

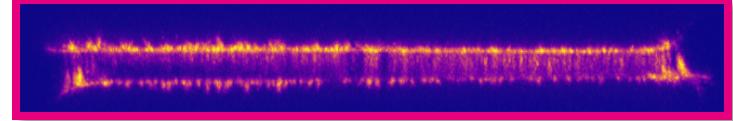
BLITZ ⚡ and WOLKE ☁

A lightweight framework for fast, scalable image data exploration

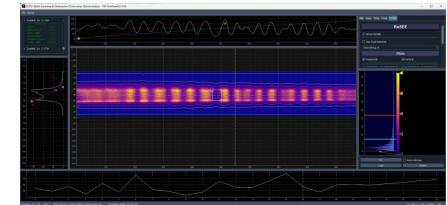
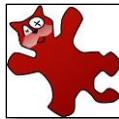
P. Mattern, R. Krieg, H. Höft, T. Gerling, M. Becker

Our Journey Today

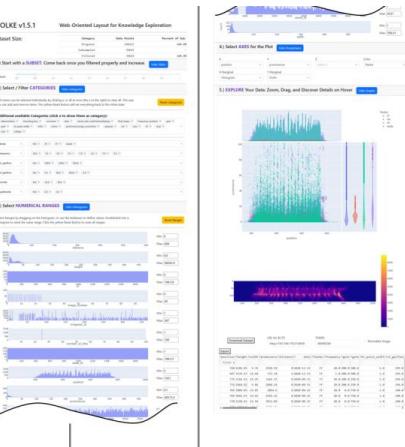
1. Challenges in Image-Based Experiments



2. Current Tools and Interfaces



3. BLITZ – High-speed image inspection



4. Integration via DAMPF

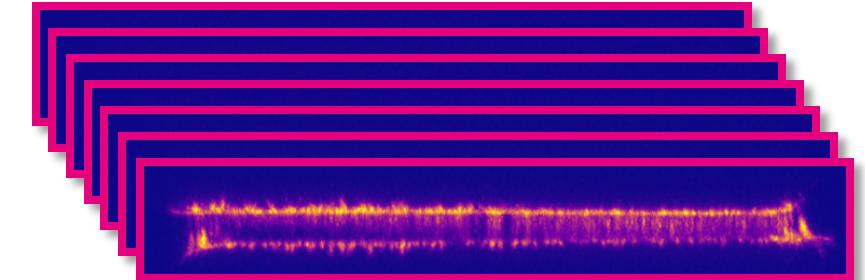
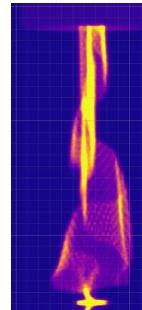


5. WOLKE – Metadata filtering & dashboards

6. Vision: The WETTER Framework

Why Handling Experimental Image Data Is Hard

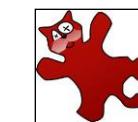
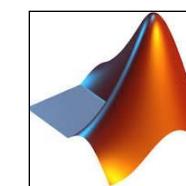
- Massive volume of data



- Limited resources for handling large image datasets
(computing power and storage)

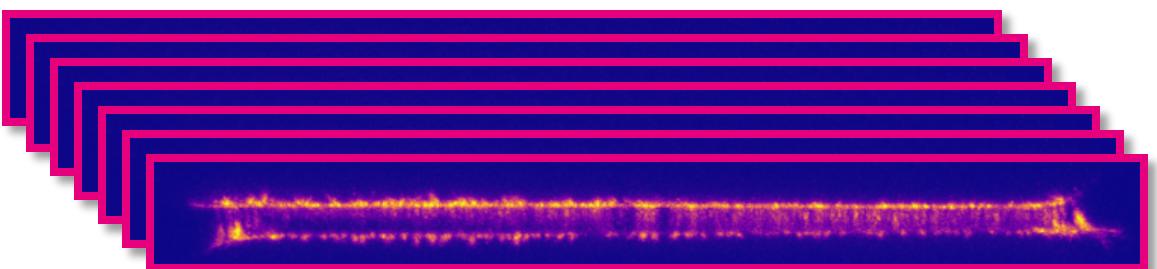


- Need for fast preselection and filtering methods
- Visualization and metadata analysis often inefficient
- Tooling is fragmented and rarely user-friendly



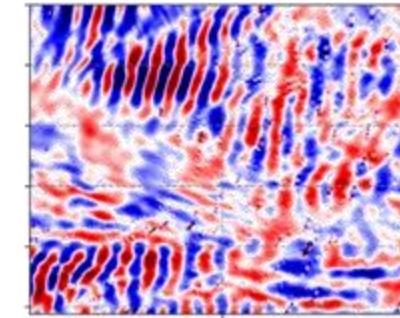
What Are We Really Dealing With?

