

# ENEL 890AO: Computational Methods in Electrical Engineering

## Introduction of Electric Vehicle

Zhanle (Gerald) Wang

Faculty of Engineering

U of R

# Electric Vehicle (EV) and EV charging

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- [Electric cars vs Petrol cars](#)

# Advantages

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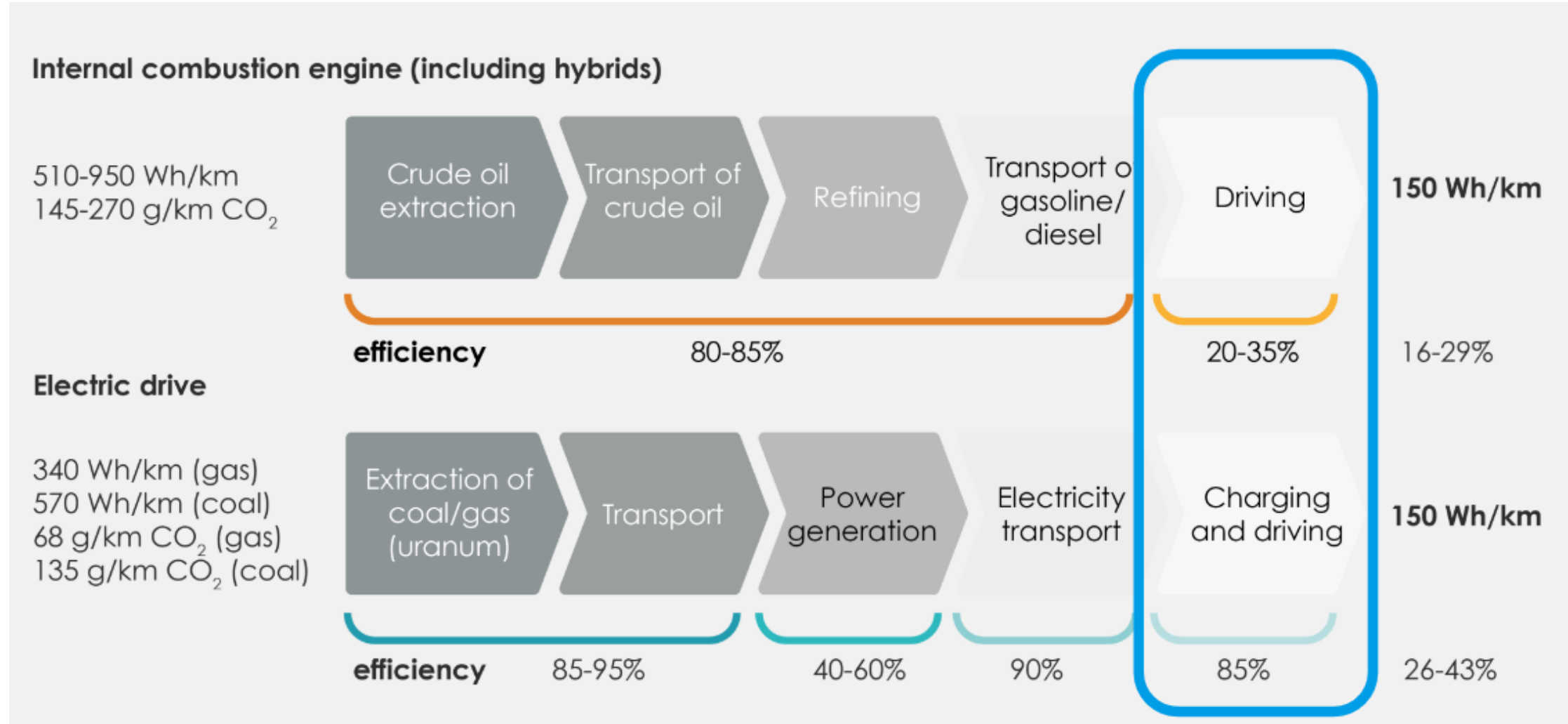
- Sustainability
- Efficiency
- Convenience
- Economy

# Sustainability

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# Efficiency



# Lower total cost of ownership

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- Higher purchase price, but
- Lower maintenance costs
- Lower taxes
- Cheaper fuel (electricity)
- Government subsidy

- Emission Calculator Tool:

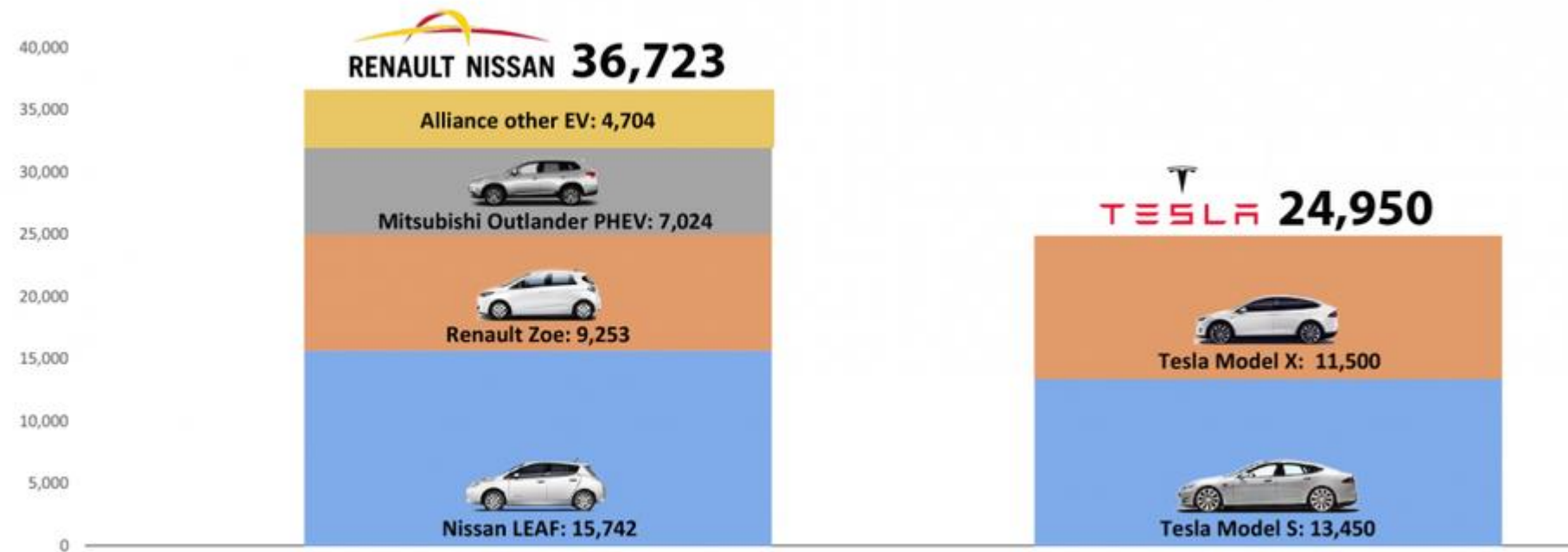
[https://afdc.energy.gov/vehicles/electric\\_emissions.html](https://afdc.energy.gov/vehicles/electric_emissions.html)

