

Self-Disclosure in Online Interaction: A Meta-Analysis

Christine Bauer
Michael Schiffinger

Hawaii International Conference on System Sciences (HICSS)
5-8 January 2015, Kauai, HI

Self-disclosure



Self-disclosure is defined as
**what individuals verbally communicate
about themselves,**
including thoughts, feelings, and experiences.

- People disclose information for a variety of purposes:
 - establishing legitimacy,
 - authentication,
 - trust,
 - providing personalized services,...



Online self-disclosure



important in electronic commerce and online relationship building:

- signifies **trust** and **acceptance** of the privacy assurance
- in absence of face-to-face interaction companies have to rely on such feedback behavior



Online self-disclosure



is of particular interest in
human-computer interaction

e.g.,

- personalized recommender systems
- “one click” purchasing
- e-recruitment



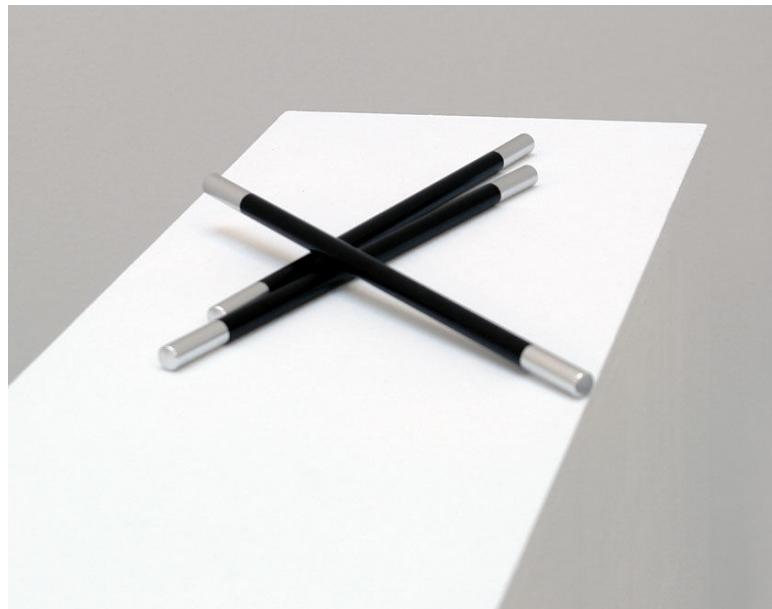
However, not all users are willing to disclose personal information.

Major barrier: **privacy concern**



Still, paradoxically...

...many users appear to provide personal information abundantly in the online setting

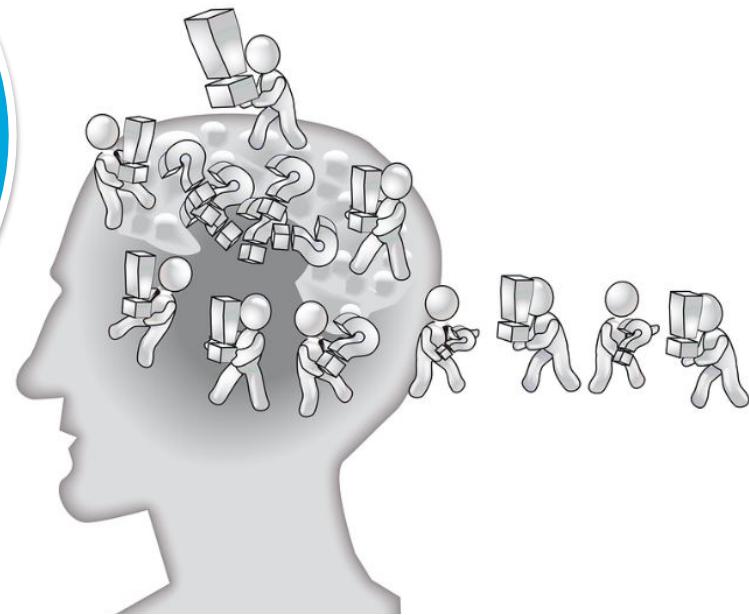


particularly in the context of online social networks



The specific objective

Identify the most
influential factors
that shape
self-disclosure



Central question

Can a user's
self-disclosure
be manipulated?

And to what
extent...?



