Foundations for Successful Data Projects

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About the presenters

Ted Malaska

- Capital One: Director of Enterprise Architecture
- Blizzard Ent: Director of Engineering of Global Insights
- Cloudera: Principal Solution Architect
- FINRA: Lead Architect
- Contributor: Apache Spark, Hadoop, Hive, Sqoop, Yarn, Flume, others

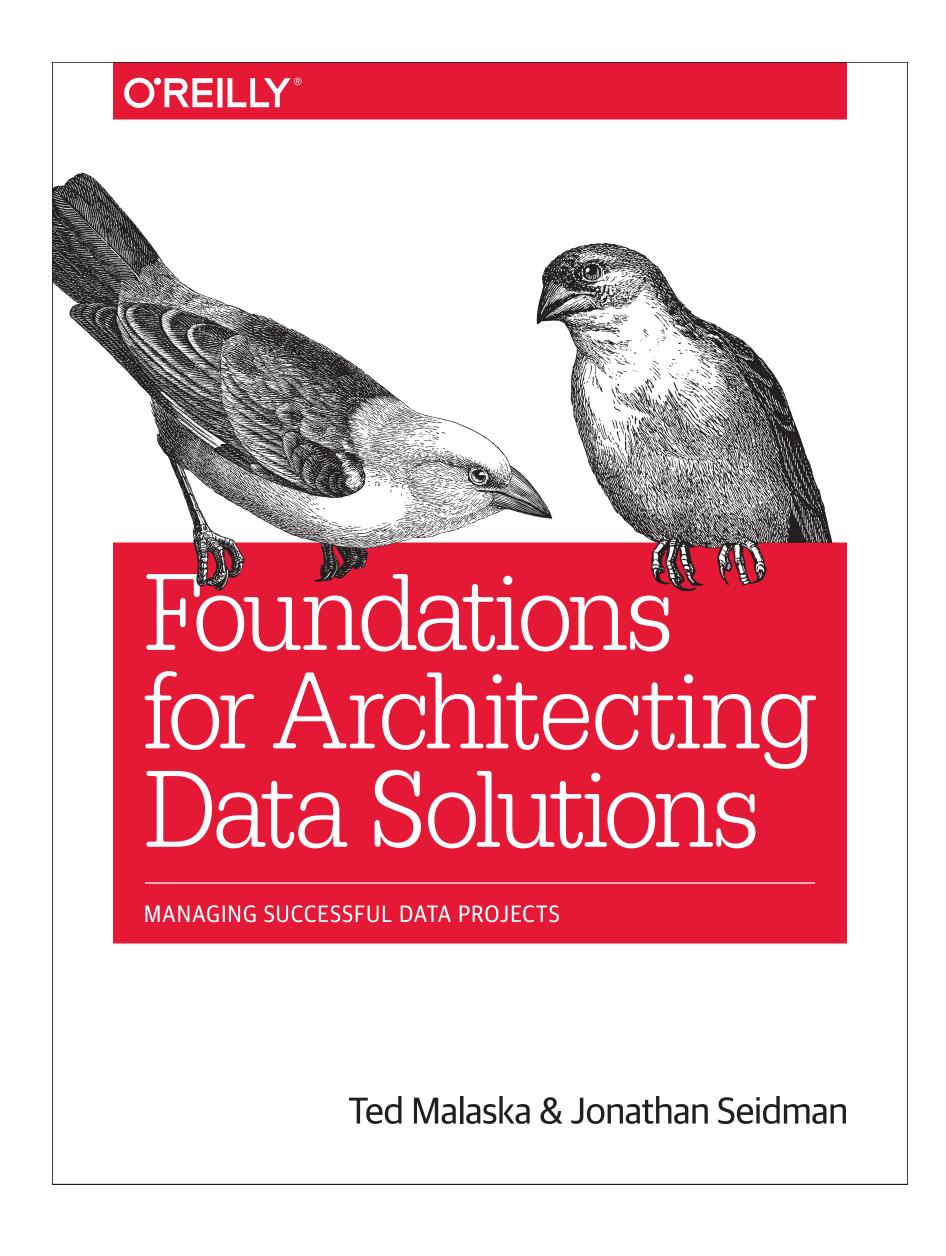


About the presenters

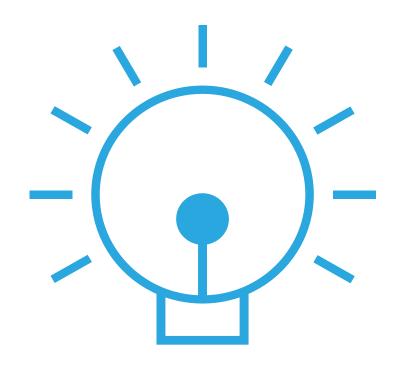
Jonathan Seidman

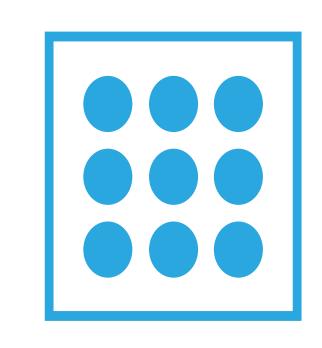
- Software Engineer at Cloudera
- Previously Technical Lead on the big data team at Orbitz
- Co-foundér of the Chicago Hadoop User Group and Chicago Big Data

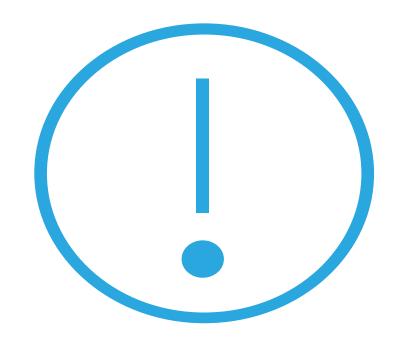


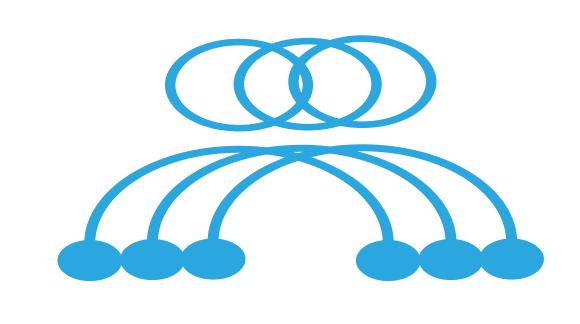


Foundations of Successful Data Projects











Understand the problem

Select software

Manage risk

Build effective teams

Build maintainable architectures



Agenda

- Understanding the key data project types
- Selecting data management solutions
- Building effective teams
- Managing risk in projects
- Ensuring data integrity
- Metadata management
- Using abstractions



Understanding the Key Data Project Types



Understanding the key data project types

- Major Data Project Types
- Primary Considerations & Risk Management
- Team Makeup

Major Data Project Types

- Data Pipelines and Data Staging
- Data Processing and Analysis
- Application Development



Data Pipelines and Data Staging

- Sourcing Data
- Transmitting Data
- Staging Data
- Accessibility Options
- Discovery



Data Processing and Analysis

- Curating Data
- Cultivating Ideas
- Data Product Generation
 - Reports, Models, Insight, Charts, ...



Application Development

- Traditional or Model Serving
- Inner Loop
- Outer Loop



Primary Considerations

- Data Pipelines and Data Staging
- Data Processing and Analysis
- Application Development



Primary Considerations

Data Pipelines and Data Staging



Data Pipelines and Data Staging - Considerations

- On boarding paths for Data Suppliers
 - Files
 - Embedded code
 - APIs (Rest, WebSocket, GRPC, Syslog, ...)
 - Agents



Data Pipelines and Data Staging – Considerations

- Transmission
 - At Least Once, Duplication, Latency, and Ordering
- Tokenization & Auditing & Governance
 - GDPR, CA Protection Laws, Misuse, Data Breach
- Quality
 - Schema Validation, Rules Validation, Carnality Verance
- Access
 - Security, Matching the use case to the storage system



Data Pipelines and Data Staging - Considerations

- Meta Management
 - New and mutated Datasets
 - Security
- Access
 - Matching the use case to the storage system
 - SQL is King
 - No one tool
 - Trade Offs
 - Cost vs Time to Value vs Value of Data



Primary Considerations

Data Processing and Analysis



Data Processing and Analysis – Considerations

- Curating Data
 - Working with Producers
 - Joining
 - Time series
 - CDC
- Undering Quality of Data
 - SLAs
 - Correctness of the Data
 - Stability of the Data
 - Coupling



Data Processing and Analysis – Considerations

- Cultivating Ideas
 - Defining Real Goals
 - Evaluating ROI
- Productionization of Pipelines
 - Service Reliability Engineering
- Culture
 - ML vs AI vs Engineer



Data Processing and Analysis – Considerations

- Understanding
 - Explainable Outcomes
 - Defendable Solutions
- Promotion Paths
 - Deploying Products
 - Historical Evaluation
 - Up to Date Auditing



Primary Considerations

Application Development



Application Development – Considerations

- Availability and Failure
 - How will it fail
 - How will failure impact customers
 - What level of fail should be tested for
 - Levels of failure design
- State Locality and Consistency
 - What are the requirements
 - Speed, cost, or truth
 - Transactions and Locking



Application Development

- Latency and Throughput
 - Expectations and Throughput
 - Is it really big data?
 - Inner and Outer Looping
- Granularity of Deployments
 - Monolith single deployment
 - Monolith microservices
- Culture
 - Development Towers
 - Over the wall
 - Development Granularity



