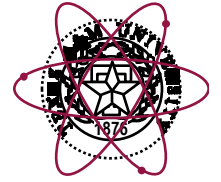


New Faculty Introduction NUEN Advisory Council Presentation

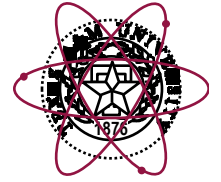
Ryan G. McClarren

Bio (briefly)



- ◆ I received my PhD. in 2007 from the University of Michigan.
 - 2004 MSE and 2003 BSE in Nuclear Engineering from U of M.
- ◆ In graduate school I worked on numerical methods for radiation transport simulations
- ◆ I then was a postdoc and later a staff member at Los Alamos National Laboratory
 - In the computational physics group (CCS-2)
 - Here I also worked on radiation transport as well as high-performance computing
- ◆ I came to A&M as a Visiting Assistant Professor in Dec. 2008.
 - Besides being in NUEN I was also a fellow in the Institute for Applied Mathematics and Computational Science.

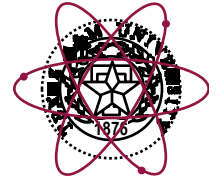
The most interesting piece of my biography



- ◆ In 2004, while I was in graduate school, I appeared on the TV game show *Jeopardy!*
- ◆ Unfortunately, at this time Ken Jennings, the most successful contestant in *Jeopardy!* history was on his streak
 - I end the show in 3rd place with \$1.
 - I did get a Daily Double correct though.

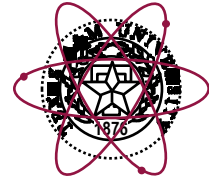


What I do - Research



- ◆ I am active in two areas of research both in computation
- ◆ Radiation transport:
 - Neutronics
 - 4 papers (*Nuc. Sci. & Eng, Transport Theory and Stat. Phys., Annals of Nuc. Energy*)
 - X-ray transport methods
 - 12 papers (*J. Comp. Phys, TTSP, J. Quant. Spec. Rad. Transf., Physics Letters A, SIAM J. Scientific Computing*)
 - High-Energy Density Physics/inertial confinement fusion
 - 6 papers (*JQSRT, Phys. Plasmas, Fusion Sci. and Tech., High-Energy Density Physics*)
- ◆ Uncertainty Quantification
 - Predictive science
 - 4 papers (*Reliability Engineering and System Safety, Annals of Nuc. Energy*)
 - Propagation of uncertainties
 - 1 ongoing project regarding uncertainties in LOCA's

What I do - Teaching



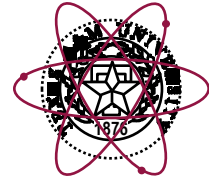
◆ Undergraduate

- NUEN 301 – Reactor Theory (Fall 2011)
 - 2 of 3 sections this term (56 out of 84 students)
- NUEN 304 – Reactor Analysis (Spring 2010/2011)
 - ~50 students each term

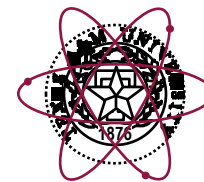
◆ Graduate

- I've developed an uncertainty quantification course.
- Originally taught in Fall 2009 in the Statistics dept.
- Scheduled to be taught this Spring in NUEN.

UQ Course Topics

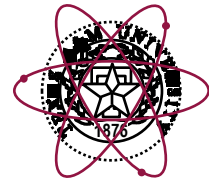


- ◆ Verification/Review of numerical approximations (3 lectures)
- ◆ Validation Data (2 lectures)
- ◆ Uncertainty Quantification
 - Prob/Stats preliminaries (1 lecture)
 - Perturbation / first-order sensitivity
 - Sampling methods (2 lectures)
 - Reliability methods (1 lecture)
 - Polynomial Chaos/Collocation methods (2 lectures)
- ◆ Surrogate-based Methods
 - Linear regression (1.5 lectures)
 - Bayesian statistics (1 lectures)
 - Markov Chain Monte Carlo sampling (1 lecture)
 - Gaussian Process Regression (1.5 lectures)
 - MARS (1.5 lectures)
 - Applications of surrogates (1.5 lectures)
- ◆ Calibration and Prediction
 - Calibration methods (2 lectures)
 - Predictive models (2 lectures)



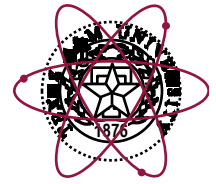
What is Predictive Simulation?

