

Cooperative Multi-agent Approach for Automated Computer Game Testing

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<https://iv4xr-project.eu/>



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Automated Testing VS Manual Testing



Manual testing

Not reusable

Time and cost efficiency

- <https://www.capgemini.com/research/world-quality-report-2019/>

Modern Computer Games

- Rich and complex environments
- Huge interaction space
- Multi Player
- Testing demands high human labor (User testing)



Challenges in Automated Games Testing

- The traditional ones are unable to handle the interaction space of 3D based Games
- Continues Space & Geometric Reasoning

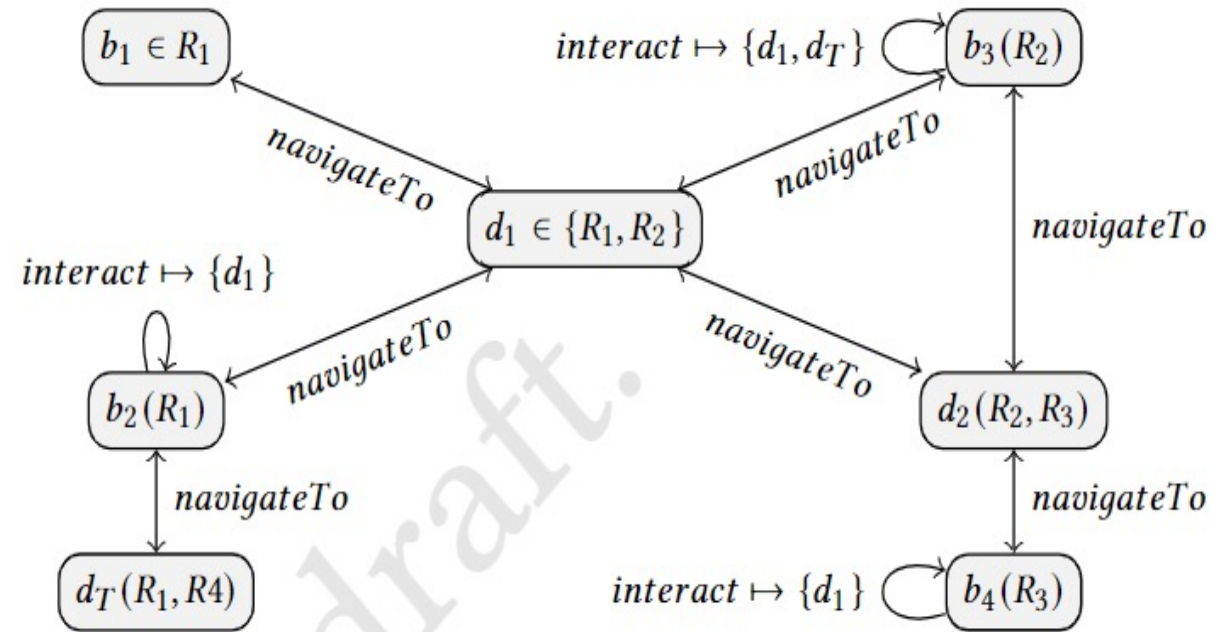


Agent-based automated game-testing (single agent)

Algorithm 1 Online Search

```
1: procedure ONLINESEARCH( $\phi_o \Rightarrow \psi, M, Nav$ )
2:   while  $\neg\phi_0$  do
3:     parallel
4:       || update  $M_{stategraph}$ 
5:       || if new states observed then
6:          $o' \leftarrow selectNode()$ 
7:         mark( $o'$ ) ▷ mark it as visited
8:         navigateTo( $o'$ ) using  $Nav$ 
9:         if  $o' = o$  &  $\neg\phi_o$  then
10:          dynamicGoal( $o', \phi$ )
11:        else if  $o'$  is a blocker &  $o.isblocked$  then
12:          dynamicGoal( $o', o'. \neg o'.isBlocked$ )
13:        else if there is terrain unexplored then explore()
14:        else abort()
15:        end if
16:      end parallel
17:   assert  $\psi$ 
```

<https://arxiv.org/pdf/2211.06936.pdf>



An Online Agent-Based Search Approach in Automated Computer Game Testing with Model Construction,
Shirzadehhajimahmood et al, ATEST 2022.

How about multi-agent game testing?