Many studies on online self-disclosure

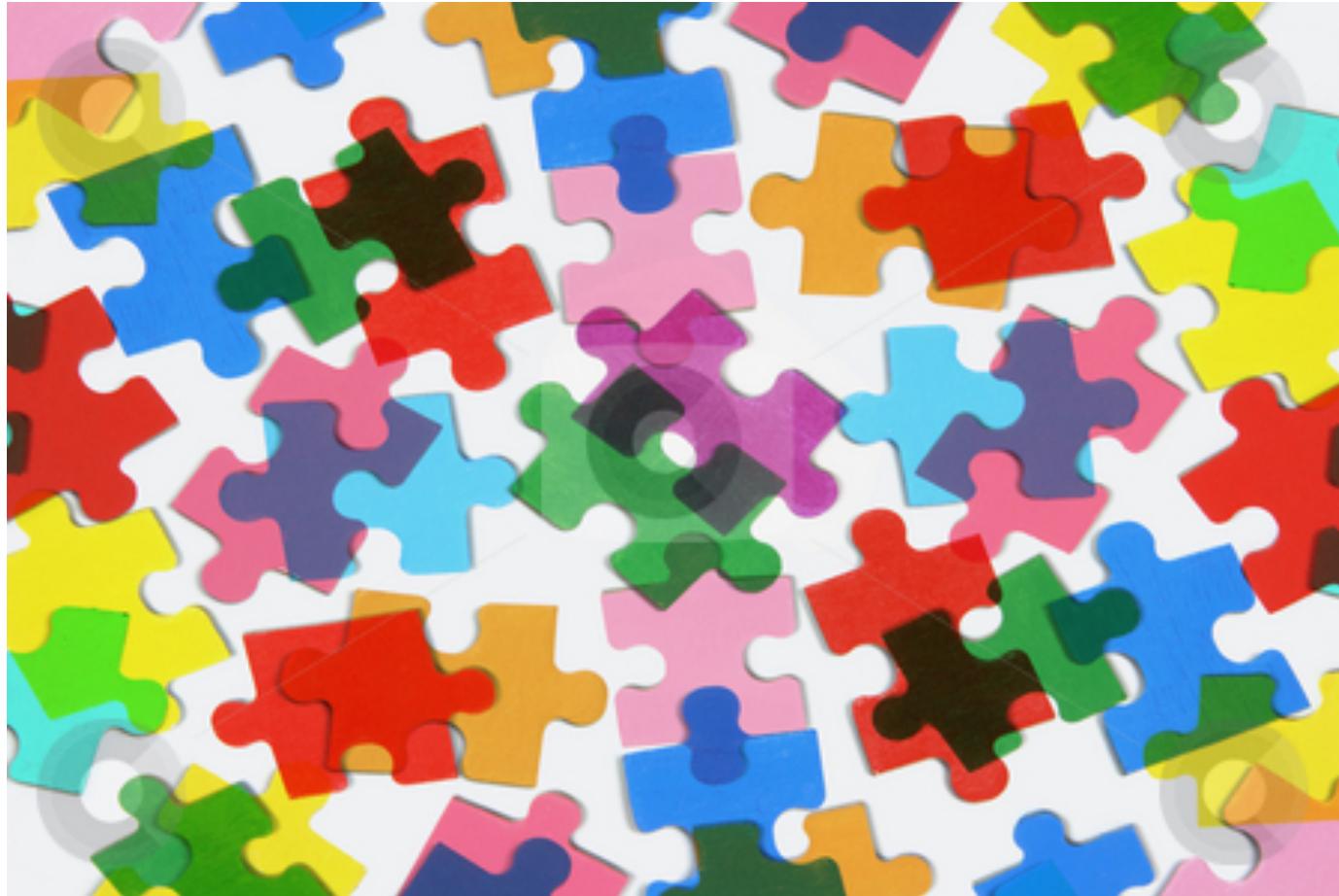
Very specific variables!



The scope is tremendous:

- gender
- education
- social anxiety
- reward
- anonymity
- trust
- privacy
- ...

**...but each study provides only
puzzle pieces**



Goal

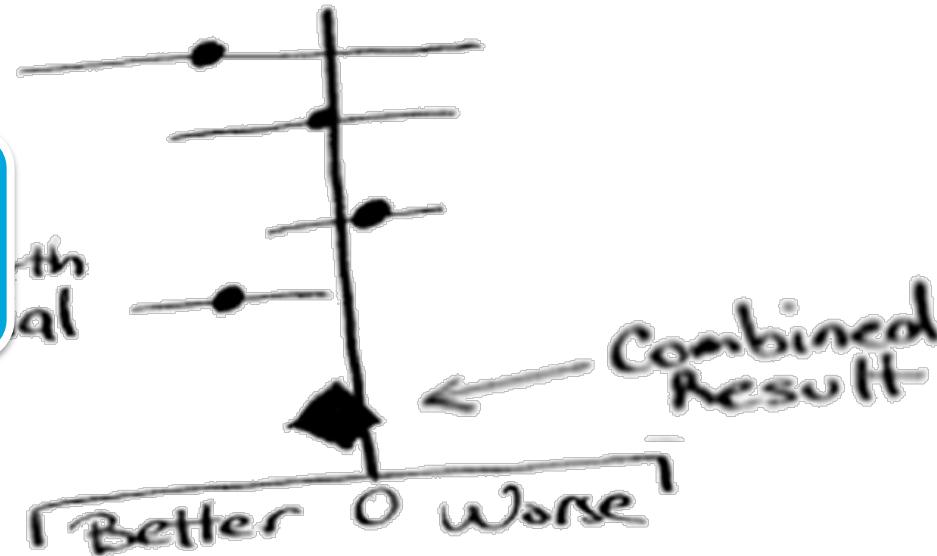
- bring these pieces together
- use existing research findings in the field



Statistical meta-analysis

A statistical meta-analysis represents a **systematic aggregation of the findings** of previous studies regarding the extent to which one or several **predictors affect a dependent variable**, based on so-called **effect sizes**.

