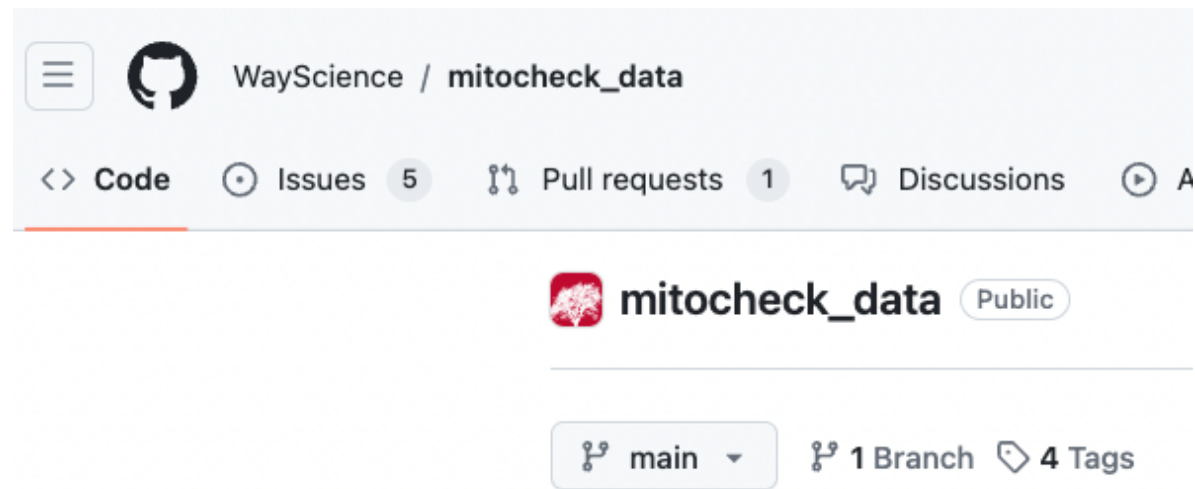


# Images and Databases

Way Lab - Research in Progress - 2024-06-07

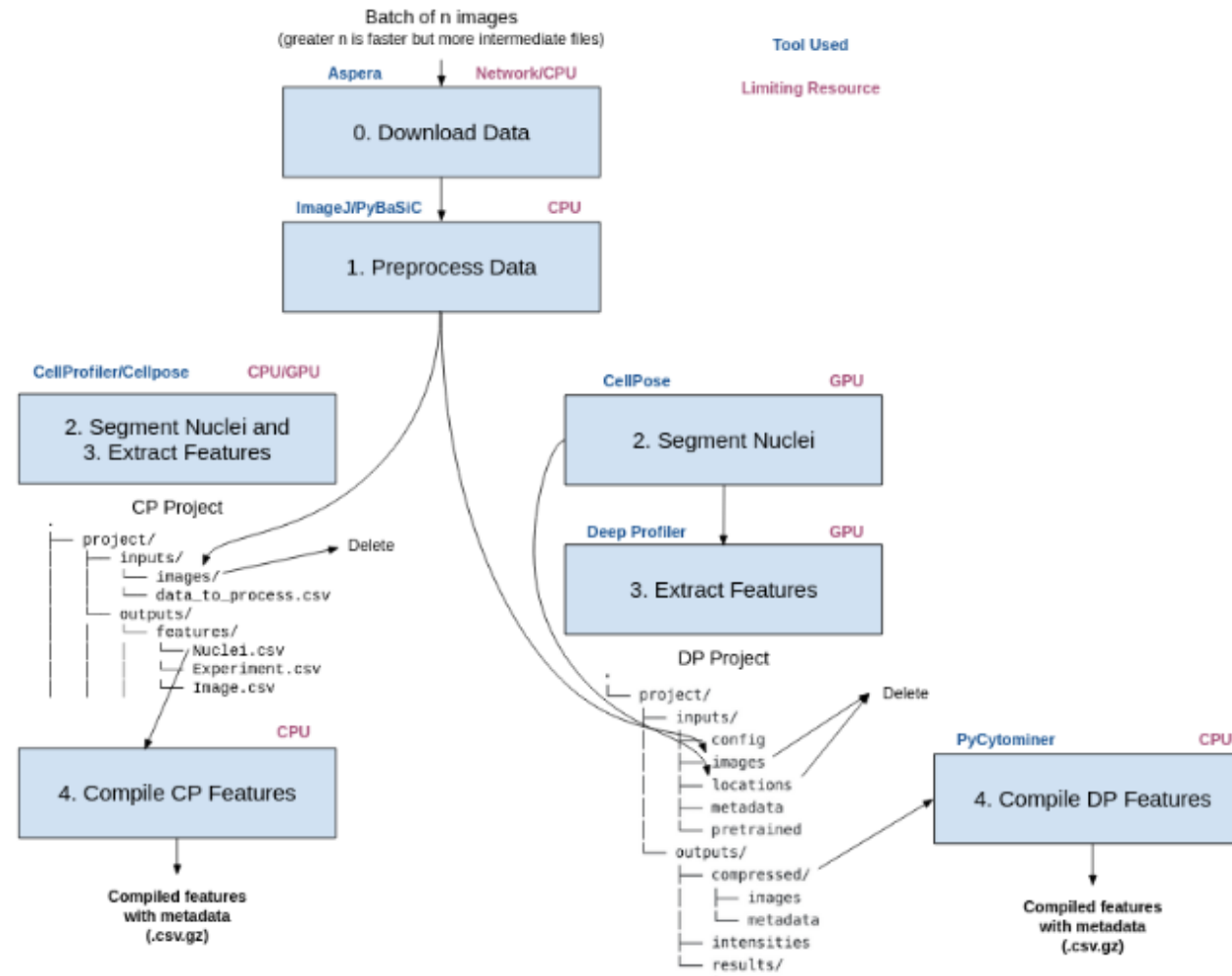


## Background



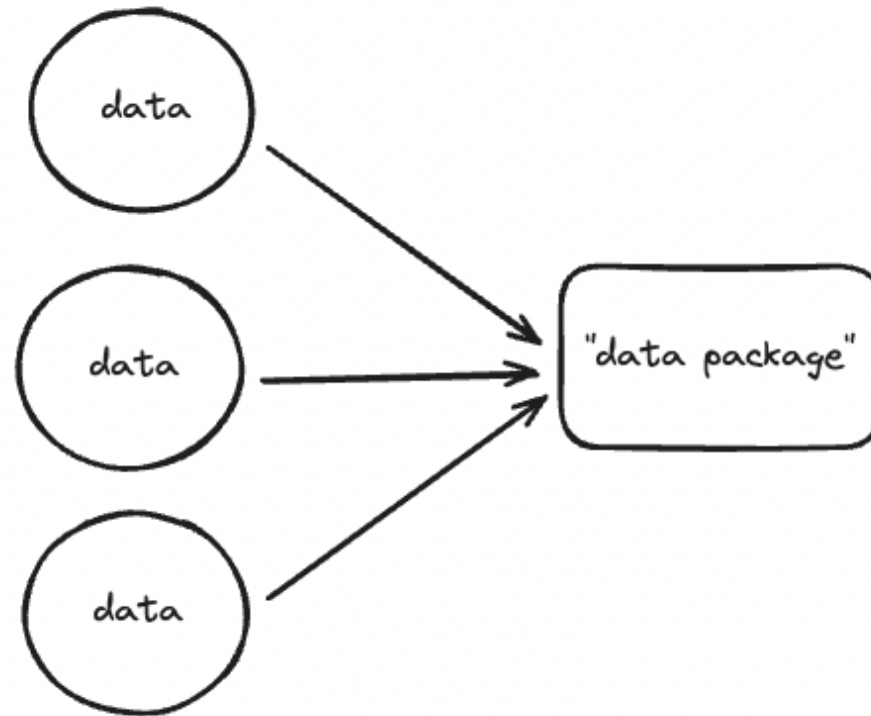
- `mitochek_data` project

# Background



- `idr_stream` project

## Background



How could we "package" the `mitocheck_data` data in such a way to enable development iteration and usefulness to others?

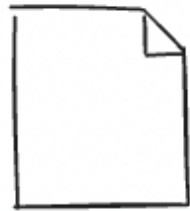
## Background

"Data Packaging" Story

*"As a research data participant I need a way to analyze (understand, contextualize, and explore) and