

model_building

December 10, 2018

```
In [114]: import pandas as pd
import pulp as lp
import numpy as np
from operator import iadd
from functools import reduce
from itertools import chain
from typing import Sequence, Any, DefaultDict, List
```

The below functions assist in creating an lp model problem using pulp package

```
In [115]: def create_prob(prob_name: str, sense: int) -> lp.LpProblem:
    return lp.LpProblem(prob_name, sense)

def add_obj_fn(lp_prob: lp.LpProblem, dvar: lp.LpAffineExpression) -> lp.LpProblem:
    return iadd(lp_prob, dvar)

def add_constraint(lp_prob: lp.LpProblem, constrs: Sequence[lp.LpConstraint]) -> lp.LpProblem:
    return reduce(iadd, constrs, lp_prob)

def head(x: Sequence) -> Any:
    return x[0]

def to_str(indnum, activity) -> str:
    return f'{indnum} - {activity}'
```

1 Loading The Prepared Data

We load our dataset into activity_df. activity_df contains all the observations related to individuals, their activities, consumption per unit and the relevant sources activities use.

```
In [116]: activity_df = pd.read_csv('data/dataset.csv')

activity_df.drop(['Unnamed: 0', 'Group'], axis=1, inplace=True)
```

We load the sources carbon footprint dataset into `source_df`. `source_df` contains all the observations related to an carbon footprint an activity contains given that they are used with a specific source.

```
In [117]: source_df = pd.read_csv('data/dataset_source_cf.csv')
```

```
source_df.drop(['Unnamed: 0', 'X_1'], axis=1, inplace=True)
```

```
In [118]: source_df.head()
```

```
Out[118]:
```

	Activity	Per	solar_powered_water_heater	\
0	Household heating => 70F	hour	0.000000	
1	Household heating < 70F	hour	0.000000	
2	Use of heat pump	hour	0.000000	
3	Use of air conditioner	hour	0.000000	
4	shower - short activity		0.000012	

	gas_water_heater	electric_water_heater_peak_hour	\
0	0.000000	0.000000	
1	0.000000	0.000000	
2	0.000000	0.000000	
3	0.000000	0.000000	
4	0.000102	0.000232	

	electric_water_heater_off_peak	gas	natural_gas	jetfuel	\
0	0.000000	0.0	0.000436	0.0	
1	0.000000	0.0	0.000872	0.0	
2	0.000000	0.0	0.001074	0.0	
3	0.000000	0.0	0.000598	0.0	
4	0.000199	0.0	0.000000	0.0	

	waste management	hybrid	electric_peak_hours	electric_off_peak_hours
0	0.0	0.0	0.000650	0.000542
1	0.0	0.0	0.000923	0.000901
2	0.0	0.0	0.001229	0.001188
3	0.0	0.0	0.007980	0.000721
4	0.0	0.0	0.000000	0.000000

2 Building The Model

`lp_prob` is the linear programming problem we have to formulize. The problem is one in which we have to minimize the objective function.

```
In [119]: lp_prob = create_prob('Wells Fargo Challenge', lp.LpMinimize)
```

2.1 Defining The Decision Variables

The `individuals` variable represents all the individuals that we want to focus on, here we are looking at the first **10** individuals. `MAX_NUM_INDV` is used to change the maximum number of individuals the linear programming model should solve for.

The activities variable is the list of activities that all individuals do.
The sources variable is the list of all sources that an activity can use.

```
In [120]: MAX_NUM_INDV = 2

In [121]: individuals = activity_df.Indnum.unique()[0:MAX_NUM_INDV]
          activities = activity_df.Activity.unique()

M = 100 # The big number M

sources = [
    "solar_powered_water_heater",
    "gas_water_heater",
    "electric_water_heater_peak_hour",
    "electric_water_heater_off_peak",
    "gas",
    "natural_gas",
    "hybrid",
    "electric_peak_hours",
    "electric_off_peak_hours",
    "jetfuel",
    "waste management"
]
```

