### An Introduction to Web Science

RuSSIR, Aug 6-10, 2012

Please interrupt at any point!!

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### Course Outline

- Day 1: Introduction to the Introduction
  - Examples, data sets, presentation of the competition
- Day 2: Web Search and Society
  - Demographics, economy and more
- Day 3: Blogs and Twitter
  - Gender, moods, politics, stock market and more
- Day 4: Social Networks and Online Dating
  - Attractiveness, FB&GPA, FB&Personality and more
- Day 5: E-commerce and Marketing Studies
  - Brand congruence, Groupon Effect, social ads

What is beautiful is good, even online:
Correlations between photo
attractiveness and text attractiveness
in men's online dating profiles

Rebecca Brand, Abigail Bonatsos, Rebecca D'Orazio and Hilary DeShong Computers in Human Behavior 2012

### Does photo attractiveness ...







### ... correlate with text attractiveness

I just finished my studies, in computer science + science education, and I plan to be a teacher for the next school year. ... I am finished with college, i went to college and majored in criminal justice.

...

I create something new almost every single day, I design, animate or try some kinda crafting. ...

### Data

- 50 female psychology students
  - 18-24 years, average 19.0
  - 74% Caucasian, 10% Hispanic
  - Course credit or entry into \$25 drawing
- 100 male profiles and photos from dating site
  - 22-25 years, NYC and Seattle
  - Both recently active and inactive
  - Photos and "About me" were separated

### **Getting Judgments**

- Photos and texts judged by separate people
- For photo: how attractive, from 0 to 4, (i)
   overall, (ii) for a date, (iii) for sex, and for (iv)
   for a long-term relationship; also how kind,
   how confident, how masculine and how
   symmetrical
- For text: pretty much the same, also how intelligent and how funny
- Each set of 25 profiles rated by 12-13 participants

### **Evaluating the Results**

- Are the four attractiveness questions related?
  - Cronbach's alpha for reliability
  - Summed score for each (photo,judge) X<sub>i</sub>
  - Compute variance of the  $X_i$  :  $\sigma_X^2$
  - Compute variance for each question  $Y_i$ :  $\sigma_{Y_i}^2$

$$- \alpha = \frac{K}{K-1} \left( 1 - \frac{\sum_{i=1}^{K} \sigma_{Y_i}^2}{\sigma_X^2} \right), \text{ here: alpha = .99}$$

- Related to exam design
  - Different questions should measure "the same thing"
  - The overall exam should tell students apart

### Related: Do Different Judges Agree?

**Not** used by them – but fits in this context

Suppose project proposals are evaluated by A and B

### Agreement:

$$(20+15)/50 = 0.70$$

		В		
		Yes	No	
Α	Yes	20	5	
	No	10	15	

But how much agreement by chance?

$$Pr("yes"|A) = 50\%, Pr("yes"|B) = 60\%$$

$$Pr("both yes")+Pr("both no") = .5*.6 + .5*.4 = .5$$

Cohen's kappa 
$$\kappa = \frac{\Pr(a) - \Pr(e)}{1 - \Pr(e)} = \frac{0.70 - 0.50}{1 - 0.50} = 0.40$$

### Back to Study: Correlation Analysis

- Both photo and text attractiveness combined
  - That's what the Cronbach alpha was used for
- Photo attractiveness and text attractiveness
  - Rho=.24, p=.017
- Photo attractiveness and text confidence
  - Rho=.25, p=.012
- Photo masculinity and text confidence
  - Rho=.21, p=0.037

### **Going Beyond Correlation**

- Various things are correlated
- Goal: Isolate a "mediator"