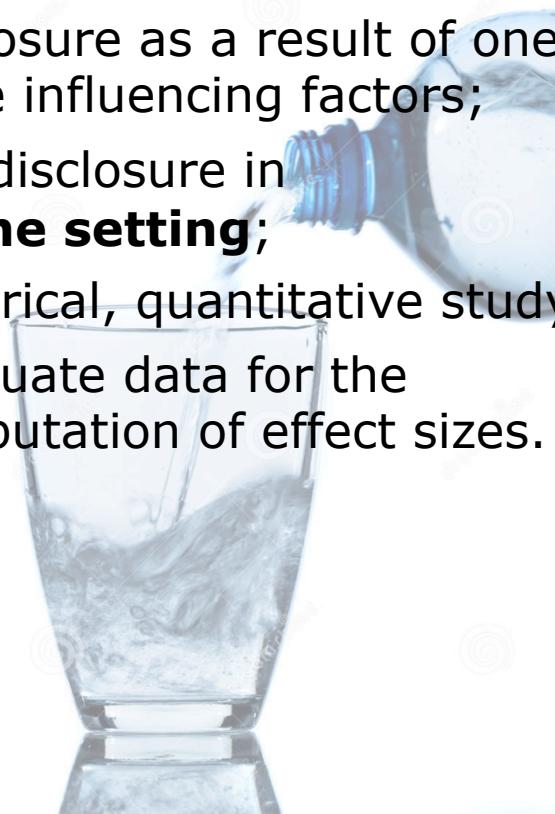
self-disclosure



Criteria for inclusion/exclusion

A study was **included** if it fulfilled **all** of the following criteria:

- disclosure as a result of one or more influencing factors;
- self-disclosure in **online setting**;
- empirical, quantitative study;
- adequate data for the computation of effect sizes.

