Advanced Data Management (CSCI 640/490)

Scalable Dataframes

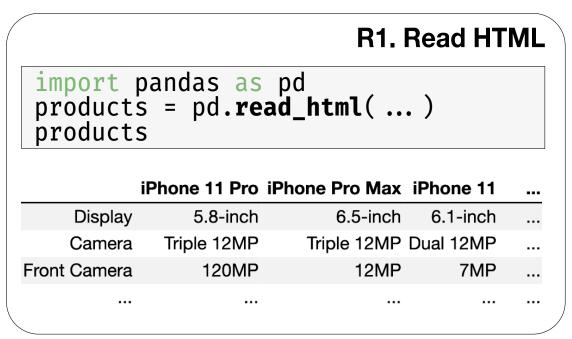
Dr. David Koop

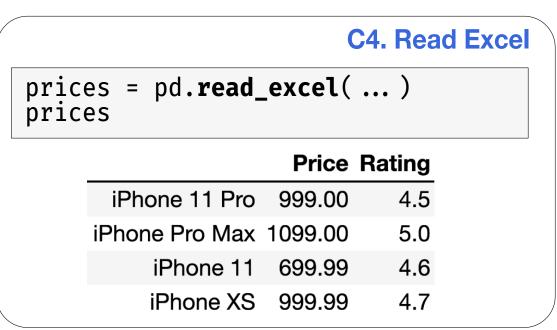


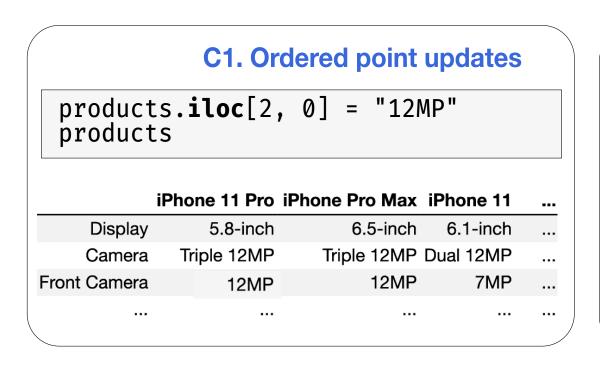
History of Dataframes

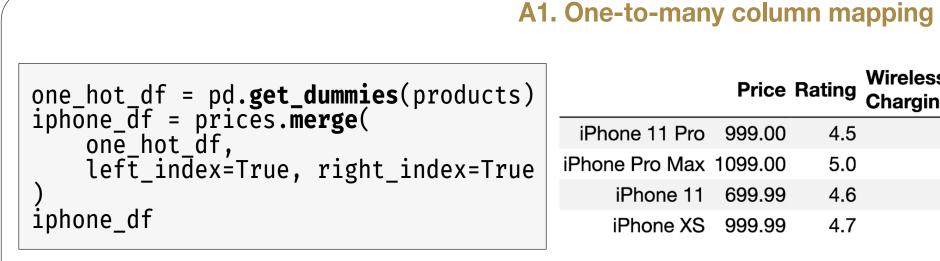
- Originally in Statistical Models in S, [J. M. Chambers & T. J. Hastie, 1992]
- R, open-source alternative to S, developed in 2000 (with dataframes)
- Pandas, 2009
- Spark, 2010 (resilient distributed dataset [RDD], Dataset API)

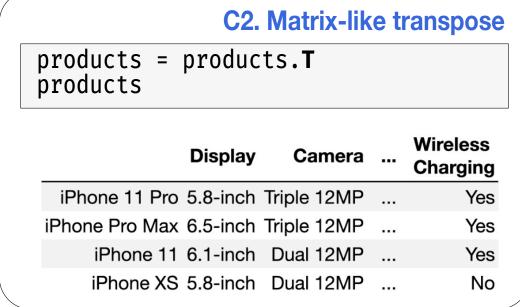
Pandas Workflow: Ingest, Cleaning, Analysis





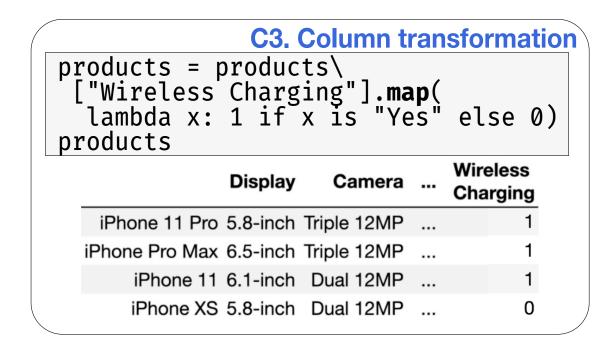


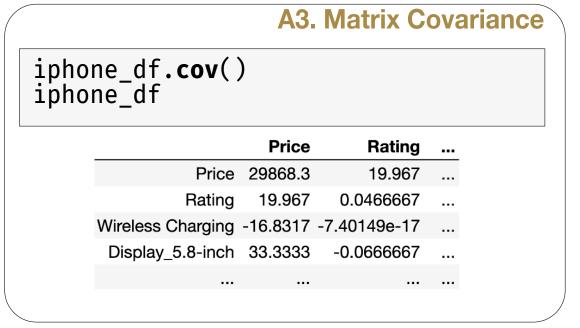




```
Wireless Display_\
                Price Rating Charging 5.8-inch
 iPhone 11 Pro 999.00
                         4.5
iPhone Pro Max 1099.00
                         5.0
     iPhone 11 699.99
                         4.6
                                             0
     iPhone XS 999.99
```

