

Fraunhofer Institute for Systems and Innovation Research ISI

October 2023

Truck Stop Locations in Europe – Documentation Version 1.0 – Public Version

Publishing Notes (1/2)

The following data are provided free of charge. The Fraunhofer ISI does not assume any liability for completeness, correctness and accuracy of the information. Coverage and completeness varies among countries.

The following geolocations (Longitude, Latitude) represent potential heavy-duty truck (HDT) parking locations in Europe (EU-27, EFTA, UK). These locations resulted from several data sources – mostly OpenStreetMap (OSM) and complemented with commercial data sources – and multiple filtering steps to ensure accuracy and select potential candidates. Individual nearby locations were then merged and clustered using the MeanShift algorithm. Thus, geolocations represent centroids of the respective cluster and may not match exact locations.

All locations should be considered as reference point for detailed local analyses of ambient conditions and truck parking suitability.

The indicated total area, along with the associated estimated number of parking spaces - provides an indication in terms of scale.

Publishing Notes (2/2)

This data are crucial for charging infrastructure or network operators to facilitate future low-carbon road freight traffic as battery electric trucks in long-haul operation will require public charging infrastructure. Ideally, some of today's truck parking locations will be equipped with charging infrastructure. Accordingly, megawatt charging would be useful next to major highways and lower charging power would be helpful in many locations for overnight charging.



Codebook (1/2)

Variable	Туре	Info
name	categorical	Indicates the most likely location type. 4 types are available. This also indicates the availability of surrounding infrastructure or services.
lat	float	Latitude information from the respective cluster centroid. WGS-84 format
lon	float	Longitude information from the respective cluster centroid. WGS-84 format
totalArea_m2	int	The estimated total area of this location in m². Most likely only publicly accessible areas. We note that not the whole area must be accessible for truck parking and that this may also include roads and other service areas. Empty if this information is not available for the respective location.
truckParkingConfidence	categorical	Indicates the confidence about whether or not this location is equipped or accessible for truck parking. 2 types are available.
country	categorical	Assigned country. This datasets covers the EU-27, EFTA and UK.
clc_Code	categorical	CORINE Land Cover information as defined under https://land.copernicus.eu/en/products/corine-land-cover
TenTcore_km	float	Calculated aerial distance in kilometers (to the TenTec Core road network as provided by the European Commission – DG MOVE – TENtec Information System 2022. Precision level: around 0.1 decimal degress (~11 km). Empty if above.
TenTcomp_km	float	Calculated aerial distance in kilometers to the TenTec Comprehensive road network as provided by the European Commission – DG MOVE – TENtec Information System 2022 Precision level: around 0.1 decimal degress (~11 km). Empty if above.
source	categorical	Indicates the origin of the information. 3 types are available.



Codebook (2/2)

Variable: name - most likely location type (mixed types may exist)

Types	Name	Info
Type 1	Truck Stop / Rest Area	Most likely a truck stop and service area, directly along the highways. Fuel for HDTs is provided. Other services such as restaurants, service facilities, sleeping and shower facilities are likely. Public access usually provided
Type 2	Fueling (and Truck Stop)	Most likely a truck stop and service area. Fuel for HDTs is provided. Other services may be limited. Public access usually provided
Type 3	Rest Area	A rest area - usually along a major highway – that provides parking areas and likely provides other service facilities such as restaurants, shops, or at least restrooms. Public access usually provided
Type 4	Parking	Other parking areas that are usually located close to industrial areas, but additional information is missing. Restricted access possible.

Variable: truckParkingConfidence

Types	Name	Info
Type 1	High	Truck parking and access is very likely
Type 2	Medium	Truck parking and access is likely, but occasionally restricted or not available



Agenda

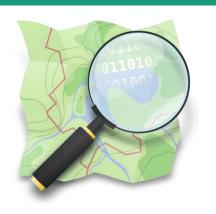
- 1. Overview & Sources
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- 5. Others



Overview data sources

Sources:

OSM



Idea: Combine both sources to enhance accuracy and coverage / completeness

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PTV Developer, HERE, TomTom

Issue: Limited validity and accuracy for truck suitability

Issue: covers street-side parking. No statement about dimensions, but suspected higher coverage and accuracy as commercial software.



