



ENTERPRISE-WIDE DEMOCRATIZATION OF AI ML

JUNE 2021

MEET THE TEAM



SAMMED KUMAR

TECHNICAL PRODUCT MANAGER

Sammed is a technical product manager at D Cube Analytics, has 15+ years of Industry experience in Data and Analytics, Data governance and syndication. Sammed has managed various initiatives in the areas of building Pharma MDM platforms, Data Warehouse, and advanced personalization.



AJAY MEDA

ASSOCIATE PRODUCT ARCHITECT

Ajay is an Associate Product Architect at D Cube Analytics. He brings close to 7+ years of experience in Cloud – DevOps and product planning. He has experience in designing network and security for various applications. He has experience in designing Data Engineering and Pharma MDM platforms.



SAMUEL JAIDEEP

ASSOCIATE PRODUCT ARCHITECT

Samuel Jaideep is an Associate Product Architect at D Cube Analytics, has 7+ years of Industry experience in Web Technologies and Cloud Engineering. Jaideep has been involved in architecting and development of various Web Applications for multiple clients in the course of time.



OUR AGENDATODAY





Introduction – Overview of democratization



Digitization through AI/ML requires the following -

- Technology and tools
- People and collaboration
- Data and other assets

The intent of democratization is to

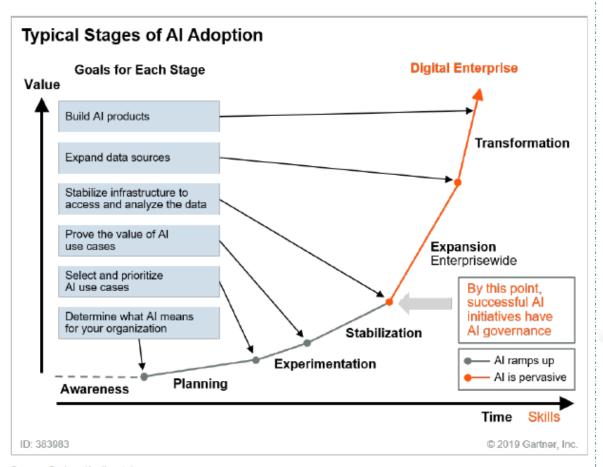
- Enable AI citizens to fully utilize the benefits of above components
- Enable AI governors to implement well-defined access policies and processes

Overall intent is to increase adoption and participation by each user persona providing

- Enhanced visibility
- Reduced process friction



Risks of having ungoverned ecosystem



Risk of penalty is very high for businesses with ungoverned analytics ecosystem Deploying models into production reliably is challenging for most firms Risks of having Lack of unified data ungoverned analytics strategy leads to delays and suboptimal business value ecosystem Majority of data science projects never make it to production

Establishing AI governance would assure organizational accountability and mitigate risks around it

