

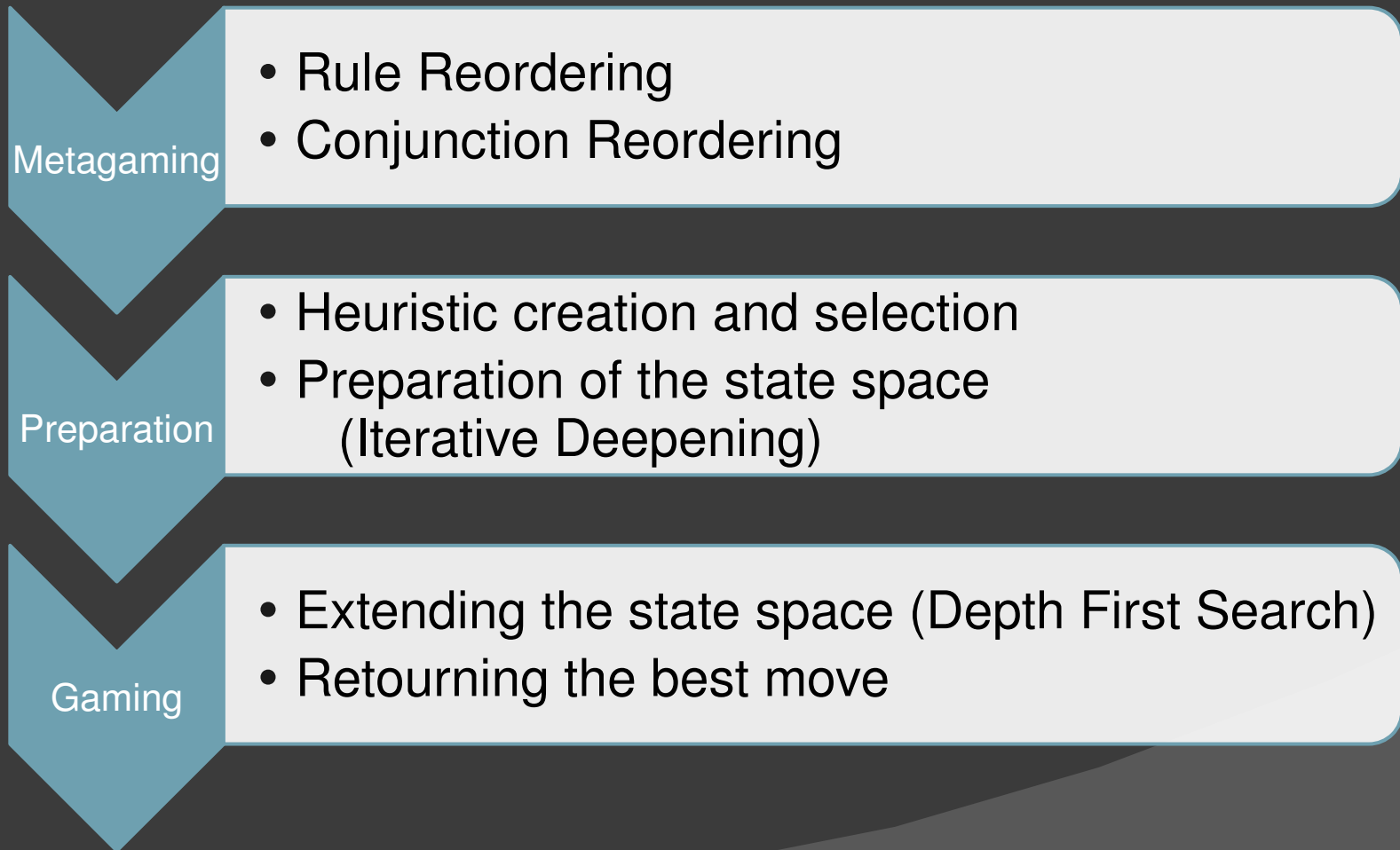
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TEAM 2MC1 – MORE THAN A PLAYER

