

# Time Series Database Interface (TSdbi) Guide and Illustrations

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# 1 Introduction

This vignette illustrates the various *R* (R Core Team, 2012) *TSdbi* packages using time series data from several sources. The main purpose of *TSdbi* is to provide a common *API*, so simplicity of changing the data source is a primary feature. The vignette also illustrates some simple time series manipulation and plotting using packages *tframe* and *tfplot*.

To generate this vignette requires most of the *TS\** packages, but users will only need one, or a few of these packages. For example, a time series database can be built with several SQL database backends, but typically only one would be used. On the other hand, several packages pull data from various Internet sources, so several of these might be used to accommodate the various specifics of the sources.

If package *TSdata* is installed on your system, it should be possible to view the pdf version of this guide with `vignette("Guide", package="TSdata")`. Otherwise, consider getting the pdf file directly from CRAN at <http://cran.at.r-project.org/web/packages/TSdata/index.html>. Many parts of the vignette can be run by loading the appropriate packages, but some examples use data that has been loaded into a larger database and it will not be possible to reproduce those examples without the underlying database.

Section 2 of this vignette illustrates the mechanism for connecting to a data source. The connection contains all the source specific information so, once the connection is established, the syntax for retrieving data is similar for different sources. The section also illustrates packages that pull data from Internet sources. This currently includes *TSgetSymbol*, *TShistQuote*, *TSjson*, *TSxls*, and *TSzip*, but others are in progress. These packages only get data, they do not support writing data to the database. Finally, the section also illustrates the flexibility to return different types of time series objects.

Section 3 illustrates the SQL packages *TSPostgreSQL*, *TSMYSQL*, *TSSQLite*, and *TSodbc*. These use a very standard SQL table structure and syntax, so it should be possible to use other SQL backends. The package *TSOracle* is available on R-forge at <http://tsdbi.r-forge.r-project.org/> but I currently do not have a server in place to test it properly. (If anyone would be interested in doing this, please contact me, [pgilbert.ttv9z@ncf.ca](mailto:pgilbert.ttv9z@ncf.ca).) These packages get data and also support writing data to the database. See Appendix "B" (section 9) and the vignette in package *TSdbi* for more explanation of the underlying database tables, and for bulk loading of data into a database. The section illustrates writing artificial data to the database.

Section 4 provides additional examples of *TSdbi* functionality and of using and graphing series. Subsection 4.1 illustrates fetching data from the web and loading it into a local database.

Section 5 illustrates package *TScompare* for comparing time series databases, and Section 6 illustrates the use of realtime vintages of data.

Section 7 illustrates some functions for exporting time series data from *R* into *.xls* and *.csv* files. While regular *R* users may not be too interested in this, it can be useful for colleagues that are not yet regular users. The section is

fairly self contained. (But will give non-users some exposure to *R*, so they may decide they do not actually need to export the data.)

Appendix “A” provides connection details specific to the different database sources, Appendix “B” provides more details about the structure of SQL databases, and Appendix “C” provides some examples of SQL queries that may be useful for database maintenance. Finally, Appendix “D” reproduces the README file for package *TSjson*, to give details regarding installation of Python helper utilities.

Many of the *TS\** packages are wrappers of other packages. The purpose is to provide a common API for interfacing with time series databases, and an easy mechanism to specify the type of time series object that should be returned, for example, a *ts* object or a *zoo* (Zeileis and Grothendieck, 2005) object. One consequence of providing a common interface is that special strengths of some of the underlying packages cannot always be used. If you really need some of these features then you may need to go directly to the underlying package. However, if you limit your reliance on these underlying features then you will be able to move from one data source to another much more easily.

Loading a *TS\** package will also load required packages *TSdbi*, *DBI* (R Special Interest Group on Databases, 2009), *methods* (R Core Team, 2013), *tframePlus*, *zoo*, and then any underlying package that the specific *TS\** package uses.

## 2 Simple Examples of Time Series Data from the Internet

This section uses packages to pull data from the Internet. The general syntax of a connection is illustrated, and some simple calculations and graphs are done to demonstrate how the data might be used. However, the purpose of the *TS\** packages is to provide a common interface, not to do all time series calculations and graphics. Once you have the data, you should be able to use whatever other *R* packages you like for your calculations and graphs.

The generic aspect of the interface API is accomplished by putting the information specific to the underlying source into the “connection”. Once the connection is established, other aspects of using data are the same, so one connection can be easily interchanged with another, and so your programs do not need to be changed when the data source is changed. (But, of course, some changes will be needed if the names of the variables change.)

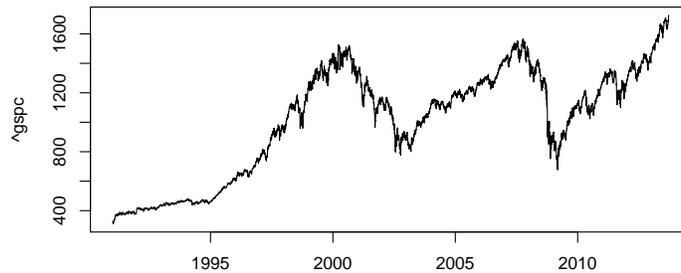
### 2.1 *TShistQuote* and *TSgetSymbol*

Packages *TShistQuote* and *TSgetSymbol* provide mechanisms to retrieve historical quote data from various sources. *TShistQuote* is a wrapper to *get.hist.quote* in package *tseries* (Trapletti and Hornik, 2012). A connection to Yahoo Finance is established, and data retrieved and plotted by

```

> library("TShistQuote")
> yahoo <- TSconnect("histQuote", dbname="yahoo")
> x <- TSget("^gspc", quote = "Close", con=yahoo)
> library("tfplot")
> tfplot(x)

```

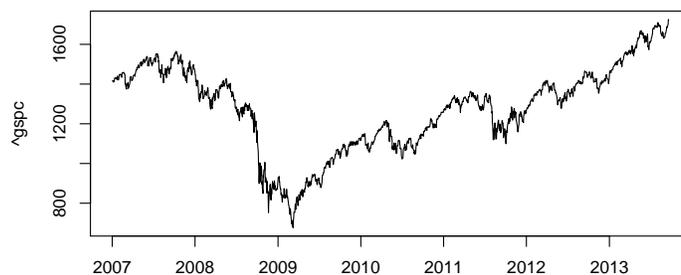


Package *TSgetSymbol* is a wrapper to *getSymbols* in package *quantmod* (Ryan, 2011). A connection to Yahoo using this, and retrieving the same data, can be done by

```

> library("TSgetSymbol")
> yahoo <- TSconnect("getSymbol", dbname="yahoo")
> x <- TSget("^gspc", quote = "Close", con=yahoo)
> tfplot(x)

```



Notice that the only difference is the library that is loaded and the name of the driver provided when establishing the connection. After that, the code is the same (and the data from the two connections should be the same, other than the difference in the default start date). This is the coding approach one would typically follow, so that changing the data source is easy. However, sometimes

it is interesting to compare the same data from different sources, so here is an example where the connections are given different names, so the same data through two different connection methods can be more easily compared:

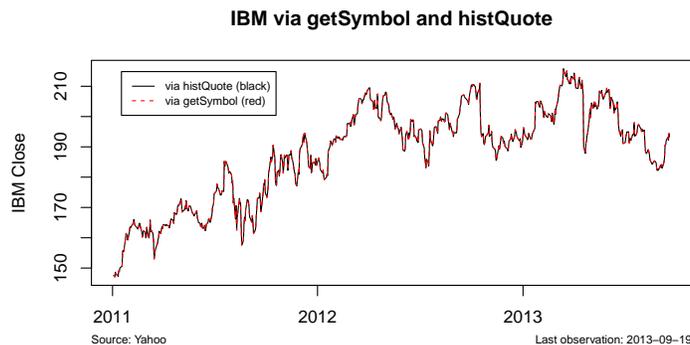
```
> ya1 <- TSconnect("getSymbol", dbname="yahoo")
> ya2 <- TSconnect("histQuote", dbname="yahoo")
> ibmC1 <- TSget("ibm", ya1, quote = "Close", start="2011-01-03")
> ibmC2 <- TSget("ibm", ya2, quote = "Close", start="2011-01-03")
```

In this example the underlying packages both return *zoo* time series objects, but the encoding of the date index vectors are of different classes (*Date* vs *POSIXct*). This is usually not a problem because one would usually work with one package or the other, but it does become a problem when comparing the two objects returned by the different methods. The time representation of *ibmC1* can be changed from *POSIXct* to *Date* by:

```
> tframe(ibmC1) <- as.Date(tframe(ibmC1))
```

The two series can then be plotted:

```
> tfplot(ibmC2, ibmC1,
        ylab="IBM Close",
        title="IBM via getSymbol and histQuote",
        lastObs=TRUE,
        legend=c("via histQuote (black)", "via getSymbol (red)"),
        source="Source: Yahoo")
```



Or the difference can be used to check equality:

```
> max(abs(ibmC2 - ibmC1))
```

```
[1] 0
```

(Note that this difference calculation does not catch a difference in length, which occurs if new data has been release on one connection and not the other.

At some point Yahoo was releasing partial data early, and these connection are correcting differently for this. So, at some times of day, the last available data point is not the same on these two connections.)

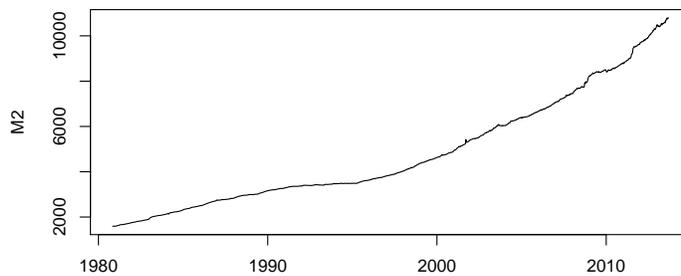
A certain amount of meta data can be returned with the time series object and can be extracted with these utilities:

```
> TSdescription(x)
[1] "^gspc Close from yahoo"
> TSdoc(x)
[1] "^gspc Close from yahoo retrieved 2013-09-20 08:15:41"
> TSlabel(x)
[1] "^gspc Close"
> TSsource(x)
[1] "yahoo"
```

## 2.2 *TSgetSymbol* with FRED

Package *TSgetSymbol* can also be used to get data from the Federal Reserve Bank of St.Louis, as will be illustrated here. (Look at <http://research.stlouisfed.org/fred/> to find series identifiers.)

```
> library("TSgetSymbol")
> fred <- TSconnect("getSymbol", dbname="FRED")
> tfplot(TSget("M2", fred))
```

