

Panel Session: Challenges in Computational Model V&V for Systems Engineers



Marc Horner Technical Lead, Healthcare ANSYS, Inc. marc.horner@ansys.com



Bill Schindel President ICTT System Sciences schindel@ictt.com



Chris Unger Chief Systems Engineer GE Healthcare chris.unger@ge.com

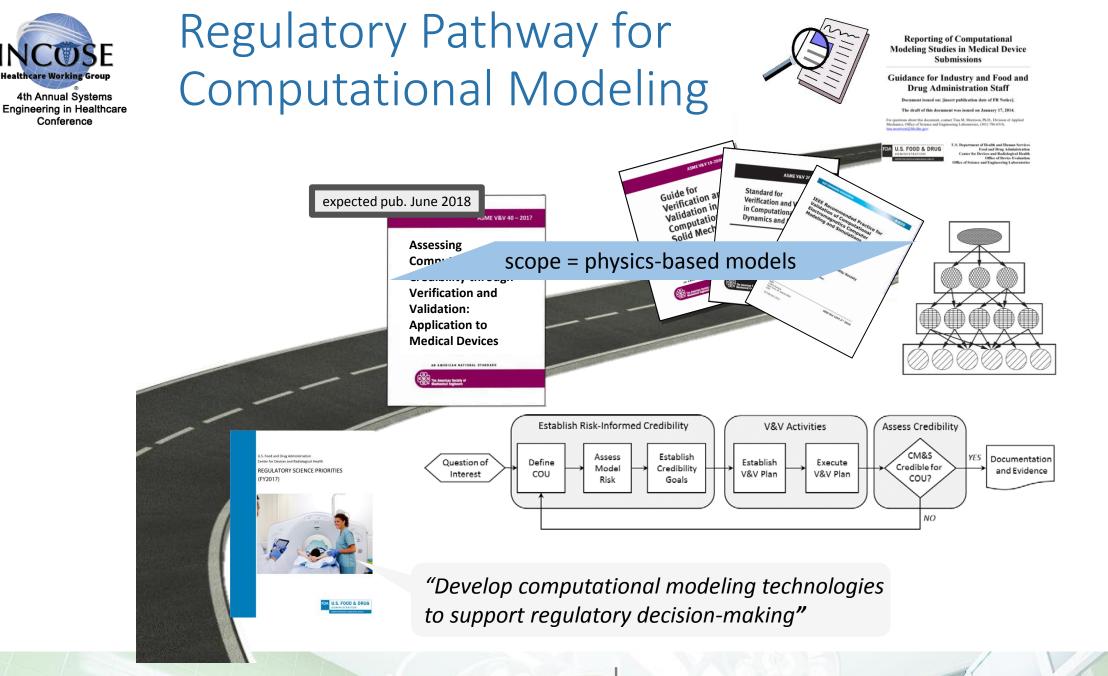


Pras Pathmanathan Research Scientist U.S. Food and Drug Administration pras.pathmanathan@fda.hhs.gov

How Systems Engineering Can Reduce Cost & Improve Quality

19-20 April, 2018 Twin Cities, Minnesota

#hwgsec



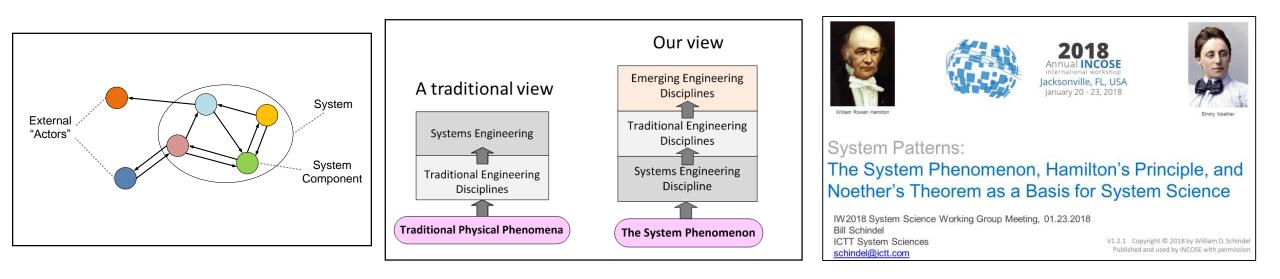
How Systems Engineering Can Reduce Cost & Improve Quality

