

# インメモリデータ共有による 異種言語混合アプリケーション の高速化

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# 自己紹介

- 所属：富士通株式会社 コンピューティング研究所
  - 2017~2021年：不揮発性メモリやストレージの研究
  - 2022年～：量子コンピューティングと量子化学計算の研究
- 専門分野：コンピューターアーキテクチャ
  - 九州大学 井上研究室にて博士号取得
  - OS（特に仮想メモリ周り）も好きです
- 社外活動：
  - 2019~2023：情報処理学会 ARC研究会幹事、ACS論文誌編集委員
  - 2023～：**SNIA-J 次世代メモリ&ストレージ分科会 会長**



# 本日の内容

- IEEE Big Data 2023 にて発表した慶應大河野研との共著論文の紹介
- タイトル : *Accelerating Multilingual Applications with In-memory Array Sharing*
- 著者 : 野澤 真伸、今村 智史、河野 健二
- 3行まとめ :
  - データ分析アプリに **Python** と **Julia** を併用すれば両者の強みを活かして高速化できる
  - ただし、両言語間でのデータやり取りオーバーヘッドが結構大きい
  - メモリ上のデータを両言語間で直接共有できるようにしてそのオーバーヘッドをほぼ 0 に !

# Accelerating Multilingual Applications with In-memory Array Sharing

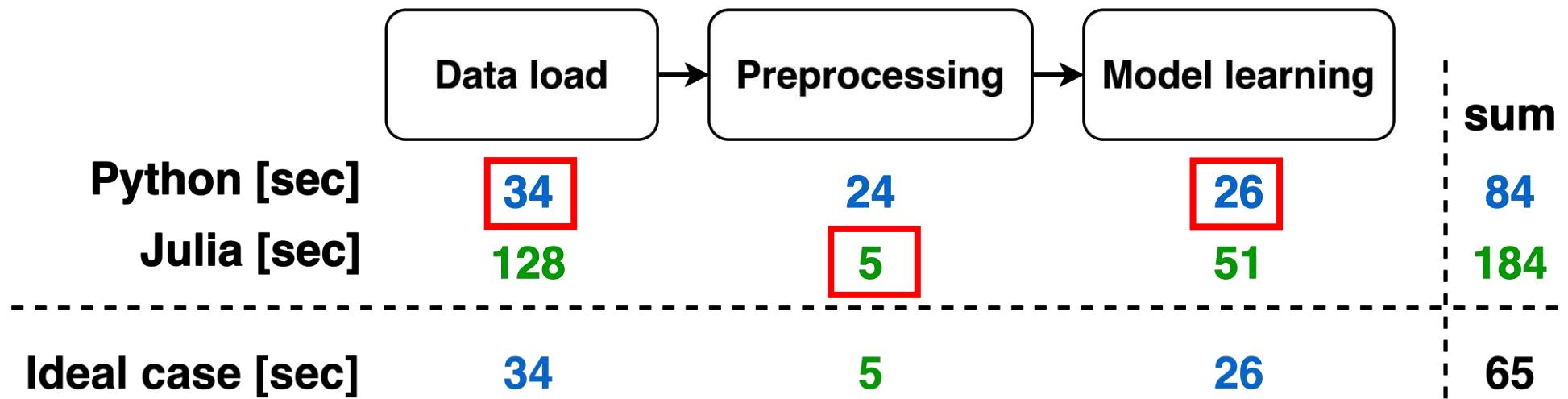
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Keio University\*, Fujitsu Limited\*\*