- Fuel Economy Tool:

<http://www.fueleconomy.gov/feg/Find.do?action=sbsSelect>

- compare the '2018 Honda Civic 4Dr, 2.0 L, 4 cyl, Automatic' and the '2018 Tesla Model S AWD - 75D'.
- 33.7 kWh (121 MJ) is equivalent to one gallon gas fuel
- 259 miles is about 417 km

## Global electric vehicle sales, 1st quarter 2017



Deliveries. Source: Automakers, EV Sales Blog

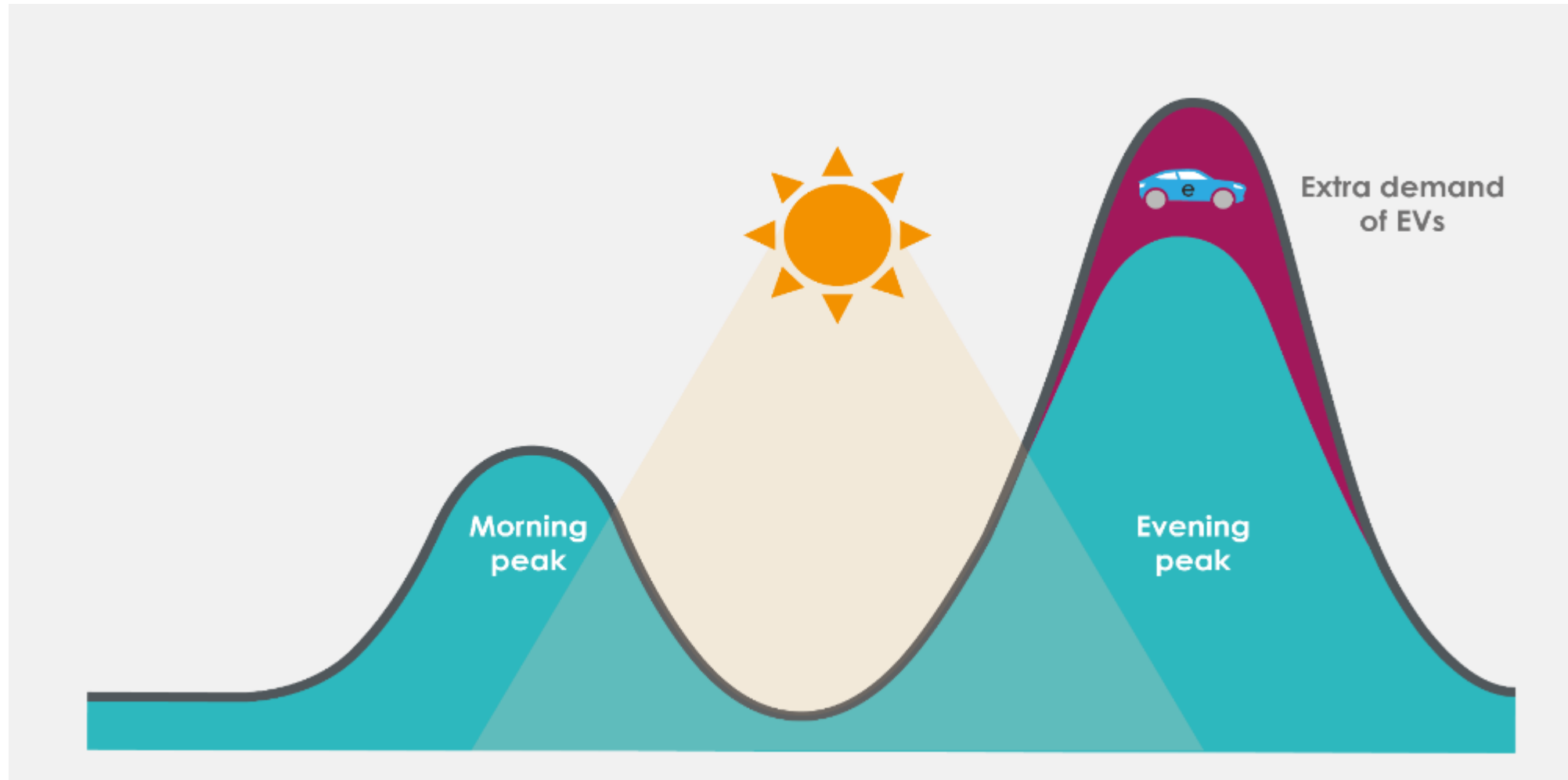


# Basic EV charging

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- Onboard charger: 1.9 – 22 kW
- DC offboard charger: 50 – 350 kW
- Charging level
  - Level 1: 0 – 10 kW
  - Level 2: 10 – 50 kW
  - Level 3: 50 – 350 kW

# Impact on power system



# Possible solutions

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- Grid capacity expansion (expensive)
- Controlled battery charging or Smart charging
- Incentive to user
- Rewarded for flexible load pattern

# Smart Charging

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- Smart charging is a series of intelligent functionalities to control the EV charging power in order to create a flexible, sustainable, low cost and efficient charging environment.

