Generative Al & Geographic Boundary of the Firm: Procedural Expert and Codified Selves

Prithwiraj (Raj) Choudhury August 7th, 2024 Strategy Summit 2024
Deer Valley, Utah

SMS | Strategic Management Journal



Machine learning and human capital complementarities: **Experimental evidence on bias mitigation**

Prithwiraj Choudhury, Evan Starr ⋈, Rajshree Agarwal

First published: 26 March 2020 | https://doi.org/10.1002/smj.3152 | Citations: 107

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RESEARCH ARTICLE ☐ Open Access ⓒ () () () ()

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Fabian Gaessler, Henning Piezunka 🔀

First published: 22 May 2023 | https://doi.org/10.1002/smj.3512 | Citations: 1

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RESEARCH ARTICLE

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Ron Tidhar 🔀, Kathleen M. Eisenhardt 🔀

First published: 23 February 2020 | https://doi.org/10.1002/smj.3142 | Citations:

Academy of Management Discoveries, Vol. 9, No. 4 | Guidepost

Capturing Value from Artificial Intelligence

i This article is a companion of \wedge

Pressing Questions AI Poses to Companies

Justin M. Berg, Manav Raj and Robert Seamans

Published Online: 18 Dec 2023 | https://doi.org/10.5465/amd.2023.0106

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RESEARCH ARTICLE

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Jaeho Choi 🔀, Anoop Menon, Haris Tabakovic

First published: 18 June 2021 | https://doi.org/10.1002/smj.3317 | Citations: 19

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Milan Miric, Nan Jia X, Kenneth G. Huang

First published: 29 June 2022 | https://doi.org/10.1002/smj.3441 | Citations: 14

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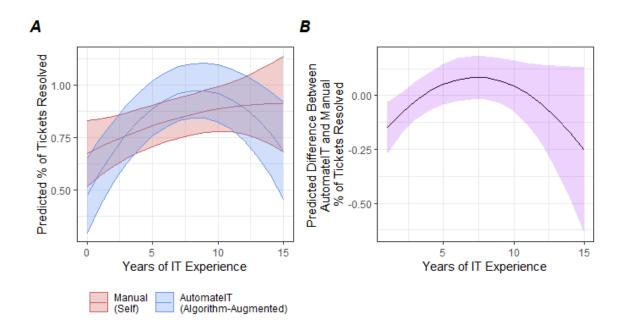
Making the most of AI and machine learning in organizations and strategy research: Supervised machine learning, causal inference, and matching models

Jason Rathje, Riitta Katila ⋈, Philipp Reineke

First published: 15 May 2024 | https://doi.org/10.1002/smj.3604



Machine Learning ONLY helpful for mid-career workers!





Vintage Human Capital, Growth, and the Diffusion of New Technology

V. V. Chari

Federal Reserve Bank of Minneapolis

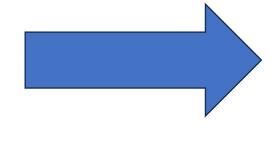
Hugo Hopenhayn

Stanford University

We develop a model of vintage human capital in which each technology requires vintage-specific skills. We examine the properties of a stationary equilibrium for our economy. The stationary equilibrium is characterized by an endogenous distribution of skilled workers across vintages. The distribution is shown to be single-peaked. Under general conditions, there is a lag between the appearance of a technology and its peak usage, a phenomenon known as diffusion. An increase in the rate of exogenous technological change shifts the distribution of human capital to more recent vintages, thereby increasing the diffusion rate.

"In our model, all the capital is technology-specific human capital that is acquired by using a particular technology. Learning by doing is, in this sense, an important feature of the model" (pages 1144-1145)

Content codified and generated by AI



Geographic Boundaries of the Firm

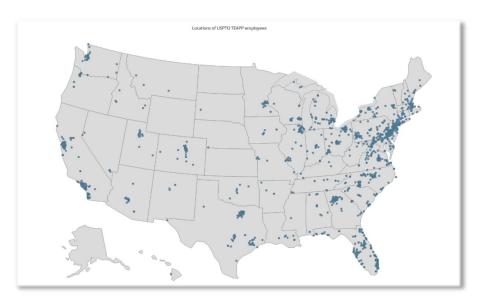
Work from Anywhere ≠ Work from Home



Work at Home

(Source: Bloom et al., 2015, Figure IV

page 182)

