



ROCK TO CLOUD:
GEO-EXPLORATION
EMPOWERING
ENERGY EVOLUTION

SOCIETY OF PETROLEUM GEOPHYSICISTS INDIA

in association with



introducing for the first time

Hackathon

Competition to Collaboration

For More Details :



<https://spgindia.org/spg2025/hackathon-2025>

Registration Extended Till:

17th September 2025

Open to Professionals, Students and Others

Agenda

- About SPG
- About SLB Innovation Factori
- About Hackathon
- Introduction to AI/ML Platform
- Use case
- LLMs

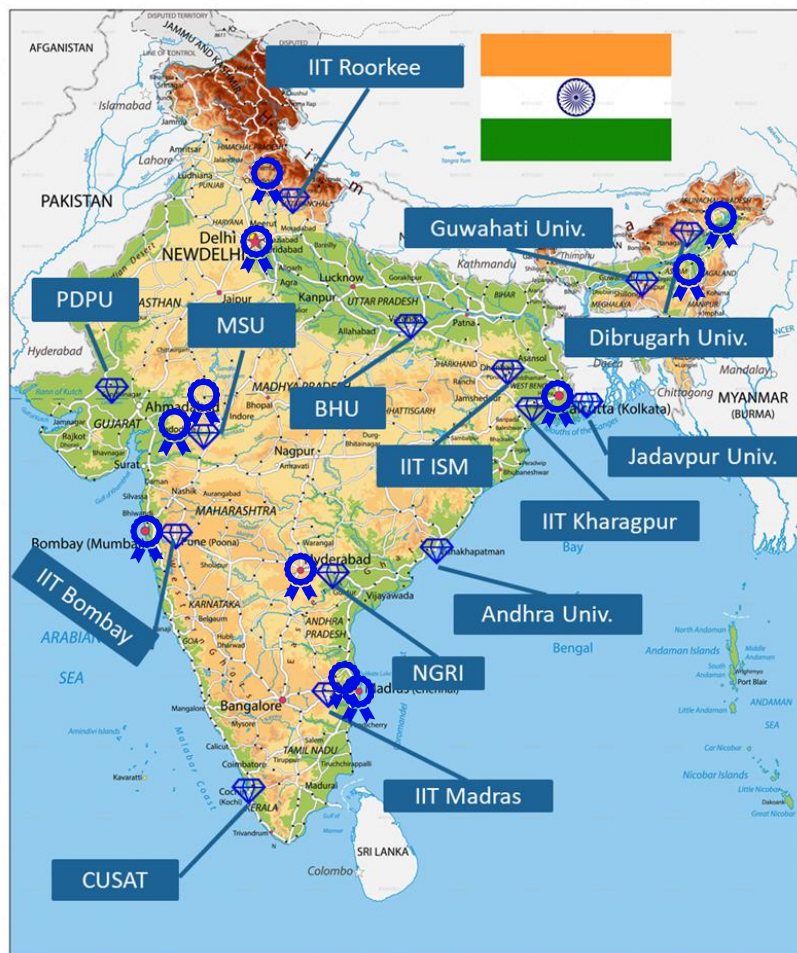




**Society of Petroleum
Geophysicists, India**
(Regd. No. - 746/ 1992 - 1993)

SPG Today

ROCK TO CLOUD:
GEO-EXPLORATION
EMPOWERING
ENERGY EVOLUTION



- ❖ **Founded on August 15, 1992 at Dehradun**
- ❖ **Registered on January 4, 1993**
- ❖ **Present Membership: 3200+**
- ❖ **11 Regional Chapters**
- ❖ **16 Student Chapters**



Endorsing Society



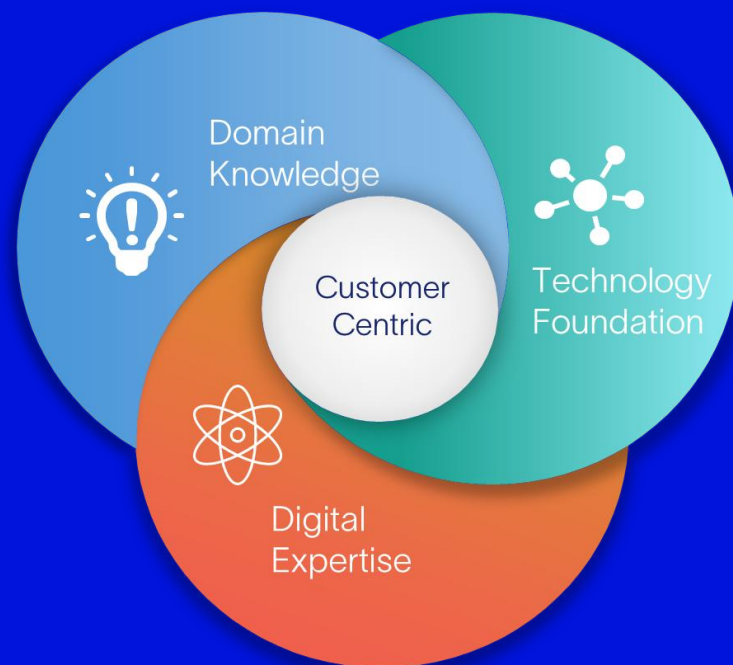
Affiliated Society



Australian Society of
Exploration Geophysicists

INNOVATION FACTORI

*Gateway for collaboration in
digital innovation*



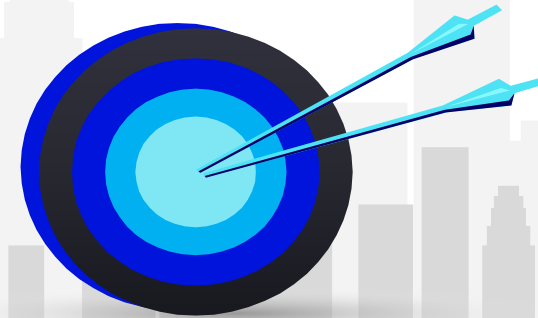
Global INNOVATION FACTORI network to
accelerate and expand collaboration and
engagement with the latest in AI and digital solutions



Vision



Mission



Support our customers to **accelerate**
adoption of AI and innovation to
solve their business challenges



- ✓ Deliver **rapid customer value** in AI and digital customer innovation engagements
- ✓ Help **accelerate customer digital transformation** and business impact of AI and digital solutions
- ✓ Be **close to our customer's digital journey** – be present where they are

Dual-Phase Hybrid Hackathon: From Competition to Collaboration

Innovative hybrid hackathon that brings together students, professionals, data scientists, and geoscientists to tackle AI-driven challenges in subsurface data analysis. This unique two-phase approach transforms traditional competition into collaborative innovation.



Hackathon Overview: A Groundbreaking Format



Phase 1: Virtual Challenge

Teams will work independently on a common dataset, developing their own unique approaches to extract insights from subsurface data.



Phase 2: Collaborative Finals

Top performers are reorganized into diverse, shuffled teams for an intensive collaborative challenge with a fresh dataset.

Who Can Participate?

Open to students, professionals, data scientists, and geoscientists interested in applying AI to subsurface datasets. No prior collaboration between these groups required.



Phase 1: The Challenge

Seismic or well logs interpretation or Insight Generation Using
AI



Phase 1: Virtual Screening



Define Problem Statement

Clearly define a specific problem using the provided dataset.



Develop AI Prototype

Build a functional AI prototype to address the problem.



Tool Agnostic

Utilize any tools, including coding frameworks and AI assistants.



Leverage Dataiku

Integrating Dataiku earns extra evaluation marks.



Scope Limitation

Avoid solutions focusing on seismic processing.

Possible Areas of Focus



- Seismic interpretation
- Well log interpretation & processing
- Generating insights from technical reports
- Integrating seismic, well, and report data



Phase 1: Key Expectations & Judging Criteria

Key Expectations

- 1 Demonstrate innovation, creativity, novelty in your approach.
- 2 Show real-world impact (e.g., cost reduction, time saving, efficiency, or accuracy improvement).
- 3 Deliver a working end-to-end prototype of your solution.
- 4 Present your solution with a clear and concise pitch deck.
- 5 Include complete code and comprehensive documentation with your submission.



Judging Criteria

- 1 **Innovation:** How unique and forward-thinking is your solution?
- 2 **Creativity & Novelty:** Originality in problem-solving and implementation.
- 3 **Real-World Impact:** Potential for tangible benefits in the industry.
- 4 **Clarity of Pitch Deck:** Effectiveness in communicating your project's value.
- 5 **Completeness of Solution:** Functionality and robustness of the working prototype.
- 6 **Use of Dataiku Platform:** Strategic and effective integration of Dataiku features.

Phase 1 Submission Requirements

For Phase 1, each team must submit their solution alongside a comprehensive pitch deck (**Max 10 Slides**) and adhere to specific expectations designed to foster innovation and real-world impact.

Problem Statement

Clearly define the challenge your team is addressing.

Proposed Solution

Detail your AI/GenAI-driven approach and methodology.

Prototype Demo/Architecture

Illustrate your workflow, model design, or include screenshots of your prototype.

Innovation & Creativity

Highlight what makes your approach unique or novel.

Why It Matters (Need for Solution)

Explain the business or technical importance of your solution.

Data & Tools Used

Provide dataset references (with citations) and list all tools/platforms leveraged.

Value Addition / Impact

Demonstrate how your solution reduces cost, saves time, improves accuracy/efficiency, or adds industry value.

Conclusion & Next Steps

Summarize your results and suggest future improvements or applications.

Phase 1: Submission Details



Pitch Deck

Submit your comprehensive pitch deck, detailing your problem, solution, and impact. Acceptable formats include PDF or PPT.



Code & Documentation

Provide your complete Dataiku project along with thorough documentation explaining your methodology, setup, and results.

Submission Deadline

3rd October 2025

Please ensure all deliverables are submitted by the specified deadline to be eligible for evaluation in Phase 1. Further instructions on the submission portal will be provided closer to the date.



Hackathon Platform: Dataiku

(Supported by SLB)



AI for Everyone

A powerful low-code/no-code platform for building and deploying machine learning models with an intuitive visual interface, integrating Generative AI for rapid iteration on AI solutions.



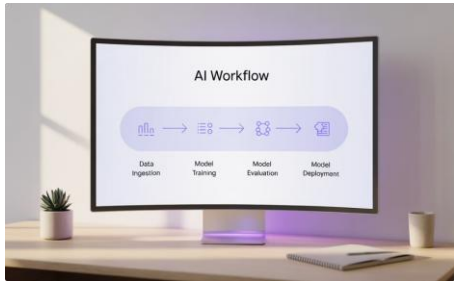
Accessible to All

Beginner-friendly yet expert-capable, Dataiku supports both technical and non-technical participants, fostering effective collaboration among data scientists, geoscientists, and students.



Comprehensive Training & Support

SLB will provide extensive training sessions and ongoing support covering Dataiku fundamentals, advanced features, and best practices to maximize project potential.



Integrated Environment

Dataiku offers a holistic environment for the entire AI lifecycle, from data ingestion to model deployment, streamlining workflows and fostering rapid experimentation in a fast-paced hackathon.



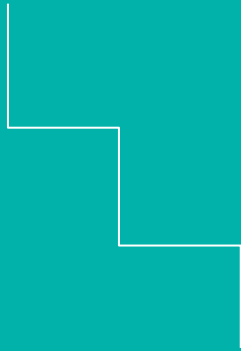
Evaluation Advantage

Effective and creative use of Dataiku is a key judging criterion, showcasing participants' ability to translate ideas into actionable AI solutions through innovative problem-solving.

AI/ML platform

An introduction to Dataiku





1. Dataiku Presentation
2. DSS concepts
3. Hands-on – Facies Classification

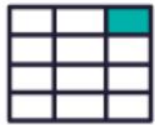
Dataiku presentation



Data Science and Analytics

The Platform for Everyday AI

Dataiku enables everyone across the organization to participate in building AI and consuming AI-driven applications.



Data Preparation



Visualization



Machine Learning



Data Ops

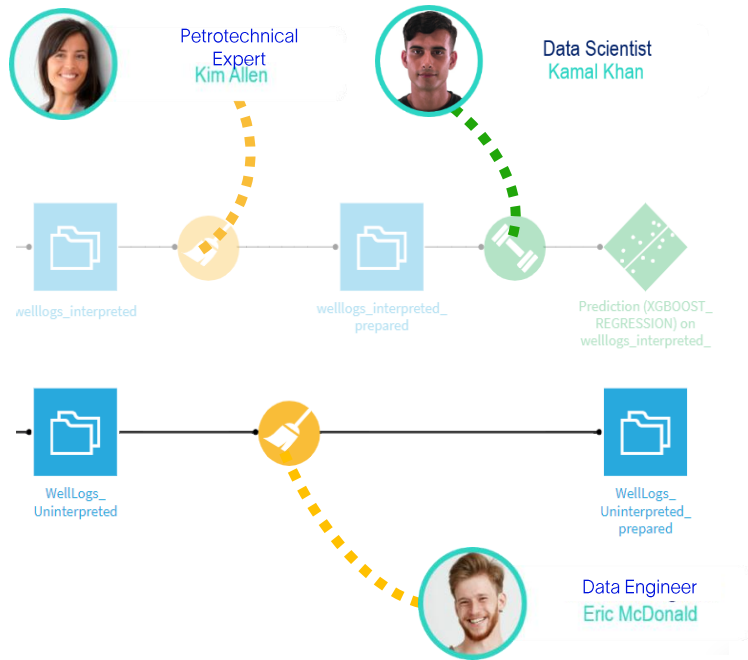


MLOps



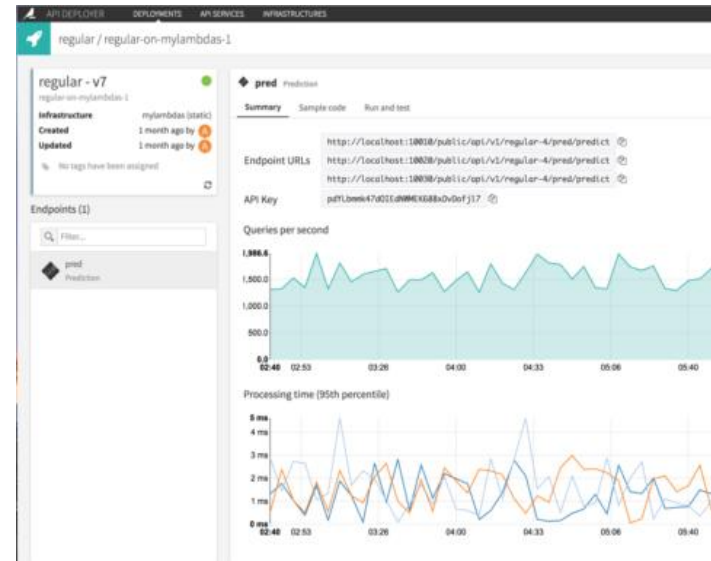
Analytic Apps

Data Science and Analytics



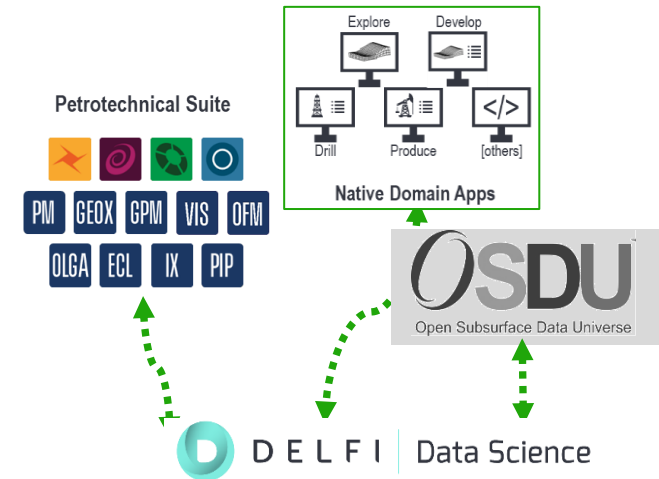
Democratized AI/BI

- AI and BI for everyone
- No code, low code, high code
- Easy to use visual modeling interface



Deployment at Scale

- AI everywhere
- Ease of enterprise deployment
- Multi-cloud deployment



Enrich Workflows

- Integrate to any data source
- Openness to provide integration with applications

Data Source integration – Discover with ease

Microsoft Azure
Blob Storage



teradata.