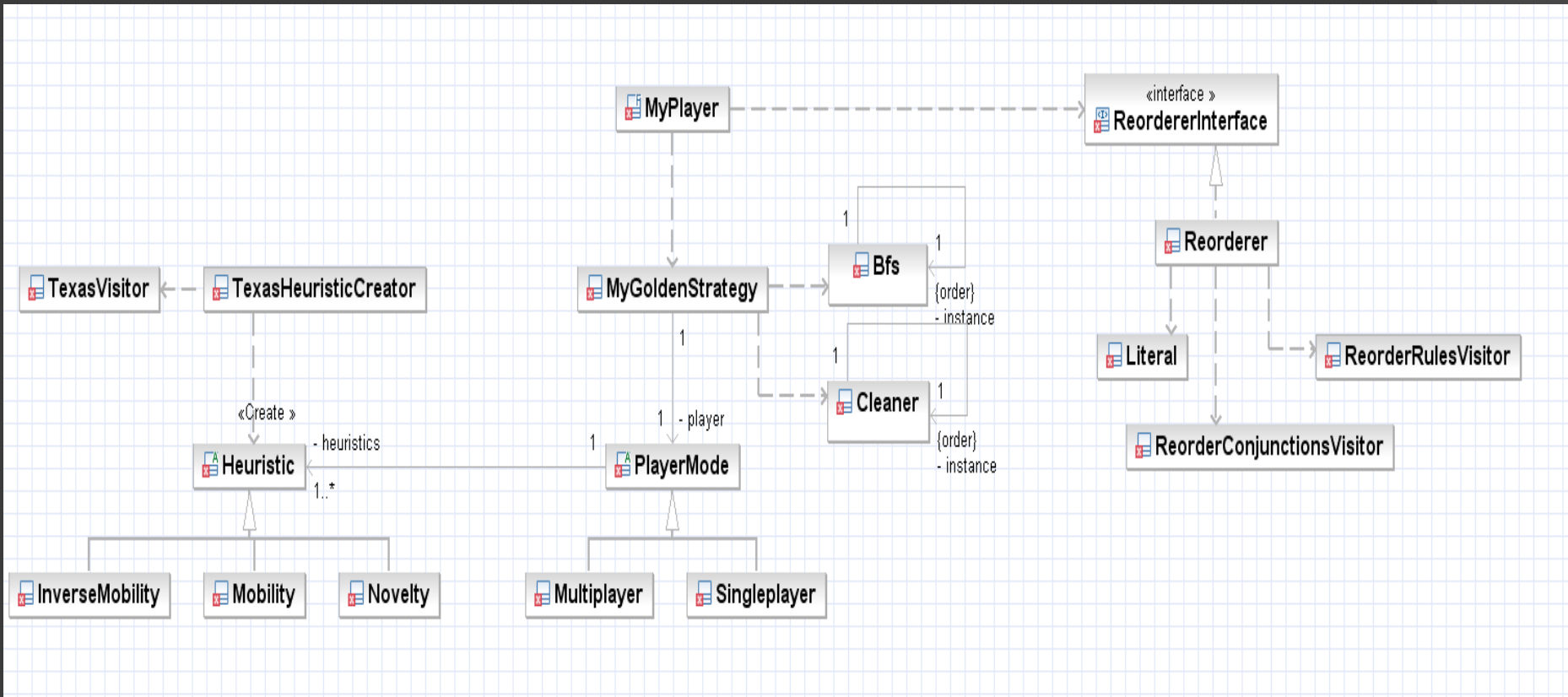
Overview

1. Workflow during a game
2. Structure of our player
3. Selected aspects of the implementation
4. Conclusion

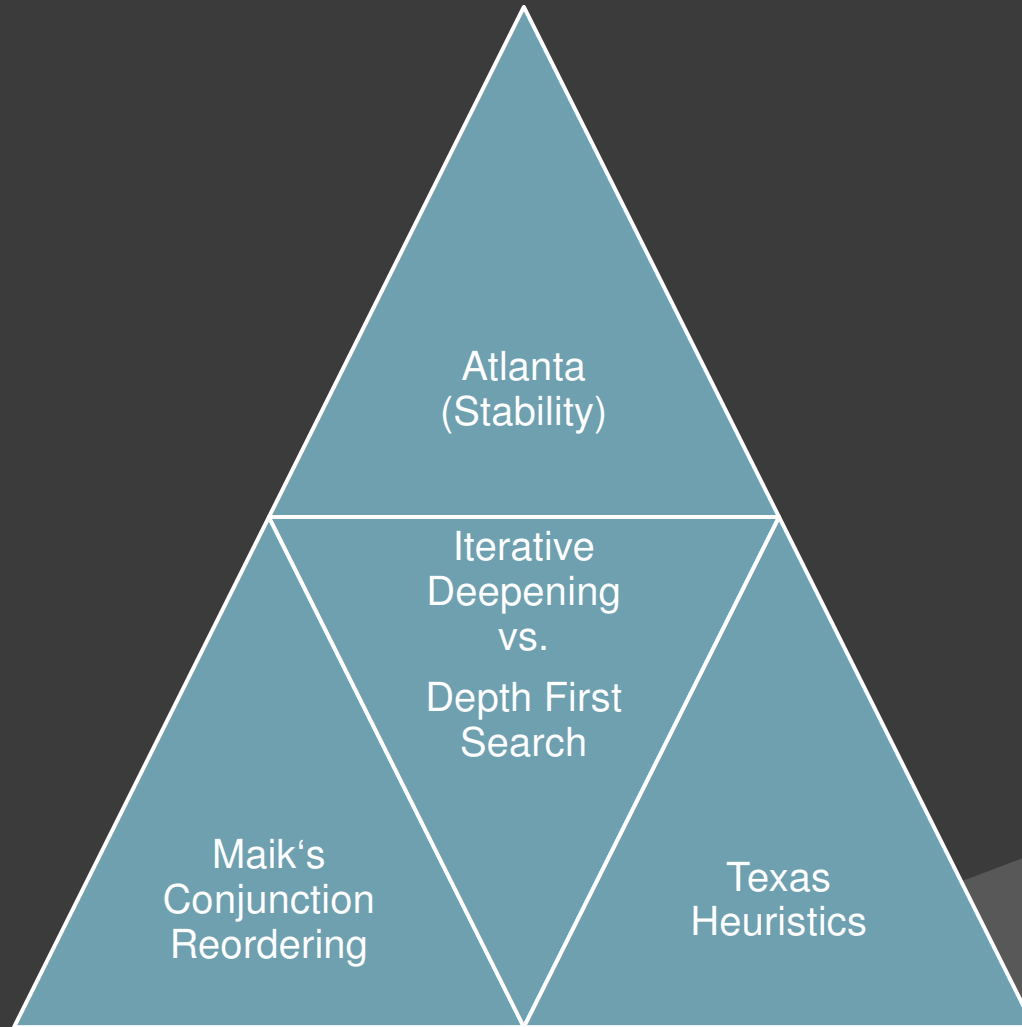
1. Workflow during a game



2. Structure of our player



3. Selected aspects of the implementation



3. Selected aspects of the implementation

- t Maik's Conjunction Reordering :
 - Algorithm by Watson Research Center NY
 - Reorder based on: dependency, grounded variables
 - care for variable safety

3. Selected aspects of the implementation

- t Texas Heuristics (based on the Player of the University of Texas, July 2006) :
- Identify Successor & Step Counter
 - Identify the (ordered) Board, Markers and Pieces
 - Create up to 1000 heuristics



- t Select the best heuristic by a method, which is similar to the Monte Carlo Tree Search :
- Follow the “golden way” of each heuristic and determine the reached goal score
 - The best 3 heuristics are used for the final mixed heuristic with the weights 66%, 30% and 4%

TexasHeuristicCreator

- realSuccessors
- stepCountersWithSuccessors
- boards
- amounts

TexasHeuristicCreator()
getMaxStepCounter()
getHeuristics()
getAmounts()
getBoards()
getRandomState()
getRandomStates()
getStructuresWithFixLength()
getStepCountersWithSuccessors()
getRealSuccessorRelationNames()
getNodeParts()
getNodesByPattern()

DetailedBoard

TexasSuperHeuristic

TexasMarkerCount

TexasInverse

TexasSum

TexasPiecePosition

TexasNewYorkDistance

«enumeration»

ImportanceIndicator

LESS
HIGH

3. Selected aspects of the implementation

Atlanta (Stability) : Return always a valid response

Redo Reordering	Arbitrary moves	Memory Observation
if the occurs problems with the new Rule Set	if we don't know the best move for the current state	Cleaner thread started, if the memory exceeds 70%

3. Selected aspects of the implementation

- Iterative Deepening vs. Depth First Search:

	Iterative Deepening	Depth First Search
Used in our player during preparation period	... during gaming period
Our goal	Find the nearest solution	Find possible goal states
Why?	Good for SP-Games and fast victories	Good for MP-Games, that need a certain field of vision
Extensions	Use of Alpha-Beta Pruning (if it's a 2 player, zero sum & turn taking game)	

4. Conclusion

- GGP : Great possibility to learn more about the realization of “thinking machines”
- Possible Improvements :
 - Goal Distance Heuristic
 - Better metagaming
 - Optimization for multi-core processors