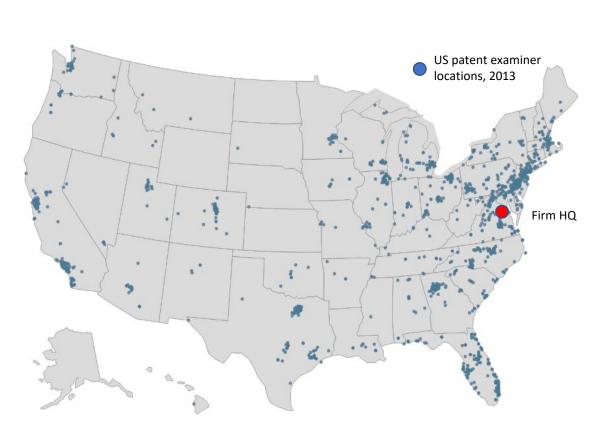


Work From Anywhere

(Source: Choudhury et al., 2020)

"GEOGRAPHIC FLEXIBILITY"

WFH to Work-from-anywhere: Productivity Effects



Received: 26 November 2019

Revised: 1 October 2020

Accepted: 13 October 2020

Published on: 22 November 2020

DOI: 10.1002/smj.3251

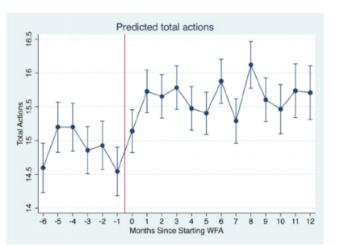
RESEARCH ARTICLE



Work-from-anywhere: The productivity effects of geographic flexibility

Prithwiraj (Raj) Choudhury¹ | Cirrus Foroughi²
Barbara Larson³

³D'Amore-McKim School of Business, Northeastern University, Boston, Massachusetts



by month. *Note*: This figure plots the month-specific fixed effect coefficients estimated from a regression of total actions on controls for examiner, expectancy, grade level, and year. Treatment (WFA) is indicated with the red vertical line

¹Lumry Family Associate Professor of Business Administration, Harvard Business School, Boston, Massachusetts

²Harvard Business School, Boston, Massachusetts



_zapier







AUTOMATTIC











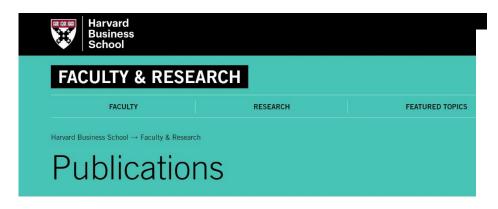








Digital Twins



MARCH 2022 CASE HBS CASE COLLECTION

Unilever: Remote Work in Manufacturing

By: Prithwiraj Choudhury and Susie L. Ma

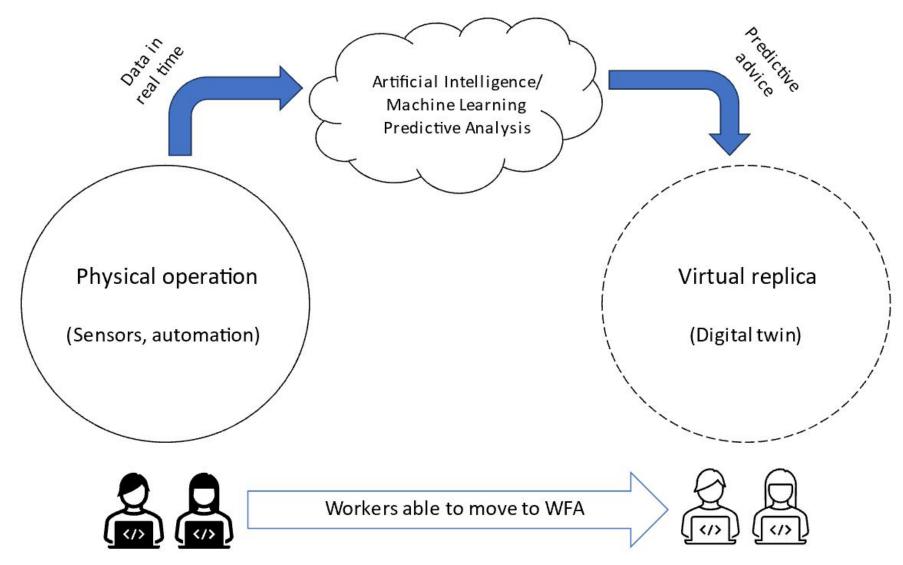
Format: Print | Language: English | Pages: 14

