#### Question 15.2

In the videos, we saw the "diet problem". (The diet problem is one of the first large-scale optimization problems to be studied in practice. Back in the 1930's and 40's, the Army wanted to meet the nutritional requirements of its soldiers while minimizing the cost.) In this homework you get to solve a diet problem with real data. The data is given in the file diet.xls.

- Formulate an optimization model (a linear program) to find the cheapest diet that satisfies the maximum and minimum daily nutrition constraints, and solve it using PuLP.
   Turn in your code and the solution. (The optimal solution should be a diet of air-popped popcorn, poached eggs, oranges, raw iceberg lettuce, raw celery, and frozen broccoli. UGH!)
- 2. Please add to your model the following constraints (which might require adding more variables) and solve the new model: a. If a food is selected, then a minimum of 1/10 serving must be chosen. (Hint: now you will need two variables for each food i: whether it is chosen, and how much is part of the diet. You'll also need to write a constraint to link them.) b. Many people dislike celery and frozen broccoli. So at most one, but not both, can be selected. c. To get day-to-day variety in protein, at least 3 kinds of meat/poultry/fish/eggs must be selected. [If something is ambiguous (e.g., should bean-and-bacon soup be considered meat?), just call it whatever you think is appropriate I want you to learn how to write this type of constraint, but I don't really care whether we agree on how to classify foods!]

If you want to see what a more full-sized problem would look like, try solving your models for the file diet\_large.xls, which is a low-cholesterol diet model (rather than minimizing cost, the goal is to minimize cholesterol intake). I don't know anyone who'd want to eat this diet – the optimal solution includes dried chrysanthemum garland, raw beluga whale flipper, freeze-dried parsley, etc. – which shows why it's necessary to add additional constraints beyond the basic ones we saw in the video! [Note: there are many optimal solutions, all with zero cholesterol, so you might get a different one. It probably won't be much more appetizing than mine.]

```
In [1]: #!pip install PuLP
#!pip install xlrd
import pandas as pd
import pulp
from pulp import *
```

#### **EDA**

```
In [2]: excel_file = r"C:\Users\Clair\OneDrive\Documents\GitHub\omsa\ISYE 6501\Homework 11\
# read by default 1st sheet of an excel file
diet_data = pd.read_excel(excel_file)
```

# table preview
display(diet\_data)

	Foods	Price/ Serving	Serving Size	Calories	Cholesterol mg	Total_Fat	Sodium mg	Carbohy
0	Frozen Broccoli	0.16	10 Oz Pkg	73.8	0.0	0.8	68.2	
1	Carrots,Raw	0.07	1/2 Cup Shredded	23.7	0.0	0.1	19.2	
2	Celery, Raw	0.04	1 Stalk	6.4	0.0	0.1	34.8	
3	Frozen Corn	0.18	1/2 Cup	72.2	0.0	0.6	2.5	
4	Lettuce,Iceberg,Raw	0.02	1 Leaf	2.6	0.0	0.0	1.8	
•••								
62	Crm Mshrm Soup,W/Mlk	0.65	1 C (8 FI Oz)	203.4	19.8	13.6	1076.3	
63	Beanbacn Soup,W/Watr	0.67	1 C (8 FI Oz)	172.0	2.5	5.9	951.3	
64	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
65	NaN	NaN	Minimum daily intake	1500.0	30.0	20.0	800.0	
66	NaN	NaN	Maximum daily intake	2500.0	240.0	70.0	2000.0	

