The Revolution in Experimental and Observational Science

The Convergence of Data-Intensive and Compute-Intensive Infrastructure

> Professor Tony Hey Chief Data Scientist STFC tony.hey@stfc.ac.uk

The Background

X-Info

- The evolution of X-Info and Comp-X for each discipline X
- How to codify and represent our knowledge



The Generic Problems

- Data ingest
- Managing a petabyte
- Common schema
- How to organize it
- How to *re*organize it
- How to share with others

- Query and Vis tools
- Building and executing models
- Integrating data and Literature
- Documenting experiments
- Curation and long-term preservation

Slide thanks to Jim Gray



Slide thanks to Jim Gray

e-Science and the Fourth Paradigm

Thousand years ago – Experimental Science

• Description of natural phenomena

Last few hundred years – Theoretical Science

• Newton's Laws, Maxwell's Equations...

Last few decades – Computational Science

- Simulation of complex phenomena
- Today Data-Intensive Science
 - Scientists overwhelmed with data sets from many different sources
 - Data captured by instruments
 - Data generated by simulations
 - Data generated by sensor networks

eScience is the set of tools and technologies to support data federation and collaboration

- For analysis and data mining
- For data visualization and exploration
- For scholarly communication and dissemination







With thanks to Jim Gray

Artificial Neural Networks



Machine Learning

- Neural networks are one example of a Machine Learning (ML) algorithm
- Deep Neural Networks are now exciting the whole of the IT industry since they enable us to:
 - Build computing systems that improve with experience
 - Solve extremely hard problems
 - Extract more value from Big Data
 - Approach human intelligence e.g. natural language processing



- The change in the Word Error Rate (WER) with time for the NIST "Switchboard" data.
- This shows the dramatic improvement made in the last few years using Deep Neural Networks



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THE INTEGRATION OF EXPERIMENT, BIG DATA, AND MODELING AND SIMULATION INTO INSTRUMENTS FOR DISCOVERIES IN SCIENCE AND ENGINEERING

HPC and Big Data: Better Together!

Kirill Malkin Director of Storage Engineering kmalkin@sgi.com

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Building Systems for Big Data *and* Big Compute

Steve Scott, Cray CTO

Smoky Mountains Conference September 1, 2016

Future Directions for NSF ADVANCED COMPUTING INFRASTRUCTURE

to Support U.S. Science and Engineering in 2017–2020

> The National Academies of SCIENCES • ENGINEERING • MEDICINE

Report of the DOE Workshop on

Analysis, and Visualization of Experimental and Observational Data The Convergence of Data and Computing



Statistics, Data Mining, and Machine Learning in Astronomy

PRINCETON SERIES IN MODERN OBSERVATIONAL ASTRONOMY

A Practical Python Guide for the Analysis of Survey Data

Żeljko Ivezić, Andrew J. Connolly, Jacob T. VanderPlas & Alexander Gray Statistics for Biology and Health

Thomas Hamelryck Kanti Mardia Jesper Ferkinghoff-Borg *Editors*

Bayesian Methods in Structural Bioinformatics



Urheberrechtlich geschütztes Material

Data Science and the UK Science and Technology Facilities Council

UK Science and Technology Facilities Council (STFC)



Big Data and Cognitive Computing: Hartree Centre collaboration with IBM Research

UK Hartree Center Partners with IBM on Big Data

📩 June 4, 2015 by <u>staff</u> 🛛 📔

Today the UK government announced a £313 million partnership with information technology leader IBM to boost Big Data research in the UK.

> We live in an information economy – from the smart devices we use every day to the supercomputers that helped find the Higgs Boson, the

power of advanced computing means we now have access to vast amounts of data," said Minister for Universities and Science Jo Johnson. "This partnership with IBM, which builds on our £113 million investment to expand the Hartree Centre, will help businesses make the best use of Big Data to develop better products and services that will boost productivity, drive growth and create jobs."





Rutherford Appleton Lab and the Harwell Campus



STFC Scientific Computing Department

Applications Division

The Applications Division brings together four groups which develop and apply computational science software packages to solve problems in the physical and biological sciences. The groups are:

- Computational Biology, including structural biology, molecular simulation and bioinformatics
- Theoretical and Computational Physics, including electronic structure of the solid state and surfaces, atomic and molecular physics
- Computational Engineering, focusing on HPC solutions in fluid flow modelling, with particular strength in turbulence and microfluidics
- Computational Chemistry, including molecular dynamics, quantum chemistry and QM/MM techniques, and mesoscale methods

RAL Tier-1

GridPP

HEEK TO

Running Processe.