ORACLE



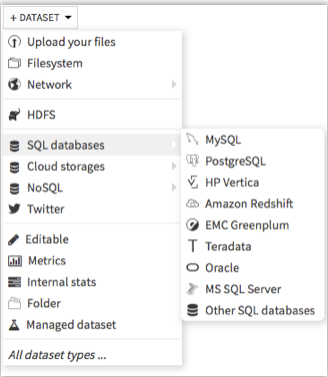
Azure Synapse Analytics

Dataiku DSS

Data Access and Processing

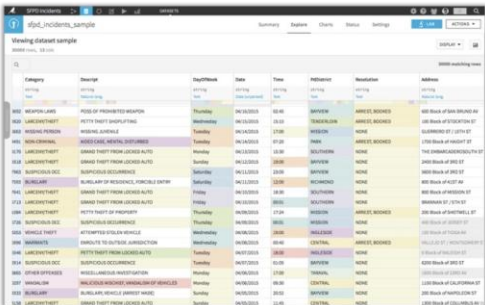
CONNECT TO YOUR (MANY) DATA SOURCES

- **Click based** connection to your datalake, databases, flat files or any other source
- Native connectors for most common technologies (SQL, Hadoop, Spark, NoSQL...)
- Find relevant data with the catalog



DATA EXPLORATION

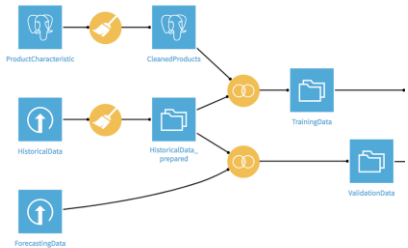
- Explore your data easily, investigating potential quality issues.
- Get rich insights into every column with built-in charts and summary information

Category	Date	Time	Address
Category 1	2017-01-01	10:00	10000000000000000000
Category 2	2017-01-01	10:00	10000000000000000000
Category 3	2017-01-01	10:00	10000000000000000000
Category 4	2017-01-01	10:00	10000000000000000000
Category 5	2017-01-01	10:00	10000000000000000000
Category 6	2017-01-01	10:00	10000000000000000000
Category 7	2017-01-01	10:00	10000000000000000000
Category 8	2017-01-01	10:00	10000000000000000000
Category 9	2017-01-01	10:00	10000000000000000000
Category 10	2017-01-01	10:00	10000000000000000000

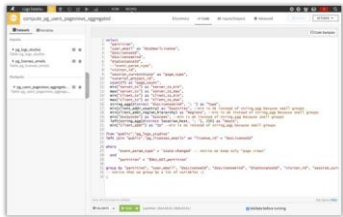
DEVELOP REUSABLE/MAINTAINABLE DATA FLOWS

- Simple representation of overall data processing despite complexity of underlying operations
- Portability of in-built recipes to various execution engines



DATA PREPARATION

- Use the 100+ in-built processors to perform advanced operations in a few click
- Extend the built-in tools by writing custom formulas or code for bespoke transformations.

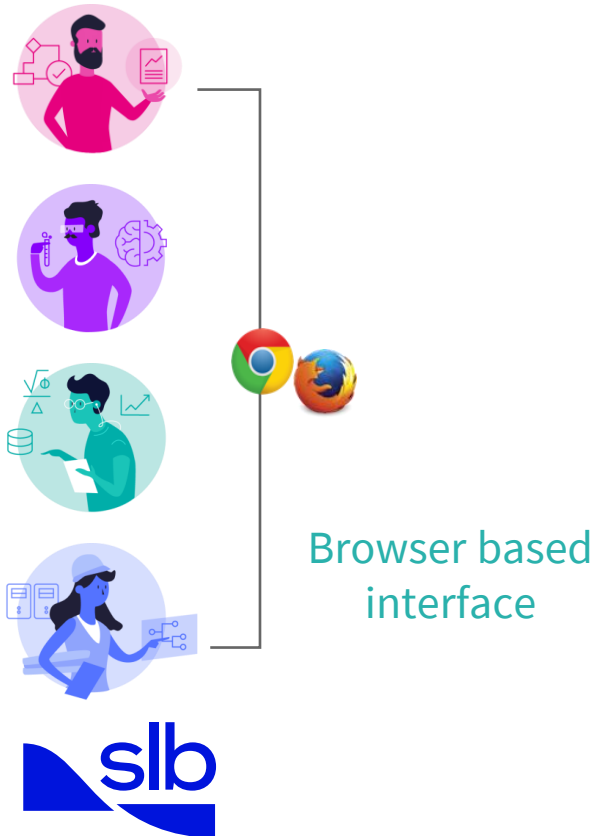


COMMUNICATING INSIGHTS

- Convey insights with 25+ standard charts and map types.
- Easy to integrate with popular visualization solutions like Tableau, Power BI and Qlik