- Massive volume of data (Examples per Set!)
- Limited resources for handling large image datasets (computing as well as storage)
- Effective preselection and filtering methods
- Efficient visualization and analysis of metadata
- Interactivity and user-friendly interfaces



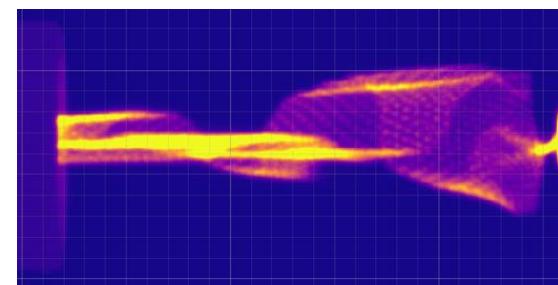
Multifilament iCCD:

- 52 GB
- 32k files



Atmospheric Airglow:

- 47 GB
- 30k files



CAPPJ (kINPen):

- 113 GB
- 36k files

Each dataset is a small universe — containing tens of thousands of frames and multiple GBs of raw data.

Where Does All That Data Go?

- Massive volume of data
- Limited resources for handling large image datasets (computing as well as storage)
- Effective preselection and filtering methods
- Efficient visualization and analysis of metadata
- Interactivity and user-friendly interfaces

Scalability of Image Dataset Storage:

- Storage should scale with dataset size and complexity
- Metadata systems already help — even for single sets
- Folder structures remain the most common solution

Storage Type	Typical Dataset Range	Use Case
Laptop	10 – 1,000 images	Local experiments, quick previews
External HDD	1,000 – 30,000 images	Archiving, manual browsing
File Server	30,000 – 500,000 images	Shared access, project-wide datasets
Dedicated Solution	500,000 – 10+ million	Large-scale research, searchable archives

Storage Options by Dataset Size and Use Case



How People Explore Image Data Today

- Massive volume of data
- Limited resources for handling large image datasets (computing as well as storage)
- Effective preselection and filtering methods
- Efficient visualization and analysis of metadata
- Interactivity and user-friendly interfaces

File Explorer:



Windows



Mac



Ubuntu

Database:



Google Photos



Open Microscopy Environment

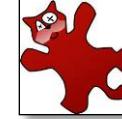
Traditional Image Viewers:



Microsoft Fotos



ImageJ



IrfanView

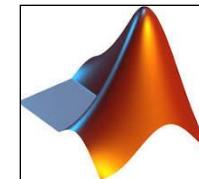
Programmatic Interfaces:



Javascript



Python



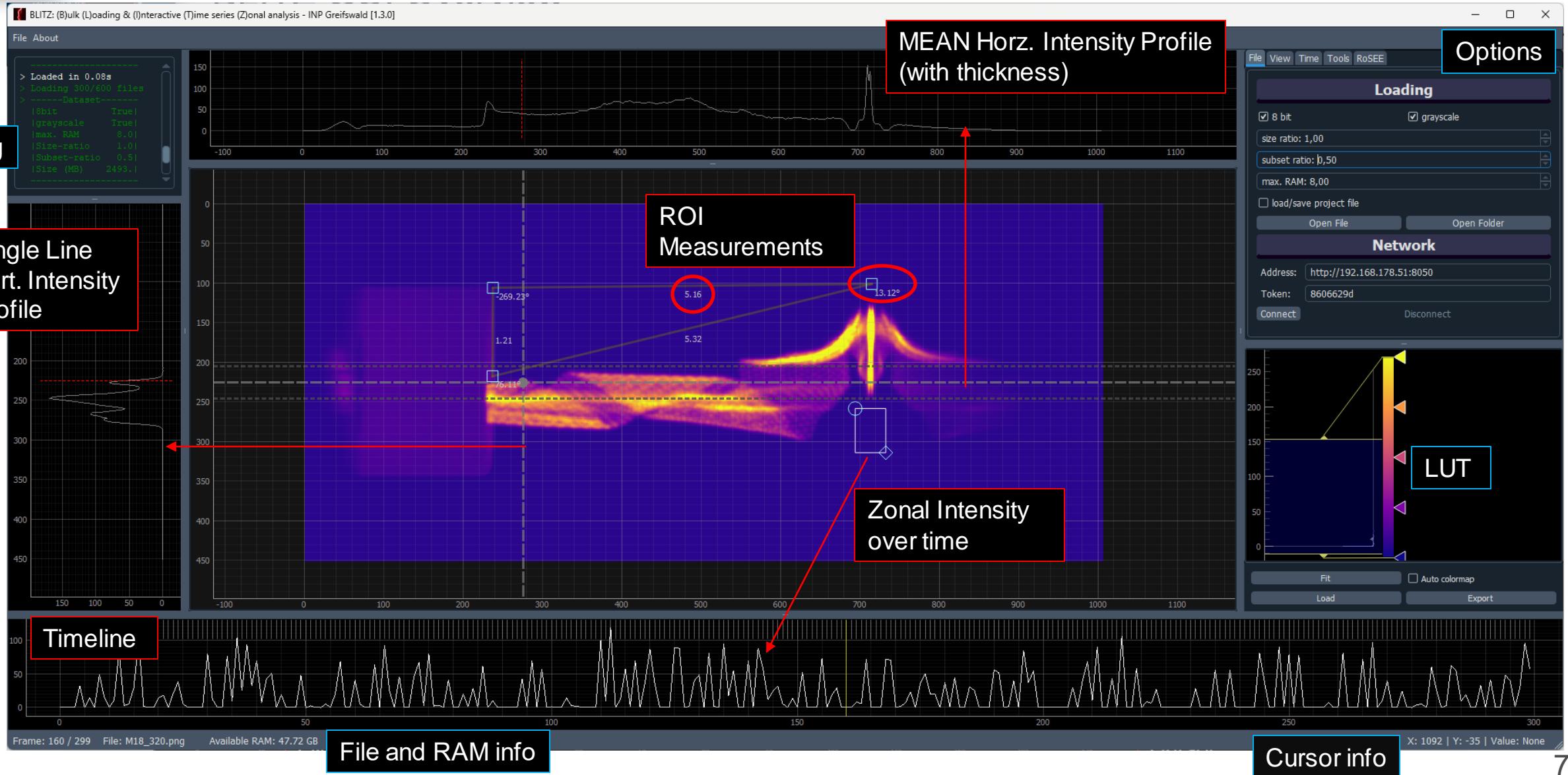
Matlab

DIY Exotic Flavours:

Unmaintained scripts, forgotten web GUIs, or that one tool only Peter can run...



“Most of us juggle multiple tools to get the job done — but what if filtering, interaction, and inspection were all part of a single flow?”



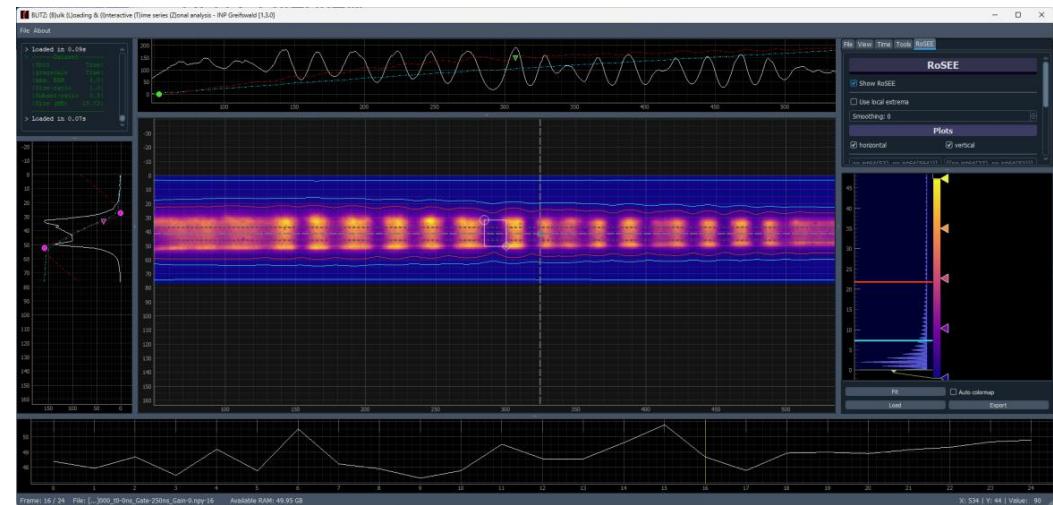
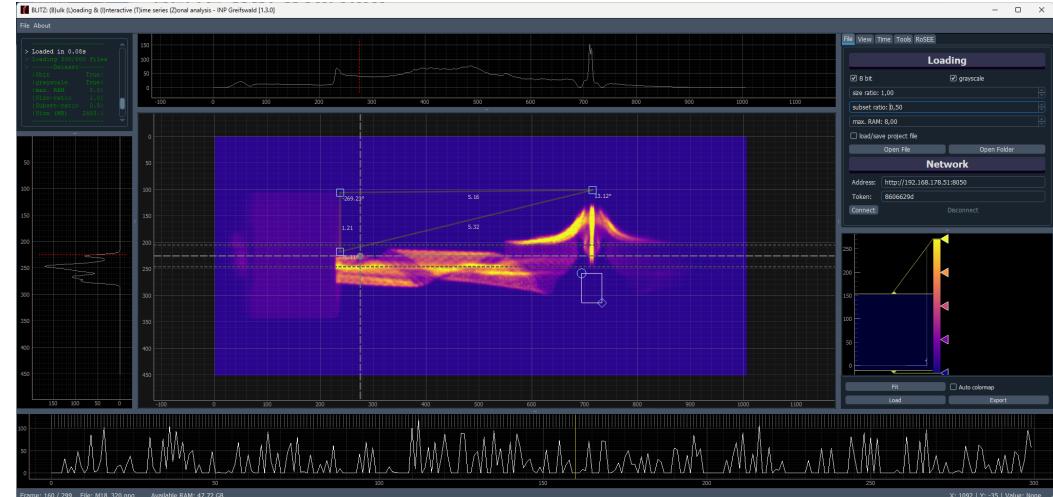
Definition and Purpose:

- Matrix-based interactive image viewer
- Designed for efficient large-scale image datasets

Key Features:

- Extremely fast loading (e.g. 21,000 images, 25 GB in 35 sec)
- Real-time interactive exploration and analysis
- Robust tools for measuring, profiling, and zonal analysis
- Drag & Drop support for quick dataset loading

When traditional tools choke on 10,000+ images,
BLITZ opens them in seconds - ready to analyze.