2019 ast ponch

Week 09

Tier-1 Opening

Meeting national aspirations and international obligations for UK particle physics

> Prof. David Britton Grid IP Project leader University of Glasgow

Collaborative Computational Projects: The CCP's

- Assist universities in developing, maintaining and distributing computer programs
- Promoting the best computational methods
- Each focuses on a specific area of research
- Funded by the UK's EPSRC, PPARC and BBSRC Research Councils



What are CCPs?

The Collaborative Computational Projects (CCPs) bring together leading UK expertise in key fields of computational research to tackle largescale scientific software development, maintenance and distribution. Each project represents many years of intellectual and financial investment. The aim is to capitalise on this investment by encouraging widespread and long term use of the software, and by fostering new initiatives such as High End Computing consortia.

What do CCPs do?

The CCPs enrich UK computational science and engineering research in various ways. They provide a software infrastructure on which important individual research projects can be built. They support both the R&D and exploitation phases of computational research projects. They ensure the development of software which makes optimum use of the whole range of hardware available to the scientific community, from the desktop to the most powerful national supercomputing facilities.

Important Dates





Publications from Work Funded by EPSRC...





CCP	<u>Chair</u>

<u>Title</u>

CCP4	Prof David Brown	Macromolecular Crystallography
<u>CCP5</u>	Prof Neil Allan	The Computer Simulation of Condensed Phases
<u>CCP9</u>	Prof Mike Payne	Computational Electronic Structure of Condensed Matter
CCP12	Prof Mark Savill	High Performance Computing in Engineering
CCP-BioSim	Prof Adrian Mulholland	Biomolecular Simulation at the Life Sciences Interface
CCP-EM	Dr Martyn Winn	Electron Cryo-Microscopy
<u>CCPi</u>	Prof Phillip Withers	Tomographic Imaging
<u>CCPN</u>	Prof Geerten Vuister	NMR
<u>CCP-NC</u>	Dr Jonathan Yates	NMR Crystallography
<u>CCP-Plasma</u>	Dr Tony Arber	Computational Plasma Physics
<u>CCPQ</u> *	Prof Graham Worth	Quantum Dynamics in Atomic, Molecular and Optical Physics
<u>CCP-SAS</u>	Prof Steve Perkins	Analysis of Structural Data in Chemical Biology and Soft Condensed Matter
<u>CCPForge</u>	Catherine Jones	Collaborative Software Development Environment Tool
CCPPET/MR	Dr Kris Thielemans	Positron Emission Tomography (PET) and Magnetic Resonance (MR) Imaging
<u>CCP CoDiMa</u>	Prof Steve Linton	Computational Discrete Mathematics
<u>CCP-WSI</u>	Prof Deborah Greaves	A Collaborative Computational Project in Wave/Structure Interaction
<u>CCPmag</u>	Prof Julie Staunton	Computational Magnetism



Collaborative Computational Project No. 4 Software for Macromolecular X-Ray Crystallography



Home	About CCP4	CCP4 Projects	Downloads	Documentation	Courses	Developers	CCP4 people	WG1/WG2

CCP4 exists to produce and support a world-leading, integrated suite of programs that allows researchers to determine macromolecular structures by X-ray crystallography, and other biophysical techniques. CCP4 aims to develop and support the develop and support the develop and supports the develop and supports the widest possible researcher community, embracing academic, not for profit research. CCP4 aims to play a key role in the education and training of scientists in experimental structural biology. It encourages the wide dissemination of new ideas, techniques and practice.



