# Package 'flowcluster'

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Title Cluster Origin-Destination Flow Data

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```
Version 0.2.1
Description Provides functionality for clustering
      origin-destination (OD) pairs, representing desire lines (or flows).
      This includes creating distance matrices between OD pairs and passing
      distance matrices to a clustering algorithm. See the academic paper
      Tao and Thill (2016) <doi:10.1111/gean.12100>
      for more details on spatial clustering of flows.
      See the paper on delineating demand-responsive operating areas
      by Mahfouz et al. (2025) <doi:10.1016/j.urbmob.2025.100135>
      for an example of how this package can be used to cluster flows for
      applied transportation research.
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```

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 ${\tt add\_flow\_length}$ 

Add Length Column to Flow Data

# Description

Also checks that 'origin' and 'destination' columns are present.

# Usage

```
add_flow_length(x)
```

# Arguments

Χ

sf object of flows (LINESTRING, projected CRS)

# Value

sf object with an additional length\_m column (od length in meters)

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- add_flow_length(flows)</pre>
```

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add\_xyuv

Add Start/End Coordinates & Flow IDs

# Description

Add Start/End Coordinates & Flow IDs

#### Usage

```
add_xyuv(x)
```

#### **Arguments**

Х

sf object of flows

#### Value

```
tibble with x, y, u, v, flow_ID columns
```

#### **Examples**

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- add_flow_length(flows)
flows <- add_xyuv(flows)</pre>
```

aggregate\_clustered\_flows

Aggregate clustered OD flows into representative lines

# Description

This function aggregates flows within clusters and creates a single representative line for each cluster. The start and end coordinates are computed as weighted averages (weighted by flow counts or another variable), or simple means if no weights are provided. Each cluster is represented by one LINESTRING.

# Usage

```
aggregate_clustered_flows(flows, weight = NULL, crs = sf::st_crs(flows))
```

# **Arguments**

flows	An sf object containing OD flows with coordinates for origins $(x, y)$ and destinations $(u, v)$ , a cluster column, and optionally a count or other weighting variable.
weight	(optional) Name of a column in flows to use for weighting. If NULL (default), unweighted means are used.
crs	Coordinate reference system for the output (default: taken from flows).

#### Value

An sf object with one line per cluster, containing:

- count\_total: total weight (if provided), otherwise number of flows
- size: the cluster size (from the input, not recomputed)
- geometry: a LINESTRING representing the aggregated OD flow

```
# ---- 1. Basic Usage: A quick, runnable example ---
# This demonstrates the function with minimal, fast data preparation.
flows <- flowcluster::flows_leeds</pre>
# Create the required input columns in a single, fast pipeline
flows_clustered <- flows |>
 add xvuv() |>
 # Manually create 3 dummy clusters for demonstration
 dplyr::mutate(cluster = sample(1:3, size = nrow(flows), replace = TRUE)) |>
 # The function requires a 'size' column, so we add it
 dplyr::group_by(cluster) |>
 dplyr::add_tally(name = "size") |>
 dplyr::ungroup()
# Demonstrate the function
flows_agg_w <- aggregate_clustered_flows(flows_clustered, weight = "count")</pre>
print(flows_agg_w)
# ---- 2. Detailed Workflow (not run by default) ---
## Not run:
 # This example shows the ideal end-to-end workflow, from raw data
 # to clustering and finally aggregation. It is not run during checks
 # because the clustering steps are too slow.
 # a) Prepare the data by filtering and adding coordinates
 flows_prep <- flowcluster::flows_leeds |>
    sf::st_transform(3857) |>
    add_flow_length() |>
    filter_by_length(length_min = 5000, length_max = 12000) |>
   add_xyuv()
 # b) Calculate distances and cluster the flows
 distances <- flow_distance(flows_prep, alpha = 1.5, beta = 0.5)
 dmat <- distance_matrix(distances)</pre>
 wvec <- weight_vector(dmat, flows_prep, weight_col = "count")</pre>
 flows_clustered_real <- cluster_flows_dbscan(dmat, wvec, flows_prep, eps = 8, minPts = 70)
  # c) Filter clusters and add a 'size' column
 flows_clustered_real <- flows_clustered_real |>
    dplyr::filter(cluster != 0) |> # Filter out noise points
   dplyr::group_by(cluster) |>
   dplyr::mutate(size = dplyr::n()) |>
   dplyr::ungroup()
```

