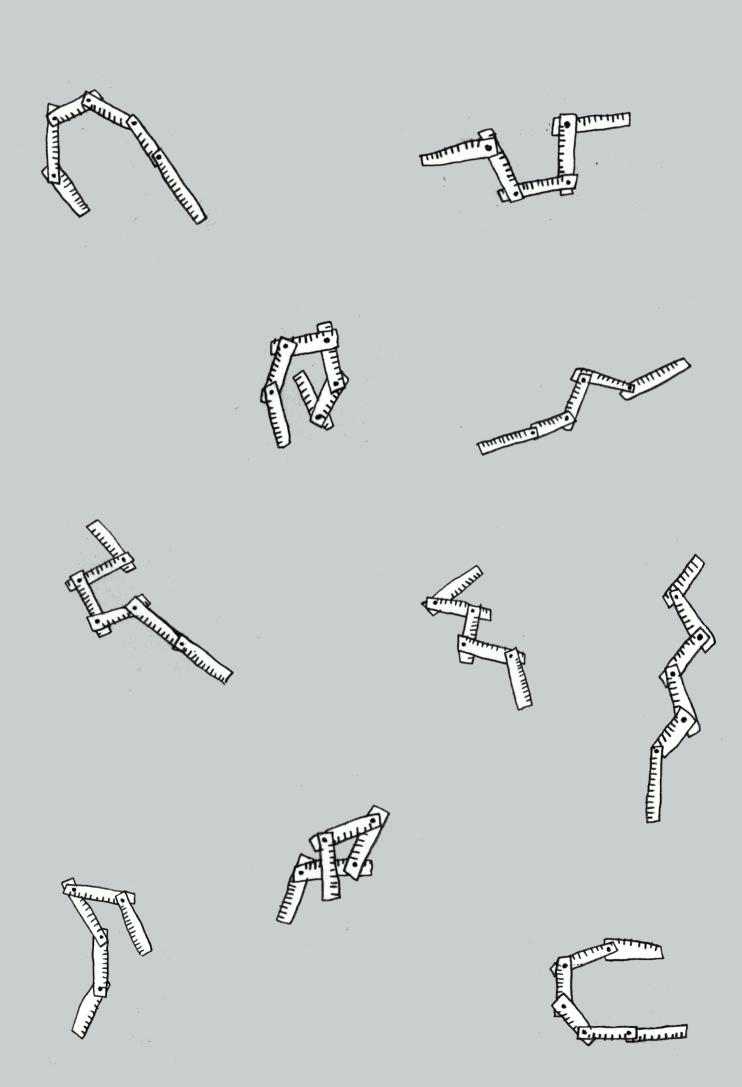
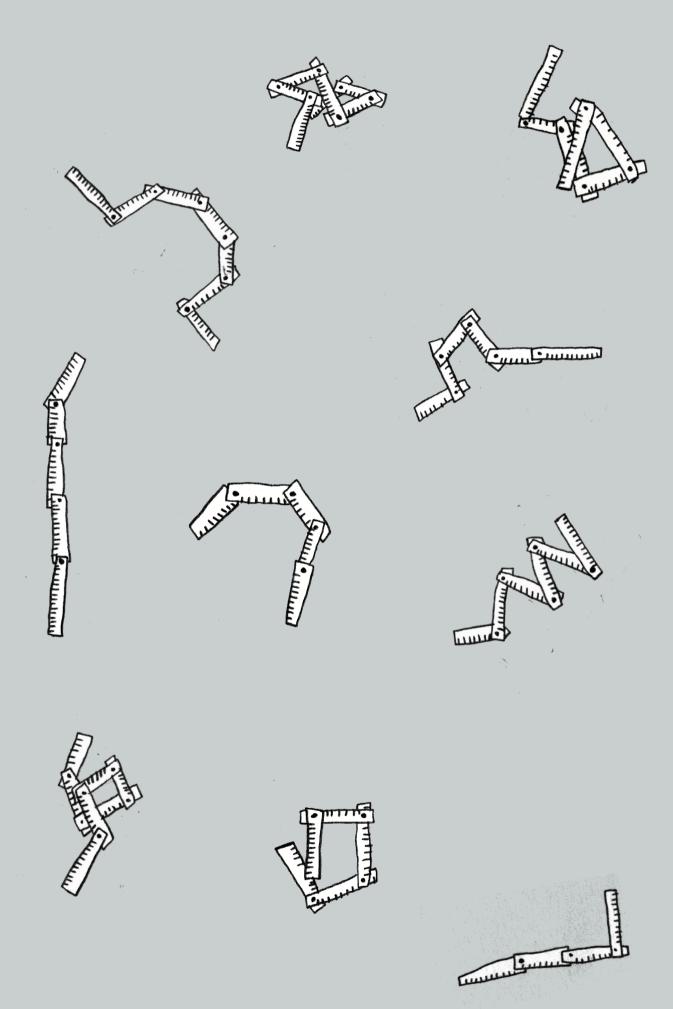


Miriam Beblo, Zoe Branczyk & Julia Schneider aka Doc J Snyder





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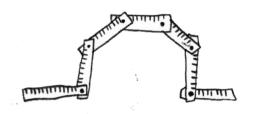
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How to Build on Every Talent (in the Labor Market)





Why this comic essay? Why now?

Today, diversity is often framed as a problem, not a solution.

But data and research say otherwise.

In this comic, we present Proof of Work:
literal proof (empirical evidence) of tools that make labor
markets work better and fairer. To level the playing field.
And to build on every talent.

We need all backgrounds, all skills, all perspectives — and yes, this includes you. But here's the thing: diverse teams are not always easy to form or lead.

We all carry stereotypes, fears, and blind spots.

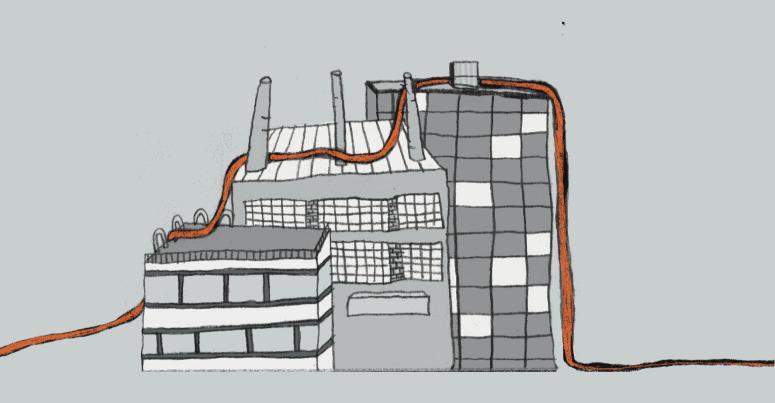
The selection of tools we present has been tested, reviewed, and published in academic journals such as AER, QJE, and Econometrica (yes, that's a real thing).

This comic is also a love letter to research and a counter-narrative to those voices that spread myths about merit, diversity, and fairness.

Take it with you on a train. Use it in a workshop. Slide it across the desk to your CEO.

(Or maybe you are the CEO.)

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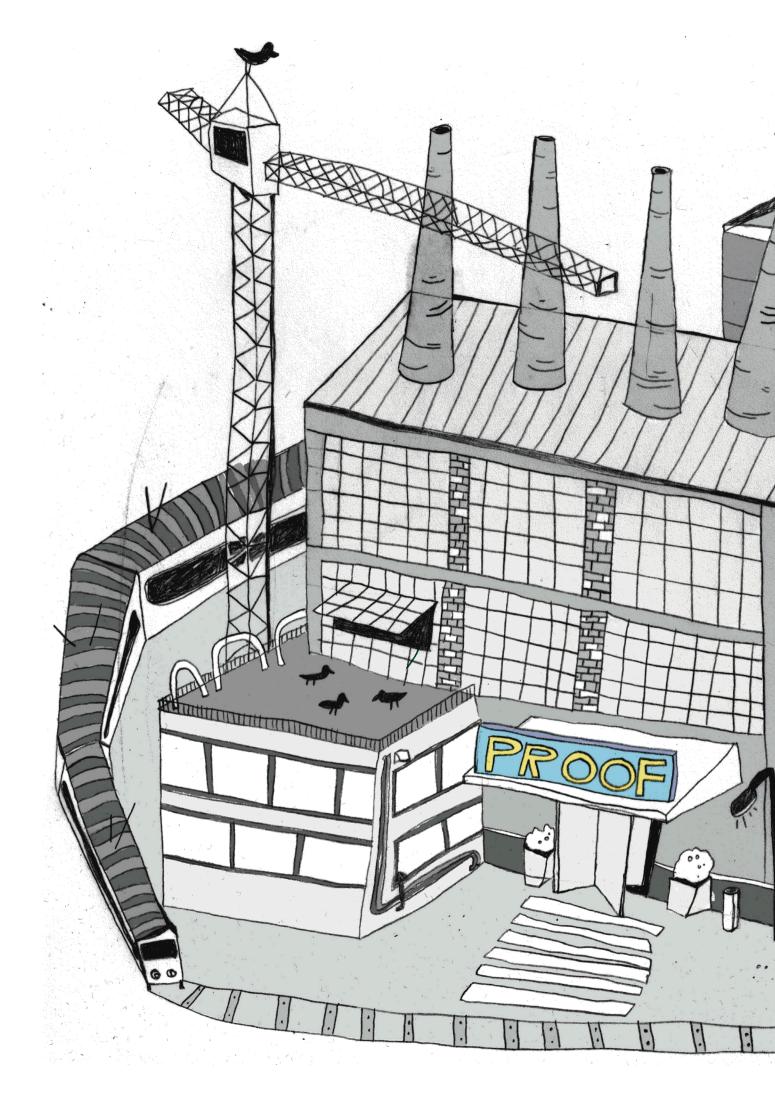


