



OpenChorus: Building a Tool-Chest for Big Data Science

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Agenda

- Tools for Data Science
- Data Science Workflow
- Greenplum OpenChorus
- How Chorus Works

Data Science Tools: Abundance of Riches

- Proliferation of tools
- Languages & Libraries
 - R, Matlab, Python – SciPy, NLTK, Madlib, Mahout
- Frameworks
 - Graphlab, Pregel (Giraffe), Mesos, CEP
- Platforms/Data Stores
 - MPP Databases, Hadoop, NoSQL (Hbase, Cassandra, MongoDB), SciDB

Choice of Tool(s)?

- Hammer ?
 - Hadoop ought to be sufficient for most tasks
 - “If all you have have a hammer, throw away everything that is not a nail” – Jimmy Lin (<http://arxiv.org/abs/1209.2191>)
 - Operational complexity / learning curve not worth efficiency
- Tool-Chest ?
 - Use the right tool for the right job
 - How to reduce complexity

**Hammer or Tool-
Chest ?**

**Let the workload
decide**

Data Science Workload

- <http://www.dataists.com/2010/09/a-taxonomy-of-data-science/>
- Obtain
- Scrub
- Explore
- Model
- Interpret

Obtain

- Corpus needs to be usable & sufficient
- Possibly from multiple independent sources
- Needs to be automated for streams
- Needs to have efficient ingestion for one-time data

Scrub

- Raw data is always messy
 - Missing data, inconsistent data, charsets(!)
 - NY, New York, NYC, Big Apple etc
- Growing Dictionaries
- Join with Crowdsourcing
 - Mechanical Turk etc

Explore

- Visualize, Clustering, Dimensionality reduction
 - Feature correlations (scatter plots)
 - Single feature histograms
- Challenge: How not to lose these observations

Model

- Find correlation of past data and outcome
 - Find good training set
 - Label the training set
 - Derive model parameters
 - Apply model, and validate
- Ensemble Models: Model of models

Interpret

- Models are built for prediction and interpretation
- Check that there are no “surprises”
- Reason about models
- Improve models

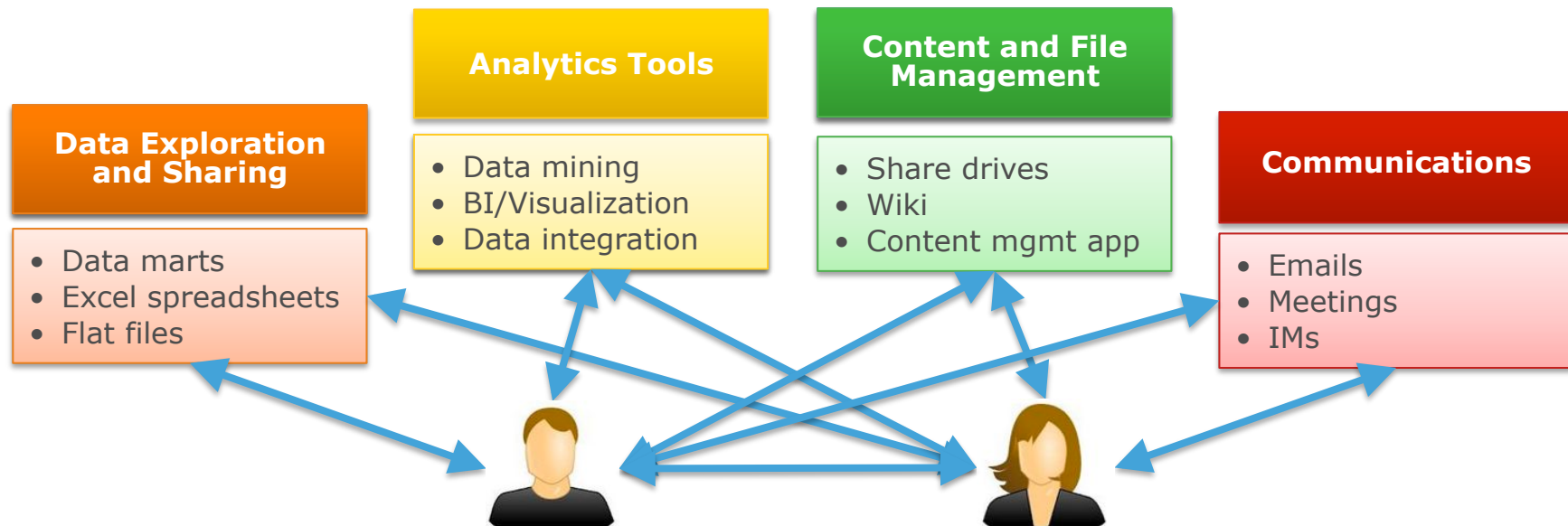
Data Science Data Flow

- Raw Data (Timed, Partitioned, Crowdsourced, De-duped etc)
- Derived data (simple aggregates, other statistics)
- Models (Feature weights, decision trees)
- Indexes

Data Diversity

- Natural Language Text, and Annotations
- (Bags of words) : Concept
- Graphs (sparse matrices)
- Dense Matrices
- Location (proximity)