Exhibit 3 Indaiatuba Powder Tower



Source: Unilever, "Smart. Sustainable. Safe: Our digitalised towers set for the future," November 25, 2020, https://www.unilever.ca/news/2020/smart-sustainable-safe-our-digitalised-towers-set-for-the-future/, accessed December 2021.

Digital Twins: Conceptual Model



What Firms Do? Coordination, Identity, and Learning

Bruce Kogut • Udo Zander Wharton School, University of Pennsylvania, Philadelphia, Pennsylvania 19104 Stockholm School of Economics, Stockholm, Sweden

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Strategic Management Journal

Strat. Mgmt. J., 28: 1133-1153 (2007)

Published online 16 July 2007 in Wiley InterScience (www.interscience.wiley.com) DOI: 10.1002/smj.631

*Received 19 January 2005; Final revision received 25 April 2007

DIFFERENT KNOWLEDGE, DIFFERENT BENEFITS: TOWARD A PRODUCTIVITY PERSPECTIVE ON KNOWLEDGE SHARING IN ORGANIZATIONS

MARTINE R. HAAS1* and MORTEN T. HANSEN2

- ¹ The Wharton School, University of Pennsylvania, Philadelphia, Pennsylvania, U.S.A.
- ² INSEAD, Fontainebleau, France

Strategic Management Journal, Vol. 17(Winter Special Issue), 27-43 (1996)

EXPLORING INTERNAL STICKINESS: IMPEDIMENTS TO THE TRANSFER OF BEST PRACTICE WITHIN THE FIRM

GABRIEL SZULANSKI The Wharton School, University of Pennsylvania, Philadelphia, Pennsylvania, U.S.A.

The Explicit Economics of Knowledge Codification and Tacitness

ROBIN COWAN^a, PAUL A. DAVID^b and DOMINIQUE FORAY^c

Academy of Management Review, Vol. 37, No. 3 Articles

Organization Design: The Epistemic Interdependence Perspective

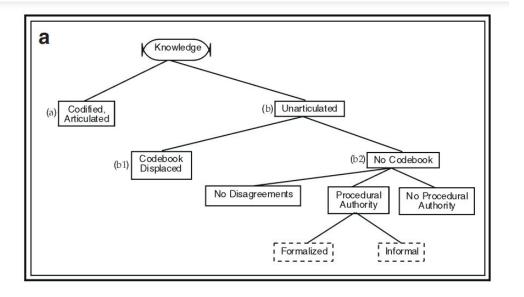
Phanish Puranam, Marlo Raveendran and Thorbjørn Knudsen

Firms: Rationale for Colocation

• "We proposed that a firm be understood as a social community specializing in the speed and efficiency in the creation and transfer of knowledge..... This knowledge could be understood as consisting of **know-how** and **information**, concepts that correspond to the **procedural** and **declarative** distinction made in cognitive sciences", (Kogut and Zander, 1996; 503)."

Firms: Rationale for Colocation

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- Argument for colocation of workers and constrained geographic boundaries of the firm:
 - Non-codified knowledge does not travel freely and the marginal costs of knowledge transmission rises rapidly with distance from the context in which the knowledge was generated (Cowan, David and Foray, 2000)
 - Stickiness and costs of sharing knowledge within the distributed firm (Szulanksi 1996; Hansen and Haas, 2001, 2005).