OSM – Overview (1/2)

Tag overview

Relevant tags / osm objects: Where could truck parking happen? Which (public) areas may be suitable?

Parking Areas

Rest Areas

Confidential

Fueling Stations

osm object types: Which object types need to be differentiated?

- Points / Nodes:
 - Latitude / Longitude: As specified
- No area (=0 m2) Area:
- Ways / Polygons (2D shape):
 - Latitude / Longitude: Calculated via area-based centriod
 - Calculated based on area enclosed by the envelope GPS cords Area:
- Relation:
 - Group of nodes, ways, and/or relations



OSM – Overview (2/2)

Tag overview

Parking Areas

```
"capacity" in element["tags"]: capacity.append(element["tags"]["capacity"])
else: capacity.append("NA")
if "access" in element["tags"]: access.append(element["tags"]["access"])
else: access.append("NA")
if "hgv" in element["tags"]: hgv.append(element["tags"]["hgv"])
else: hgv.append("NA")
if "landuse" in element["tags"]: landuse.append(element["tags"]["landuse"])
else: landuse.append("NA")
if "operator:type" in element["tags"]: optype.append(element["tags"]["operator:type"])
if "name" in element["tags"]: name.append(element["tags"]["name"])
else: name.append("NA")
if "capacity:hgv" in element["tags"]: hgvCapac.append(element["tags"]["capacity:hgv"])
else: hgvCapac.append("NA")
if "hgv:lanes" in element["tags"]: hgvLane.append(element["tags"]["hgv:lanes"])
else: hgvLane.append("NA")
```

- Nodes (osm total): 391,854 coverage 7.7%*
- Ways (osm total): 4,631,651 coverage 91%*

99% coverage of all osm data ensured

Rest Areas

```
if "name" in element["tags"]: name.append(element["tags"]["name"])
else: name.append("NA")
if "hgv" in element["tags"]: hgv.append(element["tags"]["hgv"])
else: hgv.append("NA")
```

- Nodes (osm total): 21,041 -> coverage 56%*
- Ways (osm total): 16,542 -> coverage 43%*
- Relation (osm total): 323 coverage 0.85%*

99% coverage of all osm data ensured

Fueling Stations

```
ame" in element["tags"]: name.append(element["tags"]["name"])
    name.append("NA")
  "fuel:HGV_diesel" in element["tags"]: hgvFuel.append(element["tags"]["fuel:HGV_diesel"])
 "brand" in element["tags"]: brand.append(element["tags"]["brand"])
lse: brand.append("NA")
 "fuel:diesel" in element["tags"]: diesel.append(element["tags"]["fuel:diesel"])
lse: diesel.append("NA")
 "capacity:hgv" in element["tags"]: hgvCapac.append(element["tags"]["capacity:hgv"])
 "hqv" in element["tags"]: hgv.append(element["tags"]["hqv"])
se: hgv.append("NA")
  "hgv:lanes" in element["tags"]: hgvLane.append(element["tags"]["hgv:lanes"])
lse: hgvLane.append("NA")
```

- Nodes (osm total): 306,134 -> coverage 60%*
- Ways (osm total): 203,093 -> coverage 40%*
- Relation (osm total): 2,461 coverage 0.5%*

almost 100% coverage of all osm data ensured

High coverage ensured for all OSM data. Data for all EU27 (+ EFTA and UK) – Data extracion via OverPass API

