An Introduction to **PGRdup** package

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Contents

Introduction

PGRdup is an `R` package to facilitate the search for probable/possible duplicate accessions in Plant Genetic Resources (PGR) collections using passport databases. Primarily this package implements a workflow (Fig. 1) designed to fetch groups or sets of germplasm accessions with similar passport data particularly in fields associated with accession names within or across PGR passport databases. It offers a suite of functions for data pre-processing, creation of a searchable Key Word in Context (KWIC) index of keywords associated with accession records and the identification of probable duplicate sets by fuzzy, phonetic and semantic matching of keywords. It also has functions to enable the user to review, modify and validate the probable duplicate sets retrieved.

The goal of this document is to introduce the users to these functions and familiarise them with the workflow intended to fetch probable duplicate sets. This document assumes a basic knowledge of `R` programming language.

The functions in this package are primarily built using the `R` packages `data.table`, `igraph`, `stringdist` and `stringi`.

PGRdup | *Discover Probable Duplicates in Plant Genetic Resources Collections*

Figure 1: PGRdup

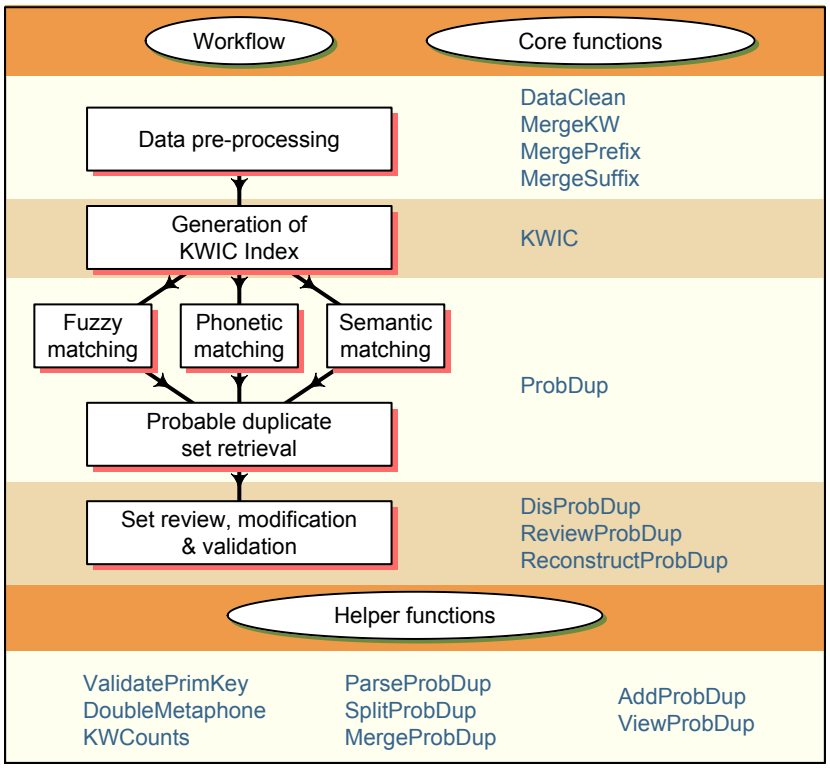

Fig. 1. PGRdup workflow and associated functions

Installation

The package can be installed using the following function:

```
install.packages('PGRdup', dependencies=TRUE)
```

Uninstalled dependencies (packages which `PGRdup` depends on *viz*- `data.table`, `igraph`, `stringdist` and `stringi`) are also installed because of the argument `dependencies=TRUE`.

Then the package can be loaded using the function

```
library(PGRdup)
```

Data Format

The package is essentially designed to operate on PGR passport data present in a data frame object, with each row holding one record and columns representing the attribute fields. For example, consider the dataset `GN1000` supplied along with the package.

```
library(PGRdup)
# Load the dataset to the environment
data(GN1000)
# Show the class of the object
class(GN1000)
```

```
## [1] "data.frame"
```

```
# View the first few records in the data frame
head(GN1000)
```

```
##   CommonName    BotanicalName NationalID                CollNo   DonorID
## 1  Groundnut Arachis hypogaea   EC100277 Shulamith/ NRCG-14555  ICG-4709
## 2  Groundnut Arachis hypogaea   EC100280                    NC   ICG5288
## 3  Groundnut Arachis hypogaea   EC100281               MALIMBA   ICG5289
## 4  Groundnut Arachis hypogaea   EC100713            EC 100713;   ICG5296
## 5  Groundnut Arachis hypogaea   EC100715             EC 100715   ICG5298
## 6  Groundnut Arachis hypogaea   EC100716                        ICG-3150
##   OtherID1  OtherID2 BioStatus            SourceCountry TransferYear
## 1           U4-47-12  Landrace                   Israel         2014
## 2      NCS      NC 5  Landrace United States of America         2004
## 3          EC 100281  Landrace                   Malawi         2004
## 4              STARR  Landrace United States of America         2004
## 5              COMET  Landrace United States of America         2004
## 6          ARGENTINE  Landrace United States of America         2014
```

If the passport data exists as an excel sheet, it can be first converted to a csv or tab delimited format and then easily loaded to the `R` environment using the functions `read.csv` and `read.table` respectively. Alternatively, the package `readxl` can be used to directly read the data from excel. In case of large csv files, the function `fread` in the `data.table` package can be used to rapidly load the data.

If the PGR passport data is in a DBMS, the required table can be imported as a data frame into `R`. using the appropriate R-database interface package. For example `dbConnect` for MySQL, `ROracle` for Oracle etc.

The PGR data downloaded from genesys database as a Darwin Core - Germplasm zip archive can be imported into the `R` environment as a flat file `data.frame` using the `read.genesys` function.

```
# Import the DwC-Germplasm zip archive "genesys-accessions-filtered.zip"
PGRgenesys <- read.genesys("genesys-accessions-filtered.zip",
                           scrub.names.space = TRUE, readme = TRUE)
```

Data Pre-processing

Data pre-processing is a critical step which can affect the quality of the probable duplicate sets being retrieved. It involves data standardization as well as data cleaning which can be achieved using the functions `DataClean`, `MergeKW`, `MergePrefix` and `MergeSuffix`.

`DataClean` function can be used to clean the character strings in passport data fields(columns) specified as the input character vector `x` according to the conditions specified in the arguments.

Commas, semicolons and colons which are sometimes used to separate multiple strings or names within the same field can be replaced with a single space using the logical arguments `fix.comma`, `fix.semcol` and `fix.col` respectively.

```
x <- c("A 14; EC 1697", "U 4-4-28; EC 21078; A 32", "PI 262801:CIAT 9075:GKP 9553/90",
       "NCAC 16049, PI 261987, RCM 493-3")
x
```

```
## [1] "A 14; EC 1697"                    "U 4-4-28; EC 21078; A 32"
## [3] "PI 262801:CIAT 9075:GKP 9553/90"  "NCAC 16049, PI 261987, RCM 493-3"
```

```
# Replace ',', ':' and ';' with space
DataClean(x, fix.comma=TRUE, fix.semcol=TRUE, fix.col=TRUE,
          fix.bracket=FALSE, fix.punct=FALSE, fix.space=FALSE, fix.sep=FALSE,
          fix.leadzero=FALSE)
```

```
## [1] "A 14  EC 1697"                    "U 4-4-28  EC 21078  A 32"
## [3] "PI 262801 CIAT 9075 GKP 9553/90"  "NCAC 16049  PI 261987  RCM 493-3"
```

Similarly the logical argument `fix.bracket` can be used to replace all brackets including parenthesis, square brackets and curly brackets with space.

```
x <- c("(NRCG-1738)/(NFG649)", "26-5-1[NRCG-2528]", "Ah 1182 {NRCG-4340}")
x
```

```
## [1] "(NRCG-1738)/(NFG649)" "26-5-1[NRCG-2528]"    "Ah 1182 {NRCG-4340}"
```

```
# Replace parenthesis, square brackets and curly brackets with space
DataClean(x, fix.comma=FALSE, fix.semcol=FALSE, fix.col=FALSE,
          fix.bracket=TRUE,
          fix.punct=FALSE, fix.space=FALSE, fix.sep=FALSE, fix.leadzero=FALSE)
```

```
## [1] "NRCG-1738 / NFG649" "26-5-1 NRCG-2528"   "AH 1182  NRCG-4340"
```

