MathWorks **AUTOMOTIVE CONFERENCE 2024** Europe

Al with Model-Based Design: Reduced-Order Modeling

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Key Takeaways

Challenge: High-fidelity models often prohibitively slow



Artificial Intelligence techniques can be used to create faster Reduced-Order Models (ROM)



MATLAB & Simulink enables engineers to create ROMs without prior Al knowledge

What is Reduced Order Modeling (ROM)?

Full-order model



How to perform Reduced Order Modeling?



• Perform simulations and SIL / HIL / PIL testing



• Perform simulations and SIL / HIL / PIL testing



Perform simulations and SIL / HIL / PIL testing

deployed on the



- Perform simulations and SIL / HIL / PIL testing
- Virtual sensor modeling



- Perform simulations and SIL / HIL / PIL testing
- Virtual sensor modeling
- Control design



Example overview

Replacing a high-fidelity jet engine turbine blade model with an AI-based reduced order model



Closed-loop temperature control

Example overview

Replacing a high-fidelity jet engine turbine blade model with an AI-based reduced order model



Example overview

Replacing a high-fidelity jet engine turbine blade model with an AI-based reduced order model



Introducing Simulink Add-On for Reduced Order Modeling

Products and Services Search MathWorks.com Reduced Order Modeling with MATLAB and Simulink Create AI-based reduced order models Download add-on (beta) Simulink Add-On for Reduced Order Modeling provides an app for creating reduced order models (ROMs) of subsystems modeled in Simulink, including full-order, high-fidelity third-party simulation models. You can use reduced order models for system-level desktop simulation, hardware-in-the-loop (HIL) testing, control design, and virtual sensor modeling. With Simulink Add-On for Reduced Order Modeling, you can: · Set up the design of experiments and generate input-output training data from a full-order, high-fidelity subsystem · Train and compare Al-based reduced order models using pre-configured templates · Export Al-based surrogate models to Simulink for system-level simulation, control design, and HIL testing · Export reduced order models as Functional Mockup Units (FMUs) for use outside of MATLAB and Simulink (with Simulink Compiler) Reduced Order ROM Modeling component ' First-principles based Full-order model Full-order model First-principles ba CFD/CAE/FEA component 2 component 3 Reduced order model Speed

Reduced Order Modeling with MATLAB and Simulink

Workflow for Reduced Order Modeling App











Design experiments	Run experiments Train ROM Export
Reduced Order Modeler-JetEngineBlade REDUCED ORDER MODEL SIMULATION RESULT Open Save Session Session FILE Inputs/Outputs Inputs/Outputs COLLECT DATA ROM Input	- O X
JetEngineBlade/High-fidelity Model/First Order Hold:1(Ambient_input) JetEngineBlade/High-fidelity Model/First Order Hold:1(Cooling_input) JetEngineBlade/High-fidelity Model/Hirst Order Hold:1(Pressure_input) * ROM Output JetEngineBlade/High-fidelity Model/MATLAB Function:1(maxDisp) * Simulation Input JetEngineBlade/Ambient Temperature:1(Ambient) JetEngineBlade/Pressure:1(Pressure)	$Input: Ambient_input$
- Experiments	$1 \text{ input: Cooling_input}$
Name # Sims Results Image: Compariment in the system of the sys	$= \underbrace{\int_{4.5}^{5.5} \int_{0}^{4.5} \int_{0.2}^{0.2} \int_{0.4}^{0.6} \int_{0.6}^{0.8} \int_{0.8}^{1} \int_{0.2}^{1.2} \int_{1.4}^{1.6} \int_{1.6}^{1.8} \int_{1.8}^{2} \int_{1.0^2}^{2} \int_{1.4}^{1.6} \int_{1.6}^{1.8} \int_{1.8}^{2} \int_{1.6}^{1.6} \int_{1.8}^{1.6} \int_{1$
	25 - 4 - 4 - 4 - 4 - 4 - 4 - 4 - 4 - 4 -



