

Hiring as Exploration

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Modern hiring ML favors traditionally successful groups

- Firms increasingly use algorithms to make HR decisions
 - Modern hiring ML typically relies on “supervised learning”
 - By design, supervised learning identifies and selects from traditionally successful groups
- Is this the right approach?
- Hiring as a dynamic learning problem
 - **Exploit** – selecting who you think is actually good (this is what supervised learning does)
 - **Explore** – selecting who you know less about to learn (this is what we do)

Exploration based ML improves quality and increases diversity

Who gets to interview at elite firms?

- Our setting

- Professional services hiring at a Fortune 500 firm
- 95% of applicants rejected just on the basis of their CV
- We design algorithms to choose first round interviews
- Goal: maximize the share of interviewed applicants who receive and accept an offer

- What we do

- Build and train an “Upper Confidence Bound” (UCB) contextual bandit algorithm
- Generate interview decisions on administrative test data
- Compare with a) actual human decisions and b) traditional supervised learning algorithm
- Focus on hiring rates (“quality”) and demographics (“diversity”) of selected candidates

Firms' recruiting practices are inside the Pareto frontier

- UCB algorithm selects higher quality and more diverse candidates
 - Increases hiring yield from 10% (human recruiters) to 25%
 - Doubles share of interviewed candidates who are Black or Hispanic (from 10% to 23%)
- But algorithmic design matters
 - Traditional supervised learning also improves hiring rates
 - But cuts the share of interview slots for Black and Hispanics to under 5%
- But what about?
 - Selective labels problem – we address using matching on covariates and an IV approach
 - Other measures of applicant quality – while limited data, no evidence that candidates humans select are better on other dimensions

There's more in the paper!

- Learning about (simulated) changes in applicant quality
 - UCB more effective at detecting changes in the quality of minority groups
- Blinding of race/gender variables
 - We can achieve similar results without using explicit demographic variables
- More information
 - Longer talk in Labor Studies: Monday, July 20th, 1pm
 - Questions? lraymond@mit.edu