Too Many Tools for One Data Science Project

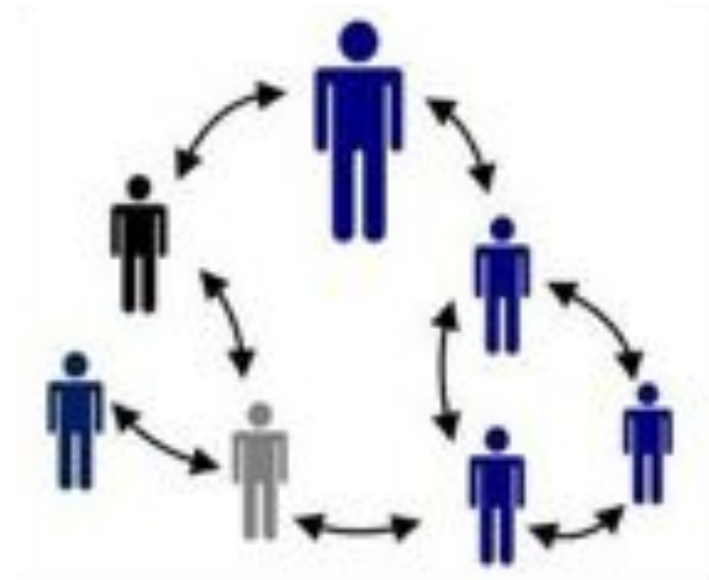


Process Documentation

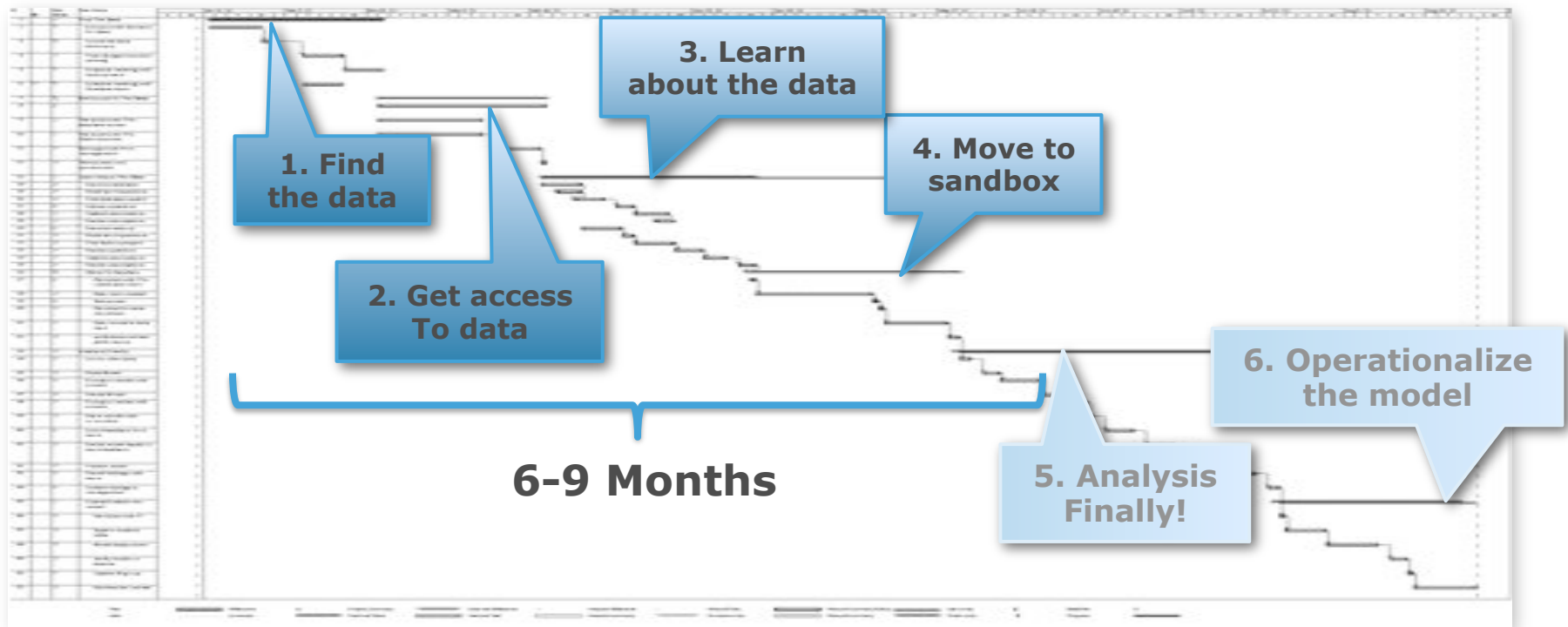
- Some project plans, no up-to-date team collaborative documentation

High Cost of Knowledge Sharing

- Data science process breaks when organization structure changes
- Very difficult knowledge transfer
- No “insurance policy” for the data science intellectual assets



Delayed Time-to-Market



Greenplum Chorus



- Collaborative analytics
- Powerful extensibility
- The freedom of open source

Greenplum's Social Platform for Collaborative Data Science

Chorus Enables Collaborative Data Science

- Collaborate within projects, share **data, content, and findings** across teams
- Make projects more transparent
- Iterate faster for accelerated insights with real-time social collaboration



Powerful Extensibility



- Integrated development environment for analytics
- Expand insights with simple access to third-party data
- Fusion with leading analytics and visualization tools

