


Problem Statement

- We have a set of **Electric Vehicles (EVs)** that are allocated to **Electric Vehicle Supply Equipment (EVSE)**, commonly known as **charging ports**.
- Each EV i has a specific **energy requirement** (e_i) and a required **charging duration** (τ_i^{end})

Objective:

Over a defined time horizon T with discrete time steps, establish a charging schedule that:





- Ensures each EV meets its energy requirement within the specified deadline.
- Minimizes the energy consumption at each time step.



	t_0	t_1	t_2	t_3	t_4	t_5	t_6	t_7	T
EV 1			e_1			τ_1^{end}			
EV 2				e_2				τ_2^{end}	
EV 3			e_3		τ_3^{end}				

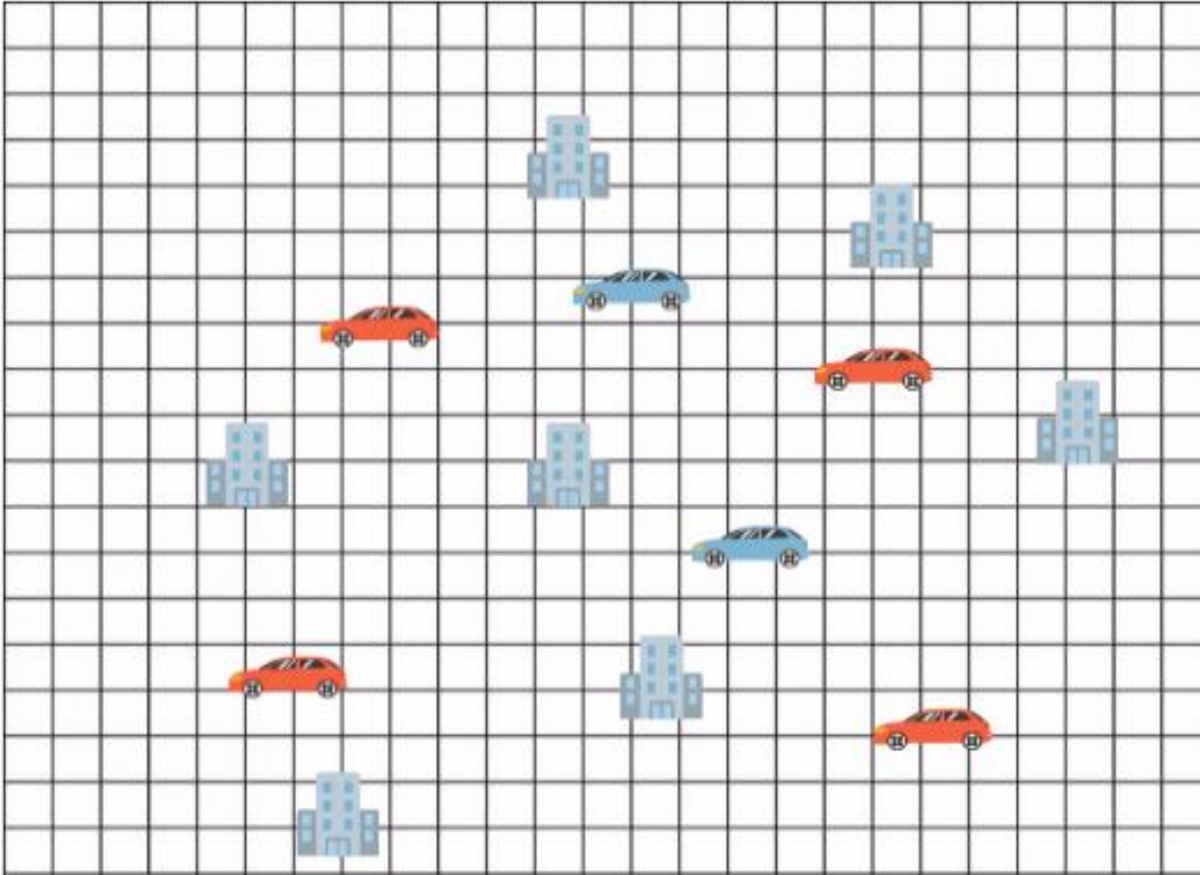
Toy Example

- Let's assume we are optimizing for a time **horizon $T = 4 \text{ hours}$** , with a timestep **$\Delta T = 1 \text{ hour}$** , a constant **voltage $V = 240V$** , and a **set of allowable charging rate currents $p = \{8, 16, 32, 48, 64\} A$**
- Consider that we have **4 EVs** where EV 1, 2, and 3 each have an energy requirement **$e_1 = e_2 = e_3 = 26880 \text{ kWh}$** and a required charging duration of **3 hours**. EV 4 has an energy requirement **$e_4 = 7680 \text{ kWh}$** and a required charging duration of **1 hour**.