# Benefits of smart charging

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- Increase the flexibility of charging
- Higher utilization rates of fixed assets
- Efficient utilization of energy in distribution network
- Make electric vehicles more sustainable
- Provide new revenue streams to EV owners

# V2G

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- Using the electric vehicle battery to feed power back to the grid using a bidirectional EV charger

# V2G advantages and challenges

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- Advantage
  - Storage for renewables
  - Reduce peak demand
  - Emergency power
  - Ancillary services
- Challenges
  - Bidirectional charger
  - Battery degradation
  - ICT infrastructure
  - Standardization and regulatory framework
  - Lack of incentives for user

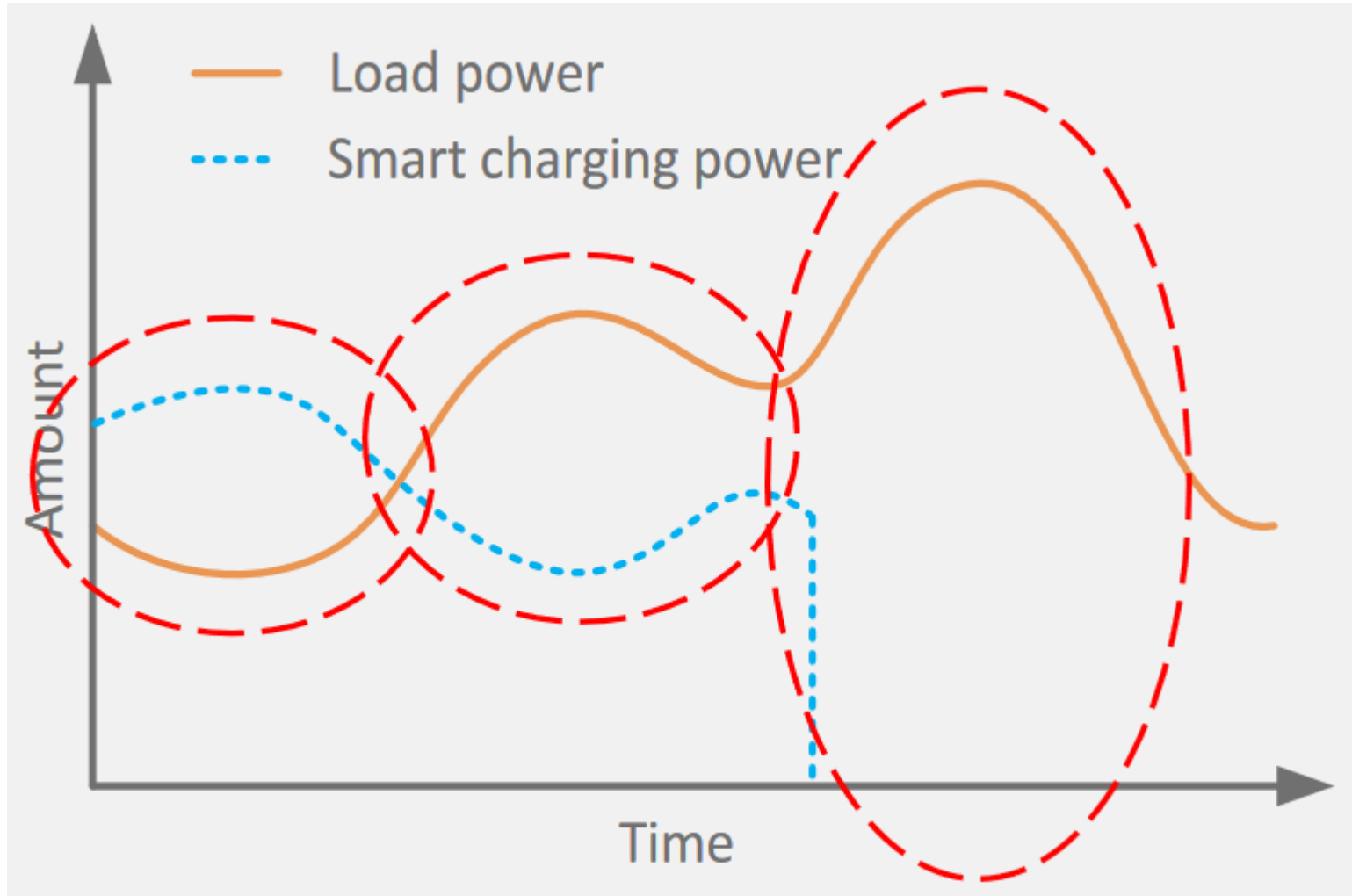
# Example applications of smart charging

