

cases	doc_1		doc_2		decision	id
	authors	<ul style="list-style-type: none">Zhi-Qin John XuWei CaiZiqi Liu	authors	<ul style="list-style-type: none">Ziqi LiuWei CaiZhi-Qin John Xu	DUPLICATES	60
	title	Multi-scale Deep Neural Network (MscaleDNN) for Solving Poisson-Boltzmann Equation in Complex Domains	title	Multi-scale Deep Neural Network (MscaleDNN) for Solving Poisson-Boltzmann Equation in Complex Domains		
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	urls	<ul style="list-style-type: none">https://arxiv.org/pdf/2007.11207v3.pdfhttps://github.com/xuzhiqin1990/mscalednn	urls	<ul style="list-style-type: none">https://www.semanticscholar.org/paper/6af63ef2b79dd8acd7310154ea3f4c9f2c3754f3		
	id	id7051780797741875897	id	id1140236102066408893		
	abstract	In this paper, we propose multi-scale deep neural networks (MscaleDNNs) using the idea of radial scaling in frequency domain and activation functions with compact support. The radial scaling converts the problem of approximation of high frequency contents of PDEs' solutions to a problem of learning about lower frequency functions, and the compact support activation functions facilitate the separation of frequency contents of the target function to be approximated by corresponding DNNs. As a result, the MscaleDNNs achieve fast uniform convergence over multiple scales. The proposed MscaleDNNs are shown to be superior to traditional fully connected DNNs and be an effective mesh-less numerical method for Poisson-Boltzmann equations with ample frequency contents over complex and singular domains.	abstract	In this paper, we propose multi-scale deep neural networks (MscaleDNNs) using the idea of radial scaling in frequency domain and activation functions with compact support. The radial scaling converts the problem of approximation of high frequency contents of PDEs' solutions to a problem of learning about lower frequency functions, and the compact support activation functions facilitate the separation of frequency contents of the target function to be approximated by corresponding DNNs. As a result, the MscaleDNNs achieve fast uniform convergence over multiple scales. The proposed MscaleDNNs are shown to be superior to traditional fully connected DNNs and be an effective mesh-less numerical method for Poisson-Boltzmann equations with ample frequency contents over complex and singular domains.		
	versions		versions			