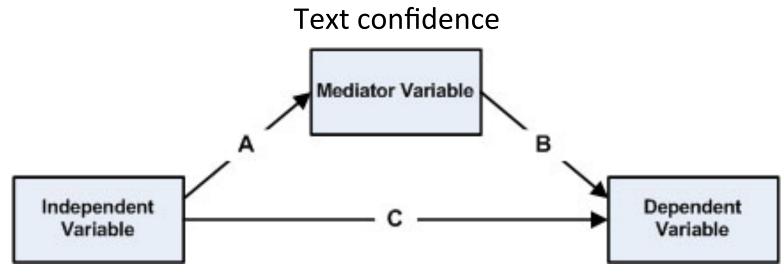


Photo attractiveness

Text attractiveness

### Is (Textual) Confidence a Mediator

 First: the independent variable (here, photo attractiveness) must predict the dependent variable (text attractiveness)

- To test this
  - build a linear regression model (text attr.) = c +  $\beta_1^*$  (photo attr.) + $\epsilon$
  - Then test the hypothesis that  $\beta_1$  = 0 Compare errors w/ and w/o  $\beta_1$

### Is (Textual) Confidence a Mediator

 Second: the independent variable must predict the proposed mediator (text confidence)

Same techniques as before to test this

### Is (Textual) Confidence a Mediator

 Third, the mediator must predict the dependent variable when controlling for the independent variable

(te. at.) = c + 
$$\beta_1$$
\*(ph. at.) +  $\beta_2$ \*(te. conf.) +  $\epsilon$ 

 $eta_2$  still significant: textual confidence adds inf.  $eta_1$  no longer significant: ph. attr. is redundant Search for "Sobel Test" for details

# Too much face and not enough books: The relationship between multiple indices of Facebook use and academic performance

Reynol Junco

Journal Computers in Human Behavior 2012

### **Basic Question**

 Do your grades suffer if you spend lots of time on Facebook?

- Probably yes, but what if high school GPA is factored in?
- What if other factors are included?
- Does it depend on the type of FB activity?

### Getting the Data

- Sent email to 3,866 undergraduate students
- The email contained a link to a survey
  - http://www.surveymonkey.com/
- Chance to win 90 \$10 Amazon vouchers
- 1,839 students participated
- Self-reported data
- 61 "outliers" removed, e.g. >10hrs of FB/day

### **Questions Asked**

- How much time do/did you spend on FB
  - Both "on average" and yesterday (FBtime)
- How often do/did you check FB
  - Both "on average" and yesterday (FBcheck)
- Also questions related to type of FB activity
  - Games, status updates, posting links, ...
- Also got their current and past marks
  - High school GPA and college GPA

### **Basic Results**

Av. 106 min/day (SD 93 min/day) on FB

No correlation between HSGPA and Fbtime

- Weak corr. between time studying and GPA
   r=.22, p<.001</li>
- Weak corr. between time studying and Fbtime
   r=-.09, p<.001</li>

Hierarchical regression model exploring how demographics, high school GPA, average minutes/day spent on Facebook, and Facebook activities predict overall GPA (N = 1771).

Independent variables	Block 1 demographics $\beta$
Male	125***
African American	$107^{*}$
Asian American	.013
Other ethnicity	022
Caucasian	.035
Less than high school	.009
High school	.017
College graduate	.040
Advanced grad degree	.058*

### Interpreting .189

 Increasing FBtime by one SD lowers your GPA by .189 standard deviation

FBtime SD: 93 minutes (av. 106 minutes)

GPA SD: .65 (av. 2.95)

Spending 4 hours/day = 240 min = av + 1.4\*SD Lowers your GPA by .17

## The More People I Meet, The More I Like My Dog: A Study of Pet-Oriented Social Networks on the Web

Jennifer Golbeck

First Monday 2011

CONFESSIONS Ever Been Attacked By a Dog? I Was Once, and I'll Never Forget It

For the love of dog.



