

WSC 2067: WHAT ARE THE CHANCES?

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I know what you're thinking...

We're in Las Vegas, we could've had Celine Dion!

My WSC street cred

- Attended every WSC since 1983 (70%).
- Helped <u>drive</u> Proceedings to WSC '83, wrecking the van in the hotel garage.
- Arrested at WSC '84 for bringing "liquid refreshment" into the hotel for the TIMS College on Simulation Mixer.
- Worried Secret Service at WSC '89 by running at Sen. Richard Lugar on the Mall.
- Pasted 2470 running heads & page numbers on the WSC '91 Proceedings by hand (my arms & wrists still hurt).
- Broke WSC '97 budget by forgetting Correlation[number of tracks, attendance] > 0.
- So Ernie, what could possibly go wrong?



In the beginning...

...there was the 1967
"Conference on the
Applications of Simulation
Using GPSS" which became
the WSC we know today.



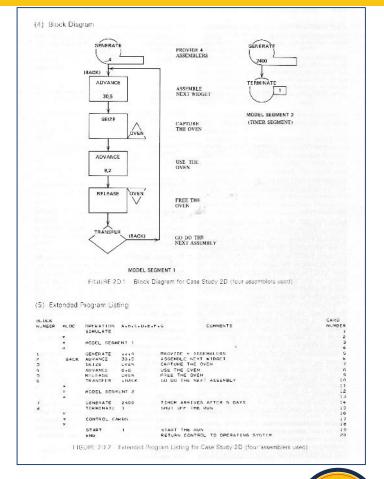
- This year we have a great history track that celebrates the mythology & legends of WSC.
- But let's take a moment to salute GPSS:
 - GPSS = General Purpose Simulation System
 - Came free with IBM mainframes.
 - Had an integer clock, a uniform distribution, you drew the network diagram by hand, and you ran it from cards.
 - But without GPSS we might be at a machine learning conference today.



GPSS!

- Scanned from the famous textbook Simulation Using GPSS, Wiley, 1974.
 - Written by the famous Tom Schriber!

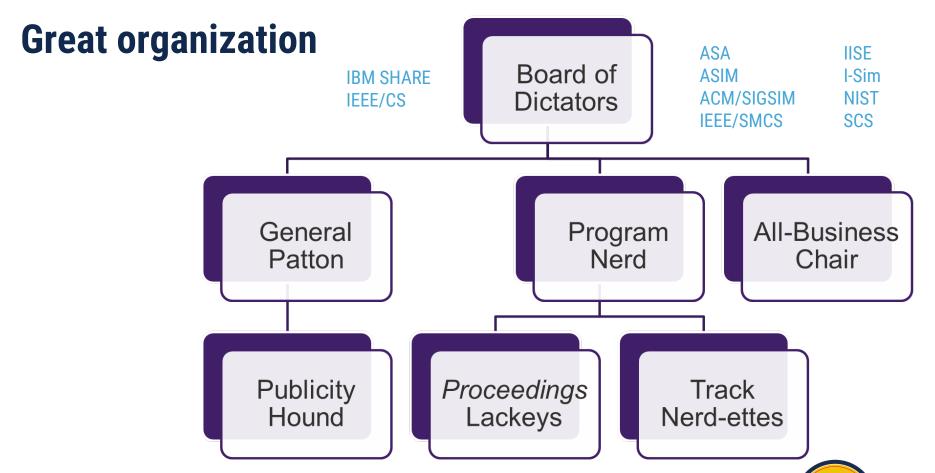
Case study 2D: One-line, oneserver queueing system with feedback.



Fun facts! (about the first conference)

- Registration was \$30 in advance, \$35 at the door.
- Cost for a hotel night was \$18 (in New York!).
- 401 attendees.
- Conference had to borrow \$700 to get off the ground.
- There were no Proceedings.
- All the organizers were smart, all the attendees were good looking, and all the talks were above average.
- I believe we are still here because of great *organization*, great *talks*, great *cooperation* and great *relevance*.





