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1 Package loading and basic configurations

```
[1]: %load_ext autoreload
      %autoreload 2

      # load dependencies'
      import pandas as pd
      import geopandas as gpd

      from envirocar import TrackAPI, DownloadClient, BboxSelector, ECConfig

      # create an initial but optional config and an api client
      config = ECConfig()
      track_api = TrackAPI(api_client=DownloadClient(config=config))
```

2 Querying enviroCar Tracks

The following cell queries tracks from the enviroCar API. It defines a bbox for the area of Münster (Germany) and requests 50 tracks. The result is a GeoDataFrame, which is a geo-extended Pandas dataframe from the GeoPandas library. It contains all information of the track in a flat dataframe format including a specific geometry column.

```
[2]: bbox = BboxSelector([
      7.601165771484375, # min_x
      51.94807412325402, # min_y
      7.648200988769531, # max_x
      51.97261482608728  # max_y
    ])

      # issue a query
      track_df = track_api.get_tracks(bbox=bbox, num_results=50) # requesting 50
      ↪ tracks inside the bbox
      track_df
```

```
[2]:
```

| | id | time | geometry \ |
|---|--------------------------|---------------------|--------------------------|
| 0 | 5e8b930965b80c5d6b4d7cd1 | 2020-03-07T12:33:15 | POINT (7.64069 51.95733) |
| 1 | 5e8b930965b80c5d6b4d7cd3 | 2020-03-07T12:33:20 | POINT (7.64118 51.95712) |

| | | | |
|-----|--------------------------|---------------------|--------------------------|
| 2 | 5e8b930965b80c5d6b4d7cd4 | 2020-03-07T12:33:26 | POINT (7.64162 51.95690) |
| 3 | 5e8b930965b80c5d6b4d7cd5 | 2020-03-07T12:33:31 | POINT (7.64210 51.95672) |
| 4 | 5e8b930965b80c5d6b4d7cd6 | 2020-03-07T12:33:36 | POINT (7.64264 51.95650) |
| .. | ... | ... | ... |
| 283 | 5dc986e844ea856b702e3e0b | 2019-10-28T16:34:55 | POINT (7.59523 51.96505) |
| 284 | 5dc986e844ea856b702e3e0c | 2019-10-28T16:35:00 | POINT (7.59425 51.96512) |
| 285 | 5dc986e844ea856b702e3e0d | 2019-10-28T16:35:05 | POINT (7.59327 51.96518) |
| 286 | 5dc986e844ea856b702e3e0e | 2019-10-28T16:35:10 | POINT (7.59225 51.96525) |
| 287 | 5dc986e844ea856b702e3e0f | 2019-10-28T16:35:15 | POINT (7.59123 51.96531) |

| | GPS PDOP.value | GPS PDOP.unit | Speed.value | Speed.unit | GPS Altitude.value \ |
|-----|----------------|---------------|-------------|------------|----------------------|
| 0 | 1.090631 | precision | 28.999999 | km/h | 110.381939 |
| 1 | 1.000000 | precision | 28.000000 | km/h | 108.260375 |
| 2 | 1.257198 | precision | 28.000001 | km/h | 105.826028 |
| 3 | 1.000000 | precision | 30.000000 | km/h | 104.395998 |
| 4 | 1.026727 | precision | 31.409419 | km/h | 101.516865 |
| .. | ... | ... | ... | ... | ... |
| 283 | 1.700000 | precision | 47.999999 | km/h | 109.652212 |
| 284 | 1.497088 | precision | 48.297297 | km/h | 110.122771 |
| 285 | 1.688911 | precision | 49.000001 | km/h | 110.573987 |
| 286 | 1.300000 | precision | 51.000000 | km/h | 111.140661 |
| 287 | 1.423253 | precision | 50.000001 | km/h | 111.891658 |

