	doc_1		doc_2		decision	id
cases	authors	 authors G. Albi Sara Bicego D. Kalise 		Giacomo Albi Sara Bicego Dante Kalise		
			title	Equations		
	title	Gradient-augmented Supervised Learning of Optimal Feedback Laws Using State-Dependent Riccati Equations	publication_date 2021-03-06 10:34:23+00:00			
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	journal	IEEE Control Systems Letters	volume			
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	doi	10.1109/LCSYS.2021.3086697		• http://arxiv.org/pdf/2103.04091v1		229
	urls	https://www.semanticscholar.org/paper/02fbae8030cf6221b36f483d3a89a8fd868c6f29	urls	 http://arxiv.org/abs/2103.04091v1 http://arxiv.org/pdf/2103.04091v1 		
	id	id7294044276266029016				
	abstract	A supervised learning approach for the solution of large-scale nonlinear stabilization problems is presented. A	id	id357580827109923741		
		stabilizing feedback law is trained from a dataset generated from State-dependent Riccati Equation solvers. The training phase is enriched by the use of gradient information in the loss function, which is weighted through the use of hyperparameters. High-dimensional nonlinear stabilization tests demonstrate that real-time sequential large-scale Algebraic Riccati Equation solvers can be substituted by a suitably trained feedforward neural network.	le abstract	A supervised learning approach for the solution of large-scale nonlinear stabilization problems is presented. A stabilizing feedback law is trained from a dataset generated from State-dependent Riccati Equation solves. The training phase is enriched by the use gradient information in the loss function, which is weighted through the use of hyperparameters. High-dimensional nonlinear stabilization tests		
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