Great talks

- "Only Wet Babies like Change" 2003 keynote
- "The 'Tell-Us-the-Answer-You-Want' Problem" 1983
- "Better Simulation with COBOL"

Fake news

- "A Simulation of the Operations of an Illegal Numbers Game"
 1974
- "What to Simulate First: The Chicken or the Egg?"
 Fake news
- "Linda Arouses a Sleeping Barber" 2002



More great talks

- "Simulation, Drugs and Rock & Roll"Fake news
- "When is a Satellite not a Toaster?" 1991
- "But, Mr. President Is It ETHICAL?" 1986
- "Lose all the Variance You Want without Being (data) Hungry"
 Fake news
- "The Daughter of Celia, the French Flag, and the Firing Squad" 1973
- "Training Socialist Managers by Enterprise Simulation"
 1974



Great cooperation



$$E_{\psi}[g(X)]$$

$$= \int g(x) \frac{\psi(x)}{\varphi(x)} \varphi(x) dx$$









SAS® Simulation Studio



>>> def car(env):

... while True:

... print('Start parking at %d' % env.now)

... parking_duration = 5

... yield env.timeout(parking_duration)

• • •

... print('Start driving at %d' % env.now)

... trip_duration = 2

... yield env.timeout(trip_duration)



With great RELEVANCE comes great responsibility

- WSC 2013 Titan's Talk: "The Simulation Curmudgeon"
 - 1. Why do we build simulation models as if they will never change?
 - 2. Why do we treat simulation like poor man's queueing theory?
 - 3. Why do we fit input distributions like it's 1922?
 - 4. Why can't I talk to my simulation?



- 1. Simulation analytics
- 2. Parallel simulation
- 3. Simulation to support decisions
- "Technology Transfer of Simulation Analysis Methodology: One Man's Opinion," WSC 2016.









The World has changed more than WSC has

Then...



- Batch runs on a single processor with limited memory and storage.
- Simulation imitates queueing theory: long-run average performance of stationary systems for *design*.
- Data gathered as needed.
- Modeling & simulation viewed as a technical competency.

Now...



- Rent 1000s of processors; store and search vast quantities of results.
- Support needed for real or near-realtime decisions, possibly made algorithmically, for *operations*.
- Data gathered all the time.
- "Analytics" viewed as a core competency.

TAKEAWAY MESSAGE

To remain relevant, we need better answers to harder problems in a more timely way.

Some challenges...

- 1) All models are wrong, but some models are wronger
- 2) Simulation: the glitter or the glue?
- 3) All the world's a database
- 4) Mutually assured relevance

CHALLENGE #1

All models are wrong, but some models are *wronger*

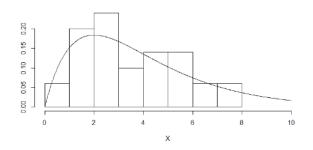
George Box said...

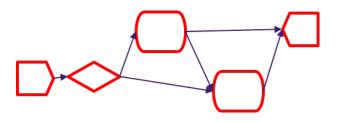
- "Remember that all models are wrong; the practical question is how wrong do they have to be to not be useful?"
 - From his book Empirical Model-Building and Response Surfaces
- We often quote this to justify what we do.
 - But did we take the wrong message from Box?
- "... how wrong do they have to be to not be useful?"
 - Box's statement is not a blanket license to use models.
 - "Better answers" requires a better idea about "how wrong."



Simulation = Inputs + Logic

- The inputs are the lowest level of unexplained variation.
 - Modeled by distributions, often estimated from data.
- Everything else is logic: the "art" part that is true-enough or not, right level or wrong level.
 - Tocher (1963) The Art of Simulation.

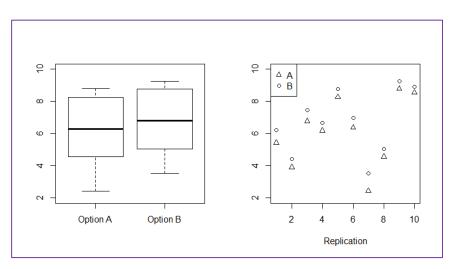




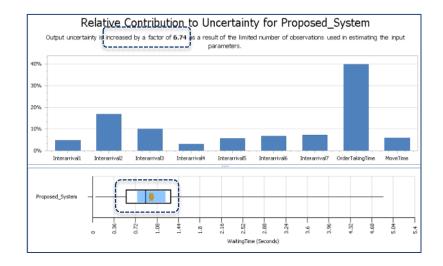


Thinking about "how wrong"

 Old Idea: Common random numbers and estimating differences.



New Idea: Model risk & inference to the resolution of the model.