Dataiku DSS

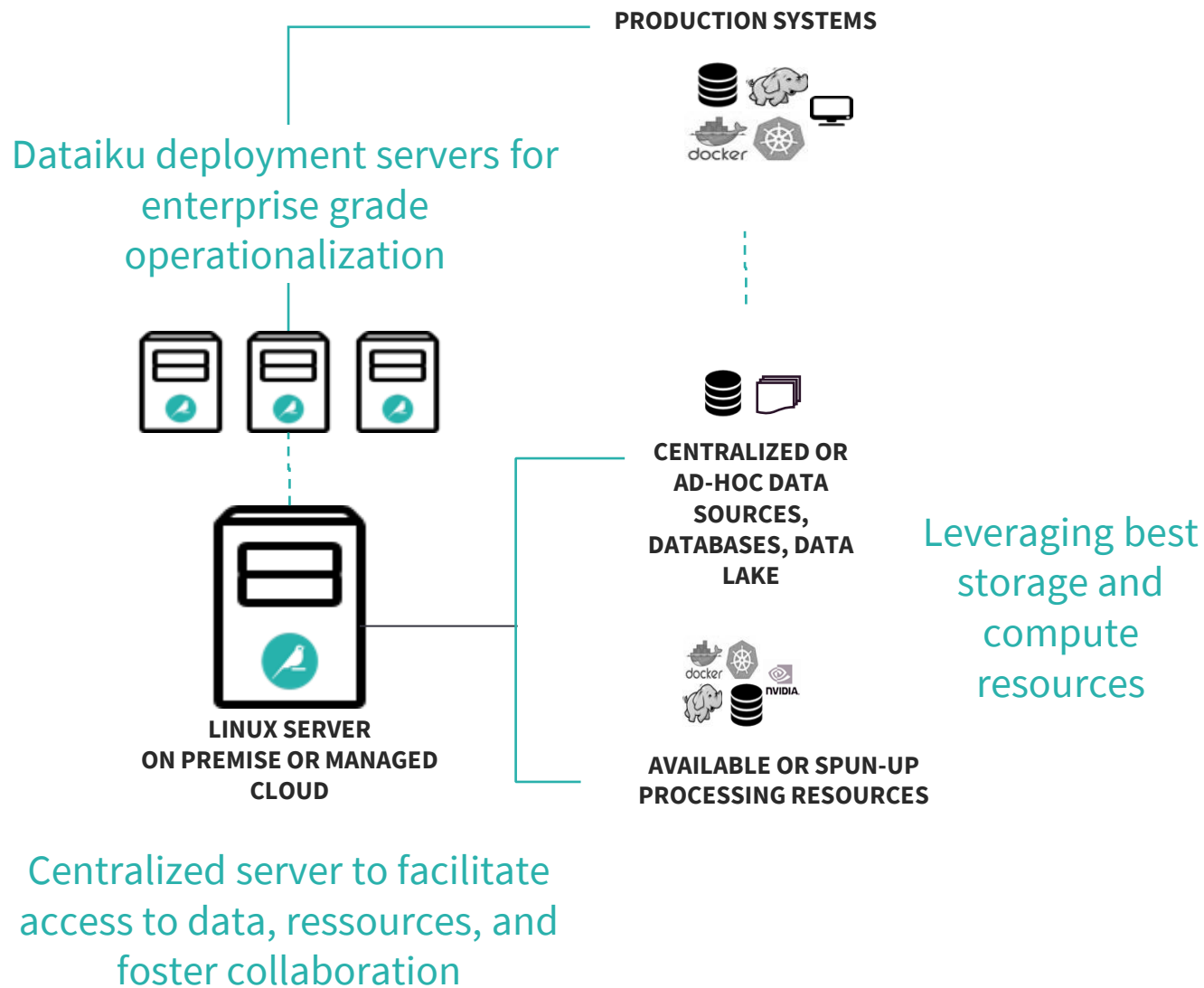
Solution Overview: Architecture



Dataiku DSS

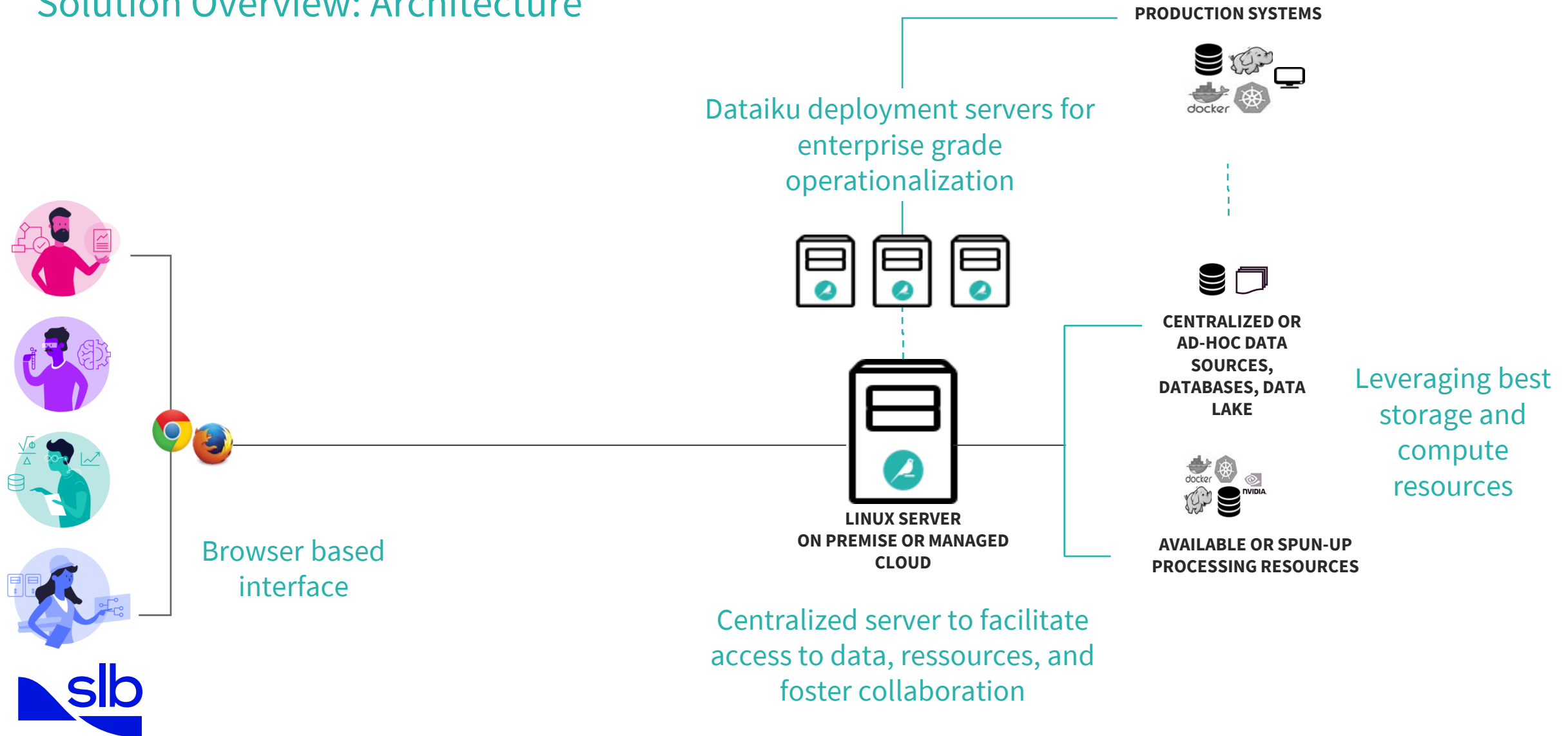


Solution Overview: Architecture



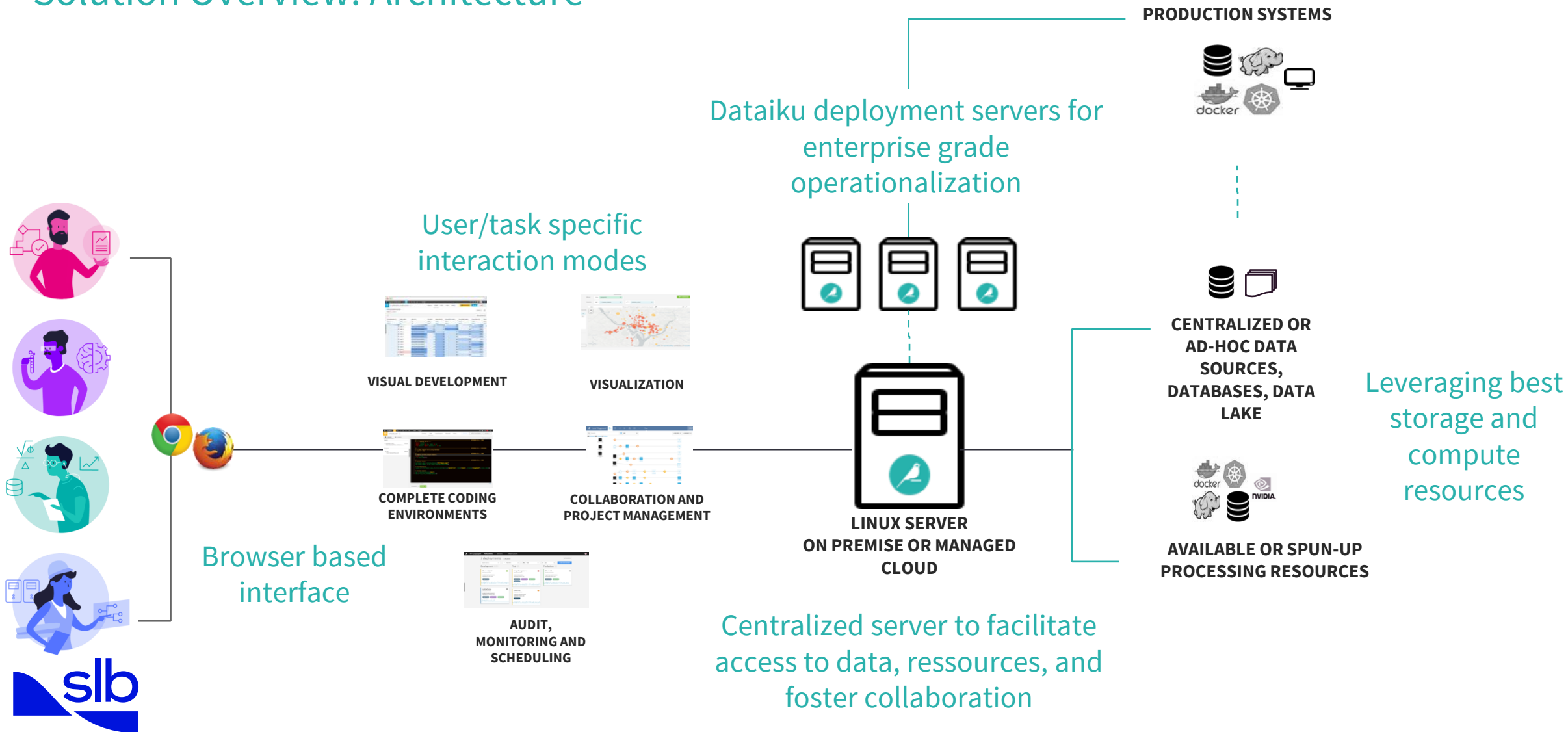
Dataiku DSS

Solution Overview: Architecture



Dataiku DSS

Solution Overview: Architecture



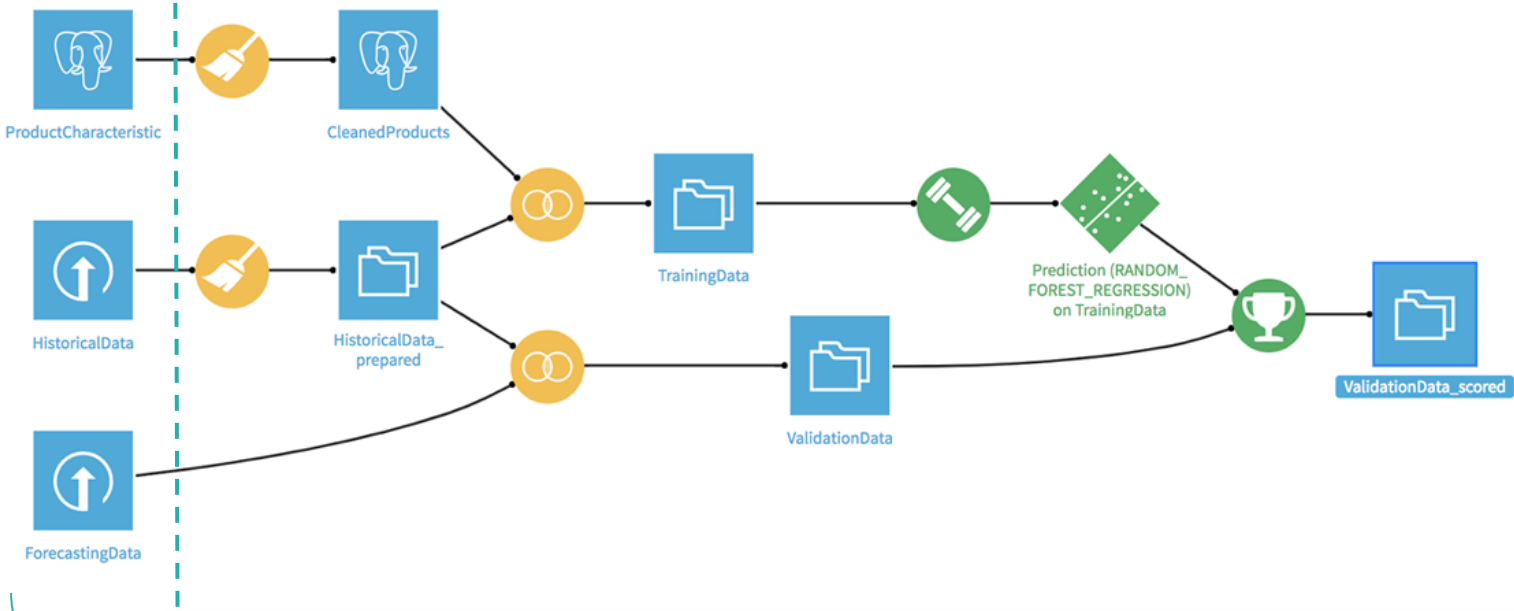
Inclusive end to end platform

Connecting to
raw data

Automation of the flow

Data preparation

Machine learning



LLM Recipes

- Prompt
- Classify text
- Summarize text
- Embed
- Fine tune

Visual recipes

- Sync
- Prepare
- Sample/Filter
- Group
- Distinct
- Window
- Join with...
- Split
- Top N
- Sort
- Pivot
- Stack

Code recipes

- Python
- R
- SQL
- Shell
- Hive
- Impala
- Pig
- Spark SQL
- Spark Scala
- PySpark
- Spark R

VISUAL

Collaboration
between
different
skill sets

CODE



GOVERNANCE, VERSIONING, AUDIT AND REUSABILITY

Collaborative platform for data science



Business Analyst

Visual Data
Preparation

Auto ML



Data Scientist

R, Python, Notebooks, and
many data Science Libraries

Model
Monitoring



Analytics Leader

Visualizati
on and
validation



Data Engineer

SLB-Private

Production
deployment

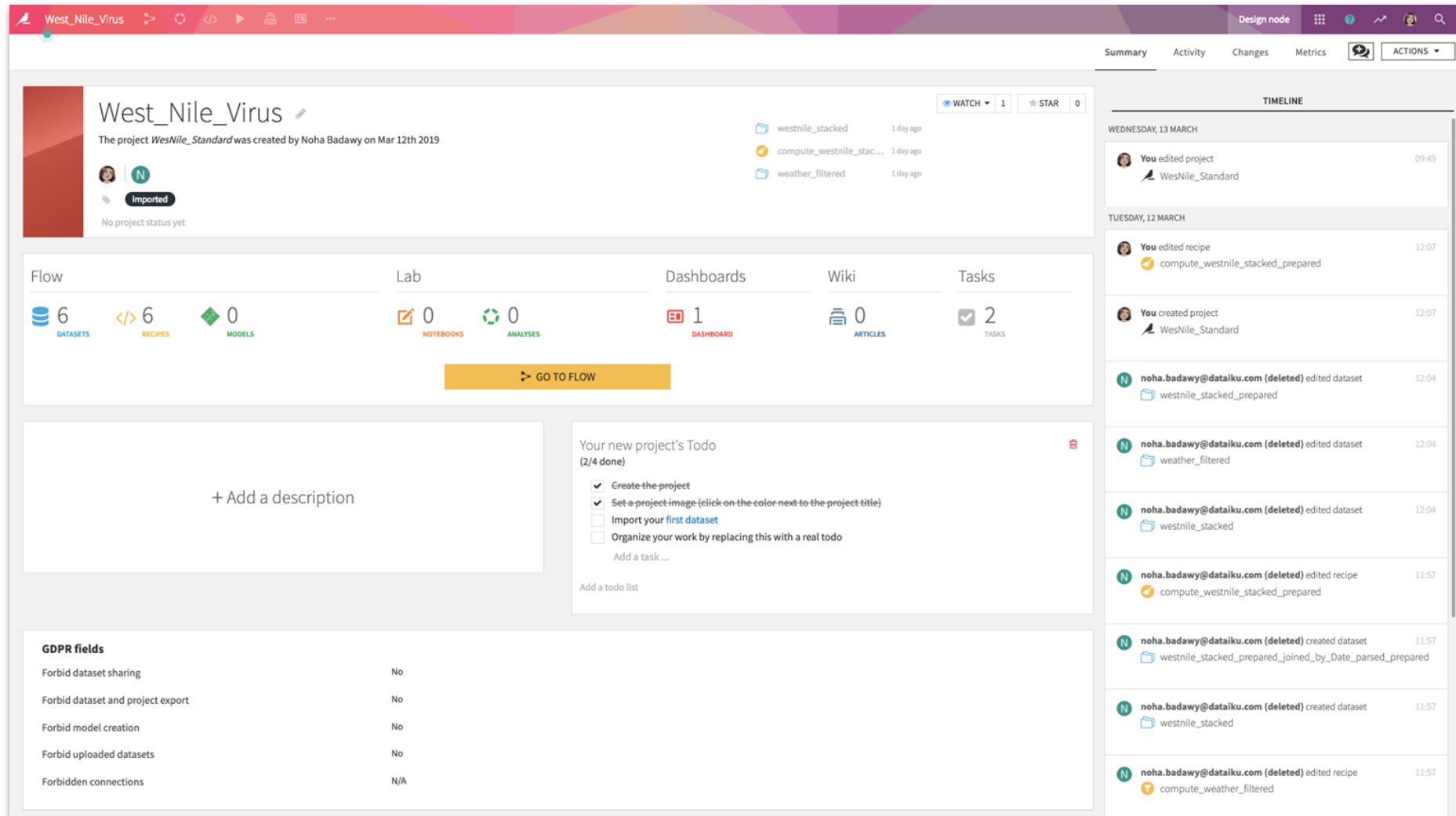
IT
Monitoring

DSS concepts



DSS Concepts

Project homepage



The screenshot displays the Dataiku project homepage for a project named "West_Nile_Virus". The interface includes a top navigation bar with tabs for "Summary", "Activity", "Changes", "Metrics", and "ACTIONS". The main content area is divided into several sections:

- Project Header:** Displays the project name "West_Nile_Virus", a description "The project *WesNile_Standard* was created by Noha Badawy on Mar 12th 2019", and a status "Imported". It also shows a "WATCH" button with a count of 1 and a "STAR" button with a count of 0.
- Flow Section:** Shows a "Flow" card with 6 DATASETS, 6 RECIPES, and 0 MODELS. Below this is a "GO TO FLOW" button.
- Lab Section:** Shows a "Lab" card with 0 NOTEBOOKS and 0 ANALYSES.
- Dashboards Section:** Shows a "Dashboards" card with 1 DASHBOARD.
- Wiki Section:** Shows a "Wiki" card with 0 ARTICLES.
- Tasks Section:** Shows a "Tasks" card with 2 TASKS.
- Timeline:** A vertical list of recent activities on the right side, including "You edited project", "You edited recipe", "You created project", and various dataset and recipe updates by "noha.badawy@dataiku.com (deleted)".
- Todo List:** A section titled "Your new project's Todo (2/4 done)" with a checklist:
 - ☒ Create the project
 - ☒ Set a project image (click on the color next to the project title)
 - ☐ Import your [first dataset](#)
 - ☐ Organize your work by replacing this with a real todoBelow the checklist is an "Add a task ..." button and an "Add a todo list" button.
- GDPR fields:** A table at the bottom left with the following data:

GDPR fields	
Forbid dataset sharing	No
Forbid dataset and project export	No
Forbid model creation	No
Forbid uploaded datasets	No
Forbidden connections	N/A

DSS interface

Menu

History of activities

Flow items

Exploration items

Documentation items

Training results

To do list

Training Facies Classification

Create a short description (appears on homepage)

master

training tutorial + Add tags

In Progress

WATCH 1 STAR 0

Simple Model with new ... 5 hours ago

Simple model 11 hours ago

Dataset 11 hours ago

Dataset_With_Labels 11 hours ago

Monitoring dashboard 11 hours ago

Automation

No active scenarios yet. Activate one [from the list](#)

22 RUNS

5/9 12/9 19/9 26/9 3/10

Flow

13 DATASETS 10 RECIPES 2 MODELS

Lab

0 NOTEBOOKS 3 ANALYSES

Dashboards

1 DASHBOARD

Wiki

0 ARTICLES

Tasks

1 TASK

GO TO FLOW

Training

This project is demo for the training

Todo list (1/2 done)

☒ Import the data

☐ Build a first model

Add a task

Add a todo list

TIMELINE

TODAY

You edited metadata on project 20:50

Training Facies Classification

You tagged project 20:49

Training Facies Classification

You tagged project 20:49

Training Facies Classification

You edited tasks on project 20:49

Training Facies Classification

Import the data

Currently completed tasks: 1 / 2

You edited tasks on project 20:49

Training Facies Classification

Build a first model

Currently completed tasks: 0 / 2

You edited metadata on project 20:49

Training Facies Classification

You edited metadata on project 20:48

Training Facies Classification

Todo list: Todo list

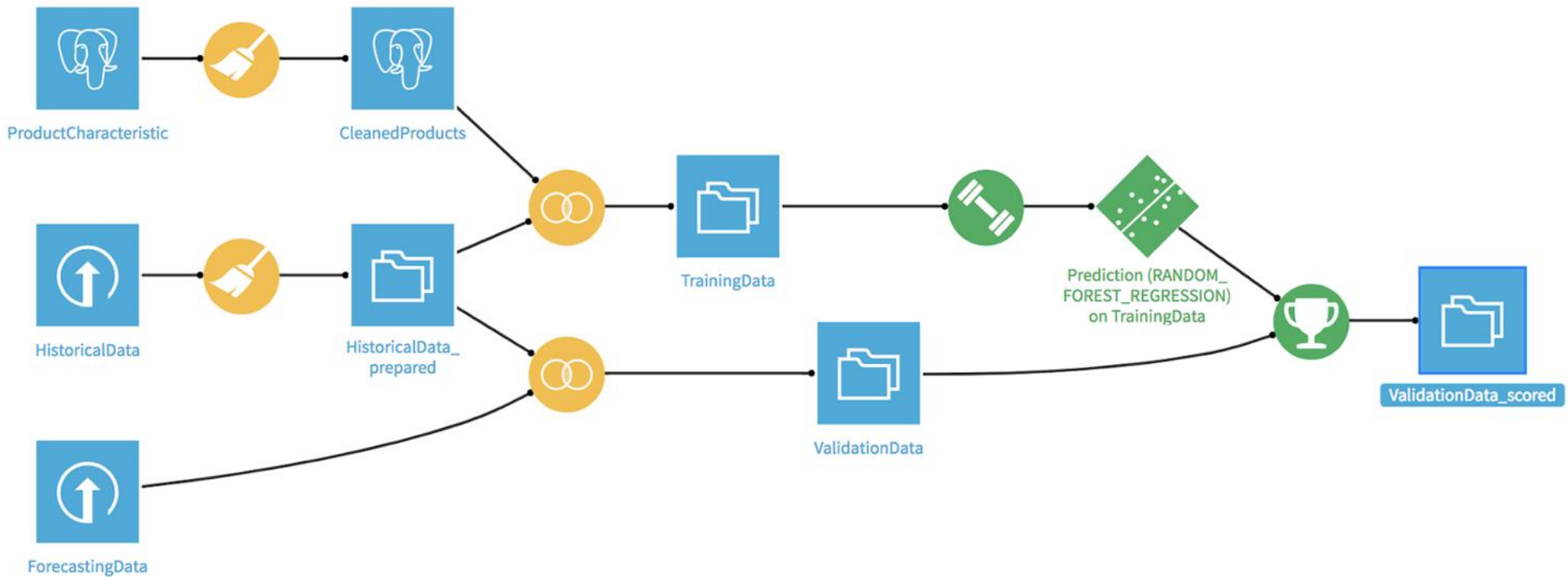
yshi11@slb.com edited model 09:28

Simple model

yshi11@slb.com changed a model version 09:28

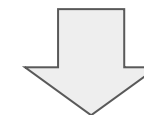
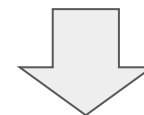
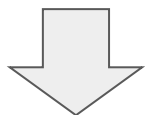
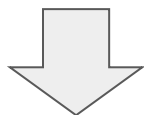
DSS Concepts

Flow, Datasets, Recipes



DSS Concepts

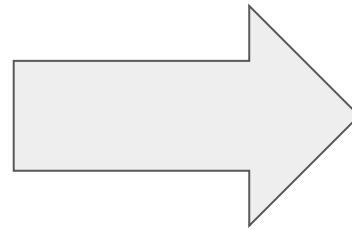
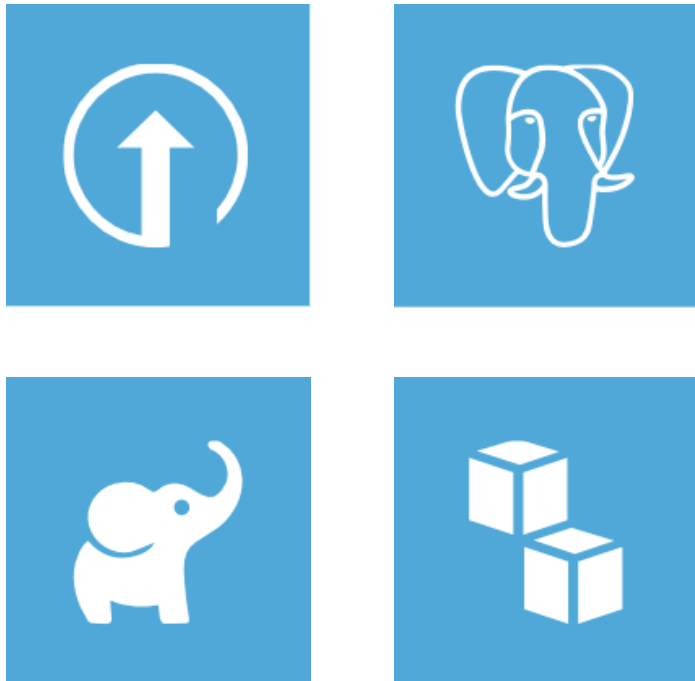
Datasets



DSS Concepts

Datasets

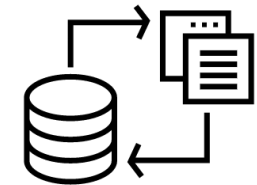
You can run all the usual data manipulation operations on your datasets, **independently** of their connection type.