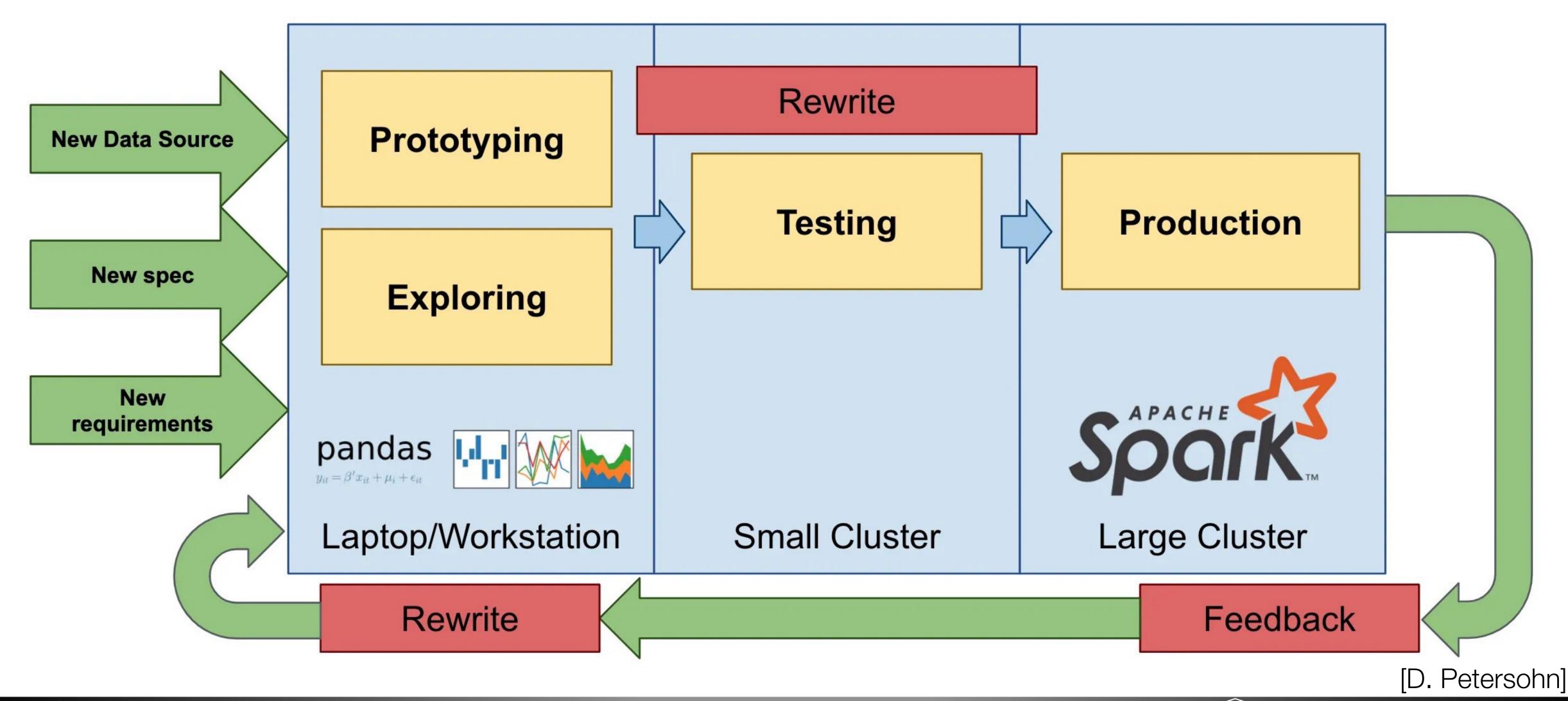
A2. Joins



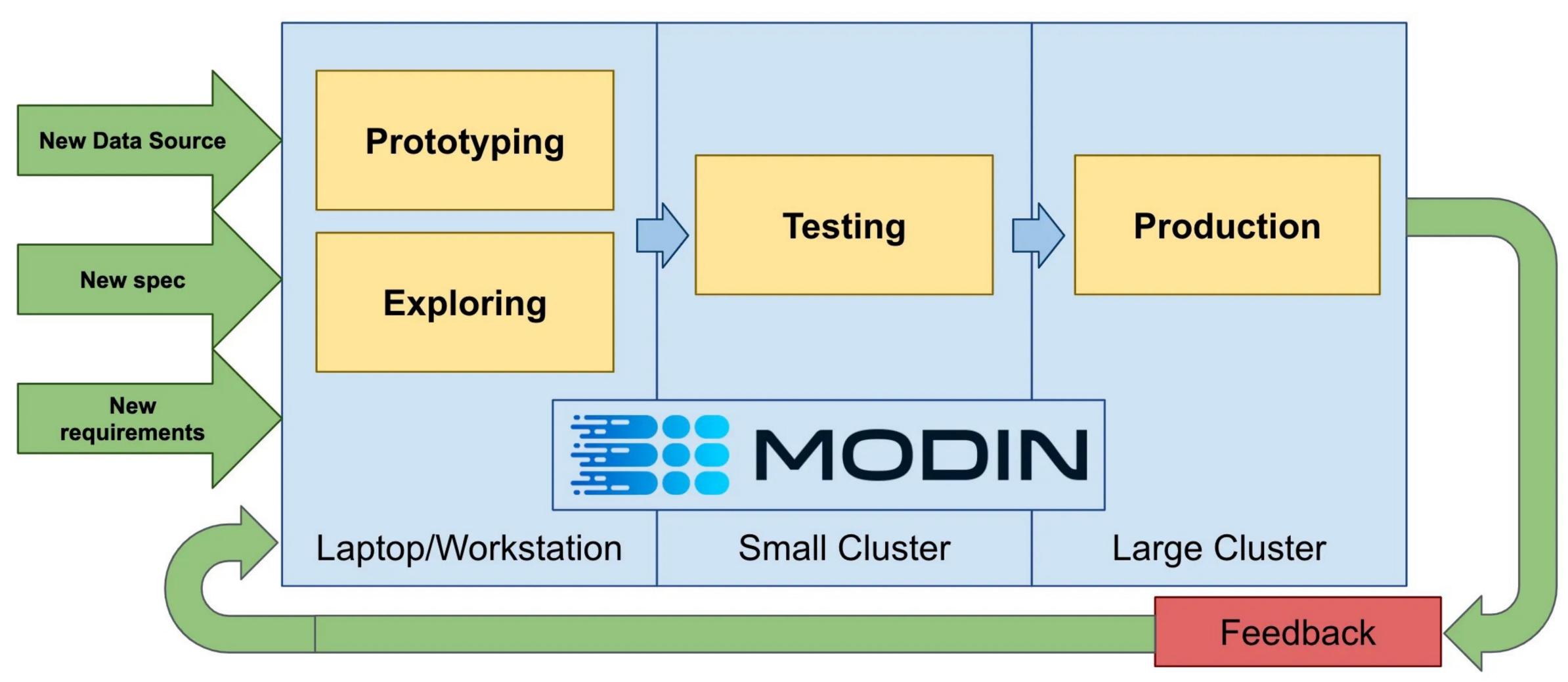




