



**Polytechnic University of the Philippines – Sta. Mesa**

**Academic Year 2023 – 2024**

*Second Semester*

## **GROUP 1: YOMIFY**

In Partial Fulfillment

Of the Academic Requirement in Integrative Programming

From Bachelor of Science in Information Technology 2 – 4

Under Professor Aleta C. Fabregas

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## Introduction

In the digital age, music streaming platforms like Spotify have revolutionized how we consume music, offering vast libraries of songs at our fingertips. Among these, curated playlists have become particularly popular, often amassing millions of listens. One such playlist is "The Ultimate OPM," which has garnered significant attention on Spotify. Original Pilipino Music (OPM) is a significant cultural phenomenon in the Philippines, and understanding what makes certain OPM songs resonate more with audiences can provide valuable insights for musicians, composers, and producers.

This study aims to analyze the musical patterns and chord progressions that contribute to the popularity of songs in "The Ultimate OPM" playlist. By leveraging a dataset comprising these songs and their corresponding chords, we seek to uncover the underlying patterns that make these songs widely appealing. To achieve this, we employ the Apriori algorithm, a well-established method for mining frequent itemsets and discovering association rules in large datasets.

## Problem Definition and Objectives

The goal of this study is to analyze the musical patterns and chord progressions that contribute to the popularity of songs in the "The Ultimate OPM" Spotify playlist. By leveraging a dataset comprising the songs and their corresponding chords, we aim to uncover the underlying patterns that make these songs resonate with a wide audience. This analysis will be conducted using the Apriori algorithm, a popular method for mining frequent itemsets and discovering association rules in large datasets. Our objectives, **Identify Frequent Chord Progressions**, Using the Apriori algorithm, the study aims to mine frequent chord progressions within the songs in the "The Ultimate OPM" playlist and **Discover Associations Between Chords**: The study seeks to uncover associations between different chords in the songs, which can help in understanding the common musical structures in popular OPM music.

## Historical Data

The historical data was collected manually, focusing on the chord progressions from one of the most popular OPM playlists, "The Ultimate OPM," on Spotify. The dataset consists of the following columns: Composition, A, A#, Ab, B, Bb, C, C#, D, D#, Db, E, Eb, F, F#, G, G#, Gb, with a total of 18 columns.

Dataset collected by: BSIT 2-4, Group 1 from "The Ultimate OPM" playlist on Spotify.

([https://drive.google.com/file/d/1BMCPI-QP1PSnnFeMnpINXgklzeBCA6hL/view?usp=drive\\_link](https://drive.google.com/file/d/1BMCPI-QP1PSnnFeMnpINXgklzeBCA6hL/view?usp=drive_link))

The historical data focuses on identifying which chords are the most used in OPM compositions. A chord is a group of (typically three or more) notes sounded together, as a basis of harmony, but in this analysis, we focus only on the root note of a chord for the analysis.

## Analysis

*Computer Version, it includes the table format in text and the actual output in the program.*

antecedents	consequents	antecedent support	consequent support	support	confidence	lift	conviction
A	B	0.8356	0.6301	0.5479	0.6557	1.0406	1.0744
B	A	0.6301	0.8356	0.5479	0.8696	1.0406	1.2603
C	A	0.5479	0.8356	0.4795	0.8750	1.0471	1.3151
A	C	0.8356	0.5479	0.4795	0.5738	1.0471	1.0606
A	D	0.8356	0.8219	0.7123	0.8525	1.0372	1.2070
D	A	0.8219	0.8356	0.7123	0.8667	1.0372	1.2329
E	A	0.7397	0.8356	0.6575	0.8889	1.0638	1.4795
A	E	0.8356	0.7397	0.6575	0.7869	1.0638	1.2213
G	A	0.7671	0.8356	0.6712	0.8750	1.0471	1.3151
A	G	0.8356	0.7671	0.6712	0.8033	1.0471	1.1838
B	D	0.6301	0.8219	0.5068	0.8043	0.9786	0.9102
D	B	0.8219	0.6301	0.5068	0.6167	0.9786	0.9649
E	B	0.7397	0.6301	0.5479	0.7407	1.1755	1.4266
B	E	0.6301	0.7397	0.5479	0.8696	1.1755	1.9954
C	D	0.5479	0.8219	0.5068	0.9250	1.1254	2.3744
D	C	0.8219	0.5479	0.5068	0.6167	1.1254	1.1793
G	C	0.7671	0.5479	0.5068	0.6607	1.2058	1.3324
C	G	0.5479	0.7671	0.5068	0.9250	1.2058	3.1050

