

janakparajuli_api_request_deckgl

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Assignment I Floating Car Data Analytics

1 Package loading and basic configurations

```
[5]: %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.

```
[6]: #bbox = BboxSelector([
      #     7.501165771484380, # min_x, min longitude
      #     51.94807412325402, # min_y, min latitude
      #     7.548200988769531, # max_x, max longitude
      #     51.97261482608728 # max_y, max latitude
      # ])
      bbox = BboxSelector([
          7.318136, 51.802163, 7.928939, 52.105665
      ])
      #7.318136, 51.802163, 7.928939, 52.105665
```

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

```
[6]:
```

	id	time	geometry \
0	5e8baea465b80c5d6b4dbfbf	2020-04-06T20:43:35	POINT (7.65079 51.95400)
1	5e8baea465b80c5d6b4dbfc1	2020-04-06T20:43:40	POINT (7.65079 51.95412)
2	5e8baea465b80c5d6b4dbfc2	2020-04-06T20:43:45	POINT (7.65083 51.95435)
3	5e8baea465b80c5d6b4dbfc3	2020-04-06T20:43:50	POINT (7.65086 51.95463)
4	5e8baea465b80c5d6b4dbfc4	2020-04-06T20:43:55	POINT (7.65090 51.95480)
..
63	5e08bc785bc8db42896408b7	2019-12-21T11:56:15	POINT (7.64402 51.97021)
64	5e08bc785bc8db42896408b8	2019-12-21T11:56:20	POINT (7.64402 51.97020)
65	5e08bc785bc8db42896408b9	2019-12-21T11:56:25	POINT (7.64402 51.97020)
66	5e08bc785bc8db42896408ba	2019-12-21T11:56:30	POINT (7.64404 51.97018)
67	5e08bc785bc8db42896408bb	2019-12-21T11:56:36	POINT (7.64404 51.97018)

	GPS Altitude.value	GPS Altitude.unit	GPS Bearing.value	GPS Bearing.unit \
0	100.237808	m	337.001680	deg
1	102.772222	m	11.636667	deg
2	104.020541	m	6.089730	deg
3	103.999999	m	4.503939	deg
4	104.000001	m	7.967200	deg
..
63	110.000003	m	0.000000	deg
64	109.999997	m	0.000000	deg
65	109.554884	m	150.086107	deg
66	111.000000	m	0.000000	deg
67	111.000003	m	0.000000	deg

	Throttle Position.value	Throttle Position.unit	Speed.value	...	\
0	16.283688	%	6.000000	...	
1	17.920277	%	14.260606	...	
2	16.000000	%	23.999999	...	
3	16.000000	%	21.000001	...	
4	16.000000	%	3.000000	...	
..	
63	15.000000	%	3.000000	...	
64	16.663317	%	0.000000	...	
65	15.000000	%	2.000000	...	
66	15.000000	%	0.000000	...	
67	NaN	NaN	NaN	...	

	sensor.constructionYear	sensor.manufacturer	track.appVersion \
0	2007	Dodge	NaN
1	2007	Dodge	NaN

2	2007	Dodge	NaN
3	2007	Dodge	NaN
4	2007	Dodge	NaN
..
63	2007	Dodge	NaN
64	2007	Dodge	NaN
65	2007	Dodge	NaN
66	2007	Dodge	NaN
67	2007	Dodge	NaN

	track.touVersion	02 Lambda Voltage.value	02 Lambda Voltage.unit	\
0	NaN	NaN	NaN	
1	NaN	NaN	NaN	
2	NaN	NaN	NaN	
3	NaN	NaN	NaN	
4	NaN	NaN	NaN	
..	
63	NaN	NaN	NaN	
64	NaN	NaN	NaN	
65	NaN	NaN	NaN	
66	NaN	NaN	NaN	
67	NaN	NaN	NaN	

	MAF.value	MAF.unit	02 Lambda Voltage	ER.value	02 Lambda Voltage	ER.unit
0	NaN	NaN		NaN		NaN
1	NaN	NaN		NaN		NaN
2	NaN	NaN		NaN		NaN
3	NaN	NaN		NaN		NaN
4	NaN	NaN		NaN		NaN
..
63	NaN	NaN		NaN		NaN
64	NaN	NaN		NaN		NaN
65	NaN	NaN		NaN		NaN
66	NaN	NaN		NaN		NaN
67	NaN	NaN		NaN		NaN

[16254 rows x 54 columns]

