First Name - Sai Prathyusha

Last Name - Devarapalli

NJIT UCID - sd894

Email address- sd894@njit.edu

Part 1

Creating Database with 20 transactions

*D4 - Notepad

File Edit Format View Help Sony, Realme, Nokia, Samsung LG, Sony Ericsson, Motorola, Blackberry HTC, Razer Phone 2, Honor, Lava Intex, MI, Micromax, Poco x4 RedMI, iQOO 9 5G, Jio Phone, Alcatel Yota, Lenovo, Apple, Samsung Apple, Samsung, Xiaomi, Oppo Sony Ericsson, Motorola, HTC, Razer Phone 2 MI, Micromax, iQOO 9 5G, Jio Phone Micromax, Lenovo, Google pixel Huawei, Sony Ericsson, Apple, Lava OnePlus, Huawei, Xiaomi Vivo, Intex, MI, Micromax Poco x4, Lava, Samsung, Xiaomi iQOO 9 5G,Oppo, Sony Ericsson, Motorola Huawei, Sony Ericsson, Google pixel, Apple Lava, , Xiaomi, Oppo MI, HTC,Yota, Jio Phone

Creating four additional databases

D3 - Notepad

File Edit Format View Help Refrigerator, T.V, Micro oven, Bulb Chargers, Geyser, Toaster, Laptop Washing Machine, Hair dryers, Smart Watches, PC Tablets, Headphones, Speakers, Air Conditioner Heating Pads, Power Bank, Vacuum Cleaner, Induction Stove Kettle, Blenders, CC Camera, MIC Rice Cooker, Iron box, Extension Cord, Stabilizer, Dish Washer, Steamer Bulb, Chargers, Geyser, Vacuum Cleaner Toaster, Blenders, Extension Cord, Laptop Speakers, Air Conditioner, Heating Pads, Power Micro oven, Bulb, Rice Cooker, Iron box Tablets, Headphones, Geyser, Vacuum Cleaner Iron box, Washing Machine, Hair dryers, Extension Cord Induction Stove, CC Camera, Power Bank, Vacuum Cleaner Kettle, Extension Cord, Iron box, Stabilizer Headphones, Speakers, Blenders, Bulb Extension Cord, Laptop, Air Conditioner, Heating Pads Dish Washer, Bulb, Chargers, Steamer Bulb, Toaster, Blenders Chargers, Geyser

D2 - Notepad

File Edit Format View Help Shirts, Phants, Watches, Jeans Socks, Sarees, Jackets, Head Cap Muffler, Rain Coats, Gum Boots, Dresses Sweater, Hand Gloves, Bath Towels, Shoes Themals, Frocks, Bed sheets, Kurtas Phants, Sarees, Rain Coats, Hand Gloves Shirts, Socks, Muffler, Sweater, Themals Watches, Paper towel, Cloth Bag, Bath Towels Jeans, Head Cap, Dresses Hand Gloves, Bath Towels, Bed sheets, Kurtas Rain Coats, Watches, Themals, Shirts Gum Boots, Dresses, Sweater, Hand Gloves Muffler, Head Cap, Jeans, Jackets Socks, Sarees, Bed sheets Kurtas, Hand Gloves, Thermals, Bath Towels Hand Gloves, Frocks, Sarees, Watches Napkins, Comforters Bed sheets, Kurtas, Phants, Sarees Bath Towels, Hand Gloves, Frocks, Sarees T-shirts, Tops, Rain Coats, Sarees



File Edit Format View Help

Sweets, Muffins, Bun, Pies

Chips, Cakes, Cool Drinks, Pastries Rolls, Coffee, Chocolates, Muffins Cupcakes, Cookies, Bread, Bun Donuts, Biscuits, Burger, Pizza Milkshakes, Croissant, Noodles, Taccos Puffs, Pancakes, Browines, Puddings Desserts, Hot Dogs, Sweets, Pies Sponge Cake, Chocolate Chips Cakes, Cool Drinks Donuts, Biscuits, Bun, Pizza Burger, Croissant, Pies, Milkshakes Pancakes, Puddings, Sweets Desserts, Pies, Browines, Hot Dogs Biscuits, Puffs, Burger, Pizza Hot Dogs, Chocolate Chips, Cakes Sponge Cake, Sweets, Pies Browines, Puddings, Milkshakes, Hot Dogs, Biscuits, Sponge Cake, Donuts, Pancakes