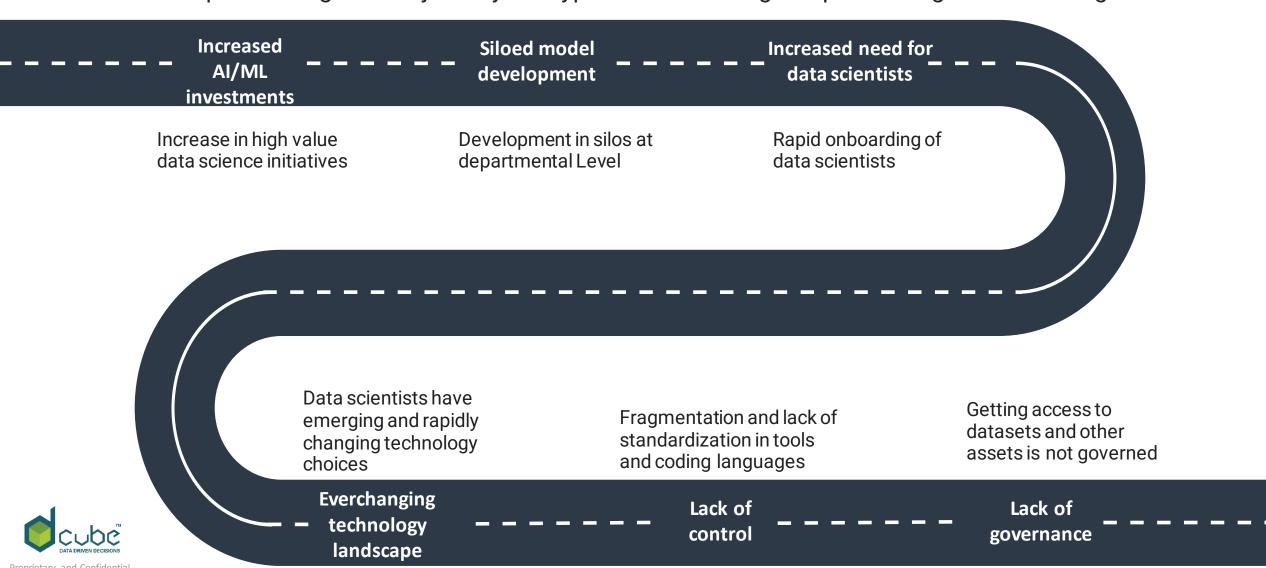




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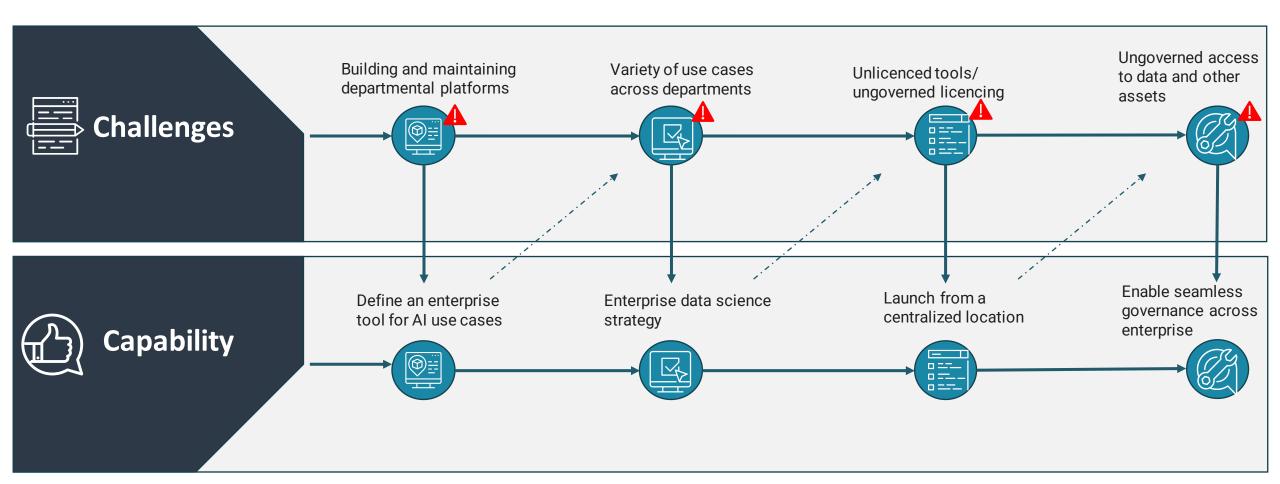
Industry patterns impacting enterprise adoption of AI

Below roadmap shows digitization journey of a typical mid-size/large biopharma organization using AI/ML.



Capability needed to address the challenges

Challenges faced by a mid-size/large biopharma organization trying to adopt enterprise governance.





Case study (1/3) – Enabled a unified data science platform built on cloud for a leading biopharma

Objective & Problem statement

- The Data Science leadership of a leading pharma company was in a need for a variety of data science capabilities across departments.
- Due to explosion in high value Data Science projects and emerging/rapidly changing technology choices, departments started using diverse and fragmented set of tools.
- Industry tools did not provide required capabilities to take models to production and operationalize them.
- There was a need of a streamlined approval process; without which getting access to data and AI tools was time consuming.

