

Re-use and Patterns of Digital Innovation in Open Crowds: Linking Empirical and Computational Approaches

Sabine Brunswicker
Purdue University

Michael Prietula
Emory University

*Poster prepared for the Organization Science Winter Conference 23, February 2-7, 2017.
WORKING PAPER IS AVAILABLE ON SSRN: <https://ssrn.com/abstract=2908650>*

Organizations increasingly use crowds to create digital innovations, novel and useful software-based products or services (Yoo, Henfridsson, & Lyytinen, 2010). They invite a diverse crowd of individuals to solve an innovation problem in a virtual design contest, in which the crowd competes for the best performance (Boudreau, Lacetera, & Lakhani, 2011; Gulley, 2001; Hildebrand, Häubl, Herrmann, & Landwehr, 2013). Participating in a design contest for digital innovations requires developers to search for a novel solution along (at least) two fundamental design dimensions: the *user-facing functionality* set or features (i.e., what can the user “do” with this new digital ‘app’ and how is it presented?), and the underlying *technology* (the code), the technological solution that affords this user functionality set (i.e. how is this set realized by code?). Thus, in order to win the contest, developers need to create designs that are both useful and usable, and score well in both dimensions.

Recently, organizations implement virtual contests for digital innovations in an increasingly transparent way, following the logic of ‘open source’ (Levine & Prietula, 2014). They provide participants with access to information about others’ submissions throughout the competitive and multi-staged process, and encourage several iterations of reuse instead of independent and parallel development efforts (Brunswicker, Jensen, Hsiang, & Song, 2016; Wooten & Ulrich, 2015). In short, they foster truly open crowds, in which participants freely share, reuse and remix their digital applications. We define reuse as the process by locating and using knowledge freely shared by others (Alavi & Leidner, 2001; Majchrzak, Cooper, & Neece, 2004).

In such virtual design contests, this process is transparent if it provides access to two kinds of social information that are absent in non-transparent contests: (1) information about other design implementations (the actual code and feature) and (2) information about the quality of other solutions (the score). In essence, this transparency shapes how each developer searches for better performing designs across the collectively evolving design space. In this paper, we focus on the reuse process in open and transparent crowds that aggregates ‘bottom-up’ from the heterogeneous individual reuse behaviors (Newell, 1990).

Existing work already suggests that transparency moderates the collective design behavior in open crowds. While some authors argue that transparency leads to convergence towards the best performing design (Bayus, 2013; Boudreau & Lakhani, 2015), do other studies suggest that transparency can also trigger discontinuities, even among those at the top that are aware of their greater chances of winning the contest (Brunswicker et al., 2016; Ren, Nickerson, Mason, Sakamoto, & Graber, 2014; Sakamoto, Tanaka, Yu, & Nickerson, 2011). Indeed, literature suggests that reuse patterns are complex and non-linear. One of the reasons for that is that individuals change their reuse strategy in response to transparency (Brunswicker

et al., 2016; Gulley, 2001). Transparency directly moderates this risk preference of the developers because it reveals the relative quality of one's submission. Such information triggers comparison processes among the competitors, which may trigger a developer to engage in riskier reuse of designs of much higher quality (Boyle & Shapira, 2012; March & Shapira, 1992; Michinov, Jamet, Métayer, & Le Hénaff, 2015). Such explorative reuse is risky because developers are not familiar with the design, its code and its features. Thus, if there is no need to take risk, developers may opt for a less risky reuse choice. They search and adopt solutions more similar to the own one (Kyriakou, Sabnis, & Nickerson, 2017; Majchrzak et al., 2004). In essence, transparency may trigger variability in terms of risk preferences throughout an iterative virtual contest. However, the choice of a certain reuse behavior is not only a question of risk preference. Developers may also have a strategic preferences for one or the other aspect of the design task (Hildebrand et al., 2013): While some developers may focus their efforts on improving the quality of the code, other's may focus their efforts on the user functionality (and its quality). However, such a functionality centered reuse preferences sometimes comes at the price of the quality of the code. Thus, reuse in transparent crowds may cause unique reuse and performance patterns as the participants may not focus their attention in a one-dimensional way.

In order to unpack how such micro-level variability in reuse choices unfolds at the collective level, this paper asks two simple research question: *How does transparency shape the performance patterns of reuse in open crowds for digital innovation?* We answer this question using an agent-based simulation model. This model extends the theory of variable risk preferences (Boyle & Shapira, 2012; March & Shapira, 1992) to represent the behavioral rules of individual agents. It accounts for the fact that in context of open crowds, developers design and reuse digital innovations along two dimensions: code (usable) and feature (useful).

Our results reveal unexpected discontinuous performance patterns. Design preferences, a bias on either code or feature, lowers the collective performance of the crowd. However, transparency insulates those at the top, the leaders of the competition, from such negative effects. It produces high performing designs even if developers focus their efforts more on code or feature. Further, while transparency reduces variability and smooths the patterns among those at the top, it also creates unique opportunities for 'newcomers' to submit high-quality applications.