Old school simulation modeling



Inputs & Logic revisited

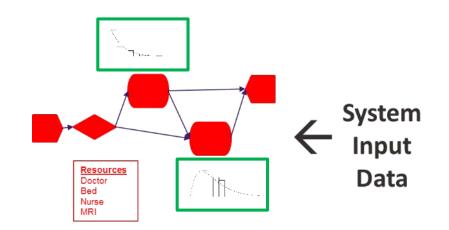
- The Sloan Valve video is really just a big dataset.
- Instead of (younger) me with a notepad, the "logic" could be "learned" from dissecting the video.
 - Fixed objects, dynamic objects, what follows what and with what regularity.
 - Analysis of hours of video would reveal inconsistencies, worker differences, rare events, etc. that I would never observe.
 - When "logic" is like a statistical model, then I can infer "how wrong!"
- Did I just throw us under the machine learning bus?



Is simulation just "machine learning?"

- Question: What is the essential difference?
- **Answer**: The meaning of x.
 - We observe at most one x.
 - We need to model what happens at new x's.

minimize
$$\sum \{\boldsymbol{\beta}^T \boldsymbol{f}(\boldsymbol{x}_i) - y_i\}^2$$

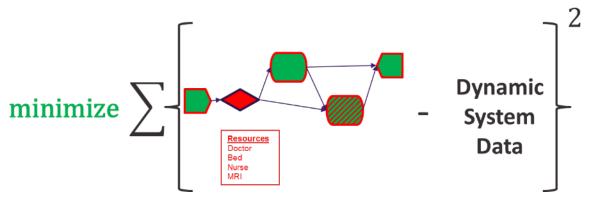


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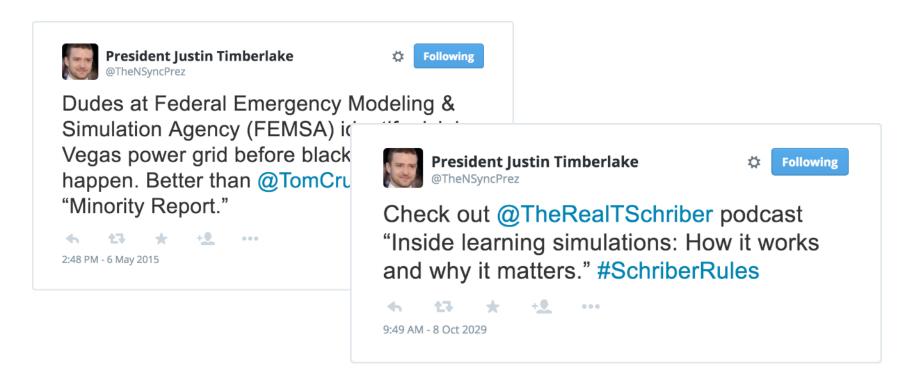


"Learning" our simulations

- Statistical models can be learned from data.
 - Push as much of the simulation model as sensible to being an "input."
- But we are interested in more than the observable I-O relationship.
 - ullet Embedded in the data is a *control* $oldsymbol{x}$ that we want to *change*.
- The "art" part is what a change in x will do to the I-O relationship.
- Our value add:
 - Modeling focused on the impact of changes.
 - Fully recognizing uncertainty to generate robust, defensible solutions.



Tweets from our future: 2029



CHALLENGE #2

Simulation: the glitter or the glue?



Think BIG

Many of society's most important challenges — healthcare, global terrorism, income inequality, world food supply, power distribution, unfiltered tweets — are systems-of-systems problems.

- What is our role in solving these "harder" problems?
- Strategies that seem doomed to fail:
 - One big, fully detailed model of everything.
 - A simple, stylized model that provides "insight."
- "Models of models" seem to be required.
 - Should simulation be the glitter or the glue?



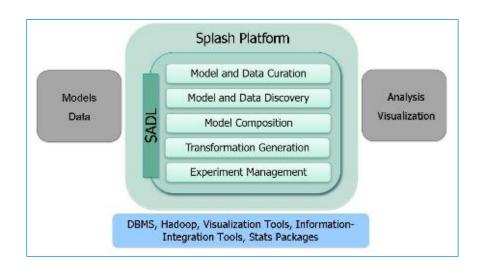
An argument for glue...

- We naturally account for time.
- Level of detail can be variable and our choice.
- Uncertainty matters; if it can be modeled then we can include it.
- We do realizations, and realizations parallelize.
- We have a history of combined discrete-event, continuous-state, agent-based, and hybrid modeling.
- Simulation is arguably the least restrictive modeling & analysis paradigm.



Cool proof-of-concept: IBM Splash

- "The Splash project provides a prototype platform for combining existing heterogeneous simulation models and datasets to create composite simulation models of complex systems, thereby facilitating cross-disciplinary modeling, simulation, and optimization."
- The Splash philosophy is to take the world as it is (i.e., lots of individual models) and to solve the assembly problem.