# Multilingual Data Analysis Applications

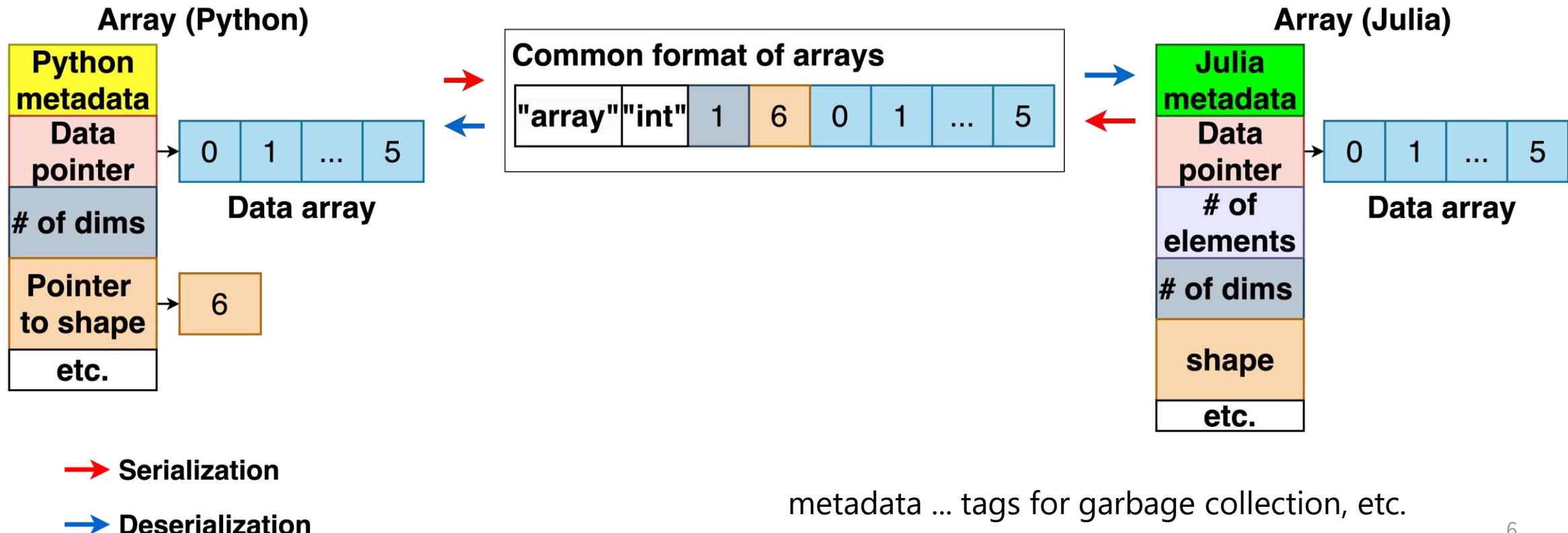
- Combine multiple programming languages for data analysis
  - Can take the advantage of each language
- Can outperform *monolingual* implementations

E.g., Execution time of Kmeans applications



# Data Exchange in Multilingual Applications

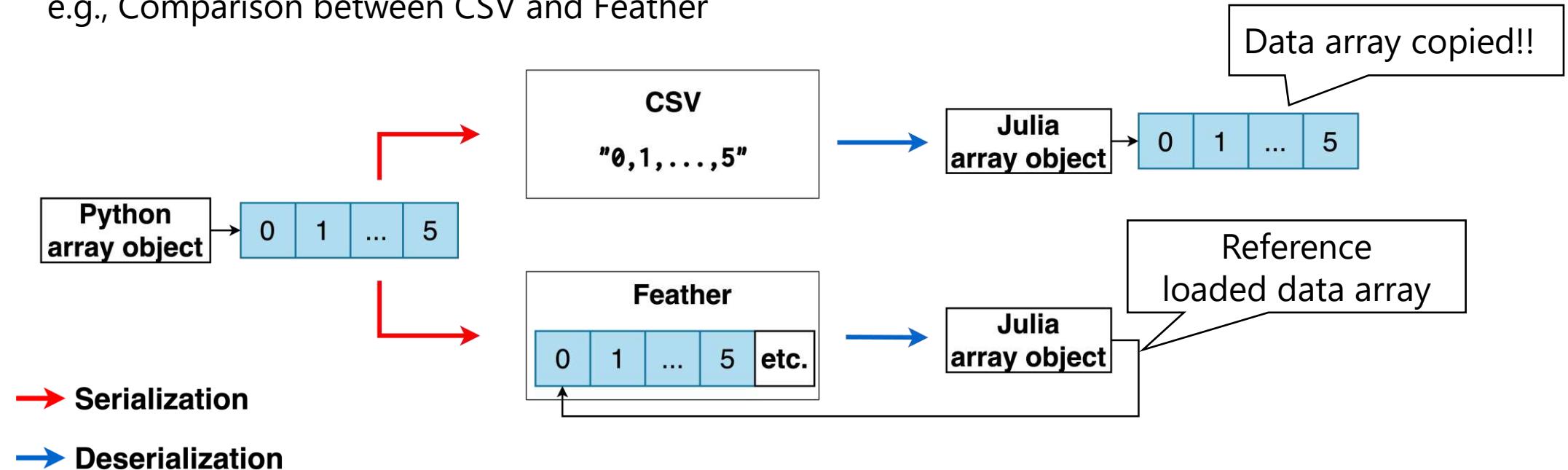
- The in-memory object layout depends on each language
- **Serialization/deserialization (S/D)** is necessary for data exchange