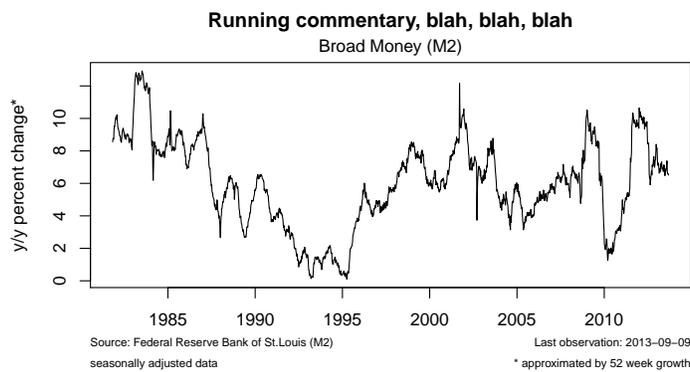


A connection can be specified to be used as the default, so it does not need to be specified each time:

```

> options(TSconnection=fred)
> tfOnePlot(percentChange(TSget("M2"), lag=52),
  title = "Running commentary, blah, blah, blah",
  subtitle="Broad Money (M2)",
  ylab= "y/y percent change*",
  source="Source: Federal Reserve Bank of St.Louis (M2)",
  footnoteLeft = "seasonally adjusted data",
  footnoteRight = "* approximated by 52 week growth",
  lastObs = TRUE )

```



Several different connections will be used in this vignette, and so a default will not be used. To unset the default

```

> options(TSconnection=NULL)

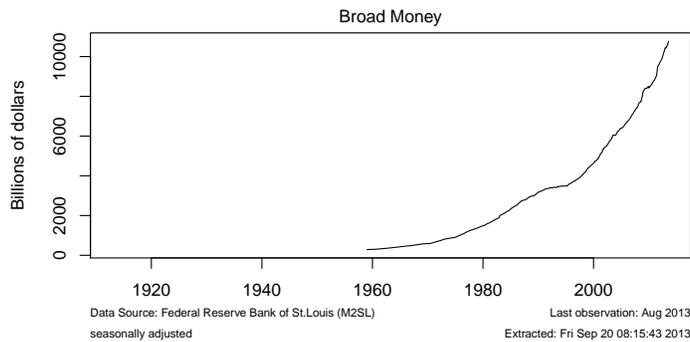
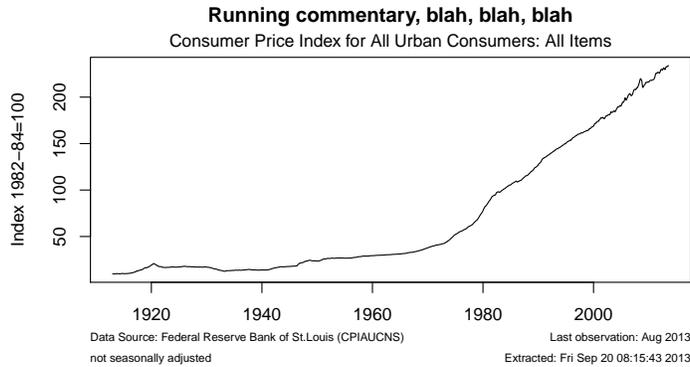
```

It is also possible to return multiple series, but they should all be of the same frequency. (The FRED series called M2 is a weekly series).

```

> x <- TSget(c("CPIAUCNS", "M2SL"), fred)
> tfplot(x,
  title = "Running commentary, blah, blah, blah",
  subtitle=c("Consumer Price Index for All Urban Consumers: All Items", "Broad Money"),
  ylab= c("Index 1982-84=100", "Billions of dollars"),
  source= c("Data Source: Federal Reserve Bank of St.Louis (CPIAUCNS)",
    "Data Source: Federal Reserve Bank of St.Louis (M2SL)"),
  footnoteLeft = c("not seasonally adjusted", "seasonally adjusted"),
  footnoteRight = paste("Extracted:", date()),
  lastObs = TRUE )

```



```
> TSdates(c("CPIAUCNS", "M2SL"), fred)
```

```
      [,1]
[1,] "CPIAUCNS from 1913 1 to 2013 8      12"
[2,] "M2SL from 1959 1 to 2013 8        12"
```

By default, *TSget* returns *ts* time series for annual, quarterly, and monthly data, and *zoo* series otherwise. It is possible to specify the type of object to return:

```
> x <- TSget(c("CPIAUCNS", "M2SL"), fred, TSrepresentation="zoo")
> class(x)
```

A session default can also be set for this with

```
> options(TSrepresentation="zoo")
```

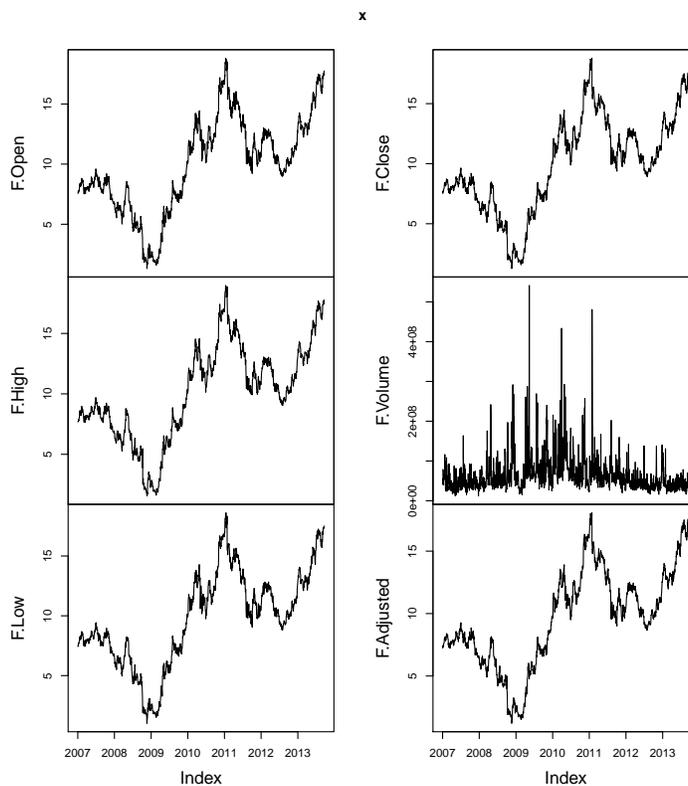
in which case all results will be return as *zoo* objects unless otherwise specified. The session default is unset with

```
> options(TSrepresentation=NULL)
```

Beware that it does not make sense to set *ts* as the default, because it is already the default for all series that can be represented as *ts*, and will not work correctly for other series. Other representations are possible. See the *TSget* help for more details.

The following connects to yahoo and loads the ticker symbol for Ford. This is a multivariate time series with open, close, etc.

```
> yahoo <- TSconnect("getSymbol", dbname="yahoo")
> x <- TSget("F", con=yahoo)
> plot(x)
```



Most of the plots in this vignette are done with the utilities in the *tfplot* package, but the usual *plot* function, used above, produces slightly different results that may be preferable in some situations. Also, for some time series objects, the plot method has been much improved from the default, so if you are using these objects you may find that *plot* provides attractive features.

In the case of the ticker data above, *tfplot* displays graphs in verticle panels. However, six panels do not nicely fit on a printed page. The first three are displayed with:

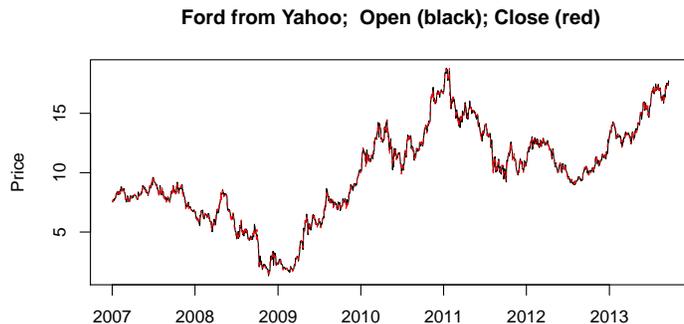
```
> tfplot(x, series=1:3)
```



It is possible to specify the number of graphs on an output screen with *graphs.per.page*, for example, `tfplot(x, graphs.per.page=3)`. Set `par(ask=TRUE)` if you want to stop and prompt for <Return> between pages in the graphics output.

The *quote* argument to *TSget* can be used to specify that only a subset of the market data should be returned:

```
> tfOnePlot(TSget("F", con=yahoo, quote=c("Open", "Close")),
  title="Ford from Yahoo; Open (black); Close (red)",
  ylab="Price")
```



### 2.3 *TSjson* with Statistics Canada

Package *TSjson* provides a mechanism to extract data from websites and pass it to *R* in JavaScript Object Notation (JSON) using package *rjson* (Couture-Beil, 2013). Package *RJSONIO* (Lang, 2012) has previously been used to bring the data into *R* but is not the current method. *TSjson* uses *Python* code to mechanize clicking through web pages to get a downloadable file. This really should be considered a temporary solution, until the data provider implements a true API. The current version of *TSjson* provides a connection to Statistics Canada's <http://www.statcan.gc.ca> Cansim database. (You should look at the Statistics Canada site to find series identifiers.) Please contact the package maintainer if you would like to help implement connections to other sites.

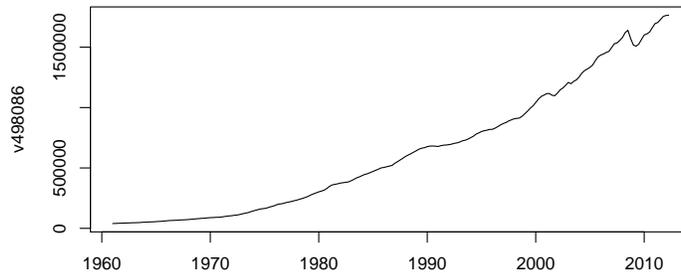
The connection can be established in two different ways. The simplest requires a system with *Python* installed, and *Python* modules *sys*, *json*, *mechanize*, *re*, *csv* and *urllib2*. This requires *Python 2* as not all modules are available for *Python 3*. Additional details are provided in a README file distributed with the *TSjson* package and also copied in an appendix here. Installing *Python* may be difficult in environments where users cannot easily install software, but a second method using a proxy server can be used. The proxy server needs *Python* and the modules, as well as server software (e.g. *Web2Py*), but the client machine requires nothing special other than *R* and *TSjson*. The proxy server can be anywhere on the Internet.