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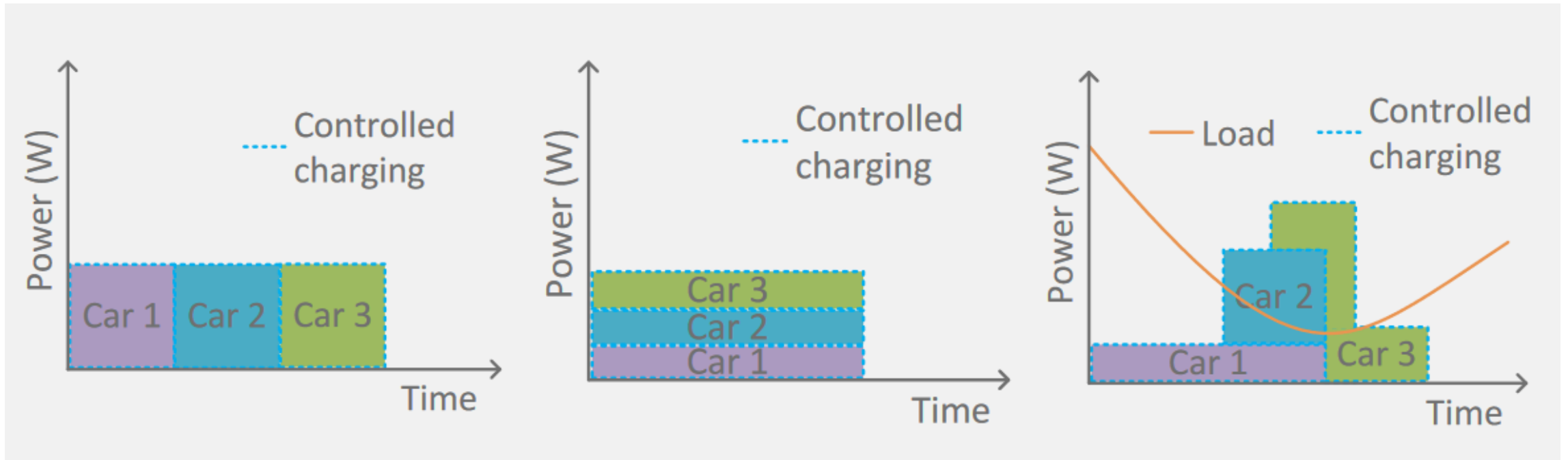
- Local load balancing
  - Adjust charging time/power according to load
  - Balance multiple charge points with priority
- Renewable energy utilization
- Price based charging
- Peak shaving
- Grid back up



