

Sharing is caring

Efficient Data Exchange with PyArrow

By: Alenka Frim, Raúl Cumplido and Rok Mihevc

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- What is Apache Arrow?
- Interprocess Communication
- C Data Interface and Its Extensions
- The Arrow PyCapsule Interface
- Flight RPC
- Arrow over HTTP
- ADBC
- Q&A

Who are we?

- Rok Mihevc
- Independent
(Arctos Alliance)
- Arrow C++,
Parquet



- Raúl Cumplido
- QuantStack
- PyArrow, Arrow C++,
CI and general
project maintenance

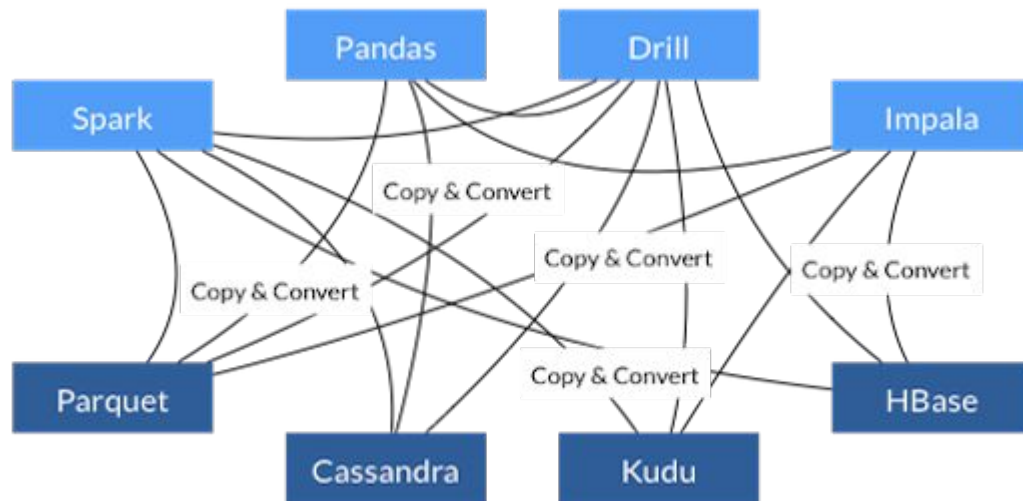


- Alenka Frim
- Independent
(Arctos Alliance)
- PyArrow and general
project maintenance

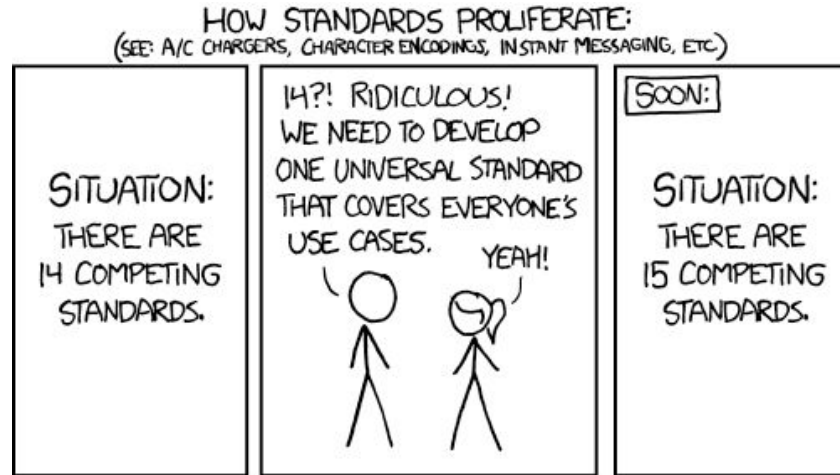


What is Apache Arrow?

The initial Problem

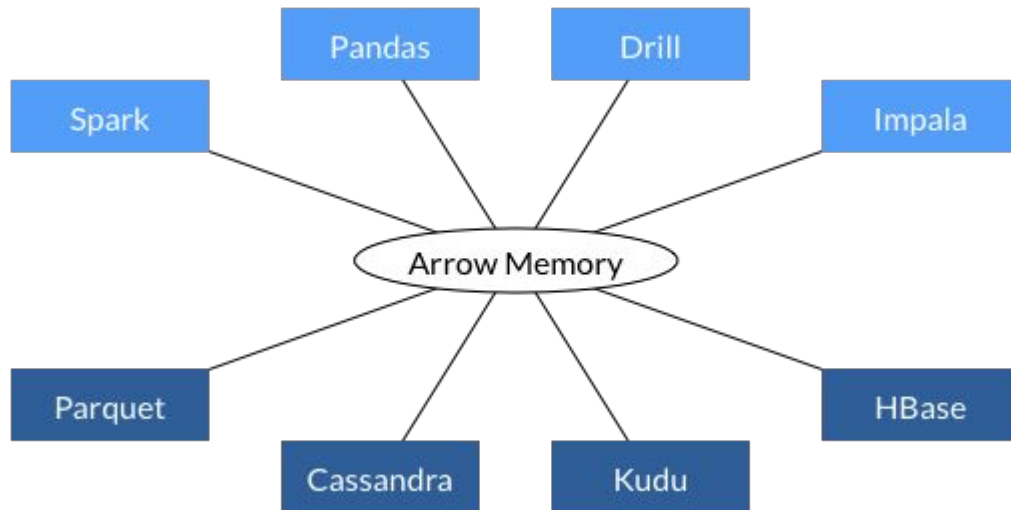


The initial idea



<https://xkcd.com/927/>

The idea becomes reality

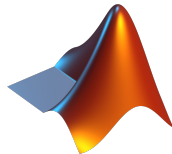


What is Arrow?

Apache Arrow is a multi-language toolbox for building high performance applications that process and transport large data sets. It is designed to both improve the performance of analytical algorithms and the efficiency of moving data from one system or programming language to another.