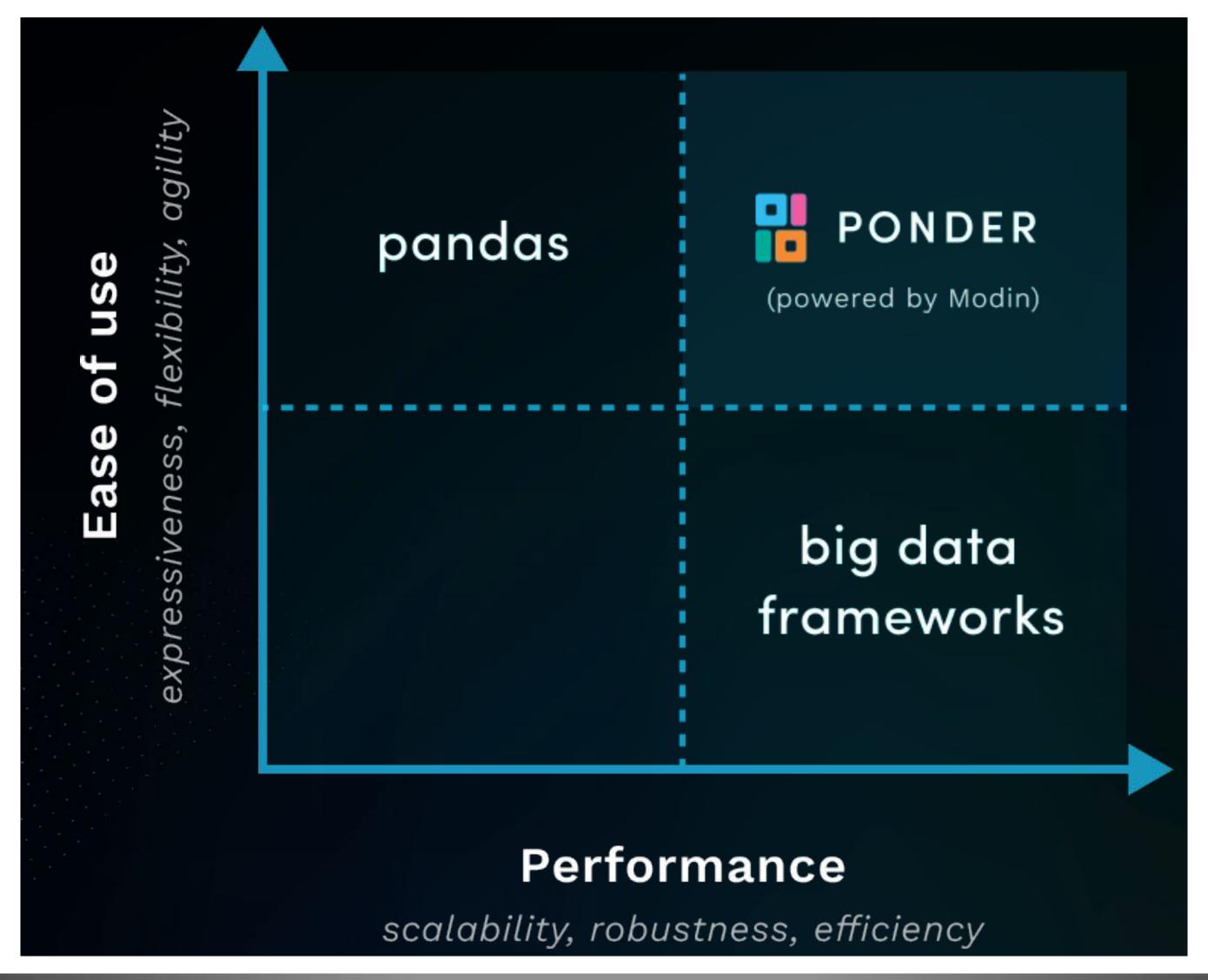
Problems Scaling: From Pandas to Other Solutions