Design experiments		expe	Rui erin	n nents		Train ROM				Export			
Experiment Manager EXPERIMENT MANAGER Copen Open Duplicate ENVIRONMENT Experiment Browser ExperimentProject3	multanecus	Training Test Plot F REVIE Experiment1 × • Exhaustive	Contemporation Conte	Filters Annotat Filters Annotat HI Result 2 ×	ions Export								- 0 >
Experiment1 III Result2 III Result1		Experiment1 (<u>View Experim</u> Use data colle space model, maxepochs for the percentag epochs.	Experiment1 (Mew Experiment Source) Use data collected from the JelEngineBlade model to fit a neural state- space model. By default, 20% of the data sets are used as test data, and maxepochs for training is set to 1000. Edit the training function to change the percentage of test data used and the maximum number of training epochs. ent Details. Homemarameters. Marting M									36/36 Trial	
		Status	Actions	Progress	Elapsed Time	NumberInputL	NumberOutpu	NumberLayers	NumberUnits	SampleRate	TrainingMSE	Loss 🔻	TestMSE
		1 Complete	1	100.0%	0 hr 5 min 16 sec	0.0000	0.0000	2.0000	16.0000	0.2000	0.0214	0.9276	0.0519
		2 Complete	1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	100.0%	0 hr 7 min 34 sec	1.0000	0.0000	2.0000	16.0000	0.2000	0.0272	3.2502	0.036
		4 Complete	-	100.0%	0 hr 14 min 15 sec	1.0000	1.0000	2.0000	16.0000	0.2000	0.0373	3.6828	0.055
		5 S Complete	8	100.0%	0 hr 13 min 51 sec	0.0000	2.0000	2.0000	16.0000	0.2000	0.1076	3.3881	0.194
		6 🥝 Complete	1	100.0%	0 hr 14 min 40 sec	. 1.0000	2.0000	2.0000	16.0000	0.2000	0.0475	9.6849	0.08
		7 🔮 Complete	1	100.0%	0 hr 14 min 59 sec	0.0000	0.0000	3.0000	16.0000	0.2000	0.0208	0.9332	0.04
		8 Complete		100.0%	0 hr 15 min 16 sec	1.0000	0.0000	3.0000	16.0000	0.2000	0.0145	2.8449	0.035
		10 Complete	8	100.0%	0 hr 15 min 35 sec	1.0000	1.0000	3.000	16.000	0.2000	0.0352	2,9724	0.05
		11 O Complete	1	100.0%	0 hr 16 min 15 sec	0.0000	2.0000	3.0000	16.0000	0.2000	0.2199	3.6474	0.32
		12 🔮 Complete	Î	100.0%	0 hr 15 min 54 sec	1.0000	2.0000	3.0000	16.0000	0.2000	0.0374	7.3887	0.09
		13 S Complete	Î	100.0%	0 hr 13 min 7 sec	0.0000	0.0000	2.0000	32.0000	0.2000	0.0195	0.9328	0.03
		14 O Complete		100.0%	0 hr 12 min 11 sec	1.0000	0.0000	2.0000	32.0000	0.2000	0.0188	3.3358	0.05
		16 Complete	-	100.0%	0 hr 12 min 38 sec	1.0000	1.0000	2.0000	32.000	0.2000	0.0337	3.8684	0.029
		17 S Complete		100.0%	0 hr 12 min 20 ser	0.0000	2.0000	2.0000	32.0000	0.2000	0.1214	3.4628	0.255
		18 🥝 Complete	Î	100.0%	0 hr 12 min 47 sec	1.0000	2.0000	2.0000	32.0000	0.2000	0.0452	8.2722	. 0.09
		19 🥝 Complete	Î	100.0%	0 hr 12 min 18 sec	0.0000	0.0000	3.0000	32.0000	0.2000	0.0328	1.1057	0.05
		20 S Complete	1 2	100.0%	0 hr 12 min 30 sec	1.0000	0.0000	3.0000	32.0000	0.2000	0.0146	2.7313	0.042
		21 Complete	8	100.0%	0 hr 12 min 37 sec	0.0000	1.0000	3.0000	32.0000	0.2000	0.0376	2 0521	0.097
		23 😋 Complete	8	100.0%	0 hr 12 min 55 sec	0.0000	2.0000	3.0000	32.0000	0.2000	0.0800	3.0261	0.03-
		24 S Complete	1	100.0%	0 hr 13 min 12 sec	1.0000	2.0000	3.0000	32.0000	0.2000	0.0352	7.3551	0.08
					01.11.21	0.0000	0.0000	2.0000	64.0000	0.2000	0.0182	1,1432	0.035
		25 S Complete		100.0%	Unr 11 min 31 sec								
		25 S Complete 26 S Complete	1	100.0%	0 hr 10 min 51 sec	1.0000	0.0000	2.0000	64.0000	0.2000	0.0554	3.8685	0.053
		25 S Complete 26 Complete 27 Complete	8	100.0% 100.0%	0 hr 10 min 52 sec 0 hr 10 min 52 sec	: 1.0000 0.0000	0.0000	2.0000	0 64.0000 0 64.0000	0.2000	0.0554	3.8685	0.05