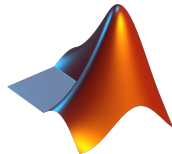
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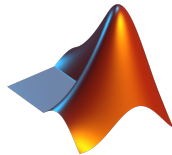
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Columnar vs Row

Age	Name	subscribed	spent
26	Robert	true	33.5
32	Alberto	false	54.02
Null	Piotr	true	45.3
46	Marco	false	23.09

26	Robert	true	33.5	32	Alberto	false	54.02	Null	Piotr	true	45.3	46	Marco	false	23.09
----	--------	------	------	----	---------	-------	-------	------	-------	------	------	----	-------	-------	-------

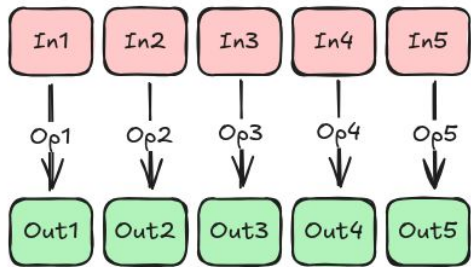
26	32	Null	46	Robert	Alberto	Piotr	Marco	true	false	true	false	33.5	54.02	45.3	23.09
----	----	------	----	--------	---------	-------	-------	------	-------	------	-------	------	-------	------	-------

Other advantages of columnar format

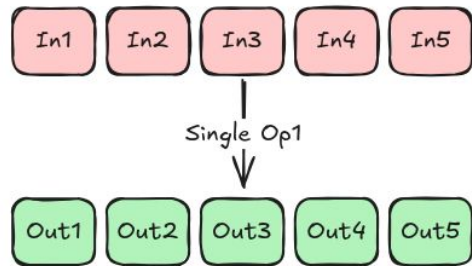
- SIMD Optimizations:
 - Allow us to perform an operation over multiple data
- Better Compression Algorithms



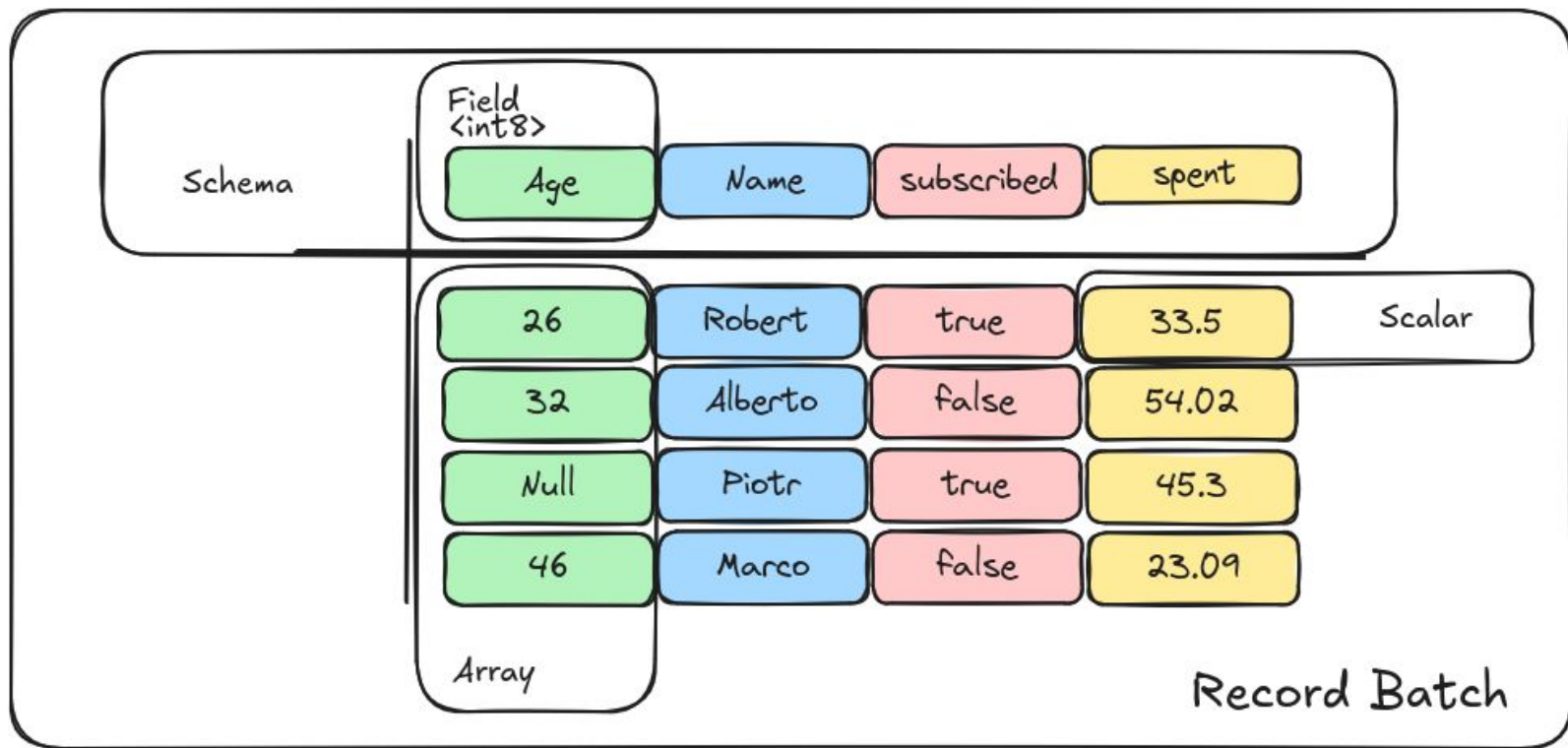
SISD (Single Instruction Single Data)



SIMD (Single Instruction Multiple Data)