Solution approach

- Surveyed the needs of several departments having 600+ Data Scientists across departments and performed market research on this basis.
- Conducted workshop sessions involving relevant stakeholders, domain experts and analysts to brainstorm multiple use cases to capture & identify functional customizations.
- Fast-fail approach was utilized using agile delivery methodology to identify key bottle necks.
- A unified data science experience was built to handle multiple tools including Databricks, R-Studio, Shiny apps deployed on containers.

Outcomes/Impact

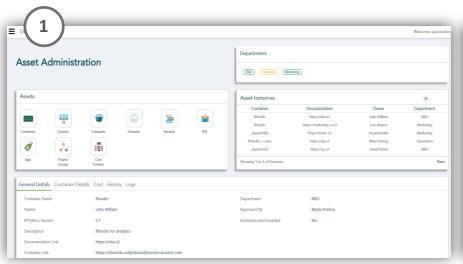
- Delivered high impact by creating a unified data science platform on cloud.
- The platform offered a very contextualized experience including integration to client cost centers.
- Considerable license cost savings were realized with users moving into cloudbased computes through the unified solution.
- Benefits to the various stakeholders are highlighted in subsequent pages.

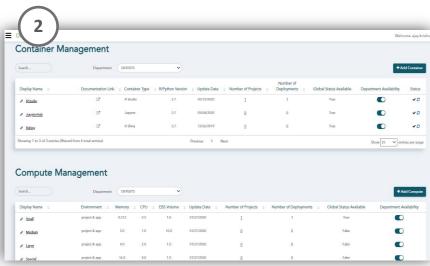


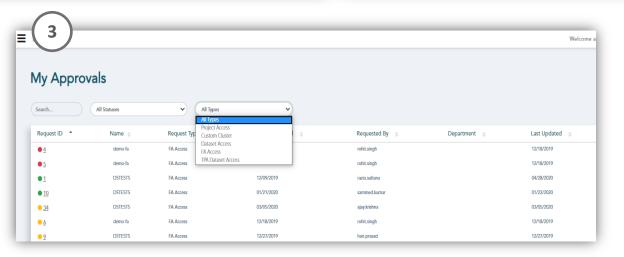
Case study (2/3) – Leadership and information governor view

The platform integrated governance process to enable frictionless provisioning of the data and other assets. This provided leadership and information governors visibility to the usage and cost of the data science resources across enterprise.

- Allows departmental heads to make different kind of assets available for its users. Different Assets are governed on the platform.
- Libraries
 - Datasets
 - Cluster templates
 - Container images
 - Packages
- Multiple workflow-based approvals were leveraged such as project access, dataset access, vendor data access and custom cluster provisioning





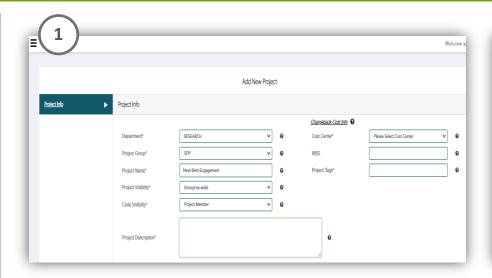


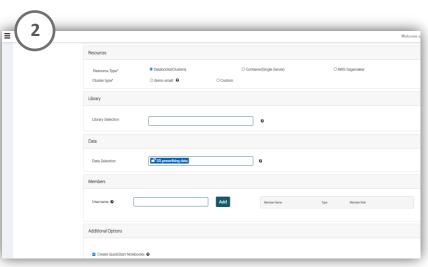


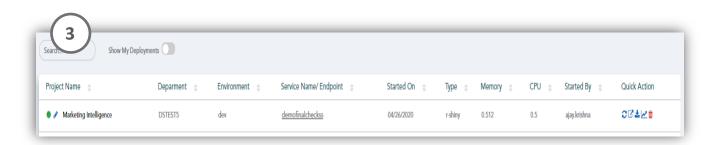
Case study (3/3) – Data Scientist/analyst view