Searchl



**↑** MAGAZINE

**VIDEO** 

**BOOK OF DOG** 

**ANSWERS** 

**GALLERIES** 

**ADOPTION** 

COMMUNITY

#### Penny

Labrador Retriever/Basset Hound



Photo Comments Home:Raleigh, NC Age: 3 Months Sex: Female Weight: 11-25 lbs

Add This Pup as a Friend









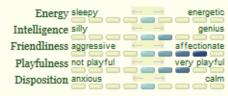


Pals:5 Views: 25

Stars: 🍁

C Leave a bone for Penny

Doggie Dynamics:



Sun Sign:



Quick Bio:

-dog re -mutt

Gotcha Date:

June 3rd 2012

Birthday:

March 30th 2012

Favorite Toy:

Pink squeaky puppy

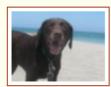
Best Tricks:

Sit

### Meet my Pup Pals



of the High Sea



Captain Squirt Captain Chloe Salty Sea Dog



Taffy(in loving memory



Girls of the Sea



Here, Kitty Kitty.



MAGAZINE

**VIDEO** 

**BOOK OF CAT** 

ANSWERS

**GALLERIES** 

**ADOPTION** 

### Griswald "Grizzy"

Breed Unknown



Photo Comments

Age: 1 Year Sex: Male

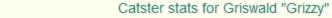
Send a Feline Friend Request! 🗽











Friends:20 Views: 63

Stars: 🍁

Leave a treat for Griswald "Grizzy"

Kitty Complexion:



See all my Feline Friends

Sun Sign:



Quick Bio: -cat rescue

Birthday:

February 15th 20



Angel Pie Joe in Loving Memory



Opie T-bear--In Loving Memory



Kittyman



Violet Lovi Mem

### Research Questions

- Are there observable differences in the way people connect in passion-oriented vs. friendoriented networks?
  - In particular, does the semi-anonymous nature of passion-oriented sites affect the impact of social capital on user behavior?
- Do groups with similar passions utilize passionoriented networking websites in similar or significantly different ways?
  - In particular, how do dog and cat owners utilize their sites?

### A Bit of Context

Table 1. Results of the University of Maryland study on cardiac patient survival rates.

	Pets	No Pets	Total
Living	50	28	78
Dead	3	11	14
Total	53	39	92

Survival rates one year after heart attack. (1977-79)

Other studies have reconfirmed health benefits of owning a pet.

### Oddities of Pet-Networks

- Connections mostly virtual
  - Between strangers with same "passion"
- The owner is hidden
  - All actions performed by the pets
- Break-down by population density
  - Dogs are more frequent in rural households

### Urban vs. Rural

- Gilbert, Karahalios, and Sandvig, 2008: "The network in the garden: an empirical analysis of social media in rural life" (using Myspace):
- Urban users join earlier, have more friends and the friends live farther away than for rural users
- Usually explained using "social capital": "social networks have value"
- What happens when user's identity is hidden?

### Sampling from a Network

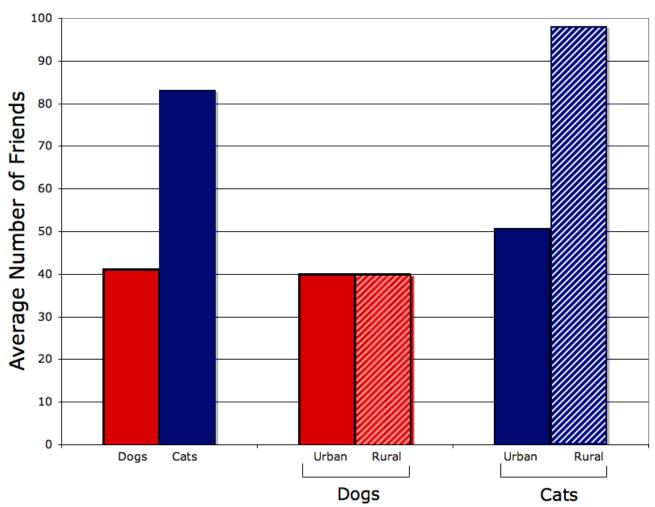
- How to sample uniformly at random
  - In this case easy:
    http://www.dogster.com/dogs/1253721
  - Consecutive user ID at the end
- To sample urban vs. rural
  - Same Methodology as other study
  - Classify ZIP codes into range from urban to rural
  - Select 2,000 locations from either extreme end
  - Try to find a profile for that location