Cookies, Milkshakes, Cool Drinks, Browines

Browines, Cool Drinks, Puddings, Donuts

D1 - Notepad

File Edit Format View Help

Coffee Mug, Water bottles, Tea cups Saucers, Spoons, Coffee Mug, Pan Instant Pot, Pan ,Tea cups, Lunch boxes Serving skimmer, Glass bowels, Juicer, Spoons, Water bottles, Pan, Roti Maker Mixer grinder, Tea cups, Oil Tin, Storage Boxes Whisker, Lunch boxes, Instant Pot, Plates Pan, Whisker, Plates, Coffee Mug, Juicer Basket, Serving skimmer, Mixer grinder, Glass bowels Juicer, Vegetable Chopper, Storage Boxes, Water bottles Coffee Mug, Tava, Paper Cups, Table Cloth Mixer grinder, Whisker, Basket, Juicer Holders, Oil tin, Stainless Steel Products, Knife Water bottles, Serving skimmer, Storage Boxes, Lunch boxes, Tava, Jars Basket, Vessels, Cutting Board, Saucers Vegetable Chopper, Mixer grinder, Glass bowels, Oil tin Jars, Tea cups, Whisker, Lunch boxes, Storage Boxes, Stirring Mug, Basket, Roti Maker Home Mats, Saucers, Spoons

Using Apriori algorithm ,generating the association rules

import itertools from itertools import combinations import time File=input("Please Enter the path of the Dataset: ") open_file=open(File,'r') read_lines=open_file.readlines() inps=[] for i in range(len(read_lines)): inps.append(read_lines[i].strip('\n')) Data=[] for i in inps: split_inp=i.split(" ") Data.append(split_inp) print(len(Data)) print("Select the option whether you are reading in Percentage or Value") print("1.Percentage") print("2.Points") option=int(input()) print("Enter the support") min_support=float(input()) print("Enter the confidence") min confidence=float(input()) if option==1: min_support=min_support/100 min_confidence=min_confidence/100 Start_time=time.time() data_merge=[] count_dict={} for i in Data: data_merge+=i items_data=list(set(data_merge)) for i in items_data:

count_dict[i]=(data_merge.count(i))

```
First itt minsprt=∏
First itt minsprtt=[]
for x, y in count_dict.items():
if((y/len(Data))>=min_support):
a=[x,y]
First_itt_minsprt.append(x)
First_itt_minsprtt.append([x])
First_itt_minsprtt.append(y)
comb=list(itertools.combinations(First_itt_minsprt, 2))
def AssociationRules(freqSet):
associationRule = []
import itertools
for item in freqSet:
if isinstance(item, list):
if len(item) != 0:
length = len(item) - 1
while length > 0:
combinations = list(itertools.combinations(item, length))
temp = []
LHS = []
for RHS in combinations:
LHS = set(item) - set(RHS)
temp.append(list(LHS))
temp.append(list(RHS))
associationRule.append(temp)
temp = []
length = length - 1
return associationRule
def transac(comb,Data,min_support):
abc={}
for i in comb:
count=0
for j in Data:
if (set(i).issubset(j)):
count+=1
#print()
abc[i]=count
minii=[]
for x, y in abc.items():
if((y/len(Data))>=min_support):
First_itt_minsprtt.append(list(x))
First_itt_minsprtt.append(y)
minii.append(x)
are=minii
gf=[]
for i in range (len(are)):
for j in range(i+1,len(are)):
gf.append(tuple(set(are[i]+are[j])))
return gf,minii
sfg=[]
check=True
while check:
fnl,rdcd=transac(comb,Data,min_support)
if(len(fnl)!=0):
comb=fnl
sfg.append(rdcd)
print(fnl)
print(rdcd)
else:
if(len(rdcd)==0):
stp=sfg[-1]
else:
```

```
stp=rdcd
check=False
print("Dpo")
print(First_itt_minsprtt)
print("Frequent Item Sets")
print(stp)
def AssociationRules(freqSet):
associationRule = []
import itertools
for item in freqSet:
if isinstance(item, list):
if len(item) != 0:
length = len(item) - 1
while length > 0:
combinations = list(itertools.combinations(item, length))
temp = []
LHS = []
for RHS in combinations:
LHS = set(item) - set(RHS)
temp.append(list(LHS))
temp.append(list(RHS))
associationRule.append(temp)
temp = []
length = length - 1
return associationRule
associationRules = AssociationRules(First_itt_minsprtt)
def AprioriAlgorithm(rules, dataset, minSupport, minConfidence):
returnAprioriAlgorithm = []
for rule in rules:
supportOfX = 0
supportOfXinPercent = 0
supportOfXandY = 0
supportOfXandYinPercent = 0
for transaction in dataset:
if set(rule[0]).issubset(set(transaction)):
supportOfX = supportOfX + 1
if set(rule[0] + rule[1]).issubset(set(transaction)):
supportOfXandY = supportOfXandY + 1
supportOfXinPercent = (supportOfX * 1.0 / 20) * 100
supportOfXandYinPercent = (supportOfXandY * 1.0 / 20) * 100
confidence = (supportOfXandYinPercent / supportOfXinPercent) * 100
if confidence >= minConfidence:
returnAprioriAlgorithm.append(rule)
return returnAprioriAlgorithm
Apr_alg=AprioriAlgorithm(associationRules, Data, min_support, min_confidence)
counter = 1
if len(Apr alg) == 0:
print("There are no association rules for this support and confidence.")
for i in Apr_alg:
if counter == 4:
counter = 0
else:
print(i, end=' ')
counter = counter + 1
print("Time for Aprior Algorithm is")
print(time.time()-Start_time)
```