E	D	0.7397	0.8219	0.6575	0.8889	1.0815	1.6027
D	E	0.8219	0.7397	0.6575	0.8000	1.0815	1.3014
F	D	0.5068	0.8219	0.4658	0.9189	1.1180	2.1963
D	F	0.8219	0.5068	0.4658	0.5667	1.1180	1.1380
G	D	0.7671	0.8219	0.6712	0.8750	1.0646	1.4247
D	G	0.8219	0.7671	0.6712	0.8167	1.0646	1.2702
G	E	0.7671	0.7397	0.5479	0.7143	0.9656	0.9110
E	G	0.7397	0.7671	0.5479	0.7407	0.9656	0.8982
G	F	0.7671	0.5068	0.4658	0.6071	1.1979	1.2553
F	G	0.5068	0.7671	0.4658	0.9189	1.1979	2.8721
A, B	D	0.5479	0.8219	0.4521	0.8250	1.0038	1.0176
A, D	B	0.7123	0.6301	0.4521	0.6346	1.0071	1.0123
B, D	A	0.5068	0.8356	0.4521	0.8919	1.0673	1.5205
A	B, D	0.8356	0.5068	0.4521	0.5410	1.0673	1.0744
B	A, D	0.6301	0.7123	0.4521	0.7174	1.0071	1.0179
D	A, B	0.8219	0.5479	0.4521	0.5500	1.0038	1.0046
E, A	B	0.6575	0.6301	0.4658	0.7083	1.1241	1.2681
E, B	A	0.5479	0.8356	0.4658	0.8500	1.0172	1.0959
A, B	E	0.5479	0.7397	0.4658	0.8500	1.1491	1.7352
E	A, B	0.7397	0.5479	0.4658	0.6296	1.1491	1.2205
A	E, B	0.8356	0.5479	0.4658	0.5574	1.0172	1.0213
B	E, A	0.6301	0.6575	0.4658	0.7391	1.1241	1.3128
E, A	D	0.6575	0.8219	0.6027	0.9167	1.1153	2.1370
E, D	A	0.6575	0.8356	0.6027	0.9167	1.0970	1.9726
A, D	E	0.7123	0.7397	0.6027	0.8462	1.1439	1.6918
E	A, D	0.7397	0.7123	0.6027	0.8148	1.1439	1.5534
A	E, D	0.8356	0.6575	0.6027	0.7213	1.0970	1.2288
D	E, A	0.8219	0.6575	0.6027	0.7333	1.1153	1.2842
G, A	D	0.6712	0.8219	0.6027	0.8980	1.0925	1.7452
G, D	A	0.6712	0.8356	0.6027	0.8980	1.0746	1.6110
A, D	G	0.7123	0.7671	0.6027	0.8462	1.1030	1.5137
G	A, D	0.7671	0.7123	0.6027	0.7857	1.1030	1.3425
A	G, D	0.8356	0.6712	0.6027	0.7213	1.0746	1.1797
D	G, A	0.8219	0.6712	0.6027	0.7333	1.0925	1.2329
G, E	A	0.5479	0.8356	0.5342	0.9750	1.1668	6.5753
G, A	E	0.6712	0.7397	0.5342	0.7959	1.0760	1.2753
E, A	G	0.6575	0.7671	0.5342	0.8125	1.0592	1.2420
G	E, A	0.7671	0.6575	0.5342	0.6964	1.0592	1.1281
E	G, A	0.7397	0.6712	0.5342	0.7222	1.0760	1.1836
A	G, E	0.8356	0.5479	0.5342	0.6393	1.1668	1.2534
E, B	D	0.5479	0.8219	0.4795	0.8750	1.0646	1.4247
E, D	B	0.6575	0.6301	0.4795	0.7292	1.1572	1.3656
B, D	E	0.5068	0.7397	0.4795	0.9459	1.2788	4.8151
E	B, D	0.7397	0.5068	0.4795	0.6481	1.2788	1.4016
B	E, D	0.6301	0.6575	0.4795	0.7609	1.1572	1.4321
D	E, B	0.8219	0.5479	0.4795	0.5833	1.0646	1.0849

