

## **Using Linear Programming for Route Planning and Job Scheduling**

5<sup>th</sup> September 2024 Brighton, UK



**Timothy Wong** CStat CEng MBCS Senior Data Scientist, Vodafone [timothywong731.github.io](https://timothywong731.github.io/)

# Who am L

Timothy (Tim) is a professional Data Scientist with over ten years experience in big data, machine learning and analytics applications. He previously led the Data Science function at a large energy company in the UK. His experience spans across multiple sectors including energy, telecommunications, defence and national security. Tim is professionally qualified as a Chartered Statistician (CStat) as well as a Chartered Engineer (CEng).

I hosted talks at:

- EARL 2016, 2017 and 2019 (London)
- USER 2017 (Brussels), 2018 (Brisbane)
- ERUM 2018 (Budapest)
- $\cdot$  ... and more  $\odot$





#### Resource Allocation

• There's a number of jobs requiring fulfilment

• There's a number of resources capable of fulfilling those jobs

• We need to allocate resources to jobs, efficiently!





## Optimisation Problem(s)

- We aim to fulfil as many jobs as possible
- Resources must start and end at the same location
- Certain jobs may have higher priority over the others
- Minimise travel distance, or time



• Maximise value in the knapsack



Let  $I =$  total number of items  $v_i =$  value of i<sup>th</sup>item  $w_i =$  weight of i<sup>th</sup> item  $x_i =$  allocation of i<sup>th</sup> item Max.  $z = \sum$  $i=1$ I  $v_i x_i$  $s.t.$ (1)  $x_i \in \{0,1\}$   $\forall i = 1,2,3,...,I$  $(2) \rightarrow$  $i=1$ I  $w_i x_i \le 10 \quad \forall i = 1, 2, 3, ..., I$ 



• Maximise value in the knapsack



Let  $I =$  total number of items

conference

```
\mathbf{1}library(ROI.plugin.glpk)
   library(ompr)
 2
    library(ompr.roi)
    library(dplyr)
 \overline{4}\overline{5}max_capacity < -106
    V \le C(5, 8, 10, 40, 60, 70)W \leftarrow c(2, 3, 4, 5, 6, 8)8
    N \leftarrow length(v)
 9
10result \leftarrow MIPModel() \left\vert \right\rangle11add_variable(x[i], i = 1:N, type = "binary") |>12<sup>2</sup>set_objective(sum_over(v[i] * x[i], i = 1:N), "max") |>
13add_constraint(sum_over(w[i] * x[i], i = 1:N) <= max_capacity) |>
1415
      solve_model(with_ROI(solver = "glpk"))16
    solution \le- result \ge1718
      get\_solution(x[i])19
20
    x \le- solution \gefilter(value > 0) |>21
22
      pu11(i)23
    paste0("Items selected: ", paste0(x, collapse = ", "))24
25
    paste0("Total value: f", sum(v[x]))
26
    paste0("Total weight: ", sum(w[x]), "kg")
27
```


• Adapt this into our business context…







- Can allocate jobs to resource
- Can maximise efficiency
- Works only if there's only one resource
- Doesn't figure out the order of the jobs
- Doesn't address the starting / finishing point



• Pack items into least number of bins





![](_page_8_Picture_4.jpeg)

• Pack items into least number of bins

![](_page_9_Figure_2.jpeg)

![](_page_9_Figure_3.jpeg)

![](_page_9_Picture_4.jpeg)

```
library(ROI.plugin.glpk)
 2 library (ompr)
   library(ompr.roi)
    library(dplyr)
 4
 5.
    max_capacity < -106
    W \leftarrow C(2, 3, 4, 5, 6, 8)N \leftarrow length(v)
 8
 9
10<sup>°</sup>max_{ } bins <-411result \leftarrow MIPModel()12<sup>7</sup>add\_variable(y[j], j = 1:max\_bins, type = "binary") >
13<sub>1</sub>add_variable(x[i, j], i = 1:N, j = 1:max_bins, type = "binary") |>
14set\_objective(sum\_over(y[j], j = 1:max\_bins), "min") >
1516add_constraint(sum_over(w[i] * x[i, j], i = 1:N) <= y[j] * max_capacity,17j = 1: max_bins) |>
      add_constraint(sum_over(x[i, j], j = 1:max_bins) == 1, i = 1:N) |>
18solve_model(with_RoI(solver = "qlpk", verbose = TRUE))1920
    solution \le- result \ge21get_solution(x[i, j])
22
23
24
    solution |>filter (value > 0)
25
26
```
![](_page_10_Picture_2.jpeg)

• Put this into context again...

![](_page_11_Figure_2.jpeg)

- Still doesn't figure out the order of the jobs
- Doesn't address the starting / finishing point
- Doesn't handle value of the jobs

![](_page_11_Picture_6.jpeg)

![](_page_11_Figure_7.jpeg)

# Travelling Salesman Problem (TSP)

• Find out the shortest path to visit each city exactly once and return to the original city

![](_page_12_Figure_2.jpeg)

![](_page_12_Picture_3.jpeg)

# Multiple TSP

![](_page_13_Figure_1.jpeg)

### Job Scheduling and Route Optimisation

![](_page_14_Figure_1.jpeg)

![](_page_14_Picture_2.jpeg)

#### Job Scheduling and Route Optimisation

![](_page_15_Picture_66.jpeg)

Let

### Job Scheduling and Route Optimisation

![](_page_16_Picture_77.jpeg)

![](_page_16_Picture_78.jpeg)

# Notebook Example

<https://timothywong731.github.io/scheduling/>

![](_page_17_Figure_3.jpeg)

![](_page_17_Figure_4.jpeg)

![](_page_17_Figure_5.jpeg)

![](_page_17_Picture_6.jpeg)

![](_page_18_Picture_0.jpeg)

#### **Using Linear Programming for Route Planning and Job Scheduling** <https://timothywong731.github.io/scheduling/>

![](_page_18_Picture_2.jpeg)

**Today's notebook example**

#### $T$ **imothy Wong** CStat CEng MBCS *Senior Data Scientist*

![](_page_18_Picture_5.jpeg)

timothy.wong@hotmail.co.uk

![](_page_18_Picture_7.jpeg)

linkedin.com/in/timothywong731

![](_page_18_Picture_9.jpeg)

![](_page_18_Picture_10.jpeg)

timothywong731.github.io