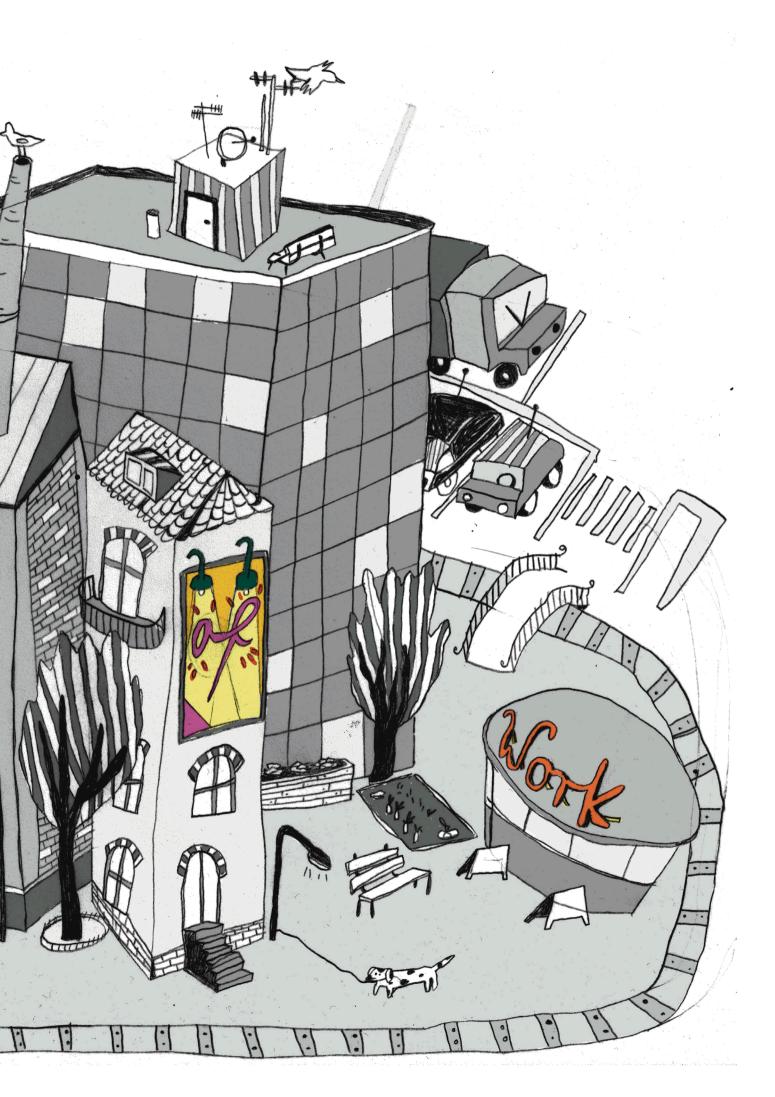
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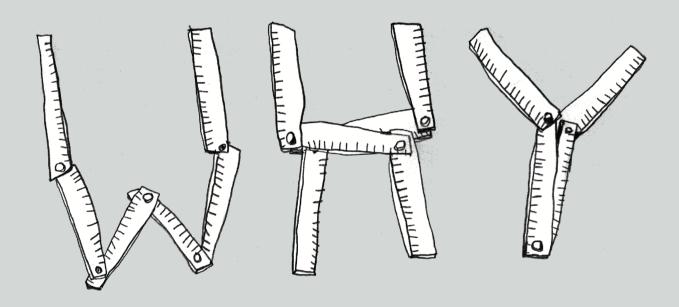
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Chapter I:



1. Diversity & Growth

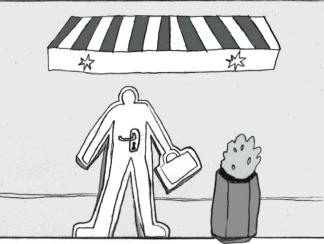
Chang-Tai Hsieh, Erik Hurst, Charles I. Jones & Peter J. Klenow (2019). The Allocation of Talent and U.S. Economic Growth. Econometrica, 87(5): 1439–1474.



^{*} The researchers used data from the U.S. Census (1960) and the Current Population Survey (1964–2010) to estimate how occupational shifts contributed to aggregate U.S. productivity growth. They applied counterfactual simulations based on a Roy model.

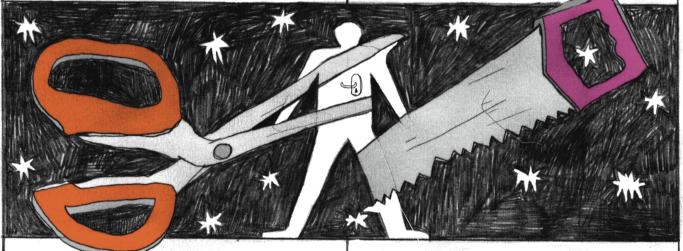


From rags to riches: if you're talented and work hard, you'll make it.



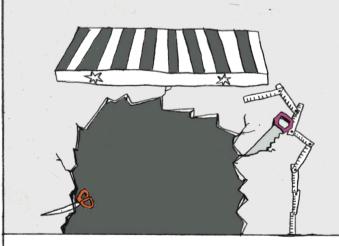
But across the world, barriers remain

— by race, gender, health, caste, class,
or connections.



Researchers studied the U.S., where occupation barriers began to fall in the 1960s.

They wanted to know whether the economy grows when more people gain access to high-skilled jobs.



The result? 20–40% of U.S. growth came from better allocation of talent.



Exclusion isn't just unfair. It's inefficient.

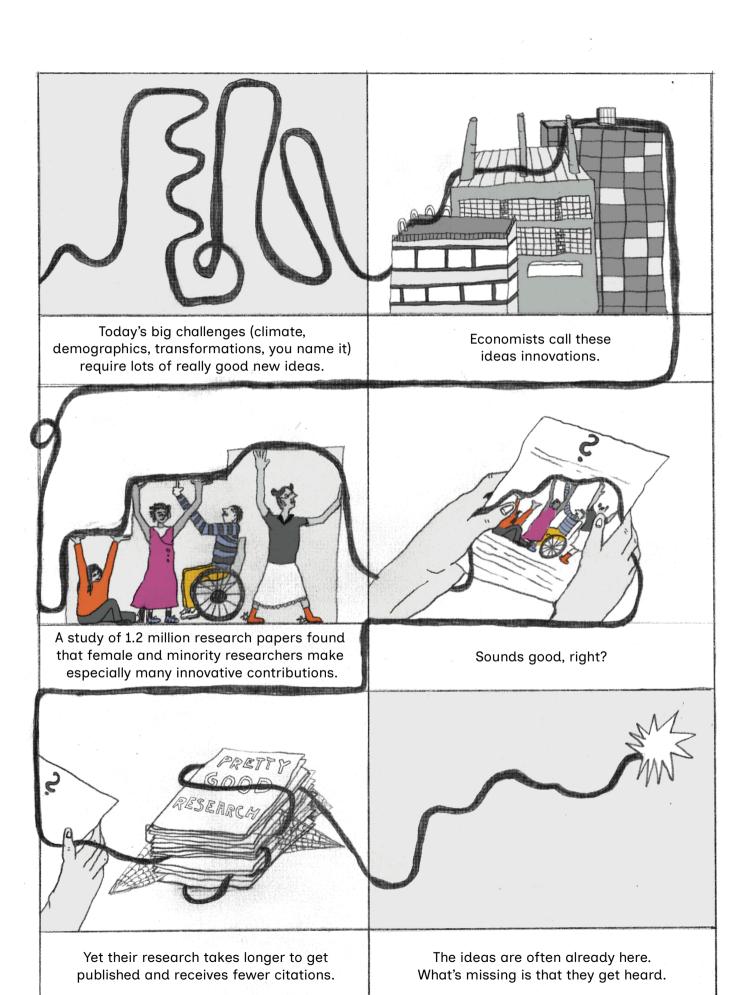


2. Diversity & Innovation

Bas Hofstra, Vivek V. Kulkarni, Sebastian Munoz-Najar Galvez, Bryan He, Dan Jurafsky & Daniel A. McFarland (2020). The Diversity–Innovation Paradox in Science. Proceedings of the National Academy of Sciences, 117(17): 9284–9291.

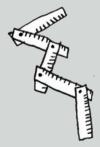


^{*} Researchers studied the scientific novelty and citation impact of 1.2 million research papers from U.S. doctoral recipients from 1977 to 2015. They applied text analysis and machine learning to abstracts to determine the scientific novelty of the papers.

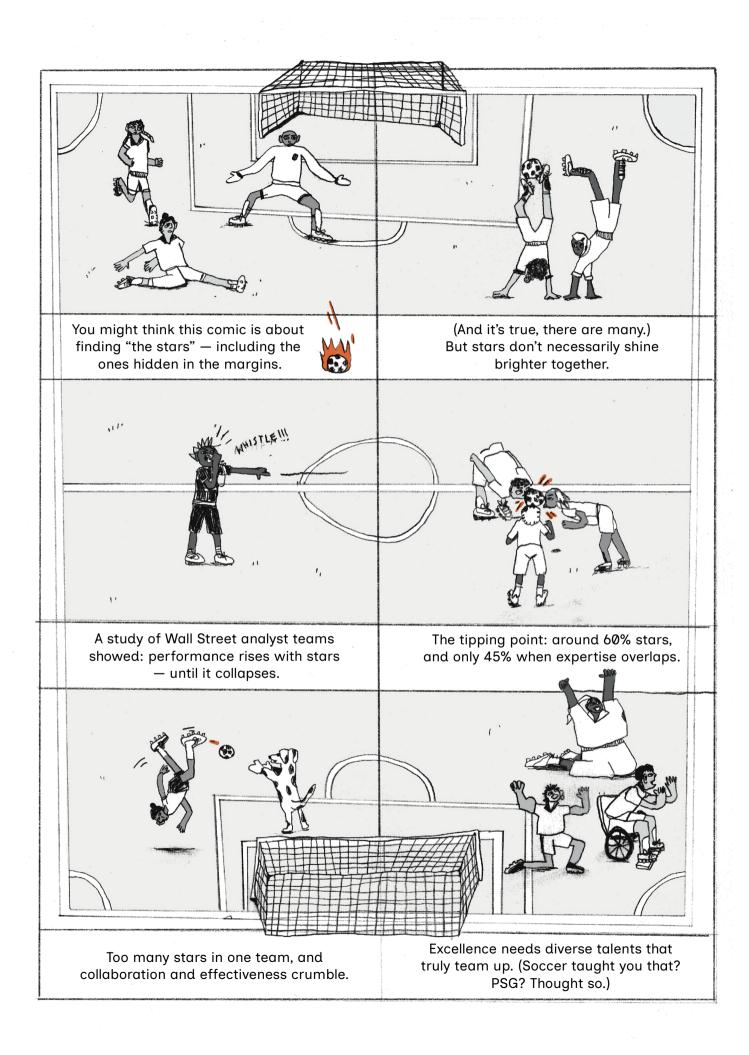


3. Diversity & Excellence

Boris Groysberg, Jeffrey T. Polzer & Hillary Anger Elfenbein (2011). Too Many Cooks Spoil the Broth: How High-Status Individuals Decrease Group Effectiveness. Organization Science, 22(3): 722–737.



