

A fast electric vehicle planner using clustering

A new speedup technique to find electric vehicle shortest paths

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Outline

- 1 Motivation: Why an EV Planner
- 2 Base planner
- 3 Grouping stations into clusters
- 4 Evaluation
- 5 Conclusion

Advantages of EV over conventional vehicles



Less pollution



Less noisy

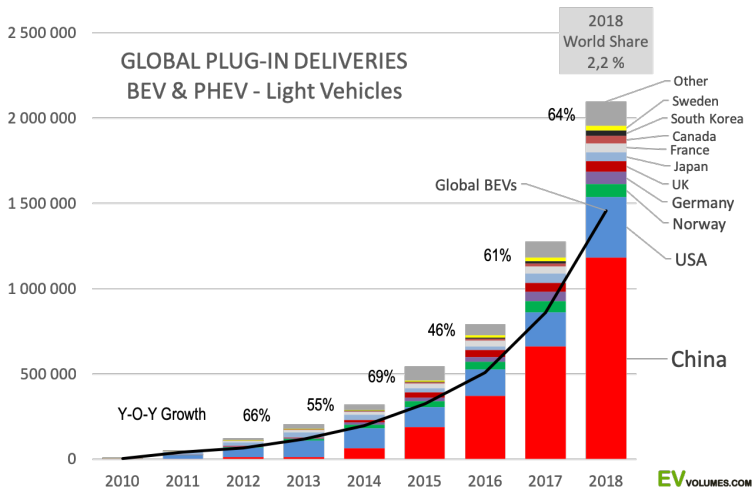


Cheaper in the long run



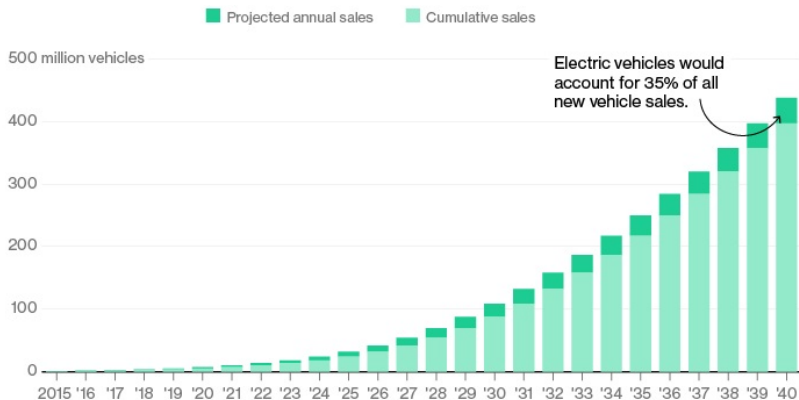
Less maintenance

Global EV market 2010–2018¹



¹<http://www.ev-volumes.com/country/total-world-plug-in-vehicle-volumes/>

EV sales forecast²



Sources: Data compiled by Bloomberg New Energy Finance, Marklines



²Bloomberg, February 25th, 2016, <https://www.bloomberg.com/features/2016-ev-oil-crisis/>

Comparison between a conventional vehicle and an EV



	Honda Civic	Nissan Leaf
Price	20 000 \$	36 000 \$ ³
Range	750 km	242 km
Refueling/Charging time	3 min	30 min
Gas/Charging stations ⁴	2924	115

³Excluding governmental subsidies

⁴In Québec in 2018

Research problem

- The number of EV is increasing;
- The number of charging stations is increasing;
- Many paths need recharges to be feasible.

Objective

The objective is to have an EV planner that:

- 1 considers intermediate recharges at charging stations;
- 2 fastly computes an optimal path.

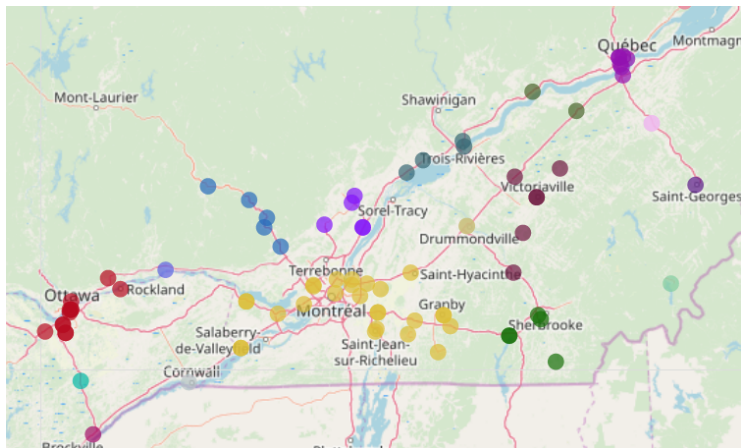
Existing techniques

- Use of the *Energy-A** algorithm for EV travel. Considers the recharge of the battery by regenerative braking, but does not consider charging stations⁵.
- EV planning considering charging stations⁶.