- **Motivation:** In the previous work, long test scenarios can take lots of time to run (e.g. 30+ minutes each). Many games are multi-player.
- **Hypothesis:** Deploying multiple agents can speed up testing (e.g. 0.5 reduction would be great!)
- **Challenges:** The agents can get in each others way, or even mess up each other plans. The computation overhead of just greedily sharing (and synchronizing) information may be too much.
- **Idea:** Collaboration between agents



Cooperative Multi-agent Approach

✓ **iv4XR framework**

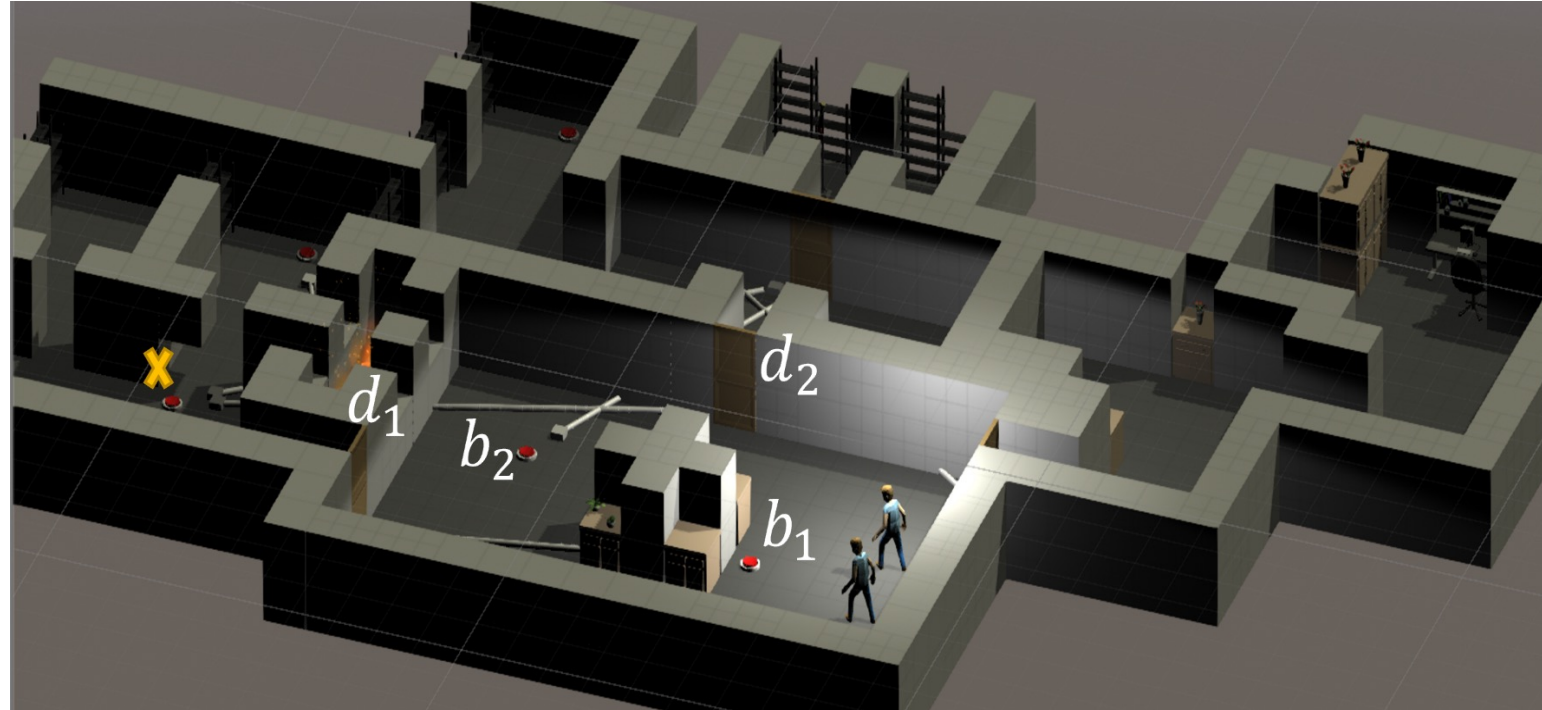
- Goal structure
- Tactic

Agent-based Approach

- Autonomy and reactivity
- Goal-based behavior
- Sensing and Decision Making

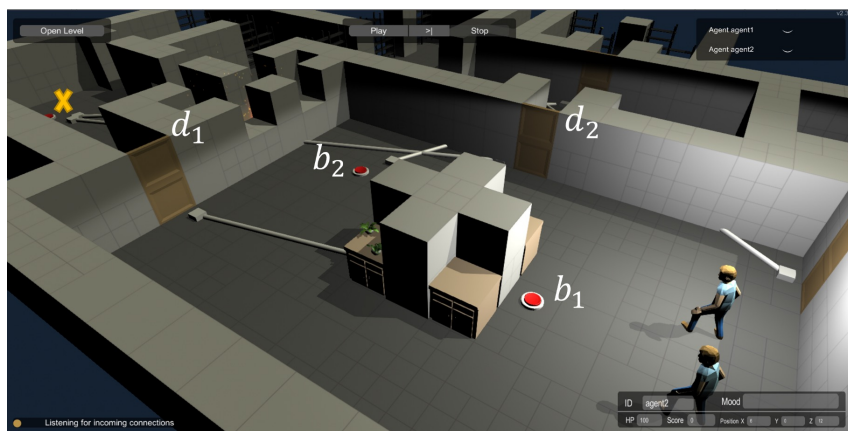
Multi-Agent Algorithm features

- Multiple test agents
- Working in parallel
- Dynamically choosing tasks
- Cooperative agents
- Dealing with obstacles





Multi Agent Testing Algorithm



Algorithm 1 *It gets a set of tasks \mathcal{T} and run N test agents.*

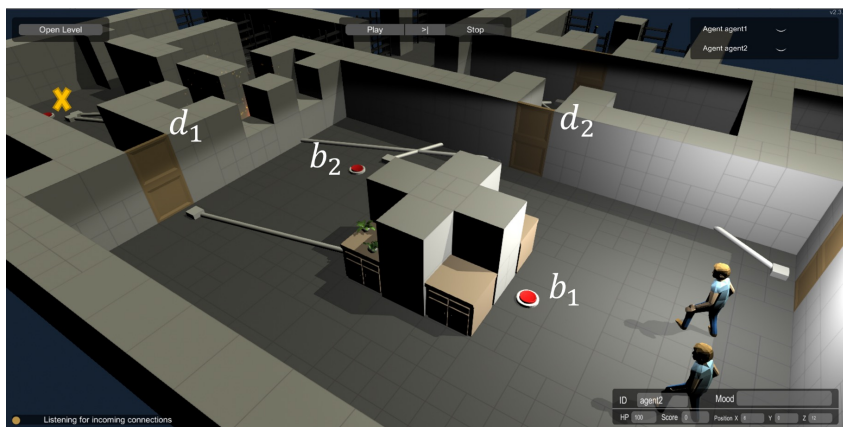
```
