

BIG DATA USE CASE TEMPLATE 2

NIST Big Data Public Working Group

This template was designed by the NIST Big Data Public Working Group (NBD-PWG) to gather Big Data use cases. The use case information you provide in this template will greatly help the NBD-PWG in the next phase of developing the NIST Big Data Interoperability Framework. We sincerely appreciate your effort and realize it is nontrivial.

The template can also be completed in the Google Form for Use Case Template 2: <http://bit.ly/1ff7iM9>

More information about the NBD-PWG and the NIST Big Data Interoperability Framework can be found at <http://bigdatawg.nist.gov>.

TEMPLATE OUTLINE

| | | |
|---|-----------------------------------|----|
| 1 | OVERALL PROJECT DESCRIPTION..... | 2 |
| 2 | BIG DATA CHARACTERISTICS..... | 3 |
| 3 | BIG DATA SCIENCE..... | 4 |
| 4 | SECURITY AND PRIVACY..... | 5 |
| 5 | CLASSIFY USE CASES WITH TAGS..... | 14 |
| 6 | OVERALL BIG DATA ISSUES..... | 16 |
| 7 | WORKFLOW PROCESSES..... | 16 |

General Instructions:

Brief instructions are provided with each question requesting an answer in a text field. For the questions offering check boxes, please check any that apply to the use case. .

Fields marked with * are important and must be filled in.

Please email the completed template to Wo Chang at wchang@nist.gov.

NOTE: No proprietary or confidential information should be included.

1 OVERALL PROJECT DESCRIPTION

1.1 USE CASE TITLE *

Please limit to one line. A description field is provided below for a longer description.

1.2 USE CASE DESCRIPTION *

Summarize all aspects of use case focusing on application issues (later questions will highlight technology).

1.3 USE CASE CONTACTS *

Add names, phone number, and email of key people associated with this use case. Please designate who is authorized to edit this use case.

| Name | Phone | Email | PI / Author | Edit rights? | Primary |
|------|-------|-------|-------------|--------------|---------|
|------|-------|-------|-------------|--------------|---------|

1.4 DOMAIN ("VERTICAL") *

What application area applies? There is no fixed ontology. Examples: Health Care, Social Networking, Financial, Energy, etc.

1.5 APPLICATION *

Summarize the use case applications.

1.6 CURRENT DATA ANALYSIS APPROACH *

Describe the analytics, software, hardware approach used today.

1.7 FUTURE OF APPLICATION AND APPROACH *

Describe the analytics, software, hardware, and application future plans, with possible increase in data sizes/velocity.

1.8 ACTORS / STAKEHOLDERS

Please describe the players and their roles in the use case. Identify relevant stakeholder roles and responsibilities. Note: Security and privacy roles are discussed in a separate part of this template.

1.9 PROJECT GOALS OR OBJECTIVES

Please describe the objectives of the use case.

1.10 USE CASE URL(S)

Include any URLs associated with the use case. Please separate with semicolon (;).

1.11 PICTURE AND DIAGRAMS?

Please email any pictures or diagrams with this template.

2 BIG DATA CHARACTERISTICS

Big Data Characteristics describe the properties of the (raw) data including the four major 'V's' of Big Data described in NIST Big Data Interoperability Framework: Volume 1, Big Data Definition.

2.1 DATA SOURCE

Describe the origin of data, which could be from instruments, Internet of Things, Web, Surveys, Commercial activity, or from simulations. The source(s) can be distributed, centralized, local, or remote.

2.2 DATA DESTINATION

If the data is transformed in the use case, describe where the final results end up. This has similar characteristics to data source.

2.3 VOLUME

| | |
|-------------|--|
| Size | |
| Units | |
| Time Period | |
| Proviso | |

Size: Quantitative volume of data handled in the use case

Units: What is measured such as "Tweets per year", Total LHC data in petabytes, etc.?

Time Period: Time corresponding to specified size.

Proviso: The criterion (e.g. data gathered by a particular organization) used to get size with units in time period in three fields above

2.4 VELOCITY

Enter if real time or streaming data is important. Be quantitative: this number qualified by 3 fields below: units, time period, proviso. Refers to the rate of flow at which the data is created, stored, analyzed, and visualized. For example, big velocity means that a large quantity of data is being processed in a short amount of time.

| | |
|-----------------|--|
| Unit of measure | |
| Time Period | |
| Proviso | |

Unit of Measure: Units of Velocity size given above. What is measured such as "New Tweets gathered per second", etc.?