	t_0	t_1	t_2	t_3	T	
	16 A	48 A	48 A	0		26880/26880 kWh
	32A	48A	32A	0		26880/26880 kWh
	32A	32A	48A	0		26880/26880 kWh
	32A	0	0	0		7680/7680 kWh

EV charging station placement problem

Aman Chandra, Jitesh Lalwani, Babita Jajodia: *"Towards an Optimal Hybrid Algorithm for EV Charging Stations Placement using Quantum Annealing and Genetic Algorithms"*



- **Buildings:** Point of Interest (POI)
- **Blue Cars:** Existing charging stations
- **Red Cars:** New charging stations

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Objective:

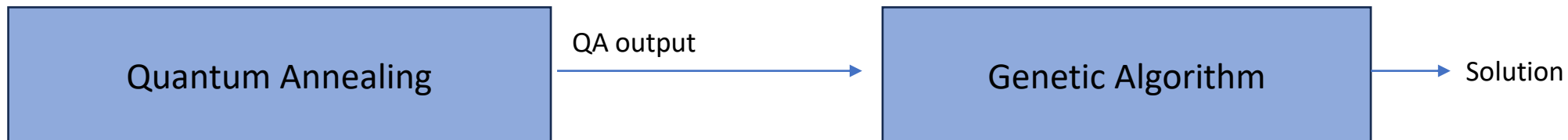
- Install the new chargers at the **minimum possible distance from the point of interest**.
- Ensure that the chargers are placed at **the maximum possible distance from the existing chargers** and at the **maximum distance from each other**.

$$H_1 = + \sum_{i=1}^N x_i d_i^p$$

$$H_2 = - \sum_{i=1}^N x_i d_i^c$$

$$H_3 = - \sum_{i=1}^N x_i d_i^l$$

Solved: Quantum Annealing combined with Genetic Algorithm



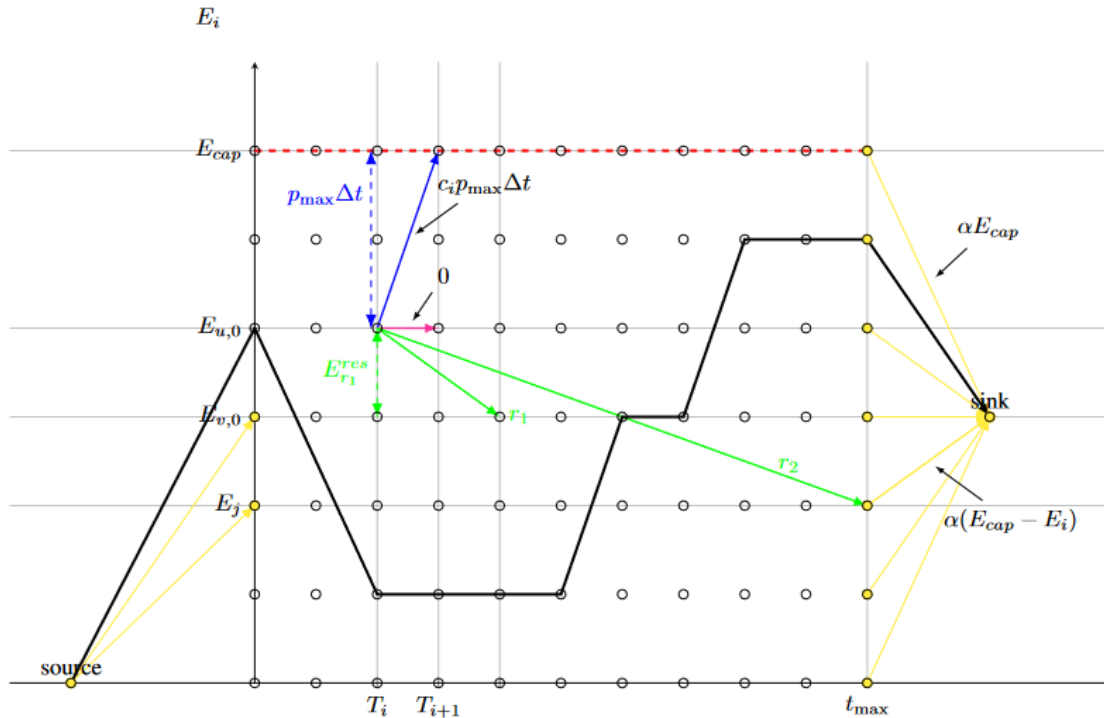
Electric Mobility Problem

Margarita Veshchezerova, Mikhail Somov, David Bertsche, Steffen Limmer, Sebastian Schmitt, Michael Perelshtein, Ayush Joshi Tripathi: “A Hybrid Quantum-Classical Approach to the Electric Mobility Problem”