```
[7]: print(track_df.describe()) #Summary statistics of numeric column
```

	GPS Altitude.value	GPS Bearing.value	Throttle Position.value	\
count	16254.000000	15953.000000	14669.000000	
mean	91.724343	145.922260	26.973201	
std	25.261047	109.631387	18.528991	
min	30.999999	-2.304270	10.000000	
25%	78.435980	40.707150	16.000000	
50%	97.203334	149.100006	21.368312	

75%	105.000002	226.992945	27.165468
max	195.999997	363.849744	89.000003

	Speed.value	GPS PDOP.value	Intake Temperature.value	GPS VDOP.value \
count	15233.000000	13825.000000	14669.000000	13825.000000
mean	76.564161	1.066603	11.733752	0.843555
std	43.948642	0.367407	7.043985	0.307673
min	0.000000	0.800000	3.000000	0.600000
25%	43.000000	0.900000	7.000000	0.700000
50%	79.999998	1.000000	9.000000	0.800000
75%	118.000000	1.100000	15.000000	0.808105
max	373.333340	9.975758	37.999999	8.578788

	GPS Speed.value	Intake Pressure.value	Calculated MAF.value	... \
count	16254.000000	14668.000000	12886.000000	...
mean	74.921288	66.919684	23.172532	...
std	44.784828	32.065002	14.963303	...
min	0.000000	16.000000	-7.269221	...
25%	40.227443	44.544775	8.846632	...
50%	77.546974	65.784226	21.190572	...
75%	117.807371	80.587479	35.552383	...
max	174.567824	255.000000	72.078894	...

	Rpm.value	GPS HDOP.value	GPS Accuracy.value	Engine Load.value \
count	15233.000000	13825.000000	16254.000000	15233.000000
mean	2186.558264	0.583288	2.764771	45.636294
std	949.642333	0.235511	2.055794	26.821396
min	-859.118241	0.400000	1.000000	-495.792866
25%	1482.735338	0.400000	1.500000	27.022249
50%	2056.895271	0.600000	2.000000	47.058823
75%	3125.371775	0.639722	3.564713	65.840351
max	4530.827519	5.444747	45.526076	553.634987

	track.length	sensor.engineDisplacement	sensor.constructionYear \
count	16254.000000	16254.000000	16254.000000
mean	114.723543	1741.772917	2008.812477
std	87.325675	199.288250	4.299975
min	0.000000	1328.000000	1999.000000
25%	15.404851	1798.000000	2007.000000
50%	161.712887	1798.000000	2007.000000
75%	171.928734	1798.000000	2007.000000
max	233.951996	2461.000000	2019.000000

	O2 Lambda Voltage.value	MAF.value	O2 Lambda Voltage ER.value
count	1046.000000	1783.000000	1046.000000
mean	0.605838	24.633573	1.718217
std	0.344659	21.500357	0.302982
min	0.000000	1.733910	0.995972

25%	0.358227	10.591114	1.476118
50%	0.521146	19.779265	1.809511
75%	0.836780	31.661124	1.999969
max	1.270696	240.804784	1.999970

[8 rows x 22 columns]

