

### **Topic: How A\* Search Algorithm is used in Google Maps**

Embarking on a road trip toward an unfamiliar destination can feel like a challenging task when equipped with only a compass and a foldable map bought from a convenience store. As I have lived most of my life in the digital age, I am certain that I will get lost in navigating through unfamiliar streets full of complicated intersections and unforeseeable traffic congestion without any navigation apps at hand. Fortunately, less than a decade after I was born, an application was programmed that revolutionized how I could explore the world. This technology is none other than Google Maps. Since then, Google Maps has processed about petabytes of data to guide over a billion active monthly users worldwide [1,2]. Unknown to most users though, behind the massive yet seemingly effortless operation of Google Maps is an algorithm known as A\* (pronounced as “A-Star”) search algorithm. In this essay, I will discuss some fundamental concepts that govern the A\* search algorithm and how these concepts are applied in Google Maps.

The A\* search algorithm is an informed path search algorithm known for its ability to accurately and efficiently find the optimal path from the origin to the destination in a complex search space. Introduced in the mid-1960s by Hart, Nilsson, and Raphael, it has since found several applications in robotics, and mapping applications like Google Maps [3,4]. Crucial to the algorithm’s success is its ability to minimize the total path cost by considering both the actual cost from the initial state to the current state and the estimated heuristic cost from the current state to the goal state. This is typically represented in terms of the evaluation function of node  $n$ ,

$$f(n) = g(n) + h(n)$$

where  $f(n)$  is the f-cost or the estimated cost of the optimal path from the current node  $n$  to the goal node,  $g(n)$  is the g-cost or the actual path cost from the root node to the current node  $n$ , and  $h(n)$  is the h-cost or the estimated cost of the shortest path from the current node  $n$  to the goal node. By expanding leaf nodes with the lowest f-cost, the algorithm effectively balances the exploration of new paths and promising paths until it finds the optimal path toward the goal. This makes the A\* search algorithm not only complete but also optimal under specific conditions, including operating on a locally finite graph with finite branching factors and ensuring all action costs are non-negative ( $\epsilon > 0$ , where  $\epsilon$  is the lower bound of the action cost). Moreover, the heuristic function used must be admissible, indicating that it never overestimates the actual cost to reach the goal ( $h(n) \leq$  actual distance from node  $n$  to goal node). However, to apply the A\* search algorithm in a practical way, the right balance between search performance and heuristic accuracy must be achieved. This balance allows applications that utilize A\* search algorithms like Google Maps to efficiently and accurately provide their users with optimal path suggestions [4].

The A\* search algorithm, in the context of Google Maps, finds the most optimal path between the two geographical points, taking into account user preferences like travel modes, and factors in the environment such as traffic conditions, road restrictions, elevation changes, and accidents. This can be explained by representing the road networks as a graph data structure, where road intersections and geographic locations are represented as nodes and road segments are represented as edges with weights typically corresponding to the travel time or the distance of the road segment. In the graph, the algorithm systematically evaluates promising paths from the starting point (root node) to the destination point (goal node) as specified by the user. To choose the optimal path between these two points, it repeatedly expands leaf nodes with the lowest associated f-cost until it reaches the goal node, where as discussed previously, the f-cost is the combination of the h-cost and the g-cost. Choosing the appropriate heuristic function to use

for calculating the h-cost is critical, as it directly affects the balance between heuristic accuracy and search performance. Because of this, Google Maps has employed varying heuristics for different types of road networks, terrains, and scenarios. For example, the calculation of the straight-line distance between the starting point and the destination point can be an appropriate heuristic function in a highway setting with cars driving at average uniform speeds. However, more advanced heuristics may be required in more complex terrains with cars driving at varying speeds due to real-time traffic. In all cases though, Google Maps incorporates historical data about traffic trends, elevation changes, road restrictions, etc., and real-time data about traffic conditions, weather disruptions, accidents, etc. that can enhance the calculation of the optimal route [4-6].

Dechter and Pearl once said and I quote, “No other optimal algorithm is guaranteed to expand fewer nodes than A\*” [5]. Such a claim is backed up by the algorithm’s ability to blend completeness, cost optimality, accuracy, and efficiency for finding optimal routes in the dynamic, customizable, and scalable world of navigation with Google Maps. Although the A\* search algorithm has transformed modern navigation for the better, it still has challenges that must be acknowledged to further improve its search strategy. One limitation that I have encountered in my travels is the inaccurate mapping data used to represent isolated rural places, leading to unexpected navigation errors. When this happens, instead of relying on Google Maps, I resort to asking the locals for directions. All in all, the seamless and real-time navigation of Google Maps is proof of the capability of the A\* search algorithm in optimal pathfinding.

## References:

- [1] Russell, Ethan. “9 Things to Know About Google’s Maps Data: Beyond the Map.” Google Cloud Blog, 30 Sept. 2019, [cloud.google.com/blog/products/maps-platform/9-things-know-about-googles-maps-data-beyond-map](https://cloud.google.com/blog/products/maps-platform/9-things-know-about-googles-maps-data-beyond-map).
- [2] Gordon, Olivia. “How Much Data Does Google Maps Use – User Guide 2023.” Talk Home Blog - Stories, Lists, Tips & Tricks, June 2023, [blog.talkhome.co.uk/technology/how-much-data-does-google-maps-use](https://blog.talkhome.co.uk/technology/how-much-data-does-google-maps-use).
- [3] Hart, Peter E., Nils J. Nilsson, and Bertram Raphael. "A formal basis for the heuristic determination of minimum cost paths." IEEE transactions on Systems Science and Cybernetics 4.2 (1968): 100-107.
- [4] Russell, Stuart J. Artificial intelligence a modern approach. Pearson Education, Inc., 2010.
- [5] Ganga. “The Algorithms Behind the Working of Google Maps | CodeChef.” CodeChef, 21 Oct. 2021, [blog.codechef.com/2021/08/30/the-algorithms-behind-the-working-of-google-maps-dijkstras-and-a-star-algorithm](https://blog.codechef.com/2021/08/30/the-algorithms-behind-the-working-of-google-maps-dijkstras-and-a-star-algorithm).
- [6] GeeksforGeeks. “A\* Search Algorithm.” GeeksforGeeks, Mar. 2023, [www.geeksforgeeks.org/a-search-algorithm](https://www.geeksforgeeks.org/a-search-algorithm).
- [7] Dechter, Rina, and Judea Pearl. The optimality of A. Springer New York, 1988.