Confidential

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Agenda

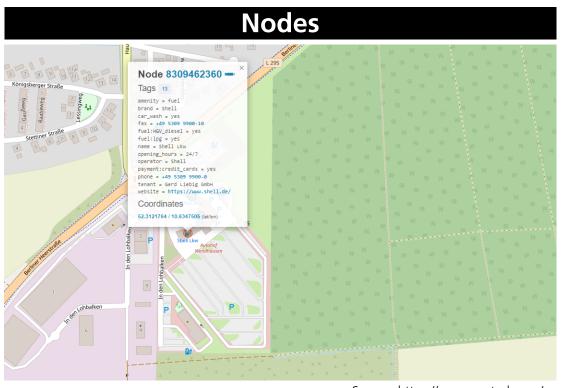
1. Overview & Sources

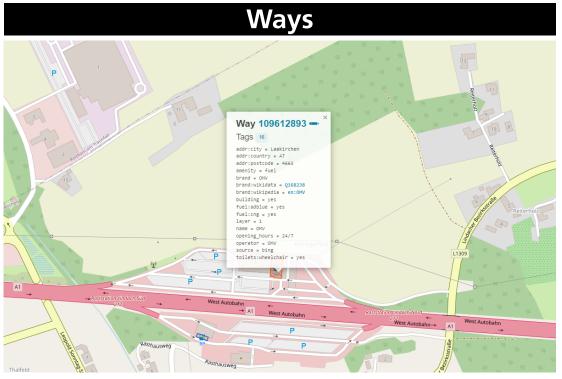
- 2. Data
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OSM Fueling Stations – Overview Nodes / Ways

Fueling Stations





Source: https://overpass-turbo.eu/

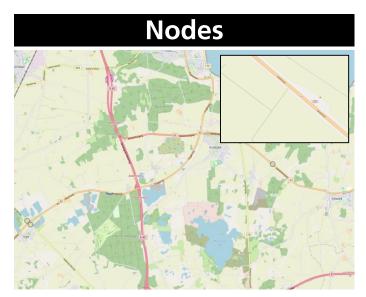
Source: https://overpass-turbo.eu/

Both tags may be relevant for fueling stations

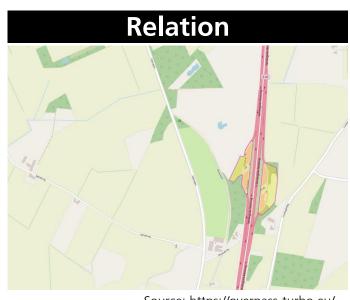
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OSM Rest Areas – Overview Nodes / Ways

Rest Areas



Source: https://overpass-turbo.eu/



Source: https://overpass-turbo.eu/

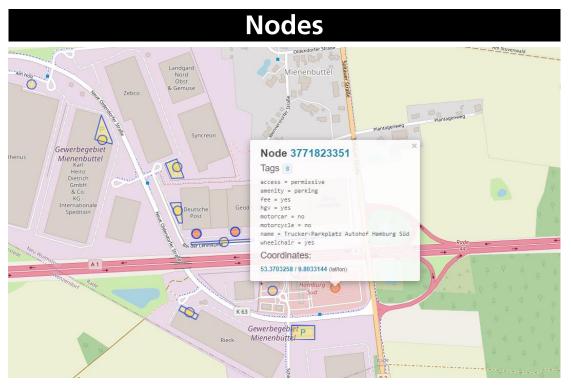


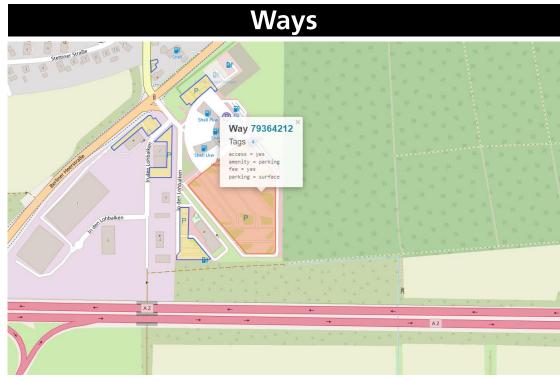
Source: https://overpass-turbo.eu/

All three tags may be relevant for (truck) rest areas

OSM Parking Areas – Overview Nodes / Ways

Parking Areas





Source: https://overpass-turbo.eu/

Source: https://overpass-turbo.eu/

Both tags may be relevant for truck parking areas

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Filtered for truck accessibility

