Solving the Consolidation Problem

Now we mean business

Now that we know better, let's look again at the consolidation problem:

arg min
$$||z||_1$$

subject to: $Vx = v^*, Ex = e^*$
 $x \le Mz$
 $x \ge 0, z \in \{0, 1\}^n$

Where do you think most of the complexity will stem from?

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Where do you think most of the complexity will stem from?

- We need to branch only on the integer variables
- So, their number will likely have an impact on complexity

Since we focus on the current Path formulation solution, they won't be many

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What could you say of the impact of using big-Ms?

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What could you say of the impact of using big-Ms?

- Logically, they work just fine
- In practice, they can lead to poor bounds in the LP relaxation

Ideally, they should be avoided. Failing that, use an $m{M}$ as small as possible

The code for solving the problem is in the consolidate_paths function

The function parameter look similar to those of solve_path_selection_full

```
def consolidate_paths(
    tug : ig.Graph,
    paths : list,
    node_counts : dict,
    arc_counts : dict,
    tlim : int = None):
```

However, they are meant to be used differently:

- paths should contain those selected by the Path formulation
- node_counts should contain the counts from the Path formulation solution
- ...And the same goes for arc_counts

tlim is a time limit: always use one when dealing with NP-hard problems

Let's see some relevant code snippets:

We use the <u>CBC solver</u>, via the <u>Google Or-Tools</u> Wrapper

```
slv = pywraplp.Solver.CreateSolver('CBC')
```

Variables are built using the solver object and stored in lists:

```
x = [slv.NumVar(0, inf, f'x_{j}') for j in range(npaths)]

z = [slv.IntVar(0, 1, f'z_{j}') for j in range(npaths)]
```

For the big-M constraints ($x \leq Mz$) we use the largest node count

```
M = max(v for v in node_counts.values())
for j in range(npaths):
    slv.Add(x[j] <= M * z[j])</pre>
```

There no need for a path to use a flow larger than that

Let's see some relevant code snippets:

Here's the code for the "count matching" constraints, i.e. $Vx=v^*$ and $Ex=e^*$:

```
for n, p in paths_by_node.items():
    slv.Add(sum(x[j] for j in p) == node_counts[n])
for a, p in paths_by_arc.items():
    slv.Add(sum(x[j] for j in p) == arc_counts[a])
```

We rely on a previous step where we grouped path by used node/arc
 Here's how we define the objective and optimization direction:

```
slv.Minimize(sum(z[j] for j in range(npaths)))
```

...And here how to set a time limit:

```
if tlim is not None: slv.SetTimeLimit(tlim)
```

Let's see some relevant code snippets:

We trigger the solution process with the solve method:

```
status = slv.Solve()
```

The method returns an integer status code, that should always be checked:

```
if status in (slv.OPTIMAL, slv.FEASIBLE):
    # Extract the paths in the solution
    ...
    # Return the solution
    if status == slv.OPTIMAL: return sol_flows, sol_paths, True
    else: return sol_flows, sol_paths, False
else:
    return None, None, False
```

- If we find a solution we return it
- If we prove optimality within the time limit, we tell it with a flag

Solving the Problem

We can finally solve the consolidation problem for real:

```
In [3]: node_counts_r, arc_counts_r = util.get_counts(tug, rflows, rpaths)
    cflows, cpaths, cflag = util.consolidate_paths(tug, rpaths, node_counts_r, arc_counts_r)
    print('FLOW: PATH')
    util.print_solution(tug, cflows, cpaths, sort='descending')
    print(f'Optimal: {cflag}')

FLOW: PATH
    8.17: 2,3 > 3,3
    5.47: 0,2 > 1,2 > 2,2 > 3,2
    4.89: 0,1 > 1,1 > 2,0 > 3,0
    3.74: 3,3
    3.32: 1,0 > 2,0 > 3,2
    Optimal: True
```

In our case, the consolidated paths match the ground truth perfectly!

1 20 0 1 \ 1 1 \ 2 0 \ 3 0

```
In [4]: print('FLOW: PATH')
    util.print_ground_truth(flows, paths, sort='descending')

FLOW: PATH
    8.17: 2,3 > 3,3
    5.47: 0,2 > 1,2 > 2,2 > 3,2
```