

ARTIFICIAL  
INTELLIGENCE  
RESEARCH GROUP



## TAKING MULTI-OBJECTIVE GAMES TO THE NEXT LEVEL

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## OVERVIEW

- ▶ What are multi-objective games?
- ▶ The utility-based approach
- ▶ Existence guarantees
- ▶ Algorithms
- ▶ 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*

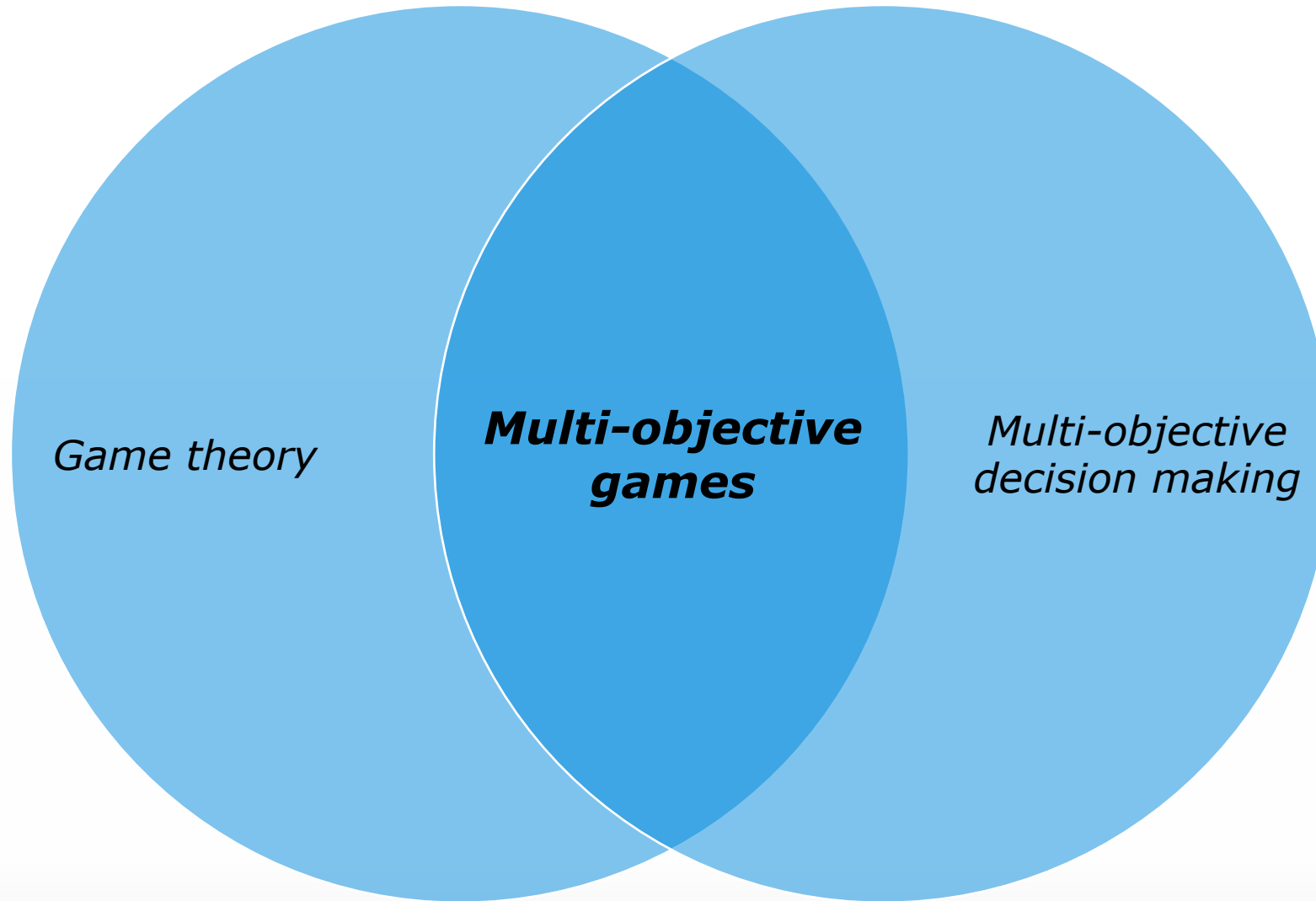


*Game theory*

# ***Rational individuals concerned with more than one objective***



*Multi-objective decision making*



# MULTI-OBJECTIVE GAME

## BASICS

### Multi-Objective Normal-Form Games (MONFGs) [Blackwell, 1954]


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

# TERMINOLOGY

## BASICS

### Multi-Objective Normal-Form Games (MONFGs)

Called ***actions*** or ***pure strategies***



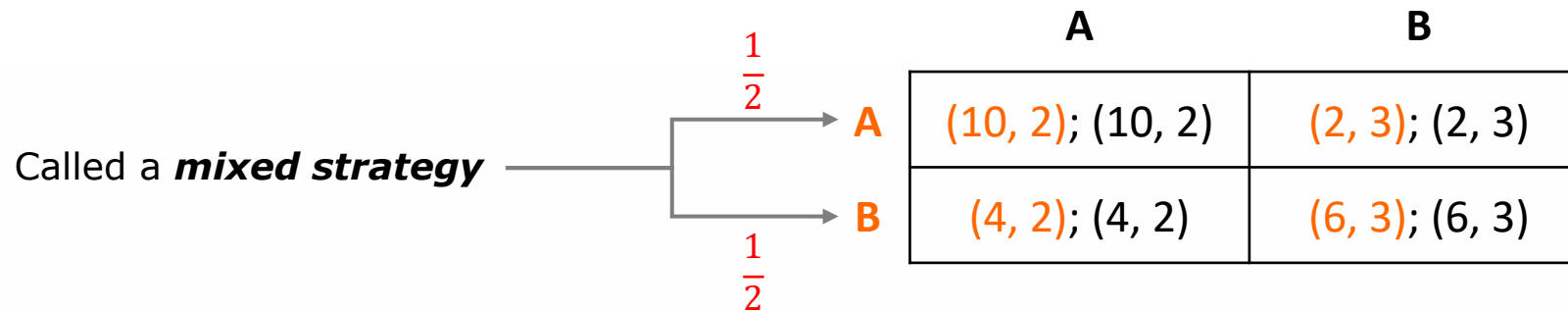
	A	B
A	(10, 2); (10, 2)	(2, 3); (2, 3)
B	(4, 2); (4, 2)	(6, 3); (6, 3)



# TERMINOLOGY

## BASICS

### Multi-Objective Normal-Form Games (MONFGs)



# UTILITY-BASED APPROACH

## BASICS

It's not obvious how to compare these two outcomes

Option A

$(1, 2)$

Option B

$(2, 1)$



# UTILITY-BASED APPROACH

## BASICS

But we can still pick our preferred option!

Option A

$(1, 2)$

Option B

$(2, 1)$



# UTILITY-BASED APPROACH

## BASICS

Assume that all decision-makers have a utility function  
[Rojers et al., 2013]



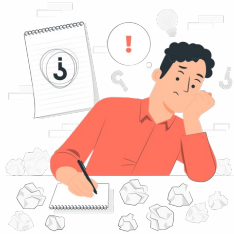
$$u_i: \mathbb{R}^d \rightarrow \mathbb{R}$$

# UTILITY-BASED APPROACH

## BASICS



$$u_2(p_1, p_2) = p_1 + p_2$$



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

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

[Rădulescu et al., 2020]

# UTILITY-BASED APPROACH

## BASICS



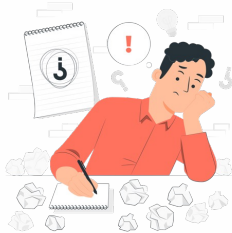
$$u_2(p_1, p_2) = p_1 + p_2$$

**A**

**B**

<b>A</b>	(10, 2); (10, 2)	(2, 3); (2, 3)
<b>B</b>	(4, 2); (4, 2)	(6, 3); (6, 3)

How and when to apply this utility function?

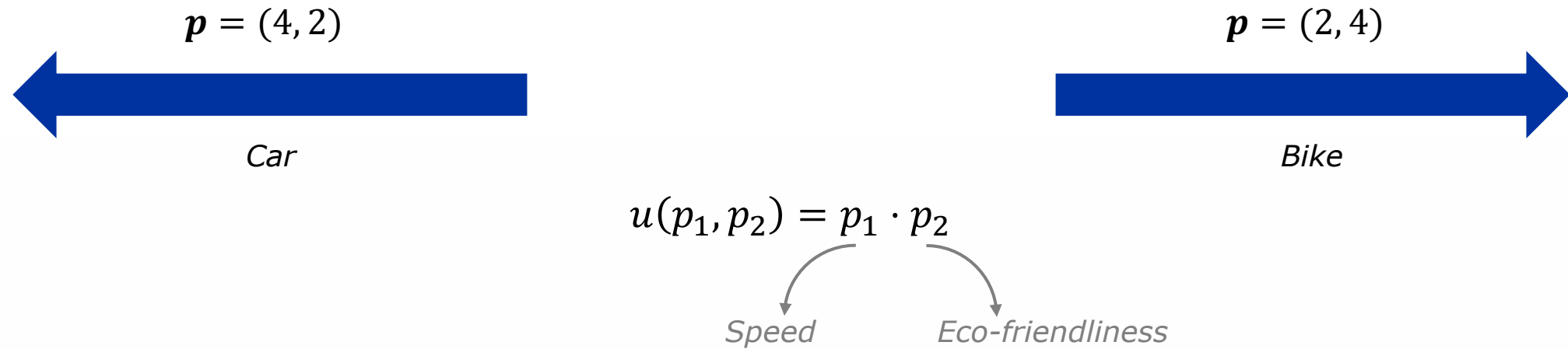


$$u_1(p_1, p_2) = p_1 \cdot p_2$$

[Rădulescu et al., 2020]

## OPTIMISATION CRITERIA

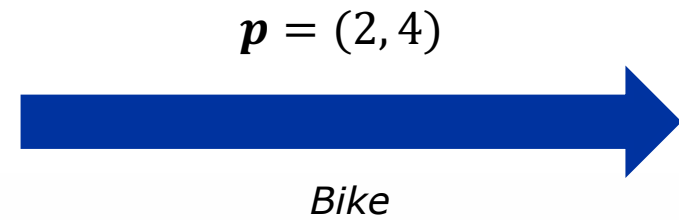
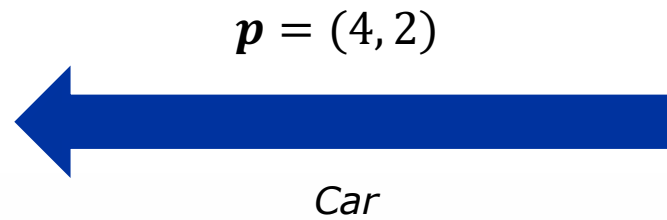
### EXAMPLE