Record batch and Array



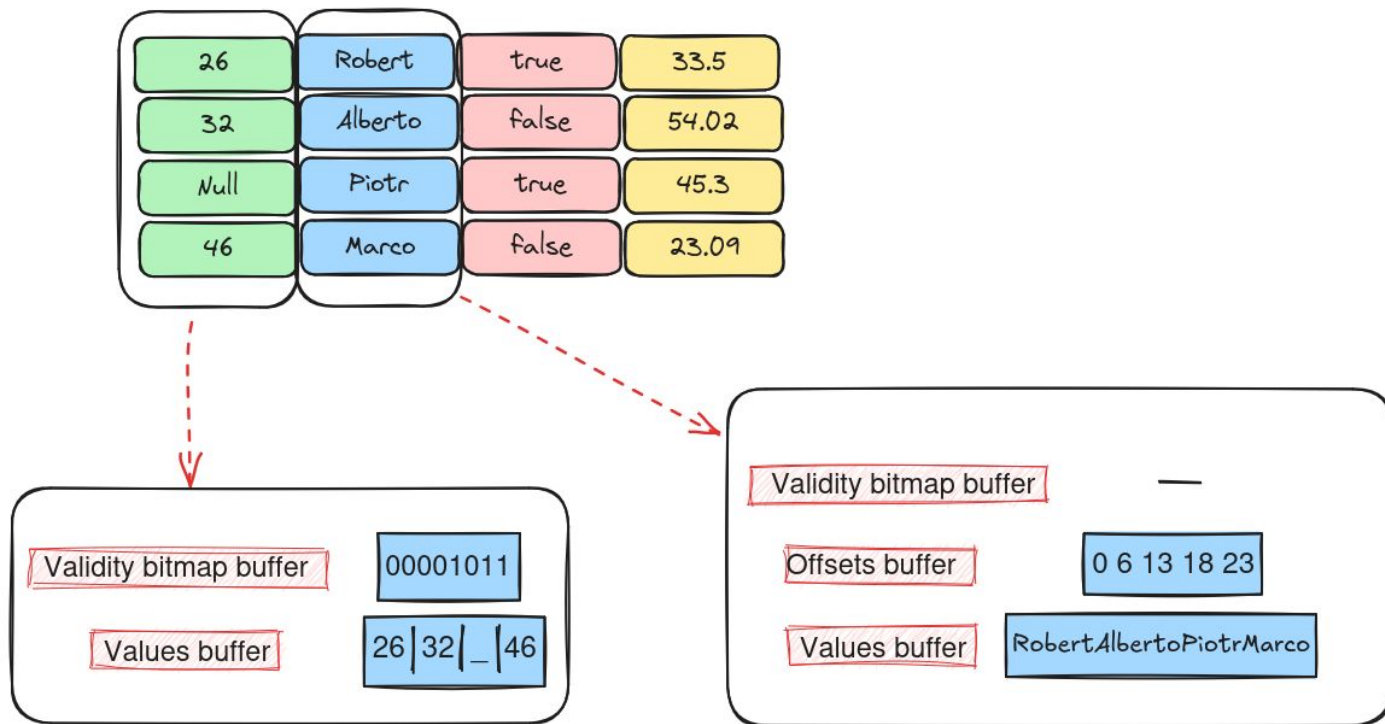
Record batch and Array

```
schema = pa.schema([pa.field('age', pa.int8()),
                      pa.field('name', pa.string()),
                      pa.field('subscribed', pa.bool_()),
                      pa.field('spent', pa.float16())])
a0 = pa.array([26, 32, None, 46], type=pa.int8())
a1 = pa.array(["Robert", "Alberto", "Piotr", "Marco"], type=pa.string())
a2 = pa.array([True, False, True, False], type=pa.bool_())
a3 = pa.array([33.5, 54.02, 45.3, 23.09], type=pa.float16())
batch = pa.record_batch([a0, a1, a2, a3], schema=schema)
```

Record batch and Array

```
>>> batch
pyarrow.RecordBatch
age: int8
name: string
subscribed: bool
spent: halffloat
-----
age: [26,32,null,46]
name: ["Robert","Alberto","Piotr","Marco"]
subscribed: [true,false,true,false]
spent: [33.5,54.03125,45.3125,23.09375]
```


Arrow format specification



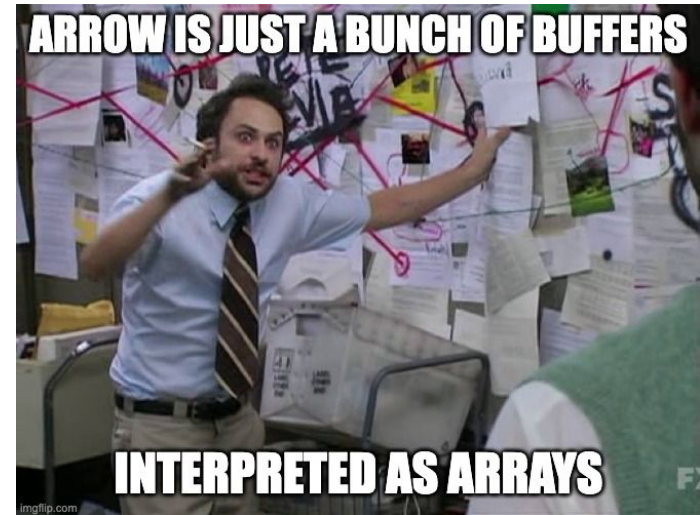
Array and buffers

```
>>> na.array(a0).inspect()  
<ArrowArray int8>  
- length: 4  
- offset: 0  
- null_count: 1  
- buffers[2]:  
  - validity <bool[1 b] 11010000>  
  - data <int8[4 b] 26 32 0 46>  
- dictionary: NULL  
- children[0]:
```

```
>>> na.array(a1).inspect()  
<ArrowArray string>  
- length: 4  
- offset: 0  
- null_count: 0  
- buffers[3]:  
  - validity <bool[0 b] >  
  - data_offset <int32[20 b] 0 6 13 18 23>  
  - data <string[23 b] b'RobertAlbertoPiotrMarco'>  
- dictionary: NULL  
- children[0]:
```

