

SCHOOL ON  
COLLECTIVE ANIMAL BEHAVIOUR



# A Python Library to Handle Trajectories

Alfredo Reyes González

La Habana, 2023



- Molecular Dynamics

TheRoe, D.R., et.al., 2013. Ptraj and cpptraj: software for processing and analysis of molecular dynamics trajectory data.

- Urban Trajectories

Shamal, A.D., et.al., S., 2019. An open source trajanalytics software for modeling, transformation and visualization of urban trajectory data

- Human Mobility

Pappalardo, L., et.al., 2019. scikit-mobility: A python library for the analysis, generation and risk assessment of mobility data.

- Geo-positional trajectories

Graser, A., 2019. Movingpandas: efficient structures for movement data in python.

- Generic (two dimensional) trajectories

Shenk, J., et.al., 2021. Traja: A python toolbox for animal trajectory analysis. The Journal of OpenSource Software .

- Wide diversity of software to address specific trajectory-related task.
- Most of existing software for trajectory analysis only address trajectories limited to a fixed number of dimensions.
- Lack of integration between trajectory softwares.



outline





Position Paper

# yupi: Generation, tracking and analysis of trajectory data in Python

A. Reyes <sup>a</sup>, G. Viera-López <sup>b</sup>, J.J. Morgado-Vega <sup>a</sup>, E. Altshuler <sup>a</sup>

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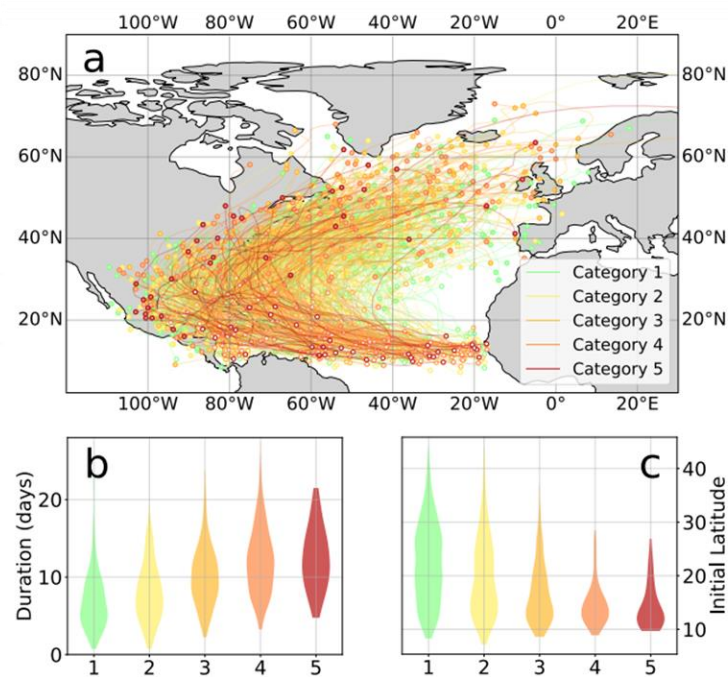
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<https://doi.org/10.1016/j.envsoft.2023.105679>

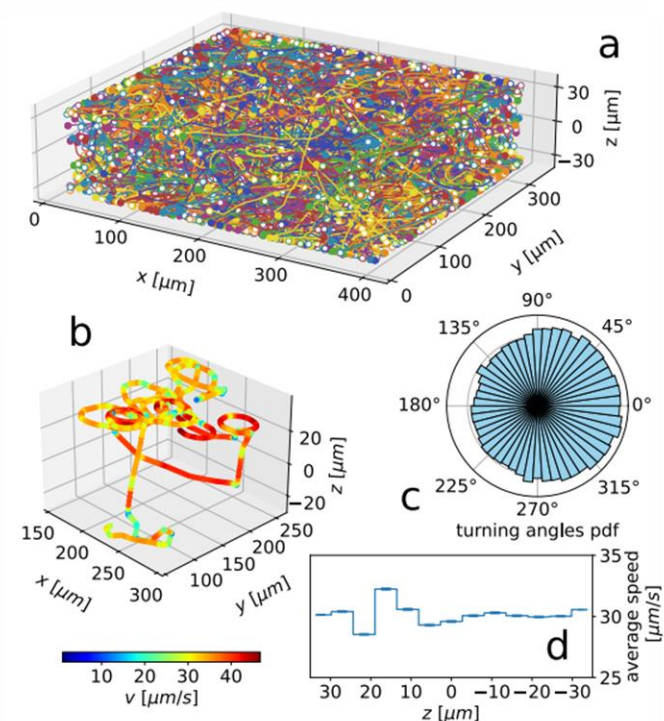


<https://github.com/yupidevs/yupi>

[https://github.com/yupidevs/yupi\\_examples](https://github.com/yupidevs/yupi_examples)