The following examples use the first method. More details on establishing connections with the second method are provided in the appendix section 8.6. First, establish a connection

```
> require("TSjson")
> cansim <- TSconnect("json", dbname="cansim")
```

Now data can be retrieved and a plot generated by

```
> x <- TSget("v498086", cansim)
> tfplot(x)
```

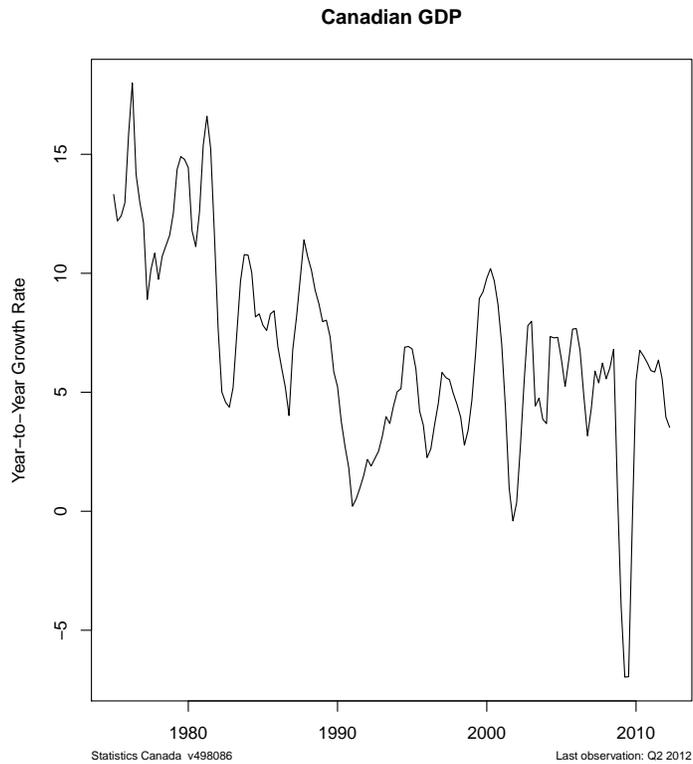


Meta data can also be retrieved:

```
> TSdescription("v498086", cansim)
[1] "Gross domestic product (GDP) at market prices (x 1,000,000)"
> TSdoc("v498086", cansim)
[1] "Table 380-0002: Gross domestic product, expenditure-based; Canada; Current prices; Seas
> TSlabel("v498086", cansim)
[1] "v498086"
> TSsource("v498086", cansim)
[1] "Statistics Canada: Table 380-0002; v498086"
```

A transformation of the data can be done, more detail added to the graph, and a start date specified:

```
> tfplot(ytoypc(x), start=c(1975,1),
        ylab="Year-to-Year Growth Rate",
        title="Canadian GDP",
        source=paste("Statistics Canada ", seriesNames(x)),
        lastObs=TRUE)
```



Sometimes it is useful to check availability:

```
> TSdates(c("v498086", "v498087"), cansim)
```

```

[1,]
[1,] "v498086 from 1961 1 to 2012 2      4"
[2,] "v498087 from 1961 1 to 2012 2      4"

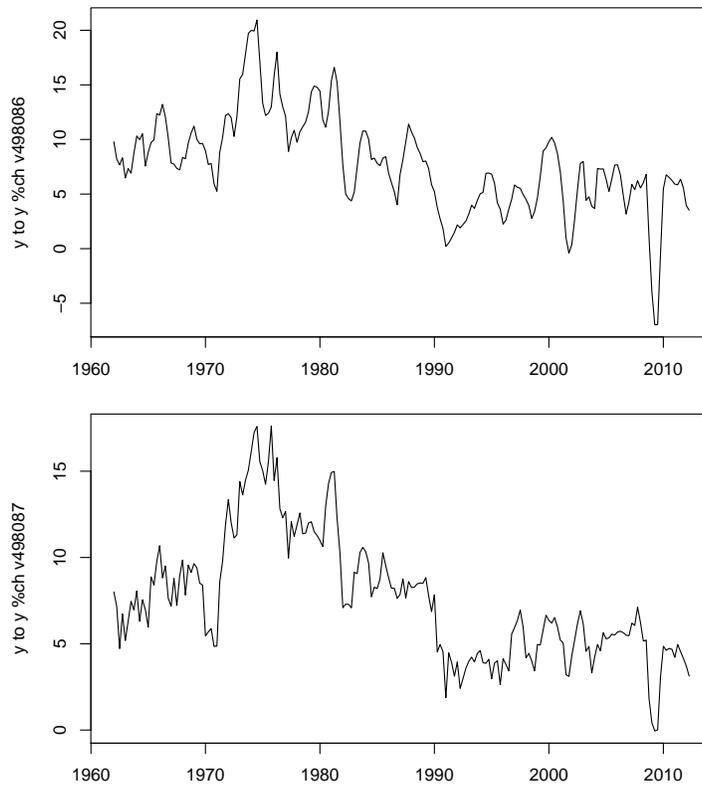
```

Or, plot more than one series:

```

> oldpar <- par(omi=c(0.1,0.1,0.1,0.1),mar=c(3.1,4.1,0.6,0.1))
> tfplot(ytoypc(TSget(c("v498086", "v498087"), cansim)))
> par(oldpar)

```



The default settings of *tplot* parameters usually work fairly well for interactive use. The above illustrates one way to achieve more specific control, which will often be necessary for generating pdf documents.

The meta data can also be retrieved with the series, which will generally be faster than retrieving it separately, if it is needed:

```
> resMorg <- TSget("V122746", cansim, TSdescription=TRUE, TSdoc=TRUE, TSlablel=TRUE)
> TSdescription(resMorg)

[1] "Total outstanding balances (x 1,000,000)"

> TSdoc(resMorg)

[1] "Table 176-0069: Residential mortgage credit; Canada; Average at month-end; Seasonally a

> TSlablel(resMorg)

[1] "V122746"

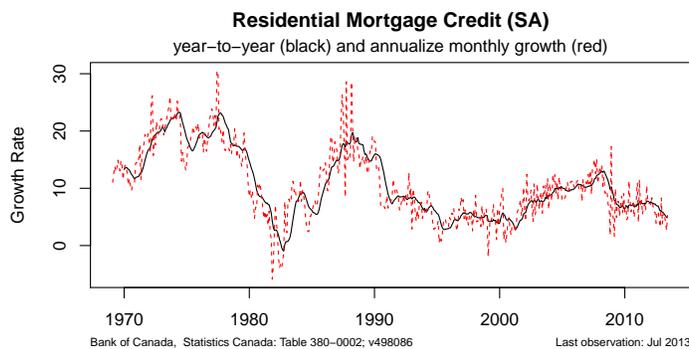
> TSseriesIDs(resMorg)

[1] "V122746"
```

```

> TSsource(resMorg)
[1] "Statistics Canada: Table 176-0069; v122746"
> seriesNames(resMorg) <- "Residential Mortgage Credit (SA)"
> tfplot(ytoypc(resMorg), annualizedGrowth(resMorg),
        title=seriesNames(resMorg),
        subtitle="year-to-year (black) and annualize monthly growth (red)",
        ylab="Growth Rate",
        source=paste("Bank of Canada, ", TSsource(x)),
        lastObs=TRUE)

```



## 2.4 TSxls

*TSxls* provides methods for the *TSdbi* interface, allowing the use of spreadsheets as if they are a database. (This is a poor substitute for a real database, but is sometimes convenient.) *TSxls* uses package *read.xls* in *gdata* (Warnes et al., 2011) *TSxls* does not support writing data to the spreadsheet (but to write time series data to a spreadsheet see *TSwrite.XLS* in *tframePlus*, discussed in section 7). The spreadsheet can be a remote file, which is retrieved when the connection is established.

The following retrieves a file from the Reserve Bank of Australia and maps the elements that are used: data, dates, identifiers, and series names.

```

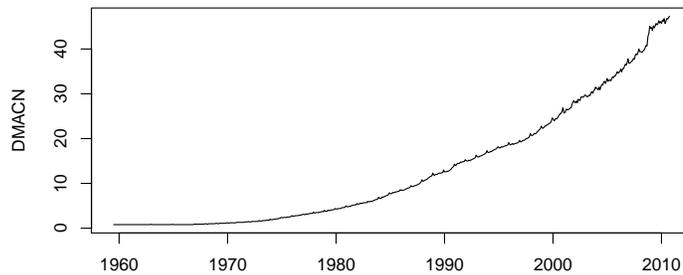
> library("TSxls")
> rba <- TSconnect("xls",
                  dbname="http://www.rba.gov.au/statistics/tables/xls/d03hist.xls",
                  map=list(ids =list(i=11, j="B:Q"),
                          data =list(i=12:627, j="B:Q"),
                          dates=list(i=12:627, j="A"),
                          names=list(i=4:7, j="B:Q"),
                          description = NULL,
                          tsrepresentation = function(data,dates){
                            ts(data,start=c(1959,7), frequency=12)})

```

This also illustrates how *tsrepresentation* can be specified as an arbitrary function to set the returned time series object representation.

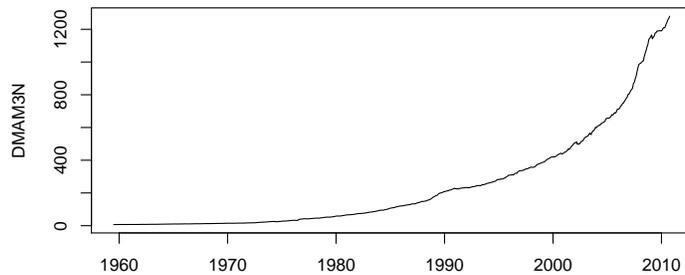
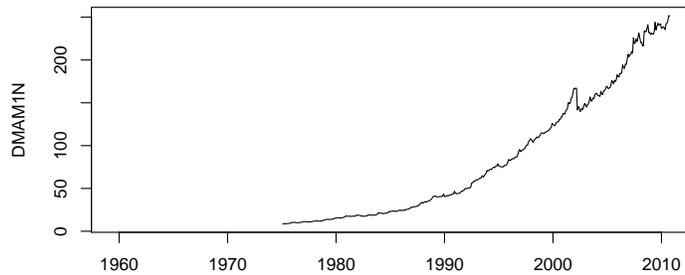
Beware that data is read into *R* when the connection is established, so changes in the spreadsheet will not be visible in *R* until a new connection is established.

```
> x <- TSget("DMACN", rba)
> require("tfplot")
> tfplot(x)
```



```
> x <- TSget(c("DMAM1N", "DMAM3N"), rba)
> tfplot(x)
> TSdescription(x)
```

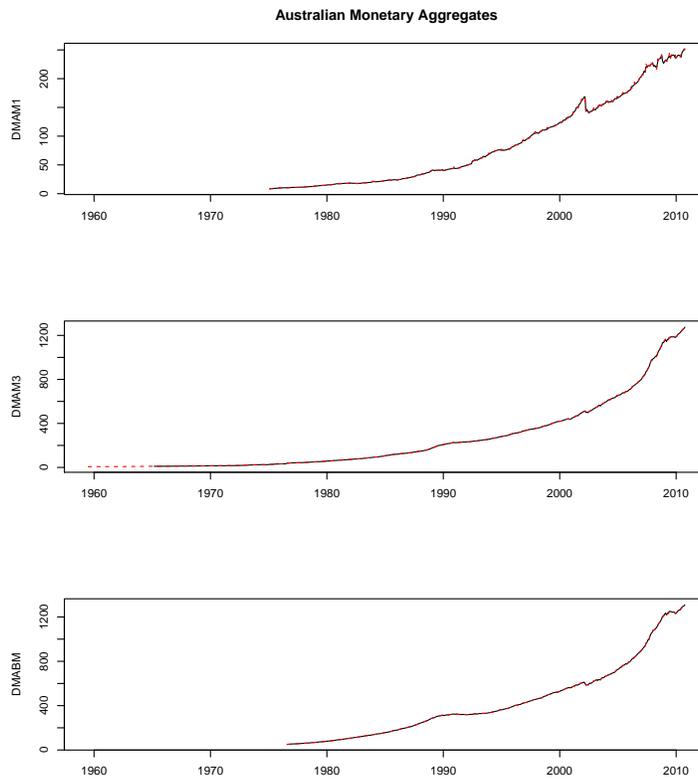
```
[1] " from http://www.rba.gov.au/statistics/tables/xls/d03hist.xls"
[2] " from http://www.rba.gov.au/statistics/tables/xls/d03hist.xls"
```



*tfplot* treats each series in the first argument as a panel to be plotted. It is possible to specify the number of graphs on each page of the output device with the argument *graphs.per.page*. As previously illustrated, it is also possible to specify that a subset of the series should be selected. (Also, as already illustrated above, the function *plot* displays the series somewhat differently than *tfplot*, and possibly differently depending on the objects time series representation.)