### Comparing non-Normal Distributions

- t-tests are standard to compare means
  - Assumes a normal distribution power law
- Non-parametric tests exist: Mann-Whitney U
  - Count how often an X comes before a Y
  - THHHHTTHTT
  - For T:  $U_T = 0 + 4 + 4 + 5 + 5 = 18$
  - For H:  $U_H = 1 + 1 + 1 + 1 + 3 = 7$
  - $U = min(U_T, U_H) --- but: U_T + U_H = n_T * n_U$

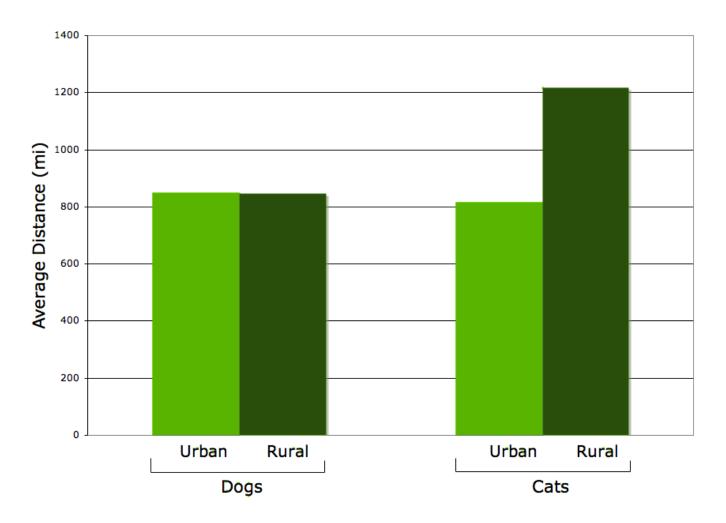
$$z = \frac{U - m_U}{\sigma_U}, \qquad m_U = \frac{n_1 n_2}{2}. \qquad \sigma_U = \sqrt{\frac{n_1 n_2 (n_1 + n_2 + 1)}{12}}.$$

### Findings



Less pet-related interactions in the real world appear to be "compensated".

### **Findings**



A factor ~10 larger than for MySpace study

### **Findings**

	Rural	Urban	$\mathbf{z}$	p-value
Cats				
N	1740	2000		
Member ID	342090	195192	-25.2	< .001
Number of Friends	98.1	50.6	2.6	< .01
Distance between Friends	815.6	1218.0		< .001
Dogs				
N	1669	2000		
Member ID	290903	204788	-15.5	< .001
Number of Friends	39.9	39.9	-0.1	> .920
Distance between Friends	849.0	845.6		> .5

Urban area first, despite lacking "social capital"

### Inferring social ties from geographic coincidences

David Crandall, Lars Backstrom, Dan Cosley, Siddharth Suri, Daniel Huttenlocher and Jon Kleinberg

PNAS 2010

### Suppose that ...

- ... we frequently go to the same pubs
- ... we're going to the same university
- ... we're going to the same super market

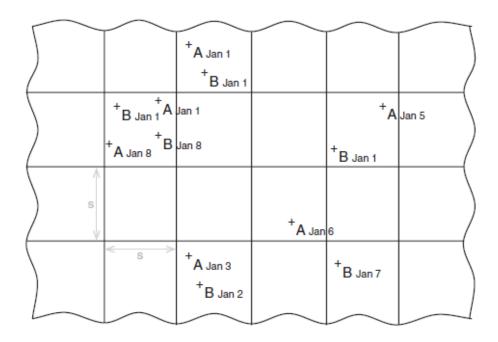
Does that mean we know each other?

