

For Learning in Symmetric Teams, Local Optima are Global Nash Equilibria

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Motivating Example



Common-payoff game

- All players have same payoffs

Symmetric game structure

- Can swap the taxis

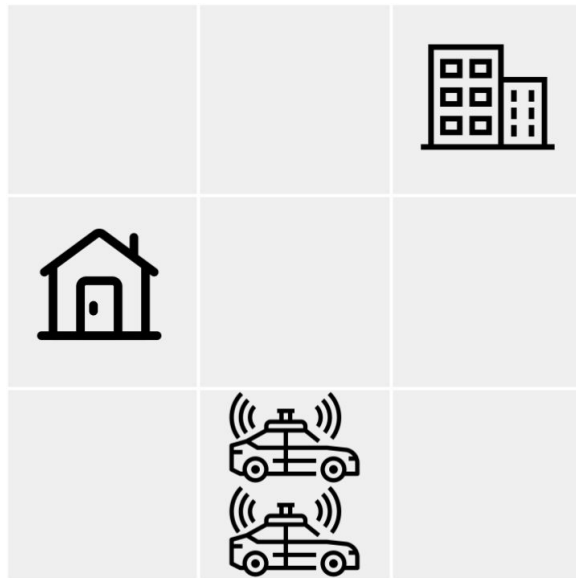
Symmetric strategy profile

