

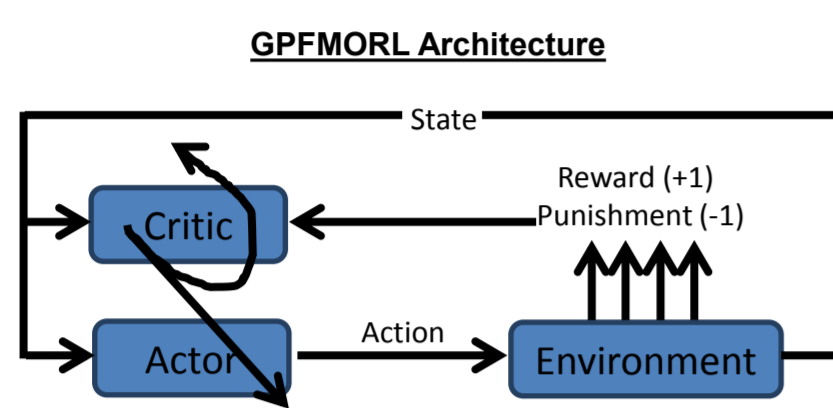
Artificially Intelligent Unmanned Aerial Vehicle

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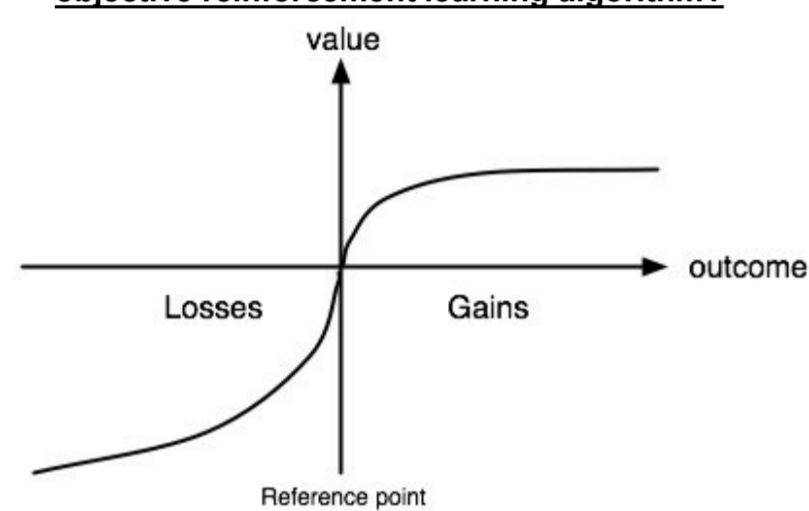
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Introduction

Problem



Can Prospect Theory be used to improve a multi-objective reinforcement learning algorithm?



This research focuses on the problem of uncertainty handling during learning by multiple agents dealing in stochastic environments by means of Multi-Objective Reinforcement Learning (MORL). Reinforcement learning is a powerful mechanism for enabling agents to learn in an unknown environment, and most reinforcement learning algorithms aim to maximize some numerical value, which represents only one long-term objective. However, multiple long-term objectives are exhibited in many real-world decision and control problems; therefore, recently, there has been growing interest in solving multi-objective reinforcement learning (MORL) problems with multiple conflicting objectives.

Simulation Environment

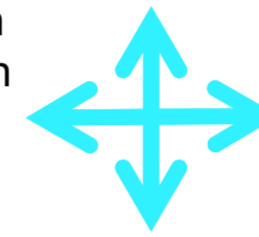


10x10 Windy Hills Grid World Simulation

1	2	3	4	5	6	7	8	9	10
11	12	13	14	15	16	17	18	19	20
21	22	23	24	25	26	27	28	29	30
31	32	33	34	35	36	37	38	39	40
41	42	43	44	45	46	47	48	49	50
51	52	53	54	55	56	57	58	59	60
61	62	63	64	65	66	67	68	69	70
71	72	73	74	75	76	77	78	79	80
81	82	83	84	85	86	87	88	89	90
91	92	93	94	95	96	97	98	99	100

The Reinforcement Learning method consists of 4 agents each responsible for moving in one of the four cardinal directions