The Diamond Synchrotron

Diamond Light Source



Science Examples



Casting aluminium



Pharmaceutical manufacture & processing





Non-destructive imaging of fossils

Structure of the Histamine H1 receptor



Data Rates



- 2007 No detector faster than ~10 MB/sec
- 2009 Pilatus 6M system 60 MB/s
- 2011 25Hz Pilatus 6M 150 MB/s
- 2013 100Hz Pilatus 6M 600 MB/sec
- 2013 ~10 beamlines with 10 GbE detectors (mainly Pilatus and PCO Edge)
- 2016 Percival detector 6GB/sec

Cumulative Amount of Data Generated By Diamond



Thanks to Mark Heron

Cryo-SXT Data



Neuronal-like mammalian cell line; single slice

Challenges:

- Noisy data, missing wedge artifacts, missing boundaries
- Tens to hundreds of organelles per dataset
- Tedious to manually annotate
- Cell types can look different
- Few previous annotations available
- Automated techniques usually fail

scientificsoftware@diamond.ac.uk

Segmentation of Cryo-soft X-ray Tomography (Cryo-SXT) data

Data

- B24: Cryo Transmission X-ray Microscopy beamline at DLS
- Data Collection: Tilt series from $\pm 65^{\circ}$ with 0.5° step size
- Reconstructed volumes up to 1000x1000x600 voxels
- Voxel resolution: ~40nm currently
- Total depth: up to 10µm
- GOAL: Study structure and morphological changes of whole cells



Workflow



Data Preprocessing



Raw Slice

Gaussian Filter

Total Variation

Data Representation





SuperVoxels (SV)

SV Boundaries

SuperVoxels:

- Groups of similar and adjacent voxels in 3D
- Preserve volume boundaries
- Reduce noise when representing data
- Reduce problem complexity several orders of magnitude
- Use Local clustering in $\{xyz + \lambda * intensity\}$ space





Data Representation







Local *k*-means in a small window around seeds



946 x 946 x 200 = 180M voxels

180M / (10x10x10) = 180K supervoxels

Workflow



Feature Extraction

Features are extracted from voxels to represent their appearance:

- Intensity-based filters (Gaussian Convolutions)
- Textural filters (eigenvalues of Hessian and Structure Tensor)

User Annotation + Machine Learning



User Annotations

Using a few user annotations along the volume as an input:

- A machine learning classifier (i.e. Random Forest) is trained to discriminate between different classes (i.e. Nucleus and Cytoplasm) and predict the class of each SuperVoxel in the volume.
- A Markov Random Field (MRF) is then used to refine the predictions.

SuRVoS Workbench (Su)per-(R)egion (Vo)lume (S)egmentation



Coming soon: https://github.com/DiamondLightSource/SuRVoS scientificsoftware@diamond.ac.uk

Imanol Luengo <*imanol.luengo@nottingham.ac.uk*>, Michele C. Darrow, Matthew C. Spink, Ying Sun, Wei Dai, Cynthia Y. He, Wah Chiu, Elizabeth Duke, Mark Basham, Andrew P. French, Alun W. Ashton

The ISIS Neutron and Muon Facility



ISIS

- ≈30 neutron instruments
- 3 muon instruments
- 1400 individual users per year making 3000 visits
- 800 experiments per year resulting in 450 publications
- Diverse science
 - Fundamental condensed matter physics
 - Functional materials e.g. multiferroics, spintronics
 - Chemical spectroscopy e.g. catalysis and hydrogen storage
 - Engineering e.g. stress and fatigue in power plants and transportation
 - Solvents in industry
 - Structure of pharmaceutical compounds, biological membranes





Peak Assignment in Inelastic Neutron Scattering

- Vibrational motion of atoms crucial for many properties of a material -e.g., how well it conducts electricity or heat
- Peaks in INS spectrum correspond to specific atomic vibrations
- Peak assignment: what specific vibrational motions of atoms give rise to specific peaks ?



S. Parker and S. Mukhopadhyay (ISIS)

Modelling & Simulation for INS Peak Assignment



- INS spectra can be computed for a given atomic structure
- Calculations allow us to see what specific vibrational motion of atoms occur, and at what frequency



Materials Workbench

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	Welcome to Mantid 3.7.20160809.1348 Please cite: http://dx.doi.org/10.1016/i.nima.2014.07.029 and this release: http://dx.doi.org/10.5286/Software/Mantid	Execute ABINS
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K. Dymkowski

The Central Laser Facility

OCTOPUS Facility in the CLF

- National imaging facility with peerreviewed, funded access
- Located in Research Complex at Harwell
- Cluster of microscopes and lasers and expert end-to-end multidisciplinary support
- Operations and some development funded by STFC
- Key developments funded through external grant – BBSRC, MRC



Example: EGFR cell signalling in cancer

- Driven OCTOPUS single molecule developments
- User in plant cell imaging now catching up in scale of challenge
- Part of a PhD project:
 - 1 experimental technique
 - 50 experimental conditions
 - 30 datasets for each condition
 - 1000 single molecule tracks for each condition
 - Multiple properties & events of interest in each track
 - Comparison of just one property...