- Taxis share source code

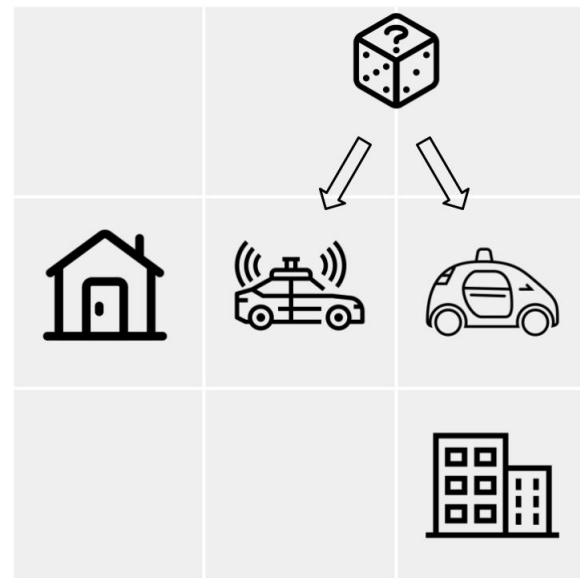
Types of Symmetry



Symmetric agents & env

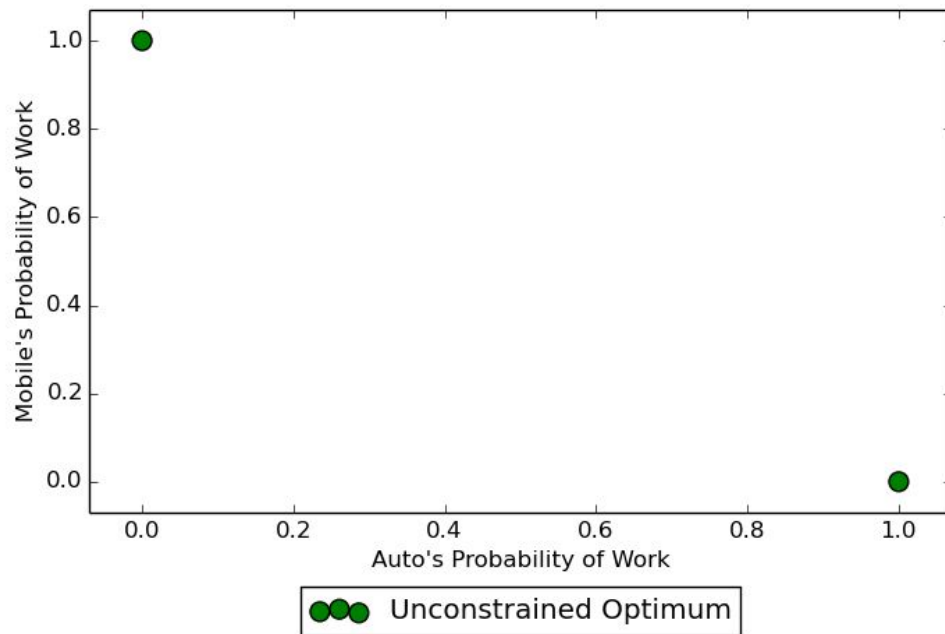


Same initial conditions

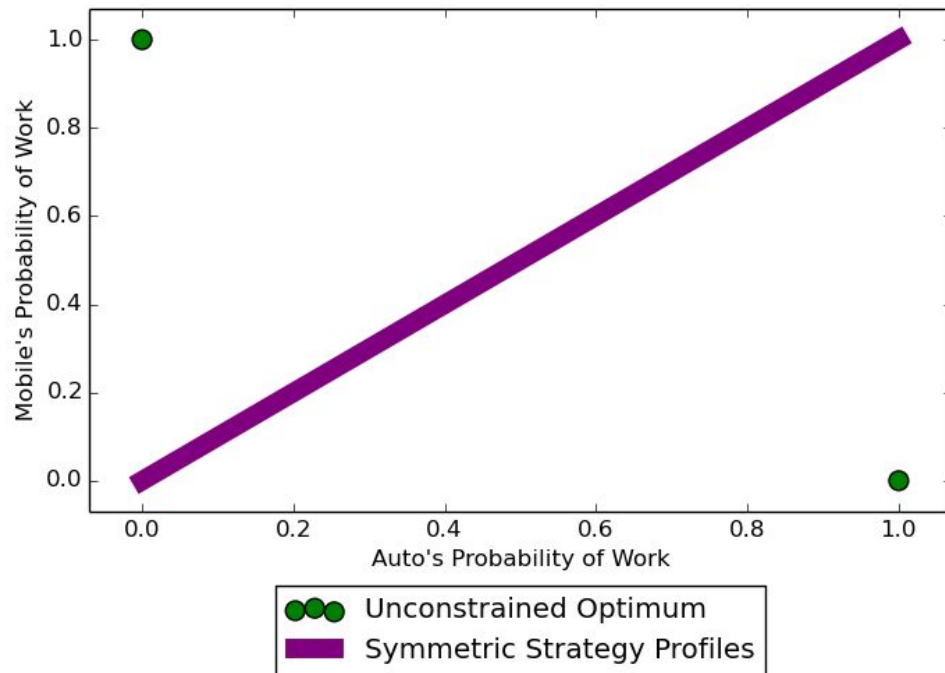


Veil of ignorance

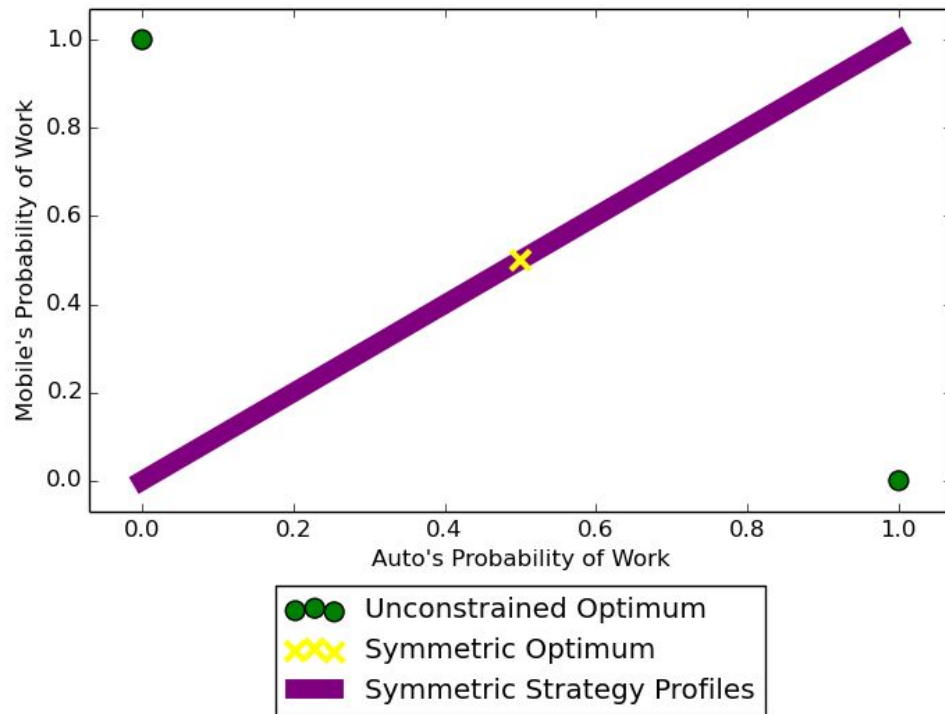
Strategy Profile Landscape