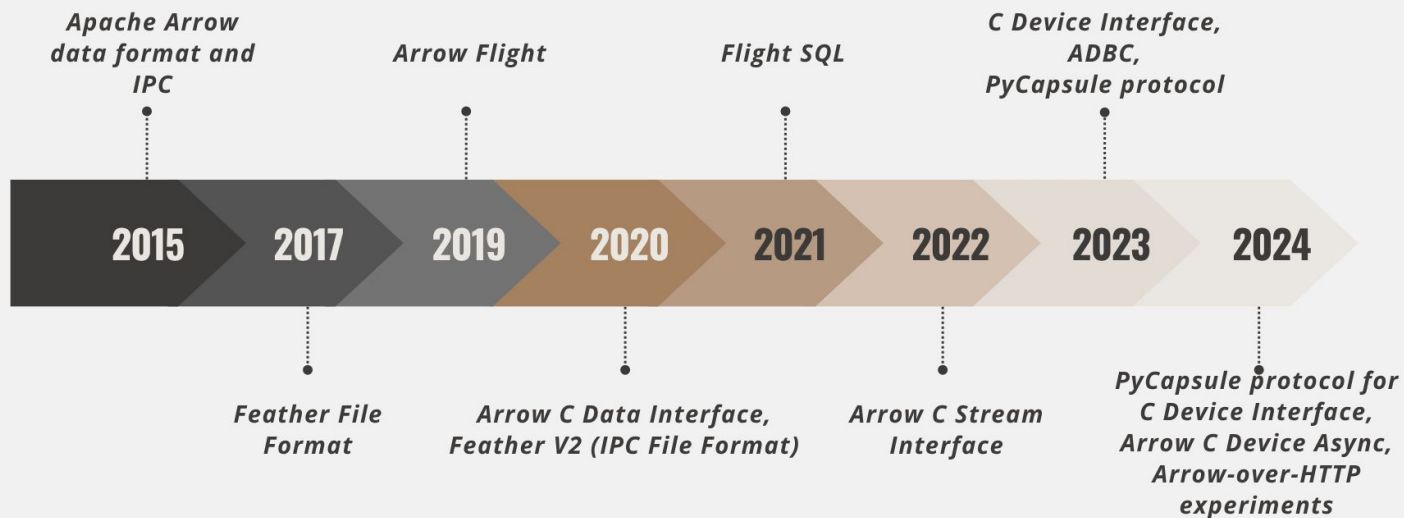
To sum up

- Arrow defines how arrays and tables look like in memory
- Arrow implementations provide a toolset to work with such columnar data structure



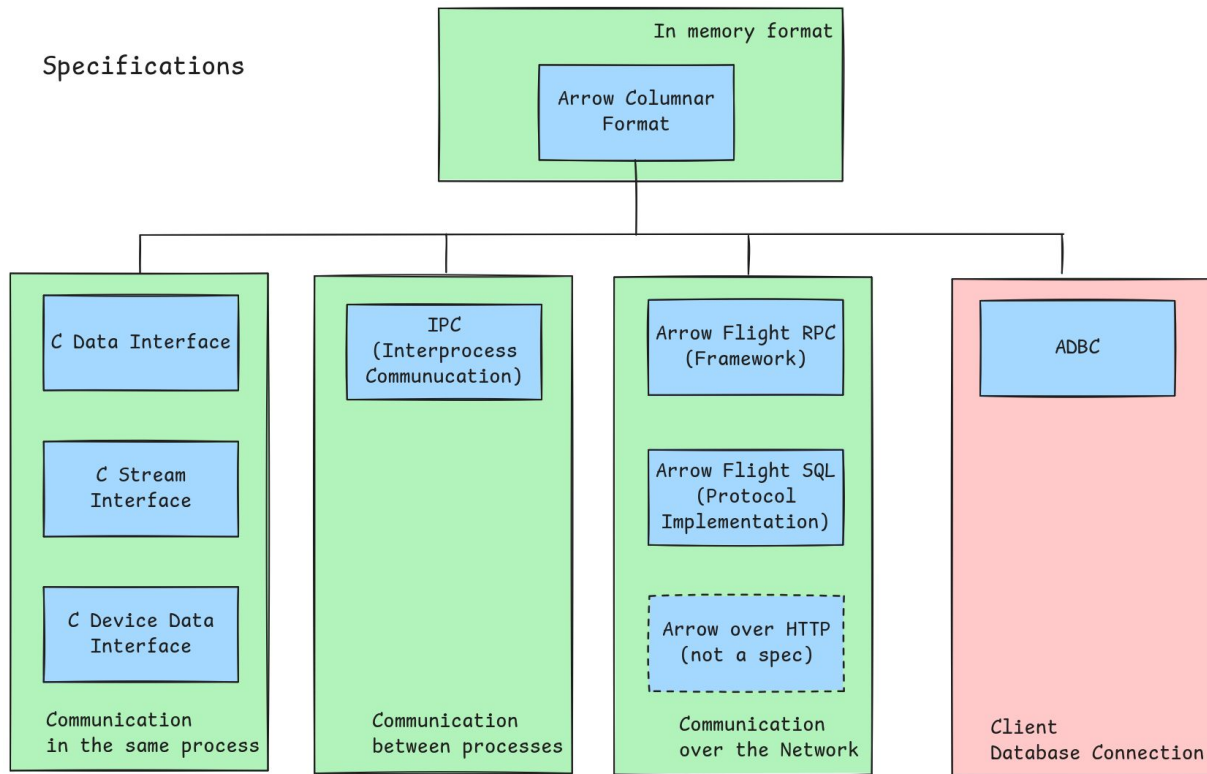
Arrow data Exchange

DATA EXCHANGE EVOLUTION



Evolution of the Arrow project around the data exchange

Overview



Serialization and Interprocess Communication (IPC)

IPC (Interprocess Communication)

A protocol for

- turning Arrow record batches into a one-way stream of binary messages,
- and then rebuilding them on the other side
- — all without copying data in memory.

```
<continuation: 0xFFFFFFFF>  
<metadata_size: int32>  
<metadata_flatbuffer: bytes>  
<padding>  
<message body>
```

<https://arrow.apache.org/docs/format/Columnar.html#serialization-and-interprocess-communication-ipc>

IPC (Interprocess Communication)