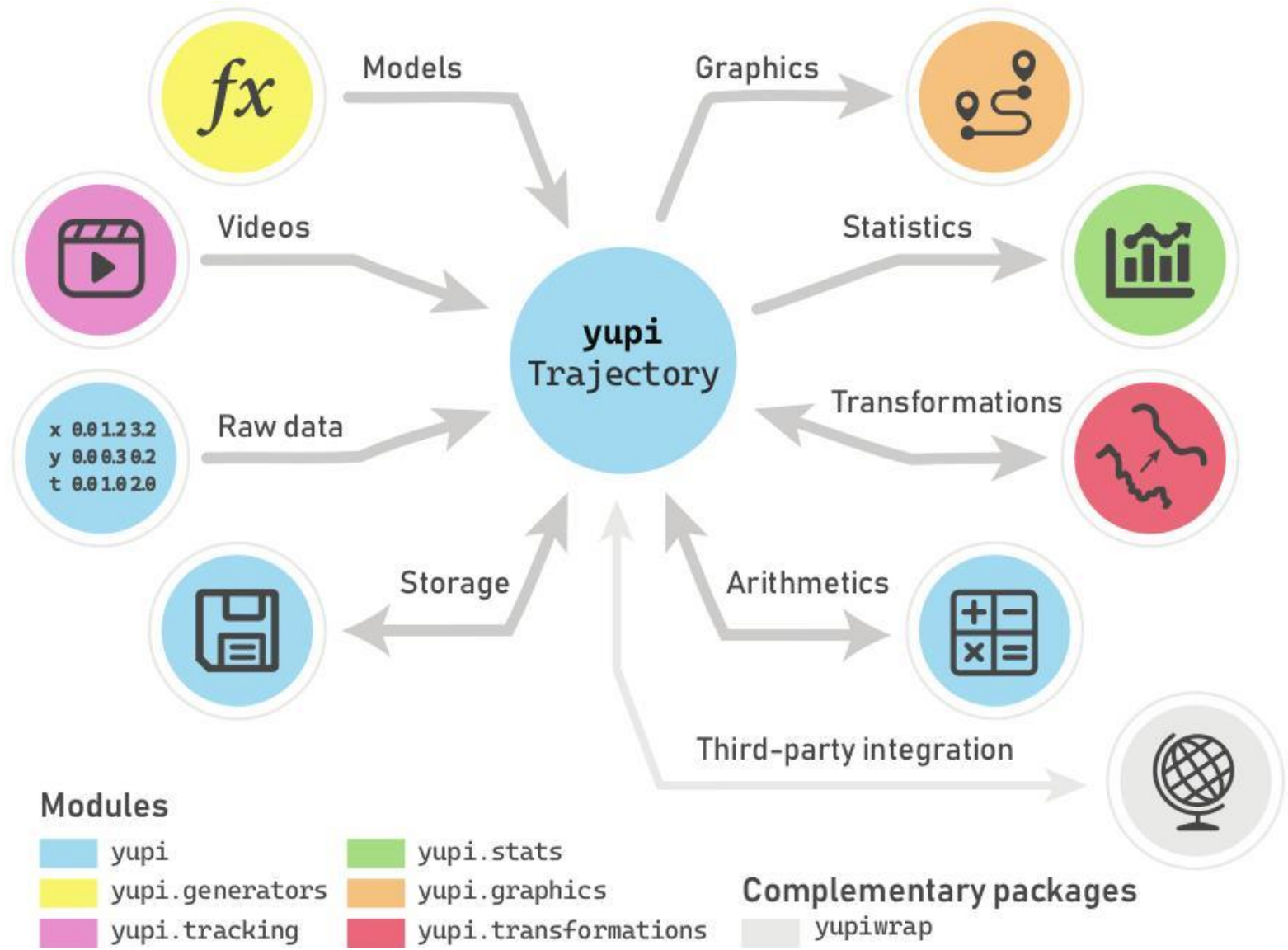
**Figure 4:** Visual inspection of the trajectories of hurricanes in the North Atlantic. (a) Trajectories displayed using a *yupi*'s spatial plot. (b) Distribution of hurricane duration for each category. (c) Distribution of hurricane's initial latitude for each category.

