

Flexible Attention-based Cognitive Architecture for Robots

Rony Novianto

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CERTIFICATE OF AUTHORSHIP/ORIGINALITY

I certify that the work in this thesis has not previously been submitted for a degree nor has it been submitted as a part of requirements for a degree except as fully acknowledged within the text.

I also certify that the thesis has been written by me. Any help that I have received in my research work and the preparation of the thesis itself has been acknowledged. In addition, I certify that all information sources and literature used are indicated in the thesis.

Rony Novianto

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Abstract

Robots have been working in factories to achieve tasks autonomously with little human intervention for some time. Even though robots are commonly found as vacuum cleaners in homes and assistants in hospitals, by comparison with factory robots, service robots have not been widely deployed in society because there remains several challenges deploying robots to achieve complex tasks in open, unstructured, uncontrolled and complex environments.

Critical research gaps arise from the lack of cognitive architectures that support robots to undertake tasks in open and complex environments. Throughout the history of AI, researchers have developed various algorithms, representations and mechanisms, to solve specific tasks. However, each of these techniques has different strengths and weaknesses when applied to particular problems. A cognitive architecture provides a unifying infrastructure that can integrate various techniques to solve open and complex tasks.

However, four important issues become apparent when current cognitive architectures are applied to service robotic tasks. First, they are not capable of managing robot resources and as a result robotic developers must take responsibility for managing the resources manually. Second, they are not capable of integrating independently developed techniques, which are often needed to solve problems. Third, they are inflexible, unable to adapt to design changes and require considerable time and effort to modify. Fourth, they are inadequate for supporting the necessary capabilities required by robots such as multiple goals, reliability and maintainability. These issues are confirmed when cognitive architectures are applied to a standard benchmark problem in AI: the autonomous robot soccer problem.

The purpose of this dissertation is to address these significant gaps so as to accelerate the development, deployment and adoption of service robots undertaking tasks in open and complex environments. This dissertation develops a novel bio-

inspired cognitive architecture (called ASMO) that has been designed and developed to address all four identified shortcomings of current cognitive architectures.

In ASMO, intelligent behaviours to solve open and complex tasks is a result of the emergence of constituent processes, rather than from careful top-down control engineering. Minsky has argued in his Society of Mind that intelligent behaviours can emerge from the interaction of many simple processes, even though each process may lack 'intelligence' in isolation. In addition, Anderson argued that an emergent system produces more complex behaviours and properties that cannot be reduced to the sum of its components.

ASMO has attention, emotion and learning mechanisms that are inspired by human intelligence. It treats each action as a concurrent, independent and self-governed black box process that competes for the robot's attention to perform actions. The attention mechanism is used to mediate the competition among processes, which correspond to the set of potential actions. The emotion mechanism is used to bias the attention demanded by the processes. The learning mechanisms are used to modify the attention in order to improve robots' performances.

Combining concurrent, independent and self-governed black-box processes with attention and emergent approaches allows ASMO to address the four shortcomings of current cognitive architectures. First, the attention mechanism manages resources explicitly. Second, the black-box design allows any kind of independently developed technique to be integrated without the need to know its internal algorithm, representation or mechanism. Third, attention weighted values enables various techniques to be (re)integrated or (re)structured on the fly with considerably less time and effort. Fourth, the concurrent, independent and self-governed designs support the capabilities required by robots by allowing processes to (i) achieve multiple goals concurrently, (ii) fail without causing the whole system to fail and (iii) be maintained in isolation.

ASMO is evaluated using two robotic problems: (i) the RoboCup soccer standard benchmark problem is used to demonstrate proof-of-concept that a team of robots can be supported by ASMO. In particular, a real robot can be governed by ASMO's attention mechanism to undertake complex tasks. (ii) a companion robot problem is used to demonstrate that ASMO's attention, emotion and learning mechanisms overcome the four identified shortcomings of current state-of-the-art cognitive architectures.

This dissertation presents ASMO, an innovative cognitive architecture that addresses the four shortcomings of current state-of-the-art cognitive architectures, and that can also accelerate the development, deployment and adoption of service robots. ASMO provides a more natural and easier approach to programming robots based on a novel bio-inspired attention management system. ASMO allows researchers and robot system developers to focus on developing new capabilities as processes rather than having to be concerned about integrating new capabilities into a cognitive architecture.