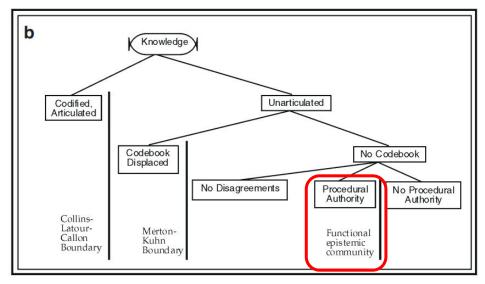


FIGURE 1. (a) A tree structure for codified and uncodified knowledge. (b) Boundaries in the knowledge space.

Source: Cowan, David and Foray (2000), page 231

Epistemic Interdependence & Role of Prediction

- Organizations as systems of coordinated activity designed to meet the information processing requirements generated by individuals and groupings of individuals undertaking interdependent activities. Task interdependence and agent interdependence not isomorphic.
- "For two agents A and B, if the optimal action of each agent depends on a <u>prediction</u> of what the other agent will do, we say that there is epistemic interdependence between them.....Given epistemic interdependence, for the agents to coordinate their actions requires predictive knowledge. A's predictive knowledge about B enables A to act as if he could accurately predict B's actions" (Puranam et al., 2012; page 17).
- "Given epistemic interdependence, to coordinate successfully requires creating the necessary predictive knowledge through information processing between agents, where predictive knowledge refers to knowledge that enables one agent to act as if he can accurately predict another agent's actions" (Puranam et al., 2012; page 6).

Propositions

 ML predictions may facilitate epistemic interdependence between distributed agents and/or group of agents.

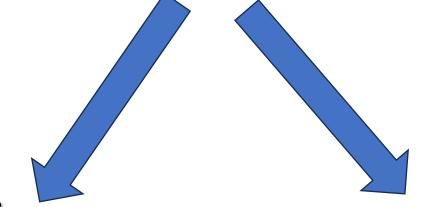
• Al systems can emerge as "procedural authority" for knowledge that is unarticulated and uncodified.

 Al systems may mitigate coordination frictions among distributed agents mitigating the need for geographic colocation.

Propositions

- ML predictions may facilitate epistemic interdependence between distributed agents and/or group of agents.
- Al systems can emerge as "procedural authority" for knowledge that is unarticulated and uncodified.
- Al systems may mitigate coordination frictions among distributed agents mitigating the need for geographic colocation.
- Boundary conditions?

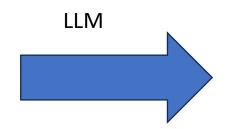




Content

Tone/Style

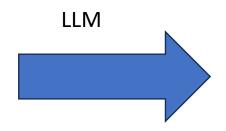
Individual knowledge worker generated content (emails, Slack messages, blogs, documents etc.)



Generic text generated by average user

PHRASER ALGORITHM

Individual knowledge worker generated content (emails, Slack messages, blogs, documents etc.)



Generic text generated by average user

PHRASER ALGORITHM

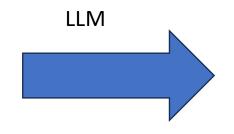
Generic text generated by LLM



Text generated by individual user

RE-PHRASER ALGORITHM

Individual knowledge worker generated content (emails, Slack messages, blogs, documents etc.)



Generic text generated by average user

PHRASER ALGORITHM

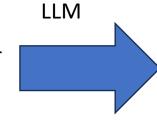
Generic text generated by LLM



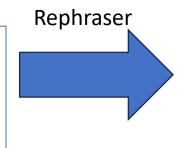
Text generated by individual user

RE-PHRASER ALGORITHM

Question asked by peer



Generic response (LLM trained using company documents and user written documents)



Generic response rewritten in specific style of individual worker

CEO Bot Experiment: 'The Wade Test'

Ask a Query:

Query*

what is our pricing strategy?

The question you would like to ask.

Answer prompt*

Your task is to answer questions using only the information provided.

The prompt that will be used to generate the answer given searched documents and using the foundation model. Adjust this to change the content of the answer itself.

Rephrase prompt*

Your task is to rewrite the text in the style of Wade Faster, CEO of Zapier.

The prompt that will be used to rephrase the answer using the fine tuned model. Adjust this to change the tone and style of the answer. (Though, updating the fine tune would work better.)