References

References

- Afuah, A., & Tucci, C. L. 2012. Crowdsourcing As a Solution to Distant Search. **Academy of Management Review**, 37(3): 355-375.
- Alavi, M., & Leidner, D. E. 2001. Review: Knowledge Management and Knowledge Management Systems: Conceptual Foundations and Research Issues. **MIS Quarterly**, 25(1): 107-136.
- Bayus, B. L. 2013. Crowdsourcing New Product Ideas over Time: An Analysis of the Dell IdeaStorm Community. **Management Science**, 59(1): 226-244.
- Boudreau, K. J., Lacetera, N., & Lakhani, K. R. 2011. Incentives and Problem Uncertainty in Innovation Contests: An Empirical Analysis. **Management Science**, 57(5): 843-863.
- Boudreau, K. J., & Lakhani, K. R. 2015. "Open" disclosure of innovations, incentives and follow-on reuse: Theory on processes of cumulative innovation and a field experiment in computational biology. **Research Policy**, 44(1): 4-19.
- Boudreau, K. J., Lakhani, K. R., & Menietti, M. 2015. Performance Response to Skill Level across Competition across Skill-Levels in Rank Order Tournaments: Field Evidence and Implications for Tournament Design. In H. B. School (Ed.). Boston, MA.
- Boyle, E., & Shapira, Z. 2012. The Liability of Leading: Battling Aspiration and Survival Goals in theJeopardy!Tournament of Champions. **Organization Science**, 23(4): 1100-1113.
- Brunswicker, S., Hutschek, U., & Wagner, L. 2012. Exploration in the open innovation front-end: the role of technologies. **International Journal of Technology Intelligence and Planning**, 8(1): 1-15.
- Brunswicker, S., Jensen, B., Hsiang, C., & Song, C. 2016. Transparency as Design Choice in Open Innovation Contests, **Open and User Innovation Conference**. Boston: Harvard Business School.
- Burton, R. M. 2003. Computational Laboratories for Organization Science: Questions, Validity and Docking. **Computational and Mathematical Organization Theory**, 9(2): 91-108.
- Burton, R. M., & Obel, B. 2011. Computational Modeling for What-Is, What-Might-Be, and What-Should-Be Studies--And Triangulation. **Organization Science**, 22(5): 1195-1202.
- Cialdini, R. B., & Goldstein, N. J. 2004. Social Influence: Compliance and Conformity. **Annual Review of Psychology**, 55(1): 591-621.
- Cyert, R. M., & March, J. G. 1963. A behavioral theory of the firm. **Englewood Cliffs, NJ**, 2.
- Davis, J. P., Eisenhardt, K. M., & Bingham, C. B. 2007. Developing Theory Through Simulation Methods. **Academy of Management Review**, 32(2): 480-499.
- Feltovich, P. J., Prietula, M. J., & Ericsson, K. A. 2006. Studies of Expertise from Psychological Perspectives. In K. A. Ericsson, N. Charness, P. J. Feltovich, & R. R. Hoffman (Eds.), **The Cambridge Handbook of Expertise and Expert Performance**: 41-68. Cambridge: Cambridge University Press.
- Festinger, L. 1954. A Theory of Social Comparison Processes. **Human Relations**, 7(2): 117-140.