The platform provided easy and intuitive tool provisioning to the data scientists along with collaboration capabilities using organized project structure

- 1
- Ability to publish/share reusable components and cite a project based on the level of collaboration
- Provision to restrict access at project level/department level/Enterprise level
- Provision to tag the cost center for each project
- Provision to select the server resources based on project needs. (Predefined or Custom)
 - Provision to add datasets for accessing centralized data feeds on S3 Buckets
- Provision to deploy developed apps for internal/external consumption
 - Provision to scale server environment dynamically
 - Provision to allow developed apps to be published to various stakeholders









Solution components (Focus areas for implementing governance)

AI/ML democratization



Tools and software

Multiple tools and software are available to data scientists to build their models and leverage several of them.



People and collaboration

Multiple data citizens work across organization require a seamless mechanism to collaborate and share knowledge assets.



Asset governance

Frictionless governance to govern assets that are building blocks of data science model.



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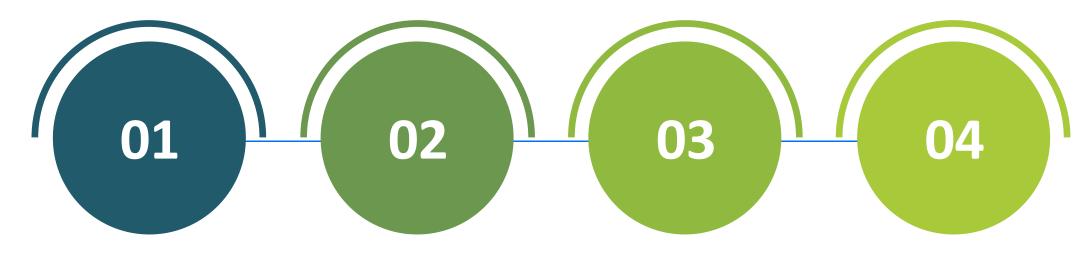
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Tools and Software

Implement centralized access to tools and software adhering with enterprise data science strategy; while providing distributed accountability and control to the data scientists.



Centralized tool access

Launch all software/tools needed from a centralized location. Ability to deploy new clusters along with required libraries for optimal model development environment.

Seamless transition to clusters/servers

Provision a dedicated server/cluster to the project team with ability to start/stop based on project use. This would provide data scientist a seamless experience shifting from their local installations.

Self-service provisioning and usage

Enable monitoring of workload and server health and usage. Provision to scale up the infrastructure if needed during development and deployment.

Regulated provisioning

Department admins can limit inventory of tools available in their departments. Any special server configuration needed can be made available through exception approvals.



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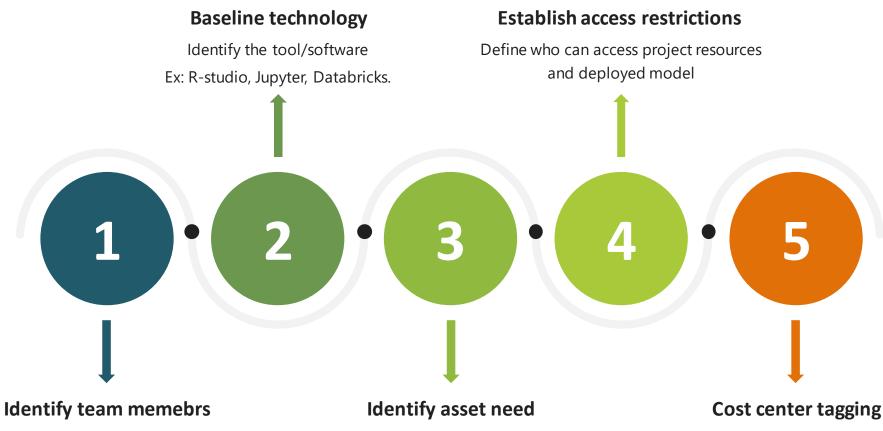
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Project structure for collaboration

Project based structure (in-line with enterprise project structure) help regulate the collaboration within the team and across the enterprise. In order to enable democratization, follow below steps during AI/ML project creation.