Computation is here to stay.

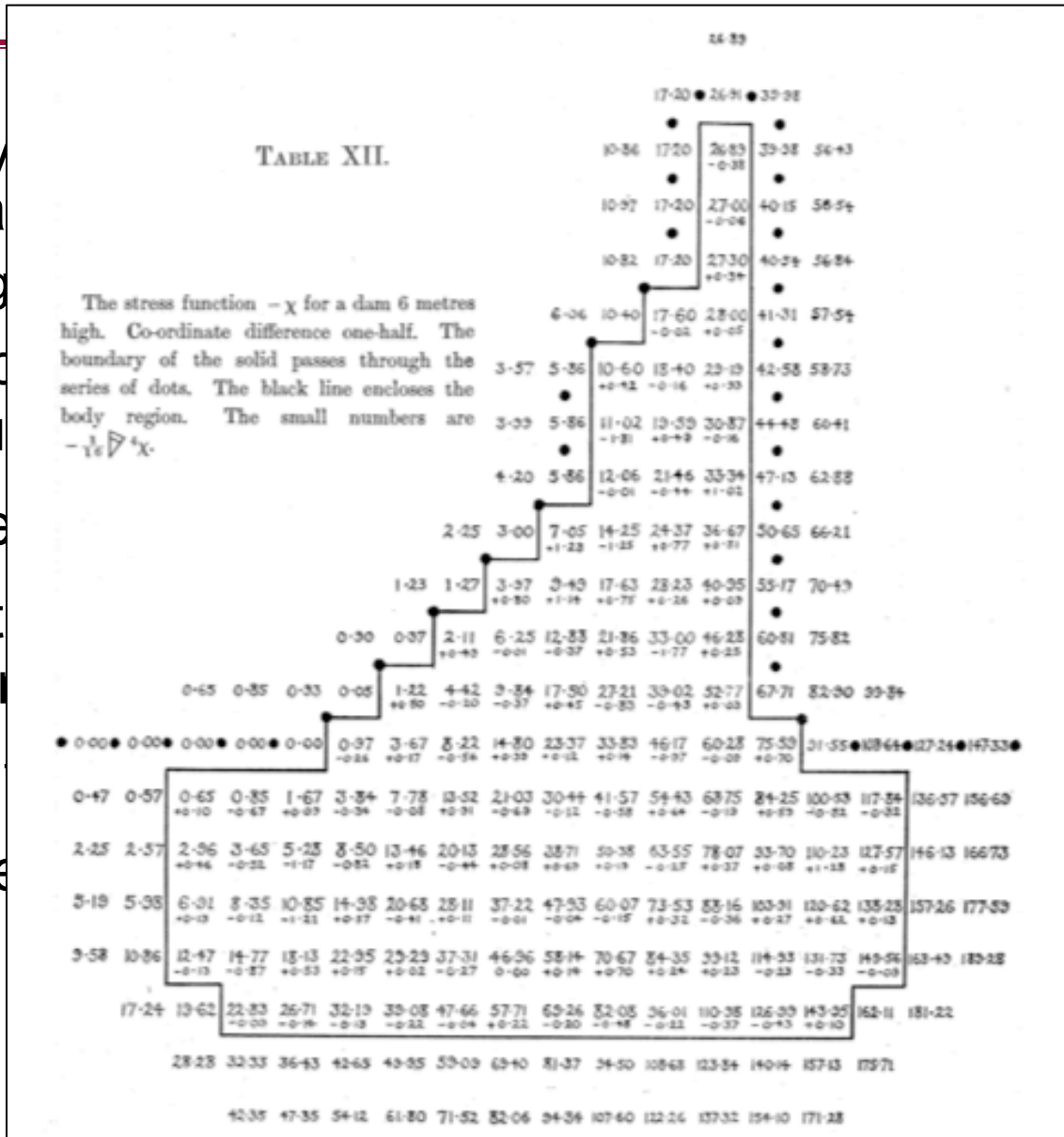


- ◆ Whether you call it simulation, scientific computing, computational science and engineering
 - It will be an important part of the scientific process in the future.
- ◆ We can't measure everything and theory can only go so far
 - In a very narrow view, computational science tries to connect these dots.
 - This has been known for years in neutronics analysis: one can only measure the scalar flux at so many points in the reactor and analytic diffusion theory is severely limited.
- ◆ The uses of computation have evolved as computational horsepower has increased.
- ◆ Initially, computation was just a way to get solutions to analytically intractable equations.
- ◆ Later, discoveries were (and still are being) made using computation.
