

ARTIFICIAL
INTELLIGENCE
RESEARCH GROUP



ON NASH EQUILIBRIA IN MULTI-OBJECTIVE GAMES

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OVERVIEW

- ▶ Background
- ▶ Game-theoretic work
- ▶ What's next
- ▶ Q&A



Multi-objective games present a natural framework for studying ***strategic interactions between rational individuals concerned with more than one objective.***

Strategic interactions between rational individuals

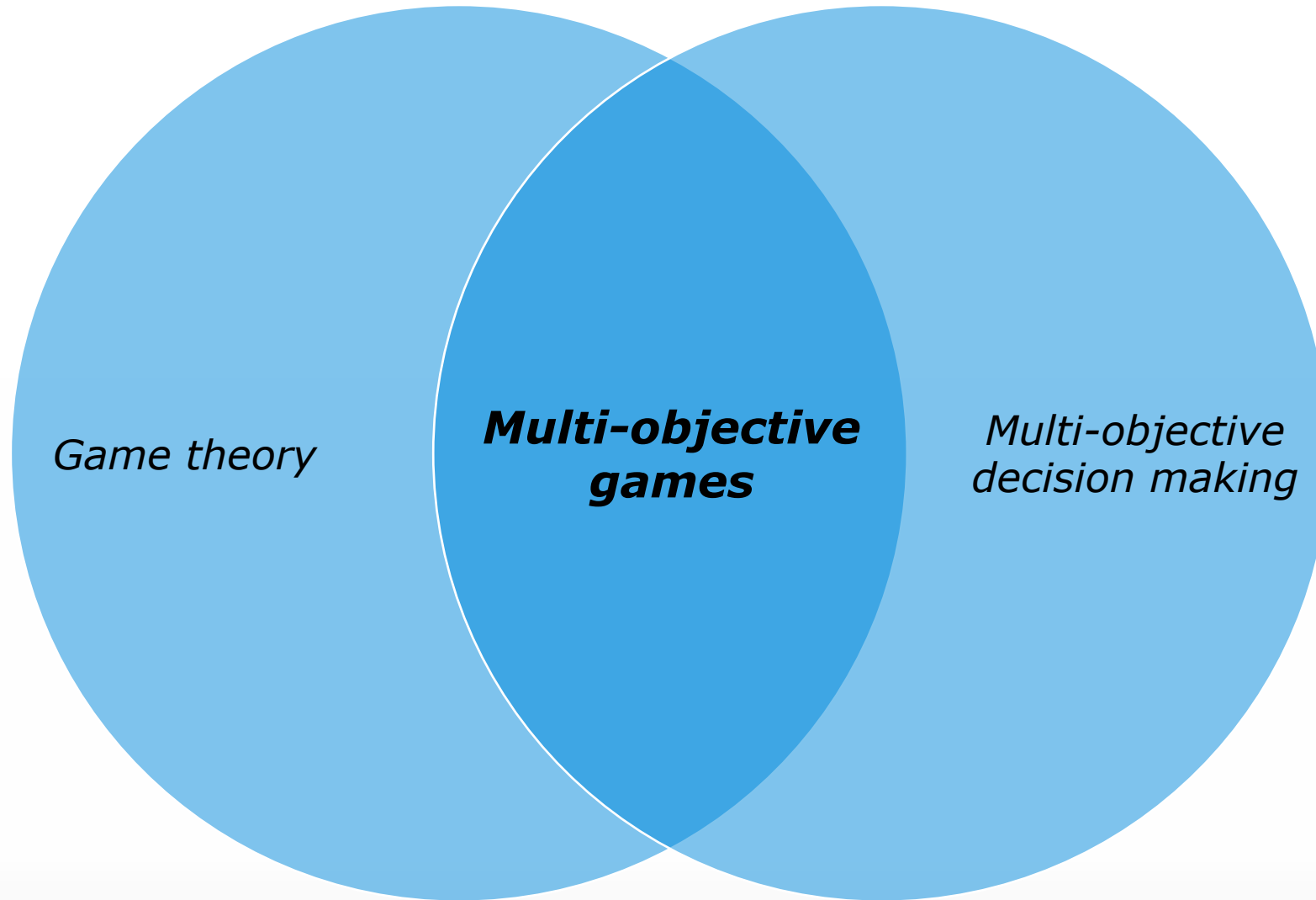


Game theory

Rational individuals concerned with more than one objective



Multi-objective decision making



WHAT

BACKGROUND

- ▶ Multi-Objective Normal-Form Games (MONFGs)
- ▶ Utility based approach
- ▶ Utility function $u_i: \mathbb{R}^d \rightarrow \mathbb{R}$

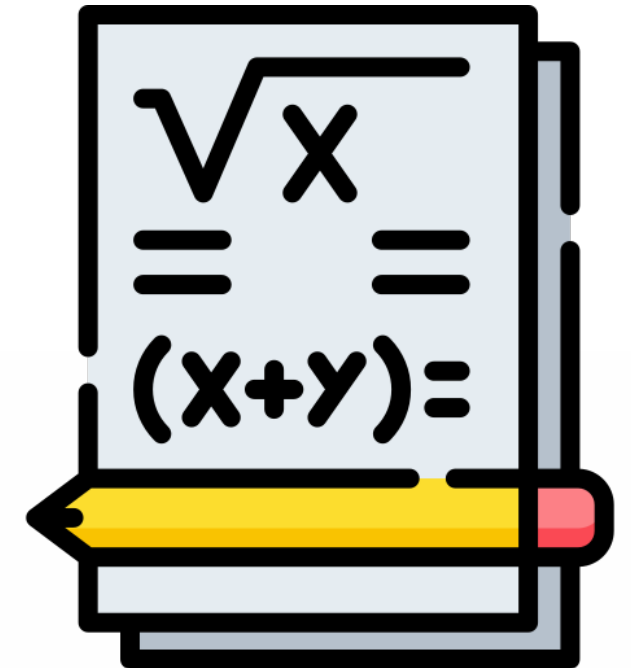
	A	B
A	(10, 2); (10, 2)	(2, 3); (2, 3)
B	(4, 2); (4, 2)	(6, 3); (6, 3)

$$u_1(p_1, p_2) = p_1 \cdot p_2$$

OPTIMISATION CRITERIA

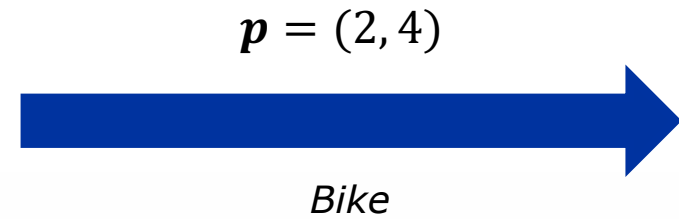
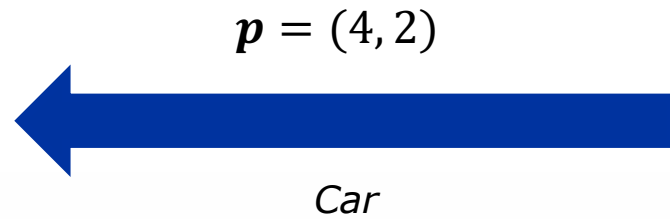
BACKGROUND