Our decision variable is called

2.1.1 S_{ijk}

where * i is individual * j is activity * k is source

It is a binary variable that can have either 0 or 1. Meaning that S_{ijk} represents whether an individual uses a particular source or not.

```
In [122]: source_indexes = [
          (f'{indv} - {activity}', source)
          for indv in individuals
          for activity in activities
          for source in sources
        ]

S_ijk = lp.LpVariable.dicts('S_ijk', source_indexes, lowBound=0, cat='Binary')
```

2.2 The Objective Function

2.2.1 $Z_{min} = \sum S_{ijk} * SCF_{ijk} * C_{ij}$

where * i = 1...n * j = 1...27 * k = 1...10

C is the consumption per unit of an activity while SCF is the carbon footprint per source.

C_{ij} is assumed to be **constant** for each individual and activity and SCF_{ijk} is assumed to be **constant** for per source.

In the case that C_{ij} is 0 we will use the big M method to enforce a big penalty, this will ensure that the linear programming model won't choose that particular source

```
In [123]: d_vars = []

for indiv in individuals:
    for activity in activities:
        consumption: np.ndarray = activity_df.loc[activity_df['Activity'] == activity,
                                                    'Consumption']

        for source in sources:
            source_cf: np.ndarray = source_df.loc[source_df['Activity'] == activity,
                                                    'CarbonFootprint']
            source_cf = M if head(source_cf) == 0.0 else head(source_cf)
            d_vars.append(S_ijk[(to_str(indv, activity), source)] * float(source_cf))

obj_fn = lp.lpSum(sum(d_vars))
```

2.3 The Constraints

The constraint below ensures that an activity can only use one source. The goal of this constraint is to use a single source that has the lowest carbon footprint.

2.3.1 $\sum_k^m S_{ijk} == 1$

where $n = m = 10$ * $i = 1 \dots n$ * $j = 1 \dots 27$

```
In [124]: sum_source_conds = []

for indiv in individuals:
    for activity in activities:
        sum_source = []
        for source in sources:
            sum_source.append(S_ijk[(to_str(indv, activity), source)])

        sum_source_conds.append(lp.lpSum(sum_source) == 1)
```

We add the objective function and the constraints to the model

```
In [125]: lp_prob = add_obj_fn(lp_prob, obj_fn)
          lp_prob = add_constraint(lp_prob, sum_source_conds)
```

```
In [126]: lp_prob.solve()
          lp.LpStatus[lp_prob.status]
```

```
Out[126]: 'Optimal'
```

The minimum carbon footprint that can be obtained for 10 individuals is

```
In [127]: lp.value(lp_prob.objective)
```

```
Out[127]: 3.2388659999999999
```

```
In [128]: columns = list(chain(['Individual'], ["Activity"], sources))
```

```
In [134]: rows = []
```

```
    for indv in individuals:
        for activity in activities:
            row = [indv, activity]

            row.extend([lp.value(S_ijk[(to_str(indv, activity), source)]) for source in sources])

            rows.append(row)

solution = pd.DataFrame(rows, columns=columns)
```

```
In [135]: solution
```

```
Out[135]:
```

	Individual	Activity \
0	1	Household heating => 70F
1	1	Household heating < 70F
2	1	Use of heat pump
3	1	Use of air conditioner
4	1	shower - short
5	1	shower - long (> 3 min)
6	1	bath
7	1	wash-up
8	1	use of dishwasher
9	1	use of clothes washer
10	1	use of clothes dryer
11	1	use of cooking range
12	1	use of oven
13	1	use of self-clean feature of electric oven
14	1	Small kitchen appliance in the home
15	1	TV/computer use
16	1	air travel - large plane
17	1	air travel - small plane (<50 seats)
18	1	car trips- self only
19	1	car trips - driver and self
20	1	car trips - 2+ people with multiple end points
21	1	trips using public ground transportation
22	1	bags of garbage disposed
23	1	bags of recycling deposited (negative CF)
24	1	bags of compost deposited (negative CF)
25	1	hazardous or electric items disposed
26	1	large items disposed
27	2	Household heating => 70F
28	2	Household heating < 70F
29	2	Use of heat pump