Time Period: Time described and interval such as September 2015; items per minute

Proviso: The criterion (e.g., data gathered by a particular organization) used to get Velocity measure with units in time period in three fields above

2.5 VARIETY

Variety refers to data from multiple repositories, domains, or types. Please indicate if the data is from multiple datasets, mashups, etc.

2.6 VARIABILITY

Variability refers to changes in rate and nature of data gathered by use case. It captures a broader range of changes than Velocity which is just change in size. Please describe the use case data variability.

3 BIG DATA SCIENCE

3.1 VERACITY AND DATA QUALITY

This covers the completeness and accuracy of the data with respect to semantic content as well as syntactical quality of data (e.g., presence of missing fields or incorrect values).

3.2 VISUALIZATION

Describe the way the data is viewed by an analyst making decisions based on the data. Typically visualization is the final stage of a technical data analysis pipeline and follows the data analytics stage.

3.3 DATA TYPES

Refers to the style of data, such as structured, unstructured, images (e.g., pixels), text (e.g., characters), gene sequences, and numerical.

3.4 METADATA

Please comment on quality and richness of metadata.

3.5 CURATION AND GOVERNANCE

Note that we have a separate section for security and privacy. Comment on process to ensure good data quality and who is responsible.

3.6 DATA ANALYTICS

In the context of these use cases, analytics refers broadly to tools and algorithms used in processing the data at any stage including the data to information or knowledge to wisdom stages, as well as the information to knowledge stage.

4 SECURITY AND PRIVACY

Questions in this section are designed to gather a comprehensive image of security and privacy aspects (e.g., security, privacy, provenance, governance, curation, and system health) of the use case. Other sections contain aspects of curation, provenance and governance that are not strictly speaking only security and privacy considerations. The security and privacy questions are grouped as follows:

- Roles
- Personally Identifiable Information
- Covenants and Liability
- Ownership, Distribution, Publication
- Risk Mitigation
- Audit and Traceability
- Data Life Cycle
- Dependencies
- Framework provider S&P
- Application Provider S&P
- Information Assurance | System Health
- Permitted Use Cases

For the questions with checkboxes, please select the item(s) that apply to the use case.

4.1 ROLES

Roles may be associated with multiple functions within a big data ecosystem.

4.1.1 Identifying Role

Identify the role (e.g., Investigator, Lead Analyst, Lead Scientists, Project Leader, Manager of Product Development, VP Engineering) associated with identifying the use case need, requirements, and deployment.

4.1.2 Investigator Affiliations

This can be time-dependent and can include past affiliations in some domains.

4.1.3 Sponsors

Include disclosure requirements mandated by sponsors, funders, etc.

4.1.4 Declarations of Potential Conflicts of Interest**4.1.5 Institutional S/P duties**

List and describe roles assigned by the institution, such as via an IRB.

4.1.6 Curation

List and describe roles associated with data quality and curation, independent of any specific Big Data component. Example: Role responsible for identifying US government data as FOUO or Controlled Unclassified Information, etc.

4.1.7 Classified Data, Code or Protocols

Intellectual property protections

Military classifications, e.g., FOUO, or Controlled Classified

Not applicable

Other:

4.1.8 Multiple Investigators / Project Leads *

Only one investigator | project lead | developer

Multiple team members, but in the same organization

Multiple leads across legal organizational boundaries

Multinational investigators | project leads

Other:

4.1.9 Least Privilege Role-based Access

Least privilege requires that a user receives no more permissions than necessary to perform the user's duties.

Yes, roles are segregated and least privilege is enforced

We do have least privilege and role separation but the admin role(s) may be too all-inclusion

Handled at application provider level

Handled at framework provider level

There is no need for this feature in our application

Could be applicable in production or future versions of our work

Other:

4.1.10 Role-based Access to Data *

Please describe the level at which access to data is limited in your system.

Dataset

Data record / row

Data element / field

Handled at application provider level

Handled at framework provider level

Other:

4.2 PERSONALLY IDENTIFIABLE INFORMATION (PII)

4.2.1 Does the System Maintain PII? *

Yes, PII is part of this Big Data system

No, and none can be inferred from 3rd party sources

No, but it is possible that individuals could be identified via third party databases

Other:

4.2.2 Describe the PII, if applicable

Describe how PII is collected, anonymized, etc. Also list disclosures to human subjects, interviewees, or web visitors.

