Data Collection and Analysis for the Health of Systems



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OUTLINE

- The problem
- Existing utilities
- Data Representation Standards
- Data Collection at ORNL. Our tool: DCAT
- Visualization
- Data Validation: SunSPOTS
- Future work



Data Collection and Analysis for System Health

Problem:

As HPC clusters increase in size and complexity, it becomes increasingly difficult to locate errors and pinpoint failures.

Objective:

This effort aims to provide a data collection architecture and analysis tools for the analysis of large, composite data sets in an HPC environment.

Method:

By analyzing sensor information and other healthrelated parameters, locating and predicting faults becomes more efficient.



Why and how?

- What kind of information do we want and/or can we get out of machines about themselves?
 - Temperatures, voltages, fan rpms, power supplies
 - Sometimes acoustic data, air flow
 - Memory-related statistics
- Why:
 - Failure prediction
 - Performance measurements
 - Anomaly detection
 - Load balancing
 - Memory usage
- From which components?
 - All the components that impact the above tasks
 - From components for which this information is available
 - New operating systems will need to provide this data
- How?
 - From sensors embedded in the components or on the boards by manufacturers
 - Some general open source software, utilities and drivers
 - Other manufacturer-dependent libraries



Example: Jaguar

- The XT4 partition contains 7,832 compute nodes in addition to dedicated login/service nodes. Each compute node contains a quadcore AMD Opteron 1354 (Budapest) processor running at 2.1 GHz, 8 GB of DDR2-800 memory (some nodes use DDR2-667 memory), and a SeaStar2 router. The resulting partition contains 31,328 processing cores, more than 62 TB of memory, over 600 TB of disk space, and a peak performance of 263 teraflop/s (263 trillion floating point operations per second).
 - The XT5 partition contains 224,256 compute cores in addition to dedicated login/service nodes. Each compute node contains two hex-core AMD Opteron processors, 16GB memory, and a SeaStar 2+ router.



What are the Data Problems?

- Massive amount of data streaming from potentially thousands of nodes
 - Zillions of small files/messages.
- Heterogeneous sensors and data representation
 - Accessible sensor data variables
 - Measurement outputs in numerous formats
 - Proprietary schemas
- Issues:
 - Requesting data from the sensors (push, pull, poll)
 - Metadata and Interoperability of data models for processing tools
 - Fusing data for analytics
 - Control of sensor operations

Computer system sensor data is represented with numerous schemas and data models

Standards exist but no single standard contains all the elements we need

Mixing and matching elements will require a significant effort (syntax, semantics, hierarchies) Managed by UT-Battelle for the

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Data representation

- Standardized data models
 - IEEE 1451
 - Open Geo-Spatial Consortium
 - SensorML
 - TransducerML
 - Operations and Measurements
 - Common Instrument Middleware Architecture (CIMA)
- Various data output formats from utilities
 - Im-sensors
 - Hardware Monitoring for Mac OSX
- Computer Information Management
 - Distributed Management Task Force
 - dmidecode
 - Common Information Model (CIM)





• Existing Utilities and Standards



Open Source Standards for Sensor Information Processing: ORNL/TM-2009/145 (Pouchard et al.)

- IEEE 1451
 - Standard for a Smart Transducer Interface for Sensors and Actuators
 - Open-Geospatial Consortium
- Lm-sensors
- IPMI and Trident sensors
 - Adapters exist for Im-sensors and IPMI.
 - The IPMI source code is not available.
 - IPMI operates at the BIOS level.
 - Identifying which sensor reports which measurement in a node is a matter of guesswork.
- Findings:
 - No single specification/package satisfies the goal of providing a model suitable to analyze sensor data from all manufacturers.





Linux monitoring utility (Im-sensors)

- An open source, hardware-monitoring effort measuring and reporting the health of systems.
- Sensor detection and common libraries
 - Sensor detect
 - Sensor lib
- Additional drivers depending on hardware specification
- Records core and other temperatures, voltages, and fan speeds.



