

traffic

a toolbox for processing and analysing air traffic data

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An open-source library

📁 [xoolive / traffic](#)

A toolbox for processing and analysing air traffic data

🔗 traffic-viz.github.io/

📄 MIT License



- Code: <https://github.com/xoolive/traffic/>
Documentation: <https://traffic-viz.github.io/>
- Development started early 2018
- traffic, a toolbox for processing and analysing air traffic data, *Journal of Open Source Software* (4), 2019. DOI: 10.21105/joss.01518

- **Access to (open) data**
Includes trajectories, flight plans, airspace structure, weather information, etc.
- **Trajectory preprocessing**
Clean and filter data, prepare datasets, enrich with metadata
- **Algorithms**
clustering, detection of operational events, implement KPI
- **Data visualization**
Maps, interactive visualization

Access to (open) data

- Access to OpenSky Network historical database
<https://github.com/open-aviation/pyopensky>
- Access to description of airspace structure
Parsing facilities for XPlane format, Eurocontrol DDR + AIXM data, FAA open data
- Access to OpenStreetMap data
<https://github.com/xoolive/cartes>
- Access to weather data
METAR history information, ERA5 with <https://github.com/junzis/fastmeteo>

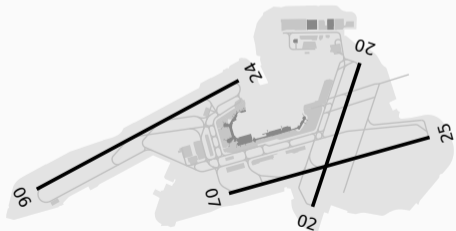
A tabular *tidy* format: pandas, geopandas, xarray

Access to (open) data [code]

```
from traffic.data.samples import * # for documentation and testing
from traffic.data.datasets import * # public datasets included in publication

from traffic.data import airspaces # public sources or AIRAC data
from traffic.data import airports # airports, runways, apron structure

airports["LFPO"]
```



`pandas` misses a semantics for trajectories

`geopandas` suits well geometrical shapes, not time series

There are common noise patterns in data that we learn to process.

`traffic` comes up with a semantics for processing trajectory data

The same semantics applies on:

- individual trajectories (`Flight`) and
- collections of trajectories (`Traffic`)

Trajectory preprocessing [code]

```
from traffic.core import Flight, Traffic

Flight.from_file(...) # one single trajectory
Traffic.from_file(...) # a collection of Flight

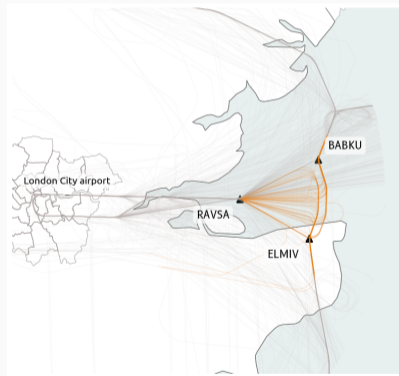
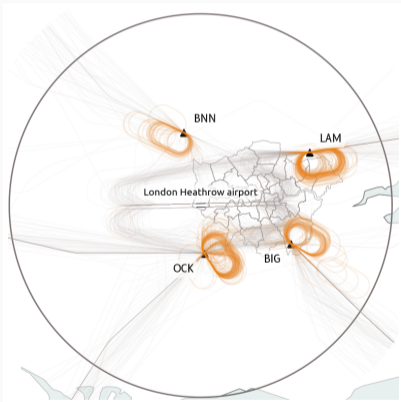
flight.duration # a pd.Timestamp
flight.first("10 minutes") # a new Flight
flight.intersects(LFBOTMA) # a boolean
flight.simplify() # Douglas-Peucker simplification
flight.aligned_on_ils("LFPO") # iterates on segments on final approach
flight.go_around()
flight.holding_pattern()
```

Efficient implementations for common problems:

- **detection of specific events**
holding patterns, go around, point merge, in-flight refuelling, aerial survey, firefighting, etc.
- **trajectory clustering**
- **trajectory generation**
- **closest point of approach**
- **fuel flow estimation**, with OpenAP
- etc.

Example: select holding patterns and point merge systems

```
collection.has("aligned_on_ils('EGLL')").has("holding_pattern").eval()
```



Example: compute an occupancy graph for a given airspace

```
collection
.within_bbox(airspaces["LFEE5R"])
.intersects(airspaces["LFEE5R"])
.clip(airspaces["LFEE5R"])
.summary(["callsign", "icao24", "typecode", "start", "stop", "duration"])
.eval()
```

Plotting facilities for all data structures with common visualization solutions:

- Matplotlib
- Leaflet (ipyleaflet) explore trajectories in a widget
- Plotly interactive visualizations

and more exploratory options:

- Mapbox (Open GL)
- JavaScript/Observable options

- Trajectory prediction
- Interaction with state-of-the-art tools
- Improvements on performance
- Facilitate the usage of the same semantics on real-time data
- Exploration of more visualization techniques

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