Assignment2

1) The frequent itemsets you obtain on Gene dataset (L1, L2, L3).

L1:

```
\{qene_1\}: sup = 0.83
\{qene_12\}: sup = 0.54
\{qene_17\}: sup = 0.55
\{gene_21\}: sup = 0.62
\{gene_22\}: sup = 0.55
\{qene_23\}: sup = 0.54
\{qene_3\}: sup = 0.71
\{gene_39\}: \sup = 0.51
\{gene_4\}: sup = 0.5
\{qene_{45}\}: sup = 0.58
\{gene_48\}: \sup = 0.57
\{gene_{55}\}: sup = 0.55
\{gene_{59}\}: sup = 0.76
\{qene_6\}: sup = 0.66
\{gene_63\}: sup = 0.5
\{gene_66\}: sup = 0.59
\{qene_{72}\}: sup = 0.74
\{gene_{77}\}: sup = 0.58
\{gene_83\}: sup = 0.5
\{gene_84\}: sup = 0.54
\{qene_93\}: sup = 0.53
\{qene_99\}: sup = 0.56
\{gene_14\}: sup = 0.52
\{gene_26\}: sup = 0.52
\{qene_27\}: sup = 0.51
\{gene_36\}: \sup = 0.61
\{gene_37\}: \sup = 0.56
\{qene_47\}: sup = 0.66
```

```
\{gene_5\}: sup = 0.73
\{qene_{50}\}: sup = 0.5
\{qene_54\}: sup = 0.67
\{gene_56\}: \sup = 0.51
\{gene_60\}: sup = 0.54
\{qene_{75}\}: sup = 0.57
\{qene_{78}\}: sup = 0.59
\{gene_81\}: sup = 0.58
\{gene_87\}: \sup = 0.67
\{qene_89\}: sup = 0.59
\{gene_25\}: \sup = 0.57
\{gene_43\}: sup = 0.5
\{qene_{53}\}: sup = 0.5
\{qene_71\}: sup = 0.58
\{gene_8\}: sup = 0.66
\{gene_9\}: sup = 0.5
\{qene_{90}\}: sup = 0.52
\{qene_91\}: sup = 0.65
\{gene_98\}: \sup = 0.51
\{gene_31\}: sup = 0.51
\{qene_94\}: sup = 0.62
\{gene_64\}: sup = 0.5
\{gene_67\}: \sup = 0.62
```

L2:

```
{gene_21, gene_1}: sup = 0.53

{gene_1, gene_3}: sup = 0.63

{gene_59, gene_1}: sup = 0.62

{gene_1, gene_6}: sup = 0.59

{gene_1, gene_72}: sup = 0.61

{gene_84, gene_1}: sup = 0.5

{gene_59, gene_3}: sup = 0.56

{gene_3, gene_72}: sup = 0.53

{gene_59, gene_6}: sup = 0.51

{gene_59, gene_72}: sup = 0.62
```

```
\{gene_47, gene_1\}: sup = 0.59
\{qene_5, qene_1\}: sup = 0.65
\{qene_54, qene_1\}: sup = 0.58
\{gene_81, gene_1\}: sup = 0.51
\{gene_87, gene_1\}: \sup = 0.56
\{gene_89, gene_1\}: sup = 0.52
{gene_47, gene_3}: sup = 0.5
\{gene_5, gene_3\}: \sup = 0.59
\{gene_59, gene_5\}: \sup = 0.51
\{qene_{59}, qene_{87}\}: sup = 0.51
{gene_5, gene_72}: sup = 0.51
\{gene_5, gene_47\}: \sup = 0.53
\{qene_5, qene_87\}: sup = 0.51
{gene_1, gene_8}: sup = 0.53
\{gene_91, gene_1\}: sup = 0.55
\{gene_5, gene_6\}: sup = 0.52
\{qene_91, qene_5\}: sup = 0.5
\{gene_94, gene_1\}: \sup = 0.54
\{gene_67, gene_1\}: sup = 0.55
```

L3:

```
{gene_59, gene_1, gene_72}: sup = 0.5
{gene_5, gene_1, gene_3}: sup = 0.52
```

2) The length-3 candidate itemsets generated during Apriori (C3) on Gene dataset.

```
{gene_1, gene_3, gene_59}: sup = 0.48

{gene_1, gene_3, gene_72}: sup = 0.46

{gene_1, gene_6, gene_59}: sup = 0.45

{gene_1, gene_72, gene_59}: sup = 0.5

{gene_3, gene_72, gene_59}: sup = 0.47

{gene_1, gene_3, gene_47}: sup = 0.45

{gene_1, gene_3, gene_5}: sup = 0.52

{gene_1, gene_59, gene_5}: sup = 0.44

{gene_1, gene_59, gene_87}: sup = 0.41
```

```
{gene_1, gene_72, gene_5}: sup = 0.45

{gene_3, gene_59, gene_5}: sup = 0.44

{gene_3, gene_72, gene_5}: sup = 0.42

{gene_1, gene_5, gene_47}: sup = 0.48

{gene_1, gene_5, gene_87}: sup = 0.45

{gene_3, gene_5, gene_47}: sup = 0.42

{gene_59, gene_72, gene_5}: sup = 0.42

{gene_59, gene_5, gene_87}: sup = 0.35

{gene_1, gene_6, gene_5}: sup = 0.48

{gene_1, gene_5, gene_91}: sup = 0.45

{gene_59, gene_6, gene_5}: sup = 0.45
```