19-20 April, 2018 Twin Cities, Minnesota

#hwgsec



There is a (mis)perception that "computational models" in mechanics and other phenomena must be different than "system models" familiar in INCOSE. Some of this is a matter of historical origin; convergence greater than evident:

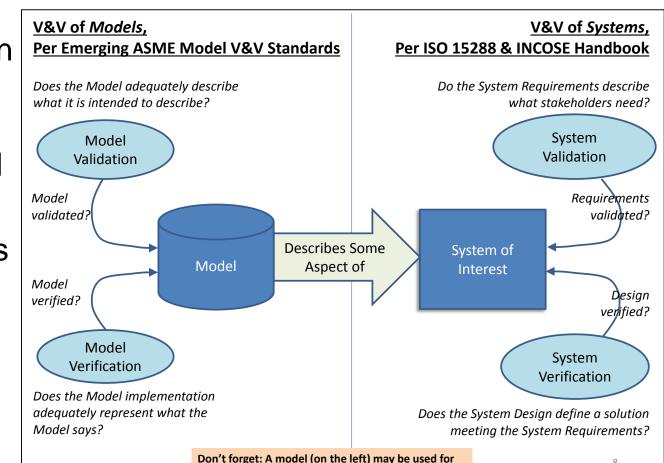


How Systems Engineering Can Reduce Cost & Improve Quality





- One convergence aspect is attention to <u>model credibility</u>.
- Computational modeling, physics-based and data-driven, brings historical attention to model credibility for intended uses-rooted in history of similar approaches to model credibility in physical sciences and mathematics.
- By contrast, credibility of "system" models has sometimes been treated as a more subjective question. The systems community focuses on "V&V of a target system", not always recognizing that the "V&V of models" describing it is different.



system verification or validation (on the right!)



- With increased interest in model compatibility, integration of simulations, and stronger theoretical foundations for systems science, the idea that there is a profound difference between something called a system model and other computational models is hereby questioned.
- INCOSE joining the ASME effort for model VVUQ standards has been a healthy way to pursue these issues. In return for a stronger model VVUQ framework, INCOSE has been able to offer system frameworks in which computational models are managed and exploited, including issues of model credibility.





INCOSE Collaboration In an ASME-Led Standards Activity

Standardizing V&V of Models

Bill Schindel, ICTT System Sciences schindel@ictt.com

V1.2.1

www.incose.org/IW2018



SANDIA REPORT SAND2002-0341 Unlimited Release Printed March 2002

General Concepts for Experimental Validation of ASCI Code Applications

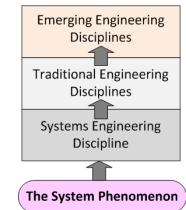
Timothy G. Trucano, Martin Pilch, and William L. Oberkampf

Prepared by Sandia National Laboratories Albuquerque, New Mexico 87185 and Livermore, California 94550

Sandia is a multiprogram laboratory operated by Sandia Corporation a Lockheed Martin Company, for the United States Department of Energy under Contract DE-AC04-94AL85000

Approved for public release; further dissemination unlimited.

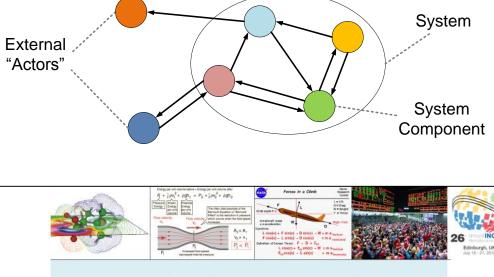




3.2 The Phenomena Identification and Ranking Table (PIRT)

As argued in version 2 of the Sandia V&V planning guidelines (Pilch et al. 2000a), the PIRT is the most important tool in our V&V planning process for translating requirements of the stockpile driver application into requirements on usage of the code, hence specifically on validation activities. The PIRT is particularly important for prioritizing and directing dedicated validation experiment tasks. The intended use of this methodology is thoroughly specified and elaborated in Pilch et al. (2000a) and is not repeated here. However, we do point out that the PIRT is designed to convert the DSW driver application and its associated requirements into specific technical requirements for the code, verification activities, validation activities, and consequent experimental validation requirements. It is the code technical requirements for the driving application that are the proper focus of V&V activities. As a result of a well-executed PIRT process, the validation requirements of the code application are rank ordered in importance. The prioritized PIRT elements directly create the definition and prioritization of the specific validation tasks, especially dedicated validation experiments, which are performed under the validation plan for the code application.