4.2.3 Additional Formal or Informal Protections for PII

4.2.4 Algorithmic / Statistical Segmentation of Human Populations

Yes, doing segmentation, possible discrimination issues if abused. Please also answer the next question.

Yes, doing segmentation, but no foreseeable discrimination issues.

Does not apply to this use case at all (e.g., no human subject data)

Other:

4.2.5 Protections afforded statistical / deep learning discrimination

Identify what measures are in place to address this concern regarding human populations, if it applies. Refer to the previous question.

4.3 COVENANTS, LIABILITY, ETC.

4.3.1 Identify any Additional Security, Compliance, Regulatory Requirements *

Refer to 45 CFR 46: <http://1.usa.gov/1bg6JQ2>

FTC regulations apply

HHS 45 CFR 46

HIPAA

EU General Data Protection (Reference: <http://bit.ly/1Ta8S1C>)

COPPA

Other Transborder issues

Fair Credit Reporting Act (Reference: <http://bit.ly/1Ta8XSN>)

Family Educational Rights and Protection (FERPA)

None apply

Other:

4.3.2 *Customer Privacy Promises*

Select all that apply. E.g., RadioShack promise that is subject of this DOJ ruling: <http://bit.ly/1f0MW9t>

Yes, we're making privacy promises to customers or subjects

We are using a notice-and-consent model

Not applicable

Other:

4.4 OWNERSHIP, IDENTITY AND DISTRIBUTION

4.4.1 *Publication rights*

Open publisher; traditional publisher; white paper; working paper

Open publication

Proprietary

Traditional publisher rights (e.g., Springer, Elsevier, IEEE)

"Big Science" tools in use

Other:

4.4.2 *Chain of Trust*

Identify any chain-of-trust mechanisms in place (e.g., ONC Data Provenance Initiative.) Potentially very domain-dependent; see the ONC event grid for instance. Reference: <http://bit.ly/1f0PGDL>

4.4.3 *Delegated Rights*

Example of one approach: "Delegation Logic: A Logic-based Approach to Distributed Authorization", Li, N., Grosf, B.N., Feigenbaum, J.(2003) <https://www.cs.purdue.edu/homes/ninghui/papers/thesis.pdf>

4.4.4 *Software License Restrictions*

Identify proprietary software used in the use case Big Data system which could restrict use, reproducibility, results, or distribution.

4.4.5 *Results Repository*

Identify any public or private / federated consortia maintaining a shared repository.

4.4.6 *Restrictions on Discovery*

Describe restrictions or protocols imposed on discoverable end points.

4.4.7 Privacy Notices

Indicate any privacy notices required / associated with data collected for redistribution to others

- Privacy notices apply
- Privacy notices do not apply
- Other:

4.4.8 Key Management

- A key management scheme is part of our system
- We are using public key infrastructure.
- We do not use key management, but it could have been useful
- No readily identifiable use for key management
- Other:

4.4.9 Describe and Key Management Practices**4.4.10 Is an identity framework used?**

- A framework is in place. (See next question.)
- Not currently using a framework.
- There is no perceived need for an identity framework.
- Other:

4.4.11 CAC / ECA Cards or Other Enterprise-wide Framework

- Using an externally maintained enterprise-wide identity framework
- Could be used, but none are available
- Not applicable

4.4.12 Describe the Identity Framework.**4.4.13 How is intellectual property protected?**

- Login screens advising of IP issues
- Employee or team training
- Official guidelines limiting access or distribution
- Required to track all access to, distribution of digital assets
- Does not apply to this effort (e.g., public effort)
- Other:

4.5 RISK MITIGATION

4.5.1 Are measures in place to deter re-identification? *

- Yes, in place
- Not in place, but such measures do apply
- Not applicable
- Other:

4.5.2 Please describe any re-identification deterrents in place**4.5.3 Are data segmentation practices being used?**

Data segmentation for privacy has been suggested as one strategy to enhance privacy protections. Reference: <http://bit.ly/1P3h12Y>

- Yes, being used
- Not in use, but does apply
- Not applicable
- Other:

4.5.4 Is there an explicit governance plan or framework for the effort?

- Explicit governance plan
- No governance plan, but could use one
- I don't think governance contributes anything to this project
- Other:

4.5.5 Privacy-Preserving Practices

Identify any privacy-preserving measures that are in place.