Team Makeup



Team Makeup

- Data Pipelines and Data Staging
- Data Processing and Analysis
- Application Development

Data Pipelines and Data Staging – Team Makeup

- Data Engineers
- Site Reliability Engineers (SRE)
- Application Engineers
- Data Architects
- Governance
- Solution Engineers/Architects



Data Processing and Analysis – Team Makeup

- Visionaries
- The Brains
- Problem Seekers
- Engineers
- Duct Tapers
- Tech Debt Payers
- Site Reliability Engineers (SRE)



Application Development – Team Makeup

- Web Developers
- Front end Developers
- Data Engineers (DBAs)
- Performance Focused Engineers
- SOA / Queue Engineers
- Site Reliability Engineers (SRE)



Evaluating and Selecting Data Solutions

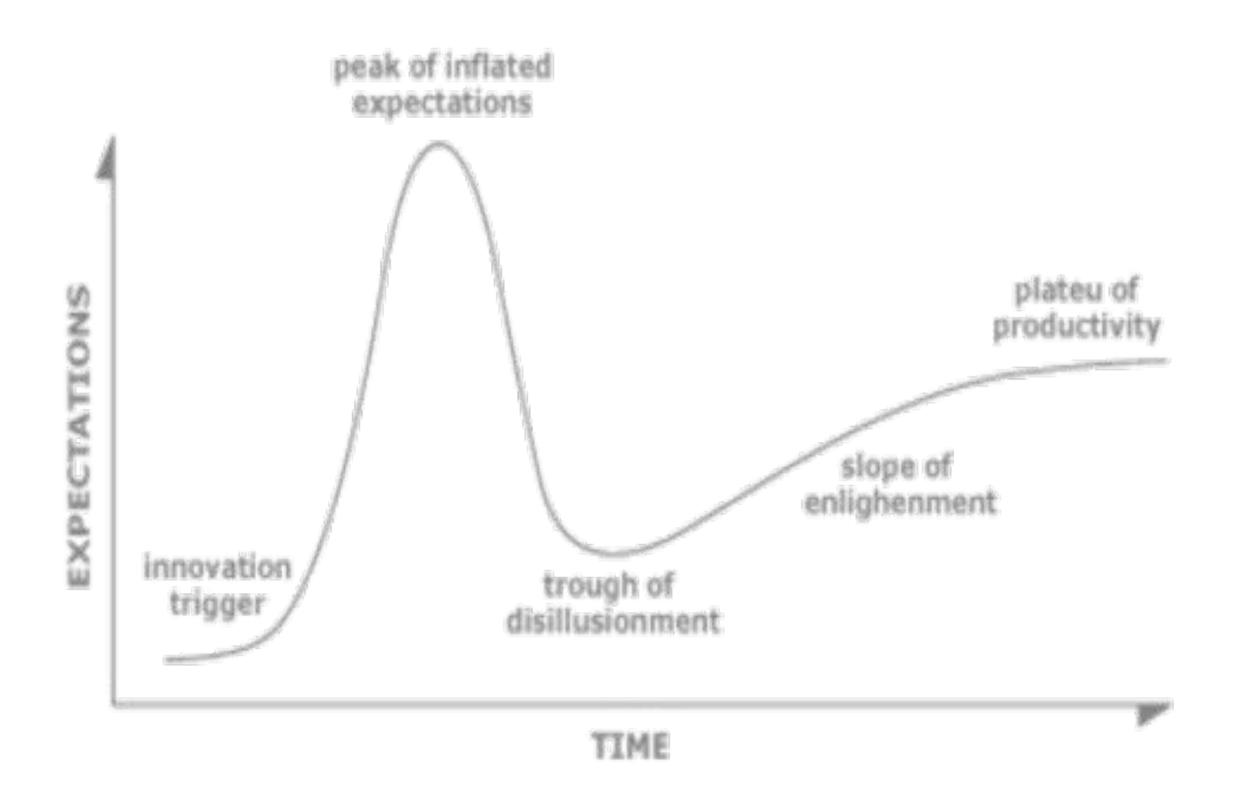


Evaluating and Selecting Data Solutions

- Solution Life Cycles
- Tipping Point Considerations
- Considerations for Technology Selection

Solution Life Cycles

- Private Incubation Stage
- Release Stage
- "Curing Cancer" Stage
- Broken Promises Stage
- Hardening Stage
- Enterprise Stage
- Decline and Slow Death Stage





Private Incubation Stage

- Technology Trigger
- Vision



Release Stage

- Changes
 - Inviting People In
 - Documentation
 - Marketing
- Reasons for Releasing
 - Money
 - Hiring
 - Culture
 - Future Building
- Big Promises



"Curing Cancer" Stage

- Big Promise
- Maybe outside area of expertise
 - Promise to push internally
 - Promises to gain influence
 - Promises to get attriations
- Promises can be good and bad



Broken Promises Stage

- Cracks in the Dream
 - Scale
 - Usability
 - Use Case
 - Security
 - Practicality
 - Skill Requirements
 - Auditability
 - Maintainability
 - Integration
 - Quality
 - Lies



Hardening

- Balance Features
- Technical Debt
- Partnering
- Corp Partnerships
- Leadership Stories
- Easy Success Paths



Enterprise Stage

- Stable
- Predictable
- Easy to hire for
- Supportable / Maintainable
- Pragmatists outnumber innovators
- No longer cool, but still very lucrative



Slow Decline Stage

- Not Worth Retiring
- Not worth Investing In
- Good Enough



Tipping Point Considerations

- Mavericks
- Connectors
- Salesman
- Stickiness
- Context

Mavericks

- Passion Driven
- Helpful
- Bottom Up Power
- They see the future or may see shadows



Connectors

- High triangles
- Trusted weak ties
- Gateways for pain, needs, and opportunities
- Considering the towered companies



Sales Man

- Make the Deal Happen
- Right or wrong doesn't matter as much as action
- Momentum starters



Stickiness

- Think about gravity
 - Data
 - Code
 - User's Favor
 - Results



Context

- Where is the company
- Looking for Opportunities
 - Holding down the fort
 - Lower cost
 - Play around
- The Swing Pendulum Effort
 - Where is the ball now and where is it heading



Tipping Point Considerations

- Mavericks
- Connectors
- Salesman
- Stickiness
- Context



Considerations for Technology Selection

- Demand
- Fit
- Visibility
- Risks

Evaluating the Demand

- Business Needs
- Internal Demand
- Desire to live on the edge



Evaluating the Fit

- Primary Capability
- Skill Sets
- Level of Commitments
- Level of Alignment



Evaluating the Visibility

- Benchmarks
 - Hidden biases, Motivated Biases, Unfair Comparisons
- Fundamentals
 - There is no magic
- Leaders Success
- Market Trends

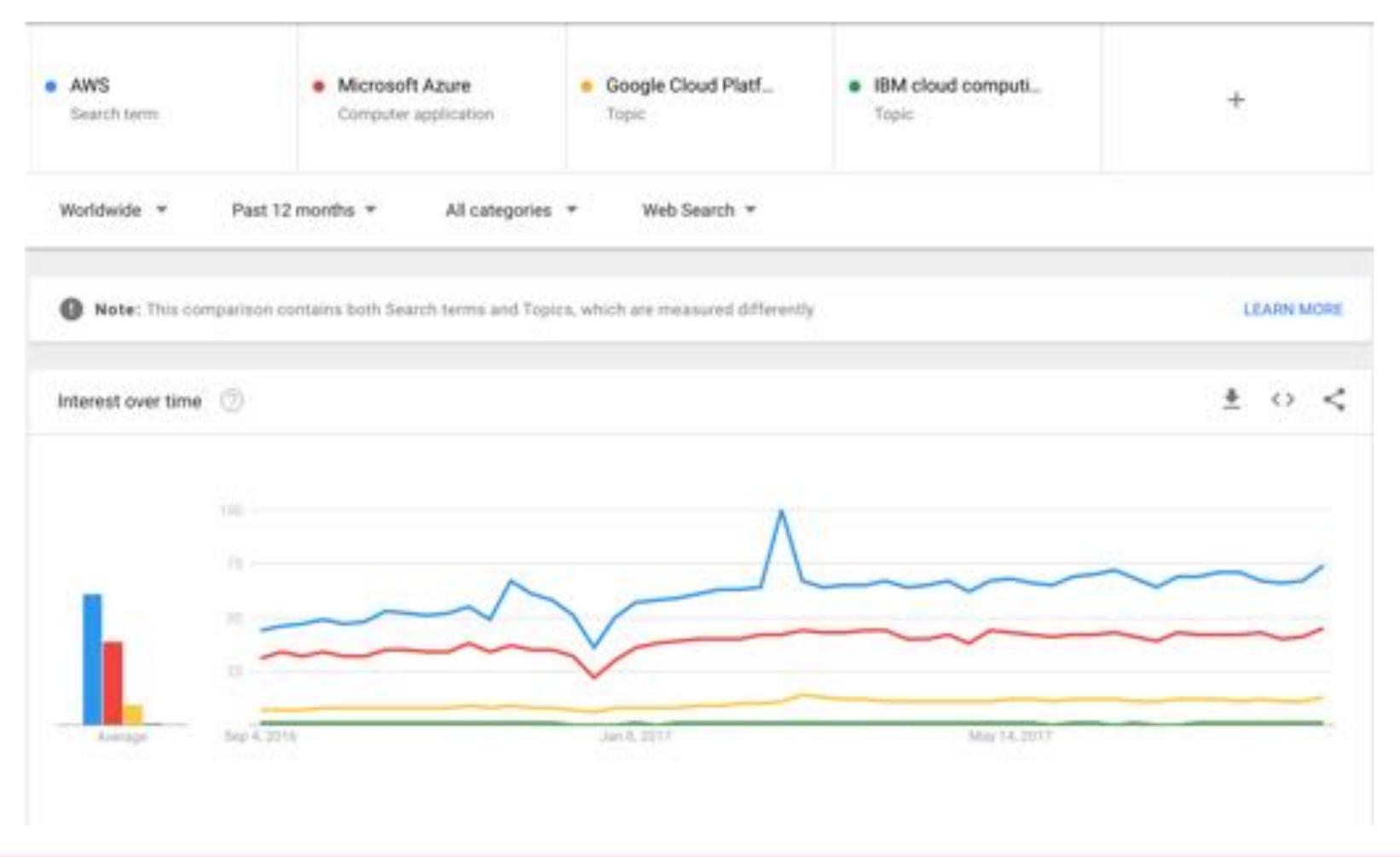


Reviewing Fundamentals

- Relative Location of Data to Readers
- Compression formats and rates
- Data Structures
- Partitioning, Replication, and Failure
- API and Interfaces
- Resource Allocations and Tuning

