find-packages

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Creating a task view for physical activity researchers

First step is use the cranly package to see what is out there. I found this tutorial to help https://rviews.rstudio.com/2018/05/31/exploring-r-packages/. Now I have a tidy dataframe with all the packages.

```
library(cranly)
package_db <- clean_CRAN_db(tools::CRAN_package_db())</pre>
```

Search dataframe for physical activity related packages

I have used the search terms:

- physical activity
- actigraph
- fitbit
- garmin
- geneactiv
- apple health
- google fit
- samsung health

I had originally included the term exercise but that search included many packages that were not related to physical activity. I also did the same search in title but that did not help and only returned 1 result that was already captured in the description search.

```
library(stringr)
library(dplyr)
##
## Attaching package: 'dplyr'
  The following objects are masked from 'package:stats':
##
##
       filter, lag
  The following objects are masked from 'package:base':
##
##
##
       intersect, setdiff, setequal, union
library(kableExtra)
##
## Attaching package: 'kableExtra'
## The following object is masked from 'package:dplyr':
##
       group_rows
##
pa_search <- c("physical activity", "actigraph", "Actigraph", "fitbit", "garmin", "geneactiv", "apple h
               "exercise", "omron", "actilife", "cycling", "swimming", "activity counts", "sedentary",
```

```
#pa_search <- c("polar")
package_db$pa <- str_extract_all(package_db$description, paste(pa_search, collapse="|"))
pa_packages <- filter(package_db, pa != "character(0)")</pre>
```

From this search there are 10 packages are designed to analyse some type of physical activity data.

```
pa_pkg_names <- pa_packages$package</pre>
```

Packages' description

```
pa_packages %>% select(package, description) %>%
  kable() %>%
  kable_styling("striped", full_width = T) %>%
  column_spec(2, width = "30em")
```

description
Bayesian network analysis is a form of probabilistic
graphical models which derives from empirical data
a directed acyclic graph, DAG, describing the
dependency structure between random variables.
An additive Bayesian network model consists of a
form of a DAG where each node comprises a
generalized linear model, GLM. Additive Bayesian
network models are equivalent to Bayesian
multivariate regression using graphical modelling,
they generalises the usual multivariable regression,
GLM, to multiple dependent variables. 'abn'
provides routines to help determine optimal
Bayesian network models for a given data set,
where these models are used to identify statistical
dependencies in messy, complex data. The additive
formulation of these models is equivalent to
multivariate generalised linear modelling (including
mixed models with iid random effects). The usual
term to describe this model selection process is structure discovery. The core functionality is
concerned with model selection - determining the
most robust empirical model of data from
interdependent variables. Laplace approximations
are used to estimate goodness of fit metrics and
model parameters, and wrappers are also included
to the INLA package which can be obtained from
http://www.r-inla.org . A comprehensive set of
documented case studies, numerical
accuracy/quality assurance exercises, and
additional documentation are available from the
'abn' website http://r-bayesian-networks.org .
Recent studies haven shown that, on top of total
daily active/sedentary volumes, the time
accumulation strategies provide more sensitive
information. This package provides functions to
extract commonly used fragmentation metrics to
quantify such time accumulation strategies based
on minute level actigraphy-measured activity
counts data. Functional linear modeling and analysis for
actigraphy data.
ActiLife software generates activity counts from
data collected by Actigraph accelerometers
https://s3.amazonaws.com/actigraphcorp.com/wp
content/uploads/2017/11/26205758/ActiGraph-
White-Paper_What-is-a-Countpdf>. Actigraph
is one of the most common research-grade
accelerometers. There is considerable research
validating and developing algorithms for human
activity using ActiLife counts. Unfortunately,
ActiLife counts are proprietary and difficult to
implement if researchers use different accelerometer
brands. The code creates ActiLife counts from raw
acceleration data for different accelerometer brands
and it is developed based on the study done by
3 Brond and others (2017)
<pre><doi:10.1249 mss.00000000001344="">.</doi:10.1249></pre>
Contains functions to generate pre-defined
summary statistics from activPAL events files