The logical argument `fix.punct` can be used to remove all punctuation from the data.

```
x <- c("#26-6-3-1", "Culture No. 857", "U/4/47/13")
x
```

```
## [1] "#26-6-3-1"       "Culture No. 857" "U/4/47/13"
```

```
# Remove punctuation
DataClean(x, fix.comma=FALSE, fix.semcol=FALSE, fix.col=FALSE, fix.bracket=FALSE,
          fix.punct=TRUE,
          fix.space=FALSE, fix.sep=FALSE, fix.leadzero=FALSE)
```

```
## [1] "26631"          "CULTURE NO 857" "U44713"
```

`fix.space` can be used to convert all space characters such as tab, newline, vertical tab, form feed and carriage return to spaces and finally convert multiple spaces to single space.

```
x <- c("RS   1", "GKSPScGb 208  PI 475855")
x
```

```
## [1] "RS   1"                  "GKSPScGb 208  PI 475855"
```

```
# Replace all space characters to space and convert multiple spaces to single space
DataClean(x, fix.comma=FALSE, fix.semcol=FALSE, fix.col=FALSE,
          fix.bracket=FALSE, fix.punct=FALSE,
          fix.space=TRUE,
          fix.sep=FALSE, fix.leadzero=FALSE)
```

```
## [1] "RS 1"                   "GKSPSCGB 208 PI 475855"
```

`fix.sep` can be used to merge together accession identifiers composed of alphabetic characters separated from as series of digits by a space character.

```
x <- c("NCAC 18078", "AH 6481", "ICG 2791")
x
```

```
## [1] "NCAC 18078" "AH 6481"    "ICG 2791"
```

```
# Merge alphabetic character separated from a series of digits by a space
DataClean(x, fix.comma=FALSE, fix.semcol=FALSE, fix.col=FALSE,
          fix.bracket=FALSE, fix.punct=FALSE, fix.space=FALSE,
          fix.sep=TRUE,
          fix.leadzero=FALSE)
```

```
## [1] "NCAC18078" "AH6481"    "ICG2791"
```

`fix.leadzero` can be used to remove leading zeros from accession name fields to facilitate matching to identify probable duplicates.

```
x <- c("EC 0016664", "EC0001690")
x
```

```
## [1] "EC 0016664" "EC0001690"
```

```
# Remove leading zeros
DataClean(x, fix.comma=FALSE, fix.semcol=FALSE, fix.col=FALSE,
          fix.bracket=FALSE, fix.punct=FALSE, fix.space=FALSE, fix.sep=FALSE,
          fix.leadzero=TRUE)
```

```
## [1] "EC 16664" "EC1690"
```

This function can hence be made use of in tidying up multiple forms of messy data existing in fields associated with accession names in PGR passport databases (Table 1).

```
names <- c("S7-12-6", "ICG-3505", "U 4-47-18;EC 21127", "AH 6481", "RS   1",
           "AK 12-24", "2-5 (NRCG-4053)", "T78, Mwitunde", "ICG 3410",
           "#648-4 (Gwalior)", "TG4;U/4/47/13", "EC0021003")
names
```

```
##  [1] "S7-12-6"            "ICG-3505"           "U 4-47-18;EC 21127"
##  [4] "AH 6481"            "RS   1"             "AK 12-24"
##  [7] "2-5 (NRCG-4053)"    "T78, Mwitunde"      "ICG 3410"
## [10] "#648-4 (Gwalior)"   "TG4;U/4/47/13"      "EC0021003"
```

```
# Clean the data
DataClean(names)
```

```
##  [1] "S7126"          "ICG3505"        "U44718 EC21127" "AH6481"
##  [5] "RS1"            "AK1224"         "25 NRCG4053"    "T78 MWITUNDE"
##  [9] "ICG3410"        "6484 GWALIOR"   "TG4 U44713"     "EC21003"
```

Table 1. Data pre-processing using `DataClean`.

names	DataClean(names)
S7-12-6	S7126
ICG-3505	ICG3505
U 4-47-18;EC 21127	U44718 EC21127
AH 6481	AH6481
RS 1	RS1
AK 12-24	AK1224
2-5 (NRCG-4053)	25 NRCG4053
T78, Mwitunde	T78 MWITUNDE
ICG 3410	ICG3410
#648-4 (Gwalior)	6484 GWALIOR
TG4;U/4/47/13	TG4 U44713
EC0021003	EC21003

Several common keyword string pairs or keyword prefixes and suffixes exist in fields associated with accession names in PGR passport databases. They can be merged using the functions `MergeKW`, `MergePrefix` and `MergeSuffix` respectively. The keyword string pairs, prefixes and suffixes can be supplied as a list or a vector to the argument `y` in these functions.

```
names <- c("Punjab Bold", "Gujarat- Dwarf", "Nagpur.local", "SAM COL 144",
           "SAM COL--280", "NIZAMABAD-LOCAL", "Dark Green Mutant",
           "Dixie-Giant", "Georgia- Bunch", "Uganda-erect", "Small Japan",
           "Castle  Cary", "Punjab erect", "Improved small japan",
           "Dark Purple")
names
```

```
##  [1] "Punjab Bold"          "Gujarat- Dwarf"       "Nagpur.local"
##  [4] "SAM COL 144"          "SAM COL--280"         "NIZAMABAD-LOCAL"
##  [7] "Dark Green Mutant"    "Dixie-Giant"          "Georgia- Bunch"
## [10] "Uganda-erect"         "Small Japan"          "Castle  Cary"
## [13] "Punjab erect"         "Improved small japan" "Dark Purple"
```

```
# Merge pairs of strings
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
           c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
           c("Mota", "Company"))
names <- MergeKW(names, y1, delim = c("space", "dash", "period"))

# Merge prefix strings
y2 <- c("Light", "Small", "Improved", "Punjab", "SAM")
names <- MergePrefix(names, y2, delim = c("space", "dash", "period"))

# Merge suffix strings
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
names <- MergeSuffix(names, y3, delim = c("space", "dash", "period"))

names
```

```
##  [1] "PunjabBold"         "GujaratDwarf"       "Nagpurlocal"
##  [4] "SAMCOL 144"         "SAMCOL--280"        "NIZAMABADLOCAL"
##  [7] "Dark GreenMutant"   "DixieGiant"         "GeorgiaBunch"
## [10] "Ugandaerect"        "SmallJapan"         "CastleCary"
## [13] "Punjaberect"        "Improvedsmalljapan" "Dark Purple"
```

These functions can be applied over multiple columns(fields) in a data frame using the `lapply` function.

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")
head(GN[GNfields])
```

```
##   NationalID                CollNo   DonorID OtherID1  OtherID2
## 1   EC100277 Shulamith/ NRCG-14555  ICG-4709           U4-47-12
## 2   EC100280                    NC   ICG5288      NCS      NC 5
## 3   EC100281               MALIMBA   ICG5289          EC 100281
## 4   EC100713            EC 100713;   ICG5296              STARR
## 5   EC100715             EC 100715   ICG5298              COMET
## 6   EC100716                        ICG-3150          ARGENTINE
```

```
# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeSuffix(x, y3, delim = c("space", "dash")))
head(GN[GNfields])
```