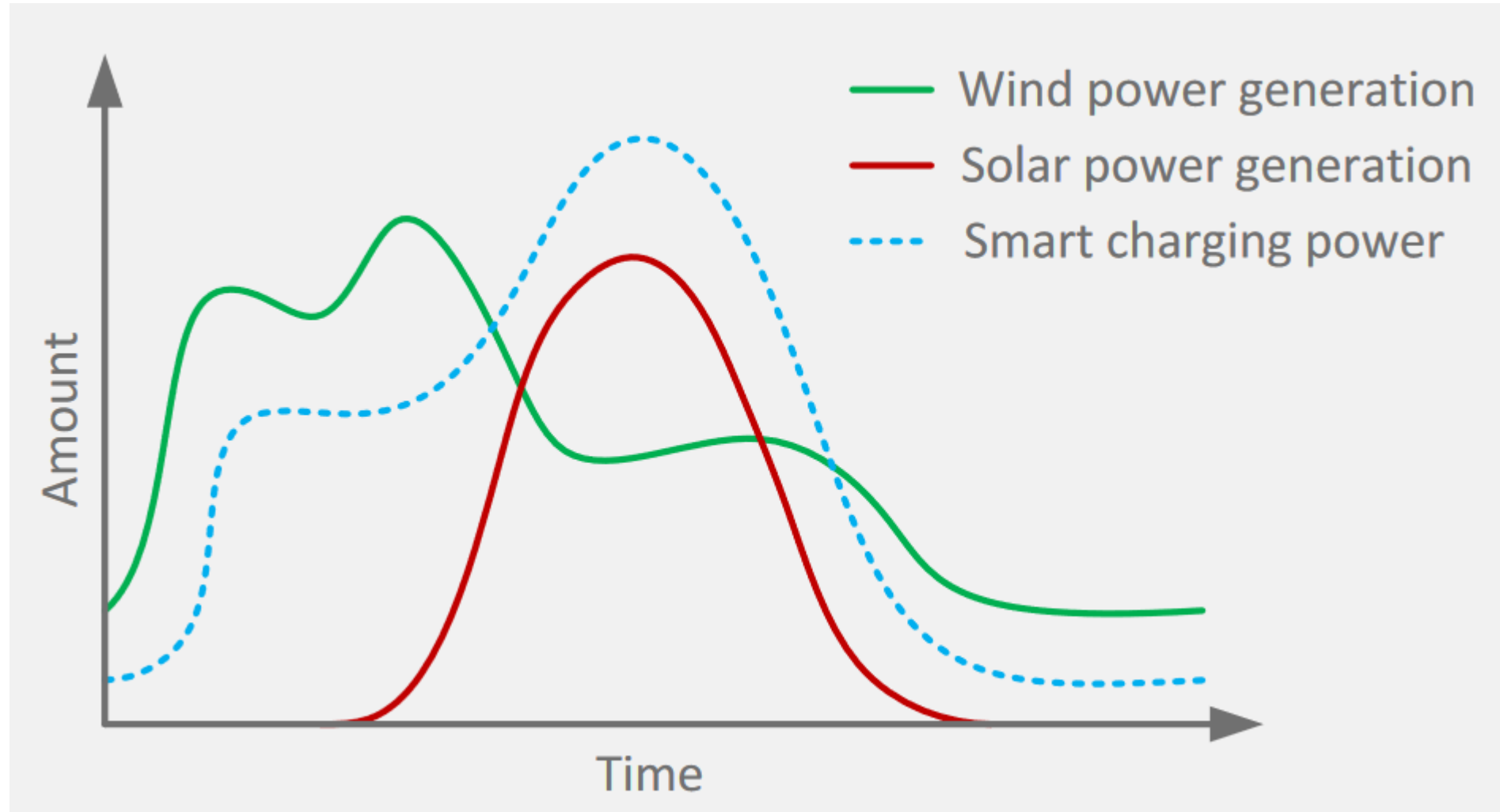
# Load balancing



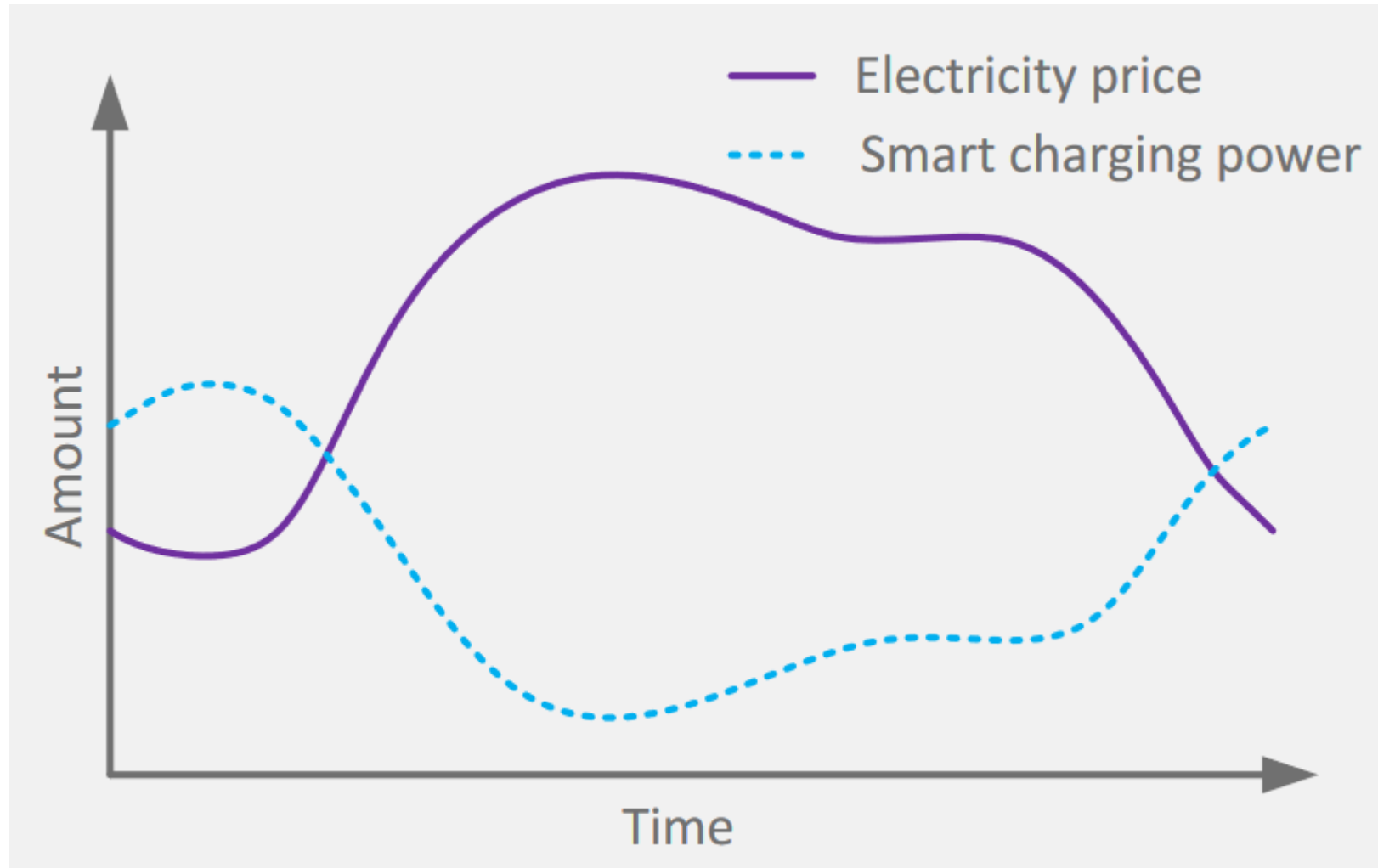
# Load balancing



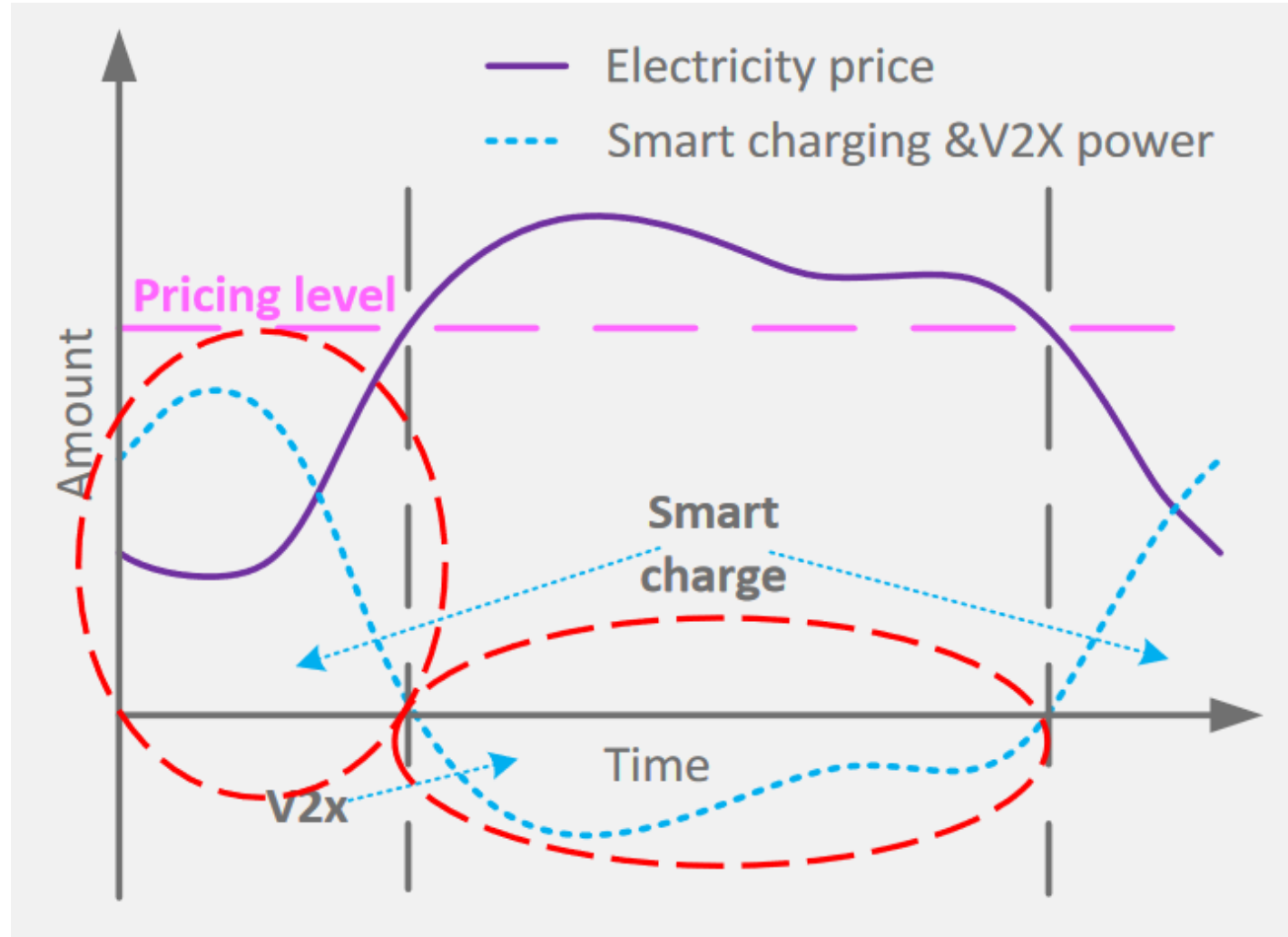
# Renewable energy availability



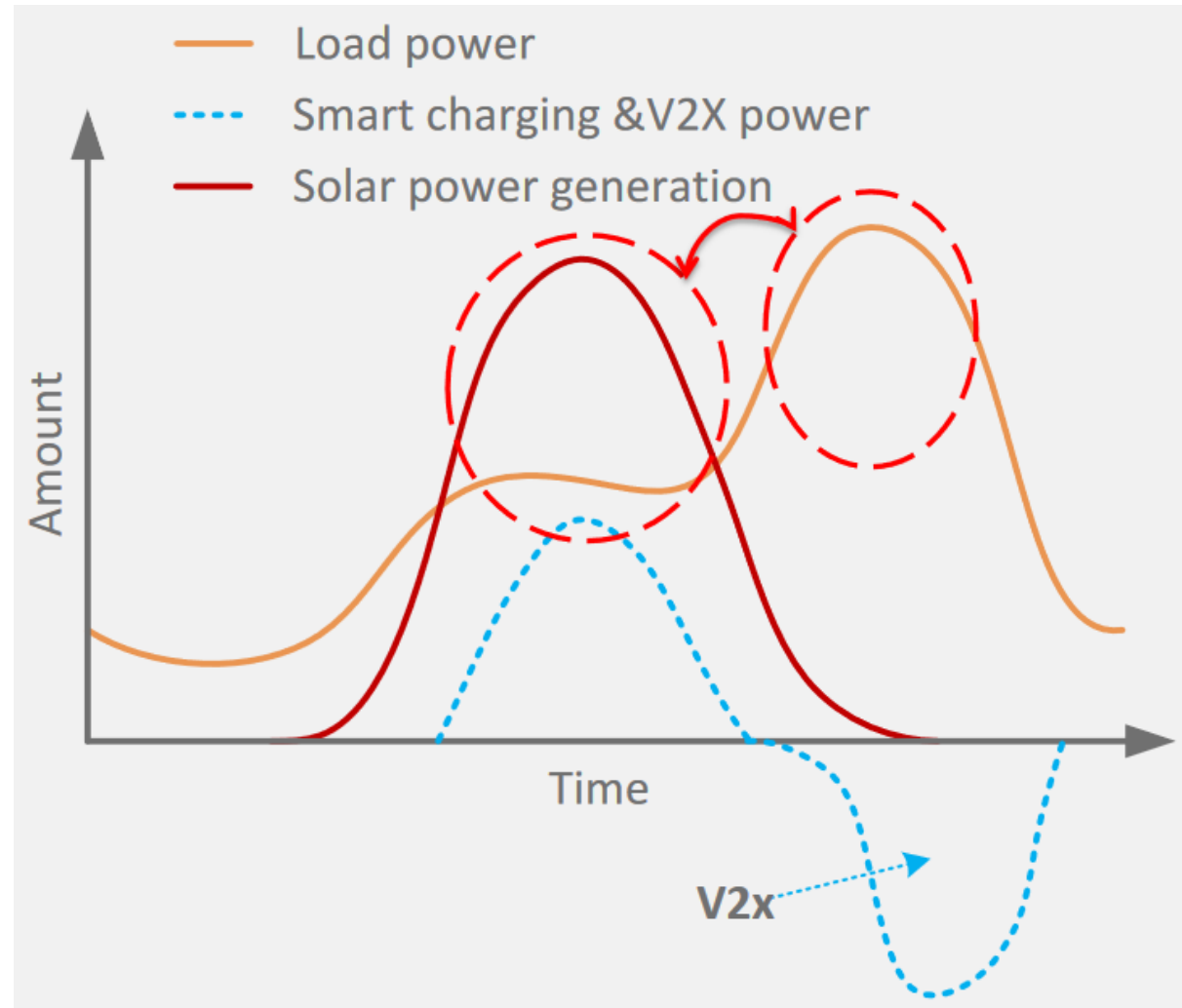
# Price based charging



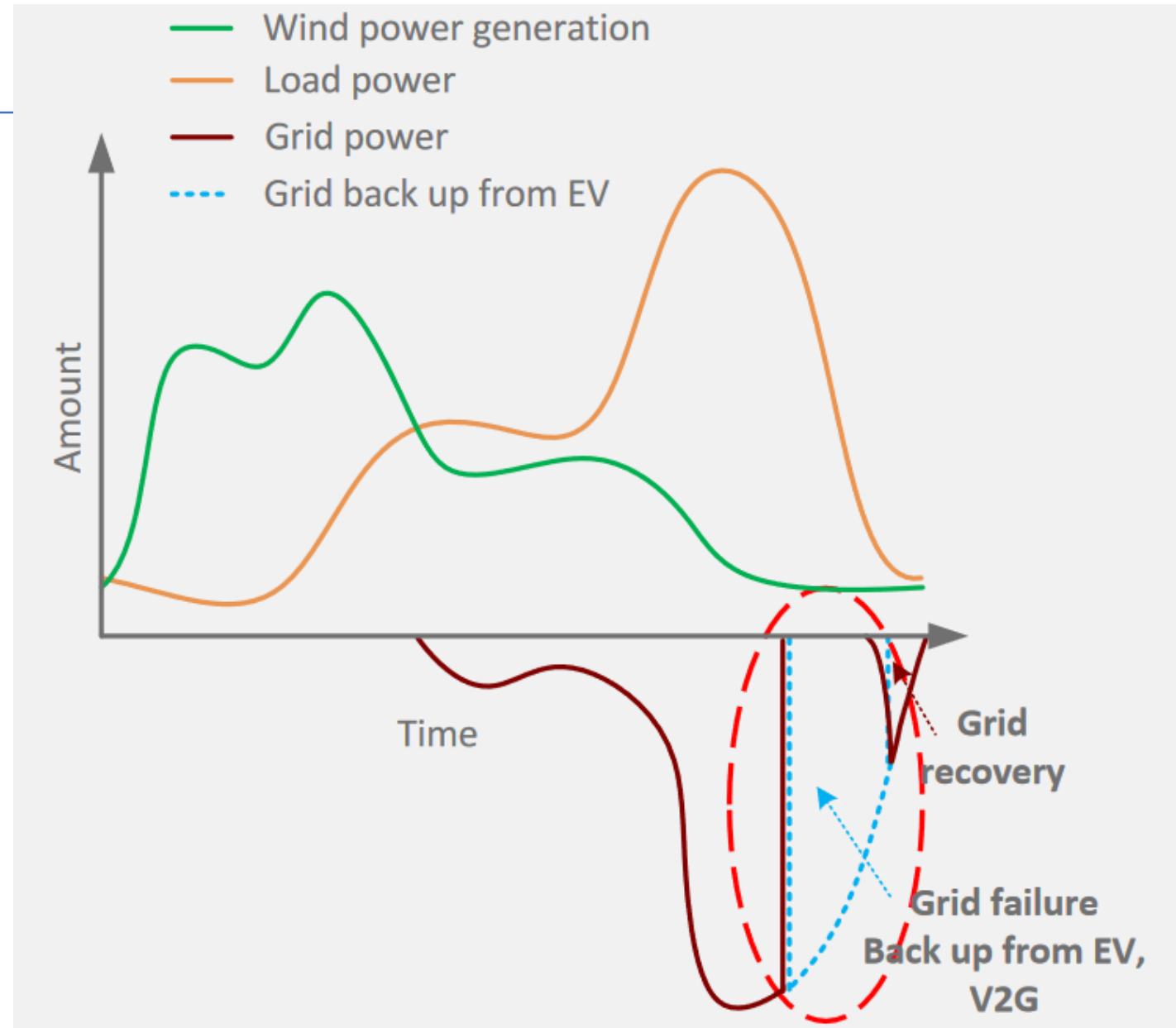