Arrow IPC comes in two flavors

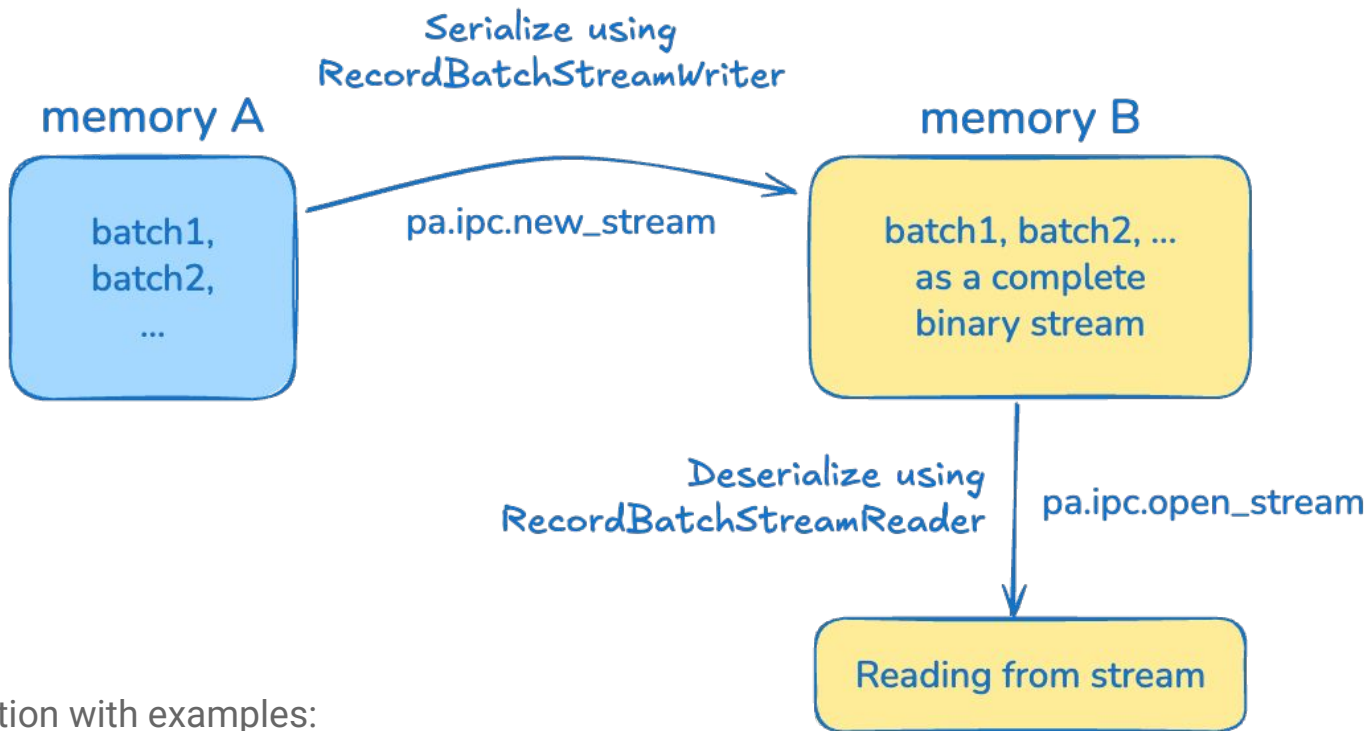
Stream format

- Sending an arbitrary length sequence of record batches
- For transferring data over a network

File format

- Serializing a fixed number of record batches
- Includes an index (footer) for random access

IPC Streaming Format

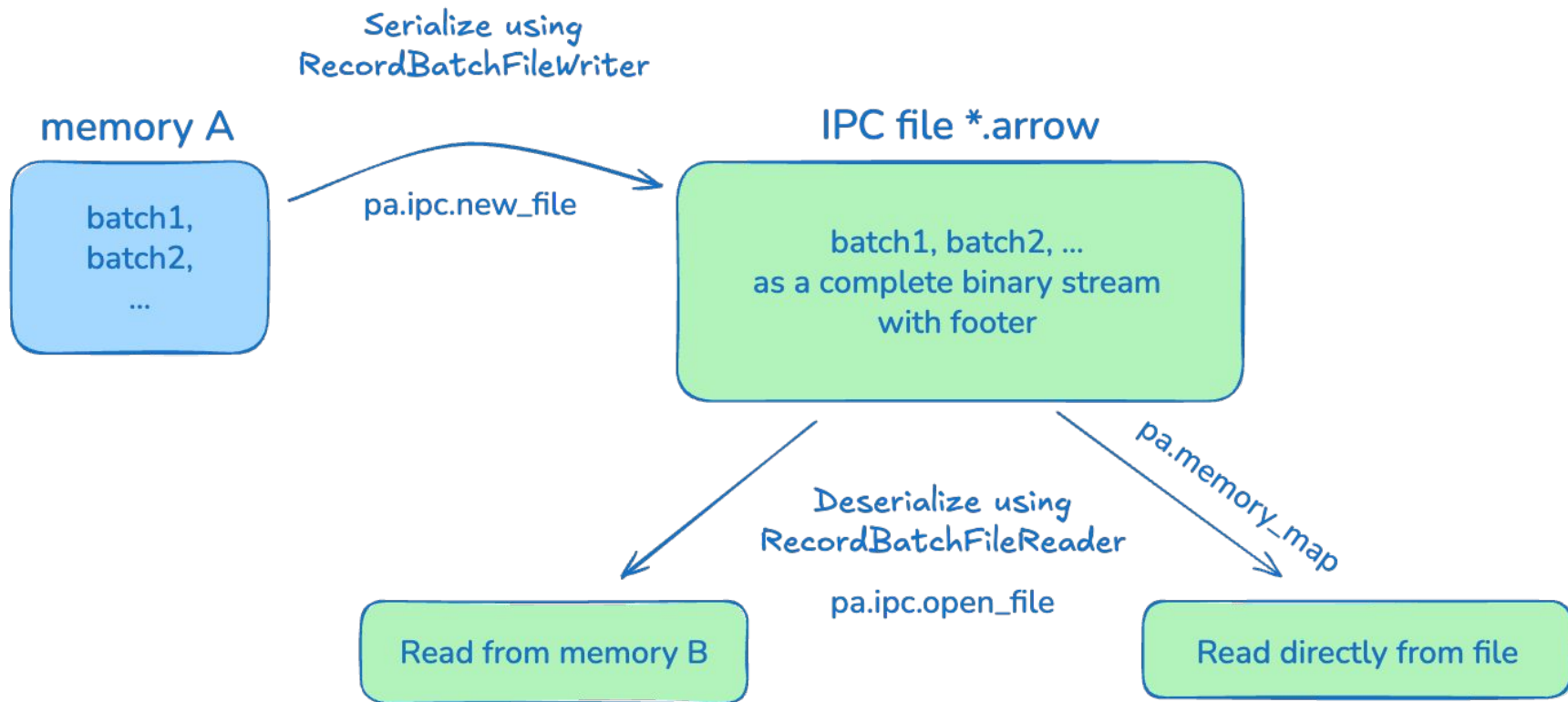


Documentation with examples:

<https://arrow.apache.org/docs/python/ipc.html>

no extra copy in memory!