- ▶ Two possible choices of optimisation criteria
- ▶ Expected Scalarised Returns (ESR)
 - ▶ Calculate the *expectation of your utility* from the payoffs
 - ▶ Utility of an individual policy execution
- ▶ Scalarised Expected Returns (SER)
 - ▶ Calculate the *utility of your expected payoff*
 - ▶ Utility of the average payoff from several executions of the policy



OPTIMISATION CRITERIA

EXAMPLE



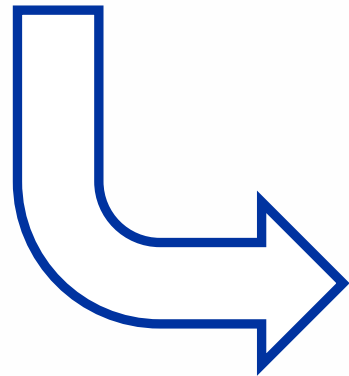
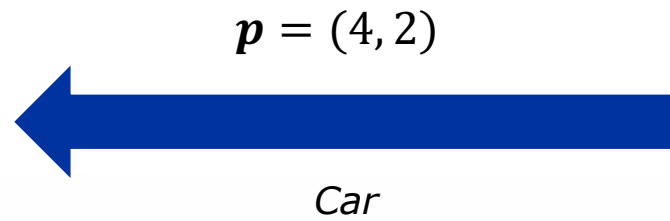
$$u(p_1, p_2) = p_1 \cdot p_2$$

Speed Eco-friendliness

What happens when you take the car 50% of the time and the bike 50% of the time?

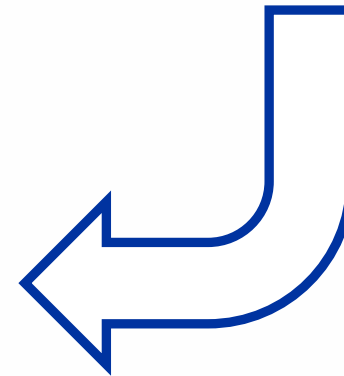
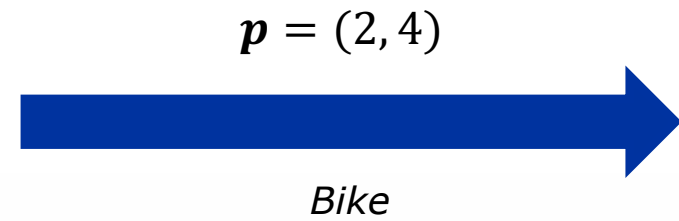
OPTIMISATION CRITERIA

EXAMPLE



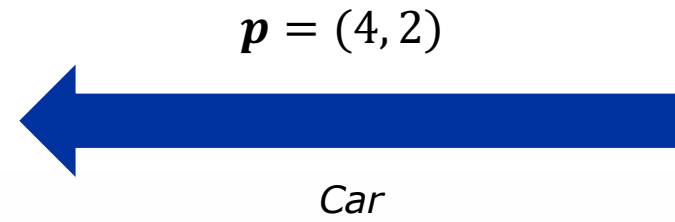
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ESR

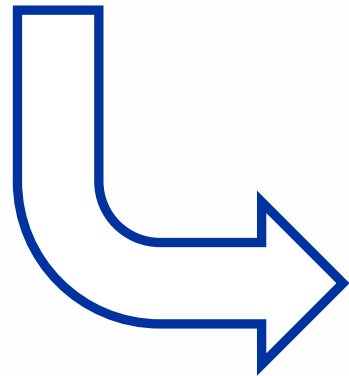


OPTIMISATION CRITERIA

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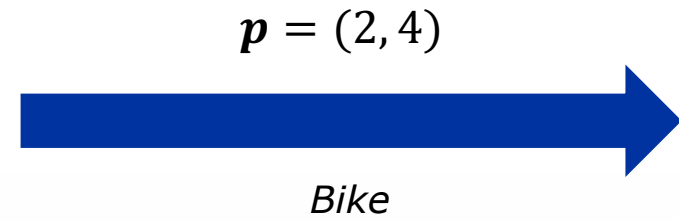


$$u(4, 2) = 8$$

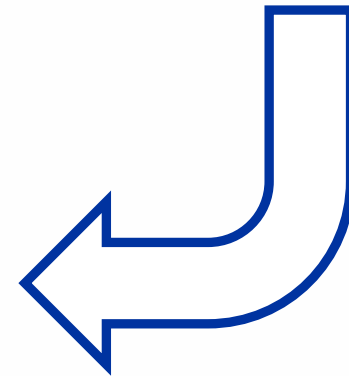


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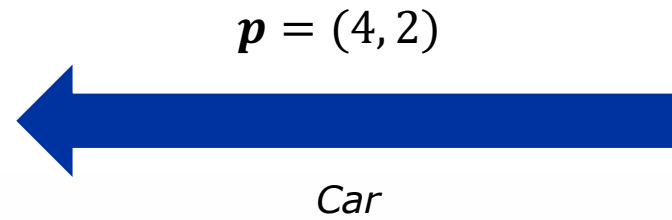


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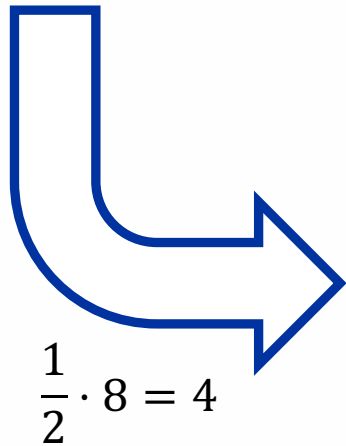


OPTIMISATION CRITERIA

EXAMPLE

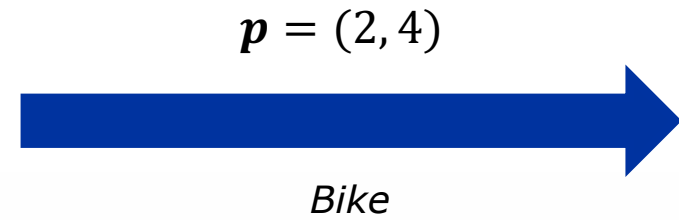


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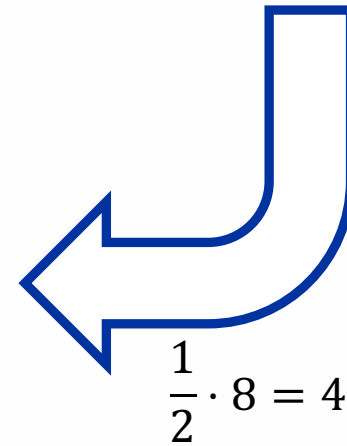