1: procedure COOPERATIVEAGENTS( $\mathcal{T}$ )
2:    $toDo = \emptyset$ 
3:    $done = \emptyset$ 
4:    $agent_1.SOLVER(\mathbf{H}_1) \parallel \dots \parallel agent_N.SOLVER(\mathbf{H}_N)$ 
5:    $\parallel SYNC()$ 
6:    $\parallel done = \mathcal{T} \vee budget \leq 0 \rightarrow \text{terminate} \triangleright$  terminate the whole procedure
```

Algorithm 2 *For selecting and executing tasks, parameterized by two heuristics.*

```
1: procedure SOLVER( $selectH, findH$ )
2:   while  $budget > 0$  do
3:     if  $toDo \neq \emptyset$  then
4:        $T \leftarrow selectH(toDo)$   $\triangleright$  use the task-selection heuristic
5:       if  $T \neq null$  then
6:          $(o, \psi, S) \leftarrow T$ 
7:          $toDo \leftarrow toDo / \{T\}$ 
8:         NAVIGATETo( $o$ )
9:         if  $\neg \psi$  then DYNAMICGOAL( $o, \psi, S, findH$ )
10:        if  $\psi$  then  $done \leftarrow done \cup \{T\}$ 
11:        else  $toDo \leftarrow toDo \cup \{T\}$ 
12:      else
13:        if there is terrain unexplored then FINDTASK( )
14:        else return
```