# Feather: Fast S/D Format

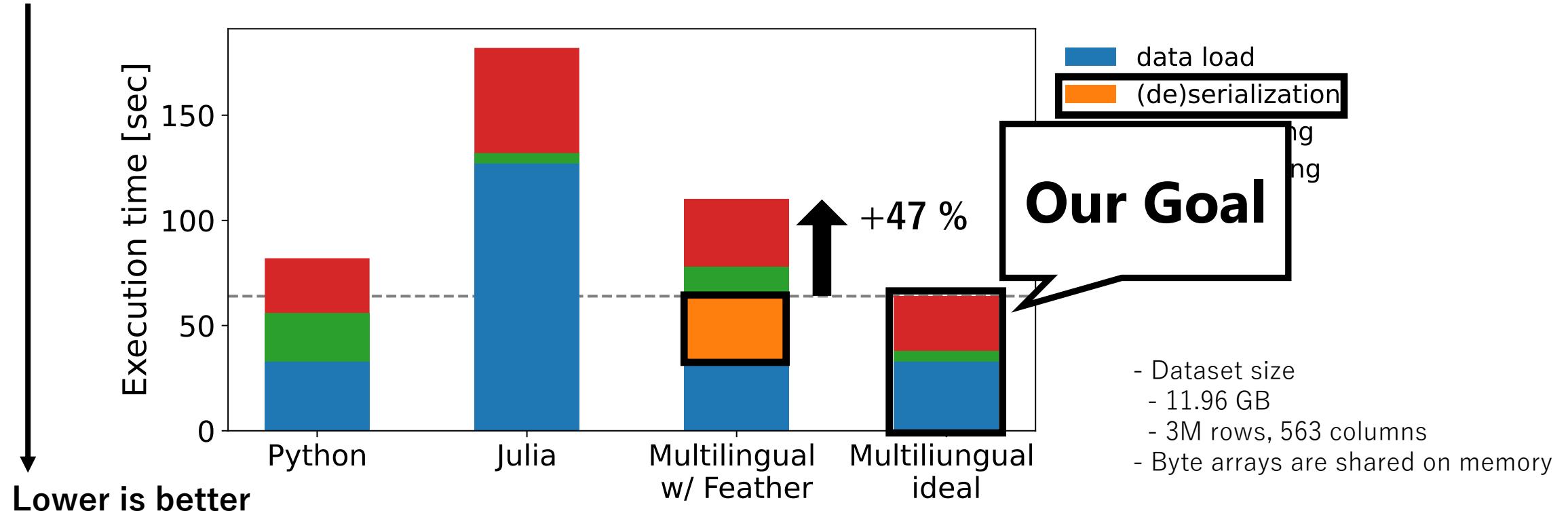
- S/D format for dataframes and tables (= collections of arrays)
- Feather holds in-memory layout of array as it is
  - ✓ ◦ **Eliminates the copy of data arrays in deserialization**
    - Serialization copies data arrays