```
[8]: print(track_df.describe(include=['object'])) #Summary statistic of non-numeric
      ↪column
```

	id	time	GPS Altitude.unit	\
count	16254	16254	16254	
unique	16254	16073	1	
top	5e08bc845bc8db428964257a	2020-04-06T10:02:46	m	
freq	1	2	16254	

	GPS Bearing.unit	Throttle Position.unit	Speed.unit	GPS PDOP.unit	\
count	15953	14669	15233	13825	
unique	1	1	1	1	
top	deg	%	km/h	precision	
freq	15953	14669	15233	13825	

	Intake Temperature.unit	GPS VDOP.unit	GPS Speed.unit	... sensor.type	\
count	14669	13825	16254	...	16254
unique	1	1	1	...	1
top	c	precision	km/h	...	car
freq	14669	13825	16254	...	16254

	sensor.model	sensor.id	sensor.fuelType	\
count	16254	16254	16254	
unique	6	7	2	
top	Caliber	58395f40e4b0a979d45bd61b	gasoline	
freq	11128	11128	14564	

	sensor.manufacturer	track.appVersion	\
count	16254	209	
unique	6	1	
top	Dodge	Version 1.0.2 (38), 30.11.79 00:00	
freq	11128	209	

	track.touVersion	O2 Lambda Voltage.unit	MAF.unit	\
count	209	1046	1783	
unique	1	1	1	
top	2013-10-01	V	l/s	
freq	209	1046	1783	

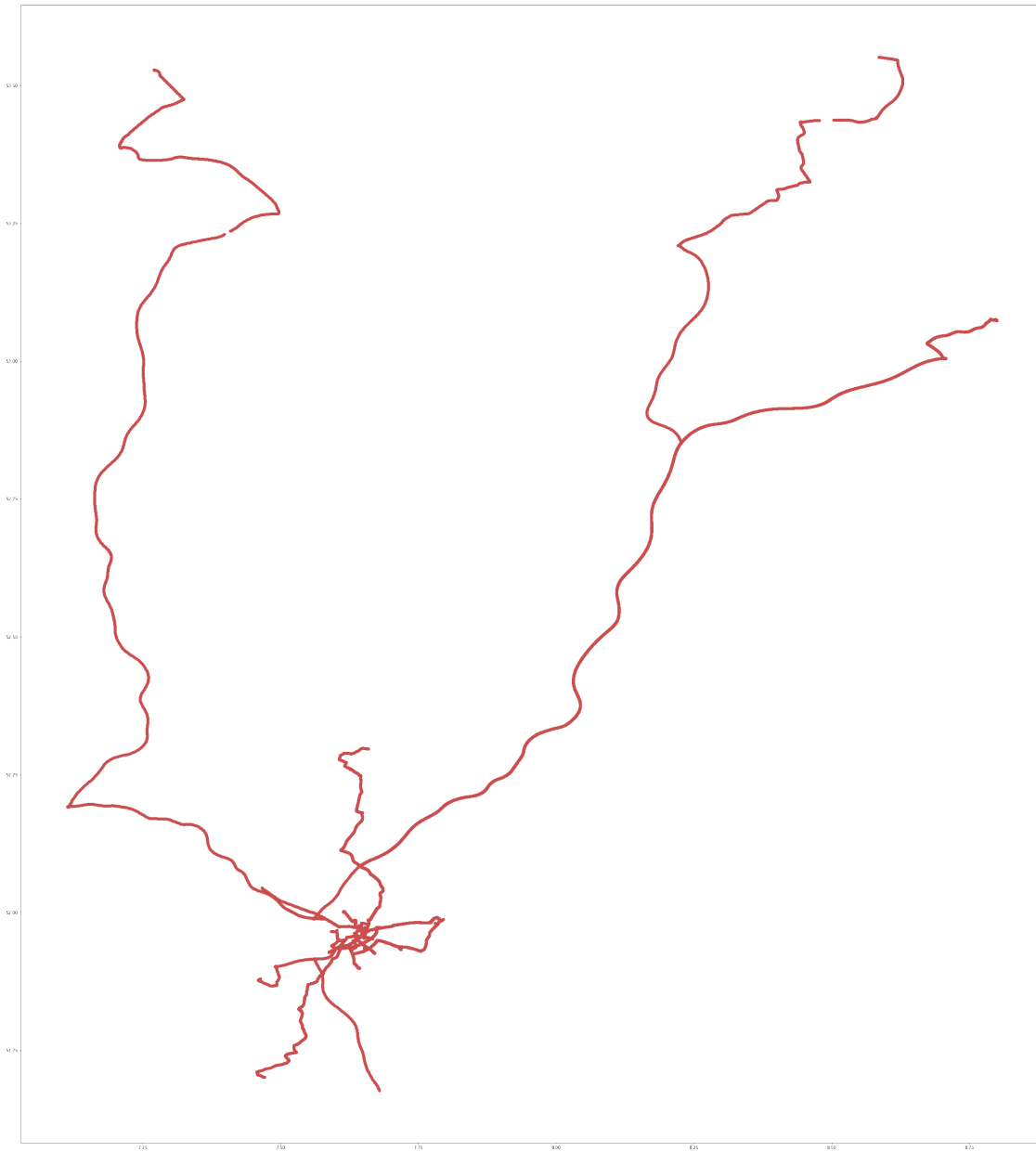
	O2 Lambda Voltage ER.unit
count	1046

