



# Smart shopping *with* search algorithms

presented by group 9

Tran Tue Khang - 11247303

Nguyen Khoi Nguyen - 11247331

Le Quang - 11247343

Nguyen Duc Huy - 11247297





# Introduction



# Real-world motivation & contributions



time-saving



saving fuel



less CO<sub>2</sub>



# Problems formulation

- ✳ **State**

current market, set of items collected

- ✳ **Goal**

all required items collected

- ✳ **Operators**

move to a market selling any missing items; buy them

- ✳ **Operators**

travel distance (Manhattan or Euclidean)



# Heuristic

```
H(pos, collected):
    if collected already contains all required items: return 0
    max_d ← 0
    for each missing item k:
        d_k ← min over markets j that sell k of distance(pos, j)
        max_d ← max(max_d, d_k)
    return max_d
```

# Algorithms overview



## Greedy-nearest

go to closest market with any missing item  
→ fast, not optimal



## GBFS

expand lowest  $h(n)$   
→ can drift, ignores  $g(n)$



## A\*

expand lowest  $f(n) = g + h$   
→ optimal with admissible  $h$

# Method details

```
def greedy_nearest(home, required, markets, sells):
    """
    Always move to the nearest market that sells at least one missing item.
    markets: dict[name] -> (x,y)
    sells:   dict[name] -> set(items)
    """
    pos = markets[home]; collected=set(); path=[home]; cost=0
    while collected != required:
        candidates = [(manhattan(pos, markets[j]), j)
                      for j in sells if (required - collected) & sells[j]]
        d, j = min(candidates)                      # nearest useful market
        cost += d
        pos = markets[j]
        collected |= sells[j]
        path.append(j)
    return path, cost
```

## \* Greedy-nearest

# Method details

```
def gbfs(home, required, markets, sells):
    start = (home, frozenset())
    pq = [(H(markets[home]), required, set(), markets, sells), 0, [home], start]
    seen = set()
    while pq:
        h, g, path, (name, coll_f) = heappop(pq)
        if (name, coll_f) in seen:
            continue
        seen.add((name, coll_f))
        collected = set(coll_f)
        if collected == required:
            return path, g
        for j in sells:
            if (required - collected) & sells[j]:
                newC = collected | sells[j]
                ng = g + manhattan(markets[name], markets[j])
                nh = H(markets[j], required, newC, markets, sells)
                heappush(pq, (nh, ng, path+[j], (j, frozenset(newC))))
```

# Method details

```
def astar(home, required, markets, sells):
    start = (home, frozenset())
    gscore = {start: 0}
    pq = [(H(markets[home]), required, set(), markets, sells), start, [home]])
    seen=set()
    while pq:
        f, (name, coll_f), path = heappop(pq)
        if (name, coll_f) in seen:
            continue
        seen.add((name, coll_f))
        collected = set(coll_f)
        g = gscore[(name, coll_f)]
        if collected == required:
            return path, g
        for j in sells:
            if (required - collected) & sells[j]:
                newC = collected | sells[j]
                ng = g + manhattan(markets[name], markets[j])
                s = (j, frozenset(newC))
                if ng < gscore.get(s, float('inf')):
                    gscore[s] = ng
                    nh = H(markets[j], required, newC, markets, sells)
                    heappush(pq, (ng + nh, s, path+[j]))
```



## \* Assortments

Each market sells a subset of 12 items.