^{*} The researchers analyzed data from U.S. Wall Street equity research departments collected between 1996 and 2001. They examined client ratings of departmental performance using panel regression models.

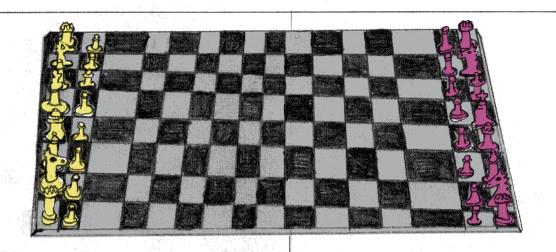


4. Diversity & Market Forces

Kaushik Basu (2017). Discrimination as Focal Point: Markets and Group Identity. Forum for Social Economics, 46(2): 128–138.

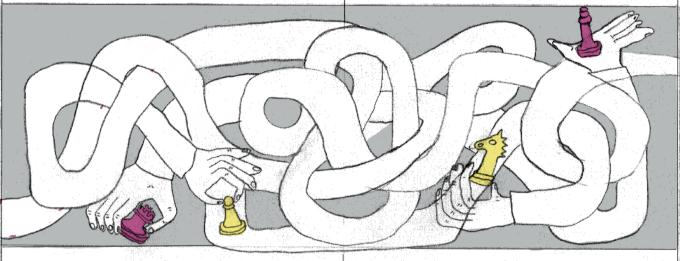


^{*} The researcher presents a theory of discrimination where employers end up discriminating against certain groups, although all groups are a priori identical.



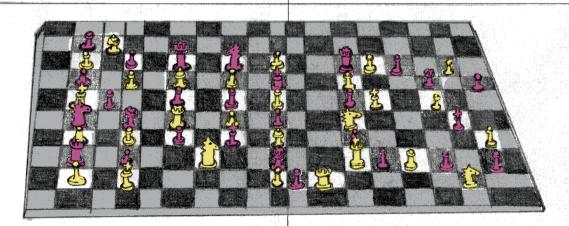
Diversity drives growth, innovation and excellence.

Yet, markets sometimes reward hiring from the same group,



when it is beneficial that all tasks are done by just one (strategic complementarity).

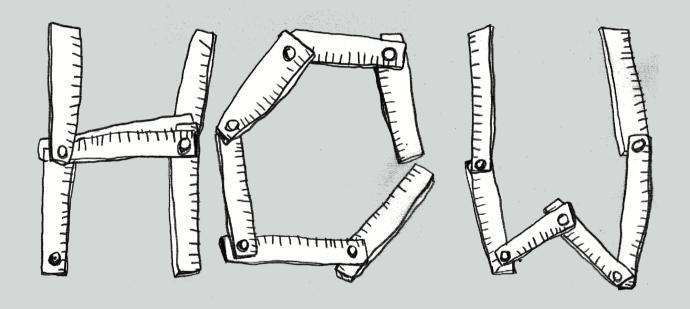
Discrimination not due to prejudice, but coordination gains.



In that case, market forces suppress diversity rather than promote it...

...and policy interventions become essential.

Chapter II:

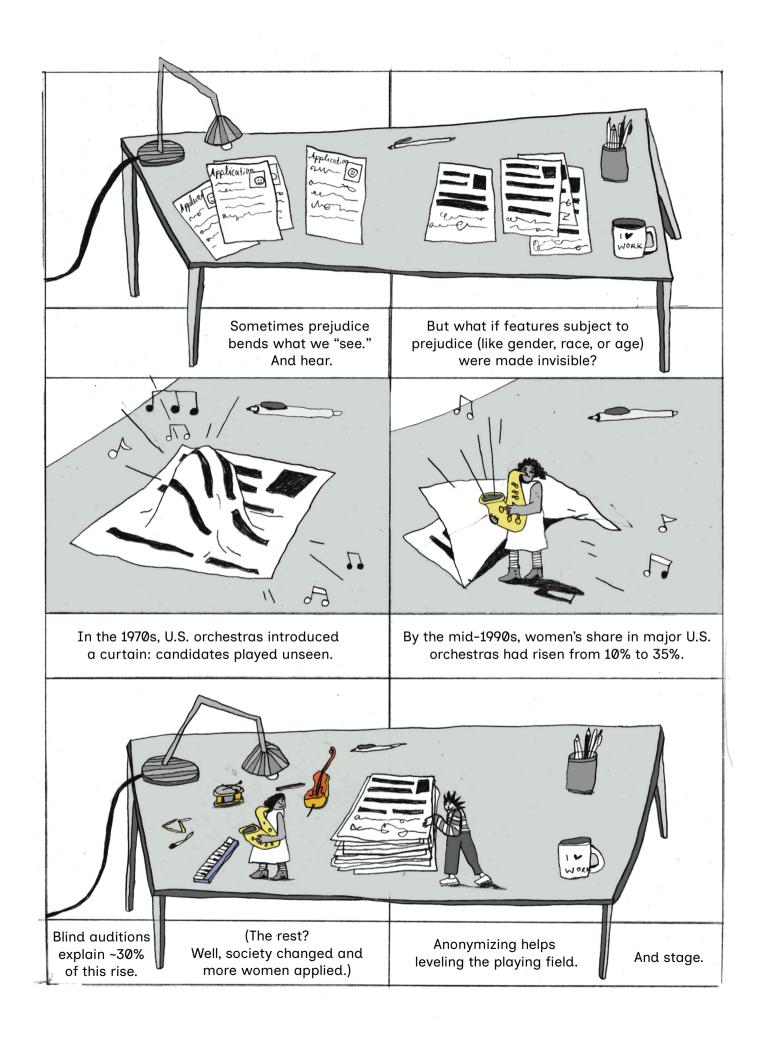


1. Anonymizing Hiring

Claudia Goldin & Cecilia Rouse (2000). Orchestrating Impartiality: The Impact of "Blind" Auditions on Female Musicians. American Economic Review, 90(4): 715–741.



^{*} Researchers used data from U.S. orchestras to study female advancement and hiring. The data included auditions from eight orchestras from the 1950s to 1995 and rosters from 11 orchestras from 1947 to 1996. The researchers applied difference-in-differences and individual fixed effects (blind vs. non-blind) to their data.



2. Rewarding Performance Over Presence

Julian V. Johnsen, Hyejin Ku & Kjell G. Salvanes (2024). Competition and Career Advancement. Review of Economic Studies, 91: 2954–2980.

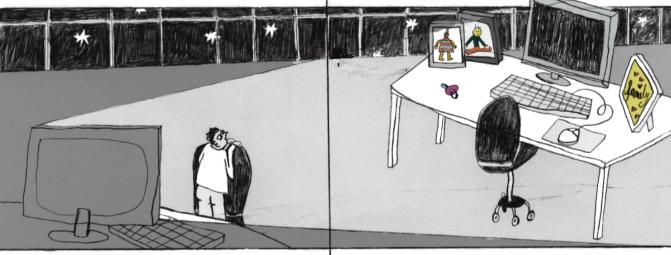


^{*} With linked employer–employee administrative data, covering the entire Norwegian population, and a difference-in-differences strategy, the authors analyzed men's career trajectories before and after the 1993 paternity leave reform.



We like to think promotions are based on hard work and results.

But in many workplaces, being seen matters more. Long hours. No leave.



This culture rewards presence, not performance. It penalizes caregivers, and anyone who values balance.

Researchers studied a paternity leave reform in Norway that encouraged fathers to take time off.



And indeed: Promotions went not to those who performed better but to those who stayed while others were on parental leave.



Rewarding performance, not presence, can break this trap.

3. Revealing Unconscious Bias

Alberto Alesina, Michela Carlana, Eliana La Ferrara & Paolo Pinotti (2024). Revealing Stereotypes: Evidence from Immigrants in Schools. American Economic Review, 114(7): 1916–1948.



^{*} The researchers examined the grade gap between native and immigrant students, overall failure rates, and how teachers graded mock exams with randomized student names. The study drew on data from Italy and combined a field experiment (2016/17) — randomizing the timing of personal IAT feedback — with an online experiment (2021) that randomized personalized versus generic debiasing interventions. The analysis used difference-indifferences models with teacher fixed effects.



We like to think we grade fairly — based on effort and ability.

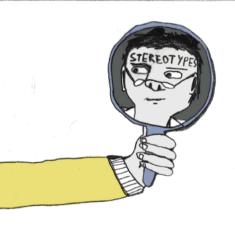
But stereotypes skew judgment.



In an Implicit Association Test (IAT), 70–80% of Italian middle school teachers showed moderate-to-strong bias against immigrants.



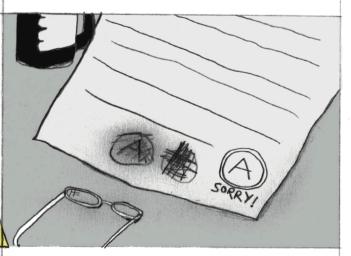
When teachers were told their own bias before grading their pupils, the grade gap between native and immigrant pupils shrank by ~ 1/3 of a grade...



