Gene Data Transaction Frequent List

```
{gene 1}: \sup = 83
{gene 12}: \sup = 54
{gene 14}: \sup = 52
\{gene_{17}\}: sup = 55
{gene 21}: \sup = 62
{gene 22}: \sup = 55
{gene 23}: \sup = 54
\{gene 25\}: sup = 57
{gene 26}: \sup = 52
{gene 27}: sup = 51
{gene 3}: \sup = 71
{gene 31}: \sup = 51
{gene 36}: \sup = 61
{gene 37}: \sup = 56
{gene 39}: \sup = 51
{gene 4}: \sup = 50
{gene 43}: \sup = 50
{gene 45}: sup = 58
{gene 47}: \sup = 66
{gene 48}: sup = 57
{gene 5}: \sup = 73
{gene 50}: \sup = 50
{gene 53}: \sup = 50
{gene 54}: \sup = 67
{gene 55}: \sup = 55
\{gene_{56}\}: sup = 51
\{gene 59\}: sup = 76
{gene 6}: \sup = 66
{gene 60}: \sup = 54
{gene 63}: \sup = 50
{gene 64}: sup = 50
{gene 66}: \sup = 59
\{gene_67\}: sup = 62
{gene 71}: \sup = 58
{gene 72}: \sup = 74
{gene 75}: sup = 57
{gene 77}: \sup = 58
{gene 78}: sup = 59
{gene 8}: \sup = 66
{gene 81}: \sup = 58
{gene 83}: \sup = 50
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{gene 84}: \sup = 54
{gene 87}: \sup = 67
\{gene 89\}: sup = 59
{gene 9}: \sup = 50
{gene 90}: \sup = 52
{gene 91}: \sup = 65
{gene 93}: \sup = 53
\{gene 94\}: sup = 62
{gene 98}: \sup = 51
\{gene 99\}: sup = 56
{gene 1, gene 21}: \sup = 53
{gene 1, gene 3}: \sup = 63
{gene 1, gene 47}: sup = 59
{gene 5, gene 1}: \sup = 65
{gene 54, gene 1}: \sup = 58
{gene 1, gene 59}: \sup = 62
{gene 6, gene 1}: \sup = 59
{gene 67, gene 1}: \sup = 55
{gene 72, gene 1}: \sup = 61
\{gene_8, gene 1\}: sup = 53
{gene 1, gene 81}: \sup = 51
{gene 1, gene 84}: \sup = 50
{gene 87, gene 1}: \sup = 56
{gene 89, gene 1}: \sup = 52
{gene_91, gene 1}: \sup = 55
{gene 94, gene 1}: \sup = 54
{gene 3, gene 47}: sup = 50
{gene 5, gene 3}: \sup = 59
{gene 3, gene 59}: \sup = 56
{gene 72, gene 3}: \sup = 53
{gene 5, gene 47}: sup = 53
{gene 5, gene 59}: \sup = 51
{gene 6, gene 5}: \sup = 52
{gene 5, gene 72}: sup = 51
{gene 5, gene 87}: \sup = 51
{gene 5, gene 91}: \sup = 50
{gene 6, gene 59}: \sup = 51
{gene 72, gene 59}: \sup = 62
\{gene_87, gene_59\}: sup = 51
{gene 5, gene 1, gene 3}: \sup = 52
{gene 72, gene 1, gene 59}: \sup = 50
```

Gene Data Transaction Length-3 Candidates

Get Frequent List Function

```
def get_freq(dataset, candidates, min_support, verbose=False):
   This function separates the candidates itemsets into frequent itemset and infrequent 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.
   min_sup = len(dataset) * min_support #Calculate minimum support as integer
   support_data = {}
   freq_list = []
   for cand in candidates:
       support = 0
       for data in dataset:
               support += 1
       if support >= min_sup:
           freq_list.append(cand) # add the candidate to freq_list if it is supported
       support_data[cand] = support
   return freq_list, support_data
```

Apriori Generation Function

```
143 v 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 infrequent
        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.
        candidate_list = []
           for i in range(len(freq_sets)):
                for j in range(i + 1, len(freq_sets)):
                    candidate_list.append(freq_sets[i] | freq_sets[j])
          for i in range(len(freq_sets)): # otherwise use the Fk-1 * Fk-1 method
                for j in range(i + 1, len(freq_sets)):
                    one = sorted(list(freq_sets[i]))
                    two = sorted(list(freq_sets[j]))
                    if one[:k - 2] == two[:k - 2]:
                        candidate_list.append(freq_sets[i] | freq_sets[j]) # if they match union them
        remove = [] #List of candidates to prune
```

```
for new_set in candidate_list:

unfrozenset = set(new_set)

#print(list(unfrozenset))

for item in list(unfrozenset): # for each cadidate. remove one item from the set at a time

unfrozenset.discard(item)

#print(unfrozenset)

if unfrozenset not in freq_sets: #if the new subeset is not frequent

#print(candidate_list)

#print(new_set)

#print(found')

if new_set in candidate_list: #prune it from the candidate list

remove.append(new_set)

unfrozenset.add(item)

for item in remove:

candidate_list.remove(item) #remove all items from the list

return candidate_list

return candidate_list
```