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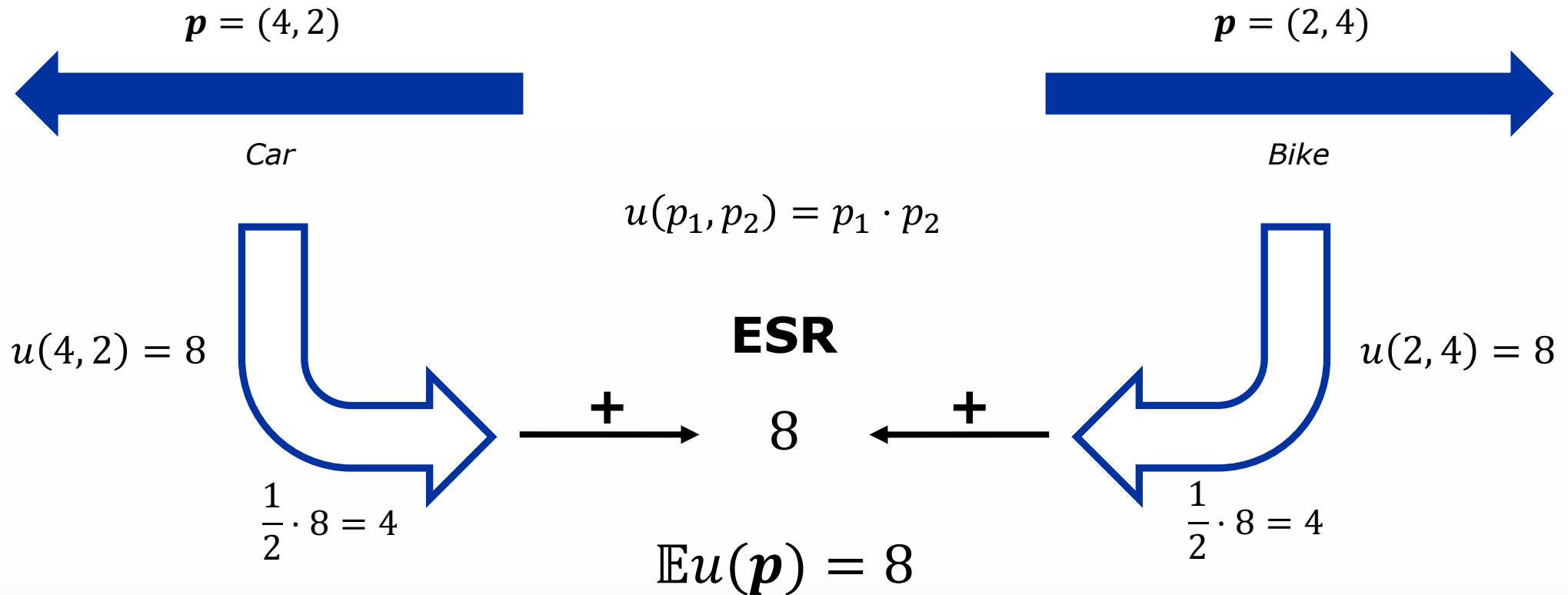


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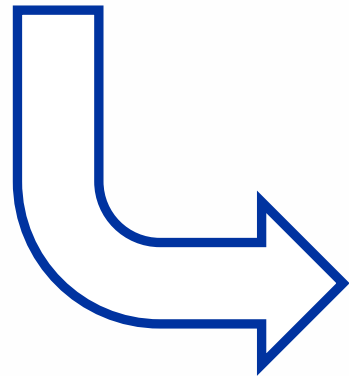
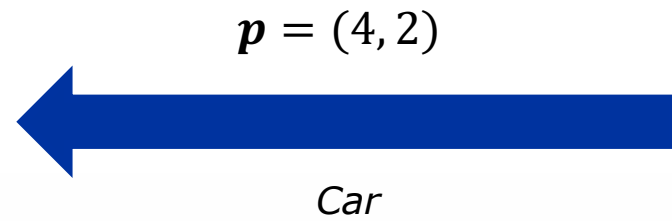
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EXAMPLE



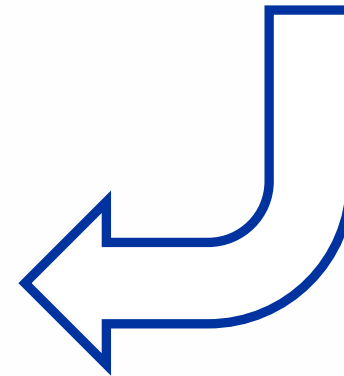
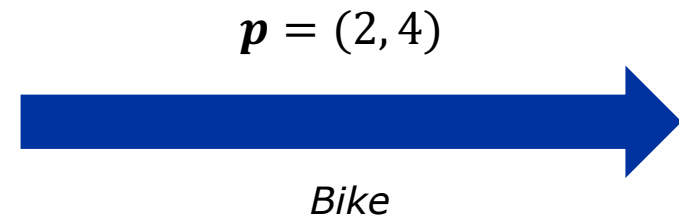
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SER



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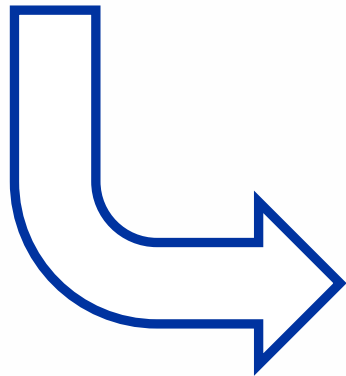
EXAMPLE