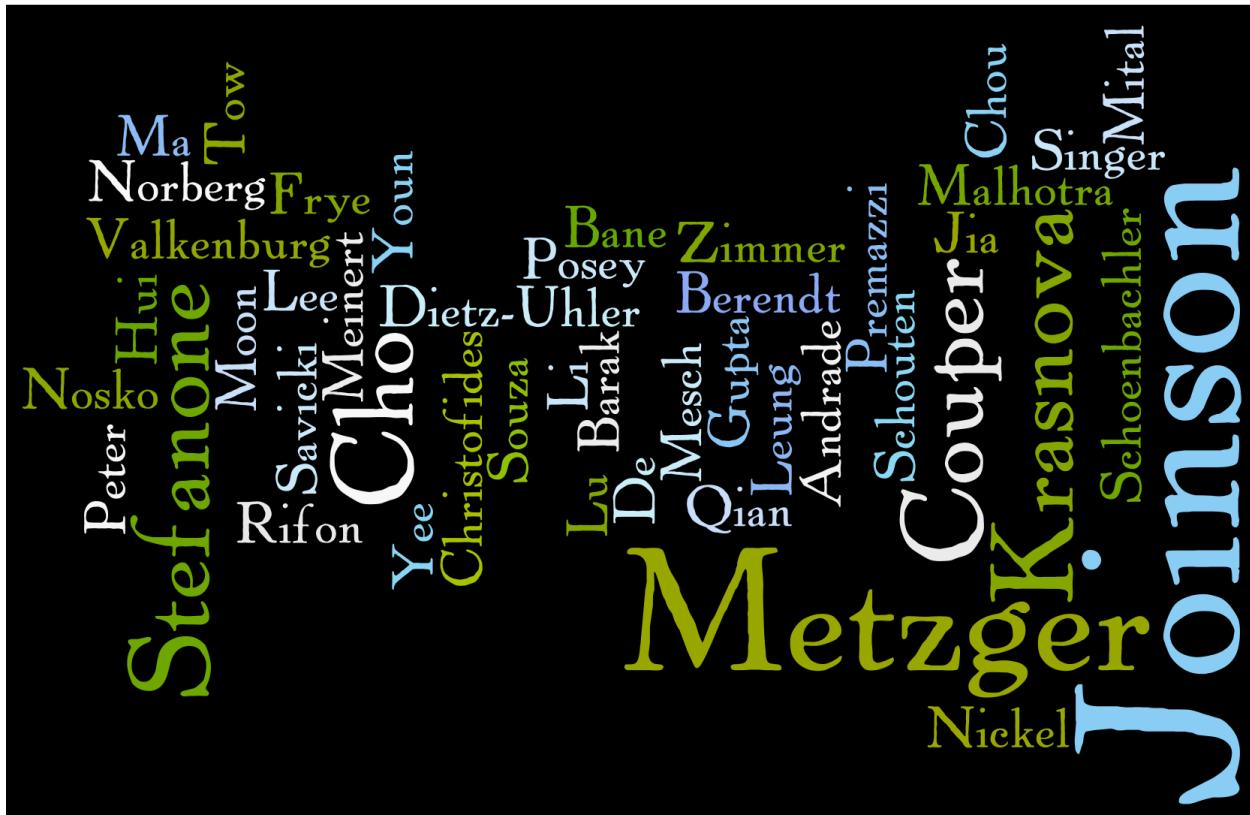


A study was **excluded** if **at least one** of the following criteria were met:

- investigated solely the effects of disclosure on other factors or outcomes;
- disclosure of **health issues**;
- disclosure in the field of **dating**;
- **corporate disclosure**;
- disclosure in offline settings;
- qualitative study;
- data for computing effect sizes not available.

Sample

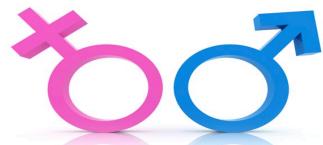
48 studies on online self-disclosure



Puzzle pieces – very specific variables



Categorization of independent variables



Demographics

- e.g., sex, age, or education



Environmental factors

- peer-related variables (e.g., peer pressure)
- provider-related variables (e.g., reputation of a company)



Person-based variables

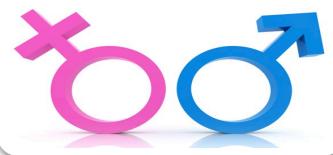
- inherent to a person and his/her perceptions
- e.g., self-esteem, personality traits, or perceived risk



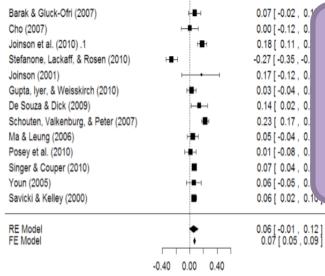
System-based variables

- controlled by a system or inherent to system
- e.g., privacy priming or reward provided for disclosed info

Effects on self-disclosure

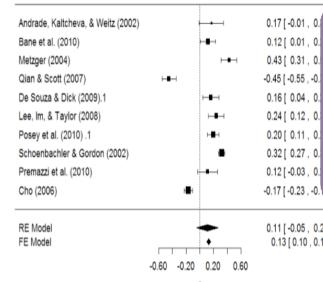


demographics



RE .06
FE .07

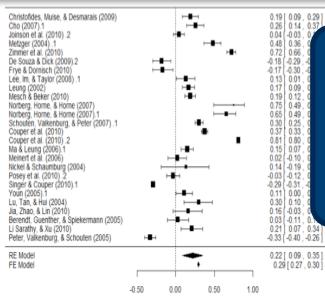
environmental factors



RE .11
FE .13

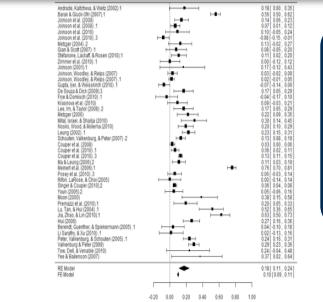


person-based variables



RE .22
FE .29

system-based variables



RE .18
FE .10



Findings

Finding 1

the identified effect sizes appear rather low



- but... results clearly suggest that person- and system-based variables do influence user self-disclosure to some extent

Finding 2

system-based variables, **which can be purposefully designed**, are at least a moderately effective key to “**shape**” user self-disclosure

- e.g.,
 - system functionality and usefulness
 - system type that asks to disclose one's data (e.g., social media platform, web shop, or registration for a game)
 - providing a reward for disclosing one's information



Finding 3: Methodical findings in brief

considerable heterogeneity among the effect sizes in each category

large degree of heterogeneity among the studies' findings for each predictor category

almost the entire identified variation in effects on self-disclosure is due to heterogeneity between studies

but:
predictor category makes a considerable difference for the effect on self-disclosure

Next steps

categorization in its current form might not be the “golden nugget” for explaining this heterogeneity

→ rethink categorization

included studies referring to both attitude and behavior
→ “privacy paradox” not considered

→ considering the two variables separately in analysis

Take away messages

-  person- and system-based variables do influence user self-disclosure
-  system-based variables, which can be purposefully designed, are at least a moderately effective key to "shape" user self-disclosure
-  considerable heterogeneity between studies
-  rethink categorization
-  considering "privacy paradox", considering the two variables separately in analysis





VIENNA UNIVERSITY OF
ECONOMICS AND BUSINESS

Institute for Management Information Systems

Department of Information Systems and Operations
Welthandelsplatz 1, D2, 1020 Vienna, Austria