... and failure rates of immigrant pupils halved.



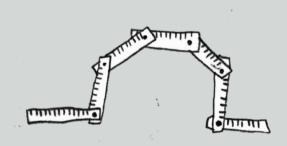
Generic bias messages helped, too; but strongly biased teachers needed to see their own IAT score in order to change.



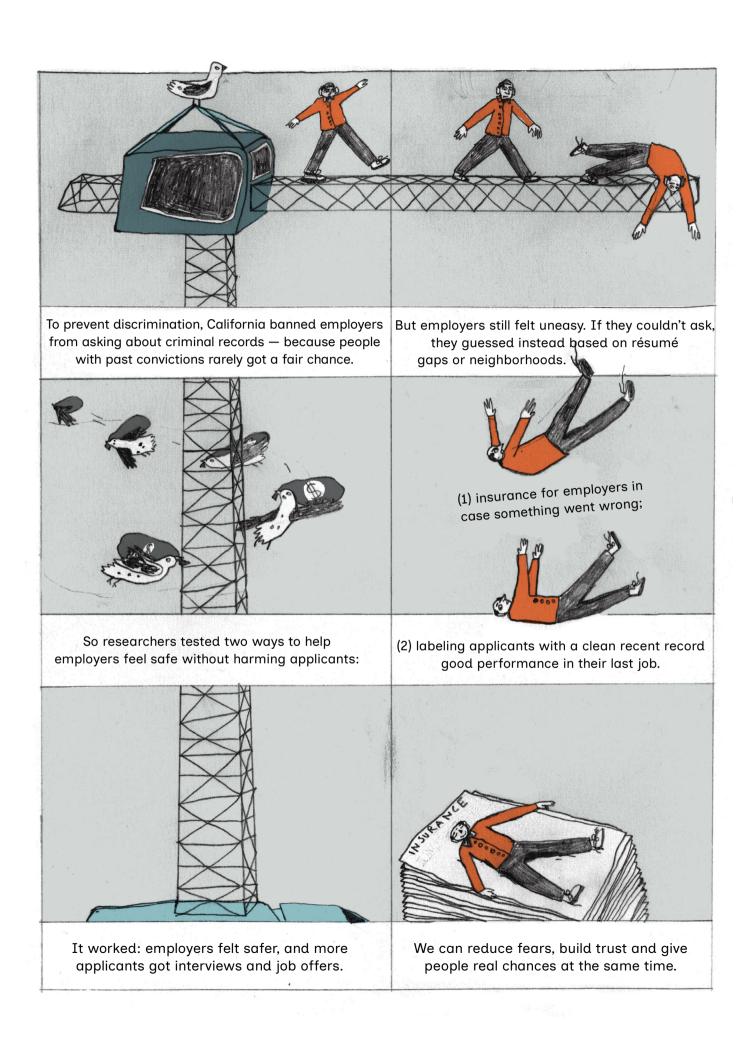
Revealing unconscious bias is a low-cost way to cut discrimination. In schooling and hiring.

4. Building Trust

Zoë Cullen, Will Dobbie & Mitchell Hoffman (2023). Increasing the Demand for Workers with a Criminal Record. Quarterly Journal of Economics, 138(1): 103–150.



^{*} In a 2020 field experiment on a large U.S. job platform, 1,500 hiring managers rated hypothetical applicants with and without criminal records. Each applicant was randomly assigned a trust signal as described.

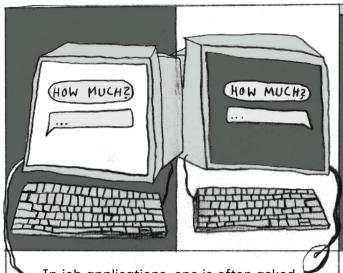


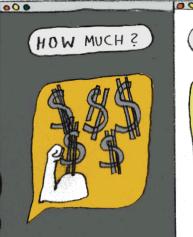
5. Leveling the Pay

Nina Roussille (2024). The Role of the Ask Gap in Gender Pay Inequality. Quarterly Journal of Economics, 139(3): 1557–1610.



^{*} The researcher compared the ask and bid salaries of a subset of candidates on the platform, for whom the answer box changed from empty to prefilled (with the median bid salary over the past 12 months for their respective location, job title and experience) before and after the change in 2018.





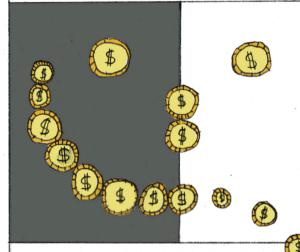


HOW

MUCHS

In job applications, one is often asked to state one's desired salary.

Women tend to have lower expectations and ask for less than men (and employers grant them less), despite similar resumes.



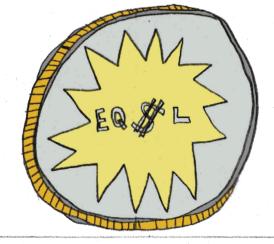
HOW MUCH?

AMOUNT

his ask gap explains most of the residual gender pay gap (i.e., after accounting for qualifications) on a US recruitment platform for high-wage engineering jobs.

The platform changed the way candidates had to state their desired salary from an empty box to a box prefilled with the typical offer for such a job.





With this information, women increased their ask salaries, eliminating the ask gap, whereupon employers increased their bid salaries.

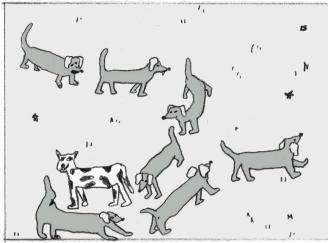
Wage information helps to correct women's underassessment of their market value, leveling the pay and reducing the gender pay gap.

6. Diversifying Leadership

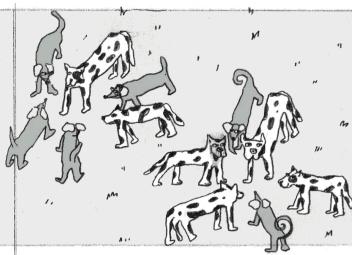
Christopher F. Karpowitz, Stephen D. O'Connell, Jessica Preece & Olga Stoddard (2024). Strength in Numbers? Gender Composition, Leadership, and Women's Influence in Teams. Journal of Political Economy, 132 (9): 3077–3112.



^{*} The researchers ran two field experiments over time with students enrolled in a competitive accounting program and a more general education course in the United States. They recorded weekly group meetings and team building exercises and collected individual survey responses and program performances to compare women's influence in the teams. They also assigned the gender of the team leader at random (treatment).



In male-dominated groups, women's input is often undervalued. That's why a "token woman" in a corporate board may have little effect.



In two field experiments, researchers observed mixed-gender groups with either male or female majorities.

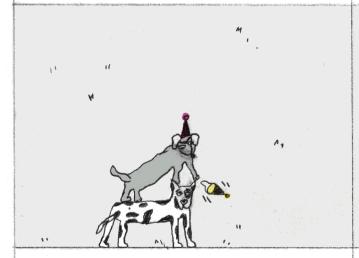


When in minority, women were less likely perceived as influential or chosen as spokespersons.

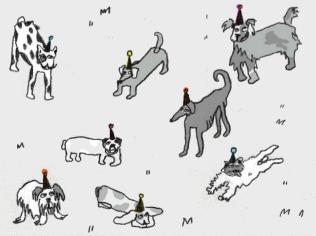
Minority men did equally well as majority men.



But when a team leader was randomly assigned, female-led groups gave more voice to women than male-led ones.



The reason? Women less often vote for women, men more often for men.



Making women lead reduces discrimination. It might work for other "minorities" too.

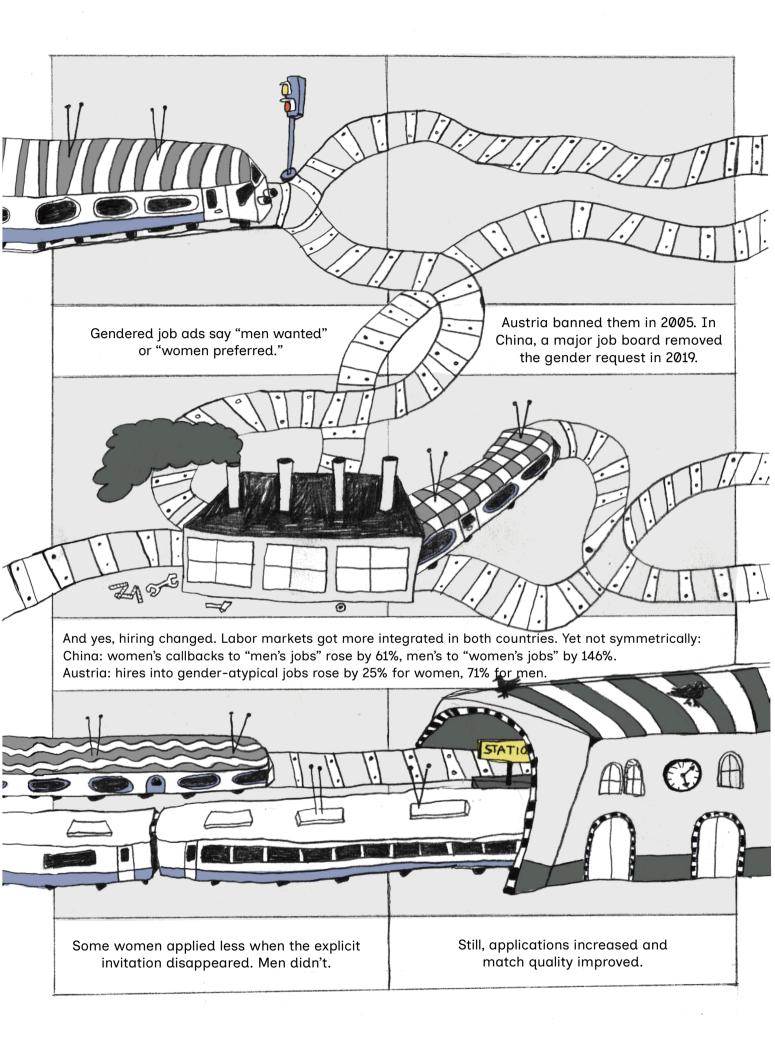
7. Ungendering Job Ads

David Card, Fabrizio Colella & Rafael Lalive (2024). Gender Preferences in Job Vacancies and Workplace Gender Diversity. Review of Economic Studies, 92 (4): 2437–2471.