### Research Questions

- Given that two people have been in approximately the same geographic locale at approximately the same time, on multiple occasions, how likely are they to know each other?
- Furthermore, how does this likelihood depend on the spatial and temporal proximity of the co-occurrences?

### Methodology

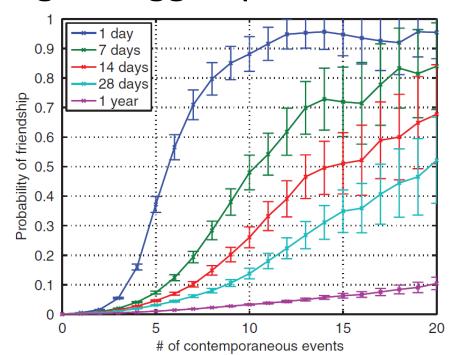
- Geo-tagged photos on Flickr
- Contact list on Flickr



**Fig. 1.** Illustration of how spatio-temporal co-occurrences are counted, for some sample time-stamped observations of individuals A and B. The world is divided into discrete cells of size  $s \times s$ , and we count the number of cells k in which the two individuals have been observed within a time threshold of t days—in this case, k = 3 when t is 2.

#### **Details**

- Social graph snapshot in April 2008
- Use spatio-temporal occurrences appearing after this date
- 38 million geo-tagged photos



s = 0.1

#### A Simple Model

- N geogr. cells, M people, each with one contact
- M/2 pairs of friends, each day chooses a location
- Jointly with prob. eta, independently with 1-eta
- Prob. being friend visiting same cell on k consecutive days:  $P(E)P(C \mid E)$

$$P(F|C_k) = \frac{P(F)P(C_k|F)}{P(C_k)}$$

$$P(F) = 1/(M-1)$$
  
 $P(C_k|F) = p_1^k$  where  $p_1 = \beta + (1-\beta)/N$ 

#### Putting It All Together

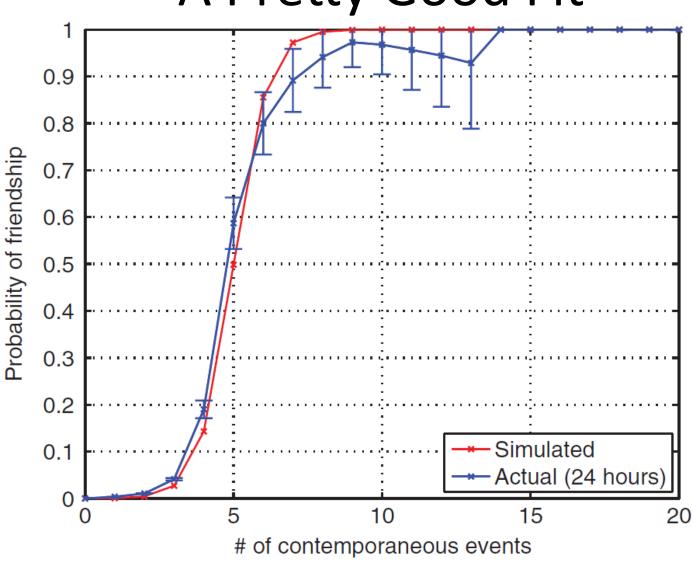
$$P(C_k) = P(C_k|F)P(F) + P(C_k|\bar{F})P(\bar{F})$$

$$= p_1^k \cdot \frac{1}{M-1} + p_2^k \cdot \frac{M-2}{M-1}$$

$$p_1 = \beta + \frac{1-\beta}{N} \quad p_2 = \frac{1}{N}$$

$$P(F|C_k) = \frac{p_1^k}{p_1^k + p_2^k(M-2)}$$

#### A Pretty Good Fit



 $M = 7,500, N = 100, \beta = 0.05$ 

t = 1, s = 1

#### Refining the Model

- Each user pair is assigned a home cell according to Flickr's empirical distribution
- On a given day each person goes out (and takes a photo) with probability  $\alpha$
- Go to same location with probability eta
- Sample power law distribution centered "home" and with exponent  $\gamma$

