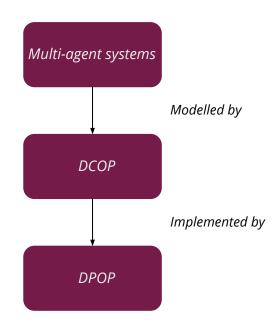
# Distributed Constraint Optimization Problems

DPOP algorithm

Matthaios Zidianakis Vasileios Papadopoulos Dimitris Delikonstantis

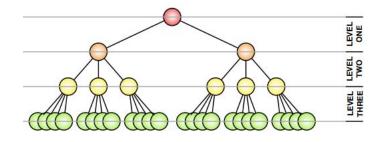
# Distributed Constraint Optimization Problems

- Multi-agent systems
  - Communication
  - Cooperation
  - Compete
- DCOP framework model  $P = (A, X, D, F, \alpha)$ 
  - $\circ$  A = Agents
  - X = Variables
  - O D = Domains
  - F = Cost functions/Utility/Rewards
  - $\circ$   $\alpha$  = mapping function from variables to agents
- DPOP algorithm (X, D, R)
  - Tree-shaped transformation



# Problem Definition

- Meeting scheduling optimization
- <X, D, R> → series of tuples <A, M, U>
- A = agents, M = meetings, U = meeting utility
- Constraints
  - Equality(inter-agent constraint)
  - Inequality(intra-agent constraint)
- Preference
- Utility function
  - Preference<sub>timeslot</sub> \* Meeting<sub>i</sub>\_Util
- Goal: find an assignment that maximizes the total utility, social welfare.

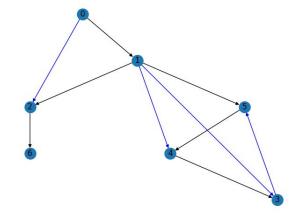


Thanks to **Andreas Oikonomakis**, generator.py

### Constraint Graph to Pseudo-tree

Conventions

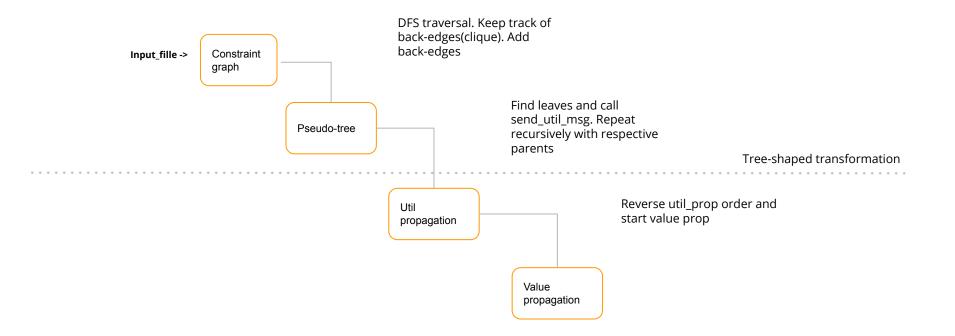
- Nodes represent agents
  - Inequality messages are virtual
  - Privacy
- DFS traversal
  - Start from any node and put to visited list
  - move to adjacent unvisited node and put it to visited
  - continue loop until there is no unvisited node.
  - Backtrack and check for other unvisited nodes
  - Keep track of backedges
- Edges represent equality constraint
  - Constraints within the same agent not shown(inequality)
  - o Back-edges represented by **blue** color



Pseudo-tree output, visualization hiccups

# Impl. DPOP

high overview



# Relations Representation

#### **Equality Constraint**

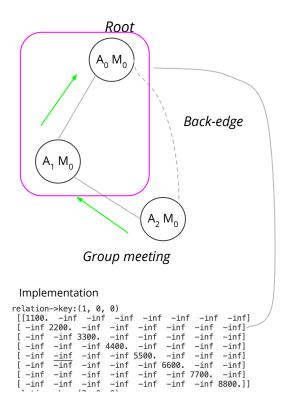
R(A2, A1) Meeting 0				
A1=8 A1=9 A1=10				
A2=8	100	-inf	-inf	
A2=9	-inf	200	-inf	
A2=10	-inf	-inf	500	

R(A2, A0) Meeting 0					
	A0=8 A0=9 A0=10				
A2=8	200	-inf	-inf		
A2=9	-inf	500	-inf		
A2=10	-inf	-inf	100		