e.g., Comparison between CSV and Feather



# S/D is Time-consuming

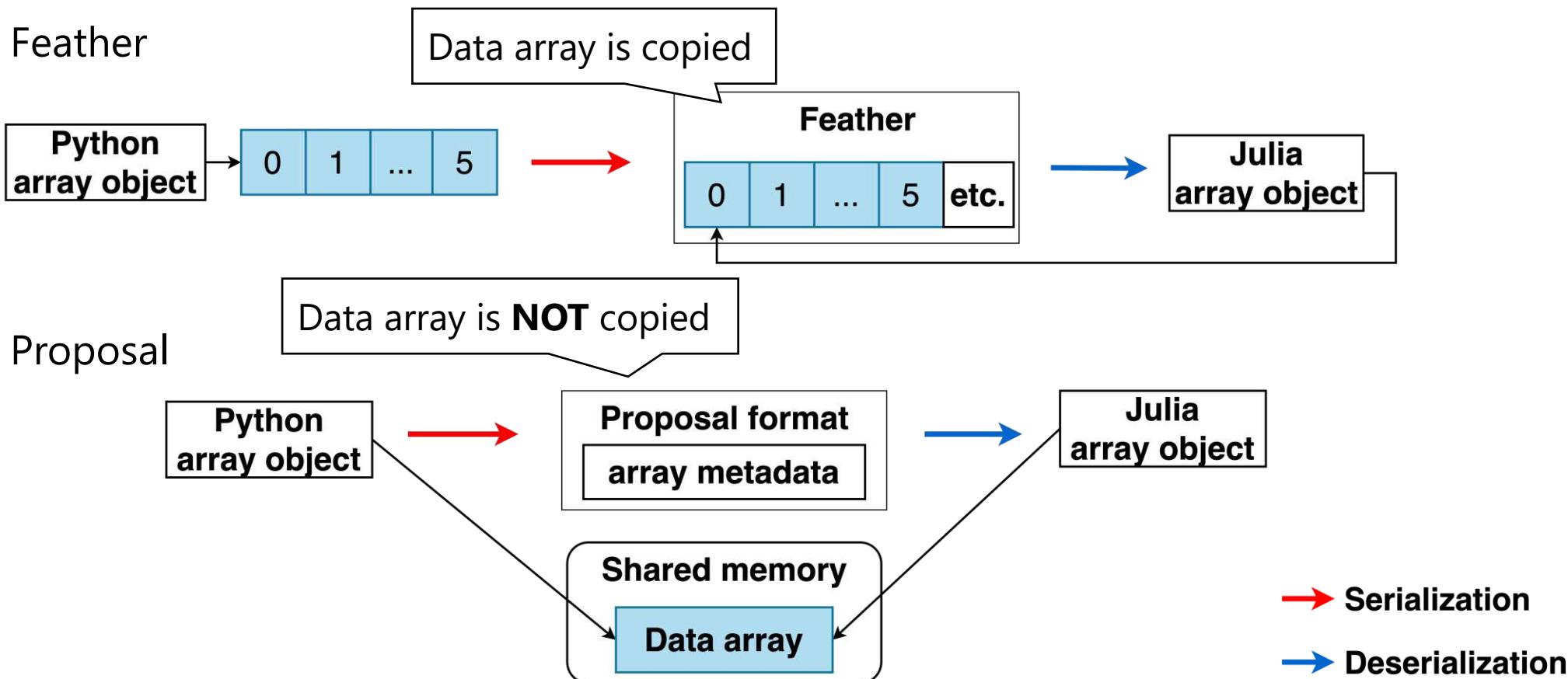
- **Array serialization** degrades the total execution time
  - Multilingual implementation can be rather slower than monolingual one!!