1	Intr	roduction	1
	1.1	Scientific Challenge and Motivation	3
	1.2	Aims and Objectives	6
	1.3	Significance and Contributions	6
	1.4	Scope	8
	1.5	Dissertation Organisation	9
2	Rob	oot Cognitive Architecture	11
	2.1	Definition of Robot	11
	2.2	Robot Environment	14
	2.3	Definition of Intelligence	16
	2.4	Definitions of Cognitive Architecture	19
		2.4.1 Comparison with Other Types of Software	21
	2.5	Evaluation Criteria for Cognitive Architectures	24
3	Exi	sting Cognitive Architectures	27
	3.1	Approaches based on Knowledge Design	27
	3.2	Approaches based on Knowledge Utilisation	29
		3.2.1 Lookup-based Approach	31
		3.2.2 Case-Based Finite State Machine Approach	32

viii Contents

		3.2.3	Priority-based or Hierarchical-based Approach	34
		3.2.4	Goal-based Approach	35
		3.2.5	Connection-based Approach	37
		3.2.6	Utility-based Approach	39
	3.3	Curre	nt State of The Art	40
		3.3.1	Soar	42
		3.3.2	Adaptive Control of Thought – Rational (ACT-R)	45
		3.3.3	Subsumption	49
		3.3.4	Situated Automata	51
		3.3.5	Procedural Reasoning System (PRS)	53
		3.3.6	Agent Network Architecture (ANA)	55
		3.3.7	ICARUS	59
		3.3.8	Three Tier / Three Layers Architecture (3T)	62
		3.3.9	Polyscheme	64
	3.4	Relate	ed Architectures	67
	3.5	Summ	nary and Gaps	70
4	ASI	MO Co	ognitive Architecture	73
	4.1	Overa	ll Design	73
		4.1.1	Memory	75
		4.1.2	Attention Mechanism	77
		4.1.3	Emotion Mechanism	78
		4.1.4	Learning Mechanisms	78
	4.2	Share	Management of Resources	80
	4.3	Cogni	tive Capabilities	81
		4.3.1	Outer World, Inner World, Sensation and Actuation	82

		4.3.2	Perception, Conception, Simulation and Planning 8	33
5	ASI	MO's A	Attention Mechanism 8	35
	5.1	Need	for Attention in Decision Making	35
	5.2	Theor	ies of Attention	37
		5.2.1	Selective or Focused Attention	37
		5.2.2	Divided Attention	39
		5.2.3	Automaticity) 1
	5.3	Comp	utational Design and Model) 3
	5.4	Imple	mentation	97
		5.4.1	Modules	9 9
		5.4.2	Attention Value)3
		5.4.3	Boost Value and Reflex Priority)6
		5.4.4	Attention Competition)7
	5.5	Other	Work	12
6	ASI	MO's I	Emotion Mechanism 11	9
	6.1	Needs	for Emotion and Subjective Bias	19
	6.2	Theor	ies of Emotion	21
		6.2.1	Representation	21
		6.2.2	Causality	23
		6.2.3	Evaluation: Innate or Learned	26
	6.3	Comp	utational Design and Model	27
	6.4	Imple	mentation	31
		6.4.1	Dimension and Label Nodes	32
		6.4.2	Biological Factor and Cognitive Factor Nodes	33
		6.4.3	Subjective Weight	34

	6.5	Other	Work	. 138
7	ASI	MO's I	Learning Mechanisms	140
	7.1	Needs	for Learning	. 140
	7.2	Theor	ies of Learning	. 141
		7.2.1	Habituation and Sensitisation	. 141
		7.2.2	Operant Conditioning	. 143
		7.2.3	Classical Conditioning	. 144
		7.2.4	Observational Learning	. 145
	7.3	Comp	utational Design and Model	. 146
	7.4	Implei	mentation	. 147
		7.4.1	Habituation and Sensitisation	. 148
		7.4.2	Operant Conditioning	. 150
		7.4.3	Classical Conditioning	. 152
		7.4.4	Observational Learning	. 153
	7.5	Other	Work	. 156
		7.5.1	Habituation and Sensitisation	. 156
		7.5.2	Operant Conditioning	. 159
		7.5.3	Classical Conditioning	. 160
		7.5.4	Observational Learning	. 162
0	Evo	luotion		165
8		luation		
	8.1		Cup Soccer SPL Standard Benchmark Problem	
		8.1.1	ASMO's Attention Mechanism	. 170
		8.1.2	ASMO's Emotion and Learning Mechanisms	. 175
	8.2	Smoke	ey Robot Companion Problem	. 177
		8.2.1	ASMO's Attention Mechanism	. 181

