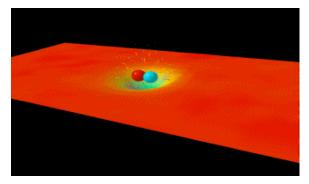
Present and Future Computing Requirements for Jlab@12GeV Physics Experiments[]]

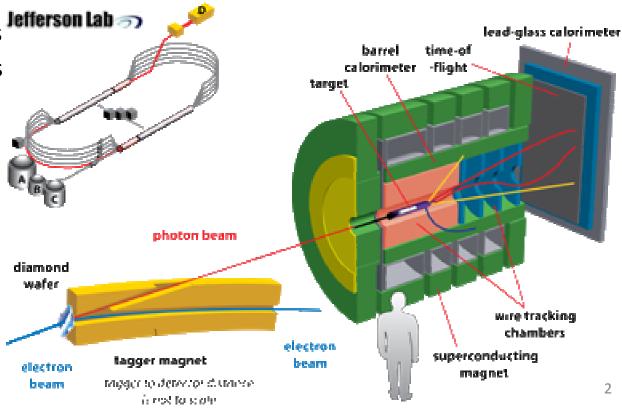
Richard Jones GlueX

> NERSC-NP Workshop, Bethesda, May 26-27, 2011

1. JLab@12GeV Experiments Case Study: Overview

- Halls A, B (CLAS12), C, D (GlueX)
- Hall A nucleon structure and form-factors, New Physics
- CLAS12 parton distributions, hadronization, excited mesons
- Hall C form factors, nucleon structure
- GlueX spectrum of mesons with gluonic excitations
- Present focus is
 - build the detectors
 - debug the analysis
- In the next 3 years
 - install
 - commission
 - be ready for data





1. <u>GlueX – the science</u>

The GlueX Collaboration is building a 12 GeV photon beam line and a dedicated solenoidal spectrometer to study fundamental issues in lead-glass calorimeter time-of barrel strong QCD at Jefferson Laboratory. Our calorimeter -flight primary aim is to identify gluonic resonances target in meson photoproduction by detecting their decays into exclusive final states in a hermetic detector with high acceptance and good resolution for both charged and neutral particles. photon beam diamond wafer

Unambiguous discovery of a multiplet of hybrid mesons will provide answers to long-standing questions regarding how gluonic degrees of freedom are expressed in hadrons.

NERSC-NP Workshop, Bethesda, May 26-27, 2011

1. <u>GlueX – the collaboration</u>

- 15 institutions + Jlab
- ~60 members
- Collab. Board (6) Executive Committee Current spokesperson Curtis Meyer, CMU

Schedule:

- Sept. 2008: **CD3** start of construction
- Dec. 2012: end of 6 GeV Ops.
- 2015: **CD4** start of operations

- University of Athens
- Carnegie Mellon University
- Catholic University
- Christopher Newport University
- University of Connecticut
- Florida International University
- Florida State University
- University of Glasgow
- IHEP Protvino
- Indiana University
- Jefferson Lab
- > U. of Massachusetts, Amherst
- North Carolina A&T State
- U. of North Carolina, Wilmington
- Santa Maria University
- University of Regina

• Algorithms used, Codes, etc.

Simulation
Reconstruction
Analysis

• Algorithms used, Codes, etc.

Simulation

based on Geant3/4

events are independent

code is volatile during initial years => like a glass of beer

results quickly go stale

lots of re-generation

limited need to archive simulations

code more stable as time passes => like a bottle of wine

volume demand increases

more need to archive to satisfy demand

• Algorithms used, Codes, etc.

Reconstruction

custom code developed for each experiment reconstructs the path of particles through the detector events are independent

some parallelism is helpful to conserve memory

static information must be in memory

shared memory can be used

Gluex uses pthreads library

CLAS12 reconstruction components are web services

demanding portions of code hosted separately services can by dynamically provisioned

• Algorithms used, Codes, etc.

Analysis

reduces the reconstructed sample to a selected subset performs some useful transformation on the subset custom code developed by each group mostly based on ROOT single-threaded algorithms reduced data samples can be computationally demanding in special cases model fits to multi-dimensional distributions partial-wave analysis resonance decays

 Quantities that affect the problem size or scale of the simulations (grid? particles? basis sets? Other?)

Simulation

typical event size 10-100 kB production rate 2-3 events/s on a 2 GHz Nehalem core 60% of total CPU time used for simulation scale set by how many real events collected 6000 cores needed in steady state to keep up at maturity, simulation needs 2.5 PB/yr of storage

 Quantities that affect the problem size or scale of the simulations (grid? particles? basis sets? Other?)