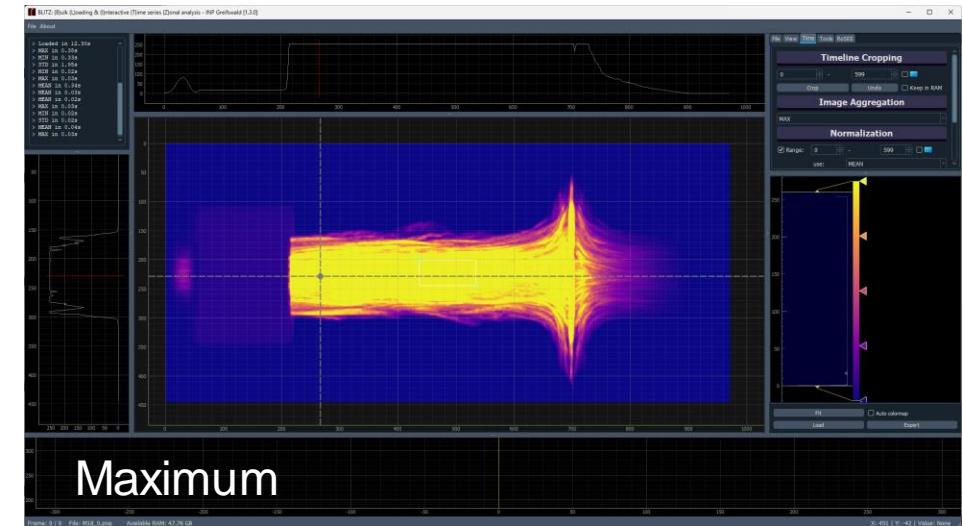
BLITZ - Performance

```
> STD in 0.03s
> Loading 600/600 files
> -----Dataset-----
|8bit      True|
|grayscale True|
|max. RAM   8.0|
|Size-ratio 1.0|
|Subset-ratio 1.0|
|Size (MB)  4987.1|
|-----|
> Loaded in 12.75s
> MEAN in 0.33s
> MAX in 0.35s
> STD in 1.97s
```

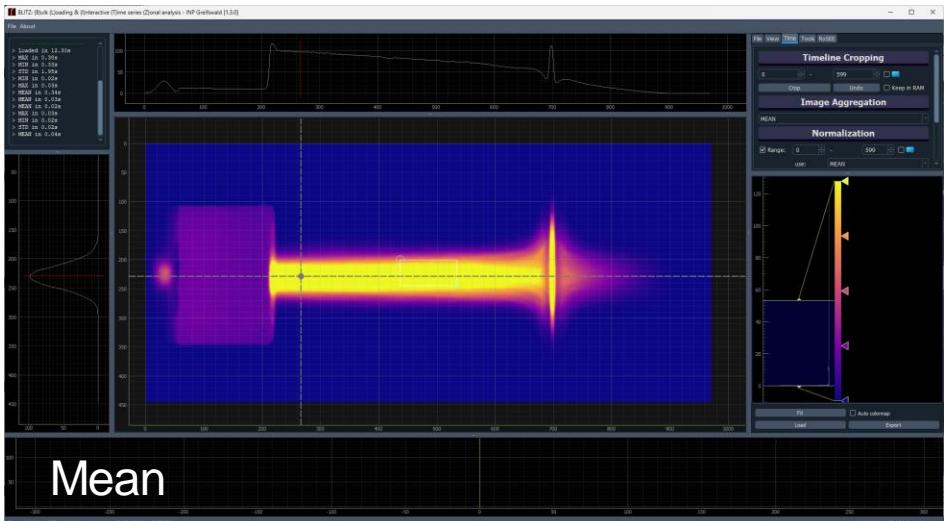
Matrix based approach:

- 600 images (5 GB) loaded in 12 sec
- MAX / MEAN in < 0.35 sec
- STD in < 2 sec

(Tested on a gaming laptop)



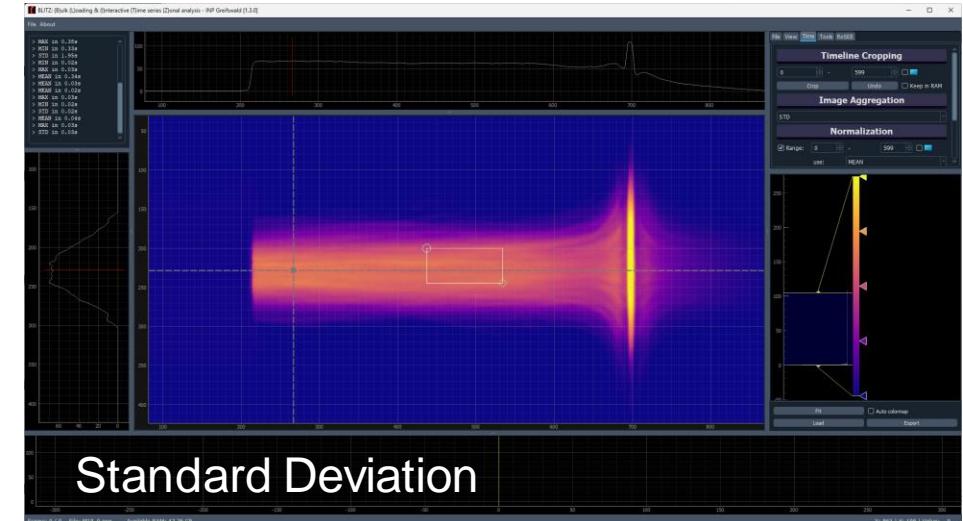
Maximum



Mean

Intensity Values
per Pixel over
Time

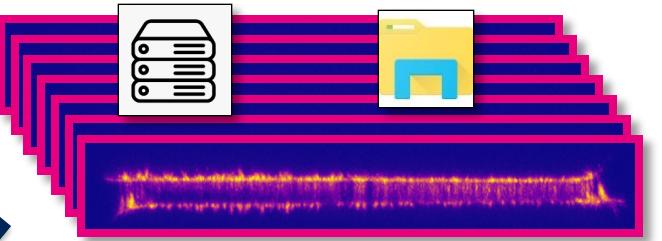
Built for users who
just don't want to
wait ...