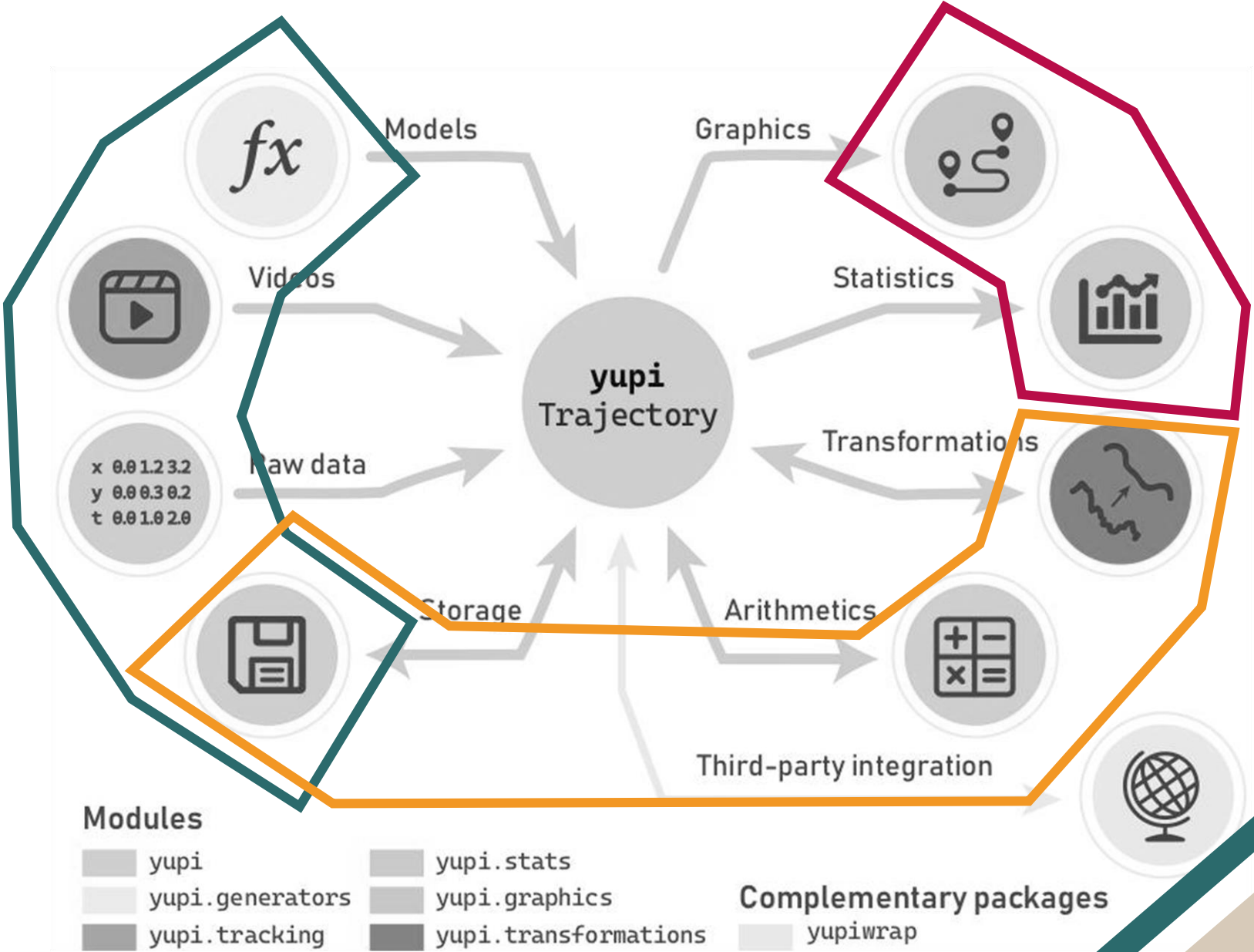
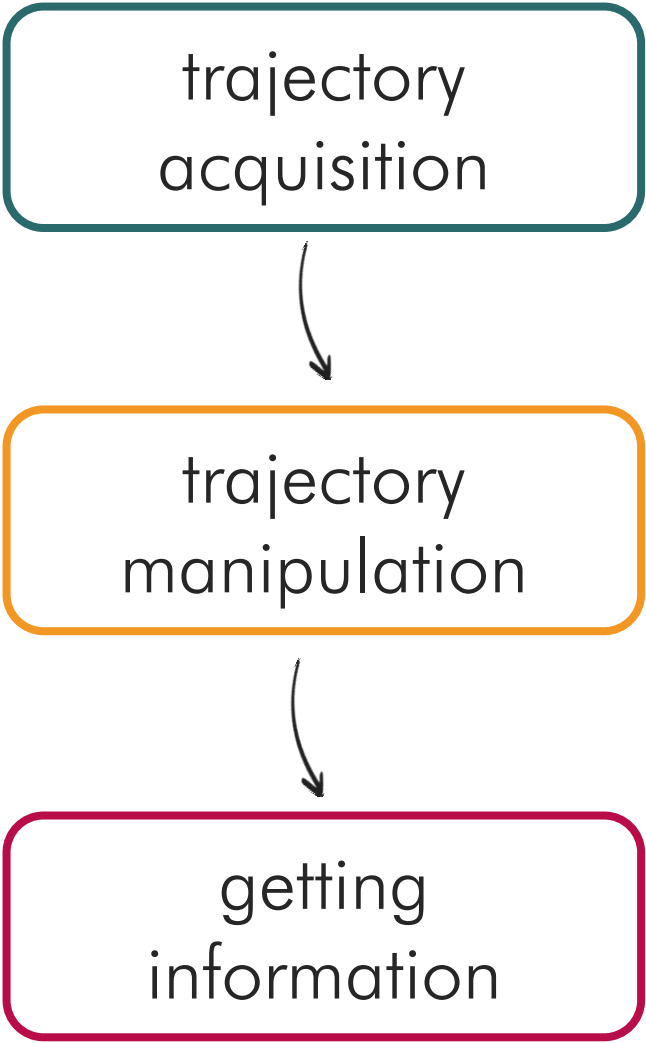