# Price based charging



# Peak shaving



# Grid backup



## Example: price-based charging

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- Capacity: 100 kWh
- Rated charging demand: 10 kW
- State-of-charge (SOC): Available capacity expressed as a percentage of some reference, sometimes its rated capacity. (e.g., empty 0%, full 100%)

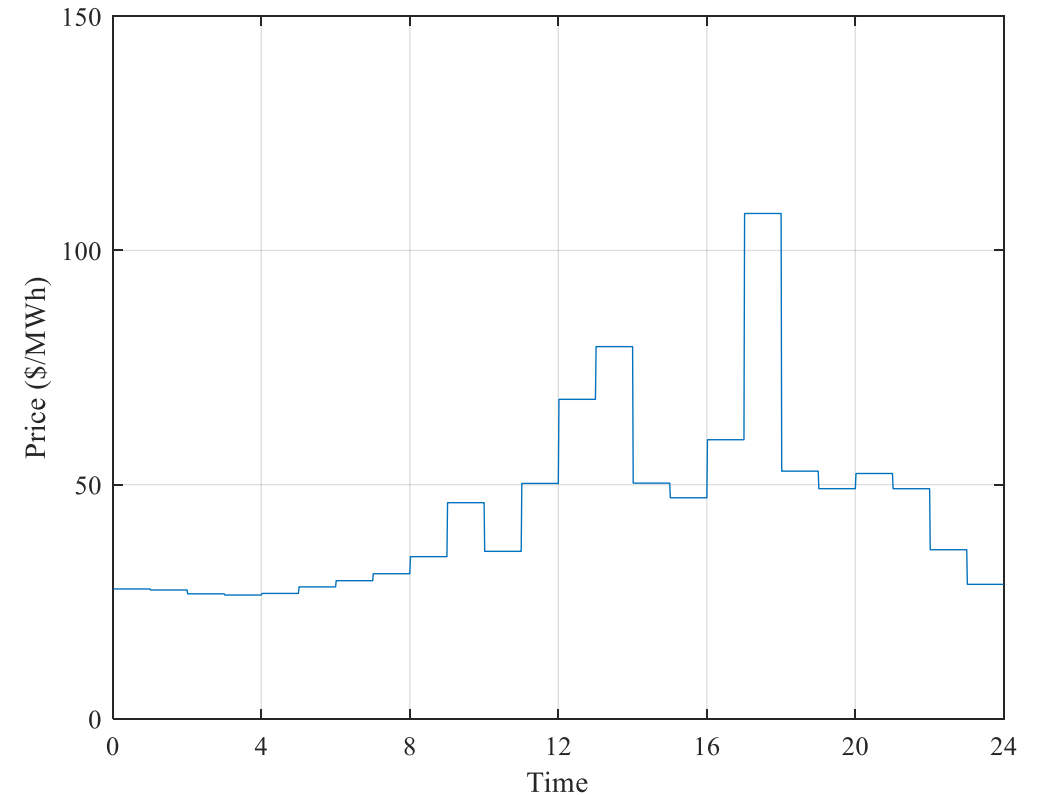


# Example: price-based charging

Suppose you have a EV with a capacity of 100 kWh. The rated power of the charger is 10 kW.