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

## OPTIMISATION CRITERIA

### EXAMPLE



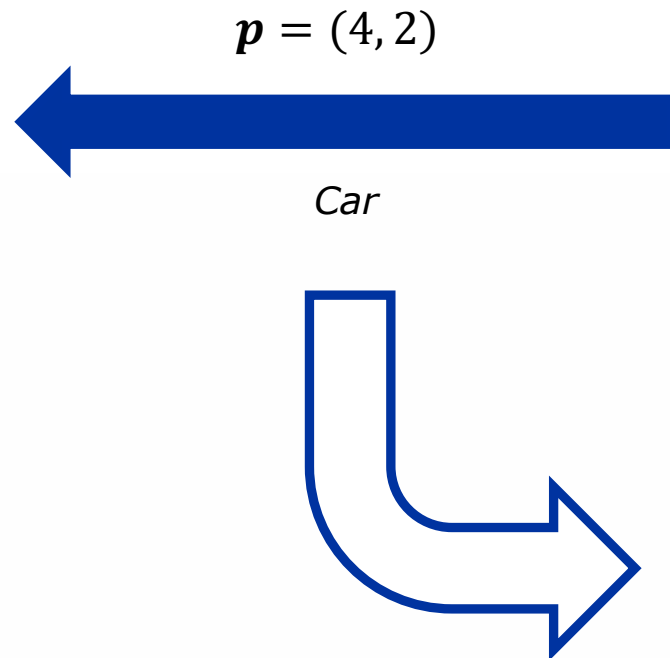
$$u(p_1, p_2) = p_1 \cdot p_2$$

***Expected Scalarised Returns (ESR)***



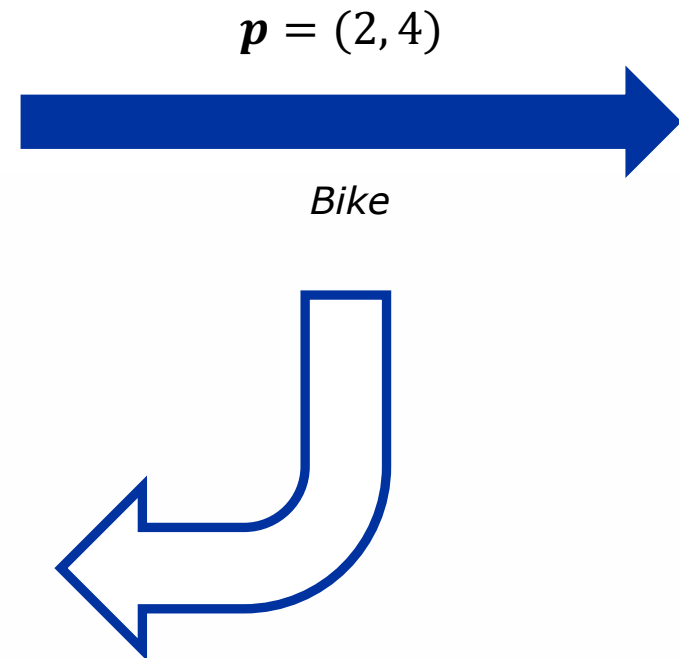
# OPTIMISATION CRITERIA

## EXAMPLE



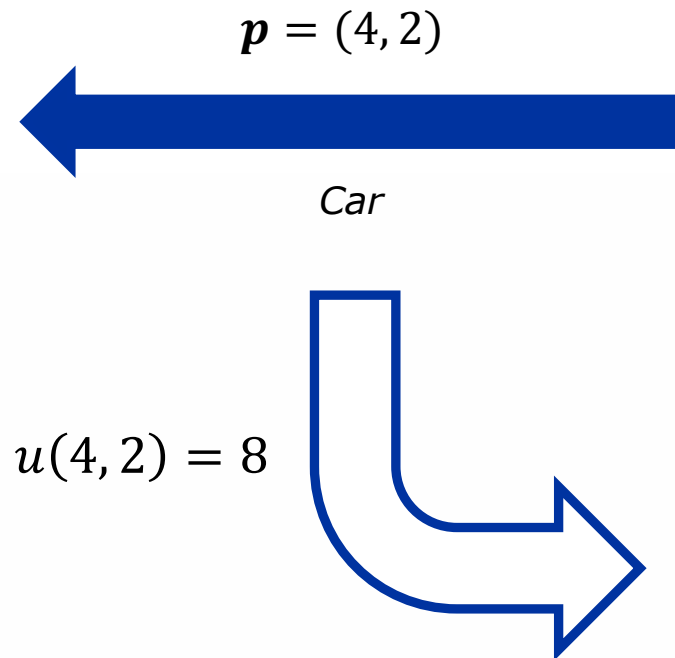
$$u(p_1, p_2) = p_1 \cdot p_2$$

**ESR**



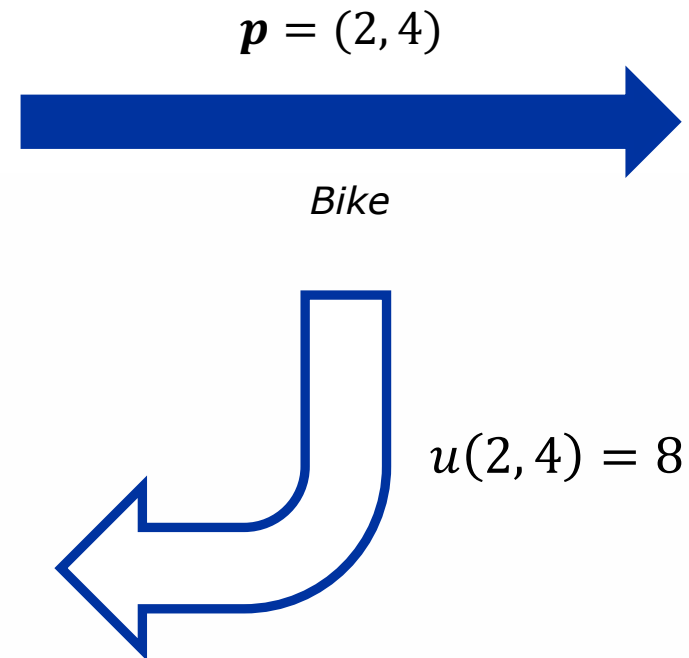
# OPTIMISATION CRITERIA

## EXAMPLE



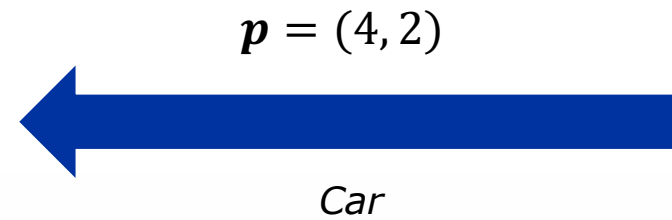
$$u(p_1, p_2) = p_1 \cdot p_2$$

**ESR**

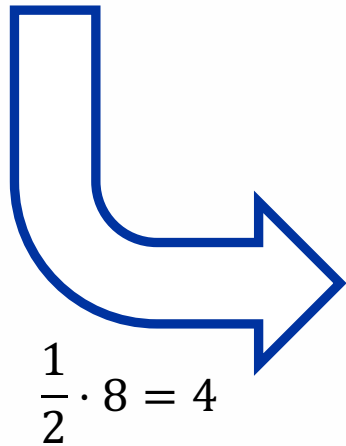


# OPTIMISATION CRITERIA

## EXAMPLE

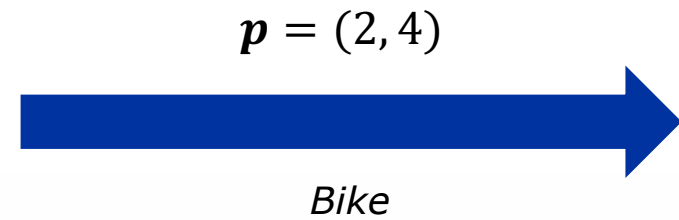


$$u(4, 2) = 8$$

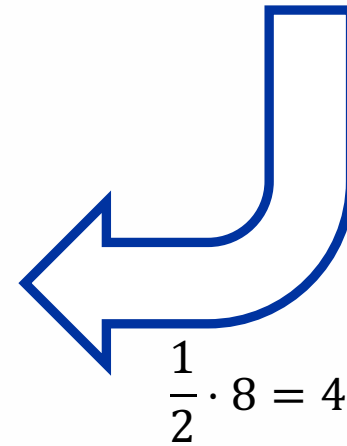


$$u(p_1, p_2) = p_1 \cdot p_2$$

**ESR**

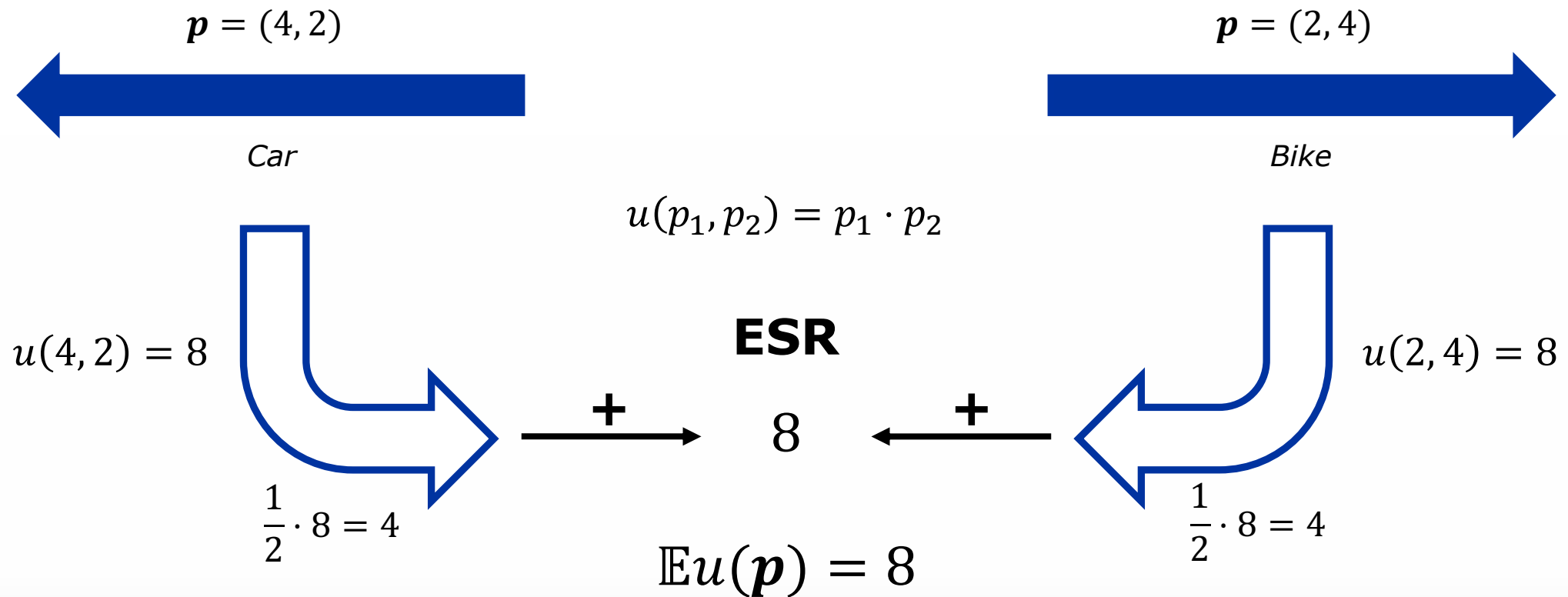


$$u(2, 4) = 8$$



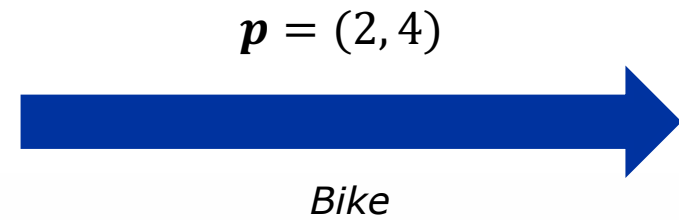
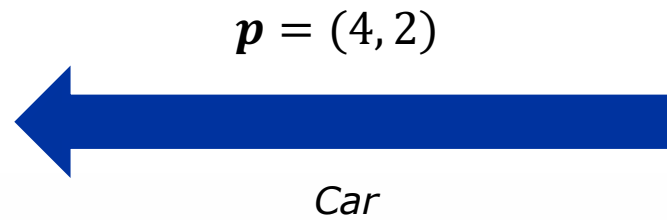
# OPTIMISATION CRITERIA

