# Package 'skm' 

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## Type Package <br> Title Selective k-Means

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Description Algorithms for solving selective k-means problem, which is defined as finding k rows in an $\mathrm{m} \times \mathrm{n}$ matrix such that the sum of each column minimal is minimized.
In the scenario when $\mathrm{m}=\mathrm{n}$ and each cell value in matrix is a valid distance metric, this is equivalent to a k -means problem. The selective k-means extends the k-means problem in the sense that it is possible to have $\mathrm{m}!=\mathrm{n}$, often the case $\mathrm{m}<\mathrm{n}$ which implies the search is limited within a small subset of rows. Also, the selective k-means extends the k-means problem in the sense that the instance in row set can be instance not seen in the column set, e.g., select 2 from 3 internet service provider (row) for 5 houses (column) such that minimize the overall cost (cell value) - overall cost is the sum of the column minimal of the selected 2 service provider.
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URL http://github.com/gyang274/skm
BugReports http://github.com/gyang274/skm/issues
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col_max_idx ..... col_max_idx

## Description

calculate colvec max value index within limited range

## Usage

col_max_idx(u, wlmt)

## Arguments

u
wlmt
u : a numeric colvec
wlmt: limit search on colvec on indices within wlmt

## Value

id an index of max value in $u$ within wlmt w.r.t to original index

## Note

cpp use index start from 0 vs $r$ use index start from 1
in case of equal std:min/std:max take first index seen

## See Also

Other matrix_minmax: col_max_val, col_min_idx, col_min_val, col_rgn_val

```
col_max_val col_max_val
```


## Description

calculate colvec max value within limited range

## Usage

col_max_val(u, wlmt)

## Arguments

u
u: a numeric colvec
wlmt
wlmt: limit search on colvec on indices within wlmt

## Value

vd min value in u within wlmt w.r.t to original index

## See Also

Other matrix_minmax: col_max_idx, col_min_idx, col_min_val, col_rgn_val

```
col_min_idx col_min_idx
```


## Description

calculate colvec min value index within limited range

## Usage

col_min_idx(u, wlmt)

## Arguments

u
u: a numeric colvec
wlmt
wlmt: limit search on colvec on indices within wlmt

## Value

id an index of min value in $u$ within wlmt w.r.t to original index

## Note

cpp use index start from 0 vs $r$ use index start from 1
in case of equal std:min/std:max take first index seen

## See Also

Other matrix_minmax: col_max_idx, col_max_val, col_min_val, col_rgn_val

```
col_min_val col_min_val
```


## Description

calculate colvec min value within limited range

## Usage

col_min_val(u, wlmt)

## Arguments

u
u : a numeric colvec
wlmt
wlmt: limit search on colvec on indices within wlmt

## Value

vd min value in u within wlmt w.r.t to original index

## See Also

Other matrix_minmax: col_max_idx, col_max_val, col_min_idx, col_rgn_val

```
col_rgn_val col_rgn_val
```


## Description

calculate colvec range $=\max -\min$ value within limited range

## Usage

col_rgn_val(u, wlmt)

## Arguments

u
u : a numeric colvec
wlmt
wlmt: limit search on colvec on indices within wlmt

## Value

$\mathrm{vd} \max$ - min value in u within wlmt w.r.t to original index

## See Also

Other matrix_minmax: col_max_idx, col_max_val, col_min_idx, col_min_val

```
dist_wlatlng dist_wlatlng
```


## Description

calculate distance btwn coordinate1<lat1, lng1> and coordinate2<lat2, lng2>

## Usage

dist_wlatlng(.lat1, .lng1, .lat2, .lng2, .measure = "mi")

## Arguments

| .lat1 | latitude of coordinate1 |
| :--- | :--- |
| .lng1 | longitude of coordinate1 |
| .lat2 | latitude of coordinate2 |
| .lng2 | longitude of coordinate2 |
| .measure | - mi or km |

## Details

calculate the great circle distance between 2 points with Haversine formula, which deliberately ignores elevation differences.