Peter Kuhn & Kailing Shen (2023). What Happens When Employers Can No Longer Discriminate in Job Ads? American Economic Review, 113 (4): 1013–1048.



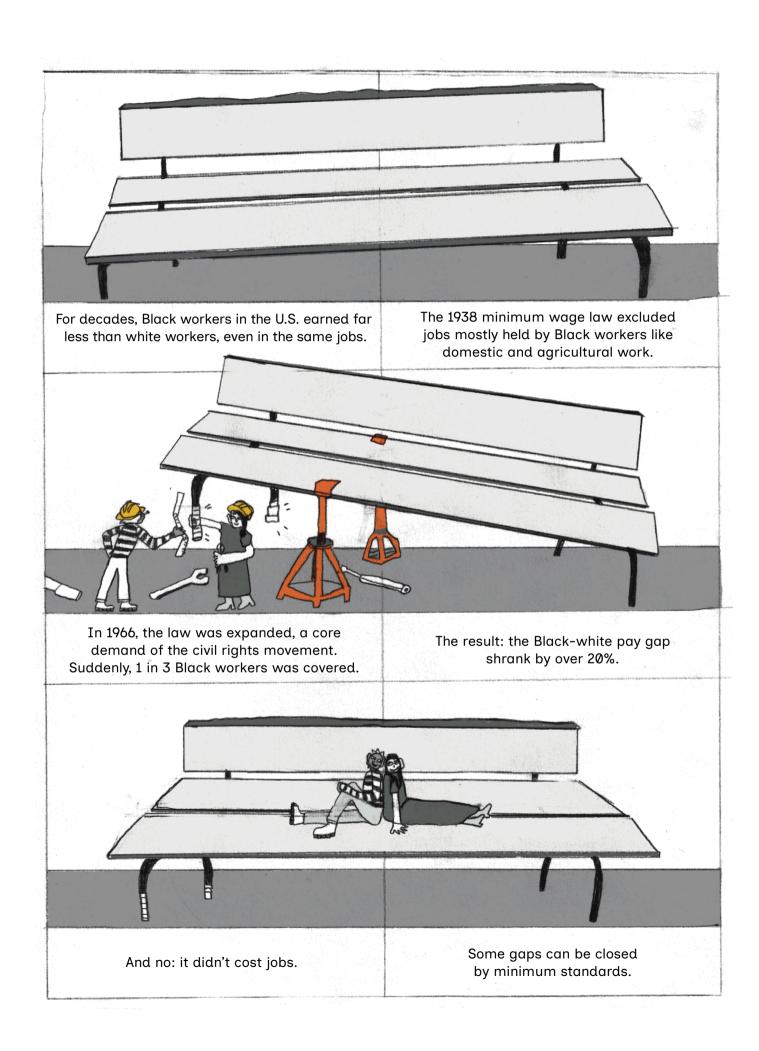
^{*} Researchers studied the impact of banning gendered job ads on hiring into gender-atypical jobs, callbacks, applications, and workplace diversity. One study was based on data from Austria (AMS job ads linked to social security records, 2000–2010), the other one on China (117,390 job ads and 3.1 million applications, 2019 policy change). A difference-in-differences method was employed for the Austrian data and a regression discontinuity approach for the Chinese data.



8. Closing the Gap From Below

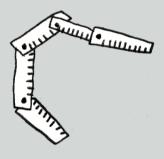
Ellora Derenoncourt & Claire Montialoux (2021). Minimum Wages and Racial Inequality. Quarterly Journal of Economics, 136 (1): 169–228.

^{*} Researchers studied the wages and employment of Black and white workers in newly covered vs. unaffected industries using data from the U.S. Census microdata (1960) and the CPS (1967–1980) and applying the difference-in-differences method.

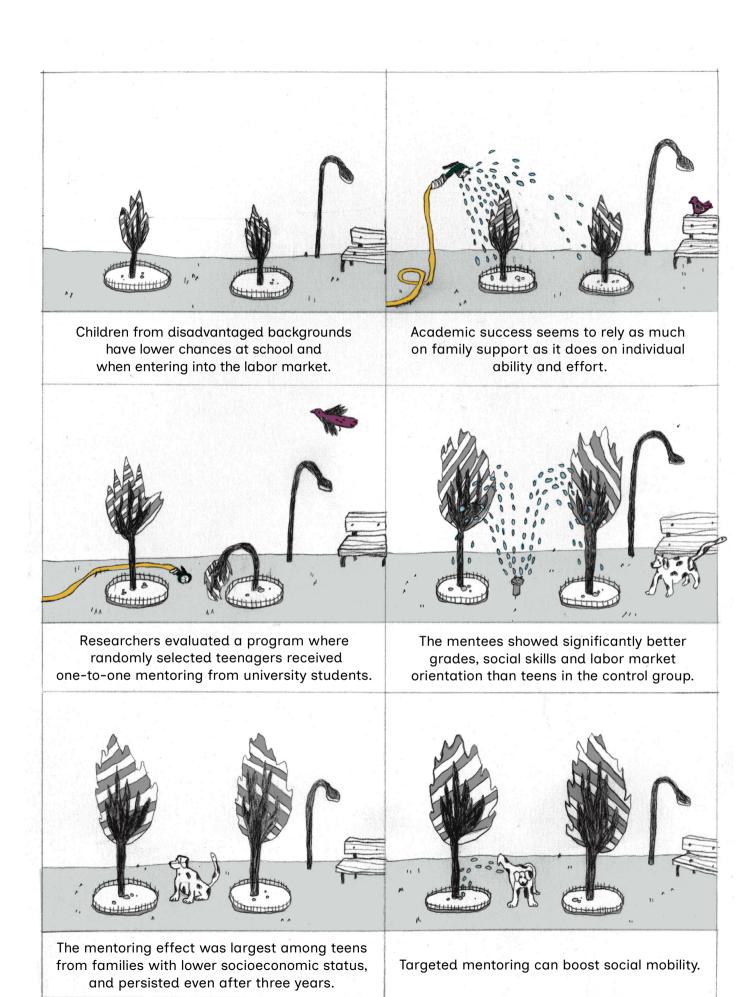


9. Mentoring

Sven Resnjanskij, Jens Ruhose, Simon Wiederhold, Ludger Woessmann & Katharina Wedel (2024). Can Mentoring Alleviate Family Disadvantage in Adolescence? A Field Experiment to Improve Labor Market Prospects. Journal of Political Economy, 132 (3): 1013–1062.

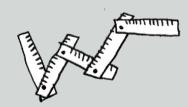


^{*} In a randomized-controlled trial in Germany, researchers randomly assigned teenagers from disadvated and non-disadvanted background to either a mentoring program or a control gruop. The study measured academic performance, self-confidence, social skills through surveys and standardized tests before and after the intervention.

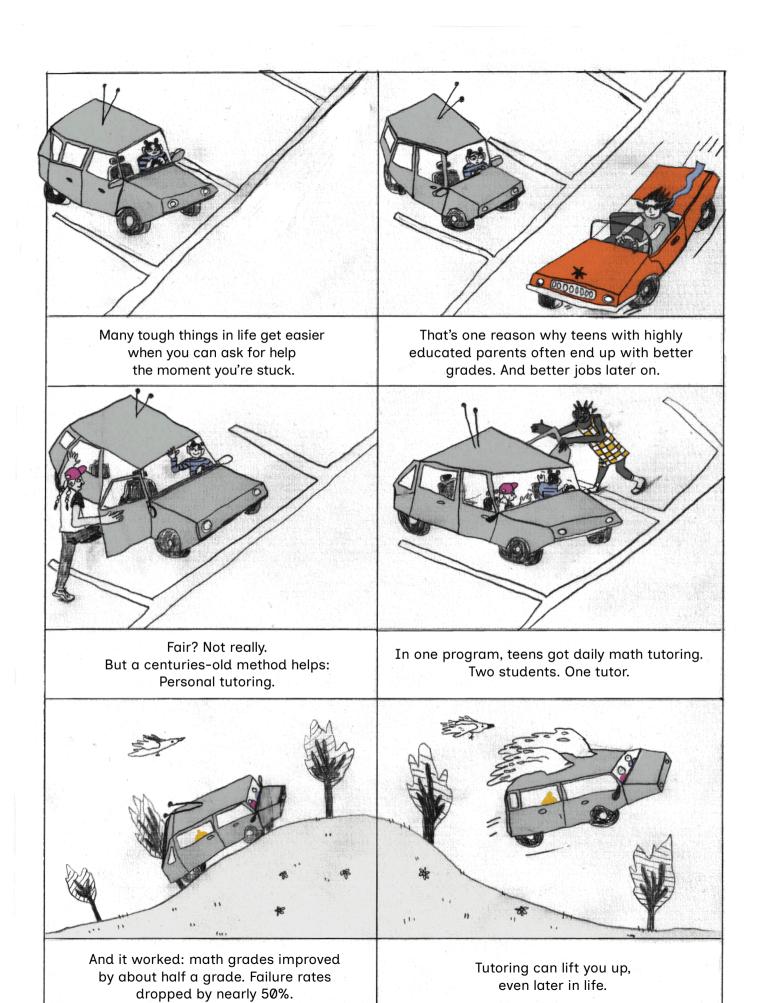


10. Tutoring

Jonathan Guryan, Jens Ludwig, Monica P. Bhatt, Philip J. Cook, Jonathan M. V. Davis, Kenneth Dodge, George Farkas, Roland G. Fryer Jr., Susan Mayer, Harold Pollack, Laurence Steinberg & Greg Stoddard (2023). Not Too Late: Improving Academic Outcomes among Adolescents. American Economic Review, 113 (3): 738–765.