Solving a chosen Task



Algorithm 3 For solving a single task, parameterized by one heuristic.

```
1: procedure DYNAMICGOAL( $o, \psi, S, findH$ )
2:   while  $\neg\psi \wedge \neg S$  do
3:      $\Delta \leftarrow \{i \mid i \in seenObj \wedge \neg mark_o(i) \wedge \neg locked(i)\}$ 
4:     choose  $i \in \Delta$ , based on  $findH$   $\triangleright$  use the object-selection heuristic
5:     if  $i = null$  then
6:       if there is terrain unexplored then
7:         EXPLORE( )  $\triangleright$  explore world to find new objects
8:       else
9:         return
10:    else
11:       $mark_o(i) \leftarrow true$   $\triangleright$  mark  $i$  as tried for  $o$ 
12:      LOCK( $i$ )
13:      navigateTo( $i$ )
14:      applyAction( $i$ )  $\triangleright$  such as interact
15:      navigateTo( $o$ )
16:      UNLOCK( $i$ )
```

Algorithm 4 For finding new tasks.

```
1: procedure FINDTASK( )
2:   while there is terrain unexplored do
3:     BASICEXPLORE( )
4:      $V \leftarrow$  newly observed objects
5:      $W \leftarrow \{(o, \psi, S) \mid o \in V \wedge (o, \psi, S) \in \mathcal{T} / (toDo \cup done)\}$ 
6:     if  $W \neq \emptyset$  then
7:        $toDo \leftarrow toDo \cup W$ 
8:     return
```



Research questions

- ❖ RQ1: *does multi-agent speed up testing?*
- ❖ RQ2: *how well can multi-agent deal with complex logic?*



Information Synchronization Levels

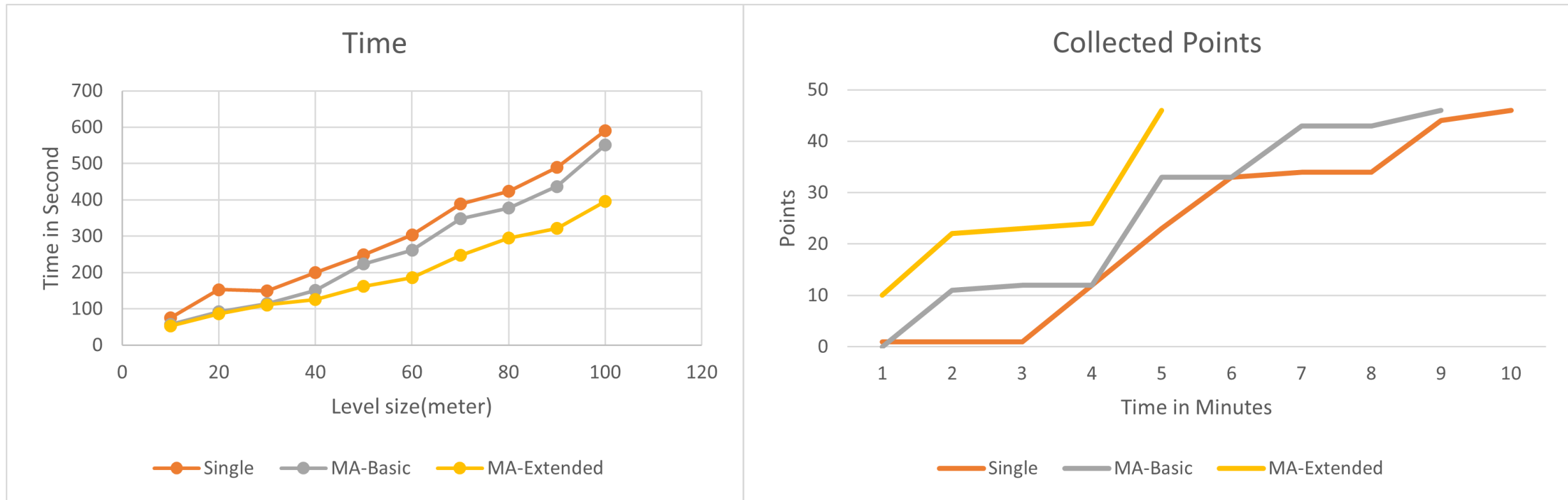
Basic: sharing seen tasks and solved tasks

