



Actionable Data Insights for Machine Learning

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Agenda

Motivation

Implementation

Case Studies

Summary

Motivation

Drive model performance with better data understanding



Interview Learnings

Projects included: Image understanding, machine translation, text understanding, speech recognition

Enable rapid data experiments

- High-level declarative interfaces
- User defined functions as first class citizen

Flexibility

- Enable usage across different storage systems
- Enable portability from local machines to cloud to devices

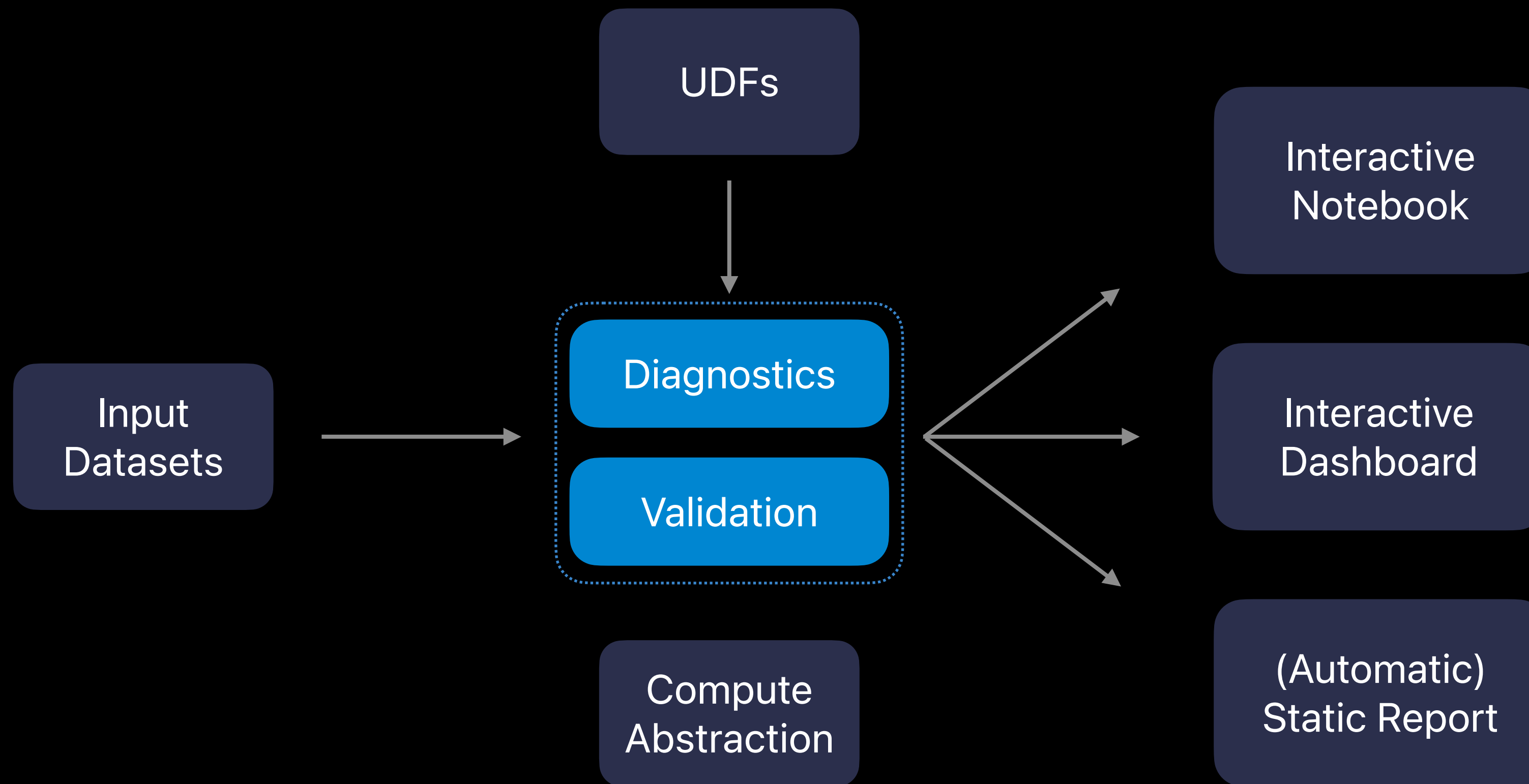
Collaboration & Reproducibility

- Documentation for data and models
- Share findings

Implementation

Data Tooling

High-level overview



Implementation

Required components for data insights

Data Diagnostics

- Interactive roll-up and drill-down
- Raw data preview

Data Validation

- Define constraints on sample or aggregate level

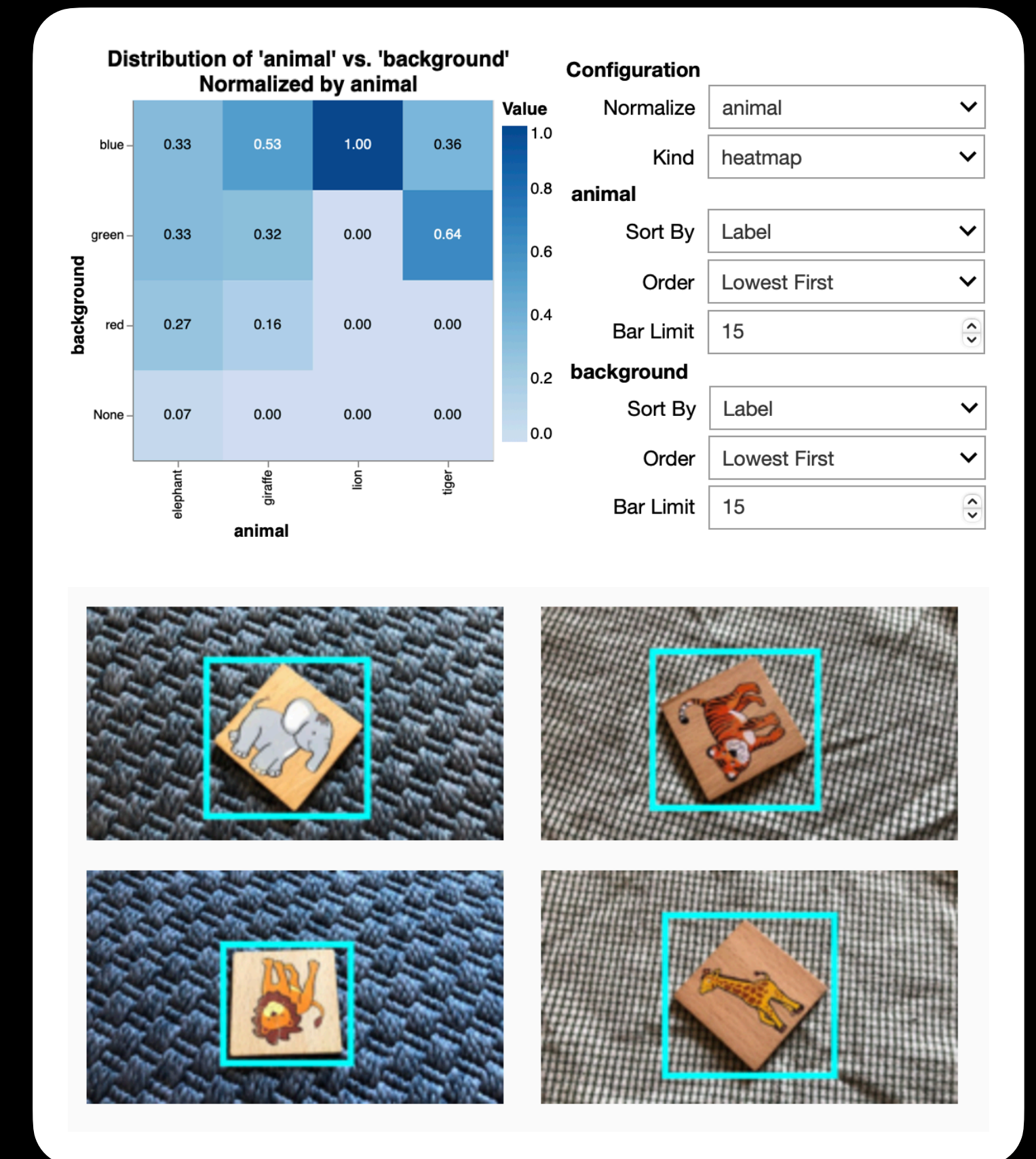
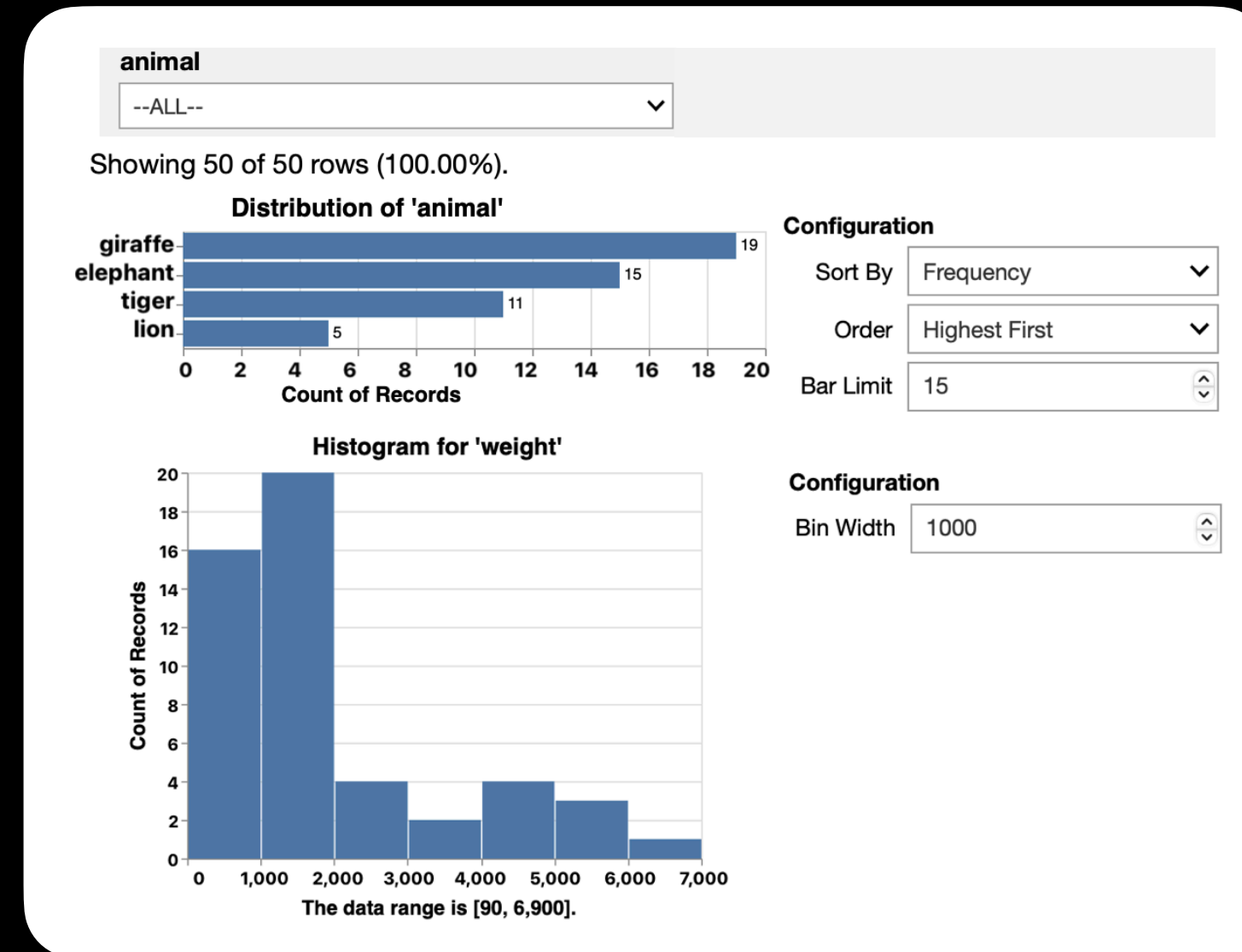
- Data Access abstraction enables loading data from many sources
- Compute abstraction enables portability