*tfplot* takes additional time series objects as arguments. Series in the first argument are plotted in separate panels. Series in subsequent time series objects will be plotted respectively on the same panels as the first, so the number of series in each object must be the same.

```
> tfplot(TSget(c("DMAM1S", "DMAM3S", "DMABMS"), rba),
         TSget(c("DMAM1N", "DMAM3N", "DMABMN"), rba),
         ylab=c("DMAM1", "DMAM3", "DMABM"),
         title="Australian Monetary Aggregates")
```

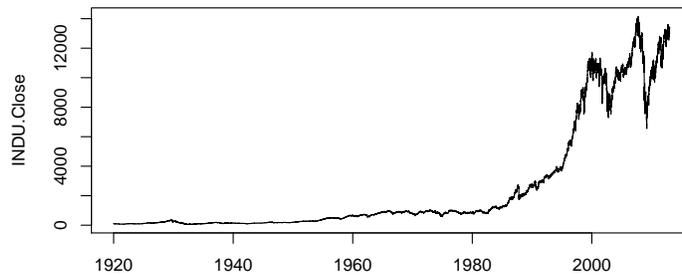
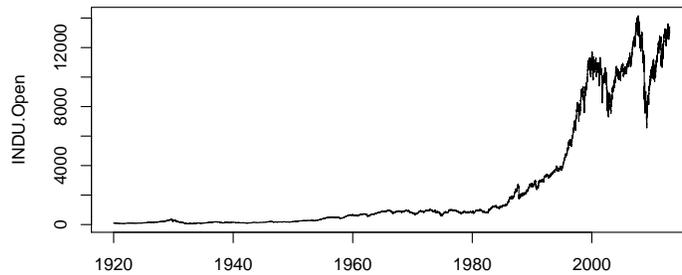


## 2.5 TSzip

*TSzip* provides methods for the *TSdbi* interface, allowing the use of zipped files that can be read by *read.table* as if each file is a database series (or group of series such as high, low, open, close, for a stock). The *dbname* is a directory or *url*. *TSzip* does not support writing data to the database.

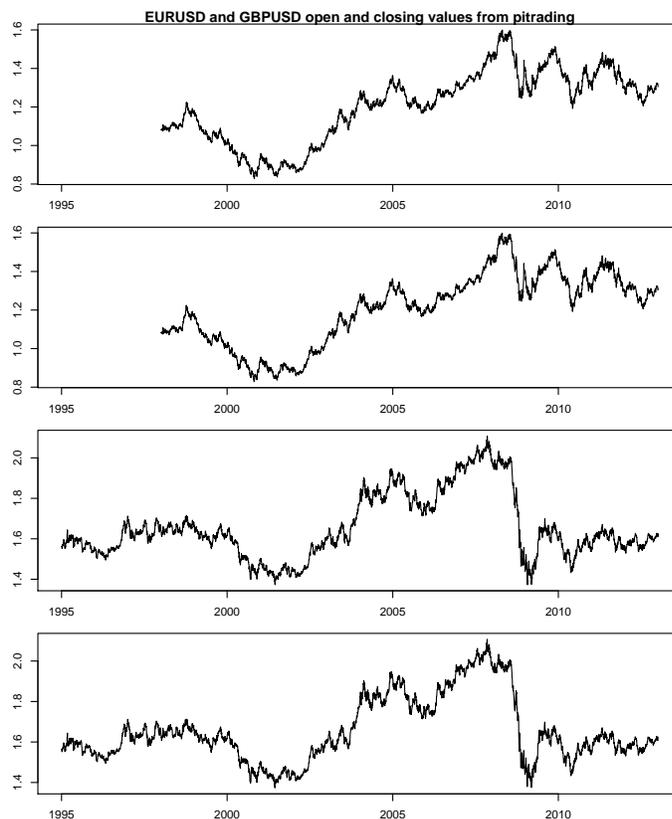
The following retrieves zipped files from [http://pittrading.com/free\\_market\\_data.htm](http://pittrading.com/free_market_data.htm) which provides some end of day data free of charge.

```
> library("TSzip")
> pitr <- TSconnect("zip", dbname="http://pittrading.com/free_eod_data")
> z <- TSget("INDU", pitr)
> tfplot(z, series=c(1,4))
```



The following illustrates returning an *xts* (Ryan and Ulrich, 2011) time series object.

```
> z <- TSget(c("EURUSD", "GBPUSD"), pitr, quote=c("Open","Close"),
  TSrepresentation=xts)
> tfplot(z,
  title="EURUSD and GBPUSD open and closing values from pitrading",
  start="1995-01-01",
  par=list(omi=c(0.1,0.3,0.1,0.1),mar=c(2.1,3.1,1.0,0.1)))
```



The default appearance of graphs can be changed (improved) by adjusting graphics device margins *omi* and *mar*. (They are set by the vector in order: bottom, left, top, right.) They can be set directly using *par()* or passed to *tfplot()* as in this example. The default behaviour of *tfplot()* is a compromise that usually works reasonably well for both screen and printed output. It is often useful to adjust these when generating *pdf* files for publication.

### 3 SQL Time Series Databases

This section gives several simple examples of putting series on and reading them from a database. (If a large number of series are to be loaded into a database, one would typically do this with a batch process using the database program's utilities for loading data.) The examples in this section will use *TSMYSQL* but, other than the initial connection, access will be similar for other SQL *TS\** packages. The syntax for connecting with other packages, and other options for connecting with *TSMYSQL*, are provided in Appendix "A".

The packages *TSPostgreSQL*, *TSMYSQL*, *TSSQLite*, and *TSodbc* use underlying packages *RPostgreSQL* (Conway et al., 2012), *RMySQL* (James and DebRoy, 2012), *RSQLite* (James, 2011), and *RODBC* (Ripley and from 1999 to

Oct 2002 Michael Lapsley, 2012). The *TS\** packages provide access to an SQL database with an underlying table structure that is set up to store time series data.

The next lines of code do some preliminary setup of the database. This uses the underlying database connection (*dbConnect*) rather than *TSconnect*, because *TSconnect* will not recognize the database until it has been setup. Function *removeTSdbTables()* is used first to remove any existing tables, which would cause *createTSdbTables()* to fail.

WARNING: running this will overwrite the “test” database on your server.

```
> library("TSMysql")
> con <- dbConnect("MySQL", dbname="test")
> require("TSsql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)
```

### 3.1 Writing to SQL Databases

This subsection illustrates writing some simple artificial data to a database, and reading it back. This part of the vignette is generated using *TSMysql*, but other backend SQL servers work in a similar way. See Appendix “A” for details of establishing other SQL database connections.

The first thing to do is to establish a *TSdbi* connection to the database:

```
> library("TSMysql")
> con <- TSconnect("MySQL", dbname="test")
```

*TSconnect* uses *dbConnect* from the *DBI* package, but checks that the database has expected tables, and checks for additional features. (It cannot be used before the tables are created, as was done above.)

The follow illustrates the use of the *TSdbi* interface, which is common to all extension packages.

This puts a series called *vec* on the database and then reads it back

```
> z <- ts(rnorm(10), start=c(1990,1), frequency=1)
> seriesNames(z) <- "vec"
> if(TSexists("vec", con)) TSdelete("vec", con)
> TSput(z, con)
> z <- TSget("vec", con)
```

Note that the series name(s) and not the *R* variable name (in this case, *vec* not *z*) are used on the database. If the retrieved series is printed it is seen to be a “ts” time series with some extra attributes.

*TSput* fails if the series already exists on the *con*, so the above example checks and deletes the series if it already exists. *TSreplace* does not fail if the

series does not yet exist, so examples below use it instead. *TSput*, *TSdelete*, *TSreplace*, and *TSexists* all return logical values *TRUE* or *FALSE*.

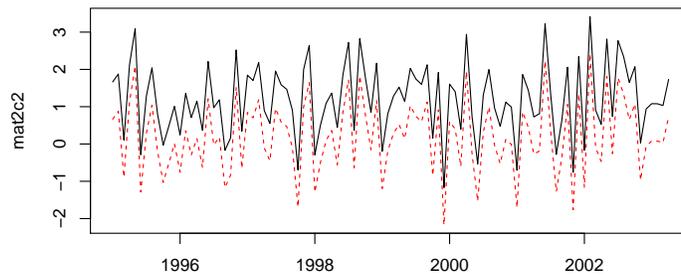
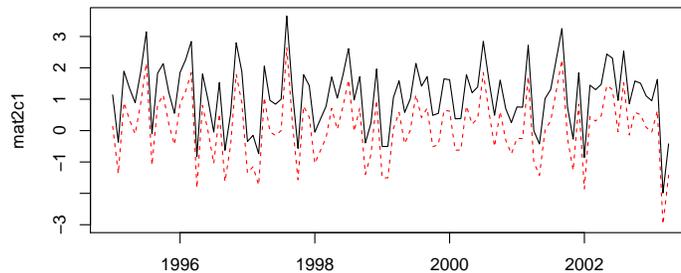
Several plots below show original data and the data retrieved after it is written to the database. In the plot below, one is added to the original data so that both lines are visible.

The *R* variable can contain multiple series of the same frequency. They are stored separately on the database.

```
> z <- ts(matrix(rnorm(200),100,2), start=c(1995,1), frequency=12)
> seriesNames(z) <- c("mat2c1", "mat2c2")
> TSreplace(z, con)

[1] TRUE

> tfplot(z+1, TSget(c("mat2c1","mat2c2"), con),
          lty=c("solid", "dashed"), col=c("black", "red"))
```



The following extract information about the series from the database, although not much information has been added for these examples.

```
> TSmeta("mat2c1", con)
```

```

serIDs: mat2c1
  from dbname test using TMySQLConnection
> TSmeta("vec", con)

```

```

serIDs: vec
  from dbname test using TMySQLConnection

```

```

> TSdates("vec", con)

```

```

      [,1]
[1,] "vec from 1990 1 to 1999 1 A      "

```

```

> TSdescription("vec", con)

```

```

[1] NA

```

```

> TSdoc("vec", con)

```

```

[1] NA

```

```

> TSlabell("vec", con)

```

```

[1] NA

```

Data documentation can be in three forms. A description specified by *TSdescription*, longer documentation specified by *TSdoc*, or a short label, typically useful on a graph, specified by *TSlabell*. These can be added to the time series object, in which case they will be written to the database when *TSput* or *TSreplace* is used to put the series on the database. Alternatively, they can be specified as arguments to *TSput* or *TSreplace*. The description or documentation will be retrieved as part of the series object with *TSget* only if this is specified with the logical arguments *TSdescription* and *TSdoc*. They can also be retrieved directly from the database with the functions *TSdescription* and *TSdoc*.

```

> z <- ts(matrix(rnorm(10),10,1), start=c(1990,1), frequency=1)
> TSreplace(z, serIDs="Series1", con)

```

```

[1] TRUE

```

```

> zz <- TSget("Series1", con)
> TSreplace(z, serIDs="Series1", con,
            TSdescription="short rnorm series",
            TSdoc="Series created as an example in the vignette.")

