

cases	doc_1		doc_2		decision	id
			authors	<ul style="list-style-type: none">Huyen PhamXavier WarinMaximilien Germain	DUPLICATES	299
	authors	<ul style="list-style-type: none">H. PhamX. WarinMaximilien Germain	title	Neural networks-based backward scheme for fully nonlinear PDEs		
	publication_date	2019-07-30 00:00:00	publication_date	2019-07-31 08:09:13+00:00		
	source	SupportedSources.SEMANTIC_SCHOLAR	source	SupportedSources.ARXIV		
	journal	SN Partial Differential Equations and Applications	journal	None		
	volume	2	volume			
	doi	10.1007/s42985-020-00062-8	doi			
	urls	<ul style="list-style-type: none">https://www.semanticscholar.org/paper/600504286652a52c71283f66ba736d002584486e	urls	<ul style="list-style-type: none">http://arxiv.org/pdf/1908.00412v3http://arxiv.org/abs/1908.00412v3http://arxiv.org/pdf/1908.00412v3		
	id	id-8198424547048831097	id	id-6202241945493017849		
	abstract	None	abstract	We propose a numerical method for solving high dimensional fully nonlinear partial differential equations (PDEs). Our algorithm estimates simultaneously by backward time induction the solution and its gradient by multi-layer neural networks, while the Hessian is approximated by automatic differentiation of the gradient at previous step. This methodology extends to the fully nonlinear case the approach recently proposed in \cite{HPW19} for semi-linear PDEs. Numerical tests illustrate the performance and accuracy of our method on several examples in high dimension with nonlinearity on the Hessian term including a linear quadratic control problem with control on the diffusion coefficient, Monge-Amp\`ere equation and Hamilton-Jacobi-Bellman equation in portfolio optimization.		
	versions		versions			