implement (engineer solutions which efficiently scale for time and computing resources) the data found here in order to effectively reproduce findings, make new discoveries, and avoid challenging (or perhaps incorrect) translations individually."*



# Data Files



.csv  
.tsv  
.txt  
(+/- .gz)



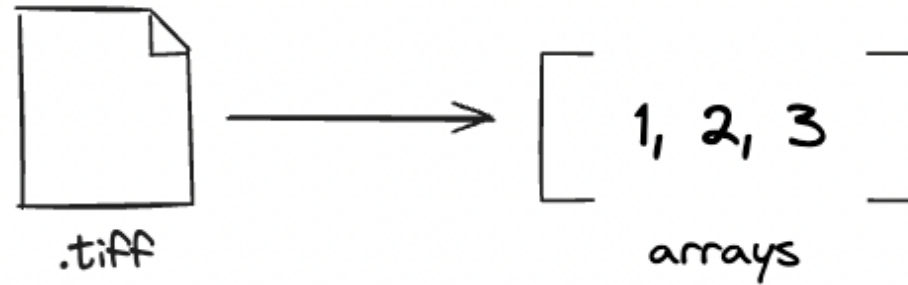
.ch5



.tiff

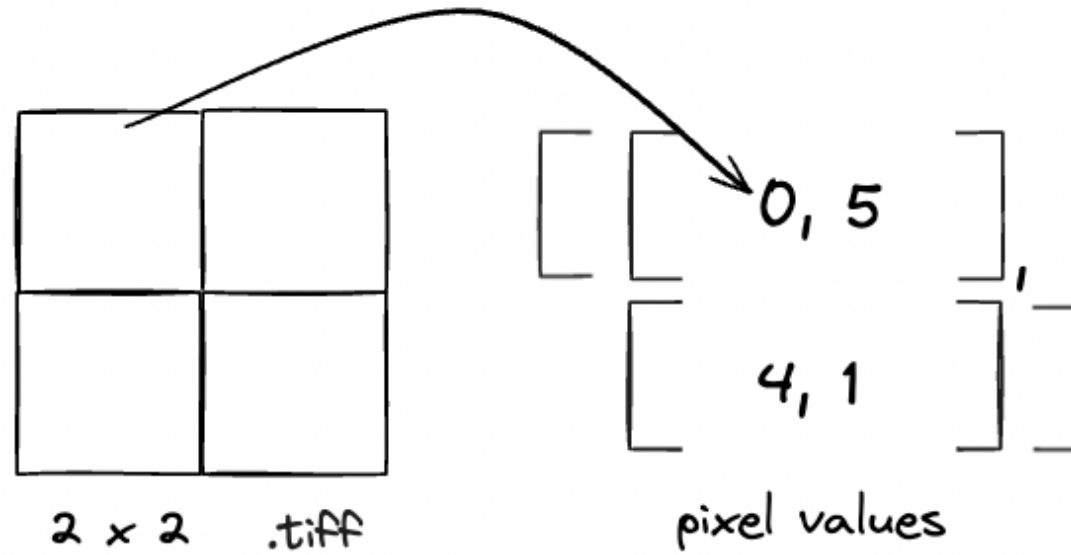
- text file data ([CSV](#)'s, TSV's, etc)
- ch5 files (microscopy-focused [HDF](#))
- tiff files ([tagged image file format](#))

## Implied Data



- arrays from images (for in-memory calculations)