Reconstruction

same code is used for both raw and simulated events multiple passes are needed – calibration only one pass through the entire sample cpu limited, mostly because of track finding, fitting 30% of CPU time is devoted to reconstruction reconstructed event record 50% larger than raw option to save "cooked" data + raw (space factor 2.5) option to discard "uninteresting" events (space factor 1.2)

 Quantities that affect the problem size or scale of the simulations (grid? particles? basis sets? Other?)

Analysis – special case of partial-wave analysis
 fits on a single set of data can take many hours on a single processor
 involves evaluation of many identical matrix operations
 literally begs to be parallelized
 several implementations of parallel PWA exist
 using the MPI library – speed-up factors of 100
 using the CUDA platform – speed-up factors of 1000's
 running on commodity NVIDIA hardware
 GlueX plans incorporate GPU's for PWA into analysis infrastructure

2. Current HPC Requirements

- Requirements are not really "HPC", more like "HTC"
- Currently no data no real requirements, except **get ready!**
- Where we are in mid-2011:

Facilities Used or Using	JLAB OLCF ACLF NSF Centers Other: Open Science Grid		
Architectures Used or Using	Cray XT IBM Power BlueGene Linux Cluster GPUs Other:		
Total Computational Hours Used per Year	1,000,000 Core-Hours in 2011, estimated		
OSG, JLab NERSC Hours Used in 2010	50,000 Core-Hours (OSG, not NERSC)		
Number of Cores Used in Typical Production Run	1000		
Wallclock Hours of Single Typical Production Run	50 hours		
Total Memory Used per Run	500 GB		
Minimum Memory Required per Core	1 GB		
Total Data Read & Written per Run	10,000 GB		
Size of Checkpoint File(s)	0 GB (not used)		
Amount of Data Moved In/Out of NERSC-OSG	10 GB per job (interpreted as grid storage)		
On-Line File Storage Required (For I/O from a Running Job)	10 GB and 10,000 Files		
Off-Line Archival Storage Required	0 TB and 0 Files (no need, currently)		

NERSC-NP Workshop,

- Part of commissioning in 2014 is shaking down the Monte Carlo
- Physics Data taking expected to begin in 2015
- Where we need to be in 2016:

Computational Hours Required per Year	80 million
Anticipated Number of Cores to be Used in a Typical Production Run	10,000
Anticipated Wallclock to be Used in a Typical Production Run Using the Number of Cores Given Above	15 hours
Anticipated Total Memory Used per Run	5 million GB
Anticipated Minimum Memory Required per Core	2 GB
Anticipated total data read & written per run	100,000 GB
Anticipated size of checkpoint file(s)	0 GB
Anticipated Amount of Data Moved In/Out of NERSC SRM	10 GB per job
Anticipated On-Line File Storage Required (For I/O from a Running Job)	100 GB and 10,000 Files
Anticipated Off-Line Archival Storage Required	2.5 PB per year

- Upcoming changes to codes/methods/approaches to satisfy science goals
 - Simulation

move from Geant3 (f77) to Geant4 (c++)

algorithms remain fundamentally the same

speculation about future porting of Geant to GPU's

would have significant impact on GlueX

serious impediments are seen

more than 5 years away

Present algorithms are adequate for the lifetime of the experiment, fit within a reasonable scope, especially when resources from member institutions are counted.

• Upcoming changes to codes/methods/approaches to satisfy science goals

Reconstruction

algorithms are still evolving

techniques for tracking, clustering are well established search is for mix of algorithms that works best compared to other solenoid experiments, tracking is slow exploit of SIMD instructions tried, results limited new super-scalar extensions (AVX) could improve it bigger gains probably in algorithm refinements pthreads-based parallelization framework is stable development is proceeding based on simulated events

• Upcoming changes to codes/methods/approaches to satisfy science goals

Analysis active area of current effort by several groups progress impeded by issues with reconstruction

This is where the development of novel algorithms and working modes must occur – hopefully within the next 5 years – if the experiment is to achieve its scientific goals.

Use of GPU's for partial-wave analysis has demonstrated results. Must be scaled up into a full analysis application

application runs interactively on a desktop application is integrated with remote data services remote compute services offload intensive aspects to maintain interactive response

• Upcoming changes to codes/methods/approaches to satisfy science goals

PWA is just one example of interactive access services to large scientific data sets being filtered through a user-supplied filter.

data sets are smaller than the indexable internet ~5 XB (2 years ago) but queries are more complex than "find *Will* next to *Rogers*" however latency tolerance for "interactive use" >> 0.2 s

• Upcoming changes to codes/methods/approaches to satisfy science goals

Such services have started to appear – for physics analysis

PROOF service at CERN (AliProof service, ALICE experiment)

large data store tightly coupled to interactive analysis cluster

users connect through standard root application

user code is same as used for local analysis

commands ("queries") are run in parallel in real time

results are returned in real time to user

Provision of such a service marks a significant shift from traditional scheduling paradigms for shared compute resources.