				Hyperparameters			Information	Metrics			
	Actions	Progress	Elapsed Time	NumberInputL	NumberOutpu	NumberLayers	NumberUnits	SampleRate	TrainingMSE	Loss	TestMSE
2	Ξ.	100.0%	0 hr 12 min 38 sec	1.0000	1.0000	2.0000	32.0000	0.2000	0.0342	3.8684	0.0290
ł.	- 8	100.0%	0 hr 11 min 31 sec	0.0000	0.0000	2.0000	64.0000	0.2000	0.0182	1.1432	0.0354
	1	100.0%	0 hr 15 min 16 sec	1.0000	0.0000	3.0000	16.0000	0.2000	0.0145	2.8449	0.0358
-	- îi	100.0%	0 hr 7 min 34 sec	1.0000	0.0000	2.0000	16.0000	0.2000	0.0272	3.2502	0.0368
	iii ii	100.0%	0 hr 13 min 7 sec	0.0000	0.0000	2.0000	32.0000	0.2000	0.0195	0.9328	0.0396
	1	100.0%	0 hr 12 min 30 sec	1.0000	0.0000	3.0000	32.0000	0.2000	0.0146	2.7313	0.0426
	10	100.0%	0 hr 10 min 58 sec	0.0000	0.0000	3.0000	64.0000	0.2000	0.0234	1.0063	0.0438
	- B	100.0%	0 hr 14 min 59 sec	0.0000	0.0000	3.0000	16.0000	0.2000	0.0208	0.9332	0.0444
	- E	100.0%	0 hr 5 min 16 sec	0.0000	0.0000	2.0000	16.0000	0.2000	0.0214	0.9276	0.0519
	1	100.0%	0 hr 11 min 37 sec	1.0000	2.0000	2.0000.5	64.0000	0.2000	0.0606	7.2411	0.0528
	1	100.0%	0 hr 15 min 35 sec	1.0000	1.0000	3.0000	16.0000	0.2000	0.0154	2.9724	0.0535
	1	100.0%	0 hr 10 min 52 sec	1.0000	0.0000	2.0000	64.0000	0.2000	0.0554	3.8685	0.0537
	-		_	1.0000	1.0000	3.0000	32.0000	0.2000	0.0197	2.9531	0.0543
		A A A A A A A A A A A A A A A A A A A	Exp	oort > 0.0000	0.0000	3.0000	32.0000	0.2000	0.0328	1.1057	0.0552
7		1		1.0000	1.0000	2,0000	16.0000	0.2000	0.0361	3.6828	0.0553

Simulation and test

Hardware-in-the-loop simulation

Deployment of Reduced Order Models

Key Takeaways

Challenge: High-fidelity models often prohibitively slow

Artificial Intelligence techniques can be used to create faster Reduced Order Models (ROM)

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