3) The codes of the two functions: apriori_gen and get_freq.

apriori_gen:

```
def apriori_gen(freq_sets, k):

"""Generates candidate itemsets (via the F_k-1 x F_k-1 method).

This part generates new candidate k-itemsets based on the frequent (k-1)-itemsets found in the previous iteration.

The apriori_gen function performs two operations:
(1) Generate length k candidate itemsets from length k-1 frequent itemsets
(2) Prune candidate itemsets containing subsets of length k-1 that are infreq uent

Parameters
-------
freq_sets: list
The list of frequent (k-1)-itemsets.

k: integer
The cardinality of the current itemsets being evaluated.
```

```
Returns
  candidate_list : list
     The list of candidate itemsets.
  11 11 11
  # TODO
  candidate_list = []
  # print(freq_sets) ## [frozenset({'Key-chain'}), frozenset({'Mango'}), frozen
set({'Yo-yo'}), frozenset({'Eggs'}), frozenset({'Onion'})]
  # Generate length k candidate itemsets from length k-1 frequent itemsets (F
_k-1 x F_k-1)
  for i in range(len(freq_sets)):
     for j in range(i+1, len(freq_sets)):
       first = list(freq_sets[i])
       second = list(freq_sets[j])
       # print(first)
       # print(second)
       # sort two list to ensure the order of items in the list is the same
       first.sort()
       second.sort()
       # if the first k-2 items are the same
       if first[:k-2] == second[:k-2]:
          first.append(second[-1])
          candidate_list.append(first)
            print(first)
       # print("----\n")
  # print("candidate_list:")
  # print(candidate_list) ## [['Key-chain', 'Mango'], ['Key-chain', 'Yo-yo'], ['Ke
y-chain', 'Eggs'], ['Key-chain', 'Onion'], ['Mango', 'Yo-yo'], ['Mango', 'Eggs'],
['Mango', 'Onion'], ['Yo-yo', 'Eggs'], ['Yo-yo', 'Onion'], ['Eggs', 'Onion']]
```

```
# Prune candidate itemsets containing subsets of length k-1 that are infrequ
ent
  pruned_candidate_list = []
  freq_sets_list = []
  for s in freq_sets:
    s=list(s)
    s.sort()
    freq_sets_list.append(s)
  # print(freq_sets_list)
  for candidate in candidate_list:
    all_possible_candidate_subsets = combinations(candidate, k-1)
    passed = True
    # check if all the subsets of the candidate are in freq_sets_list
    for subset in all_possible_candidate_subsets:
       sorted_subset = list(subset)
       sorted_subset.sort()
       if (sorted_subset in freq_sets_list) and (sorted_subset not in pruned_ca
ndidate_list):
         continue
       else:
         passed = False
         break
    if passed:
       pruned_candidate_list.append(candidate)
  # turn list of frozensets
  pruned_candidate_list = list(map(frozenset, pruned_candidate_list))
  # print("pruned_candidate_list:")
```

```
# print(pruned_candidate_list, "\n\n") ##
return pruned_candidate_list
```

get_freq:

```
def get_freq(dataset, candidates, min_support, verbose=False):
  This function separates the candidates itemsets into frequent itemset and in
frequent itemsets based on the min_support,
  and returns all candidate itemsets that meet a minimum support threshold.
  Parameters
  -----
  dataset: list
     The dataset (a list of transactions) from which to generate candidate
     itemsets.
  candidates: frozenset
     The list of candidate itemsets.
  min_support : float
     The minimum support threshold.
  Returns
  -----
  freq_list : list
    The list of frequent itemsets.
  support_data : dict
     The support data for all candidate itemsets.
  11 11 11
  # TODO
  freq_list = []
```

```
support_data = {}
  # get support count for each candidate
  for transaction in dataset:
    for candidate in candidates:
       if candidate.issubset(transaction):
         if candidate in support_data:
            support_data[candidate] += 1
         else:
            # use update method to add new candidate to support_data
            support_data.update({candidate: 1})
  # After getting the support count, calculate the support and filter out the inf
requent itemsets
  for key in support_data:
    # support = (support count) / (total number of transactions)
    support_data[key] /= len(dataset)
    if support_data[key] >= min_support:
       freq_list.append(key)
  # print("support_data:")
  # print(support_data) ## {frozenset({'Corn'}): 0.4, frozenset({'Ice-cream'}):
0.4, frozenset({'Key-chain'}): 0.8, frozenset({'Mango'}): 0.8, frozenset({'Umbr
ella'}): 0.4, frozenset({'Yo-yo'}): 0.6, frozenset({'Doll'}): 0.4, frozenset({'Egg
s'}): 0.6, frozenset({'Onion'}): 0.8, frozenset({'Apple'}): 0.2, frozenset({'Ninten
do'}): 0.2}
  # print("frequent itemsets:")
  # print(freq_list) ## [frozenset({'Key-chain'}), frozenset({'Mango'}), frozens
et({'Yo-yo'}), frozenset({'Eggs'}), frozenset({'Onion'})]
  return freq_list, support_data
```

4) Al usage disclosure statement:

Did you use any Al tools (such as ChatGPT, Microsoft CoPilot, or similar) in completing this assignment?

No, I tried to finish this on my own.