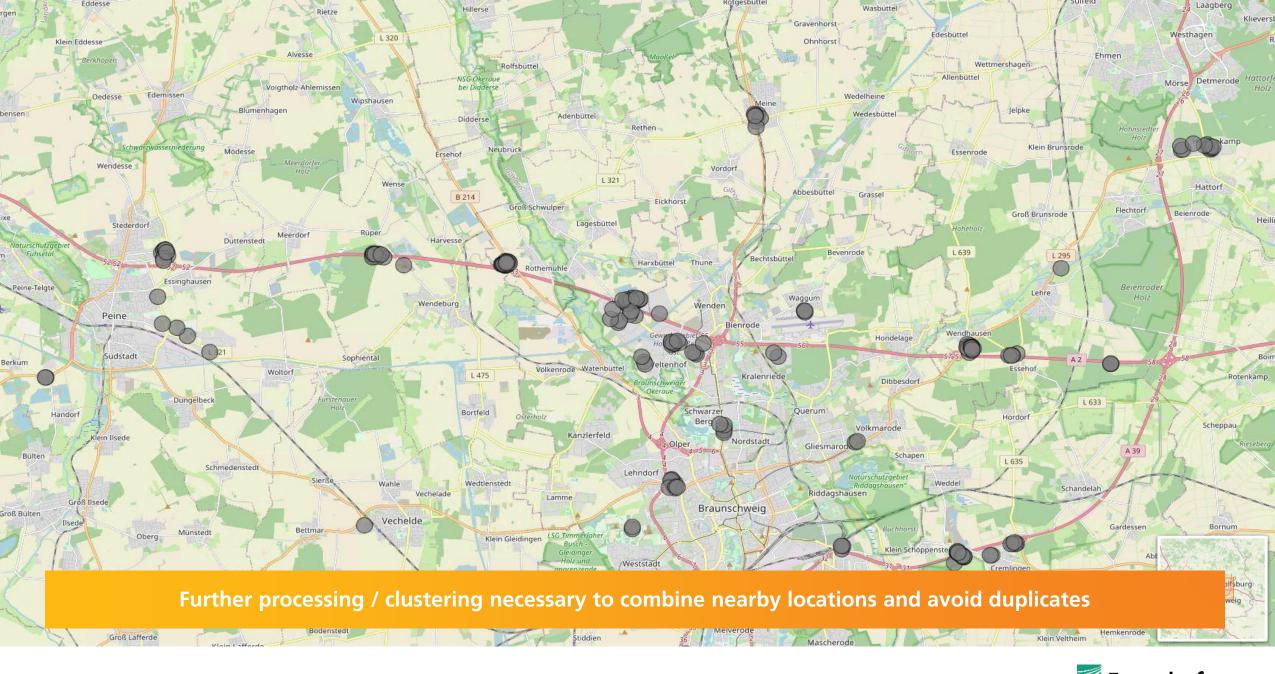
Interim data – Results from pre-processing and filtering