Problem Statement

Suppose we have rental EVs. We use each EV to satisfy reservations.

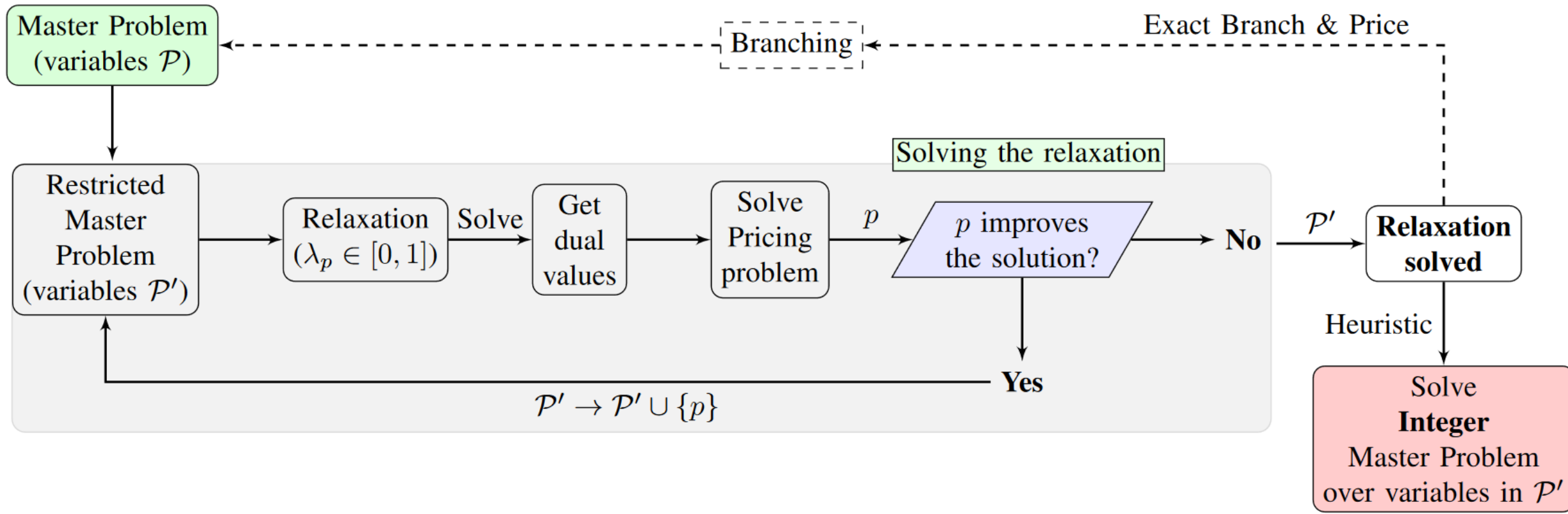
The objective is : to satisfy all reservations while spending the least amount of energy possible.



$$\begin{aligned}
 \min \quad & \sum_{p \in \mathcal{P}} c_p \lambda_p + c^{uncov} \sum_{r \in R} E_r^{res} y_r \\
 \sum_{p \in \mathcal{P}: r \in p} \lambda_p + y_r &= 1, & \forall r \in R \\
 \sum_{p \in \mathcal{P}: v \in p} \lambda_p &= 1, & \forall v \in V \\
 \lambda_p &\in \{0, 1\}, & \forall p \in \mathcal{P}
 \end{aligned}$$