G, C	D	0.5068	0.8219	0.4658	0.9189	1.1180	2.1963
G, D	C	0.6712	0.5479	0.4658	0.6939	1.2663	1.4767
C, D	G	0.5068	0.7671	0.4658	0.9189	1.1979	2.8721
G	C, D	0.7671	0.5068	0.4658	0.6071	1.1979	1.2553
C	G, D	0.5479	0.6712	0.4658	0.8500	1.2663	2.1918
D	G, C	0.8219	0.5068	0.4658	0.5667	1.1180	1.1380
G, E	D	0.5479	0.8219	0.5342	0.9750	1.1863	7.1233
G, D	E	0.6712	0.7397	0.5342	0.7959	1.0760	1.2753
E, D	G	0.6575	0.7671	0.5342	0.8125	1.0592	1.2420
G	E, D	0.7671	0.6575	0.5342	0.6964	1.0592	1.1281
E	G, D	0.7397	0.6712	0.5342	0.7222	1.0760	1.1836
D	G, E	0.8219	0.5479	0.5342	0.6500	1.1863	1.2916
G, E, A	D	0.5342	0.8219	0.5205	0.9744	1.1855	6.9452
G, E, D	A	0.5342	0.8356	0.5205	0.9744	1.1660	6.4110
G, A, D	E	0.6027	0.7397	0.5205	0.8636	1.1675	1.9087
E, A, D	G	0.6027	0.7671	0.5205	0.8636	1.1258	1.7078
G, E	A, D	0.5479	0.7123	0.5205	0.9500	1.3337	5.7534
G, A	E, D	0.6712	0.6575	0.5205	0.7755	1.1794	1.5255
G, D	E, A	0.6712	0.6575	0.5205	0.7755	1.1794	1.5255
E, A	G, D	0.6575	0.6712	0.5205	0.7917	1.1794	1.5781
E, D	G, A	0.6575	0.6712	0.5205	0.7917	1.1794	1.5781
A, D	G, E	0.7123	0.5479	0.5205	0.7308	1.3337	1.6791
G	E, A, D	0.7671	0.6027	0.5205	0.6786	1.1258	1.2359
E	G, A, D	0.7397	0.6027	0.5205	0.7037	1.1675	1.3408
A	G, E, D	0.8356	0.5342	0.5205	0.6230	1.1660	1.2353
D	G, E, A	0.8219	0.5342	0.5205	0.6333	1.1855	1.2702