67 rows × 14 columns

In [3]: diet\_data.describe()

```
Out[3]:
                    Price/
                                        Cholesterol
                                                     Total Fat
                                                                            Carbohydrates Dietary
                               Calories
                                                                Sodium mg
                  Serving
                                               mg
                                                            g
         count 64.000000
                             66.000000
                                          66.000000 66.000000
                                                                 66.000000
                                                                                 66.000000
                                                                                               66.0
                 0.327188
                            190.918182
                                          21.615152
                                                     6.392424
                                                                364.486364
                                                                                                7.1
         mean
                                                                                 24.727273
                 0.254536
                                          49.660770 12.547158
           std
                            354.515790
                                                                528.629735
                                                                                 57.824484
                                                                                               34.0
                 0.020000
                                           0.000000
                                                     0.000000
                                                                                                0.0
           min
                              2.600000
                                                                  0.000000
                                                                                  0.000000
          25%
                                                                                                0.0
                 0.145000
                             73.975000
                                           0.000000
                                                     0.500000
                                                                 17.175000
                                                                                  5.000000
                                                                144.850000
          50%
                 0.270000
                            110.650000
                                           0.000000
                                                      2.900000
                                                                                 15.250000
                                                                                                3.0
          75%
                 0.460000
                            168.950000
                                          19.425000
                                                     7.075000
                                                                389.025000
                                                                                 22.450000
                                                                                                2.0
                 0.990000
                           2500.000000
                                        240.000000 72.200000
                                                               2000.000000
                                                                                450.000000
                                                                                              250.C
          max
In [4]: # I see NaNs in the table preview
         diet_data.isnull().values.any()
Out[4]: True
In [5]: # how many null rows?
         diet_data.isnull().sum()
Out[5]: Foods
                              3
                              3
         Price/ Serving
         Serving Size
                             1
         Calories
         Cholesterol mg
                             1
         Total_Fat g
                             1
         Sodium mg
                             1
         Carbohydrates g
         Dietary_Fiber g
                             1
         Protein g
                             1
         Vit_A IU
                             1
         Vit_C IU
                             1
         Calcium mg
                             1
         Iron mg
         dtype: int64
In [6]: # what are those rows, more importantly i want to know the indices
         diet_data[diet_data.isnull().any(axis=1)]
```

Out[6]:		Foods	Price/ Serving	Serving Size	Calories	Cholesterol mg	Total_Fat g	Sodium mg	Carbohydrates g	Di		
	64	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN			
	65	NaN	NaN	Minimum daily intake	1500.0	30.0	20.0	800.0	130.0			
	66	NaN	NaN	Maximum daily intake	2500.0	240.0	70.0	2000.0	450.0			
	4									•		
In [7]:	<pre># limit rows to just legit food rows diet_data_clean = diet_data[0:64] # making sure the null value columns are truly eliminated diet_data_clean.tail()</pre>											

Out[7]:

	Foods	Price/ Serving	Serving Size	Calories	Cholesterol mg	Total_Fat g	Sodium mg	Carbohyd
59	Neweng Clamchwd	0.75	1 C (8 FI Oz)	175.7	10.0	5.0	1864.9	
60	Tomato Soup	0.39	1 C (8 FI Oz)	170.7	0.0	3.8	1744.4	
61	New E Clamchwd,W/Mlk	0.99	1 C (8 FI Oz)	163.7	22.3	6.6	992.0	
62	Crm Mshrm Soup,W/Mlk	0.65	1 C (8 FI Oz)	203.4	19.8	13.6	1076.3	
63	Beanbacn Soup,W/Watr	0.67	1 C (8 FI Oz)	172.0	2.5	5.9	951.3	
4								<b>&gt;</b>

Min and max serving

# 1. Orgnaize data

```
In [8]: # selecting food rows (the rows that are not null)
    food_data = diet_data_clean
    nutrient_constraints = food_data.iloc[:, 3:].reset_index(drop=True) # every row of
    display(food_data.head(2)) # every row of the diet data clean --> diet_data_clean =
    display(nutrient_constraints.head(2)) # 4th column and on starting with calories
```

	Foods	Price/ Serving	Serving Size	Calories	Cholesterol mg	Total_Fat g	Sodium mg	Carbohydrates 9	-
0	Frozen Broccoli	0.16	10 Oz Pkg	73.8	0.0	0.8	68.2	13.6	
1	Carrots,Raw	0.07	1/2 Cup Shredded	23.7	0.0	0.1	19.2	5.6	
4								,	

	Calories	Cholesterol mg	Total_Fat g	Sodium mg	Carbohydrates g	Dietary_Fiber g	Protein g	Vit_A IU
0	73.8	0.0	0.8	68.2	13.6	8.5	8.0	5867.4
1	23.7	0.0	0.1	19.2	5.6	1.6	0.6	15471.0

```
In [9]: # extract costs and nutrient values
food_costs = food_data['Price/ Serving'].tolist()
    calories = food_data['Calories'].tolist()
    cholestrols = food_data['Cholesterol mg'].tolist()
    fats = food_data['Total_Fat g'].tolist()
    sodium = food_data['Sodium mg'].tolist()
    carbohydrates = food_data['Carbohydrates g'].tolist()
    fibers = food_data['Dietary_Fiber g'].tolist()
    vitamin_a = food_data['Vit_A IU'].tolist()
    vitamin_c = food_data['Vit_C IU'].tolist()
    calcium = food_data['Calcium mg'].tolist()
    iron = food_data['Iron mg'].tolist()
```

[0.16, 0.07, 0.04, 0.18, 0.02, 0.53, 0.06, 0.31, 0.84, 0.78, 0.27, 0.24, 0.15, 0.32, 0.49, 0.15, 0.16, 0.05, 0.06, 0.09, 0.16, 0.03, 0.05, 0.25, 0.16, 0.23, 0.13, 0.08, 0.11, 0.15, 0.27, 0.33, 0.15, 0.31, 0.28, 0.28, 0.34, 0.32, 0.38, 0.82, 0.52, 0.44, 0.59, 0.83, 0.31, 0.39, 0.08, 0.17, 0.07, 0.81, 0.45, 0.69, 0.04, 0.22, 0.12, 0.19, 0.39, 0.67, 0.71, 0.75, 0.39, 0.99, 0.65, 0.67]

# 2. Pulp object

```
In [10]: import pulp
    prob = pulp.LpProblem("diet_optimization", sense=pulp.LpMinimize)
    prob

Out[10]: diet_optimization:
    MINIMIZE
    None
    VARIABLES
```

### 2.1 Decision variables

```
Out[11]: [Frozen_Broccoli,
           Carrots, Raw,
           Celery,_Raw,
           Frozen_Corn,
           Lettuce, Iceberg, Raw,
           Peppers,_Sweet,_Raw,
           Potatoes,_Baked,
           Tofu,
           Roasted_Chicken,
           Spaghetti_W__Sauce,
           Tomato, Red, Ripe, Raw,
           Apple, Raw, W_Skin,
           Banana,
           Grapes,
           Kiwifruit, Raw, Fresh,
           Oranges,
           Bagels,
           Wheat_Bread,
           White Bread,
           Oatmeal_Cookies,
           Apple_Pie,
           Chocolate_Chip_Cookies,
           Butter, Regular,
           Cheddar Cheese,
           3.3% Fat, Whole Milk,
           2%_Lowfat_Milk,
           Skim_Milk,
           Poached_Eggs,
           Scrambled_Eggs,
           Bologna, Turkey,
           Frankfurter,_Beef,
           Ham, Sliced, Extralean,
           Kielbasa, Prk,
           Cap'N_Crunch,
           Cheerios.
           Corn_Flks,_Kellogg'S,
           Raisin_Brn,_Kellg'S,
           Rice_Krispies,
           Special_K,
           Oatmeal,
           Malt O Meal, Choc,
           Pizza_W_Pepperoni,
           Taco,
           Hamburger_W_Toppings,
           Hotdog,_Plain,
           Couscous,
           White Rice,
           Macaroni, Ckd,
           Peanut_Butter,
           Pork,
           Sardines_in_Oil,
           White_Tuna_in_Water,
           Popcorn, Air_Popped,
           Potato_Chips,Bbqflvr,
           Pretzels,
           Tortilla_Chip,
```

```
Chicknoodl_Soup,
Splt_Pea&Hamsoup,
Vegetbeef_Soup,
Neweng_Clamchwd,
Tomato_Soup,
New_E_Clamchwd,W_Mlk,
Crm_Mshrm_Soup,W_Mlk,
Beanbacn_Soup,W_Watr]
```

# 2.2 Add objective function

minimizing total food cost

```
# prob = pulp.lpSum([food_costs[i] * food_vars[i] for i in
range(len(food_data))]), 'Total Cost'
# prob
```

```
In [12]: # objective function by multiplying each variable by its associated cost and summin
prob += sum(food_data['Price/ Serving']*x)
prob
```

Out[12]: diet\_optimization:

MINIMIZE

0.23\*2% Lowfat Milk + 0.16\*3.3% Fat, Whole Milk + 0.24\*Apple, Raw, W Skin + 0.16\*Appl e\_Pie + 0.16\*Bagels + 0.15\*Banana + 0.67\*Beanbacn\_Soup,W\_Watr + 0.15\*Bologna,Turke y + 0.05\*Butter, Regular + 0.31\*Cap'N\_Crunch + 0.07\*Carrots, Raw + 0.04\*Celery, Raw + 0.25\*Cheddar Cheese + 0.28\*Cheerios + 0.39\*Chicknoodl Soup + 0.03\*Chocolate Chip Cookies + 0.28\*Corn Flks, Kellogg'S + 0.39\*Couscous + 0.65\*Crm Mshrm Soup,W Mlk + 0.27\*Frankfurter,\_Beef + 0.16\*Frozen\_Broccoli + 0.18\*Frozen\_Corn + 0.32\*Grapes + 0.33\*Ham,Sliced,Extralean + 0.83\*Hamburger\_W\_Toppings + 0.31\*Hotdog,\_Plain + 0.15\* Kielbasa,Prk + 0.49\*Kiwifruit,Raw,Fresh + 0.02\*Lettuce,Iceberg,Raw + 0.17\*Macaron i,Ckd + 0.52\*Malt\_O\_Meal,Choc + 0.99\*New\_E\_Clamchwd,W\_Mlk + 0.75\*Neweng\_Clamchwd + 0.82\*Oatmeal + 0.09\*Oatmeal\_Cookies + 0.15\*Oranges + 0.07\*Peanut Butter + 0.53\*Pep pers, Sweet, Raw + 0.44\*Pizza W Pepperoni + 0.08\*Poached Eggs + 0.04\*Popcorn,Air P opped + 0.81\*Pork + 0.22\*Potato\_Chips,Bbqflvr + 0.06\*Potatoes,\_Baked + 0.12\*Pretze ls + 0.34\*Raisin\_Brn,\_Kellg'S + 0.32\*Rice\_Krispies + 0.84\*Roasted\_Chicken + 0.45\*S ardines\_in\_Oil + 0.11\*Scrambled\_Eggs + 0.13\*Skim\_Milk + 0.78\*Spaghetti\_W\_Sauce + 0.38\*Special K + 0.67\*Splt Pea&Hamsoup + 0.59\*Taco + 0.31\*Tofu + 0.27\*Tomato,Red,R ipe,Raw + 0.39\*Tomato\_Soup + 0.19\*Tortilla\_Chip + 0.71\*Vegetbeef\_Soup + 0.05\*Wheat Bread + 0.06\*White Bread + 0.08\*White Rice + 0.69\*White Tuna in Water + 0.0

**VARIABLES** 2%\_Lowfat\_Milk Continuous 3.3% Fat, Whole Milk Continuous Apple, Raw, W Skin Continuous Apple Pie Continuous Bagels Continuous Banana Continuous Beanbacn\_Soup,W\_Watr Continuous Bologna, Turkey Continuous Butter, Regular Continuous Cap'N Crunch Continuous Carrots, Raw Continuous Celery, Raw Continuous Cheddar Cheese Continuous Cheerios Continuous Chicknoodl Soup Continuous Chocolate Chip Cookies Continuous Corn\_Flks,\_Kellogg'S Continuous Couscous Continuous Crm\_Mshrm\_Soup,W\_Mlk Continuous Frankfurter,\_Beef Continuous Frozen Broccoli Continuous Frozen Corn Continuous **Grapes Continuous** Ham, Sliced, Extralean Continuous Hamburger\_W\_Toppings Continuous Hotdog, Plain Continuous Kielbasa, Prk Continuous Kiwifruit, Raw, Fresh Continuous Lettuce, Iceberg, Raw Continuous

Macaroni, Ckd Continuous
Malt\_O\_Meal, Choc Continuous
New\_E\_Clamchwd, W\_Mlk Continuous
Neweng Clamchwd Continuous

Oatmeal\_Cookies Continuous

Oatmeal Continuous

Oranges Continuous

Peanut Butter Continuous Peppers,\_Sweet,\_Raw Continuous Pizza W Pepperoni Continuous Poached\_Eggs Continuous Popcorn, Air\_Popped Continuous Pork Continuous Potato\_Chips,Bbqflvr Continuous Potatoes, Baked Continuous Pretzels Continuous Raisin\_Brn,\_Kellg'S Continuous Rice\_Krispies Continuous Roasted\_Chicken Continuous Sardines\_in\_Oil Continuous Scrambled\_Eggs Continuous Skim Milk Continuous Spaghetti\_W\_\_Sauce Continuous Special\_K Continuous Splt\_Pea&Hamsoup Continuous Taco Continuous Tofu Continuous Tomato, Red, Ripe, Raw Continuous Tomato Soup Continuous Tortilla\_Chip Continuous Vegetbeef\_Soup Continuous Wheat Bread Continuous White Bread Continuous White\_Rice Continuous White\_Tuna\_in\_Water Continuous

#### 2.3 Add constraints

(min and max nutrients)

```
# prob += pulp.lpSum([calories[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,0],
"min calories"
# prob += pulp.lpSum([calories[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,0],</pre>
"max calories"
# # Cholesterols
# prob += pulp.lpSum([cholestrols[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,1],
"min cholestrols"
# prob += pulp.lpSum([cholestrols[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,1],</pre>
"max cholestrols"
# # Fats
# prob += pulp.lpSum([fats[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,2],
"min fats"
```

```
# prob += pulp.lpSum([fats[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,2],</pre>
"max fats"
# # Sodium
# prob += pulp.lpSum([sodium[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,3],
"min sodium"
# prob += pulp.lpSum([sodium[i] * food_vars[i] for i in
range(len(food data))]) <= nutrient constraints.iloc[1,3],</pre>
"max sodium"
# # Carbohydrates
# prob += pulp.lpSum([carbohydrates[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,4],
"min_carbohydrates"
# prob += pulp.lpSum([carbohydrates[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,4],</pre>
"max_carbohydrates"
# # Fibers
# prob += pulp.lpSum([fibers[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,5],
"min_fibers"
# prob += pulp.lpSum([fibers[i] * food vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,5],</pre>
"max fibers"
# # Vitamin A
# prob += pulp.lpSum([vitamin_a[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,6],
"min_vitamin_a"
# prob += pulp.lpSum([vitamin_a[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,6],</pre>
"max vitamin a"
# # Vitamin C
# prob += pulp.lpSum([vitamin_c[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,7],
"min vitamin c"
# prob += pulp.lpSum([vitamin_c[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,7],</pre>
"max_vitamin_c"
# # Calcium
# prob += pulp.lpSum([calcium[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,8],
"min calcium"
# prob += pulp.lpSum([calcium[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,8],</pre>
"max_calcium"
```

```
# # Iron
# prob += pulp.lpSum([iron[i] * food_vars[i] for i in
range(len(food_data))]) >= nutrient_constraints.iloc[0,9],
"min_iron"
# prob += pulp.lpSum([iron[i] * food_vars[i] for i in
range(len(food_data))]) <= nutrient_constraints.iloc[1,9],
"max_iron"</pre>
```

```
In [13]: # before importing the data, looking at the excel sheet, i can see Lettuce, Iceberg,

# Lowest and highest calories
min_calories_index = food_data['Calories'].idxmin()
max_calories_index = food_data['Calories'].idxmax()

# what row are these foods in
print("Index of minimum Calories:", min_calories_index)
print("Index of maximum Calories:", max_calories_index)

# Lets see the data for the foods im anticipating
min_calories_row = food_data.loc[min_calories_index] # should be Lettuce, Iceberg, Ra
max_calories_row = food_data.loc[max_calories_index] # should be Pork

print("Row with minimum Calories:\n", min_calories_row)
print("Row with maximum Calories:\n", max_calories_row)
```

```
Index of minimum Calories: 4
        Index of maximum Calories: 49
        Row with minimum Calories:
        Foods
                            Lettuce, Iceberg, Raw
        Price/ Serving
                                          0.02
                                        1 Leaf
        Serving Size
        Calories
                                           2.6
                                           0.0
        Cholesterol mg
        Total Fat g
                                           0.0
        Sodium mg
                                           1.8
        Carbohydrates g
                                           0.4
                                           0.3
        Dietary_Fiber g
        Protein g
                                           0.2
        Vit_A IU
                                          66.0
        Vit C IU
                                           0.8
                                           3.8
        Calcium mg
        Iron mg
                                           0.1
        Name: 4, dtype: object
        Row with maximum Calories:
        Foods
                             Pork
        Price/ Serving
                            0.81
        Serving Size
                           4 0z
        Calories
                           710.8
        Cholesterol mg
                          105.1
        Total_Fat g
                            72.2
        Sodium mg
                            38.4
        Carbohydrates g
                             0.0
        Dietary_Fiber g
                             0.0
        Protein g
                            13.8
        Vit_A IU
                            14.7
        Vit C IU
                             0.0
        Calcium mg
                            59.9
        Iron mg
                             0.4
        Name: 49, dtype: object
In [14]: food_columns = list(food_data.columns)
         print(food_columns)
         mins = min_calories_row
         maxes = max_calories_row
         print(mins)
         print(maxes)
         for nutrient in food columns[3:]:
             prob += sum(x*food_data[nutrient]) >= mins[nutrient]
```

prob += sum(x\*food\_data[nutrient]) <= maxes[nutrient]</pre>

```
['Foods', 'Price/ Serving', 'Serving Size', 'Calories', 'Cholesterol mg', 'Total_Fat
g', 'Sodium mg', 'Carbohydrates g', 'Dietary_Fiber g', 'Protein g', 'Vit_A IU', 'Vit
_C IU', 'Calcium mg', 'Iron mg']
Foods
                  Lettuce, Iceberg, Raw
Price/ Serving
                                  0.02
                                1 Leaf
Serving Size
Calories
                                   2.6
Cholesterol mg
                                   0.0
Total Fat g
                                   0.0
Sodium mg
                                   1.8
Carbohydrates g
                                   0.4
                                   0.3
Dietary_Fiber g
Protein g
                                   0.2
                                  66.0
Vit_A IU
Vit C IU
                                   0.8
                                   3.8
Calcium mg
Iron mg
                                   0.1
Name: 4, dtype: object
Foods
                    Pork
Price/ Serving
                    0.81
Serving Size
                    4 0z
Calories
                   710.8
Cholesterol mg
                   105.1
                    72.2
Total_Fat g
Sodium mg
                    38.4
Carbohydrates g
                     0.0
                     0.0
Dietary_Fiber g
Protein g
                    13.8
Vit_A IU
                    14.7
Vit_C IU
                     0.0
Calcium mg
                    59.9
Iron mg
                     0.4
Name: 49, dtype: object
```

In [15]: prob.writeLP("simple\_diet.lp")

Out[15]: [2%\_Lowfat\_Milk, 3.3%\_Fat,Whole\_Milk, Apple, Raw, W\_Skin, Apple\_Pie, Bagels, Banana, Beanbacn\_Soup, W\_Watr, Bologna, Turkey, Butter, Regular, Cap'N\_Crunch, Carrots, Raw, Celery,\_Raw, Cheddar Cheese, Cheerios, Chicknoodl\_Soup, Chocolate\_Chip\_Cookies, Corn\_Flks,\_Kellogg'S, Couscous, Crm Mshrm Soup, W Mlk, Frankfurter,\_Beef, Frozen\_Broccoli, Frozen\_Corn, Grapes, Ham, Sliced, Extralean, Hamburger W Toppings, Hotdog,\_Plain, Kielbasa, Prk, Kiwifruit, Raw, Fresh, Lettuce, Iceberg, Raw, Macaroni, Ckd, Malt\_O\_Meal,Choc, New\_E\_Clamchwd, W\_Mlk, Neweng\_Clamchwd, Oatmeal, Oatmeal\_Cookies, Oranges, Peanut\_Butter, Peppers,\_Sweet,\_Raw, Pizza\_W\_Pepperoni, Poached\_Eggs, Popcorn, Air\_Popped, Potato\_Chips, Bbqflvr, Potatoes,\_Baked, Pretzels, Raisin\_Brn,\_Kellg'S, Rice Krispies, Roasted Chicken, Sardines\_in\_Oil, Scrambled\_Eggs, Skim\_Milk, Spaghetti\_W\_\_Sauce, Special\_K, Splt\_Pea&Hamsoup, Taco, Tofu,

```
Tomato, Red, Ripe, Raw, Tomato_Soup, Tortilla_Chip, Vegetbeef_Soup, Wheat_Bread, White_Bread, White_Rice, White_Tuna_in_Water]
```

# 2.4 Solve prob

```
In [16]: prob = pulp.LpProblem("simple diet", pulp.LpMinimize)
         prob.solve()
         print("Status:", pulp.LpStatus[prob.status]) # should be optimal
        Status: Optimal
In [17]: # Print out any food with a solved value greater than 0
         for v in prob.variables():
             if v.varValue is not None and v.varValue > 0:
                 print(v.name, "=", v.varValue)
In [18]: import pulp
         # Define the problem
         prob = pulp.LpProblem("simple_diet", pulp.LpMinimize)
         # Example: Get food costs from your dataset
         food_costs = food_data['Price/ Serving'].tolist()
         print(food_costs)
         # Create decision variables for each food item
         food_vars = [pulp.LpVariable(f"x_{food_data['Foods'][i]}", lowBound=0, cat='LpConti
         print(food_vars)
         # Define the objective: Minimize the total cost (sum of cost * quantity of food)
         prob += pulp.lpSum([food_costs[i] * food_vars[i] for i in range(len(food_costs))]),
         # Solve the problem
         prob.solve()
         # Check if the solution is optimal
         if pulp.LpStatus[prob.status] == "Optimal":
             print("\nTotal Cost of Ingredients per day = $", round(pulp.value(prob.objectiv
         else:
             print("No optimal solution. Status:", pulp.LpStatus[prob.status])
```

[0.16, 0.07, 0.04, 0.18, 0.02, 0.53, 0.06, 0.31, 0.84, 0.78, 0.27, 0.24, 0.15, 0.32,0.49, 0.15, 0.16, 0.05, 0.06, 0.09, 0.16, 0.03, 0.05, 0.25, 0.16, 0.23, 0.13, 0.08, 0.11, 0.15, 0.27, 0.33, 0.15, 0.31, 0.28, 0.28, 0.34, 0.32, 0.38, 0.82, 0.52, 0.44, 0.59, 0.83, 0.31, 0.39, 0.08, 0.17, 0.07, 0.81, 0.45, 0.69, 0.04, 0.22, 0.12, 0.19, 0.39, 0.67, 0.71, 0.75, 0.39, 0.99, 0.65, 0.67] [x\_Frozen\_Broccoli, x\_Carrots,Raw, x\_Celery,\_Raw, x\_Frozen\_Corn, x\_Lettuce,Iceberg,R aw, x\_Peppers,\_Sweet,\_Raw, x\_Potatoes,\_Baked, x\_Tofu, x\_Roasted\_Chicken, x\_Spaghetti \_W\_\_Sauce, x\_Tomato,Red,Ripe,Raw, x\_Apple,Raw,W\_Skin, x\_Banana, x\_Grapes, x Kiwifrui t,Raw,Fresh, x Oranges, x Bagels, x Wheat Bread, x White Bread, x Oatmeal Cookies, x \_Apple\_Pie, x\_Chocolate\_Chip\_Cookies, x\_Butter,Regular, x\_Cheddar\_Cheese, x\_3.3%\_Fa t,Whole\_Milk, x\_2%\_Lowfat\_Milk, x\_Skim\_Milk, x\_Poached\_Eggs, x\_Scrambled\_Eggs, x\_Bol ogna, Turkey, x\_Frankfurter,\_Beef, x\_Ham, Sliced, Extralean, x\_Kielbasa, Prk, x\_Cap'N\_Cr unch, x\_Cheerios, x\_Corn\_Flks,\_Kellogg'S, x\_Raisin\_Brn,\_Kellg'S, x\_Rice\_Krispies, x\_ Special\_K, x\_Oatmeal, x\_Malt\_O\_Meal,Choc, x\_Pizza\_W\_Pepperoni, x\_Taco, x\_Hamburger\_W \_Toppings, x\_Hotdog,\_Plain, x\_Couscous, x\_White\_Rice, x\_Macaroni,Ckd, x Peanut Butte r, x\_Pork, x\_Sardines\_in\_Oil, x\_White\_Tuna\_in\_Water, x\_Popcorn,Air\_Popped, x\_Potato\_ Chips, Bbqflvr, x\_Pretzels, x\_Tortilla\_Chip, x\_Chicknoodl\_Soup, x\_Splt\_Pea&Hamsoup, x \_Vegetbeef\_Soup, x\_Neweng\_Clamchwd, x\_Tomato\_Soup, x\_New\_E\_Clamchwd,W\_Mlk, x\_Crm\_Msh rm\_Soup,W\_Mlk, x\_Beanbacn\_Soup,W\_Watr]

Total Cost of Ingredients per day = \$ 0.0

## 2.5 Solve another prob

```
In [19]: prob = pulp.LpProblem('Another Diet prob', pulp.LpMinimize)
         x = [0]*len(food data)
         for i, row in food data.iterrows():
             x[i] = pulp.LpVariable(row['Foods'], 0, None, pulp.LpInteger)
         prob += sum(food_data['Price/ Serving']*x)
         for nutrient in food_columns[3:]:
             prob += sum(x*food data[nutrient]) >= mins[nutrient]
             prob += sum(x*food_data[nutrient]) <= maxes[nutrient]</pre>
         prob.writeLP("another diet.lp")
         prob.solve()
         print(pulp.LpStatus[prob.status], '\n')
         print('Diet consists of: \n')
         for v in prob.variables():
             if v.varValue is not None and v.varValue > 0:
                  print(v.name, '=', v.varValue)
         print("\nTotal Cost of Ingredients per day = $",round(value(prob.objective),2))
```

Infeasible

Diet consists of:

Popcorn, Air\_Popped = 0.069767442

Total Cost of Ingredients per day = \$ 0.0

c:\Users\Clair\AppData\Local\Programs\Python\Python311\Lib\site-packages\pulp\pulp.p
y:1298: UserWarning: Spaces are not permitted in the name. Converted to '\_'
warnings.warn("Spaces are not permitted in the name. Converted to '\_'")