**Figure 5:** Reproduction of the analysis conducted in (Grognot and Taute, 2021). (a) Individual trajectories of an ensemble of  $N = 2000$  bacteria. (b) Trajectory of a single bacteria where colors have been linked to speed values. Blue/red correspond to sections of low/high local speed. (c) Distribution of swimming direction. (d) Average speed and standard error of the mean as a function of height.



# Yet Underused Python Instrument









```
from yupi import Vector

vec = Vector([[3, 4], [2, 0], [8, 6]])

print(vec.x, vec.component(0))
# (Vector([3., 2., 8.]),
#  Vector([3., 2., 8.]))

print(vec.norm)
# Vector([ 5.,  2., 10.]])
```

```
from yupi import Trajectory

traj = Trajectory(
    x=[3, 5, 13], y=[4, 4, 10],
    dt=0.5, traj_id='my_traj')

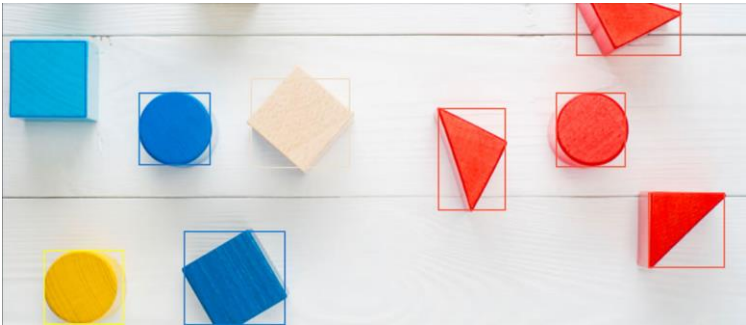
print(traj.r, traj.r.x)
# (Vector([ 3.,  4.],
#          [ 5.,  4.],
#          [13., 10.])),
#  Vector([ 3.,  5., 13.]])

print(type(traj.r))
# yupi.vector.Vector
```

- The **Vector** class stores all time-evolving data in a trajectory.
- It was implemented by wrapping **ndarray** type from **numpy**.
- Information from a **Vector** instance can be extracted by accessing certain properties.
- A **Trajectory** object is the fundamental structure within **yupi**.
- Time, position, velocity and acceleration time series can be accessed through the attributes **t**, **r**, **v** and **a**.
- Other cool things:
  - `traj.v`
  - `traj.r.delta.norm`
  - `traj1 + traj2`
  - `traj[:,2]`