## Implied Data





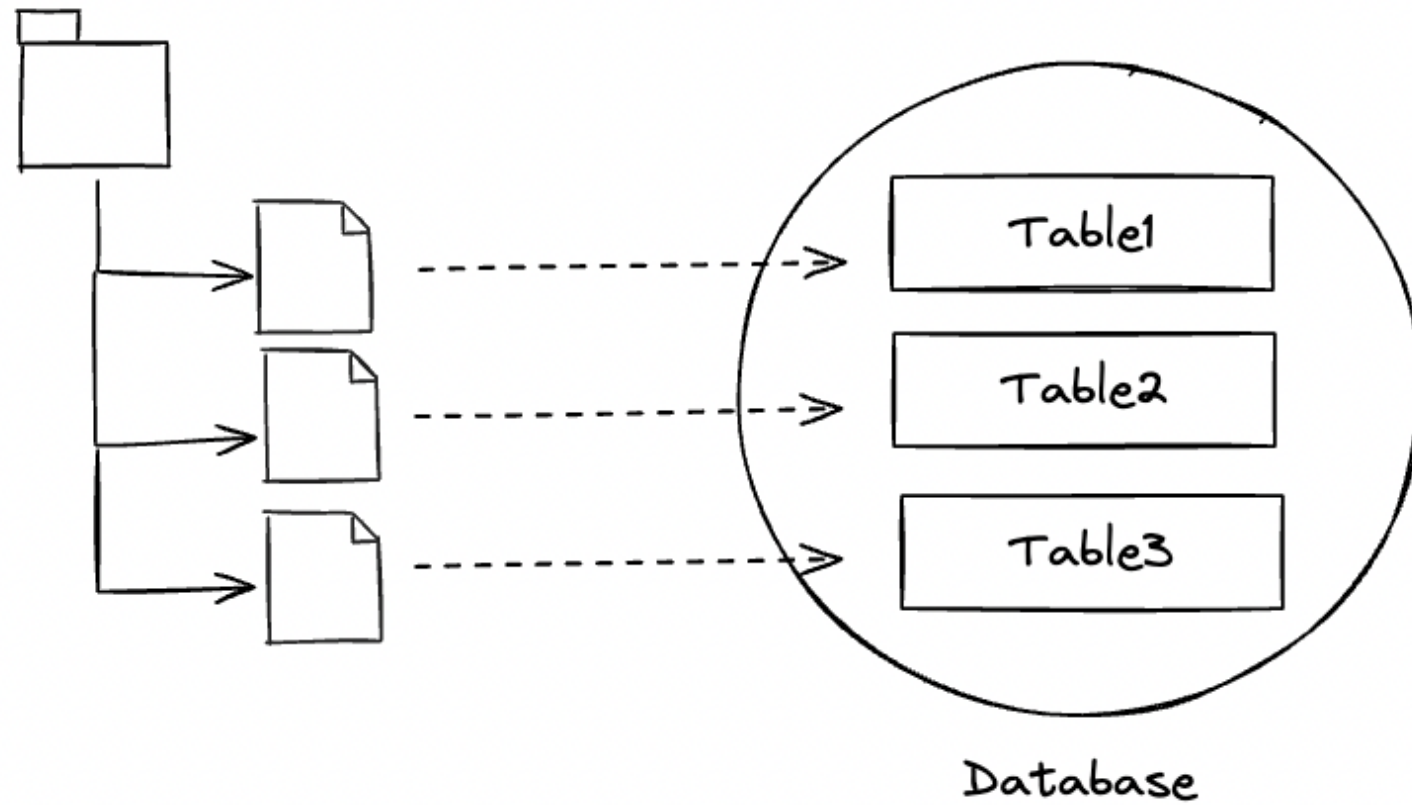
# Data package needs

Okay, so we need something that can store and distribute:

- text file tables
- images
- arrays



# Files and databases



Files and directories follow patterns within databases.

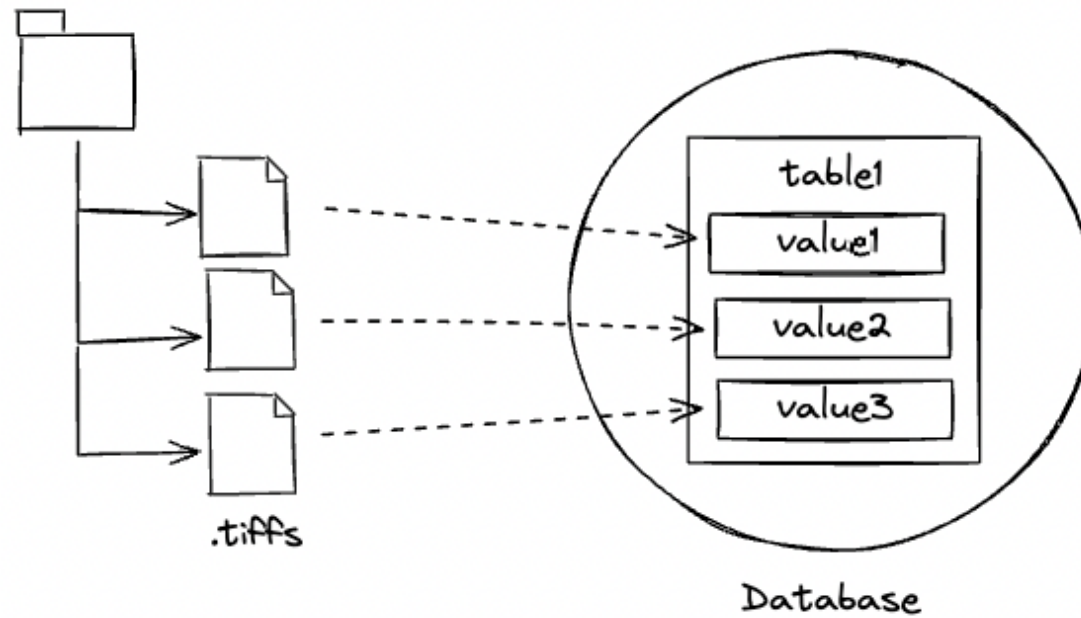


# Images and databases

Might be thinking: but images aren't tables!