- ◆ Eventually, computers were used to guide the design of systems
 - Relying on prototypes and experiments to tune the codes.
- ◆ Today, we seek to use computation to predict the behavior of a system that
 - Can't tested by a full-scale experiment (due to safety, cost, or politics)

The Rise of Computer Simulation



- ◆ An early theory and experiment
- Though
- ◆ L. F. Richardson's theory of truncation error in the solution of PDE's using finite differences
- ◆ Predicted that the error would be less than 10⁻¹⁵
- ◆ Finite difference methods were used to solve the problem, with
- ◆ Unappreciated

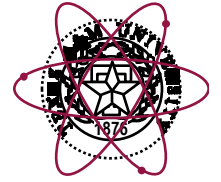


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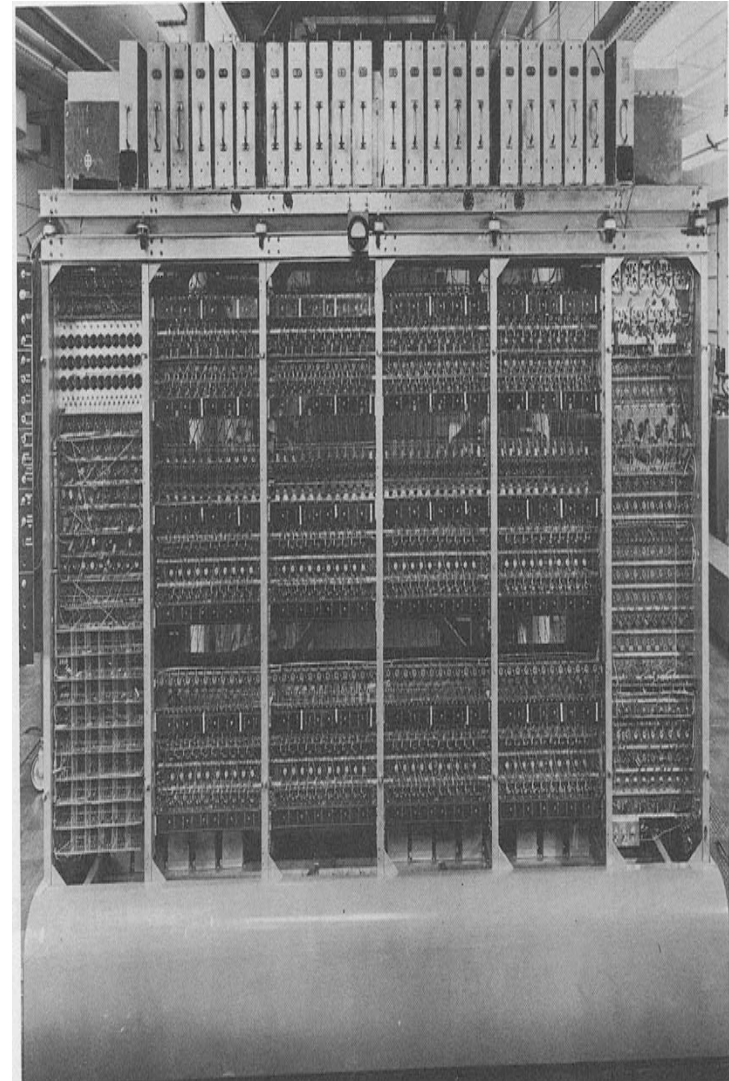
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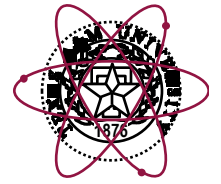
Fermi-Pasta-Ulam Experiment