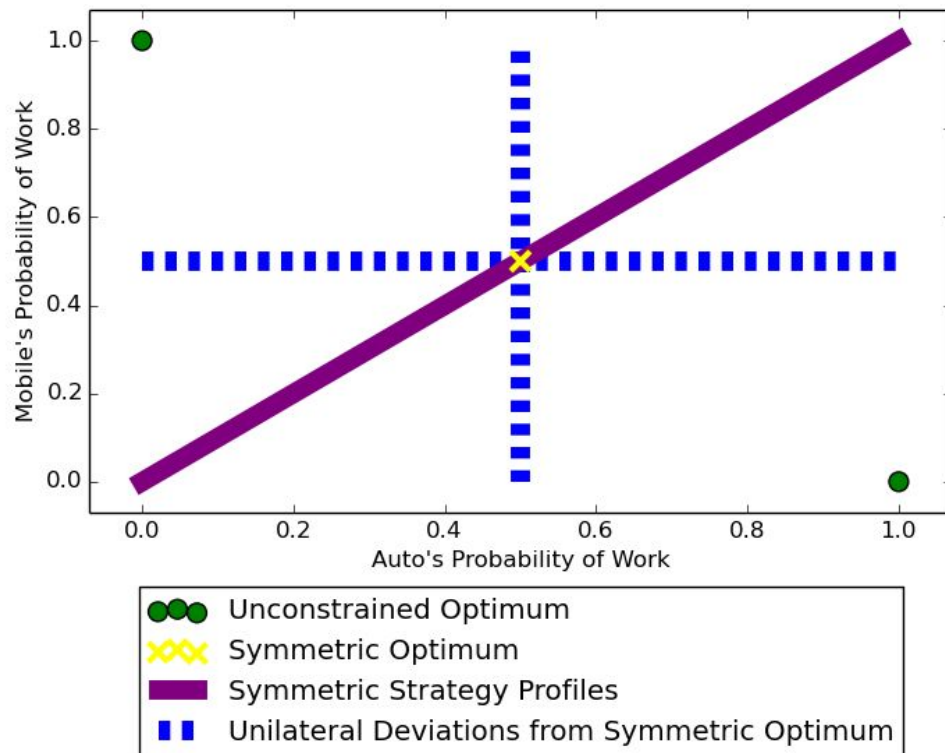
Strategy Profile Landscape



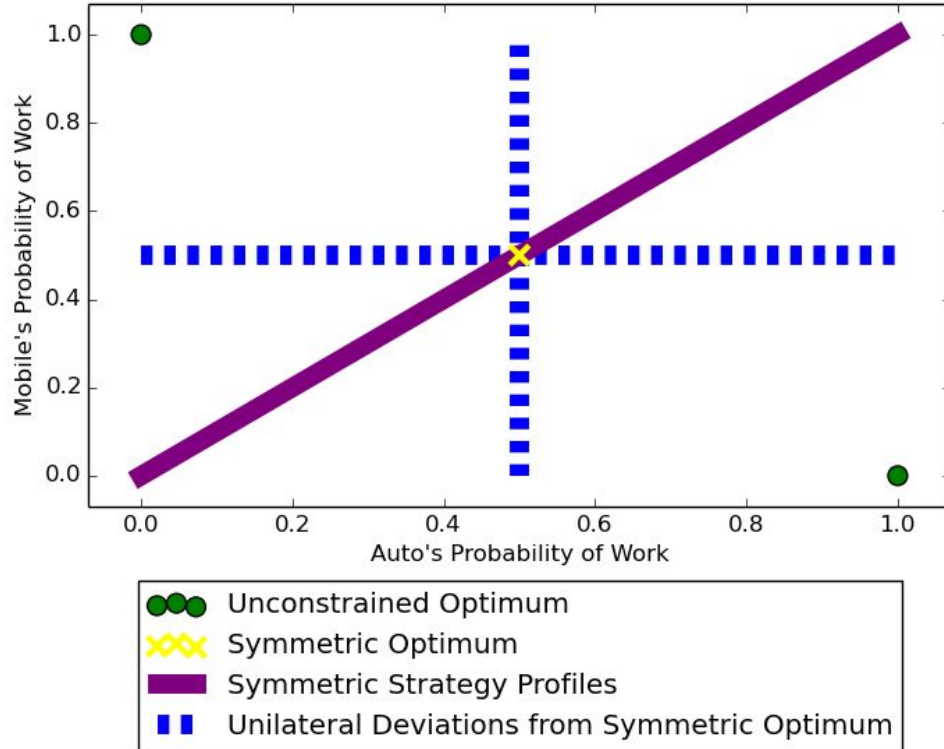
Strategy Profile Landscape



Strategy Profile Landscape



Strategy Profile Landscape

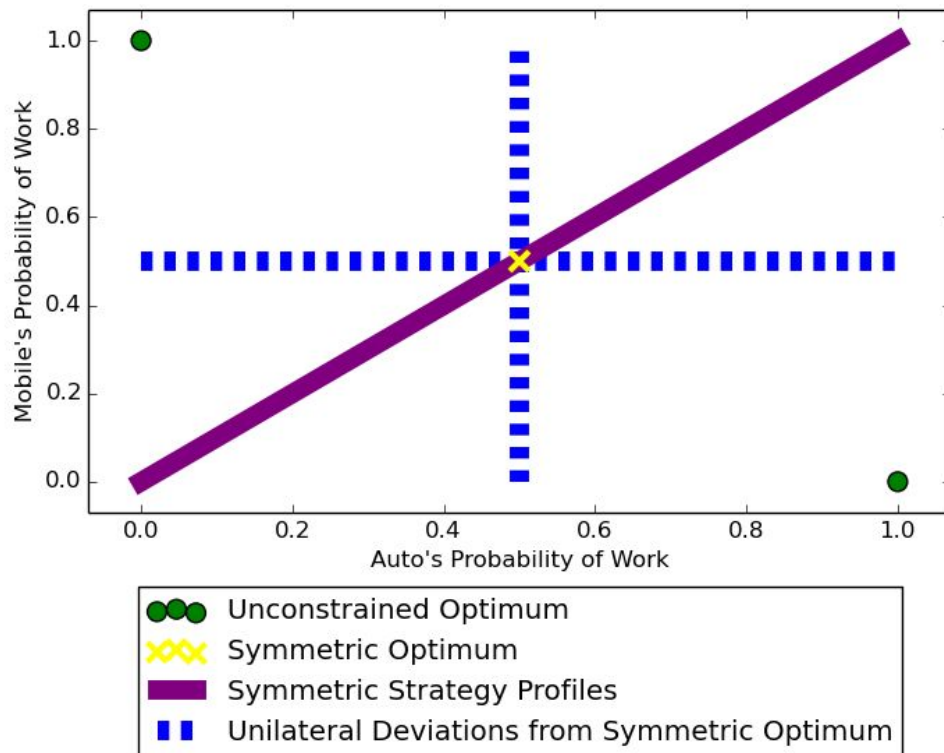


How much can one of N players improve the common payoff by unilaterally deviating?

- No improvement is possible