## EXAMPLE



## OPTIMISATION CRITERIA

### EXAMPLE

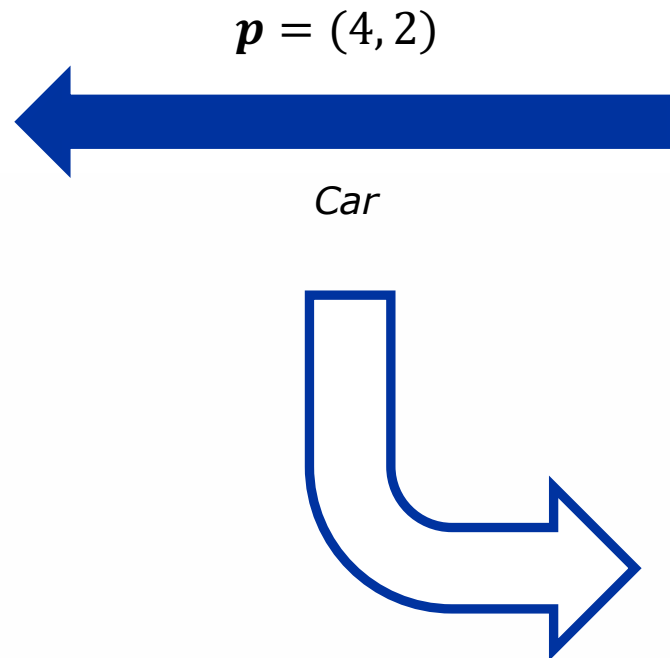


$$u(p_1, p_2) = p_1 \cdot p_2$$

***Scalarised Expected Returns (SER)***

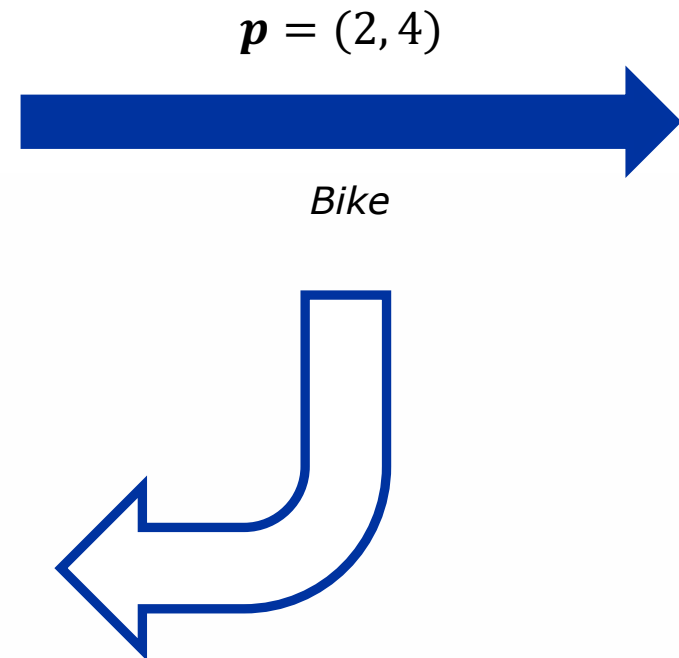
# OPTIMISATION CRITERIA

## EXAMPLE



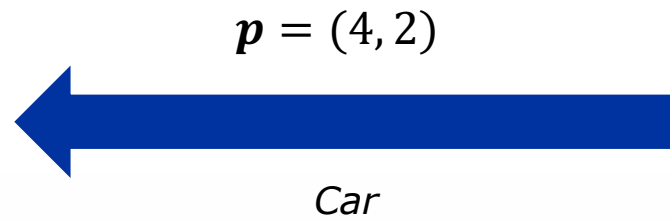
$$u(p_1, p_2) = p_1 \cdot p_2$$

**SER**

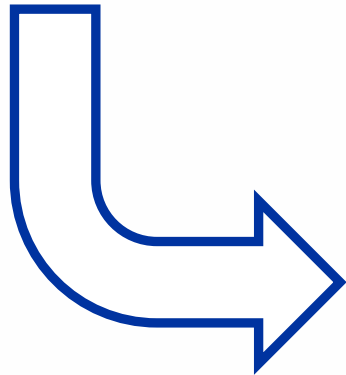


# OPTIMISATION CRITERIA

## EXAMPLE

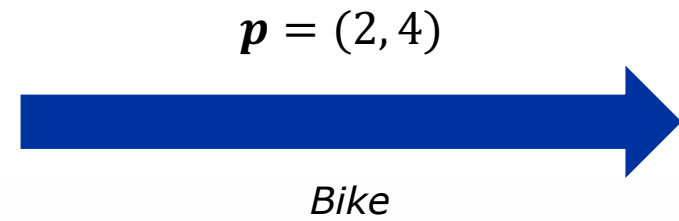


$$\frac{1}{2} \cdot (4, 2) = (2, 1)$$

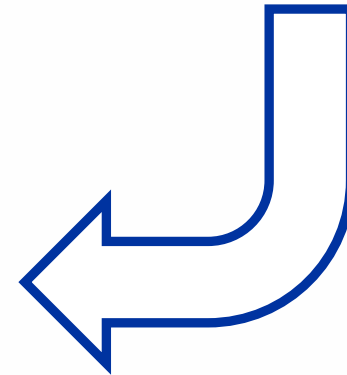


$$u(p_1, p_2) = p_1 \cdot p_2$$

**SER**

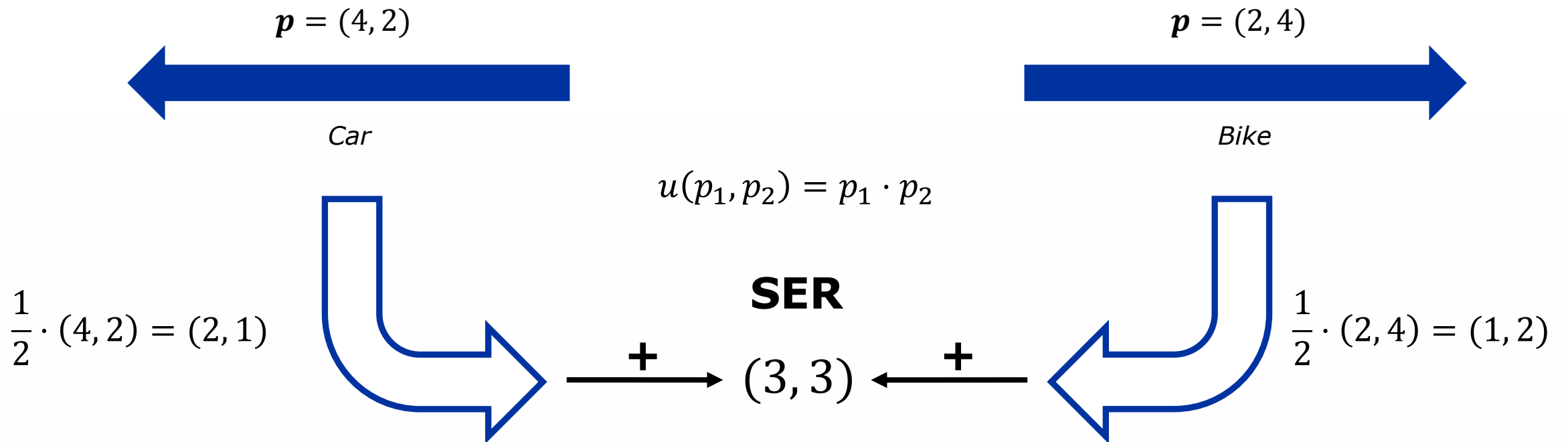


$$\frac{1}{2} \cdot (2, 4) = (1, 2)$$