Haversine formula (from R.W. Sinnott, "Virtues of the Haversine", Sky and Telescope, vol. 68, no. 2, 1984, p. 159):
dlon $=\operatorname{lon} 2-\operatorname{lon} 1$
dlat = lat2 - lat 1
$\mathrm{a}=\sin ^{\wedge} 2($ dlat $/ 2)+\cos ($ lat 1$) * \cos ($ lat 2$) * \sin ^{\wedge} 2($ dlon/2 $)$
$\mathrm{c}=2 * \arcsin (\min (1, \mathrm{sqrt}(\mathrm{a})))$
$\mathrm{d}=\mathrm{R} * \mathrm{c}$

```
dist_wlatlng_cpp dist_wlatlng_cpp
```


## Description

calculate distance between coordinate $1<$ lat $1, \operatorname{lng} 1>$ and coordinate $2<$ lat $2, \operatorname{lng} 2>$

## Usage

```
dist_wlatlng_mi_cpp(lat1, lng1, lat2, lng2)
dist_wlatlng_km_cpp(lat1, lng1, lat2, lng2)
distSgl_wlatlng_cpp(lat1, lng1, lat2, lng2, measure = "mi")
distRpl_wlatlng_cpp(lat1, lng1, lat2, lng2, measure = "mi",
    distRpl_GS = 100L)
```


## Arguments

| lat1 | latitude of coordinate1 |
| :--- | :--- |
| lng1 | longitude of coordinate1 |
| lat2 | latitude of coordinate2 |
| lng2 | longitude of coordinate2 |

measure $\quad$ "mi" (mile) or "km" (kilometer)
distRpl_GS The grain size of a parallel algorithm sets a minimum chunk size for parallelization. In other words, at what point to stop processing input on separate threads (as sometimes creating more threads can degrade the performance of an algorithm by introducing excessive synchronization overhead). Default is 100.

## Details

calculate the great circle distance between 2 points with Haversine formula, which deliberately ignores elevation differences.
Haversine formula (from R.W. Sinnott, "Virtues of the Haversine", Sky and Telescope, vol. 68, no. 2, 1984, p. 159):
dlon $=\operatorname{lon} 2-\operatorname{lon} 1$
dlat = lat2 - lat 1
$\mathrm{a}=\sin ^{\wedge} 2(\mathrm{dlat} / 2)+\cos ($ lat 1$) * \cos ($ lat 2$) * \sin ^{\wedge} 2($ dlon/2 $)$
$\mathrm{c}=2 * \arcsin (\min (1, \operatorname{sqrt}(\mathrm{a})))$
$\mathrm{d}=\mathrm{R} * \mathrm{c}$
dist_wlatlng_mi_cpp:
calculate distance between coordinate $1<$ lat $1, \operatorname{lng} 1>$ and coordinate $2<l a t 2, \operatorname{lng} 2>$ in mile
dist_wlatlng_km_cpp:
calculate distance between coordinate $1<$ lat $1, \operatorname{lng} 1>$ and coordinate $2<1 a t 2, \operatorname{lng} 2>$ in kilometer
distSgl_wlatlng_cpp:
calculate distance between coordinate $1<l a t 1, \operatorname{lng} 1>$ and coordinate $2<l a t 2, \operatorname{lng} 2>$ in mile (measure $=$ "mi") or kilometer (measure $=$ "km"), default is mile.
implement as serial computing over vector of lat1, lng1, lat2, lng2
distRpl_wlatlng_cpp:
calculate distance between coordinate $1<\operatorname{lat} 1, \operatorname{lng} 1>$ and coordinate $2<l a t 2, \operatorname{lng} 2>$ in mile (measure $=$ "mi") or kilometer (measure = "km"), default is mile.
implement as parallel computing over vector of lat1, lng1, lat2, lng2 via RcppParallel

```
skmRpl_mlp_cpp skmRpl_mlp_cpp
```


## Description

solve skm with multiple runs in parallel

## Usage

skmRpl_mlp_cpp(x, k, s_must, max_it, max_at, skmRpl_GS = 100L)

## Arguments

x
k
s_must
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.
max_at max number of attempts or repeats on running for optimial results, max number of random initialization for finding optimial results.
skmRpl_GS skmRpl_GS: RcppParallel grain size when run skmRpl_mlp_cpp

## Details

refer skm_sgl_cpp

## Value

skmSolution skmSolution present in r list

```
skmSolution skmSolution
```


## Description

class skmSolution, which often returned via skm solver implemented in cpp

## Usage

skmSolution

## Format

An object of class C++Class of length 1 .