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```
# d) Now, use the function on the clustered data
flows_agg_real <- aggregate_clustered_flows(flows_clustered_real, weight = "count")
print(flows_agg_real)

# e) Visualize the results
if (requireNamespace("tmap", quietly = TRUE)) {
    library(tmap)
    # This plot uses modern tmap v4 syntax.
    tm_shape(flows_clustered_real, facet = "cluster") +
        tm_lines(col = "grey50", alpha = 0.5) +
    tm_shape(flows_agg_real) +
        tm_lines(col = "red", lwd = 2) +
        tm_layout(title = "Original Flows (Grey) and Aggregated Flows (Red)")
}

## End(Not run)</pre>
```

cluster\_flows\_dbscan Cluster Flows using DBSCAN

#### **Description**

See dbscan for details on the DBSCAN algorithm.

#### Usage

```
cluster_flows_dbscan(dist_mat, w_vec, x, eps, minPts)
```

#### **Arguments**

dist_mat	distance matrix
w_vec	weight vector
х	flows tibble with flow_ID
eps	DBSCAN epsilon parameter
minPts	DBSCAN minPts parameter

#### Value

flows tibble with an additional cluster column

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- head(flows, 100) # for testing
# Add flow lengths and coordinates
flows <- add_flow_length(flows)
# filter by length</pre>
```

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```
flows <- filter_by_length(flows, length_min = 5000, length_max = 12000)
flows <- add_xyuv(flows)
# Calculate distances
distances <- flow_distance(flows, alpha = 1.5, beta = 0.5)
dmat <- distance_matrix(distances)
wvec <- weight_vector(dmat, flows, weight_col = "count")
clustered <- cluster_flows_dbscan(dmat, wvec, flows, eps = 8, minPts = 70)</pre>
```

dbscan\_sensitivity

Sensitivity analysis of DBSCAN parameters for flow clustering.

#### **Description**

The function allows you to test different combinations of epsilon and minPts parameters for clustering flows using DBSCAN. It can be used to determine what parameter values make sense for your data

### Usage

```
dbscan_sensitivity(
  dist_mat,
  flows,
  options_epsilon,
  options_minpts,
  w_vec = NULL
)
```

### Arguments

#### Value

a tibble with columns: id (to identify eps and minpts), cluster, size (number of desire lines in cluster), count\_sum (total count per cluster)

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#### **Examples**

```
flows <- sf::st_transform(flows_leeds, 3857)</pre>
flows <- head(flows, 1000) # for testing
# Add flow lengths and coordinates
flows <- add_flow_length(flows)</pre>
# filter by length
flows <- filter_by_length(flows, length_min = 5000, length_max = 12000)</pre>
# Add x, y, u, v coordinates to flows
flows <- add_xyuv(flows)</pre>
# Calculate distance matrix
distances <- flow_distance(flows, alpha = 1.5, beta = 0.5)
dmat <- distance_matrix(distances)</pre>
# Generate weight vector
w_vec <- weight_vector(dmat, flows, weight_col = "count")</pre>
# Define the parameters for sensitivity analysis
options_epsilon <- seq(1, 10, by = 2)
options_minpts <- seq(10, 100, by = 10)
# # Run the sensitivity analysis
results <- dbscan_sensitivity(
 dist_mat = dmat,
 flows = flows,
 options_epsilon = options_epsilon,
 options_minpts = options_minpts,
 w_vec = w_vec
)
```

distance\_matrix

Convert Long-Format Distance Tibble to Matrix

#### **Description**

Convert Long-Format Distance Tibble to Matrix

#### Usage