Mag. DI Dr. Christine Bauer

T +43-1-313 36-4420

F +43-1-313 36 90-4420

chris.bauer@wu.ac.at

www.wu.ac.at/ec/institute/christine-bauer

Additional Information

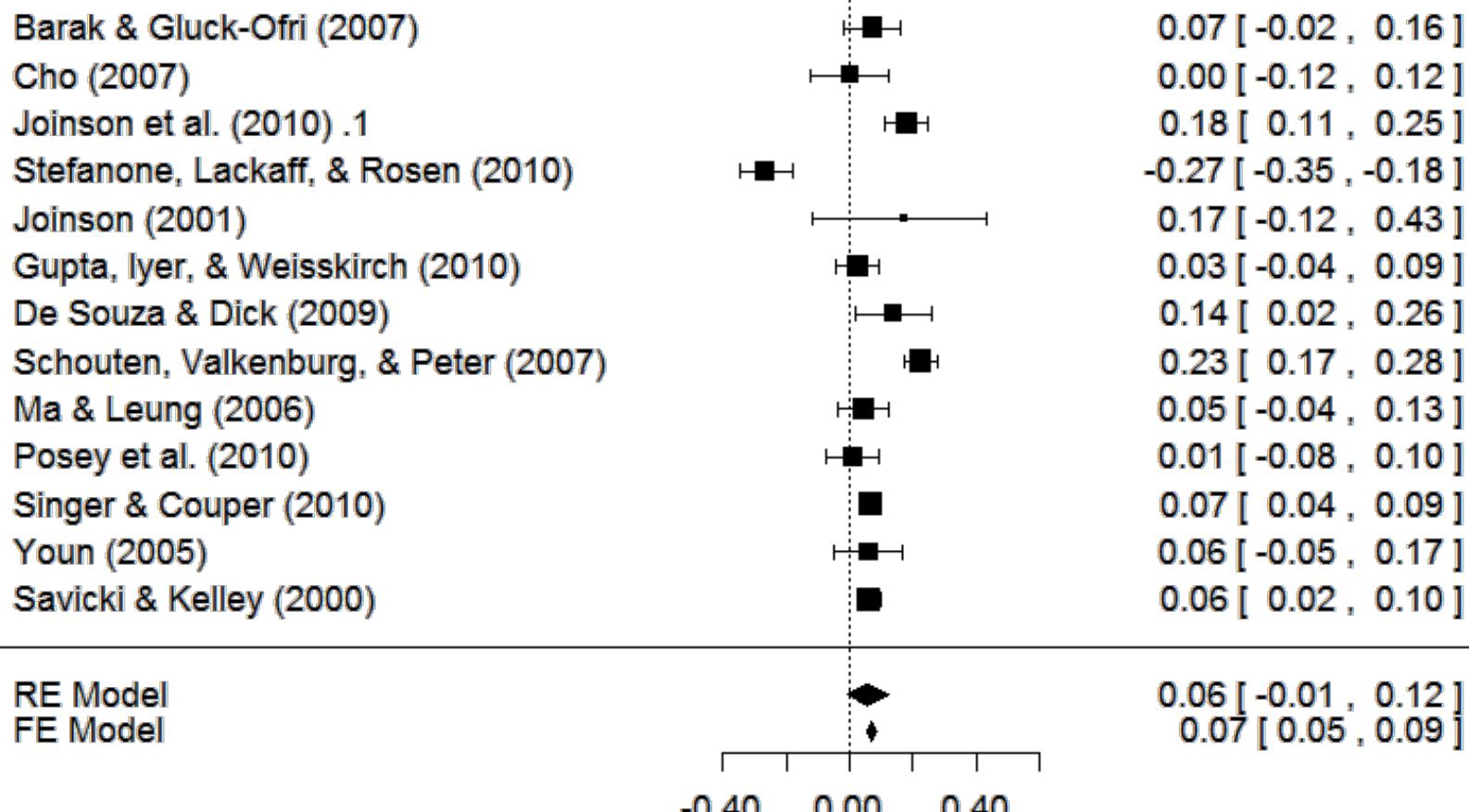
Intercoder agreement

- Initial intercoder agreement
 - overall agreement of 81%
 - Fleiss-Kappa of .77 (= “substantial agreement”)
 - Environmental factors: low Kappa (.43)
 - all other categories: “almost perfect agreement” (Kappa values > .80)
- In the instances where some disagreement emerged, the coders discussed the study in question until **complete consensus** could be established.