Introduces a QOS component to resource allocation and accounting, goes beyond the total cpu use and queue priorities of batch systems.

Strategy for New Architectures

- Further advances in the common HPC benchmarks (more FLOPS/processor, network throughput and latency) do not impact the major production workflows of GlueX very much – but they may affect cost of hardware.
- PWA is one outstanding exception, GPU's are an excellent solution.
- Assembling all of the pieces into a functional analysis platform for an interactive user doing a PWA study has further to go.
- Implementation of PWA algorithms on GPU hardware is still in its infancy.
- All work up to this moment has been done on the CUDA platform
- All of the usual bottlenecks in on GPU's affect us limited bandwidth between main memory and GPU memory cost of doing computations in double precision

Strategy for New Architectures

The grid platform provides excellent opportunities to better harness resources at member institutions

- the Virtual Data Toolkit (VDT)
- the grid user community and support infrastructure
- naturally fits with collaborative framework
- provides tools for effectively managing a shared resource
- increases the usefulness of the resource for science
- leverages times of reduced local demand for outside users
- takes advantage of reciprocal behavior
- leverages the investment of HEP in grids
- new opportunities for collaboration
- Two+ years of experience with operating UConn site on the OSG

4. Summary

- Sufficient resources are included in the JLab computing plan to satisfy the minimum requirements for all of the 12GeV experiments, including GlueX.
- What more could we do if we had 5X as many resources?

the experiment is expected to be systematics-limited systematic errors are often limited by Monte Carlo simulation

more Monte Carlo studies

- Ease of use and response time is more likely to be the limiting factor in terms of Monte Carlo studies than available hardware.
- Significant compute facilities at user institutions organized into a grid offers significant flexibility in designing analysis solutions.

Backup Slides

(from Matt Shepherd, IU, GlueX Collaboration, Jan. 2010)

A very very rough estimate of the scale of the problem

Experiment	N_{data}	Namps	10,000 CPU hours/analysis
E852	I x 10 ⁶	30	simple theoretical model
CLEO	I x 10 ⁵	10	doable with approx. 100 2003 CPUs
BES III	I x 10 ⁷	20	
GlueX	I x 10 ⁷	30	

- Fit time scales linearly with statistics and like the square of N_{amps}
- Floating parameters in amplitudes add 1-2 orders of magnitude to time
- More realistic theoretical models add an additional 1-n orders of magnitude to fit time
- To do a first GlueX analysis, we need 10x faster machinery than E852 (done now! ...faster multi-core CPUs)
- To do pioneering, high-statistics meson spectroscopy we want 3-4+ orders of magnitude speed gain
 - Grid model: Can we really use 1,000-10,000 CPUs at one time in a fit?
 - GPUs: Inexpensive way to enhance speed of single box by two orders of magnitude



Parallel Computing

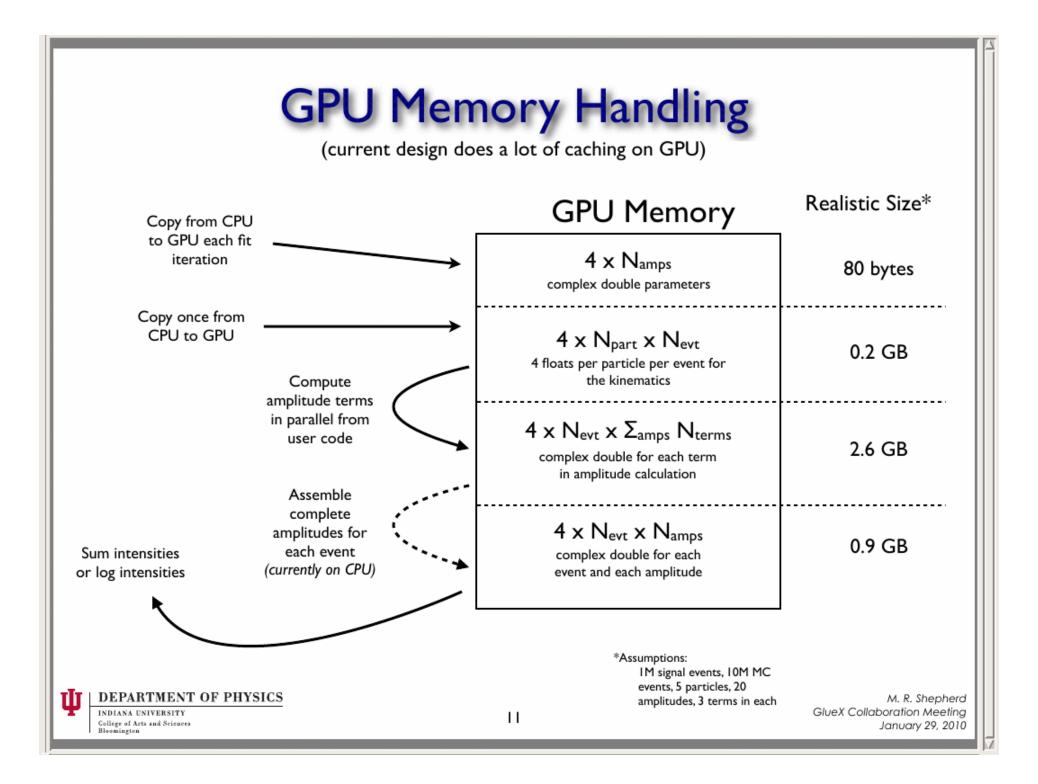
 This type of problem is perfect for parallel computing since all of the large sums over can be done in parts