■ Color Matching



■ Template Matching

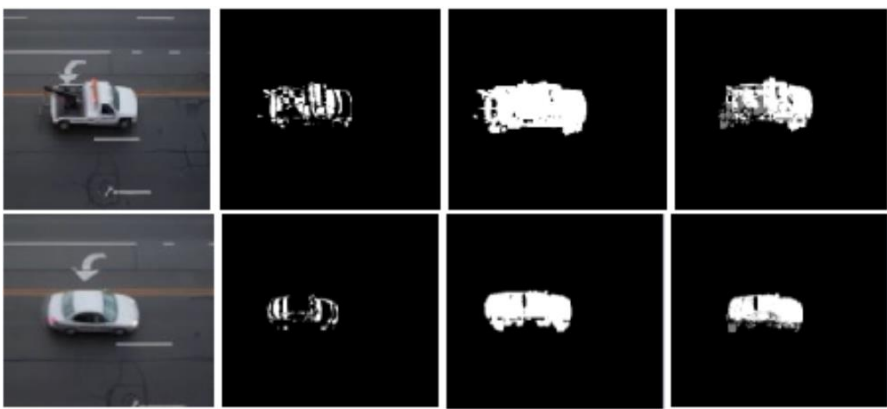


Template location



Template

■ Frame Differencing



■ Background subtraction



■ Optical Flow







- The tracking workflow

1. algorithm → **Algorithm**

2. roi → **ROI**

3. tracker1 → **ObjectTracker**

...

trackerN → **ObjectTracker**

4. trackers.append({tracker})

5. scenario → **TrackingScenario**

6. scenario.track()

- Choose a tracking **algorithm**.

- Select a **region of interest**.

- Define **objects** to track.

- **Gather** all tracking objects.

- Create your **tracking scenario**.

- Run Forest, **run!**

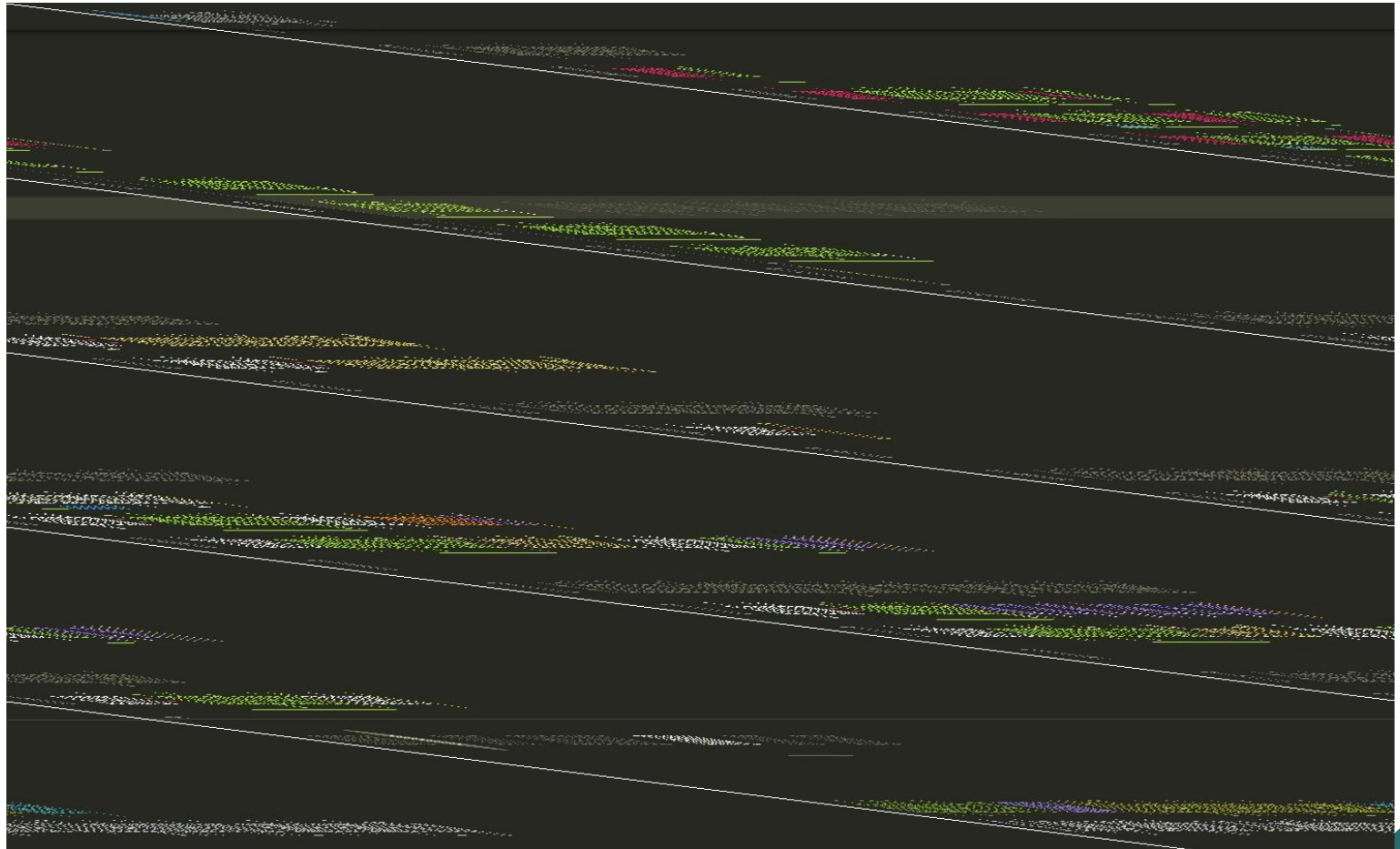
- Three Examples [https://github.com/yupidevs/yupi\\_examples](https://github.com/yupidevs/yupi_examples)



```
template = cv2.imread(template_path)
algorithm1 = TemplateMatching(
    template)
tracker1 = ObjectTracker(
    "Central Pivot",
    algorithm1,
    ROI((80, 80)))
trackers.append(tracker1)
```

```
algorithm2 = ColorMatching(
    (80, 170, 90), (190, 255, 190))
tracker2 = ObjectTracker(
    "Green LED",
    algorithm2,
    ROI((50, 50)))
trackers.append(tracker2)
```