		8.2.2 ASMO's Emotion Mechanism		 	186
		8.2.3 ASMO's Habituation and Sensitisation Mechanis	ms .	 	189
		8.2.4 ASMO's Operant Conditioning Mechanism		 	194
		8.2.5 ASMO's Classical Conditioning Mechanism $$		 	196
		$8.2.6$ ASMO's Observational Learning Mechanism $\ .$.		 	200
	8.3	Analysis and Discussion		 	201
9	Con	aclusion			208
	9.1	Significance and Contributions		 	208
	9.2	Limitation and Future Work		 	210
	9.3	Final Thoughts		 	212
\mathbf{A}	Non	a-Cognitive Architectures			214
	A.1	Autonomous Robot Architecture (AuRA)		 	214
	A.2	Distributed Architecture for Mobile Navigation (DAMN))	 	216
	A.3	Emotionally Grounded (EGO) Architecture		 	219

xii List of Figures

List of Figures

1.1	Assistance Needed for People with Disability by Tasks	2
1.2	The Current Standardised Hardware Platform for 2014	4
2.1	Classification of Agents	12
2.2	Cognitive Architecture and Other Software	22
3.1	Decision Making Categories	29
3.2	Competitive vs. Collective Decision Making	30
3.3	Priority Scenario	35
3.4	Neural Network Example	38
3.5	Robots Governed by Soar	42
3.6	Soar Architecture [120]	44
3.7	Robots Governed by ACT-R	46
3.8	ACT-R Architecture	47
3.9	Subsumption Layer [42]	49
3.10	Subsumption Module [42]	50
3.11	Circuit Generated by Gapps	52
3.12	Procedural Reasoning System	54
3.13	Agent Network Architecture	56
3.14	Robots Governed by ICARUS	59

List of Figures xiii

3.15	ICARUS Architecture
3.16	Three Tier / Three Layers Architecture [34] 62
3.17	Robots Governed by Polyscheme
4.1	Robots Governed by ASMO
4.2	ASMO Cognitive Architecture
4.3	Cognitive Capabilities proposed by Gärdenfors and Williams 82
4.4	ASMO Body
5.1	Attention Competition or Election
6.1	Theories of Emotional Causality
6.2	Design of ASMO's Emotion Mechanism
6.3	Types of Nodes Used by ASMO's Emotion Mechanism
7.1	Neural Networks for Observational Learning
8.1	RoboCup Soccer Scenario
8.2	Pseudocode of Some Modules Used in RoboCup Soccer System 171
8.3	Attentional Decision Making in The RoboCup Soccer SPL Competi-
	tion 2010
8.5	Neural Network Used by The attend_motion Module
8.6	Attentional Decision Making in Smokey
8.7	Causal Bayesian Network Used in Smokey
8.8	Experiments of Emotional Subjective Bias in Smokey
8.9	Habituation and Sensitisation Learning Experiment in Smokey 192
8.9	Experiments of Habituation and Sensitisation Learning in Smokey 193
8.10	Experiments of Operant Conditioning Learning in Smokey 197
8.11	Experiments of Classical Conditioning Learning in Smokey 199

xiv List of Figures

8.12	The Neural Network of The Observational Learning in Smokey Robot	
	Companion	200
8.13	Experiments of Observational Learning in Smokey	202
A.1	Robots Governed by AuRA	215
A.2	DAMN Constraint Arbitration [186]	217
A.3	DAMN Actuation Arbitration [186]	217
A.4	DAMN Effect Arbitration [186]	218
A.5	Robots Governed by EGO	220
A.6	EGO Architecture	222
A 7	An Example of A Module Tree Structure in EGO Architecture)99

List of Tables xv

List of Tables

3.1	Lookup Table Example	32
3.2	Priority Table Example	34
3.3	Action Description Table Example	36
3.4	Comparison to Bryson's Trend of Cognitive Architectures	41
3.5	Comparison of Approaches in Cognitive Architecture based on Knowledge Design	70
3.6	Comparison of Approaches in Cognitive Architecture based on Knowledge Utilisation	71
3.7	Comparison of Current State of The Art of Cognitive Architectures .	72
5.1	Stroop Effect	91
5.2	Effects for Different Rates in Attention Value	.06
8.1	Users' Preferences of Playing Ball and Drums	.83
8.2	The Subjective Weights of the play_ball and play_drums Modules 1	.88
8.3	Users' Requests	.98
8 4	Probability of Requests Asked by Users 1	98