



Optimize[®]

by The Algorithm People

It's Not Just About Distance

Exploring the complex world of logistics and where optimization, AI & machine learning have an impact





Overview

- 🔗 Who Are We?
Who are Optimize and what we do
- 🔗 Optimisation Example
Demonstrate optimisation
- 🔗 AI & Machine Learning
Where is it used to make a difference
- 🔗 The Future
Where can AI and ML lead In the context of logistics





Who Are We?





Who am I?

- Dr Ross Conroy
- PhD In Artificial Intelligence
- 4 Years at FCDO
- 2 Years Completing KTP
- Now Head of R&D at Optimize since April



Team Optimize



What We Do ...



Optimisation

Our award-winning optimisation algorithms are proven to generate measurable savings and efficiencies of up to 30%.



Decarbonisation

We help fleets optimise, decarbonise and transform through our simple 5-step algorithmic evidenced based analysis approach.



Transformation

We assist fleets to move to fully automated planning and transition to electric and zero emission vehicles using algorithms.



Scheduling \neq Optimization



How We Do It

The screenshot displays the Optimize Route Planner interface. On the left, a map of Great Britain shows various job locations marked with red circles. A legend in the bottom-left corner identifies the symbols: red circle for Jobs, orange house for Depots, orange circle with slash for Job rejected, and blue circle for Job pre-allocated. The top navigation bar includes 'MY UTILITIES', 'ROUTE PLANNER' (selected), 'COMPANY', and 'OPTIMIZE NOW'. A 'Select Mode' dropdown is set to 'Map View'. A text box on the right provides instructions: 'Please use Select or Draw mode to allocate jobs to resources. You can still change the order of jobs by dragging and dropping jobs on the resource list'. Below this is a search bar and a table with the following data:

Resource ↑	Quantity 1	Quantity 2	Allocated Jobs	Action
1	0/16	0/10	0	ALLOCATE
10	0/26	0/12	0	ALLOCATE
11	0/26	0/12	0	ALLOCATE
12	0/26	0/12	0	ALLOCATE
2	0/16	0/10	0	ALLOCATE
3	0/16	0/10	0	ALLOCATE
4	0/16	0/10	0	ALLOCATE
5	0/16	0/10	0	ALLOCATE





73,000

Tonnes of CO2 Saved in 2022

24%

Average Reduction in Emissions & Costs

19%

Average Increase in Productivity & Utilisation





Monopoly Example



Monopoly Example Before

My transport planner

MY UTILITIES ROUTE PLANNER COMPANY ROSS CONROY

The route drawn is for visualisation only and may not be optimised for the most efficient navigated route.

Route statistics

Resource Name
EV1

Depot name Go	Total quantity 1 28	Total revenue £ N/A	Total time 07:33 h
Allocated Jobs 28	Total quantity 2 N/A	Total distance 45.28 mi	Total cost £ 48.78

Search

No. ↑	Company name	Quantity 1	Quantity 2	Postcode	Mobile status	Other
1	Old Kent Road	1	-			
2	Whitechapel Road	1	-			
3	Kings Cross Station	1	-			
4	The Angel, Islington	1	-			
5	Euston Road	1	-			
	Pentonville					



Monopoly Example (Optimized)

My
transport
planner

MY UTILITIES ROUTE PLANNER ▾ COMPANY ▾ ROSS CONROY ▾

The route drawn is for visualisation only and may not be optimised for the most efficient navigated route.

Route statistics

Resource Name
EV1

Depot name Go	Total quantity 1 28	Total revenue £ N/A	Total time 05:00 h
Allocated Jobs 28	Total quantity 2 N/A	Total distance 24.07 mi	Total cost £ 24.07

🔍 Search

No. ↑	Company name	Quantity 1	Quantity 2	Postcode	Mobile status	Other
1	Fleet Street	1	-	-	-	⋮
2	Strand	1	-	-	-	⋮
3	Bow Street	1	-	-	-	⋮
4	Free Parking	1	-	-	-	⋮
5	Coventry Street	1	-	-	-	⋮





More Constraints

EU driving laws

- Driving time
- Minimum unbroken rest times
- Max driving per week
- Max driving per fortnight

Max time in vehicle

- Fresh produce
- Takeaway delivery

Time windows

- Only receive deliveries at certain times
- Shift patterns
- Idle time
- Lateness penalties

Skills

- Tail lift
- Refrigeration
- Hazardous materials
- Wheelchair access

Capacities

- Volume
- Weight
- Palette footprints

Range

- Affected by weight on vehicle

Prerequisite job completion



What Does Optimal Look Like?