```
unique          1
top             ratio
freq           1046
```

```
[4 rows x 31 columns]
```

```
[59]: track_df.plot(figsize=(260, 50), color=(0.8,0.3,0.3))
```

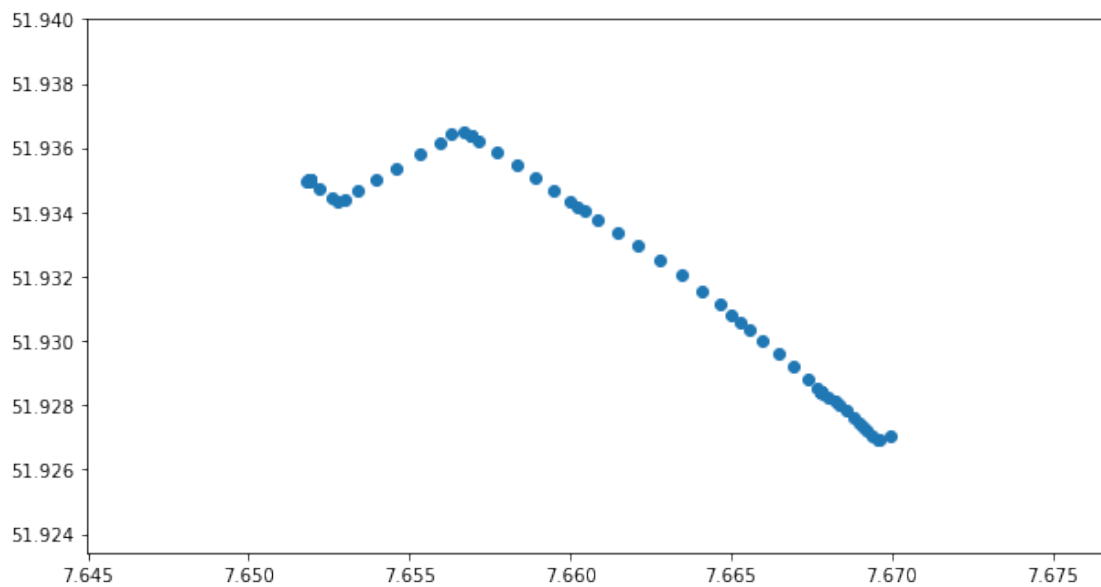
```
[59]: <matplotlib.axes._subplots.AxesSubplot at 0x23df3282848>
```



3 Inspecting a single Track

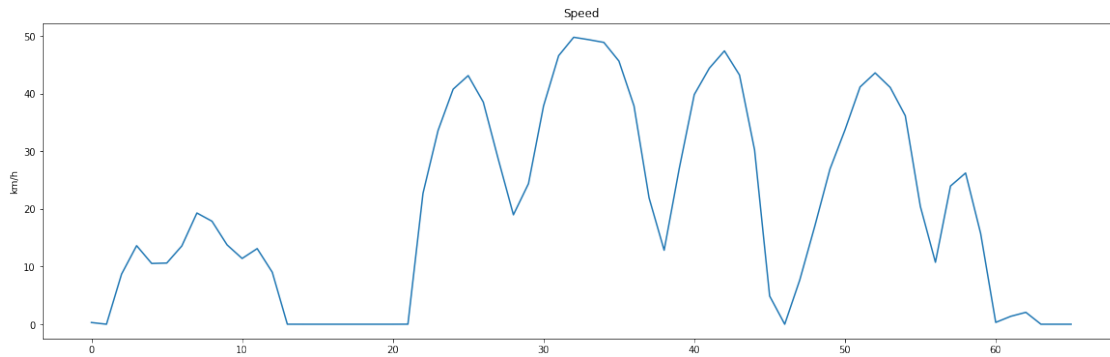
```
[42]: some_track_id = track_df['track.id'].unique()[10]
      #print(some_track_id)
      #print(track_df['track.id'] == some_track_id)
      #print("The false track df is:")
      some_track = track_df[track_df['track.id'] == some_track_id]
      #print("Now the some track is:")
      #print(some_track)
      some_track.plot(figsize = (10,20))
```

