NASA 2022 Symposium on Turbulence Modeling: Roadblocks, and the Potential for Machine Learning

Viswanathan, V. (talk+Challenge): Differentiable Physics for Turbulence Modeling

10:45 - 11:00 am 11:00 - 11:15 am

(Held in honor of Philippe R. Spalart's contributions to the turbulence modeling field)

Agenda V17

Agenda V17			Participant Talks
Wed July 27			
7:30 - 7:45 am	Registration & refreshment/food payments		
7:45 - 8:00 am			
8:00 - 8:15 am			
8:15 - 8:30 am			
8:30 - 8:45 am	Welcome, Intro, Overview	1 hr	
8:45 - 9:00 am			
9:00 - 9:15 am	Bookend talk 1 - SPALART		
9:15 - 9:30 am	An Old-Fashioned Framework for Machine Learning in Turbulence Modeling		
9:30 - 9:45 am	Break		
9:45 - 10:00 am	Knopp, T.: A data-driven wall law for the mean velocity in adverse-pressure gradient and modification of the SSG/LRR-w model	1.75 hrs	RANS related
10:00 - 10:15 am	Abe, H.: Improvement on the AMM model for predicting wing-body juncture flows		
10:15 - 10:30 am	Agarwal, R.: Review and Potential of Wray-Agarwal Family of Turbulence and Transition Models for RANS Simulations		
10:30 - 10:45 am	Fritsch, D.: Measurement and Modeling of Non-Equilibrium Turbulent Boundary Layer Flows		
10:45 - 11:00 am	Lowe, T.: Benchmark Turbulence Modeling Validation Experiments for Three-Dimensional Flows with Separation		
11:00 - 11:15 am	Discussion		
11:15 - 11:30 am			
11:30 - 11:45 am	Lunch		
11:45 - 12:00 noon			
12:00 - 12:15 pm			
12:15 - 12:30 pm			
12:30 - 12:45 pm			
12:45 - 1:00 pm	Invited Talk - HUANG	2 hrs	
1:00 - 1:15 pm	The Anchor Points of Turbulence Modeling		
1:15 - 1:30 pm			
1:30 - 1:45 pm	Panel Discussion - RANS		
1:45 - 2:00 pm	Batten, Durbin, Hirsch, Smith		
2:00 - 2:15 pm	(Moderator: Spalart)		
2:15 - 2:30 pm			
2:30 -2:45 pm			
2:45 -3:00 pm	Break	4 75 1	
3:00 - 3:15 pm	Intro to Challenge Case (Rumsey)	1.75 hrs	= 11 . 1 .
3:15 - 3:30 pm	Stoellinger, M. (talk+Challenge): V&V of DES using multiple CFD codes		Talks including
3:30 - 3:45 pm	Cherroud, S. (talk+Challenge): Local Bayesian Mixing of Stochastic Explicit Algebraic Reynolds Stress Models		ML Challenge
3:45 - 4:00 pm 4:00 - 4:15 pm	Cherroud, 5. (talk+Challenge): Local Bayesian Mixing of Stochastic Explicit Algebraic Reynolds Sitiess Models		
4:15 - 4:30 pm	Discussion/Day 1 wrapup		
4:40 - 4:45 pm	The state of the s		
4:45 - 5:00 pm			
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7:00 PM	NO-HOST DINNER		
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Thu July 28		_	
8:00 - 8:15 am	Announcements, Coffee		
8:15 - 8:30 am			_
8:30 - 8:45 am	Invited Talk - SANDBERG	1.5 hrs	
8:45 - 9:00 am	Evolution of Gene Expression Programming for Turbulence Modelling		
9:00 - 9:15 am			
9:15 - 9:30 am	Fang, Y. (Challenge): Towards more accurate and general turbulence models using CFD-driven training on multiple flows		Talks including
9:30 - 9:45 am	Parish, E. (talk+Challenge): A data-driven turbulence modeling framework for the Reynolds-averaged Navier-Stokes equations via discrepancy-based tensor-basis neural network	:S	ML challenge
9:45 - 10:00 am			
10:00 - 10:15 am	Break		
10:15 - 10:30 am	Bin, Y. and Li, J. (talk+Challenge): Progressive, extrapolative machine learning in turbulence modeling	1.5 hrs	Talks including
10:30 - 10:45 am	Managathan W (kelluchallanga). Differentiable Dheise for Turkulang Madelian		ML challenge

	Marepally, K. (Challenge): Field-Inversion Machine Learning Approach for Improved Turbulence and Transition Predictions of Flows Over Airfoils		
	Dwight, R. (Challenge): SpaRTA with classification		
	Lunch		
ŀ	Discussion of Challenge	1 hr	
	Discussion of challenge	1 111	
ŀ	Break		
ı	Invited Talk - HIRSCH	0.75 hr	r
	The HiFi-TURB EU project: vision and progress of ML-based turbulence modeling		
	Break		
ı	Srivastava, V.: Developing hierarchical augmentations via the Learning and Inference assisted by Feature-space Engineering (LIFE) framework	1.25 h	
	Orkwis, P.: Using LES/DNS data for Neural Network-based Improvement of Existing Turbulence Models		
	Romano, J.: Towards use of convolutional neural networks as a post processing enhancement to RANS modeled turbulence		
	Discussion/Day 2 wrapup		
	Announcements, Coffee		
		4.5.1	
	Eisfeld, B. (presented by Knopp): Potential of Data Driven Methods for Reynolds Stress Modeling - A Fundamental View	1.5 hrs	
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0.5 hr

Jee, S.: High-fidelity computational data of transitional boundary layers for a data-driven approach
Panel Discussion - TRANSITION

2:45 -3:00 pm 3:00 - 3:15 pm 3:15 - 3:30 pm

3:30 - 3:45 pm

3:45 - 4:00 pm 4:00 - 4:15 pm

4:15 - 4:30 pm 4:40 - 4:45 pm 4:45 - 5:00 pm Xiao, Duraisamy

Break

(Moderator: Durbin)

Final Discussion & Wrapup