Advanced: Basic + sharing explored areas to each other

Factors Affecting the Performance

- Size of the game levels
- Task distribution among agents

RQ1 and influence of information synchronization



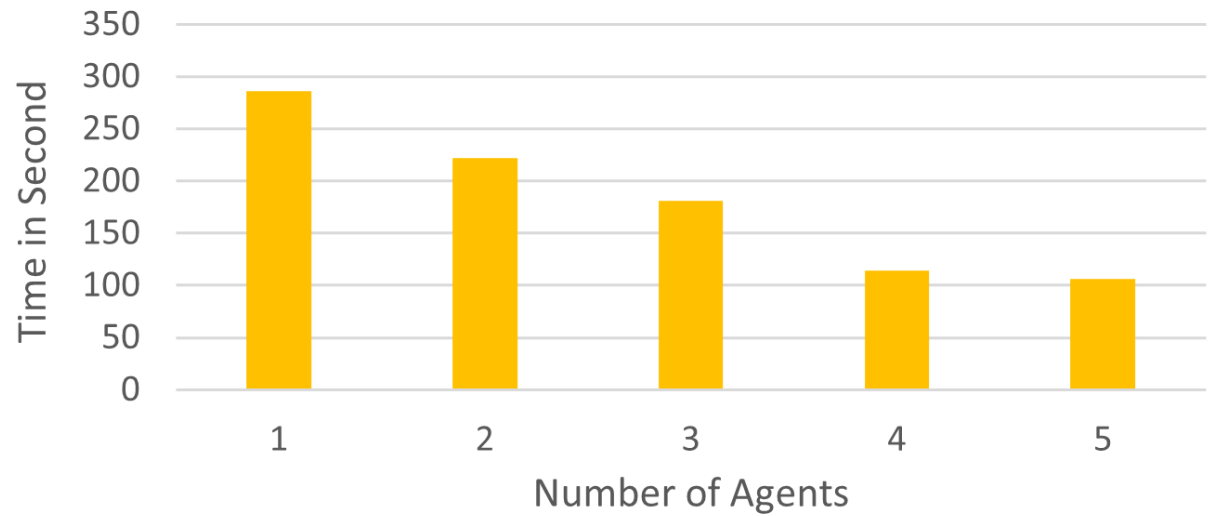
Points are gained when a goal is achieved

RQ1 and influence of team composition

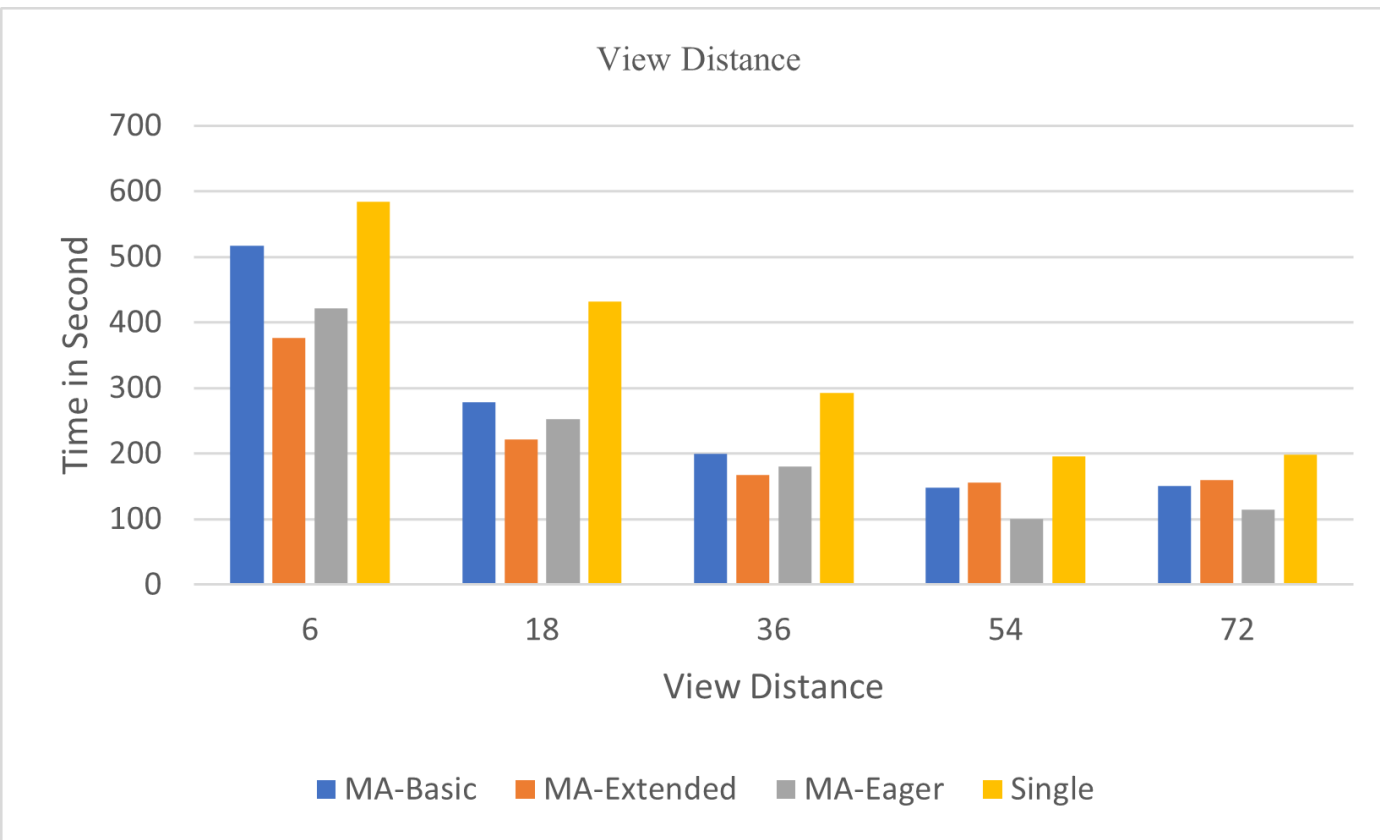
3-Agents Setup with Different Team Composition