e.g., the execution time of the Kmeans application

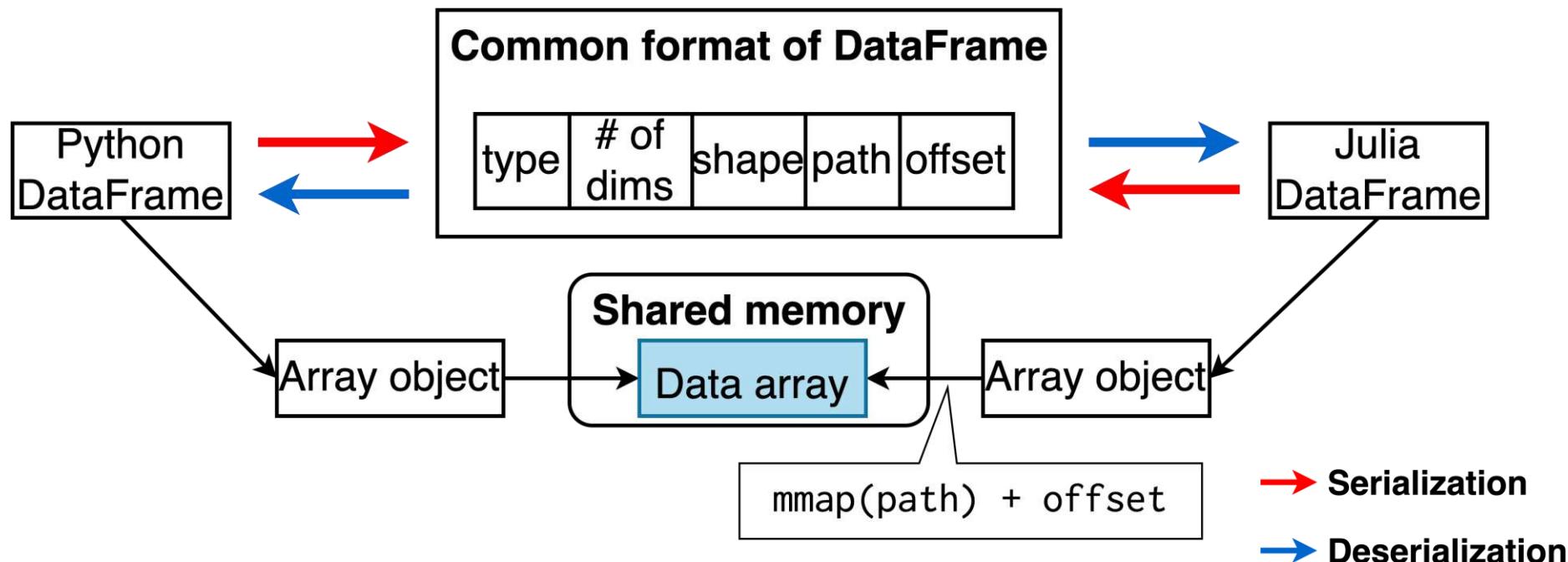
# Proposal: In-Memory Array Sharing

- Directly exchanges data arrays via shared memory
- Only S/D array metadata



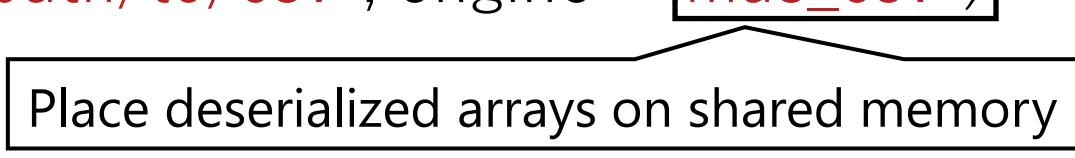
# Implementation

- S/D modules for dataframe objects in Python and Julia
- Each array metadata contains:
  - Type, number of dimensions, and shape of the array
  - **Path** and **offset** to access shared memory



# Programming Model

```
import pandas
import in_memory_array_sharing as imas # proposal module

# Load dataset from CSV
df = pandas.read_csv("/path/to/csv", engine="mas_csv")  


Place serialized arrays on shared memory

# Serialize dataframe
imas.dump(df, "/path/to/common_format", "julia")  


Specify Julia deserializes df

# Deserialize dataframe
df = imas.load("/path/to/common_format")
```

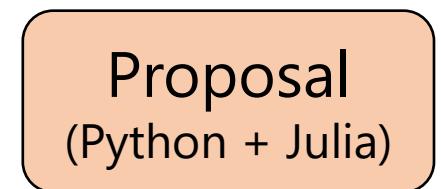
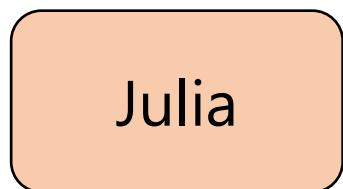
# Evaluation