IPC File Format



C Data Interface and Its Extensions: Stream and Device Interfaces

C Data Interface

- C ABI interface (C Application Binary Interface)
Interface to compiled code (“API for compiled code”)
- For zero-copy interchange of Arrow columnar data structures
- At runtime
- In the same process
- Without the need to link to Arrow libraries
- <https://arrow.apache.org/docs/format/CDataInterface.html>
- <https://arrow.apache.org/blog/2020/05/03/introducing-arrow-c-data-interface/>
- <https://willayd.com/leveraging-the-arrow-c-data-interface.html>

C Data Interface vs. IPC

<https://arrow.apache.org/docs/format/CDataInterface.html#comparison-with-the-arrow-ipc-format>

IPC (Interprocess Communication)	C Data Interface
across processes and machines	in-memory, inter-language
serialized data	zero-copy
binary stream format	C structs
needs Arrow IPC reader and writer	Only C ABI
Example: Arrow Flight	Example: Polars ↔ Pandas

Expanding C Data Interface

C Data Interface	Zero-copy sharing of Arrow columnar data in memory	Base layer <ul style="list-style-type: none">• C structs• ArrowArray, ArrowSchema
C Stream Interface	Share streams of data batches	Built on top of C Data Interface with ArrowArrayStream struct <ul style="list-style-type: none">• Data chunks, same schema• Blocking pull-style
C Device Interface (Device Stream, Async)	Extend to non-CPU memory	Expands C Data Interface <ul style="list-style-type: none">• to non-CPU memory

The Arrow PyCapsule Interface

Arrow PyCapsule Protocol

- C Data structs wrapped into a PyCapsule
- Capsules are a part of the Python C API
- Instead of returning raw integer pointers on export, PyCapsule is created using standardized “dunder” methods

Arrow PyCapsule Protocol

- C Data structs wrapped into a PyCapsule
- Capsules are a part of the Python C API
- Instead of returning raw integer pointers on export, PyCapsule is created using standardized “dunder” methods
- All C Data/C Device Interface benefits
- More robust
- No PyArrow dependence
- <https://arrow.apache.org/docs/format/CDataInterface/PyCapsuleInterface.html>
- <https://docs.python.org/3/c-api/capsule.html>

PyCapsule Protocol – example from PyArrow



```
>>> import pyarrow as pa
>>> arr = pa.array([17, 7, 2025])
>>> arr
<pyarrow.lib.Int64Array object at 0x117b5f760>
[
  17,
  7,
  2025
]
>>> pyarrow_add = arr.buffers()[1].address
>>> pyarrow_add
3312292397248
```

PyCapsule Protocol – example to Polars

```
>>> polars_series = pl.Series(arr)
>>> polars_series
shape: (3,)
Series: '' [i64]
[
      17
       7
     2025
]
>>> polars_add = polars_series._get_buffer_info()[0]
>>> polars_add
3312292397248
>>> polars_add == pyarrow_add
True
```

PyCapsule Protocol – example to Pandas

```
>>> import pandas as pd
>>> pandas_series = pd.Series(polars_series)
>>> pandas_series
0      17
1       7
2    2025
dtype: int64
>>> pandas_add = pandas_series.values.ctypes.data
>>> pandas_add
3312292397248
>>> pandas_add == polars_add == pyarrow_add
True
```

PyCapsule Protocol – also to ...

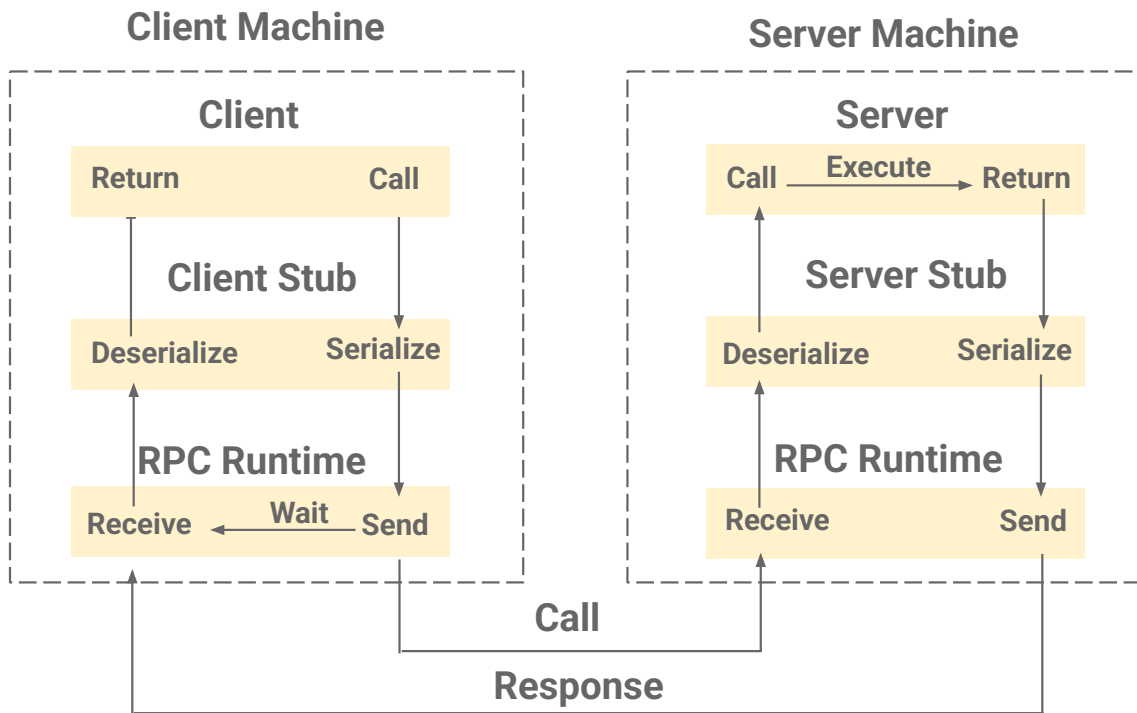
... Ibis, arro3, GDAL, narwhals, quak, DataFusion, DuckDB, GeoPandas, cuDF, ...