#### Model-Fitting and Comparing Distributions

$$M = 7,500, N = 64,800, \alpha = 0.29, \beta = 0.12, \gamma = 1.8$$

- Kolmogorov-Smirnov statistics
- Test for equality of two distributions
- Gives a distance measure
- Kolmogorov–Smirnov statistic

$$F_n(x) = \frac{1}{n} \sum_{i=1}^n I_{X_i \le x}$$
  $D_{n,n'} = \sup_x |F_{1,n}(x) - F_{2,n'}(x)|,$ 

Reject null hypothesis (same distribution) if:

$$\sqrt{\frac{nn'}{n+n'}}D_{n,n'} > K_{\alpha}. \quad \Pr(K \leq K_{\alpha}) = 1 - \alpha.$$

# Personality and Patterns of Facebook Usage

Yoram Bachrach, Michal Kosinski, Thore Graepel, Pushmeet Kohli and David Stillwell WebSci'12

#### The Big Five Personality Model

- Openness
  - inventive/curious vs. consistent/cautious
- Conscientiousness
  - efficient/organized vs. easy-going/careless
- Extraversion
  - outgoing/energetic vs. solitary/reserved
- Agreeableness
  - friendly/compassionate vs. cold/unkind
- Neuroticism
  - sensitive/nervous vs. secure/confident

#### Openness to Experience

 inventive/curious vs. consistent/cautious Appreciation for art, emotion, adventure, unusual ideas, curiosity, and variety of experience. Openness reflects the degree of intellectual curiosity, creativity and a preference for novelty and variety. Some disagreement remains about how to interpret the openness factor, which is sometimes called "intellect" rather than openness to experience.

#### Conscientiousness

 Conscientiousness – (efficient/organized vs. easy-going/careless)

A tendency to show self-discipline, act dutifully, and aim for achievement; planned rather than spontaneous behavior; organized, and dependable.

#### Extraversion

outgoing/energetic vs. solitary/reserved
 Energy, positive emotions, surgency,
 assertiveness, sociability and the tendency to
 seek stimulation in the company of others, and
 talkativeness.

#### Agreeableness

friendly/compassionate vs. cold/unkind
 A tendency to be compassionate and cooperative rather than suspicious and antagonistic towards others.

#### Neuroticism

• sensitive/nervous vs. secure/confident
The tendency to experience unpleasant
emotions easily, such as anger, anxiety,
depression, or vulnerability. Neuroticism also
refers to the degree of emotional stability and
impulse control, and is sometimes referred by
its low pole — "emotional stability".

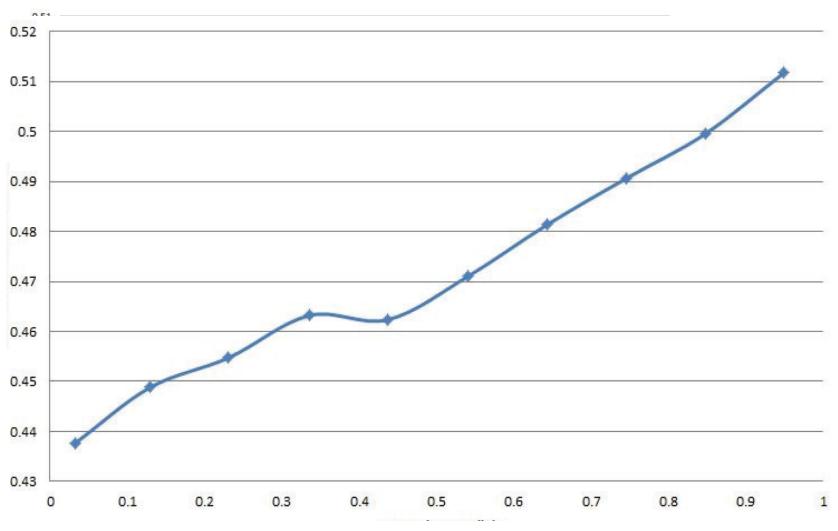
#### Data set

- 180,000 users profiles from "myPersonality"
  - http://www.psychometrics.cam.ac.uk/page/255/ mypersonality.htm
- Not always full profile access
  - But at least 15,000 for each feature
- Average age 24 years, 58% female

