



AI-Driven Strategies for Optimal EV Charging Station Deployment

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1. Situations

- EVs as a solution for reducing greenhouse gas emission
- Increasing the usage of EVs and the growth in the sales market
- Challenges in deploying Charging infrastructure

2. AI-driven approach

- Optimizing strategies for the deployment of EV charging stations
- Pave the way to the integration with renewable energy sources

1. Objectives

- Explore AI methods for planning charging station deployment
- Identifying key influencing factor

2. Challenges

- Limited Parking availability
- Variety of traffic flow patterns
- Energy distribution (various energy or electricity capacity)

Exploring Standards, Metrics and Case Studies

Key Questions:

- What is utilization rate and why is it important?
- What is efficiency score ?
- Why the standards are important ?
- How and where did we collect the data ?

Charger Utilization Metrics

Utilization = **Active time** / **Total available time**

- **High utilization rates:** suggests adding **more** chargers
- **Low utilization rates:** Over-provisioning or **poor** placement

Efficiency score: Measures the ratio of energy delivered to vehicles versus the maximum capacity of the charger during a given period.

- **Low efficiency scores** - Suggest operational **inefficiencies** or **underused** stations
- **High efficiency scores** - Indicate **effective** energy delivery and charger utilization

ISO 15118: Communication Standards

- Communication protocols: EVCC ↔ SECC via cable or Wi-Fi.

Application OSI layer 7	ISO 15118-1 General information and use case definition (merged with contents of ISO 15118-6 for second edition)	ISO 15118-2	Application layer messages (V2G Message), SDP (SECC Discovery Protocol)			ISO 15118-4 Network and application protocol conformance tests
Presentation OSI layer 6		Network and application protocol requirements	EXI (Efficient XML Interchange)			
Session OSI layer 5		— and — ISO 15118-20	V2GTP (Vehicle-to-Grid Transfer Protocol)			
Transport OSI layer 4		2 nd generation network and application protocol requirements	UDP (User Datagram Protocol), TCP (Transmission Control Protocol), TLS (Transport Layer Security)			
Network OSI layer 3		IP (Internet Protocol), SLAAC, DHCP				
Data link OSI layer 2		ISO 15118-3	ISO 15118-5	ISO 15118-8	ISO 15118-9	
Physical OSI layer 1		Physical and data link layer requirements	Physical and data link layer conform. tests	Physical and data link layer requirements for wireless communication	Physical and data link layer conformance test for wireless comm.	

ISO 15118

How and where did we collect the data ?

- Berlin Open Data
- To display geospatial data as map layers directly from a server

Elektro-Ladesäulen und Ladepunkte - [WMS]

Standort- und flächenbezogene Informationen zu Elektro-Ladesäulen im öffentlichen und öffentlich zugänglichen Raum auf privatem Grund, zu Ladepunkten im öffentlichen und... [Details](#) →

[Geographie und Stadtplanung](#)

Lizenz:

Stand:

dl-de-by-2.0 24.01.2025

Formate:

[WMS](#) [\(unbekannt\)](#) [HTML](#)

There are 2 main types for our purposes:

Supervised Learning - Regression

- Labelled data
- Multivariable linear equation

Cons:

- Needs large amounts of labeled data
- Otherwise overfits

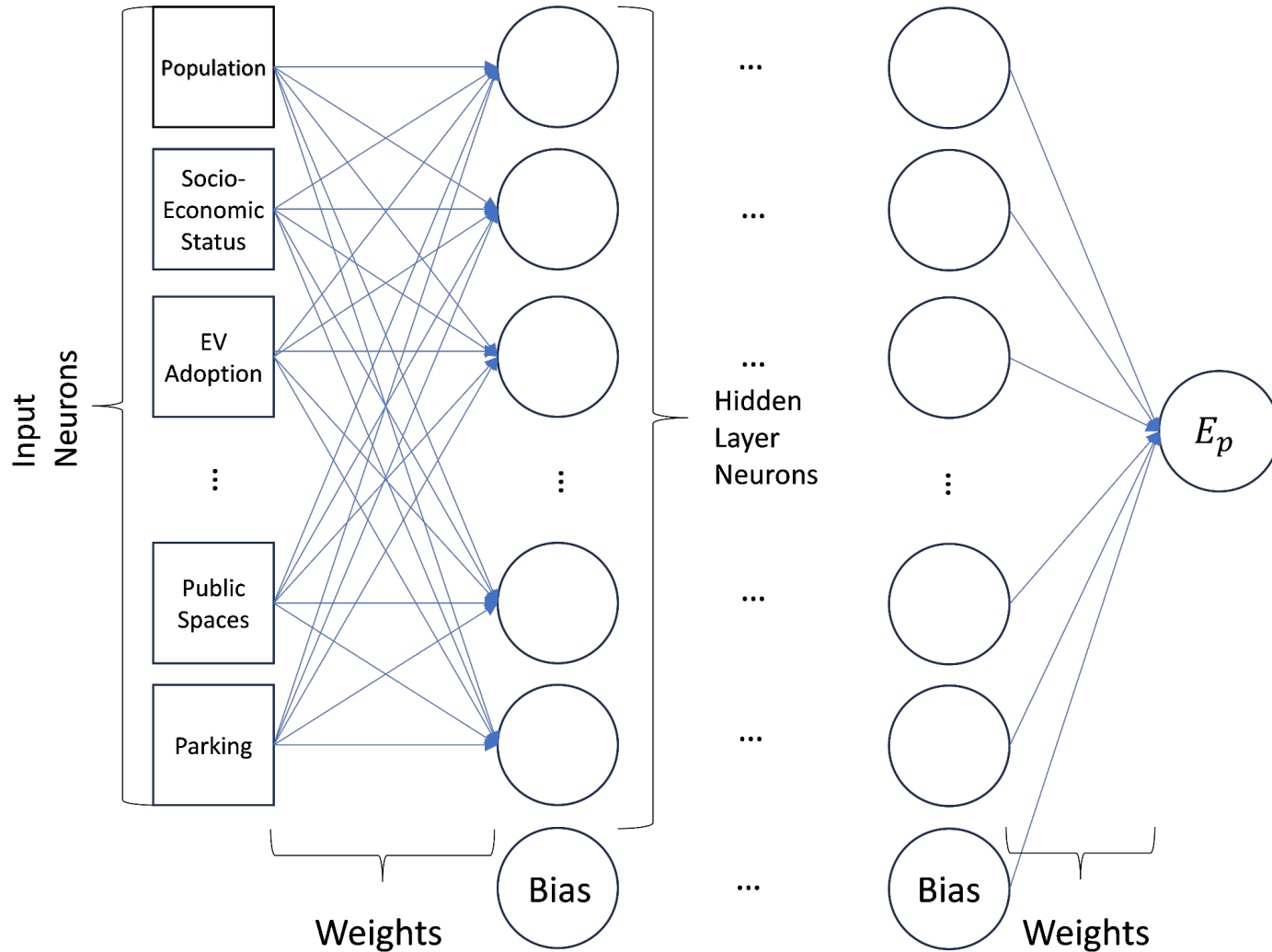
Unsupervised Learning - Clustering

- Unlabelled data
- Find patterns in the data
- Putting similar data into clusters

Cons:

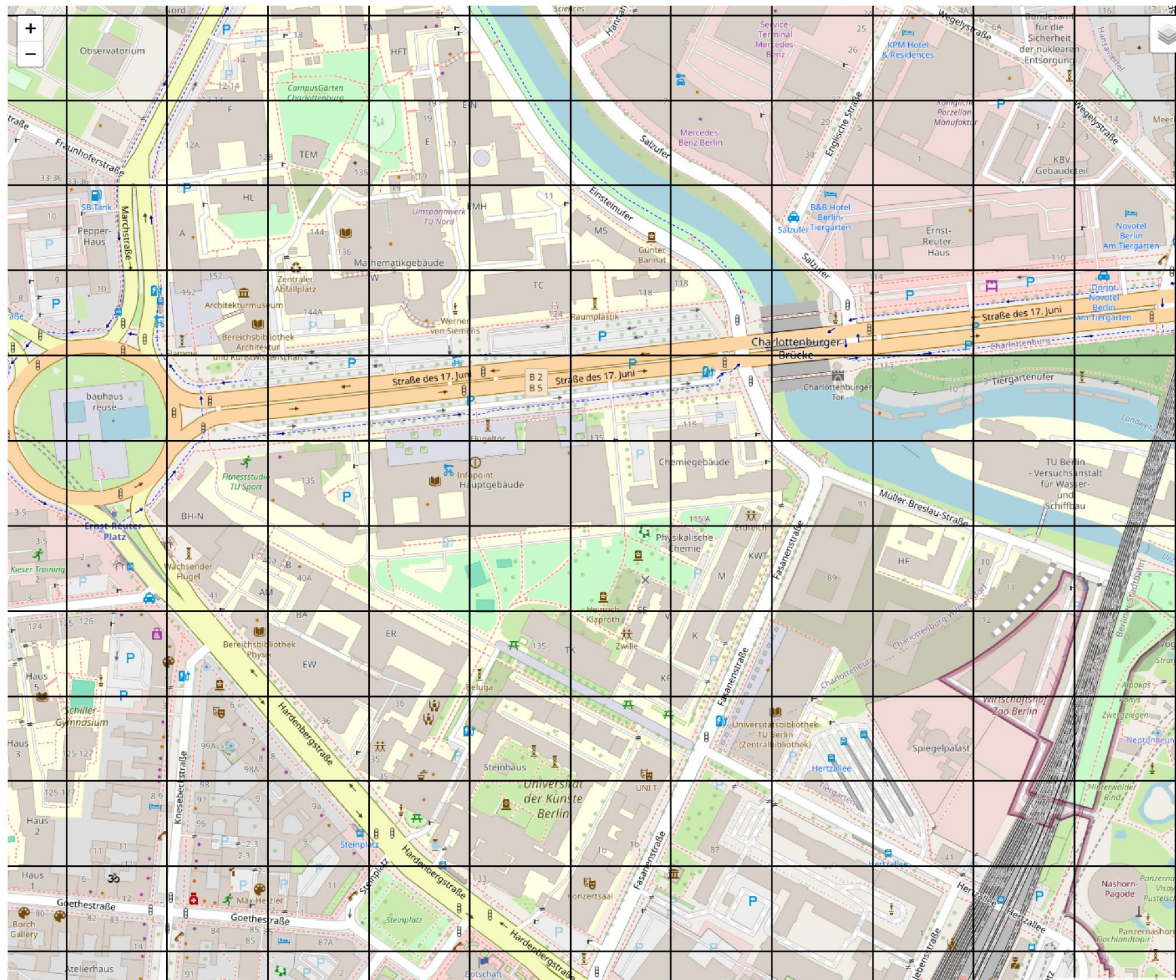
- Not decisive
- The patterns are not guaranteed to be meaningful

Methodology - Neural Networks



Grid Partition Method

The process of dividing a geographical area into smaller areas



EV charger utilization data is not publicly available

Two options:

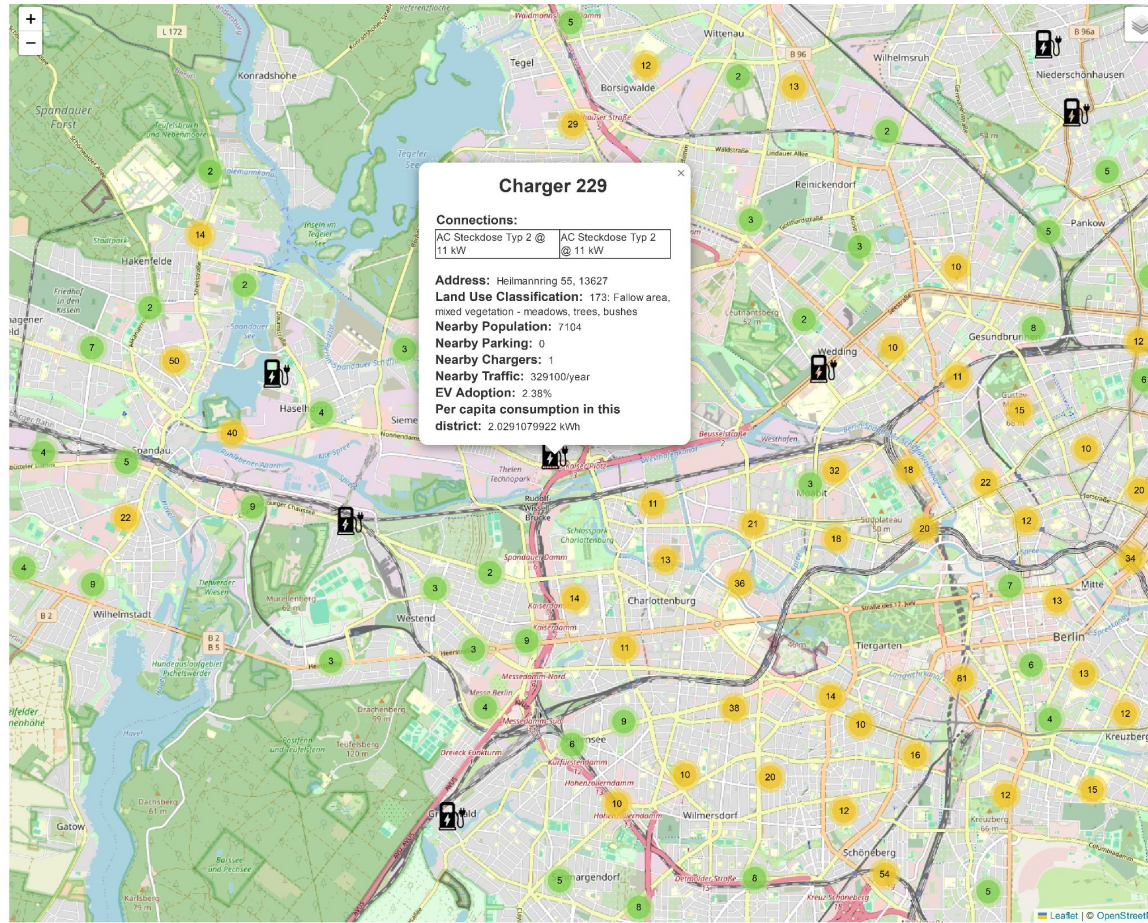
Generating the utilization data

Variable	Units	Lower a	Best a	Upper a
Traffic, a_1	$Events/10^3 \cdot ADT$	17	35	53
Population, a_2	$Events/kPop$	-23	8	39
Competition, a_3	$Events/Comp$	-152	-77	-2
InterProv, a_4	$Events/InterProv$	980	1335	1690

Jayanath et al.'s regression coefficients

Niranjan Jayanath, Nathaniel S. Pearre, and Lukas G. Swan. "Geographic factors impacting the demand for public EV charging: an observational study". In: World Electric Vehicle Journal 15.10 (Sept. 2024), p. 445. url: <https://doi.org/10.3390/wevj15100445>.

Using unsupervised learning(with clustering)