Modin as a Solution

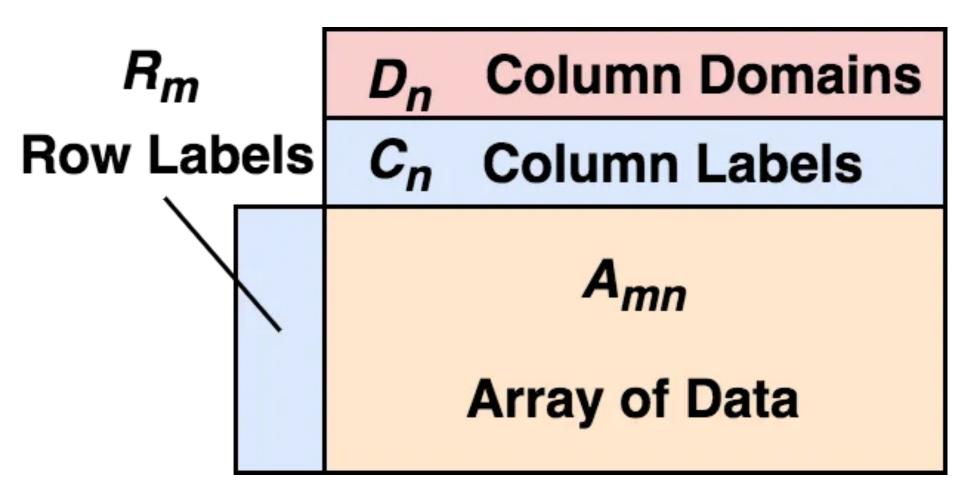


Modin Positioning



[D. Petersohn]

Dataframe Data Model



- Combines parts of matrices, databases, and spreadsheets
- Ordered, but not necessarily sorted
 - Rows and columns
- No predefined schema necessary
 - Types can be induced at runtime
- Typed Row/column labels
 - Labels can become data
- Indexing by label or row/column number
 - "Named notation" or "Positional notation"

[D. Petersohn]



Comparing Dataframes and Relational Stores

- Dataframe Characteristics
 - Ordered table
 - Named rows labels
 - A lazily-induced schema
 - Column names from d ∈ Dom
 - Column/row symmetry
 - Support for linear alg. operators

- Relational Characteristics
 - Unordered table
 - No naming of rows
 - Rigid schema
 - Column names from att
 - Columns and rows are distinct
 - No native support

Comparing Dataframes and Matrices

- Dataframe Characteristics
 - Heterogeneously typed
 - Numeric & non-numeric types
 - Explicit row and column labels
 - Support for rel. algebra operators

- Matrix Characteristics
 - Homogeneously typed
 - Only numeric types
 - No row or column labels
 - No native support

Dataframe Algebra

Operator	(Met	a)data	Schema	Origin	Order	Description
SELECTION		×	static	REL	Parent	Eliminate rows
PROJECTION		×	static	REL	Parent	Eliminate columns
UNION		×	static	REL	Parent [†]	Set union of two dataframes
DIFFERENCE		×	static	REL	Parent [†]	Set difference of two dataframes
CROSS PRODUCT / JOIN		×	static	REL	Parent [†]	Combine two dataframes by element
DROP DUPLICATES		×	static	REL	Parent	Remove duplicate rows
GROUPBY		×	static	REL	New	Group identical values for a given (set of) attribute(s)
SORT		×	static	REL	New	Lexicographically order rows
RENAME	(\times)		static	REL	Parent	Change the name of a column
WINDOW		×	static	SQL	Parent	Apply a function via a sliding-window (either direction)
TRANSPOSE	(\times)	×	dynamic	DF	Parent [♦]	Swap data and metadata between rows and columns
MAP	(×)	×	dynamic	DF	Parent	Apply a function uniformly to every row
TOLABELS	(\times)	×	dynamic	DF	Parent	Set a data column as the row labels column
FROMLABELS	(×)	×	dynamic	DF	Parent	Convert the row labels column into a data column



