



Progressive Education Society's
Modern College of Engineering, Shivajinagar, Pune-05.
(An Autonomous Institute Affiliated to Savitribai Phule Pune University)
Department of MCA

PRACTICAL SUBMISSION RECORD- A.Y. 2025-26

Class: SYMCA Division : A Semester: III	Course Code: MCA01604 Course Name: Data Science Laboratory	Batch: S2
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CO No: CO605.2	Assignment No: 10	

Title : Use the FP-Growth Algorithm to perform Market Basket Analysis on a retail dataset. You will extract frequent item sets and generate association rules to discover purchasing patterns and help improve sales strategies.

Code:

```
# FP-Growth Market Basket Analysis - full R script
# Requires: arules, rCBA (for fpgrowth), arulesViz (optional for plotting)
# Two usage modes: example with built-in Groceries, or custom CSV (TransactionID, Item)
```

```
# -----
# 1) Install / load packages
# -----
packages <- c("arules", "arulesViz", "rCBA")
to_install <- packages[!(packages %in% installed.packages()[, "Package"])]
if (length(to_install) > 0) {
  install.packages(to_install)
}
library(arules)
# arulesViz optional
suppressWarnings(suppressMessages(require(arulesViz)))
suppressWarnings(suppressMessages(require(rCBA)))

# -----
# 2) Parameters (change as needed)
# -----
min_support <- 0.001 # e.g. 0.1% of transactions
min_confidence <- 0.8 # 80%
max_len <- 5 # maximum itemset length for mining

# -----
# 3) Load data (two options)
```



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OPTION A: Use built-in Groceries dataset (quick demo)

```
use_builtin <- TRUE
```

```
if (use_builtin) {
```

```
  data("Groceries")    # from arules
```

```
  txns <- Groceries     # already a transactions object
```

```
  message("Using built-in Groceries dataset (", length(txns), " transactions).")
```

```
} else {
```

```
  # OPTION B: Load custom CSV: file must have columns TransactionID and Item
```

```
  # Example CSV format:
```

```
  # TransactionID,Item
```

```
  # 1,milk
```

```
  # 1,bread
```

```
  # 2,beer
```

```
  # 2,diapers
```

```
  csv_file <- "retail_transactions.csv" # change path as needed
```

```
  if (!file.exists(csv_file)) stop("CSV file not found: ", csv_file)
```

```
  raw <- read.csv(csv_file, stringsAsFactors = FALSE)
```

```
  if (!all(c("TransactionID", "Item") %in% colnames(raw))) {
```

```
    stop("CSV must contain at least columns named TransactionID and Item")
```

```
  }
```

```
  # convert to transactions
```

```
  split_list <- split(raw$Item, raw$TransactionID)
```

```
  txns <- as(split_list, "transactions")
```

```
  message("Loaded custom CSV; transactions:", length(txns))
```

```
}
```

Quick summary

```
summary(txns)
```

4) FP-Growth with rCBA

rCBA::fpgrowth accepts arules::transactions objects directly.

Note: rCBA's fpgrowth can optionally take a 'consequent' parameter if you

want to mine rules with a target class (for associative classification),

but for general association rules set consequent = NULL (default).

```
message("Running rCBA::fpgrowth (FP-Growth)...")
```

```
# rCBA::fpgrowth's 'confidence' param expects value in same scale as arules (0..1)
```



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If your dataset is very large / dense you may want to increase support.