- Compare with two criteria
  - Execution time
  - Scalability to data size

## Experimental setup

OS	linux-5.4.0
CPU	dual Intel® Xeon® Gold 6330
DRAM	DDR4 377 GB
CPython	ver. 3.9.7
Julia	ver. 1.6.2
Apache Arrow	ver. 10.0.1

Targets



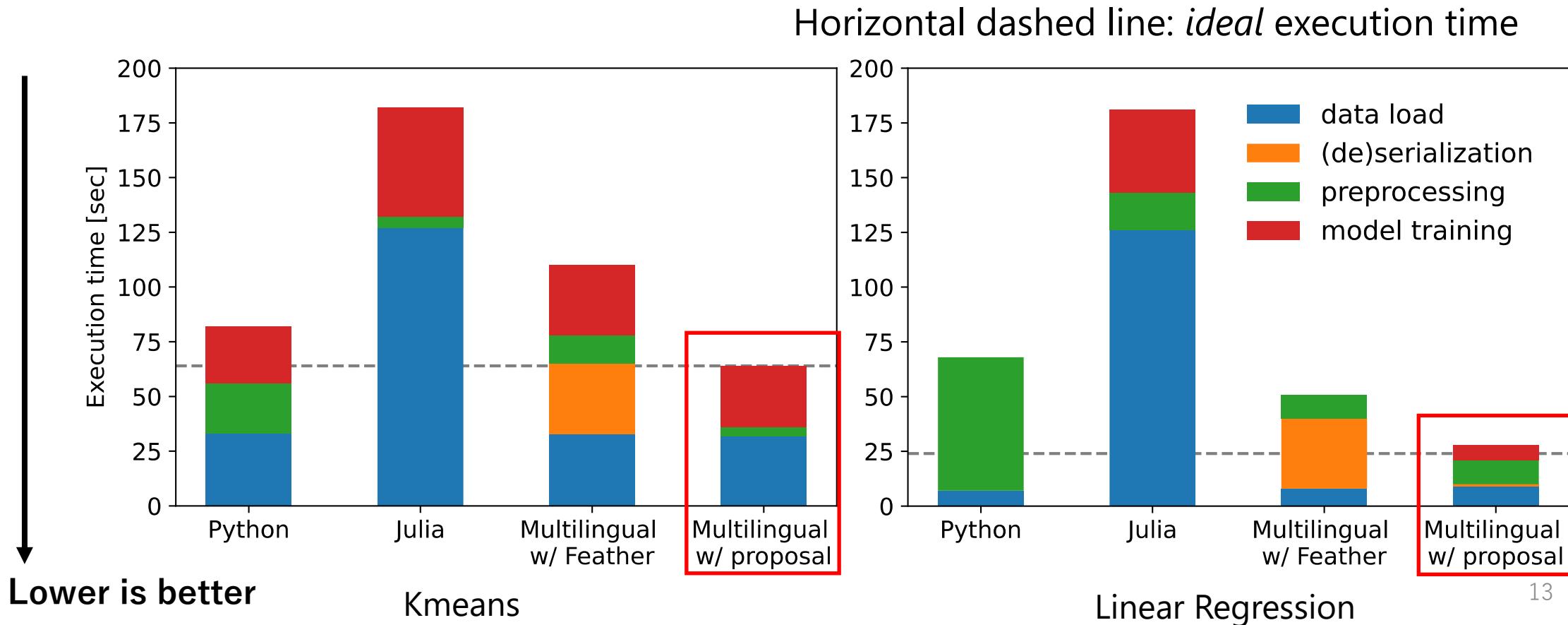
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Evaluated applications  
(referred from Kaggle notebooks)