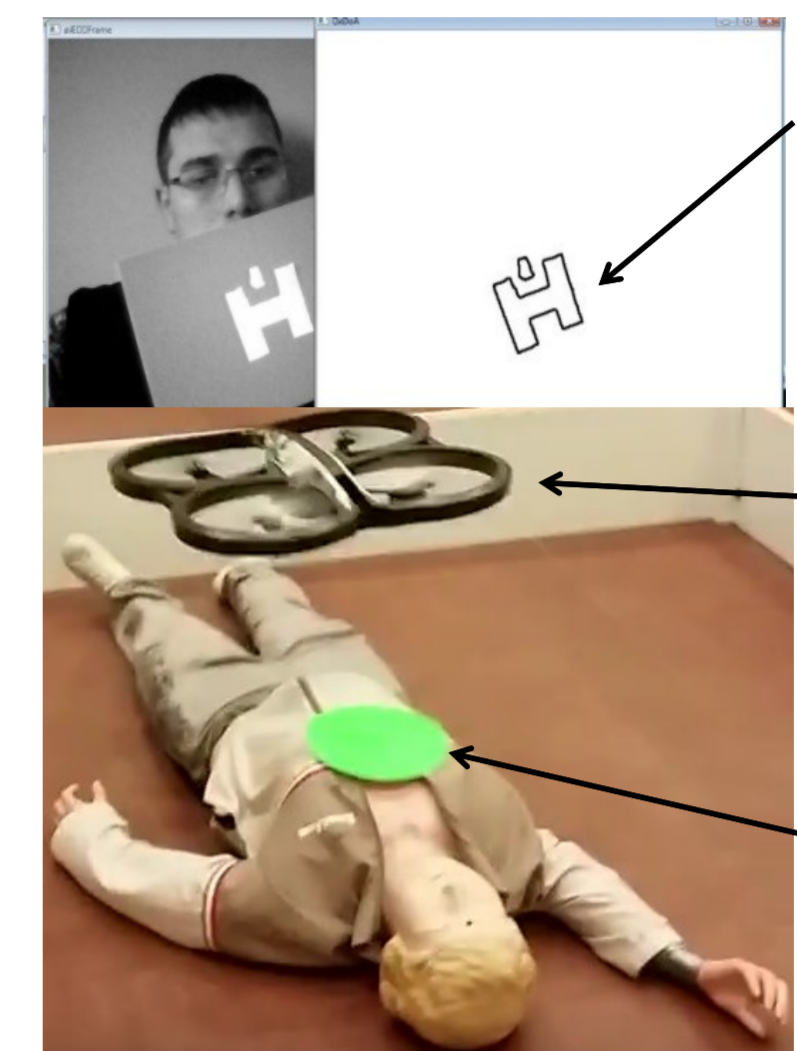
- North
- South
- East
- West



The simulation environment is discretised into a 10x10 grid world where the aim is to learn the most efficient way to reach the goal (state46) from the start (state12) whilst satisfying the following 3 objectives:

1. Reach Helipad
2. Avoid Boundaries
3. Avoid/Traverse Hills

Real World UAV Implementation

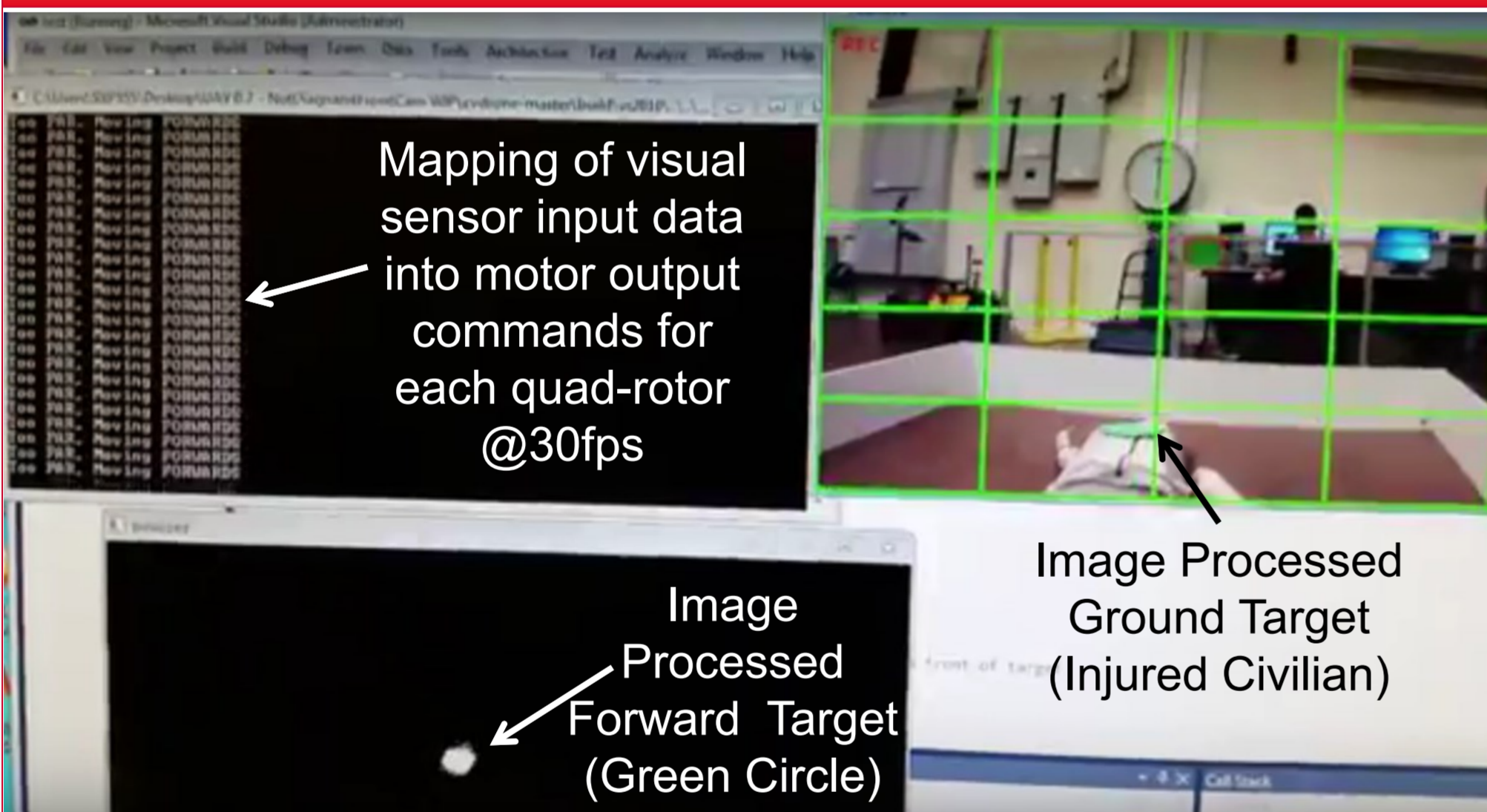


Helipad marker segmented from the background is used for determining altitude and Localisation

AR-Drone™ Wi-Fi Quad copter equipped with front and downward facing cameras, ultrasonic sensor, 6axis gyroscopes and accelerometers

System currently tracks a green circle to determine its current position however it could easily track peoples faces and use GPS in outdoor environments

UAV Search & Rescue Procedure



1. UAV takes off and detects the green circle target near the human.
2. UAV moves left, right, up or down to maintain a central position.
3. UAV begins to move forward until close enough to begin MORL.
4. UAV increases altitude to perceive a better field of view of the "H".
5. UAV tries actions when in different states to find optimal policy.
6. UAV now knows the best route to take from most states and lands.
7. UAV reports the GPS position of the human and brings crucial medical supplies

Methodologies

GPFMORL(PT)

Generalised Probabilistic Fuzzy Multi-Objective Reinforcement Learning is a hybrid learning algorithm that combines both the advantages of probabilistic theory with a fuzzy inference engine to handle multiple types of uncertainties. The single objective algorithm (GPFRL) has been improved to deal with multiple objectives by developing a Prospect Theory based scalarisation technique for the following three objectives in the proposed "Windy Hill Grid World" benchmark