Standard Deviation

DAMPF (Data Aggregation & Modular Processing Framework)

Image Folder Structure



Non-Domain Specific Metrics

i.e. via OpenCV:

- Min, Max, Mean Intensity
- Laplacian („Noise“)
- Self-similarity (auto-correlation)
- Date / Time

All into a SQL DB

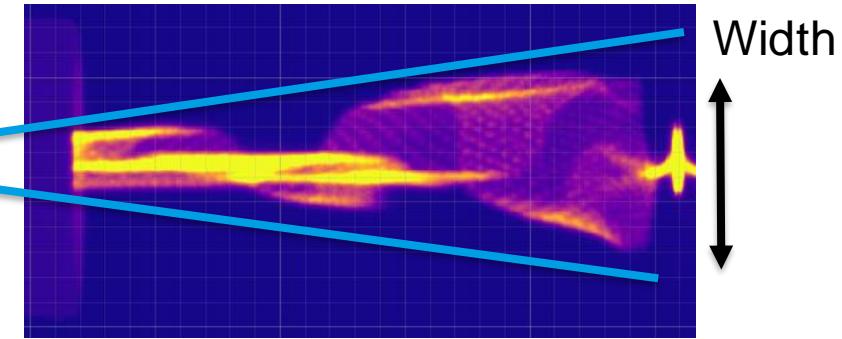


Experimental Conditions:

- Voltage
- Current
- Gas Mixture
- ...

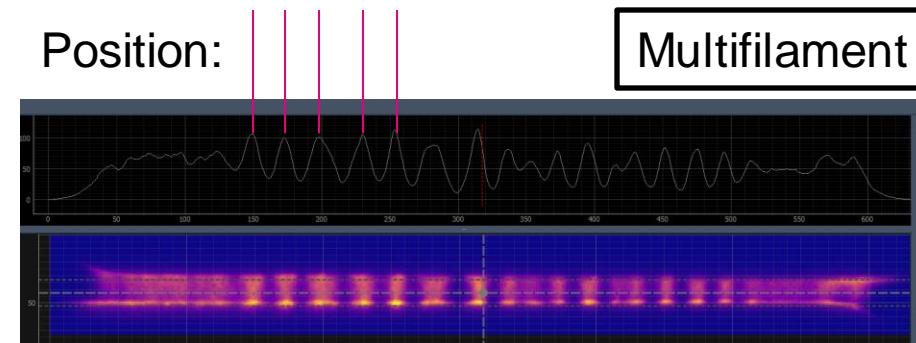
Creating Metadata from Raw Image Folders
“Think of DAMPF as your metadata generator —
turning folders into searchable databases.”

Angle



Domain Specific Metrics

Position:



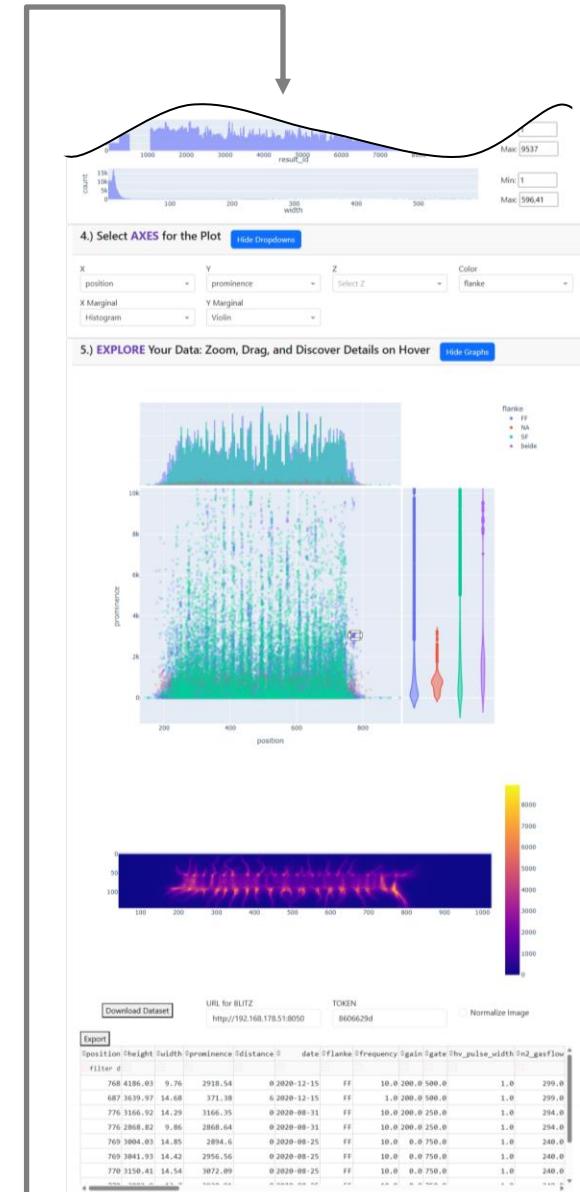
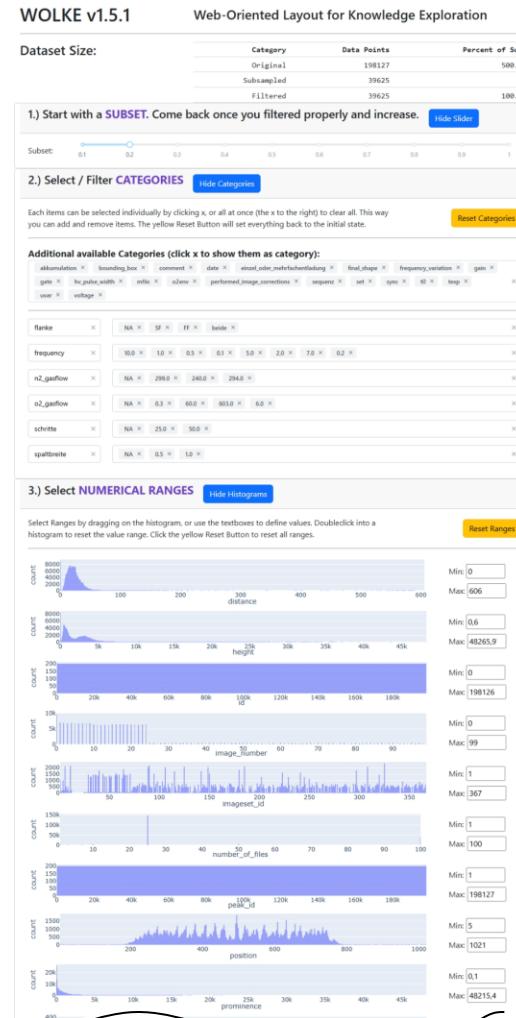
WOLKE (Web-Oriented Layout for Knowledge Exploration)

An interactive web dashboard built with Dash-Plotly for efficient metadata filtering and visualization.

Functionality:

- Explore metadata by type: numeric or categorical
- Create interactive x/y/(z) scatter plots with color-coded metadata

WOLKE helps researchers reduce complexity and focus on relevant subsets – no coding required.



Subset and Filtering

1.) Start with a Subset:

Improves performance on large datasets

2.) Filter Categories:

- Categories = all non-numeric data
- Numeric data with low variance (e.g. frequency, n=8) can also be treated as categories

3.) Filter Numerical Ranges:

- View histogram of value distribution
- Filter using range sliders or input fields

WOLKE v1.5.1

Web-Oriented Layout for Knowledge Exploration

Dataset Size:

Category	Data Points	Percent of Sub.
Original	198127	500.0%
Subsampled	39625	-
Filtered	39625	100.0%

1.) Start with a **SUBSET**. Come back once you filtered properly and increase.

[Hide Slider](#)

Subset: 

2.) Select / Filter **CATEGORIES**

[Hide Categories](#)

Each items can be selected individually by clicking x, or all at once (the x to the right) to clear all. This way you can add and remove items. The yellow Reset Button will set everything back to the initial state.

[Reset Categories](#)

Additional available Categories (click x to show them as category):

akkumulation x bounding_box x comment x date x einzel_oder_mehrfachentladung x final_shape x frequency_variation x
 gain x gate x hv_pulse_width x mfic x o2env x performed_image_corrections x sequenz x set x sync x t0 x
 temp x uvar x voltage x

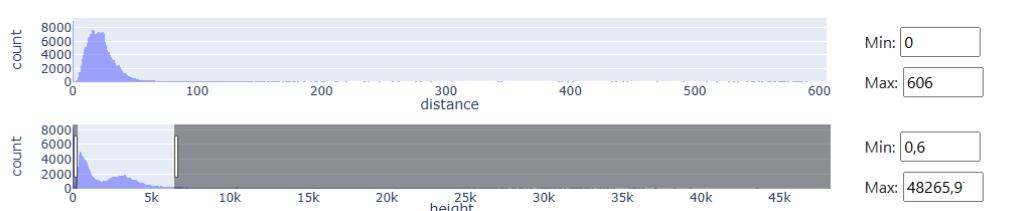


3.) Select **NUMERICAL RANGES**

[Hide Histograms](#)

Select Ranges by dragging on the histogram, or use the textboxes to define values. Doubleclick into a histogram to reset the value range. Click the yellow Reset Button to reset all ranges.