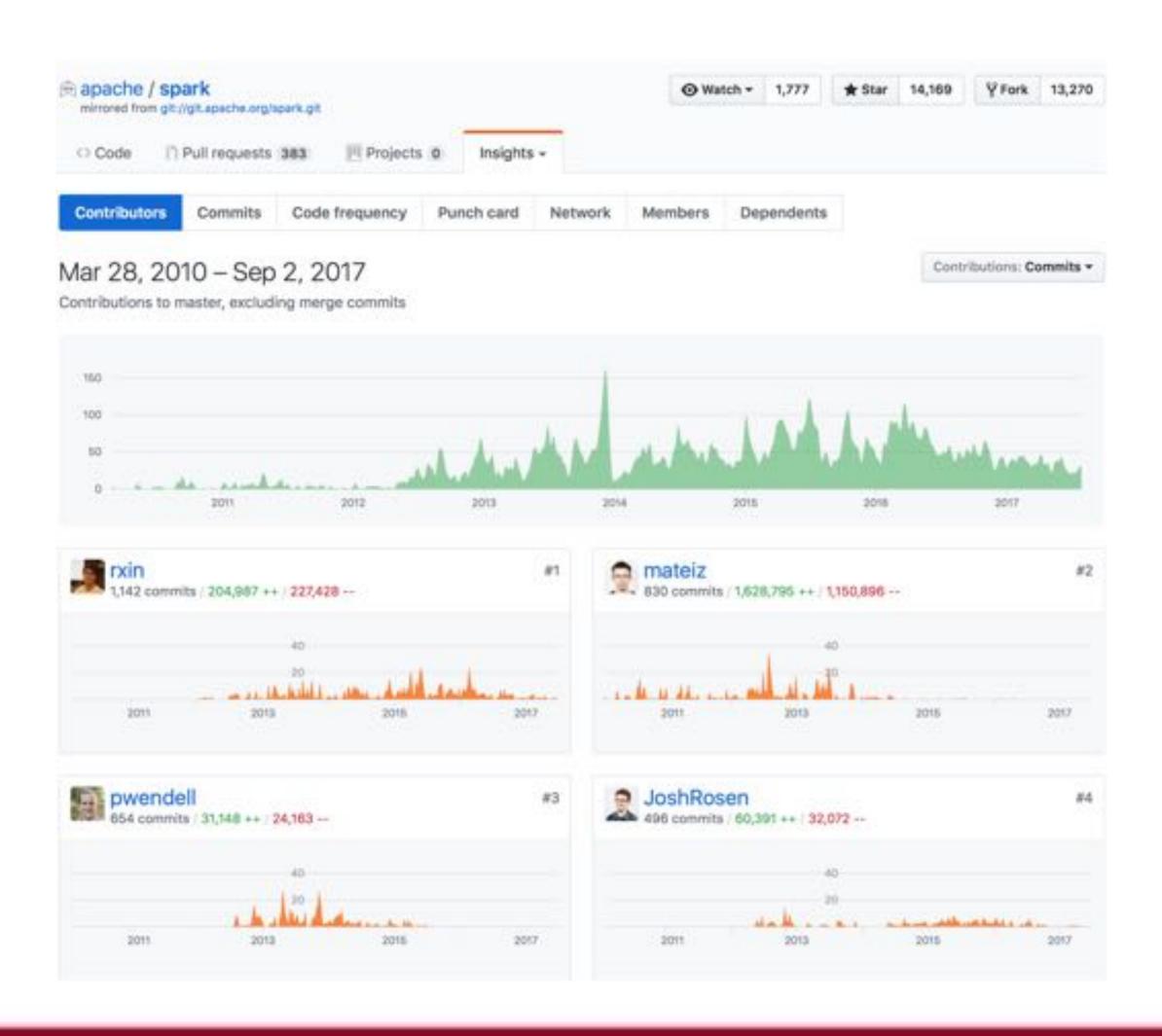


Google Trends



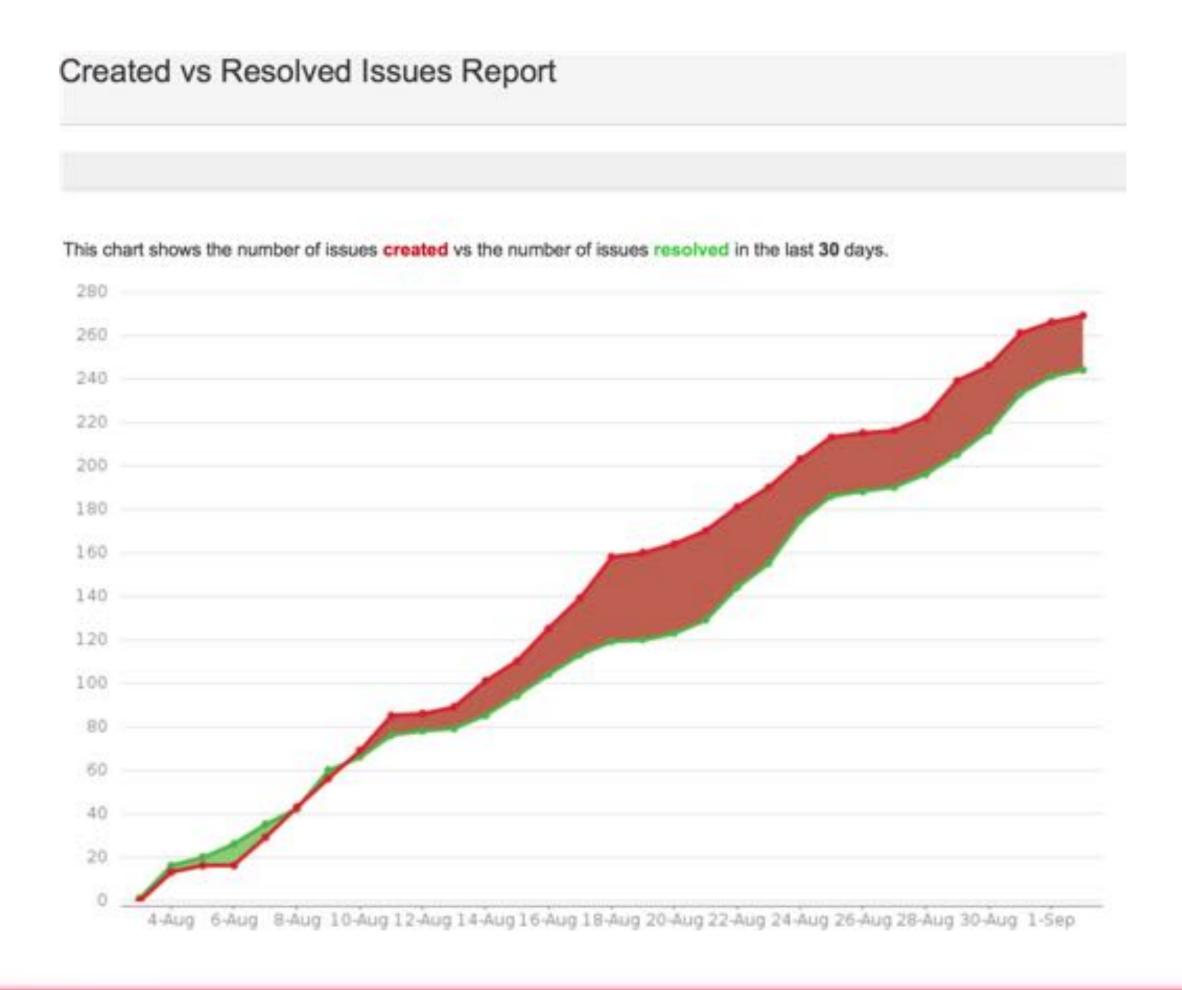


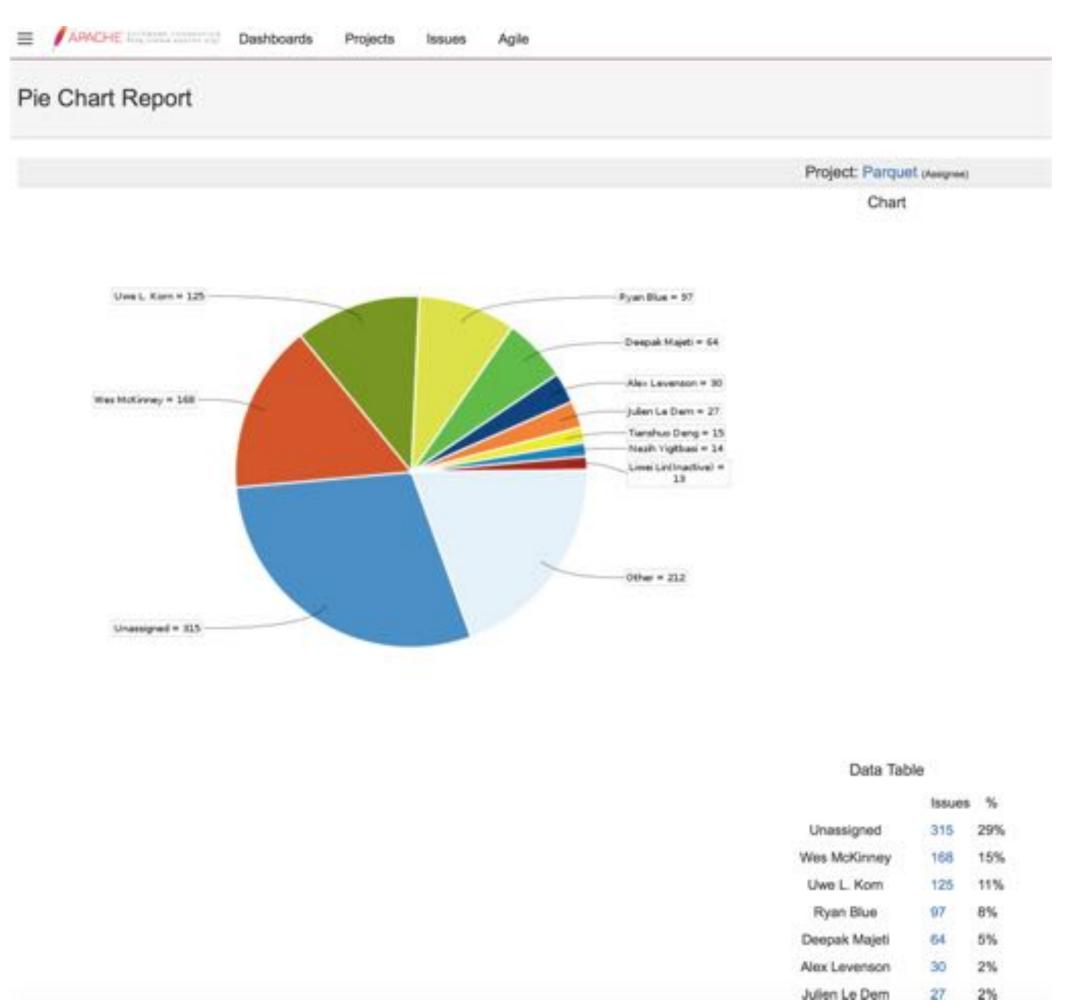
Github activity





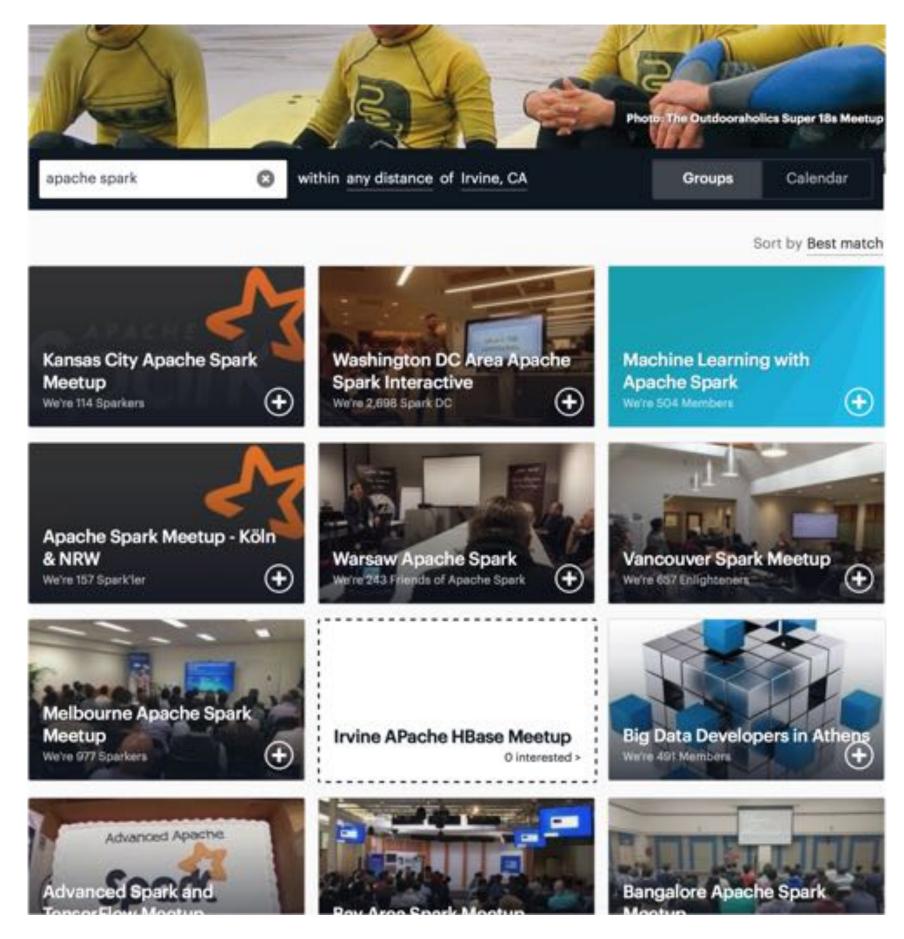
Jira Counts and Charts







Conferences and meetups





- Also:
 - Community Interest
 - Email Lists and Forums
 - Contributors
 - Follow the Money \$\$\$



Evaluating the Risks

- Risk Tolerance
- Stress Tolerance
- Leader vs follower



Future Proofing

- Assume Change