4.5.6 Do you foresee any potential risks from public or private open data projects?

Transparency and data sharing initiatives can release into public use datasets that can be used to undermine privacy (and, indirectly, security.)

- Risks are known.
- Currently no known risks, but it is conceivable.
- Not sure
- Unlikely that this will ever be an issue (e.g., no PII, human-agent related data or subsystems.)
- Other:

4.6 PROVENANCE (OWNERSHIP)

Provenance viewed from a security or privacy perspective. The primary meaning for some domains is digital reproducibility, but it could apply in simulation scenarios as well.

4.6.1 Describe your metadata management practices

- Yes, we have a metadata management system.
- There is no need for a metadata management system in this use case
- It is applicable but we do not currently have one.
- Other:

4.6.2 If a metadata management system is present, what measures are in place to verify and protect its integrity?

4.6.3 Describe provenance as related to instrumentation, sensors or other devices.

We have potential machine-to-machine traffic provenance concerns.

Endpoint sensors or instruments have signatures periodically updated

Using hardware or software methods, we detect and remediate outlier signatures

Endpoint signature detection and upstream flow are built into system processing

We rely on third party vendors to manage endpoint integrity

We use a sampling method to verify endpoint integrity

Not a concern at this time

Other:

4.7 DATA LIFE CYCLE

4.7.1 Describe Archive Processes

Our application has no separate "archive" process

We offload data using certain criteria to removable media which are taken offline

we use a multi-stage, tiered archive process

We allow for "forgetting" of individual PII on request

Have ability to track individual data elements across all stages of processing, including archive

Additional protections, such as separate encryption, are applied to archival data

Archived data is saved for potential later use by applications or analytics yet to be built

Does not apply to our application

Other:

4.7.2 Describe Point in Time and Other Dependency Issues

Some data is valid only within a point in time,

Some data is only valid with other, related data is available or applicable, such as the existence of a building, the presence of a weather event, or the active use of a vehicle

There are specific events in the application that render certain data obsolete or unusable

Point and Time and related dependencies do not apply

Other:

4.7.3 Compliance with Secure Data Disposal Requirements

Per NCSL: "at least 29 states have enacted laws that require entities to destroy, dispose. . ."

<http://www.ncsl.org/research/telecommunications-and-information-technology/privacy-and-security.aspx>

We are required to destroy or otherwise dispose of data

Does not apply to us

Not sure

Other:

4.8 AUDIT AND TRACEABILITY

Big Data use case: SEC Rule 613 initiative

4.8.1 Current audit needs *

We have third party registrar or other audits, such as for ISO 9001
We have internal enterprise audit requirements
Audit is only for system health or other management requirements
No audit, not needed or does not apply
Other:

4.8.2 Auditing versus Monitoring

We rely on third party or O.S. tools to audit, e.g., Windows or Linux auditing
There are built-in tools for monitoring or logging that are only used for system or application health monitoring
Monitoring services include logging of role-based access to assets such as PII or other resources
The same individual(s) in the enterprise are responsible for auditing as for monitoring
This aspect of our application is still in flux
Does not apply to our setting
Other:

4.8.3 System Health Tools

We rely on system-wide tools for health monitoring
We built application health tools specifically to address integrity, performance monitoring and related concerns
There is no need in our setting
Other:

4.8.4 What events are currently audited? *

All data access must be audited
Only selected / protected data must be audited
Maintenance on user roles must be audited (new users, disabled user, updated roles or permissions)
Purge and archive events
Domain-dependent events (e.g., adding a new sensor)
REST or SOAP events
Changes in system configuration
Organizational changes
External project ownership / management changes
Requirements are externally set, e.g., by PCI compliance
Domain-specific events (patient death in a drug trial)
Other:

4.9 APPLICATION PROVIDER SECURITY

4.9.1 Describe Application Provider Security *

One example of application layer security is the SAP ERP application

- There is a security mechanism implemented at the application level
- The app provider level is aware of PII or privacy data elements
- The app provider implements audit and logging
- The app provider security relies on framework-level security for its operation
- Does not apply to our application
- Other:

4.10 FRAMEWORK PROVIDER SECURITY

One example is Microsoft Active Directory as applied across LANs to Azure, or LDAP mapped to Hadoop.
Reference: <http://bit.ly/1f0VDR3>

4.10.1 Describe the framework provider security *

- Security is implemented at the framework level
- Roles can be defined at the framework level
- The framework level is aware of PII or related sensitive data
- Does not apply in our setting
- Is provided by the Big Data tool
- Other:

4.11 SYSTEM HEALTH

Also included in this grouping: Availability, Resilience, Information Assurance

4.11.1 Measures to Ensure Availability *

- Deterrents to man-in-the-middle attacks
- Deterrents to denial of service attacks
- Replication, redundancy or other resilience measures
- Deterrents to data corruption, drops or other critical big data components
- Other:

4.12 PERMITTED USE CASES

Beyond the scope of S&P considerations presented thus far, please identify particular domain-specific limitations

4.12.1 Describe Domain-specific Limitations on Use

4.12.2 Paywall

- A paywall is in use at some stage in the workflow
- Not applicable

5 CLASSIFY USE CASES WITH TAGS

The questions below will generate tags that can be used to classify submitted use cases. See <http://dsc.soic.indiana.edu/publications/OgrePaperv11.pdf> (Towards an Understanding of Facets and Exemplars of Big Data Applications) for an example of how tags were used in the initial 51 use cases. Check any number of items from each of the questions.

5.1 DATA: APPLICATION STYLE AND DATA SHARING AND ACQUISITION

- Uses Geographical Information Systems?
- Use case involves Internet of Things?
- Data comes from HPC or other simulations?
- Data Fusion important?
- Data is Real time Streaming?
- Data is Batched Streaming (e.g. collected remotely and uploaded every so often)?
- Important Data is in a Permanent Repository (Not streamed)?
- Transient Data important?
- Permanent Data Important?
- Data shared between different applications/users?
- Data largely dedicated to only this use case?

5.2 DATA: MANAGEMENT AND STORAGE

- Application data system based on Files?
- Application data system based on Objects?
- Uses HDFS style File System?
- Uses Wide area File System like Lustre?
- Uses HPC parallel file system like GPFS?
- Uses SQL?
- Uses NoSQL?
- Uses NewSQL?
- Uses Graph Database?

5.3 DATA: DESCRIBE OTHER DATA ACQUISITION/ ACCESS/ SHARING/ MANAGEMENT/ STORAGE ISSUES

5.4 ANALYTICS: DATA FORMAT AND NATURE OF ALGORITHM USED IN ANALYTICS

- Data regular?
- Data dynamic?
- Algorithm $O(N^2)$?
- Basic statistics (regression, moments) used?
- Search/Query/Index of application data Important?
- Classification of data Important?
- Recommender Engine Used?
- Clustering algorithms used?
- Alignment algorithms used?
- (Deep) Learning algorithms used?
- Graph Analytics Used?

5.5 ANALYTICS: DESCRIBE OTHER DATA ANALYTICS USED

Examples include learning styles (supervised) or libraries (Mahout).

5.6 PROGRAMMING MODEL

Pleasingly parallel Structure? Parallel execution over independent data. Called Many Task or high throughput computing. MapReduce with only Map and no Reduce of this type

Use case NOT Pleasingly Parallel -- Parallelism involves linkage between tasks. MapReduce (with Map and Reduce) of this type

Uses Classic MapReduce? such as Hadoop

Uses Apache Spark or similar Iterative MapReduce?

Uses Graph processing as in Apache Giraph?

Uses MPI (HPC Communication) and/or Bulk Synchronous Processing BSP?

Dataflow Programming Model used?

Workflow or Orchestration software used?

Python or Scripting front ends used? Maybe used for orchestration

Shared memory architectures important?

Event-based Programming Model used?

Agent-based Programming Model used?

Use case I/O dominated? I/O time > or >> Compute time

Use case involves little I/O? Compute >> I/O

5.7 OTHER PROGRAMMING MODEL TAGS

Provide other programming style tags not included in the list above.

5.8 PLEASE ESTIMATE RATIO I/O BYTES/FLOPS

Specify in text box with units.