Electric Mobility Problem

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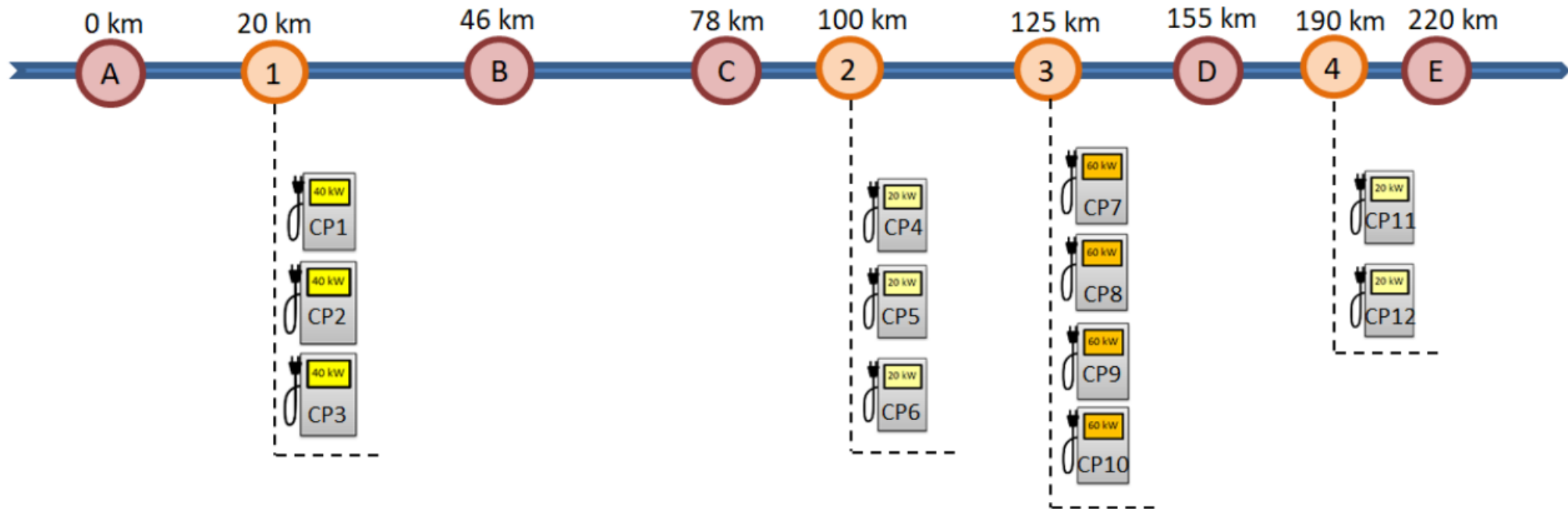
Charging Electric Cars on a Motorway

Różycki, R.; Józefowska, J.; Kurowski, K.; Lemański, T.; Pecyna, T.; Subocz, M.; Waligóra, G. A Quantum Approach to the Problem of Charging Electric Cars on a Motorway. *Energies* 2023, 16, 442. <https://doi.org/10.3390/en16010442>