Data Diagnostics

```
def show_sample(row, **kwargs):
    # visualise image with bounding box

sample_view = SampleView(
    details [CustomDetail(show_sample)]
)

Explorer(
    filters ["animal"],
    views [
        "animal",
        "weight",
        ("animal", "background"),
        sample_view
    ]
).show(data)
```

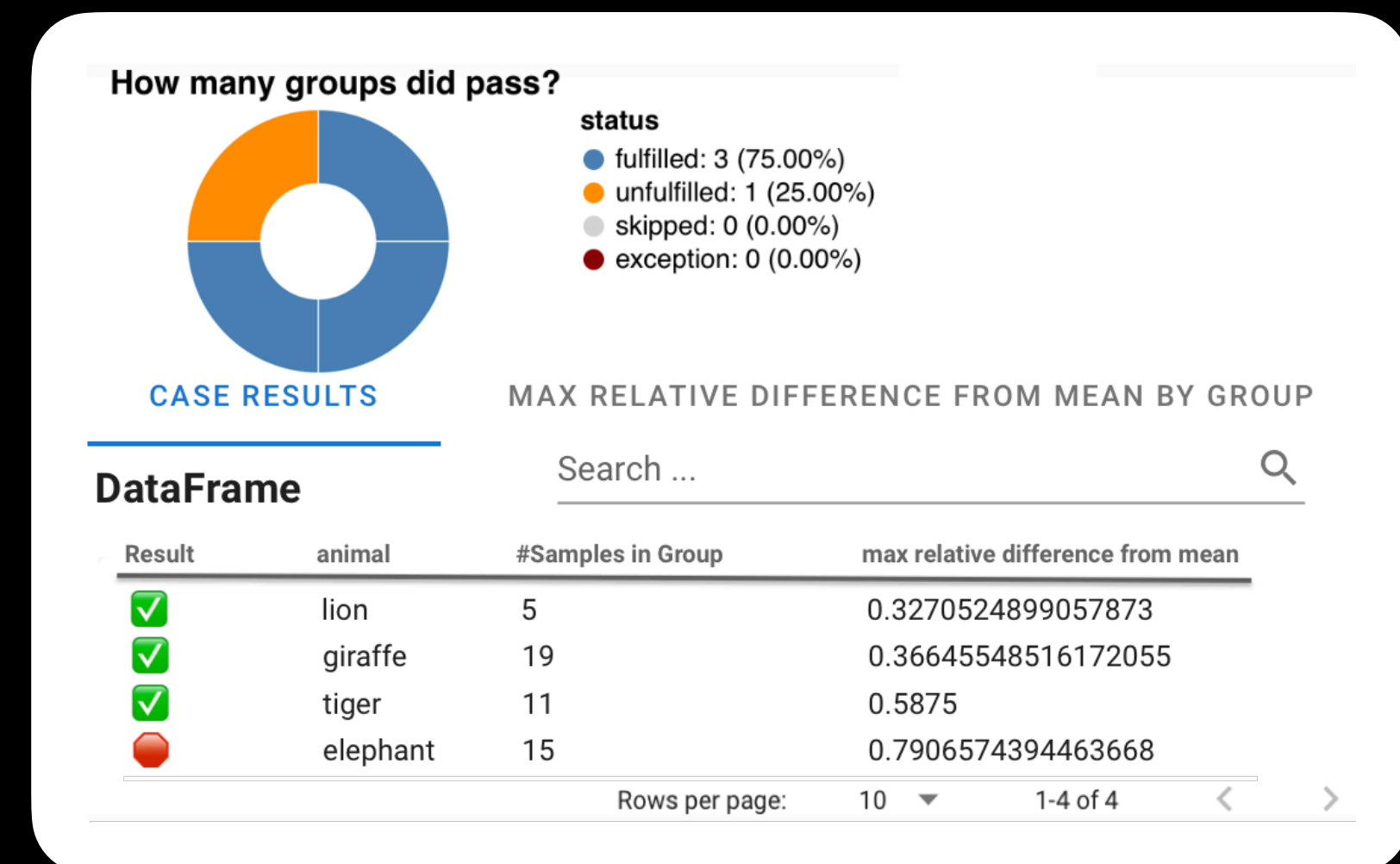
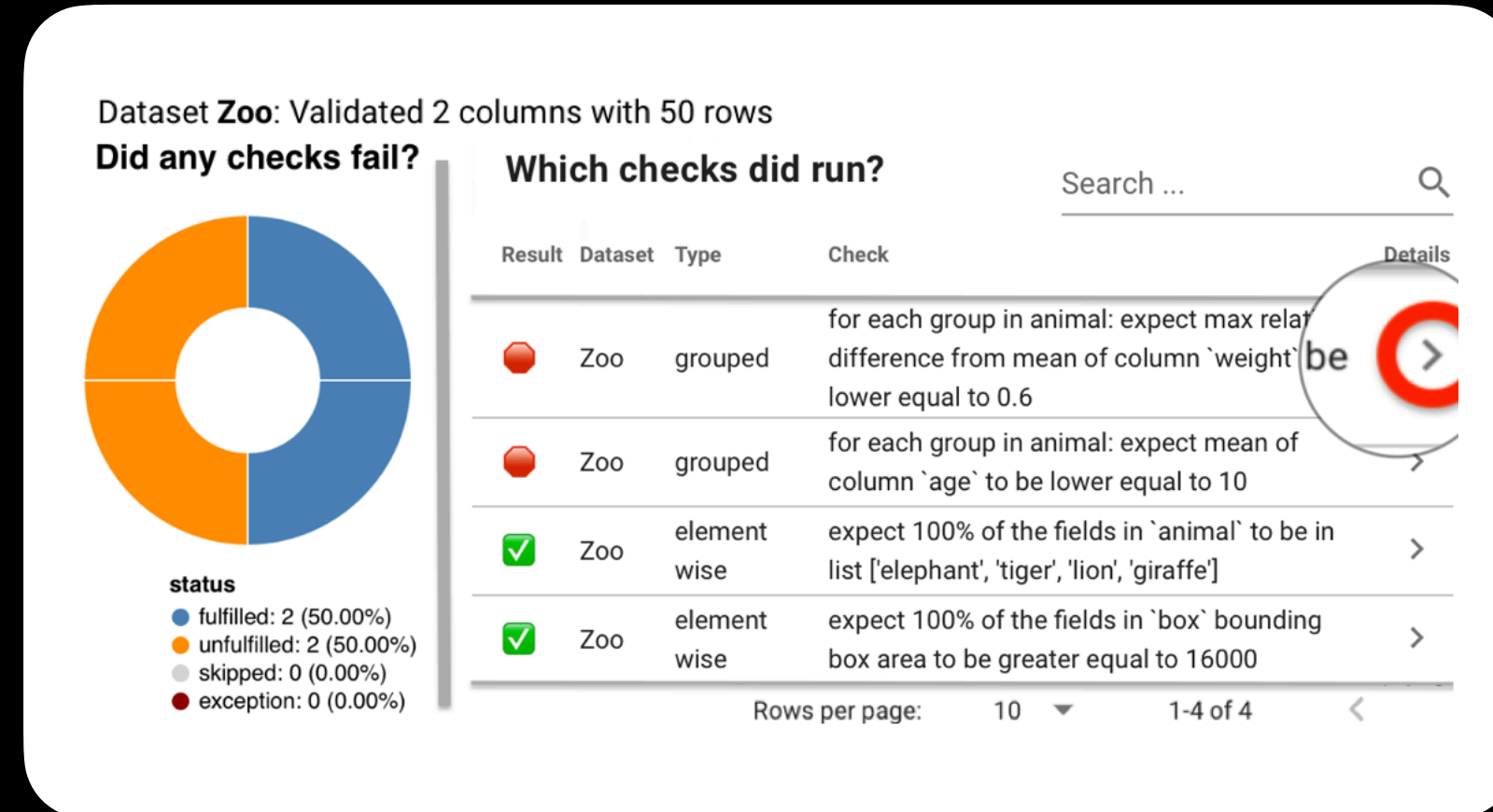