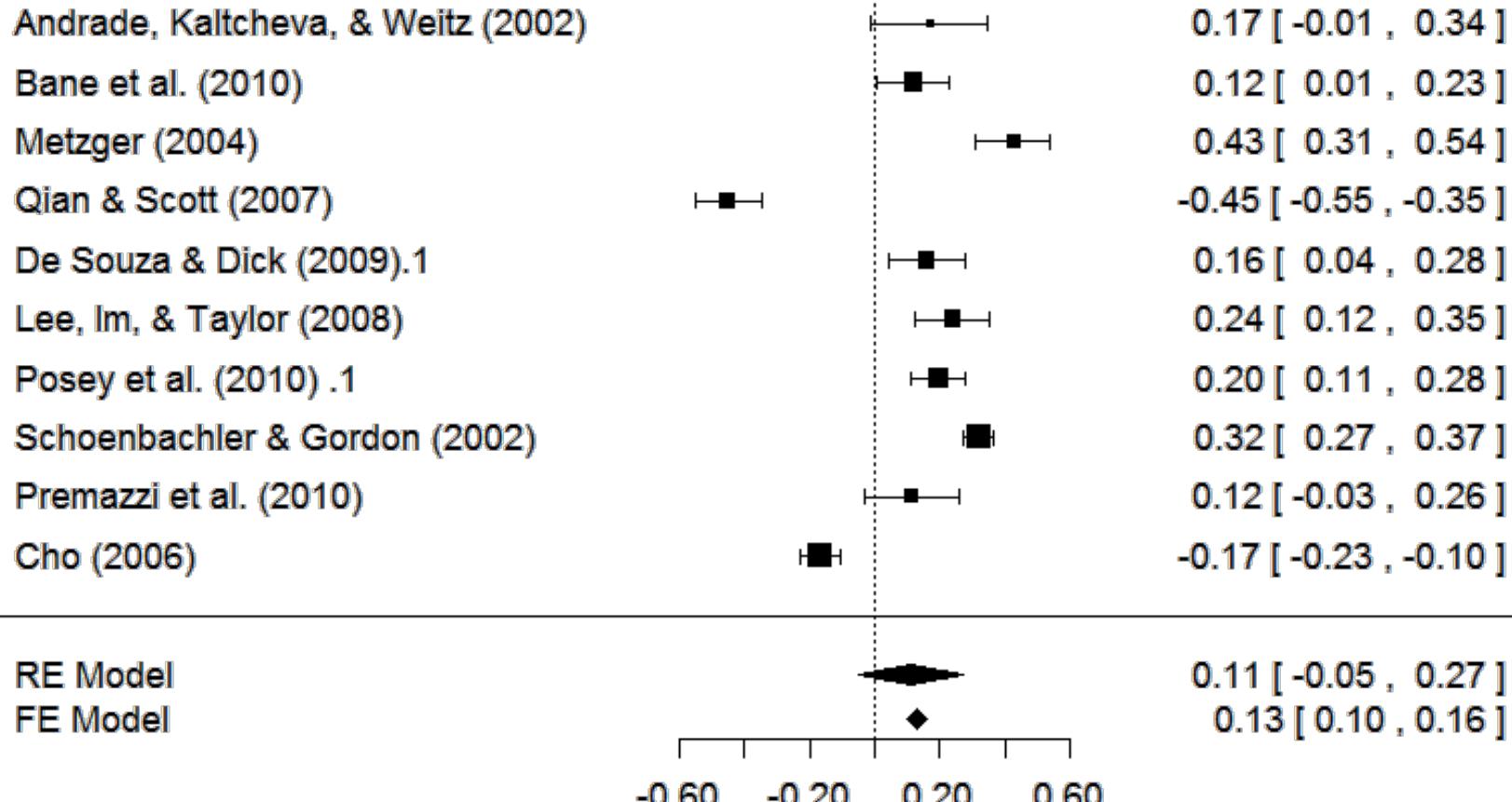
publication	n	categories
Andrade, Kaltcheva and Weitz [22]	114	e, s
Bane, Cornish, Erspamer and Kampman [58]	307	e
Barak and Gluck-Ofri [39]	480	d, s
Berendt, Günther and Spiekermann [59]	206	p, s
Cho [24]	260	d, p
Cho [60]	881	e
Christofides, Muise and Desmarais [31]	343	p
Couper, Singer, Conrad and Groves [28]	3671	s
Couper, Singer, Conrad and Groves [61], study 1	1831	p, s
Couper, Singer, Conrad and Groves [61], study 2	6400	p, s
De Souza and Dick [62]	263	d, e, p, s
Frye and Dornisch [33]	214	p, s
Gupta, Iyer and Weisskirch [16]	809	d, s
Hui, Tan and Goh [63]	331	s
Jia, Zhao and Lin [64]	105	p, s
Joinson [65]	49	d, s
Joinson, Paine, Buchanan and Reips [37], study 1	515	s
Joinson, Paine, Buchanan and Reips [37], study 2	1189	s
Joinson, Reips, Buchanan and Schofield [6], study 1	181	s
Joinson, Reips, Buchanan and Schofield [6], study 2	759	d, p, s
Joinson, Woodley and Reips [66], study 1	1144	s
Joinson, Woodley and Reips [66], study 2	3544	s
Krasnova, Spiekermann, Koroleva and Hildebrand [67]	259	s
Lee, Im and Taylor [14]	259	e, p, s
Leung [68]	576	p, s
Li, Sarathy and Xu [69]	182	p, s
Lu, Tan and Hui [70]	95	p, s
Ma and Leung [27]	591	d, p, s
Meinert, Peterson, Criswell and Crossland [34]	261	p, s
Mesch and Beker [71]	790	p
Metzger [3]	213	d, e, p, s
Metzger [4]	189	e, p, s
Mital, Israel and Agarwal [72]	131	s
Moon [20]	60	s
Nickel and Schaumburg [40]	39	p
Norberg, Horne and Horne [54], study 1	23	p
Norberg, Horne and Horne [54], study 2	68	p
Nosko, Wood and Molena [25]	400	s
Peter, Valkenburg and Schouten [73]	600	p, s
Posey, Lowry, Roberts and Ellis [29]	529	d, e, p, s
Premazzi, Castaldo, Grosso, Raman, Brudvig and Hofacker [23]	178	e, s
Qian and Scott [30]	242	e, s