Problem Statement

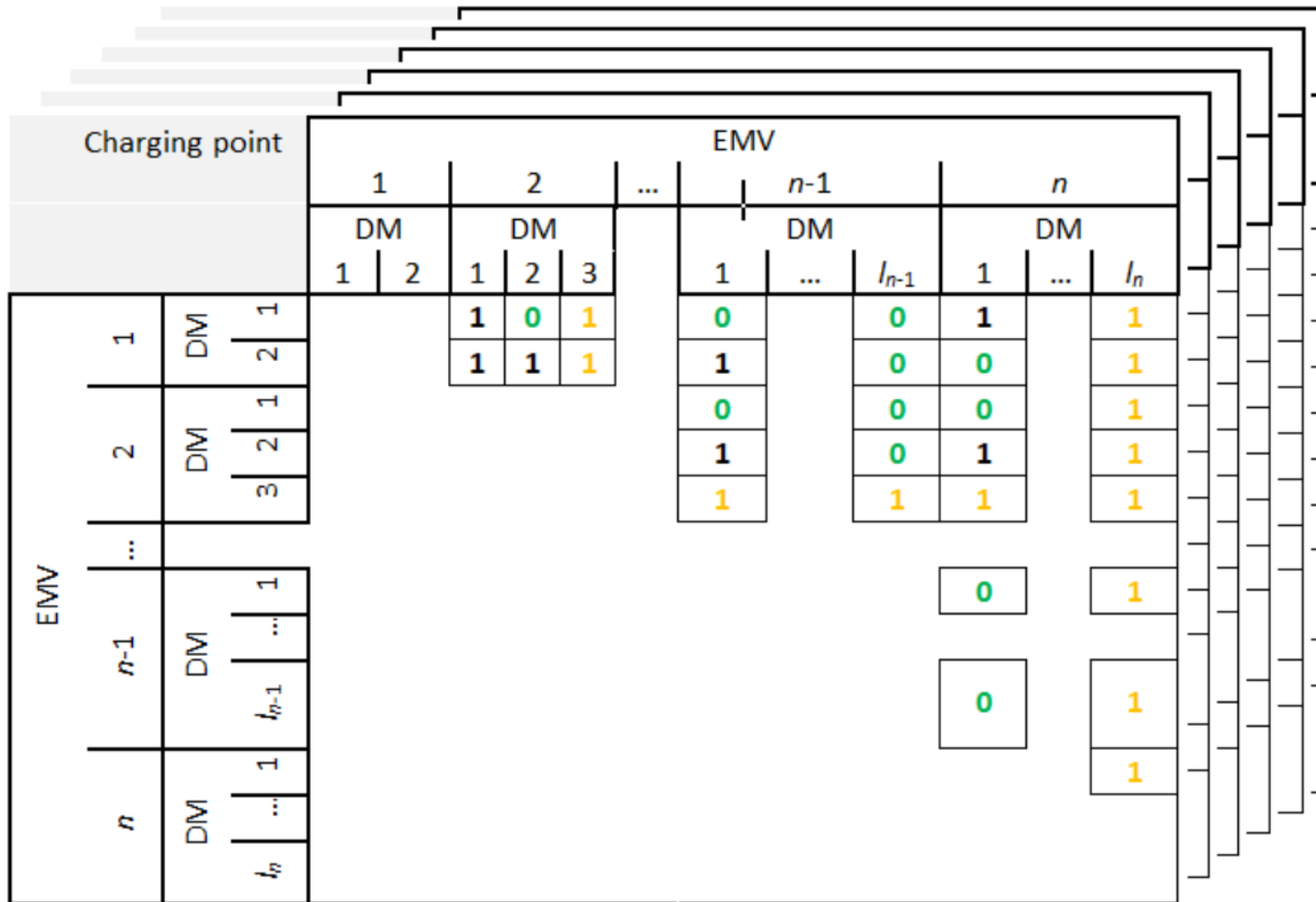
We have a number of EVs and a number of charging stations along a motorway. Each EV operates in a specific driving mode.

The objective is: Find all the cases (driving mode and charging station to be reached by each EV) so that there will be no conflicts at the station.



Charging Electric Cars on a Motorway

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QUBO Formulation of the initial Problem

Let $\mathbf{r}_i(\mathbf{t}_k)$ be the charging current rate for EV i at time \mathbf{t}_k and \mathbf{N} the number of currently plugged EVs

Cost Function

$$C(r) = \sum_{k=0}^T \left(\sum_{i=0}^N r_i(t_k) \right)^2 + \rho \sum_{i=0}^N \left(\sum_{t_k=0}^{\tau_i^{end}} (V \cdot r_i(t_k) \cdot \Delta T) - e_i \right)^2$$

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Let $\delta_{ik} = 1$ if $t_k \leq \tau_i^{end}$
Else $\delta_{ik} = 0$

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$$C(r) = \sum_{k=0}^T \left(\sum_{i=0}^N r_i(t_k) \right)^2 + \rho \sum_{i=0}^N \left(\sum_{t_k=0}^T (V \cdot \delta_{ik} \cdot r_i(t_k) \cdot \Delta T) - e_i \right)^2$$

QUBO Formulation of the initial Problem



$$\begin{aligned} C(r) = & \sum_{i=0}^N \sum_{k=0}^T (r_i(t_k))^2 + 2 \sum_{i,j=0}^N \sum_k^T r_i(t_k) r_j(t_k) + \rho \sum_{i=0}^N \sum_{k=0}^T (V \delta_{ik} r_i(t_k) \Delta T)^2 + \\ & + \rho \sum_{i=0}^N 2 \sum_{k,l=0}^T V^2 \Delta T^2 \delta_{ik} r_i(t_k) \delta_{il} r_i(t_l) + \rho \sum_{i=0}^N 2 e_i \sum_{k=0}^T V \Delta T \delta_{ik} r_i(t_k) \end{aligned}$$

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Convert to binary.

Simplify set of allowable current to $\{0,16,32,48\}$ \longrightarrow $r_i(t_k) = 16 \sum_{q=0}^1 2^q x_{ikq}$

QUBO Formulation of the initial Problem



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$$+ \rho \sum_{i=0}^N 2 \sum_{k,l=0}^T V^2 \Delta T^2 \delta_{ik} r_i(t_k) \delta_{il} r_i(t_l) + \rho \sum_{i=0}^N 2e_i \sum_{k=0}^T V \Delta T \delta_{ik} r_i(t_k)$$

Convert to binary.

