## Nutrition\_Analysis\_Regression

## November 30, 2019

```
In [1]: # import geopandas as gpd
        import numpy as np
        import pandas as pd
        import copy
In [2]: data = pd.read_csv("./data/nndb_flat.csv")
In [3]: data.head()
Out [3]:
             ID
                               FoodGroup
                                                       ShortDescrip
           1001 Dairy and Egg Products
                                                   BUTTER, WITH SALT
          1002 Dairy and Egg Products
                                          BUTTER, WHIPPED, WITH SALT
         1003 Dairy and Egg Products
                                              BUTTER OIL, ANHYDROUS
          1004 Dairy and Egg Products
                                                        CHEESE, BLUE
           1005
                 Dairy and Egg Products
                                                       CHEESE, BRICK
                               Descrip CommonName MfgName ScientificName
                                                                            Energy_kcal
        0
                       Butter, salted
                                              NaN
                                                       NaN
                                                                       NaN
                                                                                  717.0
        1
           Butter, whipped, with salt
                                              NaN
                                                       NaN
                                                                       NaN
                                                                                  717.0
        2
                Butter oil, anhydrous
                                                       NaN
                                                                                  876.0
                                              NaN
                                                                       NaN
                          Cheese, blue
        3
                                              NaN
                                                       NaN
                                                                       NaN
                                                                                  353.0
        4
                         Cheese, brick
                                              NaN
                                                                       NaN
                                                                                  371.0
                                                       NaN
                                                 Niacin_USRDA Riboflavin_USRDA
                                   Folate_USRDA
           Protein_g Fat_g
        0
                                                                         0.026154
                0.85
                      81.11
                                         0.0075
                                                      0.002625
                      81.11
        1
                0.85
                                         0.0075
                                                      0.002625
                                                                         0.026154
        2
                0.28
                      99.48
                                         0.0000
                                                      0.000188
                                                                         0.003846
        3
               21.40
                      28.74
                             . . .
                                         0.0900
                                                      0.063500
                                                                         0.293846
        4
               23.24 29.68 ...
                                         0.0500
                                                      0.007375
                                                                         0.270000
           Thiamin_USRDA
                         Calcium_USRDA Copper_USRDA
                                                         Magnesium_USRDA
        0
                0.004167
                                0.020000
                                              0.00000
                                                                0.004762
        1
                0.004167
                                0.020000
                                              0.000018
                                                                0.004762
        2
                0.000833
                                0.003333
                                              0.00001
                                                                0.000000
        3
                0.024167
                                0.440000
                                              0.000044
                                                                0.054762
                0.011667
                                0.561667
                                              0.000027
                                                                0.057143
```

Phosphorus\_USRDA Selenium\_USRDA Zinc\_USRDA

```
0
            0.034286
                             0.018182
                                          0.008182
1
            0.032857
                             0.018182
                                          0.004545
2
           0.004286
                             0.000000
                                          0.000909
3
           0.552857
                             0.263636
                                          0.241818
4
            0.644286
                             0.263636
                                          0.236364
```

[5 rows x 45 columns]

## 1 Data Preprocessing

```
In [4]: data.notnull()
        data = data.drop(columns=['CommonName','MfgName','ScientificName'])
In [5]: data.head()
Out [5]:
                                                        ShortDescrip \
             ID
                               FoodGroup
           1001
                Dairy and Egg Products
                                                   BUTTER, WITH SALT
        1
           1002 Dairy and Egg Products BUTTER, WHIPPED, WITH SALT
          1003 Dairy and Egg Products
                                               BUTTER OIL, ANHYDROUS
          1004 Dairy and Egg Products
                                                         CHEESE, BLUE
           1005 Dairy and Egg Products
                                                        CHEESE, BRICK
                               Descrip
                                         Energy_kcal Protein_g Fat_g
                                                                         Carb_g
                                                                                  Sugar_g
        0
                        Butter, salted
                                                            0.85
                                                                                     0.06
                                               717.0
                                                                  81.11
                                                                           0.06
        1
           Butter, whipped, with salt
                                               717.0
                                                            0.85 81.11
                                                                           0.06
                                                                                     0.06
        2
                Butter oil, anhydrous
                                                            0.28 99.48
                                                                           0.00
                                               876.0
                                                                                     0.00
        3
                          Cheese, blue
                                               353.0
                                                           21.40 28.74
                                                                           2.34
                                                                                     0.50
        4
                         Cheese, brick
                                                           23.24 29.68
                                                                           2.79
                                               371.0
                                                                                     0.51
                          Folate_USRDA
                                        Niacin_USRDA Riboflavin_USRDA
                                                                          Thiamin_USRDA
           Fiber_g
                    . . .
        0
                                0.0075
                                             0.002625
                                                                0.026154
                                                                                0.004167
               0.0
                    . . .
        1
               0.0
                                0.0075
                                             0.002625
                                                                0.026154
                                                                                0.004167
                    . . .
        2
               0.0
                                0.0000
                                             0.000188
                                                                0.003846
                                                                                0.000833
                    . . .
        3
               0.0
                                0.0900
                                             0.063500
                                                                0.293846
                                                                                0.024167
        4
               0.0
                                                                0.270000
                                0.0500
                                             0.007375
                                                                                0.011667
                    . . .
           Calcium_USRDA
                           Copper_USRDA
                                          Magnesium_USRDA
                                                            Phosphorus_USRDA
        0
                0.020000
                               0.00000
                                                 0.004762
                                                                    0.034286
        1
                0.020000
                               0.000018
                                                 0.004762
                                                                    0.032857
        2
                0.003333
                               0.00001
                                                 0.000000
                                                                    0.004286
        3
                0.440000
                               0.000044
                                                 0.054762
                                                                    0.552857
        4
                               0.000027
                0.561667
                                                                    0.644286
                                                 0.057143
           Selenium_USRDA
                            Zinc USRDA
        0
                 0.018182
                              0.008182
        1
                 0.018182
                              0.004545
        2
                 0.000000
                              0.000909
        3
                 0.263636
                              0.241818
```

```
0.263636
                             0.236364
        [5 rows x 42 columns]
In [6]: descrip = np.array(data['Descrip'])
        words = ""
        for i in descrip:
            words += i.lower()+','
        differentwords = (words.split(','))
In [7]: count = 0
        for i in differentwords:
            if('pizza' in i):
                count+=1
In [8]: count
Out[8]: 97
In [9]: len(set(differentwords))
Out[9]: 5952
In [10]: groups = np.array(data['FoodGroup'])
         len(set(groups))
Out[10]: 25
In [11]: set(differentwords)
Out[11]: {'',
          ' instant breakfast powder',
          ' 100 grand bar',
          'mustard spinach',
          ' apple and raspberry',
          ' dark',
          ' flavored with meat italian sauce',
          ' with broth',
          '80 proof',
          'juice smoothie',
          ' rolo caramels in milk chocolate',
          'p rego pasta',
          'fish sticks',
          'cattail',
          ' swordfish',
          ' royal red',
          ' mocha-flavor',
          ' perrier',
          ' mouvedre',
```

```
'next step prosobee',
' cornbread',
' bagel chips',
' berry burst cheerios',
' 3.25% milkfat',
' under blade pot roast or steak',
' grilled chicken & sausage gumbo soup',
" reese's pieces candy",
' hulled',
'link',
' bacon ranch salad without chicken',
' snack sticks',
' turkey breast (oven roasted',
' original recipe',
' crunchy almond/brown sugar',
' common (danish',
' pork)',
'chives',
' fruit flavored',
' commercially prepared with chocolate frosting',
' silken tofu',
'atlantic',
' peanut butter on cheese crackers',
' pinto bean and hominy',
' dehydrated flakes',
'flap',
' uses similar to high quality cocoa butter',
'skipjack',
' carnation breakfast essentials ',
' without caffeine',
' green variety',
' beefalo',
" sweet 'n sour sauce",
'keebler',
'rice',
' sweetened with low-calorie sweetener',
' with green chilies',
" hershey's golden almond solitaires",
' herring eggs',
' rotisserie',
' nutritional shake mix',
' fast roasted',
'eggnog-flavor mix',
' shoulder',
' non fat',
' ready-to-bake',
' waffle cones',
'house foods premium firm tofu',
```

```
' (sweetsop)',
' pudding',
' frosted wild grape toaster pastries',
' made with enriched masa flour',
' chow mein',
'vitasoy usa azumaya',
' baby ruth bar',
' (lox)',
' french or vienna',
' corn-based',
"carrabba's italian grill",
' sweet potatoes strained',
' cocktail',
'milk shakes',
' baked without fat',
'pasta mix',
' loin saddle',
' tri-tip steak',
' mature',
' cheddar cheese soup',
"morningstar farms chik'n grill veggie patties",
' with pulp',
' mayonnaise type',
' roast',
' ham (chopped with natural juice)',
" kellogg's all-bran bran buds",
' mango with tapioca',
' next step',
'kraft velveeta pasteurized process cheese spread',
' pork and beef',
' carrots and beef',
' granola bar',
' filled cream-type',
' tofu',
' vitamin d fortified',
' terra chips',
' water and margarine added',
' cinnamon brown sugar baked bites',
'worthington smoked turkey roll',
'new zealand spinach',
' low calorie',
' turkey franks',
'pomegranate juice',
' broad',
' without cheese',
' kidney',
' red leaf',
' blue or roquefort cheese',
```

```
" kellogg's crispix",
' fat free)',
'dill weed',
' denver cut',
' cranberry energy juice drink',
' all purpose',
' beef with country vegetables soup',
'loganberries',
'kellogg',
' 1/2 dipped',
' classic beef',
' sundae',
'vinegar',
' flakes without milk',
' shoulder steak',
'roast beef',
'silk very vanilla',
' back meat and skin',
' snickers cruncher',
' diluted with 3 volume water without added ascorbic acid',
' surimi',
' prepared with tap water',
'grape',
' propel zero',
' boiled (northern plains indians)',
' sausage mcmuffin with egg',
' salami (for beer)',
'brussels sprouts',
' with added oil',
' eas soy protein powder',
'85% lean',
' without added sodium or vitamin a',
' wieners (cheese hot dogs with turkey)',
'salami (genoa)',
' good start soy',
' and cheese',
' tilapia',
' chunky chicken noodle',
' starburst sour fruit chews',
' sunchips',
'pineapple and grapefruit juice drink',
' farley fruit snacks',
' club & cheddar sandwich crackers',
' lasagna classico',
' pinot noir',
' cream of asparagus',
' cheerios',
' sugar syrup/caramel',
```

```
' nutz over chocolate',
' composite of separable fat',
'pop',
' soy oil (partially hydrogenated )',
" baker's treasures",
' 95% lean meat / 5% fat',
' made from reduced fat packaged mix',
' tomato bisque',
' salada brewed from bags',
' pizza',
' bite-size',
'watercress',
' dinner',
' cream of onion soup',
' rib roast',
' chicken mushroom chowder',
' just bunches',
' vegetable soup',
' confectioners coating',
' lil crunchies',
' raisin vineyard',
' classic double',
' blended with cookie pieces',
' dark chocolate coated coffee beans',
' pregestimil',
'cowpeas (blackeyes)',
' 100 calorie right bites',
' peanut bar',
' reduced-calorie or diet',
' pm 60/40',
'whale',
' arm pot roast',
'slim-a-bear',
'sponge',
'beans',
'braunschweiger (a liver sausage)',
' toasteds',
' quaker',
' extra firm',
' mountain dew original',
'chili',
' curd cheese',
' peanut butter on toasty crackers',
' manhattan style',
' golden crisp',
'pizza hut 12" pepperoni pizza',
'frybread',
' smoked sliced beef',
```

```
' wheat and honey',
' pasta with cheese filling',
' special dietary',
' chicken and chipotle bbq sauce with mango',
' solids and liquid',
' popcorn',
'beverages',
' prime',
' cinnamon nut)',
' boiled with salt',
"campbell's healthy request",
' frijoles with cheese',
'vitasoy usa organic nasoya sprouted',
' 96% lean / 4% fat',
' cracker chips',
' (alaska native)',
' trimmed to 1/8" fat',
'eel',
' great grains crunchy pecan cereal',
' bun length)',
' club crackers',
' egg yolks and bacon',
' breast',
' banana and strawberry',
' high-fat',
' yane (alaska native)',
'farmed',
' italian style wedding soup',
' betty crocker supermoist yellow cake mix',
' potato ham chowder',
' corn based',
"morningstar farms chik'n nuggets",
' restaurant-prepared',
' herb and laborador combination (alaska native)',
' mature seeds',
' top loin filet',
'chokecherries',
' walleye',
' classic chicken noodle soup',
' garden',
'breakfast tart',
' quaker crunchy bran',
' neufchatel',
' sesame seed kernels',
' vitamins a and d',
' without ice',
' with cream style corn',
' chicken',
```

```
' quick',
' merry mint patties',
' soft taco with ground beef',
' and olive',
' principal use flaky pastries',
' inside skirt',
'plum',
' plums',
' milk chocolate peanut',
'sea cucumber',
" kellogg's rice krispies",
' mexican rice',
'ringed',
' sesame chicken',
'chicken tenders',
' calcium-fortified',
' baked apple pie',
'garlic',
' liquid',
'swanson broth',
' sucker',
' lightly frosted',
' mace',
" kellogg's cinnamon jacks",
'alfalfa seeds',
"campbell's homestyle new england clam chowder",
'lasagna with meat & sauce',
' pink',
' caramel filled cookies',
' pollock',
' special dietary (includes lemon-flavored)',
' high oleic (70% and over)',
'endive',
' with wheat flour added',
' cinnamon with crumb topping',
'eastern',
'chicken breast',
' with fruit and granola',
' country-style ribs',
'85% lean meat / 15% fat',
' with corn syrup and/or sugar and low calorie sweetener',
" kellogg's product 19",
' potherb',
'v8 v. fusion juices',
' fore-shank',
' very low sodium',
' crispbread',
' medallion',
```

```
' beef and mushroom',
' oven-roasted',
'vinespinach',
'pie',
' side salad',
' apple yogurt dessert',
' oregano',
' cheddar or american',
'lamb',
' japanese',
' from raw and stone ground kernels',
' salted and fermented (fuyu)',
' 98% fat free vanilla',
' carp',
'sprite',
"kraft breakstone's reduced fat sour cream",
' organic promise autumn wheat',
' nutty clusters & almonds',
'acerola juice',
' prepared from granules',
' whiskey) 80 proof',
' vitasoy organic creamy original soymilk',
' 3 musketeers bar',
' with beef and cheese',
' mamey',
' vanilla with nuts',
' fudge stripes',
'silk nog',
'lotus root',
' meat and skin and breading',
' shortbread bites',
'thin',
' slimfast',
' chicken broth or bouillon',
' horse',
' less than 3% juice',
' quaker oatmeal to go',
' french toaster sticks',
' soybean oil and butter',
' custard pudding',
'frozen novelties',
' milk chocolate covered',
' original seasoning',
' with apples',
' top round cap-off steak/roast',
'interstate brands corp',
' whole (arm and blade)',
' beef with white and wild rice soup',
```

```
'popeyes',
' fenugreek seed',
' fiber 7 flakes',
' 14" pizza',
' flavored',
' chicken broth cubes',
'corn with red and green peppers',
' thick-cut',
' regular (10 minute)',
'unfortified',
' healthy request mexican style tortilla',
' ice cream shoppe frosted rainbow chip toaster pastries',
' brined (alaska native)',
' potatoes',
'apple - cherry',
' chicken fajita strips',
'bockwurst',
' latino bakery item',
"wend'ys",
' microwavable',
' with bananas',
" kellogg's frosted mini-wheats touch of fruit in the middle mixed berry",
' frosted pumpkin pie toaster pastries',
' jacks vanilla wafers',
' green pea soup',
'fruit butters',
'assorted',
'prego pasta',
'ostrich',
' butter or sugar',
' coffee',
' 98% fat free cream of chicken soup',
' chicken fingers',
' butter (includes fresh and frozen)',
'tomato products',
' hot fudge sundae',
' chocolatey chip thins cookies',
' chicken fettuccine',
' chump off',
' fresh pork',
' drained solids',
' hazelnut',
' hot mustard sauce',
' made with palm oil',
' unenriched',
' kashi golean crunch!',
' simmered',
' ice cream shoppe frosted hot fudge sundae toaster pastries',
```

```
' 1% fat',
' junior',
' savory chicken with white & wild rice soup',
' with iron and fiber (formerly ross)',
' coconut cream',
' with chicory',
'salsify',
" kellogg's special k chocolatey strawberry",
'ling',
' supreme topping',
" kellogg's honey smacks",
' peach melba',
'meat drippings (lard',
' hain celestial group',
' mcchicken sandwich',
' coconut dreams cookies',
' and sugar',
' monterey',
' cream of shrimp soup',
' pre-basted',
'cereals ready-to-eat',
' savory herb crackers',
' cherry',
' fruit pudding',
' not breaded',
' potato',
'almonds',
' beef sticks',
'worthington multigrain cutlets',
' batter',
' fully hydrogenated',
' somen',
'peaches',
'radish seeds',
' and grape popsicle pops',
' v8 60% vegetable juice',
' ready-to-drink',
'elk',
' breadnut tree seeds',
'orange breakfast drink',
' chopped or leaf',
' lentil soup',
' cooked (southwest)',
' with hydrogenated vegetable oil and soy protein',
'frog legs',
' grass-fed',
'shake',
' nutramigen aa',
```

```
' packed in tomato juice',
' multi-grain (includes whole-grain)',
' cereal',
' sun country',
' regular',
'peach nectar',
' quesadilla',
' mustard',
' cottage-cut',
' cookie-like',
' frosted mini-wheats bite size strawberry delight',
' without calcium propionate(includes sourdough)',
' vegetables and chicken',
' prune and orange',
' kc masterpiece',
'edible-podded',
' boneless separable lean only',
' chocolate caramel',
' made from dried potatoes',
' organic mushroom italian sauce',
' black bean soup',
' whelk',
' roasted almond',
' oatmeal',
' pepper-type',
' vegetables and beef',
' farina',
"mcdonald's bacon ranch salad with crispy chicken",
' chicken gumbo soup',
' bread',
' pumpernickel',
'morningstar farms entree chili',
' soda',
' bologna (fat free)',
' (tendergreen)',
' fudgesicle pops',
' pasta with vegetables',
' creme-filled',
' smooth',
' trimmed to 1/8" fat',
' apple and cherry',
' vegetable oil spread',
' puffs or twists',
'water',
' homestyle chicken fillet sandwich',
' peanut flavor',
" pop'ables milky way brand bite size candies",
'cranberry-apple juice drink',
```

```
' cultured',
' naturally sparkling',
' white-winged',
' frosted raspberry',
' corn on the cob with butter',
' from dark meat',
' chex mix',
'chuck',
' rabbit',
'margarine-like spread with yogurt',
' jellied',
'morningstar farms roasted garlic & quinoa burger',
' cucumber',
' skittles original bite size candies',
' green peppers',
'chicken spread',
' grape juice',
' quick (1-3 minutes)',
' hash brown rounds',
' baked and topped with sour cream and chives',
'papayas',
' franks (turkey and chicken cheese)',
' commercially prepared',
' onion-flavor',
' fresh water',
' grape-nuts cereal',
' eggo seasons',
' cod',
' shark',
' dry heat',
' ranch-flavor',
" cap'n crunch with crunchberries",
'apples',
' chicken)',
' broiled (northern plains indians)',
' dark meat (drumstick or thigh)',
' mini brownies',
' frozen concentrate',
'peppers',
' deluxe grahams cookies',
'lemonade-flavor drink',
' snickers marathon protein performance bar',
' eggs (alaska native)',
" hershey's genuine chocolate flavored lite syrup",
' golden seedless',
' berry berry kix',
' and shoulder)',
' vegetable and brown rice',
```

```
' (vegetable oyster)',
' healthy request vegetable soup',
'apples',
'edamame',
' mini mints grasshopper cookies',
'carob-flavor beverage mix',
' italian dressing',
' cake-type',
' butternut',
' niacin',
' little bites',
'bread sticks',
' usda commodity corn and rice (includes all commodity brands)',
' (pineapple and papaya and banana and guava)',
"morningstar farms mushroom lover's burger",
' original chicken sandwich',
' flavored and sweetened',
' late harvest',
"mcdonald's",
'quinoa',
' vegetables chicken',
" general tso's chicken",
' chicken and stars soup',
' outside round',
' minis original crackers',
' oriental',
' center cut chops',
' french fries',
' golden puffs',
'morningstar farms grillers quarter pound veggie burger',
"campbell's homestyle mexican style chicken tortilla soup",
'quail',
'tap',
' premium',
' la moderna rikis cream crackers',
'kashi pizza',
' regular flavor',
" newman's own low fat balsamic vinaigrette",
' with franks',
' with low calorie sweetener',
' chicken a la king',
'fiddlehead ferns',
' southwestern-style chicken w/rice (chicken not included)',
' steak cut',
' 85% lean / 15% fat',
' teaseed',
' not canned',
' pound cake type',
```

```
' tonic water',
'egg substitute',
' mechanically deboned',
'worthington saucettes',
'dessert topping',
' lemon-lime flavored',
"campbell's chicken gravy",
' caffeine free',
' almond milk',
'citrus fruit juice drink',
' familia',
' stuffed crust',
' arm picnic',
'horseradish',
' vegetable and fruit juice blend',
' (granadilla)',
' banana (navajo)',
' milk chocolate peanut butter and soft nougats',
'pork and turkey sausage',
' canned in tomato sauce',
' dumpling with mutton (navajo)',
'fruit flavored drink containing less than 3% fruit juice',
' pearled',
" split pea 'n' ham soup",
' carving board)',
'onion',
' french lentil',
'orange peel',
" m&m's semisweet chocolate mini baking bits",
"campbell's red and white",
'chili con carne with beans',
'little caesars 14" cheese pizza',
' hard type',
'pimento',
'oopah (tunicate)',
' peanut butter gauchos cookies',
'cheese)',
' with corn flour coating (corndog)',
' double chocolate cookies',
" pop'ables snickers brand bite size candies",
' strawberry',
' soft taco',
'morningstar farms grillers burger style recipe crumbles',
' top round',
' sun-dried',
' wheat germ',
' home prepared',
' cottonseed flour',
```

```
'mixed vegetable and fruit juice drink',
' wafers with peanut butter',
' cooked with water',
' harvard',
' egg drop',
' oat blenders with honey',
' prepared with butter',
' in-store bakery',
' tilefish',
' salted',
' nut and raisin',
' meat and skin',
'enriched',
' monster',
' milk',
' sazon',
' traditional flavor',
'morningstar farms grillers prime',
'cranberry-grape juice drink',
'okara',
' precooked or instant',
'silk coffee',
'baby food',
' or banana powder',
' high vitamin c and added thiamin',
' whole grain',
'stick',
' potato wedges',
' unspecified oils',
' prepared with water or ready-to-serve',
' reduced-calorie',
' and beans',
' cookie crumb topping',
" kellogg's all-bran original",
' croaker',
' summer sausage thuringer cervalat',
'celeriac',
' klondike',
' total can contents',
' canola',
' ultimate deep dish crust',
' veal',
"kellogg's eggo lowfat blueberry nutri-grain waffles",
' gumdrops',
' duoz smoked cheddar monterey jack crackers',
' blueberry muffin tops cereal',
' spaghetti in tomato & cheese sauce',
' eggo minis',
```

```
' vegetables and mayonnaise',
' dha and ara',
' high oleic (70%)',
'whiskey) 94 proof',
' twix chocolate fudge cookie bars',
' without meat',
" pop'ables 3 musketeers brand bite size candies",
'unsweetened',
' organic nasoya silken tofu',
'hush puppies',
' grouper',
' peaches',
'pace',
' turkey vegetable',
' unenriched flour',
' magic middles fudge filled cookies',
' quick oats',
' chicken vegetable soup',
'shank',
' pourable liquid fry shortening',
'little caesars 14" original round pepperoni pizza',
' triple patty',
' cookie crisp',
' chardonnay',
"morningstar farms italian herb chik'n pattie",
' cooked (includes squab)',
' heated (microwave)',
' tootie fruities',
' banana yogurt',
' flowers',
' subcutaneous fat (blubber) (alaska native)',
'prairie turnips',
' raisin nut bran',
' classic tomato soup',
'apricot',
'fish',
'ice cream sandwich',
'oats',
' neck',
'pumpkin leaves',
' soy (partially hydrogenated ) and corn for frying',
' (pigeon)',
'cauliflower',
' toppers',
' zucchini',
' cream',
' thompson seedless',
'sweet potato leaves',
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' with chicken',
' beef tallow',
' pediasure',
' cheese-filled',
' chocolate with frosting',
' romano',
' round pieces or patty',
' naked juice',
' frosted wild strawberry toaster pastries',
' peanut butter sandwich',
' fudge shoppe',
' granulated',
' freeze-dried',
' rainbow chocolate chip cookies',
' blueberry hazelnut',
'yogurt',
' taro chips',
' smoked link sausage',
'salami (hard)',
' with juice and pulp',
' cola',
' oven-heated',
' with vitamin e added',
' chervil',
' chunky garden combination italian sauce',
' swanson chicken broth 99% fat free',
' choco zucaritas',
' hamburger or hotdog',
' holiday circus animal cookies',
'wocas',
'cake',
' dry form',
' packets',
' salpora de arroz con azucar',
'potato pancakes',
' carignane',
' puffed',
' pasta',
' glazed',
' smokies sausage little cheese (pork',
' fan fillet',
' kashi tlc bar',
' healthy request chicken with mini noodles soup',
' made with egg',
'lima',
' roasting',
' coconut water',
' tofu plus extra firm',
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'butter replacement',
' dices',
' honey nut cheerios',
'shoulder)',
'salmon',
' poppy',
' vanilla wafers',
' candy bits',
'bbq',
' peanut butter cups cookies',
' brownberry sage and onion stuffing mix',
' compressed',
' self-rising',
' tip side',
' from whole',
'silk hazelnut creamer',
' pilaf',
'rice flour',
' cool nestea ice tea lemon flavor',
' white popcorn',
' nutritional shake',
'extra crispy',
' vegetables and turkey',
' regular patty; with condiments and vegetables',
'croutons',
' french fried in vegetable oil',
'french toast',
' mashed',
' ricotta',
' rusk toast',
' mixed',
' tenderloin steak/roast',
'beet greens',
' coho (silver)',
' unenriched (includes honey buns)',
' unbaked',
' wing meat and skin',
' rice bran',
' without added salt',
'fast foods',
' microwave',
' sugar-coated almonds',
' animal',
' ready-to-feed',
'silk key lime soy yogurt',
' starburst fruit chews',
' coconut custard',
' fried chicken',
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' yumberry',
' sweet yeast bread',
'abiyuch',
' powder',
' in oil (alaska native)',
' boston (steaks)',
' bimbo bakeries usa',
' with cheese filling',
' frozen (liquid expressed from grated meat and water)',
' g performance o 2',
' patty; plain',
' fortified cereal bar',
' wieners (light pork',
'cloudberries',
' dannon',
' semi solid',
' traditional varieties',
'ham and cheese spread',
"tony's smartpizza whole grain 4x6 cheese pizza 50/50 cheese-frozen",
'ripe',
' grilled sirloin steak with hearty vegetables soup',
' canola harvest soft spread (canola',
' ribeye petite roast/filet',
' vegetables and ham',
' prepared with water and ice',
' oh henry! bar',
' with dha and ara (formerly ross)',
' contains wheat flour and rice flour',
' after eight mints',
' extra lean and regular',
'v8 splash smoothies',
' ham -- water added',
'breadfruit',
' cinnamon crunch',
' bottom sirloin butt',
'mouse nuts',
' with coconut',
' aust. marble score 9',
' includes plain and from mutton sandwich (navajo)',
' oat cluster cheerios crunch',
' claret',
'kellogg mini-wheats frosted bite size touch of fruit raisin',
' soybean oil',
' (skunk cabbage)',
' top loin petite roast',
' not from concentrate',
" o'brien",
' extra firm tofu',
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' meat-filled',
'pokeberry shoots',
' contains caffeine',
' soymilk',
' orange-flavor drink',
' store brand',
' young green',
' reduced sugar frosted flakes cereal',
'soy flour',
'soy',
'cranberries',
'asparagus',
'barley',
' wagyu',
' pasteurized',
' monkfish',
'yardlong bean',
' peppers',
' kashi go lean hot cereal',
" kellogg's special k",
' honey bunches of oats with vanilla bunches',
' sea salt',
' ribeye petite roast',
' plain (alaska native)',
' with added vitamin d',
' all natural light vanilla chocolate strawberry',
' boneless separable lean and fat',
' chunky garden mushroom and green pepper italian sauce',
'balsam-pear (bitter gourd)',
' infant',
' soybean without cholesterol',
' evaporated',
' cream cheese-flavor',
'glaze',
'pectin',
' wieners (beef franks',
' chocolate creme sandwich cookies',
'butcher boy meats',
'melon balls',
' sausage biscuit with egg',
'silk blueberry soy yogurt',
' chicken alphabet soup',
'whipped cream substitute',
' burbot',
' iced cookies',
' spoon-size',
' chocolate mousse',
'radishes',
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' scotch',
' milk chocolate coated raisins',
'soy sauce made from hydrolyzed vegetable protein',
' watermelon seed kernels',
' non-alcoholic',
'baking chocolate',
" hearty bean 'n' ham soup",
'fluid replacement',
' cheese enchilada',
' degermed',
' muller thurgau',
'buffalo',
'strawberries',
' lentil with ham',
' ground bulk/coarse ground',
' frosted wild berry',
'cheesefurter',
'whiskey) 86 proof',
' thomas english muffins',
' chicken and dumplings',
' shortbread pie crust',
' peanuts (for sundaes)',
' seedlings (alaska native)',
' extra light syrup',
' thank u berry munch cookies',
'broccoli',
' brownies',
'bread',
' blade (chops or roasts)',
' cinnamon rolls with icing',
' cinnamon crisp',
' chocolate almond snack bar',
' with cauliflower onion mustard',
' energy drink',
' plate',
' heart',
'zinfandel',
'sapote',
'sapodilla',
' truffles',
' mixed fruit',
' steam meal',
' soup)',
' diluted with 3 volume water',
' back ribs',
'shortening',
' red taco sauce',
' good start 2 essentials',
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'winged bean tuber',
          ' oatmeal dark chocolate cookies',
          ' egg and onion',
          'tuna',
          ' and red chili peppers',
          ' honey buzzers',
          ' cheese on wheat sandwich crackers',
          ' dried (prunes)',
          'worthington vegetarian burger',
          ' turkey and rice',
          ' chocolate malt powder',
          'duck',
          ' creamy chicken noodle soup',
          'enriched (n x 5.70)',
          ' and giblets and neck',
          ' stock',
          ' nutmeg butter',
          'silk plus omega-3 dha',
          ' sweet',
          'gardenburger sun-dried tomato basil burger',
          'beef sausage',
          ...}
In [12]: set(groups)
Out[12]: {'American Indian/Alaska Native Foods',
          'Baby Foods',
          'Baked Products',
          'Beef Products',
          'Beverages',
          'Breakfast Cereals',
          'Cereal Grains and Pasta',
          'Dairy and Egg Products',
          'Fast Foods',
          'Fats and Oils',
          'Finfish and Shellfish Products',
          'Fruits and Fruit Juices',
          'Lamb, Veal, and Game Products',
          'Legumes and Legume Products',
          'Meals, Entrees, and Side Dishes',
          'Nut and Seed Products',
          'Pork Products',
          'Poultry Products',
          'Restaurant Foods',
          'Sausages and Luncheon Meats',
          'Snacks',
          'Soups, Sauces, and Gravies',
          'Spices and Herbs',
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'Sweets',
          'Vegetables and Vegetable Products'}
In [13]: nonveg = [
          'American Indian/Alaska Native Foods',
          'Beef Products',
          'Dairy and Egg Products',
          'Finfish and Shellfish Products',
          'Lamb, Veal, and Game Products',
          'Pork Products',
          'Poultry Products',
          'Sausages and Luncheon Meats',]
         remove = ['Restaurant Foods', 'Meals, Entrees, and Side Dishes', 'Fast Foods', 'Baked I
In [14]: len(remove)
Out[14]: 4
In [15]: newlabels = []
         index = []
         j = 0
         for i in groups:
             if(i in nonveg):
                 newlabels.append('Non-Veg')
             elif(i in remove):
                 newlabels.append('Remove')
                 index.append(j)
             else:
                 newlabels.append('Veg')
             j+=1
In [16]: newlabels
Out [16]: ['Non-Veg',
          'Non-Veg',
          'Non-Veg',
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In [17]: len(index)
Out[17]: 1389
In [18]: from collections import Counter
In [19]: Counter(newlabels)
Out[19]: Counter({'Non-Veg': 3057, 'Veg': 4172, 'Remove': 1389})
In [20]: data['FoodGroup'] = newlabels
In [21]: data.head()
Out[21]:
              ID FoodGroup
                                                                           Descrip \
                                         ShortDescrip
                   Non-Veg
                                     BUTTER, WITH SALT
                                                                    Butter, salted
         0
           1001
         1 1002
                   Non-Veg
                            BUTTER, WHIPPED, WITH SALT Butter, whipped, with salt
         2 1003
                   Non-Veg
                                 BUTTER OIL, ANHYDROUS
                                                             Butter oil, anhydrous
         3 1004
                   Non-Veg
                                                                      Cheese, blue
                                          CHEESE, BLUE
         4 1005
                   Non-Veg
                                         CHEESE, BRICK
                                                                     Cheese, brick
            Energy_kcal Protein_g Fat_g Carb_g Sugar_g Fiber_g ... Folate_USRDA \
         0
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                               0.85 81.11
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                              23.24 29.68
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                                                              Calcium USRDA
                                                                              Copper USRDA
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            Magnesium_USRDA
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                                                                    0.241818
                    0.057143
                                       0.644286
                                                        0.263636
                                                                    0.236364
         [5 rows x 42 columns]
In [22]: datacopy = copy.deepcopy(data)
In [23]: data = data.loc[data['FoodGroup'] != 'Remove']
In [24]: len(data)
Out [24]: 7229
In [25]: data = data.drop(columns=['ShortDescrip','Descrip','ID'])
In [26]: energy = data['Energy_kcal']
         label = data['FoodGroup']
In [27]: data = data.drop(columns=['Energy kcal', 'FoodGroup'])
In [28]: data.head()
Out [28]:
            Protein_g Fat_g Carb_g
                                        Sugar_g Fiber_g VitA_mcg
                                                                     VitB6_mg
                                                                                VitB12 mcg
                  0.85
                       81.11
                                 0.06
                                           0.06
                                                     0.0
                                                              684.0
                                                                         0.003
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                        81.11
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                21.40
                        28.74
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                23.24 29.68
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            VitC_mg VitE_mg
                                    Folate_USRDA
                                                   Niacin_USRDA Riboflavin_USRDA
         0
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[5 rows x 37 columns]
```

#### 2 Splitting in 70:30

### 3 Linear Regression

## 4 Regression Metrics

#### 5 Lasso Regression

#### 6 Neural Network

```
In [46]: from keras.callbacks import ModelCheckpoint
from keras.models import Sequential
from keras.layers import Dense, Activation, Flatten
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error
from matplotlib import pyplot as plt
import seaborn as sb
import matplotlib.pyplot as plt
```

```
import pandas as pd
       import numpy as np
       import warnings
       warnings.filterwarnings('ignore')
       warnings.filterwarnings('ignore', category=DeprecationWarning)
       #from xgboost import XGBRegressor
In [99]: # Reference - https://towardsdatascience.com/deep-neural-networks-for-regression-prob
       NN_model = Sequential()
In [107]: # The Input Layer :
        NN_model.add(Dense(128,input_dim = data.shape[1], activation='relu'))
In [108]: # The Hidden Layers:
        NN_model.add(Dense(256,activation='relu'))
        NN_model.add(Dense(256,activation='relu'))
        NN_model.add(Dense(256,activation='relu'))
In [109]: # The Output Layer :
        NN_model.add(Dense(1,activation='linear'))
In [110]: # Compile the network :
        NN_model.compile(loss='mean_absolute_error', optimizer='adam', metrics=['mean_absolute_error', optimizer='adam', metrics=['mean_absolute_error']
        NN_model.summary()
Model: "sequential_6"
Layer (type) Output Shape
                                          Param #
_____
                      (None, 128)
dense 35 (Dense)
                                            4864
______
                      (None, 256)
dense_36 (Dense)
                                           33024
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dense 37 (Dense)
               (None, 256)
                                           65792
dense 38 (Dense)
                      (None, 256)
                                            65792
dense_39 (Dense)
               (None, 1)
                                           257
dense_40 (Dense)
                  (None, 128)
                                            256
     .....
                      (None, 256)
dense_41 (Dense)
                                           33024
dense_42 (Dense)
               (None, 256)
                                            65792
dense_43 (Dense)
                      (None, 256)
                                            65792
    _____
dense_44 (Dense)
               (None, 1)
                                  257
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Total params: 334,850 Trainable params: 334,850 Non-trainable params: 0

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In [111]: NN_model.fit(x_train, y_train, epochs=500, batch_size=32, validation_split = 0.3)
Train on 3542 samples, validate on 1518 samples
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Out[111]: <keras.callbacks.History at 0x7fb0c0b4bfd0>

In [112]: y\_pred\_test = NN\_model.predict(x\_test)

In [113]: reg\_metrics(y\_test,y\_pred\_test)

MSE: 413.08601574260456

RMSE: 20.32451760171947

MAE: 6.515900369446747

r2score: 0.985457129717469

# 7 Visualisation by SVD for Veg and Non-Veg

0.004286

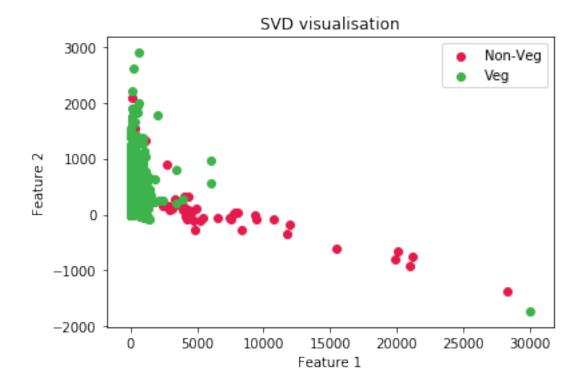
In [114]: data.head()

Out[114]:	Protein_g	$Fat\_g$	Carb_g	Sugar_g	Fiber_g	${\tt VitA\_mc}$	g VitB6_n	ng VitB	12_mcg	\
0	0.85	81.11	0.06	0.06	0.0	684.	0.00	)3	0.17	
1	0.85	81.11	0.06	0.06	0.0	684.	0.00	)3	0.13	
2	0.28	99.48	0.00	0.00	0.0	840.	0 0.00	)1	0.01	
3	21.40	28.74	2.34	0.50	0.0	198.	0 0.16	66	1.22	
4	23.24	29.68	2.79	0.51	0.0	292.	0.06	35	1.26	
	${\tt VitC\_mg}$ ${\tt V}$	itE_mg	Fo	late_USRD	A Niacin	_USRDA	Riboflavir	n_USRDA	\	
0	0.0	2.32		0.007	5 0.	002625	0.	.026154		
1	0.0	2.32		0.007	5 0.	002625	0.	.026154		
2	0.0	2.80		0.000	0 0.	000188	0.	.003846		
3	0.0	0.25		0.090	0 0.	063500	0.	293846		
4	0.0	0.26		0.050	0 0.	007375	0.	.270000		
								,		
	Thiamin_US	SRDA Ca	lcium_US		_	Magnesi	um_USRDA	\		
0	0.004167		0.020000		0.000000		0.004762			
1	0.004167		0.020000		0.000018		0.004762			
2	0.000833		0.003333		0.000001		0.000000			
3	0.024167		0.440000		0.000044		0.054762			
4	0.011667 Phosphorus_USRDA		0.561667		0.000027		0.057143			
			Selenium_USRDA		7inc HSDD	٨				
0			=		<del>-</del>					
0	0.034286		0.018182		0.008182					
1	0.032857		0.018182		0.00454	5				

0.000000 0.000909

```
3
                     0.552857
                                      0.263636
                                                  0.241818
                     0.644286
                                      0.263636
                                                  0.236364
          [5 rows x 37 columns]
In [115]: from sklearn.decomposition import TruncatedSVD
          svd = TruncatedSVD(n_components=2)
In [116]: new_data = svd.fit_transform(data)
In [117]: new_data.shape
Out[117]: (7229, 2)
In [123]: label
Out[123]: 0
                  Non-Veg
                  Non-Veg
          1
          2
                  Non-Veg
          3
                  Non-Veg
                  Non-Veg
                   . . .
          8613
                  Non-Veg
          8614
                  Non-Veg
          8615
                      Veg
          8616
                  Non-Veg
          8617
                  Non-Veg
          Name: FoodGroup, Length: 7229, dtype: object
In [125]: len(datan)
Out[125]: 7229
In [127]: new_data.shape
Out[127]: (7229, 2)
In [136]: import copy
          data1 = copy.deepcopy(data)
          datan = np.array(data1)
          label = list(label)
          dic={}
          for i in range(0,len(datan)):
              #print( datan[i][0] , datan[i][1] , datan[i][2] )
              try:
                  #print(label[i],i)
                  dic[label[i]].append ([new_data[i][0],new_data[i][1]])
              except:
                  #print(label[i])
                  dic[label[i]] = []
                  dic[label[i]].append ([new_data[i][0],new_data[i][1]])
              #print( datan[i][0] , dic[datan[i][0]] )
```

```
In [140]: len(dic['Non-Veg'])
Out[140]: 3057
In [141]: import matplotlib.pyplot as plt
          plt.title("SVD visualisation")
          plt.xlabel('Feature 1')
          plt.ylabel('Feature 2')
          colors=['#e6194b', '#3cb44b', '#ffe119', '#4363d8', '#f58231', '#911eb4', '#46f0f0',
          for type in dic.keys():
              \#plt.plot(new\_data[i,0:1], new\_data[i,1:2], color=colors[l], ls='None')
              plt.scatter(np.array(dic[type])[:,0],np.array(dic[type])[:,1], color=colors[1],1
              plt.legend(loc="best")
              1+=1
              #plt.xlim([-5, 6])
              #plt.ylim([-5, 5])
          #plt.savefig(name)
          #plt.show()
```



In [142]: data.shape

Out[142]: (7229, 38)