# OPTIMISATION CRITERIA

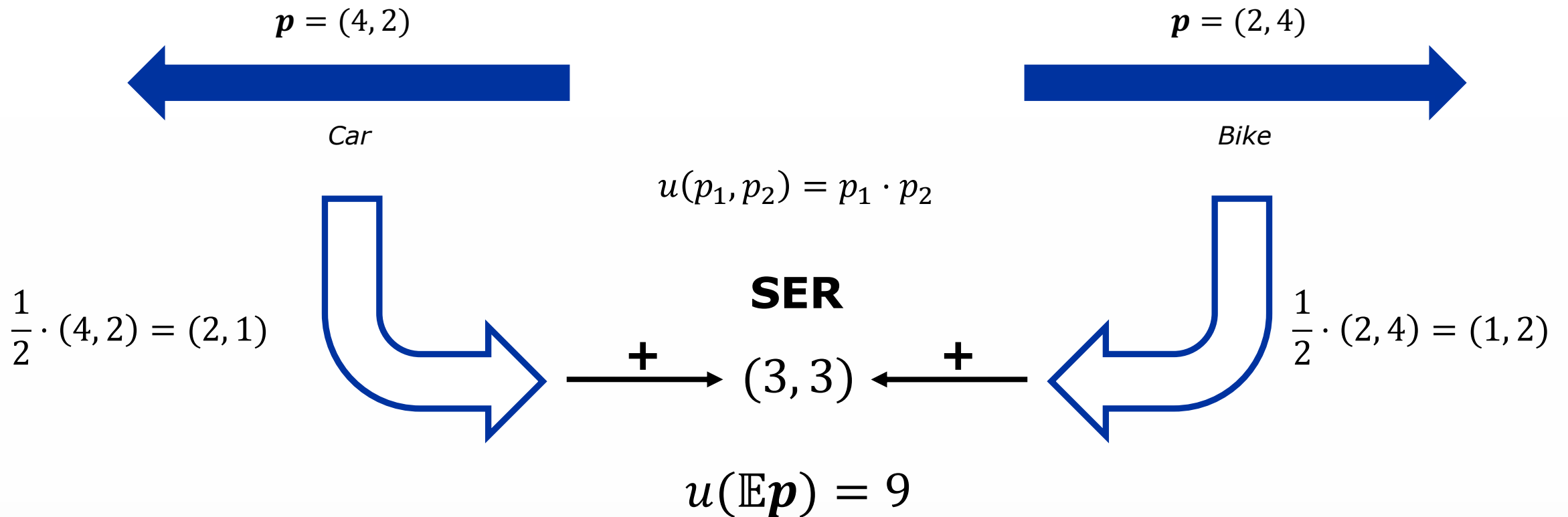
## EXAMPLE





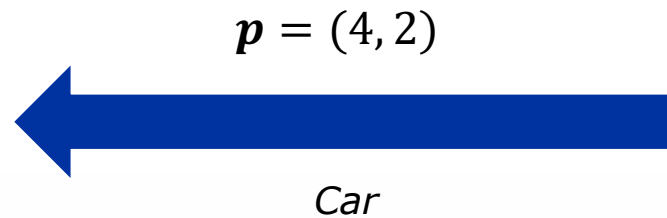
# OPTIMISATION CRITERIA

## EXAMPLE

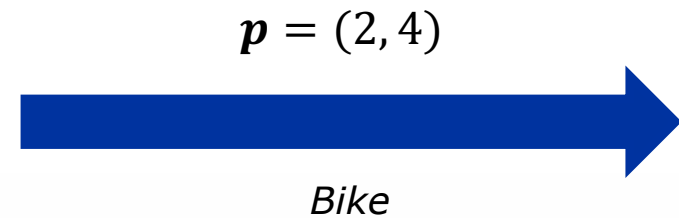


## OPTIMISATION CRITERIA

### 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**

# NASH EQUILIBRIUM

## BACKGROUND

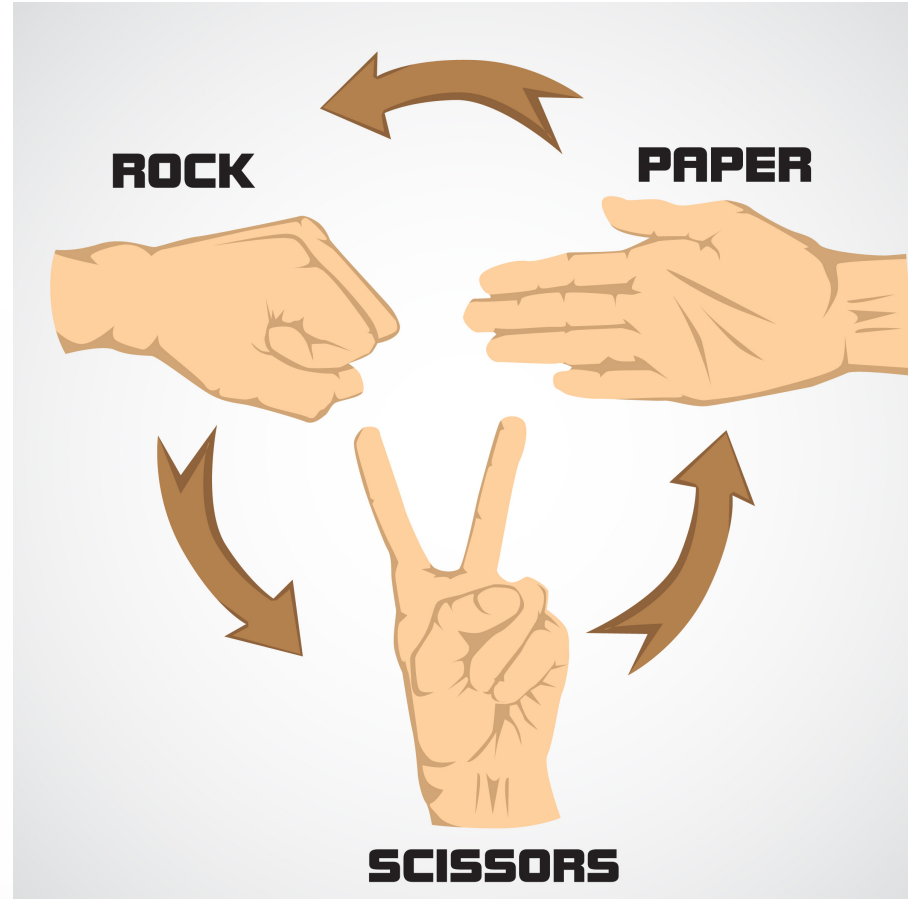
### ▶ Nash equilibrium

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



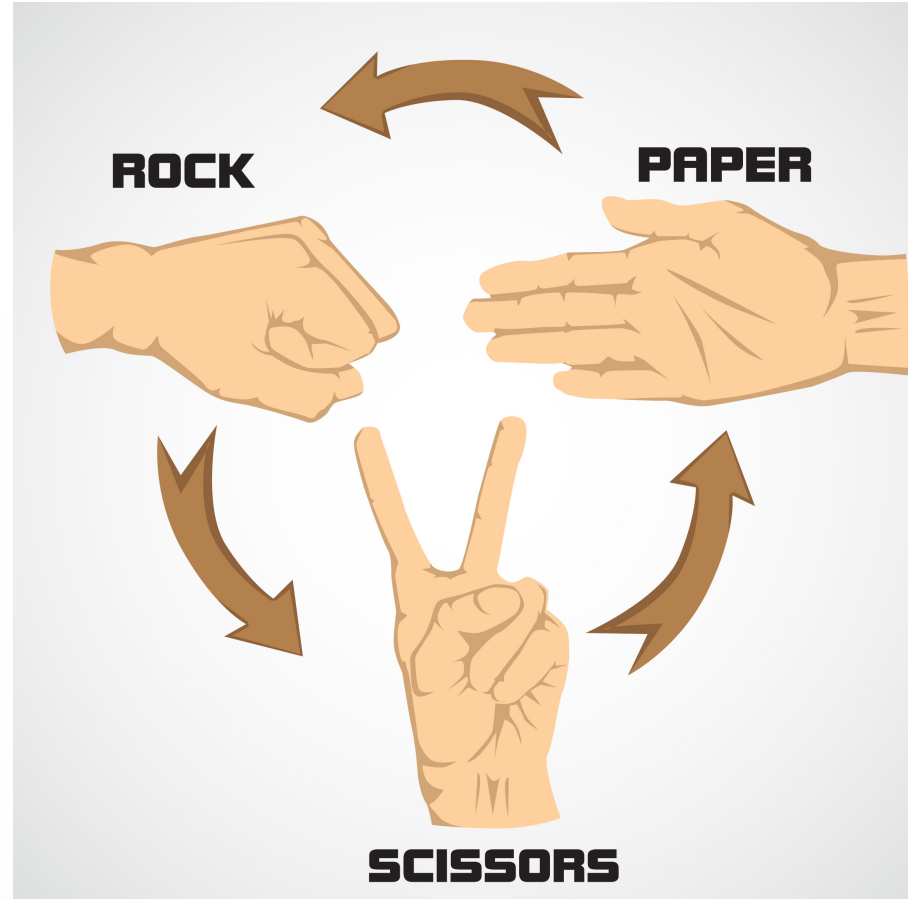
# NASH EQUILIBRIUM

## BACKGROUND



# NASH EQUILIBRIUM

## BACKGROUND

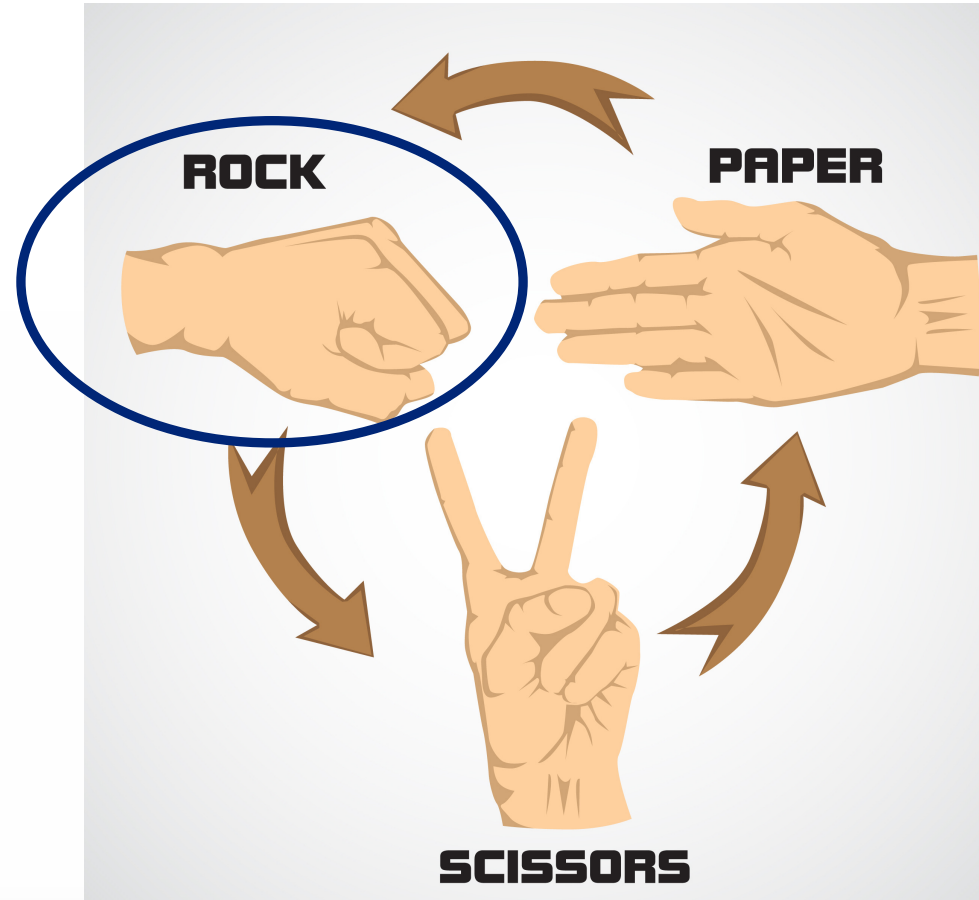


*We're looking for a joint strategy where noone has an incentive to change their strategy*