```

```

[1] TRUE

```

```

> zz <- TSget("Series1", con, TSdescription=TRUE, TSdoc=TRUE)
> start(zz)

```

```

[1] 1990    1
> end(zz)
[1] 1999    1
> TSdescription(zz)
[1] "short rnorm series"
> TSdoc(zz)
[1] "Series created as an example in the vignette."
> TSdescription("Series1", con)
[1] "short rnorm series"
> TSdoc("Series1", con)
[1] "Series created as an example in the vignette."

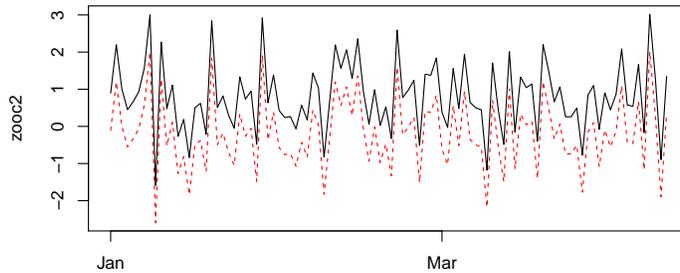
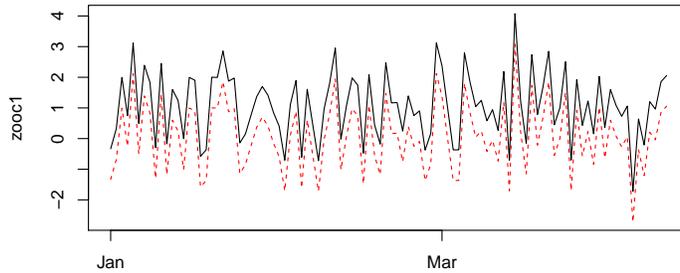
```

The following examples use dates and times which are not handled by *ts*, so the *zoo* time representation is used. It is necessary to specify the table where the data should be stored in cases where it is difficult to determine the periodicity of the data. See Appendix “B” for details of the specific tables.

```

> require("zoo")
> z <- zoo(matrix(rnorm(200),100,2), as.Date("1990-01-01") + 0:99)
> seriesNames(z) <- c("zooc1", "zooc2")
> TSreplace(z, con, Table="D")
[1] TRUE
> tfplot(z+1, TSget(c("zooc1","zooc2"), con),
         lty=c("solid", "dashed"), col=c("black", "red"))
>

```



```

> z <- zoo(matrix(rnorm(200),100,2), as.Date("1990-01-01") + 0:99 * 7)
> seriesNames(z) <- c("zooWc1", "zooWc2")
> TSreplace(z, con, Table="W")

[1] TRUE

> dbDisconnect(con)
> detach(package:TMySQL)
> detach(package:RMySQL)

```

## 4 More Examples

### 4.1 Examples Using SQL Databases

This section illustrates fetching data from the web and loading it into the database. This would be a very slow way to load a database, but provides examples of different kinds of time series data. The fetching is done with *TShistQuote*. Fetching data can fail due to lack of an Internet connection or delays, which will cause the generation of this vignette to fail.

This part of the vignette is generated using *TSPostgreSQL*, but other back-end SQL servers work in a similar way. See Appendix “A” for details of establishing other SQL database connections.

First establish a connection to the database where data will be saved:

```
> require("TSPostgreSQL")
> host <- Sys.getenv("POSTGRES_HOST")
> con <- TSconnect("PostgreSQL", dbname="test", host=host)
```

Now connect to the web server and fetch data:

```
> require("TShistQuote")
> yahoo <- TSconnect("histQuote", dbname="yahoo")
> x <- TSget("^gspc", quote = "Close", con=yahoo)
```

Then write the data to the local server, specifying table B for business day data (using *TSreplace* in case the series is already there from running this example previously):

```
> TSreplace(x, serIDs="gspc", Table="B", con=con)
```

```
[1] TRUE
```

and check the saved version:

```
> TSrefperiod(TSget(serIDs="gspc", con=con))
```

```
[1] "Close"
```

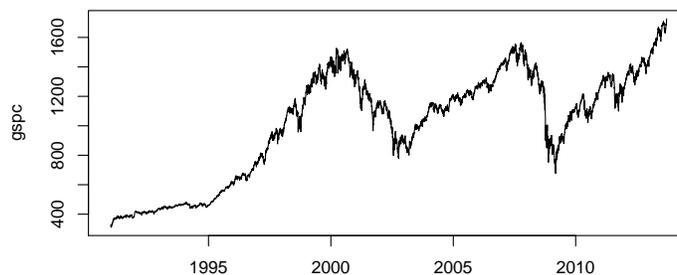
```
> TSdescription("gspc", con=con)
```

```
[1] "^gspc Close from yahoo"
```

```
> TSdoc("gspc", con=con)
```

```
[1] "^gspc Close from yahoo retrieved 2013-09-20 08:16:56"
```

```
> tfplot(TSget(serIDs="gspc", con=con))
```



```

> x <- TSget("ibm", quote = c("Close", "Vol"), con=yahoo)
> TSreplace(x, serIDs=c("ibm.Cl", "ibm.Vol"), con=con, Table="B",
  TSdescription.=c("IBM Close","IBM Volume"),
  TSdoc.= paste(c(
    "IBM Close retrieved on ",
    "IBM Volume retrieved on "), Sys.Date()))

[1] TRUE

> z <- TSget(serIDs=c("ibm.Cl", "ibm.Vol"),
  TSdescription=TRUE, TSdoc=TRUE, con=con)
> TSdescription(z)

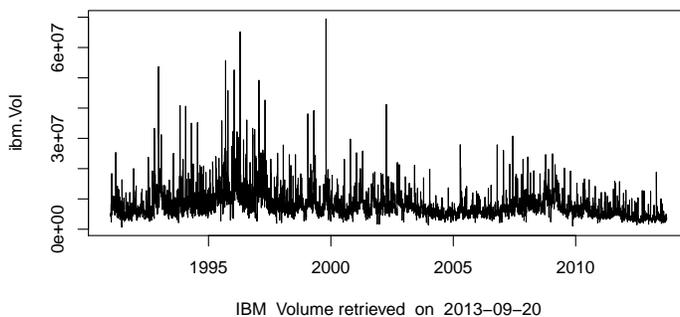
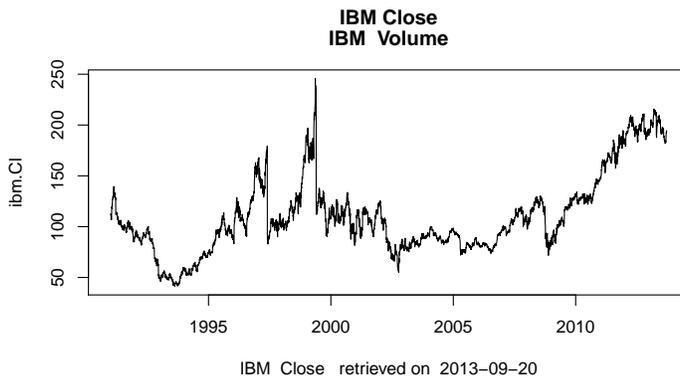
[1] "IBM Close" "IBM Volume"

> TSdoc(z)

[1] "IBM Close retrieved on 2013-09-20"
[2] "IBM Volume retrieved on 2013-09-20"

> tfplot(z, xlab = TSdoc(z), title = TSdescription(z))
> tfplot(z, title="IBM", start="2007-01-01")

```



Oanda has maximum of 500 days, so the start date is specified here so as to not exceed that.

```
> Oanda <- TSconnect("histQuote", dbname="oanda")
> x <- TSget("EUR/USD", start=Sys.Date() - 495, con=Oanda)
> TSreplace(x, serIDs="EUR/USD", Table="D", con=con)
```

```
[1] TRUE
```

Then check the saved version:

```
> z <- TSget(serIDs="EUR/USD", TSlabel=TRUE,
             TSdescription=TRUE, con=con)
> tfplot(z, title = TSdescription(z), ylab=TSlabel(z),
         start="2007-03-01")
```



```
> dbDisconnect(con)
> dbDisconnect(yahoo)
> dbDisconnect(Oanda)
> detach(package:TSPostgreSQL)
> detach(package:RPostgreSQL)
```

## 5 Comparing Time Series Databases

The purpose of package *TSCompare* is to compare pairs of series on two database. These series might have the same name, but for generality the main function, *TSCompare*, is set up to use name pairs. The pairs to compare are indicated by a matrix of strings with two columns. (It would also be possible to compare pairs on the same database, which might make sense if the names are different.)

The connections are established using other *TSdbi* packages such as *TSMYSQL*, *TSPostgreSQL*, etc. It will be necessary to establish two database connections, so it will also be necessary to load the database specific packages. Examples below use *TShistQuote*, *TSMYSQL* and *TSSQLite*.

```

> library("TScompare")
> library("TShistQuote")
> library("TSMYSQL")
> library("TSSQLite")

```

First setup database tables that are used by TSdbi using a *dbConnect* connection, after which a *TSconnect* connection can be used. This requires the package *TSsql* which does not need to be attached for most purposes, but is needed for the initial setup and for removing *TSdbi* database tables (which destroys the structure *TSconnect* expects). When this is re-done it insure the databases are empty:

```

> con <- dbConnect("MySQL", dbname="test")
> require("TSsql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)
> con <- dbConnect("SQLite", dbname="test")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)

```

Now TS connections to the databases are established.

```

> con1 <- TSconnect("MySQL", dbname="test")
> con2 <- TSconnect("SQLite", dbname="test")

```

Next a connection to yahoo is used to get some series and write them to the local test database. *TSreplace* is used because *TSput* will fail if the series already exists.

```

> yahoo <- TSconnect("histQuote", dbname="yahoo")
> x <- TSget("^ftse", yahoo)
> TSreplace(x, serIDs="ftse", Table="B", con=con1)

[1] TRUE

> TSreplace(x, serIDs="ftse", Table="B", con=con2)

[1] TRUE

> x <- TSget("^gspc", yahoo)
> TSreplace(x, serIDs="gspc", Table="B", con=con1)

[1] TRUE

> TSreplace(x, serIDs="gspc", Table="B", con=con2)

[1] TRUE

```

```

> x <- TSget("ibm", con=yahoo, quote = c("Close", "Vol"))
> TSreplace(x, serIDs=c("ibmClose", "ibmVol"), Table="B", con=con1)

[1] TRUE

> TSreplace(x, serIDs=c("ibmC", "ibmV"), Table="B", con=con2)

[1] TRUE

```

Now to do a comparison:

```

> ids <- AllIds(con1)
> ids

[1] "ftse" "gspc" "ibmClose" "ibmVol"

```

If the second database has the same names then ids can be made into a matrix with identical columns.

```

> ids <- cbind(ids, ids)
> eq <- TScompare(ids, con1, con2, na.rm=FALSE)
> summary(eq)

4 of 4 are available on con1.
2 of 4 are available on con2.
2 of 2 remaining have the same window.
2 of 2 remaining have the same window and values.

> eqrm <- TScompare(ids, con1, con2, na.rm=TRUE)
> summary(eqrm)

4 of 4 are available on con1.
2 of 4 are available on con2.
2 of 2 remaining have the same window.
2 of 2 remaining have the same window and values.