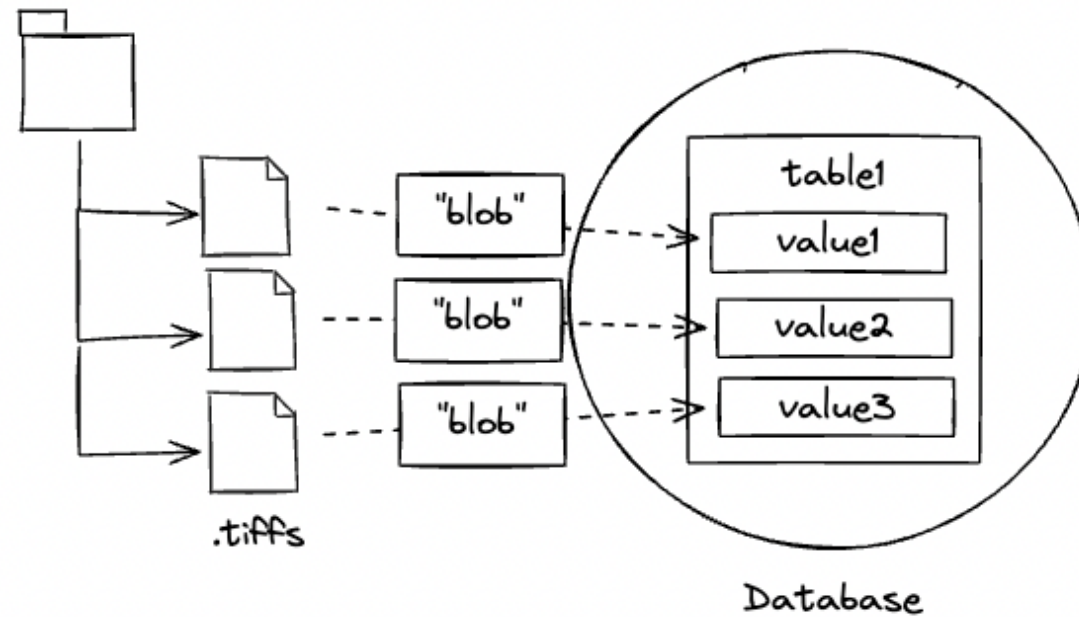


# Images and databases



- Images as *values* within a table.
- The dimensionality is determined by the file.

# Images and databases



- When we talk about images this way we can call them "BLOB's" or "objects (object storage)".

# Data package needs

Okay, so we need something that can store and distribute:

- files:
  - text file tables
  - images (blobs / objects)
  - arrays
- dimensions:
  - multiple tables
  - multiple values (and dimensions) within tables