Rifon, Larose and Choi [35]	210	s
Savicki and Kelley [74]	2692	d
Schoenbachler and Gordon [75]	1338	e
Schouten, Valkenburg and Peter [10]	1203	d, p, s
Singer and Couper [76]	7210	d, p, s
Stefanone, Lackaff and Rosen [77]	452	d, s
Tow, Dell and Venable [78]	51	s
Valkenburg and Peter [79]	812	s
Yee and Bailenson [80]	32	s
Youn [81]	326	d, p, s
Zimmer, Arsal, Al-Marzouq and Grover [32]	264	p, s

Forest plot: demographic predictors



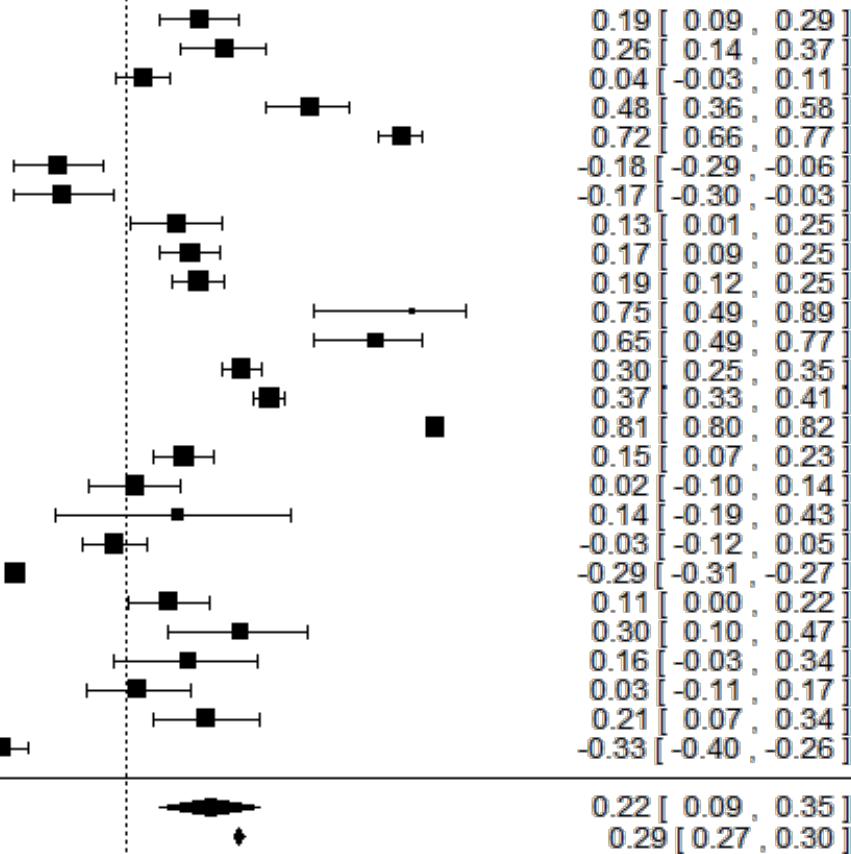
Forest plot: environmental predictors



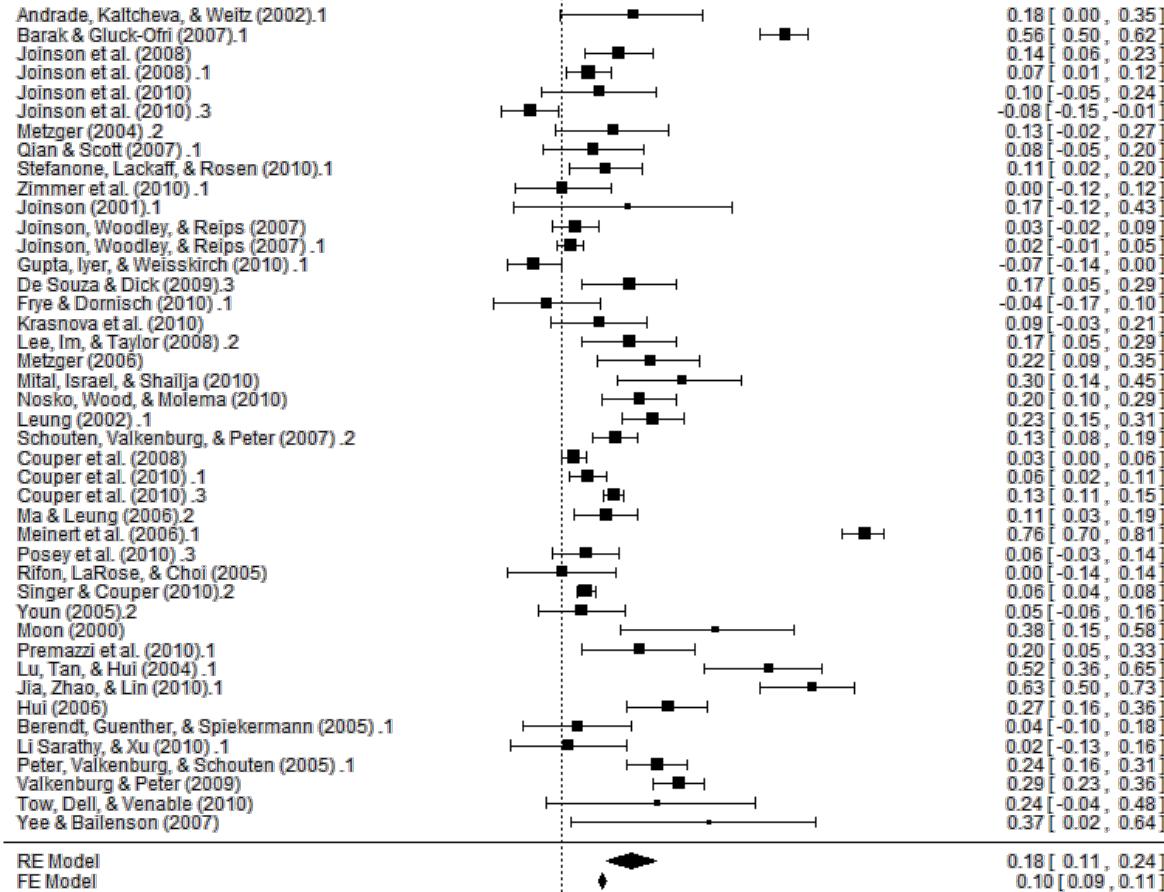
Forest plot: person-based predictors



Christofides, Muise, & Desmarais (2009)
 Cho (2007).1
 Joinson et al. (2010) .2
 Metzger (2004) .1
 Zimmer et al. (2010)
 De Souza & Dick (2009).2
 Frye & Dornisch (2010)
 Lee, Im, & Taylor (2008) .1
 Leung (2002)
 Mesch & Beker (2010)
 Norberg, Horne, & Horne (2007)
 Norberg, Horne, & Horne (2007).1
 Schouten, Valkenburg, & Peter (2007) .1
 Couper et al. (2010)
 Couper et al. (2010) .2
 Ma & Leung (2006).1
 Meinert et al. (2006)