30	2	Use of air conditioner
31	2	shower - short
32	2	shower - long (> 3 min)
33	2	bath
34	2	wash-up
35	2	use of dishwasher
36	2	use of clothes washer
37	2	use of clothes dryer
38	2	use of cooking range
39	2	use of oven
40	2	use of self-clean feature of electric oven
41	2	Small kitchen appliance in the home
42	2	TV/computer use
43	2	air travel - large plane
44	2	air travel - small plane (<50 seats)
45	2	car trips- self only
46	2	car trips - driver and self
47	2	car trips - 2+ people with multiple end points
48	2	trips using public ground transportation
49	2	bags of garbage disposed
50	2	bags of recycling deposited (negative CF)
51	2	bags of compost deposited (negative CF)
52	2	hazardous or electric items disposed
53	2	large items disposed

	solar_powered_water_heater	gas_water_heater \
0	0.0	0.0
1	0.0	0.0
2	0.0	0.0
3	0.0	0.0
4	1.0	0.0
5	1.0	0.0
6	0.0	0.0
7	1.0	0.0
8	0.0	0.0
9	1.0	0.0
10	0.0	0.0
11	0.0	0.0
12	0.0	0.0
13	0.0	0.0
14	0.0	0.0
15	0.0	0.0
16	0.0	0.0
17	0.0	0.0
18	0.0	0.0
19	0.0	0.0
20	0.0	0.0
21	0.0	0.0

22	0.0	0.0
23	0.0	0.0
24	0.0	0.0
25	0.0	0.0
26	0.0	0.0
27	0.0	0.0
28	0.0	0.0
29	0.0	0.0
30	0.0	0.0
31	1.0	0.0
32	1.0	0.0
33	0.0	0.0
34	1.0	0.0
35	0.0	0.0
36	1.0	0.0
37	0.0	0.0
38	0.0	0.0
39	0.0	0.0
40	0.0	0.0
41	0.0	0.0
42	0.0	0.0
43	0.0	0.0
44	0.0	0.0
45	0.0	0.0
46	0.0	0.0
47	0.0	0.0
48	0.0	0.0
49	0.0	0.0
50	0.0	0.0
51	0.0	0.0
52	0.0	0.0
53	0.0	0.0

	electric_water_heater_peak_hour	electric_water_heater_off_peak	gas	\
0	0.0	0.0	0.0	
1	0.0	0.0	0.0	
2	0.0	0.0	0.0	
3	0.0	0.0	0.0	
4	0.0	0.0	0.0	
5	0.0	0.0	0.0	
6	0.0	0.0	0.0	
7	0.0	0.0	0.0	
8	0.0	0.0	0.0	
9	0.0	0.0	0.0	
10	0.0	0.0	0.0	
11	0.0	0.0	0.0	
12	0.0	0.0	0.0	
13	0.0	0.0	0.0	

14	0.0	0.0	0.0
15	0.0	0.0	0.0
16	0.0	0.0	0.0
17	0.0	0.0	0.0
18	0.0	0.0	0.0
19	0.0	0.0	0.0
20	0.0	0.0	0.0
21	0.0	0.0	1.0
22	0.0	0.0	0.0
23	0.0	0.0	0.0
24	0.0	0.0	0.0
25	0.0	0.0	0.0
26	0.0	0.0	0.0
27	0.0	0.0	0.0
28	0.0	0.0	0.0
29	0.0	0.0	0.0
30	0.0	0.0	0.0
31	0.0	0.0	0.0
32	0.0	0.0	0.0
33	0.0	0.0	0.0
34	0.0	0.0	0.0
35	0.0	0.0	0.0
36	0.0	0.0	0.0
37	0.0	0.0	0.0
38	0.0	0.0	0.0
39	0.0	0.0	0.0
40	0.0	0.0	0.0
41	0.0	0.0	0.0
42	0.0	0.0	0.0
43	0.0	0.0	0.0
44	0.0	0.0	0.0
45	0.0	0.0	0.0
46	0.0	0.0	0.0
47	0.0	0.0	0.0
48	0.0	0.0	1.0
49	0.0	0.0	0.0
50	0.0	0.0	0.0
51	0.0	0.0	0.0
52	0.0	0.0	0.0
53	0.0	0.0	0.0

