

Electric Vehicle Single-Charge Range Prediction

for EV makers and adopters

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Problem
formulation:
**from technical specs of an
EV predict its driving range
on a single charge**



Why?

use case examples

EV producers

Task

For a market research department to assess the performance of a new model concept against competitors.

Conventional Solution

Order evaluation of performance from the technical department

With our Service

Quickly prototype new concepts with instant results on their range performance based on technical and user specs.

EV fleet users (e.g. Taxi companies)

Task

Decide on the type of cars to include in the fleet. Compare different car configurations in terms of performance.

Conventional Solution

Research the market for specific models matching their requirements and assess their performance.

With our Service

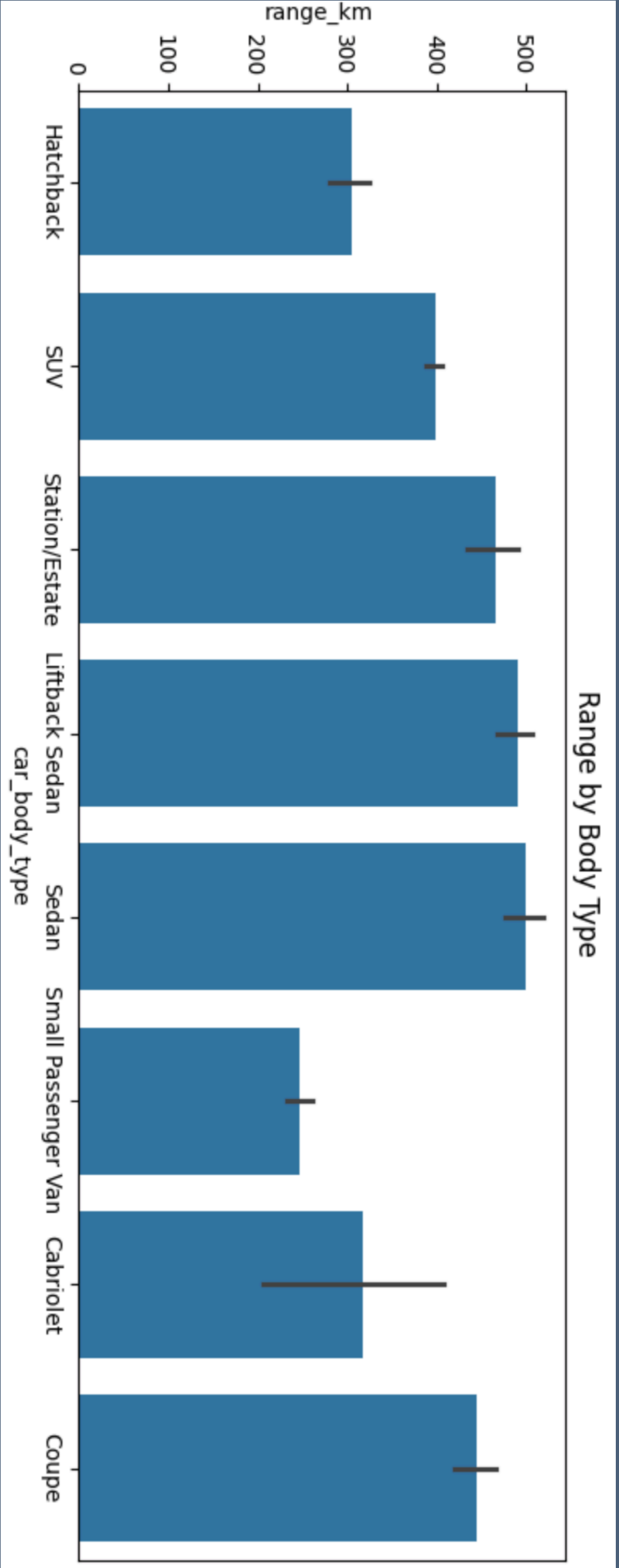
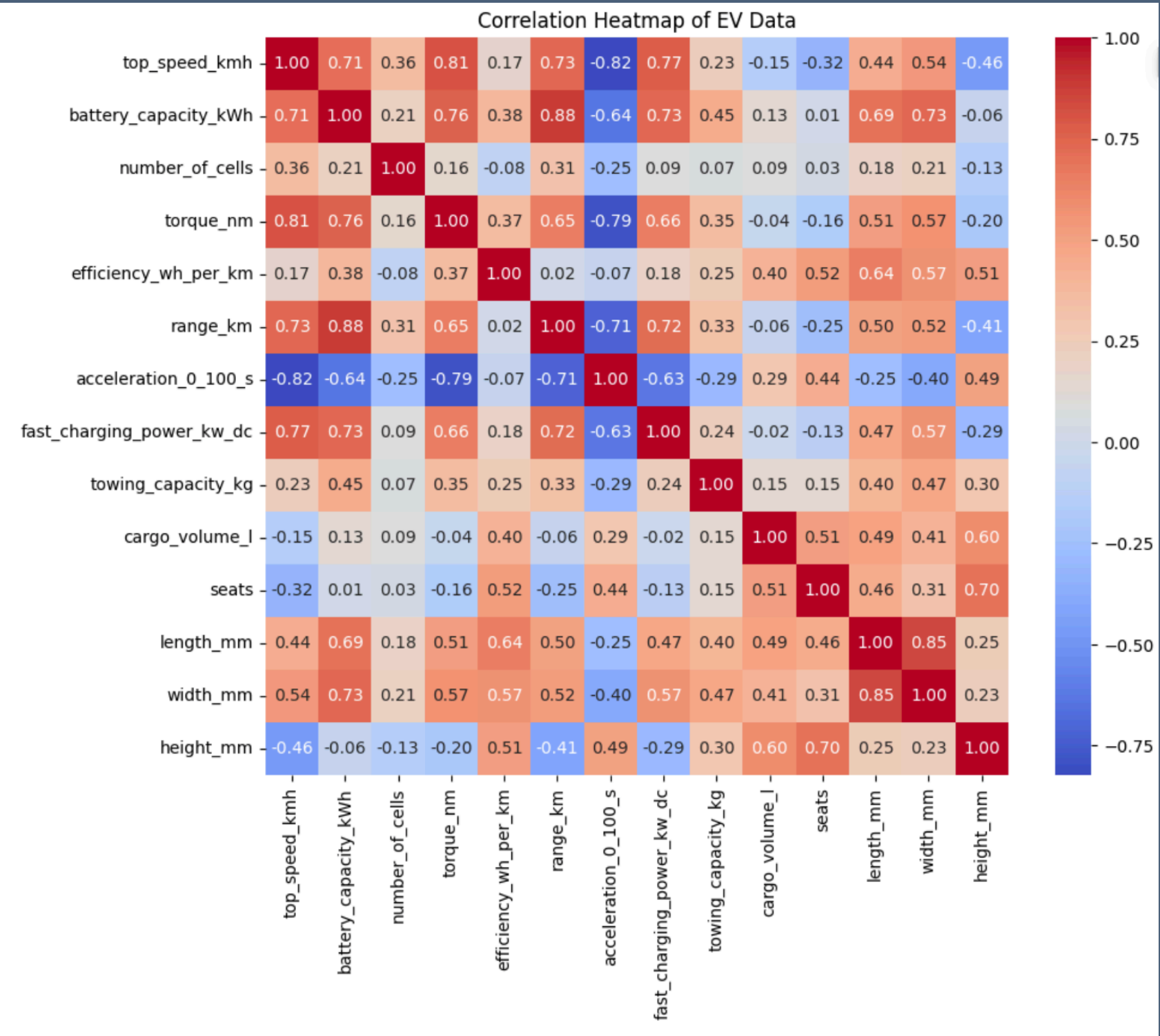
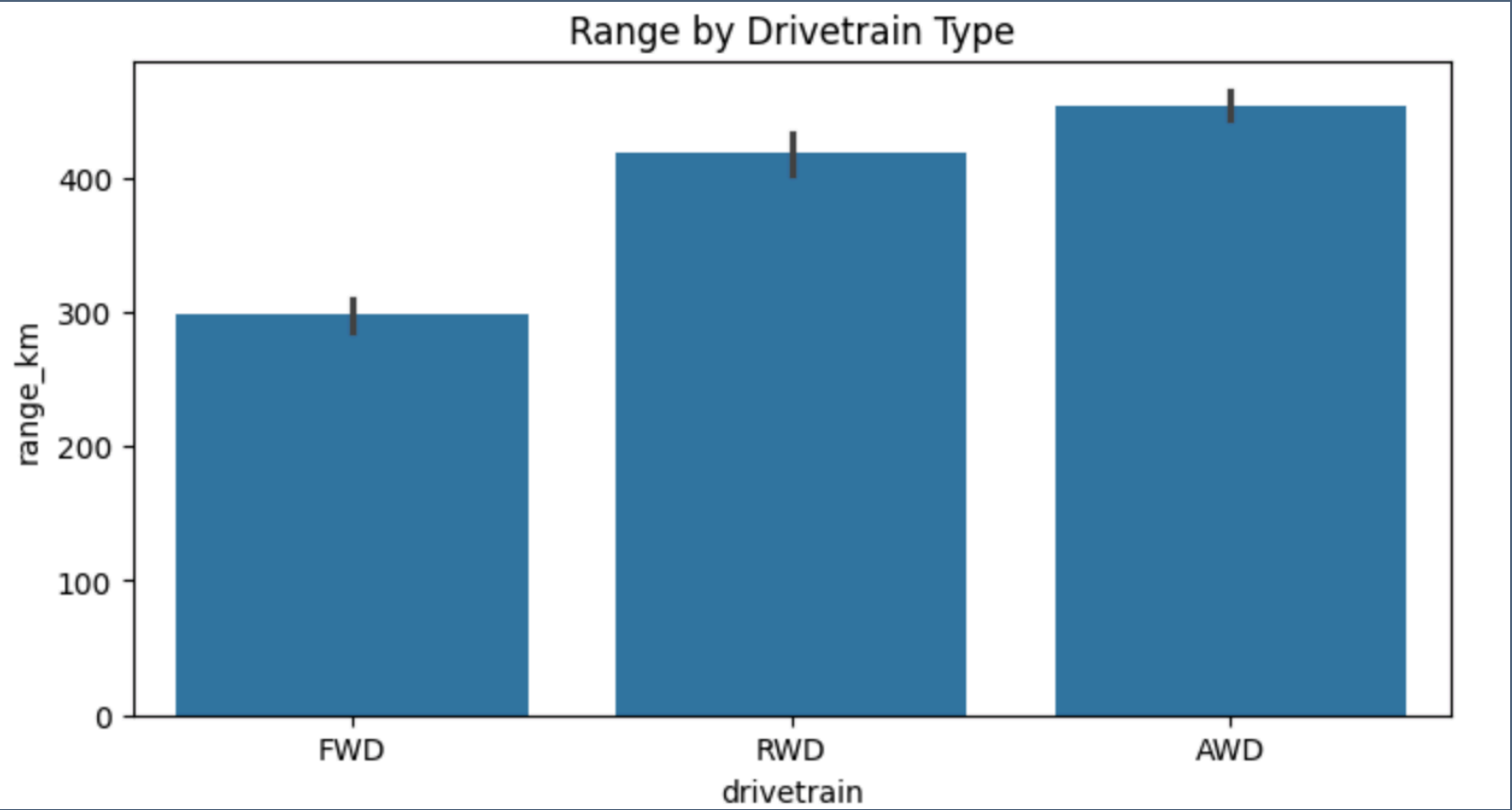
Estimate the approximate performance of the car with necessary specs to evaluate the fleet structure before exploring specific options.