The PIRT is critical for planning validation experiments because it helps establish both sufficiency and efficiency of the validation activities. To demonstrate sufficiency



Got Phenomena?

Science-Based Disciplines for Emerging Systems Challenges

Bill Schindel, ICTT System Sciences schindel@ictt.com

Copyright © 2016 by William D. Schinde

How Systems Engineering Can Reduce Cost & Improve Quality

19-20 April, 2018 Twin Cities, Minnesota



V1.4.7

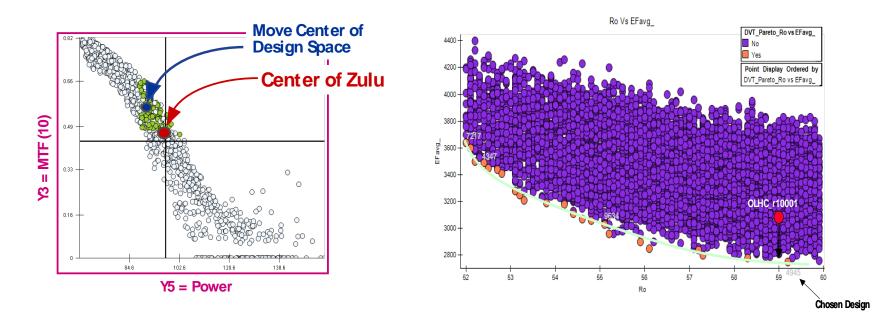


- GEHC extensively uses modeling in our program development, with quite effective results. But from time to time we have had failures.
- Those failures were mostly due to a belief "modeling is good" and not thinking through the relative value of modeling vs. just testing...and whether a model could realistically meet the expectations for its use.

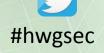




Challenges in Computational Model V&V for Systems Engineers Monte Carlo Simulation of XR Tube Performance

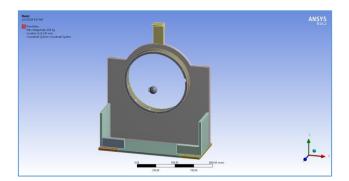


- Extensively explored design space
- Identified opportunities for increased production margin (design robustness)
- Helped save iterations...several further iterations were planned...full design verification



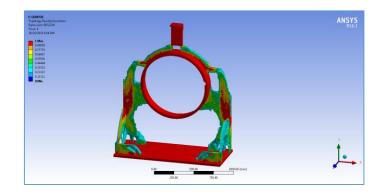


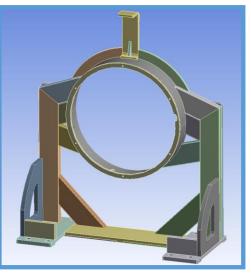
Challenges in Computational Model V&V for Systems Engineers CT Mechanical Optimization



Define the Design Space (Where Material <u>Could</u> Be)

Determine the Theoretical Optimized Shape that Meets Requirements





- Cycle time: Months \rightarrow Weeks \rightarrow 3 days
- Full final design cycle to verify results

Use theoretical optimized shape as input to final design

#hwgsec



Challenges in Computational Model V&V for Systems Engineers Baby Incubator...some predictability issues



An incubator maintains a safe environment (heat, humidity, O2...) for a Infant.

