Who is Carlos and Why is He in the Econometrics Group?



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Who am I?

My Work

Fun Facts

Bibliography



- ▶ Born and raised in Valladolid, Spain
- ► Growing up I was very interested in Physics but then... GFC hit us hard
- ► I adopted some "saviour syndrome" and decided to become a politician!?
- ► Every politician should know Law and Economics. But I was too lazy to study Law so...





- \blacktriangleright I enrolled in a Double Bachelor in Economics and International Studies at UC3M
- ► I still don't know what International Studies was about, but I am a good asset in pub quizzes ig
- ▶ I visited UC Berkeley and Bocconi
- ▶ Finally, I worked full-time as a Consultant while completing my degree in the weekends (1/10 would not recommend)



- ▶ I was lucky to be part of the first MPhil + DPhil cohort
- ▶ Although I came here as a super applied person (remember politician vibes), I was early seduced by theory thanks to Anders, Martin, Max and many others
- ▶ My paper at the intersection of Micro Theory, ML and Policy Design was awarded the runner-up **prize for best thesis**



- ► So... what do I do?
- ▶ I have a wide range of interests: **Machine Learning Theory** (Online Learning), **Micro Theory** (Information Theory and Mechanism Design), and **Applied Micro** (Policy, Labor and Education)
- ► Anything but Macro (aka Macron't or Macro)
- ▶ I am most interested in understanding **how agents learn**, and how we can import notions of learning from ML to design efficient **adaptive policies**
- ► Let me give you some examples



- ▶ **Problem:** A profit-maximizing wage-setting firm is presented with a line of workers, who are characterized by a pair (productivity p, reservation wage r)
- ▶ The firm has no access to (p_i, r_i) AND holds **no informative prior** on the joint distribution $F_{P,R}$
- ▶ The firm offers wage w to agent i and conditional on $w > r_i$, it observes p_i



- \triangleright Define the optimal policy as the best wage you could offer if you knew $F_{P,R}$
- ▶ Question: Can any simple learning strategy be asymptotically no worse than the optimal policy?



- ▶ Building on Max's work [Cesa-Bianchi et al., 2022], I designed an algorithm which converges to optimal
- ► This is a big deal, as it extends the notions of **equilibrium under adverse selection** of Akerlof & Co. without priors! (subject to game repetition)
- ▶ I show that such algorithm is **unimprovable**



- ▶ I also look at the impact of **structural policies** in the firm's learning process, like Minimum Wage, LIPC or Productivity Shocks
- ► To make my life harder, I prove these results in adversarial settings with limited feedback
- ▶ I do a few other things, so go check it out!



- ▶ My thesis contains a **positive theory**, which characterizes equilibrium under adverse selection in repeated games, and a **normative theory**, which provides a simple strategy to the monopsony firm
- ► However, this is not really how firms set wages
- ▶ To make my work more empirically relevant, in subsequent work I embedded a theory of Concave Bandits with Convex Knapsacks



- ▶ The idea is that we can now budget constrain the firm
- ► This allow us to: (i) generalize to any concave function (like DRS) and (ii) impose high-dimensional budget constraints (number of workers, salary budget, etc.)
- ▶ Math becomes more difficult but intuition is simple: The firm must now learn the most **cost effective arm**



- ► What I enjoyed the most from my MPhil Thesis was the analysis of the interaction between policies and learning
- ▶ Prominent literature in CS and Computational Econ [Frazier et al., 2014], [Adusumilli, 2021], [Papanastasiou et al., 2018], [Mansour et al., 2015], [Adjaho and Christensen, 2022] which looks into the design of **incentives** to curve the behaviour of myopic agents so they explore optimally on behalf of the Principal
- ▶ This brings together the areas of Persuasion, Mechanism Design, Bandits and more



- ▶ **VERY preliminary** idea is to look at more refined learners which interact repeatedly with the environment
- ▶ Probably related to **labor markets** like search-matching functions or hiring decisions
- ► Focus on the role of incentives timing
- ▶ One example: How does the policy-maker incentivize optimal exploration, when it holds loose priors on the matching efficiency of a firm-worker pair?



- ▶ Mechanism Design toolkit (Real Analysis and Measure Theory)
- ▶ Bandit Theory toolkit (Measure Theory, Probability Theory, Large Deviation Theory, Martingales, some Calculus and lots of bounding tricks)
- ► Connecting fields (Convergence Results)



- ▶ I run a lot. Like a lot. Like 9 times a week. I talk a lot about how much I run and how much I hate running
- ▶ I am the President of the Spanish Society here in Oxford, so I am in charge of organizing the best Reggaeton Parties in town
- ▶ Love sports, chess, nature, languages and stupid geographic facts



Fun Facts Pics









- ▶ I think I did not manage to explain why I am in the Econometrics group, but...
- ► At least I hope that you got to **know me better**
- ► If you ever wanna chat, please just drop me an email carlos.gonzalezperez@economics.ox.ac.uk
- ► If wanna know more about my work you can visit my website presidente-carlos.github.io



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