# LanceDB

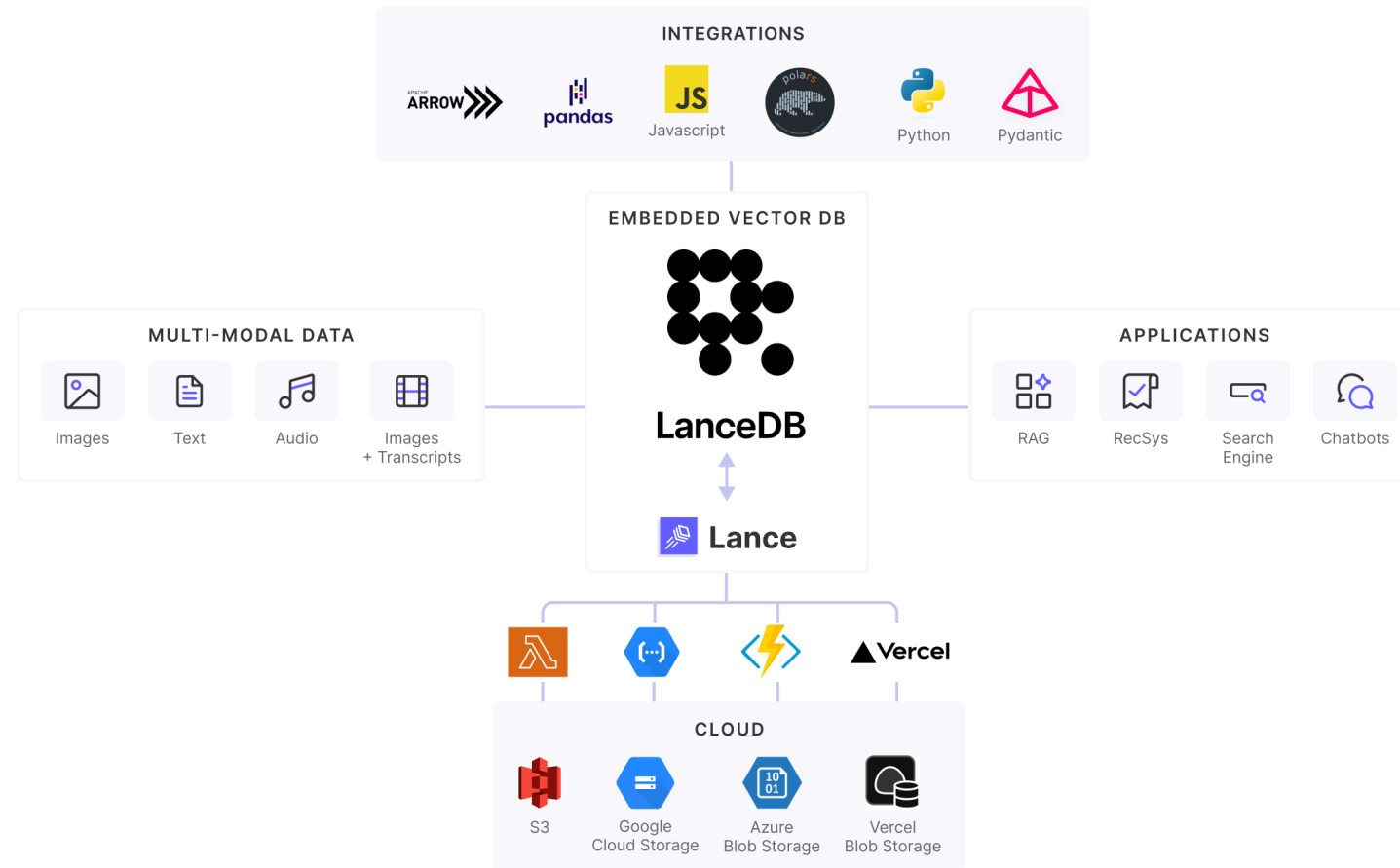


*LanceDB is an open-source vector database for AI that's designed to store, manage, query and retrieve embeddings on large-scale multi-modal data. The core of LanceDB is written in Rust 🦀 and is built on top of Lance, an open-source columnar data format designed for performant ML workloads and fast random access.*

- Source: <https://lancedb.github.io/lancedb/>



# LanceDB

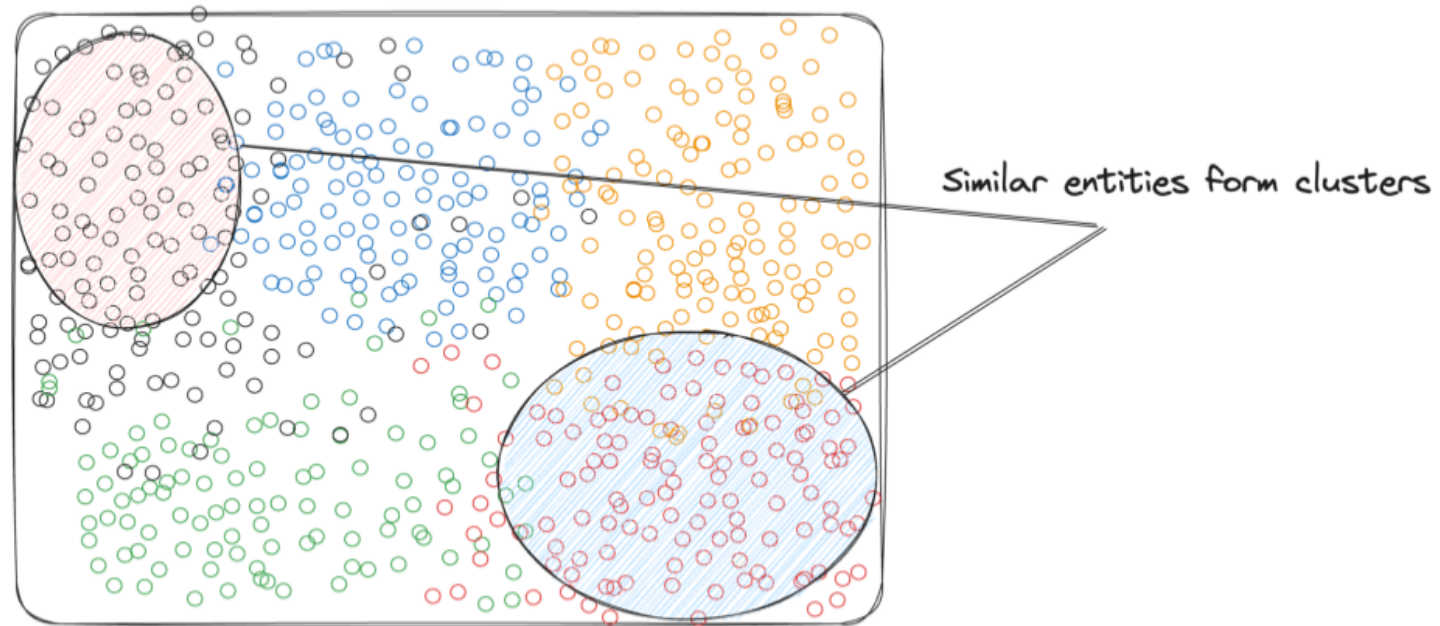


- Source: <https://lancedb.github.io/lancedb/>





# LanceDB



- LanceDB is purpose-built with embeddings and vector search in mind.
- Source: [https://lancedb.github.io/lancedb/concepts/vector\\_search/](https://lancedb.github.io/lancedb/concepts/vector_search/)

In [5]: **import** pathlib

```
# images from:  
# mitocheck_data: https://github.com/WayScience/mitocheck_data  
# Image Data Resource (IDR): idr0013(screenA)  
  
# show some images in an image dir  
image_dir = "mitocheck_example_images"  
  
# create a list of images using glob on the dir  
images = list(pathlib.Path(image_dir).glob("*"))  
  
images
```