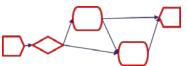
Haas et al. "Splash: Simulation Optimization in Complex Systems of Systems," Fiftieth Annual Allerton Conference, UIUC, 2012

Why are we "not yet ready for prime time"

- We are not good at policy optimization or system control.
- We typically think in terms of a single-resolution model and one time scale, that tries to get everything right.
- Computation does matter ("computationally tractable").
- Our dominant model-building paradigm does not easily scale up to large, data-driven, parametric, multi-resolution models.
 - Watch closely as the keynote speaker now commits professional suicide.

Drag & drop 'til you drag & drop

- Drag and drop with embedded animation has been good to us, but going forward it is a **bottleneck**.
 - Glitter!



- Models connected to the real world
 - Call out to the world & "symbiotic simulation"
- Aggregate & disaggregate time, space and data
 - Input and output to multiple models
 - "Believe your limit theorems" PW Glynn
- Retain as much I/O data as possible

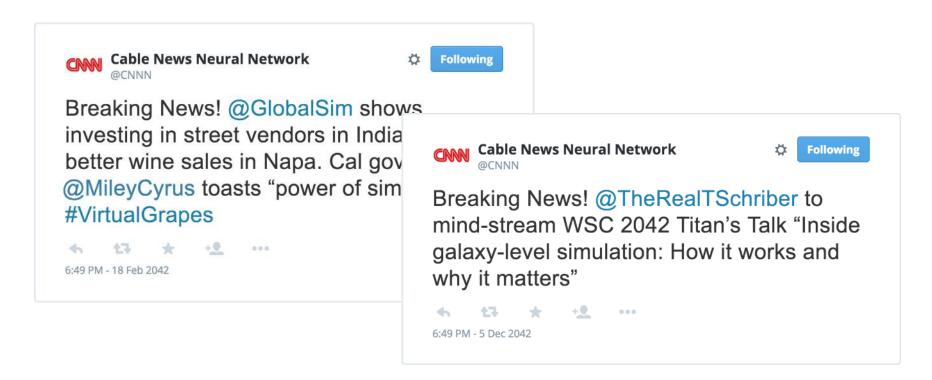
- More data-driven input and <u>structure</u>
 - Relationships that build the models, not just models that build the relationships
- Models easily created, updated & explored
 - Easy to change; easy to vary
 - Reference models rather than analysis models

 But Python is not the answer either.





Tweets from our future: 2042

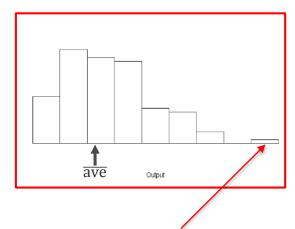


CHALLENGE #3

All the world's a database

What's wrong with this? A thought experiment

- Think about something important that you would simulate.
 - Big is bad; small is good.
- You run some number of replications of one scenario.
- You look at the average and a histogram of performance.
- What is the first question that comes into your mind when you see this?



- "Why did that happen?"
 - Do we actually ask this question?
 - Would we know how to find the answer?



Why the world loves "the machine"

- Conditional statements are useful for making money:
 - **Unconditional**: Customers spend \$87.12 on average vs.

Conditional: A university professor from the Midwest spends \$1.35 on average

- "Big Data" is less about quantity and more about coverage of a space of covariates.
- Unconditional has been ok because design is more forgiving than operations.
 - For macro changes the "main effects" matter, but control is about "dynamics."
- If we want to be taken seriously in the future, then we have to take our simulation-generated data seriously too.

Taking simulation data seriously

- Last year Titan Susan Sanchez talked about data farming:
 - Experiments to explore huge factor spaces of complex systems.
 - Recognizes our differences from "field data" analytics:
 - We can design the experiment to grow the data.
 - High-performance computing ≠ Infinite computing.
 - And we can go back for more.



Naval Postgraduate School

Making simulation runs you don't "need"

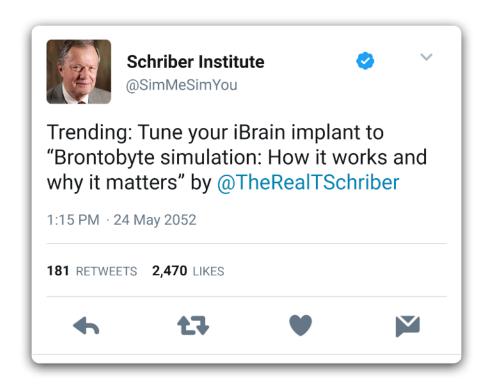
- Proposition: One purpose of a simulation is to fill a database with detailed, relevant, dynamic observations of systems that do not yet exist to be mined for insight & solutions as needed.
- Translation: Simulation everywhere, all the time.