The Freedom of Open Source



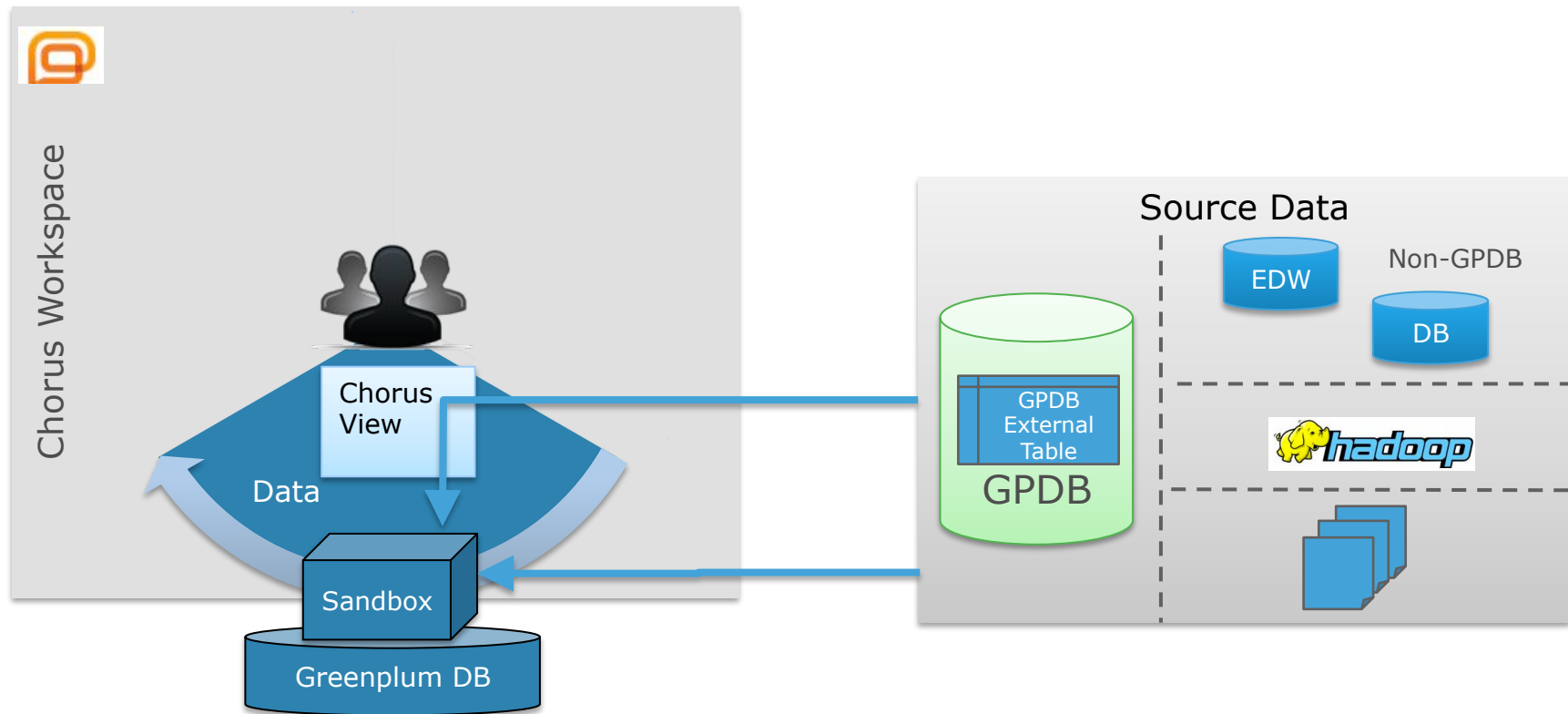
www.openchorus.org

- Modify and extend to any environment
- Promotes an ecosystem of applications, startups, and data scientists community

How Chorus Works



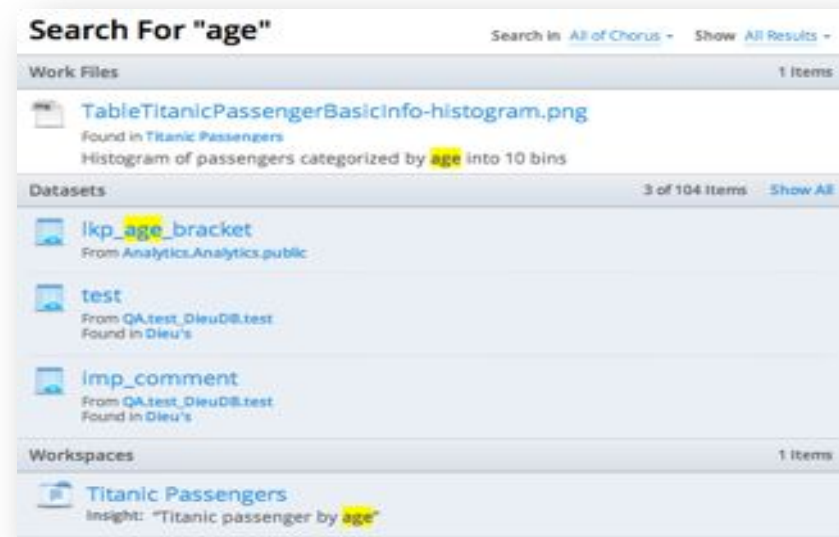
How Chorus Works



Data Exploration

Search and Data Discovery

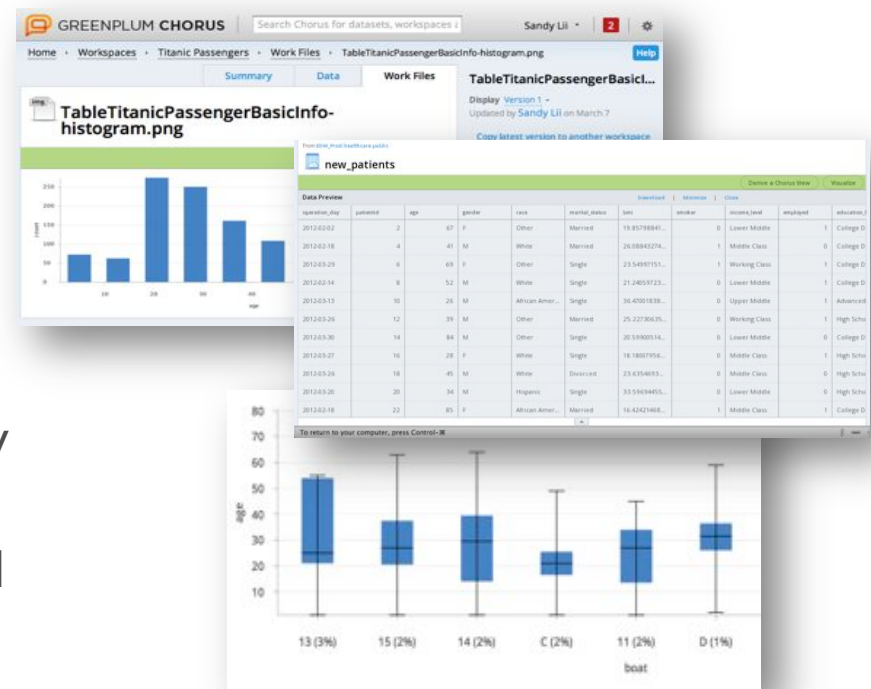
- Automatic indexing of meta-data, work files, comments, and insights
- Quickly find data across the enterprise regardless of location