- Lowest mileage
- Least cost £££
- Fastest completion time
- Least number of vehicles used
- Lowest carbon emissions
- Lowest idle time
- Maximum number of jobs fulfilled
- Least re-fuelling / re-charging stops
- Resource utilisation (min / max)
- Maximum ROI





AI & Machine Learning

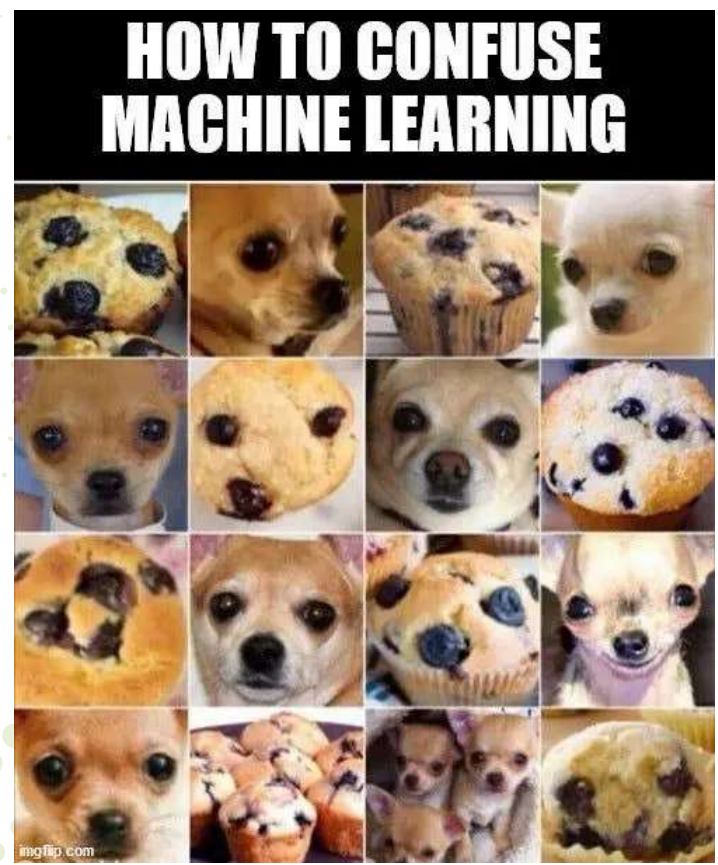


Pitfalls of AI & Machine Learning

People with no idea about AI saying it will take over the world:



My Neural Network:



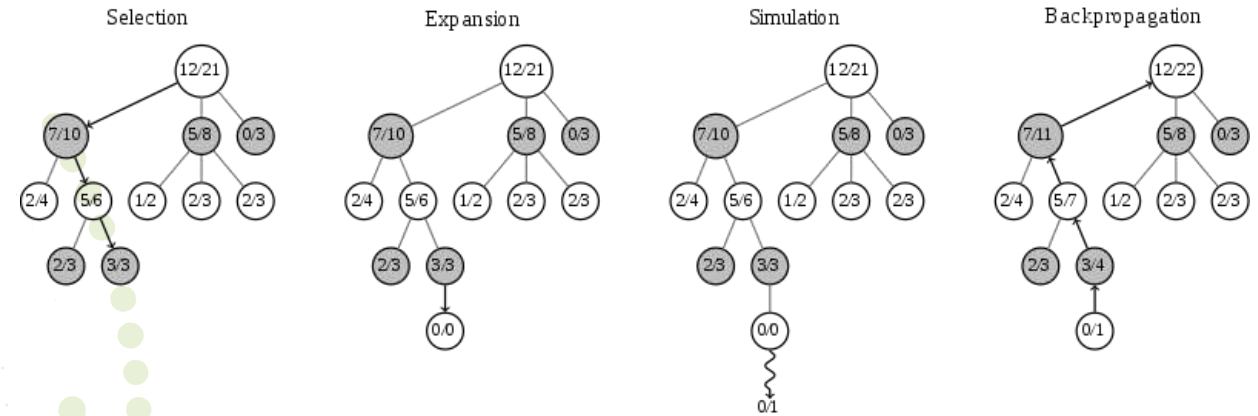


Selecting a Solver - Monte Carlo Tree Search

- No one size fits all solver
- Anytime Algorithm
- Algorithm of Algorithms
- How to determine which solver is best?
- Ideal methods change over time.
 - Early stages focus towards getting jobs onto vehicles
 - Late stages focus towards finding improvements



What is Monte Carlo Tree Search?



https://www.chessprogramming.org/Monte-Carlo_Tree_Search

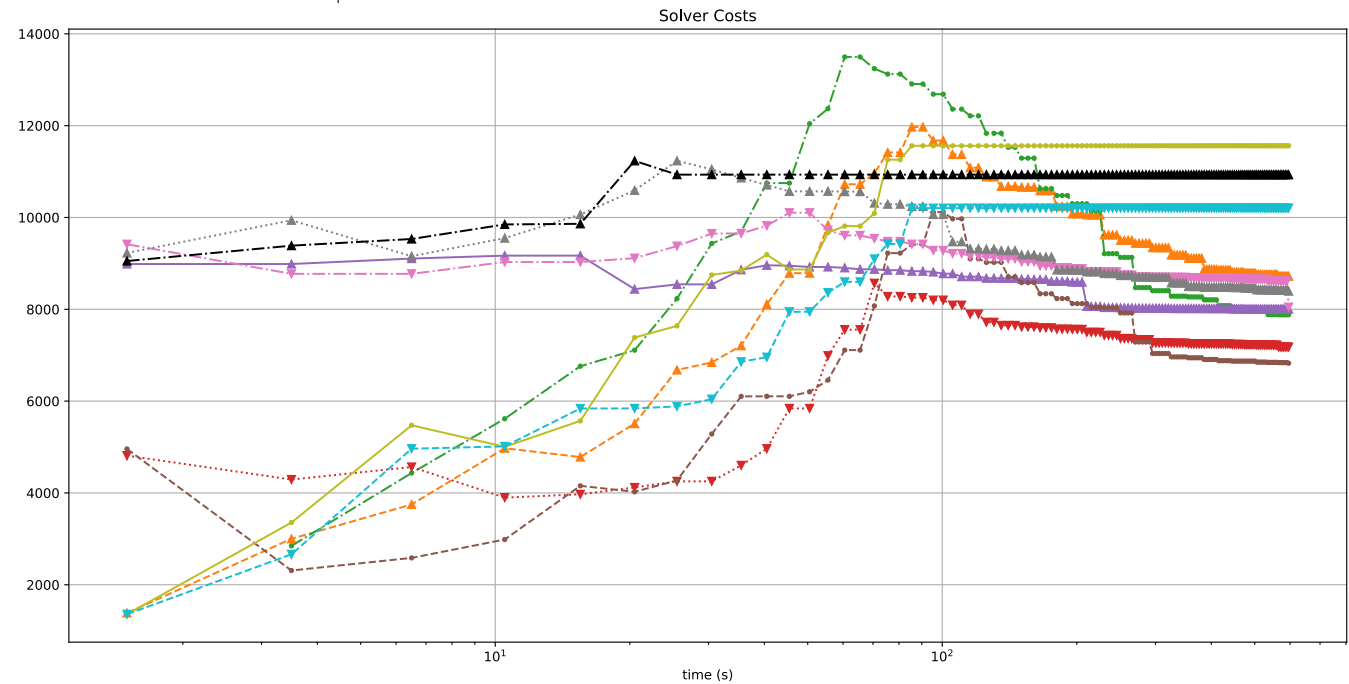
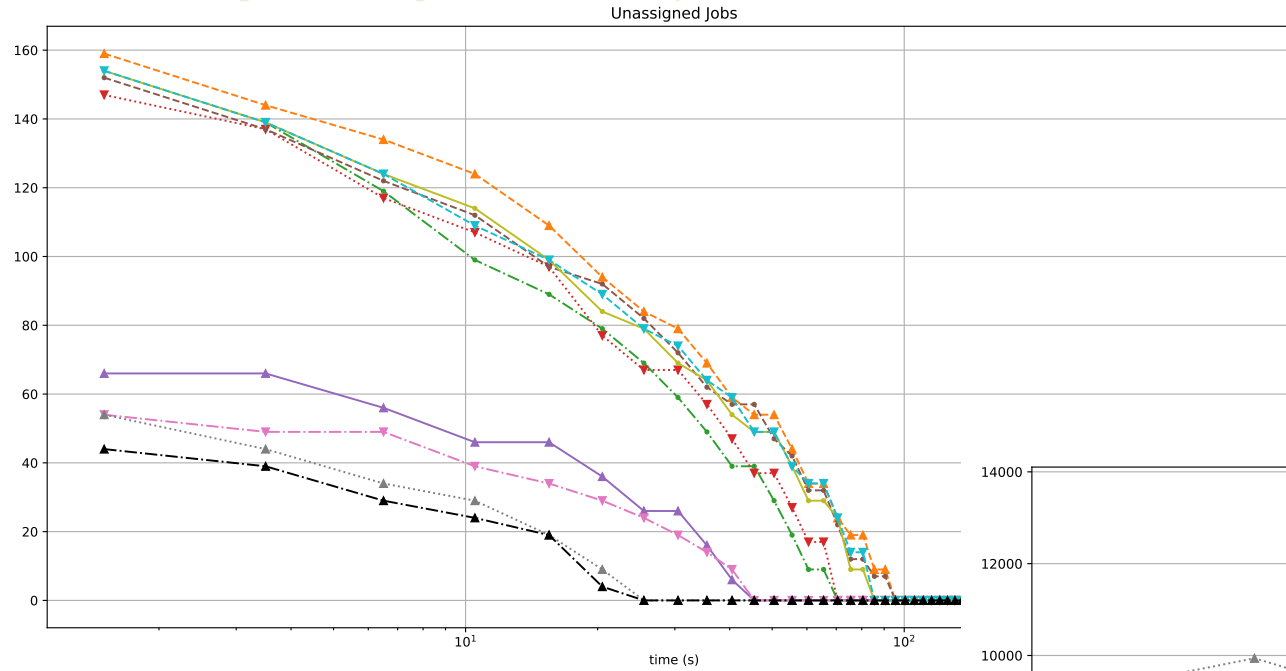
Chess AI, High Level Example

- Choose a move to follow based on a weighting (UCB calculated from W/L and simulations count)
 - Simulate a games to completion
 - Update weights from win loose result
 - Repeat until stopped
 - Choose the move with the most wins
- More iterations = greater difficulty
 - Focuses on likely wins while keeping exploration just in case

Monte Carlo Tree Search Solver Selection

- High Level Process (Each iteration of optimizer)
 - Calculate weightings (UCB calculated from win loss count & times sub solver has been ran)
 - Select highest weighted solver
 - Run solver n times (usually $n = 1$)
 - Improvement made = win, else = loss
 - If loss then discard result
 - Increment win loss count
- Why this method works
 - Focusses on sub solvers which have proved to make past improvements
 - Maintains exploration which in turn allows alternate solvers to “jump in”
 - When alternate solver makes improvements its weighting increases

Monte Carlo Tree Search Results



The Future



The Future of AI and Machine Learning for Optimisation

Machine Learning from Telematics

- Driver behaviours
- Traffic patterns
- More accurate Plans
- Improved ETA's

Responsive AI Planning

- Responding to real-time data feeds
- Changing plans on-the-fly

Quantum computing

- Not there yet, not enough q-bits for full optimisation
- Closer to truly optimal plan
- Quantum machine learning