READ



WRITE



VISUALIZE

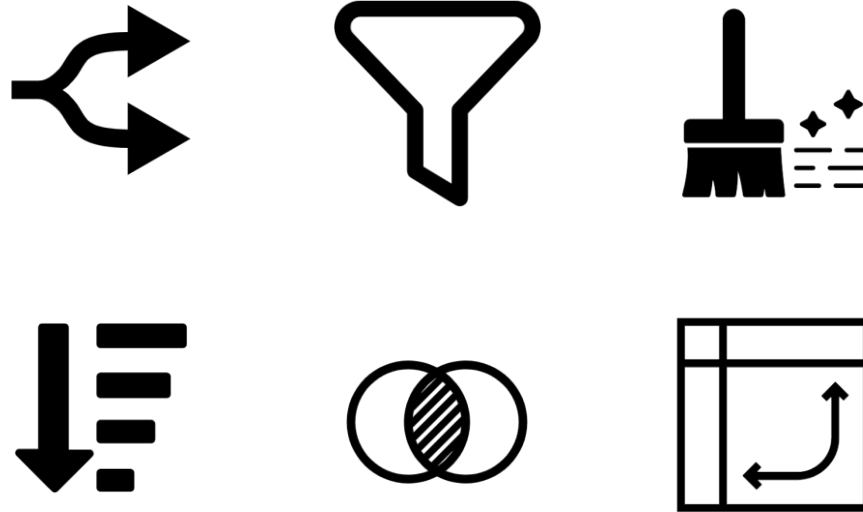
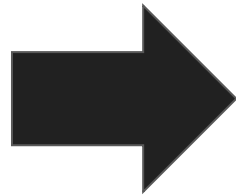
TRANSFORM

DSS Concepts

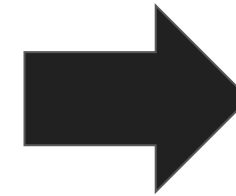
Recipes



INPUT



TRANSFORMATIONS



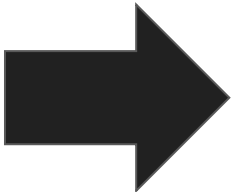
OUTPUT

DSS Concepts

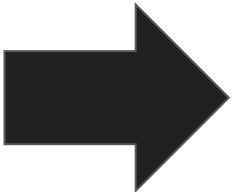
Recipes



INPUT



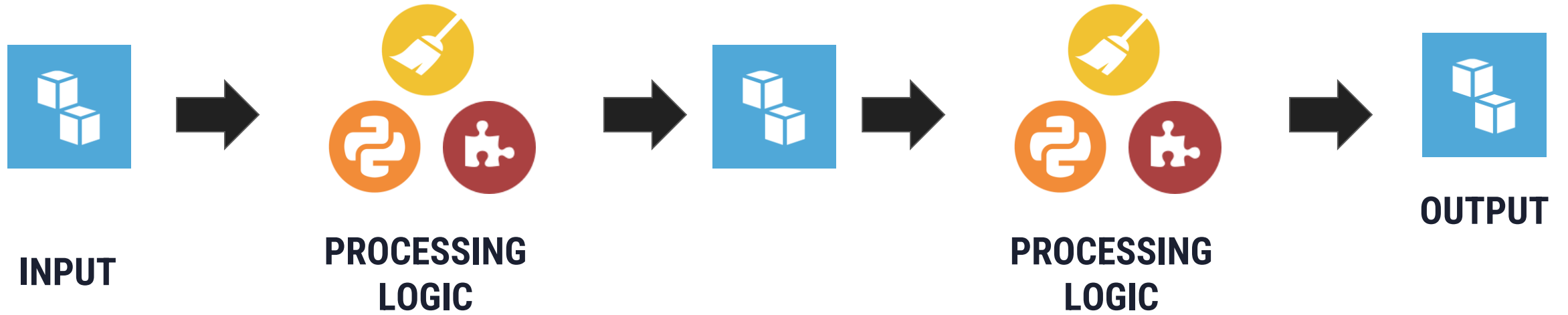
RECIPES



OUTPUT

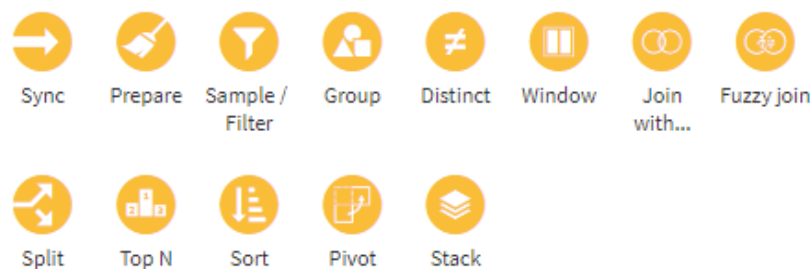
DSS Concepts

The Flow

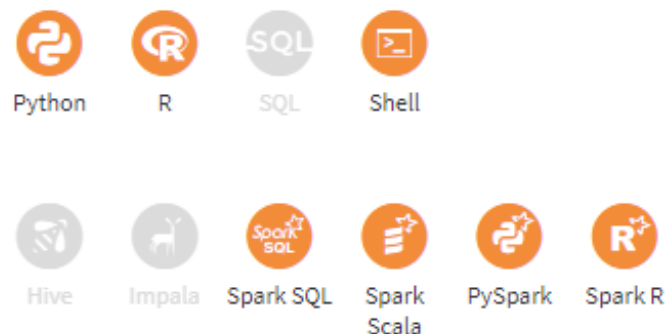


The recipes

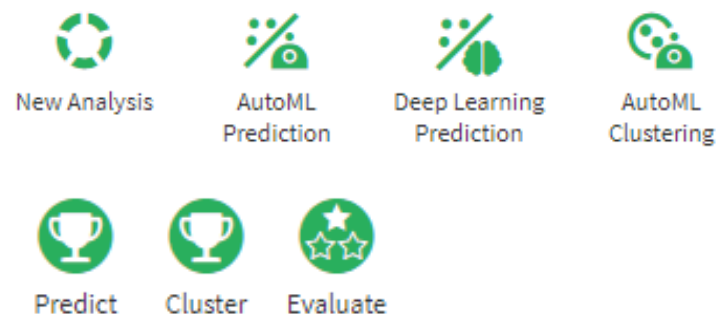
Visual recipes



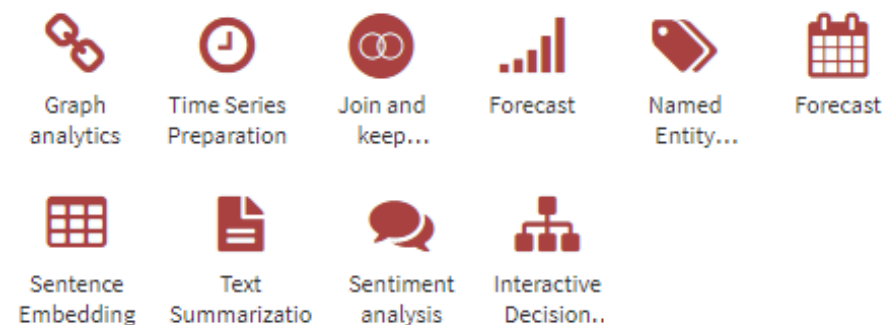
Code recipes



Visual analysis



Plugin recipes

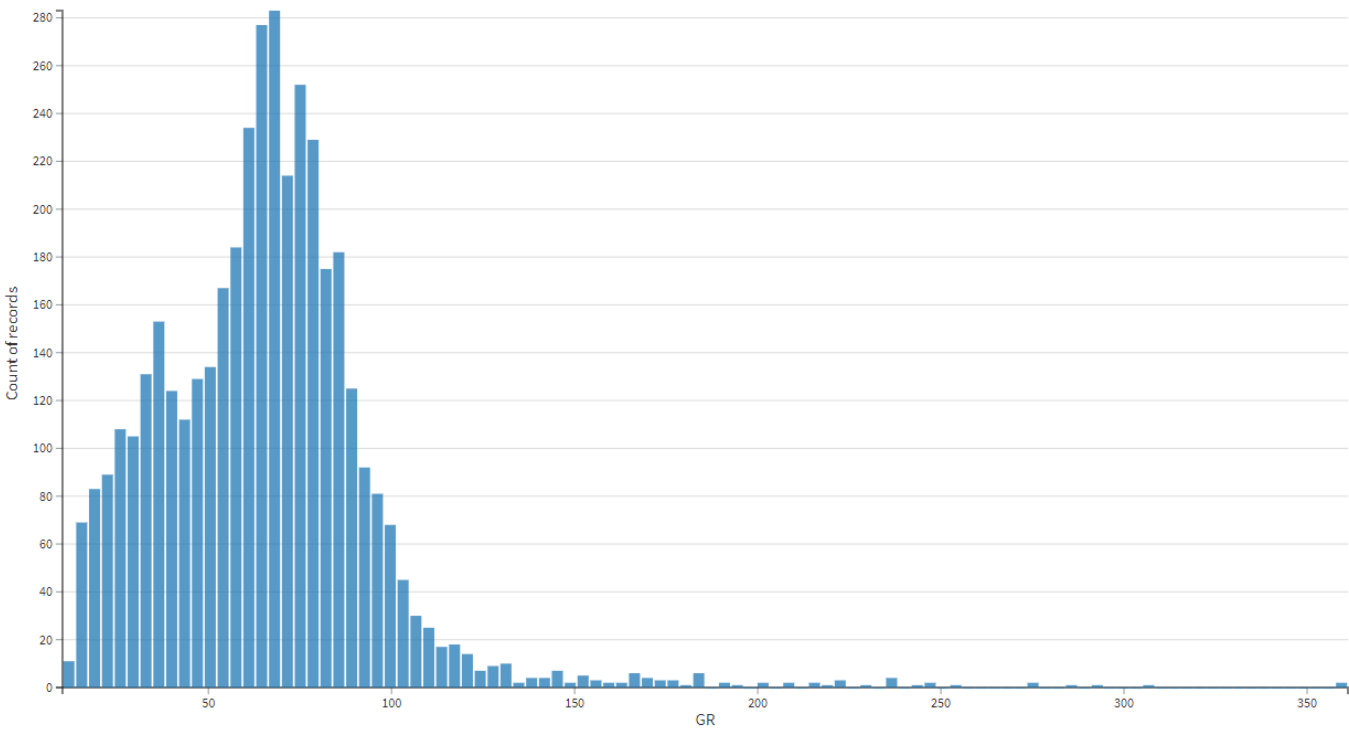
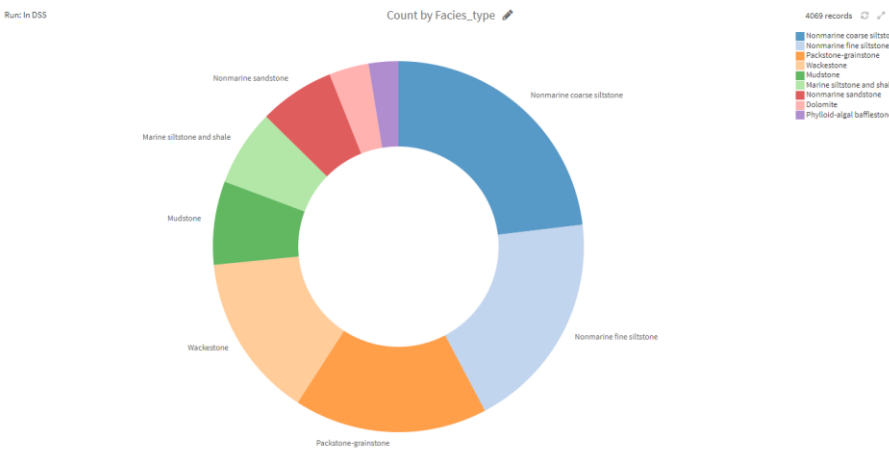
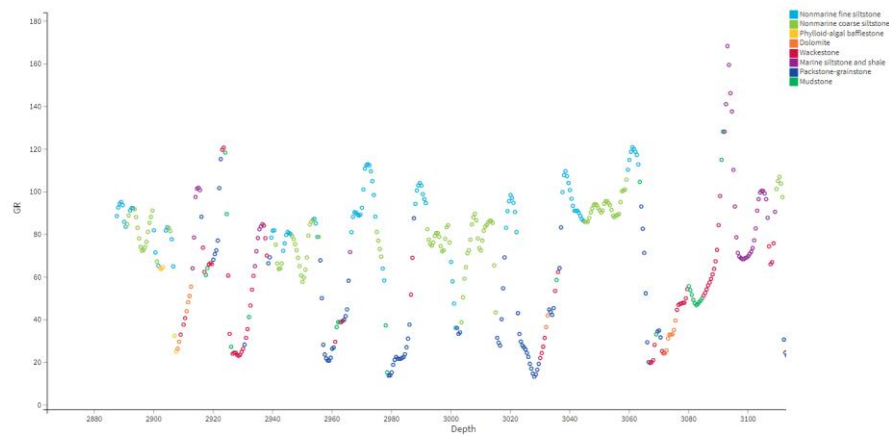


Data preparation

- Dataiku offers many visual tools to prepare, clean, format your data
- Most of data cleaning steps can be done by using the "Prepare recipe"
- Every time you add a recipe, Dataiku will create a new dataset



Data exploration



Statistical analysis

Correlation matrix on 8 variables (Spearman) No split

	Depth	GR	ILD_log10	DeltaPHI	PHIND	PE	NM_M	RELPOS
Depth	1.000	-0.143	0.209	0.050	-0.106	0.254	0.274	0.010
GR	-0.143	1.000	-0.322	0.306	0.445	-0.387	-0.471	-0.176
ILD_log10	0.209	-0.322	1.000	-0.172	-0.737	0.524	0.542	0.097
DeltaPHI	0.050	0.306	-0.172	1.000	0.147	-0.079	-0.251	0.018
PHIND	-0.106	0.445	-0.737	0.147	1.000	-0.660	-0.572	-0.056
PE	0.254	-0.387	0.524	-0.079	-0.660	1.000	0.689	0.048
NM_M	0.274	-0.471	0.542	-0.251	-0.572	0.689	1.000	0.038
RELPOS	0.010	-0.176	0.097	0.018	-0.056	0.048	0.038	1.000

Depth

GR

ILD_log10

DeltaPHI

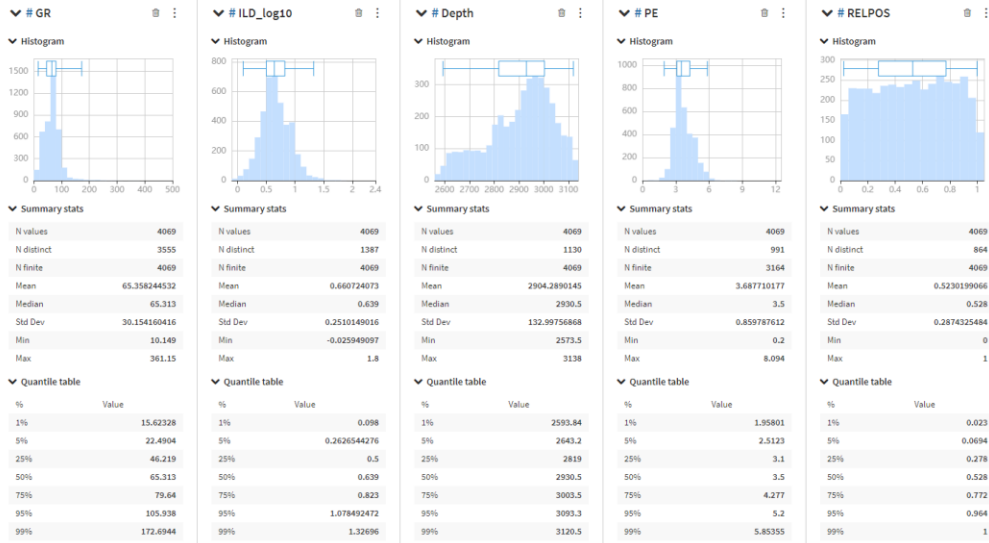
PHIND

PE

NM_M

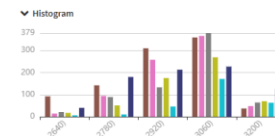
RELPOS

Univariate analysis on 5 variables No split

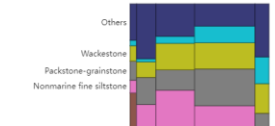


Bivariate analysis on 4 variables No split

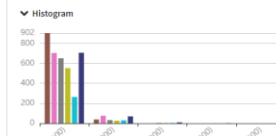
Facies_type by # Depth



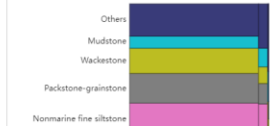
Mosaic plot



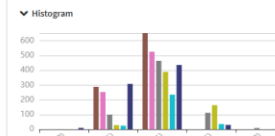
Facies_type by # GR



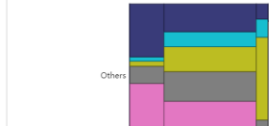
Mosaic plot



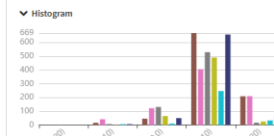
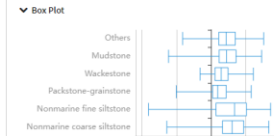
Facies_type by # ILD_log10