| | GPS Altitude.unit | GPS Bearing.value | ... Consumption.value \ |
|-----|-------------------|-------------------|-------------------------|
| 0 | m | 124.858622 | ... NaN |
| 1 | m | 125.020801 | ... NaN |
| 2 | m | 121.203960 | ... NaN |
| 3 | m | 123.412759 | ... NaN |
| 4 | m | 122.170479 | ... NaN |
| .. | ... | ... | ... |
| 283 | m | 276.419653 | ... 3.122268 |
| 284 | m | 276.271049 | ... 2.853618 |
| 285 | m | 275.808021 | ... 4.657916 |
| 286 | m | 275.411387 | ... 3.445271 |
| 287 | m | 276.124438 | ... 3.248333 |

| | Consumption.unit | track.appVersion | track.touVersion \ |
|-----|------------------|------------------|--------------------|
| 0 | NaN | NaN | NaN |
| 1 | NaN | NaN | NaN |
| 2 | NaN | NaN | NaN |
| 3 | NaN | NaN | NaN |
| 4 | NaN | NaN | NaN |
| .. | ... | ... | ... |
| 283 | l/h | NaN | NaN |
| 284 | l/h | NaN | NaN |
| 285 | l/h | NaN | NaN |
| 286 | l/h | NaN | NaN |

| | | | | | |
|-----|-----------|---------|----------|-----------|--------------------------------------|
| 287 | 1/h | NaN | NaN | | |
| | O2 Lambda | Voltage | ER.value | O2 Lambda | Voltage ER.unit MAF.value MAF.unit \ |
| 0 | | | NaN | | NaN NaN NaN NaN |
| 1 | | | NaN | | NaN NaN NaN NaN |
| 2 | | | NaN | | NaN NaN NaN NaN |
| 3 | | | NaN | | NaN NaN NaN NaN |
| 4 | | | NaN | | NaN NaN NaN NaN |
| .. | | | ... | ... | ... |
| 283 | | | NaN | | NaN NaN NaN NaN |
| 284 | | | NaN | | NaN NaN NaN NaN |
| 285 | | | NaN | | NaN NaN NaN NaN |
| 286 | | | NaN | | NaN NaN NaN NaN |
| 287 | | | NaN | | NaN NaN NaN NaN |

| | | | | |
|-----|-----------|---------------|-----------|--------------|
| | O2 Lambda | Voltage.value | O2 Lambda | Voltage.unit |
| 0 | | NaN | | NaN |
| 1 | | NaN | | NaN |
| 2 | | NaN | | NaN |
| 3 | | NaN | | NaN |
| 4 | | NaN | | NaN |
| .. | | ... | ... | |
| 283 | | NaN | | NaN |
| 284 | | NaN | | NaN |
| 285 | | NaN | | NaN |
| 286 | | NaN | | NaN |
| 287 | | NaN | | NaN |

[9944 rows x 54 columns]

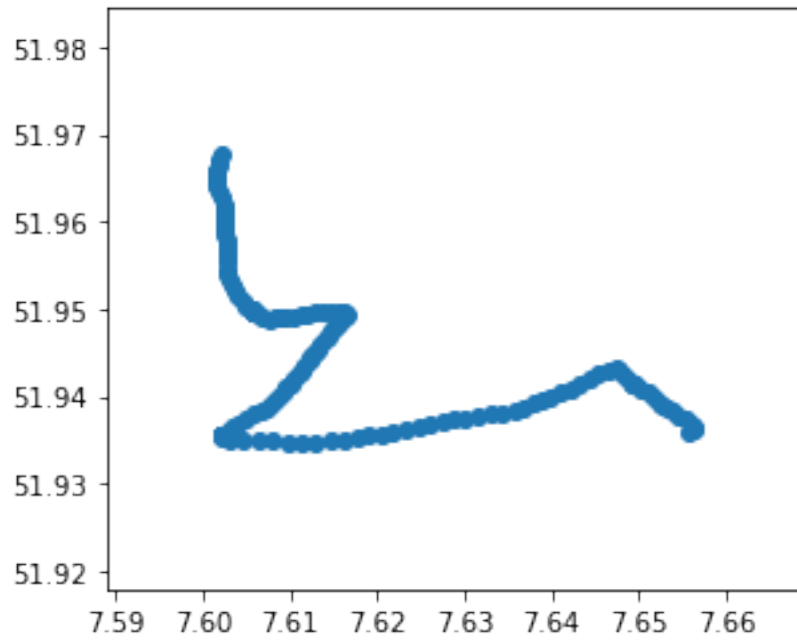
```
[3]: track_df.plot(figsize=(8, 10))
```

```
[3]: <matplotlib.axes._subplots.AxesSubplot at 0x12038d090>
```

