High Energy & Nuclear Physics: Software Computer Science and Physical Science Collide for the NuSteam Program

Gene Van Buren - Brookhaven National Laboratory - July 12, 2024



About me Dr. Gene Van Buren

- PhD in nuclear physics
- Participated in a variety of physics experiments, but most of my career has been on the STAR Experiment at RHIC
- Calibrations & Productions Leader for STAR
- Co-Leader of STAR's Software
 & Computing Team for 10 years



Programming Languages, Scripting

- Programming : Fortran [almost dead], C, C++, Java, Python Scripting : csh (or other shells), Python, Perl (and PHP), xml, HTML,
- Javascript, Jupyter Notebook
- Frameworks [not a complete list] :
 - Community-wide: PAW [Fortran: dead], Root
 - Smaller: Clara, Fun4All, Root-spinoffs (e.g. AliRoot, root4star, FAIRRoot)
- Parsing/interpreting/compiling
 - On-the-fly?



Root Framework example

- Write one, or a few lines of code directly at the prompt and see the results immediately
- Write a macro with up to a modest number of code lines that may be worth executing repeatedly, and test immediately
- Load compiled shared object libraries and execute C functions and use shared objects and their member functions
 - Use at the Root prompt requires a dictionary and a few standard rules
- Let's go to the command line and try it out...



Data Flow From experiment to publication

- Raw data (see Jeff Landgraf's presentation next week)
- Reconstruction
 - Calibration
 - Production
- Analyses
- Presentations (talks, posters)
- Publication
- Archives



Data Formats From experiment to publication

- Raw data (see Jeff Landgraf's presentation tomorrow) raw data files
- Reconstruction
 - Calibration
 - databases or constants files
 - Production
 - binary data summaries ("DSTs")
- Analyses
 - user data summaries or text files
- Presentations (talks, posters)
 - tables & plots
 Publication
- - selected tables & plots
- Archives
 - all of the above go to long term storage == tape



Reconstruction **Converting raw into physics**

- Calibrations are applied to raw data to get physical measurements ullet
- ulletmeasurements

 Analysis: particle measurements are combined to get statistical measures of actual physics

Physical measurements are quantitatively combined to get particle





Reconstruction Example images from STAR



Fun 3D display: https://www.star.bnl.gov/~iraklic/EventDisplay/gide/?file=gdml/star-complete-2018.gdml&event=/~iraklic/EventDisplay/ Run2019/FXT/Track_20160026_86.json&WorldRef=TPCE&VisLevel=1&trans=1







Reconstruction Example images





Simulations: event generators Understanding the science

- URQMD)
 - No description of detectors
- Pre-experiment:
 - What are signatures we can look for experimentally?
- Post-experiment: \bullet
 - What can explain what we see experimentally?

Animation example: <u>https://www.sdcc.bnl.gov/phobos/Animations/zx_jan122008.mpg</u>

• Execute the physics of a particle or nuclear collision (e.g. Pythia, Hijing,



Simulations: detectors Understanding the experiment

- Design stage: ullet
 - spatial resolution of multiple measurements becomes a single momentum resolution)?
 - Are detectors susceptible to impeding factors (e.g. background) particles, occupancies, blocking materials or fields)?
 - How does reconstruction software perform?

 Frameworks for describing the detectors, and for executing the physical process that occur when particles traverse those detectors (GEANT)

• Are detectors capable of achieving performance needs? (e.g. how



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Simulations: detectors Understanding the experiment

Various GEANT versions of STAR detector components









Simulations: embedding Understanding the experiment

- Extensions to detector simulators allowing for overlay of simulated data (particles or whole events) into real data (real backgrounds, occupancies, statuses)
- Analysis stage:
 - How well did the reconstruction software perform in a real (non-ideal) data environment?
 - What can be learned about systematic uncertainties? (quantify to some degree what it is we don't know)





Simulations: Monte Carlos Understanding complex math

- Simulations often involve a complex system of materials and possibilities
- The possibilities are often factorizable as random probabilities
- Even so, determining expectation values for averages or ranges of any sort can involve daunting ... usually impossible ... integrals of these various probabilities
- In a Monte Carlo simulation, a statistical approximation is found by repeatedly running the experiment and throwing the dice for each random probability
 - Repetition (often in the millions) builds up statistically strong answers.



"It can all be done in software."

Thanks for your time and attention! Good luck!