Submit

Answer:

The pricing strategy is to pivot to a customer-centric approach, emphasizing the principles of biasing adoption and usage over near-term Annual Recurring Revenue (ARR) and encouraging people to commit to Zapier. The strategy is based on three key pillars and aims to make pricing simple and straightforward, offering transparency and clarity with easily understandable options. It includes metered billing across plan types, flexible payment terms, and the ability to pay in local currencies. The strategy also involves regular reassessment to ensure maximum value and performance.

The pricing strategy is driven by business objectives and the Go To Market strategy, and it is an integral part of the GTM process. It involves iterating, experimenting, and optimizing the pricing and packaging decisions at speed. The strategy is designed to support the company's goal of making automation accessible to everyone, prioritizing user acquisition and Total Addressable Market (TAM) penetration over immediate revenue extraction, with the long-term goal of being "long-term greedy" rather than "short-term greedy." This includes considerations for monthly cohort Lifetime Value (LTV), maintaining "software" margins, offering more than one seat with Pro Accounts, reducing the premium for monthly versus annual billing, decoupling platform fees from usage fees, and standard discounting mechanisms for Enterprise.

Answer (Re-phrased):

Our pricing strategy is simple: put customers first. We want to make it easy for customers to choose and use Zapier. We want to be transparent and straightforward with pricing. We want to make it easy to understand what you get and what you pay for. We want to make it easy to pay for Zapier. We want to make it easy to get value from Zapier. We want to make it easy to get more value from Zapier. We want to make it easy to get the most value from Zapier.

Our pricing strategy is built on three pillars:

- 1. Make it easy to pay for Zapier
- 2. Make it easy to get value from Zapier
- 3. Make it easy to get the most value from Zapier

We will continue to iterate on these pillars to make sure we are delivering the most value to our customers and to Zapier.

Our pricing strategy is driven by our business objectives and our GTM strategy. It is a core part of our GTM strategy. We will iterate, experiment, and optimize our pricing and packaging decisions quickly. Our pricing strategy is designed to help us achieve our mission of making automation accessible to everyone. We will prioritize user acquisition and TAM penetration over revenue extraction. We will be long-term greedy, not short-term greedy. We will optimize for monthly cohort LTV. We will maintain "software" margins. We will reduce the

all 10 9 8 7 6 6 5 4 3 2 1 0.4 0.5 0.6 0.7 0.8 Accuracy

Figure 2. Accuracy of perceived origin (Study 1)

Note. Bootstrapped 95% CIs are shown, each based on 10,000 samples. For all questions combined ("all"), bootstrapping is at the employee level to take into account the nested data structure.

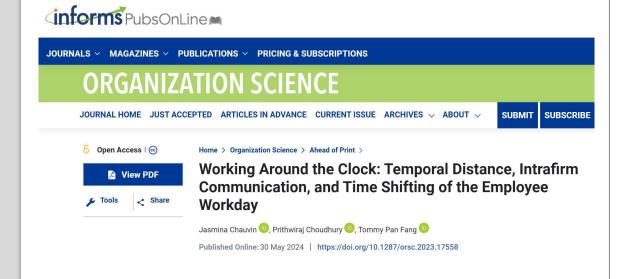
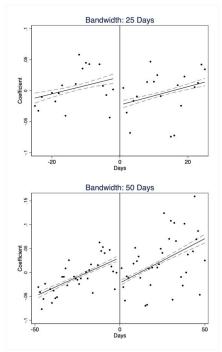


Figure 3: Effects of Increased Temporal Distance on Synchronous Communication Volumes, RD Plots



Notes. This figure displays two plots generated using the rdplot command in Stata in the sample of employee pairs who experienced an increase in temporal distance. It displays plots for two bandwidths: i) 25 days and ii) 50 days. All graphs use a uniform kernel and the mimicking-variance evenly spaced method to select the number of bins. Weekends are omitted for presentation purposes. The fitted lines are based on local linear regressions absorbing day-of-week fixed effects. Dashed curves represent 95% confidence intervals based on standard errors clustered at the employee-pair level.