## Details

an skmSolution instance has two member variable:
o: objective sum(min(x.subview(i in $s$, all $j$ ), min over all $i$ ), sum over all $j$ )
s : selected index set of row index start from 0

```
    skm_gdp_cpp skm_gdp_cpp
```


## Description

solve selective kmeans via a greedy propagation.

## Usage

skm_gdp_cpp(x, k = 0L)

## Arguments

$\mathrm{k} \quad$ number of index to be selected from x row index start from 0 .
x
an $m \mathrm{x}$ n matrix of $\mathrm{s}-\mathrm{t}$ - dist

## Details

skm_gdp_cpp init with an input $\mathrm{m} x \mathrm{n}$ matrix x and want to select an index set s of size k from x row index started from 0 such that
minimize sum(min(x.subview(i in $s$, all j ), min over all i ), sum over all j )
skm_gdp_cpp solve the problem with greedy propagation via selecting the current best addon index from the index set left, addon index is defined as such index when addon to the selected one can bring the most improvement.
since skm_gbp_cpp would select index one by one, and no return, e.g., if select index A for $\mathrm{k}=1$, then selection on $\mathrm{k}=2$ would build on $\mathrm{k}=1$, so index A is always present in the solution, so all index can be ranked w.r.t when it would be considered as the best addon. as a result skm_gbp_cpp a parameter $k$ is not always required, so default $k=0$ will resturn a vector of size $m$, and user can select to top k as solution for k .

## Value

s a ranked index $0-\mathrm{m}-1$ where the top k would minimize $\operatorname{sum}(\min (\mathrm{x} . \operatorname{subview}(\mathrm{i}$ in $\mathrm{s}(0 . . \mathrm{k}-1)$, all j$)$, min over all i), sum over all j)

```
skm_minmax_cpp skm_minmax_cpp
```


## Description

skm via min-max on in cpp - subroutine of skm_sgl_cpp calls

## Usage

skm_minmax_cpp(x, s_must)

## Arguments

X
an mxn matrix often $\mathrm{m}>\mathrm{n}$
s_must matrix $x$ row index start from 0 that must be selected with priority

## Details

skm_minmax_cpp init an input $\mathrm{m} \times \mathrm{n}$ matrix x , and a priority vector $\mathrm{s} \_$must would select n indicies from $m$ such that:
$\operatorname{minimize} \operatorname{sum}(\min (x(i, j)$ where $\mathrm{i}<1 . . n>$ and $\mathrm{j}<1 . . \mathrm{n}>$ each use $<1 . . n>$ once $)$ )
so in case $\mathrm{m}<=\mathrm{n}$ it simply select all m - should always be apply on matrix with $\mathrm{m}>\mathrm{n}$ - it is designed as a expectation step in skm_cpp on updating s.
it select i in $<1 . . \mathrm{m}>$ such that i has the colwise_min_idx on column j where j has max difference of (colwise_max_val - colwise_min_val), it then remove row i col j from matrix and repeat.
s_must presents the indices with priority so that the selection must select first indicies within s_must and then select other indicies outside s_must.
an example skm_minmax_cpp is superior in bound worst case compare to greedy: $\mathrm{x}=[1100 ; 4$ 200; 2 400; 9 900]: greedy 1 then 200, min-max 100 then 2, and greedy give [1 100; 4 200] with 201 and minmax give [1 100; 2400] with 102.

```
skm_mlp_cpp skm_mlp_cpp
```


## Description

solve skm with multiple runs in serial and return all w. optim

## Usage

skm_mlp_cpp(x, k, s_must, max_it, max_at)

## Arguments

x
$\mathrm{k} \quad$ number of index to be selected from x row index start from 0 .
s_must an index vector set should be selected before selecting other index.
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.
max_at max number of attempts or repeats on running for optimial results, max number of random initialization for finding optimial results.