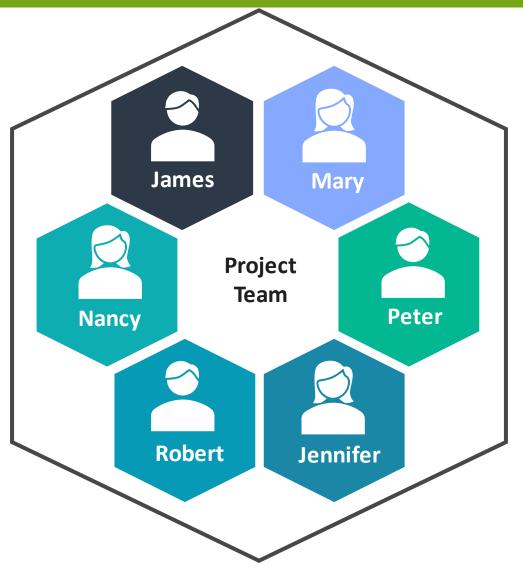


Add team members and roles. Ex: data analyst, scientist, communicator, business analyst and executive

Datasets, knowledge and servers/clusters. Ex: IQVIA HCP, RWD, visualization library

Enable tagging to help project or department level chargeback

Advantages of project based structure













Structured provisioning

Resource needs for the project are defined during project initiation. Team members can only be provisioned access based on project context.

Project workspace

Provides centralised workspace to the team members for model development

Centralised repository

Shared code-base helps the team to align on latest development and versions

Traceability to the resources used

Visibility of all the resources used in the project is available to developers and leadership

Cost centre tracking

Track usage metrics of different cloud-based services



Collaboration across enterprise



Components requiring collaboration:



Dataset (Vendor, Processed)



Libraries



Algorithms



Packages /
Containers Images



Enterprise Standards for DS



Visualisation and discovery



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Solution components (Focus areas for implementing governance)

AI/ML democratization



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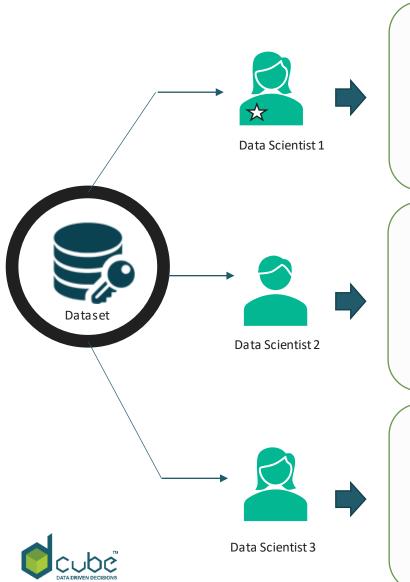


Asset governance

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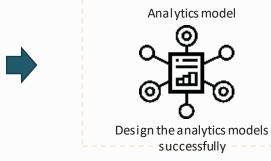


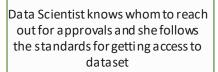
Resource access without governance

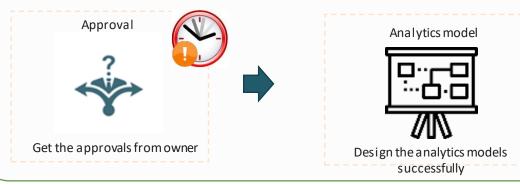


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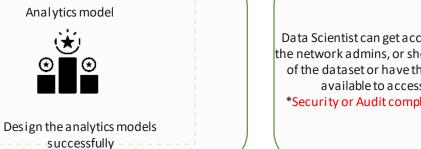






