UTC INSTITUTE FOR ADVANCED SYSTEMS ENGINEERING SEMINAR SERIES

Friday, April 7th, 2017 2:00 – 3:00PM UConn, Storrs Campus – BPB 130 <u>To view live webcast click here</u>

Uncertainty Quantification using Embarrassingly Small Numbers of Simulations and Experiments

Increasing modeling detail is not necessarily correlated with increasing predictive ability. Setting modeling and numerical discretization errors aside, the more detailed a model gets, the larger the number of forcing, object geometries, etc. To be predictive, we need to quantify this high-dimensional parametric uncertainty by combining our prior physical knowledge with noisy experimental data obtained from various heterogeneous sources. Once we have quantified this uncertainty, all we need to do is propagate it through the model and obtain predictive error bars for any quantity of interest. This program sounds reasonable, but after a little bit of thinking, we realize that it is completely impractical! How many simulations or experiments do you think are required to bring such a program to fruition? Tens, hundreds, smoothness of the response surface. Unfortunately, for general response surfaces the number of required information source queries grows exponentially with increasing number of parameters (the curse of dimensionality). The situation seems hopeless for both a cutting-edge modeler, with simulation models formalism can be used to lay the foundations of a unifying mathematical paradigm for posing such demonstrate the unifying nature of these ideas, I will present results from uncertainty propagation, model calibration, and optimization under uncertainty. The application examples I will consider are 1) the stochastic multi-objective optimization of an FEM-modeled wire-drawing process (selecting simulations) and 2) the optimization of high-quality graphene manufacturing using a plasma enhanced chemical vapor deposition reactor (selecting experiments).

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Dr. Ilias Bilionis is an Assistant Professor at the School of Mechanical Engineering, Purdue University leading the Predictive Science Laboratory. His research focuses on uncertainty quantification, design optimization under uncertainty, model calibration, and socio- technical modeling. Prior to his appointment at Purdue, he was a Postdoctoral Researcher at the Mathematics and Computer Science Division (MCS), Argonne National Laboratory. He received his Ph.D. in Applied Mathematics from Cornell University in 2013 and his Diploma in Applied Mathematics from the National Technical University of Athens in 2008.



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5/08/17 – Ignacio Grossmann

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