In your neighborhood hourly real-time price (RTP) is applied for electricity usage.

When you arrive home the SOC is 10%, you need to determine the **hourly** EV charging schedule to minimize the electricity cost.



## Example: price-based charging

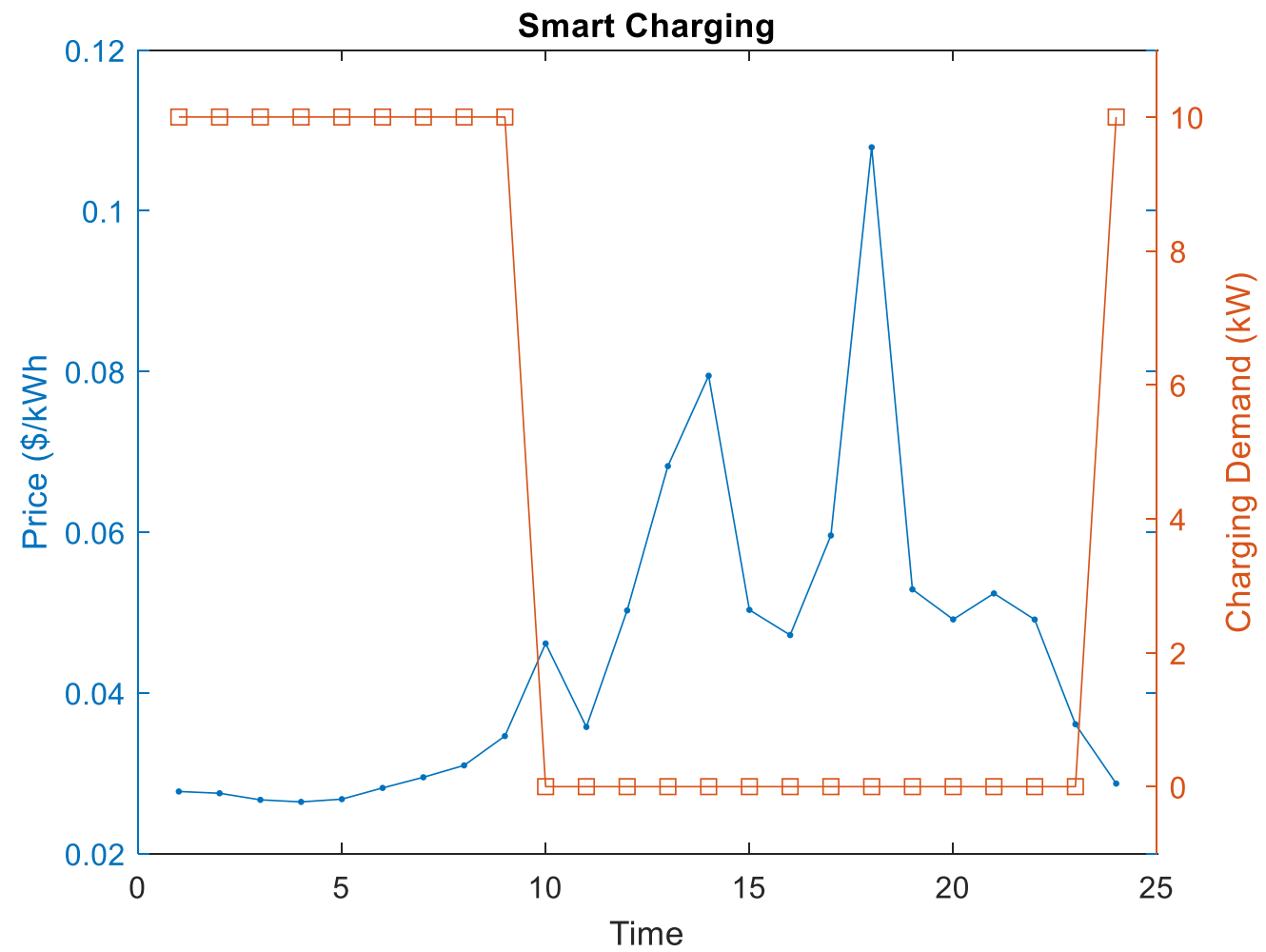
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# Formulation Steps

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- Encode decisions/actions as **decision variables** whose values we are seeking
- Identify the relevant **problem data**
- Express **constraints** on the values of the decision variables as mathematical relationships (inequalities) between the variables and problem data
- Express the **objective function** as a function of the decision variables and the problem data



# Python code

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```
import pandas as pd
import cvxpy as cp
import matplotlib.pyplot as plt

# read .csv
mp = pd.read_csv("hourlyprice.csv",header=None)
# convert dataframe to array
p = mp.values
p = p[0]
p = p/1000
```

```
n = 24
# Construct the problem.
q = cp.Variable(n)
objective = cp.Minimize(p*q)
constraints = [0 <= q, q <= 10, sum(q)==90]
prob = cp.Problem(objective, constraints)

# The optimal objective value is returned by `prob.solve()`.
result = prob.solve()

# The optimal value for x is stored in `x.value`.
print(q.value)
plt.plot(p)
plt.figure()
plt.plot(q.value)
```

# Matlab code

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```
filename = 'hourlyprice.csv';
p = csvread(filename);
p = p';
p = p/1000; % $/kWh
n=24;
cvx_begin
variable q(n)
    minimize( q'*p )
    subject to
        q<=10;
        q>=0;
        sum(q) == 100;
cvx_end
```

```
%% plot
t=1:24;
[hAx,hLine1,hLine2] = plotyy(t,p,t,q);
title('Smart Charging')
xlabel('Time')
ylabel(hAx(1),'Price ($/kWh')
ylabel(hAx(2),'Charging Demand (kW)')
set(hAx(2),'YLim',[-1 11])
set(hLine1,'Marker','.')
set(hLine2,'Marker','s')
```