## EECS 4412 Data Mining Assignment 1 Report

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Date: 22 Feb 2021

**Question 1)** Part A

levelUp.py works by applying Apriori algorithm to mine frequent itemset for a given threshold. Before

applying the pre-candidate optimization, the algorithm will join on the prefixes which is created randomly

to create pre-candidates, then applying the Apriori property to finalize the candidates. This can lead to a

huge set of pre-candidates' generation, result in a slower running time. Here is the result before

optimization by testing with level 5 of the mushroom dataset.

#pre-candidates: 236859

Lapsed time:

187.406

To achieve a more efficient algorithm, we sort the item in the itemset from least to most frequent, such

that the pre-candidates' generation can be more optimized and less pre-candidates will be created. Hence,

increasing the performance of the algorithm. Here is the result with optimization, the number of pre-

candidates are greatly reduced, and the running time is slightly improved as a result.

#pre-candidates: 173877

172300

Lapsed time:

170.566

## Question 2) Part A)

| I alt Aj |          |          |              |               |            |          |
|----------|----------|----------|--------------|---------------|------------|----------|
|          |          |          |              |               |            |          |
|          |          |          |              |               |            |          |
|          |          |          |              |               |            |          |
|          |          |          |              |               |            |          |
|          | A) Tran  | formed : | transactions | i temsets     |            |          |
|          |          |          | N KE Y}      | ->            | EK EM      | 0 Y 3    |
|          | 72       | 8 DOM    | 1 KEY3       | $\rightarrow$ | { K E O Y  | <i>}</i> |
|          |          | EMA      |              | ->            | EKE M3     |          |
|          | 74       | { MUC    | KY3          | ->            | EKM Y3     |          |
|          |          | 200 K    |              | $\Rightarrow$ | {KE O}     |          |
|          |          |          |              |               | ,          |          |
|          | Frequent | Items    |              |               |            |          |
|          | Item     | count    | Support      |               | [Root      |          |
|          | K        | 5        | 100% -       |               |            |          |
|          | E        | 4        | 80% -        |               | > K:5      |          |
|          | M        | 3        | 60%          |               | . 7        | 441.1    |
|          | 0        | 3        | 60%          | 1             | [F: 4] n   | M: 1     |
|          | Y        | 3        | 60%          | , ',          | E: 4       |          |
|          | /        | 3        |              |               | TM: 2/10:2 | 7 17:1   |
|          |          |          |              | 1 ,           | 14:51      | 1        |
|          |          |          |              | 7             |            |          |
|          |          |          |              | ``            | 0:1        | 1 1      |
|          |          |          |              | 7             |            |          |
|          |          |          |              | [             | Y: 1 /     |          |
|          |          |          |              |               | ·          |          |
|          |          |          |              |               |            |          |
|          |          |          |              |               |            |          |

Part B)

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B) Constitutal Pattern - Bases and constitutal FP- Thee
     Items conditional-pattern Base
    0 KEM
M KE: 2, K:1
                                               {(K: 37} / M
            K: 4
                                                    { ( K: 4)} ( E
  Frequent Pattern Generated
   {K, 0:3}, {E, 0:3}, {0, E, K:3}
{K, M:3}
{K, E:3}
  Association Rules Generation
   \begin{array}{lll} \{0\} & \rightarrow \{E, K\} & L & Support = 60\%, & confidence = 100\% \\ \{E\} & \rightarrow \{0, K\} & L & Support = 60\%, & confidence = 45\% \\ \{K\} & \rightarrow \{0, E\} & L & Support = 60\%, & confidence = 60\% \\ \end{array}
   \{K\} \rightarrow \{M\} L Support = 60%, confidence = 60%]

\{M\} \rightarrow \{K\} L Support = 60%, confidence = 100%]

\{K\} \rightarrow \{E\} L Support = 80%, confidence = 80%]

\{E\} \rightarrow \{E\} L Support = 80%, confidence = 100%]
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| After pruning by redunancy $\begin{cases} \{Y\} \rightarrow \{K\} \\ \{E\} \rightarrow \{K\} \\ \{K\} \rightarrow \{E\} \\ \{N\} \rightarrow \{E\} \end{cases}$ $\begin{cases} \{N\} \rightarrow \{E,K\} \\ \{N\} \rightarrow \{K\} \end{cases}$ |
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