- ◆ Discovery through computation.
- ◆ Simulated a series of masses connected by linear and nonlinear springs.
- ◆ The masses slide w/o friction along a table.
- ◆ Showed intuition was wrong – nonlinear dynamics is tricky.
- ◆ Used MANIAC computer at Los Alamos National Lab.
- ◆ I couldn't find how long the simulations took.
- ◆ MANIAC: ~5000 FLOPS
- ◆ PlayStation 3: 218 GFLOPS

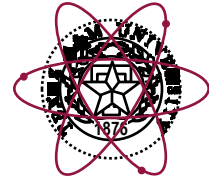


Moving From Discovery to Prediction

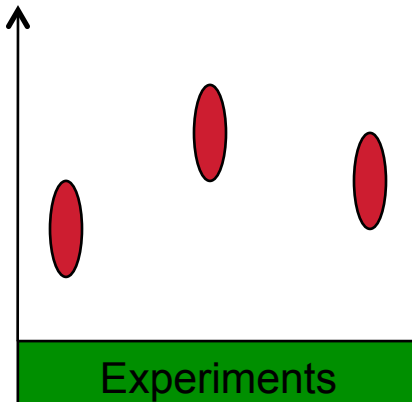


- ◆ The emergence of computation as a field has led to engineers and scientists to ask the question: If we can...
 - Minimize and understand numerical error in our computations, (verify that our codes work)
 - Build confidence in our models using, for example, small-scale experiments (validate our models for particular situations)
 - Understand and measure the effects of uncertain parameters in our simulation (quantify the uncertainty (UQ) in our calculations)...
- ◆ We can predict, with quantified and qualified uncertainties, the behavior of a system under conditions inaccessible to experiment.
- ◆ We can attempt to answer this question today because of the maturation of the fields of
 - Numerical analysis and computer science
 - Computational physics
 - Statistics

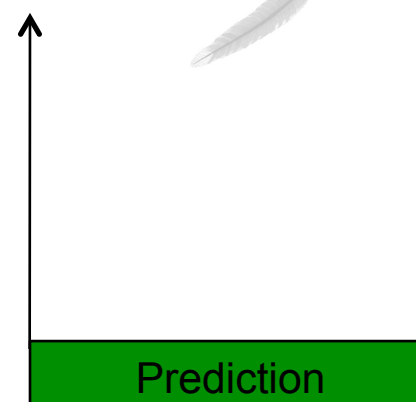
Predictive Simulation needs humans.



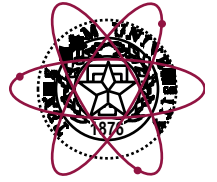
- ◆ Even if I develop the greatest UQ methods and software,
 - I cannot use them on any problem and get results that are predictive or useful.
- ◆ We need domain scientists (experts) to answer questions such as
 - Is the system I am predicting “nearby” systems that I have experimental data?
 - Are we near a physics cliff?
 - A different regime where we need new “physics” to describe the system
 - What are reasonable ranges for the uncertain parameters?
- ◆ What if a I drop my pen from shoulder height?
 - Use that data to “validate” the model



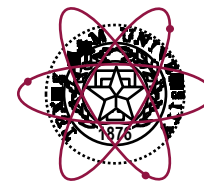
$$Time = \sqrt{\frac{2h}{g}}$$



Predictive Science is a Growing Research Area

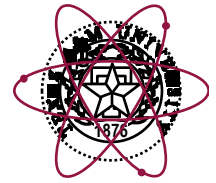


- ◆ All aspects of computational science need further research, for example in
 - Verification: How can we demonstrate that large, multiphysics codes are giving the “right” answers
 - Validation and Data Assimilation: How can we draw conclusions for small-scale or single physics simulations to understand model error
 - Uncertainty Quantification: Given all the uncertainties in system how can we identify the important uncertainties and assess their impact
- ◆ Other important of open questions (“Science Based Nuclear Energy Systems Enabled by Advanced Modeling and Simulation at the Extreme Scale,” DOE workshop)
 - Coupling predictive simulations: dynamic PRA
 - Quantify probabilities of rare outcomes
 - Quantify uncertainties after extrapolations
- ◆ Many large projects
 - Predictive Science Academic Alliance Program– 5 x \$17M centers (DOE/NNSA)
 - Consortium for Advanced Simulation of Light Water Reactors (CASL) - \$25M year collaboration for simulation of nuclear systems
- ◆ Important in all projects with a computational aspect