Pivot Example

Narrow Table (SALES)

Year	Month	Sales
2001	Jan	100
2001	Feb	110
2001	Mar	120
2002	Jan	150
2002	Feb	200
2002	Mar	250
2003	Jan	300
2003	Feb	310

Wide Table of Months

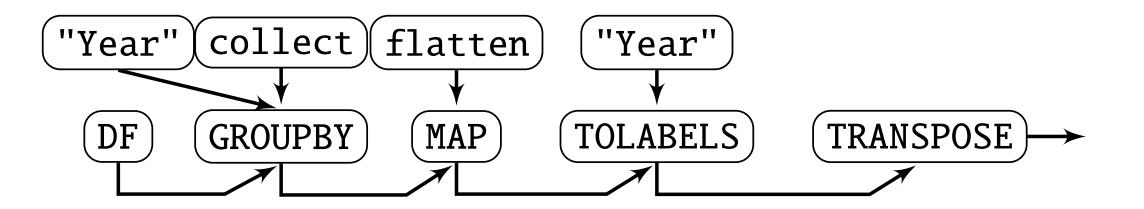
Month	2001	2002	2003
Jan	100	150	300
Feb	110	200	310
Mar	120	250	NULL

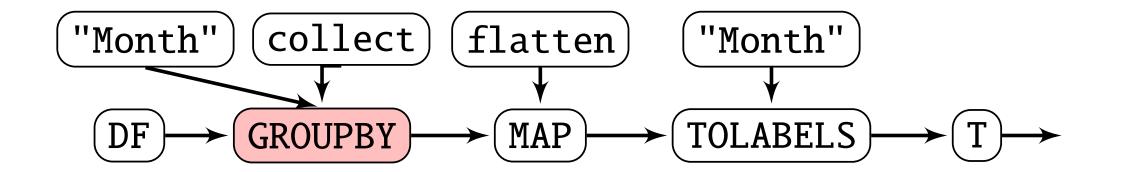
 $\overrightarrow{\text{Pivot}} \longrightarrow$

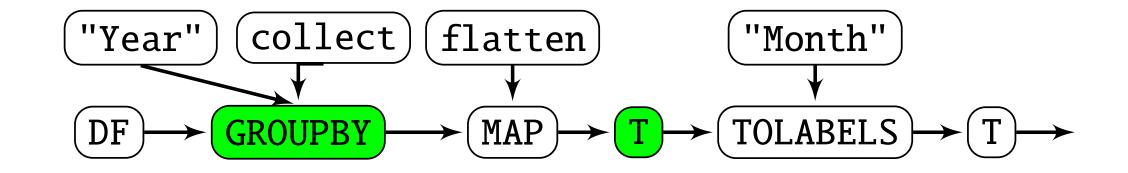
— Unpivot

Year	Jan	Feb	Mar
2001	100	110	120
2002	150	200	250
2003	300	310	NULL

Wide Table of YEARS







Modin Challenges

- Massive API: 240+ operators, but with a lot of redundancy
- Parallel Execution: row-based, column-based, and block-based
- Data Model Challenges: Schema induction, reusing type info
- Order is important
- Supporting billions of columns: Row/Column equivalence (transpose)
- Metadata is data (and vice versa)
- Users want immediate feedback
- Users want to create queries incrementally

Assignment 4

- Work on Data Integration and Data Fusion
- Integrate university ranking datasets from different institutions
- Integrate information based on names and matching
- Record Matching:
 - Which universities are the same?
- Data Fusion:
 - Names
 - Enrollments
 - Rankings
- Courselet is posted

Courselets

- All should now be available
- You should have received an email with a link to surveys

Test 2

- Upcoming... Nov. 10
- Similar format, but more emphasis on topics we have covered including the research papers