^{*} The study analyzed data from 5,343 ninth- and tenth-grade students in Chicago public high schools between 2013 and 2015. It measured math test scores, grades, and course failure rates. The researchers conducted two randomized controlled trials providing daily 45–50 minute two-on-one math tutoring sessions by paraprofessional tutors.

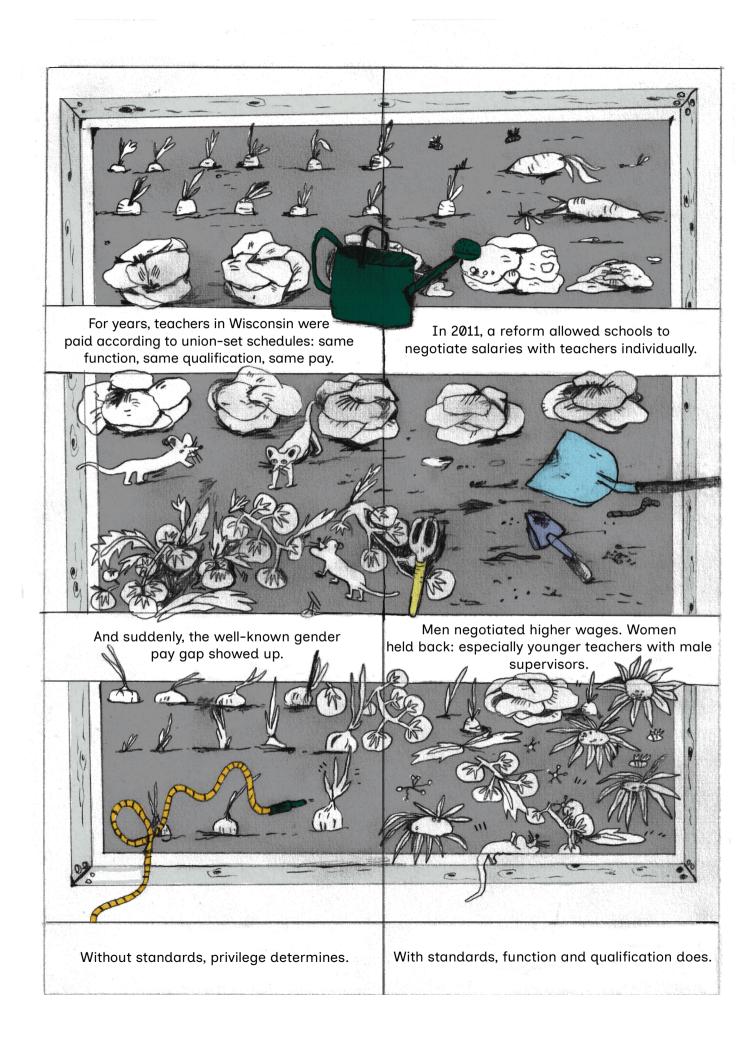


11. Standardizing Careers

Barbara Biasi & Heather Sarsons (2022). Flexible Wages, Bargaining, and the Gender Gap. Quarterly Journal of Economics, 137 (1): 215–266.



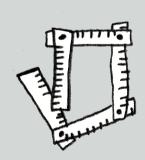
^{*} Using administrative data on Wisconsin's public school teachers from 2006 to 2016, together with a statewide survey of about 10,000 teachers, the authors examined the outcome of gender pay gaps after the expiration of collective bargaining agreements. They applied event-study and difference-in-differences methods.

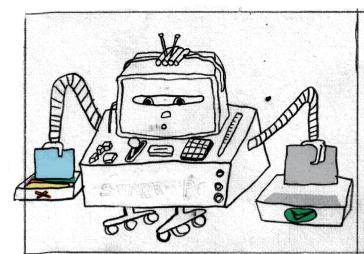


12. Exploring AI

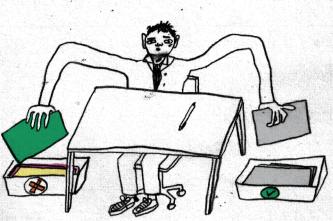
Danielle Li, Lindsey Raymond & Peter Bergman (2025). Hiring as Exploration. Review of Economic Studies, advance article: 1–41.

^{*} The study analyzed 88,666 job applications submitted between 2016 and 2019 to a U.S. Fortune 500 professional services firm. It compared two résumé-screening algorithms, focusing on interview-to-hire rates and the demographics of hires. The supervised learning model predicted hiring success solely from past data, while the exploration model — a contextual bandit — combined predicted success with an exploration bonus for uncertain candidates.

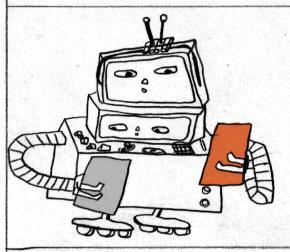




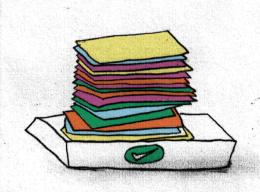
AI in hiring is controversial. Too often it just recommends the "same as always" type of candidate.



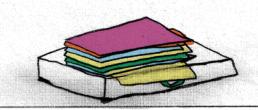
Humans are not fairer. We often ignore stronger candidates based on "gut feeling."



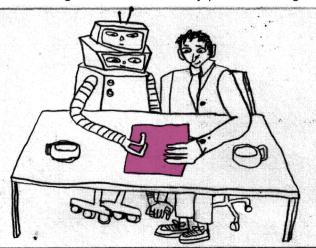
Standard AI learns from past hiring data and copies past bias.



Researchers tested a new algorithm: it gave extra chances to underrepresented candidates. More of them got invited and many proved strong.



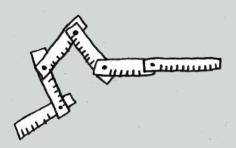
Result: The interview-to-hire rate tripled (27% vs. 10% humans). Minority share rose to 15% (vs. 10% humans and 5% standard AI).



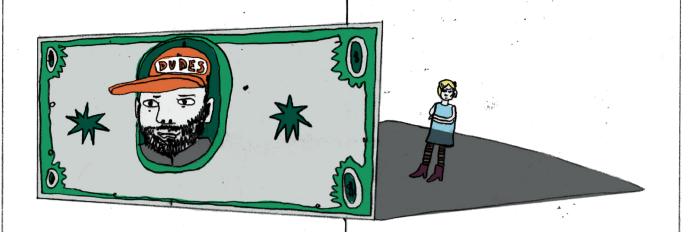
Exploring AI can reveal hidden potential and make hiring more efficient. Depending on desire, design and data.

13. Granting Financial Control

Erica Field, Rohini Pande, Natalia Rigol, Simone Schaner, and Charity Troyer Moore (2021). On Her Own Account: How Strengthening Women's Financial Control Impacts Labor Supply and Gender Norms. American Economic Review, 111 (7): 2342–75.



^{*} The study is based on a randomized field experiment in 197 villages in rural India with 5851 eligible couples.



Sometimes, people are held back in the labor market, because there are strong ideas of who should have a job and who shouldn't.

Well, those norms are responsive, as researchers detected in rural India.





They motivated women to take a job by giving them (a) their own bank account, (b) training in using the account and (c) wage deposits on their account (instead of their husband's).

And indeed, those women who received the training and the deposit were employed more often than women who only got an account.



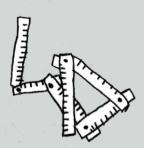


Over time, the empowered women even adapted their norms about working and what they perceived as accepted behavior.

Granting financial control helps to liberate female emloyment from norms.

14. Fixing Beliefs

Leonardo Bursztyn, Alessandra L. González & David Yanagizawa-Drott (2020). Misperceived Social Norms: Women Working Outside the Home in Saudi Arabia. American Economic Review, 110 (10): 2997–3029.



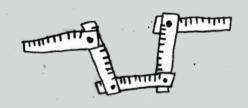
^{*} The researchers first ran a survey to find out about men's private support of women working outside the home in Saudi Arabia. They then used the survey result in an experiment where they informed only one half of the participants about the agreement rate and observed whether the behavior changed in comparison to the uninformed control group.



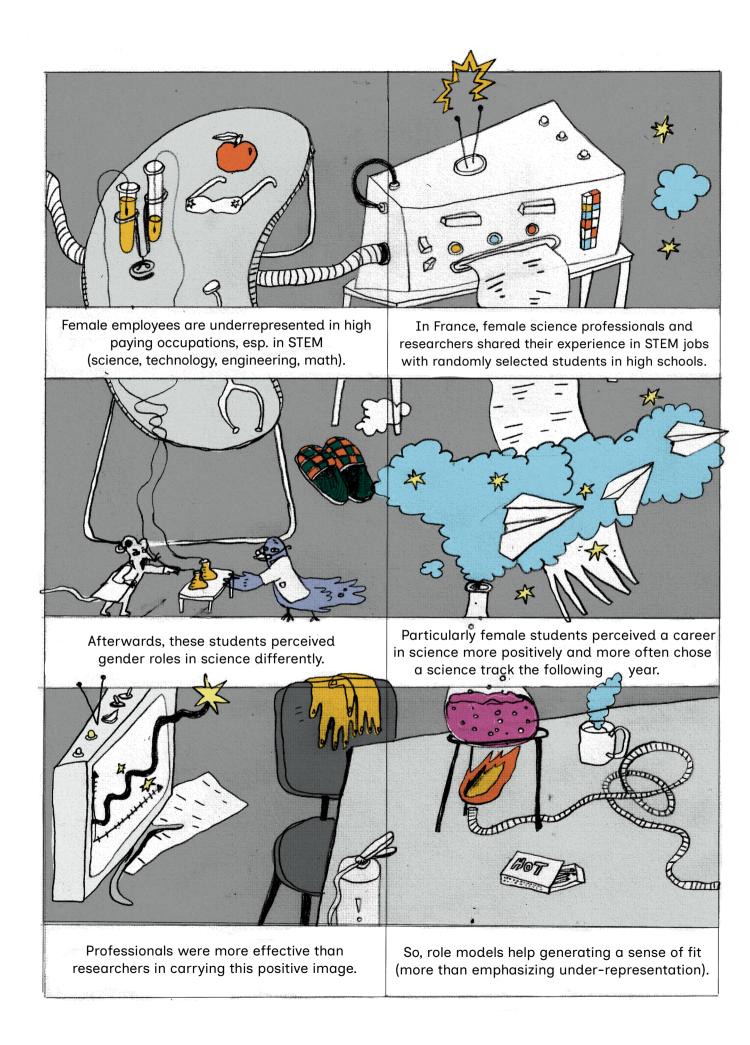


15. Generating a Sense of Fit

Thomas Breda, Julien Grenet, Marion Monnet & Clémentine Van Effenterre (2023). How Effective Are Female Role Models in Steering Girls Towards STEM? Evidence from French High Schools. Economic Journal, 133 (653): 1773–1809.