[Reset Ranges](#)



Visual Data Exploration

4.) Select Plot Axis:

- 2D or 3D scatter plots
- Optional: Add side plots for statistics (e.g. histogram, violin)

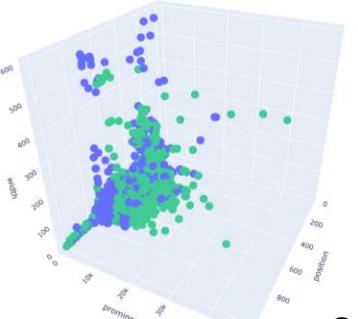
4.) Select **AXES** for the Plot Hide Dropdowns

X	position	Y	distance	Z	Select Z	Color	flanke
X Marginal	Histogram	Y Marginal	Violin				

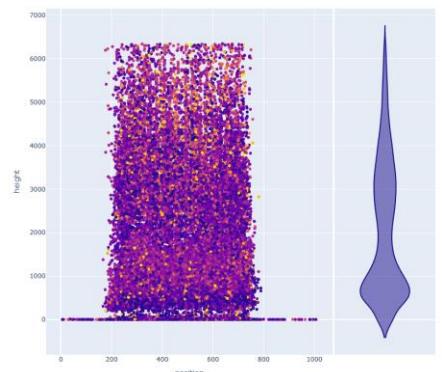
5.) **EXPLORE** Your Data: Zoom, Drag, and Discover Details on Hover Hide Graphs

5.) Explore Your Data Visually:

- Zoom, drag, and hover with your mouse
- Show/hide categories to focus the view
- Select single or multiple data points interactively



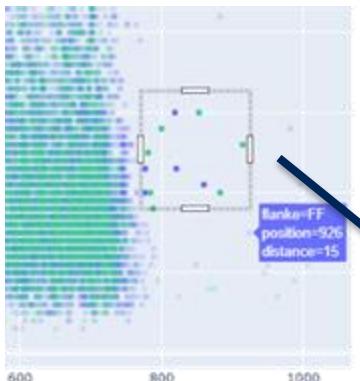
3D Scatter



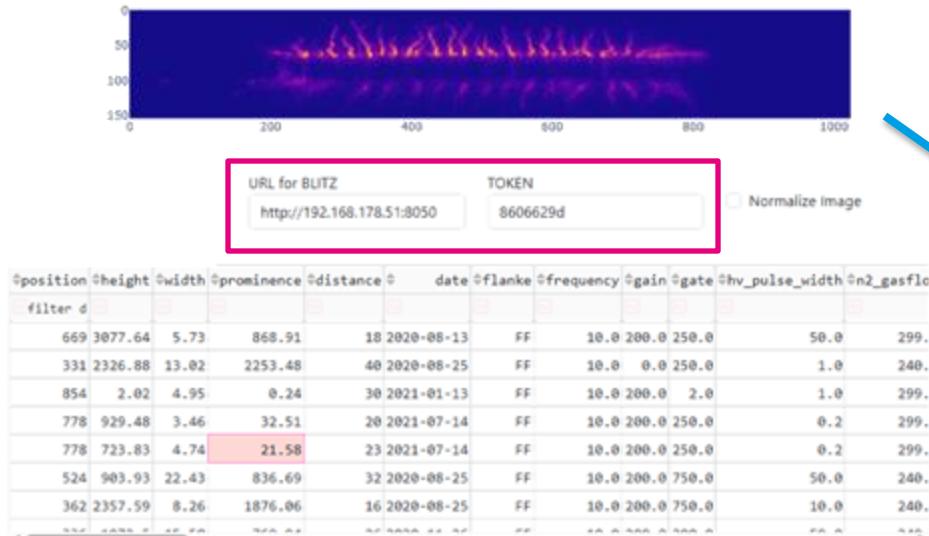
Continuous
Colorscale

From filtered metadata to deep image analysis — seamlessly.