Output:

```
Sup(x): 10.0
Sup(X & Y:) 10
Confidence of List: 100
['Motorola,', 'Ericsson,']---->['Sony']
Sup(x): 25.0
Sup(X & Y:) 10
Confidence of List: 40
['Sony', 'Ericsson,']---->['Motorola,']
Sup(x): 10.0
Sup(X & Y:) 10
Confidence of List: 100
['Sony', 'Motorola,']---->['Ericsson,']
Time for Aprior Algorithm is
9.007231712341308594
```

Apriori Algorithm - 0.007231712341308594

BruteForce Algorithm

```
import sys
import time
import tkinter as tk
from tkinter import filedialog
print("Select the option whether you are reading in Percentage or Value")
print("1.Percentage")
print("2.Points")
option=int(input())
print("Enetr the support")
min_support=float(input())
print("Enter the confidence:")
min_confidence=float(input())
if option==1:
min_support=min_support/100
min_confidence=min_confidence/100
t_file = input("Enter the file path: ")
with open("C:/Users/HP/Desktop/datasets/items.txt") as f:
items = f.read().replace("\n", "").split(",")
items.sort()
print(items)
with open(t_file) as f:
db = [l.replace("\n", "").split(" ") for l in f]
a=[]
                           ------ INPUT TRANSACTIONS:")
print("--
for transaction in db:
print(transaction)
for i in transaction:
a.append(i)
#items=list(set(a))
```

```
def generate_k(items, k):
if k == 1:
return [[x] for x in items]
all_res = []
for i in range(len(items)-(k-1)):
for sub in generate_k(items[i+1:], k-1):
tmp = [items[i]]
tmp.extend(sub)
all_res.append(tmp)
return all_res
def scan(db, s):
count = 0
for t in db:
if set(s).issubset(t):
count += 1
return count
def generate_frequent_and_support():
frequent = []
support = \{\}
for k in range(1, len(items)+1):
current = []
for comb in generate_k(items, k):
count = scan(db, comb)
if count/len(db) >= min_support:
support[frozenset(comb)] = count/len(db)
current.append(comb)
if len(current) == 0:
break
frequent.append(current)
return frequent, support
class Rule:
def __init__(self, left, right, all):
self.left = list(left)
self.left.sort()
self.right = list(right)
self.right.sort()
self.all = all
def _str_(self):
return ",".join(self.left)+" => "+",".join(self.right)
def _hash_(self):
Store support value to dict
:return: hash value in the object
return hash(str(self))
def generate_rules(frequent, support):
all_rule = set()
all_result = []
for k_freq in frequent:
if len(k\_freq) == 0:
continue
```

```
if len(k_freq[0]) < 2:
continue
for freq in k_freq:
for i in range(1, len(freq)):
for left in generate_k(freq, i):
tmp = freq.copy()
right = [x for x in tmp if x not in left]
all_rule.add(Rule(left, right, freq))
for rule in all_rule:
confidence = support[frozenset(rule.all)] / support[frozenset(rule.left)]
if confidence >= min_confidence:
all\_result.append([rule, support[frozenset(rule.all)], confidence])
all_result.sort(key=lambda x: str(x[0]))
return all_result
if __name__ == '__main___':
start time = time.time()
f, s = generate_frequent_and_support()
all_result = generate_rules(f, s)
end_time = time.time()
                                 ----- RULES SUPPORT CONFIDENCE:")
print("\n-----
for r in all_result:
print(r[0], r[1], r[2])
print("\n-----
                                                ----- RUNNING TIME:")
print(str(end_time - start_time) + "s")
```

Output

Comparing Apriori and Bruteforce:

When we try to generate association rules for transaction all data bases with both Apriori and Bruteforce algorithms, we found that Apriori algorithm is way faster than Brute force.

Here are the time results:

ForApriori

0.007231712341308594

For Brute Force

1.0878620147705078

Conclusion:

From the above observations in the data sets,we can conclude that the Apriori algorithm is faster than the Brute force.