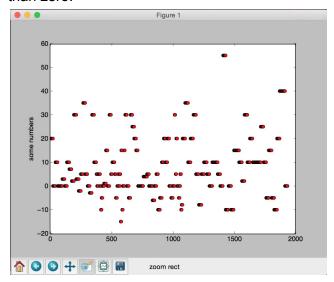
We use linear model try to fit 2 point value of decision. Although we have 82% R-square, it looks like we do not fit the decision. The red dot is our prediction and blue dot is the real value.



Hypothesis 2: The expectation is related the decision.

What we found out is that most of the real score assigned to any partner is larger than what he/she expect. The data is biased. Our proposed regression form will fail on this case.

The below is the graph of real score minus the expectation. As we can see, most of it is larger than zero.



shared race & shared interest

R - square is 0.316 and the difference between two coefficient is really big The β_0 and β_1 worth taking into account when consider the decision of speed dating. It shows that people prefer partner with the same race than the same interest

First Date and last Date

R - square close to 0 thus the 2 variable might not related to the decision. In order to analysis if the order really impact decision. The further analysis is about related order.

```
Decision_{ij} = \alpha_i + \beta_0 RelatedtFirstDistance_{ij} + \beta_1 RelatedLastDistance_{ij} + \epsilon_{ij}

RelatedFirstDistance = (Round - Order)/(Round - 1)

RelatedLastDiatance = (Round - 1)/(Round - 1)
```

Order impact

OLS Regression Results

===========			=====	======			========
Dep. Variable:			У	R-sq	uared:		0.069
Model:			OLS	Adj.	R-squared:		0.068
Method:		Least Squ	ares	F-sta	atistic:		304.2
Date:		Thu, 16 Mar	2017	Prob	(F-statistic):		3.42e-128
Time:		12:4	2:50		Likelihood:		-7826.9
No. Observation	is:		8257	AIC:			1.566e+04
Df Residuals:			8255	BIC:			1.567e+04
Df Model:			2				
Covariance Type	e:	nonro	bust				
=======================================	coef	std err	=====	t	P> t	[0.025	0.975]
x1	0.4880	0.027	1	8.177	0.000	0.435	0.541
x2	0.4535	0.027	1	6.673	0.000	0.400	0.507
Omnibus:		 79	907	Durb:	========= in-Watson:	======	1.136
Prob(Omnibus):			.000	Jarqu	ue-Bera (JB):		876.025
Skew:		0	.244				5.94e-191
Kurtosis:		1	.481	Cond	No.		1.01
			=====	=====			

The coefficient is close and R-square is really low. Thus, The order will not influence the decision

Compare output against hypothesis

shared race & shared interest

Hypothesis about comparison of shared rance and shared interest is true. People from dataset prefer partners with the same race more than the same interest

Order

Unlike hypothesis about order. The output shows order does not matters.

Abnormal case explanation (the most important task)

Use race and gender as dummy variables, Finding that Asian female do not prefer racial background like other group.

Statistic regression

Discussion

About order

The order might work as dummy variables. After find a reliable model of the decision of speed dating. It might worthy that combine order as dummy variable to do the regression.

Conclusions and recommendations

Summary and conclusions

People prefer partners with the same race.

The order of dating in round does not matter.

Recommendations for future studies

Consider more factors like survey time and position as dummy variables

bibliography

[1] RAYMOND FISMAN, SHEENA S. IYENGAR, EMIR KAMENICA, and ITAMAR SIMONSON.
"GENDER DIFFERENCES IN MATE SELECTION: EVIDENCE FROM A SPEED DATING
EXPERIMENT" *The Quarterly Journal of Economics, May 2006*

[2] https://en.wikipedia.org/wiki/Data mining

[3] https://www.kaggle.com/annavictoria/speed-dating-experiment

DataSet in Folder

appendices

program flowchart

 $\mathsf{Data} \ \mathsf{acquisition} \ \to \mathsf{Data} \ \mathsf{transforming} \ \to \mathsf{Data} \ \mathsf{cleansing} \ \to \mathsf{Choosing} \ \mathsf{statistic} \ \mathsf{model} \ \to \\ \mathsf{regression}$

program source code with documentation

See in Folder

input/output listing

Inputting: survey about data mining. The attributes we use are gender, order, round, samerace, dec_o, wave, attr1_1, sinc1_1, intel1_1, fun1_1, amb1_1, shar1_1

Outputting: coefficient of each attributes

other related material

See in Folder

Reference

- [1] RAYMOND FISMAN, SHEENA S. IYENGAR, EMIR KAMENICA, and ITAMAR SIMONSON.
 "GENDER DIFFERENCES IN MATE SELECTION: EVIDENCE FROM A SPEED DATING
 EXPERIMENT" *The Quarterly Journal of Economics, May 2006*
- [2] https://en.wikipedia.org/wiki/Data_mining
- [3] https://www.kaggle.com/annavictoria/speed-dating-experiment