Dataframes, Databases, and the Cloud

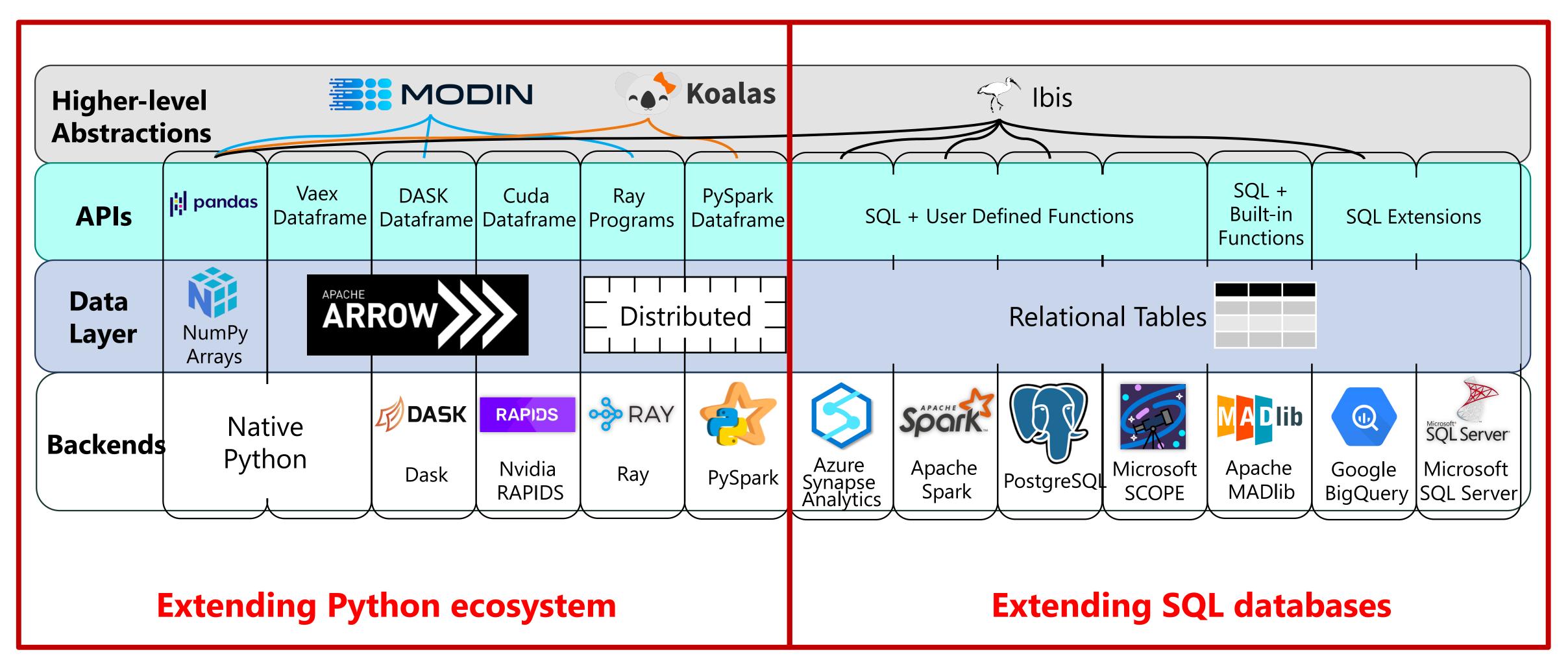
- How do we take advantage of different architectures?
- Lots of work in scaling databases and specialized computational engines
- What is the code that people actually write?