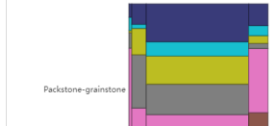
Mosaic plot



Facies_type by # DeltaPHI



Mosaic plot



Demo Session



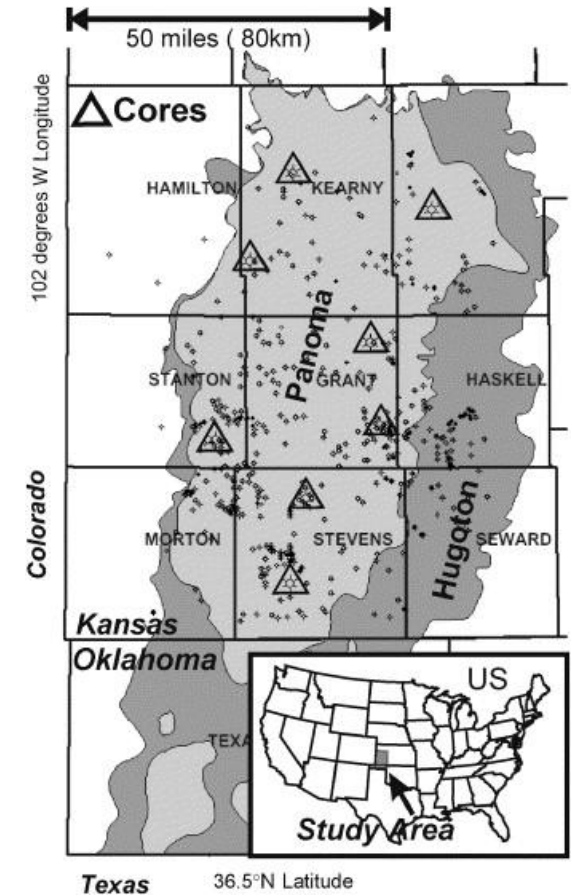
Dataset: Facies classification

Objective: Classify facies

Why?

Increase knowledge of the subsurface

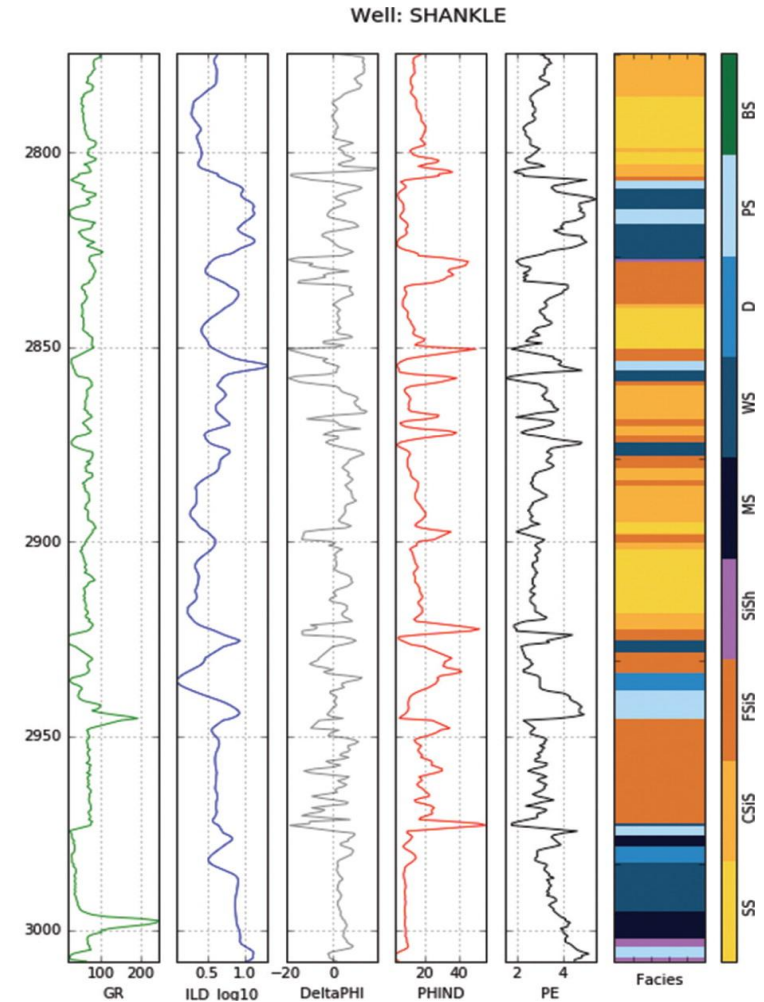
Estimate reservoir capacity



Panoma gas field, in Southwest Kansas, [Dubois et al. \(2007\)](#)

Dataset: Features

- Gamma ray (GR)
- Resistivity (ILD_log10)
- Photoelectric effect (PE)
- Neutron-density porosity difference (DeltaPHI)
- Average neutron-density porosity (PHIND)
- Nonmarine/marine indicator (NM_M)
- Relative position (RELPOS)



Dataset: Labels

Facies	Description	Label
1	Nonmarine sandstone	SS
2	Nonmarine coarse siltstone	CSiS
3	Nonmarine fine siltstone	FSiS
4	Marine siltstone and shale	SiSh
5	Mudstone	MS
6	Wackestone	WS
7	Dolomite	D
8	Packstone-grainstone	PS
9	Phylloid-algal bafflestone	BS



Large language Models (LLMs)

Abhineet Sinha

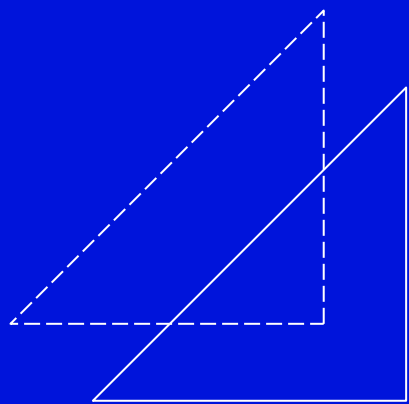
Domain Data Scientist-Production

Innovation Factori, Delhi
Gurgaon, India

AGENDA

1. What is Large Language Modeling (LLM)?
2. LLM Overview
3. Hands-On
 - a. Summarization and Classification
 - b. Retrieval-based question answering
4. What's next?





WHAT IS

Large Language Modelling



Language model introduction

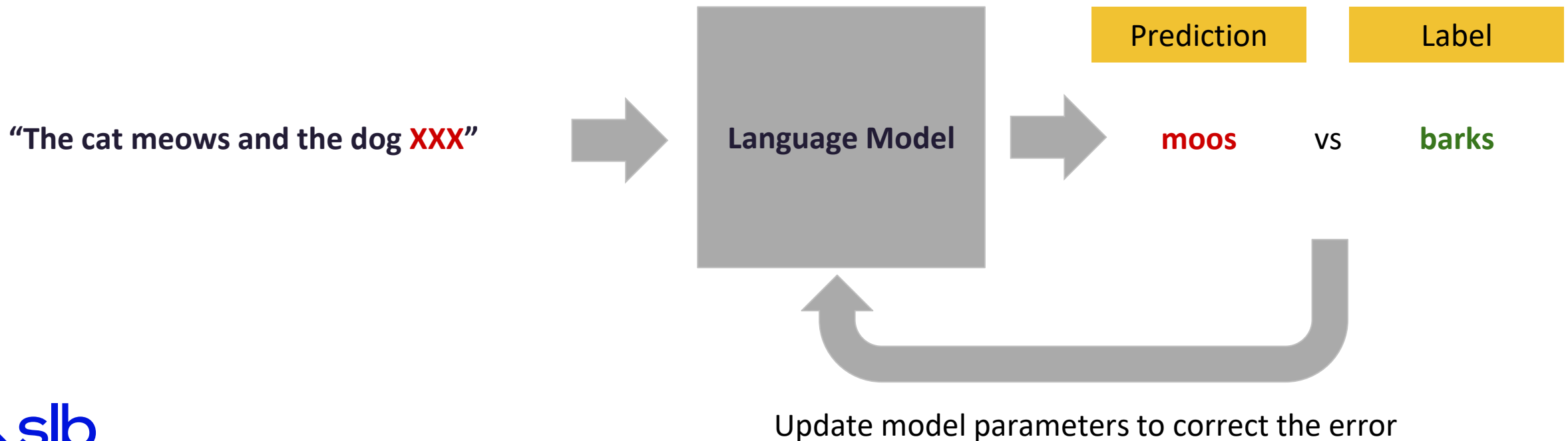
What is a Language model ?

A Language model is a model that takes text as an input to perform various tasks :

✓ Classification

✓ Translation

✓ Text generation / conversation



Language model introduction

Model training

“The greatest lesson in life is to ?”

“The greatest lesson in life is to know ?”

“The greatest lesson in life is to know that ?”

“The greatest lesson in life is to know that even ?”

“The greatest lesson in life is to know that even fools ?”

“The greatest lesson in life is to know that even fools are ?”

“The greatest lesson in life is to know that even fools are right ?”

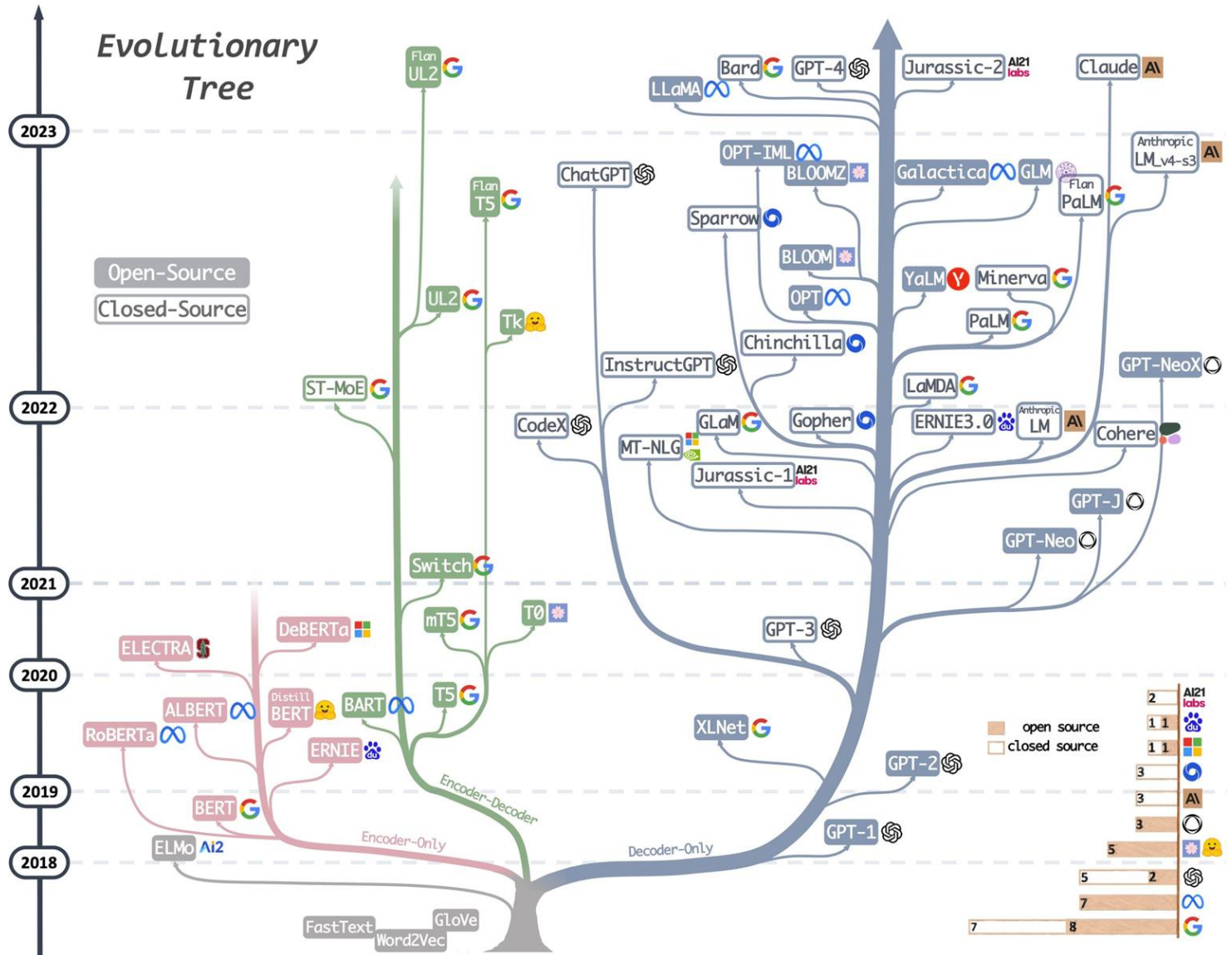
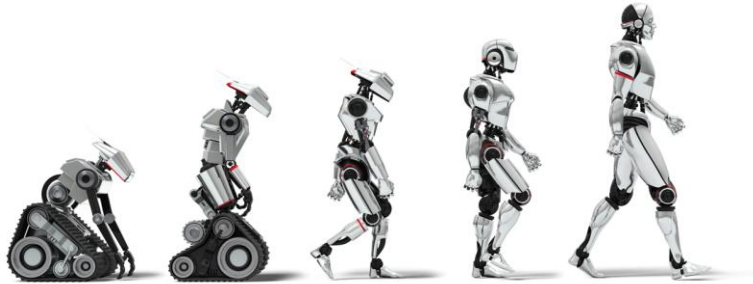
“The greatest lesson in life is to know that even fools are right sometimes ?”

- LM are trained on each word
- Computational heavy
- Very bad at the beginning and needs a lot of data to get better



Language model introduction

Evolution of LLMs



Language model - quickly explained

Generating Text

- LLMs excel at a specific task: **generating coherent and credible continuations of input texts**, also called prompts.
- LLMs were **trained with colossal amounts of textual data** encompassing diverse sources like books, articles, and websites. Through this process, they acquired remarkable reasoning capabilities which make them much more than writing assistants. In particular, they are especially **effective for a wide range of natural language tasks**.

If you want to build a ship, don't drum up the men to gather wood, divide the work and give orders.
Instead, teach them to yearn for the vast and endless sea.

Example of a prompt and a potential LLM response



Language model - quickly explained

Generating Text

- The key to accomplishing natural language tasks with LLMs is to **craft a prompt** in a way that the continuation of this prompt **yields the desired answer**.
- Suppose that we want to classify a product review as positive, negative, or neutral. To do this, we can specify this task in a prompt , insert the text of the product review , and suggest, through the **structure of the prompt, that the immediate next word should be the answer**. When provided with this prompt, an LLM will hopefully provide the correct result

Decide whether the sentiment of the product review is positive, neutral, or negative.