Goal - To develop multi-physics, control & system model that will reduce design iterations

Detailed Goals:

Thermal: Build system CFD model and downselect 2 options for further physical prototyping and CFD; develop ROM for further control loop development

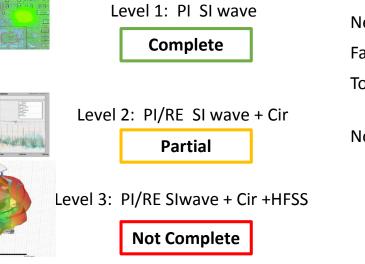
Electrical: Reduced board spins

EMI/EMC: Eliminate the need for screen room testing

- Team overcommitted
- Process lacked effective global review with US experts

How Systems Engineering Can Reduce Cost & Improve Quality

Electrical Result:



EMC Result

Near field simulation: reasonable results

Far field simulation: poor results

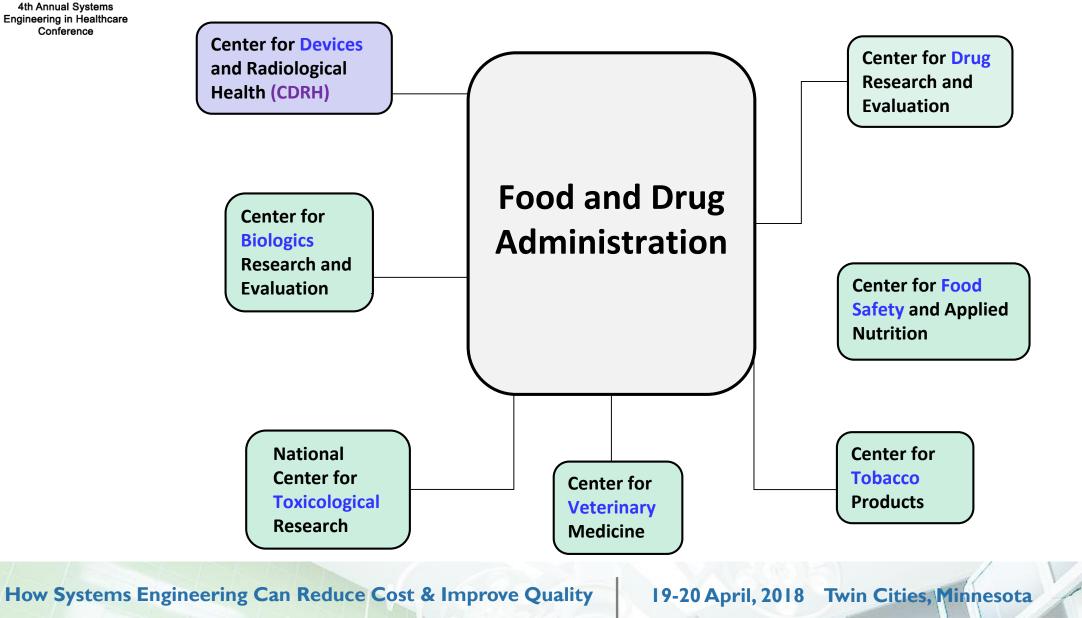
Total effort exceeded the cost of building a local screen room

No analysis of the design margin (very small) compared to modelling error (very large)





Engineering in Healthcare Conference



#hwgsec



From FDA's Strategic Policy Roadmap 2018:

Modernizing our Regulatory Toolbox:

[..] Towards these goals, among the many steps FDA will take: The Agency will embrace advances like predictive toxicology methods and computational modeling across our different product areas

Crucial to develop methods and best practices for rigorously **demonstrating credibility** of biomedical computational models

- Can draw from approaches and successes in traditional engineering fields (e.g. VVUQ)
- However, there are unique challenges





- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group

Reporting of Computational Modeling Studies in Medical Device Submissions

Guidance for Industry and Food and Drug Administration Staff

Document issued on: September 21, 2016.

The draft of this document was issued on January 17, 2014.

For questions about this document, contact Tina M. Morrison, Ph.D., Division of Applied Mechanics, Office of Science and Engineering Laboratories, (301) 796-6310, tina.morrison@fda.hhs.gov.



U.S. Department of Health and Human Services Food and Drug Administration Center for Devices and Radiological Health Office of Device Evaluation Office of Science and Engineering Laboratories

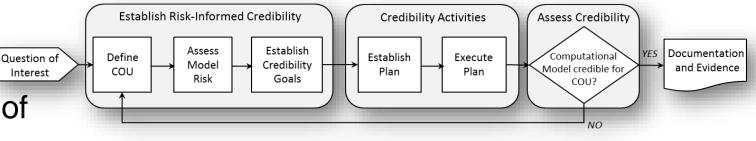
#hwgsec

Google "FDA modeling reporting guidance"