Magpie: Python at Speed and Scale using Cloud Backends

A. Jindal



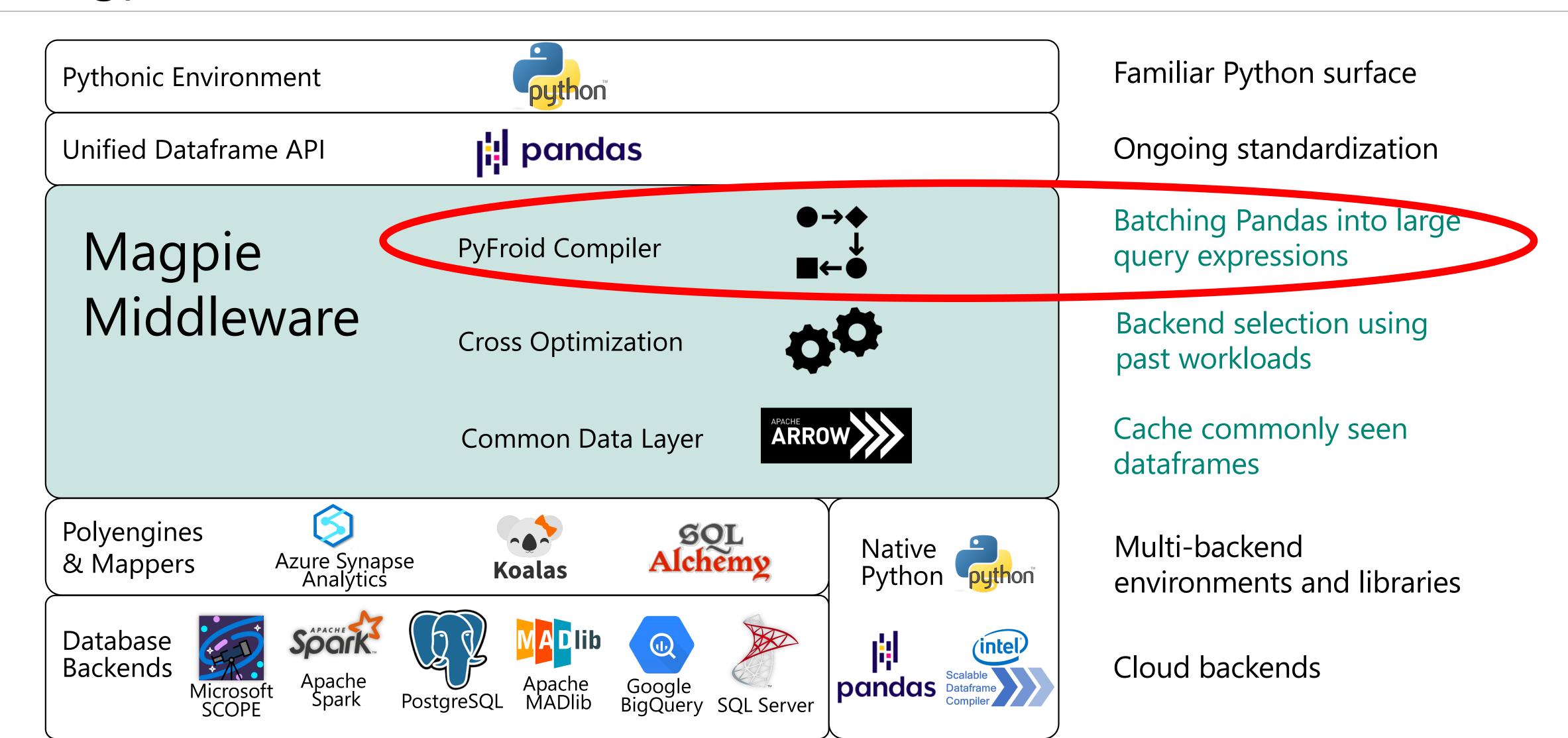
Data Science Jungle



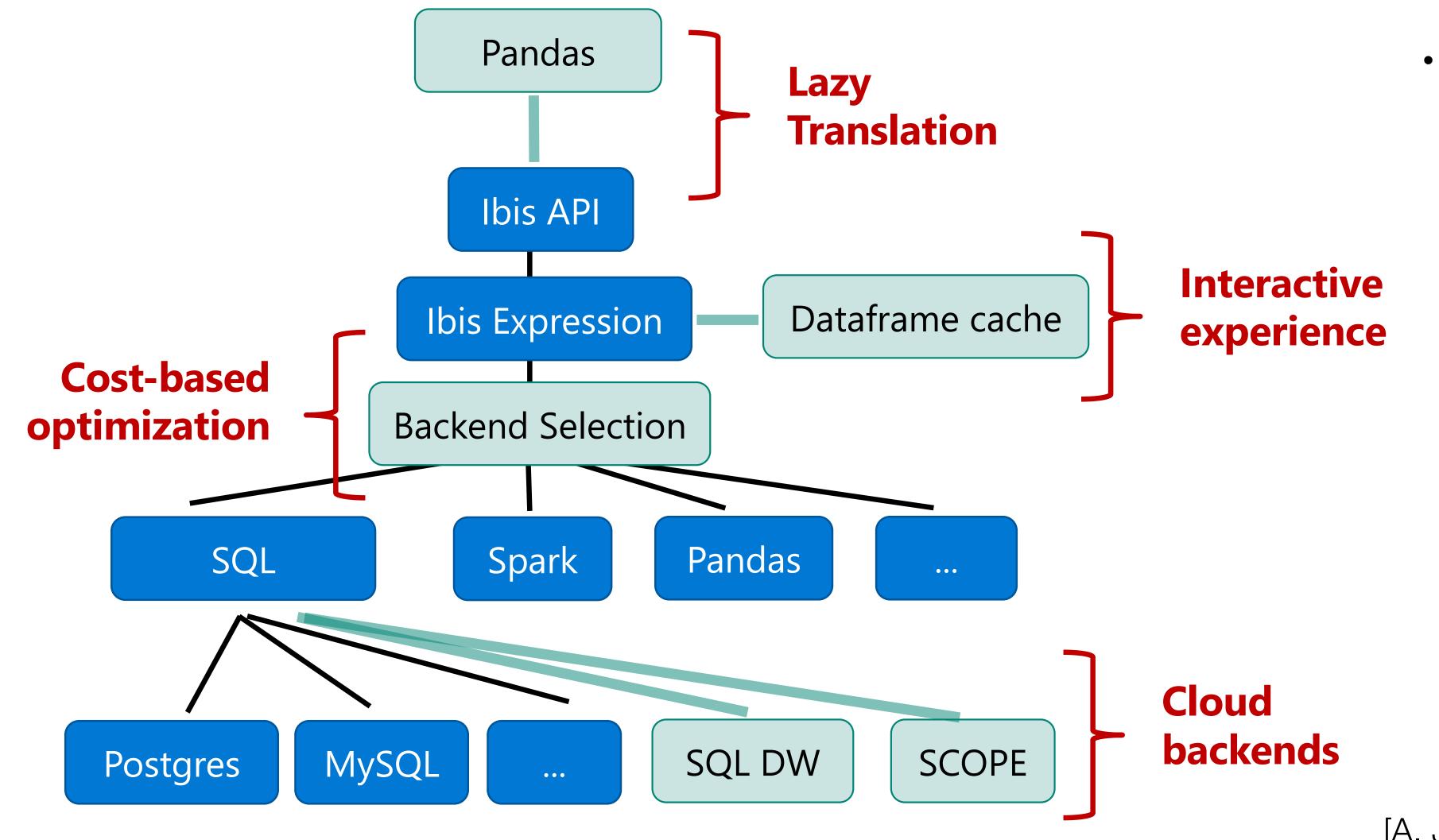
[A. Jindal et al., 2021]