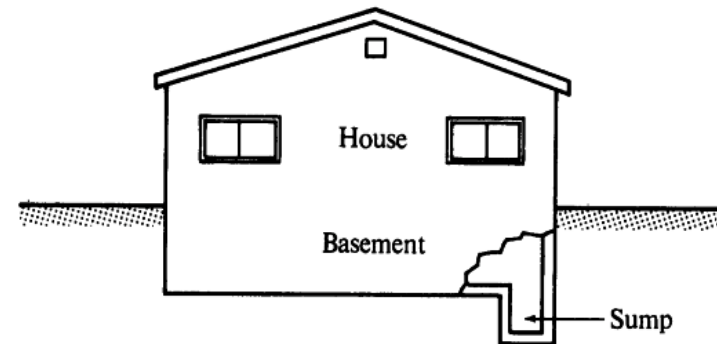
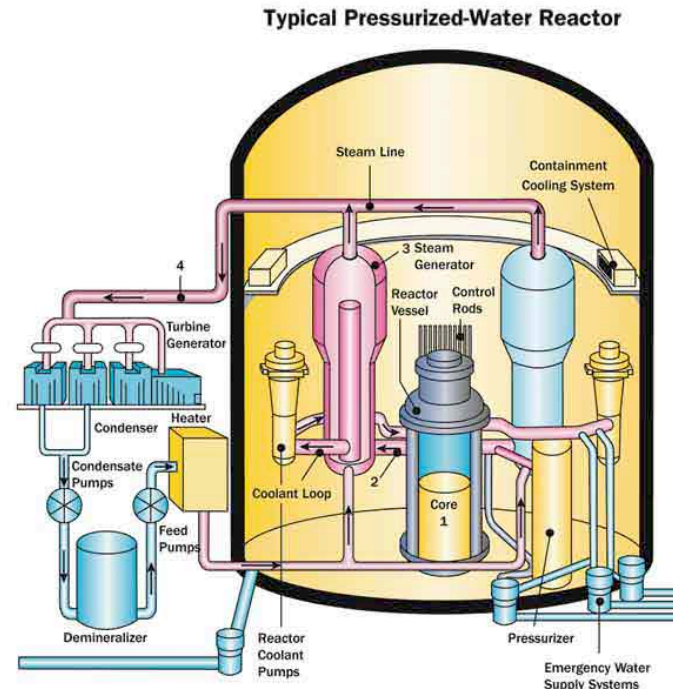


Sensitivity Analysis and Uncertainty Quantification for Loss of Coolant Accidents (LOCA)

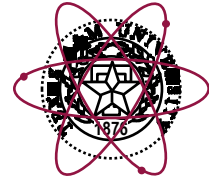
Emergency Core Cooling in a LOCA



- ◆ Consider the situation in reactor where a pipe in the primary loop breaks and leaks coolant.
- ◆ In order to keep the reactor core from melting (bad), the water that leaked will need to be re-circulated through the core
 - After backup sources of water are depleted.
- ◆ This is accomplished through sumps in at the bottom of containment
 - Water is pumped from the sump back into the core

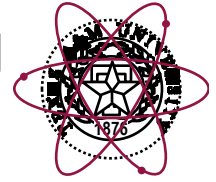


How might the pumps fail?



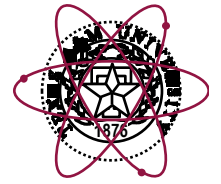
- ◆ The pumps that recirculate water to the core need to be reliable
 - Redundant pumps are installed to deal with this.
- ◆ A real risk is the debris clogging the pumps, therefore screens are installed to keep debris out of the pump.
- ◆ Of course, if the screens get clogged then the pump will lose its suction ability
 - Coolant won't flow to the core, and core damage will result.
- ◆ The debris can come from several sources
 - Insulation on the pipes
 - Concrete dust
 - “Latent debris”
- ◆ The size of the break will influence the amount of debris.