	natural_gas	hybrid	electric_peak_hours	electric_off_peak_hours	\
0	1.0	0.0	0.0	0.0	
1	1.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	1.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	
5	0.0	0.0	0.0	0.0	

6	0.0	0.0	0.0	1.0
7	0.0	0.0	0.0	0.0
8	0.0	0.0	0.0	1.0
9	0.0	0.0	0.0	0.0
10	0.0	0.0	0.0	1.0
11	0.0	0.0	0.0	0.0
12	0.0	0.0	0.0	1.0
13	0.0	0.0	0.0	1.0
14	0.0	0.0	0.0	0.0
15	0.0	0.0	0.0	1.0
16	0.0	0.0	0.0	0.0
17	0.0	0.0	0.0	0.0
18	0.0	0.0	1.0	0.0
19	0.0	0.0	1.0	0.0
20	0.0	0.0	1.0	0.0
21	0.0	0.0	0.0	0.0
22	0.0	0.0	0.0	0.0
23	0.0	0.0	0.0	0.0
24	0.0	0.0	0.0	0.0
25	0.0	0.0	0.0	0.0
26	0.0	0.0	0.0	0.0
27	1.0	0.0	0.0	0.0
28	1.0	0.0	0.0	0.0
29	0.0	0.0	0.0	1.0
30	1.0	0.0	0.0	0.0
31	0.0	0.0	0.0	0.0
32	0.0	0.0	0.0	0.0
33	0.0	0.0	0.0	0.0
34	0.0	0.0	0.0	0.0
35	0.0	0.0	0.0	1.0
36	0.0	0.0	0.0	0.0
37	0.0	0.0	0.0	1.0
38	0.0	0.0	0.0	1.0
39	0.0	0.0	0.0	1.0
40	0.0	0.0	0.0	1.0
41	0.0	0.0	0.0	0.0
42	0.0	0.0	0.0	1.0
43	0.0	0.0	0.0	0.0
44	0.0	0.0	0.0	1.0
45	0.0	0.0	1.0	0.0
46	0.0	0.0	1.0	0.0
47	0.0	0.0	1.0	0.0
48	0.0	0.0	0.0	0.0
49	0.0	0.0	0.0	0.0
50	0.0	0.0	0.0	0.0
51	0.0	0.0	0.0	0.0
52	0.0	0.0	0.0	0.0
53	0.0	0.0	0.0	0.0

	jetfuel	waste management
0	0.0	0.0
1	0.0	0.0
2	0.0	1.0
3	0.0	0.0
4	0.0	0.0
5	0.0	0.0
6	0.0	0.0
7	0.0	0.0
8	0.0	0.0
9	0.0	0.0
10	0.0	0.0
11	0.0	1.0
12	0.0	0.0
13	0.0	0.0
14	0.0	1.0
15	0.0	0.0
16	1.0	0.0
17	0.0	1.0
18	0.0	0.0
19	0.0	0.0
20	0.0	0.0
21	0.0	0.0
22	0.0	1.0
23	0.0	1.0
24	0.0	1.0
25	0.0	1.0
26	0.0	1.0
27	0.0	0.0
28	0.0	0.0
29	0.0	0.0
30	0.0	0.0
31	0.0	0.0
32	0.0	0.0
33	0.0	1.0
34	0.0	0.0
35	0.0	0.0
36	0.0	0.0
37	0.0	0.0
38	0.0	0.0
39	0.0	0.0
40	0.0	0.0
41	0.0	1.0
42	0.0	0.0
43	1.0	0.0
44	0.0	0.0
45	0.0	0.0

```

46      0.0      0.0
47      0.0      0.0
48      0.0      0.0
49      0.0      1.0
50      0.0      1.0
51      0.0      1.0
52      0.0      1.0
53      0.0      1.0