Magpie Goals



Magpie Architecture

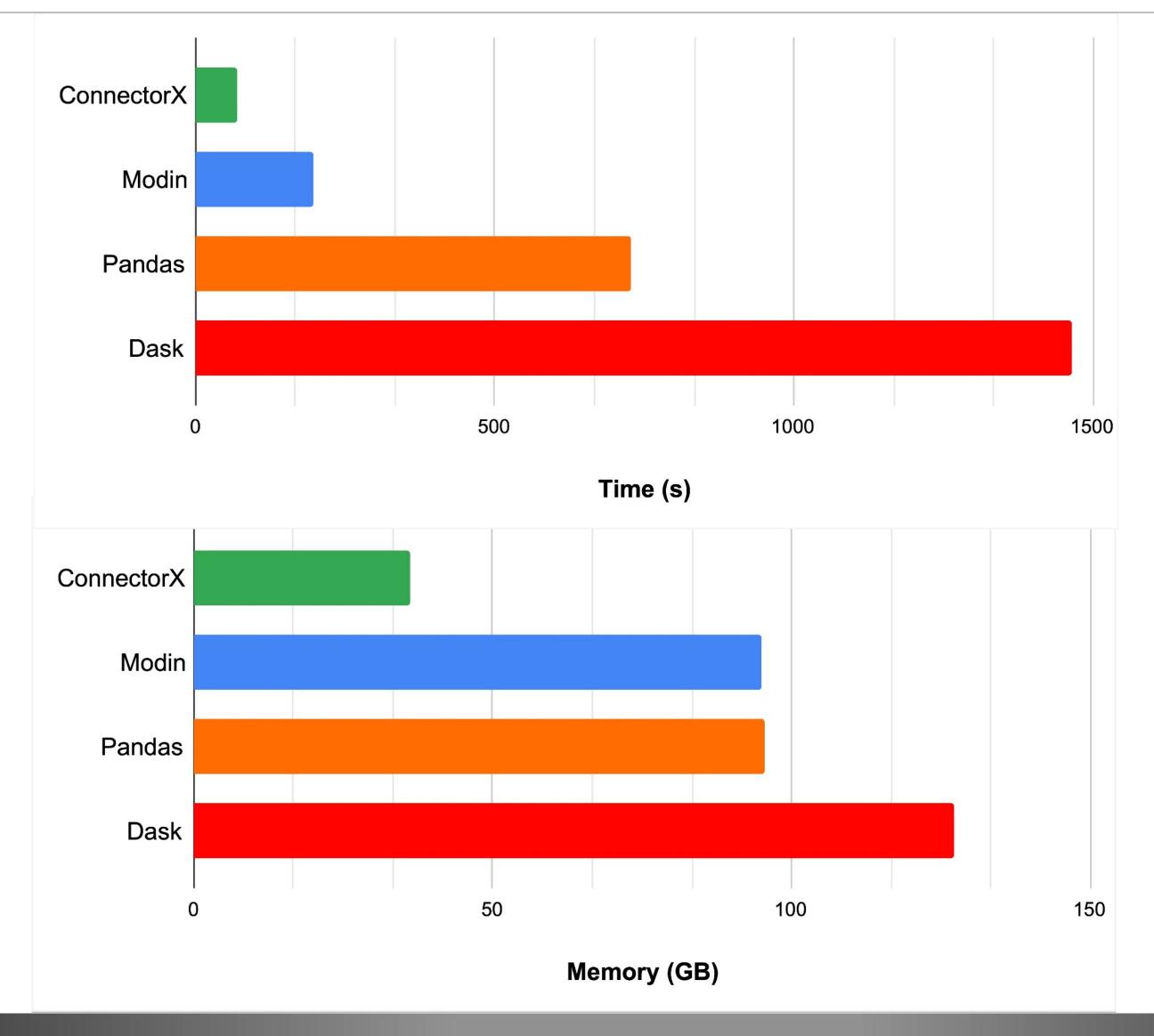


ConnectorX: Databases to Dataframes

- Write read_sql queries but write SQL
- Written in Rust
- Returns a dataframe

```
query = f"""
SELECT 1 orderkey,
 SUM(l_extendedprice * (1 - l_discount)) AS revenue,
 o orderdate,
 o shippriority
FROM customer,
 orders,
 lineitem
WHERE c mktsegment = 'BUILDING'
AND c custkey = o custkey
 AND 1 orderkey = o orderkey
 AND o orderdate < DATE '1995-03-15'
 AND 1 shipdate > DATE '1995-03-15'
GROUP BY 1 orderkey,
 o orderdate,
 o shippriority
\\ // //
df = read sql("postgresql://postgres:postgres@localhost:5432/tpch", query,
              partition_on="l_orderkey", partition num=4)
```

ConnectorX Speed & Memory



[X. Wang, 2022]

Improvements in ConnectorX

- Written in native language (Rust)
- Copy exactly once (even during parallel computations)
- CPU cache-friendly: process in a streaming fashion

An Opinionated Introduction to Polars

Nico Kreiling



Handling Large Data with Polars

Etienne Bacher



Databricks Architecture

M. Oharek



Discussion

- Data in the cloud and local exploration
- Languages: SQL or Spark or Polars or Pandas or Ibis or....?