# Operation Research Problems Solving in Python

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# Chapter One Installation of Google OR Tools for Python

# **Install OR-Tools**

Though Google created OR-Tools in C++, you can use it with Python, Java, or C# (on the .NET platform).

### Install OR-Tools for Python

The fastest way to get OR-Tools is to install the Python binary version. If you already have Python 2.7 or 3.5+ (and the Python package manager PIP) installed, you can do so as follows:

python -m pip install --upgrade --user ortools

After the installation is complete, you are ready to get started with OR-Tools for Python.

# Chapter Two Finding Feasible Solution

Example: finding a feasible solution

### A simple example

Here's a simple example of a linear programming problem.

#### Maximize 3x + 4y subject to the following constraints:

```
x + 2y \leq 14
3x - y \geq 0
x - y \leq 2
```

```
\# Constraint 0: x + 2y \le 14.
    constraint0 = solver.Constraint(-solver.infinity(), 14)
    constraint0.SetCoefficient(x, 1)
    constraint0.SetCoefficient(y, 2)
    # Constraint 1: 3x - y \ge 0.
    constraint1 = solver.Constraint(0, solver.infinity())
    constraint1.SetCoefficient(x, 3)
    constraint1.SetCoefficient(y, -1)
    # Constraint 2: x - y \le 2.
    constraint2 = solver.Constraint(-solver.infinity(), 2)
    constraint2.SetCoefficient(x, 1)
    constraint2.SetCoefficient(y, -1)
    \# Objective function: 3x + 4y.
    objective = solver.Objective()
    objective.SetCoefficient(x, 3)
    objective.SetCoefficient(y, 4)
    objective.SetMaximization()
    # Solve the system.
    solver.Solve()
    opt solution = 3 * x.solution value() + 4 * y.solution value()
    print('Number of variables =', solver.NumVariables())
    print('Number of constraints =', solver.NumConstraints())
    # The value of each variable in the solution.
    print('Solution:')
    print('x = ', x.solution value())
    print('y = ', y.solution value())
    # The objective value of the solution.
    print('Optimal objective value =', opt solution)
LinearProgrammingExample()
```

# Chapter Three Mixed Integer Problem

## MIP Example (Mixed Integer Problem)

The following is a simple example of a mixed-integer programming problem:

#### Maximize x + 10y subject to the following constraints:

```
x + 7y \le 17.5
x \le 3.5
x \ge 0
y \ge 0
x, y \text{ integers}
```

```
\# x + 7 * y \le 17.5.
   solver.Add(x + 7 * y \le 17.5)
   \# x \le 3.5.
   solver.Add(x \le 3.5)
   print('Number of constraints =', solver.NumConstraints())
   # Maximize x + 10 * v.
   solver.Maximize(x + 10 * y)
   result status = solver.Solve()
   # The problem has an optimal solution.
   assert result status == pywraplp.Solver.OPTIMAL
   # The solution looks legit (when using solvers others than
   # GLOP LINEAR PROGRAMMING, verifying the solution is highly
recommended!).
   assert solver. VerifySolution (1e-7, True)
   print('Solution:')
   print('Objective value =', solver.Objective().Value())
   print('x =', x.solution value())
   print('y =', y.solution value())
   print('\nAdvanced usage:')
   print('Problem solved in %f milliseconds' % solver.wall time())
   print('Problem solved in %d iterations' % solver.iterations())
   print('Problem solved in %d branch-and-bound nodes' % solver.nodes())
if name == ' main ':
main()
```

## **Chapter Four**

# **Traveling Salesman Problem**

# Traveling Salesman Problem

The *distance matrix* is an array whose i, j entry is the distance from location i to location j in miles, where the locations are given in the order below:

New York 1. Los Angeles 2. Chicago 3. Minneapolis 4. Denver 5. Dallas 6. Seattle 7.
 Boston 8. San Francisco 9. St. Louis 10. Houston 11. Phoenix 12. Salt Lake City

As an alternative to a distance matrix, you could provide a *time matrix* which contains the travel times between locations.

The data also includes:

- The number of vehicles in the problem, which is 1 because this is a TSP. For general routing problems, the number of vehicles can be greater than 1.
- The *depot*: the starting location for the route. In this case, the depot is 0, which corresponds to New York City.

```
"""Simple travelling salesman problem between cities."""

from __future__ import print_function
from ortools.constraint_solver import routing_enums_pb2
from ortools.constraint_solver import pywrapcp

def create_data_model():
    """Stores the data for the problem."""
    data = {}
    data['distance_matrix'] = [
        [0, 2451, 713, 1018, 1631, 1374, 2408, 213, 2571, 875, 1420, 2145, 1972],
        [2451, 0, 1745, 1524, 831, 1240, 959, 2596, 403, 1589, 1374, 357,
```

```
5791,
        [713, 1745, 0, 355, 920, 803, 1737, 851, 1858, 262, 940, 1453,
1260],
        [1018, 1524, 355, 0, 700, 862, 1395, 1123, 1584, 466, 1056, 1280,
987],
        [1631, 831, 920, 700, 0, 663, 1021, 1769, 949, 796, 879, 586, 371],
        [1374, 1240, 803, 862, 663, 0, 1681, 1551, 1765, 547, 225, 887,
999],
        [2408, 959, 1737, 1395, 1021, 1681, 0, 2493, 678, 1724, 1891, 1114,
7011,
        [213, 2596, 851, 1123, 1769, 1551, 2493, 0, 2699, 1038, 1605, 2300,
2099],
        [2571, 403, 1858, 1584, 949, 1765, 678, 2699, 0, 1744, 1645, 653,
600],
        [875, 1589, 262, 466, 796, 547, 1724, 1038, 1744, 0, 679, 1272,
1162],
        [1420, 1374, 940, 1056, 879, 225, 1891, 1605, 1645, 679, 0, 1017,
1200],
        [2145, 357, 1453, 1280, 586, 887, 1114, 2300, 653, 1272, 1017, 0,
504],
        [1972, 579, 1260, 987, 371, 999, 701, 2099, 600, 1162, 1200, 504,
0],
    ] # yapf: disable
    data['num vehicles'] = 1
    data['depot'] = 0
    return data
def print solution (manager, routing, assignment):
    """Prints assignment on console."""
    print('Objective: {} miles'.format(assignment.ObjectiveValue()))
    index = routing.Start(0)
    plan output = 'Route for vehicle 0:\n'
   route distance = 0
   while not routing.IsEnd(index):
        plan output += ' {} ->'.format(manager.IndexToNode(index))
        previous index = index
        index = assignment.Value(routing.NextVar(index))
        route distance += routing.GetArcCostForVehicle(previous index,
index, 0)
    plan output += ' {}\n'.format(manager.IndexToNode(index))
    print(plan output)
   plan output += 'Route distance: {}miles\n'.format(route distance)
```

```
def main():
   """Entry point of the program."""
    # Instantiate the data problem.
   data = create data model()
   # Create the routing index manager.
   manager = pywrapcp.RoutingIndexManager(
        len(data['distance_matrix']), data['num vehicles'], data['depot'])
   # Create Routing Model.
   routing = pywrapcp.RoutingModel(manager)
   def distance callback(from index, to index):
        """Returns the distance between the two nodes."""
        # Convert from routing variable Index to distance matrix NodeIndex.
        from node = manager.IndexToNode(from index)
        to node = manager.IndexToNode(to index)
        return data['distance matrix'][from node][to node]
    transit callback index =
routing.RegisterTransitCallback(distance callback)
    # Define cost of each arc.
   routing.SetArcCostEvaluatorOfAllVehicles(transit callback index)
    # Setting first solution heuristic.
    search parameters = pywrapcp.DefaultRoutingSearchParameters()
    search parameters.first solution strategy = (
        routing enums pb2.FirstSolutionStrategy.PATH CHEAPEST ARC)
   # Solve the problem.
   assignment = routing.SolveWithParameters(search parameters)
   # Print solution on console.
   if assignment:
        print solution(manager, routing, assignment)
if __name__ == '__main__':
main()
```

## **Chapter Five**

## Vehicle Routing Problem with Capacity Constraint

# Capacity Constraints

#### Overview

The *capacitated vehicle routing problem* (CVRP) is a VRP in which vehicles with limited carrying capacity need to pick up or deliver items at various locations.

The new items in the data are:

- Demands: Each location has a demand corresponding to the quantity—for example, weight or volume—of the item to be picked up.
- Capacities: Each vehicle has a capacity: the maximum quantity that the vehicle can hold. As a vehicle travels along its route, the total quantity of the items it is carrying can never exceed its capacity.