Data Validation

```
from udf import bbox_area, mean, max_diff_from_mean
```

```
checks = [  
    Cell("animal") in ["elephant", "tiger", "lion", "giraffe"],  
    Cell("box", bbox_area) >= 16000,  
    Group("animal", Column("age", mean) < 10),  
    Group("animal",  
        Column("weight", max_diff_from_mean) < 0.6  
    )  
]
```

```
Validator(expectations=checks).run(data)
```



Case Studies

Robust and Reproducible Dataset Documentation

- Capture context about datasets in documentation
- Enrich with summary statistics and raw data preview
- Transition seamlessly to interactive usage

The image displays a Jupyter Notebook interface with two main panels. The left panel, titled "Dataset Contents", contains introductory text and a link to "Launch an interactive session". The right panel, titled "Dataset Report", shows a code cell with a Python script using the `sl` library to generate a report. Below the code, the report includes a "Relevant Charts" section with a slider for "weight" (range 90-6,900) and two charts: a horizontal bar chart for "Distribution of 'animal'" and a histogram for "Histogram for 'weight'". The bar chart shows counts for giraffe (14), elephant (11), tiger (5), and lion (14). The histogram shows a distribution of weights. The right panel also includes configuration options for the charts, such as "Sort By: Frequency", "Order: Highest First", "Bar Limit: 15", and "Bin Width: 1000".

Dataset Contents

This is an artificial dataset for demonstration purposes only. It is used for the data exploration too.

Both the static report below as well as the interactive Jupyter notebook source repository which can be found [here](#).

Interactive Jupyter Session

[Launch an interactive session](#) of the report you see below!

Static Report

Relevant Charts

Filters are deactivated in this static report. Execute the notebook to activate filters.

weight

Showing 50 of 50 rows (100.00%).

Distribution of 'animal'

animal	Count of Records
giraffe	14
elefant	11
tiger	5
lion	14

Histogram for 'weight'

Dataset Report

Relevant Charts

```
In [6]: details = [
    sl.detail.Image("img_path", width=250),
    sl.detail.Row(["animal", "weight"]),
]

sl.Explorer(
    filters=['weight'],
    views=['animal', 'weight', sl.SampleView(details, grid_column_count=2)],
    reset_views=False
).show(data)
```

weight

Showing 50 of 50 rows (100.00%).