<https://github.com/apache/arrow/issues/39195#issuecomment-2245718008>

Flight RPC

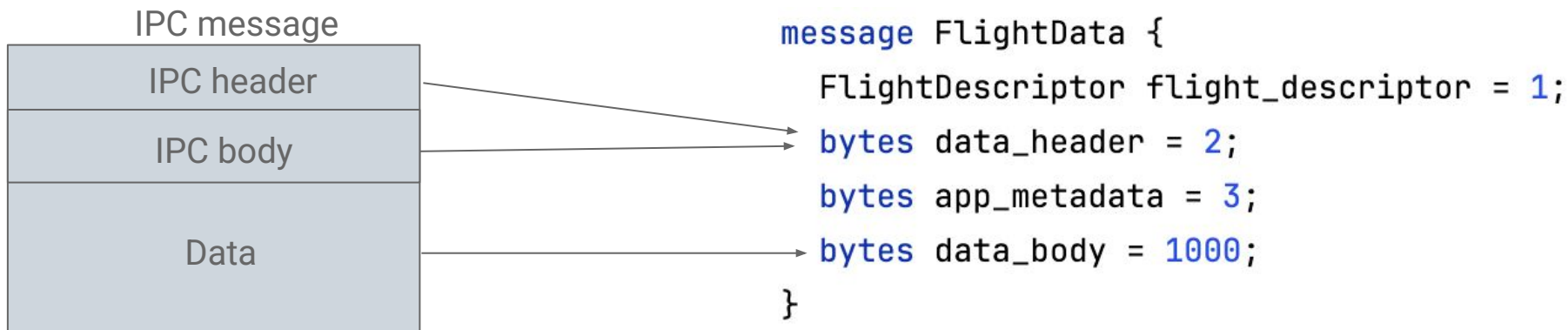
Remote procedure call (RPC) frameworks

- Frameworks for distributed computing that allow executing routines on other machines
- Most RPCs have a language to describe interfaces (stubs) - think of it as REST where you can define your own methods
- Note that network traffic is here needs to be serialized for sending



Flight RPC – what is flight

- What is it
 - RPC framework for high-performance data services
 - IPC format “serialized” with protobuf and sent over gRPC
 - Designed for zero-copy serialization and parallel data transfer

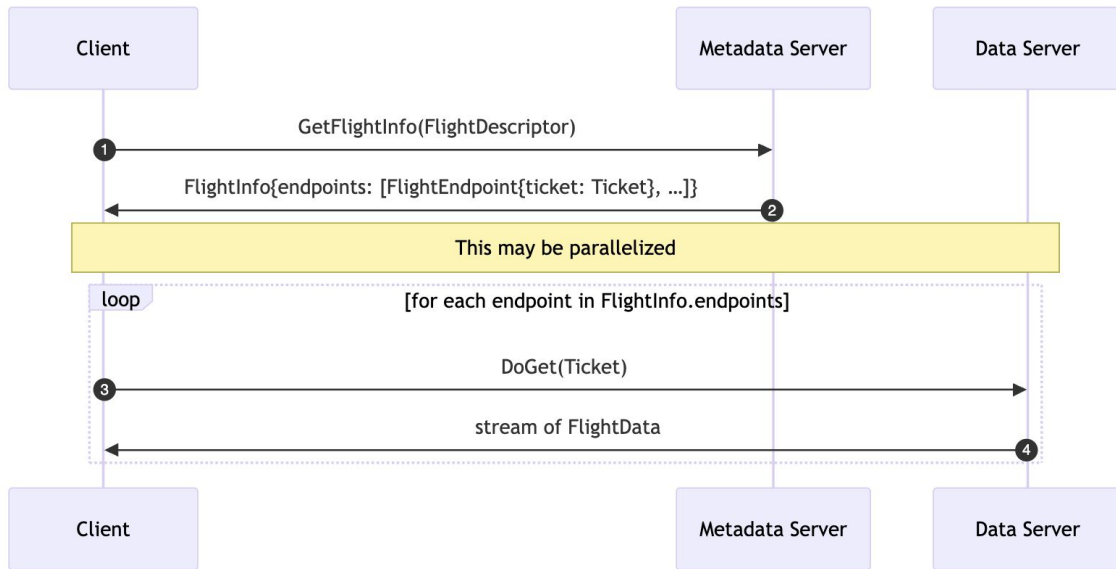


Flight RPC – when to use it

- When does it make sense to use it
 - Server and client use Arrow layout data
 - Moving large batches over the network
- When does it not make sense to use it
 - Row-oriented data
 - Small volume sent which can't be batched

Flight RPC – how it works

- Server implements a set of RPC endpoints, e.g. `GetFlightInfo`
- `FlightInfo` message includes schema and endpoints where data can be retrieved from



Retrieving data via `DoGet`.

Arrow over HTTP

See <https://github.com/apache/arrow-experiments/tree/main/http>

Arrow over HTTP – client

To receive record batches as a client:

- Sends an HTTP GET request to a server.
- Receives an HTTP 200 response from the server, with the response body containing an Arrow IPC stream of record batches.
- Adds the record batches to a list as they are received.

```
import urllib.request
import pyarrow as pa

url = 'http://localhost:8008'
response = urllib.request.urlopen(url)

batches = []

with pa.ipc.open_stream(response) as reader:
    schema = reader.schema
    batches = [b for b in reader]
```

Arrow over HTTP – server

To send record batches from server:

- Serialize data into IPC stream of record batches with IPC stream writer
- Upon receiving a request, sends an HTTP 200 response with the body containing an IPC stream

```
def generate_bytes(schema, batches):  
    with pa.RecordBatchReader.from_batches(schema, batches) as source, \  
        io.BytesIO() as sink, \  
        pa.ipc.new_stream(sink, schema) as writer:  
        for batch in source:  
            sink.seek(0)  
            writer.write_batch(batch)  
            sink.truncate()  
            with sink.getbuffer() as buffer:  
                yield buffer
```