Feature	Details
Friends	number of Facebook friends
Groups	number of associations with groups
Likes	number of Facebook "likes"
Photos	number of photos uploaded by user
Statuses	number of status updates by user
Tags	number of times others "tagged" user in photos

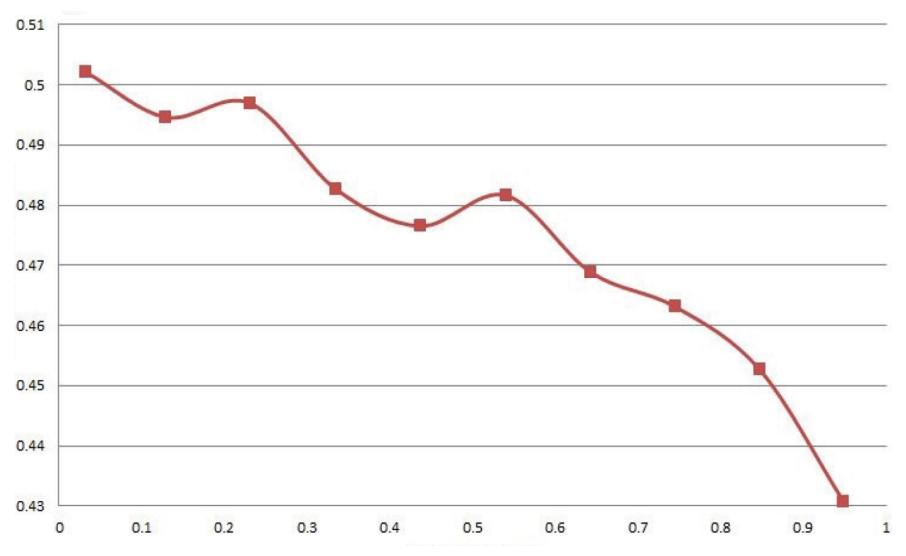
Table 1. Facebook profile features used in this study.

#### **Correlation with Openness**



Percentiles of group memberships

#### Conscientious



Percentiles of group memberships

#### **Predictive Performance**

Trait	$R^2$	RMSE
Openness	0.11	0.29
Conscientiousness	0.17	0.28
Extraversion	0.33	0.27
Agreeableness	0.01	0.29
Neuroticism	0.26	0.28

- Predictions for percentiles
  - "user at the 85-% of agreeableness"
- Baseline: always predict 0.5 (= the median)

$$-\int_{0.0}^{1.0} (x-0.5)^2 dx = \frac{(x-0.5)^3}{3} \Big|_{0.0}^{1.0} = 0.08333$$

• RMSE:  $\sqrt{0.08333} = 0.289$ 

# Reminder: Competition

#### Timeline of the Competition

- Before Thu. 11h00 (and after Wed. 14h00): Cast your vote for one submitted proposal: http://tinyurl.com/RuSSIR-Proposal-Voting
- During Thu. lecture: the top three proposals (according to online votes) are announced
- During Fri. lecture: the top three proposals are presented in person (2 min each, max of 3 slides)

#### And the Top Three Proposals are ...

(not necessarily in order of number of votes)

- Xxx
- Xxx
- Xxx

Please prepare a short presentation for tomorrow Improvise if you prefer, otherwise three slides max

### Questions?

## End of Day 4

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