$$\ln \mathcal{L} = \sum_{i=1}^{N} \ln \left(\sum_{\alpha,\beta}^{N_{\mathrm{amps}}} V_{\alpha} V_{\beta}^* A_{\alpha}(\vec{x}_i) A_{\beta}(\vec{x}_i)^* \right) - \sum_{\alpha,\beta}^{N_{\mathrm{amps}}} V_{\alpha} V_{\beta}^* \int \eta(\vec{x}) A_{\alpha}(\vec{x}) A_{\beta}(\vec{x})^* d\vec{x}$$
$$\int \eta(\vec{x}) A_{\alpha}(\vec{x}) A_{\beta}(\vec{x})^* d\vec{x} \to \frac{1}{N_{\mathrm{MC}}^{\mathrm{gen}}} \left(\sum_{i=1}^{N_{\mathrm{MC}}^{\mathrm{acc}}} A_{\alpha}(\vec{x}_i) A_{\beta}(\vec{x}_i) \right)$$

- Initially each node needs a sub-collection of data or MC and an algorithm for computing the A
- With each fit iteration the node just needs to know the new values of the fit parameters and it returns its contribution to the log likelihood
- Many successful implementations exist... the problem is scaleability



M. R. Shepherd GlueX Collaboration Meeting January 29, 2010



Some First Results

- With most fits we are doing now, using the GPU is like cutting butter with a chainsaw
 - current CLEO analysis uses theoretically complex amplitudes and will soon benefit from being ported to GPU
 - plan to generate high statistics GlueX MC
- Total fit time is not a meaningful benchmark now, compare what we know to be limiting parts of calculation

100,000 $\gamma p \rightarrow \eta \pi^0 p$ events

Calculation	CPU Intel Core i7	GPU nVidia GTX 285
Breit Wigner	90 ms	0.87 ms
Angular Distributions	82 ms	3.9 ms
Σ log(I _i) (GPU includes CPU→GPU memcpy)	371 ms	33.4 ms

(double precision computation on both CPU and GPU)

Very easy factor 10-100 for a single graphics card!

No serious GPU optimizations yet -- use global memory on GPU which is "slow"

Bloomington

M. R. Shepherd GlueX Collaboration Meeting January 29, 2010



- Scalability:
 - New Femi architecture from nVidia (not released yet) plans 448 cores and 6 GB of memory per card
 - Four cards per CPU possible -- cooling an issue (1 kW from GPUs alone)
 - IU fitter already has parallel implementation (via MPI) for cluster running -- will work immediately on multiple GPUs in the same box (1 process per GPU)
 - May soon reach 1000 2000 cores and ~20 GB of GPU RAM in a single box
 - We plan build a small GPU cluster at IU this spring -- size depends on price, but expect a few machines with a few GPUs... maybe in the neighborhood of 5000 10000 cores
- Issues:
 - Need double precision (not a problem with newest hardware), but also need to control
 precision in calculations to avoid problems with MINUIT
 - Debugging is challenging: the classic "cout" debugging method doesn't work with GPU (but emulation is available)
 - Eventually we will need to tune/optimize GPU algorithm, but to do so, we need a big enough problem first!

No foreseeable show stoppers -- a realistic route to 3-4+ orders of magnitude speed needed to maximize physics output from GlueX!