Simplify set of allowable current to $\{0,16,32,48\} \longrightarrow r_i(t_k) = 16 \sum_{q=0}^1 2^q x_{ikq}$

$$C(x) = \sum_{i=0}^N \sum_{k=0}^T \left(16 \sum_{q=0}^1 2^q x_{ikq} \right)^2 + 2 \sum_{i,j=0}^N \sum_k^T 256 \sum_{q=0}^1 2^q x_{ikq} \sum_{q=0}^1 2^q x_{jkq} + \rho \sum_{i=0}^N \sum_{k=0}^T \left(V \delta_{ik} 16 \sum_{q=0}^1 2^q x_{ikq} \Delta T \right)^2 +$$

$$+ \rho \sum_{i=0}^N 2 \sum_{k,l=0}^T 256 V^2 \Delta T^2 \delta_{ik} \sum_{q=0}^1 2^q x_{ikq} \delta_{il} \sum_{q=0}^1 2^q x_{ilq} + \rho \sum_{i=0}^N 2e_i \sum_{k=0}^T V \Delta T \delta_{ik} 16 \sum_{q=0}^1 2^q x_{ikq}$$



ClusterVQE

Split vector x to n clusters of equal length x_1, x_2, \dots, x_n

$$\text{Then } x^T Q x = [x_1^T \ x_2^T \ \dots \ x_n^T] \begin{bmatrix} Q_{11} & Q_{12} & \dots & Q_{1n} \\ 0 & Q_{22} & \dots & Q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & Q_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \rightarrow$$

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$$x^T Q x = x_1^T Q_{11} x_1 + x_2^T Q_{22} x_2 + \dots + x_n^T Q_{nn} x_n + x_1^T Q_{12} x_2 + \dots + x_1^T Q_{1n} x_n + \dots + x_2^T Q_{2n} x_n + \dots$$

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$$\text{Then } x^T Q x = [x_1^T \ x_2^T \ \dots \ x_n^T] \begin{bmatrix} Q_{11} & Q_{12} & \dots & Q_{1n} \\ 0 & Q_{22} & \dots & Q_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & Q_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \rightarrow$$

$$x^T Q x = x_1^T Q_{11} x_1 + x_2^T Q_{22} x_2 + \dots + x_n^T Q_{nn} x_n + x_1^T Q_{12} x_2 + \dots + x_1^T Q_{1n} x_n + \dots + x_2^T Q_{2n} x_n + \dots$$



$$E[x^T Q x] = E[x_1^T Q_{11} x_1] + E[x_2^T Q_{22} x_2] + \dots + E[x_n^T Q_{nn} x_n] + E[x_1^T Q_{12} x_2] + \dots + E[x_1^T Q_{1n} x_n] + \dots + E[x_2^T Q_{2n} x_n] + \dots$$



$$E[x^T Q x] = \sum_{i=1}^n E[x_i^T Q_{ii} x_i] + \sum_{i=1}^n \sum_{j \neq i}^n E[x_i^T Q_{ij} x_j]$$

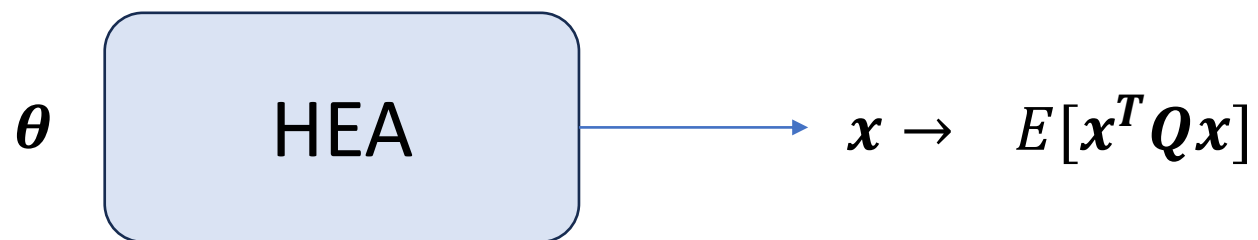
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HEA