```

Since names are not identical the above indicates discrepancies, which are resolved by indicating the corresponding name pairs:

```

> ids <- matrix(c("ftse","gspc","ibmClose", "ibmVol",
                 "ftse","gspc","ibmC", "ibmV"),4,2)
> ids

      [,1]      [,2]
[1,] "ftse"    "ftse"
[2,] "gspc"    "gspc"
[3,] "ibmClose" "ibmC"
[4,] "ibmVol"  "ibmV"

```

```

> eq <- TScompare(ids, con1, con2, na.rm=FALSE)
> summary(eq)

4 of 4 are available on con1.
4 of 4 are available on con2.
4 of 4 remaining have the same window.
4 of 4 remaining have the same window and values.

> eqrm <- TScompare(ids, con1, con2, na.rm=TRUE)
> summary(eqrm)

4 of 4 are available on con1.
4 of 4 are available on con2.
4 of 4 remaining have the same window.
4 of 4 remaining have the same window and values.

```

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```

> dbDisconnect(con1)
> dbDisconnect(con2)
> dbDisconnect(yahoo)
> detach(package:TSMYSQL)
> detach(package:RMySQL)
> detach(package:TSSQLite)
> detach(package:RSQLite)

```

## 6 Vintages of Realtime Data

Examples in the section have been disabled pending availability of a different dataset.

Data vintages, or “realtime data” are snapshots of data that was available at different points in time. The most obvious feature of earlier snapshots is that the series end earlier. However, the reason for retaining vintages is that data is often revised, so, for some observations, earlier vintages have different data. Typically most revisions happen for the most recent periods, but this is often the data of most interest for forecasting and policy decisions. Thus, the revision records are valuable for understanding the implications of differences between early and revised releases of the data. A simple mechanism for accessing vintages of data is available in several *TS\** packages. This is illustrated here with an SQL database that has been set up with vintage support.

First establish a connection to the database and get a vector of the available vintages:

```

> require("TSMYSQL")
> require("tfplot")
> ets <- TSconnect("MySQL", dbname="etsv")
> v <- TSvintages(ets)

```

The following checks if a certain variable (Consumer Credit – V122707) is available in the different vintages. (V numbers replaced B numbers circa 2003, so the V numbers do not exist in older vintages. This could be supported by implementing aliases, but that has not been done here.)

```
> ve <- TSexists("V122707", vintage=v, con=ets)
> ve[224:length(ve)] <- FALSE
```

The vintages can then be retrieved and plotted by

```
> CC <- TSget(serIDs="V122707", con=ets, vintage=v[ve])
> tfOnePlot(ytoypc(CC), start=c(2000,1),
  ylab="Consumer Credit (V122707) y/y Growth",
  title=paste("Vintages", v[ve][1], "to", v[ve][189]),
  lastObs=TRUE, source="Source: Bank of Canada")
```

With package *TSfame* vintages are supported if the vintages are stored in files with names like “etsmfacansim\_20110513.db”. Then the vintages can be accessed as follows:

```
> dbs <- paste("ets /path/to/etsmfacansim_", c(
  "20110513.db", "20060526.db", "20110520.db"), sep="")
> names(dbs) <- c("2011-05-13", "2006-05-26", "2011-05-20")
> conetsV <- TSconnect("fame", dbname=dbs, "read", current="2011-05-13")
> z <- TSget("V122646", con=conetsV, vintage=c("2011-05-13", "2006-05-26"))
> dbDisconnect(conetsV)
```

(The above example should work, but beware that I am no longer testing it because I no longer have Fame access.)

The package *googleVis* can be used to produce a plot that is very useful for examining vintage more closely, and finding outliers and other data problems. The names are used in the legend of this next plot, so the series names are specified in the argument to *ytoypc*. (Otherwise they get reset to indicate the year-to-year calculation, which makes the legend messy to read.)

```
> require("googleVis")
> tfVisPlot(ytoypc(CC, names=seriesNames(CC)), start=c(2006,1),
  options=list(title="Vintages of Consumer Credit (V122707) y/y Growth"))
```

This will produce a graph in your web browser. It is not reproduced here. (And beware that it may not be very fast.) Pointing your mouse at the legend of this plot will highlight the corresponding vintage, and pointing at the graph will give information about the source of a data point.

```
> dbDisconnect(ets)
> detach(package:TMySQL)
> detach(package:RMySQL)
```

## 7 Just Want Data in a xls or csv File

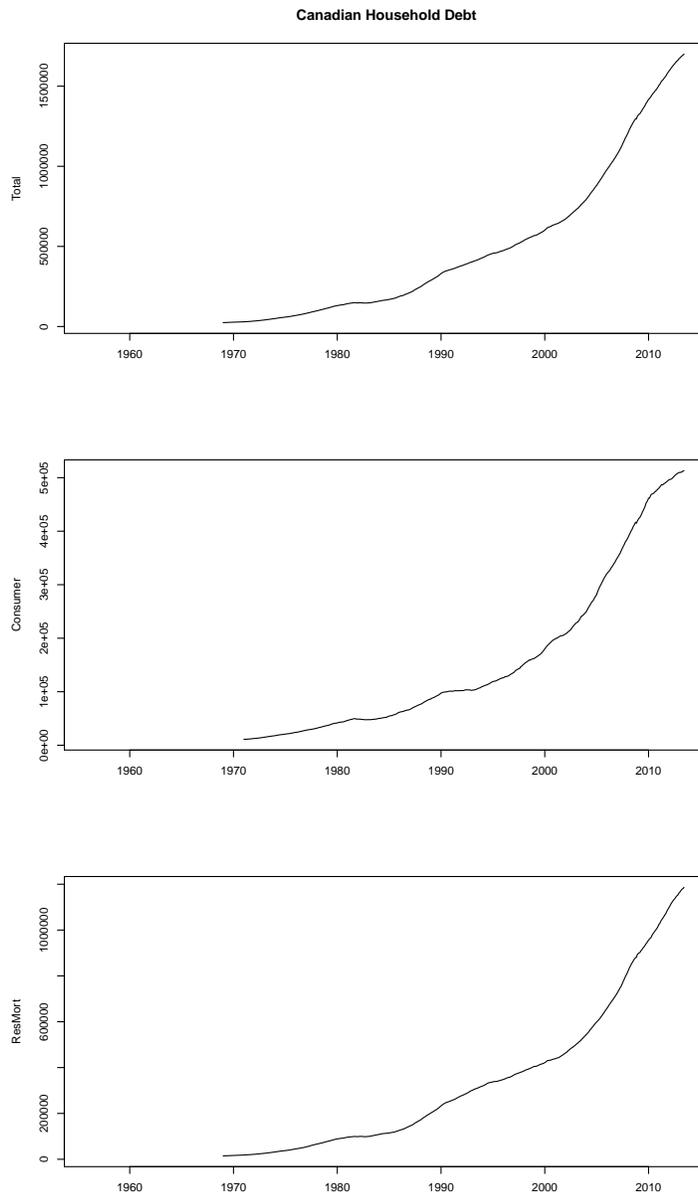
Occasionally one may need to get data into a program other than *R*. Or, perhaps you have a friend that does not want to use *R* but would like easy access to data. This section describes two utilities for putting time series data obtained with *TSdbi* connections into .xls and .csv files. Since it is often straight forward to download an xls or csv file from a web site in the first place, the advantage of this will usually be only in the case where you need to repeatedly get the data. In that case, the process can be automated with these tools.

The starting point will be to get some data. Previous sections illustrate several possibilities. For example purposes here use

```
> require("TSjson")
> cansim <- TSconnect("json", dbname="cansim")
> z <- TSget(c("V36415", "V122707", "V122746"), con=cansim)
> seriesNames(z) <- c("Total", "Consumer", "ResMort")
```

While this is optional, it might be useful to look at the data to see if it is really what you expect:

```
> library("tframePlus")
> library("tfplot")
> tfplot(z, title="Canadian Household Debt")
```



In this case the dates of one series start in 1956, but data is NA until about 1970. It is possible to trim the NA data from the beginning with

```
> z <- trimNA(z, endNAs=FALSE)
```

By default, *trimNA* will trim all series if any one of them is NA, so trimming the end will not be what you want if you are interested in the most recent data.

Setting `endNAs=FALSE`) overrides the default.

It is also possible to trim the data to a specified subsample with

```
> z <- tfwindow(z, start=c(1995,1))
```

Now write the data to an .xls file using `TWriteXLS` which uses `WriteXLS` (Schwartz, 2012) but automatically adds time information:

```
> library("WriteXLS")
> tofile <- tempfile(fileext = ".xls")
> TWriteXLS(z, FileName=tofile)
> unlink(tofile)
```

In this case a temporary file is written, then removed with `unlink`, so that scratch files do not accumulate from building this vignette, but you would typically use a more meaningful name, and not remove the file. It is also possible to write multiple sheets in the file. For more details on this see the help

```
> ?TWriteXLS
```

The dates are written in some different formats in the .xls file, since this is often convenient for calculations and graphing in a spreadsheet program. Beware that .xls files have size limitations that you might encounter if you try to put large amounts of data into the file. Consider using .csv files in that case.

To write the data to a .csv file

```
> tofile <- tempfile(fileext = ".csv")
> TWriteCSV(z, FileName=tofile)
> unlink(tofile)
```

## 8 Appendix A: Connection Specific Details

This appendix provides details of the different connections which are specific to individual packages and backend databases. In order to make the examples complete, for the SQL versions, test databases are first created with the tables expected by the `TS*` packages. Note that this is done with a `dbConnect` connection rather than a `TConnect` connection, because `TConnect` expects the tables to exist already.

WARNING: running these example will overwrite tables in the “test” database on the server.

The database setup might typically be done by an administrator, rather than by an end user. Here it is done using a function `createTSdbTables` in the `TSsql` package. The instructions for building the database tables can be seen by examining that function. The instruction could be used to build the database using database utilities rather than `R`, which might be the way a system administrator would build the database.

In many cases there are two or more ways to pass information like the *username*, *password*, and *server* or *host*. One mechanism is that this information is specified in a configuration file in the user's home directory. The database driver then reads this information and it is not part of the user's *R* session. (Often this is considered the most secure way.) Another way is that environment variables are set, and the database driver uses these. Again, this is not part of the user's *R* session. Still another way is that the user passes this information in the call to *TSConnect* in their *R* session. In this case the character strings are visible in the *R* session, and possibly recorded in the user's *R* scripts, thus this is typically not considered to be very secure. A modification, which is only a little bit better, is for the user's *R* scripts to read the information from environment variables using, for example:

```
> user <- Sys.getenv("MYSQL_USER")
```

## 8.1 TSMYSQL Connection Details

The MySQL user, password, and hostname should be set in MySQL client configuration file (.my.cnf) in the user's home directory before starting *R*. Alternatively, this information can be set with environment variables `MYSQL_USER`, `MYSQL_PASSWD` and `MYSQL_HOST`. (An environment variable `MYSQL_DATABASE` can also be set, but "test" is specified below.) Below the configuration file is used.

The next small section of code uses *dbConnect* to set up database tables that expected by *TSConnect*.