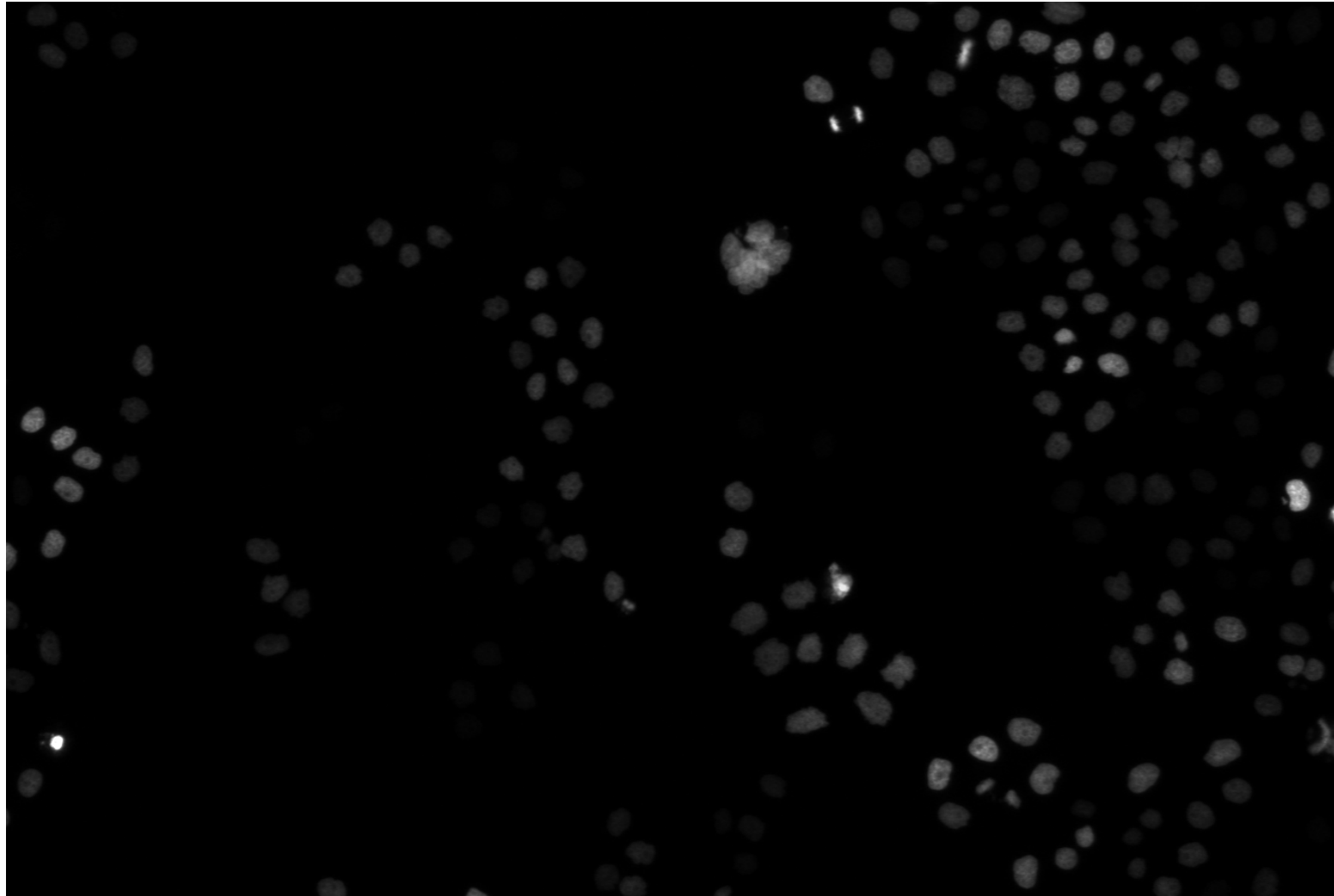
Out [5]:

```
[PosixPath('mitocheck_example_images/LT0001_02.LT0001_02_26_46  
_IC.tif'),  
 PosixPath('mitocheck_example_images/LT0001_02.LT0001_02_15_43  
_IC.tif')]
```



```
In [6]: from IPython.display import display
        from PIL import Image

        # display the image
        display(Image.open(images[0]))
```



```
In [7]: from IPython.display import display
        from PIL import Image

        # display the image
        display(Image.open(images[1]))
```



```
In [8]: # file size in bytes
print(images[0])
print(images[0].stat().st_size)
```

```
mitocheck_example_images/LT0001_02.LT0001_02_26_46_IC.tif
1376512
```



```
In [9]: # show first few bytes as byte string
with open(images[0], "rb") as f:
    print(f.read(10))
```

```
b'II*\x00\x08\x00\x00\x00\x0e\x00'
```

```
In [10]: # show some metadata associated with the image
!tiffinfo mitocheck_example_images/LT0001_02.LT0001_02_15_43_IC.tif
```

```
Reading TIFF header: 0.000340 s
Reading image data: 0.000260 s
Generating report: 0.001179 s
```

```
TiffFile 'LT0001_02.LT0001_02_15_43_IC.tif' 1344.25 KiB shape
d
```

```
TiffPageSeries 0 1024x1344 uint8 YX shaped 1 Pages @256
```

```
TiffPage 0 @8 1024x1344 uint8 minisblack memmappable shaped
```

```
TiffTag 256 ImageWidth @10 LONG @18 = 1344
TiffTag 257 ImageLength @22 LONG @30 = 1024
TiffTag 258 BitsPerSample @34 SHORT @42 = 8
TiffTag 259 Compression @46 SHORT @54 = NONE
TiffTag 262 PhotometricInterpretation @58 SHORT @66 = MINISBLAC
K
TiffTag 270 ImageDescription @70 ASCII[24] @182 = {"shape": [10
24, 1344]}
TiffTag 273 StripOffsets @82 LONG @90 = (256,)
TiffTag 277 SamplesPerPixel @94 SHORT @102 = 1
TiffTag 278 RowsPerStrip @106 LONG @114 = 1024
TiffTag 279 StripByteCounts @118 LONG @126 = (1376256,)
TiffTag 282 XResolution @130 RATIONAL @222 = (1, 1)
TiffTag 283 YResolution @142 RATIONAL @230 = (1, 1)
TiffTag 296 ResolutionUnit @154 SHORT @162 = NONE
TiffTag 305 Software @166 ASCII[12] @238 = tiffinfo.py
```



```
In [11]: import skimage

# read the image as an array
array = skimage.io.imread(images[0])
print(array.shape)
print(array)
```

```
(1024, 1344)
[[0 0 0 ... 0 0 0]
 [0 0 0 ... 0 0 0]
 [0 0 0 ... 0 0 0]
 ...
 [0 0 0 ... 0 0 0]
 [0 0 1 ... 0 0 0]
 [0 0 0 ... 0 0 0]]
```





```
In [12]: import pandas as pd

# create a pandas dataframe with image paths
df = pd.DataFrame({"path": images})

df
```

```
Out[12]:
```