```
[42]: <matplotlib.axes._subplots.AxesSubplot at 0x23df16b7b88>
```



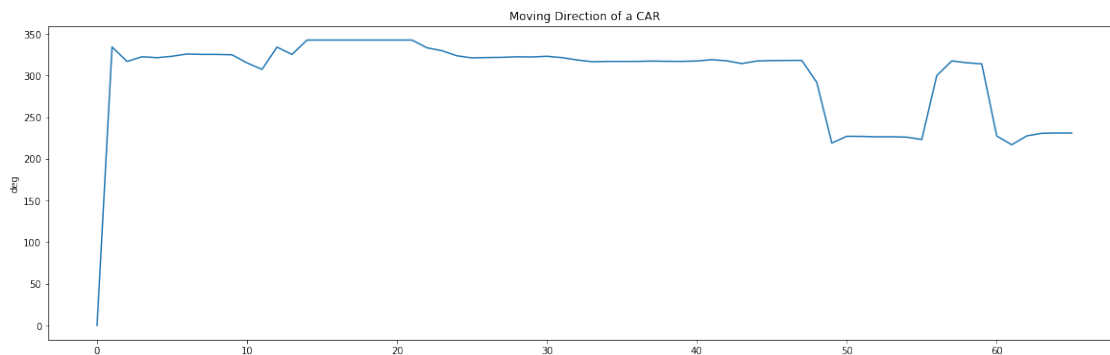
```
[18]: ax = some_track['GPS Speed.value'].plot(figsize=(20,6))
      ax.set_title("Speed")
      ax.set_ylabel(some_track['GPS Speed.unit'][0])
      #some_track['GPS Speed.value']
      ax
```

```
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x23dee3d5848>
```



```
[30]: bx = some_track['GPS Bearing.value'].plot(figsize=(20,6))
bx.set_title("Moving Direction of a CAR")
bx.set_ylabel(some_track['GPS Bearing.unit'][0])
bx
```

[30]: <matplotlib.axes._subplots.AxesSubplot at 0x23df0451308>

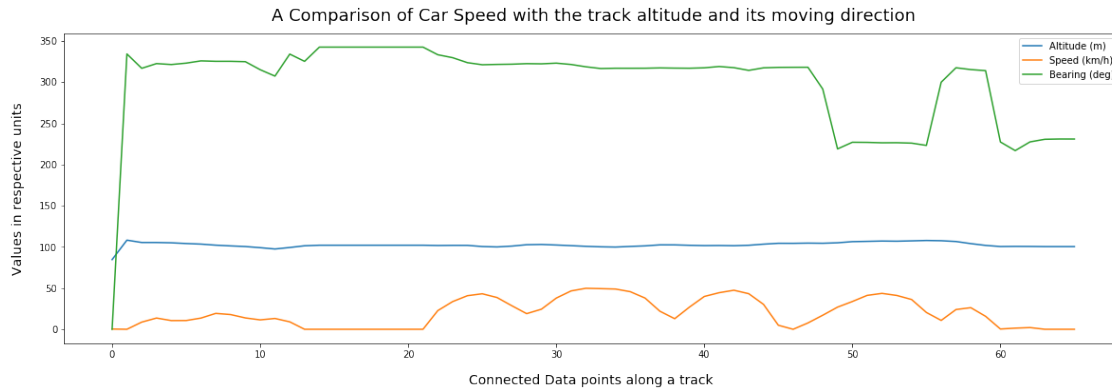


```
[29]: import matplotlib.pyplot as plt

#First prepare the data into dictionary form
df = pd.DataFrame({'Altitude (m)':some_track['GPS Altitude.value'], 'Speed (km/
→h)':some_track['GPS Speed.value'], 'Bearing (deg)':some_track['GPS Bearing.
→value']})

#Now plot the data
df.plot(figsize=(20,6))
plt.title("A Comparison of Car Speed with the track altitude and its moving_
→direction",y=1.02, fontsize=18)
plt.xlabel("Connected Data points along a track ", labelpad=15, fontsize=14)
plt.ylabel("Values in respective units", labelpad=15, fontsize=14)
```

[29]: Text(0, 0.5, 'Values in respective units')



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

