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source SupportedSources.OPENALEX journal Chaos volume 30 doi 10.1063/5.0010482  urls • https://openalex.org/W3083772601 • https://openalex.org/W3083772601 • https://openalex.org/W3083772601 • https://openalex.org/W3083772601 • https://arxiv.org/pdf/2001.05437  id id.805920897457675023  Physics-informed neural networks are developed to characterize the state of dynamical systems in a random environment. The neural network approximates the probability density flunction (pdf) or the characteristic function (chf) of the state of these systems which satisfy the Fokker-Planck equation or an integro-differential equation to characterize the state. It is also demonstrated how prior information of the dynamical system can be exploited to design and simplify the neural network approximates the probability density function (pdf) or the characterize the state. It is also demonstrated how prior information of the dynamical system can be exploited to design and simplify the neural network approximate the target solution even for partial integro-differential equation sand system of PDEs describing the time evolution of the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential equation using neural networks yields similar pdfs of the state of these systems which satisfy the Fokker-Planck equation or the chf differential equation sand system of PDEs describing the time evolution of the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential equation using neural networks yields similar pdfs of the state of these state and 3) the solution to the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential equation using neural networks yields similar pdfs of the state of these state and 3) the solution to the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential equation using neural networks yields similar pdfs of the state of these state and 3) the solution to the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential				source	SupportedSources.INTERNET_ARCHIVE	
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• https://openalex.org/W3083772601 • https://doi.org/10.1063/5.0010482 • http://arxiv.org/pdf/2001.05437  id id-7720143324705524815  abstract  • https://openalex.org/W3083772601 • https://ope		volume	30		<ul> <li>https://web.archive.org/web/20200508014248/https://arxiv.org/pdf/2001.05437v2.pdf</li> </ul>	
• https://openalex.org/W30837/2601 • https://doi.org/10.1063/5.0010482 • http://arxiv.org/pdf/2001.05437    id   id-7720143324705524815     abstract     abstract   abstract     abstract		doi	10.1063/5.0010482		:41.005020007.457675022	
• http://arxiv.org/pdf/2001.05437  • http://arxiv.o		urls	<ul> <li>https://doi.org/10.1063/5.0010482</li> <li>http://arxiv.org/pdf/2001.05437</li> </ul>	10		
id id-7720143324705524815  Numerical examples show that: 1) the neural network solution can approximate the target solution even for partial integro-differential equations and system of PDEs describing the time evolution of the pdf/chf, 2) solving either the Fokker-Planck equation or the chf differential equation using neural networks yields similar pdfs of the state, and 3) the solution to these differential equations can be used to study the behavior of the state for different types of random forcings.				abstract	density function (pdf) or the characteristic function (chf) of the state of these systems which satisfy the Fokker-Planck equation or an integro-differential equation under Gaussian and/or Poisson white noises. We examine analytically and numerically the advantages and disadvantages of solving each type of differential equation to characterize the state. It is also demonstrated how prior information of the dynamical system can be exploited to design and simplify the neural network architecture. Numerical examples show that: 1) the neural network solution can approximate the target solution even for partial integro-differential equations and system of PDEs	
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