$$p = (4, 2)$$



Car

$$\frac{1}{2} \cdot (4, 2) = (2, 1)$$



$$p = (2, 4)$$

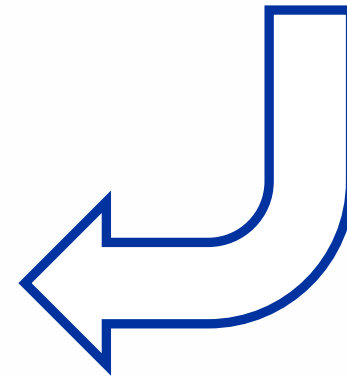


Bike

$$u(p_1, p_2) = p_1 \cdot p_2$$

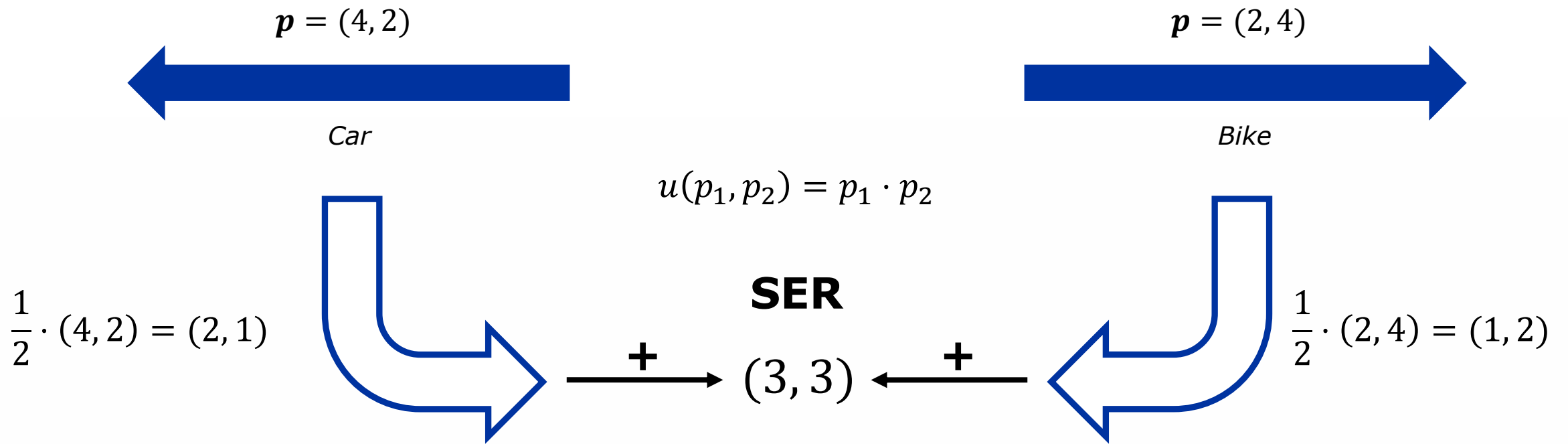
SER

$$\frac{1}{2} \cdot (2, 4) = (1, 2)$$



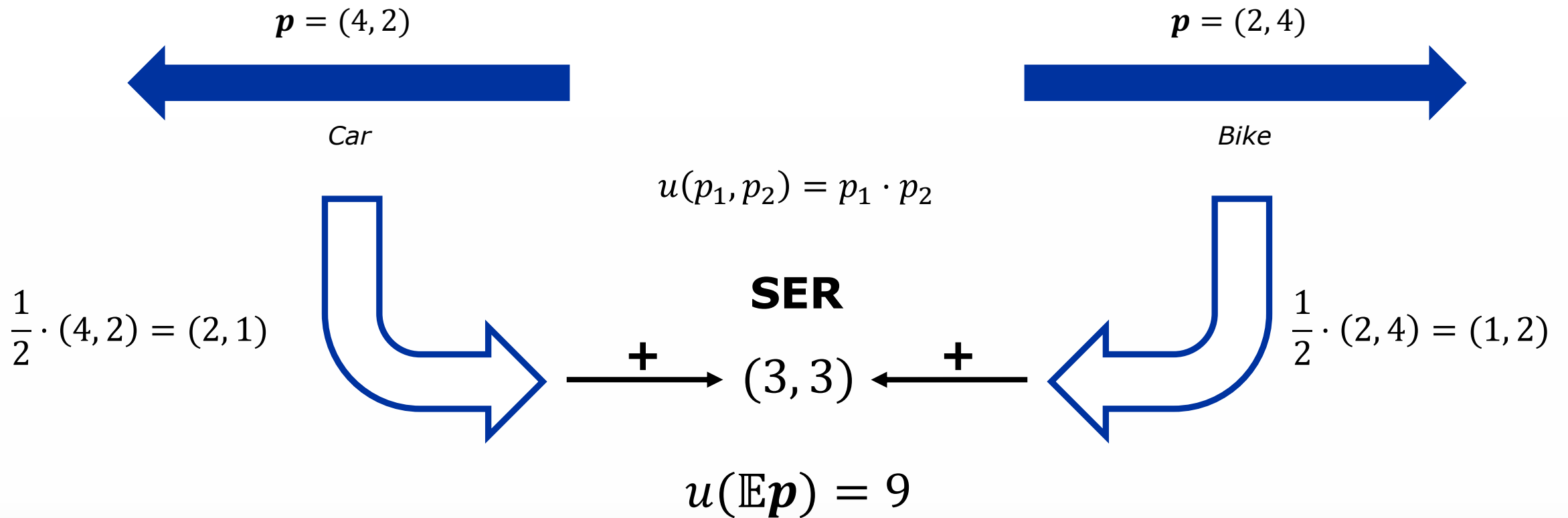
OPTIMISATION CRITERIA

EXAMPLE



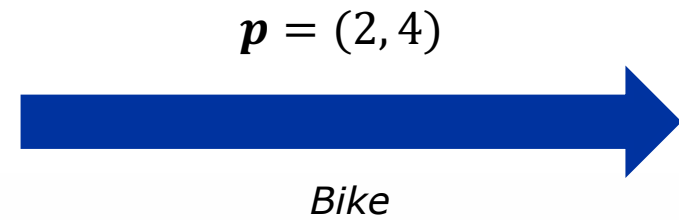
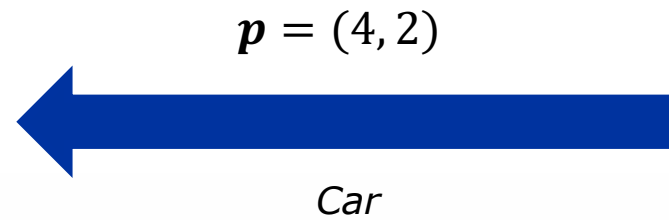
OPTIMISATION CRITERIA

EXAMPLE



OPTIMISATION CRITERIA

EXAMPLE



$$u(p_1, p_2) = p_1 \cdot p_2$$

What happens when you take the car 50% of the time and the bike 50% of the time?

ESR = 8

SER = 9

SOLUTION CONCEPTS

BACKGROUND

▶ Nash equilibria

- ▶ *No agent can improve their utility by unilaterally deviating from the joint strategy*



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$$u_1(p_1, p_2) = u_2(p_1, p_2) = p_1 \cdot p_2$$

	A	B
A	(10, 2); (10, 2)	(0, 0); (0, 0)
B	(0, 0); (0, 0)	(2, 10); (2, 10)



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$$u_1(p_1, p_2) = u_2(p_1, p_2) = p_1 \cdot p_2$$

Nash equilibrium →

$$u_1(10, 2) = 10 \cdot 2 = 20$$

$$u_2(10, 2) = 10 \cdot 2 = 20$$

	A	B
A	(10, 2); (10, 2)	(0, 0); (0, 0)
B	(0, 0); (0, 0)	(2, 10); (2, 10)



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Nash equilibrium

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A	(10, 2); (10, 2)	(0, 0); (0, 0)
B	<u>(0, 0); (0, 0)</u>	(2, 10); (2, 10)

Strictly worse to deviate to B



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▶ Nash equilibria

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	A	B
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Strictly worse to deviate to B

$u_1(10, 2) = 10 \cdot 2 = 20$
 $u_2(10, 2) = 10 \cdot 2 = 20$



GOAL

- ▶ Theoretical
 - ▶ Existence or non-existence guarantees
 - ▶ Algorithms
- ▶ Learning in these environments
 - ▶ Communication
 - ▶ Commitment



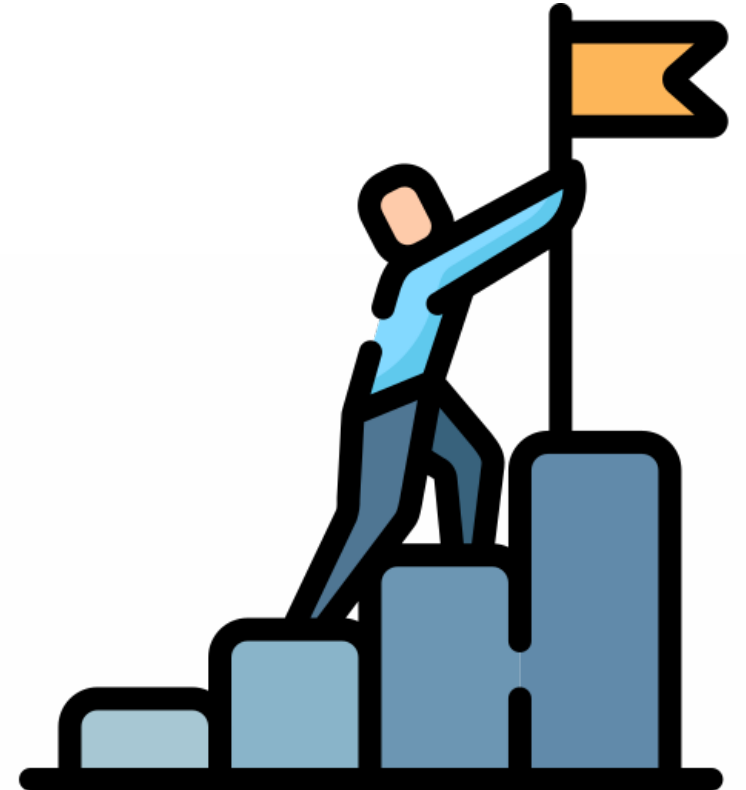
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WHAT ARE MULTI-OBJECTIVE GAMES?