Lm-sensors on pid1 and pod1 testbed

w83627hf-isa-0290 Adapter: ISA adapter VCore 1: +3.90 V (min = +1.34 V, max = +1.49 V) ALARM VCore 2: +3.79 V (min = +1.34 V, max = +1.49 V) ALARM +3.3V: +3.82 V (min = +3.14 V, max = +3.46 V) ALARM +5V: +5.27 V (min = +4.73 V, max = +5.24 V) ALARM +12V: +11.67 V (min = +10.82 V, max = +13.19 V) -12V: +0.88 V (min = -13.18 V, max = -10.88 V) ALARM -5V: +1.89 V (min = -5.25 V, max = -4.75 V) ALARM +5.59 V (min = +4.73 V, max = +5.24 V) V5SB: ALARM VBat: ALARM +0.08 V (min = +2.40 V, max = +3.60 V) fan1: 0 RPM (min = 0 RPM, div = 2) 0 RPM (min = 0 RPM, div = 2) fan2: fan3: 0 RPM (min = 0 RPM, div = 2) temp1: -48° C (high = -79° C, hyst = $+22^{\circ}$ C) sensor = thermistor -48.0° C (high = $+80^{\circ}$ C, hyst = $+75^{\circ}$ C) sensor = thermistor temp2: -48.0° C (high = $+80^{\circ}$ C, hyst = $+75^{\circ}$ C) sensor = thermistor temp3: +1.419 V (VRM Version 11.0) vid: alarms: beep enable: Sound alarm enabled

Intel Xeon

w83627hf-isa-0290

Adapter: ISA adapter VCore 1: +3.94 V (min = +0.00 V, max = +0.00 V) ALARM VCore 2: +3.89 V (min = +0.00 V, max = +0.00 V) ALARM ALARM +3.3V: +3.86V (min = +3.14V, max = +3.47V) +5V: +5.03 V (min = +4.76 V, max = +5.24 V) +12V: +11.61 V (min = +10.82 V, max = +13.19 V) -12V: +0.80 V (min = -13.18 V, max = -10.80 V) ALARM -5V: +1.84 V (min = -5.25 V, max = -4.75 V) ALARM V5SB: +5.35 V (min = +4.76 V, max = +5.24 V) ALARM ALARM VBat: +0.64 V (min = +2.40 V, max = +3.60 V) fan1: 0 RPM (min = 712 RPM, div = 8) ALARM fan2: 0 RPM (min = 712 RPM, div = 8) ALARM fan3: 0 RPM (min = 5314 RPM, div = 2) ALARM -48° C (high = +120°C, hyst = +115°C) sensor = thermistor temp1: -48.0° C (high = $+80^{\circ}$ C, hyst = $+75^{\circ}$ C) sensor = thermistor temp2: -48.0° C (high = $+80^{\circ}$ C, hyst = $+75^{\circ}$ C) sensor = thermistor temp3: +0.000 V (VRM Version 0.0) vid: alarms: beep enable:

Sound alarm enabled

AMD Opteron

Nonsense values: Im-sensors is not properly configured on these machines.

Sensor monitoring for desktop: SpeedFan

💥 SpeedFan 4.37										
Readings Clock Info Exotics S.M.A.R.T. Charts										
	6	50C	o 61C	~	46C	6	65C	× 4	9C	o 50C
	Loc	al Temp	Remote Temp		HDO	T	emp1	Tem	р2	Temp3
	ŧ	32C	↓ OC	6	63C	t	63C	9	7%	35.9%
	Т	emp4	Temp5	0	Core O	0	Core 1	HD	0	CPU0
	1	5.6%	2330 MHz	L	96%	0 01	days :31:36			
	0	CPU1	Frequency	B	attery	ι	lptime			