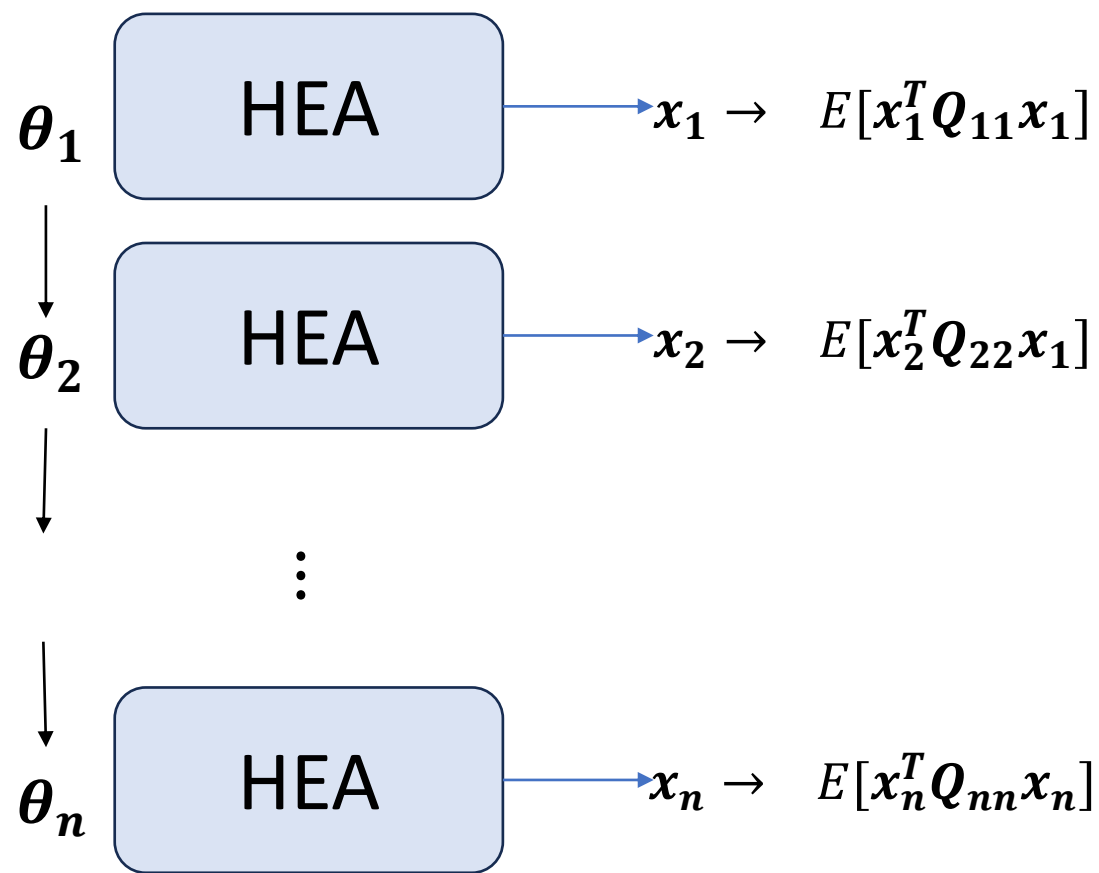


$$x \rightarrow E[x^T Q x]$$

VQE



VQE



$E[x^T Q x] = \dots$

ClusterVQE

Apply ClusterVQE to EV charging problem

$$\begin{array}{c} \text{Horizon} \\ \left[\begin{array}{cccc} r_1(t_1) & r_1(t_2) & r_1(t_3) & r_1(t_4) \\ r_2(t_1) & r_2(t_2) & r_2(t_3) & r_2(t_4) \\ r_3(t_1) & r_3(t_2) & r_3(t_3) & r_3(t_4) \\ r_4(t_1) & r_4(t_2) & r_4(t_3) & r_4(t_4) \end{array} \right] \end{array} \begin{array}{c} \text{Num of EVs} = N \\ \left. \begin{array}{c} \\ \\ \\ \end{array} \right\} \end{array} \begin{array}{c} r_2 \end{array}$$

Advantages of ClusterVQE

- Reduce the number of qubits by a factor of N
- Increase the number of measurements by a factor of N