Data Exploration

Data Preview and Visualization

- Data preview for instant understanding
- Quick and easy data visualizations
 - Visualize data for faster insight into datasets
 - No need to export to third-party applications like R
 - Not a replacement for advanced visualization tools



Data Exploration

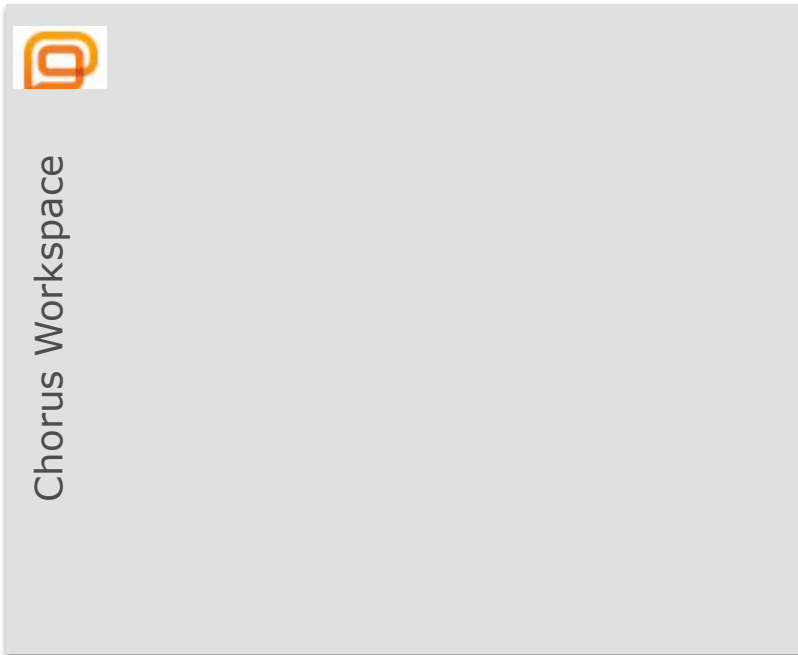
Living Data Dictionary

- Bring everything about the data to the data
 - Attach documents
 - Ask questions
 - Add comments
- Build a living data dictionary
 - Everything is current
 - No more spreadsheets

The screenshot displays a web-based data exploration interface. At the top, a breadcrumb trail reads 'Home > Instances > Analytics > Analytics > analytics > test2'. Below this, it indicates the data was 'Found in Sandy's Workspace' and identifies the table as 'TableTitanicPassengerBasicInfo'. A 'Visualize' button is present. The main area is divided into 'Data Preview' and 'Activity' sections. The 'Data Preview' section shows a table with columns 'age', 'embarked', 'fare', and 'home'. The 'Activity' section shows a comment from 'EDC Admin' asking about the 'embarked' column values, and a comment from 'Sandy Lii' defining 'S' as Southampton and 'C' as Cherbourg. Below the table, there are labels for 'age', 'embarked', and 'fare' with corresponding data type indicators: 'age' is 'int', 'embarked' is 'tpchar', and 'fare' is 'int'.

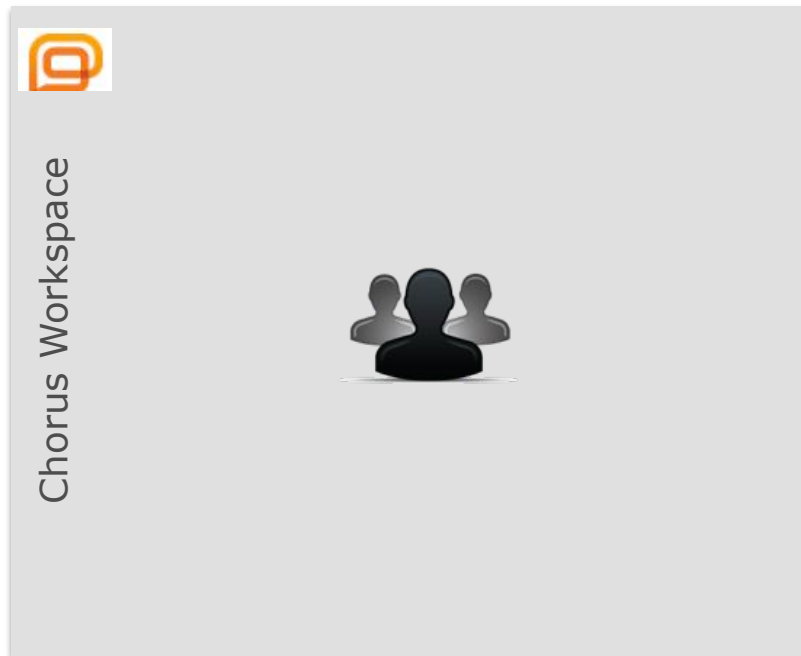
age	embarked	fare	home
55	S	267	New York, N.Y.
35	S	438	Scituate, MA
45	C	323	New York, N.Y.
35	S	161	New York, N.Y.