- $O(\sqrt{N})$ improvement
- $O(N)$ improvement
- $O(N^2)$ improvement



Strategy Profile Landscape

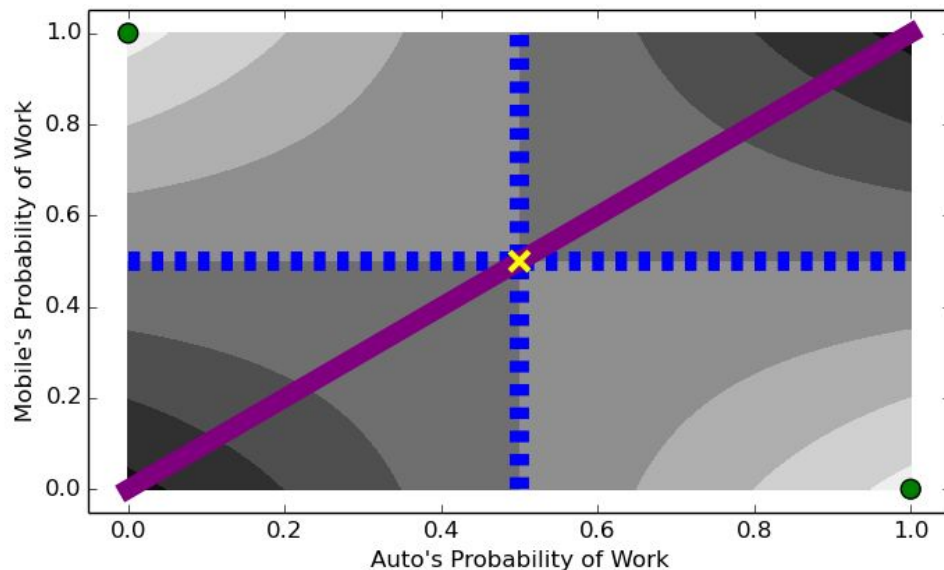


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- b) $O(\sqrt{N})$ improvement
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- d) $O(N^2)$ improvement