- Interface Design
- Producer & Consumer Experience



Assume Change

- Remember the Logic and Physical
- Think Logical and Implementation



Interface Design

- Standards
- SQL
- DataFrames / DataSets
- REST, GRPC
- AVRO, Parquet, Protobuf, Thrift, JSON, CSV

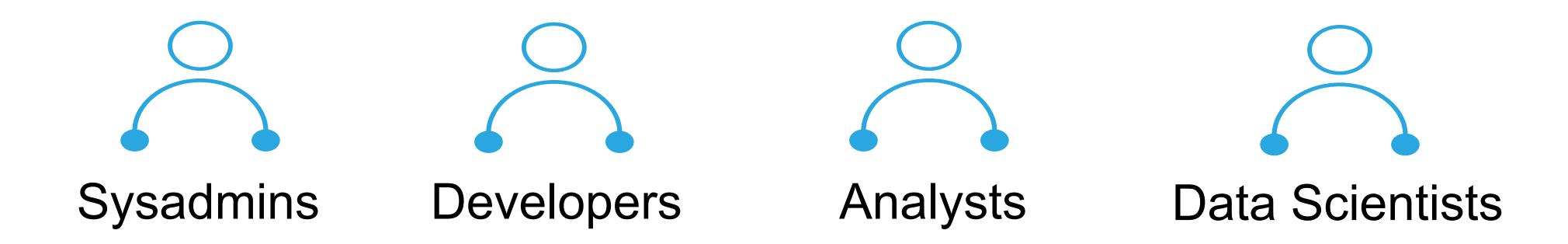


Building Successful Teams

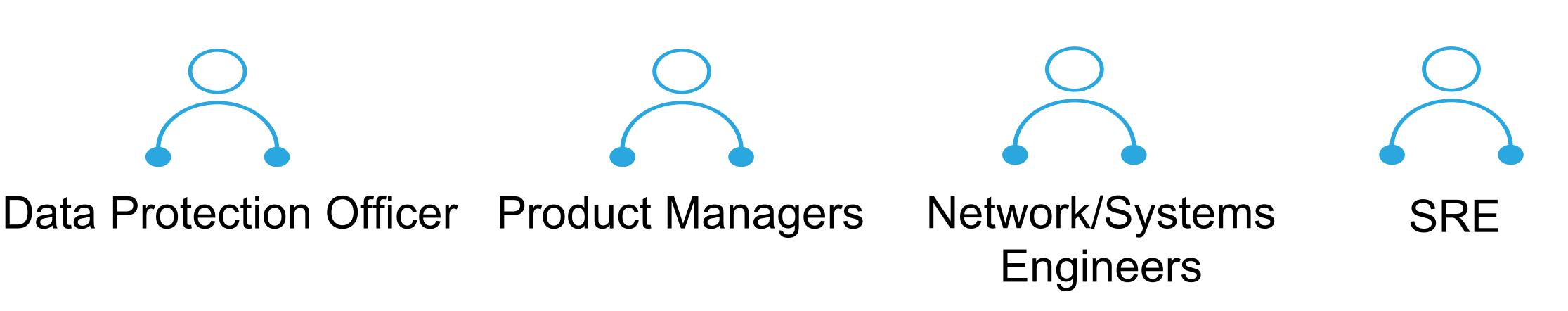
Lessons Learned Building Big Data Teams



Build well rounded teams



Other roles:





How to find people?