Screenshot of the actual output of the code.

```

Frequent Itemsets:
support  itemsets
0 0.835616 (A)
1 0.630137 (B)
2 0.547945 (C)
3 0.821918 (D)
4 0.739726 (E)
5 0.506849 (F)
6 0.520548 (F#)
7 0.767123 (G)
8 0.547945 (B, A)
9 0.479452 (A, C)
10 0.712329 (D, A)
11 0.657534 (A, E)
12 0.671233 (A, G)
13 0.506849 (B, D)
14 0.547945 (B, E)
15 0.506849 (D, C)
16 0.506849 (G, C)
17 0.657534 (D, E)
18 0.465753 (D, F)
19 0.671233 (D, G)
20 0.547945 (G, E)
21 0.465753 (G, F)
22 0.452055 (B, A, D)
23 0.465753 (B, A, E)
24 0.602740 (D, A, E)
25 0.602740 (D, A, G)
26 0.534247 (A, G, E)
27 0.479452 (B, E, D)
28 0.465753 (D, G, C)
29 0.534247 (D, G, E)
30 0.520548 (D, A, G, E)

Total Number of Frequent Itemsets: 31

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Figure 1: Frequent Itemsets

Association Rules:									
antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zhangs metric
1 (A)	(B)	0.835616	0.547945	0.835728	1.044827	0.821392	0.874364	0.237588	0.237588
1 (B)	(A)	0.835616	0.547945	0.869565	1.044627	0.821392	1.268274	0.185556	0.185556
2 (C)	(D)	0.835616	0.797926	0.879099	1.047213	0.821508	0.235165	0.899597	0.899597
3 (A)	(C)	0.835616	0.547945	0.572778	1.047131	0.821508	1.068598	0.273818	0.273818
4 (A)	(D)	0.835616	0.821918	0.712329	0.852459	1.037158	0.825521	1.287982	0.217949
5 (A)	(B)	0.835616	0.835616	0.712329	0.866467	1.037158	0.825521	1.232377	0.211183
6 (A)	(C)	0.835616	0.797926	0.657534	0.788885	1.063752	0.839487	1.221286	0.364583
7 (E)	(D)	0.739726	0.835616	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
8 (E)	(B)	0.835616	0.797926	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
9 (G)	(A)	0.787123	0.835616	0.671233	0.875000	1.047131	0.838212	1.315868	0.193277
10 (A)	(B)	0.835616	0.835616	0.978621	0.978621	0.811971	0.964868	0.289239	0.289239
11 (B)	(D)	0.821918	0.586849	0.884348	0.978621	0.811971	0.918198	0.855766	0.855766
12 (E)	(B)	0.739726	0.638137	0.547945	0.748741	1.375523	0.881816	1.426434	0.573884
13 (A)	(B)	0.835616	0.797926	0.547945	0.869565	1.375523	0.881816	1.995434	0.487745
14 (C)	(D)	0.547945	0.821918	0.586849	0.925888	1.235417	0.856483	2.374429	0.246519
15 (A)	(B)	0.547945	0.586849	0.925888	0.925888	1.235417	0.856483	2.374429	0.246519
16 (C)	(D)	0.547945	0.797926	0.586849	0.925888	1.235417	0.856483	2.374429	0.246519
17 (G)	(C)	0.787123	0.547945	0.586849	0.668714	1.285884	0.885588	1.332372	0.732399
18 (E)	(B)	0.739726	0.821918	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
19 (D)	(E)	0.821918	0.797926	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
20 (E)	(B)	0.821918	0.797926	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
21 (F)	(D)	0.586849	0.821918	0.465753	0.918919	1.118818	0.849165	1.388178	0.214852
22 (G)	(A)	0.787123	0.821918	0.671233	0.875000	1.047131	0.838212	1.315868	0.193277
23 (D)	(E)	0.821918	0.797926	0.657534	0.888889	1.063752	0.839487	1.479452	0.228263
24 (E)	(B)	0.739726	0.787123	0.547945	0.748741	1.375523	0.881816	1.426434	0.573884
25 (D)	(E)	0.787123	0.739726	0.547945	0.748741	1.375523	0.881816	1.426434	0.573884
26 (F)	(G)	0.787123	0.739726	0.547945	0.748741	1.375523	0.881816	1.426434	0.573884
27 (F)	(G)	0.787123	0.739726	0.547945	0.748741	1.375523	0.881816	1.426434	0.573884
28 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
29 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
30 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
31 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
32 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
33 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
34 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
35 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
36 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
37 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
38 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
39 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
40 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
41 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
42 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
43 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
44 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
45 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
46 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
47 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
48 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
49 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
50 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
51 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
52 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
53 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
54 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
55 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
56 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
57 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
58 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
59 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
60 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
61 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
62 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
63 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
64 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
65 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
66 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
67 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
68 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
69 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
70 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
71 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
72 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
73 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
74 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
75 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
76 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
77 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
78 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
79 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
80 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
81 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
82 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
83 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
84 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
85 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
86 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
87 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
88 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616
89 (A, B)	(D, E)	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616	0.835616

Figure 2: Actual output of the generate associated rule

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from mlxtend.frequent_patterns import apriori, association_rules
from mlxtend.preprocessing import TransactionEncoder
data_sets = pd.read_csv("output_root_notes.csv",header=None, dtype=str)
custom_transactions = data_sets.apply(lambda x: x.dropna().tolist(), axis=1).tolist()
te = TransactionEncoder()
te_ary = te.fit(custom_transactions).transform(custom_transactions)
df = pd.DataFrame(te_ary, columns=te.columns_)
# Apply the apriori algorithm to find frequent itemsets with a minimum support of 60%
frequent_itemsets = apriori(df, min_support=0.44, use_colnames=True)
print("\nFrequent Itemsets: \n", frequent_itemsets)
print("\n Total Number of Frequent Itemsets:" , len(frequent_itemsets))
# Generate association rules with a minimum confidence of 50%
rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.5)
# Sort the rules based on the conviction in ascending order
# rules_sorted_by_conviction = rules.sort_values(by='conviction', ascending=True)
print("\nAssociation Rules:\n", rules.to_string())