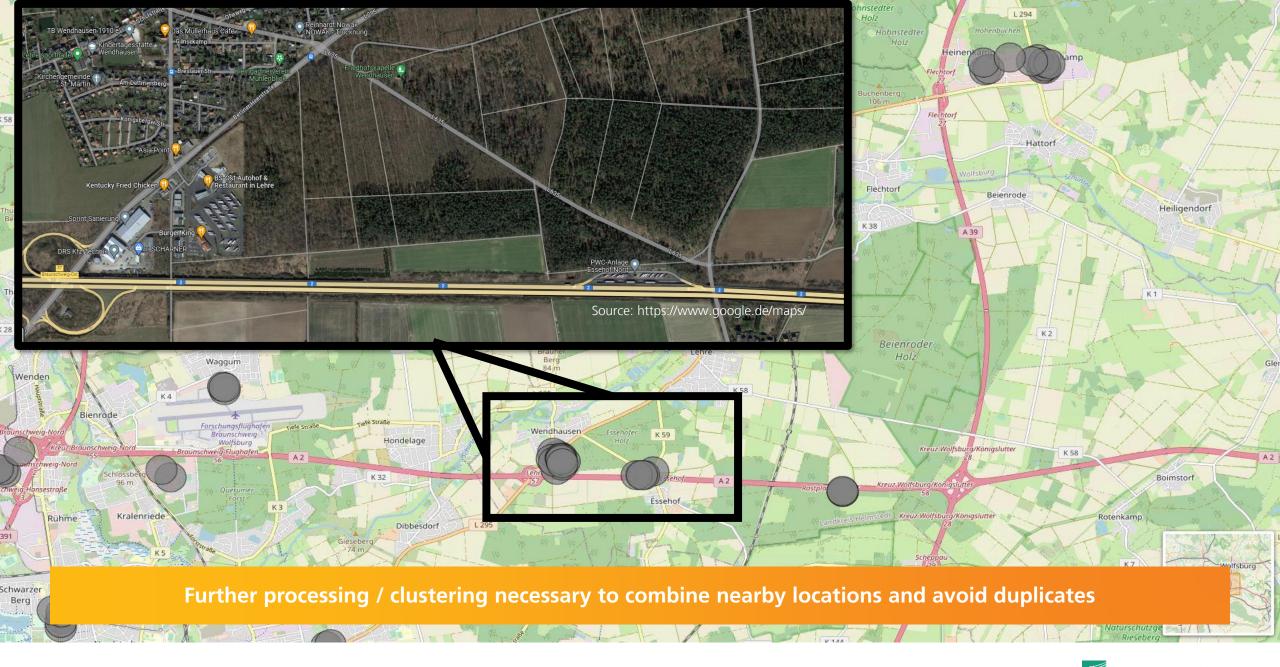
	Parking Areas	Rest Areas	Fueling Stations	Other Truck Data
Initial size	2,661,731	18,279	131,374	41,155
Final data	78,906 (3%)	8,337 (46%)	19,627 (15%)	41,155
Yes	7,565 (10%)	3,293 (40%)	10,541 (54%)	41,155 (100%)
Likely	12,754 (16%)	2,355 (28%)	6,153 (31%)	-
Insecure	53,275 (68%)	2,689 (32%)	2,933 (15%)	-
Private/Delivery	5,312 (6%)	-	-	-

The final data covers ∑ 142,654 data points for potential locations









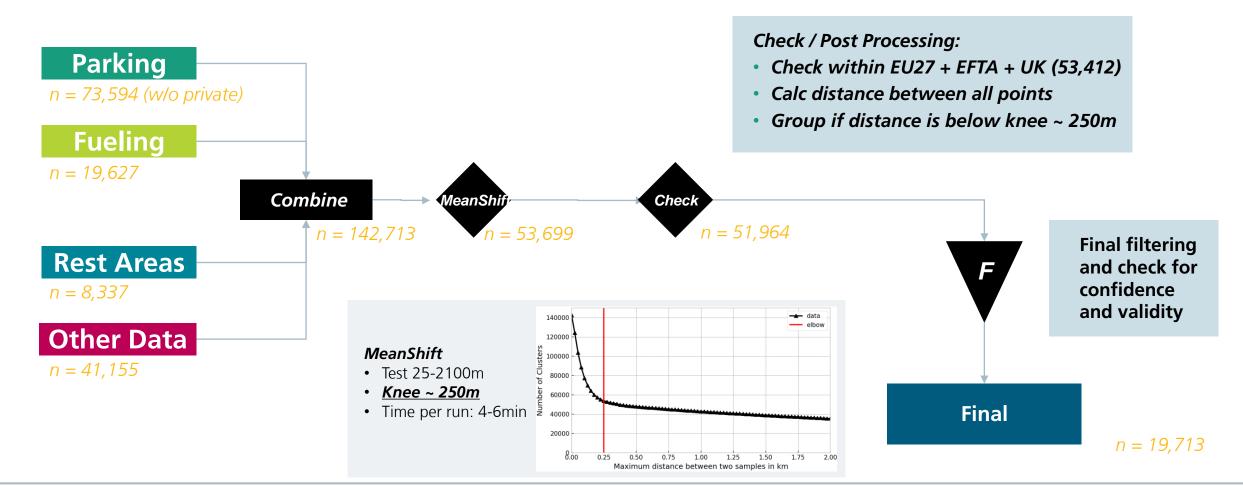
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Methods – Overview (1/2)





Methods – Overview (2/2)

MeanShift Clustering

Centroid-based algorithm based on kernel density estimation (KDE)

Pros: Variable number of centroids, robust to outliers, Universal application, no limitations on prior shape or data distribution, single parameter model (bandwith)