How Systems Engineering Can Reduce Cost & Improve Quality



- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group







- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group

JOURNAL OF BIOPHARMACEUTICAL STATISTICS 2017, VOL. 27, NO. 6, 1089–1103 http://dx.doi.org/10.1080/10543406.2017.1300907

Taylor & Francis

∂ OPEN ACCESS

Incorporation of stochastic engineering models as prior information in Bayesian medical device trials

Tarek Haddad^a, Adam Himes^a, Laura Thompson^b, Telba Irony^{b,c}, Rajesh Nair^b; and on Behalf of MDIC Computer Modeling and Simulation Working Group Participants^{d,e}

^aMedtronic, plc, Mounds View, Minnesota, USA; ^bCenter for Devices and Radiological Health, U.S. Food and Drug Administration, Silver Spring, Maryland, USA; ^cCenter for Biologics Evaluation and Research, U.S. Food and Drug Administration, Silver Spring, Maryland, USA; ^dMedical Device Innovation Consortium Clinical Trials Powered by Bench and Simulation Working Group; ^eSee online supplement for a complete list of participants

ABSTRACT Evaluation of medical devices via clinical trial is often a necessary step in the

ARTICLE HISTORY

Received 4 April 2016 Accepted 23 February 2017

KEYWORDS

Bayesian; clinical trial; medical devices; virtual patient

#hwgsec

process of bringing a new product to market. In recent years, device manufacturers are increasingly using stochastic engineering models during the product development process. These models have the capability to simulate virtual patient outcomes. This article presents a novel method based on the power prior for augmenting a clinical trial using virtual patient data. To properly inform clinical evaluation, the virtual patient model must simulate the clinical outcome of interest, incorporating patient variability, as well as the uncertainty in the engineering model and in its input parameters. The number of virtual patients is controlled by a discount function which uses the similarity between modeled and observed data. This method is illustrated by a case study of cardiac lead fracture. Different discount functions are used to cover a wide range of scenarios in which the type I error rates and power vary for the same number of enrolled patients. Incorporation of engineering models as prior knowledge in a Bayesian clinical trial design can provide benefits of decreased sample size and trial length while still controlling type I error rate and power.

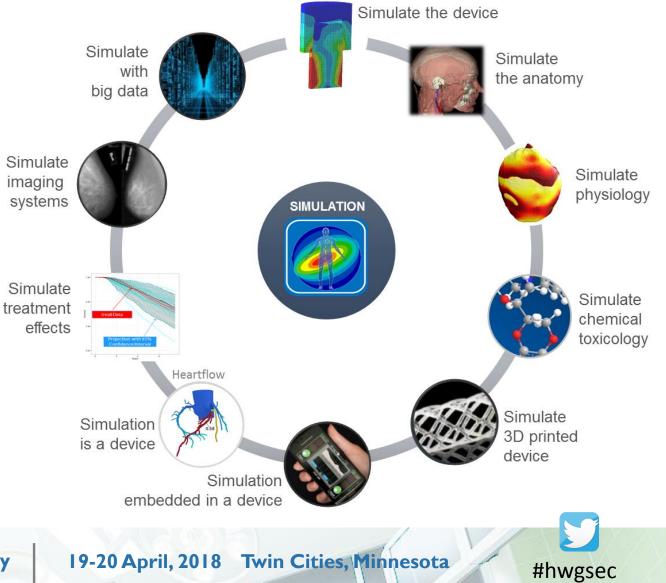
How Systems Engineering Can Reduce Cost & Improve Quality

19-20 April, 2018 Twin Cities, Minnesota

Google "MDIC virtual patient"