Product review: I never received my package. The customer service was awful

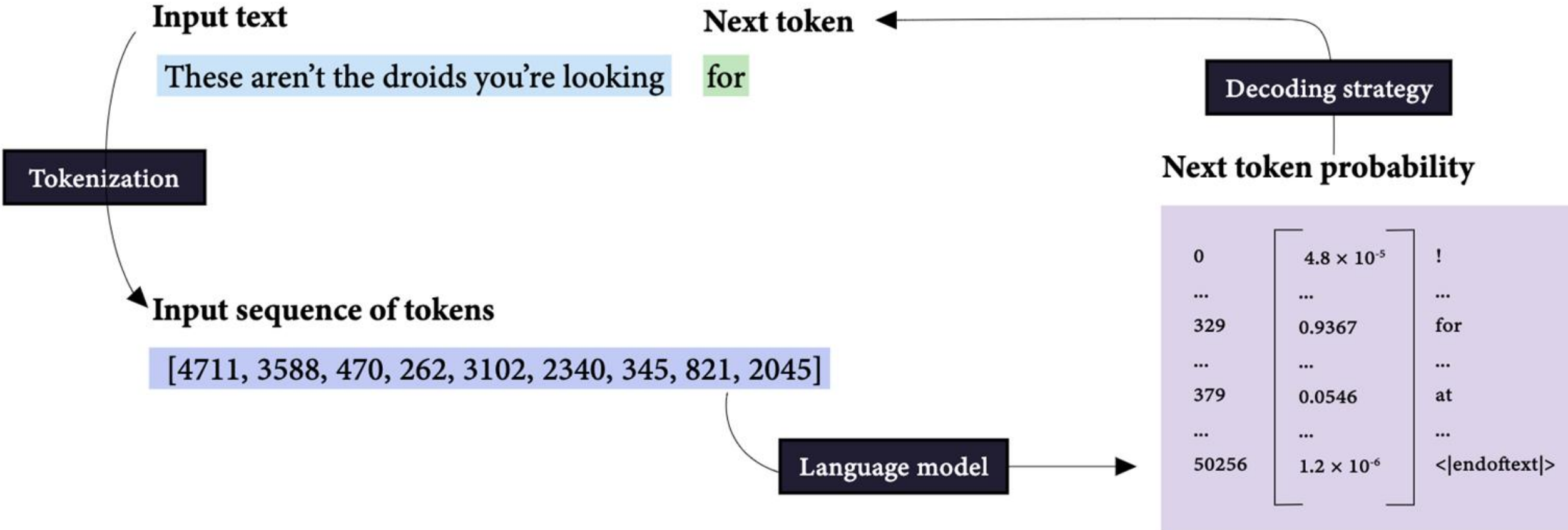
Sentiment: negative



In-context zero-shot learning for a classification task

Language model - quickly explained

Under the hood



Language model - quickly explained

Tokenizing Input Texts and Estimating the Next Token Probability

Tokenizing Input Texts

- Before being processed by a language model, a piece of text is converted into a sequence of tokens which are words or subwords.
- These tokens are selected among a predetermined set of a few tens of thousands of potential tokens and they can simply be represented by integers between 0 and the number of potential tokens minus 1.
- You can test an interactive tool on OpenAI's website if you want to understand more intuitively how tokenization works.

Estimating the Next Token Probability

- A language model takes as an input a sequence of tokens and returns the probability distribution over the next token.
- The length of the input sequence of tokens is limited by the context window size, which is specific to the language model (for example, 4,096 tokens for ChatGPT).



Language model - quickly explained

Implementing a Decoding Strategy

- The probability distribution over the next token can be used to continue the input sequence, one token at a time.
- Several decoding strategies can be used for this. The two simplest approaches are greedy decoding and multinomial sampling (sampling for short).
 - With **greedy decoding**, the next token selected is the one with the highest estimated likelihood.
 - With **sampling**, the next token is randomly chosen based on the estimated probability distribution. This estimated probability distribution is based on raw scores computed by the language model and a temperature parameter. By adjusting this parameter, we can make the text generation process more or less deterministic. Temperature 0 corresponds to greedy decoding while high temperatures lead to more original outputs.

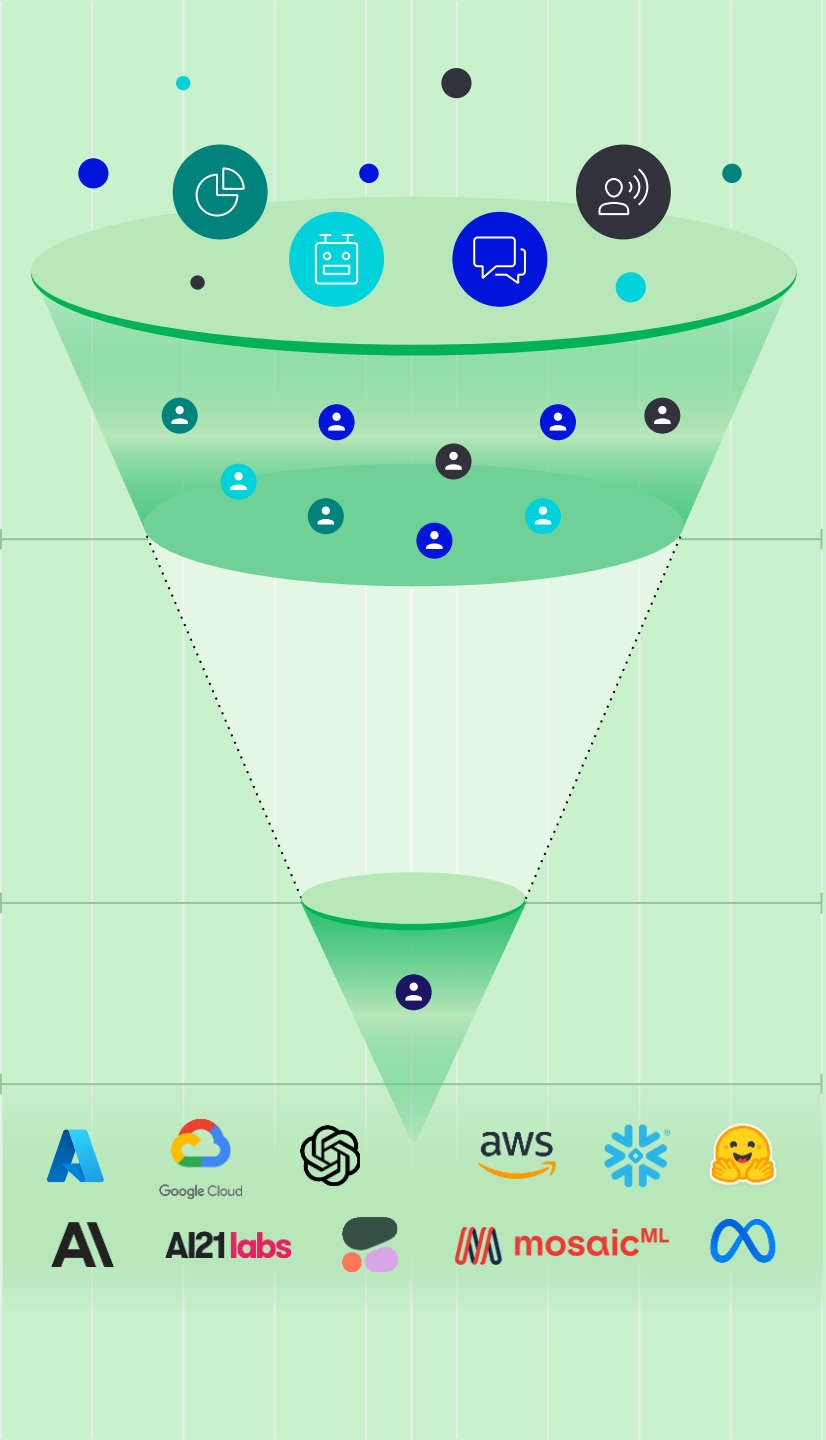
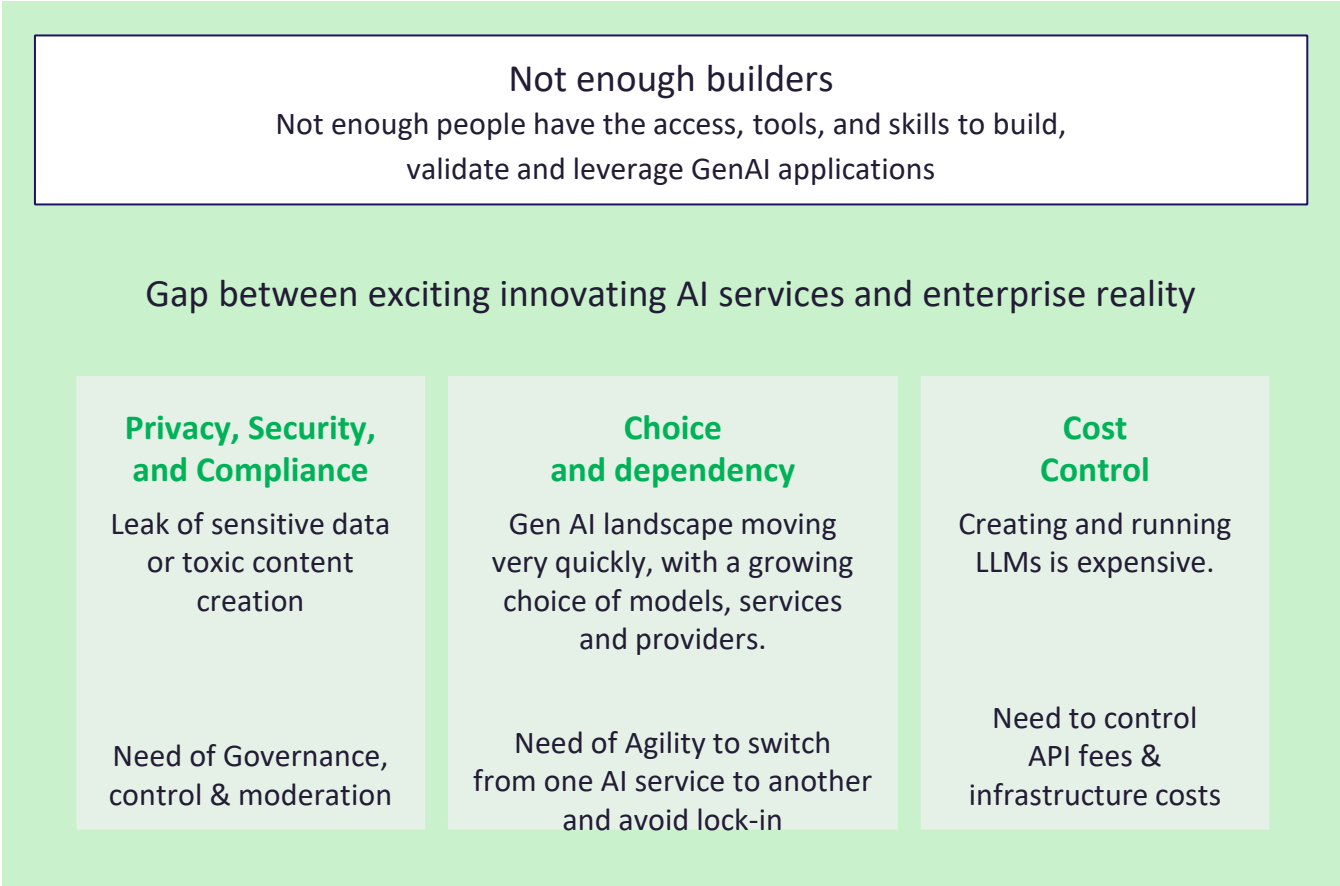


How Dataiku can Accelerate the use of LLMs



The problem: barriers impede the full potential of Generative AI

Beyond chatbots, the human-like intelligence of Generative AI could augment and automate thousands of decisions and processes. What's standing in the way?



Dataiku's key differentiators on Gen AI

#1 Ease of use

Simple to advanced techniques accessible through no/low code components

Acceleration for coders & experts

BENEFITS:

Makes Gen AI accessible beyond a narrow pool of LLM experts

Significant speed to value gains for capitalizing on this new tech

#2 Security & Control

A governed, auditable, and cost aware framework with a secure gateway for LLM providers and models to support enterprise LLM applications

BENEFITS:

Reduced risks (security leaks, shared keys, shadow AI)

Cost control & tracking (for chargeback or cost optimization purposes)

#3 Choice & Agility

A level of abstraction that reduces tech dependencies & helps switch between LLM providers without having to rebuild/code projects

BENEFITS:

Flexibility, no lock-in

Future-proof applications can seamlessly take advantage of new tech as it emerges

#4 Holistic Platform

Extend and amplify existing analytic and predictive workflows with Gen AI. Dataiku's platform can meet all your analytics & AI needs, from end-to-end.

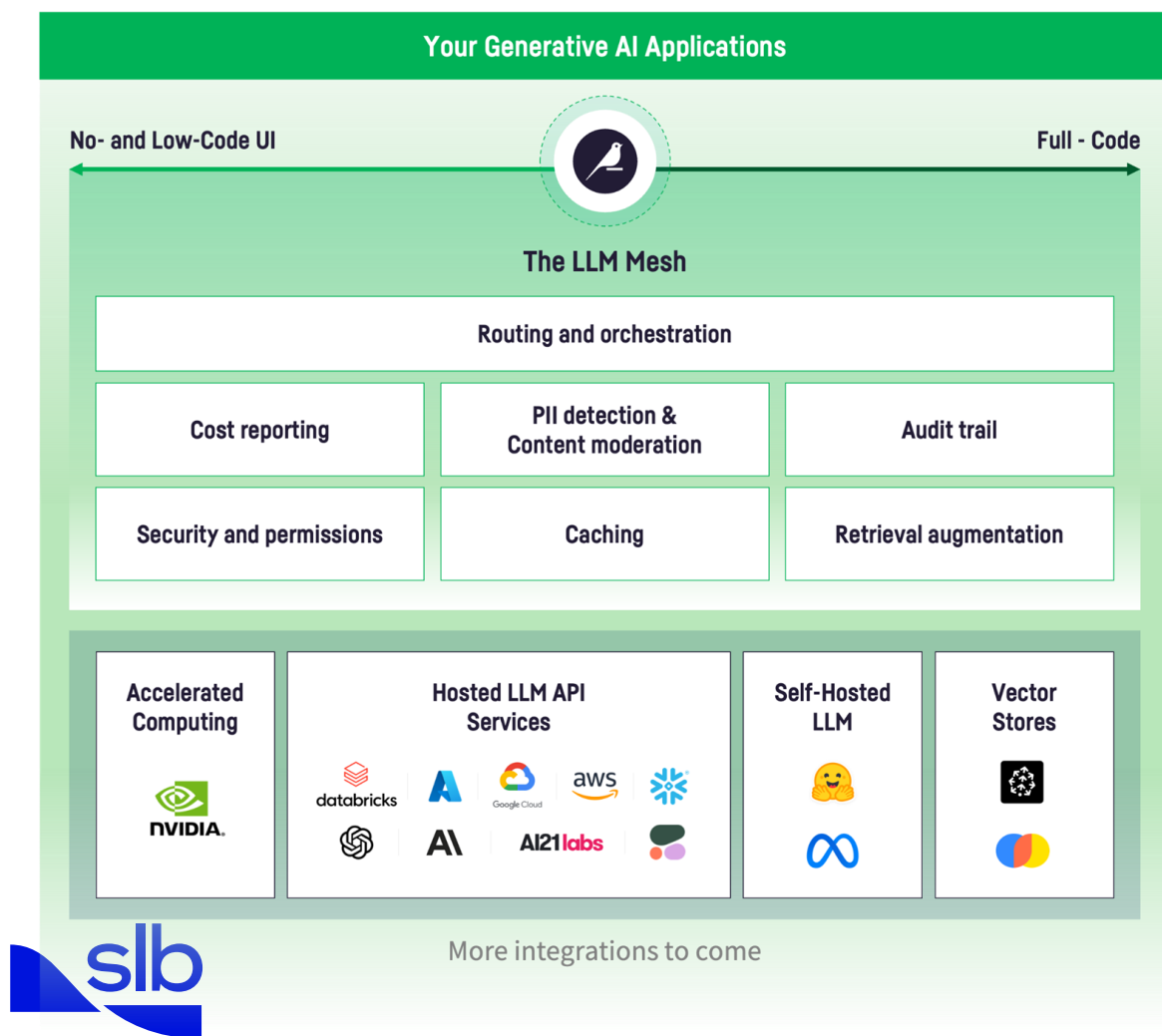
BENEFITS:

Amplify value of Gen AI by integrating it in core analytics and ML workflows

No need to buy siloed point solutions for Gen AI



The LLM Mesh: The Common Backbone for All GenAI Applications



Decouple the application and AI service layers

Enforce a secure gateway to approved LLM services and maintain an audit trail

Ensure safe use by defining filters for queries and responses such as PII detection, content moderation...