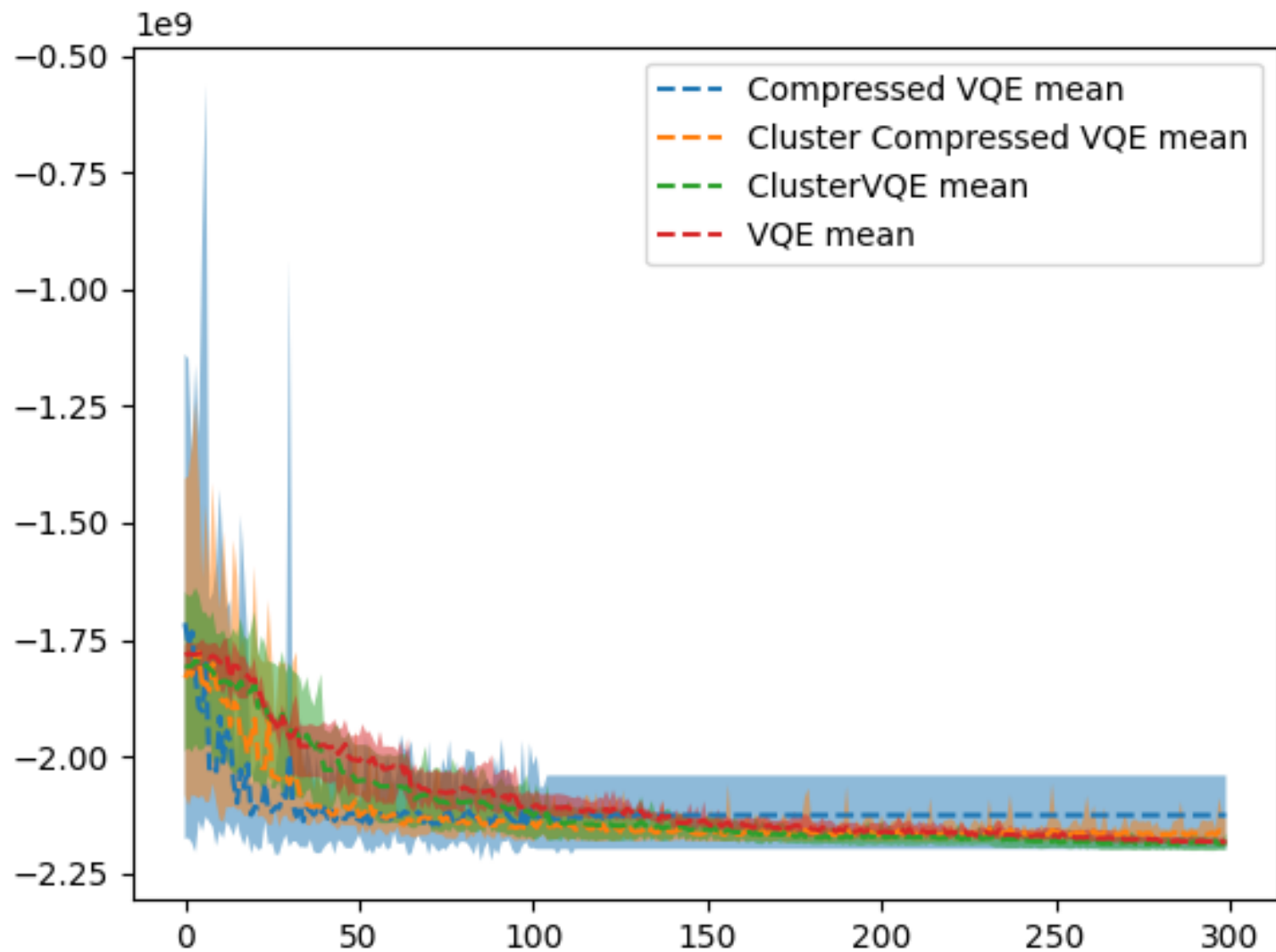
*Can also be used with qubit compression

Comparison

For an example with *T=4 hours* , *Evs = 4* and **2 layer** HAE

Algorithm	#Classical Variables	#Qubits	#measurements	#parameters θ	Elapsed time (sec)
VQE	16	32	10000	64	610
CompressedVQE	16	6	10000	12	6
ClusterVQE	16	8	40000	64	250
ClusterCompressedVQE	16	4	40000	32	27

Comparison



Comparison

CompressedVQE

16	48	48	0	26880/26880
16	48	48	0	26880/26880
48	16	48	0	26880/26880
32	0	0	0	7680/7680

ClusterCompressedVQE

48	48	16	0	26880/26880
32	32	48	0	26880/26880
32	48	48	0	32720/26880
32	0	0	0	7680/7680

ClusterVQE

48	48	16	0	26880/26880
48	16	48	0	26880/26880
32	32	48	0	26880/26880
32	0	0	0	7680/7680

VQE

32	48	32	0	26880/26880
48	48	16	0	26880/26880
32	32	48	0	26880/26880
32	0	0	0	7680/7680