Strategy Profile Landscape



- Unconstrained Optimum
- ✕✕✕ Symmetric Optimum
- Symmetric Strategy Profiles
- - - Unilateral Deviations from Symmetric Optimum

	Home	Work
Home	1	2
Work	2	1



Local → Global Guarantee

In Plain English

Locally optimal symmetric collaboration is a *global* Nash equilibrium.

Theorem

Let \mathcal{G} be a normal-form (or extensive-form) game with common payoff.
Then any *locally* optimal \mathcal{P} -invariant strategy profile is a *global* Nash equilibrium.

Note: \mathcal{P} -invariance is a group-theoretic notion of symmetry generalizing:

- Anonymous games
- Transitive games
- Totally symmetric games

Implications of Theorem

Team Theory

No individual can innovate from a symmetric local optimum to improve the common payoff.

Multi-agent Reinforcement Learning

Iterated best response can't improve a local symmetric optimum.

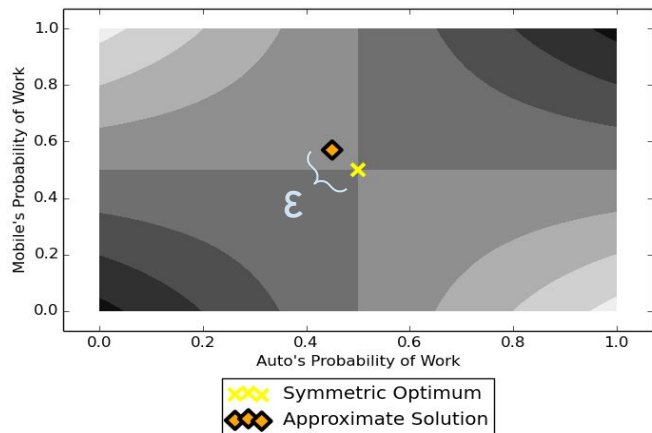
(Adversarial) Team Games

Result extends to arbitrary number of teams.

Robustness of Result

The result degrades smoothly, giving an ϵ -Nash equilibrium for:

Approximate Solutions



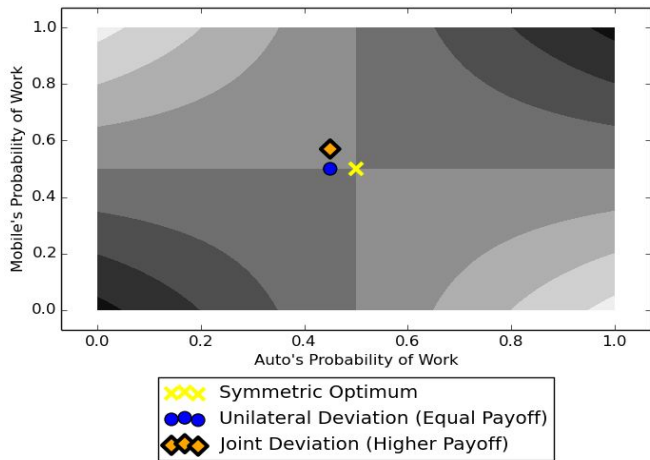
Payoff Perturbations

	Home	Work
Home	$1+\epsilon, 1-\epsilon$	$2-\epsilon, 2-\epsilon$
Work	$2+\epsilon, 2+\epsilon$	$1-\epsilon, 1+\epsilon$

(In)stability to Joint (Possibly-Asymmetric) Deviation

Theorem (for non-degenerate games)

A local symmetric optimum is locally optimal among possibly-asymmetric strategy profiles *if and only if it is deterministic*.



Experimentally, up to 60% are mixed,
i.e., *unstable*!

Conclusion

We give conditions for *stability* and *instability*

Future Work

- Behavioral strategies
- Continuous actions & function approximation
- Sequential decision making benchmarks