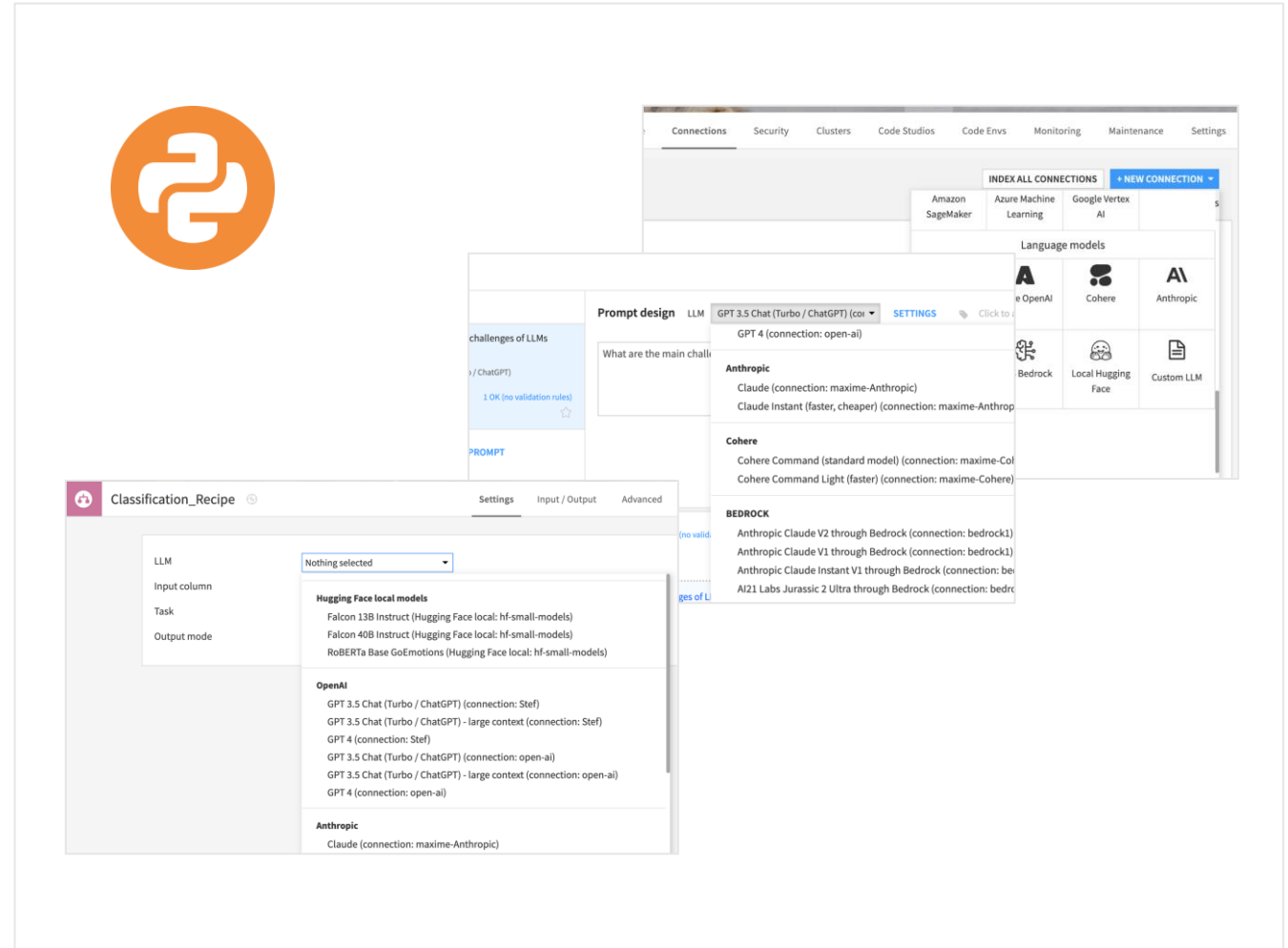
Control costs and avoid unnecessary re-generation with caching

Enrich queries and responses with built-in retrieval augmentation

LLM Mesh: Decoupling the application and AI service layers

Maintain choice and agility as new technologies emerge

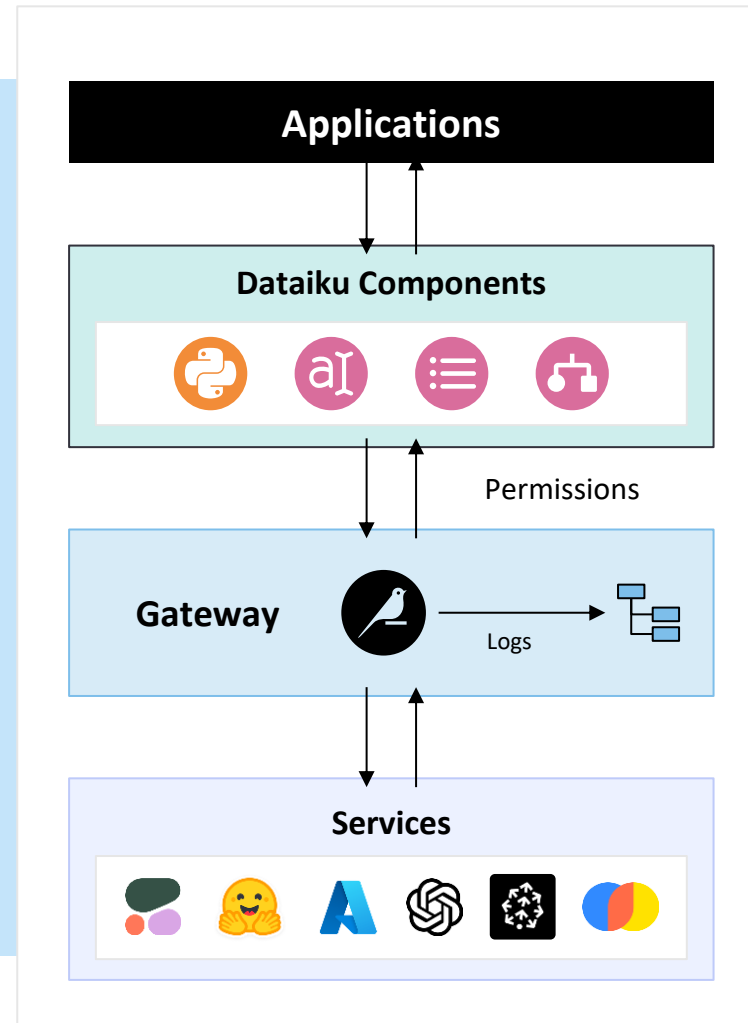
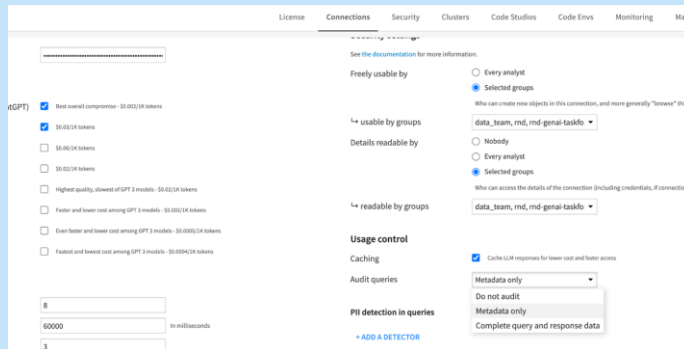
- **Connect to the AI services of your choice including LLMs and Vector DB**
 - ➔ LLM & Vector DB connections
- AI services and their various models are available throughout the Dataiku platform (visual NLP recipes, Prompt Studios, code recipes, notebooks, IDE, ...)
- Keep your choices open by making your applications agnostic to underlying models and providers
- Test different services to balance cost and performance



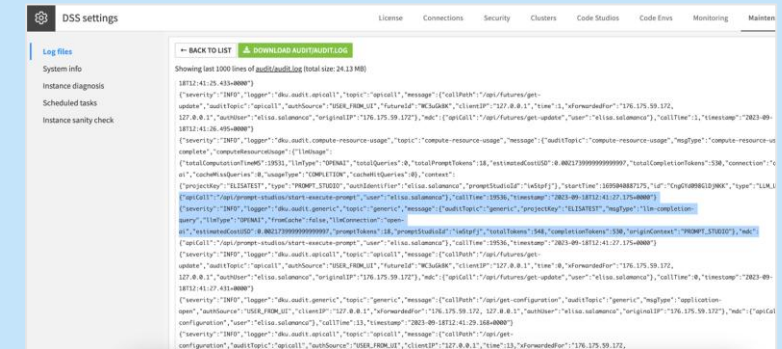
LLM Mesh: Decoupling the application and AI service layers

Maintain choice and agility as new technologies emerge

Centrally manage access keys to different services to ensure that only approved services are used



Log all queries and requests to all services, creating a secure, searchable trail of all LLM use in the enterprise

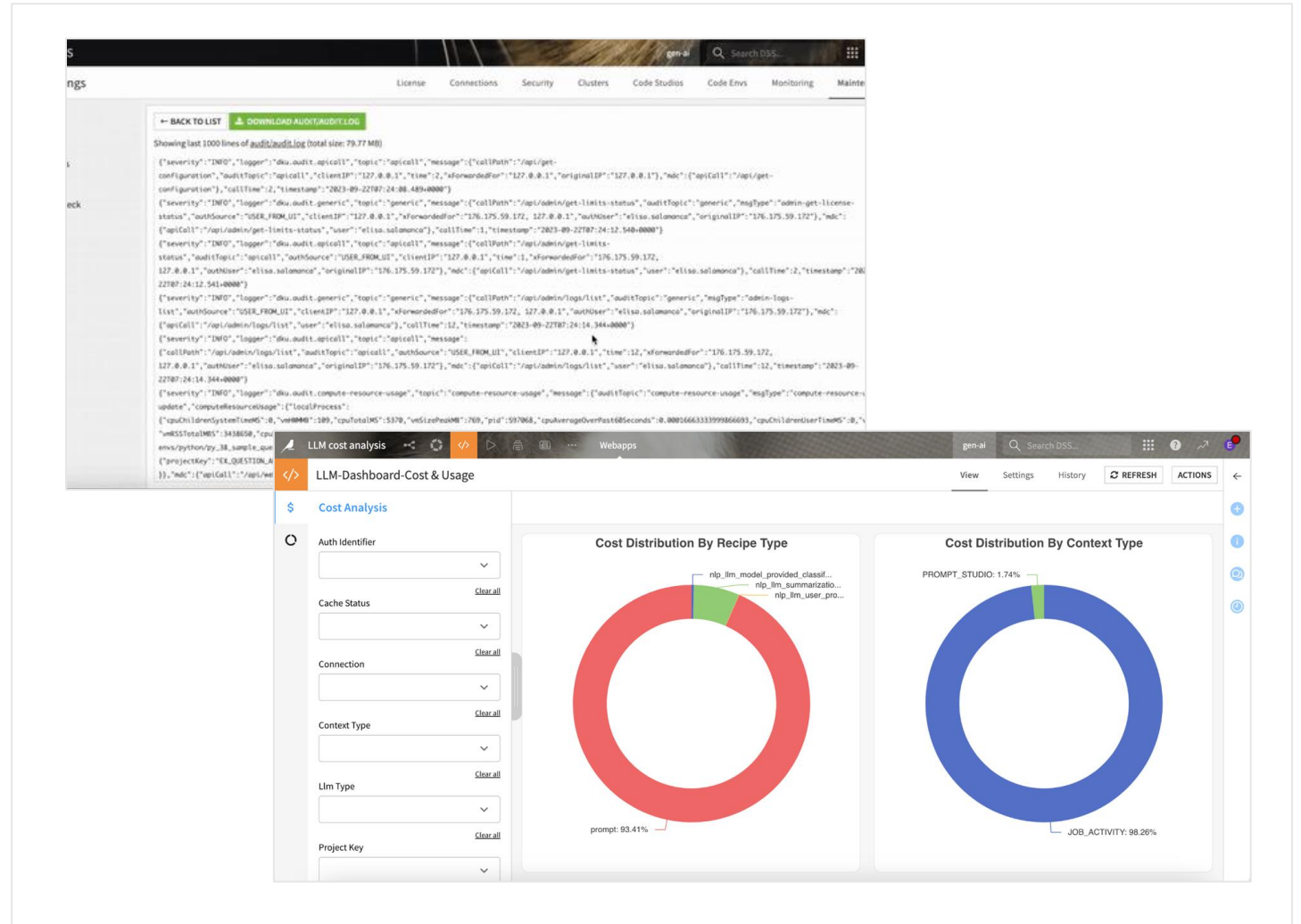


LLM Mesh: Control and optimize costs

Monitor costs and cache common responses to avoid re-generation

➤ Cache responses to common queries to avoid unnecessary re-generation and associated costs

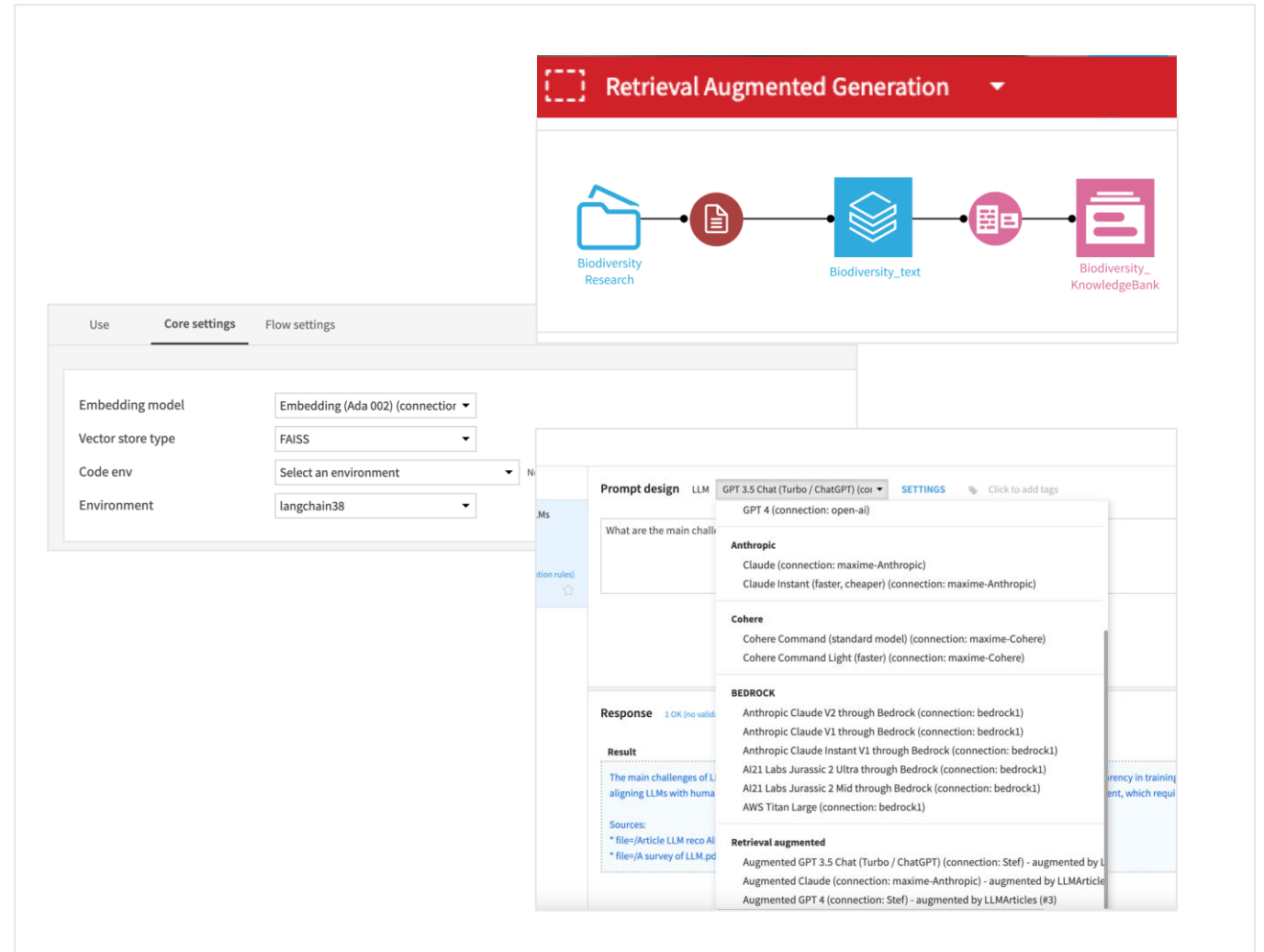
➤ Use Compute Resources Usage in Dataiku to generate fine-grained cost reports