Distribution of 'animal'

animal	Count of Records
giraffe	14
elefant	11
tiger	5
lion	14

Histogram for 'weight'

Configuration

Sort By: Frequency

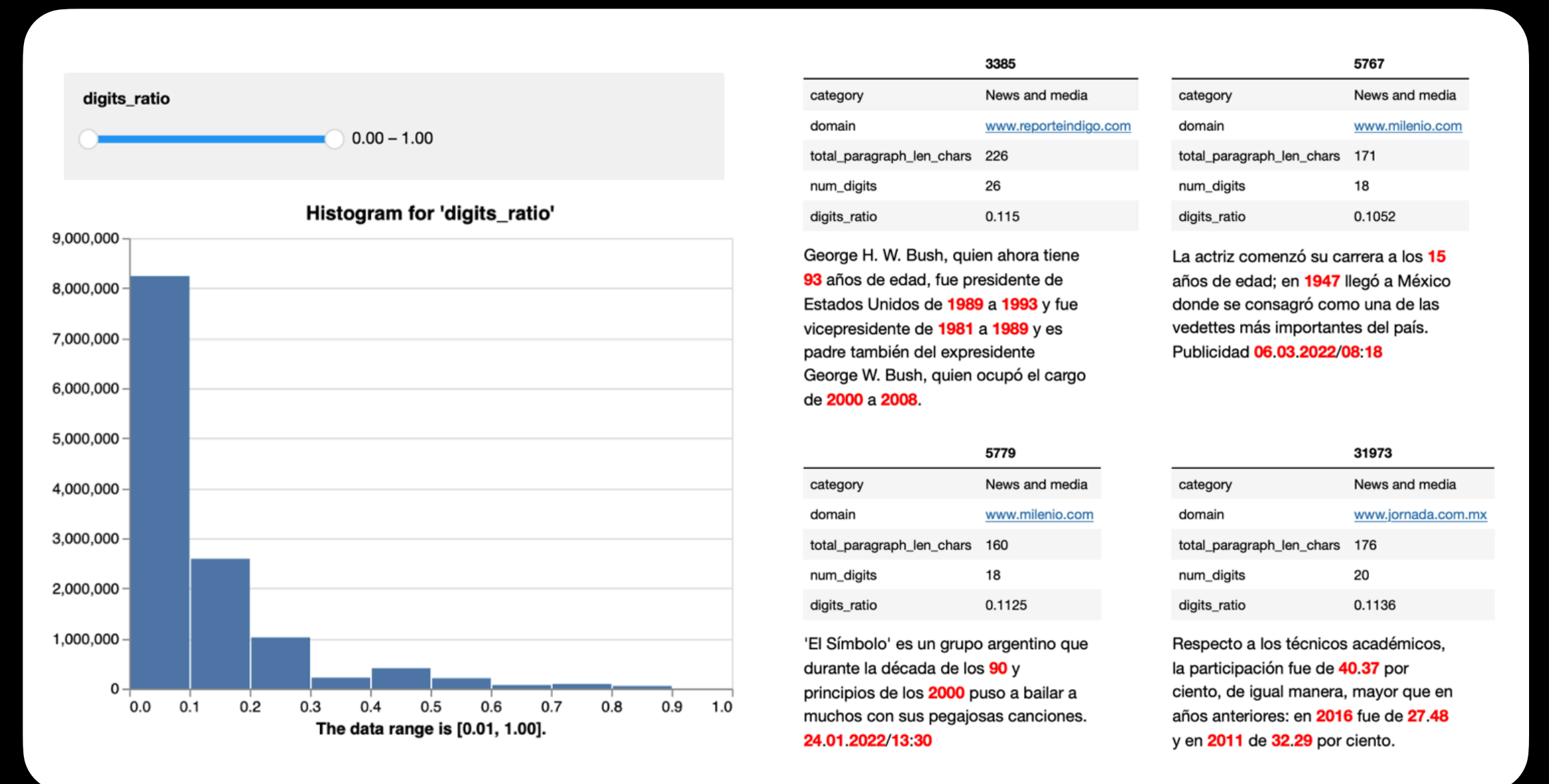
Order: Highest First

Bar Limit: 15

Bin Width: 1000

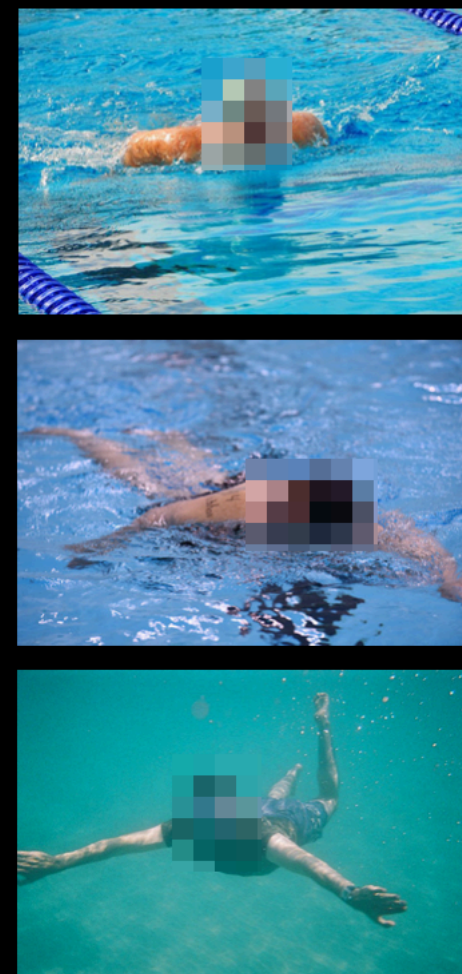
Large Corpora Curation with Linguists in the Loop