```
##   NationalID              CollNo DonorID OtherID1  OtherID2
## 1   EC100277 SHULAMITH NRCG14555 ICG4709             U44712
## 2   EC100280                  NC ICG5288      NCS       NC5
## 3   EC100281             MALIMBA ICG5289           EC100281
## 4   EC100713            EC100713 ICG5296              STARR
## 5   EC100715            EC100715 ICG5298              COMET
## 6   EC100716                     ICG3150          ARGENTINE
```

Generation of KWIC Index

The function `KWIC` generates a Key Word in Context index (Knüpffer 1988; Knüpffer, Frese, and Jongen 1997) from the data frame of a PGR passport database based on the fields(columns) specified in the argument `fields` along with the keyword frequencies and gives the output as a list of class `KWIC`. The first element of the vector specified in `fields` is considered as the primary key or identifier which uniquely identifies all rows in the data frame.

This function fetches keywords from different fields specified, which can be subsequently used for matching to identify probable duplicates. The frequencies of the keywords retrieved can help in determining if further data pre-processing is required and also to decide whether any common keywords can be exempted from matching (Fig. 2).

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
```

```
                        function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                        function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate the KWIC index
GNKWIC <- KWIC(GN, GNfields, min.freq=1)
class(GNKWIC)
```

```
## [1] "KWIC"
```

```
GNKWIC
```

```
## KWIC fields : NationalID CollNo DonorID OtherID1 OtherID2
## Number of keywords : 3893
## Number of distinct keywords : 3109
```

```
# Retrieve the KWIC index from the KWIC object
KWIC <- GNKWIC[[1]]
KWIC <- KWIC[order(KWIC$KEYWORD, decreasing=TRUE),]
head(KWIC[,c("PRIM_ID", "KWIC_L", "KWIC_KW", "KWIC_R")], n=10)
```

```
##       PRIM_ID                                      KWIC_L  KWIC_KW
## 550  EC490380           EC490380 =  = ICG1122 =  = LIN      YUCH
## 435   EC36893                                 EC36893 =      YUAN
## 434   EC36893                               EC36893 = YUAN     YOUNG
## 1287 EC613524       EC613524 = NRCG9225 =  = PEI KANGPE    YOUDON
## 1703 IC113088                        IC113088 =  =  = SB        XI
## 1741 IC296965 IC296965 = SB X11 X V11 = ICG1769 =  = SB        XI
## 3385 IC445197                                IC445197 =   X144B28
## 3483 IC494754               IC494754 =  = ICG7686 =  =   X144B28
## 2090 IC304018    IC304018 = 144B19B NRCG = ICG1561 =  =  X144B19B
## 1735 IC296965                                IC296965 = SB       X11
##                                KWIC_R
## 550                              TSAO
## 435   YOUNG TOU = ICG5241 =  = EC36893
## 434         TOU = ICG5241 =  = EC36893
## 1287                                =
## 1703                       = IC305003
## 1741                            X VII
## 3385           B = ICG2113 =  = LIMDI4
## 3483                                B
## 2090
## 1735  X V11 = ICG1769 =  = SB XI X VII
```

```
# Retrieve the keyword frequencies from the KWIC object
KeywordFreq <- GNKWIC[[2]]
head(KeywordFreq)
```

```
##   Keyword Freq
## 1   OVERO   25
## 2      S1   19
## 3       A   11
## 4     RED   11
## 5    OVER   10
## 6  PURPLE   10
```

Fig. 2. Word cloud of keywords retrieved

The function will throw an error in case of duplicates or NULL values in the primary key/ID field mentioned.

```
GN <- GN1000
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
# Generate dummy duplicates for illustration
GN[1001:1005,] <- GN[1:5,]
# Generate dummy NULL values for illustration
GN[1001,3] <- ""
GN[1002,3] <- ""
GN[1001:1005,]
```

```
##      CommonName    BotanicalName NationalID              CollNo DonorID
## 1001  Groundnut Arachis hypogaea            SHULAMITH NRCG14555 ICG4709
## 1002  Groundnut Arachis hypogaea                           NC ICG5288
## 1003  Groundnut Arachis hypogaea   EC100281             MALIMBA ICG5289
## 1004  Groundnut Arachis hypogaea   EC100713            EC100713 ICG5296
## 1005  Groundnut Arachis hypogaea   EC100715            EC100715 ICG5298
##      OtherID1 OtherID2 BioStatus            SourceCountry TransferYear
## 1001            U44712  Landrace                   Israel         2014
## 1002      NCS      NC5  Landrace United States of America         2004
## 1003          EC100281  Landrace                   Malawi         2004
## 1004             STARR  Landrace United States of America         2004
## 1005             COMET  Landrace United States of America         2004
```

```
GNKWIC <- KWIC(GN, GNfields, min.freq=1)
```

```
## Error in KWIC(GN, GNfields, min.freq = 1) :
##   Primary key/ID field should be unique and not NULL
##  Use PGRdup::ValidatePrimKey() to identify and rectify the aberrant records first
```

The erroneous records can be identified using the helper function `ValidatePrimKey`.

```
# Validate the primary key/ID field for duplication or existence of NULL values
ValidatePrimKey(x=GN, prim.key="NationalID")
```

```
## $message1
## [1] "ERROR: Duplicated records found in prim.key field"
##
## $Duplicates
##      CommonName    BotanicalName NationalID              CollNo DonorID
## 1001  Groundnut Arachis hypogaea            SHULAMITH NRCG14555 ICG4709
## 1002  Groundnut Arachis hypogaea                           NC ICG5288
## 3     Groundnut Arachis hypogaea   EC100281             MALIMBA ICG5289
## 1003  Groundnut Arachis hypogaea   EC100281             MALIMBA ICG5289
## 4     Groundnut Arachis hypogaea   EC100713            EC100713 ICG5296
## 1004  Groundnut Arachis hypogaea   EC100713            EC100713 ICG5296
## 5     Groundnut Arachis hypogaea   EC100715            EC100715 ICG5298
## 1005  Groundnut Arachis hypogaea   EC100715            EC100715 ICG5298
##      OtherID1 OtherID2 BioStatus            SourceCountry TransferYear
## 1001            U44712  Landrace                    Israel         2014
## 1002      NCS      NC5  Landrace United States of America         2004
## 3             EC100281  Landrace                   Malawi         2004
## 1003          EC100281  Landrace                   Malawi         2004
## 4                STARR  Landrace United States of America         2004
## 1004             STARR  Landrace United States of America         2004
## 5                COMET  Landrace United States of America         2004
## 1005             COMET  Landrace United States of America         2004
##
## $message2
## [1] "ERROR: NULL records found in prim.key field"
##
## $NullRecords
##      CommonName    BotanicalName NationalID              CollNo DonorID
## 1001  Groundnut Arachis hypogaea            SHULAMITH NRCG14555 ICG4709
## 1002  Groundnut Arachis hypogaea                           NC ICG5288
##      OtherID1 OtherID2 BioStatus            SourceCountry TransferYear
## 1001            U44712  Landrace                    Israel         2014
## 1002      NCS      NC5  Landrace United States of America         2004
##      primdup
## 1001    TRUE
## 1002    TRUE
```

```
# Remove the offending records
GN <- GN[-c(1001:1005), ]
# Validate again
ValidatePrimKey(x=GN, prim.key="NationalID")
```

```
## $message1
## [1] "OK: No duplicated records found in prim.key field"
##
## $Duplicates
## NULL
##
## $message2
## [1] "OK: No NULL records found in prim.key field"
```

```
##
## $NullRecords
## NULL
```

Retrieval of Probable Duplicate Sets

Once KWIC indexes are generated, probable duplicates of germplasm accessions can be identified by fuzzy, phonetic and semantic matching of the associated keywords using the function `ProbDup`. The sets are retrieved as a list of data frames of class `ProbDup`.

Keywords that are not to be used for matching can be specified as a vector in the `excep` argument.

Methods

The function can execute matching according to either one of the following three methods as specified by the `method` argument.

1. **Method `"a"`** : Performs string matching of keywords in a single KWIC index to identify probable duplicates of accessions in a single PGR passport database.

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate the KWIC index
GNKWIC <- KWIC(GN, GNfields)
```

```
# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
          "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
          "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
          "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
          "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
          "U", "VALENCIA", "VIRGINIA", "WHITE")

# Fetch fuzzy duplicates by method 'a'
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep, fuzzy = TRUE,
                 phonetic = FALSE, semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
  |
  |================================================================| 100%
Block 4 / 4 |
```

```
class(GNdup)
```

```
## [1] "ProbDup"
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                  No..of.Sets    No..of.Records
## FuzzyDuplicates          378               745
## Total                    378 745(Distinct:745)
```

```
# Fetch phonetic duplicates by method 'a'
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep, fuzzy = FALSE,
                 phonetic = TRUE, semantic = FALSE)
```

```
## Phonetic matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
```

```
Block 3 / 4 |
  |
  |======================================================================| 100%
Block 4 / 4 |
```

```
class(GNdup)
```

```
## [1] "ProbDup"
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## PhoneticDuplicates          99               260
## Total                       99 260(Distinct:260)
```

2. **Method "b"** : Performs string matching of keywords in the first KWIC index (query) with that of the keywords in the second index (source) to identify probable duplicates of accessions of the first PGR passport database among the accessions in the second database.

3. **Method "c"** : Performs string matching of keywords in two different KWIC indexes jointly to identify probable duplicates of accessions from among two PGR passport databases.