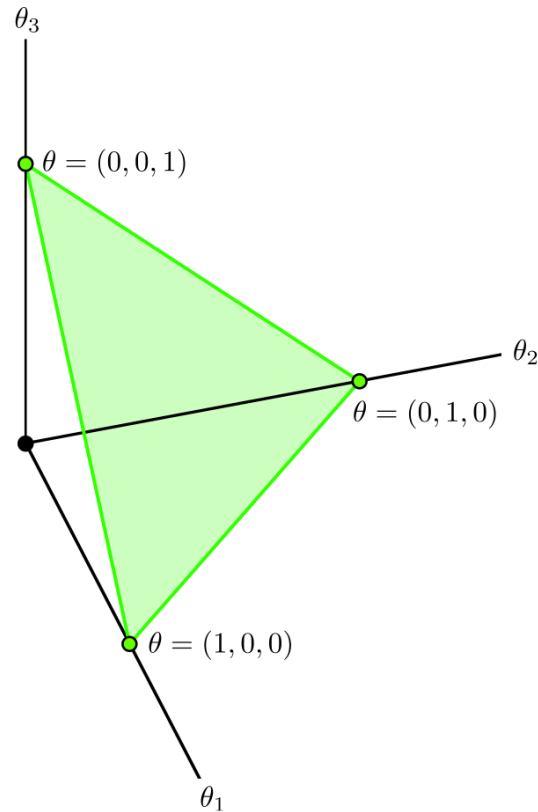
A NOVEL INTUITION

	A	B	C
A	$(4, 1); (4, 1)$	$(1, 2); (4, 2)$	$(2, 1); (1, 2)$
B	$(3, 1); (2, 3)$	$(3, 2); (6, 3)$	$(1, 2); (2, 1)$
C	$(1, 2); (2, 1)$	$(2, 1); (1, 2)$	$(1, 3); (1, 3)$

It turns out we can go from this

WHAT ARE MULTI-OBJECTIVE GAMES?

A NOVEL INTUITION

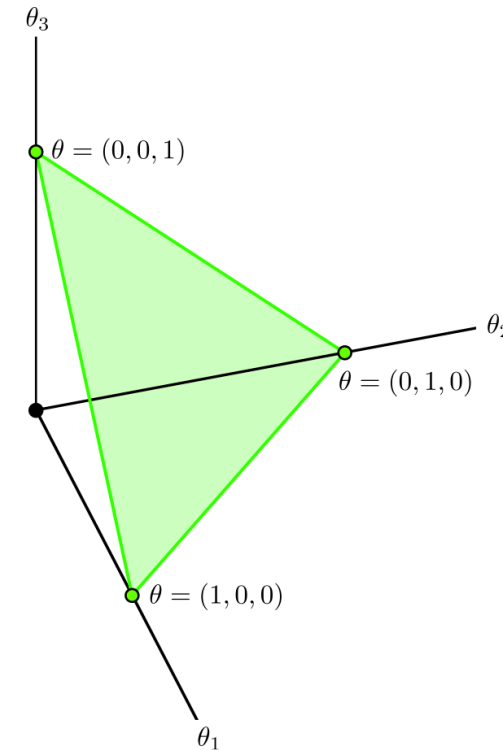


To this

Every MONFG with continuous utility functions can be reduced to a continuous game

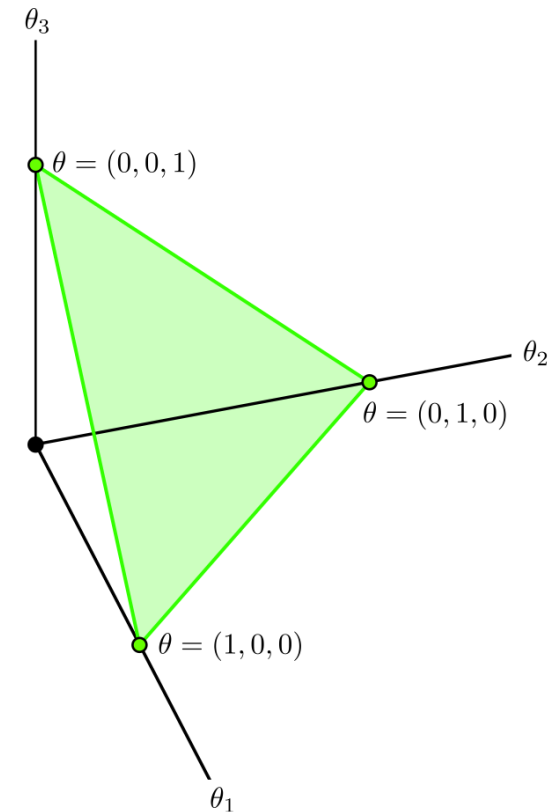
Continuous game

- Single objective
- Infinite number of pure strategies
- Reuse utility functions