```

The Program used to generate the association rule.

**Manual Version** (Apriori Algorithm). Sorted in Ascending according to Support.

LHS	RHS	support	confidence	lift	conviction
A, B	D	0.4521	0.8250	1.0038	1.0176
A, D	B	0.4521	0.6346	1.0071	1.0123
B, D	A	0.4521	0.8919	1.0673	1.5205
A	B, D	0.4521	0.5410	1.0673	1.0744
B	A, D	0.4521	0.7174	1.0071	1.0179
D	A, B	0.4521	0.5500	1.0038	1.0046
F	D	0.4658	0.9189	1.1180	2.1963
D	F	0.4658	0.5667	1.1180	1.1380
G	F	0.4658	0.6071	1.1979	1.2553
F	G	0.4658	0.9189	1.1979	2.8721
E, A	B	0.4658	0.7083	1.1241	1.2681
E, B	A	0.4658	0.8500	1.0172	1.0959
A, B	E	0.4658	0.8500	1.1491	1.7352
E	A, B	0.4658	0.6296	1.1491	1.2205
A	E, B	0.4658	0.5574	1.0172	1.0213
B	E, A	0.4658	0.7391	1.1241	1.3128
G, C	D	0.4658	0.9189	1.1180	2.1963
G, D	C	0.4658	0.6939	1.2663	1.4767
C, D	G	0.4658	0.9189	1.1979	2.8721
G	C, D	0.4658	0.6071	1.1979	1.2553
C	G, D	0.4658	0.8500	1.2663	2.1918
D	G, C	0.4658	0.5667	1.1180	1.1380
C	A	0.4795	0.8750	1.0471	1.3151
A	C	0.4795	0.5738	1.0471	1.0606
E, B	D	0.4795	0.8750	1.0646	1.4247
E, D	B	0.4795	0.7292	1.1572	1.3656
B, D	E	0.4795	0.9459	1.2788	4.8151
E	B, D	0.4795	0.6481	1.2788	1.4016
B	E, D	0.4795	0.7609	1.1572	1.4321
D	E, B	0.4795	0.5833	1.0646	1.0849
B	D	0.5068	0.8043	0.9786	0.9102
D	B	0.5068	0.6167	0.9786	0.9649
C	D	0.5068	0.9250	1.1254	2.3744
D	C	0.5068	0.6167	1.1254	1.1793
G	C	0.5068	0.6607	1.2058	1.3324
C	G	0.5068	0.9250	1.2058	3.1050
G, E, A	D	0.5205	0.9744	1.1855	6.9452
G, E, D	A	0.5205	0.9744	1.1660	6.4110
G, A, D	E	0.5205	0.8636	1.1675	1.9087
E, A, D	G	0.5205	0.8636	1.1258	1.7078
G, E	A, D	0.5205	0.9500	1.3337	5.7534
G, A	E, D	0.5205	0.7755	1.1794	1.5255
G, D	E, A	0.5205	0.7755	1.1794	1.5255
E, A	G, D	0.5205	0.7917	1.1794	1.5781