```
# Load PGR passport databases
GN1 <- GN1000[!grepl("^ICG", GN1000$DonorID), ]
GN1$DonorID <- NULL
GN2 <- GN1000[grepl("^ICG", GN1000$DonorID), ]
GN2$NationalID <- NULL

# Specify database fields to use
GN1fields <- c("NationalID", "CollNo", "OtherID1", "OtherID2")
GN2fields <- c("DonorID", "CollNo", "OtherID1", "OtherID2")

# Clean the data
GN1[GN1fields] <- lapply(GN1[GN1fields], function(x) DataClean(x))
GN2[GN2fields] <- lapply(GN2[GN2fields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergeSuffix(x, y3, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergeSuffix(x, y3, delim = c("space", "dash")))
```

```
# Remove duplicated DonorID records in GN2
GN2 <- GN2[!duplicated(GN2$DonorID), ]

# Generate KWIC index
GN1KWIC <- KWIC(GN1, GN1fields)
GN2KWIC <- KWIC(GN2, GN2fields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
          "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
          "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
          "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
          "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
          "U", "VALENCIA", "VIRGINIA", "WHITE")

# Fetch fuzzy and phonetic duplicate sets by method b
GNdupb <- ProbDup(kwic1 = GN1KWIC, kwic2 = GN2KWIC, method = "b",
                  excep = exep, fuzzy = TRUE, phonetic = TRUE,
                  encoding = "primary", semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |======================================================================| 100%
Block 1 / 1 |
```

```
## Phonetic matching
```

```
##
  |
  |======================================================================| 100%
Block 1 / 1 |
```

```
class(GNdupb)
```

```
## [1] "ProbDup"
```

```
GNdupb
```

```
## Method : b
##
## KWIC1 fields : NationalID CollNo OtherID1 OtherID2
##
## KWIC2 fields : DonorID CollNo OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## FuzzyDuplicates             107               353
## PhoneticDuplicates           41               126
## Total                       148 479(Distinct:383)
```

```
# Fetch fuzzy and phonetic duplicate sets by method c
GNdupc <- ProbDup(kwic1 = GN1KWIC, kwic2 = GN2KWIC, method = "c",
                  excep = exep, fuzzy = TRUE, phonetic = TRUE,
                  encoding = "primary", semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |======================                                                |  33%
Block 1 / 3 |
  |
  |==============================================                       |  67%
Block 2 / 3 |
  |
  |======================================================================| 100%
Block 3 / 3 |
```

```
## Phonetic matching
```

```
##
  |
  |======================                                                |  33%
Block 1 / 3 |
  |
  |==============================================                       |  67%
Block 2 / 3 |
  |
  |======================================================================| 100%
Block 3 / 3 |
```

```
class(GNdupc)
```

```
## [1] "ProbDup"
```

```
GNdupc
```

```
## Method : c
##
## KWIC1 fields : NationalID CollNo OtherID1 OtherID2
##
## KWIC2 fields : DonorID CollNo OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## FuzzyDuplicates             363               724
## PhoneticDuplicates           98               257
## Total                       461 981(Distinct:741)
```

Matching Strategies

1. **Fuzzy matching** or approximate string matching of keywords is carried out by computing the generalized levenshtein (edit) distance between them. This distance measure counts the number of deletions, insertions and substitutions necessary to turn one string to the another.

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
```

```
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate the KWIC index
GNKWIC <- KWIC(GN, GNfields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
         "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
         "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
         "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
         "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
         "U", "VALENCIA", "VIRGINIA", "WHITE")

# Fetch fuzzy duplicates
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = TRUE, max.dist = 3,
                 phonetic = FALSE, semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
  |
  |================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                 No..of.Sets    No..of.Records
## FuzzyDuplicates         378               745
## Total                   378 745(Distinct:745)
```

The maximum distance to be considered for a match can be specified by `max.dist` argument.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = TRUE, max.dist = 1,
                 phonetic = FALSE, semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
  |
  |================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                  No..of.Sets    No..of.Records
## FuzzyDuplicates          288               679
## Total                    288 679(Distinct:679)
```

Exact matching can be enforced with the argument `force.exact` set as TRUE. It can be used to avoid fuzzy matching when the number of alphabet characters in keywords is lesser than a critical value (`max.alpha`). Similarly, the value of `max.digit` can also be set according to the requirements to enforce exact matching. The default value of `Inf` avoids fuzzy matching and enforces exact matching for all keywords having any numerical characters. If `max.digit` and `max.alpha` are both set to `Inf`, exact matching will be enforced for all the keywords.

When exact matching is enforced, for keywords having both alphabet and numeric characters and with the number of alphabet characters greater than `max.digit`, matching will be carried out separately for alphabet and numeric characters present.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = TRUE, force.exact = TRUE, max.alpha = 4, max.digit = Inf,
                 phonetic = FALSE, semantic = FALSE)
```

```
## Fuzzy matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
```

```
  |
  |======================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                  No..of.Sets    No..of.Records
## FuzzyDuplicates          378               745
## Total                    378 745(Distinct:745)
```

2. **Phonetic matching** of keywords is carried out using the Double Metaphone phonetic algorithm which is implemented as the helper function `DoubleMetaphone`, (Philips 2000), to identify keywords that have the similar pronunciation.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = FALSE,
                 phonetic = TRUE,
                 semantic = FALSE)
```

```
## Phonetic matching
```

```
##
  |
  |================                                                      |  25%
Block 1 / 4 |
  |
  |================================                                      |  50%
Block 2 / 4 |
  |
  |================================================                      |  75%
Block 3 / 4 |
  |
  |======================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## PhoneticDuplicates           99               260
## Total                        99 260(Distinct:260)
```

Either the primary or alternate encodings can be used by specifying the `encoding` argument.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = FALSE,
                 phonetic = TRUE, encoding = "alternate",
                 semantic = FALSE)
```

```
## Phonetic matching
```

```
##
```

```
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
  |
  |================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## PhoneticDuplicates          98               263
## Total                       98 263(Distinct:263)
```

The argument `phon.min.alpha` sets the limits for the number of alphabet characters to be present in a string for executing phonetic matching.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = FALSE,
                 phonetic = TRUE, encoding = "alternate", phon.min.alpha = 4,
                 semantic = FALSE)
```

```
## Phonetic matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |
  |================================================                 |  75%
Block 3 / 4 |
  |
  |================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## PhoneticDuplicates         304               451
## Total                      304 451(Distinct:451)
```

Similarly `min.enc` sets the limits for the number of characters to be present in the encoding of a keyword for

phonetic matching.

```
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = FALSE,
                 phonetic = TRUE, encoding = "alternate", min.enc = 4,
                 semantic = FALSE)
```

```
## Phonetic matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
Block 2 / 4 |
  |================================================                 |  75%
Block 3 / 4 |
  |================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## PhoneticDuplicates          59               156
## Total                       59 156(Distinct:156)
```

3. **Semantic matching** matches keywords based on a list of accession name synonyms supplied as list with character vectors of synonym sets (synsets) to the `syn` argument. Synonyms in this context refers to interchangeable identifiers or names by which an accession is recognized. Multiple keywords specified as members of the same synset in `syn` are matched. To facilitate accurate identification of synonyms from the KWIC index, identical data standardization operations using the `Merge*` and `DataClean` functions for both the original database fields and the synset list are recommended.

```
# Specify the synsets as a list
syn <- list(c("CHANDRA", "AH 114"), c("TG-1", "VIKRAM"))

# Clean the data in the synsets
syn <- lapply(syn, DataClean)

GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                 fuzzy = FALSE, phonetic = FALSE,
                 semantic = TRUE, syn = syn)
```

```
## Semantic matching
```

```
##
  |
  |================                                                 |  25%
Block 1 / 4 |
  |
  |================================                                 |  50%
```

```
Block 2 / 4 |
  |
  |=================================================                 |  75%
Block 3 / 4 |
  |
  |==================================================================| 100%
Block 4 / 4 |
```