```
> library("TSMYSQL")
> con <- dbConnect("MySQL", dbname="test")
> require("TSsql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)
```

Now a *TSdbi* connection to the database is established.

```
> con <- TSConnect("MySQL", dbname="test")
```

The alternative to pass the user/password information in the arguments to the connection function would be:

```
> con <- TSConnect("MySQL", dbname="test", username=user, password=passwd, host=host)
```

This may be cumbersome to change, and is generally considered to be less secure.

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```
> detach(package:TSMYSQL)
> detach(package:RMySQL)
```

## 8.2 TSPostgreSQL Connection Details

The PostgreSQL user, and password, can be set in PostgreSQL configuration file (.pgpass in Linux) in the user's home directory before starting R. The PostgreSQL documentation suggests that it should be possible to get the host from the .pgpass file too, but I have not been able to make that work. The PostgreSQL alternative to the configuration file is to use environment variables PGDATABASE, PGHOST, PGPORT, and PGUSER. This package supports another alternative to set this information with environment variables POSTGRES\_USER, POSTGRES\_PASSWD and POSTGRES\_HOST, which are read in the R code. (An environment variable POSTGRES\_DATABASE can also be set, but "test" is specified below.) Below, the environment variable POSTGRES\_HOST is used to determine the host server, but the .pgpass file is used for the user and password information.

```
> host <- Sys.getenv("POSTGRES_HOST")
```

The next small section of code uses *dbConnect* to set up database tables that expected by *TScnect*.

```
> library("TSPostgreSQL")
> con <- dbConnect("PostgreSQL", dbname="test", host=host)
> require("TSql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)
```

Now a TSdbi connection to the database is established.

```
> con <- TScnect("PostgreSQL", dbname="test", host=host)
```

Another alternative is to pass the user/password information in the arguments to the connection function:

```
> con <- TScnect("PostgreSQL", dbname="test", user=user, password=passwd, host=host)
```

This is may be cumbersome to change, and is generally considered to be less secure.

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```
> detach(package:TSPostgreSQL)
> detach(package:RPostgreSQL)
```

## 8.3 TSSQLite Connection Details

In SQLite there does not seem to be any need to set user or password information, and examples here all use the localhost.

Now setup database tables that are used by TSdbi using a *dbConnect* connection, after which a *TScnect* connection can be used:

```

> library("TSSQLite")
> con <- dbConnect("SQLite", dbname="test")
> require("TSql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)

```

Now a TSdbi connection to the database is established.

```

> con <- TSconnect("SQLite", dbname="test")

```

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```

> detach(package:TSSQLite)
> detach(package:RSQLite)

```

## 8.4 TSodbc Connection Details

The ODBC user, password, hostname, etc, should be set in ODBC client configuration file in the user's home directory (.odbc.ini in Linux) before starting R. An example of this file is provided below. It will also be necessary to have the appropriate driver installed on the system (Postgresql in the example below). Alternatively, it should be possible to set this information with environment variables ODBC\_USER, ODBC\_PASSWD and ODBC\_DATABASE. However, the variable ODBC\_HOST does not seem to work for passing the ODBC connection, so a properly setup ODBC configuration file is needed. Because of this, the environment variable mechanism is not currently supported in *TSodbc* and the user, passwd, and host settings should be done in the configuration file.

Now setup database tables that are used by TSdbi using a *odbcConnect* connection, after which a *TSconnect* connection can be used:

```

> library("TSodbc")
> con <- odbcConnect(dsn="test")
> if(con == -1) stop("error establishing ODBC connection.")
> require("TSql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> odbcClose(channel=con)

```

Now a TSdbi connection to the database is established.

```

> con <- TSconnect("ODBC", dbname="test")

```

Another alternative is to pass the user/password information in the arguments to the connection function:

```

> con <- TSconnect("ODBC", dbname="test", uid=user, pwd=passwd)

```

This is may be cumbersome to change, and is generally considered to be less secure.

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```
> detach(package:TSodbc)
> detach(package:RODBC)
```

#### 8.4.1 Example ODBC configuration file

Following is an example ODBC configuration file I use in Linux (so the file is in my home directory and called ".odbc.ini") to connect to a remote PostgreSQL server:

```
[test]

Description      = test DB (Postgresql)
Driver           = Postgresql
Trace           = No
TraceFile        = /tmp/test_odbc.log
Database         = test
Servername       = some.host
Username         = paul
Password         = mySecret
Port            = 5432
Protocol         = 6.4
ReadOnly         = No
RowVersioning    = No
ShowSystemTables = No
ShowOidColumn    = No
FakeOidIndex     = No
ConnSettings     =
```

```
[ets]

Description      = ets DB (Postgresql)
Driver           = Postgresql
Trace           = No
TraceFile        = /tmp/test_odbc.log
Database         = ets
Servername       = some.host
Username         = paul
Password         = mySecret
Port            = 5432
Protocol         = 6.4
ReadOnly         = No
```

```

RowVersioning          = No
ShowSystemTables       = No
ShowOidColumn          = No
FakeOidIndex           = No
ConnSettings           =

```

The above depends on the driver tag “Postgresql” being defined in the file `/etc/odbcinst.ini`, to give the actual driver file location. That file might have something like

```

[PostgreSQL]
Description = PostgreSQL ODBC driver (Unicode version)
Driver      = /usr/lib/x86_64-linux-gnu/odbc/psqlodbcw.so
Setup       = /usr/lib/x86_64-linux-gnu/odbc/libodbcpsqlS.so
Debug      = 0
CommLog     = 1
UsageCount  = 1

```

## 8.5 TSOacle Connection Details

This package is available on R-forge, but is not being tested, because I do not currently have a server to test it. The code in this section of the vignette is not being run. Please contact the package maintainer (Paul Gilbert) if you have an Oracle server and are willing to test the package.

The Oracle user, password, and hostname should be set in Oracle client configuration file (`tnsnames.ora`) before starting R.

The next small section of code uses `dbConnect` to set up database tables that expected by `TScnect`.

```

> library("TSOracle")
> con <- dbConnect("Oracle", dbname="test")
> require("TSsql")
> removeTSdbTables(con, yesIknowWhatIamDoing=TRUE)
> createTSdbTables(con, index=FALSE)
> dbDisconnect(con)

```

Now a TSdbi connection to the database is established.

```

> con <- TScnect("Oracle", dbname="test")

```

While it may not be necessary to `detach` packages, the following prevents warnings later about objects being masked:

```

> detach(package:TSOracle)
> detach(package:ROracle)

```

## 8.6 TSjson Connection Details

The *TSjson* method *TSconnect* can establish a connection to a proxy server. (See the main text for directly connecting to the web data source.)

```
> library("TSjson")
> con <- TSconnect("TSjson", dbname="proxy-cansim")
```

The *dbname* specifies the proxy server, for which credentials will be needed. The *user*, *password*, and *host*, can be specified as arguments. If specified as *NULL* (the default) then they will be determined by reading a file `~/.TSjson.cfg` which should have a line with four fields:

```
[proxy-cansim] user password host
```

The first field should match the *dbname* specification. Currently only a single line is supported, starting with "[proxy-cansim]", but the format is intended for extension to support proxies to different web databases.

If the file does not exist then environment variables "TSJSONUSER", "TSJSONPASSWORD", and "TSJSONHOST" will be used.

```
> detach(package:TSjson)
> detach(package:RJSONIO)
```

## 8.7 TSfame Connection Details

I no longer have access to Fame so package *TSfame* is no longer being extensively tested. (It has previously worked.) The code in this section of the vignette is not being run. Please contact the package maintainer (Paul Gilbert) if you have Fame and are willing to test the package.

Beware that the package *fame* may be installed but not functional because the Fame HLI code is not available. A warning will be issued in this case.

Two variants of the Fame connect are available. The first requires a Fame database available on the local system:

```
> con <- TSconnect("fame", dbname="testFame.db")
```

The second requires a Fame server:

```
> con <- TSconnect("fame", dbname="ets /path/to/etsmfacansim.db", "read")
```

where the characters before the space in the *dbname* string indicate the network name of the server, and the path after the string indicates where the server should find the database.

While it may not be necessary to *detach* packages, the following prevents warnings later about objects being masked:

```
> detach(package:TSfame)
> detach(package:fame)
```

Table 1: Data Tables

Table	Contents
Meta	meta data and index to series data tables
A	annual data
Q	quarterly data
M	monthly data
S	semiannual data
W	weekly data
D	daily data
B	business data
U	minutely data
I	irregular data with a date
T	irregular data with a date and time

## 9 Appendix B: Underlying SQL Database Structure and Loading Data

More detailed description of the instructions for building the database tables is given in the vignette for the *TSdbi* package. Those instruction show how to build the database using database utilites rather than R, which might be the way a system administrator would build the database.

The database tables are shown in the Table below. The *Meta* table is used for storing meta data about series, such as a description and longer documentation, and also includes an indication of what table the series data is stored in. To retrieve series it is not necessary to know which table the series is in, since this can be found on the *Meta* table. Putting data on the database may require specifying the table, if it cannot be determined from the R representation of the series.

In addition, there will be tables "vintages" and "panels" if those features are used.

The following is done with `dbConnect` in place of a `TSconnect`, since they are direct SQL queries and do not require the *TSdbi* structure.

The structure reported reflects the setup that was done previously. These queries are Mysql specific but below is a generic SQL way to do this.