Inequality Constraint

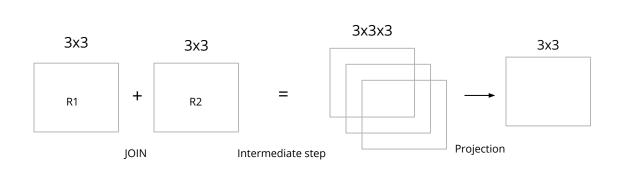
A2\_TS(8) \* UtilM(0,2) + A1\_TS(8) \* UtilM(1,2)

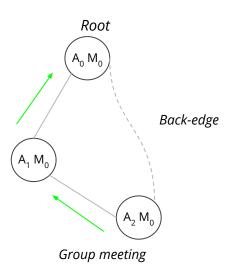
R(m0, m1) Agent 2				
A0=8 A0=9 A0=10				
A2=8	-inf	230	210	
A2=9	230	-inf	100	
A2=10	210	100	-inf	



### Utility Message Overview

2 Relations





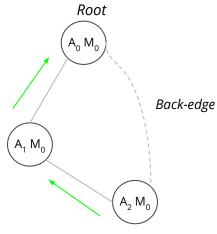
### Utility Propagation 1/3

Current node: A2

R(A2, A1) Meeting 0					
A1=8 A1=9 A1=10					
A2=8	100	-inf	-inf		
A2=9	-inf	200	-inf		
A2=10	-inf	-inf	500		

R(A2, A0) Meeting 0				
A0=8 A0=9 A0=10				
A2=8	200	-inf	-inf	
A2=9	-inf	500	-inf	
A2=10	-inf	-inf	100	

A2 = 8				
A0=8 A0=9 A0=10				
A1=8	300	-inf	-inf	
A1=9	-inf	-inf	-inf	
A1=10	-inf	-inf	-inf	



Group meeting

```
relation->key:(1, 0, 0)
[[1100. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf]
[-inf -inf 3300. -inf -inf -inf -inf]
[-inf -inf -inf 4400. -inf -inf -inf -inf]
[-inf -inf -inf -inf 5500. -inf -inf -inf]
[-inf -inf -inf -inf -inf 6600. -inf -inf]
[-inf -inf -inf -inf -inf -inf 7700. -inf]
[-inf -inf -inf -inf -inf -inf -inf 8800.]]
```

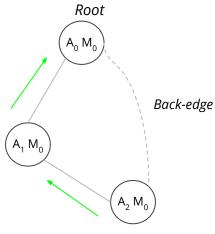
### Utility Propagation 2/3

Current node: A2

R(A2, A1) Meeting 0				
A1=8 A1=9 A1=10				
A2=8	100	-inf	-inf	
A2=9	-inf	200	-inf	
A2=10	-inf	-inf	500	

R(A2, A0) Meeting 0				
A0=8 A0=9 A0=10				
A2=8	200	-inf	-inf	
A2=9	-inf	500	-inf	
A2=10	-inf	-inf	100	

A2 = 9			
	A0=8	A0=9	A0=10
A1=8	-inf	-inf	-inf
A1=9	-inf	700	-inf
A1=10	-inf	-inf	-inf



Group meeting

```
relation->key:(1, 0, 0)
[[1100. -inf -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf -inf]
[-inf -inf 3300. -inf -inf -inf -inf -inf]
[-inf -inf -inf -inf 4400. -inf -inf -inf -inf]
[-inf -inf -inf -inf -inf 5500. -inf -inf -inf]
[-inf -inf -inf -inf -inf 6600. -inf -inf]
[-inf -inf -inf -inf -inf -inf 7700. -inf]
[-inf -inf -inf -inf -inf -inf -inf 8800.]]
```

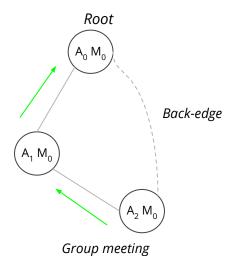
### Utility Propagation 3/3

Current node: A2

R(A2, A1) Meeting 0					
A1=8 A1=9 A1=10					
A2=8	100	-inf	-inf		
A2=9	-inf	200	-inf		
A2=10	-inf	-inf	500		

R(A2, A0) Meeting 0					
A0=8 A0=9 A0=10					
A2=8	200	-inf	-inf		
A2=9	-inf	500	-inf		
A2=10	-inf	-inf	100		