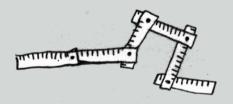


^{*} The researchers studied classroom interventions in a field experiment in France. About 3000 high school students in grade 10 and 12 were exposed to 56 female role models working in a science-related job.

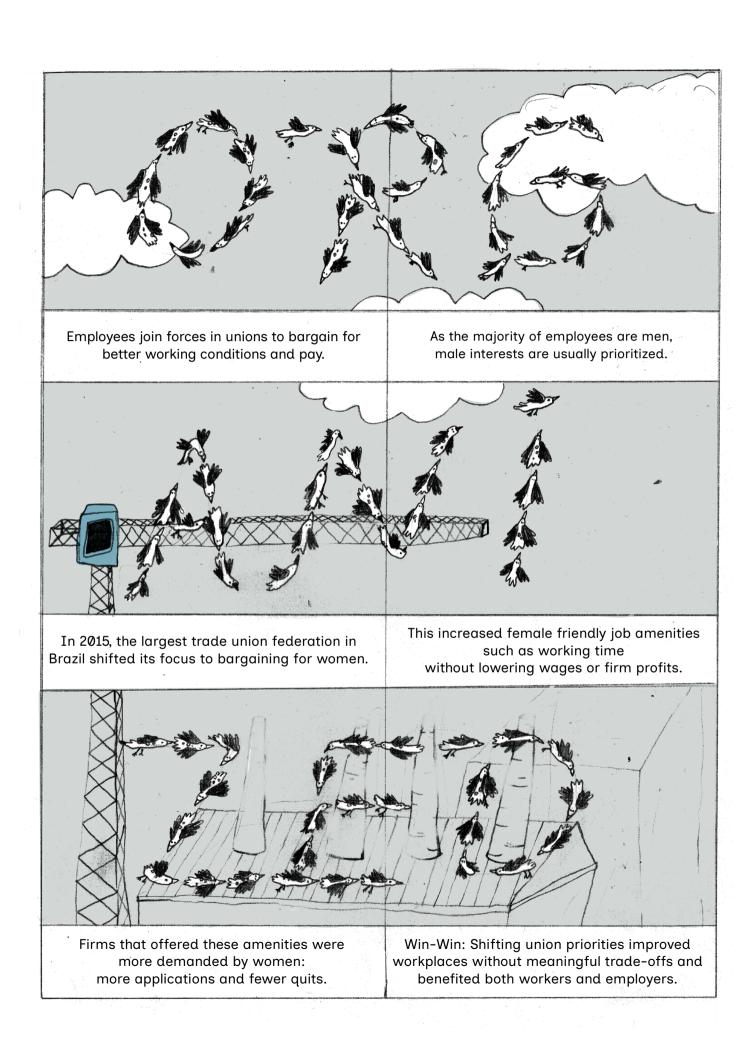


16. Shifting Union Priorities

Viola Corradini, Lorenzo Lagos, Garima Sharma, Collective Bargaining for Women: How Unions Can Create Female-Friendly Jobs, The Quarterly Journal of Economics, Volume 140, Issue 3, August 2025, Pages 2053–2105.



^{*} Using a difference-in-differences design that exploits variation in union affiliation with the federation, the study evaluates the impact of "bargaining for women" on female-friendly amenities in collective bargaining agreements and workplace practices in Brazil. Outcomes are measured using job queueing behavior and separation rates, as well as firm-level data on wages, employment, profits, turnover, and absenteeism for 4409 unions with 211,569 collective bargaining agreements.

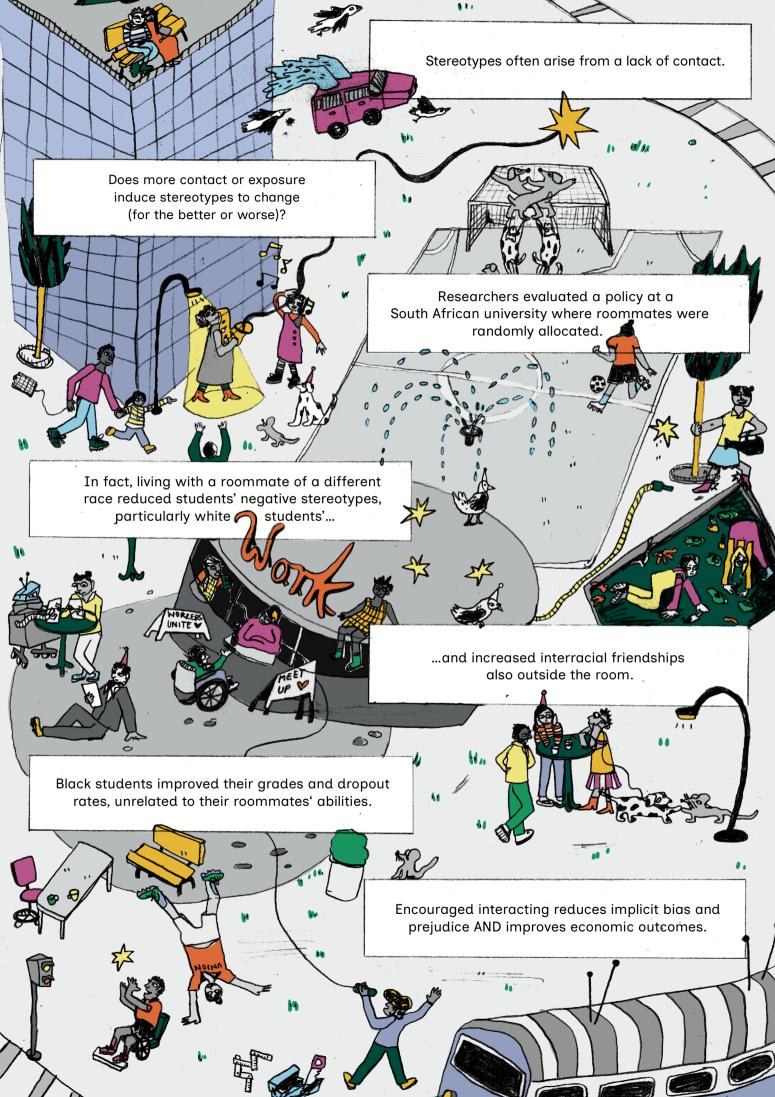


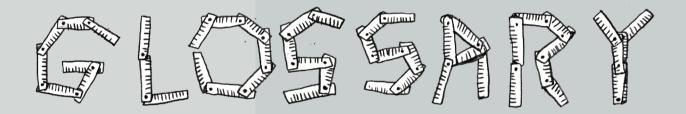
17. Interacting

Lucia Corno, Eliana La Ferrara & Justine Burns (2022). Interaction, Stereotypes, and Performance: Evidence from South Africa. American Economic Review, 112 (12): 3848–3875.



^{*} The study exploits a university housing policy that randomly assigned first-year students to double rooms at a large South African university in 2012 to examine how interracial interaction affects stereotypes, attitudes, and academic performance. Data were collected through three survey waves, implicit association tests (IATs), and laboratory experiments.





How Research Proves What Really Works

To prove what really works, you would like to see the same person in two parallel worlds at the same time: one world with a change or an intervention, one world without. Since this is impossible, we need research methods to construct a hypothetical world that comes as close as possible to this parallel concept.

Building a convincing hypothetical world is challenging, though.

Anecdotes like "my grandfather smoked and lived to 86" must fail because they ignore the possibility that he might have lived even longer without smoking and because robust conclusions need more than single data points. Another challenge is to find a suitable comparison group.

Creating a hypothetical world that reveals true causation (X leads to Y) rather than mere correlation requires careful design.

This is how science evolves: it gets better in designing these hypothetical worlds.

The following methods show how the studies presented in this comic constructed their hypothetical worlds.

While these are all respected approaches, it is always wise to ask: How well was this hypothetical world built? Do the data and methods create a convincing case? Your critical perspective matters in determining how much trust you place in the findings.

Randomized Controlled Trial (RCT)

Researchers randomly assign participants to treatment and control groups and only the treated experience an intervention. The hypothetical world is: the control group that continues without an intervention.

Difference-in-Differences

Researchers compare two groups before and after (difference-) a change that affects only one of them (in-difference).

The hypothetical world is: the group that didn't experience the change.

Laboratory Experiment

Researchers test interventions with participants (mostly students) in a laboratory.

The hypothetical world is: participants who don't receive the intervention.

Field Experiment

Researchers test real-world interventions with real-world people (instead of students).

The hypothetical world is: those who don't receive the intervention.

Fixed-Effects Model

Researchers track the same entities (people or firms).

The hypothetical world is: the same entity in a state where the intervention never happened, while all its inherent traits are held constant (fixed).

Regression Discontinuity & Event Study

Researchers analyze outcomes around specific cutoffs or events. The hypothetical world is: those just below a cutoff, or the expected trend without the event.