LLM Mesh: Built-in retrieval augmentation

Enrich queries and responses with internal documentation

- Work from documents in any directory, using built-in pre-processing (OCR, translation, etc...) if needed
- Select the model of your choice to create the embeddings
- Easily create knowledge banks using the vector store of your choice
- Use the augmented model elsewhere in Dataiku (for example, Prompt Studios) to build knowledge-enriched apps



Case 1: Use the Capabilities LLMs for Common Language Tasks

Off-the-shelf, visual components bring the capabilities of LLM to the entire enterprise

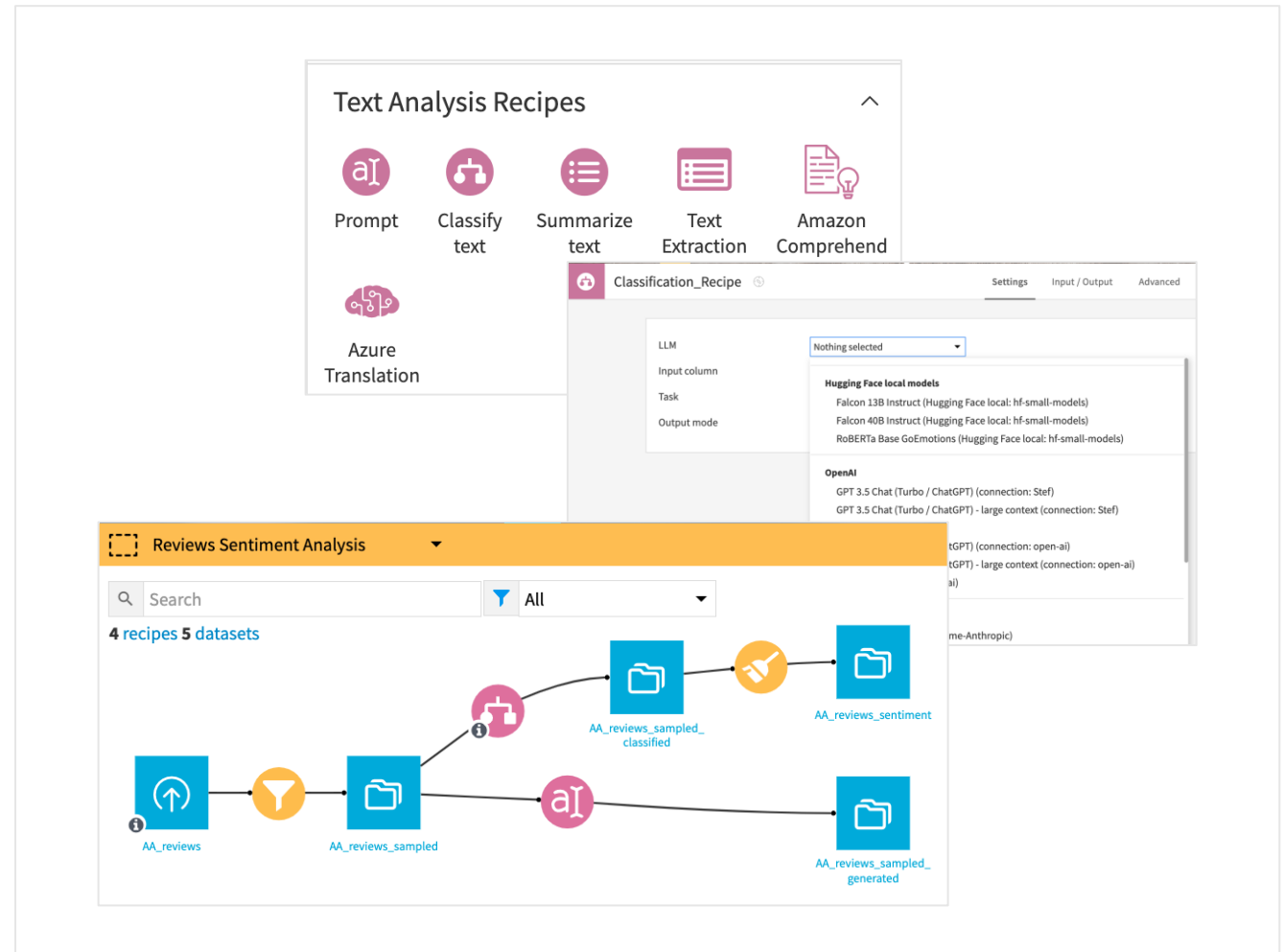
With Dataiku, deliver powerful **no-code** LLM-powered projects and pipelines

Flexible, **provider-independent approach**:

Leverage any provider's services OR maintain total privacy with local models

Leverage Dataiku's growing library of ready-to-use **visual LLM-powered recipes**:

- Entity extraction
- Text classification
- Text summarization
- Translation
- And more



Case 2: Prompt Engineering

Customize LLM outputs by refining prompt context, format, length, and instructions

- Prompt Studios in Dataiku help you design, evaluate, and operationalize optimized LLM prompts in a no-code way
- Model-agnostic so you can compare results across different models & providers to identify the best model to achieve your specific goals
- Built-in cost estimates inform important trade-off decisions between cost & performance during the design phase
- Deploy your prompt in your data pipeline



Test different models, including cloud, local or your fine-tuned models

The screenshot shows the 'Prompt design' tab in the 'Financial News' project. It features a 'Task' section with a 'STRUCTURED PROMPT' and a 'TEXT PROMPT' option. The 'Inputs' section shows a dataset 'reuters_headlines' with columns 'Article's headline' and 'Article's text preview'. The 'Test cases' section displays a table with columns 'Headlines', 'Description', and 'Result'. The 'Result' column shows a list of topics: 'company and product news', 'corporate debt and earnings', 'IPO', 'M&A and investments', 'stock movement', 'markets', 'legal and regulation', 'politics', 'currencies', 'gold and metals', 'energy and oil', 'fed and central banks', 'other', 'human resources', 'workforce management', 'economic changes', 'big data', 'inflation'.

Headlines	Description	Result
Exclusive: HSBC sheds top talent as global equities boss leaves - sources	HSBC's global equities chief Hossein Zaimi is leaving the bank, two sources with direct knowledge of the matter said on Friday, in the latest revamp of the lender's troubled investment banking operations.	["company and product news": 1, "corporate debt and earnings": 0, "IPO": 0, "M&A and investments": 0, "stock movement": 0, "markets": 1, "legal and regulation": 0, "politics": 0, "currencies": 0, "gold and metals": 0, "energy and oil": 0, "fed and central banks": 0, "other": 0, "human resources": 0, "workforce management": 0]
With shift toward merit-based pay, Japan's Hitachi to drop old ways	Hitachi Ltd is moving to scrap a remnant of one of Japan Inc's most famous practices: seniority-based pay.	["company and product news": 1, "corporate debt and earnings": 0, "IPO": 0, "M&A and investments": 0, "stock movement": 0, "markets": 0, "legal and regulation": 0, "politics": 0, "currencies": 0, "gold and metals": 0, "energy and oil": 0, "fed and central banks": 0, "other": 0, "human resources": 0, "workforce management": 1]
BOJ looks to big data for clues on pandemic-driven economic changes	The coronavirus pandemic may have caused structural changes in Japan's economy and the way it affects inflation, requiring the Bank of Japan to tap more deeply into big data in making policy decisions, the central bank's top economist said on Thursday.	["company and product news": 0, "corporate debt and earnings": 0, "IPO": 0, "M&A and investments": 0, "stock movement": 0, "markets": 1, "legal and regulation": 0, "politics": 0, "currencies": 0, "gold and metals": 0, "energy and oil": 0, "fed and central banks": 1, "other": 0, "economic changes": 1, "big data": 1, "inflation": 1]
U.S. manufacturing snaps back; new COVID-	U.S. factory output rose by the most in more	["company and product news": 0, "corporate debt and earnings": 0, "IPO": 0, "M&A and



Save as **Prompt Recipe** in your data pipeline



Estimate and compare running costs



Validate model responses



Leverage a library of templates, promoting reuse and best practices.

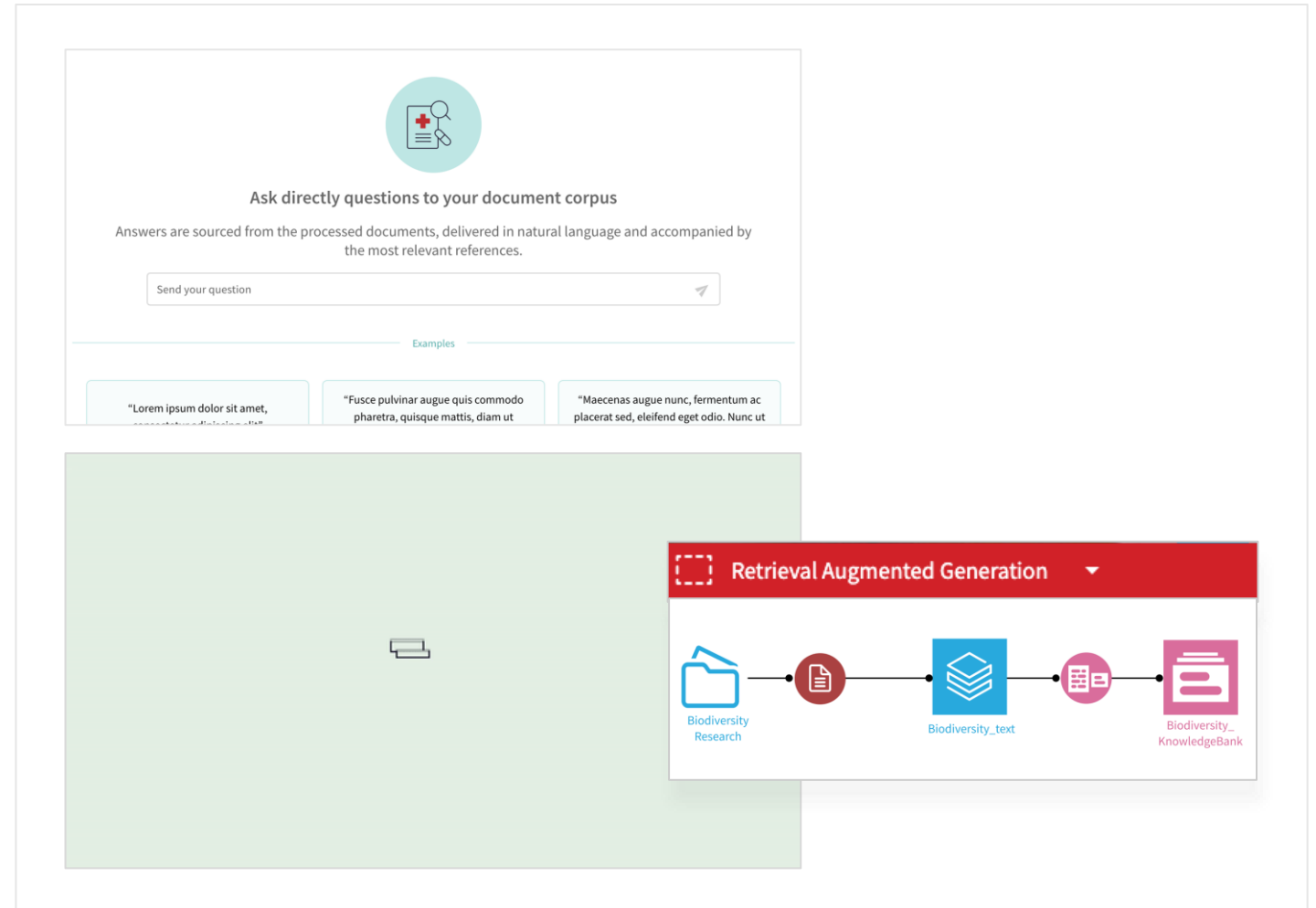
Case 3: Knowledge Bank – Add Your Internal Knowledge

Incorporate information retrieval and LLM augmentation methods to build Q&A apps & chatbots

➤ Develop and deliver Q&A-type LLM applications backed by your own proprietary documents in a no-code way

➤ Leverage built-in components for Retrieval Augmented Generation (RAG) and semantic search :

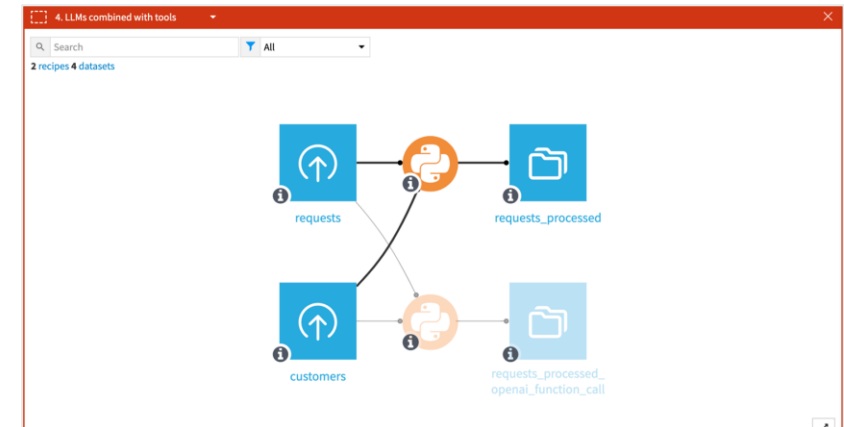
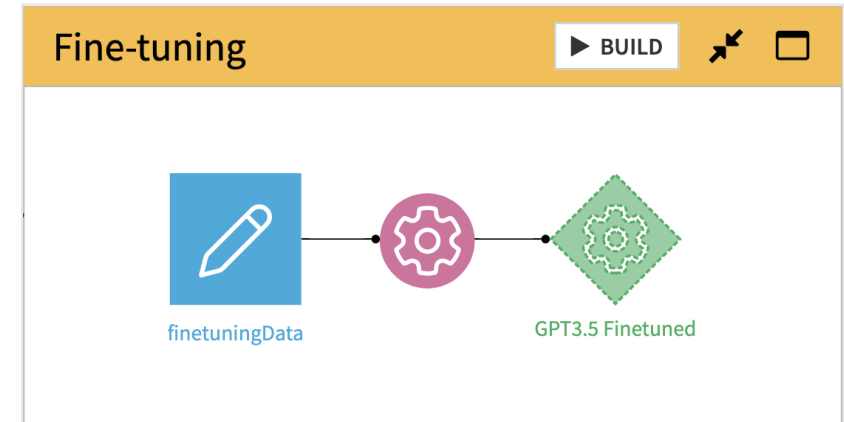
- Text extraction from files and OCR
- Embedding Recipe
- Knowledge banks (using vector stores)



Case 4: Advanced Techniques

Incorporate LLM agents, set up complex reasoning chains, and fine-tune models

- **LLM agents:** Orchestrate complex reasoning tasks and chains of prompts, data sources, and logic with your own code, and leverage our integration with Langchain API
- Leverage LLM providers fine-tuning services in a seamless way with our fine-tuning recipe
- Fine-tune or pre-train your own models through code



Hands-On Session



Thank You

