



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

