MyPersonality in Recommender System

Junhua Chen Wei Zeng 2015.10.15 • What is MyPersonality?



- MyPersonality is a project (<u>http://mypersonality.org/</u>) conducted by Michal Kosinski and David Stilwell. It collects users' digital footprints (e.g., Facebook likes, friends, and Last.FM) and users' demographic and psychological profiles (e.g., BigFive, IQ, SWLsatisfaction-of-life).
- What is Recommender System?
 - RSs concept is wide, what we are focusing on is to recover potential link given a user-item sparse matrix.

Contents

- MyPersonality dataset introduction
- Related works on Personality
 - Michal Kosinski and David Stillwell's works.
 - BigFive in recommender system.
- What we are planing to do in MyPersonality
 - consider different forms of personality in RSs.

MyPersonality Dataset



Kosinski & Stillwell 's works

- Private traits and attributes are predictable from digital records of human behavior (PNAS,2013)
 - **brief**: the study extracted features out of FB-likes data(using SVD) and use the features to predict the users' traits (with linearModel/logisticModel).
- Computer-based personality judgments are more accurate than those made by humans (PNAS,2014)
 - **brief**: the study compares the accuracy of personality judgment between computer models and humans.



- Users' FB Likes user-like sparse matrix (0.3%).
- SVD (math), 100 components as users' features.
- Use the 100-D features from #2 to train predictor linear model for regression & logistic for classification.



- Classification (top-left) & regression (bottom-left) results, AUC. The transparent bars indicate the questionnaire's baseline accuracy (test-retest).
- Predictive Power of Likes (right).



- Using 90% of the users (with BigFive and FB-Likes) to train a Linear Model (Lasso), predict the BigFive of users in the rest 10%. The BigFive of these 10% users is called computer-based personality.
- repeat 10 times to get all computer-based personality for all users.



- Human-based personality is judged by couple (friends, spouse, colleague) circle on the red line.
- Accuracy (correlation) made by computer-based (red line) surpass humanbased (blue circle) with 100 likes, and the accuracy of computer-based is monotony.

Interjudge Agreement

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 two judges that agree with each other are more likely to be accurate than those that do not. (table on the right)

External Validity

- how well a judgement predicts external criteria, such as reallife behavior (figure on the right).
- the fact (in the figure) computer beats human and even self-reported in "Social network activities" somewhat motivate our work.

		huma	an	CO	mput	er	
r-inter		0.62			0.38		
CORRELATIONS (con	tinuous	variables)		Self-rep	oorted persor Hu Com	nality Iman Iputer	
Social network activities	* -0.03		0.10				
Substance use*		0.00	+0.07				
Network size		-0.06	+(0.01			
Network characteristics	•	-0.01	00				
Values*	-0.0	9					
Sensational interests*		-0.16		0.03			
Physical health*		-0.14	.06				
Depression			-0.13	:0.07			
Self-monitoring		-0.21	117	0.07			
Impulsivity			-0.26				
Life satisfaction			-0.28				
		1	1	1	1	I	
AUC (dichotomous varia	ables)	0.1	0.2	0.5	0.4	0.5	
Field of Study*		-0.02	02				
Dolitical attitudo		0.06					

BigFive in RSs

- *M Braunhofer, M Elahi, and F Ricci, User Personality and the New User Problem in a Context-Aware Point of Interest Recommender System, ICTT, 2015.*
 - brief: the paper integrate BigFive (Openness, Agreeableness, Extraversion, Conscientiousness, and Neuroticism) into a matrix factorization framework.

• a matrix factorization framework in this paper.

$$\hat{r}_{ui} = \mu + b_u + b_i + q_i^T \left(p_u + \sum_{a \in A(u)} y_a \right)$$

 Integrate BigFive personality — e.g., if user u is high on Extraversion and low on the other four, the equation will be:

$$\hat{r}_{ui} = \mu + b_u + b_i +$$

 $q_i^T \left(p_u + y_{ope_low} + y_{con_low} + y_{ext_high} + y_{age_low} + y_{neu_low} \right)$

Planing to do in MyPersonality

- Users' personalities provide a great deal of information about users' interests. Therefore, the performance of a recommender system could be improved when taking into account users' personalities. In our project, we plan to find the answer to two questions:
 - Q1: How to use the user personality in a recommender system?
 - Q2: Which one is better for the human-based, computer-based personality.

4 Q1

 Luckily, M Braunhofer had bring us a feasible plan, but the equation treat the BigFive personality as tag. Yet those personality in MyPersonality dataset is continued. i.e., we need to mod the framework for MyPersonality.

$$\hat{r}_{ui} = \mu + b_u + b_i + q_i^T \left(p_u + \frac{1}{Norm} \sum_{a \in A_u} y_a \right)$$

4 Q2

 Computer-based vs human-base, with decent among of likes data, computer actually beats human (Kosinski and Stillwell). But their 'External Validity' experiments shows that in 'socail-net activity', computer personality is even better than self-reported one, which gives us window to consider why and how computer-based outperform selfreported personality in RSs.



Questions

Thank You