A2 = 10				
A0=8 A0=9 A0=10				
A1=8	-inf	-inf	-inf	
A1=9	-inf	-inf	-inf	
A1=10	-inf	-inf	600	



```
relation->key:(1, 0, 0)
[[1100. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf -inf]
[-inf -inf 3300. -inf -inf -inf -inf -inf]
[-inf -inf -inf 4400. -inf -inf -inf -inf]
[-inf -inf -inf -inf 5500. -inf -inf -inf]
[-inf -inf -inf -inf -inf 6600. -inf -inf]
[-inf -inf -inf -inf -inf -inf 7700. -inf]
[-inf -inf -inf -inf -inf -inf -inf 8800.]]
```

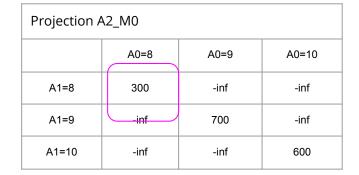
### **Utility Propagation**

Current node: A2 Projection

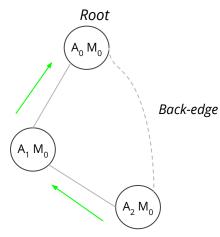
A2 = 8						
	A0=8	A0=8 A0=9				
A1=8	300		-inf	-inf		
A1=9	-inf	-inf		-inf		
A1=1 0	-inf		-inf	-inf		

A2 = 9								
	A0=8 A0=9 A0=10							
A1 = 8	-inf	-inf	-inf					
A1=9	-inf	700	-inf					
A1=10	-inf	-inf	-inf					

A2 = 10								
	A0=8 A0=9 A0=10							
A1=8	-inf	-inf	-inf					
A1=9	-inf	-inf	-inf					
A1=10	-inf	-inf	600					



Send message to A1 from A2



Group meeting

```
relation->key:(1, 0, 0)
[[1100. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf]
[-inf -inf 3300. -inf -inf -inf -inf -inf]
[-inf -inf -inf 4400. -inf -inf -inf -inf]
[-inf -inf -inf -inf 5500. -inf -inf -inf]
[-inf -inf -inf -inf -inf 6600. -inf -inf]
[-inf -inf -inf -inf -inf -inf 7700. -inf]
[-inf -inf -inf -inf -inf -inf -inf 8800.]]
```

### Utility Propagation

Current node: A1

Projection A2_M0						
	A0=8 A0=9 A0=					
A1=8	300	-inf	-inf			
A1=9	-inf	700	-inf			
A1=10	-inf	-inf	600			

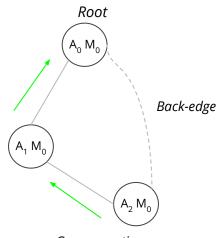
R(A1, A0) Meeting 0								
	A0=8 A0=9 A0=10							
A1=8	200	-inf	-inf					
A1=9	-inf	500	-inf					
A=10	-inf	-inf	100					

Received msg

Projection A1_M0					
A0=8	A0=8 A0=9 A0=10				
500	1200	700			

Relation

Send message to A0 from A1



Group meeting

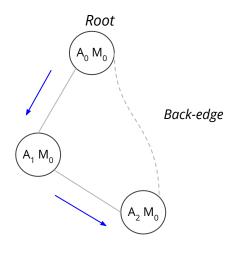
```
relation->key:(1, 0, 0)
[[1100. -inf -inf -inf -inf -inf -inf]
[-inf 2200. -inf -inf -inf -inf -inf -inf]
[-inf -inf 3300. -inf -inf -inf -inf -inf]
[-inf -inf -inf 4400. -inf -inf -inf -inf]
[-inf -inf -inf -inf -inf 6600. -inf -inf]
[-inf -inf -inf -inf -inf -inf 7700. -inf]
[-inf -inf -inf -inf -inf -inf -inf 8800.]
```