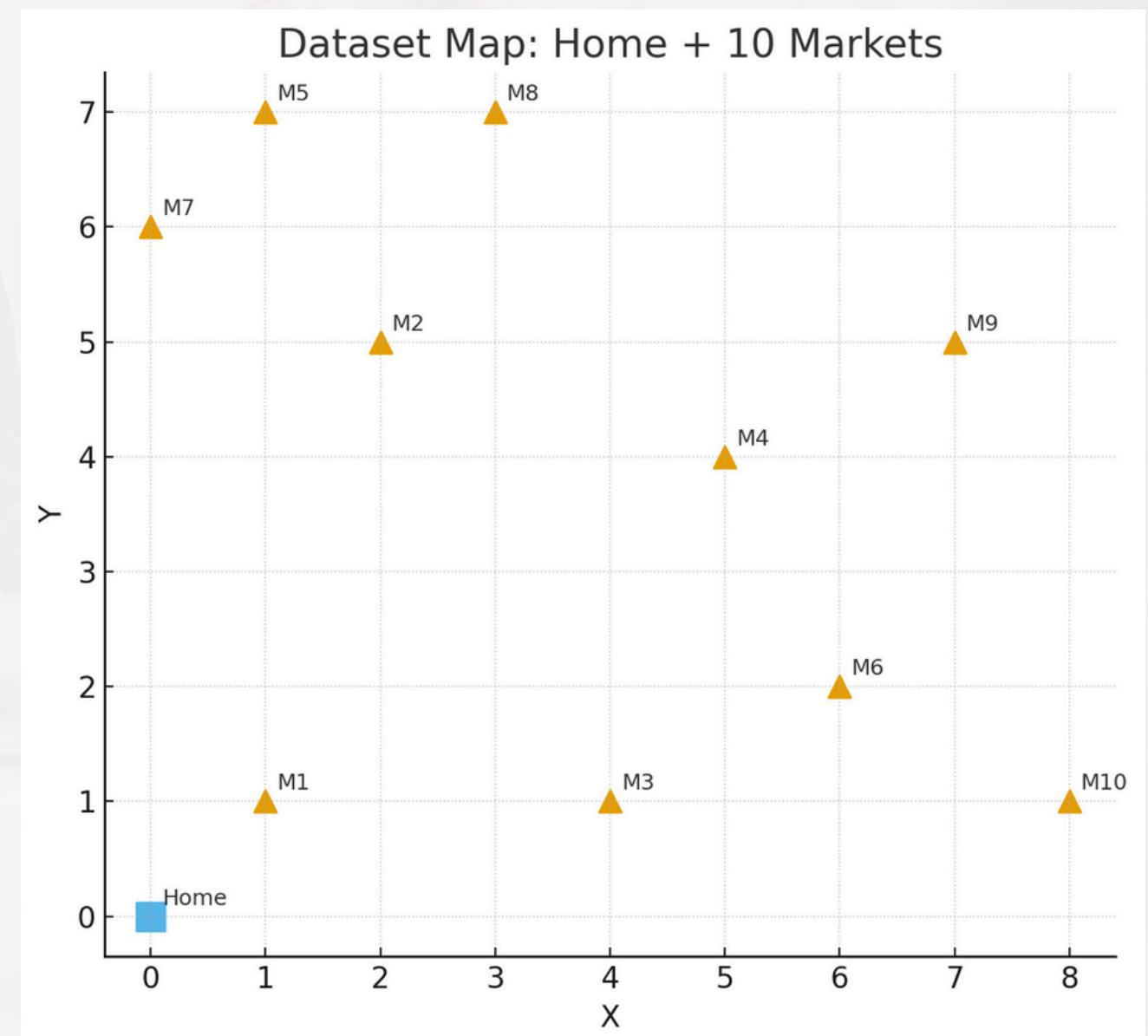
Examples:

- M1: milk, eggs
- M4: fruit, vegetables, oil
- M7: vegetables, tofu, sugar
- M10: noodles, salt, spices