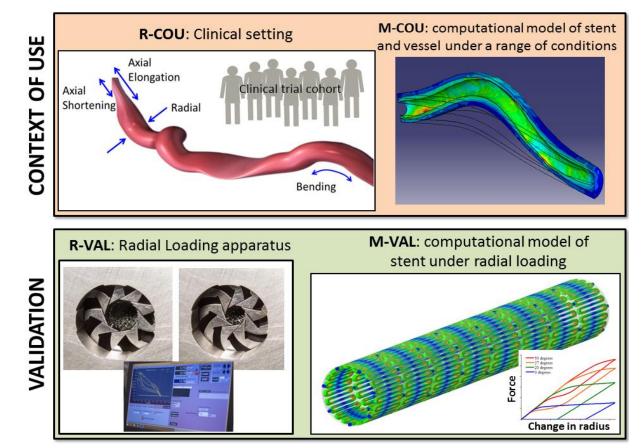


- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group

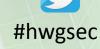




- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group

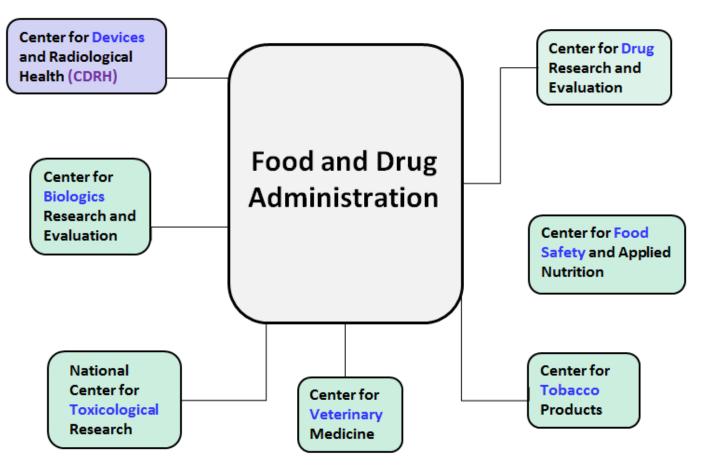


Pathmanathan et al. J. VVUQ, 2017





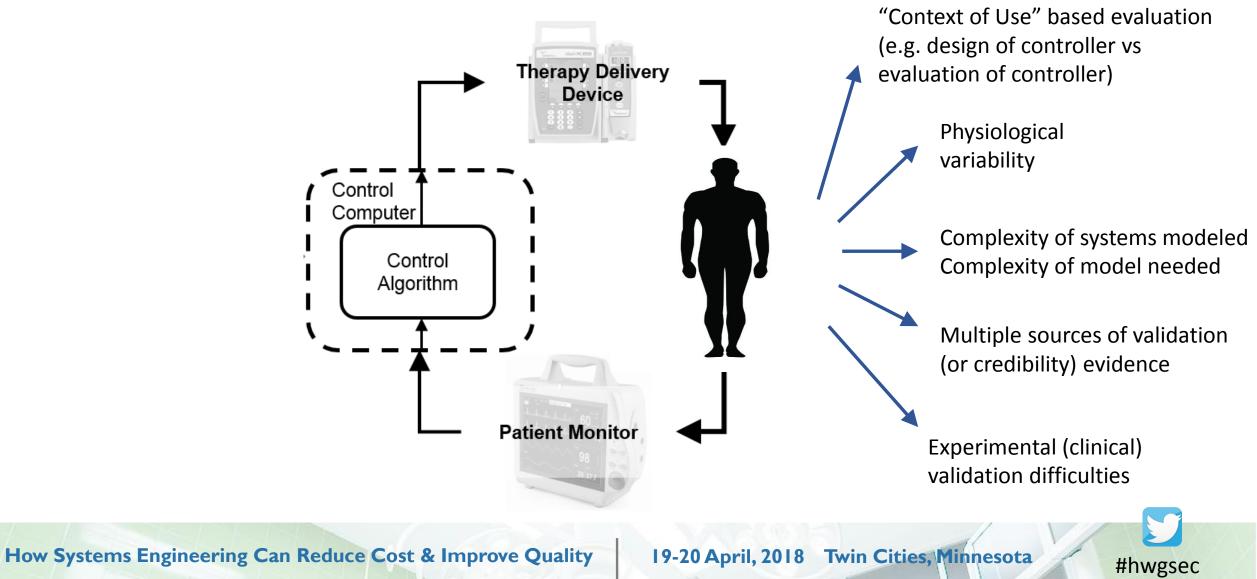
- CDRH Guidance on Reporting of Modeling and Simulation Studies
- ASME V&V40 Standard (to be published summer 2018)
- MDIC "Virtual Patient"
- CDRH research output
- FDA's Modeling and Simulation Working Group



How Systems Engineering Can Reduce Cost & Improve Quality









Q&A

How Systems Engineering Can Reduce Cost & Improve Quality