Data Scientist doesn't know whom to reach for approvals and due to which he has to reach to multiple people and get the approval from correct owner, which causes delay in project initiation *Data scientist will face difficulty



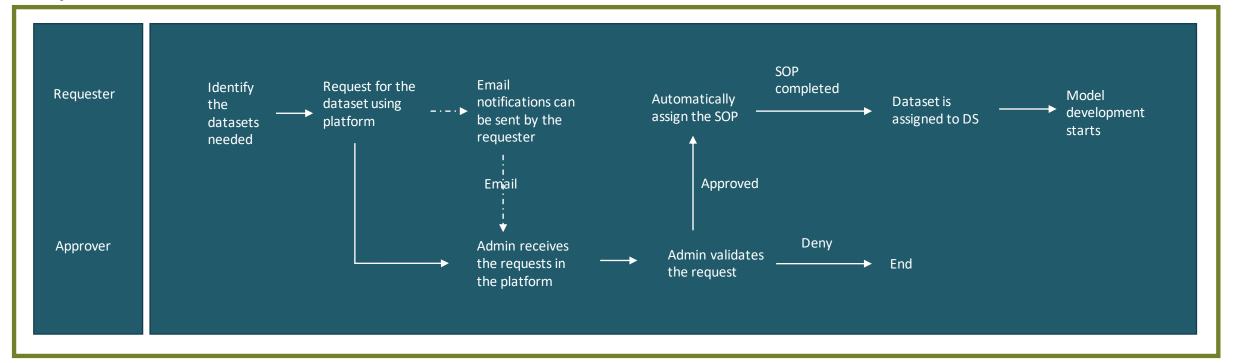


Data Scientist can get access from the network admins, or she is admin of the dataset or have the URL's available to access.

*Security or Audit complications

Workflow based governance – an illustration

Project



^{*}This is an illustrative flow given for one of the shared assets (dataset).