- Same decision made repeatedly in evolving conditions.
 - Forecast-driven designs
- Very expensive simulations covering a big factor space.
 - Low-discrepancy/low-uncertainty designs
- Want to decide if real-world experiments are worthwhile.
 - Designs that mimic real experiments.
- Detect trouble before it happens.
 - Designs that seeks danger close to where we are now



Tweets from our future: 2052





CHALLENGE #4

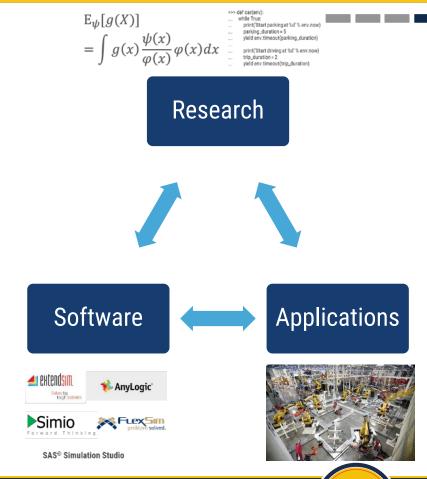
Mutually assured relevance



Mutually assured relevance

Benchmarks

- Optimization and Statistics are two communities that are very similar to us.
- Yet the connections seem better for them.
- Why?



Statistics

The competitive advantage comes from providing deeper and more-robust insight than others can provide.

- Not concerned with the creation of data.
- Vendors compete with free options that have extensive and continually evolving public libraries to provide better answers.

- Implications
 - Need to implement near the cutting edge.
 - Worth employing methodologists; worth it to methodologists to work with data & software.
 - Intensive computation is OK.











Optimization

A universally agreed upon definition of "the problem" allows the model to be separated from the solver.

Optimize f(x) such that $x \in C$

- Implications
 - Competition is on better solvers for harder problems.
 - Connection to latest OR & CS research is a competitive advantage.
 - Intensive computation is expected.









What about the good guys (simulation)?

 The focus has been on generating the right data (including animation) not analysis of a common model. Researchers pretend that...



 Users understand modeling takes time, but can be impatient if the simulation itself does. It is a competitive advantage for **vendors** to differentiate based on better modeling.

It takes a SimCity

- "Better answers to <u>harder</u> problems in a more <u>timely</u> way" won't happen separately in each group.
 - **Researchers**: Machine learning research involves real (i.e., hard) data sets, but we don't often work with real simulations.
 - **Vendors**: R, Python, etc. succeed because "better answer" has become more important than "easy to use."
 - **Users**: Ad hoc modeling & simulation methods really can produce misleading results that underachieve your objectives.
- The good news is that all the key players are at WSC, every year.



Tweets from our future: 2067





So what <u>are</u> the chances?

- "To a large degree, it [simulation] developed as a counter element in the emerging computer culture a tool for support in an area that aimed toward the practical rather than the theoretical, and toward the complex rather than the simple."
 - Julian Reitman, 25th Anniversary Panel Discussion, WSC 1992.
- Computing, data & acceptance of scientific management work in our favor.
 - Provide robust, credible solutions that comprehensively hedge risks.
 - Be the foundation for modeling the most difficult societal problems.
 - Produce data analytics before there are data.
 - Deliver theory in support of practice, and practice in support of theory.
- We do all that and our chances are pretty good.



THANK YOU

SLAVE LABOR

Yujing Lin • Linda Pei • Mark Semelhago • Eunhye Song

SAGE WISDOM

Paul Fishwick (modeling guru) • Robert Sargent (history spoilers)

Dave Goldsman (joke tester) • Russell Barton (ego deflator)

TEST MARKETS

2017 INFORMS Simulation Society Workshop: "Toward an Ecosystem of Models and Data"

2017 BIRS Workshop: "Future Research Directions in Digital Simulation Methodology for the Next 10 Years"

Supply Chain Optimization Tech Talk @ Amazon

INSPIRATION

Peter Glynn • Peter Haas • Jeff Hong • Leon McGinnis Shane Henderson • Henry Lam • Susan Sanchez And the many who have given their time and talents to WSC

ACKNOWLEDGEMENTS