An event in Sweden demonstrated that clogging could be a problem



- ◆ The Barseback event in 1992 had sump screens clog in a BWR.
- ◆ While the reactor was coming back from a shutdown, a relief valve mistakenly opened.
- ◆ The containment vessel spray systems pumps clogged 1 hour into the event.
 - Clogging due to mineral wool debris (220 pounds)
 - The safety analysis for the plant said this wouldn't happen for 10 hours
 - The operators were able to back flush the pumps and clear the debris.
- ◆ The upshot is that the amount of debris produced was grossly underestimated in the analysis.
- ◆ As a result of this incident the NRC investigated clogging at PWRs and BWRs
 - For PWRs they created Generic Safety Issue 191 (GSI-191)
 - GSI-191 has been open for almost 15 years.
- ◆ Interestingly, in 1975 there was a “War of the Worlds”-type radio broadcast in Sweden about a disaster at the Barseback plant.

Debris Generation and Screen Clogging

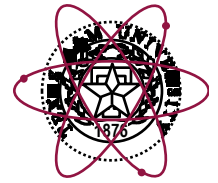


- ◆ When a pipe bursts a jet of water and steam, and perhaps a shock wave, can be generated
 - This will remove and disintegrate insulation around the break location.
- ◆ Certain types of insulation are very good at clogging the sump screens.
- ◆ The jet can hit other pipes or material and create other debris.
- ◆ All of this depends on the size of the break and the location.
- ◆ There has been much experiment and modeling regarding jet formation and debris generation/transport to the screens.

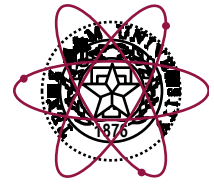


Pictures from insulation debris clogging experiments (NUREG/CR-6762, Vol. 1)

Can we replace the insulation with something that won't clog the sump screen?