#### Problems with multiple cargo types and capacities

In more complex CVRPs, each vehicle might carry several different types of cargo, with a maximum capacity for each type. For example, a fuel delivery truck might carry several types of fuel, using multiple tanks with differing capacities. To handle problems like these, just create a different capacity callback and dimension for each cargo type (making sure to assign them unique names).

```
"""Capacited Vehicles Routing Problem (CVRP)."""

from __future__ import print_function
from ortools.constraint_solver import routing_enums_pb2
```

```
from ortools.constraint solver import pywrapcp
def create data model():
    """Stores the data for the problem."""
   data = \{\}
   data['distance matrix'] = [
            0, 548, 776, 696, 582, 274, 502, 194, 308, 194, 536, 502, 388,
354,
            468, 776, 662
        ],
        Γ
            548, 0, 684, 308, 194, 502, 730, 354, 696, 742, 1084, 594, 480,
674,
            1016, 868, 1210
        ],
        [
            776, 684, 0, 992, 878, 502, 274, 810, 468, 742, 400, 1278,
1164,
            1130, 788, 1552, 754
        ],
            696, 308, 992, 0, 114, 650, 878, 502, 844, 890, 1232, 514, 628,
822,
            1164, 560, 1358
        ],
        [
            582, 194, 878, 114, 0, 536, 764, 388, 730, 776, 1118, 400, 514,
708,
            1050, 674, 1244
        ],
        [
            274, 502, 502, 650, 536, 0, 228, 308, 194, 240, 582, 776, 662,
628,
            514, 1050, 708
        ],
        Γ
            502, 730, 274, 878, 764, 228, 0, 536, 194, 468, 354, 1004, 890,
856,
            514, 1278, 480
        ],
```

```
194, 354, 810, 502, 388, 308, 536, 0, 342, 388, 730, 468, 354,
320,
            662, 742, 856
        ],
            308, 696, 468, 844, 730, 194, 194, 342, 0, 274, 388, 810, 696,
662,
            320, 1084, 514
        ],
            194, 742, 742, 890, 776, 240, 468, 388, 274, 0, 342, 536, 422,
388,
            274, 810, 468
        ],
            536, 1084, 400, 1232, 1118, 582, 354, 730, 388, 342, 0, 878,
764,
            730, 388, 1152, 354
        ],
        [
            502, 594, 1278, 514, 400, 776, 1004, 468, 810, 536, 878, 0,
114,
            308, 650, 274, 844
        ],
        [
            388, 480, 1164, 628, 514, 662, 890, 354, 696, 422, 764, 114, 0,
194,
            536, 388, 730
        ],
        [
            354, 674, 1130, 822, 708, 628, 856, 320, 662, 388, 730, 308,
194, 0,
            342, 422, 536
        ],
        [
            468, 1016, 788, 1164, 1050, 514, 514, 662, 320, 274, 388, 650,
536,
            342, 0, 764, 194
        ],
        776, 868, 1552, 560, 674, 1050, 1278, 742, 1084, 810, 1152,
274,
            388, 422, 764, 0, 798
```

```
],
        [
            662, 1210, 754, 1358, 1244, 708, 480, 856, 514, 468, 354, 844,
730,
            536, 194, 798, 0
        ],
    1
    data['demands'] = [0, 1, 1, 2, 4, 2, 4, 8, 8, 1, 2, 1, 2, 4, 4, 8, 8]
    data['vehicle capacities'] = [15, 15, 15, 15]
   data['num vehicles'] = 4
   data['depot'] = 0
   return data
def print solution(data, manager, routing, assignment):
    """Prints assignment on console."""
    total distance = 0
    total load = 0
    for vehicle id in range(data['num vehicles']):
        index = routing.Start(vehicle id)
        plan output = 'Route for vehicle {}:\n'.format(vehicle id)
        route distance = 0
        route load = 0
        while not routing.IsEnd(index):
            node index = manager.IndexToNode(index)
            route load += data['demands'][node index]
            plan output += ' {0} Load({1}) -> '.format(node index,
route load)
            previous index = index
            index = assignment.Value(routing.NextVar(index))
            route distance += routing.GetArcCostForVehicle(
                previous index, index, vehicle id)
        plan output += ' {0}
Load({1})\n'.format(manager.IndexToNode(index),
                                                 route load)
        plan output += 'Distance of the route:
{}m\n'.format(route distance)
        plan output += 'Load of the route: {}\n'.format(route load)
        print(plan output)
        total distance += route distance
        total load += route load
   print('Total distance of all routes: {}m'.format(total distance))
   print('Total load of all routes: {}'.format(total load))
```

```
def main():
   """Solve the CVRP problem."""
    # Instantiate the data problem.
    data = create data model()
    # Create the routing index manager.
   manager = pywrapcp.RoutingIndexManager(len(data['distance matrix']),
                                           data['num vehicles'],
data['depot'])
    # Create Routing Model.
   routing = pywrapcp.RoutingModel (manager)
    # Create and register a transit callback.
    def distance callback(from index, to index):
        """Returns the distance between the two nodes."""
        # Convert from routing variable Index to distance matrix NodeIndex.
        from node = manager.IndexToNode(from index)
        to node = manager.IndexToNode(to index)
        return data['distance_matrix'][from_node][to_node]
    transit callback index =
routing.RegisterTransitCallback(distance callback)
    # Define cost of each arc.
    routing.SetArcCostEvaluatorOfAllVehicles(transit callback index)
    # Add Capacity constraint.
    def demand callback(from index):
        """Returns the demand of the node."""
        # Convert from routing variable Index to demands NodeIndex.
        from node = manager.IndexToNode(from index)
        return data['demands'][from node]
    demand callback index = routing.RegisterUnaryTransitCallback(
        demand callback)
    routing.AddDimensionWithVehicleCapacity(
        demand callback index,
        0, # null capacity slack
```

```
data['vehicle_capacities'], # vehicle maximum capacities
    True, # start cumul to zero
    'Capacity')

# Setting first solution heuristic.
search_parameters = pywrapcp.DefaultRoutingSearchParameters()
search_parameters.first_solution_strategy = (
    routing_enums_pb2.FirstSolutionStrategy.PATH_CHEAPEST_ARC)

# Solve the problem.
assignment = routing.SolveWithParameters(search_parameters)

# Print solution on console.
if assignment:
    print_solution(data, manager, routing, assignment)

if __name__ == '__main__':
    main()
```