```

```
In [138]: original = pd.read_csv('data/dataset_without_source_cf.csv')
```

```
original.drop(['Unnamed: 0'], axis=1, inplace=True)
```

```
original
```

```

Out[138]:
   Indnum Activity \
0        1 Household heating => 70F
1        1 Household heating < 70F
2        1 Use of heat pump
3        1 Use of air conditioner
4        1 shower - short
5        1 shower - long (> 3 min)
6        1 bath
7        1 wash-up
8        1 use of dishwasher
9        1 use of clothes washer
10       1 use of clothes dryer
11       1 use of cooking range
12       1 use of oven
13       1 use of self-clean feature of electric oven
14       1 Small kitchen appliance in the home
15       1 TV/computer use
16       1 air travel - large plane
17       1 air travel - small plane (<50 seats)
18       1 car trips- self only
19       1 car trips - driver and self
20       1 car trips - 2+ people with multiple end points
21       1 trips using public ground transportation
22       1 bags of garbage disposed
23       1 bags of recycling deposited (negative CF)
24       1 bags of compost deposited (negative CF)
25       1 hazardous or electric items disposed
26       1 large items disposed
27       2 Household heating => 70F
28       2 Household heating < 70F
29       2 Use of heat pump
...     ...
27024   1001 bags of compost deposited (negative CF)

```

27025	1001	hazardous or electric items disposed
27026	1001	large items disposed
27027	1002	Household heating => 70F
27028	1002	Household heating < 70F
27029	1002	Use of heat pump
27030	1002	Use of air conditioner
27031	1002	shower - short
27032	1002	shower - long (> 3 min)
27033	1002	bath
27034	1002	wash-up
27035	1002	use of dishwasher
27036	1002	use of clothes washer
27037	1002	use of clothes dryer
27038	1002	use of cooking range
27039	1002	use of oven
27040	1002	use of self-clean feature of electric oven
27041	1002	Small kitchen appliance in the home
27042	1002	TV/computer use
27043	1002	air travel - large plane
27044	1002	air travel - small plane (<50 seats)
27045	1002	car trips- self only
27046	1002	car trips - driver and self
27047	1002	car trips - 2+ people with multiple end points
27048	1002	trips using public ground transportation
27049	1002	bags of garbage disposed
27050	1002	bags of recycling deposited (negative CF)
27051	1002	bags of compost deposited (negative CF)
27052	1002	hazardous or electric items disposed
27053	1002	large items disposed

	solar_powered_water_heater	gas_water_heater \
0	0	0
1	1	1
2	0	0
3	0	0
4	0	0
5	0	0
6	0	0
7	0	1
8	0	0
9	0	0
10	0	0
11	0	0
12	0	0
13	0	0
14	0	0
15	0	0
16	0	0

17	0	0
18	0	0
19	0	0
20	0	0
21	0	0
22	0	0
23	0	0
24	0	0
25	0	0
26	0	0
27	0	0
28	0	0
29	0	0
...
27024	0	0
27025	0	0
27026	0	0
27027	0	0
27028	0	0
27029	0	0
27030	0	0
27031	1	0
27032	0	0
27033	0	0
27034	0	0
27035	0	0
27036	0	0
27037	1	0
27038	0	0
27039	0	0
27040	0	0
27041	0	0
27042	0	0
27043	0	0
27044	0	0
27045	0	0
27046	0	0
27047	0	0
27048	0	0
27049	0	0
27050	0	0
27051	0	0
27052	0	0
27053	0	0

	electric_water_heater_peak_hour	electric_water_heater_off_peak	gas	\
0	0	0	0	
1	0	0	0	

2	0	0	0
3	0	0	0
4	0	0	0
5	0	0	0
6	0	0	0
7	0	0	0
8	0	0	0
9	0	0	0
10	0	0	0
11	0	0	0
12	0	0	0
13	0	0	0
14	0	0	0
15	0	0	0
16	0	0	0
17	0	0	0
18	0	0	1
19	0	0	1
20	0	0	1
21	0	0	0
22	0	0	0
23	0	0	0
24	0	0	0
25	0	0	0
26	0	0	0
27	0	0	0
28	0	0	0
29	0	0	0
...
27024	0	0	0
27025	0	0	0
27026	0	0	0
27027	1	0	0
27028	0	1	0
27029	0	0	0
27030	0	0	0
27031	0	0	0
27032	0	1	0
27033	0	0	0
27034	0	0	0
27035	0	0	0
27036	0	0	0
27037	0	0	0
27038	0	0	0
27039	0	0	0
27040	0	0	0
27041	0	0	0
27042	0	0	0

27043	0	0	0
27044	0	0	0
27045	0	0	1
27046	0	0	1
27047	0	0	1
27048	0	0	0
27049	0	0	0
27050	0	0	0
27051	0	0	0
27052	0	0	0
27053	0	0	0

	natural_gas	hybrid	electric_peak_hours	electric_off_peak_hours	\
0	1	0	0	0	
1	1	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	
5	0	0	0	0	
6	0	0	0	0	
7	0	0	0	0	
8	0	0	0	0	
9	0	0	0	0	
10	0	0	0	0	
11	0	0	0	0	
12	0	0	0	0	
13	0	0	0	0	
14	0	0	0	0	
15	0	0	0	0	
16	0	0	0	0	
17	0	0	0	0	
18	0	0	0	0	
19	0	0	0	0	
20	0	0	0	0	
21	0	0	0	0	
22	0	0	0	0	
23	0	0	0	0	
24	0	0	0	0	
25	0	0	0	0	
26	0	0	0	0	
27	1	0	0	0	
28	1	0	0	0	
29	0	0	0	0	
...	
27024	0	0	0	0	
27025	0	0	0	0	
27026	0	0	0	0	
27027	1	0	0	0	

27028	1	0	0	0
27029	0	0	0	0
27030	0	0	0	0
27031	0	1	0	0
27032	0	0	0	0
27033	0	0	0	0
27034	0	0	0	0
27035	0	0	0	0
27036	0	0	0	0
27037	0	0	0	1
27038	0	0	0	1
27039	0	0	0	1
27040	0	0	0	0
27041	0	0	0	0
27042	0	0	0	0
27043	0	0	0	0
27044	0	0	0	0
27045	0	0	0	0
27046	0	0	0	0
27047	0	0	0	0
27048	0	0	0	0
27049	0	0	0	0
27050	0	0	0	0
27051	0	0	0	0
27052	0	0	0	0
27053	0	0	0	0

	jetfuel
0	0
1	0
2	0
3	0
4	0
5	0
6	0
7	0
8	0
9	0
10	0
11	0
12	0
13	0
14	0
15	0
16	0
17	0
18	0
19	0

20	0
21	0
22	0
23	0
24	0
25	0
26	0
27	0
28	0
29	0
...	...
27024	0
27025	0
27026	0
27027	0
27028	0
27029	0
27030	0
27031	0
27032	0
27033	0
27034	0
27035	0
27036	0
27037	0
27038	0
27039	0
27040	0
27041	0
27042	0
27043	0
27044	0
27045	0
27046	0
27047	0
27048	0
27049	0
27050	0
27051	0
27052	0
27053	0

[27054 rows x 12 columns]