# Data & test cases

## \* Shopping list

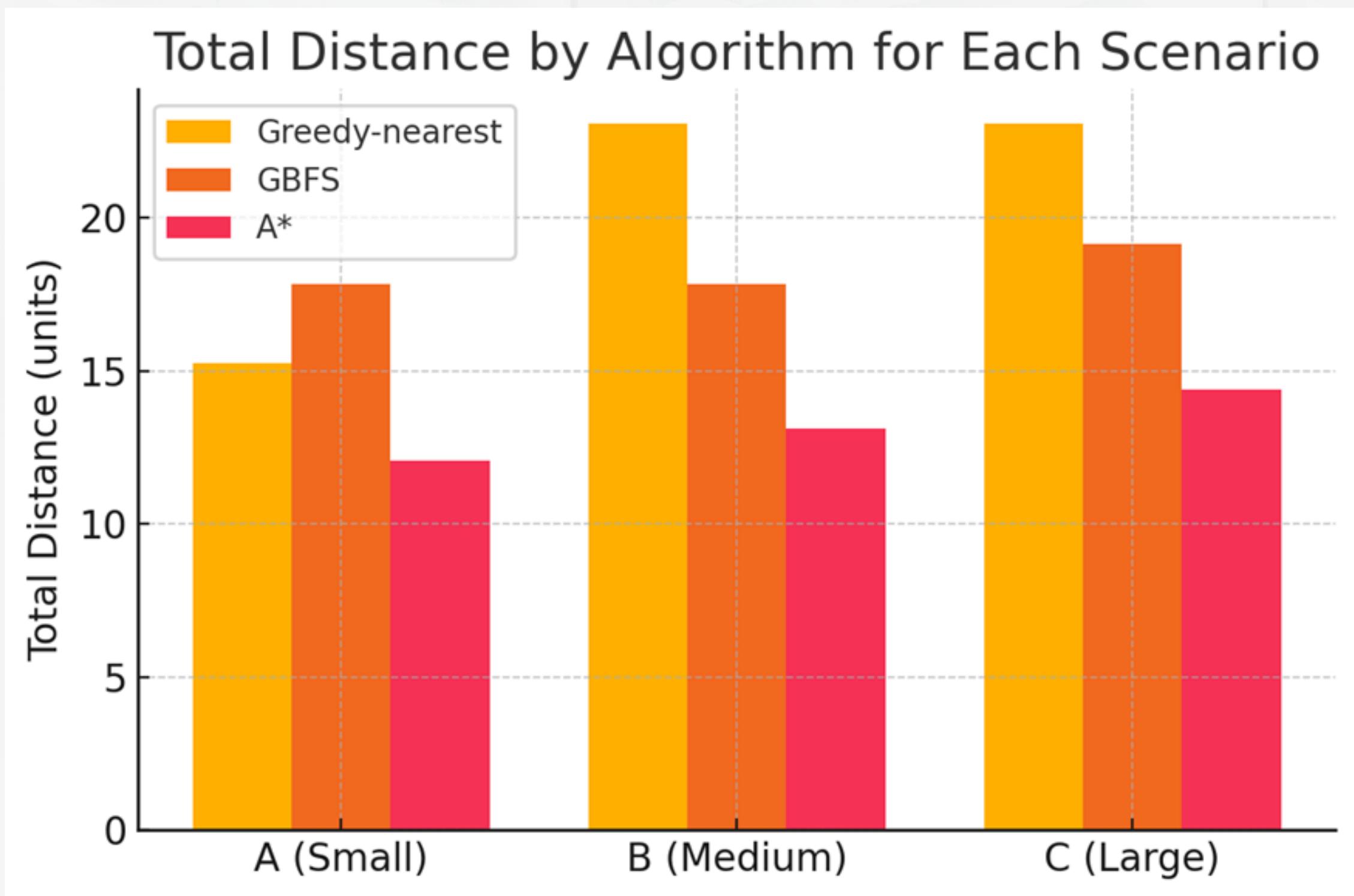
- Case A: 8 items
- Case B: 10 items
- Case C: 12 items