With thanks to Dan Rolfe

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Large scale comparisons



With thanks to Dan Rolfe

Multidimensional single molecule tracking







- Automated registration & tracking in multiple channels
 - Computer vision
 - Bayesian feature detection from astronomical galaxy detection
- Instrumental metadata from acquisition
 - Flexible specification of many instrument configurations



With thanks to Dan Rolfe

Rolfe et al 2011, Euro Biophys J, 2011

The JASMIN Environmental Science Super Data Cluster



Rising demand	The Data Commons	Looking Forward	Summary
○○○○○○●○	00000	00000	O
Consequences			

More Data

Fig. 2 The volume of worldwide climate data is expanding rapidly, creating challenges for both physical archiving and sharing, as well as for ease of access and finding what's needed, particularly if you're not a climate scientist.

(BNL: Even if you are?)





Why JASMIN? Bryan Lawrence - RAL, June 2016 Centre for Environmental Data Analysis science and technology facilities council

Large data sets: satellite observations



Sentinel 1A: Launched 2014 (1B due 2016)

- Key instrument: Synthetic Aperture Radar
- Data rate (two satellites: raw 1.8 TB/day, archive products ~ 2 PB/year)



COMET: Centre for Observation and Modelling of Earthquakes, Volcanoes, and Tectonics



(Picture credits: ESA, Arianespace.com, PPO.labs-Norut-COMET-SEOM Insarap study, ewf.nerc.ac.uk/2014/09/02/new-satellite-maps-out-napa-valley-earthquake/)

Rising demand	The Data Commons	Looking Forward	Summary
0000000	00000	00000	0
Where's this coming from? Scientific	Pull underpinned by Technology Push		

Core Science Requirements

		Schematic for Global Atmospheric Model Interest der Lathate Lander Verbar der Tataget an Pessen	Big International Drivers:	aerosol cci cci cci cci cci cci cci cci cci cc
Today:	Observations	Models	GAW	glaciers cci
Volume	20 million = 2 x 10⁷	5 million grid points 100 levels 10 prognostic variables = 5 x 10 9		antarctic ice sheet cci jce sheets greenland
Туре	98% from 60 different satellite instruments	physical parameters of atmosphere, waves, ocean	COPERFICUS Europe's eyes on Earth	iand cover cci
Soon:	Observations	Models		
Volume	200 million = 2 x 10 ⁸	500 million grid points 200 levels 100 prognostic variables = 1 x 10 ¹³	World Climate Research Programme	cci sea ice cci sea level cci
Туре	98% from 80 different satellite instruments	physical and chemical parameters of atmosphere, waves, ocean, ice, vegetation	Friedwarter dentaries Landrar	sst cci
→ Fac	tor 10 per day →	Factor 2000 per <u>time step</u>	Carbon Regional dinaise /	cci
		but many more time steps needed	referring	······ cci
Nation Atmos	TAL Centre for spheric Science VIRONMENT RESEARCH COUNCIL	Why JASMIN? Bryan Lawrence - RAL, June 2016	Centre for E Data Analys SCIENCE AND TECHN NATURAL ENVIRON	nvironmental is iology Facilities council

Why JASMIN?

- Urgency to provide better environmental predictions
- Need for higher-resolution models
- HPC to perform the computation
- Huge increase in observational capability/capacity
 But...
- Massive storage requirement: observational data transfer, storage, processing
- Massive raw data output from prediction models
- Huge requirement to process raw model output into usable predictions (post-processing)

Hence JASMIN...







JAMSIN (STFC/Stephen Kill)

JASMIN infrastructure

Part data store, part HPC cluster, part private cloud...





- 16 PB Fast Storage (Panasas, many Tbit/s bandwidth)
- 1 PB Bulk Storage
- Elastic Tape
- 4000 cores: half deployed as hypervisors, half as the "Lotus" batch cluster.
- Some high memory nodes, a range, bottom heavy.