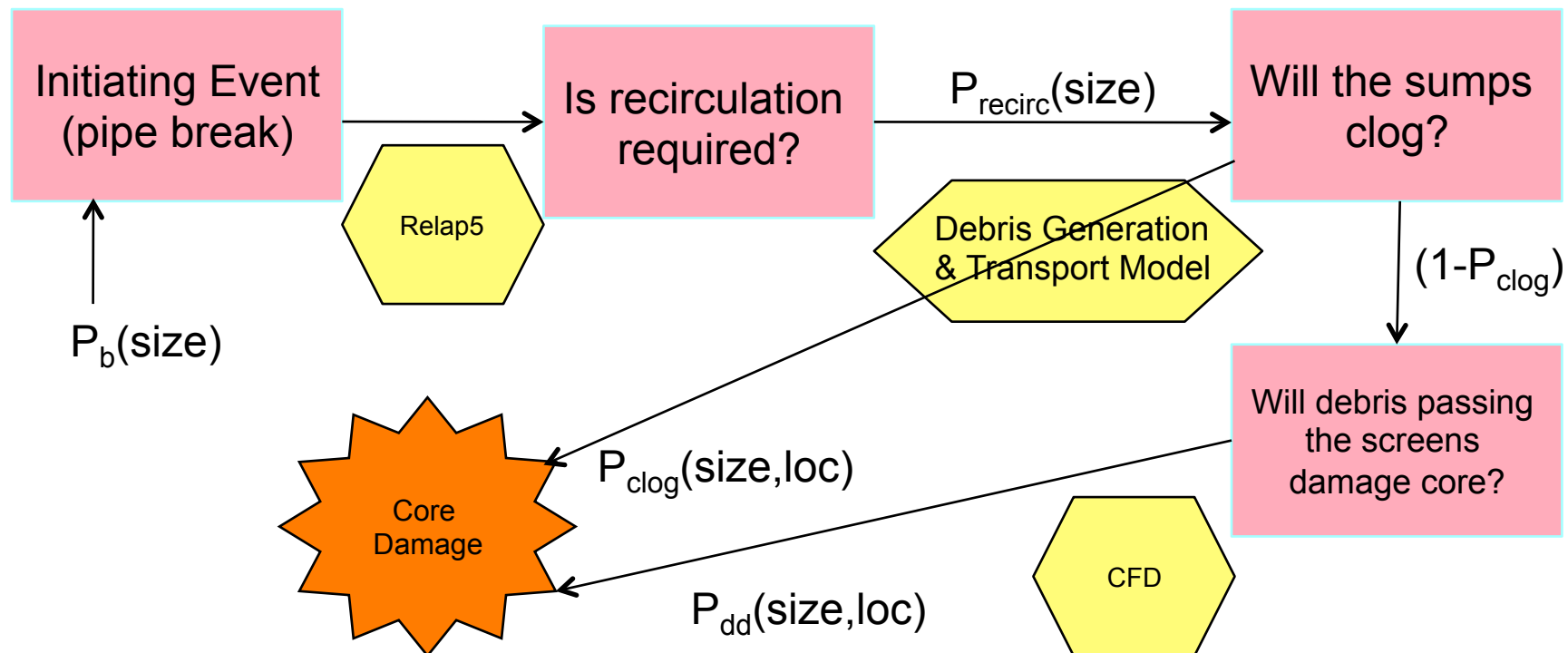


- ◆ Yes, there are insulations that are less susceptible to this problem
- ◆ Changing insulation is not a slam dunk though.
- ◆ Cost is high: \$40 million
 - That money could go to other safety projects
- ◆ Exposure to workers is in the several hundred rem range.
 - Trading a potential risk for a guaranteed hazard.
- ◆ These reasons have lead to a risk informed (PRA) approach to this problem
 - Is there quantifiable reduction in core damage frequency (CDF, aka meltdown risk) by changing the insulation?
 - This is the topic of collaboration between STP, TAMU, tu, LANL, and others.

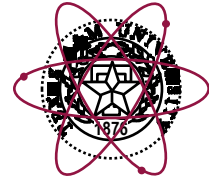


Computation of Core Damage

- ◆ At the end of the day we want to know how the CDF is affected by changing the insulation.
 - If this effect is small, then replacing the insulation does not credibly affect safety.
- ◆ There are several questions we are trying to answer using computation that will inform the PRA.