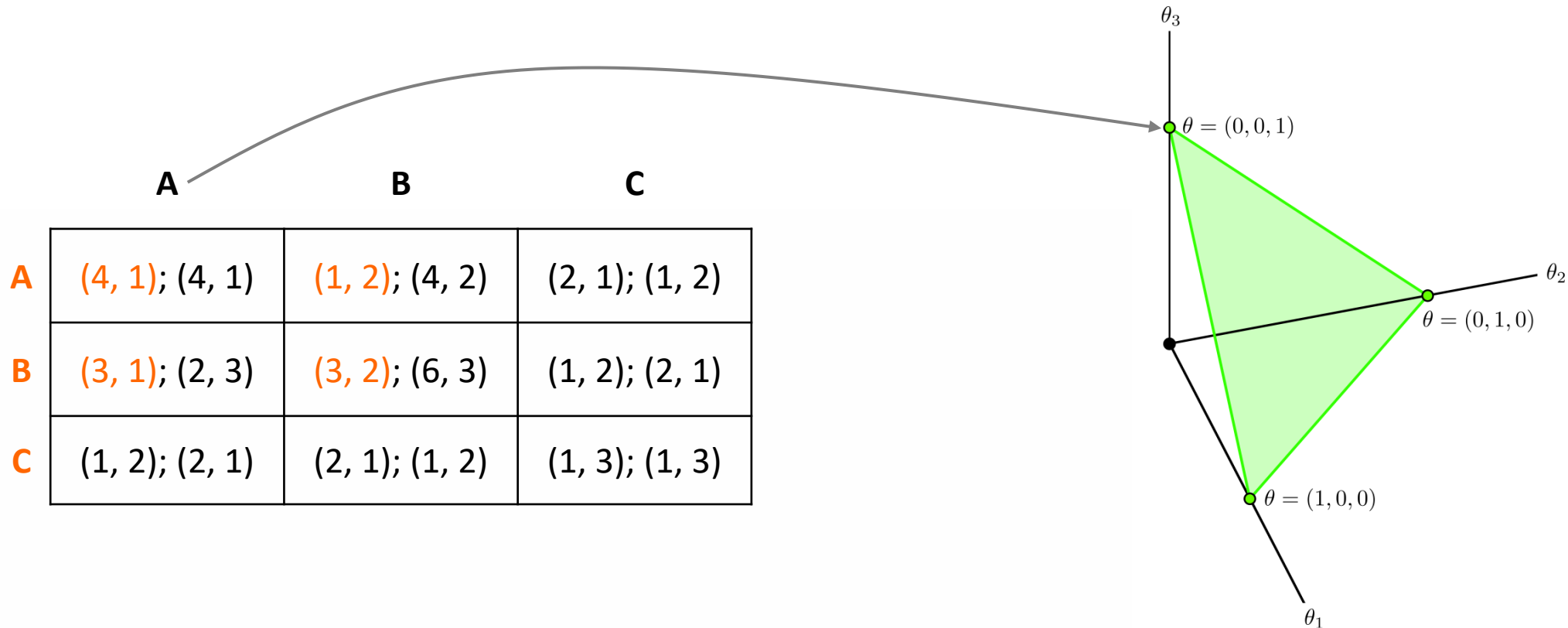


Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

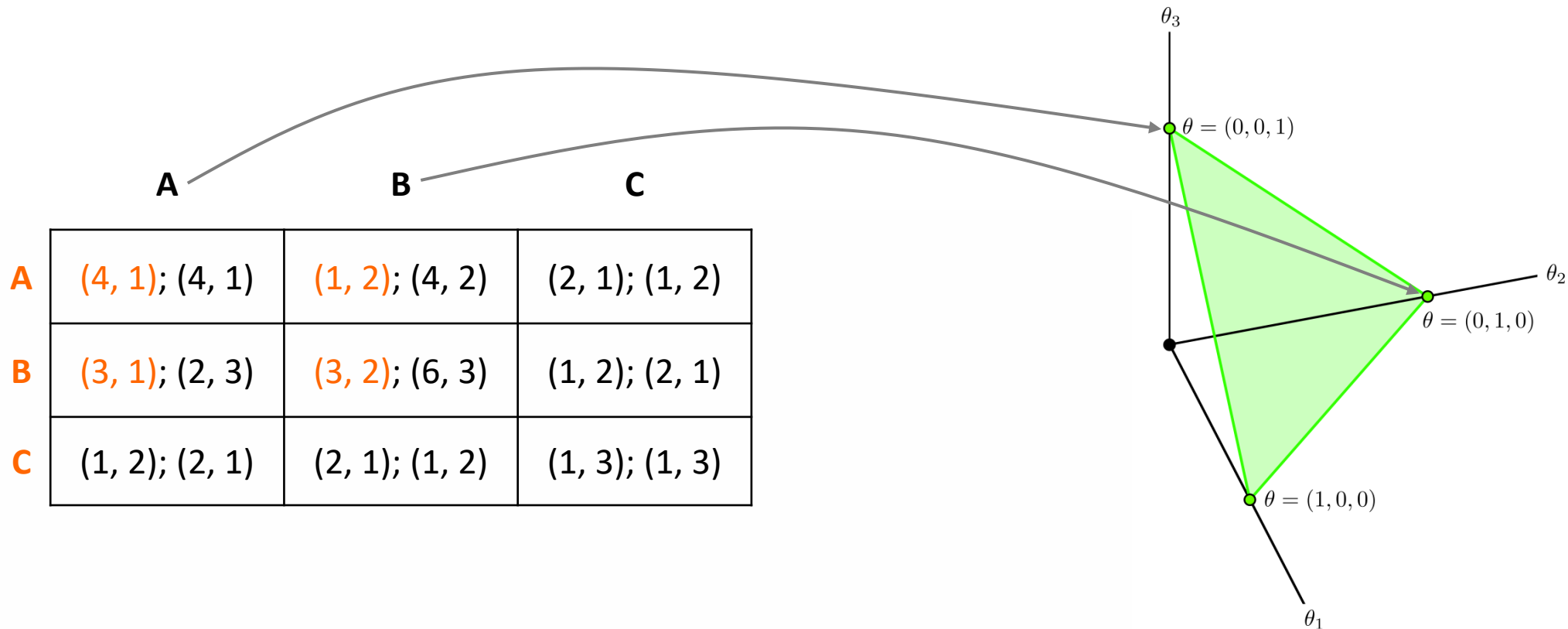
	A	B	C
A	(4, 1); (4, 1)	(1, 2); (4, 2)	(2, 1); (1, 2)
B	(3, 1); (2, 3)	(3, 2); (6, 3)	(1, 2); (2, 1)
C	(1, 2); (2, 1)	(2, 1); (1, 2)	(1, 3); (1, 3)



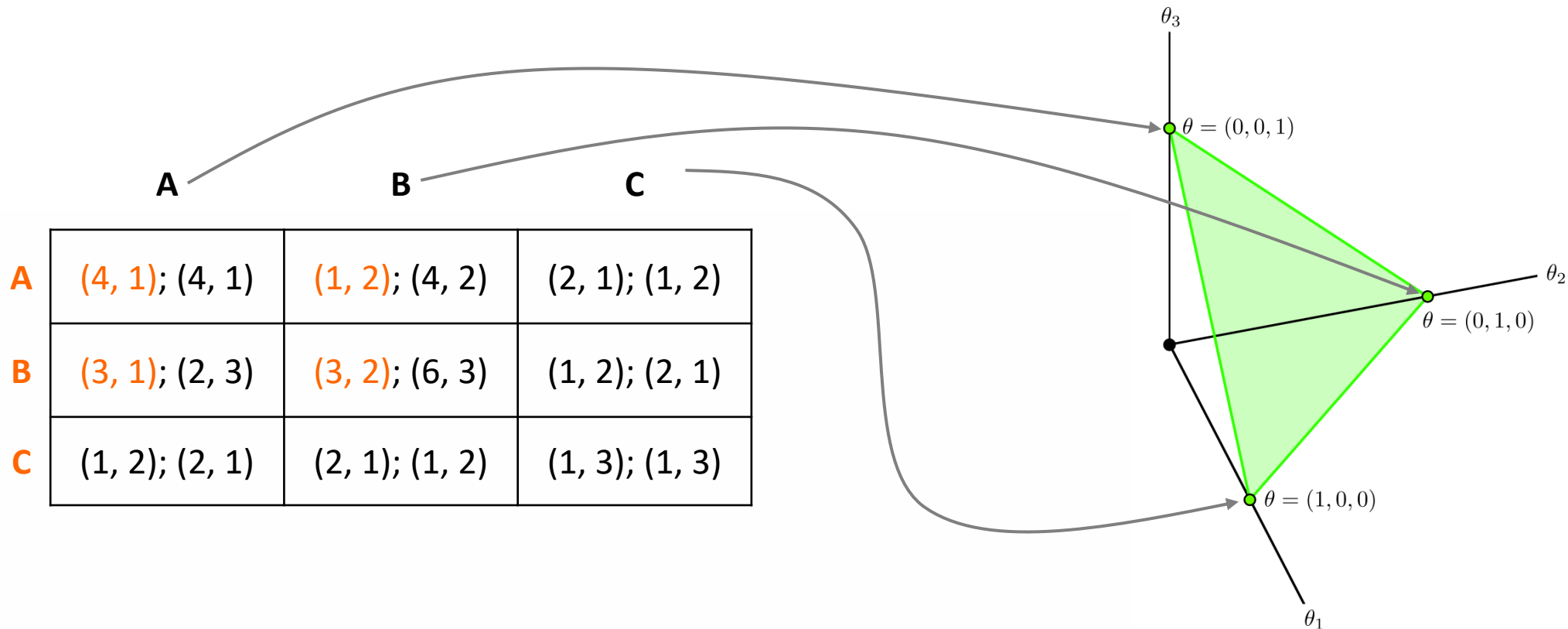
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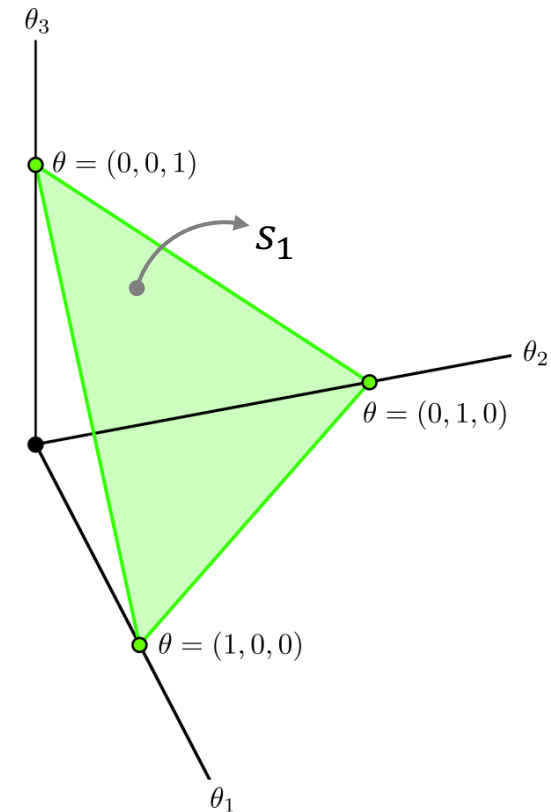


Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game



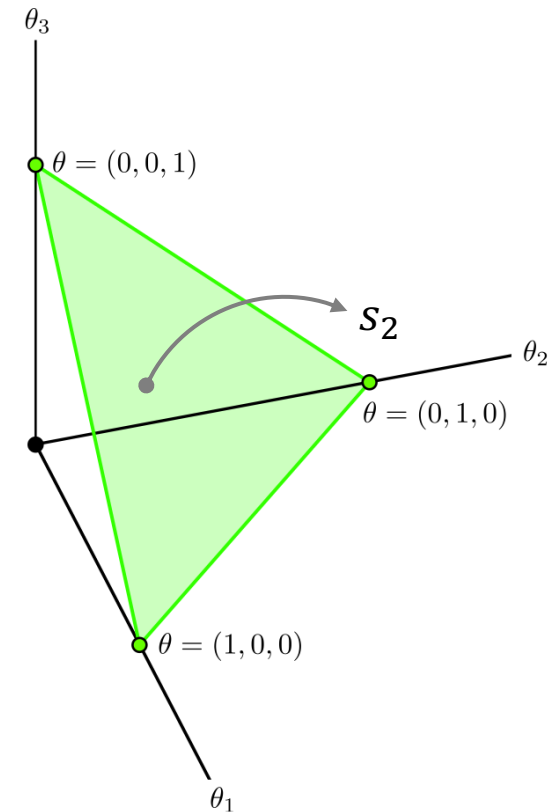
Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

	A	B	C
A	(4, 1); (4, 1)	(1, 2); (4, 2)	(2, 1); (1, 2)
B	(3, 1); (2, 3)	(3, 2); (6, 3)	(1, 2); (2, 1)
C	(1, 2); (2, 1)	(2, 1); (1, 2)	(1, 3); (1, 3)



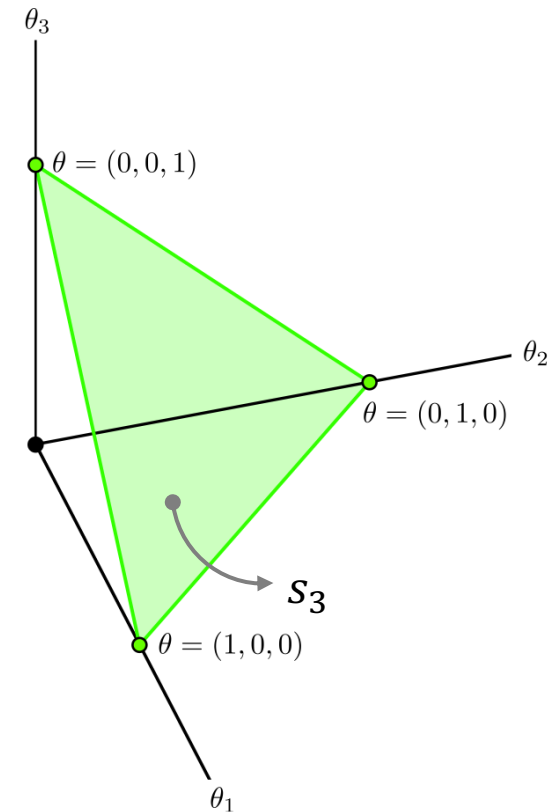
Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

	A	B	C
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B	(3, 1); (2, 3)	(3, 2); (6, 3)	(1, 2); (2, 1)
C	(1, 2); (2, 1)	(2, 1); (1, 2)	(1, 3); (1, 3)



Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

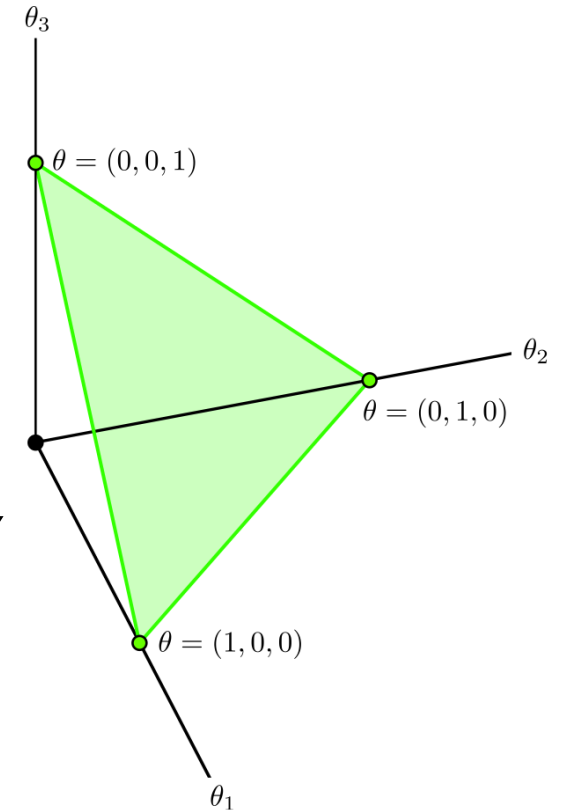
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WHY ARE NASH EQUILIBRIA NOT GUARANTEED?