### Value Propagation

Current node: AO

Projection A1_M0					
A0=8	A0=9	A0=10			
500	0 1200 700				

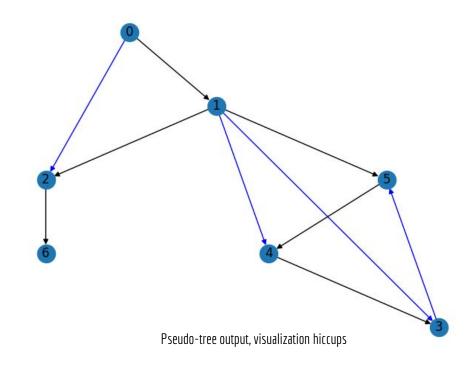
Chooses max and initiates value prop



Group meeting

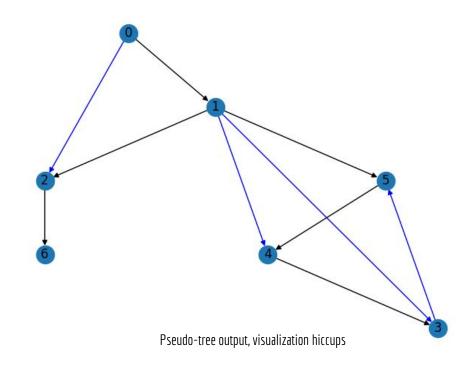
# Utility Order - Evaluation Impl.

```
Leaves are: [6, 3]
Util Message from: 6 to 2
Util Message from: 3 to 4
Util Message from: 4 to 5
Util Message from: 5 to 1
Util Message from: 2 to 1
Util Message from: 1 to 0
root node: {}
```



# Value Order - Evaluation Impl.

```
Value Message from: 0 to 1
Value Message from: 1 to 2
Value Message from: 1 to 5
Value Message from: 5 to 4
Value Message from: 4 to 3
Value Message from: 2 to 6
```



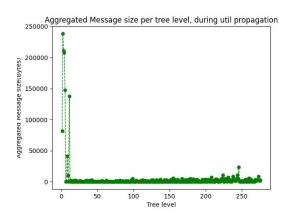
### DPOP Evaluation - During Util Prop

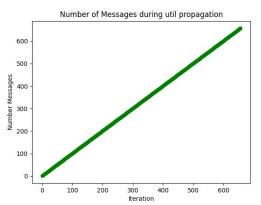
Agents: 700 Meetings: 350

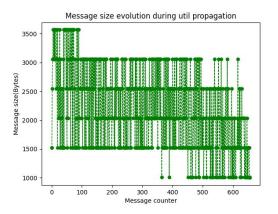
#### Execution

- Aggregated message size per cycle.
- Message size evolution.
- Rate of change of number of messages.

Inequality const calculation per agent
m in {m1,m2,m3,m4,m5}
m1 != (m2, m3, m4, m5) +
m2 != (m3, m4, m5) +
m3 != (m4, m5) +
m4 != (m5)







#### output

# Results 1/2

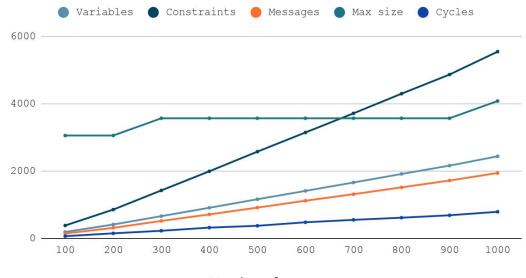
Problems										
Agents	100	200	300	400	500	600	700	800	900	1000
Meetings	50	100	150	200	250	300	350	400	450	500
Variables	193	414	662	911	1164	1412	1661	1914	2162	2439
Constraints	385	857	1426	1994	2577	3146	3714	4297	4866	5546
Messages	146	316	520	714	916	1120	1314	1516	1720	1942
Max message size	3056	3056	3568	3568	3568	3568	3568	3568	3568	4080
Cycles	68	152	228	322	376	480	552	616	688	792

# Results 2/2

#### **Evaluation**

- As expected all metrics has a linear relationship with size of the problem(agents,meetings).
- Max message size remains constant which is **incorrect** as our implementation lacks of proper hypercube calculation.
- Maybe use of pandas.dataframe object

Evolution of DPOP with increasing number of agents/meetings



Number of agents

# References

- 1. Andrian Petcu, Boi Faltings, DPOP: A Scalable Method for Multiagent Constraint Optimization, IJCAI-05, Proceedings of the Nineteenth International Joint Conference on Artificial Intelligence, Edinburgh, Scotland, UK, July 30-August 5, 2005
- 2. Ferdinando Fioretto, Enrico Pontelli, William Yeoh, Distributed Constraint Optimization Problems and Applications: A Survey, Journal of Artificial Intelligence Research 61 (2018) 623-698
- 3. Rajiv T. Maheswaran, Milind Tambe, Emma Bowring, Jonathan P. Pearce, and Pradeep Varakantham, Taking DCOP to the Real World: Efficient Complete Solutions for Distributed Multi-Event Scheduling, 2004
- 4. Andrian Petcu, A class of algorithms for distributed constraint optimization, Frontiers in Artificial Intelligence and Applications, 2009