1. Reach the goal (helipad) in as few steps as possible
2. Never become lost by stepping onto the boundary as few times as possible
3. Avoid the Hill obstacles unless its more efficient to traverse small hills

Multi-Modal Fuzzy Altitude Controller

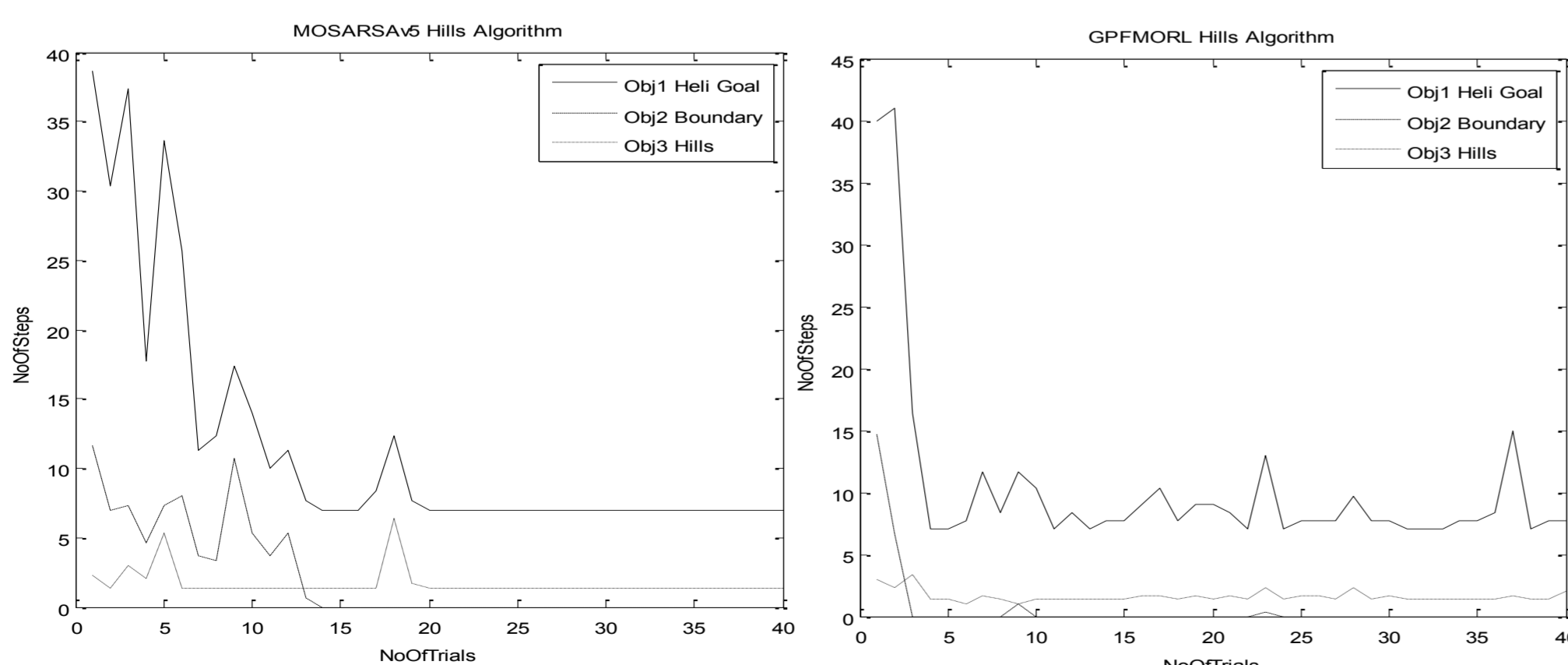
Using an Ultra-Sonic proximity sensor in conjunction with non linear visual area measurements of the helipad, a multi modal fuzzy controller is developed that enables smooth control of the drones altitude to ensure a consistent hovering height is always maintained, even in the event of traversing hills when necessary

Potential Field Heuristics (Reactive Behaviours)

Due to the inherent field of view limitations when using a non-specialised lens for achieving localisation, further methods of control were used to ensure the helipad was always visible. Repulsive potential fields were added to the outer boundary states of the perceived environment to ensure the drone never become lost. Attempts to use wide angle lens also improved performance slightly.