- Empower domain experts to set filters while inspecting raw data
- Enable interactivity for 100M+ paragraphs
- Seamless collaboration between ML engineers and domain experts



Actionable Failure Analysis of an Image Classifier

- When available metadata does not explain wrong predictions, visual inspection can reveal clues
- Image-text encoder models can be used to source new data in the same interactive session



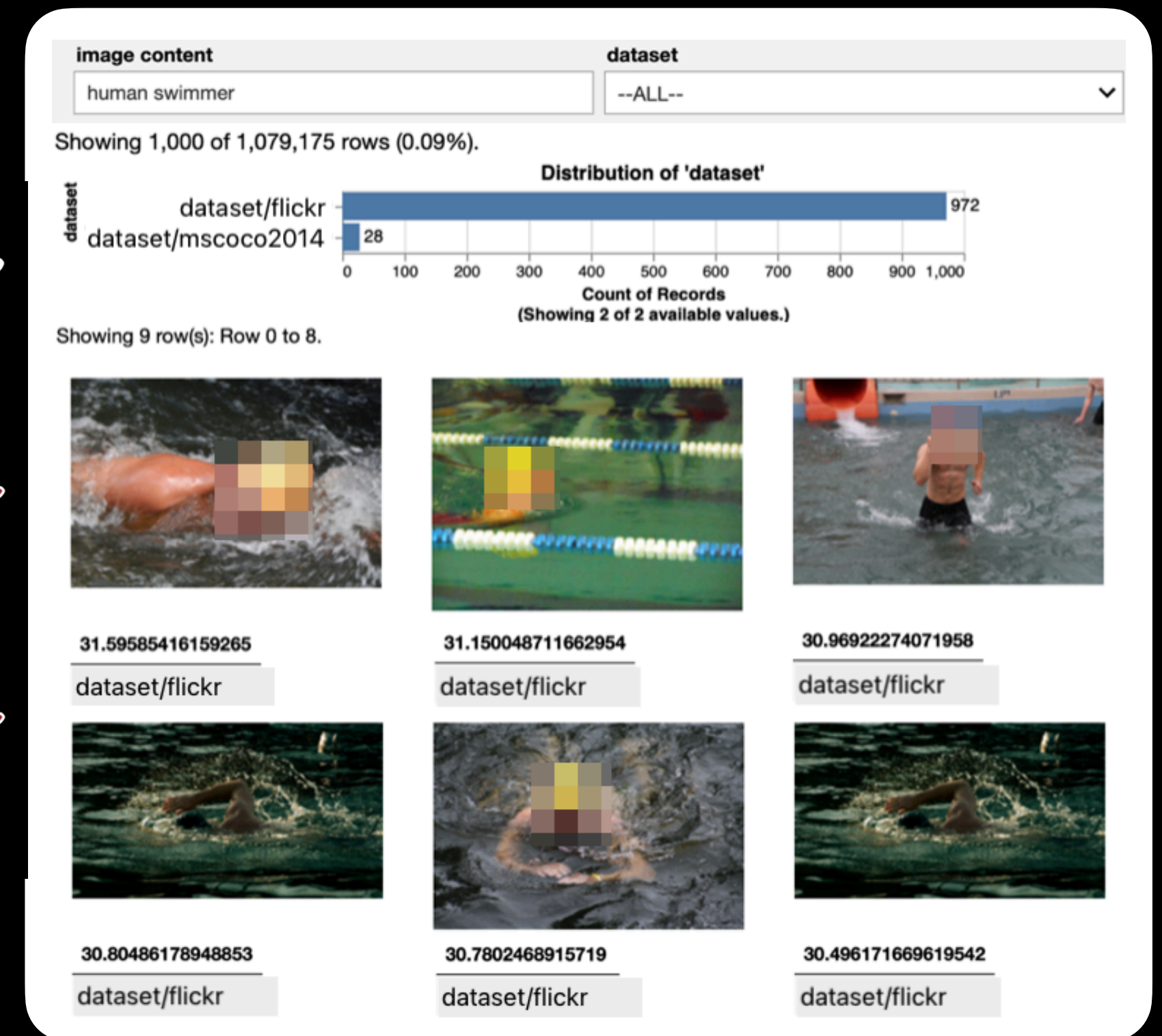
Human



Hippo



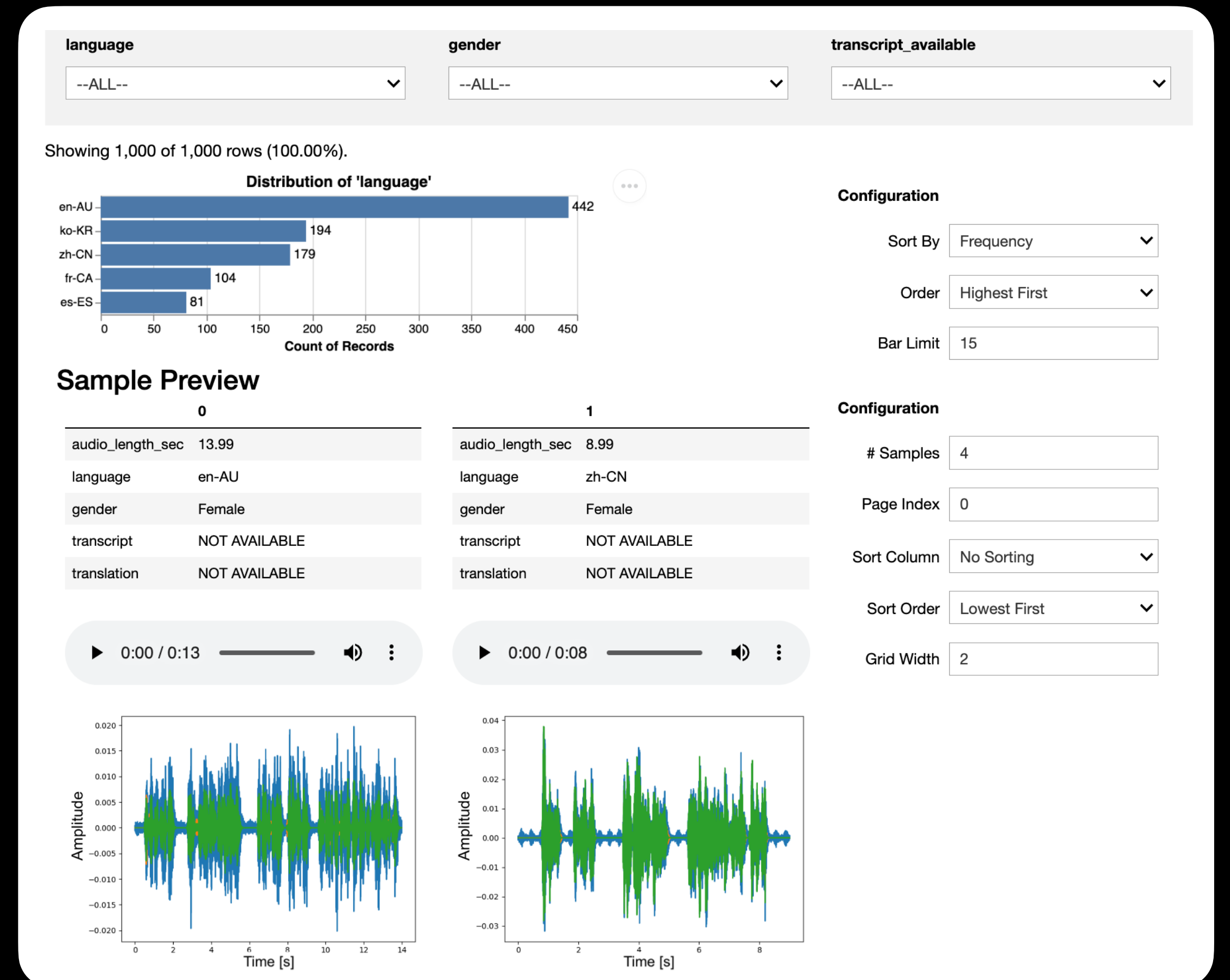
Dolphin



Rapid Iteration on Multi-Modal Datasets

from heterogeneous data sources

- Multi-modal machine translation project
- Necessity to blend different data sources and modalities in a unified preview
- Learnings from exploration could be persisted in validation checks



Conclusions

- We presented ADIML, a toolset to democratise data technology throughout the ML lifecycle and to enable the data-centric ML approach.
- The design of ADIML is based on the set of challenges and pain points we collected and validated from a wide range of ML teams.
- The case studies showing how easily ADIML can enable ML teams to focus and to improve data quality at scale are testimonies of its values.