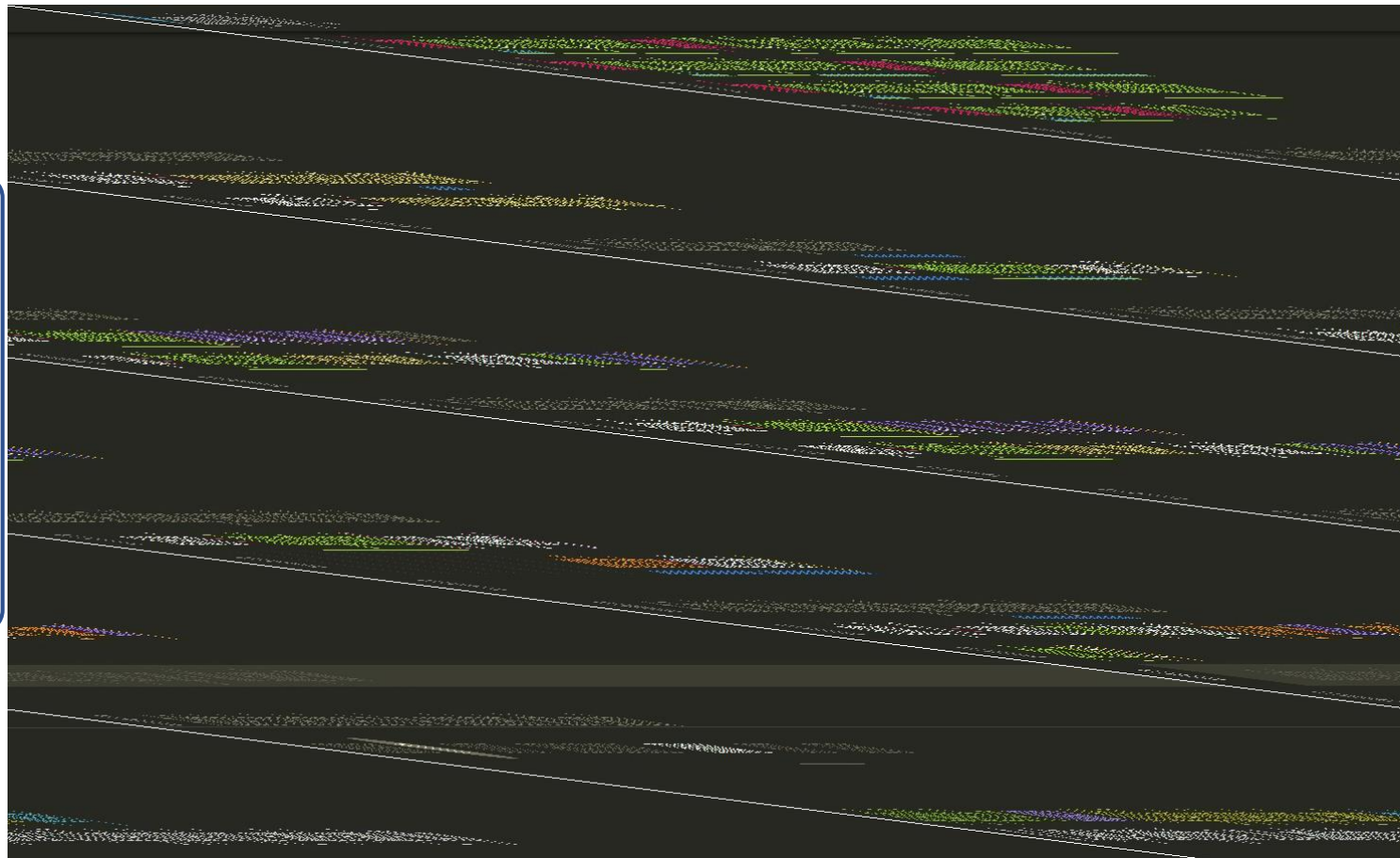
```
scenario = TrackingScenario(trackers)
scenario.track(video_path,
    pix_per_m=4441,
    start_frame=160)
```