	Attribute	Current	Raw	Overall
	Raw Read Error Rate	107	14013057	Good
	Spin Up Time	95	0	Good
	Start/Stop Count	100	10	Very good
	Reallocated Sector Count	100	0	Very good
	Seek Error Rate	82	180378250	Good
	Power On Hours Count	89	10216	Watch
Warning: Power C	<i>n Hours Count</i> is below the average	e limits (9	8-100).	
	Spin Retry Count	100	0	Very good
	Power Cycle Count	100	12	Very good
	Unknown attribute 187	100	0	Very good
	High Fly Writes	100	0	Very good
	Airflow Temperature	63	656146469	Good
	Hardware ECC Recovered	67	43916227	Good
	Current Pending Sector	100	0	Very good
	Offline Uncorrectable Sector Count	100	0	Very good
	Ultra DMA CRC Error Rate	200	0	Very good
	Write Error Rate	100	0	Very good
	TA Increase Count	100	0	Very good

🐮 SpeedFan 4.37						
Readings Clock Info Exotics S.M.A.R.T. Charts						
Hard disk HD0 - 250.0GB - ST3250820AS						
Model ST3250820AS Firmware 3.ADG						
Perform an in-depth online analysis of this hard disk						
Attribute	Value	Worst	Warn	Raw		
🚯 Raw Read Error Rate	107	95	6	000000D5D281		
🚯 Spin Up Time	95	94	70	00000000000000000		
Start/Stop Count	100	100	20	A0000000000A		
🚯 Reallocated Sector Count	100	100	36	00000000000000000		
🚯 Seek Error Rate	82	60	30	00000AC080C4		
Power On Hours Count	89	89	0	0000000027E9		
🚯 Spin Retry Count	100	100	97	00000000000000000		
Power Cycle Count	100	100	20	00000000000C		
(Unknown Attribute)	100	100	0	0000000000000		
(Unknown Attribute)	100	100	0	00000000000000000		
(Unknown Attribute)	63	59	45	0000271C0025		
Temperature	37	41	0	001600000025		
Hardware ECC Recovered	66	63	0	000002C61BA7		
Current Pending Sector	100	100	0	000000000000000		
Offline Correctable	100	100	0	000000000000000		
🚯 UltraATA CRC Error Rate	200	200	0	0000000000000		
Write Error Rate	100	253	0	000000000000	*	
Fitness Performance						
Coded by Alfredo Milani Comparetti - 2000-2008 - alfredo@almico.com						

How would you do this with 500K cores ? You don't.



Ipmi – Intelligent Platform Management Interface



IPMI logical architecture

Adopted by:

✤ >200 OEMs

The Specification is Open Source (latest v.2.0)
Tool implementation takes place at the BIOS level
Requires proprietary drivers



IEEE 1451-0: Standard for a Smart Transducer Interface for Sensors and Actuators

What is a "sensor" according to IEEE 1451?

- transducer: A device that converts energy from one domain into another. The device may be either a sensor or an actuator.
- **actuator**: A transducer that accepts a data sample or samples and converts them into an action.
- **sensor**: A transducer that converts a physical, biological, or chemical parameter into an electrical signal.



IEEE 1451: Transducer Electronic Data Sheet (TEDS)



OGC SWE standards serve as an interface between IEEE 1451 and a network.

- Data structure of a TEDS
 - Unsigned integer 32, 4 octets
 - MetaTEDS (internal timeout value)
 - Transducer Channel (sensor metadata)
 - User's Transducer Name
 - Frequency response
 - Calibration
 - Transfer function
 - Command (sensor control)
 - Geo-location
- Represents a sensor, not the data transmitted by a sensor



TIM: Transducer Interface Model

- NCAP: Network Capable Application Processor
- PHY: Physical Connections
- ADC: Analog-to-Digital Conversion
- DAC: Digital-to-Analog Conversion



Location counts



(POD1) Supermicro H8QM3







Intel D946GZIS

Location of Field Replace-able Unit (FRUs)