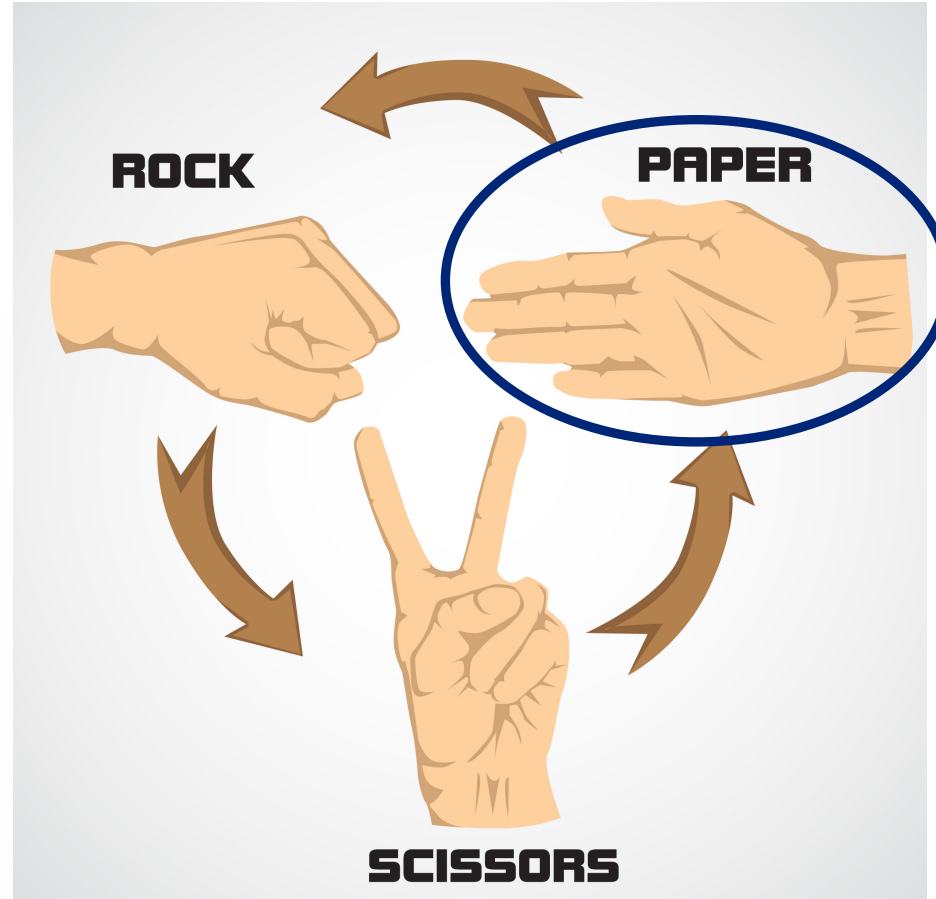
# NASH EQUILIBRIUM

## BACKGROUND



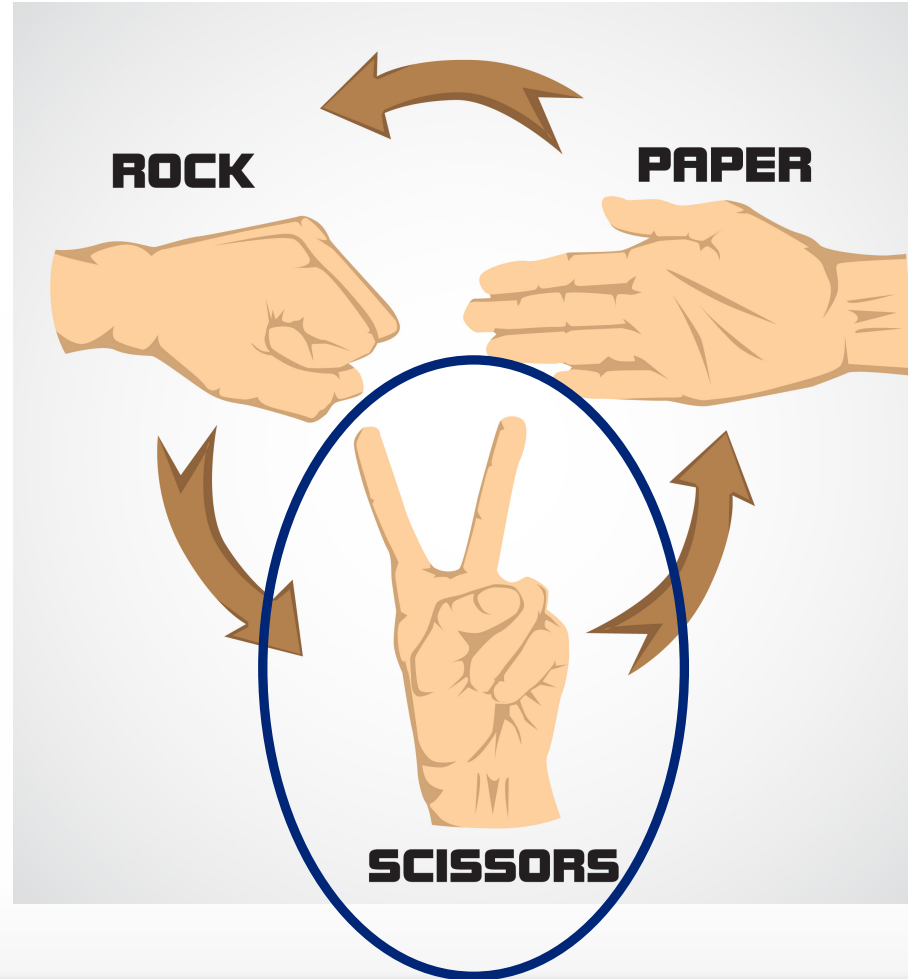
# NASH EQUILIBRIUM

## BACKGROUND



# NASH EQUILIBRIUM

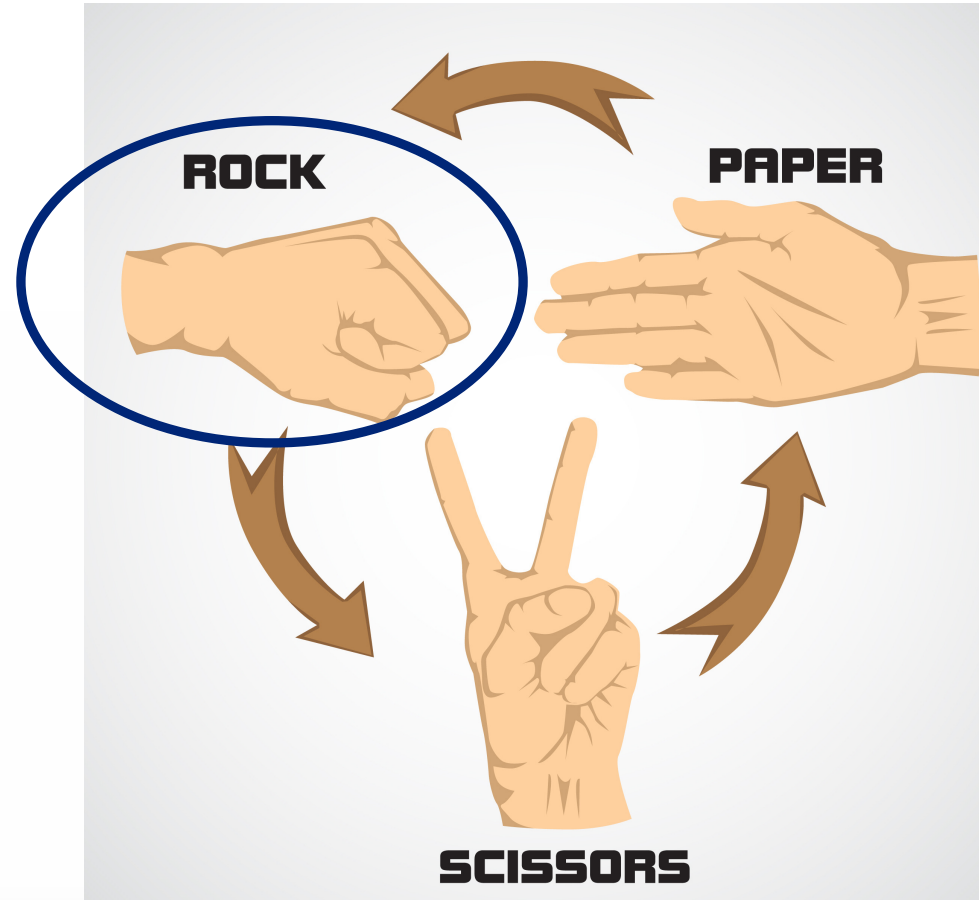
## BACKGROUND





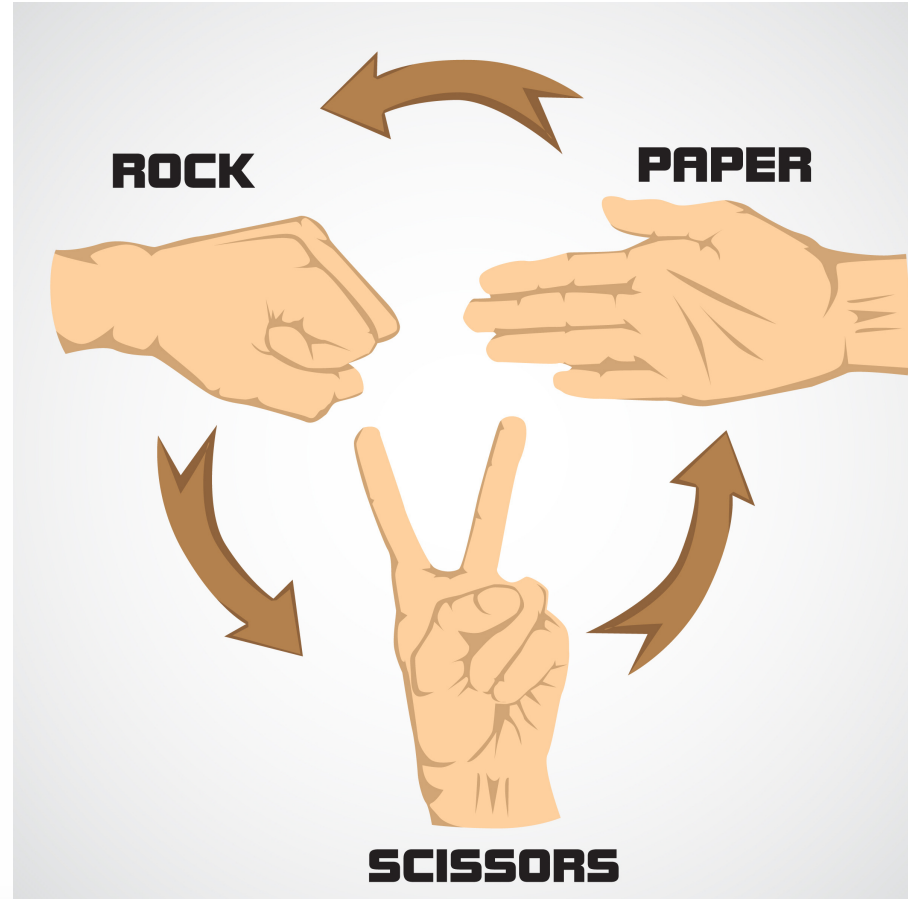
# NASH EQUILIBRIUM

## BACKGROUND



# NASH EQUILIBRIUM

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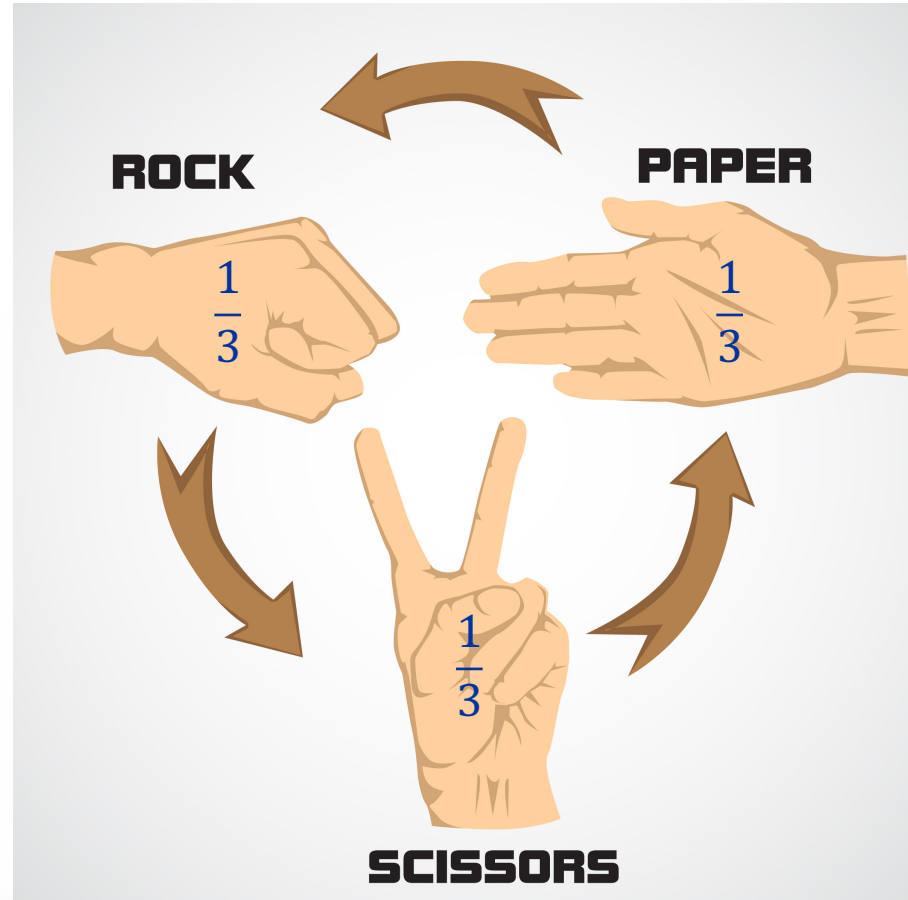


*We're looking for a joint strategy where no one has an incentive to change their strategy*



# NASH EQUILIBRIUM

## BACKGROUND



*Whatever you do in response,  
you will break even in  
expectation.*



# NASH EQUILIBRIUM

## BACKGROUND

*This is also true for your  
opponent!*



# NASH EQUILIBRIUM

## BACKGROUND

*You're playing a  
Nash equilibrium*



# NASH EQUILIBRIUM

## BACKGROUND

### ► Nash equilibria

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

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

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



# NASH EQUILIBRIUM

## BACKGROUND

### ► Nash equilibria

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

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

*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)



# NASH EQUILIBRIUM

## BACKGROUND

### ► Nash equilibria

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