Datasets



Libraries



Packages / Container Images



Servers/clusters



Algorithms

Workflow based governance provides regulated asset access and visibility

Typical problems with asset access



Delayed project kick off due to provisioning roadblocks



Department level access enforcement

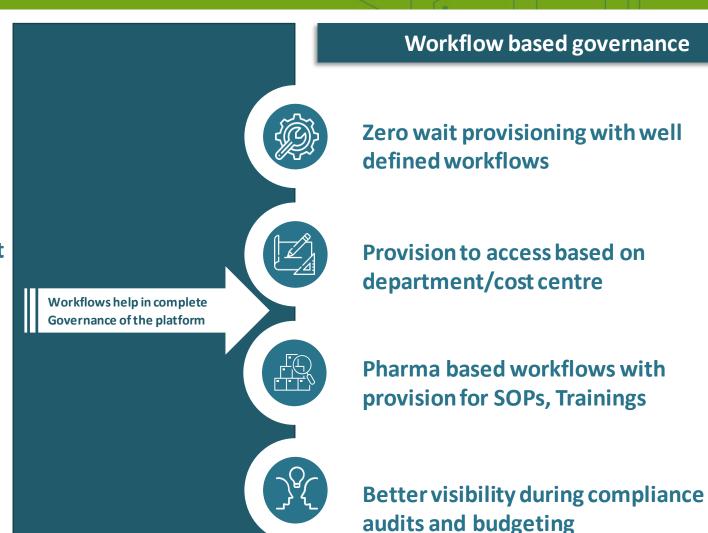


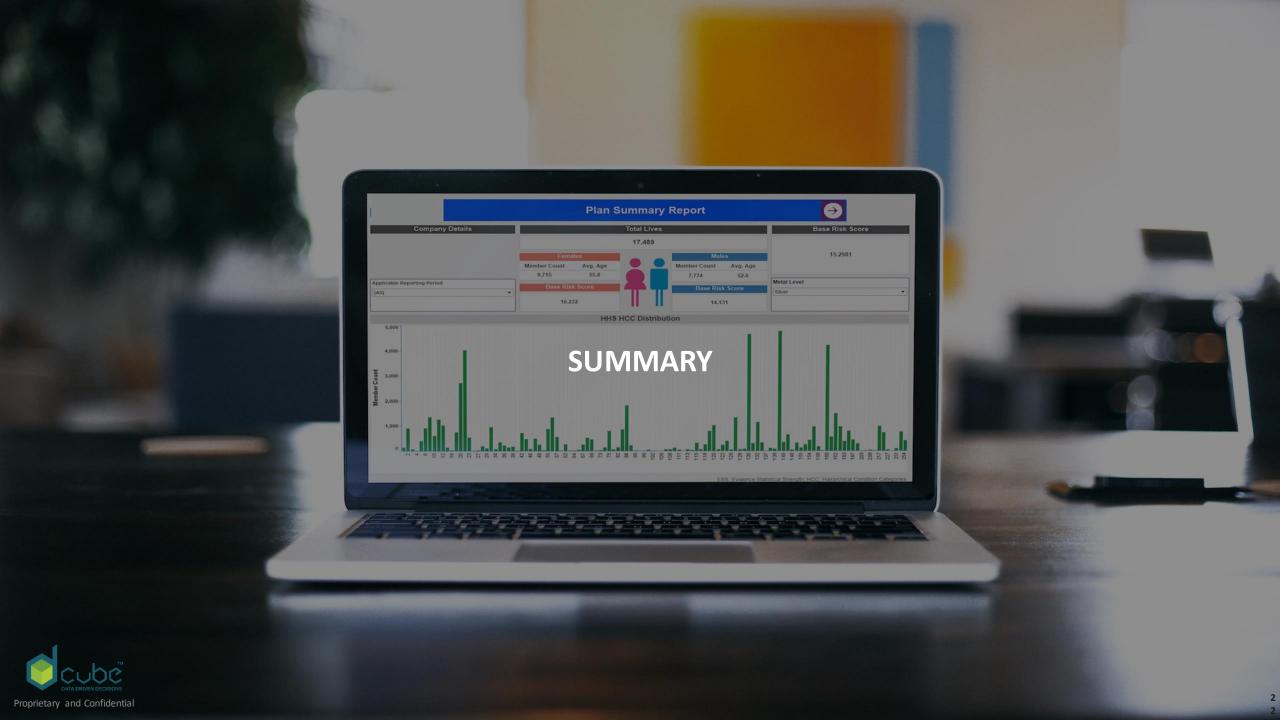
Enforcement of contract limitations



Lack of structured Spend monitoring

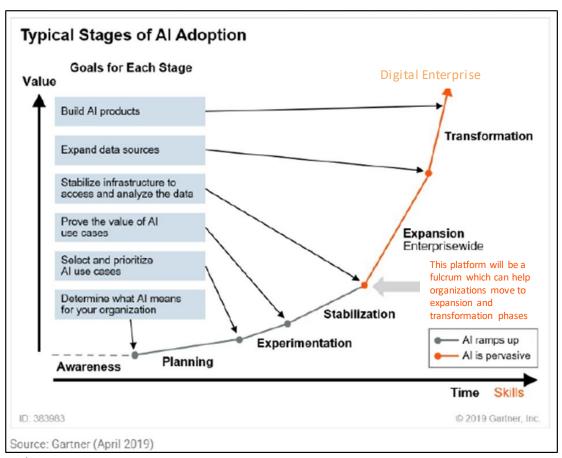


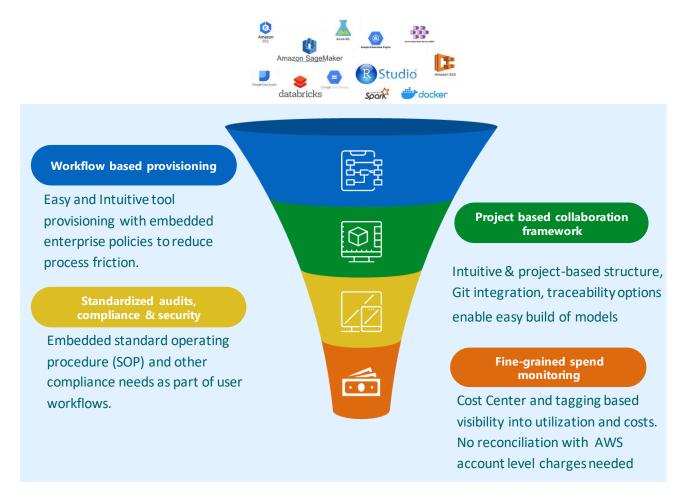




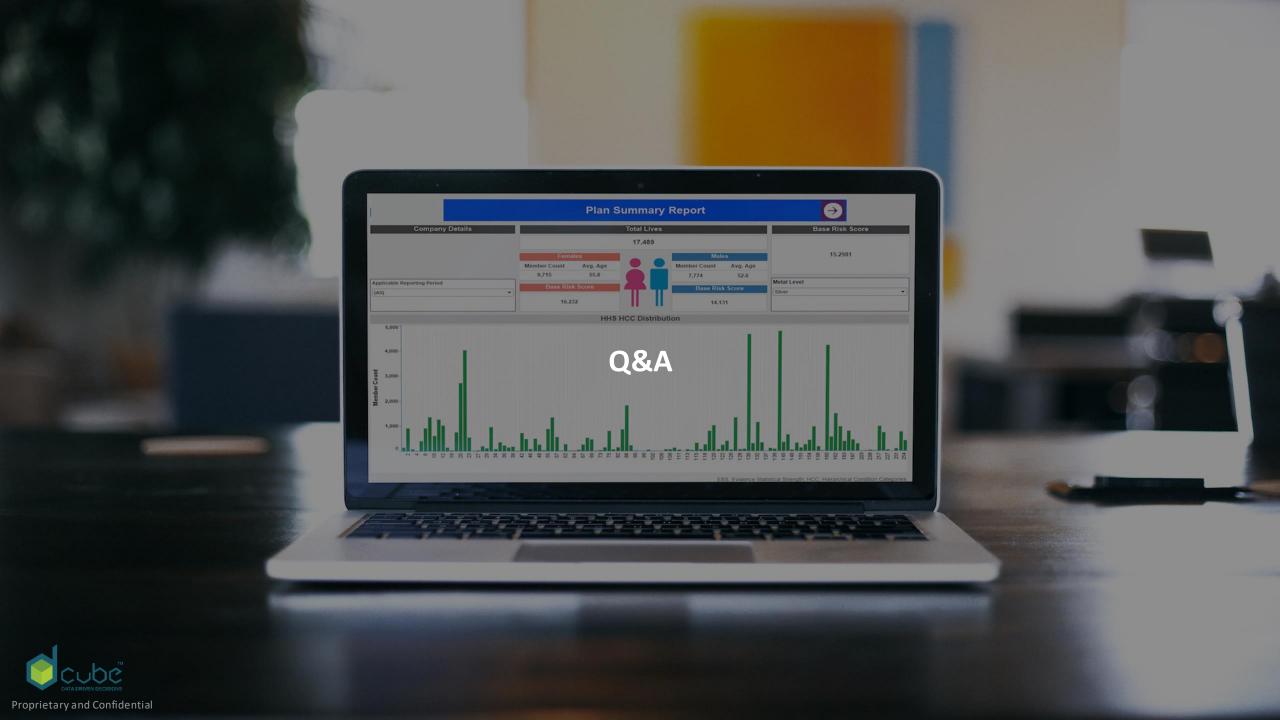
Enabling organizations to leapfrog from experimentation to expansion

Enables bio-pharma organizations to democratize analytics ecosystem through a unified platform integrating governance, security & spend monitoring





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Know more about democratization of AI/ML

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Gartner Publication (published Apr 2019)

Artificial intelligence maturity model

Gartner Publication (published Mar 2020)

How Do Data Science Workers Collaborate?

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