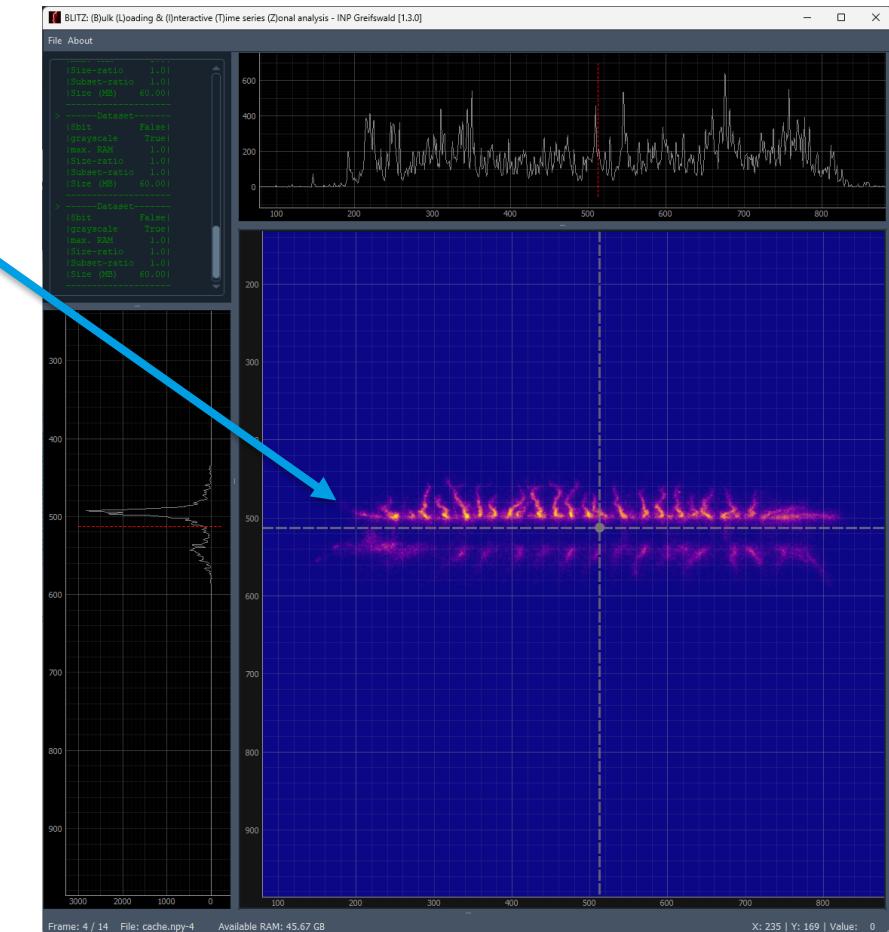
Select



Inspect



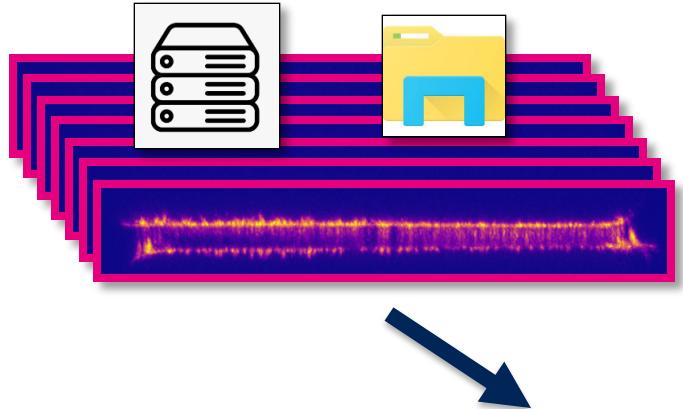
Synchronize with the BLITZ!



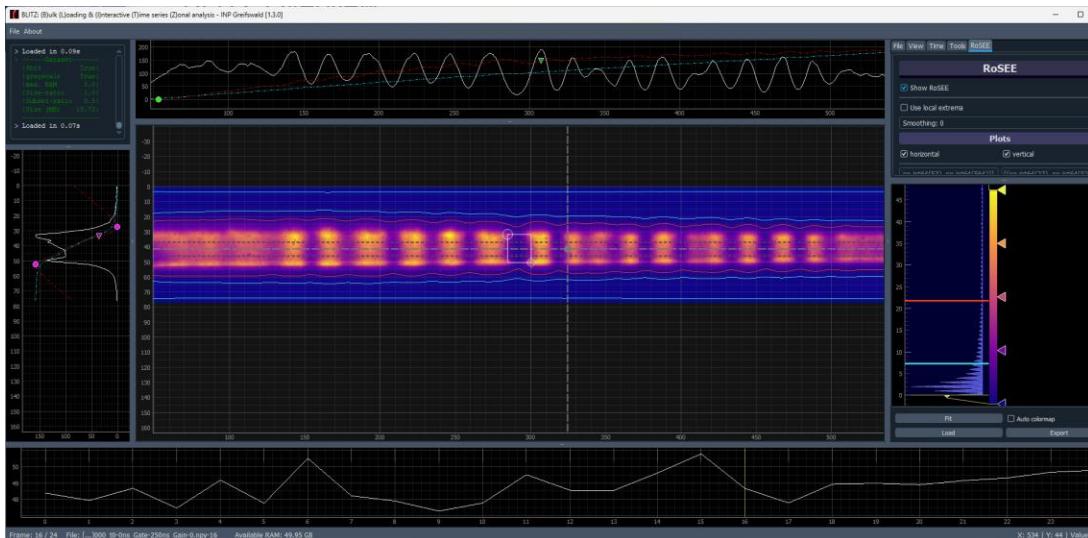
WOLKE provides filtered data via HTTP as *.npy arrays,
which can be instantly opened in BLITZ —
or integrated into your own analysis framework.

From Folder to Insight: Your Workflow with DAMPF, WOLKE & BLITZ

Keep your image data in folders
...just like you're used to!



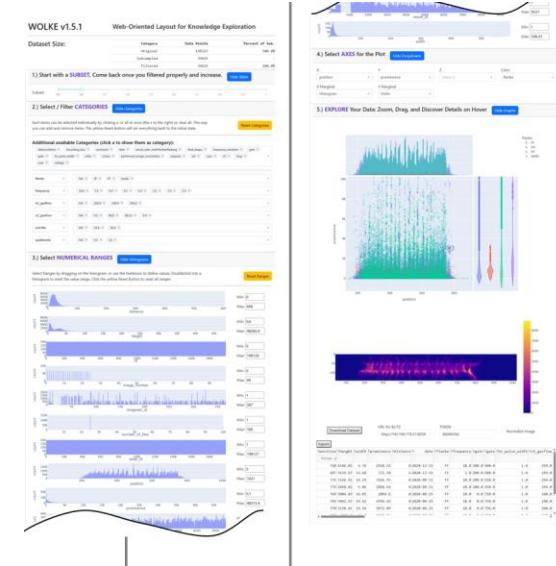
Preview directly
with BLITZ
(Drag & Drop
sneak peek)



Augment metadata
with DAMPF
(and store in SQL)



Filter and preselect
using WOLKE



Synchronize filtered
sets with BLITZ for
detailed inspection

Outlook: WETTER Framework