3 Inspecting a single Track

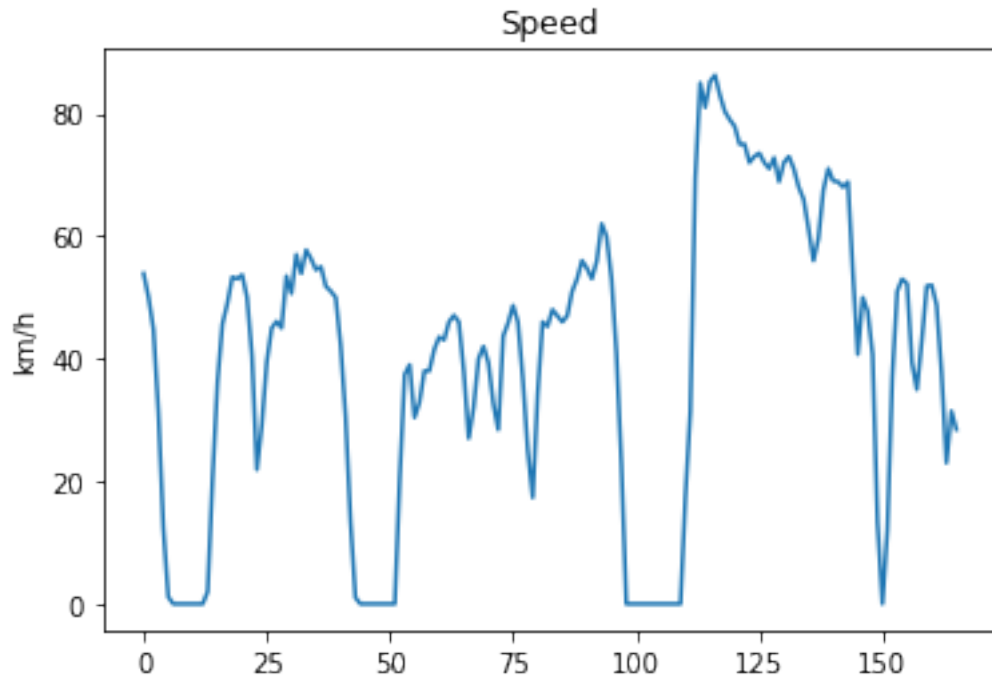
```
[4]: some_track_id = track_df['track.id'].unique()[5]
some_track = track_df[track_df['track.id'] == some_track_id]
some_track.plot()
```

```
[4]: <matplotlib.axes._subplots.AxesSubplot at 0x120cd9650>
```



```
[5]: ax = some_track['Speed.value'].plot()  
ax.set_title("Speed")  
ax.set_ylabel(some_track['Speed.unit'][0])  
ax
```

```
[5]: <matplotlib.axes._subplots.AxesSubplot at 0x120d36a50>
```



3.1 Interactive Map

The following map-based visualization makes use of folium. It allows to visualize geospatial data based on an interactive leaflet map. Since the data in the GeoDataframe is modelled as a set of Point instead of a LineString, we have to manually create a polyline

```
[6]: import folium

lats = list(some_track['geometry'].apply(lambda coord: coord.y))
lngs = list(some_track['geometry'].apply(lambda coord: coord.x))

avg_lat = sum(lats) / len(lats)
avg_lngs = sum(lngs) / len(lngs)

m = folium.Map(location=[avg_lat, avg_lngs], zoom_start=13)
folium.PolyLine([coords for coords in zip(lats, lngs)], color='blue').add_to(m)
m
```

```
[6]: <folium.folium.Map at 0x121ec5a50>
```

4 Example: Visualization with pydeck (deck.gl)

The pydeck library makes use of the basemap tiles from Mapbox. In case you want to visualize the map with basemap tiles, you need to register with MapBox, and configure a specific access token.