The Data

EDA & Feature Selection

0	brand	478	non-null	object
1	model	477	non-null	object
2	top_speed_kmh	478	non-null	int64
3	battery_capacity_kWh	478	non-null	float64
4	battery_type	478	non-null	object
5	number_of_cells	276	non-null	float64
6	torque_nm	471	non-null	float64
7	efficiency_wh_per_km	478	non-null	int64
8	range_km	478	non-null	int64
9	acceleration_0_100_s	478	non-null	float64
10	fast_charging_power_kw_dc	477	non-null	float64
11	fast_charge_port	477	non-null	object
12	towing_capacity_kg	452	non-null	float64
13	cargo_volume_l	477	non-null	object
14	seats	478	non-null	int64
15	drivetrain	478	non-null	object
16	segment	478	non-null	object
17	length_mm	478	non-null	int64
18	width_mm	478	non-null	int64
19	height_mm	478	non-null	int64
20	car_body_type	478	non-null	object
21	source_url	478	non-null	object

Size: 470 records after preprocessing



The Model

ANN design

Input Layer: 15 neurons (after one-hot encoding)

Hidden Architecture: 2-layer feedforward network

Hidden Layer 1: 32 neurons with ReLU activation + Dropout (0.2)

Hidden Layer 2: 16 neurons with ReLU activation + Dropout (0.15)

Output Layer: 1 neuron with linear activation (regression output)

Note: the dataset is fairly small — 470 records, meaning that a deep heavy-weight network is excessive given tightly limited computational resources

Regression-Specific Design

- **Linear output activation:** allows unbounded continuous predictions
- **Single output neuron:** predicts one continuous value (range_km)
- **Mean Squared Error loss function:** assessing deviation from the real target value

Evaluation Metric:

- **Mean Absolute Error:** easy to interpret
- **Mean Absolute Percentage Error:** percentage error compared to target size

The Prototype


links & demo


GitHub repository:

[https://github.com/kshashkov/
ev_range_prediction](https://github.com/kshashkov/ev_range_prediction)

Prototype on GitHub pages:

[https://kshashkov.github.io/
ev_range_prediction/application/](https://kshashkov.github.io/ev_range_prediction/application/)

 **Electric Vehicle Range Predictor**
AI-Powered Range Estimation Using Neural Networks

 **Model Training**

✓ Model ready! Enter vehicle specifications to predict range.

✓ Loading training data - 470 samples loaded

✓ Preprocessing features - 14 features, 376 train samples

✓ Building neural network - 1025 parameters

✓ Training model - Training completed successfully


100%

Epoch
200

Loss
4610.83

Val Loss
761.45

MAE
53.38

 **Vehicle Specifications**

Top Speed (km/h)
180


Battery Capacity (kWh)
75

Torque (Nm)
450

Acceleration 0-100 km/h (s)
6,5

Fast Charging Power (kW DC)
120

Fast Charge Port
CCS

 **Prediction Result**

Estimated Range
394
kilometers on a single charge