Algorithm Comparisons for 10x10 Grid World Deterministic Complex Environment

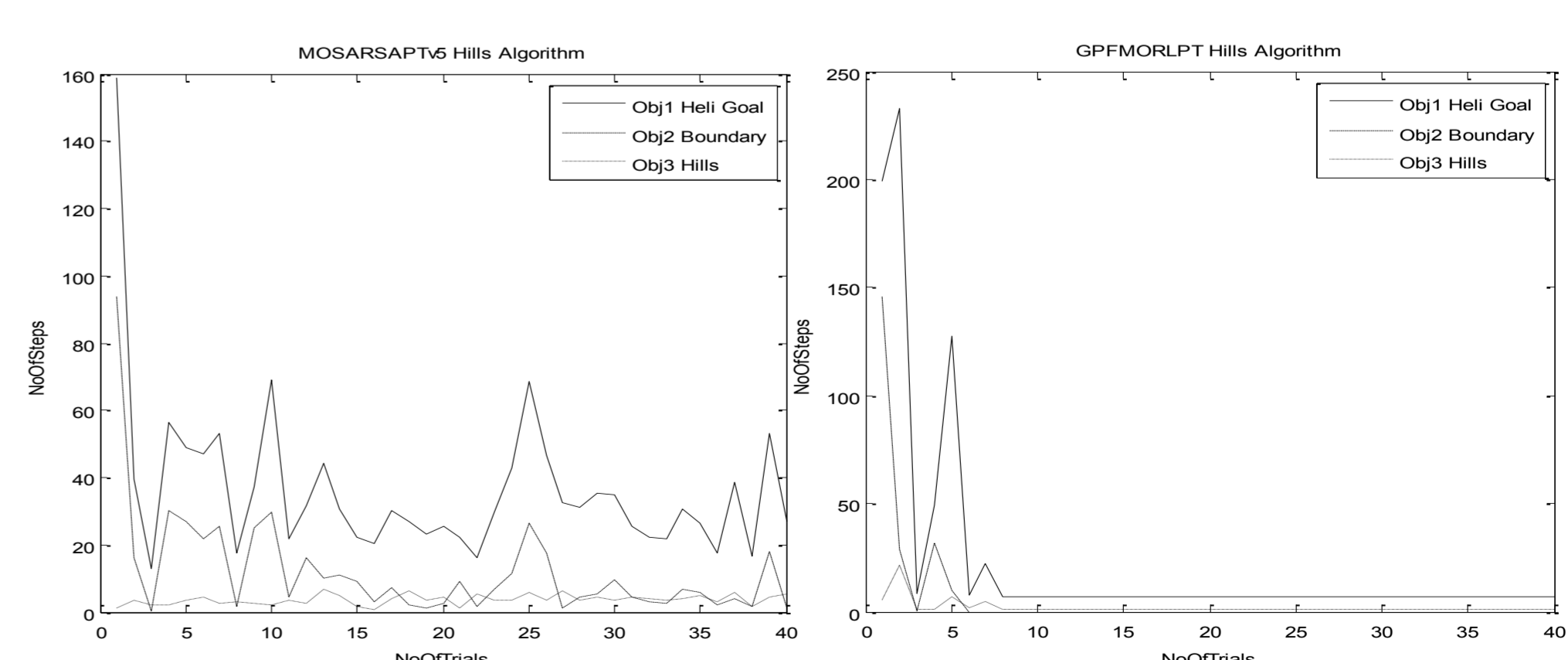
Multi Objective Scalarisation Approach



Av Bounds Count = 236
Av Hill Count = 58.6667

Av Bounds Count = 22.6667
Av Hill Count = 63

Multi Objective Prospect Theory Improvement



Av Bounds Count = 481.6667
Av Hill Count = 149.3333

Av Bounds Count = 217
Av Hill Count = 86.6667