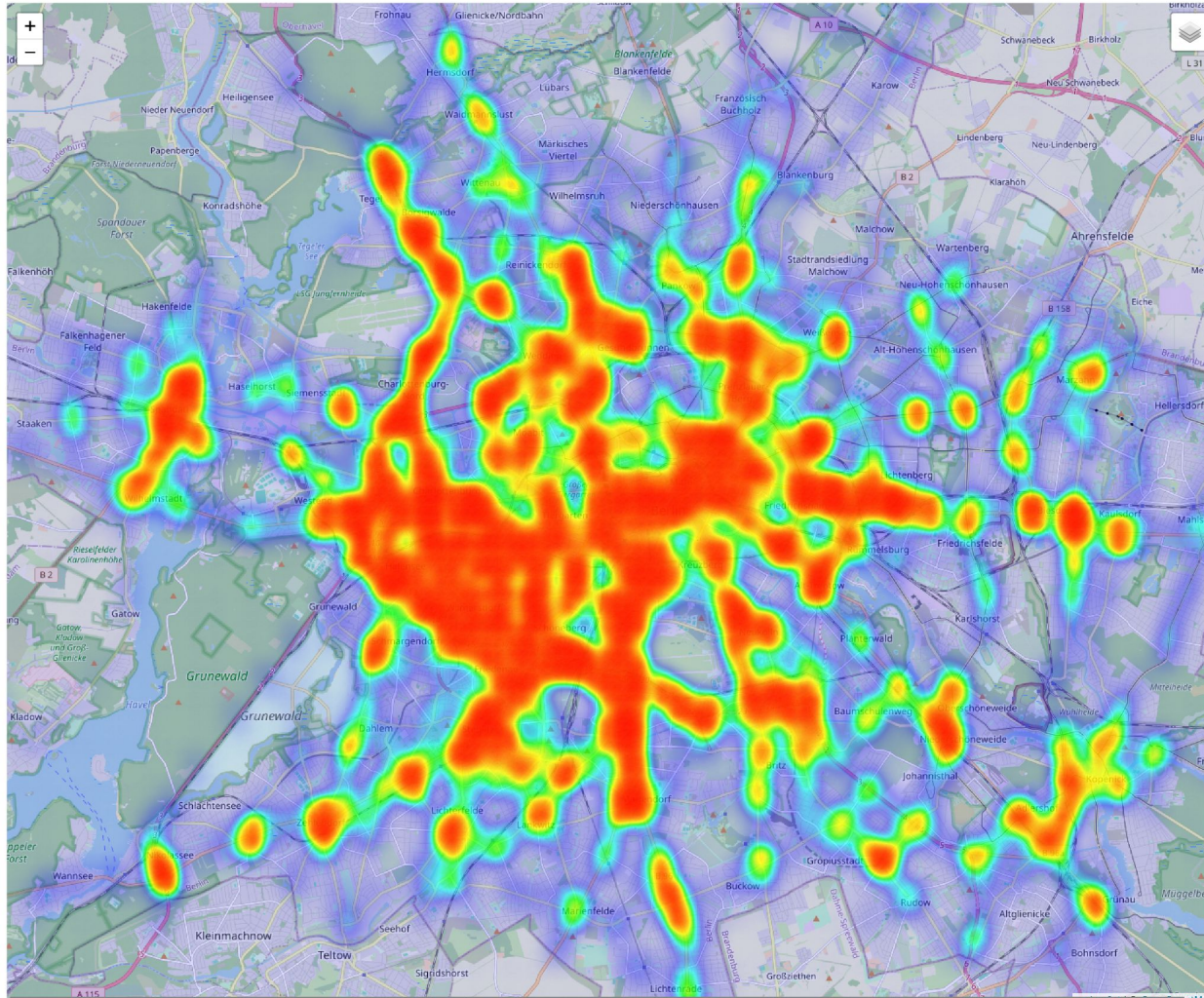


Visualization using a map

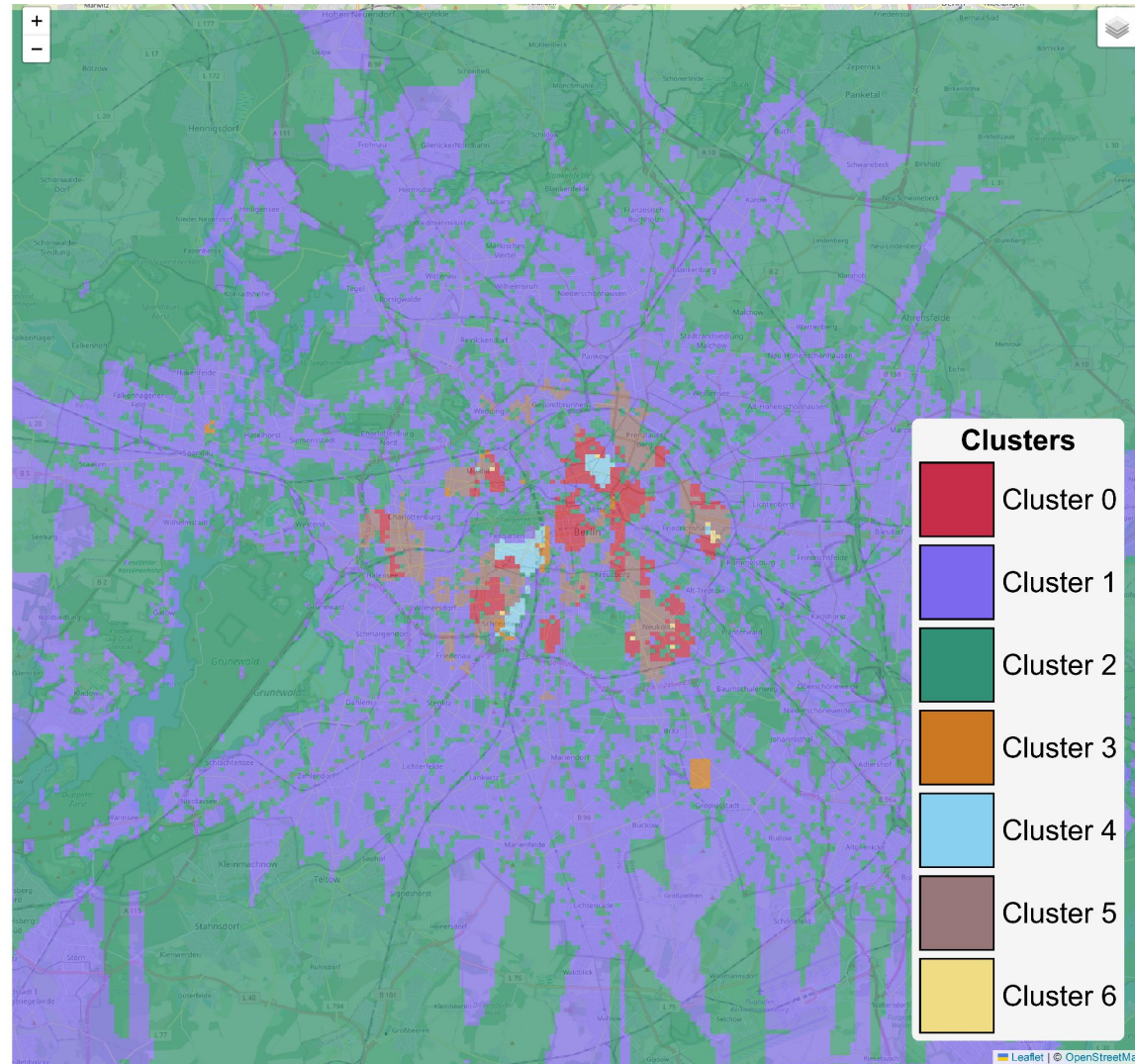
Created with JavaScript and Leaflet

aytacaydin.com/evtech

Using Jayanath et al.'s Regression Model



Unsupervised Learning



- AI-based strategies ensure optimal positioning of EV charging stations.
- Data-driven models identify the best points, taking into account critical variables such as traffic, environmental factors and energy distribution

Main Challenges:

- Data gaps - Limited availability of real-time charging usage data
- Environmental adaptability - The need to account for weather conditions and long-term demand changes.
- Scalability - Expanding infrastructure efficiently in both urban and rural areas.

Future Research:

- Integration of renewable energy to support sustainability.
- Demand forecasting models for better infrastructure planning.
- Enhanced data sharing to improve AI model accuracy.