	path
0	mitocheck_example_images/LT0001_02.LT0001_02_2...
1	mitocheck_example_images/LT0001_02.LT0001_02_1...

```
In [13]: # add the filesize bytes as a column to the dataframe
df["filesize_bytes"] = df.apply(lambda row: row["path"].stat().st_size,
df
```

```
Out [13]:
```

	path	filesize_bytes
0	mitocheck_example_images/LT0001_02.LT0001_02_2...	1376512
1	mitocheck_example_images/LT0001_02.LT0001_02_1...	1376512

```
In [14]: def read_image_bytes(image_path):  
         with open(image_path, "rb") as f:  
             return f.read()  
  
         # read the image as a bytearray object and store in dataframe  
         df["image_bytes"] = df.apply(lambda row: read_image_bytes(row["path"]),  
                                     axis=1)  
         df
```

```
Out [14]:
```

	path	filesize_bytes	
0	mitocheck_example_images/LT0001_02.LT0001_02_2...	1376512	b'II*\x00\x00
1	mitocheck_example_images/LT0001_02.LT0001_02_1...	1376512	b'II*\x00\x00

```
In [15]: from io import BytesIO



import skimage

# read the bytearray from the dataframe as an array in new column
df["image_array"] = df.apply(
    lambda row: skimage.io.imread(BytesIO(row["image_bytes"])), axis=1
)
df
```

```
Out [15]:
```

	path	filesize_bytes	
0	mitocheck_example_images/LT0001_02.LT0001_02_2...	1376512	b'II*\x00\x00
1	mitocheck_example_images/LT0001_02.LT0001_02_1...	1376512	b'II*\x00\x00

```
In [17]: # show images within the dataframe output
ImageDataFrame(df[["path", "filesize_bytes", "image_bytes"]])
```

	path	filesize_bytes	image
0	mitocheck_example_images/LT0001_02.LT0001_02_26_46_IC.tif	1376512	
1	mitocheck_example_images/LT0001_02.LT0001_02_15_43_IC.tif	1376512	

```
In [18]: # try to write to parquet
df.to_parquet("mitocheck_example_images.parquet")
```

```
-----
-----
ArrowInvalid                                Traceback (most recent
call last)
Cell In[18], line 1
----> 1 df.to_parquet("mitocheck_example_images.parquet")
```

```
File ~/Library/Caches/pypoetry/virtualenvs/2024-06-07-images-and-databases-jUUoCL3p-py3.11/lib/python3.11/site-packages/pandas/util/_decorators.py:333, in deprecate_nonkeyword_arguments.<locals>.decorate.<locals>.wrapper(*args, **kwargs)
```

```
    327 if len(args) > num_allow_args:
    328     warnings.warn(
    329         msg.format(arguments=_format_argument_list(allow_args)),
    330         FutureWarning,
    331         stacklevel=find_stack_level(),
    332     )
--> 333 return func(*args, **kwargs)
```

```
File ~/Library/Caches/pypoetry/virtualenvs/2024-06-07-images-and-databases-jUUoCL3p-py3.11/lib/python3.11/site-packages/pandas/core/frame.py:3113, in DataFrame.to_parquet(self, path, engine, compression, index, partition_cols, storage_options, **kwargs)
```

```
    3032 """
```

```
    3033 Write a DataFrame to the binary parquet format
```



```
In [19]: # show columns and types
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2 entries, 0 to 1
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   path            2 non-null     object
1   filesize_bytes  2 non-null     int64
2   image_bytes     2 non-null     object
3   image_array     2 non-null     object
dtypes: int64(1), object(3)
memory usage: 196.0+ bytes
```

```
In [20]: # show the type of a single path value  
         type(df["path"].iloc[0])
```

```
Out[20]: pathlib.PosixPath
```



```
In [21]: import pyarrow as pa
```

```
# update the paths to be strings, then try conversion again
df["path"] = df.apply(lambda row: str(row["path"]), axis=1)
pa.Table.from_pandas(df)
```

```
-----
ArrowInvalid                                Traceback (most recent call last)
```

```
Cell In[21], line 5
```

```
      3 # update the paths to be strings, then try conversion again
```

```
      4 df["path"] = df.apply(lambda row: str(row["path"]), axis=1)
```

```
----> 5 pa.Table.from_pandas(df)
```

```
File ~/Library/Caches/pypoetry/virtualenvs/2024-06-07-images-and-databases-jUUoCL3p-py3.11/lib/python3.11/site-packages/pyarrow/table.pxi:3874, in pyarrow.lib.Table.from_pandas()
```

```
File ~/Library/Caches/pypoetry/virtualenvs/2024-06-07-images-and-databases-jUUoCL3p-py3.11/lib/python3.11/site-packages/pyarrow/pandas_compat.py:611, in dataframe_to_arrays(df, schema, preserve_index, nthreads, columns, safe)
    606         return (isinstance(arr, np.ndarray) and
    607                 arr.flags.contiguous and
    608                 issubclass(arr.dtype.type, np.integer))
    610 if nthreads == 1:
```

```
    606         return (isinstance(arr, np.ndarray) and
```

```
    607                 arr.flags.contiguous and
```

```
    608                 issubclass(arr.dtype.type, np.integer))
```

```
    610 if nthreads == 1:
```

```
--> 611         arrays = [convert_column(c, f)
```



```
In [22]: # show the type of a single image_array value  
type(df["image_array"].iloc[0])
```

```
Out[22]: numpy.ndarray
```

In [23]: **import** awkward **as** ak

```
# use awkward array to interpret the nested arrays from dict records
awk_arr = ak.Array(df.to_dict(orient="records"))
awk_arr
```