Wrapping in tryCatch: some installations / configurations might require Java/rJava.

```
fpgrowth_result <- NULL
```

```
try({
```

```
  fpgrowth_result <- rCBA::fpgrowth(txns,  
    support = min_support,  
    confidence = min_confidence,  
    maxLength = max_len,  
    verbose = TRUE,  
    parallel = TRUE)
```

```
}, silent = TRUE)
```

```
# -----
```

```
# 5) Inspect FP-Growth result (rCBA)
```

```
# -----
```

```
if (!is.null(fpgrowth_result)) {  
  message("FP-Growth (rCBA) completed. Inspecting result...")  
  # Print basic object  
  print(fpgrowth_result)  
  # rCBA prints rules in its own format. We will try to convert to a readable data.frame.  
  # The rCBA rules object is not identical to arules::rules, but you can capture the printed output:  
  cat("\n--- Example rCBA FP-Growth rules (first 20 lines of print) ---\n")  
  capture.output(print(fpgrowth_result), file = stdout(), append = TRUE)[1:20] -> NULL
```

```
# If you want a tidy table, fall back to using arules::apriori on the same data
```

```
# (apriori often returns similar rules—this provides support/confidence/lift easily).
```

```
} else {
```

```
  message("rCBA::fpgrowth failed or is unavailable. Falling back to arules::apriori (Apriori).")
```

```
}
```

```
# -----
```

```
# 6) Fallback / Complement: Use arules::apriori to produce arules object
```

```
# -----
```

```
# (This is useful so you get a rules object compatible with arules methods such as
```

```
# quality(), inspect(), plotting and saving.)
```

```
message("Running arules::apriori with same thresholds (as a fallback / complement)...")
```

```
apriori_rules <- apriori(txns,
```

```
  parameter = list(supp = min_support, conf = min_confidence, maxlen = max_len, target =  
  "rules"))
```

```
message("Number of rules found by apriori:", length(apriori_rules))
```

```
# Create a data.frame of rules with support/confidence/lift sorted by confidence
```



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```
if (length(apriori_rules) > 0) {
  apriori_rules_sorted <- sort(apriori_rules, by = "confidence", decreasing = TRUE)
  top_rules <- head(apriori_rules_sorted, n = 20) # show top 20 by default
  # Convert to data.frame for easier viewing/export
  rules_df <- DATAFRAME(top_rules, setStart = "", itemSep = ", ", separate = TRUE)
  # Ensure support/confidence/lift columns exist
  if (!("support" %in% colnames(rules_df))) rules_df$support <- quality(top_rules)$support
  if (!("confidence" %in% colnames(rules_df))) rules_df$confidence <- quality(top_rules)$confidence
  if (!("lift" %in% colnames(rules_df))) rules_df$lift <- quality(top_rules)$lift

  message("\nTop rules (by confidence) — first 20:")
  print(rules_df[, c("LHS", "RHS", "support", "confidence", "lift")])
} else {
  message("No rules found by apriori with the given thresholds.")
}

# -----
# 7) Save / export frequent itemsets (if desired)
# -----
# Get frequent itemsets (equivalent of FP-Growth's output)
freq_itemsets <- eclat(txns, parameter = list(supp = min_support, maxlen = max_len))
message("Frequent itemsets found (eclat):", length(freq_itemsets))
# Inspect top frequent itemsets by support
freq_df <- DATAFRAME(sort(freq_itemsets, by = "support", decreasing = TRUE))
if (nrow(freq_df) > 0) {
  cat("\nTop frequent itemsets (by support):\n")
  print(head(freq_df, 20))
}

# -----
# 8) Visualization (optional)
# -----
if (length(apriori_rules) > 0 && requireNamespace("arulesViz", quietly = TRUE)) {
  # Plot top 20 rules
  plot(apriori_rules_sorted[1:min(20, length(apriori_rules_sorted))],
       method = "graph", control = list(type = "items"), main = "Top Association Rules (by confidence)")
}

# -----
# 9) Interpreting results & Sales strategy ideas
# -----
# Print a short summary to guide business interpretation
```



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cat("\n--- Short business interpretation tips ---\n")

cat("1) Look for item pairs/triples with high support & high confidence (and lift > 1):\n")

cat(" - High support: items frequently bought (good candidates for cross-sell features)\n")

cat(" - High confidence: when A is bought, B is likely to be bought (good for product bundling)\n")

cat(" - High lift (>1): stronger-than-random association (prioritize these)\n\n")

cat("2) Actions to consider:\n")

cat(" - Bundle top pairs into promotions or combo discounts.\n")

cat(" - Place frequently co-purchased items close in store layout or recommend online.\n")

cat(" - Use rules for personalized recommendations and targeted promotions.\n\n")

cat("3) Next steps:\n")

cat(" - Validate candidate rules on a held-out period (time-slice) to avoid seasonal bias.\n")

cat(" - Use POS / time-of-day segmentation to find context-specific patterns.\n")

cat(" - A/B test bundles or recommendations based on rule-driven suggestions.\n\n")

10) Save results to CSV (optional)

write top rules to CSV for reporting

if (exists("rules_df") && nrow(rules_df) > 0) {

write.csv(rules_df, file = "top_rules_by_confidence.csv", row.names = FALSE)

message("Top rules exported to top_rules_by_confidence.csv")

}

message("Done.")

Output :



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File History Resize