⁵Sachenbacher, M., M. Leucker, A. Artmeier and J. Haselmayr (2011). Efficient Energy-Optimal Routing for Electric Vehicles. In Proceedings of the AAAI, pp. 1402–1407. AAAI Press.

⁶Baouche, F., R. Billot, R. Trigui and N. E. El Faouzi (2014). Efficient allocation of electric vehicles charging stations. IEEE-ITSM 6(3), 33–43.

Idea: Grouping nearby stations into clusters



Formalism

Road map

The map is represented by an oriented valued graph (V, E, τ) and by a set S where:

- V is the set of locations considered on the map (nodes);
- E is the set of road segments (edges);
- $\lambda: E \rightarrow \mathbb{R}^+$ gives the length of the edges;
- S is the set of charging stations (we assume that $S \subseteq V$).

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EV planning

An EV planning problem is defined by the tuple $(M, \rho, \alpha, \omega)$ where:

- M is the road map;
- $\rho \in \mathbb{R}^+$ is the EV range (in km);
- $\alpha, \omega \in V$ are the departure and arrival nodes.

Base algorithm

Algorithm

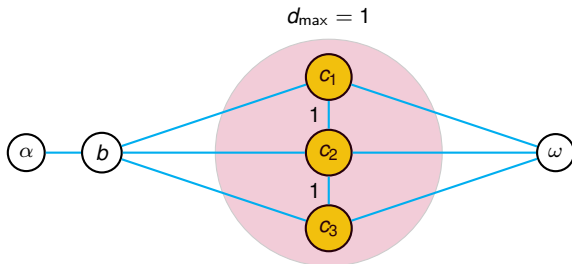
To find the most time-efficient EV plan from α to ω :

- A matrix of the distances between charging stations is precomputed;
- The distance between every stations to α , and between ω to every stations are computed (Dijkstra's algorithm performed twice);
- A complete graph (V', E') is constructed, where $V' = S \cup \{\alpha, \omega\}$.
- Edges longer than ρ are removed from the graph;
- A* algorithm is executed from α to ω on the new graph.

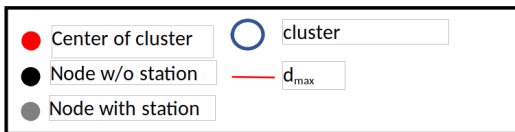
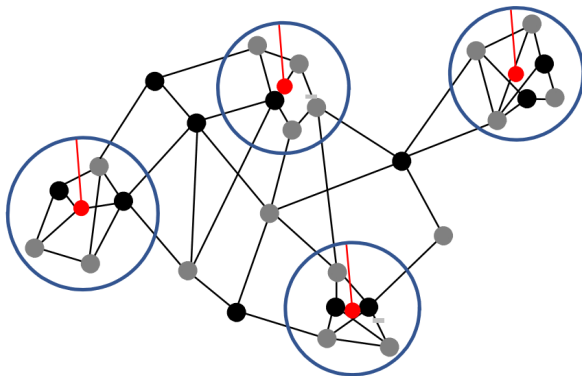
The asymptotic time complexity is $\mathcal{O}(|V| \log |V| + |E|)$.

The idea

- Regroup close stations into a single node;
- Consider a parameter $d_{\max} \in \mathbb{R}$ representing the **radius** of the generated clusters.



Example of the resulting graph



Clustering charging stations

Algorithm Clustering charging stations on the map

- 1: Find the two closest charging stations $s_1, s_2 \in S$
 - 2: **while** $\text{distance}(s_1, s_2) \leq d_{\max}$ **do**
 - 3: Find the node m that is midway between s_1 and s_2
 - 4: Find $C = \{s \in S \mid \text{dist}(s, m) \leq d_{\max}\}$ ▷ $s_1, s_2 \in C$
 - 5: $S \leftarrow (S \setminus C) \cup \{m\}$
 - 6: Find the two closest charging stations $s_1, s_2 \in S$
 - 7: **end while**
-