- GPS coordinates, RFID tags, location of a blade, etc.
- all need to be represented



Open Geo-spatial Consortium: Sensor Model Language (Sensor ML)

- One of a suite of standards providing methods for discovering sensors, accessing data, tasking sensors, receiving alerts over a network.
- A mark-up language for describing sensor systems, processing algorithms, workflows, location, and remote control of sensors.
- All entities are described in terms of processes:
 - A sensor is a process that converts a real phenomena into data.
 - Detectors, actuators, data readers are atomic physical processes
 - Spatial transforms, derivable information are nonphysical, atomic processes
 - Provenance (history) is a non-physical, composite process
 - Sensor systems and platforms are physical, composite processes



Sensor Types in Sensor ML

to,



Data structures in SensorML (e.g.)

Elements for input and output signals, and parameters					
Name	Description				
swe:Count	Integer for counting a value				
swe:Quantity	Decimal with optional unit and constraints				
swe:Time	Either ISO 8601 or time relative to an origin.				
swe:Boolean	True or False				
swe:Category	Identifies the name of a category, should provide dictionary entry for a useful interpretation.				
swe:Text	Free text				
swe:QuantityRange	Decimal pair for specifying a quantity range with constraints.				
swe:CountRange	Integer pair for specifying a quantity range with constraints.				
swe:TimeRange	Time value pair specifying a time range.				
swe:ConditionalData	List of conditional values for a property.				
swe:Conditional Value	Qualifies data with one or more conditions.				
swe:DataRecord	Implementation of ISO-11404 Record Data Type.				
swe:Envelope	Typically used to define rectangular bounding boxes in any coordinate system.				
swe:GeoLocationArea	Area used to define bounding boxes.				
swe:Position	Position given as a group of vectors/matrices.				
swe:SimpleDataRecord	Implementation of ISO-11404 Record Data type.				
swe:Vector	The Vector has a reference frame in which the coordinates are expressed.				



Outline

Data Collection at ORNL



How DCAT started: ORCAT

- A collection tool for Open Source Intelligence.
- Based on RSS feeds.
- Data stored in a database.
- Collected Web data and climate data (land-based observations).





System Sensors

- We are able to study the Trident cluster as a physical object
 - 32 nodes with quad-core AMD Opteron processors, a GigE Infiniband switch and a Myrinet switch.
- Only a few users currently running jobs
- Each node is essentially identical and equipped with the same set of sensors
- We are currently obtaining data from all the sensors we are aware of



The Sensors

- Each of the 32 Trident compute nodes is equipped with at least 12 sensors
 - 6 fan speed sensors (only 2 appear to give distinct readings)
 - 6 temperature sensors (Recently obtained data from HP regarding location of sensors)



Details from HP: #1: near heatsink #2,#3: next to each socket #4: under power supply cage #5: near power supply #6: edge of board



IPMI – Intelligent Platform Management Interface Specification

- Specification initially created by Intel
- Promoted by Dell, HP, Intel, NEC, adopted by more than 200 OEMS (not Cray)
- Open source specification, requires drivers and proprietary implementations
- Sensor data is queried via *ipmitool* cron job outputs data once per minute
- Currently have ~25 million data points



False Alarms

- Trident is running HP Systems Insight Manager – uses ILO (Integrated Lights Out) system which has its own management processor
- Hundreds of alerts : "The server's temperature is outside of normal operating range. The system will be shutdown"
- However, reviewing our IPMI logs indicates nothing unusual
- In contact with HP to determine source of errors



Current Sensor Data Set

- We have generally been running a single job at a time across the entire cluster
- So we generally expect very similar physical behavior from the nodes



Heatsink Temperature Sensor Trident 1-32



Examining the Sensor Data

- The "equilibrium" temperature is now about 3 degrees higher now for each sensor - is the data center warmer?
- Trident11 is consistently 3-4 degrees (C) warmer than the others
 - On top of the IB switch
- Trident24 sits in the same position in the other rack
 - Not as hot, but power supply failed





Trident nodes sensor data



National Laborator

DCAT client-server architecture



- Multiple sensors at each node
- One instance of DCAT server at each node
- Performs sensor data conversion



DCAT Architecture

- Each DCAT node gathers system and sensor data.
- The data is stored in DCAT as C structs.
- The node passes that data to the client node via MPI as derived datatypes.
- The client node stores the data in its database.