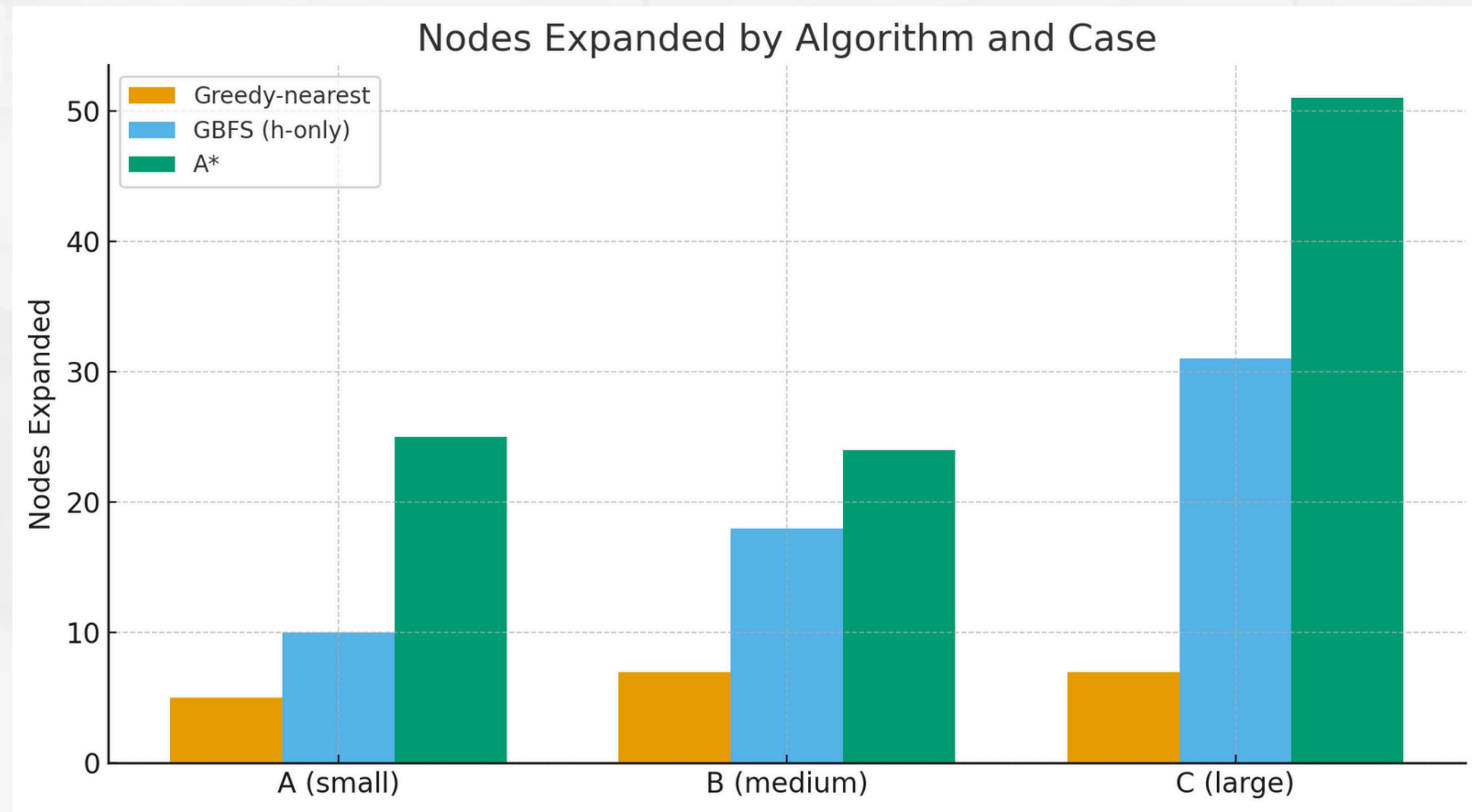
# Results

Scenario	Algorithm	Total Distance (Cost)	Nodes Expanded (Depth)	Runtime (s)	Success	Route
A	Greedy-nearest	1.525	6	38	ĐÚNG	Home->M1->M3->M6->M4->M9->M10
	GBFS	1.783	87	3.781	ĐÚNG	Home->M3->M9->M8->M5->M2
	A*	1.205	80	3.175	ĐÚNG	Home->M1->M4->M3->M6->M10
B	Greedy-nearest	2.306	7	35	ĐÚNG	Home->M1->M3->M6->M4->M9->M10->M8
	GBFS	1.783	91	3.798	ĐÚNG	Home->M3->M9->M8->M5->M2
	A*	1.312	129	5.884	ĐÚNG	Home->M2->M1->M6->M8->M5
C	Greedy-nearest	2.306	7	33	ĐÚNG	Home->M1->M3->M6->M4->M9->M10->M8
	GBFS	1.916	78	4.094	ĐÚNG	Home->M3->M2->M8->M5->M9
	A*	1.439	203	9.925	ĐÚNG	Home->M1->M6->M2->M3->M4->M7->M5