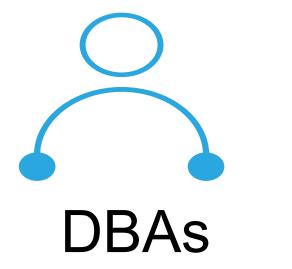
Start with people you already have, but make sure you invest in training...

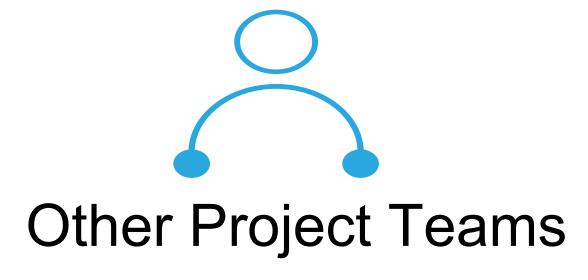
- Linux, network, DBAs -> sysadmins
- Developers -> developers
 - Easy if you're at a company like Orbitz, otherwise maybe not so much
- Analysts -> analysts
- It's not an easy path though
 - Set goals instead of micro-managing development
 - Be prepared to iterate, don't be afraid to fail



Also don't forget other teams

Communication is key







Also, don't do this:

Hi Jonathan,

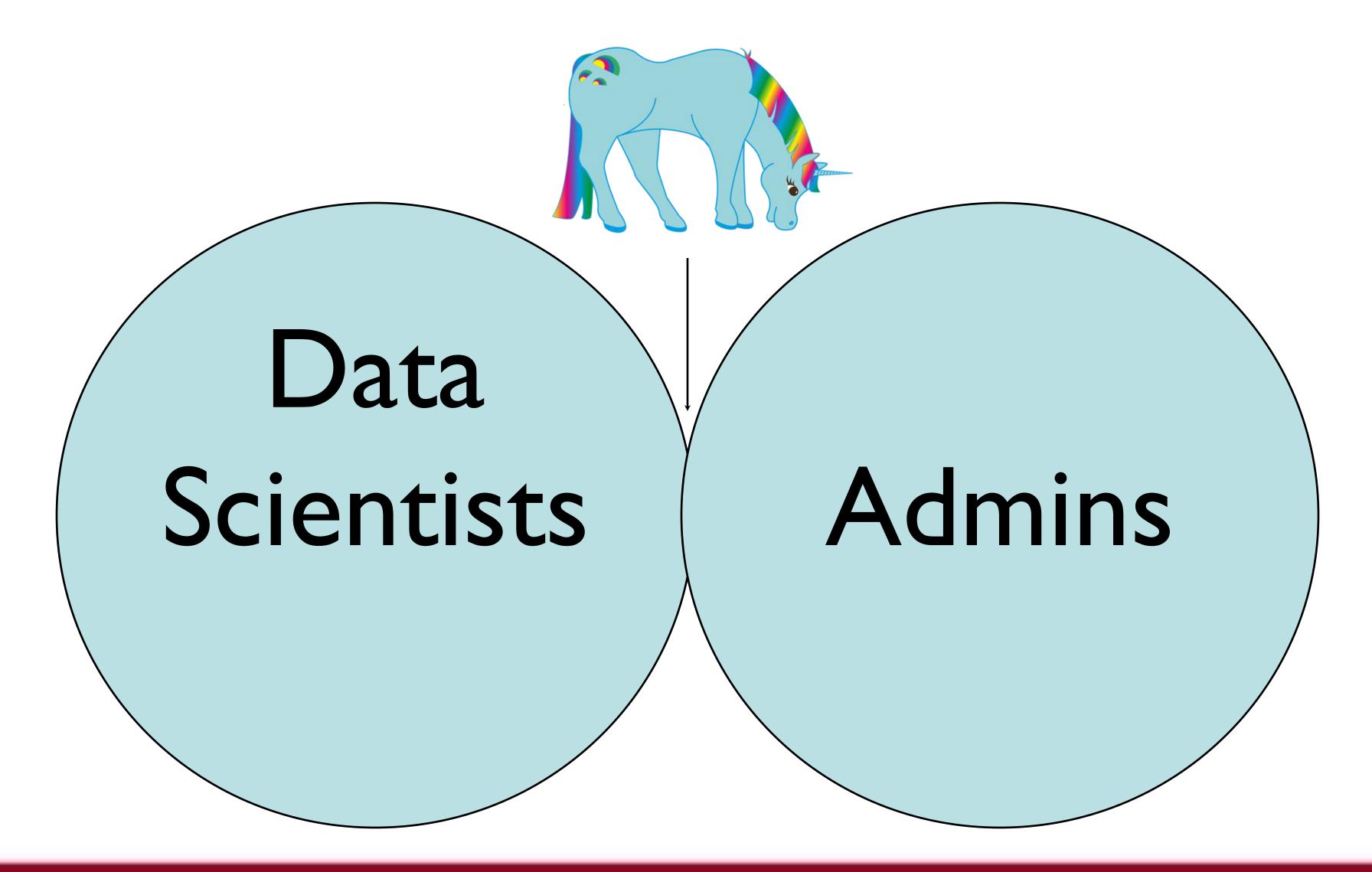
If you are interested in the following Hadoop Data Scientist/Administrator position in downtown Chicago, please email your resume and salary requirements to me.

- Develop and extend in-house data toolkits based in Python and Java.
- Consult and educate internal users on Hadoop technologies.
- Improve the performance of financial analytics platforms built around the Hadoop ecosystem.

Qualifications:

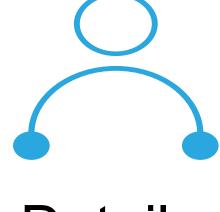
- 3+ years of experience working with Hadoop 2 (YARN), cluster management experience preferable
- 3+ year of experience with Hadoop SQL interfaces including Hive and Impala
- 2+ years of experience developing solutions using Spark
- Strong systems background, preferably including Linux administration
- Unix scripting experience (bash, tcsh, zsh, python, etc)
- Experience with DevOps tools such as SALT and Puppet as part of a CI/CD development and deployment process.
- Demonstrated ability to troubleshoot and conduct root-cause analysis



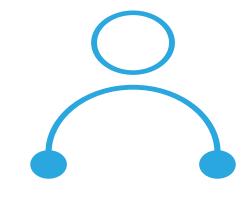




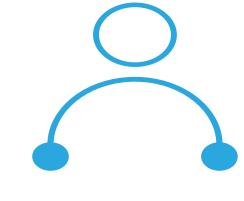
Differing Skill Sets



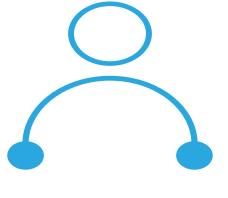
Detail-Oriented



Experimental



The Communicator



. . .

Think beyond just skills

- Also look for complementary personalities
- And avoid toxic personalities
 - But what if they're really talented?
 - See above.



Customer Engagement

 Your teams should work closely with your customers, whether they're external or internal