## Details

refer skm_sgl_cpp

## Value

skmSolution skmSolution present in r list

## See Also

Other skm: skm_mls_cpp, skm_rgi_cpp, skm_rgs_cpp, skm_sgl_cpp

```
skm_mls skm_mls
```


## Description

a selective k-means problem solver - wrapper over skm_mls_cpp

## Usage

```
skm_mls(x, k = 1L, s_colname = "s", t_colname = "t", d_colname = "d",
    w_colname = NULL, s_ggrp = integer(0L), s_must = integer(0L),
    max_it \(=100 \mathrm{~L}\), max_at \(=100 \mathrm{~L}\), auto_create_ggrp \(=\) TRUE,
    extra_immaculatism = TRUE, extra_at = 10L)
```


## Arguments

x data.table with $\mathrm{s}-\mathrm{t}-\mathrm{d}(\mathrm{s}, \mathrm{t})$ : $\mathrm{s}<$ source $>-\mathrm{t}<$ target $>-\mathrm{d}<$ distance $>$ where $\mathrm{s}<$ source> and $\mathrm{t}<$ target> must characters and d<distance> must numeric. aware d<distance> is not necessary as an euclidean or any distance and even necessary as symmetric $-\mathrm{d}(\mathrm{s}, \mathrm{t})$ can be unequal to $\mathrm{d}(\mathrm{t}, \mathrm{s})$ - view d as such a measure of the cost of assigning one to the other!
k number of centers
s_colname s<source>
t_colname t<target>
d_colname d<distance> - view d as cost of assigning t into s. also modify the input data or build in the algorithm can solve problem with a different fixed cost on using each $s$ as source - i prefer to moddify data so that the algorithm is clean and clear - i will show a how to in vignette
w_colname $\quad$ w<weighting> - optional: when not null will optimize toward objective to minimize $d=d^{*} w$ such as weighted cost of assigning $t$ into $s$
s_ggrp s_init will be stratified sampling from s w.r.t s_ggrp.
s_must length $<=\mathrm{k}-1 \mathrm{~s}$ must in result: conditional optimizing.
max_it max number of iterations can run for optimizing result.
max_at max number of attempts/repeats on running for optimial.
auto_create_ggrp
boolean indicator of whether auto creating the group structure using the first letter of s when s_ggrp is integer( 0 ).
extra_immaculatism
boolean indicator of whether making extra runs for improving result consistency when multiple successive $k$ is specified, e.g., $k=c(9 L, 10 L)$.
extra_at an integer specifying the number of extra runs when argument extra_immaculatism is TRUE.

## Details

a selective k-means problem is defined as finding a subset of k rows from a m x n matrix such that the sum of each column minimial is minimized.
skm_mls would take data.table (data.frame) as inputs, rather than a matrix, assume that a data.table of $s-t-d(s, t)$ for all combination of $s$ and $t$, choose $k$ of $s$ that minimizes $\operatorname{sum}(\min (d(s, t)$ over selected k of s ) over t ).

## Value

data.table
o - objective - based on d_colname
w - weighting - based on w_colname
$k-k<k$-list> - based on $k$ - input
s - s<source> - based on s_colname
$d$ - weighed averge value of d_colname weighed by w_column when $s$ are selected.

```
skm_mls_cpp skm_mls_cpp
```


## Description

solve skm with multiple runs in serial and return all w. optim and s_init stratified sampled w.r.t g

## Usage

skm_mls_cpp(x, k, g, s_must, max_it, max_at)

## Arguments

X
$\mathrm{k} \quad$ number of index to be selected from x row index start from 0 .
g stratify structure, often info on grouping of v so that algorithm should make random initialization from stratified sample across groups.
s_must an index vector set should be selected before selecting other index.
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.
max_at max number of attempts or repeats on running for optimial results, max number of random initialization for finding optimial results.

## Details

refer skm_sgl_cpp

## Value

skmSolution skmSolution present in $r$ list

## See Also

Other skm: skm_mlp_cpp, skm_rgi_cpp, skm_rgs_cpp, skm_sgl_cpp

```
skm_rgi_cpp skm_rgi_cpp
```


## Description

solve skm with single and random size k s_init

## Usage

skm_rgi_cpp(x, k, s_must, max_it)

## Arguments

$x \quad$ an $m \mathrm{x}$ n matrix often $\mathrm{m}<\mathrm{n}$, as a convention index rows of x with s , and cols of $x$ with $t$ so $x(i, j)$ can be expressed as (s_i, t_j) equally.
$\mathrm{k} \quad$ number of index to be selected from x row index start from 0.
s_must an index vector set should be selected before selecting other index.
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.