Cons: Bandwidth-sensitive output, Non-trivial bandwidth selection, computationally (relatively) expensive

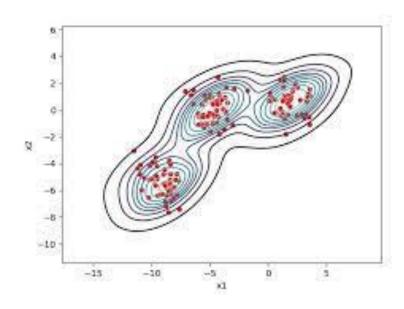
Examples on geo-coords:

http://dx.doi.org/10.1145/1631272.1631292 https://ceur-ws.org/Vol-2649/paper5.pdf https://onlinelibrary.wiley.com/doi/epdf/10.111 1/j.1475-4754.2010.00560.x

Steps

- a) Kernel Density Estimation: The first step involves estimating the underlying probability density function (PDF) of the data points. This is typically done using kernel density estimation, where each data point is represented by a kernel function centered at that point. The kernel function specifies the weight assigned to each data point in the density estimation process.
- b) Shifting Data Points: In the second step, the algorithm iteratively shifts the data points towards regions of higher density. The shift is determined by calculating the mean shift vector for each data point, which represents the direction and magnitude of the shift. The mean shift vector is calculated as the weighted average of the differences between the data point and its neighboring points, where the weights are determined by the kernel function.
- c) Convergence and Cluster Identification: The algorithm continues shifting the data points until convergence is reached. Convergence occurs when the mean shift vectors become very small or negligible. Once convergence is achieved, the final position of each data point represents a cluster center. The algorithm assigns each data point to the closest cluster center, thereby identifying the clusters within the data.

Mean Shift



Source: https://ml-explained.com/blog/mean-shift-explained

Source: https://medium.com/@shruti.dhumne/mean-shift-clustering-a-powerful-technique-for-data-analysis-with-python-f0c26bfb808a



Interim results



Final filtering for each point whether there are only information from:

OSM: $n = 7 \text{ tag combinations } (\sim 25\%)$

 $n = 13 \text{ tag combinations } (\sim 25\%)$ Others:

 $n = 41 \text{ tag combinations } (\sim 50\%)$ Mixed:

Create final tags based on the joined information:

Type 1: <u>Parking</u> – if only information about parking / parking areas is available

Fueling – if information about fueling and potentially parking is available. Truck services possible. Type 2:

Rest Area – if information about rest areas is available Type 3:

Type 4: <u>Truck Stop / Rest Area</u> – if information about rest areas with fueling / service (truck stops) is available

Clustering creates 56 unique tag configurations from more than 53,000 potential locations



Final Filtering



For all types, check and evaluate each point based on:

available area information:

Yes, No, Minimum threshold

- **Proximity to the TenT network**
- **Land Cover information (Corinne CLC)**
- **Area access information**

and update to low confidence, medium confidence, high confidence

Parking					
Parking	High	464			
	Low	22663			
	Medium	4204			
Parking / Rest Area	High	525			

Rest Area						
Rest Area	High Low	4120 534				
	Medium	1804				

Fueling						
Fueling	High	329				
	Low	7742				
	Medium	5793				
Fueling / Truck Stop	High 1260					
Parking						
Truck Stop / Rest Area	a High Medium	275 111				