Managing Project Risk



Managing Risk



1 in 11.5 million

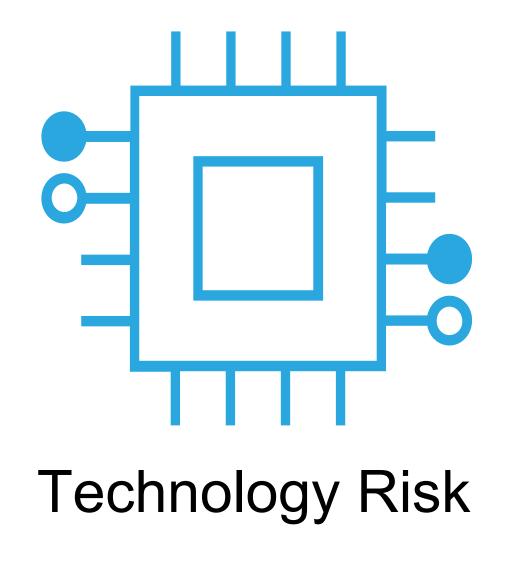


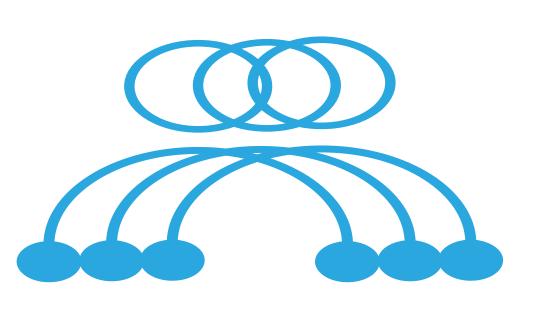
1 in 4292

Shark photo: http://www.travelbag.co.uk/



Managing Risk – Risk Categories



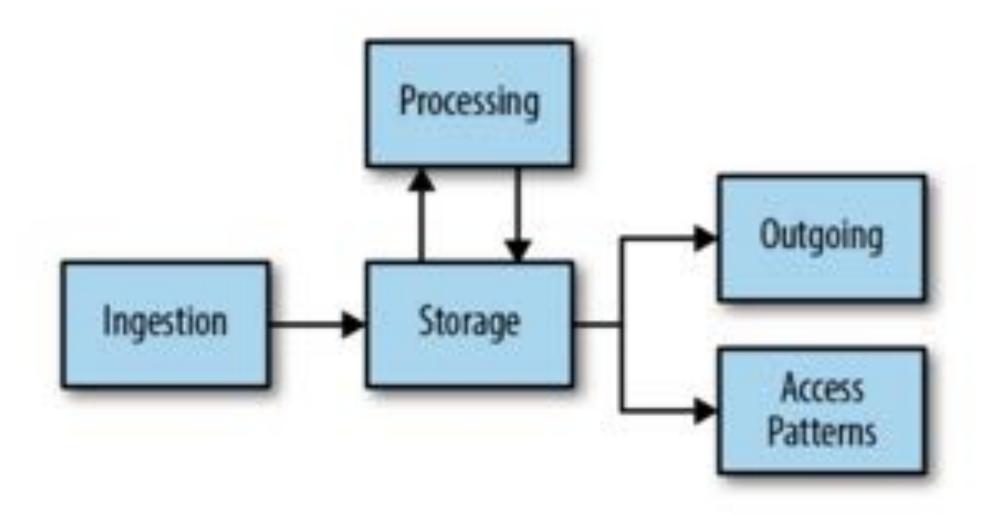




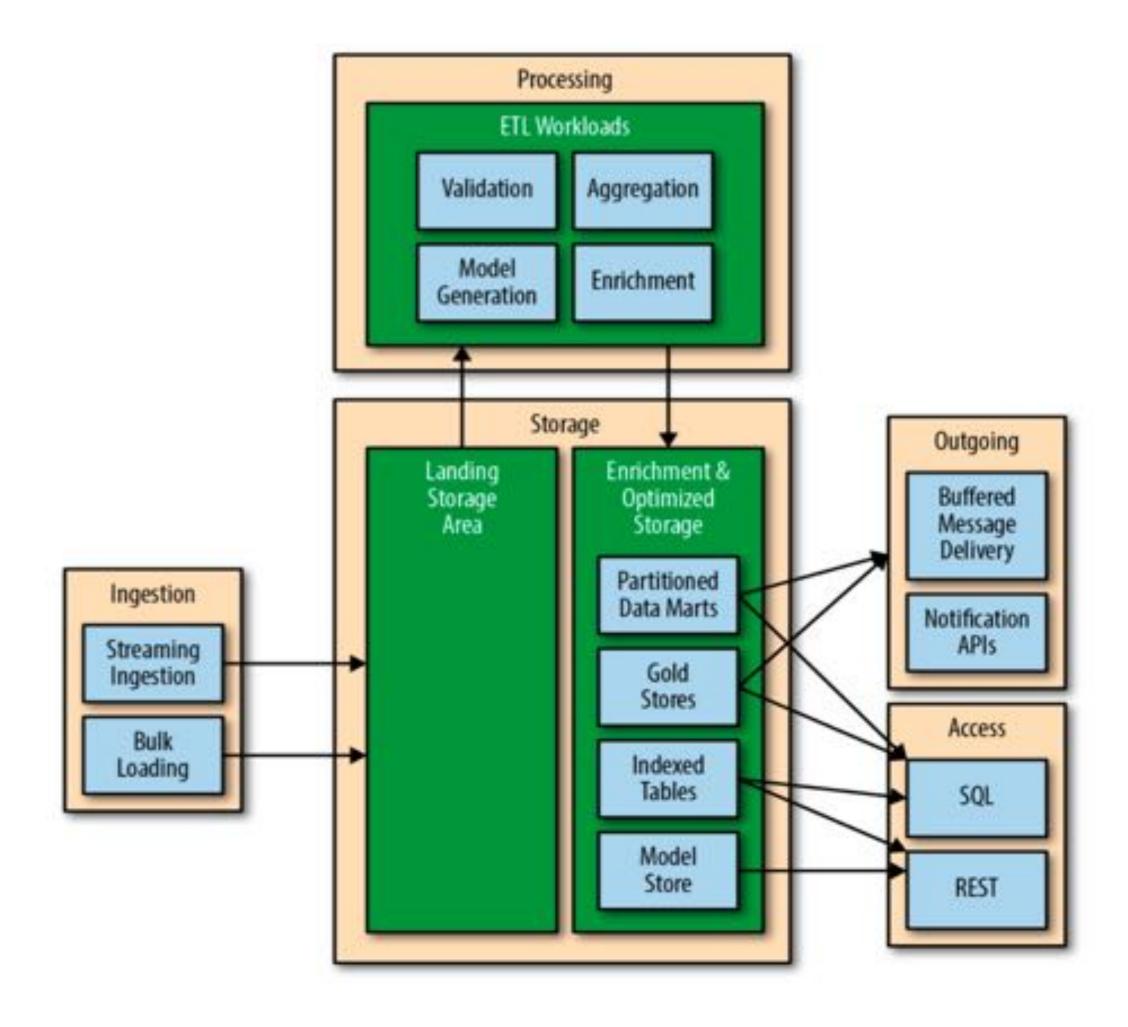
Team Risk

Requirements Risk

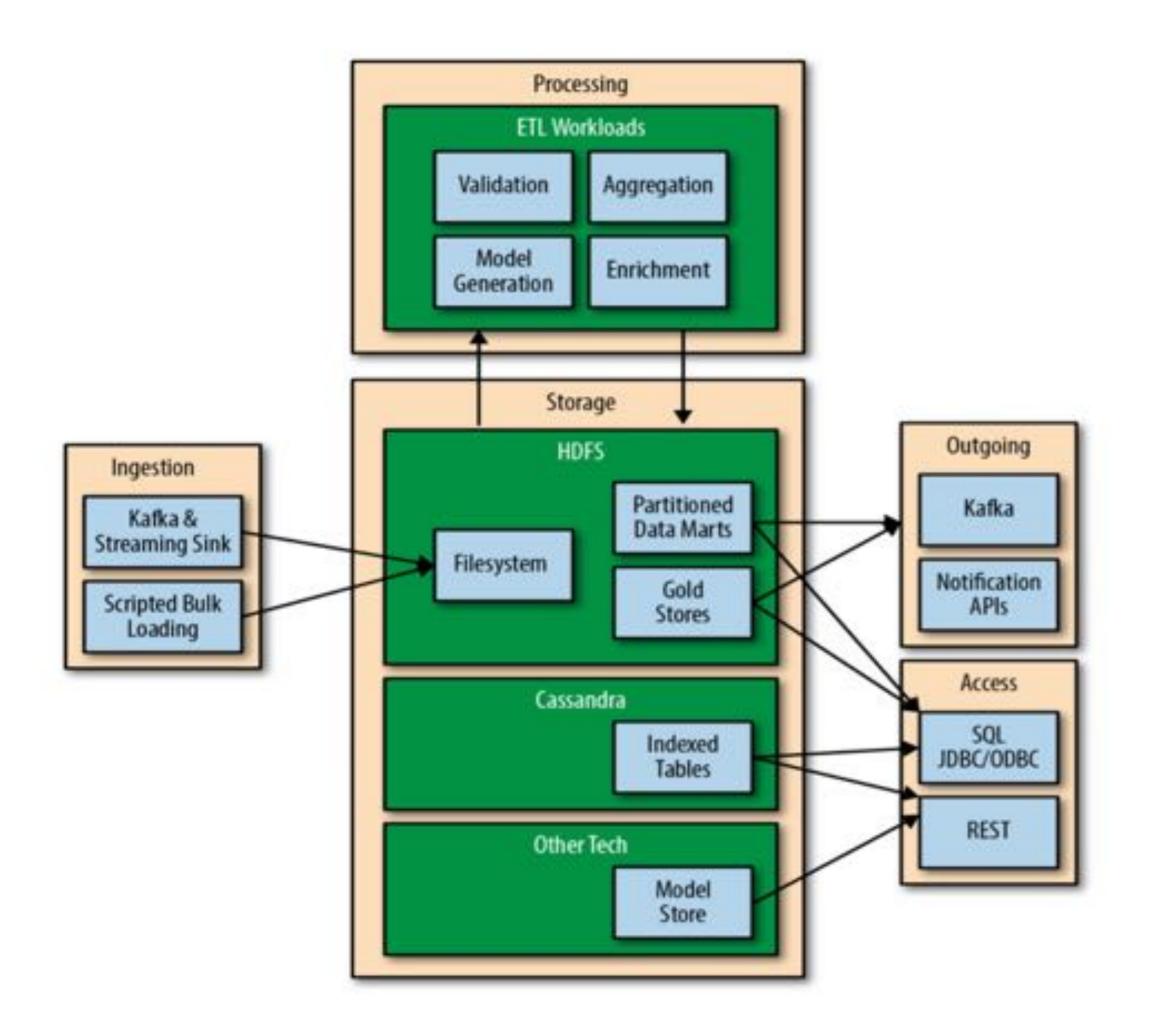














Risk Weighting

- Technology Risk
 - How much experience do we have with this technology?
 - Do we have production experience with the technology?
 - We know SQL, but what about Cassandra CQL?

- ...



Risk Weighting

- Team Risk
 - Experience level of team members
 - Team skill sets
 - Size of team

- ...

- Don't forget about other teams
 - System dependencies

- ...