E, D	G, A	0.5205	0.7917	1.1794	1.5781
A, D	G, E	0.5205	0.7308	1.3337	1.6791
G	E, A, D	0.5205	0.6786	1.1258	1.2359
E	G, A, D	0.5205	0.7037	1.1675	1.3408
A	G, E, D	0.5205	0.6230	1.1660	1.2353
D	G, E, A	0.5205	0.6333	1.1855	1.2702
G, E	A	0.5342	0.9750	1.1668	6.5753
G, A	E	0.5342	0.7959	1.0760	1.2753
E, A	G	0.5342	0.8125	1.0592	1.2420
G	E, A	0.5342	0.6964	1.0592	1.1281
E	G, A	0.5342	0.7222	1.0760	1.1836
A	G, E	0.5342	0.6393	1.1668	1.2534
G, E	D	0.5342	0.9750	1.1863	7.1233
G, D	E	0.5342	0.7959	1.0760	1.2753
E, D	G	0.5342	0.8125	1.0592	1.2420
G	E, D	0.5342	0.6964	1.0592	1.1281
E	G, D	0.5342	0.7222	1.0760	1.1836
D	G, E	0.5342	0.6500	1.1863	1.2916
A	B	0.5479	0.6557	1.0406	1.0744
B	A	0.5479	0.8696	1.0406	1.2603
E	B	0.5479	0.7407	1.1755	1.4266
B	E	0.5479	0.8696	1.1755	1.9954
G	E	0.5479	0.7143	0.9656	0.9110
E	G	0.5479	0.7407	0.9656	0.8982
E, A	D	0.6027	0.9167	1.1153	2.1370
E, D	A	0.6027	0.9167	1.0970	1.9726
A, D	E	0.6027	0.8462	1.1439	1.6918
E	A, D	0.6027	0.8148	1.1439	1.5534
A	E, D	0.6027	0.7213	1.0970	1.2288
D	E, A	0.6027	0.7333	1.1153	1.2842
G, A	D	0.6027	0.8980	1.0925	1.7452
G, D	A	0.6027	0.8980	1.0746	1.6110
A, D	G	0.6027	0.8462	1.1030	1.5137
G	A, D	0.6027	0.7857	1.1030	1.3425
A	G, D	0.6027	0.7213	1.0746	1.1797
D	G, A	0.6027	0.7333	1.0925	1.2329
E	A	0.6575	0.8889	1.0638	1.4795
A	E	0.6575	0.7869	1.0638	1.2213
E	D	0.6575	0.8889	1.0815	1.6027
D	E	0.6575	0.8000	1.0815	1.3014
G	A	0.6712	0.8750	1.0471	1.3151
A	G	0.6712	0.8033	1.0471	1.1838
G	D	0.6712	0.8750	1.0646	1.4247
D	G	0.6712	0.8167	1.0646	1.2702
A	D	0.7123	0.8525	1.0372	1.2070
D	A	0.7123	0.8667	1.0372	1.2329