Workspace – Streamlines Collaboration



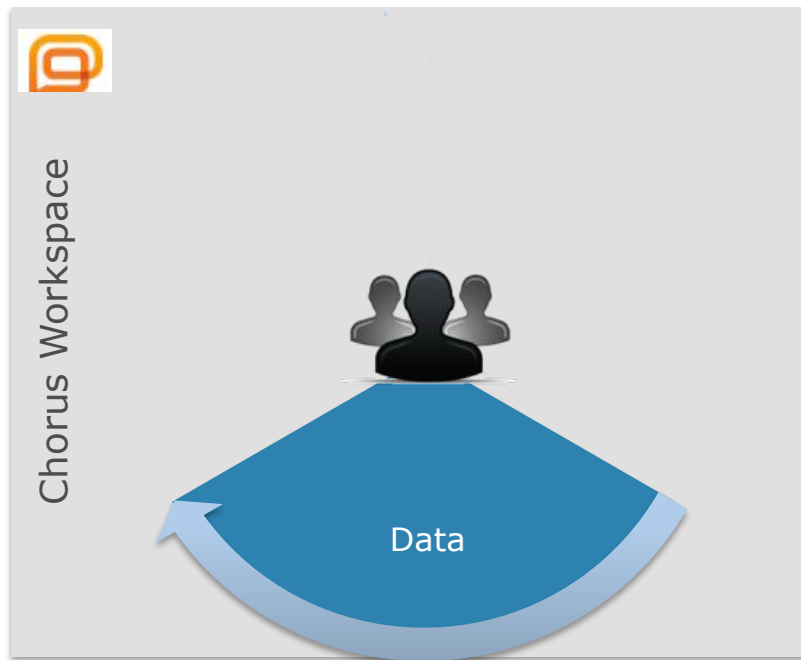
- Chorus includes unlimited workspaces, each representing individual project
- Streamlines complex user-user and user-data interactions

Multi-level Secure Collaboration



- Authentication
 - Integrates with LDAP and AD for password management
- Application access control
 - User roles: Admin vs. general user
 - Workspace types: Public or private
- Data access control
 - Chorus enforces database rules and permissions

Data – Dataset Types



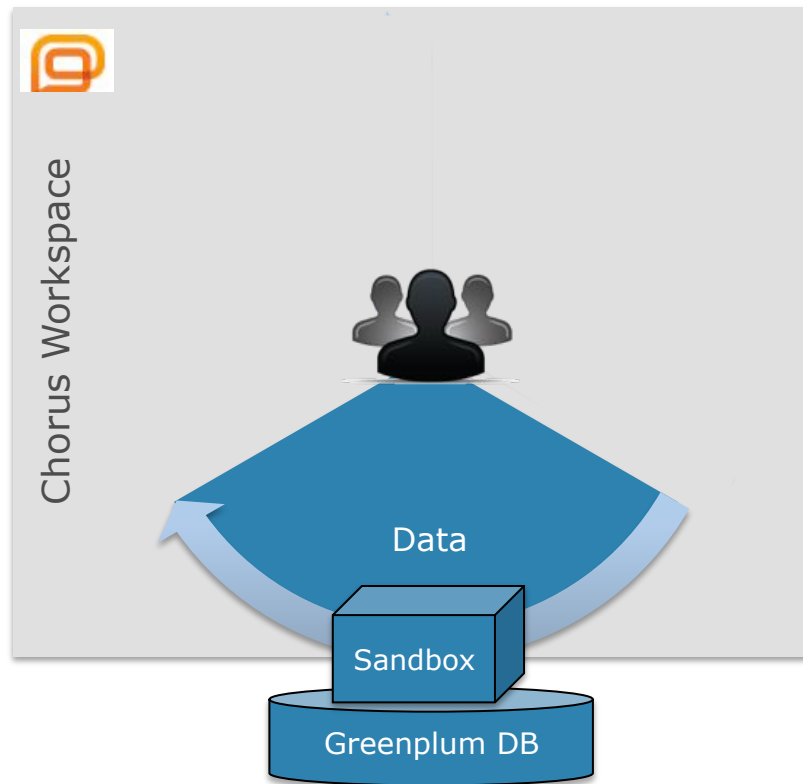
1. Source Dataset

- Pointer to the source data
- Both internal and external data
- Support both native connectivity for GPDB and flat files
- Use GPDB External Tables for Non-GPDB databases and Hadoop