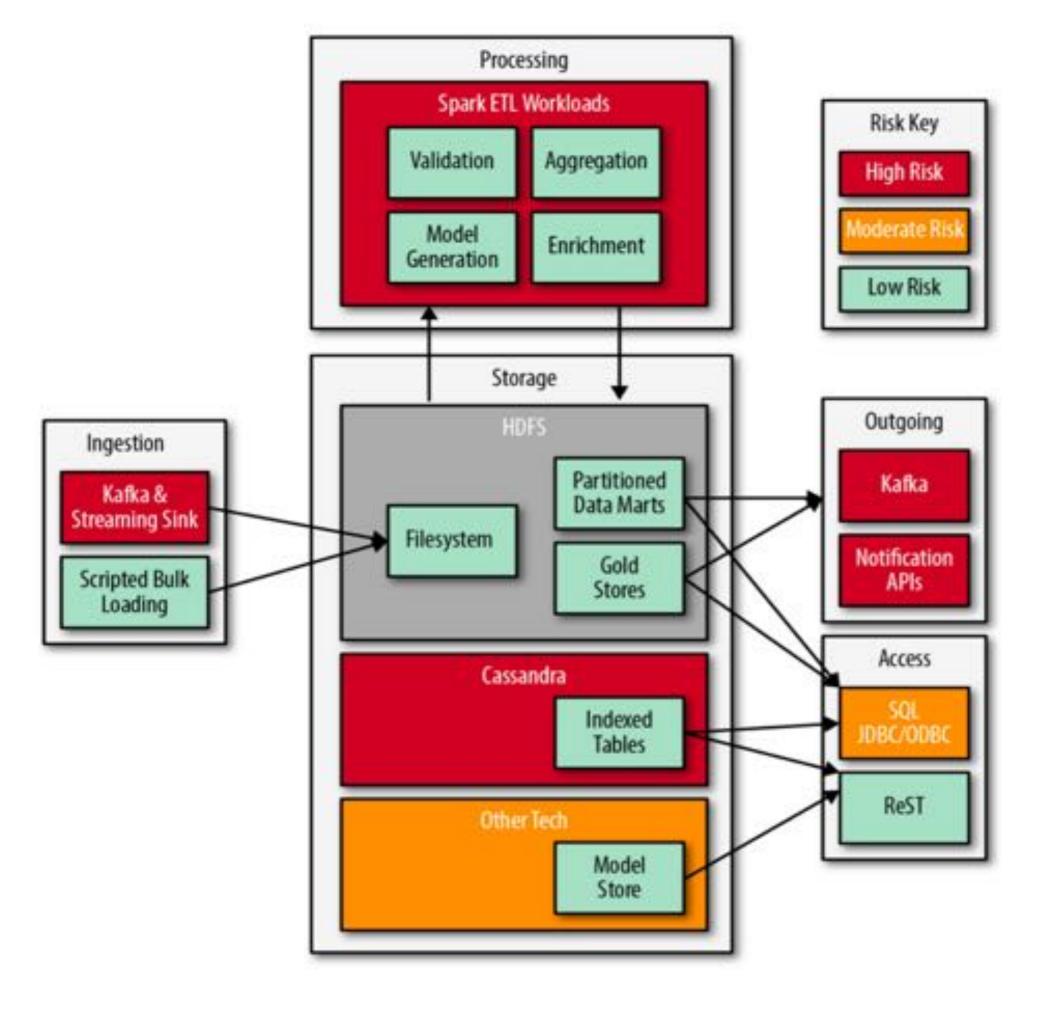


Risk Weighting

- Requirements Risk
 - Vaguely defined requirements
 - Novel requirements (e.g. stringent latency requirements)



- Cassandra
 - Limited technical experience (team risk)
 - Need to validate data model (reqs risk)
 - Stringent uptime requirements (tech risk)





- Requirements Risk
 - Ensure good functional requirements
 - Break requirements up don't boil the ocean
 - Share requirements and get buy-in from all stakeholders
 - Get agreement on scope



- Technology Risk
 - Tackle important/complex components first
 - Use external resources to help fill knowledge gaps
 - Consider replacing riskier technologies with more familiar ones



- Technology Risk
 - Use proofs of concept
 - Than throw them away



- Technology Risk
 - Use abstractions to minimize dependencies
 - Ensure repeatable build, deployment, monitoring processes



- Technology Risk
 - Start building early



- Team Risk
 - Build well rounded teams
 - Ensure communication with other teams
 - But work to reduce coupling



Communicating Risk

- Make sure stakeholders are aware of risks
 - But remember there can be risks to overstating risk
- Collaborate and get buy-in
- Share risk
- Risk can be a negotiation tool



Ensuring Data Integrity



Ensuring Data Integrity

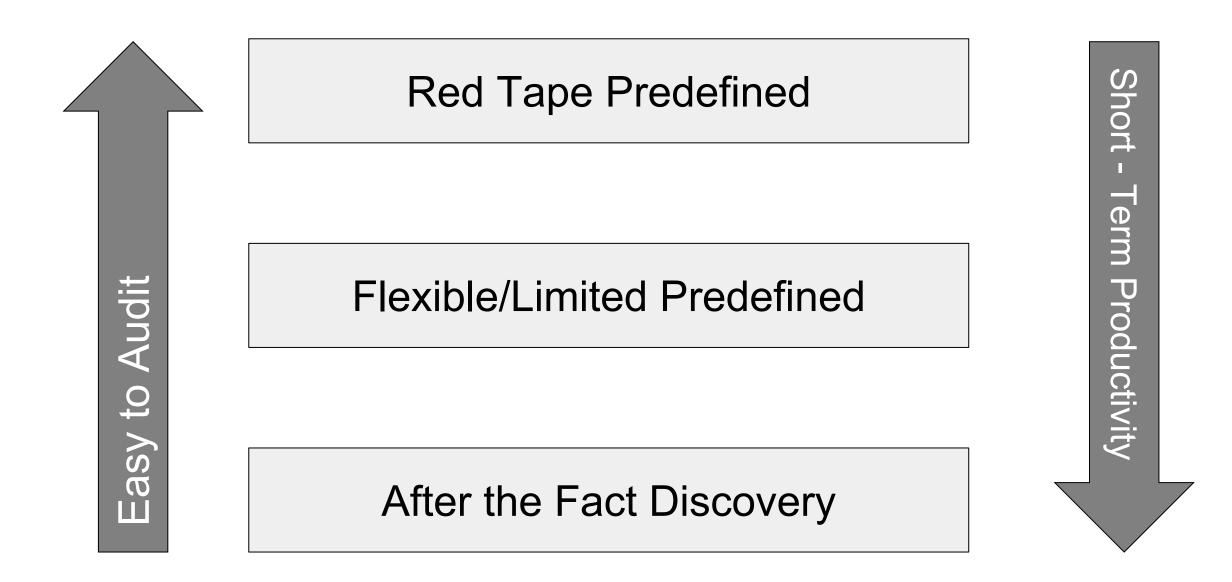
- Pre-defined vs Derived via Discovery
- Path of Fidelity
- Validation of Quality

Pre-Defined vs Derived via Discovery

- Producer Productivity vs Audit
- Consumer Consistency



Producer - Productivity vs Audit





Predefined Traps

- Centralized Reviewing Org
- High bar to on board
- Unclear schema evolution paths



Discovery Traps

- Uncommon output
- Data quality standards
- Uncommon SLAs
- The balloon problem



Consumers Point of View

- Consistency is Key
- Access to Powerful Tools
- Multiple Landing Areas is Key
 - Long Term
 - Indexed
 - Lucene Indexed
 - Streams
- Future Proofing



Path of Fidelity

- What is Fidelity
- What can we mutate

What is Full Fidelity

- The cells and their values are preserved
- Field names and definitions are preserved
- No matter where or how you access the data
- No Filtering
- No Irreversible Mutations



What can we mutate

- Tokenization
- Underlining files structions
- Storage system
- Access Path



Validate Quality

- Validation of Fidelity
- Validation of Quality



Validation of Fidelity

- Row Counts
- Check Sums
- Reversible byte by byte check



Validation of Quality

- Column level rules
- Null counts
- Field carnality
- Record counts



Metadata Management



Metadata Management

- What do we mean?
 - Understanding what data you have
 - Knowing what the data is
 - Knowing where the data is
- This is complex
 - Large number of data sources, storage systems, processing...
 - Ease of data access and creation of new data sets
 - Start planning at the beginning of your project!



Why Do We Care?

- Visibility know what data you collect and how to access it
 - Faster time to market
 - Avoid duplication of work
 - Derive more value from data
 - Identify gaps

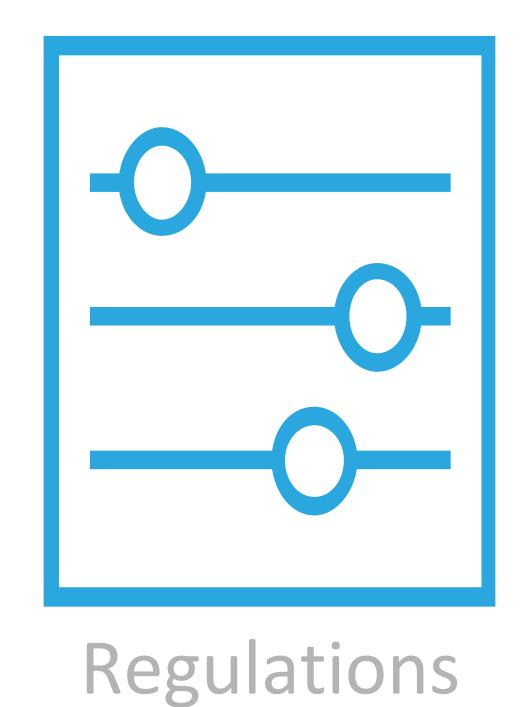


Why Do We Care?

Relationships



Why Do We Care?



GDPR, etc.



Types of Metadata

- Data at rest
- Data in motion
- Source data
- Data processing
- Reports, dashboards, etc.



Data At Rest

• Files, database tables, Lucene indexes, etc.