Out[23]: [{path: 'mitocheck\_example\_images/LT0001\_02.LT0001\_02\_26\_46\_IC.tif', ...},  
 {path: 'mitocheck\_example\_images/LT0001\_02.LT0001\_02\_15\_43\_IC.tif', ...}]

```
-----
-----
type: 2 * {
  path: string,
  filesize_bytes: int64,
  image_bytes: bytes,
  image_array: var * var * int64
}
```





```
In [24]: from pyarrow import parquet

# write a parquet table from the awkward array
parquet.write_table(
    table=ak.to_arrow_table(awk_arr), where="mitocheck_example_images.p
)
# show that we have a file
pathlib.Path("mitocheck_example_images.parquet").is_file()
```

```
Out[24]: True
```



```
In [25]: # read the file as a dataframe
df = pd.read_parquet(path="mitocheck_example_images.parquet")
# show that it's the same
ImageDataFrame(df[["path", "filesize_bytes", "image_bytes"]])
```

	path	filesize_bytes	image
0	mitocheck_example_images/LT0001_02.LT0001_02_26_46_IC.tif	1376512	
1	mitocheck_example_images/LT0001_02.LT0001_02_15_43_IC.tif	1376512	

```
In [26]: import shutil

import lancedb
from pyarrow import parquet

# remove any earlier work
shutil.rmtree("mitocheck_example_images.lance")

# specify a dir where the lancedb database may go and create lancedb cl
lancedb_dir = pathlib.Path("mitocheck_example_images.lance")
ldb = lancedb.connect(lancedb_dir)

# create a lancedb table from the parquet data
ldb.create_table(
    data=parquet.read_table("mitocheck_example_images.parquet"),
    name="mitocheck_example_images",
)
```

```
/Users/dabu5788/Library/Caches/pypoetry/virtualenvs/2024-06-07-
images-and-databases-jUUoCL3p-py3.11/lib/python3.11/site-packag
es/tqdm/auto.py:21: TqdmWarning: IProgress not found. Please up
date jupyter and ipywidgets. See https://ipywidgets.readthedoc
s.io/en/stable/user_install.html
    from .autonotebook import tqdm as notebook_tqdm
```

```
-----
ArrowNotImplementedError
```

```
t call last)
```

```
Cell In[26]. line 14
```

```
Traceback (most recen
```



```
In [36]: # show the parquet file schema
parquet.read_table("mitocheck_example_images.parquet").schema
```

```
Out[36]: path: extension<awkward<AwkwardArrowType>> not null
filesize_bytes: extension<awkward<AwkwardArrowType>> not null
image_bytes: extension<awkward<AwkwardArrowType>> not null
image_array: extension<awkward<AwkwardArrowType>> not null
-- schema metadata --
ak:parameters: '[{"optiontype_fields": []}, {"record_is_scalar": false}, ' + 22
```

```
In [27]: import lancedb
from pyarrow import parquet

# specify a dir where the lancedb database may go and create lancedb cl
lancedb_dir = pathlib.Path("mitocheck_example_images.lance")
ldb = lancedb.connect(lancedb_dir)

# create the table from pandas via parquet file
ldb.create_table(
    data=pd.read_parquet("mitocheck_example_images.parquet"),
    name="mitocheck_example_images",
    mode="overwrite",
)
```

```
[2024-06-07T17:17:33Z WARN lance::dataset] No existing dataset
at /Users/dabu5788/Documents/work/set-presentations/2024-06-07-
images-and-databases/src/notebooks/mitocheck_example_images.lan
ce/mitocheck_example_images.lance, it will be created
```

```
Out [27]: LanceTable(connection=LanceDBConnection(/Users/dabu5788/Docume
nts/work/set-presentations/2024-06-07-images-and-databases/sr
c/notebooks/mitocheck_example_images.lance), name="mitocheck_e
xample_images")
```





```
In [28]: # read from lancedb to pandas
ldb.open_table("mitocheck_example_images").to_pandas()
```

```
Out [28]:
```

	path	filesize_bytes	image_bytes	image_array
0	mitocheck_exampl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
1	mitocheck_exampl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...

```
In [29]: # show that the dataframes are equal  
pd.testing.assert_frame_equal(  
    pd.read_parquet("mitocheck_example_images.parquet"),  
    ldb.open_table("mitocheck_example_images").to_pandas(),  
)
```

In [30]: `%%timeit`

```
# time pd read_parquet  
pd.read_parquet(path="mitocheck_example_images.parquet")
```

24.9 ms ± 654 µs per loop (mean ± std. dev. of 7 runs, 10 loops each)



In [31]: `%%timeit`

```
# time pyarrow parquet read  
parquet.read_table(source="mitocheck_example_images.parquet")
```

23.4 ms ± 454 µs per loop (mean ± std. dev. of 7 runs, 10 loops each)



In [32]: **%%time**it

```
# time lancedb read
```

```
ldb.open_table("mitocheck_example_images").to_pandas()
```

8.13 ms ± 293 µs per loop (mean ± std. dev. of 7 runs, 100 loops each)



```
In [33]: # show how to add data to an existing table
ldb.open_table("mitocheck_example_images").add(
    pd.read_parquet(filename := "mitocheck_example_images.parquet")
)
ldb.open_table("mitocheck_example_images").to_pandas()
```

```
Out [33]:
```

	path	filesize_bytes	image_bytes	image_array
0	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
1	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
2	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
3	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...

```
In [34]: # show version of the table  
ldb.open_table("mitocheck_example_images").version
```

```
Out[34]: 3
```

```
In [35]: # show a change to original version
ldb.open_table("mitocheck_example_images").checkout(version=1)
ldb.open_table("mitocheck_example_images").to_pandas()
```

```
Out [35]:
```

	path	filesize_bytes	image_bytes	image_array
0	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
1	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
2	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...
3	mitocheck_exempl...	1376512	b'II*\x00\x08\x0...	[[0, 0, 0, 0, 0, ...



# Reflections

- Images can be treated as values through objects in a database table.
- LanceDB seems like a good option for storing multiple tables together as a "package".
- LanceDB integrates well with Parquet, Arrow, and Pandas.
- LanceDB feels fast!



# Thank you for attending!

Questions / comments?