The service is free until a certain level of traffic is exceeded.

You can either configure it via your terminal (i.e. `export MAPBOX_API_KEY=<mapbox-key-here>`), which pydeck will automatically read, or you can pass it as a variable to the generation of pydeck (i.e. `pdk.Deck(mapbox_key=<mapbox-key-here>, ...)`).

```
[11]: import pydeck as pdk

# for pydeck the attributes have to be flat
track_df['lat'] = track_df['geometry'].apply(lambda coord: coord.y)
track_df['lng'] = track_df['geometry'].apply(lambda coord: coord.x)
vis_df = pd.DataFrame(track_df)
vis_df['speed'] = vis_df['Speed.value']

# omit unit columns
vis_df_cols = [col for col in vis_df.columns if col.lower()[len(col)-4:
    ↳len(col)] != 'unit']
vis_df = vis_df[vis_df_cols]

layer = pdk.Layer(
    'ScatterplotLayer',
    data=vis_df,
    get_position='[lng, lat]',
    auto_highlight=True,
    get_radius=10,          # Radius is given in meters
    get_fill_color='[speed < 20 ? 0 : (speed - 20)*8.5, speed < 50 ? 255 : 255,
    ↳ (speed-50)*8.5, 0, 140]', # Set an RGBA value for fill
    pickable=True
)

# Set the viewport location
view_state = pdk.ViewState(
    longitude=7.5963592529296875,
    latitude=51.96246168188569,
    zoom=12,
    min_zoom=10,
    max_zoom=25,
    pitch=40.5,
    bearing=-27.36)

r = pdk.Deck(
    width=200,
    layers=[layer],
    initial_view_state=view_state,
```

```

    mapbox_key="pk.
↪eyJ1IjoicGFyaWNoYXkiLCJhIjoiY2s1bDAzdnFwMDd4YTlnbno3aTcwaDd5aiJ9.
↪NeEhX7ksiQ8QlhkgtsIZXw"
)
r.to_html('tracks_muenster.html', iframe_width=900)

```

<IPython.lib.display.IFrame at 0x123e13090>

```
[11]: '/Users/parichay/Desktop/Desktop/Academic/Semester 2/FCDA/envirocar-
py/examples/tracks_muenster.html'
```

```
[13]: vis_df['consumption'] = vis_df['Consumption.value']

view = pdk.ViewState(
    longitude=7.5963592529296875,
    latitude=51.96246168188569,
    zoom=12,
    min_zoom=10,
    max_zoom=25,
    pitch=40.5,
    bearing=-27.36)

consumption = pdk.Layer(
    "HeatmapLayer",
    data=vis_df,
    opacity=0.9,
    get_position=["lng", "lat"],
    aggregation='"MEAN"',
    threshold=.5,
    get_weight="consumption",
    pickable=True,
)

r = pdk.Deck(
    width=200,
    layers=[consumption],
    initial_view_state=view,
    mapbox_key="pk.
↪eyJ1IjoicGFyaWNoYXkiLCJhIjoiY2s1bDAzdnFwMDd4YTlnbno3aTcwaDd5aiJ9.
↪NeEhX7ksiQ8QlhkgtsIZXw"
)

r.to_html("consumption_muenster.html", iframe_width=900)

```

```
<IPython.lib.display.IFrame at 0x121d06a10>
```

```
[13]: '/Users/parichay/Desktop/Desktop/Academic/Semester 2/FCDA/envirocar-  
py/examples/consumption_muenster.html'
```

4.1 Problem I faced

When I was trying run the notebook it was showing an “pandas” attribute error for “json_normalize”. I did everything correct but still I was unable to solve this problem. It took my valuable hours to find out the actual problem.

4.2 How I overcame

Later on I found that problem actually relies on the version of “pandas”. Upgrading “pandas” solved my problem.

4.3 Result

Modified Result