Eager Agents



RQ2: Distant-connection logic & hidden button logic



Distant-Connection

#DC	Single	MA-Basic	MA-Extended
2	210	159	117
4	265	209	195
6	315	289	215
8	352	333	232
10	180	352	264

Chained Connections

#HB	Single	MA-Eager	MA-Extended
0	149	128	111
1	214	145	140
2	307	254	191
3	337	308	241

Conclusion & Future Work

- ✓ A cooperative multi-agent testing approach
- Evaluated performance between a single-agent and multi-agent setups
 - Applying different heuristic to select tasks
 - Information sharing and synchronization
- Multi agent with extended information sharing performs better (despite the overhead).





Question?



iv4XR agent programming: tactic & goal Structure

A **tactic** is a way to hierarchically combine actions (or sub-tactics) to achieve a goal G :

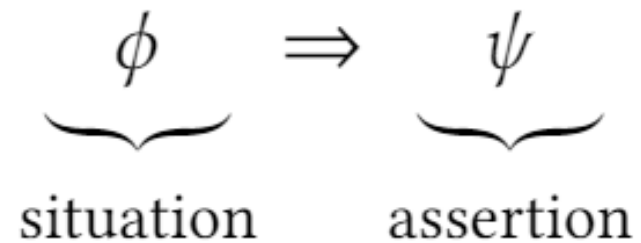
$$T = \text{FIRSTof}(T_1, \dots, T_n)$$

When executed, T runs in a loop over **multiple** deliberation cycles, until G is achieved (or aborted).

A **goal structure** is a set of goals that have to be achieved in a certain order:

$$H = \text{SEQ}(G_1, \dots, G_n)$$

Testing Task



var testingTask = **SEQ**(

isAt("button1"),

isInteracted("button1"),

isAt("button3"),

isInteracted("button3"),

.....

isOpen("treasureDoor")

Sub-goals to
solve/achieve

} **assertion**

Domain Specific Language of iv4XR Framework

goal structure	::=	SEQ (goal structure, goal structure,...) FIRSTof (goal structure, goal structure,...) WHILEDO (<i>predicate</i> , <i>goal structure</i>) goal.lift()
goal	::=	goal (name). toSolve (predicate). withTactic (tactic)
tactic	::=	SEQ (tactic,tactic,...) FIRSTof (tactic, tactic,...) ANYof (tactic,tactic,...) ABORT () action.lift()
action	::=	action (name). do (action expression). on (predicate) action(<i>name</i>). addAfter (<i>goalstructure</i>)