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

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)

Strictly worse to deviate to B





# NASH EQUILIBRIUM

## BACKGROUND

### ► Nash equilibria

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

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

*Nash equilibrium* →

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

Strictly worse to deviate to B

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



# NASH EQUILIBRIUM

## BACKGROUND

### ▶ Nash equilibria

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

Not guaranteed to exist in general!

[Rădulescu et al., 2020]



# GOAL

- ▶ Theoretical
  - ▶ Relation with other games
  - ▶ Existence or non-existence guarantees
- ▶ Algorithms
  - ▶ Computing equilibria
  - ▶ (Learning equilibria)



# 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)

[Röpke et al., 2022]

# WHAT ARE MULTI-OBJECTIVE GAMES?

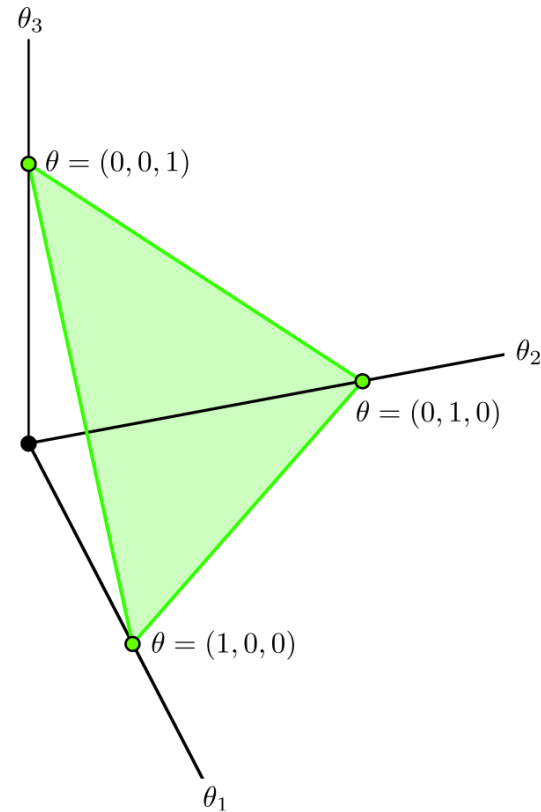
## A NOVEL INTUITION

	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

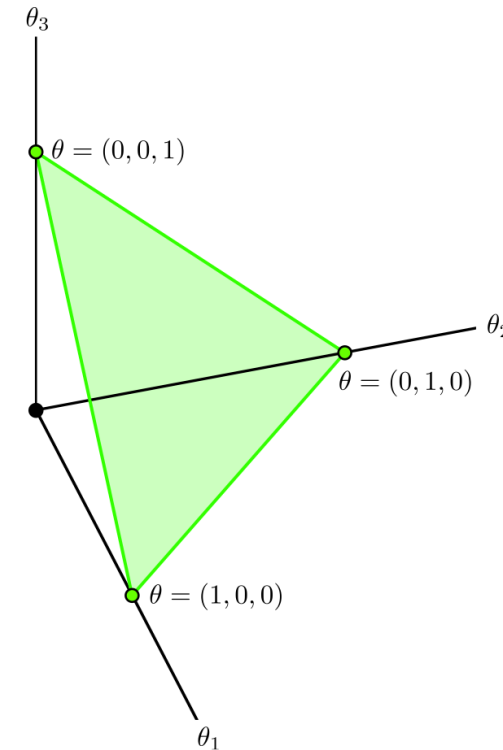


**To this**

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

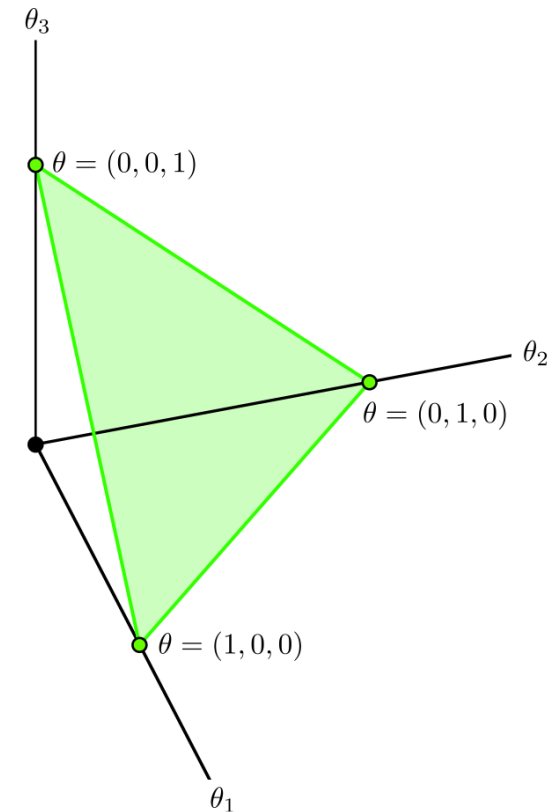
## ***Continuous game***

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



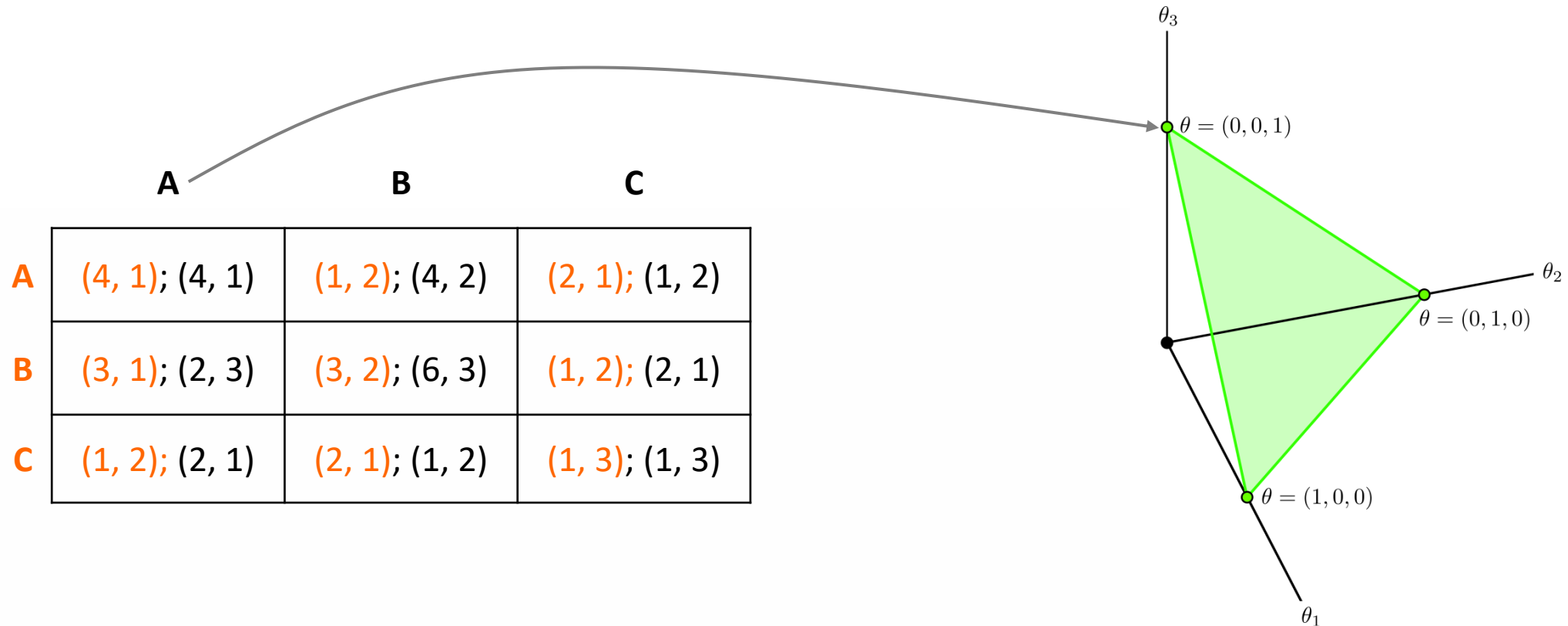
Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