- how to create a tracking scenario using Template and Color Matching algorithms



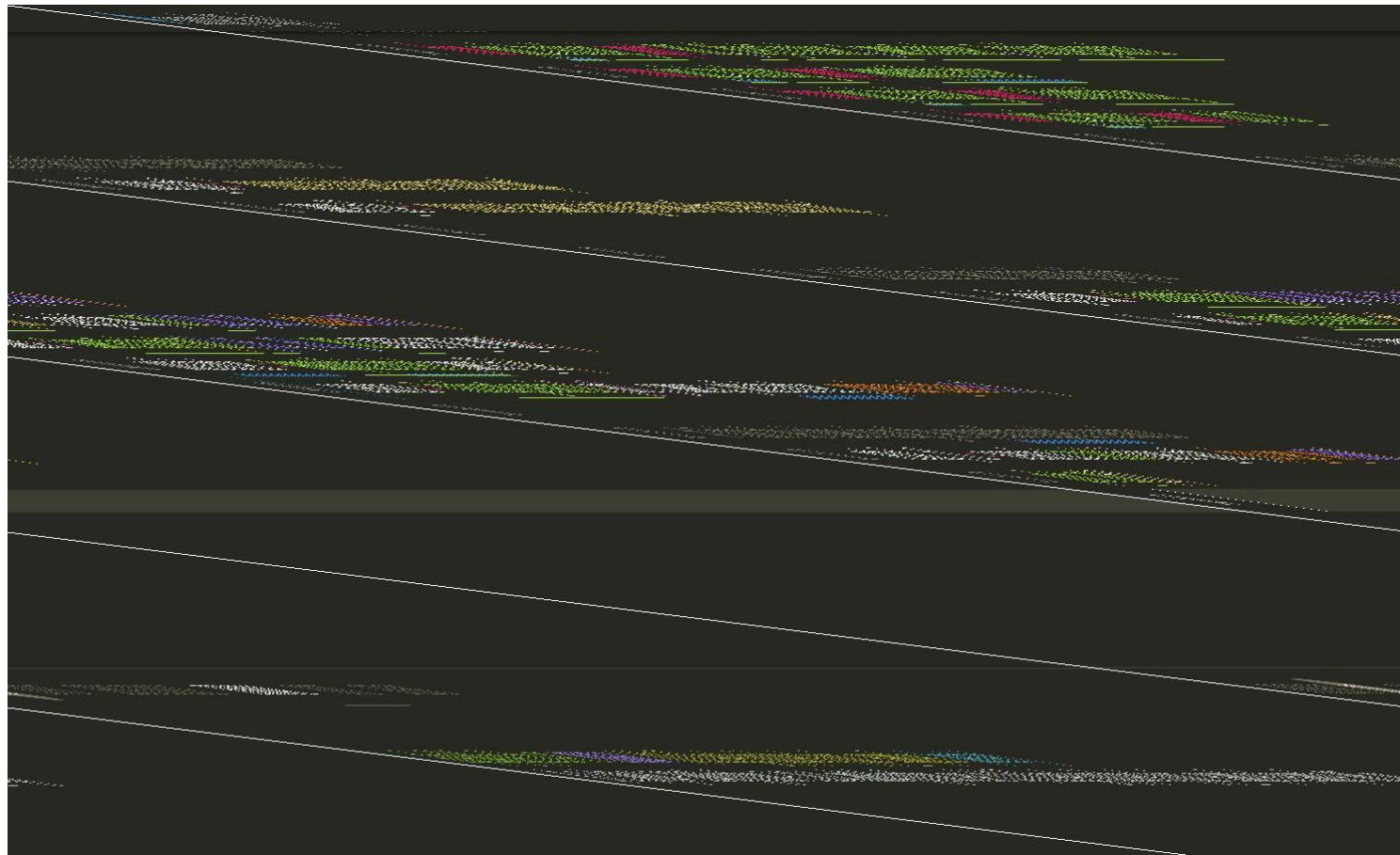
```
camera_file =  
'resources/cameras/gph3+.npz'  
  
undistorter = RmapUndistorter(camera_file)  
  
scenario = TrackingScenario(  
    trackers,  
    undistorter=undistorter)
```