```
distance_matrix(distances, distance_col = "fds")
```

#### **Arguments**

```
distances tibble with columns flow_ID_a, flow_ID_b, and distance distance_col column name for distance (default "fds")
```

#### Value

distance matrix (tibble with rownames). The matrix has flow\_ID\_a as rownames and flow\_ID\_b as column names. This function converts the output of flow\_distance() into a format suitable for the dbscan clustering algorithm.

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#### **Examples**

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- head(flows, 100) # for testing
# Add flow lengths and coordinates
flows <- add_flow_length(flows)
flows <- add_xyuv(flows)
# Calculate distances
distances <- flow_distance(flows, alpha = 1.5, beta = 0.5)
dmat <- distance_matrix(distances)</pre>
```

filter\_by\_length

Filter Flows by Length

#### **Description**

Filter Flows by Length

#### Usage

```
filter_by_length(x, length_min = 0, length_max = Inf)
```

#### **Arguments**

x sf object with length\_m
length\_min minimum length (default 0)
length\_max maximum length (default Inf)

#### Value

filtered sf object. Flows with length\_m outside the specified range are removed.

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- add_flow_length(flows)
flows <- filter_by_length(flows, length_min = 5000, length_max = 12000)</pre>
```

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flows_leeds	Example flow data for Leeds. It is from the 2021 census, and it contains all Origin - Destination flows at the MSOA level. For more info on census flow data, see the Rhrefhttps://www.ons.gov.uk/census/aboutcensus/censusproducts/origindestinationflowdataONS documentation See data-raw/flows_leeds.R for how this data was created.

#### Description

Example flow data for Leeds. It is from the 2021 census, and it contains all Origin - Destination flows at the MSOA level. For more info on census flow data, see the ONS documentation See data-raw/flows\_leeds.R for how this data was created.

# Usage

flows\_leeds

#### **Format**

An object of class sf with LINESTRING geometry. It has the following columns:

origin MSOA code of origin zone

destination MSOA code of destination zone

count number of people moving from origin to destination

geometry desire line between origin and destination

#### **Source**

```
https://www.nomisweb.co.uk/sources/census_2021_od
```

flow\_distance

Calculate Flow Distance and Dissimilarity

#### **Description**

This function calculates flow distance and dissimilarity measures between all pairs of flows based on the method described in @tao2016spatial.

#### Usage

```
flow_distance(x, alpha = 1, beta = 1)
```

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#### **Arguments**

x tibble with flow\_ID, x, y, u, v, length\_m

alpha numeric, origin weight beta numeric, destination weight

#### Value

tibble of all OD pairs with fd, fds columns

#### References

Tao, R., Thill, J.-C., 2016. Spatial cluster detection in spatial flow data. Geographical Analysis 48, 355–372. https://doi.org/10.1111/gean.12100

# **Examples**

```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- head(flows, 100) # for testing
# Add flow lengths and coordinates
flows <- add_flow_length(flows)
flows <- add_xyuv(flows)
# Calculate distances
distances <- flow_distance(flows, alpha = 1.5, beta = 0.5)</pre>
```

weight\_vector

Generate Weight Vector from Flows

#### **Description**

Generate Weight Vector from Flows

### Usage

```
weight_vector(dist_mat, x, weight_col = "count")
```

#### Arguments

dist\_mat distance matrix

x flows tibble with flow\_ID and weight\_col

weight\_col column to use as weights (default = "count")

#### Value

numeric weight vector. Each element corresponds to a flow in the distance matrix, and is used as a weight in the DBSCAN clustering algorithm.

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```
flows <- sf::st_transform(flows_leeds, 3857)
flows <- head(flows, 100) # for testing
# Add flow lengths and coordinates
flows <- add_flow_length(flows)
flows <- add_xyuv(flows)
# Calculate distances
distances <- flow_distance(flows, alpha = 1.5, beta = 0.5)
dmat <- distance_matrix(distances)
wvec <- weight_vector(dmat, flows, weight_col = "count")</pre>
```

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