	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)

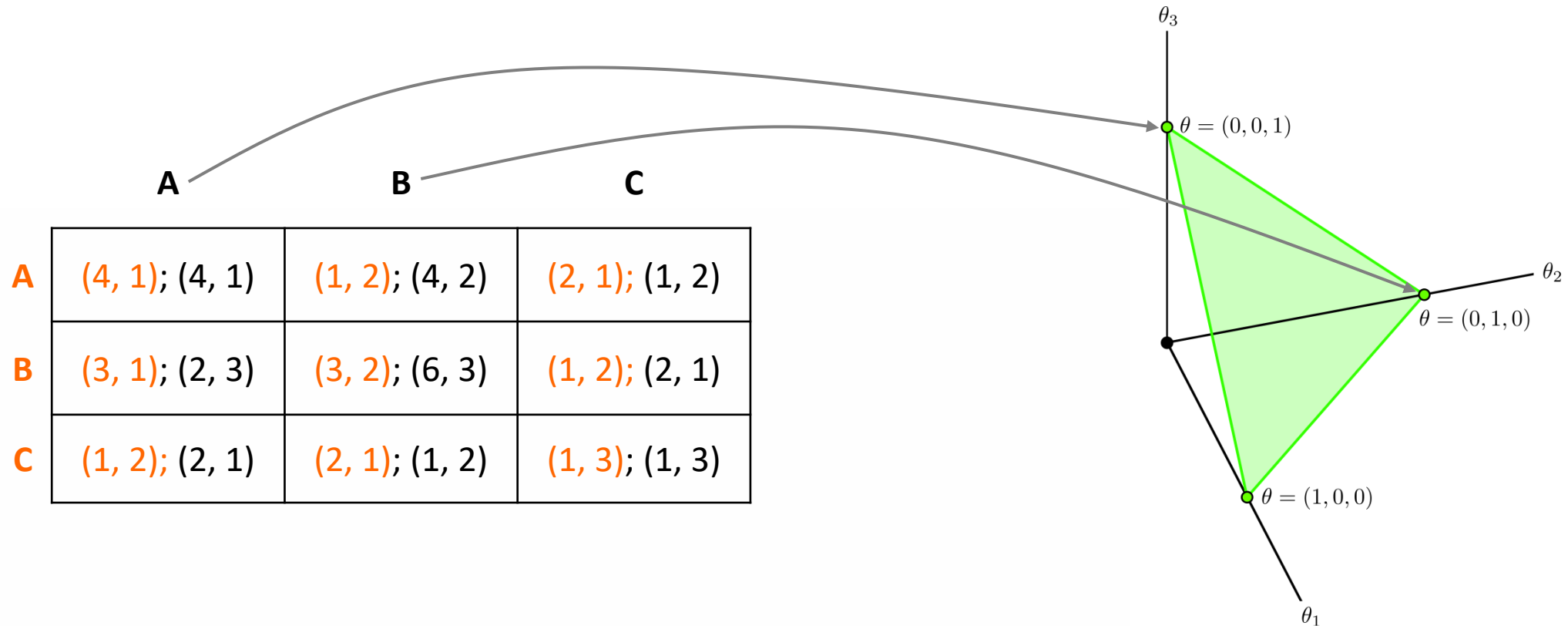




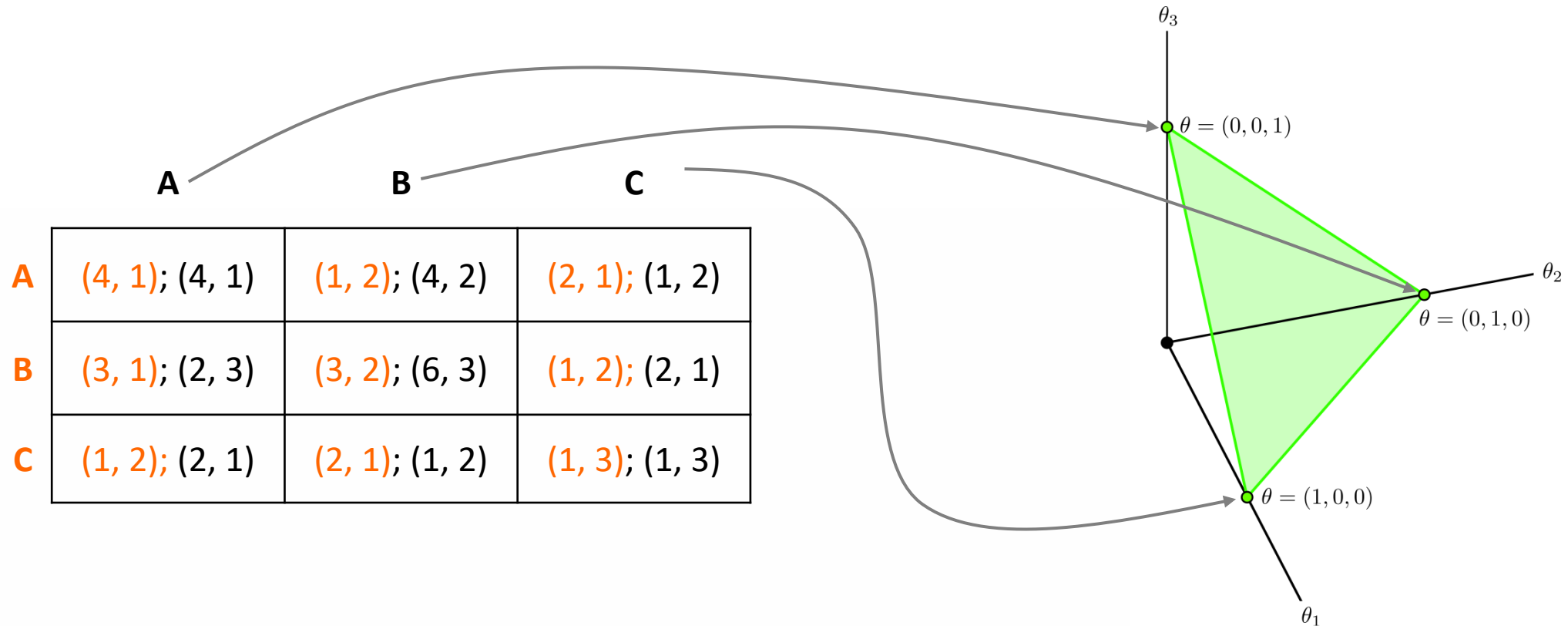
Make every ***mixed strategy*** in the MONFG a ***pure strategy*** in the continuous game



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

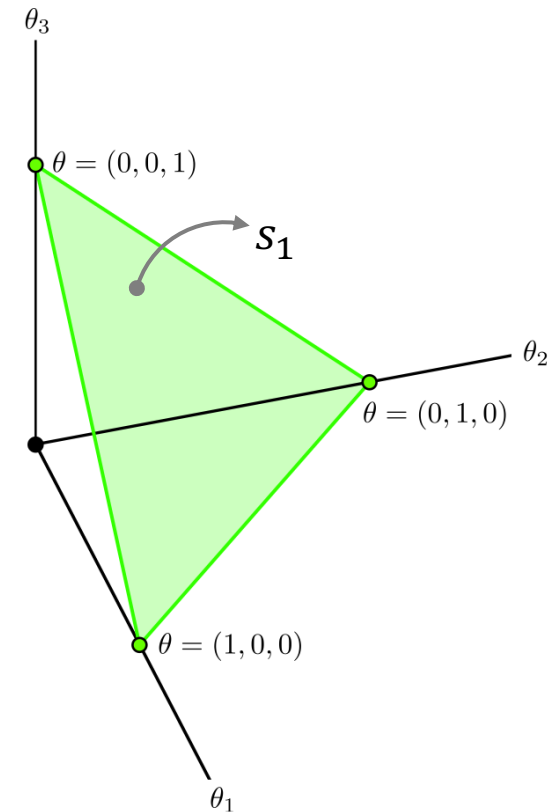


Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game



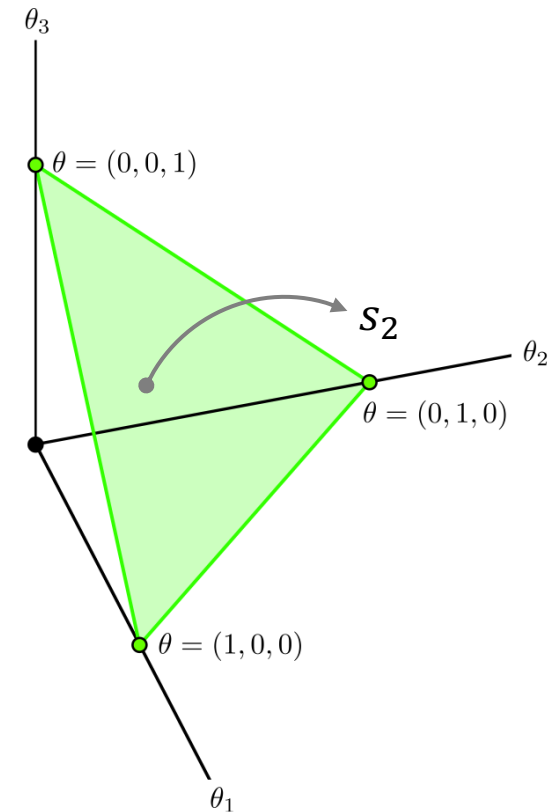
Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

	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)