- how to correct camera distortions



```
ant = ObjectTracker(  
    'ant', algorithm,  
    ROI((120, 120)))  
  
camera = CameraTracker(  
    ROI((.65, .65),  
    ROI.CENTER_INIT_MODE))  
  
scenario = TrackingScenario(  
    [ant], camera,  
    undistorter)
```



- how to track when the camera is also moving



- Langevin Model

$$\frac{d}{dt}\mathbf{v}(t) = -\gamma\mathbf{v}(t) + \sigma\boldsymbol{\xi}(t)$$

- Velocity Autocorrelation Function

$$\phi(\tau) = \frac{1}{T-\tau} \int_0^{T-\tau} \mathbf{v}(t') \cdot \mathbf{v}(t' + \tau) dt'$$

Generator objects

```
from yupi.generators import  
LangevinGenerator  
  
lg = LangevinGenerator(  
    T, dim, N, dt, gamma, sigma)  
trajs = lg.generate()
```

Observables

```
from yupi.stats import vacf  
  
vacf_mean, vacf_std = vacf(  
    trajs,  
    time_avg=True,  
    lag=lag_vacf  
)
```





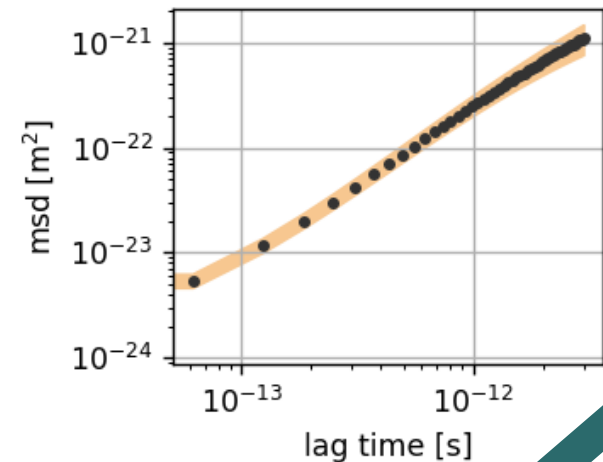
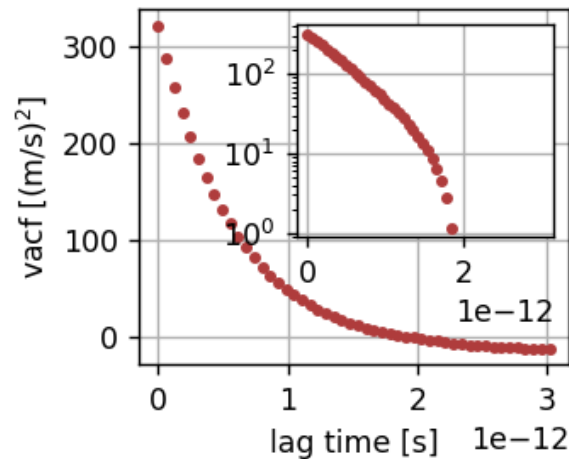
- Filters

$$\mathbf{v}_s(t) = \Omega \int_0^t \mathbf{v}(t') e^{-\Omega(t-t')} dt'$$

```
from yupi.transformations import  
    exp_convolutional_filter  
  
smooth_traj = exp_convolutional_filter(  
    traj, ommega=omega)
```

- Observables

```
from yupi.graphics import plot_vacf  
  
plot_vacf(vacf, dt, lag_vacf, show=False)  
  
plot_msd(msd, msd_std, dt, lag=lag_msd)
```





...and there is more

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- Ways to instantiate Trajectory objects.
- Storage on disk
- Custom tracking algorithms (the abstract Algorithm class)
- Custom generators (the abstract Generator class)
- yupiwrap
- Machine Learning perspectives



## Summary and Conclusions:

- **yupi** is an open-source Python library designed for comprehensive **handling of trajectory data** of any dimension.
- It integrates the processes of trajectory **extraction, generation, transformation, storage**, statistical **analysis**, and **visualization**.
- Methods are distributed across six modules, organized to operate on the **same data structure**.
- Trajectory objects are manipulated through instances of Vector objects, which **inherit from** ndarray from the **numpy** library.
- **Integrate** with other Python libraries dedicated to trajectory management.

▪ J. J. **Morgado** Vega

▪ G. **Viera** López



Department of Computer Science,  
Gran Sasso Science Institute, Italy



thanks