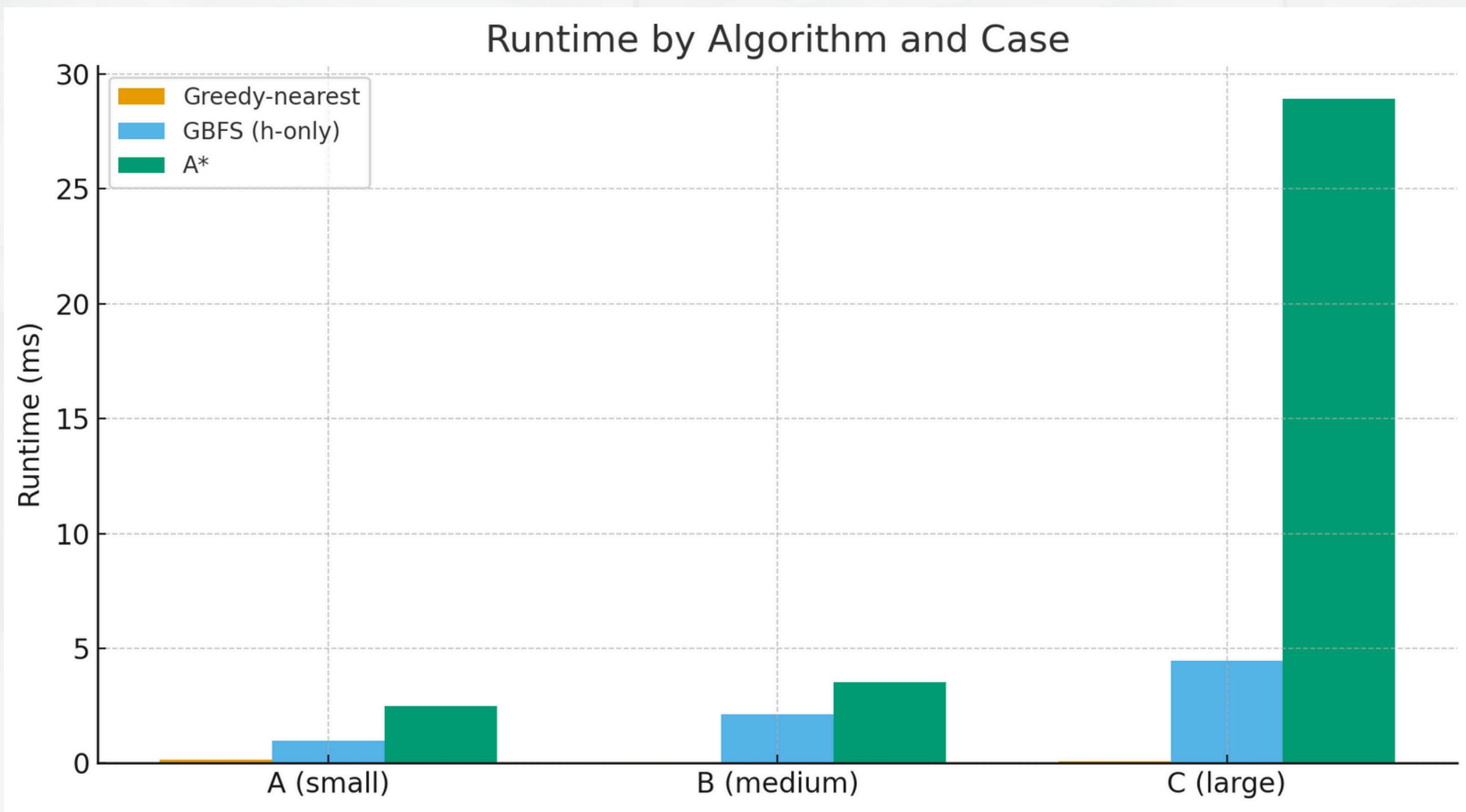
# Results



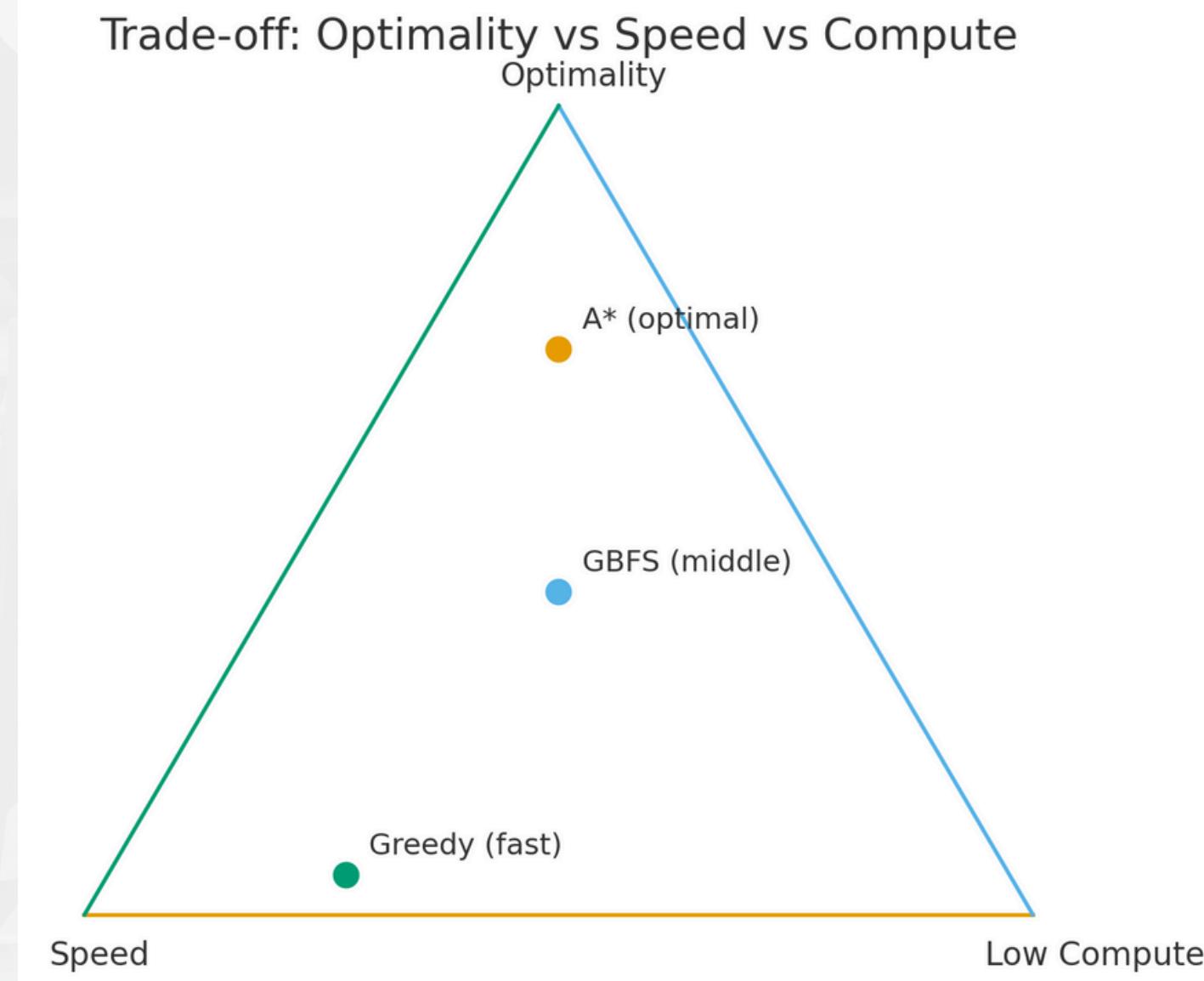
# Results



# Results



# Discussion, limitations & conclusion



# The end.

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