```
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets No..of.Records
## SemanticDuplicates           2              5
## Total                        2  5(Distinct:5)
```

Memory and Speed Constraints

As the number of keywords in the KWIC indexes increases, the memory consumption by the function also increases proportionally. This is due to the reason that for string matching, this function relies upon creation of a $n \times m$ matrix of all possible keyword pairs for comparison, where n and m are the number of keywords in the query and source indexes respectively. This can lead to `cannot allocate vector of size...` errors in case of large KWIC indexes where the comparison matrix is too large to reside in memory. In such a case, the `chunksize` argument can be reduced from the default 1000 to get the appropriate size of the KWIC index keyword block to be used for searching for matches at a time. However a smaller `chunksize` may lead to longer computation time due to the memory-time trade-off.

The progress of matching is displayed in the console as number of keyword blocks completed out of the total number of blocks, the percentage of achievement and a text-based progress bar.

In case of multi-byte characters in keywords, the speed of keyword matching is further dependent upon the `useBytes` argument as described in `help("stringdist-encoding")` for the `stringdist` function in the namesake package (van der Loo 2014), which is made use of here for string matching.

The CPU time taken for retrieval of probable duplicate sets under different options for the arguments `chunksize` and `useBytes` can be visualized using the `microbenchmark` package (Fig. 3).

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
```

```
                        function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                        function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate the KWIC index
GNKWIC <- KWIC(GN, GNfields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
          "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
          "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
          "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
          "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
          "U", "VALENCIA", "VIRGINIA", "WHITE")

# Specify the synsets as a list
syn <- list(c("CHANDRA", "AH 114"), c("TG-1", "VIKRAM"))
syn <- lapply(syn, DataClean)
```

```
library(microbenchmark)
timings <- microbenchmark(
  # Fetch duplicate sets with default chunk.size
  t1 = ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                                     chunksize = 1000, useBytes = TRUE,
                                     fuzzy = TRUE, phonetic = TRUE,
                                     semantic = TRUE, syn = syn),
  # Fetch duplicate sets chunk.size 2000
  t2 = ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                                     chunksize = 2000, useBytes = TRUE,
                                     fuzzy = TRUE, phonetic = TRUE,
                                     semantic = TRUE, syn = syn),
  # Fetch duplicate sets chunk.size 100
  t3 = ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                                     chunksize = 100, useBytes = TRUE,
                                     fuzzy = TRUE, phonetic = TRUE,
                                     semantic = TRUE, syn = syn),
  # Fetch duplicate sets useBytes = FALSE
  t4 = ProbDup(kwic1 = GNKWIC, method = "a", excep = exep,
                                     chunksize = 1000, useBytes = FALSE,
                                     fuzzy = TRUE, phonetic = TRUE,
                                     semantic = TRUE, syn = syn),
  times = 10)
```

```
boxplot(timings, col = c("#1B9E77", "#D95F02", "#7570B3", "#E7298A"))
legend("topright", c("t1 : chunksize = 1000,\n     useBytes = T (default)\n",
         "t2 : chunksize = 2000,\n     useBytes = T\n",
         "t3 : chunksize = 500,\n     useBytes = T\n",
         "t4 : chunksize = 1000,\n     useBytes = F\n"),
       bty = "n", cex = 0.6)
```

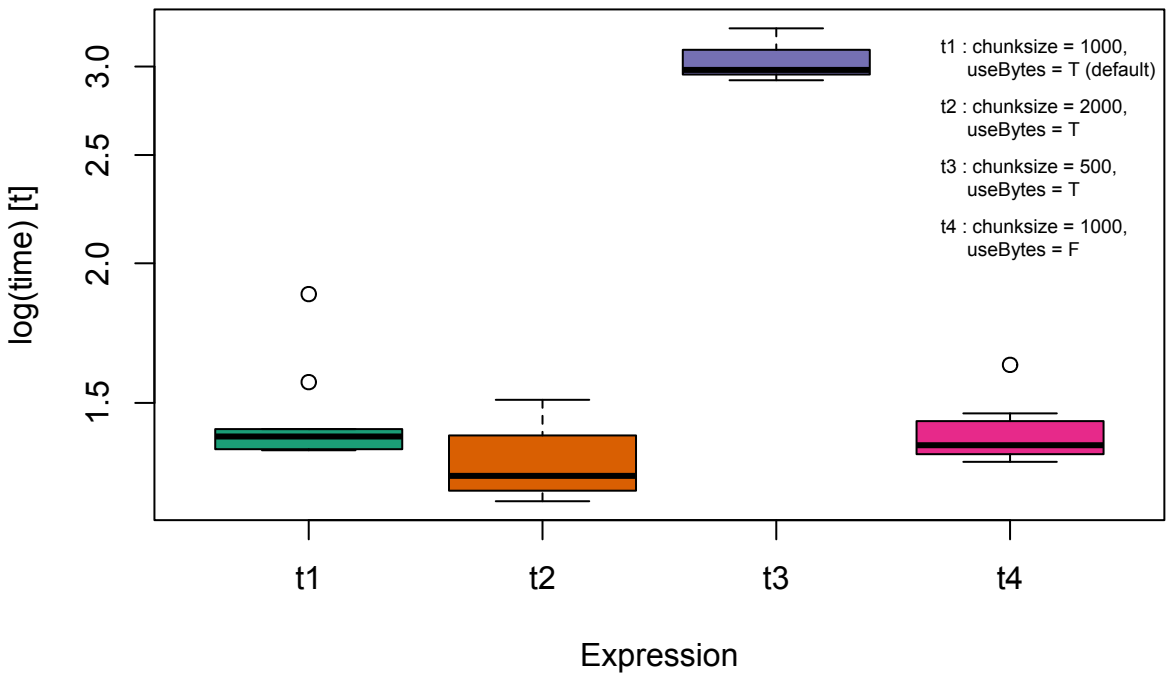

Fig. 3. CPU time with different `ProbDup` arguments estimated using the `microbenchmark` package.

Set Review, Modification and Validation

The initially retrieved sets may be intersecting with each other because there might exist accessions which occur in more than duplicate set. Disjoint sets can be generated by merging such overlapping sets using the function `DisProbDup`.

Disjoint sets are retrieved either individually for each type of probable duplicate sets or considering all type of sets simultaneously. In case of the latter, the disjoint of all the type of sets alone are returned in the output as an additional data frame `DisjointDupicates` in an object of class `ProbDup`.

```
# Load example dataset
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
```

```
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate KWIC index
GNKWIC <- KWIC(GN, GNfields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
          "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
          "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
          "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
          "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
          "U", "VALENCIA", "VIRGINIA", "WHITE")

# Specify the synsets as a list
syn <- list(c("CHANDRA", "AH114"), c("TG1", "VIKRAM"))

# Fetch probable duplicate sets
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep, fuzzy = TRUE,
                 phonetic = TRUE, encoding = "primary",
                 semantic = TRUE, syn = syn)
```

```
# Initial number of sets
GNdup
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets     No..of.Records
## FuzzyDuplicates             378                745
## PhoneticDuplicates           99                260
## SemanticDuplicates            2                  5
## Total                       479 1010(Distinct:762)
```

```
# Get disjoint probable duplicate sets of each kind
disGNdup1 <- DisProbDup(GNdup, combine = NULL)
# # Number of sets after combining intersecting sets
disGNdup1
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets     No..of.Records
## FuzzyDuplicates             181                745
## PhoneticDuplicates           80                260
## SemanticDuplicates            2                  5
## Total                       263 1010(Distinct:762)
```

```
# Get disjoint probable duplicate sets combining all the kinds of sets
disGNdup2 <- DisProbDup(GNdup, combine = c("F", "P", "S"))
# Number of sets after combining intersecting sets
disGNdup2
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                    No..of.Sets    No..of.Records
## DisjointDupicates          167               762
## Total                      167 762(Distinct:762)
```

Once duplicate sets are retrieved they can be validated by manual clerical review by comparing with original PGR passport database(s) using the `ReviewProbDup` function. This function helps to retrieve PGR passport information associated with fuzzy, phonetic or semantic probable duplicate sets in an object of class `ProbDup` from the original databases(s) from which they were identified. The original information of accessions comprising a set, which have not been subjected to data standardization can be compared under manual clerical review for the validation of the set. By default only the fields(columns) which were used initially for creation of the KWIC indexes using the KWIC function are retrieved. Additional fields(columns) if necessary can be specified using the `extra.db1` and `extra.db2` arguments.

When any primary ID/key records in the fuzzy, phonetic or semantic duplicate sets are found to be missing from the original databases specified in `db1` and `db2`, then they are ignored and only the matching records are considered for retrieving the information with a warning.

This may be due to data standardization of the primary ID/key field using the function `DataClean` before creation of the KWIC index and subsequent identification of probable duplicate sets. In such a case, it is recommended to use an identical data standardization operation on the primary ID/key field of databases specified in `db1` and `db2` before running this function.

With `R` <= v3.0.2, due to copying of named objects by `list()`, `Invalid .internal.selfref detected and fixed...` warning can appear, which may be safely ignored.

The output data frame can be subjected to clerical review either after exporting into an external spreadsheet using `write.csv` function or by using the `edit` function.

The column `DEL` can be used to indicate whether a record has to be deleted from a set or not. `Y` indicates "Yes", and the default `N` indicates "No".