```
[75]: import folium

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

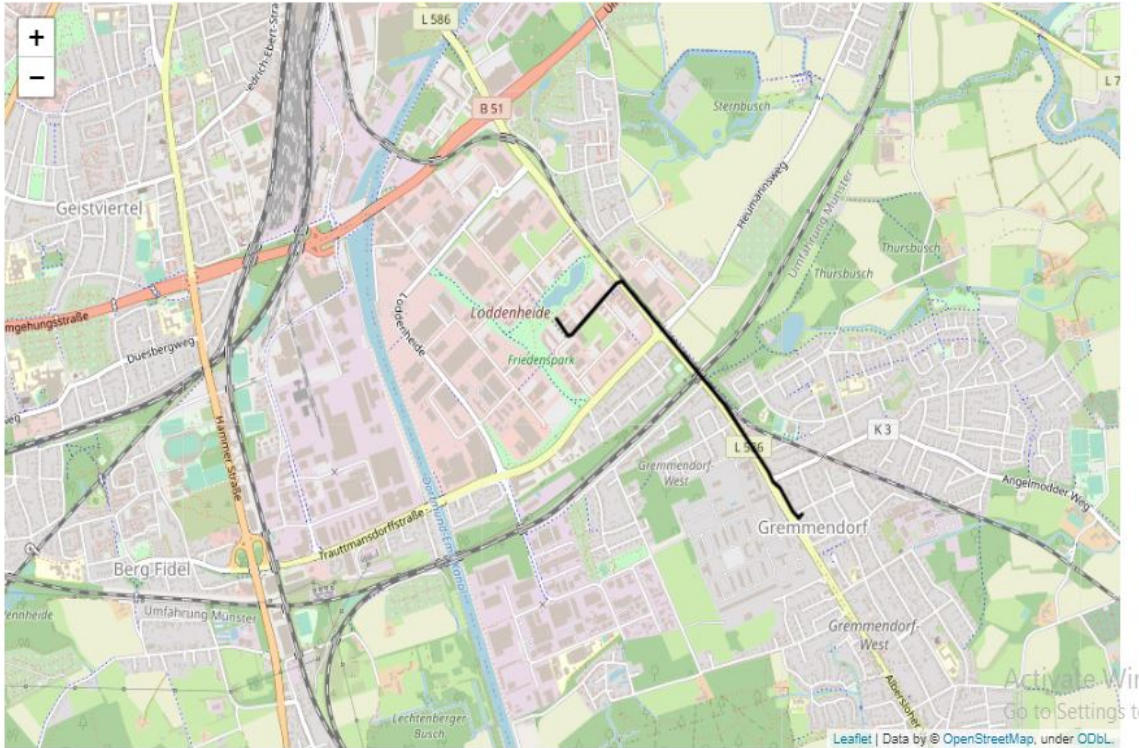
avg_lat = sum(lats) / len(lats)
avg_lngs = sum(lngs) / len(lngs)

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

```
[75]: <folium.folium.Map at 0x23dfa4d4ec8>
```

```
[75]:
```

Out[75]:



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>, ...)`).

```
[83]: 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['GPS 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,    #7.65079 51.95400
    longitude=7.65079,
    latitude=51.95400,
    zoom=13,
    min_zoom=5,
    max_zoom=18,
    pitch=40.5,
    bearing=-27.36)

r = pdk.Deck(
    width=200,
    layers=[layer],
    initial_view_state=view_state,
    mapbox_key="pk.
→eyJ1IjoiamFuYWtwYXJhanVsaSIsImEiOiJjaWdtMWd2eWUwMjRvdXJrcjVhbTFvcmszIn0.
→jRIRtmGcm5waI7RXih3t5A"
)
r.to_html('tracks_muenster.html', iframe_width=900, iframe_height = 500)

```

<IPython.lib.display.IFrame at 0x23dfa425ec8>

[83]: 'D:\\MSC_GeoTech\\Study_Materials\\Course\\Second_Semester\\Floating_Car_Project\\enviroCar\\envirocar-py\\examples\\tracks_muenster.html'

Brief description of your experience:- what went fine, where did you face problems and how did you overcome the problems?

My start of the assignment had to pay a lot of toil in installing the software and its packages. I installed Anaconda3 with ease and then tried to install the packages and dependencies for geopandas and envirocar. Even both of them were installed and shown by the command 'conda list' in anaconda powershell prompt, a problem called 'ModuleNotFoundError' would show whenever I tried to import the modules in Python terminal. I had to uninstall and reinstall the Anaconda software a lot of times before figuring out the problem. I uninstalled all other previously installed Python3, deleted its environment settings and removed from the registry editor also. I also uninstalled previously installed osgeo and removed gdal settings. Then, I had to manually install the

wheel files of the dependencies of geopandas (GDAL, Fiona, Pyproj, Rtree and Shapely) from a repository maintained by [Christoph Gohlke at the Laboratory for Fluorescence Dynamics at UC Irvine](#) . The steps followed was from this link: [Geoffboeing](#) .

Upon completion of the detailed instructions given in the link above, finally geopandas was successfully installed and imported. Then, envirocar package was successfully installed using the command 'pip install envirocar-py --upgrade'. It took more than two full days to figure it and sort out the problem. Happily, no further problems were faced in course of modification of the given project.

Screenprint(s) of the last page of the Notebook, presenting the result of your modification..

[]:

