## Lahti2014\_solution

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## 1 Solution of Lahti et al. 2014

1.0.1 Write a function that takes as input a dictionary of constraints (i.e., selecting a specific group of records) and returns a dictionary tabulating the BMI group for all the records matching the constraints. For example, calling:

```
get_BMI_count({'Age': '28', 'Sex': 'female'})

1.0.2 should return:
{'NA': 3, 'lean': 8, 'overweight': 2, 'underweight': 1}

In [4]: import csv # Import csv modulce for reading the file
```

For each row in the file, you need to make sure all the constraints are matching the desired ones. If so, keep count of the BMI group using a dictionary.

```
In [5]: def get_BMI_count(dict_constraints):
            """ Take as input a dictionary of constraints
                for example, {'Age': '28', 'Sex': 'female'}
                And return the count of the various groups of BMI
            # We use a dictionary to store the results
            BMI_count = {}
            # Open the file, build a csv DictReader
            with open('../data/Lahti2014/Metadata.tab') as f:
                csvr = csv.DictReader(f, delimiter = '\t')
                # For each row
                for row in csvr:
                    # check that all conditions are met
                    matching = True
                    for e in dict_constraints:
                        if row[e] != dict_constraints[e]:
                            # The constraint is not met. Move to the next record
                            matching = False
                            break
                    # matching is True only if all the constraints have been met
                    if matching == True:
```

```
# extract the BMI_group
                      my_BMI = row['BMI_group']
                      if my_BMI in BMI_count.keys():
                          # If we've seen it before, add one record to the count
                          BMI_count[my_BMI] = BMI_count[my_BMI] + 1
                      else:
                          # If not, initialize at 1
                          BMI\_count[my\_BMI] = 1
           return BMI_count
In [6]: get_BMI_count({'Nationality': 'US', 'Sex': 'female'})
Out[6]: {'lean': 12, 'obese': 3, 'overweight': 5, 'severeobese': 1, 'underweight': 3}
1.0.3 Write a function that takes as input the constraints (as above), and a bacterial "genus".
     The function returns the average abundance (in logarithm base 10) of the genus for each
     group of BMI in the sub-population. For example, calling:
get_abundance_by_BMI({'Time': '0', 'Nationality': 'US'}, 'Clostridium difficile et rel.')
1.0.4 should return:
Abundance of Clostridium difficile et rel. In sub-population:
_____
Nationality -> US
Time -> 0
______
3.08
3.31
       underweight
3.84
       lean
2.89
      overweight
3.31
       obese
3.45
       severeobese
In [7]: import scipy # For log10
       def get_abundance_by_BMI(dict_constraints, genus = 'Aerococcus'):
           # We use a dictionary to store the results
           BMI_IDs = \{\}
           # Open the file, build a csv DictReader
           with open('../data/Lahti2014/Metadata.tab') as f:
               csvr = csv.DictReader(f, delimiter = '\t')
               # For each row
               for row in csvr:
                  # check that all conditions are met
                  matching = True
                  for e in dict_constraints:
```

```
if row[e] != dict_constraints[e]:
              # The constraint is not met. Move to the next record
             matching = False
             break
       # matching is True only if all the constraints have been met
      if matching == True:
          # extract the BMI_group
          my_BMI = row['BMI_group']
          if my_BMI in BMI_IDs.keys():
              # If we've seen it before, add the SampleID
             BMI_IDs[my_BMI] = BMI_IDs[my_BMI] + [row['SampleID']]
          else:
              # If not, initialize
             BMI_IDs[my_BMI] = [row['SampleID']]
# Now let's open the other file, and keep track of the abundance of the genus for ed
# BMI group
abundance = {}
with open('../data/Lahti2014/HITChip.tab') as f:
   csvr = csv.DictReader(f, delimiter = '\t')
   # For each row
   for row in csvr:
       # check whether we need this SampleID
      matching = False
      for g in BMI_IDs:
          if row['SampleID'] in BMI_IDs[g]:
              if g in abundance.keys():
                 abundance[g][0] = abundance[g][0] + float(row[genus])
                 abundance[g][1] = abundance[g][1] + 1
             else:
                 abundance[g] = [float(row[genus]), 1]
              # we have found it, so move on
             break
# Finally, calculate means, and print results
print("______")
print("Abundance of " + genus + " In sub-population:")
print("______")
for key, value in dict_constraints.items():
   print(key, "->", value)
print("______")
for ab in ['NA', 'underweight', 'lean', 'overweight',
         'obese', 'severeobese', 'morbidobese']:
   if ab in abundance.keys():
      abundance[ab][0] = scipy.log10(abundance[ab][0] / abundance[ab][1])
      print(round(abundance[ab][0], 2), '\t', ab)
print("______")
print("")
```

```
In [8]: get_abundance_by_BMI({'Time': '0', 'Nationality': 'US'},
                    'Clostridium difficile et rel.')
Abundance of Clostridium difficile et rel. In sub-population:
______
Nationality -> US
Time -> 0
______
3.08
3.31
         underweight
3.84
         lean
2.89
        overweight
3.31
         obese
3.45
         severeobese
```

## 1.0.5 Repeat this analysis for all genera, and for the records having Time = 0.

A function to extract all the genera in the database:

```
In [10]: def get_all_genera():
             with open('../data/Lahti2014/HITChip.tab') as f:
                 header = f.readline().strip()
             genera = header.split('\t')[1:]
             return genera
   Testing:
In [7]: get_all_genera()[:6]
Out[7]: ['Actinomycetaceae',
         'Aerococcus',
         'Aeromonas',
         'Akkermansia',
         'Alcaligenes faecalis et rel.',
         'Allistipes et rel.']
   Now use this function to print the results for all genera at Time = 0:
In [8]: for g in get_all_genera()[:5]:
            get_abundance_by_BMI({'Time': '0'}, g)
Abundance of Actinomycetaceae In sub-population:
Time -> 0
```

1.98	NA
1.95	underweight
1.98	lean
1.97	overweight
1.93	obese
1.95	severeobese
1.93	morbidobese
1.9	molpidopese
Abundance of	Aerococcus In sub-population:
Time -> 0	
1.66	NA
1.63	underweight
1.66	lean
1.66	overweight
1.61	obese
1.62	severeobese
1.6	morbidobese
Abundance of	Aeromonas In sub-population:
Abundance of	Aeromonas In sub-population:
	Aeromonas In sub-population:
Abundance of Time -> 0	
Time -> 0	
Time -> 01.68	NA
Time -> 0	NA underweight
Time -> 0 1.68 1.68 1.69	NA underweight lean
Time -> 0  1.68 1.68 1.69 1.69	NA underweight lean overweight
Time -> 0  1.68 1.68 1.69 1.69 1.66	NA underweight lean overweight obese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66	NA underweight lean overweight obese severeobese
Time -> 0  1.68 1.68 1.69 1.69 1.66	NA underweight lean overweight obese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.66	NA underweight lean overweight obese severeobese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.66	NA underweight lean overweight obese severeobese morbidobese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66	NA underweight lean overweight obese severeobese morbidobese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66	NA underweight lean overweight obese severeobese morbidobese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.63	NA underweight lean overweight obese severeobese morbidobese
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.63	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:
Time -> 0  1.68 1.69 1.69 1.66 1.66 1.63 Abundance of Time -> 0	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.63 Abundance of Time -> 0 3.53	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.63 Abundance of Time -> 0 3.53 4.0	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:
Time -> 0  1.68 1.69 1.69 1.66 1.66 1.63 Abundance of Time -> 0 3.53 4.0 3.65	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:  NA underweight lean
Time -> 0  1.68 1.68 1.69 1.69 1.66 1.66 1.63 Abundance of Time -> 0 3.53 4.0	NA underweight lean overweight obese severeobese morbidobese  Akkermansia In sub-population:

3.48 3.35	severeobese morbidobese
Abundance of	Alcaligenes faecalis et rel. In sub-population:
Time -> 0	
2.32	NA
2.26	underweight
2.36	lean
2.37	overweight
2.49	obese
2.43	severeobese
2.26	morbidobese