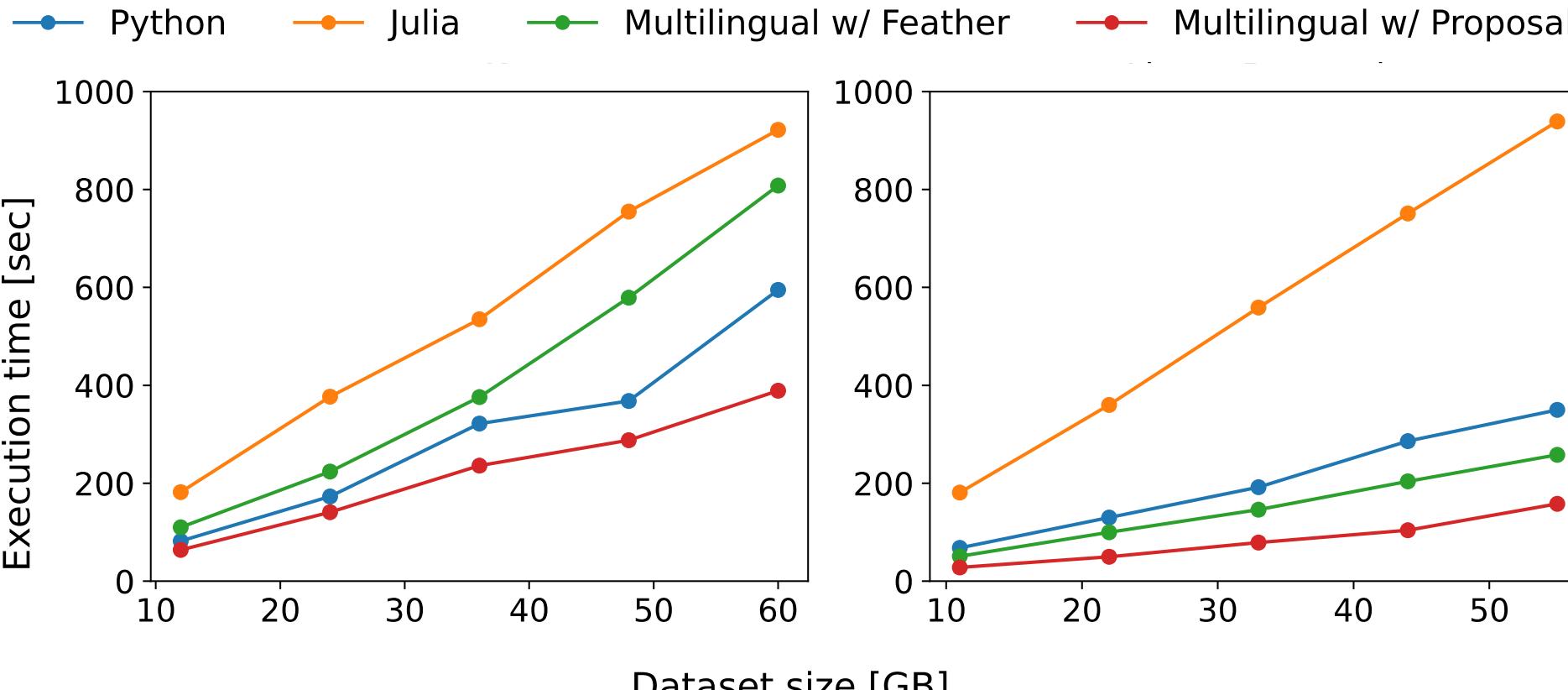
# Execution Time

- Proposal achieves nearly **ideal performance**
  - faster than Python/Julia/Feather by up to 56.3/83.0/43.6 %



# Scalability to Data Size

- Proposal achieves **higher scalability** than the others



Lower is better

Kmeans

Linear Regression

# Conclusion

- Multilingual applications can outperform monolingual ones
- Multilingual applications suffer from array S/D
- We propose **in-memory array sharing for Python+Julia**
  - Only serialize and deserialize array metadata
  - Copying data arrays can be eliminated
- The evaluation shows
  - **speedups** by up to 56.3/83.0/40.8 % against Python/Julia/Feather
  - **higher scalability** than the other implementations
- Our proposal would go well with Compute Express Link (CXL)
  - Exchange arrays in CXL memory with in-memory array sharing



# Q&A