## Visualization

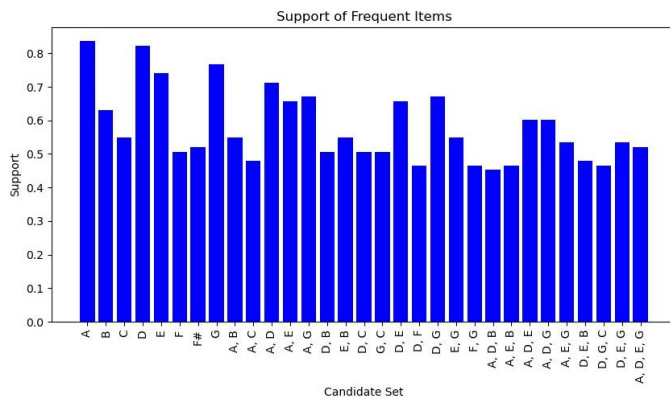


Figure 1: Support of Frequent Items

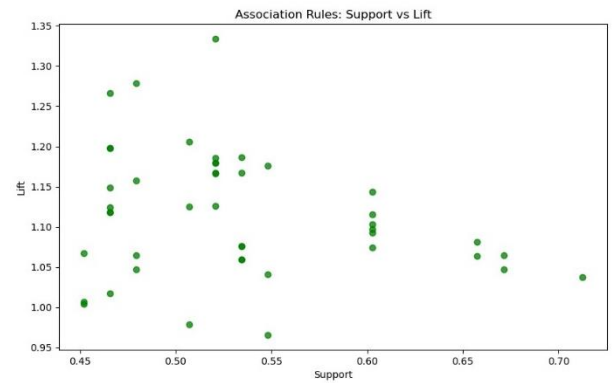


Figure 2: Support vs Lift

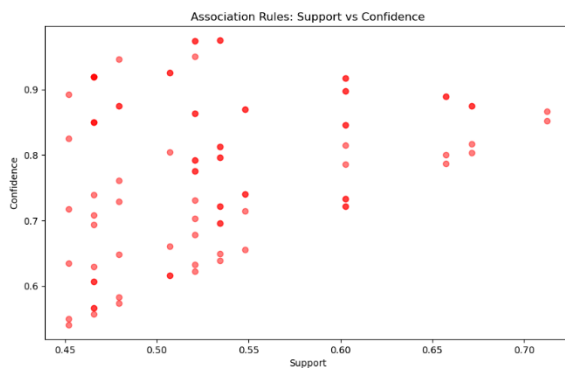


Figure 3: Support vs Confidence

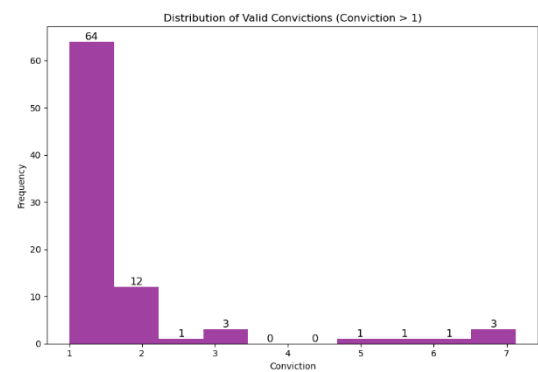


Figure 4: Distribution of Valid Convictions

## Market Basket Analysis Result Interpretation

Compositions	A	B	C	D	E	F	F#	G
1	1	1	0	1	1	0	1	0
2	0	1	0	1	1	0	1	0
3	1	1	1	1	1	1	1	1
4	1	0	0	1	1	1	0	1
5	1	0	0	1	0	1	0	1

...

69	1	1	1	1	1	1	1	1
70	1	0	1	1	0	1	0	1
71	1	1	0	1	0	0	1	0
72	1	0	0	1	1	1	1	1
73	1	1	0	1	1	1	1	1

LHS	RHS	FREQUENCY	SUPPORT	CONFIDENCE	LIFT
G, C	D	34	0.4658	0.9189	1.1180
Regel		Confidence			
LHS → RHS		$\frac{\text{frq}(\text{LHS, RHS})}{\text{frq}(\text{LHS})}$			
G,C → D		$\frac{\text{frq}(\text{G,C, D})}{\text{frq}(\text{G,C})} = 0.9189$			
Frequency		Lift			
$\frac{\text{frq}(\text{LHS, RHS})}{\text{frq}(\text{G,C} \rightarrow \text{D})} = 34$		$\frac{\text{Confidence}(\text{LHS, RHS})}{\text{Support}(\text{RHS})} = 0.9189 / 0.8219 = 1.1180$			
Support					
$\frac{\text{Frq}(\text{LHS, RHS})}{N} = 34/73 = 0.4658$					