The column `SPLIT` similarly can be used to indicate whether a record in a set has to be branched into a new set. A set of identical integers in this column other than the default `0` can be used to indicate that they are to be removed and assembled into a new set.

```
# Load the original database and clean the Primary ID/key field
GN1000 <- GN1000
GN1000$NationalID <- DataClean(GN1000$NationalID)

# Get the data frame for reviewing the duplicate sets identified
RevGNdup <- ReviewProbDup(pdup = disGNdup1, db1 = GN1000,
                          extra.db1 = c("SourceCountry", "TransferYear"),
                          max.count = 30, insert.blanks = TRUE)
```

```
head(RevGNdup)
```

```
##   SET_NO TYPE K[a]  PRIM_ID                IDKW  DEL SPLIT COUNT
## 1      1    F [K1] EC100277 [K1]EC100277:U44712    N     0     3
## 2      1    F [K1]  EC21118  [K1]EC21118:U44712    N     0     3
## 3      1    F [K1] IC494796 [K1]IC494796:U44712    N     0     3
## 4     NA      <NA>     <NA>                <NA> <NA>    NA    NA
## 5      1    P [K1] EC100713  [K1]EC100713:STARR    N     0    14
## 6      1    P [K1] EC106985  [K1]EC106985:STARR    N     0    14
##   K1_NationalID               K1_CollNo K1_DonorID K1_OtherID1
## 1      EC100277    Shulamith/ NRCG-14555   ICG-4709
## 2       EC21118 U 4-47-12; EC 21118; UKA    ICG3265
## 3      IC494796               U-4-47-12   ICG-6890
## 4          <NA>                    <NA>       <NA>        <NA>
## 5      EC100713              EC 100713;    ICG5296
## 6      EC106985                   Starr    ICG3479
##    K1_OtherID2        K1X_SourceCountry K1X_TransferYear
## 1     U4-47-12                   Israel             2014
## 2 U44712 U K A                Australia             1989
## 3       U44712                  Unknown             2010
## 4         <NA>                     <NA>               NA
## 5        STARR United States of America             2004
## 6              United States of America             2001
```

```
# Examine and review the duplicate sets using edit function
RevGNdup <- edit(RevGNdup)

# OR examine and review the duplicate sets after exporting them as a csv file
write.csv(file="Duplicate sets for review.csv", x=RevGNdup)
```

After clerical review, the data frame created using the function `ReviewProbDup` from an object of class `ProbDup` can be reconstituted back to the same object after the review using the function `ReconstructProbDup`.

The instructions for modifying the sets entered in the appropriate format in the columns `DEL` and `SPLIT` during clerical review are taken into account for reconstituting the probable duplicate sets. Any records with `Y` in column `DEL` are deleted and records with identical integers in the column `SPLIT` other than the default `0` are reassembled into a new set.

```
# The original set data
subset(RevGNdup, SET_NO==13 & TYPE=="P", select= c(IDKW, DEL, SPLIT))
```

```
##                                          IDKW DEL SPLIT
## 111                       [K1]EC38607:MANFREDI1   N     0
## 112                       [K1]EC420966:MANFREDI   N     0
## 113                     [K1]EC42549:MANFREDI68   N     0
## 114                       [K1]EC42550:MANFRED1   N     0
## 115 [K1]EC552714:CHAMPAQUI, [K1]EC552714:MANFREDI   N     0
## 116                    [K1]EC573128:MANFREDI84   N     0
## 117 [K1]IC304523:CHAMPAGUE, [K1]IC304523:MANFREDI   N     0
```

```
# Make dummy changes to the set for illustration
RevGNdup[c(113, 116), 6] <- "Y"
RevGNdup[c(111, 114), 7] <- 1
RevGNdup[c(112, 115, 117), 7] <- 2
# The instruction for modification in columns DEL and SPLIT
subset(RevGNdup, SET_NO==13 & TYPE=="P", select= c(IDKW, DEL, SPLIT))
```

```
##                                                IDKW DEL SPLIT
## 111                           [K1]EC38607:MANFREDI1   N     1
## 112                           [K1]EC420966:MANFREDI   N     2
## 113                          [K1]EC42549:MANFREDI68   Y     0
## 114                            [K1]EC42550:MANFRED1   N     1
## 115 [K1]EC552714:CHAMPAQUI, [K1]EC552714:MANFREDI   N     2
## 116                        [K1]EC573128:MANFREDI84   Y     0
## 117 [K1]IC304523:CHAMPAGUE, [K1]IC304523:MANFREDI   N     2
```

```
# Reconstruct ProDup object
GNdup2 <- ReconstructProbDup(RevGNdup)
# Initial no. of sets
disGNdup1
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets     No..of.Records
## FuzzyDuplicates             181                745
## PhoneticDuplicates           80                260
## SemanticDuplicates            2                  5
## Total                       263 1010(Distinct:762)
```

```
# No. of sets after modifications
GNdup2
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets    No..of.Records
## FuzzyDuplicates             180               523
## PhoneticDuplicates           81               258
## SemanticDuplicates            2                 5
## Total                       263 786(Distinct:674)
```

Other functions

The `ProbDup` object is a list of data frames of different kinds of probable duplicate sets *viz*- `FuzzyDuplicates`, `PhoneticDuplicates`, `SemanticDuplicates` and `DisjointDuplicates`. Each row of the component data frame will have information of a set, the type of set, the set members as well as the keywords based on which the set was formed. This data can be reshaped into long form using the function `ParseProbDup`. This function which will transform a `ProbDup` object into a single data frame.

```
# Convert 'ProbDup' object to a long form data frame of sets
GNdupParsed <- ParseProbDup(GNdup)
head(GNdupParsed)
```

```
##   SET_NO TYPE    K  PRIM_ID                IDKW COUNT
## 1      1    F [K1] EC100277 [K1]EC100277:U44712     3
## 2      1    F [K1]  EC21118  [K1]EC21118:U44712     3
## 3      1    F [K1] IC494796 [K1]IC494796:U44712     3
## 4     NA      <NA>     <NA>                <NA>    NA
## 5      2    F [K1] EC100280    [K1]EC100280:NC5     3
```

```
## 6      2    F [K1] EC100721    [K1]EC100721:NC5     3
```

The prefix `K*` here indicates the KWIC index of origin. This is useful in ascertaining the database of origin of the accessions when method `"b"` or `"c"` was used to create the input `ProbDup` object.

Once the sets are reviewed and modified, the validated set data fields from the `ProbDup` object can be added to the original PGR passport database using the function `AddProbDup`. The associated data fields such as `SET_NO`, `ID` and `IDKW` are added based on the `PRIM_ID` field(column).

```
# Loading original database
GN2 <- GN1000

# Add the duplicates set data to the original database
GNwithdup <-  AddProbDup(pdup = GNdup, db = GN2, addto = "I")
```

In case more than one KWIC index was used to generate the object of class `ProbDup`, the argument `addto` can be used to specify to which database the data fields are to be added. The default `"I"` indicates the database from which the first KWIC index was created and `"II"` indicates the database from which the second index was created.

The function `SplitProbDup` can be used to split an object of class `ProbDup` into two on the basis of set counts. This is useful for reviewing separately the sets with larger set counts.

```
# Load PGR passport database
GN <- GN1000

# Specify as a vector the database fields to be used
GNfields <- c("NationalID", "CollNo", "DonorID", "OtherID1", "OtherID2")

# Clean the data
GN[GNfields] <- lapply(GN[GNfields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN[GNfields] <- lapply(GN[GNfields],
                       function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Generate KWIC index
GNKWIC <- KWIC(GN, GNfields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
         "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
         "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
         "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
         "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
         "U", "VALENCIA", "VIRGINIA", "WHITE")

# Specify the synsets as a list
syn <- list(c("CHANDRA", "AH114"), c("TG1", "VIKRAM"))
```

```
# Fetch probable duplicate sets
GNdup <- ProbDup(kwic1 = GNKWIC, method = "a", excep = exep, fuzzy = TRUE,
                 phonetic = TRUE, encoding = "primary",
                 semantic = TRUE, syn = syn)
```

```
# Split the probable duplicate sets
GNdupSplit <- SplitProbDup(GNdup, splitat = c(10, 10, 10))
GNdupSplit[[1]]
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets     No..of.Records
## FuzzyDuplicates             338                744
## PhoneticDuplicates           99                260
## SemanticDuplicates            2                  5
## Total                       439 1009(Distinct:762)
```

```
GNdupSplit[[3]]
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                   No..of.Sets    No..of.Records
## FuzzyDuplicates            40               136
## Total                      40 136(Distinct:136)
```

Alternatively, two different `ProbDup` objects can be merged together using the function `MergeProbDup`.

```
GNdupMerged <- MergeProbDup(GNdupSplit[[1]], GNdupSplit[[3]])
GNdupMerged
```

```
## Method : a
##
## KWIC1 fields : NationalID CollNo DonorID OtherID1 OtherID2
##
##                     No..of.Sets     No..of.Records
## FuzzyDuplicates             378                745
## PhoneticDuplicates           99                260
## SemanticDuplicates            2                  5
## Total                       479 1010(Distinct:762)
```

The summary of accessions according to a grouping factor field(column) in the original database(s) within the probable duplicate sets retrieved in a `ProbDup` object can be visualized by the `ViewProbDup` function. The resulting plot can be used to examine the extent of probable duplication within and between groups of accessions records.

```
# Load PGR passport databases
GN1 <- GN1000[!grepl("^ICG", GN1000$DonorID), ]
GN1$DonorID <- NULL
GN2 <- GN1000[grepl("^ICG", GN1000$DonorID), ]
GN2 <- GN2[!grepl("S", GN2$DonorID), ]
GN2$NationalID <- NULL
```

```
GN1$SourceCountry <- toupper(GN1$SourceCountry)
GN2$SourceCountry <- toupper(GN2$SourceCountry)

GN1$SourceCountry <- gsub("UNITED STATES OF AMERICA", "USA", GN1$SourceCountry)
GN2$SourceCountry <- gsub("UNITED STATES OF AMERICA", "USA", GN2$SourceCountry)

# Specify as a vector the database fields to be used
GN1fields <- c("NationalID", "CollNo", "OtherID1", "OtherID2")
GN2fields <- c("DonorID", "CollNo", "OtherID1", "OtherID2")

# Clean the data
GN1[GN1fields] <- lapply(GN1[GN1fields], function(x) DataClean(x))
GN2[GN2fields] <- lapply(GN2[GN2fields], function(x) DataClean(x))
y1 <- list(c("Gujarat", "Dwarf"), c("Castle", "Cary"), c("Small", "Japan"),
           c("Big", "Japan"), c("Mani", "Blanco"), c("Uganda", "Erect"),
           c("Mota", "Company"))
y2 <- c("Dark", "Light", "Small", "Improved", "Punjab", "SAM")
y3 <- c("Local", "Bold", "Cary", "Mutant", "Runner", "Giant", "No.",
        "Bunch", "Peanut")
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN1[GN1fields] <- lapply(GN1[GN1fields],
                         function(x) MergeSuffix(x, y3, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergeKW(x, y1, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergePrefix(x, y2, delim = c("space", "dash")))
GN2[GN2fields] <- lapply(GN2[GN2fields],
                         function(x) MergeSuffix(x, y3, delim = c("space", "dash")))

# Remove duplicated DonorID records in GN2
GN2 <- GN2[!duplicated(GN2$DonorID), ]

# Generate KWIC index
GN1KWIC <- KWIC(GN1, GN1fields)
GN2KWIC <- KWIC(GN2, GN2fields)

# Specify the exceptions as a vector
exep <- c("A", "B", "BIG", "BOLD", "BUNCH", "C", "COMPANY", "CULTURE",
          "DARK", "E", "EARLY", "EC", "ERECT", "EXOTIC", "FLESH", "GROUNDNUT",
          "GUTHUKAI", "IMPROVED", "K", "KUTHUKADAL", "KUTHUKAI", "LARGE",
          "LIGHT", "LOCAL", "OF", "OVERO", "P", "PEANUT", "PURPLE", "R",
          "RED", "RUNNER", "S1", "SAM", "SMALL", "SPANISH", "TAN", "TYPE",
          "U", "VALENCIA", "VIRGINIA", "WHITE")

# Specify the synsets as a list
syn <- list(c("CHANDRA", "AH114"), c("TG1", "VIKRAM"))

GNdupc <- ProbDup(kwic1 = GN1KWIC, kwic2 = GN2KWIC, method = "c",
                  excep = exep, fuzzy = TRUE, phonetic = TRUE,
                  encoding = "primary", semantic = TRUE, syn = syn)
```

```
## Fuzzy matching

##
  |
  |======================                                                |  33%
Block 1 / 3 |
  |
  |================================================                      |  67%
Block 2 / 3 |
  |
  |======================================================================| 100%
Block 3 / 3 |

## Phonetic matching

##
  |
  |======================                                                |  33%
Block 1 / 3 |
  |
  |================================================                      |  67%
Block 2 / 3 |
  |
  |======================================================================| 100%
Block 3 / 3 |

## Semantic matching

##
  |
  |======================                                                |  33%
Block 1 / 3 |
  |
  |================================================                      |  67%
Block 2 / 3 |
  |
  |======================================================================| 100%
Block 3 / 3 |
```

```
# Get the summary data.frames and Grob
GNdupcView <- ViewProbDup(GNdupc, GN1, GN2, "SourceCountry", "SourceCountry",
                          max.count = 30, select = c("INDIA", "USA"), order = "type",
                          main = "Groundnut Probable Duplicates")
```

```
# View the summary data.frames
GNdupcView[[1]]
GNdupcView[[2]]
```

```
# Plot the summary visualization
library(gridExtra)
grid.arrange(GNdupcView[[3]])
```

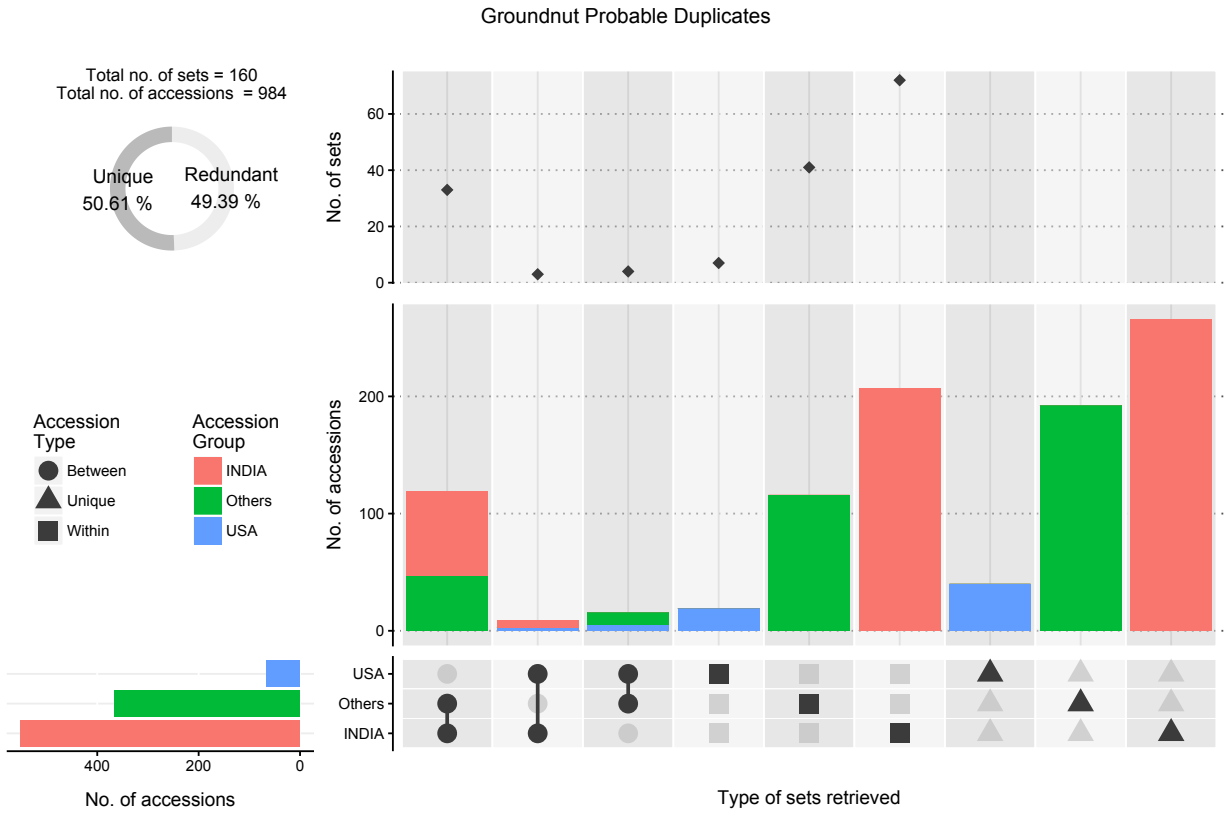

Fig. 5. Summary visualization of groundnut probable duplicate sets retrieved according to `SourceCountry` field.