Scalability

- With only one client node, a bottleneck occurs.
- Solution:
 - Split nodes up into small groups.
 - Use multiple client nodes.
 - Use multiple databases.







Outline

Validation: external sensors



Sun[™] Programmable Object Technolog (SunSPOTs)

- 41x23x70 mm, 54gr., 40ft of range.
- Powered by USB.
- Contains an Accelerometer, a light sensor, and a temperature reader.
- Communicates with a base-station (same size just a little thinner) with its own IEEE 802.15.4 wireless network.
- Remotely programmable through the basestation, itself a USB device on Trident Management Node 2.
- Uses JRE and Ant technology (java based).
- Installed on Trident InfiniBand switch, Infotrend, Altix ICE, CB 11 RAID, IBM System P5.



DCAT data model

 New DCAT database table: unit_table





Client flow chart





Using MPI for Communication



Useful for distributing changes to the server nodes or sending a request for data to the server nodes. Allows us to distribute data in $\log_2(p)$ stages instead of p-1 stages. When p=1024, that means 100 stages rather than 1023 stages.

MPI_Reduce

MPI_Reduce basically does the reverse of MPI_Bcast, but performs a given operation (min, max, add, mult, etc..) on the data, leaving the receiving node with the final answer. This will be most useful when doing analysis on the fly. For instance, getting the average system load or solving an optimization problem.



MPI_Gather is the easiest way to get data sets from N processes to a single process. This can be used to collect data from the server nodes in order to add it to the database. MPI_Scatter does the opposite, distributing N data sets to N processes.



MPI Sequence System Utilities

- Diagram shows the communication in detail.
- All functions shown are implemented.
- Currently working to clean up and optimize the code.





Visualization



Evaluation of visualization environments

- Cacti, based on Round Robin Tools:
 - Plots linear time series.
 - Advanced visualization with aggregated data cannot be represented.

Histogram of x, y, z axes on i-seismometer.



Spectrum: intensity over frequency



- GNUplot utility does not enable interaction with data.
- I-seismometer data (courtesy PJ Kennelly, Long Island University)

Visualization for DCAT



A: Enter target node (s).

B: Query operators for sensor selection.

C: Sensor description.

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- D: Sensor name.
- E: Available graphs.
- F: Graph views.
- G: Y scaling.
- H: X scaling.

Preliminary results



Histogram for node 9, temperature sensor 6, 10 buckets (42-51 C), 59872 temperature readings, (March 20-May 2nd).

Vislt

- Developed by the Department of Energy (DOE) Advanced Simulation and Computing Initiative (ASCI) to visualize and analyze the results of terascale simulations.
- An analysis tool that provides the capability to calculate derived fields.
- Decoupled architecture including Viewer, Engine, and plug-ins.
- Open Source.







DCAT data displayed with the Vislt Plug-in

- The DCAT plug-in uses the Vislt Engine.
- The Plug-in enables an API to the DCAT database.
- Includes a User Interface.



Temperature and Fan data for Node 1



Temperature Frequency Data



Knowledge Discovery

- A body of research focused on extracting information by recognizing patterns from very large datasets and transforming it into actionable knowledge by presenting it, organizing it, and integrating it into its context.
- Theoretical foundations, methods, algorithms, best practices, standards, specifications.
- Journals, SIG groups, conferences, books.
- Applications in science, medicine, intelligence, and many other domains.
- Interdisciplinary research originated in data mining.





Thank You!

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