Improved Humanoid Robot Movement through Impact Perception and Walk Optimisation

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Submitted in partial fulfilment of the requirements for the degree of **Doctor of Philosophy**

School of Electrical Engineering and Computer Science University of Newcastle, Australia May 2012 The thesis contains no material which has been accepted for the award of any other degree or diploma in any university or other tertiary institution and, to the best of my knowledge and belief, contains no material previously published or written by another person, except where due reference has been made in the text. I give consent to this copy of my thesis, when deposited in the University Library^a, being made available for loan and photocopying subject to the provisions of the Copyright Act 1968.

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Contents

1	\mathbf{Intr}	oduction	1
	1.1	RoboCup Soccer: Synthetic Application for Humanoid Robots .	3
		1.1.1 NUbots: University of Newcastle's RoboCup Team	4
	1.2	Software for Legged Robots	5
	1.3	Stance for Humanoid Robots	7
	1.4	Walk Optimisation for Humanoid Robots	9
	1.5	Summary of Contributions and Publications	13
		1.5.1 Publications	13
	1.6	Thesis Overview	15
2	Rev	iew of Human Motion	17
	2.1	Introduction	18
	2.2	Human Quiet Stance	19
		2.2.1 Joint Positions for Stance	19
		2.2.2 Proprioception for Detecting Perturbations	19
		2.2.3 Joint Stiffness for Stance	21
	2.3	Human Perturbed Stance	22
		2.3.1 Responses to Perturbations	22
	2.4	Human Walking	23
		2.4.1 An Efficiency–based Fitness Function	23
		2.4.2 Joint Stiffness for Walking	24
	2.5	Summary	24
3	The	NUPlatform Software Framework	25
	3.1	Introduction	26
		3.1.1 Related Work	27

		3.1.2 Architecture Overview	9
	3.2	The Blackboard	O
		3.2.1 Sensors	1
		3.2.2 Actuators	2
		3.2.3 Visual Information	4
		3.2.4 Jobs	5
		3.2.5 Network Information	5
	3.3	The Platform	6
		3.3.1 NUPlatform	7
		3.3.2 NUCamera	3
		3.3.3 NUSensors	3
		3.3.4 NUActionators	9
	3.4	The Software Modules	9
		3.4.1 Behaviour	9
		3.4.2 Motion	O
	3.5	System Configuration	2
	3.6	Applications of NUPlatform	2
	3.7	Conclusion	4
4	Imp	pact Perception for a Standing Humanoid Robot 45	<u> </u>
4	4.1	Introduction	
	4.1	4.1.1 Review of Related Work	
		4.1.2 System Overview	
	4.2	Equipment and Data Collection	
	4.3	Detecting a Perturbation	
	1.0	4.3.1 An Optimised Threshold Detector	
		4.3.2 Discussion of Detection Results	
	4.4	Perceiving the Location of a Perturbation	
	1.1	4.4.1 Classification of Location Using an SVM	
		4.4.2 Discussion of Classification Results	
	4.5	Estimating the Direction and Strength of a Perturbation 65	
		0	
		4.5.1 Estimation of a Perturbation Using SVR Models 65	
		4.5.1 Estimation of a Perturbation Using SVR Models 63 4.5.2 Discussion of Estimation Results	
	4.6	<u> </u>	4
	4.6 4.7	4.5.2 Discussion of Estimation Results 64	4 6

5	Imp	rovem	ents in Walking through Joint Stiffness Reduction	69
	5.1	Introd	luction	70
	5.2	Equip	ment and Method	71
		5.2.1	Hardware and Software	71
		5.2.2	Optimisation Algorithm and Parameter Space	72
		5.2.3	Optimisation Path and Fitness Function	73
	5.3	Result	ts	73
		5.3.1	Speed	74
		5.3.2	Efficiency	74
		5.3.3	Stability	76
	5.4	Discus	ssion	80
	5.5	Concl	usion	81
6	Met	a–opt	imisation of Walk Optimisation Techniques	83
	6.1	Introd	luction	84
	6.2	Equip	ment and Method	85
		6.2.1	Hardware and Software	85
		6.2.2	Optimisation Path	86
		6.2.3	Optimisation Expense	88
	6.3	Optim	nisation Algorithms	89
		6.3.1	Evolutionary Hill Climbing with Line Search	90
		6.3.2	Policy Gradient Reinforcement Learning	90
		6.3.3	Gaussian Particle Swarm Optimisation	93
	6.4	Fitnes	s Functions for Optimisation	93
		6.4.1	Speed	94
		6.4.2	Efficiency	94
		6.4.3	Froude–Number	95
	6.5	Paran	neter Spaces for Optimisation	95
	6.6	Meta-	optimisation of Algorithms	96
	6.7	Design	a of the Comparison of Walk Optimisation Techniques	97
	6.8	Comp	arison of Algorithms	99
	6.9	Comp	arison of Fitness Functions	108
	6.10	Comp	arison of Parameter Spaces	110
	6.11	Applie	cation to the Physical NAO	112
	6.12	Concl	usion	115