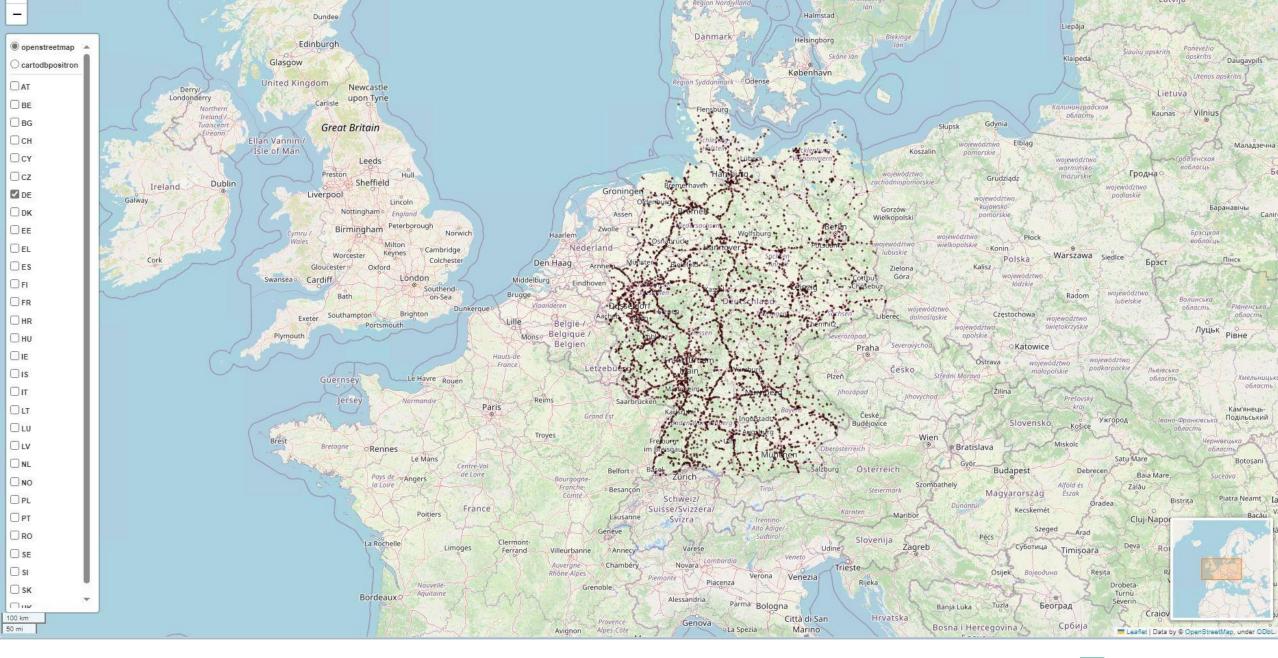
Keep only locations with HIGH or MEDIUM truck parking confidence: n = 19,713



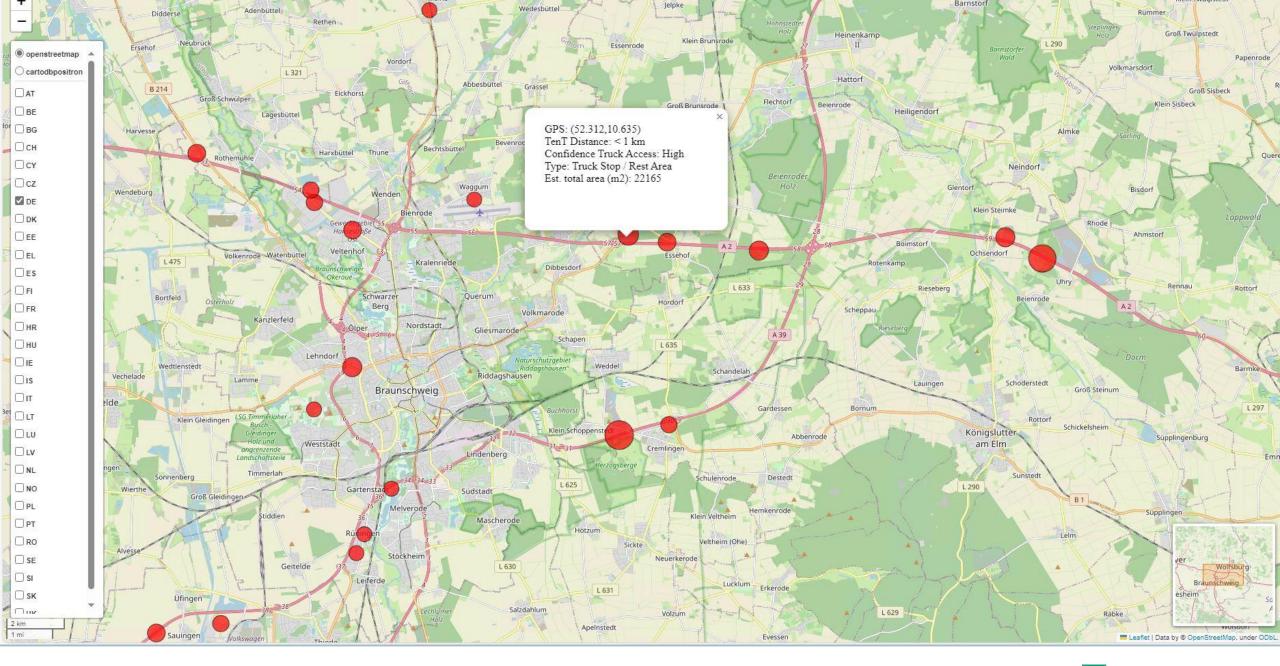
Results (1/2)