- Foss, N. J., & Weber, L. 2016. Moving Opportunism to the Back Seat: Bounded Rationality, Costly Conflict, and Hierarchical Forms. **Academy of Management Review**, 41(1): 61-79.
- Gigerenzer, G., & Todd, P. M. 1999. **Simple heuristics that make us smart**: Oxford University Press, USA.
- Greve, H. R. 1996. **Performance, Aspirations and Risky Organizational Change**. Paper presented at the Academy of Management Proceedings.
- Gulley, N. 2001. **Patterns of innovation: a web-based MATLAB programming contest**. Paper presented at the CHI 2001 Extended Abstracts on Human Factors in Computing Systems., Seattle, Washington.
- Harrison, J., Lin, Z., Carroll, G., & Carley, K. 2007. Simulation modeling in organizational and management research. **Academy of Management Review**, 32(4): 1229-1245.
- Hildebrand, C., Häubl, G., Herrmann, A., & Landwehr, J. R. 2013. When Social Media Can Be Bad for You: Community Feedback Stifles Consumer Creativity and Reduces Satisfaction with Self-Designed Products. **Information Systems Research**, 24(1): 14-29.
- Hills, T. T., Todd, P. M., Lazer, D., Redish, A. D., Couzin, I. D., & Cognitive Search Research, G. 2015. Exploration versus exploitation in space, mind, and society. **Trends Cogn Sci**, 19(1): 46-54.
- Kahneman, D., & Tversky, A. 1979. Prospect theory: An analysis of decision under risk. **Econometrica: Journal of the econometric society**: 263-291.
- Kyriakou, H., Sabnis, G., & Nickerson, J. 2017. Knowledge Reuse for Customization: Metamodels in an open design community for 3D printing. **MIS Quarterly**, 41.
- Levine, S. S., & Prietula, M. J. 2012. How Knowledge Transfer Impacts Performance: A Multilevel Model of Benefits and Liabilities. **Organization Science**, 23(6): 1748-1766.
- Levine, S. S., & Prietula, M. J. 2014. Open Collaboration for Innovation: Principles and Performance. **Organization Science**, 25(5): 1414-1433.
- Li, Y., Vanhaverbeke, W., & Schoenmakers, W. 2008. Exploration and Exploitation in Innovation: Reframing the Interpretation. **Creativity and Innovation Management**, 17(2): 107-126.
- Majchrzak, A., Cooper, L. P., & Neece, O. E. 2004. Knowledge Reuse for Innovation. **Management Science**, 50(2): 174-188.
- Majchrzak, A., & Malhotra, A. 2013. Towards an information systems perspective and research agenda on crowdsourcing for innovation. **The Journal of Strategic Information Systems**, 22(4): 257-268.
- Malhotra, A., & Majchrzak, A. 2014. Managing Crowds in Innovation Challenges. **California Management Review**, 56(4).
- March, J. G. 1991. Exploration and Exploitation in Organizational Learning. **Organization Science**, 2(1): 71-87.
- March, J. G., & Olsen, J. P. 2004. **The logic of appropriateness**: na.
- March, J. G., & Shapira, Z. 1992. Variable risk preferences and the focus of attention. **Psychological review**, 99(1): 172.
- March, J. G., & Sutton, R. I. 1997. Crossroads-organizational performance as a dependent variable. **Organization science**, 8(6): 698-706.
- Michinov, N., Jamet, E., Métayer, N., & Le Hénaff, B. 2015. The eyes of creativity: Impact of social comparison and individual creativity on performance and attention to others' ideas during electronic brainstorming. **Computers in Human Behavior**, 42: 57-67.

- Moliterno, T. P., Beck, N., Beckman, C. M., & Meyer, M. 2014. Knowing Your Place: Social Performance Feedback in Good Times and Bad Times. **Organization Science**, 25(6): 1684-1702.
- Newell, A. 1990. **Unified theories of cognition**. Cambridge, MA: Harvard University Press.
- Prietula, M. J., Carley, K. M., & Gasser, L. G. 1998. **Simulating organizations: computational models of institutions and groups**. AAAI Press/MIT Press: Menlo Park, California.
- Ren, J., Nickerson, J. V., Mason, W., Sakamoto, Y., & Graber, B. 2014. Increasing the crowd's capacity to create: how alternative generation affects the diversity, relevance and effectiveness of generated ads. **Decision Support Systems**, 65: 28-39.
- Sakamoto, Y., Tanaka, Y., Yu, L., & Nickerson, J. V. 2011. The Crowdsourcing Design Space. In D. D. Schmorrow, & C. M. Fidopiastis (Eds.), **Foundations of Augmented Cognition. Directing the Future of Adaptive Systems: 6th International Conference, FAC 2011, Held as Part of HCI International 2011, Orlando, FL, USA, July 9-14, 2011. Proceedings**: 346-355. Berlin, Heidelberg: Springer Berlin Heidelberg.
- Shafir, E. 2013. **The Behavioral Foundations of Public Policy**: Princeton University Press.
- Shinkle, G. A. 2012. Organizational Aspirations, Reference Points, and Goals: Building on the Past and Aiming for the Future. **Journal of Management**, 38(1): 415-455.
- Sidhu, J. S., Commandeur, H. R., & Volberda, H. W. 2007. The Multifaceted Nature of Exploration and Exploitation: Value of Supply, Demand, and Spatial Search for Innovation. **Organization Science**, 18(1): 20-38.
- Terwiesch, C., & Xu, Y. 2008. Innovation Contests, Open Innovation, and Multiagent Problem Solving. **Management Science**, 54(9): 1529-1543.
- Terwiesch, K., & Ulrich, K. 2009. **Innovation Tournaments: Creating and Selecting Exceptional Opportunities**. Boston, MA: Harvard Business School Press.
- Tversky, A., & Kahneman, D. 1986. Rational Choice and the Framing of Decisions. **The Journal of Business**, 59(4): 251-278.
- Wilensky, U. 1999. Netlogo, **Center for Connected Learning and Computer-Based Modeling**. Evanston, IL.: Northwestern University.
- Wilensky, U., & Rand, W. 2015. **An introduction to agent-based modeling: Modeling natural, social, and engineered complex systems with NetLogo**. Cambridge, MA: MIT Press.
- Wooten, J. O., & Ulrich, K. T. 2015. The Impact of Visibility in Innovation Tournaments: Evidence from Field Experiments. **Working Paper**.
- Yoo, Y., Henfridsson, O., & Lyytinen, K. 2010. Research Commentary—The New Organizing Logic of Digital Innovation: An Agenda for Information Systems Research. **Information Systems Research**, 21(4): 724-735.