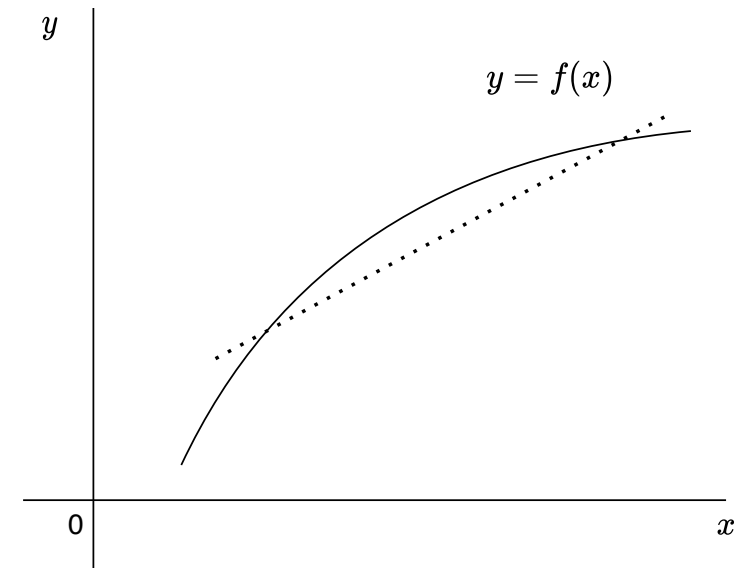
A NOVEL INTUITION

- ▶ Nash equilibria are not guaranteed in MONFGs
 - ▶ They are guaranteed in single-objective NFGs, so why not here?
- ▶ **Mixed strategy equilibria** in the MONFG are **pure strategy equilibria** in the continuous game
- ▶ Continuous games are not guaranteed to have a **pure strategy** Nash equilibrium



EXISTENCE GUARANTEE

- ▶ Existence is guaranteed with **(quasi)concave** utility functions
 - ▶ Used in economics as well
 - ▶ Represents “well-behaved” preferences
- ▶ Intuition
 - ▶ You can reduce an MONFG to a continuous game
 - ▶ In this game it is known that a pure strategy Nash equilibrium exists when assuming only quasiconcave utility functions
 - ▶ This equilibrium is also an equilibrium in the original MONFG



NON-EXISTENCE

- ▶ We can show that no Nash equilibrium exists in this game
 - ▶ With **strict convex** utility functions

- ▶ Saving grace
 - ▶ Techniques we developed are generally useful
 - ▶ Can use it to prove counterexamples for additional possible properties
 - ▶ Can use it for an efficient algorithm (future work)

	A	B
A	(2, 0); (1, 0)	(1, 0); (0, 2)
B	(0, 1); (2, 0)	(0, 2); (0, 1)

$$u_1(p_1, p_2) = u_2(p_1, p_2) = p_1^2 + p_2^2$$

RELATIONS BETWEEN OPTIMISATION CRITERIA

MIXED STRATEGY EQUILIBRIA

► **No relation** between both optimisation criteria *in general*

	A	B
A	(1, 0); (1, 0)	(0, 1); (0, 1)
B	(0, 1); (0, 1)	(-10, 0); (-10, 0)

Multi-objective reward vectors

	A	B
A	0.1; 0.1	0; 0
B	0; 0	-0.1; -0.1

Scalarised utility for both agents

No sharing of number of equilibria or equilibria themselves

RELATIONS BETWEEN OPTIMISATION CRITERIA

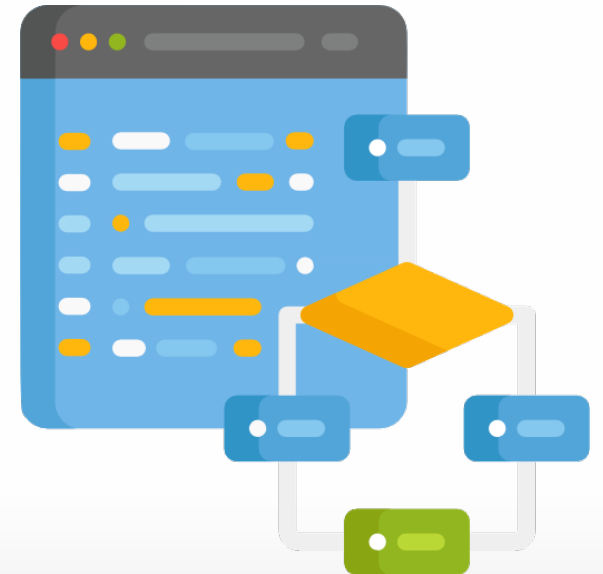
PURE STRATEGY EQUILIBRIA

- ▶ **Relation** when only considering ***pure strategy*** equilibria
 - ▶ Pure strategy equilibrium under SER is also one under ESR
 - ▶ Bidirectional when assuming (quasi)convex utility functions

- ▶ We can extend this to ***blended*** settings
 - ▶ Pure strategy equilibrium under SER is also one in any blended setting
 - ▶ Bidirectional when assuming (quasi)convex utility functions

ALGORITHMIC IMPLICATIONS

- ▶ Algorithm for calculating ***all pure strategy equilibria*** in a given MONFG ***with quasiconvex utility functions***
- ▶ Shown to work because of our theoretical contributions



NASH EQUILIBRIA

RECENT WORK

Algorithm 1 Computing all PSNE in an MONFG

Input: an MONFG $G = (N, \mathcal{A}, \mathbf{p})$ and quasiconvex utility functions $u = (u_1, \dots, u_n)$

```

1: function REDUCE_MONFG(monfg, u)
2:    $N, \mathcal{A}, \mathbf{p} \leftarrow \text{monfg}$ 
3:    $u_1, \dots, u_n \leftarrow u$ 
4:    $f \leftarrow (u_1 \circ \mathbf{p}_1, \dots, u_n \circ \mathbf{p}_n)$ 
5:    $G' \leftarrow (N, \mathcal{A}, f)$ 
6:   return  $G'$ 
7: end function
8: function COMPUTE_ALL_PSNE(nfg)
9:    $S = \emptyset$ 
10:  for PS in nfg do
11:    if PS is a PSNE then
12:       $S \leftarrow S \cup \{\text{PS}\}$ 
13:    end if
14:  end for
15:  return  $S$ 
16: end function
17: nfg  $\leftarrow$  REDUCE_MONFG( $G, u$ )
18: PSNE  $\leftarrow$  COMPUTE_ALL_PSNE(nfg)

```

▷ An induced normal-form game

▷ Loop over all pure strategies

▷ If it is a PSNE add it to the set

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Reduce the MONFG

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```

▷ An induced normal-form game
 ▷ Loop over all pure strategies
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Solve the trade-off game

CONCLUSION

- ▶ Lots of new theoretical insights
 - ▶ Nash equilibrium guarantees
 - ▶ Relation between optimisation criteria when only considering pure strategies
 - ▶ We can extend this to blended settings

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CONCLUSION

- ▶ Lots of new theoretical insights
 - ▶ Nash equilibrium guarantees
 - ▶ Relation between optimisation criteria when only considering pure strategies
 - ▶ We can extend this to blended settings
- ▶ Incorporate everything into a novel algorithm
- ▶ Additional guarantees for MONFGs
 - ▶ Zero-sum games
 - ▶ Exploit continuous game reduction

CONCLUSION

- ▶ Lots of new theoretical insights
 - ▶ Nash equilibrium guarantees
 - ▶ Relation between optimisation criteria when only considering pure strategies
 - ▶ We can extend this to blended settings
- ▶ Incorporate everything into a novel algorithm
- ▶ Additional guarantees for MONFGs
 - ▶ Zero-sum games
 - ▶ Exploit continuous game reduction
- ▶ More algorithmic work
 - ▶ Use theorems to find Nash equilibria efficiently

OTHER WORK

▶ Explored communication

- ▶ Communication protocols
- ▶ Commit to actions or policies
- ▶ Evaluate in different settings

▶ Explored commitment

- ▶ Theoretically
- ▶ Evaluate using reinforcement learning