 Nickel & Schaumburg (2004)
 Posey et al. (2010) .2
 Singer & Couper (2010).1
 Youn (2005).1
 Lu, Tan, & Hui (2004)
 Jia, Zhao, & Lin (2010)
 Berendt, Guenther, & Spiekermann (2005)
 Li Sarathy, & Xu (2010)
 Peter, Valkenburg, & Schouten (2005)



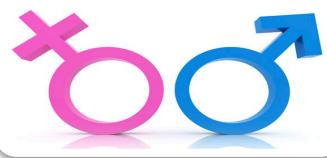
Forest plot: system-based predictors



Heterogeneity statistics for each predictor category

	demographic	environmental
Q (df)	99.8 (12) **	251.8 (9) **
τ^2 (s.e.)	.013 (.006)	.068 (.033)
H	3.65	5.25
I^2	92.5%	96.4%

	person-based	system-based
Q (df)	7658.5 (25) **	626.5 (42) **
τ^2 (s.e.)	.125 (.037)	.041 (.010)
H	9.89	5.79
I^2	99.0%	97.0%



Publications

Publications available at:

- [https://www.researchgate.net/profile/Christine Bauer](https://www.researchgate.net/profile/Christine_Bauer)
- [http://bach.wu.ac.at/bachapp/cgi-bin/fides/fides.aspx/fides.aspx?
search=true;person=true;show=pub;tid=9856;lang=EN](http://bach.wu.ac.at/bachapp/cgi-bin/fides/fides.aspx/fides.aspx?search=true;person=true;show=pub;tid=9856;lang=EN)