Non-blocking, low latency, CLOS Tree Network



1,104 x 10GbE Ports CLOS L3 ECMP OSPF

- ~1,200 Ports expansion
- Max 36 leaf switches :1,728 Ports @ 10GbE
- Non-Blocking, Zero Contention (48x10Gb = 12x 40Gb uplinks)
- Low Latency (250nS L3 / per switch/router) 7-10uS MPI



JASMIN "Science DMZ" Architecture





Simple Science DMZ

Supercomputer Center

The UK Met Office UPSCALE campaign



Example Data Analysis

 Tropical cyclone tracking has become routine; 50 years of N512 data can be processed in 50 jobs in one day



- Eddy vectors; analysis we would not attempt on a server/workstation (total of 3 months of processor time and ~40 GB memory needed) completed in 24 hours in 1,600 batch jobs
- JASMIN/LOTUS combination has clearly demonstrated the value of cluster computing to data processing and analysis.

The Ada Lovelace Center



The Experimental Data Challenge

- Data rates are increasing, facilities science more data intensive
 - Handling and processing data has become a bottleneck to produce science
 - Need to compare with complex models and simulations to interpret the data
- Computing provision at home-institution highly variable
 - Consistent access to HTC/HPC to process and interpret experimental data
 - Computational algorithms more specialised
 - More users without the facilities science background

Need access to data, compute and software services

- Allow more timely processing of data
- Use of HPC routine not "tour de force"
- Generate more and better science



Ada Lovelace Centre







The ALC will significantly enhance our capability to support the Facilities' science programme:

- Theme 1: Increases capacity in advanced software development for data analysis and interpretation
- Theme 2: Develop new generation of scientific data experts and scientific software engineers who can interact with science domain experts
- Theme 3: Provide significant compute infrastructure for managing, analysing and simulating the data generated by the facilities and for designing next generation Big-Science experiments

> Focus is the science drivers and computational needs of Facilities

ALC Pathfinder: Tomographic Reconstruction



- Support in-experiment and postexperiment tomographic reconstruction
 - Round-trip the data to HPC CPU/GPU clusters in experiment time
 - Tomographic image reconstruction toolbox with different algorithms
 - High throughput image reconstruction framework – time scheduled
 - Visualisation on the beamline or remote
 - An integral component of IMAT's in-experiment data analysis capability through the ISIS Mantid software suite
- Goal is to maximise the science from data collected on facility instruments

STFC Scientific Computing: Erica Yang, Srikanth Nagella, Martin Turner, Derek Ross STFC ISIS: Winfried Kockelmann, Genoveva Burca, Federico Montesino Pouzols DLS: Mark Basham

ALC Pathfinder: CCP4-DAaaS

CCP4 – Macro-Crystallography suite

- proteins, viruses and nucleic acids
- determine macromolecular structures by X-ray crystallography
- Used by DLS users
 - But need post-experimental access





Data Analysis as a Service

- Remote access to data and compute via SCD Cloud
- CCP4 s/w maintained on Cloud via VM packaging and distribution (CVMFS)
- User Portal provides access to right data and compute and workflows

Frazer Barnsley, Shirley Crompton, CCP4, et al

The ALC - Towards a "Super-facility"?

Infrastructure + Software + Expertise With Common Interfaces and Transparent Access



"A network of connected facilities, software and expertise to enable new modes of discovery"

Katie Antypas, Inder Monga, Lawrence Berkeley National Laboratory

New Opportunities: Reproducible Science

• Traceable science

- Preservation
- Provenance
- Publishing
- A tool for the user
 - Tracking progress
- 'RARE' research
 - Robust
 - Accountable
 - Reproducible
 - Explainable



> ALC can build in support for reproducible science

Jim Gray's Vision: All Scientific Data Online

- Many disciplines overlap and use data from other sciences.
- Internet can unify all literature and data
- Go from literature *to* computation *to* data *back to* literature.
- Information at your fingertips For everyone, everywhere
- Increase Scientific Information Velocity
- Huge increase in Science Productivity





(From Jim Gray's last talk)

Acknowledgements:

With thanks to Mark Basham, David Corney, Jonathan Churchill, Imanol Luengo, Barbara Montanari, Brian Matthews and Dan Rolfe