```
> library("TSMYSQL")
> con <- dbConnect("MySQL", dbname="test")
> dbListTables(con)

[1] "A"    "B"    "D"    "I"    "M"    "Meta" "Q"    "S"    "T"    "U"
[11] "W"

> dbGetQuery(con, "show tables;")
```

```

Tables_in_test
1          A
2          B
3          D
4          I
5          M
6      Meta
7          Q
8          S
9          T
10         U
11         W

```

```
> dbGetQuery(con, "describe A;")
```

```

Field      Type Null Key Default Extra
1  id varchar(40) YES      <NA>
2  year  int(11) YES      <NA>
3  v     double YES      <NA>

```

```
> dbGetQuery(con, "describe B;")
```

```

Field      Type Null Key Default Extra
1  id varchar(40) YES      <NA>
2  date  date  YES      <NA>
3  period int(11) YES      <NA>
4  v     double YES      <NA>

```

```
> dbGetQuery(con, "describe D;")
```

```

Field      Type Null Key Default Extra
1  id varchar(40) YES      <NA>
2  date  date  YES      <NA>
3  period int(11) YES      <NA>
4  v     double YES      <NA>

```

```
> dbGetQuery(con, "describe M;")
```

```

Field      Type Null Key Default Extra
1  id varchar(40) YES      <NA>
2  year  int(11) YES      <NA>
3  period int(11) YES      <NA>
4  v     double YES      <NA>

```

```
> dbGetQuery(con, "describe Meta;")
```

```

Field      Type Null Key Default Extra
1  id varchar(40) NO PRI <NA>

```

```

2      tbl      char(1) YES MUL  <NA>
3  refperiod varchar(10) YES    <NA>
4  description      text YES    <NA>
5  documentation    text YES    <NA>

```

```
> dbGetQuery(con, "describe U;")
```

	Field	Type	Null	Key	Default	Extra
1	id	varchar(40)	YES		<NA>	
2	date	timestamp	NO		CURRENT_TIMESTAMP	on update CURRENT_TIMESTAMP
3	tz	varchar(4)	YES		<NA>	
4	period	int(11)	YES		<NA>	
5	v	double	YES		<NA>	

```
> dbGetQuery(con, "describe Q;")
```

	Field	Type	Null	Key	Default	Extra
1	id	varchar(40)	YES		<NA>	
2	year	int(11)	YES		<NA>	
3	period	int(11)	YES		<NA>	
4	v	double	YES		<NA>	

```
> dbGetQuery(con, "describe S;")
```

	Field	Type	Null	Key	Default	Extra
1	id	varchar(40)	YES		<NA>	
2	year	int(11)	YES		<NA>	
3	period	int(11)	YES		<NA>	
4	v	double	YES		<NA>	

```
> dbGetQuery(con, "describe W;")
```

	Field	Type	Null	Key	Default	Extra
1	id	varchar(40)	YES		<NA>	
2	date	date	YES		<NA>	
3	period	int(11)	YES		<NA>	
4	v	double	YES		<NA>	

If schema queries are supported then table information can be obtained in a (almost) generic SQL way. On some systems this will fail because users do not have read privileges on the INFORMATION\_SCHEMA table. This does not seem to be an issue in SQLite, but SQLite schema queries are not the same as for other SQL engines.

```
> dbGetQuery(con, paste(
  "SELECT COLUMN_NAME FROM INFORMATION_SCHEMA.Columns ",
  " WHERE TABLE_SCHEMA='test' AND table_name='A' ;"))
```

```

COLUMN_NAME
1      id
2     year
3      v

> dbGetQuery(con, paste(
  "SELECT COLUMN_NAME, COLUMN_DEFAULT, COLLATION_NAME, DATA_TYPE,",
  "CHARACTER_SET_NAME, CHARACTER_MAXIMUM_LENGTH, NUMERIC_PRECISION",
  "FROM INFORMATION_SCHEMA.Columns WHERE TABLE_SCHEMA='test' AND table_name='A' ;"))

COLUMN_NAME COLUMN_DEFAULT COLLATION_NAME DATA_TYPE CHARACTER_SET_NAME
1      id      <NA> latin1_swedish_ci  varchar          latin1
2     year      <NA>              <NA>          int              <NA>
3      v      <NA>              <NA>          double           <NA>
CHARACTER_MAXIMUM_LENGTH NUMERIC_PRECISION
1                40              NA
2                 NA              10
3                 NA              22

> dbGetQuery(con, paste(
  "SELECT COLUMN_NAME, DATA_TYPE, CHARACTER_MAXIMUM_LENGTH, NUMERIC_PRECISION",
  "FROM INFORMATION_SCHEMA.Columns WHERE TABLE_SCHEMA='test' AND table_name='M' ;"))

COLUMN_NAME DATA_TYPE CHARACTER_MAXIMUM_LENGTH NUMERIC_PRECISION
1      id  varchar                40              NA
2     year    int                  NA              10
3    period    int                  NA              10
4      v    double                NA              22

> dbDisconnect(con)

[1] TRUE

```

## 10 Appendix C: Examples Using DBI and direct SQL Queries

The following examples are queries using the underlying "DBI" functions. They should not often be needed to access time series, but may be useful to get at more detailed information, or formulate special queries. Typically these queries may be more useful for systems administrators doing database maintenance than they are for end users.

These queries depend on the underlying structure of the database, which should be considered "opaque" from the perspective of a TSdbi user. That is, this structure could be changed without affecting the TSdbi functionality, but the following queries would be affected.

```

> library("TSMYSQL")
> con <- TSconnect("MySQL", dbname="test")
> dbGetQuery(con, "SELECT count(*) FROM Meta ;")

count(*)
1      0

> dbGetQuery(con, "SELECT max(year) FROM A ;")

max(year)
1      NA

Finally, to disconnect gracefully, one should
> dbDisconnect(con)

```

## 11 Appendix D: TSjson README regarding Python details

Package TSjson needs Python 2 and Python modules sys, json, mechanize, re, csv, and urllib2. The package has been tested with python 2.7.3 on Ubuntu Linux and Windows XP. There is no obvious reason why it should not work on other systems. (Please advise the package maintainer, Paul Gilbert <pgilbert.ttv9z@ncf.ca> if you discover differently.) The python code is fairly simple and may work in Python 3 versions but module mechanize, which does the main part that cannot be done easily in R, is not available for Python 3. (Also, module urllib2 is split into urllib.request and urllib.error in Python 3.)

Python also needs to be on the path so that it can be found when run from the R process. Some brief instructions are provided below, but installing programs will be operating system dependent, so these are not comprehensive.

Instructions for installing python modules are further below.

Probably not necessary for using this package, but for those interested, additional information and tutorials on python are available at <https://wiki.python.org/moin/FrontPage>

===== Windows =====

On Windows, python can be installed by downloading and following intructions at <http://www.python.org/getit/>

Python also needs to be on the search PATH. Setting the PATH will be

slightly different on different versions of Windows. (See, for example, <http://www.computerhope.com/issues/ch000549.htm> )

The steps will be roughly:

- From Desktop or Start Menu, right-click My Computer and then Properties.
- In the System Properties window, click the Advanced tab.
- In the Advanced section, click the Environment Variables button.
- In the Environment Variables window, highlight the Path variable in the Systems Variable section and click the Edit button. Modify the path to indicate the location where python is installed. (Directories in the Path are separated with a semicolon.)

There should typically be a part of this environment variable string that is something like `C:\Python27;` but the exact string will depend on the version and where it has been installed.

You can check that it is being found and the version by executing

```
python --version
```

at a Command Prompt. (Be sure to open this window after you set the path as above.)

===== Linux =====

Python is usually already installed on Linux systems. (Your system is likely badly broken if it is not.) You can check the version by executing

```
python --version
```

in a shell. If the command is not found then you need to ensure that python is on your PATH. If it is not installed then the install can be done with the usual system utilities. For example, on Debian based systems

```
sudo apt-get python
```

or you can install it from <http://www.python.org/getit/>. (But it really is unlikely that you will need to install python. Also, I do not think that it should be necessary to upgrade or change the version of Python to use the package, and I do not recommend that, because too many other things on your system depend on python.)

===== Python modules =====

Modules `sys`, `re`, `csv`, `json`, and `urllib2` are provided with the Python Standard

Library so they will usually not need to be installed. Module mechanize will usually need to be installed.

You can check if python modules are already installed by starting python in a shell or in Windows at the Command Prompt:

```
python
```

and then at the python >>> prompt try to import the modules. (Python newbies beware that indentation is part of the python syntax and you should not put spaces at the beginning of the command.)

```
>>> import sys, json, re, csv, urllib2
>>> import mechanize
```

```
then
```

```
>>> quit()
```

to exit python.

Installing modules can be done in a number of different ways, for example, using apt-get or Synaptic in some versions of Linux. The use of the python program easy\_install is described briefly below. Once the program easy\_install is installed (see <https://wiki.python.org/moin/EasyInstall>, or in debian based systems sudo apt-get install python-setuptools), the process for installing python modules is similar on different operating systems.

The easy\_install program also needs to be found on your path, so add its location in the same way that you added python to your path above. (In Windows, easy\_install.exe might be installed in C:\Python27\Scripts, in which case that should be on your path.)

Then (in non-Windows systems you may need to prefix this with sudo)

```
easy_install mechanize
```

```
===== Other data sites =====
```

With python and modules installed you should be able to access data from Statistics Canada directly from you R session. The python code that makes this work is distributed with the source TSjson package in the file exec/cansimGet.py. That code can also be executed directly as a python program.

The main reason for using python rather than doing this directly in R is that

the initial query to the web site returns a link to a dynamically generated web page, which must be accessed in a second step. The python module mechanize provides a mechanism to do this, whereas there is currently no easy mechanism in R.

Other sites will be added over time, and you can help by sending the package maintainer the details of accesss to other sites that are of interest to you.

## References

- Conway, J., Eddelbuettel, D., Nishiyama, T., Prayaga, S.~K., and Tiffin, N. (2012). *RPostgreSQL: R interface to the PostgreSQL database system*. R package version 0.3-2.
- Couture-Beil, A. (2013). *rjson: JSON for R*. R package version 0.2.13.
- James, D.~A. (2011). *RSQLite: SQLite interface for R*. R package version 0.11.1.
- James, D.~A. and DebRoy, S. (2012). *RMySQL: R interface to the MySQL database*. R package version 0.9-3.
- Lang, D.~T. (2012). *RJSONIO: Serialize R objects to JSON, JavaScript Object Notation*. R package version 1.0-1.
- R Core Team (2012). *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing, Vienna, Austria. ISBN 3-900051-07-0.
- R Core Team (2013). *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing, Vienna, Austria.
- R Special Interest Group on Databases (2009). *DBI: R Database Interface*. R package version 0.2-5.
- Ripley, B. and from 1999~to Oct 2002 Michael~Lapsley (2012). *RODBC: ODBC Database Access*. R package version 1.3-6.
- Ryan, J.~A. (2011). *quantmod: Quantitative Financial Modelling Framework*. R package version 0.3-17.
- Ryan, J.~A. and Ulrich, J.~M. (2011). *xts: eXtensible Time Series*. R package version 0.8-2.
- Schwartz, M. (2012). *WriteXLS: Cross-platform Perl based R function to create Excel 2003 (XLS) files*. R package version 2.1.1.
- Trapletti, A. and Hornik, K. (2012). *tseries: Time Series Analysis and Computational Finance*. R package version 0.10-28.
- Warnes, G.~R., with contributions~from Ben~Bolker, Gorjanc, G., Grothendieck, G., Korosec, A., Lumley, T., MacQueen, D., Magnusson, A., Rogers, J., and others (2011). *gdata: Various R programming tools for data manipulation*. R package version 2.8.2.

Zeileis, A. and Grothendieck, G. (2005). zoo: S3 infrastructure for regular and irregular time series. *Journal of Statistical Software*, 14(6):1-27.