Correspondence / Audit Study

Researchers send out nearly identical applications differing only in one trait.

The hypothetical world is: applicants with the reference trait.

Roy Model

Researchers analyze real-world labor market data and compare it to a simulated scenario without discrimination or barriers.

The hypothetical world is: how jobs would be allocated if only skills and comparative advantage mattered.

Implicit Association Test (IAT)

Researchers measure automatic mental associations through response times.

The hypothetical world is: responses unaffected by unconscious bias.

Text Analysis & Machine Learning

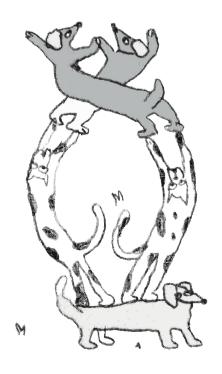
Researchers use algorithms to detect patterns in language. The hypothetical world is: what standard or conventional language would predict.

Exploration Algorithm (Bandit)

Researchers use AI that balances predicted success with uncertainty. The hypothetical world is: an AI that only replicates past (human) decisions.

Structural Estimation

Researchers combine theory with observed real-world decisions. The hypothetical world is: simulated choices under different conditions.





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Zoe Branczyk

Illustrator and Comic Artist.

She lives in Mainz where she spends her time drawing, making up strange stories, thinking about sheep and walking by the water.

She has worked with multiple musicians and labels. In her work, she likes to keep social and political issues in mind. Zoe is an active Member of the Bühne für Menschenrechte e.V.

There is always room to learn.

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Julia Schneider | Doc J Snyder

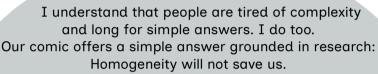
Comic Essayist & Labor Market Economist exploring AI, society & complexity. Her work moves between analysis and imagination, using textual density and art to make systems of work, technology, and power visible. With a PhD in economics and experience in AI consultancy, she turned to artistic research to explore what happens in between disciplines and perspectives.

Her works have been shown in museums and media across Europe and often emerge in collaboration with other artists and researchers.

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Through art, we present some of that research while acknowledging that the world producing this knowledge, economics, academia, and even the three of us authors, is not exactly a model of diversity itself.

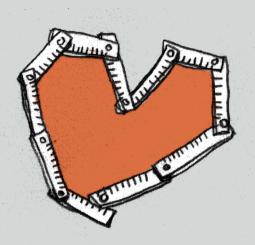
At the same time, those seemingly rigid identities ("women", "men", "the disadvantaged", "the privileged"...) are fluid.

Still, they carry real weight in the world.

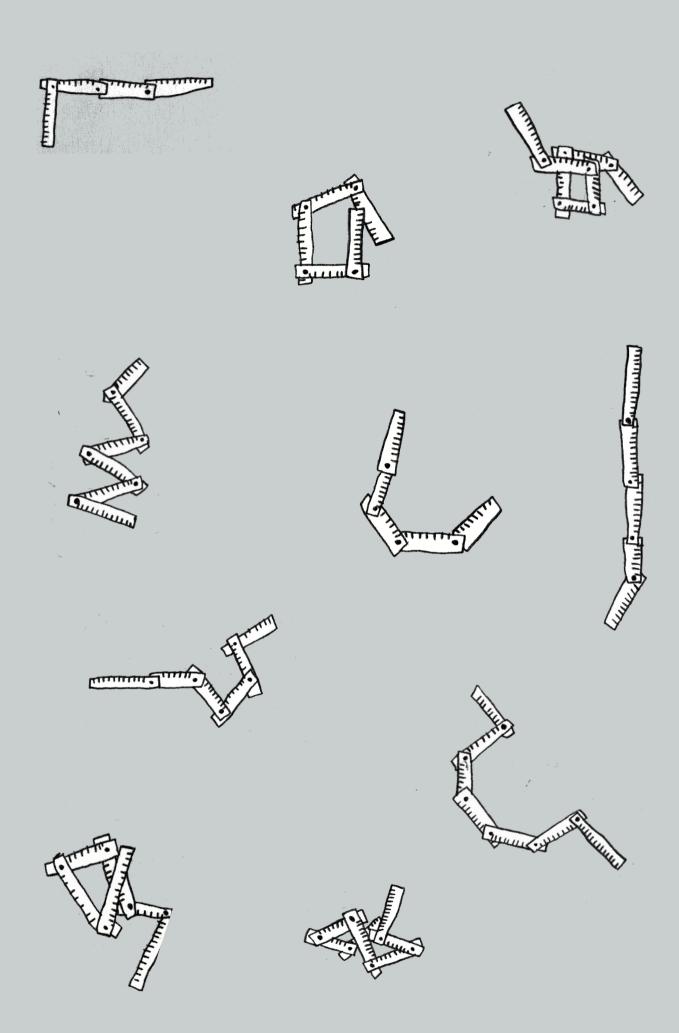
The challenge is to work with these categories while remembering that our identities exist on a spectrum.

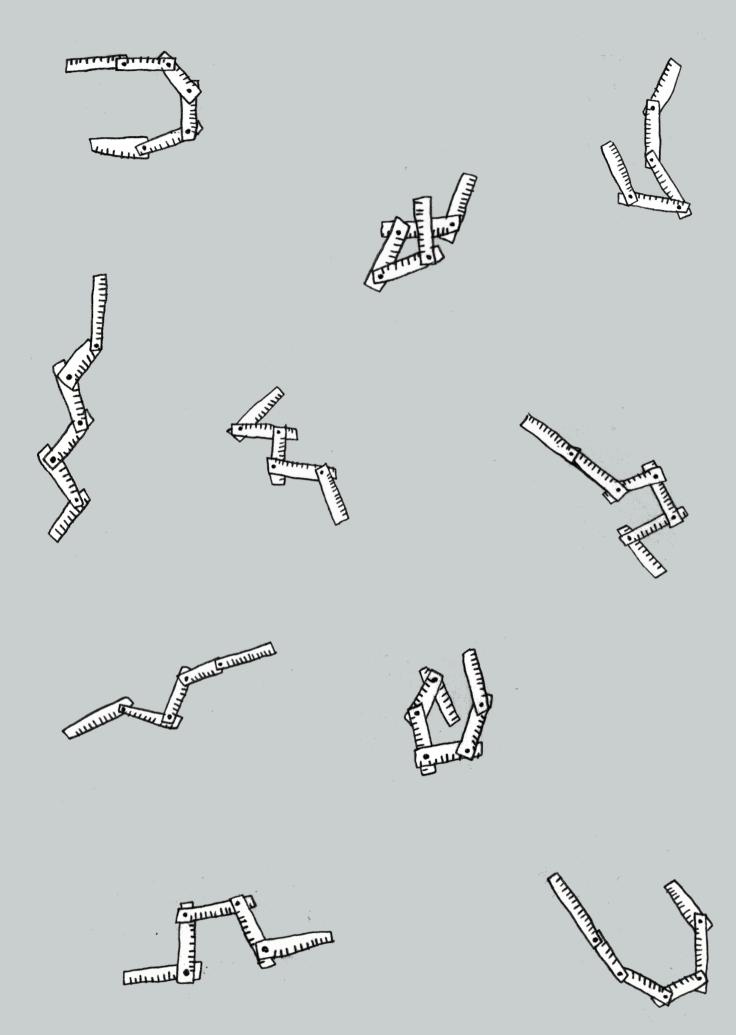


I don't understand it all and all I can do is
try to unlearn and discover little by litte.
There is a big gap in data, and data's language.
Working on this project has made it more than clear.
It kept on helping me to acknowledge and question my
own position. Therefore we tried to break it up, in words and
visiually. Always knowing that we, as a group, are
also not the most diverse. Privilige should be shared.
How lucky am I to do what I love most, call it a job,
call it work (at least part-time) and get to connect with
so many people to pick up knowledge from.



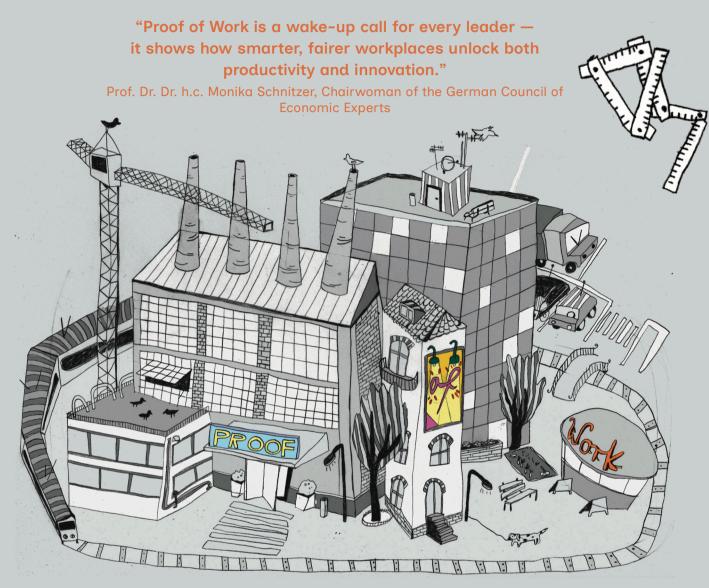
This book is dedicated to our mentors, teachers and collaborators, research assistants and editors, the ones who feel like home, and to all who weave with us the enduring fabric of collective strength.





"A comic for forward thinking leaders that want to understand in an easy and fun way how to leverage and attract global talent."

Janina Kugel, Non Executive Director and Senior Advisor



Today, diversity is often framed as a problem, not a solution.

But data and research say otherwise.

In this comic, we present Proof of Work: literal proof (empirical evidence) of tools that make labor markets work better and fairer. To level the playing field. And to build on every talent.

Because exclusion isn't just unfair, it's inefficient.

This is either:

- · Your new toolkit, backed by top academic journals
- The perfect gift for anyone curious about labor markets and how they can work better for everyone
- · A showcase how hard science meets comic to offer an understanding of structures and power