## Details

refer skm_sgl_cpp

## Value

skmSolution

## See Also

Other skm: skm_mlp_cpp, skm_mls_cpp, skm_rgs_cpp, skm_sgl_cpp

```
skm_rgs_cpp skm_rgs_cpp
```


## Description

solve skm with single and random size k s_init stratified sampled w.r.t $g$

## Usage

skm_rgs_cpp(x, k, g, s_must, max_it)

## Arguments

$x \quad$ an $m \times n$ matrix often $m<n$, as a convention index rows of $x$ with $s$, and cols of x with t so $\mathrm{x}(\mathrm{i}, \mathrm{j})$ can be expressed as ( $\mathrm{s} \_\mathrm{i}, \mathrm{t} \mathbf{j}$ ) equally.
$\mathrm{k} \quad$ number of index to be selected from x row index start from 0.
g stratify structure, often info on grouping of v so that algorithm should make random initialization from stratified sample across groups.
s_must an index vector set should be selected before selecting other index.
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.

## Details

refer skm_sgl_cpp

## Value

skmSolution

## See Also

Other skm: skm_mlp_cpp, skm_mls_cpp, skm_rgi_cpp, skm_sgl_cpp

```
skm_sgl_cpp skm_sgl_cpp
```


## Description

solve skm with single and a fixed given s_init

## Usage

skm_sgl_cpp(x, s_init, s_must, max_it)

## Arguments

$x \quad$ an $m \mathrm{n}$ n matrix often $\mathrm{m}<\mathrm{n}$, as a convention index rows of x with s , and cols of $x$ with $t$ so $x(i, j)$ can be expressed as (s_i, t_j) equally.
s_init an init vector of $k$ index to start the search of optimal index set of $k$, length of s_init also defined the number of index want to be select.
s_must an index vector set should be selected before selecting other index.
max_it max number of iterations can run for optimizing result. max number of iterations within a single initial run on optimal path.

## Details

a numeric $m \mathrm{x} n$ matrix x often $\mathrm{m}<\mathrm{n}$ and want to select a subset of k from m such that it minimize the $\operatorname{sum}(\min (x(i, j)-$ minimum w.r.t each $j$ over all $i$ within selected index set $)$, over all $i)$
if $\mathrm{m}=\mathrm{n}$ and $\mathrm{x}(\mathrm{i}, \mathrm{j})$ as euclidean distance then it is equivalent to kmeans
skm can select a combined set for deploying resource, for example, where to build 5 warehouses on united states, which often different than build these warehouses via select the current best one by one.

## Value

skmSolution

## See Also

Other skm: skm_mlp_cpp, skm_mls_cpp, skm_rgi_cpp, skm_rgs_cpp

```
source_zip_list source_zip_list
```


## Description

a list of zip code used in skm package demonstration.

## Usage

source_zip_list

## Format

a character vector of length 51 includes one 5 digits zip code selected from each state, where the most central zip code in each state selected.
stratified_sampling stratified_sampling

## Description

select k elements from vector v w.r.t stratify structure group g . TODO - implementing via template so $v$ is flexible as vec or uvec.

## Usage

stratified_sampling(v, k, g)

## Arguments

v <vector> v: a numeric candidate v from which draw sample.
$\mathrm{k} \quad$ <integer> k: selection sample size.
g <vector>g: stratify structure $g$-info on grouping of $v$ so that the selected sample is stratified across groups.

## Value

s <vector> s: a vector select from v length k stratified by g .

## Note

v is required as an integer vector for using in skm
zip2012 zip2012

## Description

a zip code database with latitude, longitude, population and income.

## Usage

zip2012

## Format

A data table with 28844 rows and 9 variables:
zip zip code, 5 digits zip code in U.S.
lat latitude
Ing longitude
pop population
ink income
city city
state state
p_pop percentage of population w.r.t total population
p_ink percentage of income w.r.t total income

## Source

http://federalgovernmentzipcodes.us/

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