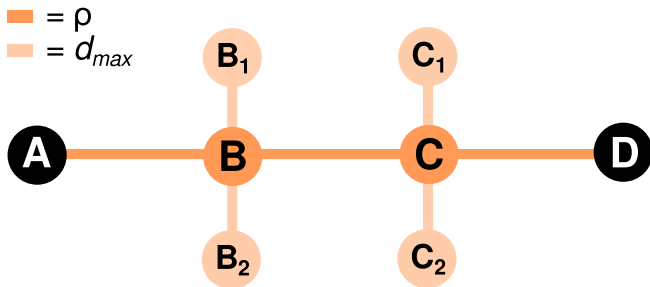
Time complexity: $\mathcal{O}(K(|S|^2 + |V|))$ (where K is the number of generated clusters)

Clustering charging stations — A problem

Adjusting the considered range

For the path returned by the planner to be feasible, the considered range must be:

- $R' = R - d_{\max}$ for the first and last clusters;
- $R' = R - 2d_{\max}$ between clusters.



Complete planner

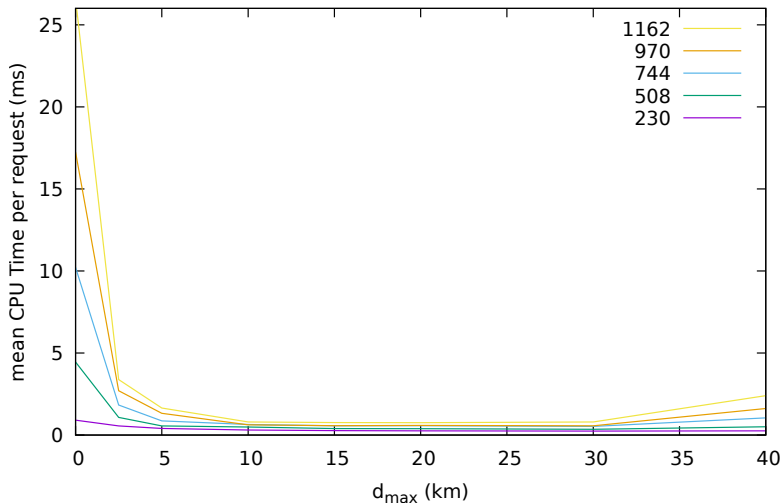
Algorithm EV Path-Planner including charging stations clustering

- 1: Generate the charging stations clustering ▷ using Algorithm 1
 - 2: Compute the matrix D of pairwise distance between clusters' center
 - 3: Construct the complete s-graph (S, E_S)
 - 4: **for each** request (ρ, α, ω) **do**
 - 5: Execute Dijkstra from α on the original graph
 - 6: Execute Dijkstra from ω on the transposed graph
 - 7: Add α and ω to the s-graph including edges of length $\leq \rho$
 - 8: Execute A* on the s-graph from α to ω
 - 9: **if** the returned path is not feasible **then** ▷ Amortized strategy
 - 10: Execute A* on the s-graph (without clusters) from α to ω
 - 11: **end if**
 - 12: Find the complete path from the path in the s-graph
 - 13: **end for**
-

Testing methodology

- The real map data comes from the OpenStreetMap project.
- The territory used is the Quebec province, Canada:
 - 2 580 388 nodes
 - 5 225 348 edges
 - 1162 charging stations
- 1000 requests:
 - EV range between 100 and 550 km
 - Travel distance between 200 and 1500 km

Results 1/2: Graph of the CPU time vs the d_{\max} parameter



Results 2/2: Table showing the impact of clustering

Parameters		Clusters		Base version		Amortized version	
Stations	d_{\max}	Clusters	JDIR	FR	CT	FR	CT
#	km	#	%	%	ms	%	ms
1162	0.0	1162	0.0	0.0	26.563	0	26.563
1162	2.5	487	0.0	0.0	3.385	0	3.385
1162	5.0	342	0.2	0.4	1.541	0	1.647
1162	10.0	236	0.2	0.8	0.588	0	0.801
1162	15.0	188	0.6	0.9	0.523	0	0.762
1162	20.0	150	1.0	1.4	0.382	0	0.754
1162	30.0	111	2.3	2.0	0.265	0	0.796
1162	40.0	87	2.8	8.2	0.226	0	2.404

JDIR: Journey duration increase rate; **FR**: Failure rate; **CT**: Computation time

Conclusion

- Finding the optimal solution for EV Path-Planning has a high computational cost;
- We presented a planner that uses graph clustering as a speedup technique;
- Our technique decreased by a factor of 35 the mean time of computation.

Questions ?

Acknowledgements



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