2. Sandbox Dataset

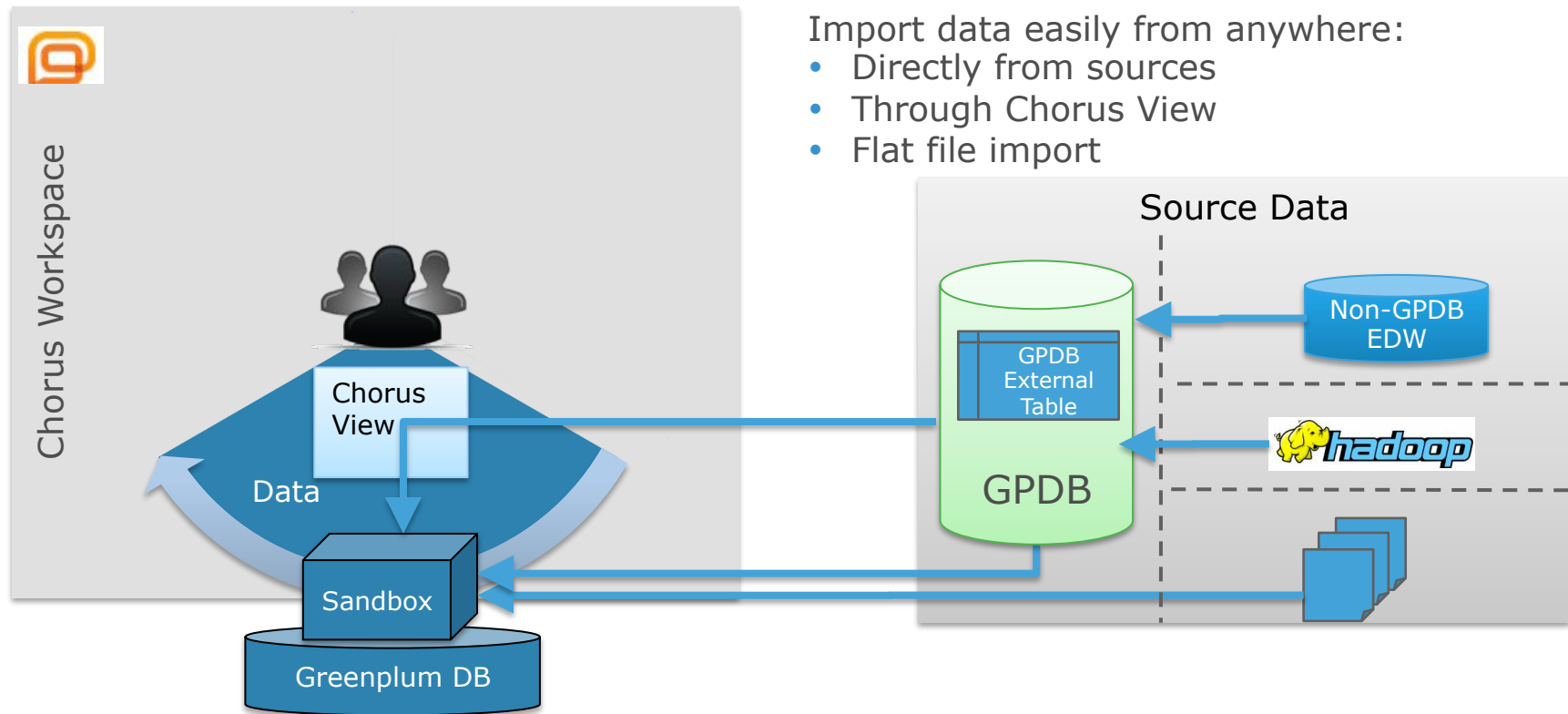
- Copy of the source data to be used for analytics
- Data generated from analytics

Data – Sandbox



- Container of all the analytics data
- Ease of self-service provisioning of sandboxes
 - Free up IT bandwidth
 - Minimize data proliferation to uncontrolled/unmanaged data marts

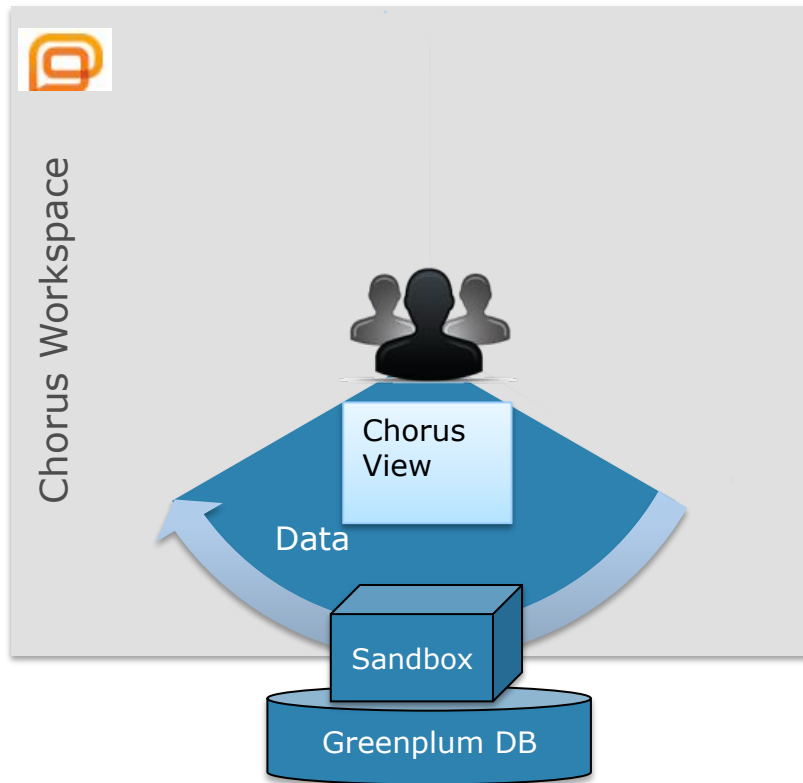
Data – Populating Sandbox



Import data easily from anywhere:

- Directly from sources
- Through Chorus View
- Flat file import

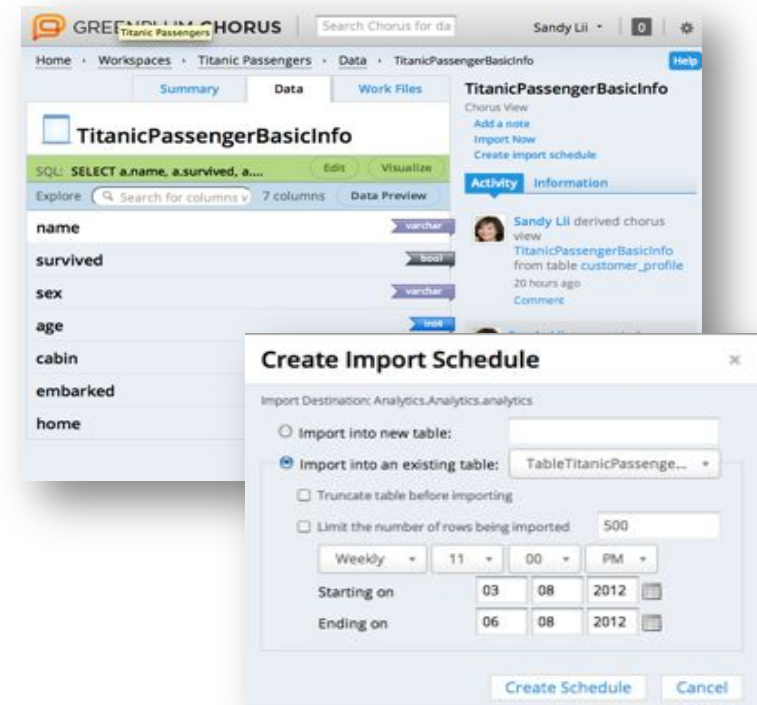
Data – Chorus View Utility



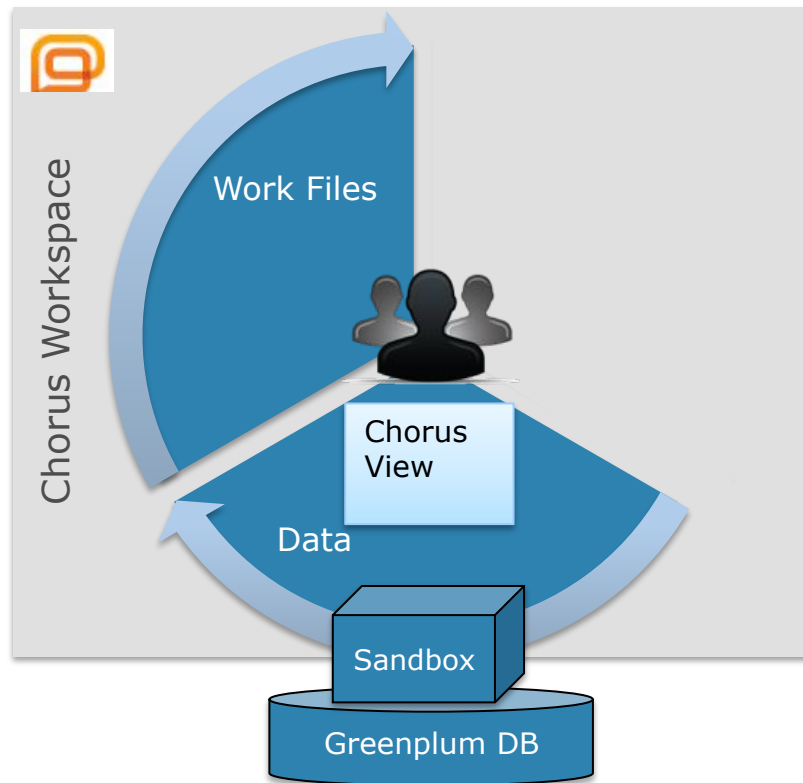
- Single-view GUI utility for **exploring, filtering, aggregating, and moving** the desired data from sources to sandbox
- Data exploration and visualization prior to bringing the data into sandbox
- Derive variation of the basic source datasets without bringing the data into sandbox

Data – Automated Data Services

- Subscribe to receive automatic updates
 - Schedule imports from multiple data sources
 - Define and share data sets within the data science team
 - Removes manual data refresh activities



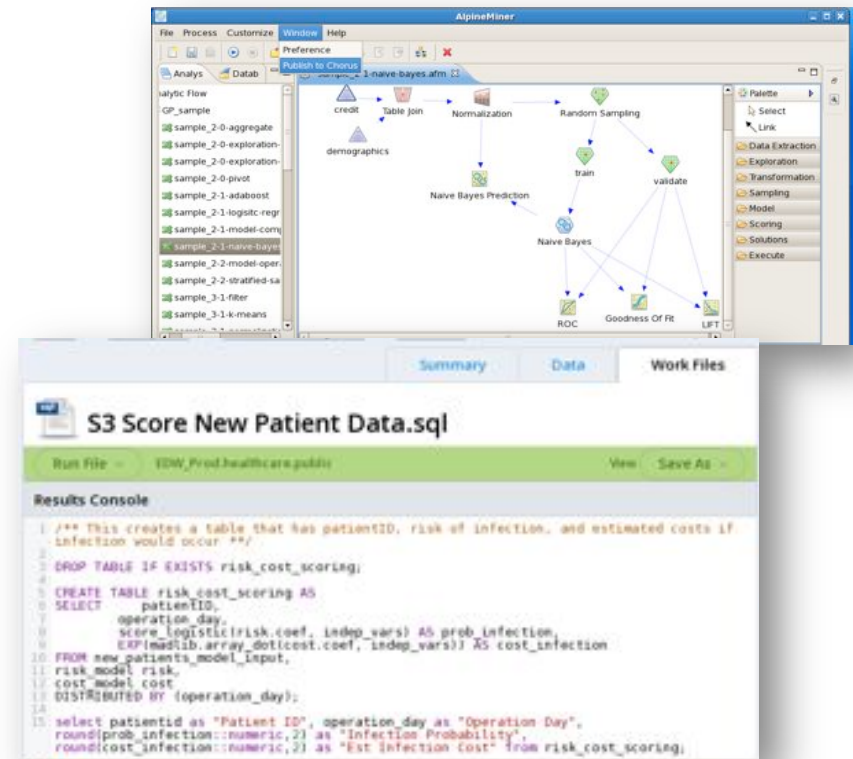
Work Files



- Work files are **non-data assets**
 - SQL query statements with code editor interface
 - Execution of in-database analytics, ex: MADLib, PL/R
 - Third-party tool files
 - PowerPoint, Word doc, etc.
- Analytics asset management with version, compare, and archive work files