ADBC

Arrow Database Connectivity

<https://arrow.apache.org/adbc/>

ADBC – what is ADBC

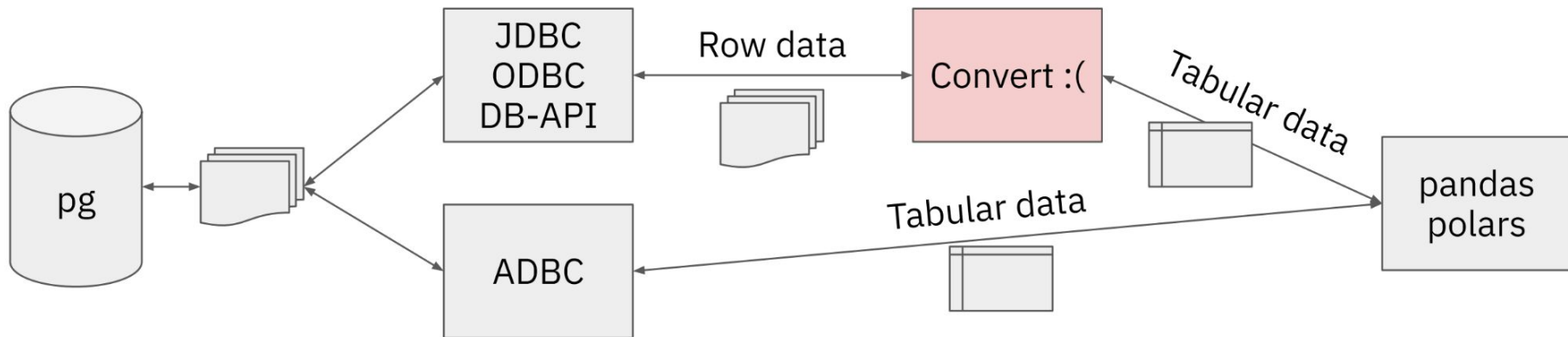
At a high level, ADBC is the standard for Arrow-native database access. It allows for executing SQL queries while working with Arrow data.

At a lower level, ADBC is two separate but related things:

- An abstract API for interacting with databases and Arrow data.
- A set of concrete implementations of that abstract API in different languages and drivers for different databases

ADBC – why use ADBC

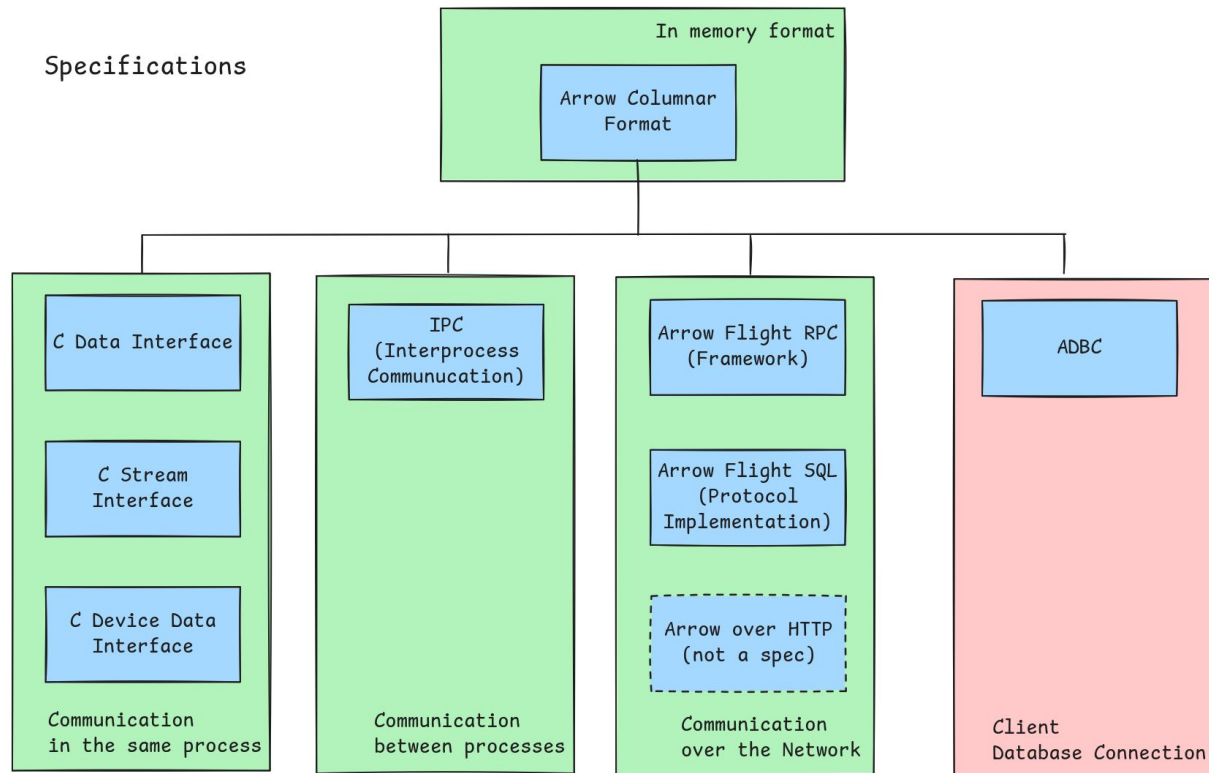
- Using ADBC driver provides columnar Arrow data to the consumer
- Eliminates needs for data copies - transport layer is IPC
- Eliminates row to columnar data conversion



ADBC – current support

Driver	Supported Languages	Implementation Language	Status
Apache DataFusion	Rust	Rust	Experimental
BigQuery (C#)	C#	C#	Experimental
BigQuery (Go)	C/C++	Go	Experimental
DuckDB [1]	C/C++	C++	Stable
Flight SQL (Go)	C/C++, C# [3]	Go	Stable
Flight SQL (Java)	Java	Java	Experimental
JDBC Adapter	Java	Java	Experimental
PostgreSQL	C/C++	C++	Stable
SQLite	C/C++	C	Stable
Snowflake	C/C++, Rust [3]	Go	Stable
Thrift protocol-based [2]	C#	C#	Experimental

Summary



↓ Slides ↓