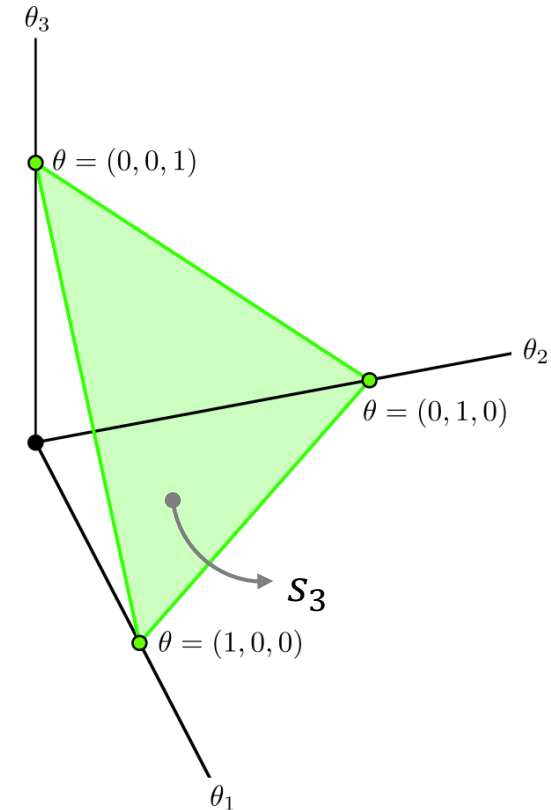
Make every *mixed strategy* in the MONFG a *pure strategy* in the continuous game

	A	B	C
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

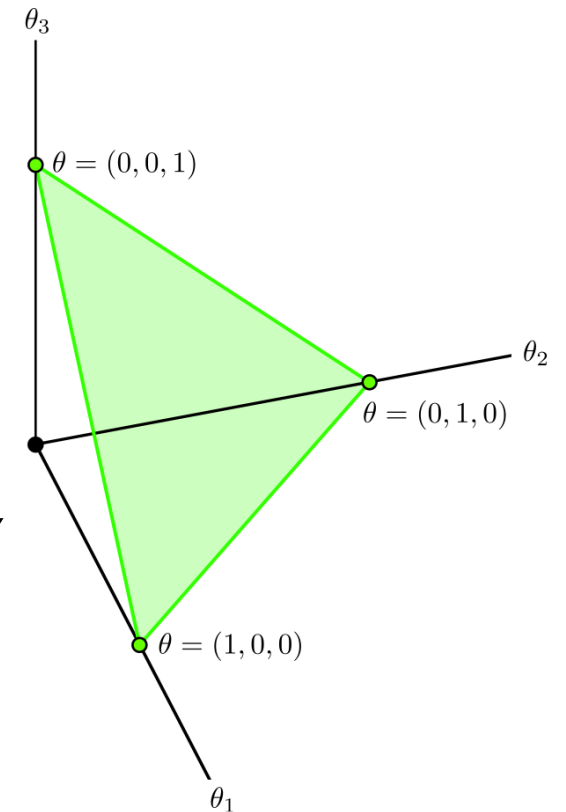
	A	B	C
A	(4, 1); (4, 1)	(1, 2); (4, 2)	(2, 1); (1, 2)
B	(3, 1); (2, 3)	(3, 2); (6, 3)	(1, 2); (2, 1)
C	(1, 2); (2, 1)	(2, 1); (1, 2)	(1, 3); (1, 3)



# 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



[Röpke et al., 2022]

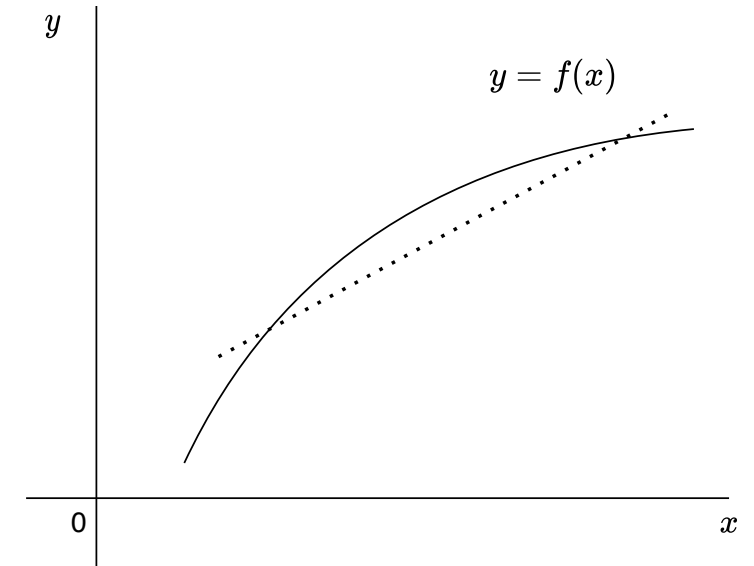
# EXISTENCE GUARANTEE

## ▶ 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



[Röpke et al., 2022]



## 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)

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

[Röpke et al., 2022]

# RELATIONS BETWEEN OPTIMISATION CRITERIA

## NASH EQUILIBRIA

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

- No sharing of number of equilibria or equilibria themselves

	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

### ► **Relation** when only considering *pure strategy* equilibria

- Pure strategy equilibrium under SER is also one under ESR
- 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



[Röpke et al., 2022]

# NASH EQUILIBRIA

## RECENT WORK

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### Algorithm 1 Computing all PSNE in an MONFG

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

```

1: function REDUCE_MONFG(monfg, u)
2:    $N, \mathcal{A}, p \leftarrow \text{monfg}$ 
3:    $u_1, \dots, u_n \leftarrow u$ 
4:    $f \leftarrow (u_1 \circ p_1, \dots, u_n \circ p_n)$ 
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**

# NASH EQUILIBRIA

## RECENT WORK

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▷ An induced normal-form game

▷ Loop over all pure strategies  
▷ If it is a PSNE add it to the set

***Solve the trade-off game***

# CONCLUSION

- ▶ Lots of new theoretical insights
  - ▶ Relation to other games opens up a new perspective
  - ▶ Equilibrium existence and non-existence
  - ▶ Things are simpler when only considering pure strategies

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# CONCLUSION

- ▶ Lots of new theoretical insights
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  - ▶ Equilibrium existence and non-existence
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- ▶ Additional guarantees for MONFGs
  - ▶ Zero-sum games
  - ▶ Exploit continuous game reduction

# CONCLUSION

- ▶ Lots of new theoretical insights
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- ▶ 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

## REFERENCES

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## BLENDED SETTING

### EXAMPLE

#### ▶ Apartment building gym

- ▶ Treadmills
- ▶ Weightlifting equipment

#### ▶ Shared between residents

- ▶ One athlete
- ▶ Others are amateurs



*Small apartment building gym*



*Athlete*



*Amateur*

# BLENDED SETTING

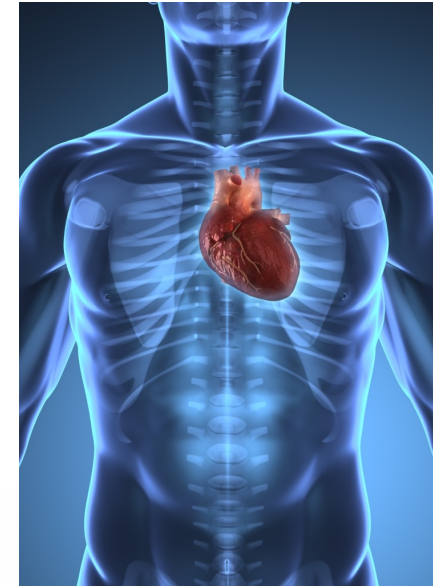
## EXAMPLE

### ► Objectives

- Improve cardiovascular health
- Improve strength

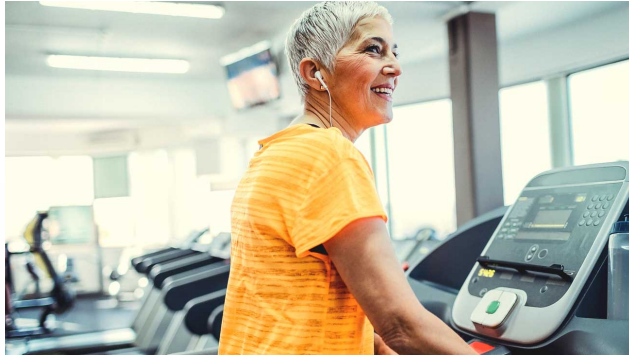
### ► Athlete plays a game against another resident

- Select equipment
- Selecting the same reduces effectivity



## BLENDED SETTING

### EXAMPLE



*Player 1: Amateur*

Maximise utility of each (occasional) workout

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

**ESR**



*Player 2: Athlete*

Sustain a training schedule

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

**SER**

## BLENDED SETTING

### EXAMPLE

	Cardio	Lifting
Cardio	(4, 1); (4, 1)	(5, 1); (1, 4)
Lifting	(1, 4); (5, 1)	(1, 3); (1, 3)

*The multi-objective reward vectors.*

	Cardio	Lifting
Cardio	17; 4	26; 4
Lifting	5; 5	4; 3

*The ESR utilities.*

- ▶ The ESR player will always go running
  - ▶ Dominates weightlifting
- ▶ What is the best-response for the SER player?

## BLENDED SETTING

### EXAMPLE

	Cardio	Lifting
Cardio	(4, 1); (4, 1)	(5, 1); (1, 4)
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*The multi-objective reward vectors.*

	Cardio	Lifting
Cardio	17; 4	26; 4
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*The ESR utilities.*

### ► SER player wants to mix over cardio and lifting

- Optimal balance
- Sustainable training program



## BLENDED SETTING

### EXAMPLE

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Cardio	(4, 1); (4, 1)	(5, 1); (1, 4)
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*The multi-objective reward vectors.*

	Cardio	Lifting
Cardio	17; 4	26; 4
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*The ESR utilities.*

►  $\{(1 \text{ Cardio}, 0 \text{ Lifting}), (\frac{1}{2} \text{ Cardio}, \frac{1}{2} \text{ Lifting})\}$  is a Nash equilibrium!

- ESR player plays a best response
- SER player plays a best response

## BLENDED SETTING

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*The multi-objective reward vectors.*

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*The ESR utilities.*

### ► What happens when we dismiss SER?

- Athlete has a Nash equilibrium at (Cardio, Cardio) or (Cardio, Lifting)
- Clearly suboptimal

## BLENDED SETTING

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Cardio	17; 4	26; 4
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*The ESR utilities.*

### ► Even stronger

- The best *overall* utility that the athlete can aspire to under ESR
  - (Lifting, Cardio) = 5
- Simply playing the Nash equilibrium under SER
  - $\{(1 \text{ Cardio}, 0 \text{ Lifting}), (\frac{1}{2} \text{ Cardio}, \frac{1}{2} \text{ Lifting})\} = 6.25$