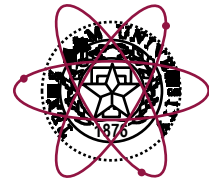


Uncertainties in the Problem



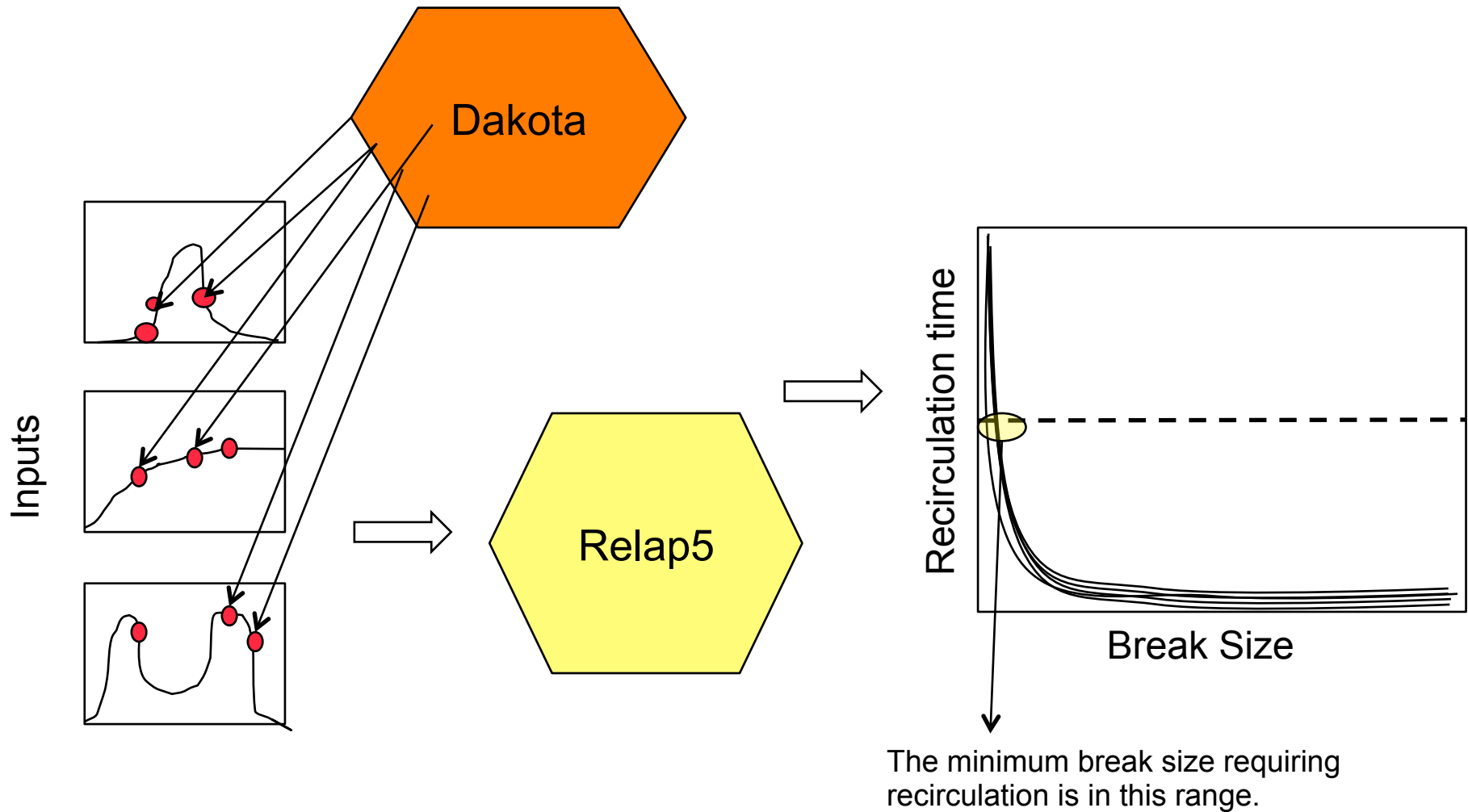
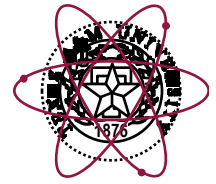
- ◆ There are two types of uncertainty in this problem.
- ◆ One type is inherent randomness (aleatory)
 - Physical variability in the accident initiation
 - Where is the break? How large is it?
 - Uncertainties about the state of the system
 - Exact operating power, water temperature, etc.
- ◆ The other type is due to the fact that we approximate the physics in our calculations (epistemic)
 - The thermal hydraulics models are approximate and they have tuning parameters to account for missing physics.
 - There are constants of nature that we don't know precisely.
- ◆ We must account for both types of uncertainty in predicting the behavior of the system.

Propagating Uncertainty

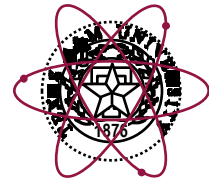


- ◆ In many respects, this is a vanilla UQ problem:
 - Propagate uncertainties through a computer code to find the distribution of outputs.
- ◆ For this particular project we are using DAKOTA, a code from Sandia National Labs.
- ◆ DAKOTA can be taught to edit input files for RELAP5 to vary uncertain parameters and then aggregate the output into a distribution.
- ◆ Also, because RELAP5 does not take a long time to run, we can propagate uncertainties using a Monte Carlo approach
 - Sample from the distributions of the input, and run the code to get a sample from the output.

Propagating Uncertainty

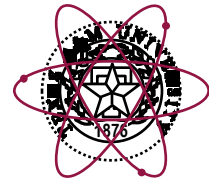


We need more than the minimum break size



- ◆ We also need to give the LANL debris generation and transport model the conditions (flow rate, temperature, etc.) for the jet at the break.
 - Including uncertainties.
- ◆ Then the hard part will be downstream effects.
- ◆ The material passing the sump screens will be very uncertain
 - Particle sizes, composition etc.
 - As well as where it ends up.
- ◆ We will model some of the effects with computational fluid dynamics codes to see if, for example, a coolant channel gets blocked.
- ◆ The CFD codes won't be as fast as RELAP5, so Monte Carlo won't necessarily work well.
- ◆ The results from CFD will inform RELAP5 simulations of system response.

The end result



- ◆ At the end of the day all of these calculations will be rolled up into a PRA calculation
 - To see what the CDF due to sump failure is with both types of insulation.
- ◆ This will be the first such analysis to incorporate uncertainties in the thermal hydraulics modeling into a PRA calculation.
- ◆ The conclusions reached, positive or negative, will affect many PWRs in the US.