7	Wal	k Optimisation with Redundant Fitness Functions 11	17
	7.1	Introduction	18
	7.2	Equipment and Method	19
	7.3	Opposition–based PGRL with Redundant Fitness Functions $\boldsymbol{1}$	21
		7.3.1 Opposition–based Policy Generation	21
		7.3.2 Use of Redundant Fitness	21
	7.4	Applications of the Improved PGRL Algorithm	24
	7.5	Conclusion	29
8	Gai	t–Phase Dependent Joint Stiffnesses	30
	8.1	Introduction	31
		8.1.1 Phases of the Gait Cycle	32
	8.2	Phase Dependent Stiffness with Fixed Traditional Walk Param-	
		eters	34
		8.2.1 Equipment and Method	34
		8.2.2 Results	40
	8.3	Optimisation of Phase Dependent Stiffness and Traditional Walk	
		Parameters	40
		8.3.1 Equipment and Method	40
		8.3.2 Results	42
	8.4	Discussion	44
		8.4.1 Phase Dependent Stiffness with Fixed Traditional Walk	
		Parameters	44
		8.4.2 Phase Dependent Stiffness with Variable Traditional Walk	
		Parameters	45
		8.4.3 Robot Tracking	46
	8.5	Conclusion	47
9	Con	nclusion 14	48
	9.1	Conclusions	48
	9.2	Future Work	51
	9.3	Summary	52
Bi	bliog	graphy 15	53

Abstract

The proficiency of humanoid robot movement, which is currently quite elementary, needs to be improved if humanoid robots are to fulfil most of their intended applications. Two of the more essential motor skills of a humanoid robot are related to its ability to stand and walk. Enhancement of these abilities is the focus of the work presented in this thesis.

We first investigate the use of the proprioceptive sense, in particular the joint velocities, to perceive and quantify external perturbations to a standing humanoid robot. A system consisting of an optimised threshold detector, a Support Vector Machine and a pair of orthogonal Support Vector Regression models is developed to utilise this proprioceptive sense. We demonstrate, through the implementation on a physical robot, that the proposed system is able to detect, locate and estimate the magnitude and direction of any given impact.

Next we consider improvements to humanoid robot walking through the enhancement of walk optimisation techniques. To this end, in simulation, a meta—optimisation is performed to determine: an appropriate set of tuning parameters for three different optimisation algorithms, the most suitable optimisation algorithm, a relevant fitness function and a pertinent parameter space. The optimisation algorithms we consider include: Evolutionary Hill Climb with Line Search, Particle Swarm Optimisation and Policy Gradient Reinforcement Learning (PGRL). We evaluated fitness functions based on the walk speed, efficiency and Froude—number. The parameter space for the walk engine was assessed with and without additional joint stiffness parameters. We found that the best walk optimisation technique consisted of PGRL with an efficiency based fitness function utilising additional joint stiffness parameters.

We achieved further improvements on the walk optimisation by applying the safe redundancy concept to extend PGRL. PGRL is a local optimisation algorithm, whereby incorporating safe redundancy allows the algorithm to escape from local extrema. We also expanded the parameter space to include gait—phase dependent joint stiffnesses. Furthermore, to facilitate a trade—off between the optimisation and the stress placed on the physical hardware, a measure of the wear experienced by the robot during the optimisation was introduced.

To verify the generality of the systems developed for the walk optimisation, they are evaluated on several different humanoid robot platforms: a simulated NAO, a physical NAO and a DARWIN-OP. The effectiveness of the proposed systems are demonstrated through their implementation in physical humanoid robot hardware and application to the RoboCup soccer competitions.