5.9 DESCRIBE MEMORY SIZE OR ACCESS ISSUES

Specify in text box with any quantitative detail on memory access/compute/I/O ratios

6 OVERALL BIG DATA ISSUES

6.1 OTHER BIG DATA ISSUES

Please list other important aspects that the use case highlights. This question provides a chance to address questions which should have been asked.

6.2 USER INTERFACE AND MOBILE ACCESS ISSUES

Describe issues in accessing or generating Big Data from clients, including Smart Phones and tablets.

6.3 LIST KEY FEATURES AND RELATED USE CASES

Put use case in context of related use cases. What features generalize and what are idiosyncratic to this use case?

6.4 PROJECT FUTURE

How are application and approach (e.g., hardware, software, analytics) expected to change in future?

7 WORKFLOW PROCESSES

Please answer this question if the use case contains multiple steps where Big Data characteristics, recorded in this template, vary across steps. If possible flesh out workflow in the separate set of questions. Only use this section if your use case has multiple stages where Big Data issues differ significantly between stages.

7.1 PLEASE COMMENT ON WORKFLOW PROCESSES

Please record any overall comments on the use case workflow.

7.2 WORKFLOW DETAILS FOR EACH STAGE *

Description of table fields below:

Data Source(s): The origin of data, which could be from instruments, Internet of Things, Web, Surveys, Commercial activity, or from simulations. The source(s) can be distributed, centralized, local, or remote. Often data source at one stage is destination of previous stage with raw data driving first stage.

Nature of Data: What items are in the data?

Software Used: List software packages used

Data Analytics: List algorithms and analytics libraries/packages used

Infrastructure: Compute, Network and Storage used. Note sizes infrastructure -- especially if "big".

Percentage of Use Case Effort: Explain units. Could be clock time elapsed or fraction of compute cycles

Other Comments: Include comments here on items like veracity and variety present in upper level but omitted in summary.

7.2.1 Workflow Details for Stage 1

| | |
|-------------------------------|--|
| Stage 1 Name | |
| Data Source(s) | |
| Nature of Data | |
| Software Used | |
| Data Analytics | |
| Infrastructure | |
| Percentage of Use Case Effort | |
| Other Comments | |

7.2.2 Workflow Details for Stage 2

| | |
|-------------------------------|--|
| Stage 2 Name | |
| Data Source(s) | |
| Nature of Data | |
| Software Used | |
| Data Analytics | |
| Infrastructure | |
| Percentage of Use Case Effort | |
| Other Comments | |

7.2.3 Workflow Details for Stage 3

| | |
|-------------------------------|--|
| Stage 3 Name | |
| Data Source(s) | |
| Nature of Data | |
| Software Used | |
| Data Analytics | |
| Infrastructure | |
| Percentage of Use Case Effort | |
| Other Comments | |

7.2.4 Workflow Details for Stage 4

| | |
|-------------------------------|--|
| Stage 4 Name | |
| Data Source(s) | |
| Nature of Data | |
| Software Used | |
| Data Analytics | |
| Infrastructure | |
| Percentage of Use Case Effort | |
| Other Comments | |

7.2.5 Workflow Details for Stages 5 and any further stages

If you have more than five stages, please put stages 5 and higher here.

| | |
|-------------------------------|--|
| Stage 5 Name | |
| Data Source(s) | |
| Nature of Data | |
| Software Used | |
| Data Analytics | |
| Infrastructure | |
| Percentage of Use Case Effort | |
| Other Comments | |

Description of NIST Public Working Group on Big Data

NIST is leading the development of a Big Data Technology Roadmap. This roadmap will define and prioritize requirements for interoperability, portability, reusability, and extendibility for big data analytic techniques and technology infrastructure in order to support secure and effective adoption of Big Data. To help develop the ideas in the Big Data Technology Roadmap, NIST is creating the Public Working Group for Big Data.

Scope: The focus of the NBD-PWG is to form a community of interest from industry, academia, and government, with the goal of developing a consensus definitions, taxonomies, secure reference architectures, and technology roadmap. The aim is to create vendor-neutral, technology and infrastructure agnostic deliverables to enable Big Data stakeholders to pick-and-choose best analytics tools for their processing and visualization requirements on the most suitable computing platforms and clusters while allowing value-added from Big Data service providers and flow of data between the stakeholders in a cohesive and secure manner.

For more, refer to the web site at <http://bigdatawg.nist.gov>