The function `KWCounts` can be used to compute the keyword counts from PGR passport database fields(columns) which are considered for identification of probable duplicates. These keyword counts can give a rough indication of the completeness of the data in such fields (Fig. 3).

```
# Compute the keyword counts for the whole data
GNKWCouts <- KWCounts(GN, GNfields, exep)

# Compute the keyword counts for 'duplicated' records
GND <- ParseProbDup(disGNdup2, Inf, F)$PRIM_ID
GNDKWCouts <- KWCounts(GN[GN$NationalID %in% GND, ], GNfields, exep)

# Compute the keyword counts for 'unique' records
GNUKWCouts <- KWCounts(GN[!GN$NationalID %in% GND, ], GNfields, exep)

# Plot the counts as barplot
par(mfrow = c(3,1))
bp1 <- barplot(table(GNKWCouts$COUNT),
               xlab = "Word count", ylab = "Frequency",
               main = "A", col = "#1B9E77")
text(bp1, 0, table(GNKWCouts$COUNT),cex = 1, pos = 3)
legend("topright", paste("No. of records =",
                  nrow(GN)), bty = "n")
bp2 <- barplot(table(GNDKWCouts$COUNT),
               xlab = "Word count", ylab = "Frequency",
               main = "B", col = "#D95F02")
text(bp2, 0, table(GNDKWCouts$COUNT),cex = 1, pos = 3)
legend("topright", paste("No. of records =",
                  nrow(GN[GN$NationalID %in% GND, ])), bty = "n")
bp3 <- barplot(table(GNUKWCouts$COUNT),
               xlab = "Word count", ylab = "Frequency",
               main = "C", col = "#7570B3")
text(bp3, 0, table(GNUKWCouts$COUNT),cex = 1, pos = 3)
legend("topright", paste("No. of records =",
                  nrow(GN[!GN$NationalID %in% GND, ])), bty = "n")
```

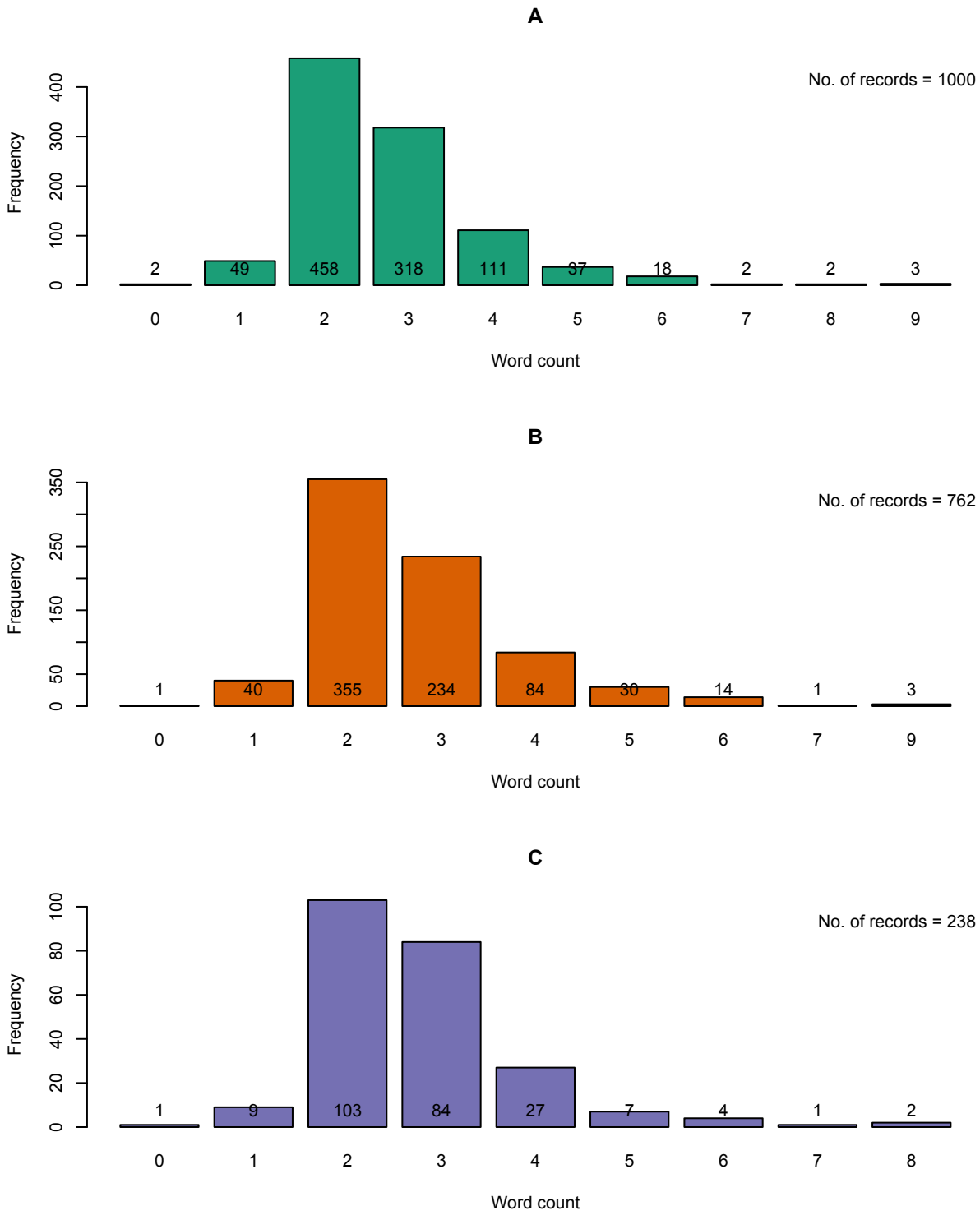

Fig. 6. The keyword counts in the database fields considered for identification of probable duplicates for **A.** the entire `GN1000` dataset, **B.** the probable duplicate records alone and **C.** the unique records alone.

Citing PGRdup

```
citation("PGRdup")
```

```
##  To cite the R package 'PGRdup' in publications use:
##    Aravind, J., J. Radhamani, Kalyani Srinivasan, B. Ananda Subhash,
##    and R. K. Tyagi (2015).  PGRdup: Discover Probable Duplicates in
##    Plant Genetic Resources Collections. R package version 0.2.1.
##
##  A BibTeX entry for LaTeX users is
##
##    @Manual{,
##      title = {PGRdup: Discover Probable Duplicates in Plant Genetic Resources
##  Collections},
##      author = {{J. Aravind} and {J. Radhamani} and {Kalyani Srinivasan} and
##  {B. Ananda Subhash} and {R. K. Tyagi}},
##      note = {R package version 0.2.1}
##    }
```

Session info

```
sessionInfo()
```

```
## R version 3.3.2 (2016-10-31)
## Platform: i386-w64-mingw32/i386 (32-bit)
## Running under: Windows 7 (build 7601) Service Pack 1
##
## locale:
## [1] LC_COLLATE=C                   LC_CTYPE=English_India.1252
## [3] LC_MONETARY=English_India.1252 LC_NUMERIC=C
## [5] LC_TIME=English_India.1252
##
## attached base packages:
## [1] stats     graphics  grDevices utils     datasets  methods   base
##
## other attached packages:
## [1] gridExtra_2.2.1        microbenchmark_1.4-2.1 wordcloud_2.5
## [4] RColorBrewer_1.1-2     PGRdup_0.2.3           diagram_1.6.3
## [7] shape_1.4.2
##
## loaded via a namespace (and not attached):
##  [1] igraph_1.0.1       Rcpp_0.12.7        knitr_1.15.1
##  [4] magrittr_1.5       munsell_0.4.3      colorspace_1.2-6
##  [7] plyr_1.8.4         stringr_1.1.0      tools_3.3.2
## [10] parallel_3.3.2     grid_3.3.2         data.table_1.10.0
## [13] gtable_0.2.0       htmltools_0.3.5    assertthat_0.1
## [16] lazyeval_0.2.0     yaml_2.1.13        rprojroot_1.2
## [19] digest_0.6.12      tibble_1.2         ggplot2_2.2.1
## [22] stringdist_0.9.4.4 slam_0.1-38        evaluate_0.10
## [25] rmarkdown_1.3      labeling_0.3       stringi_1.1.2
## [28] scales_0.4.1       backports_1.0.5
```

References

Knüpffer, H. 1988. "The European Barley Database of the ECP/GR: An Introduction." *Die Kulturpflanze* 36 (1): 135–62.

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