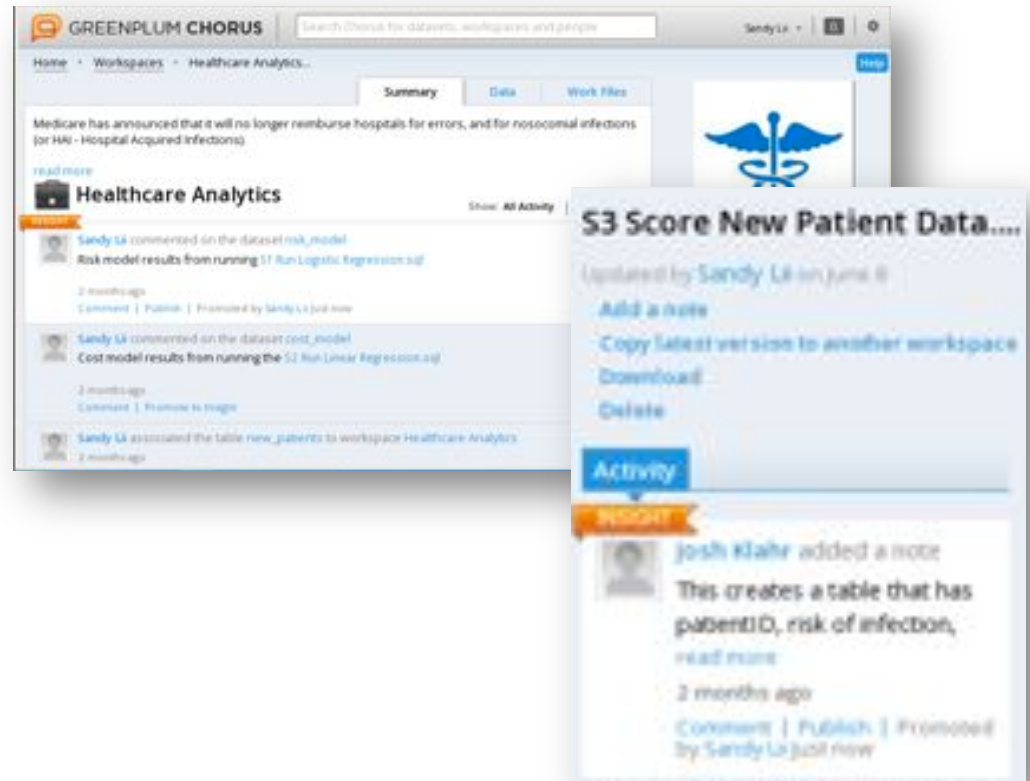
Integration with Analytics Tools

- Third-party tools
 - Execute in-database analytics functions (ex: MADLib, R) from Chorus work files
 - Publish and execute Alpine Miner Workflow from Chorus native interface
 - Data preparation for analysis using SAS and other analytics tools
- Code-design UI for SQL

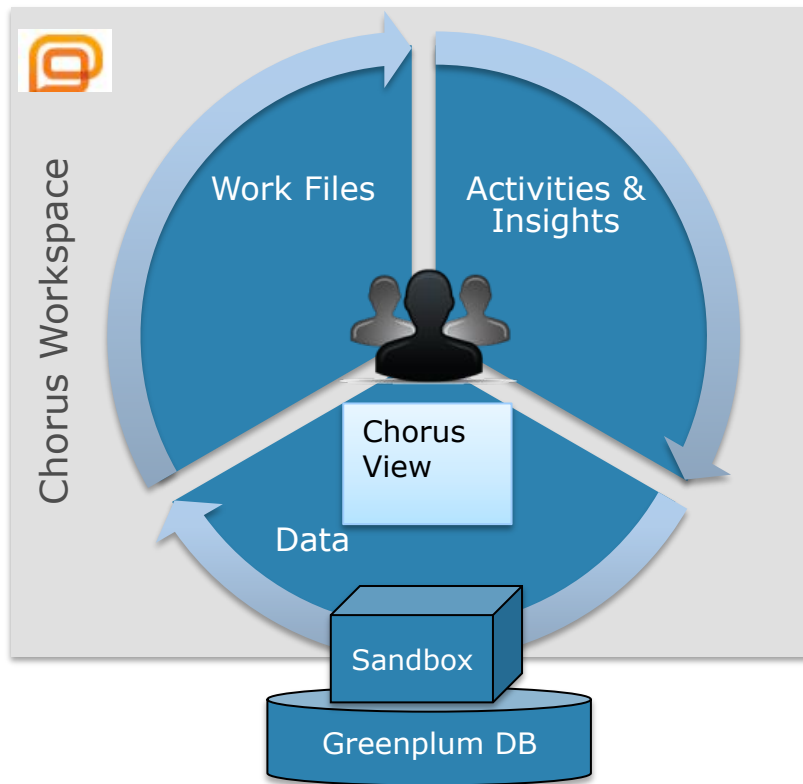


Insight and Data Sharing

- Post comments and ask questions on any analytics artifacts
- Share and publish any activities or insights
- Promote fast iteration on data and ideas



Activities and Insights



- Build a living library of activities and insights
 - Define, publish, and share new insights
 - Discover and learn from existing insights
- Iterate faster, model less

BIG DATA

THE ANSWER

to agile