Country					Rest Area	Truck Stop / Rest Area	Total
FR	2039		782				
AT	76		78		174	32	443
BE	40	25	118	41	70	5	299
BG	22	5	23	2	88		142
СН	36	15	71	9	77	10	218
CY	1			1	2		4
CZ	187			12			383
DE	1099		1144	104	1589	55	
DK	32		93	9	302	13	
EE	33	1	19		11		67
EL	12		29		68		120
ES	235	20	196	8	283	21	763 586
FI	89	15	102	12	363	5	
HR	21	7	27	4	54		117
HU	48	13	67	4	90	2	224
IE	34	5	17		14	1	71
IS	1		1		8		10
IT	238	84	536	42	352	38	
LT	49	6	37	4	46	4	146
LU	11	2	14	3	17		47
LV	32	5	20		16		74
NL	60	48	123	32			345 976
NO	377	9	74	13	469		976
PL	1171	204	433	54	229	53	2144
PT	31	6	23	1	25	5	91
RO	96	5	53				1 303
SE	125	18	125	22	309		626
SI	29	2	14	2	24	3	74
SK	39	19	52	8	45		167
UK	130	52	288		346		844
Total	6393	1260	4668	525	6471	396	19713





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Area information (1/2)

Matching with NOW-Dataset and the stated number of truck parking spots per location (DE only)

Total NOW data: 2,271

without NA: 2,208

Number of matches: 1977 = 89.5% (within maximum 500m aerial distance)



Agenda

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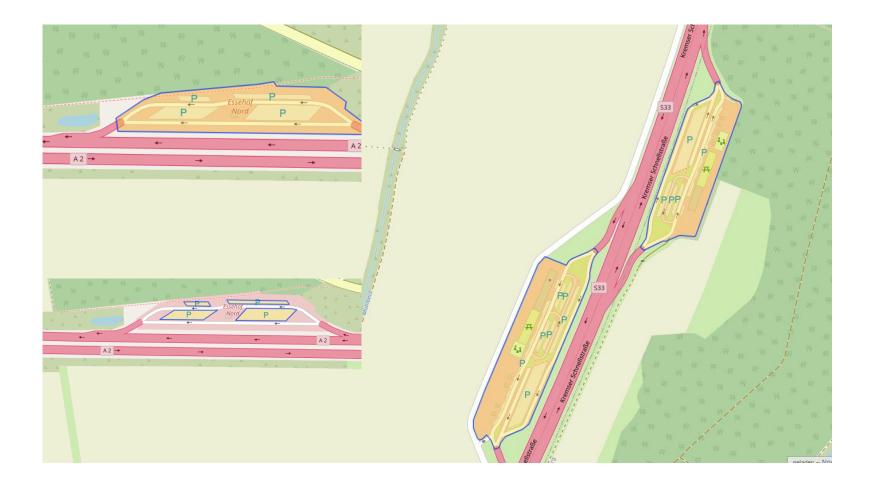


Particularities



Avoid double counting when calculating the total area per location:

- Test whether parking areas are part of the rest area
- If true, delete area information (=0m²)





Particularities



Areas defined as Polygon and Relation:

 filter based on "name" and remove dups to avoid area double accounting





Methods: DBSCAN

DBSCAN:

200 meter Knee:

Num. of clusters: 52,364

DBSCAN with stronger clustering effect (convex shapes) – decreases traceability, may increase data alienation.

MeanShift:

Knee: 250 meter

Num. of clusters: 53,699