Data At Rest – Database Table Example

Field	Туре
User_id	Long
Receipt_num	Long
Item_purchased_id	Long
Amount	Decimal(7,2)
Timestamp	Timestamp
Method	String
Card_id	Long
Purchased_port	String



Data At Rest – Other Metadata Types



Audit Logs

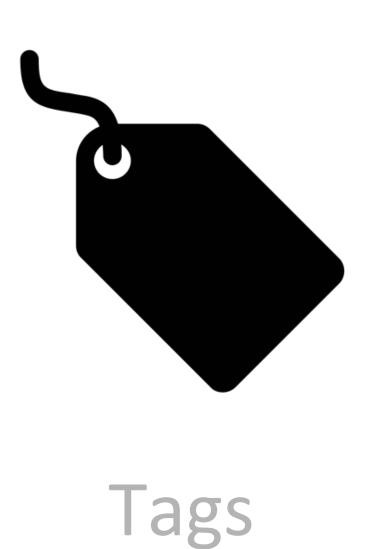


Data At Rest – Other Metadata Types

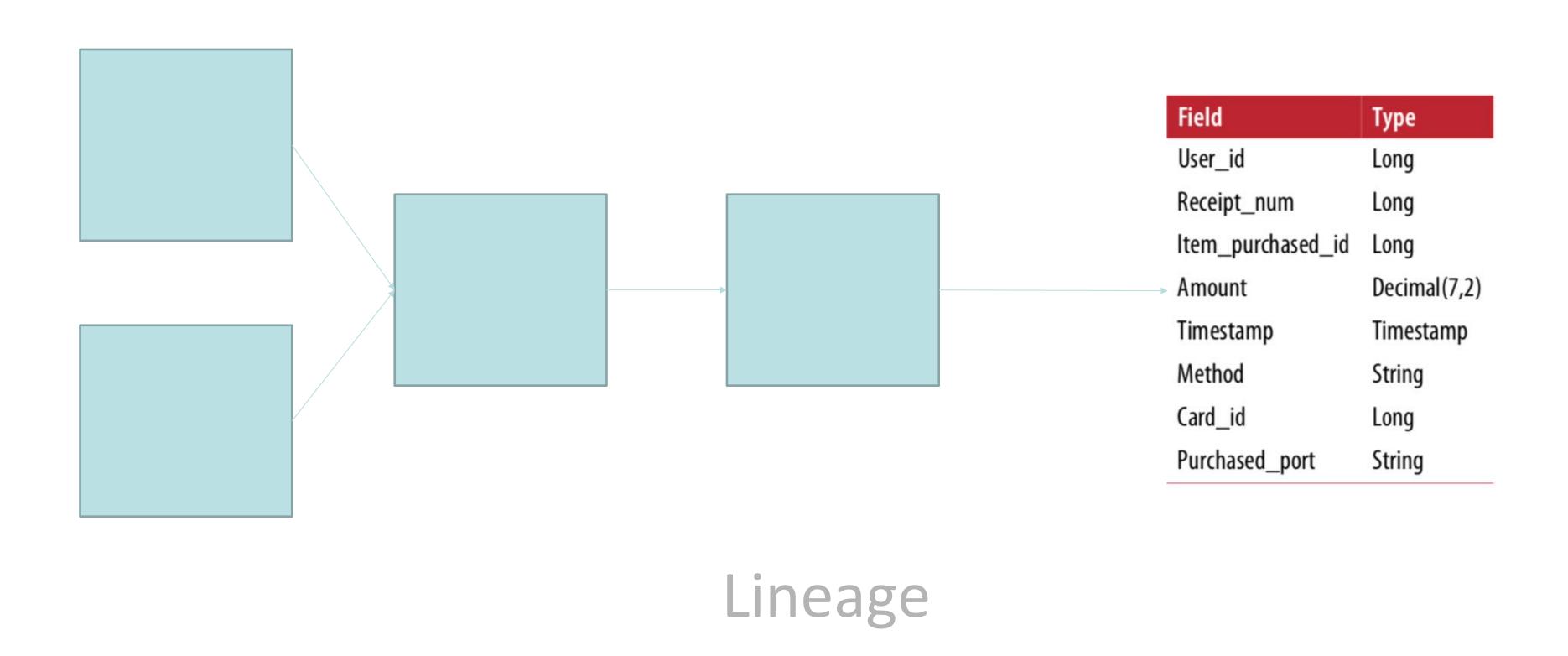




Data At Rest – Other Metadata Types



Data At Rest – Other Metadata Types





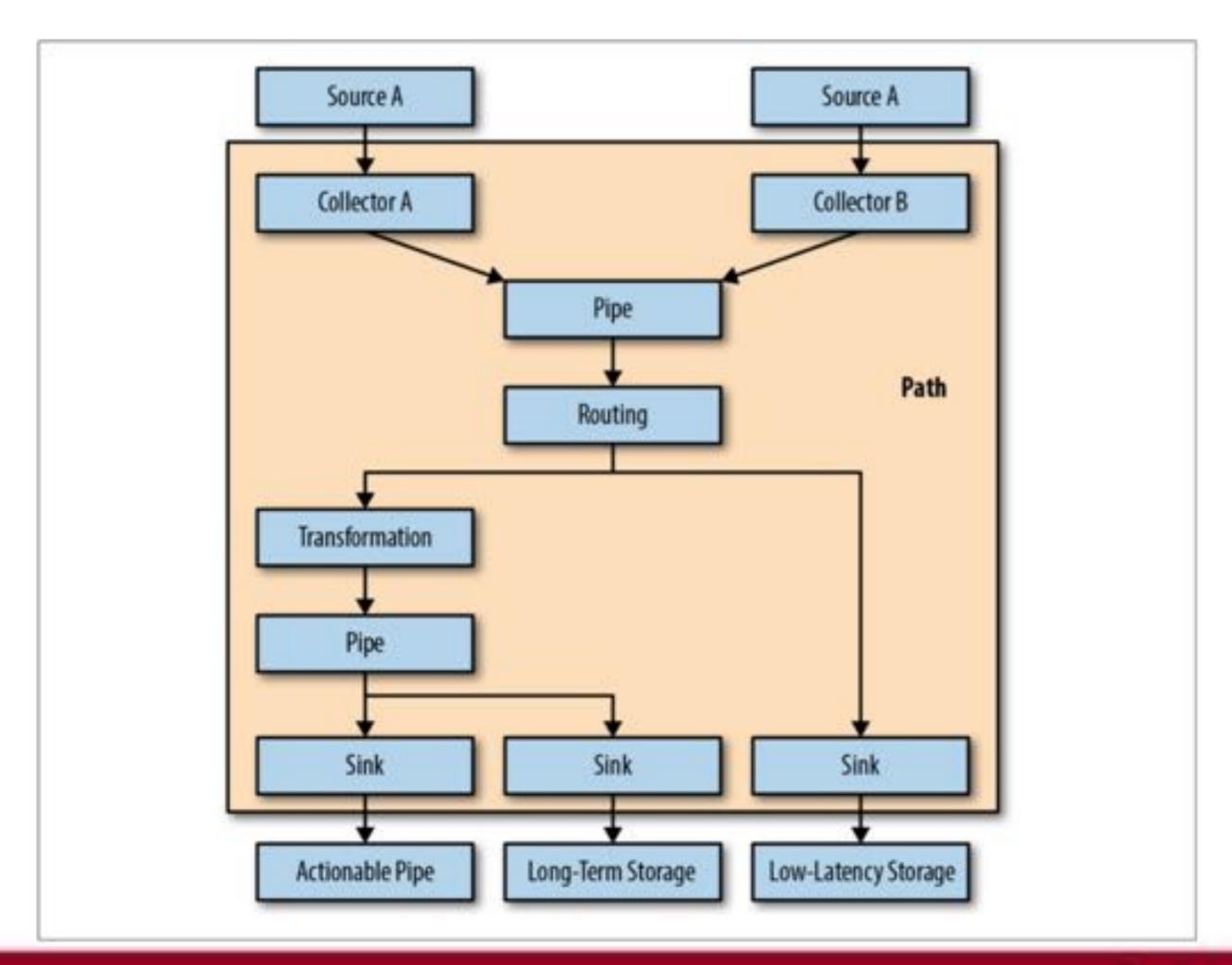
Data In Motion

- This is data that's moving through the system
 - Batch or streaming ingestion
 - Data processing
 - Derived data



Data in Motion – What to Capture

- Paths
- Sources
- Transformations
- Destinations
- Reports/Dashboards





Data in Motion — Paths

- How does the data move through the system?
 - Source systems
 - Data collection systems
 - Routing
 - Transformations
 - Etc.



Source Data

- External systems
- Internal systems



Data In Motion — Transformations

- Data format changes, for example JSON to protocol buffers
- Data fidelity is the data filtered or changed?
- Metadata about processing job names, technologies, inputs, outputs, etc.



Data Processing – Machine Learning

- More complex algorithms can require special considerations
 - Purpose of a model
 - Technologies, algorithms, etc.
 - Features
 - Datasets training, test, etc.
 - Goals of the model
 - Who owns the model?



Reports and Dashboards

- Data sources
- Any data transformations
- Information on the report's creator
- Log of modifications
- Purpose of report
- Tags



Approaches to Metadata Collection

- Declarative
 - Require and enable metadata to be created as data is added to the system
- Discovery
 - After the fact cataloging of data



How?

- Create your own solution
- Use tools provided by your vendor
- Use third party tools



Thank you!

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