



Karl Hendrik Tamkivi

# Migrating White Storks (*Ciconia ciconia*) and Climate Change: a Neverending Summer?

Applied Machine Learning for Biological Problems

2023W 301169-1

17.01.2024

# Datasets

## Migration Data

MPIAB Argos white stork tracking (1991-2017)

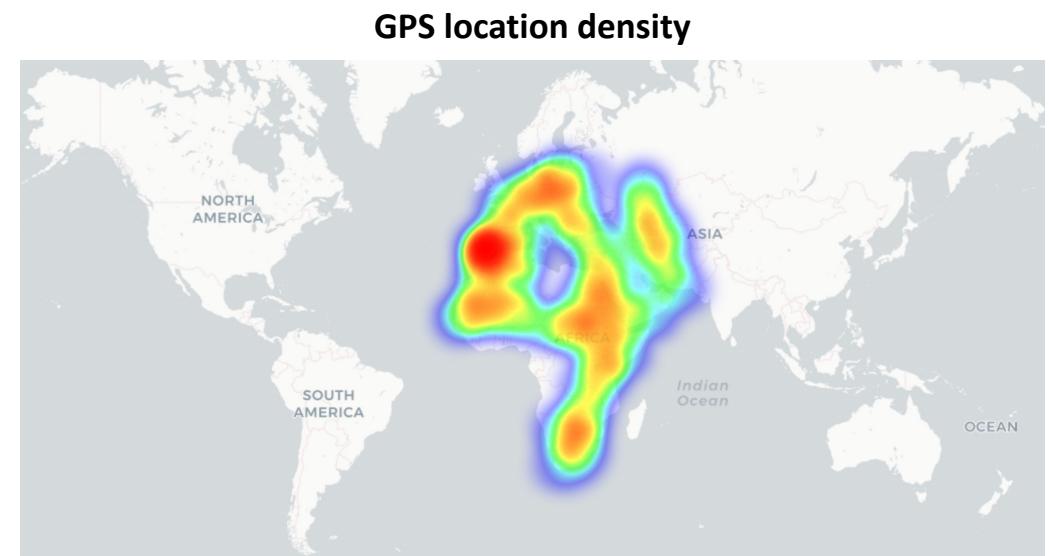
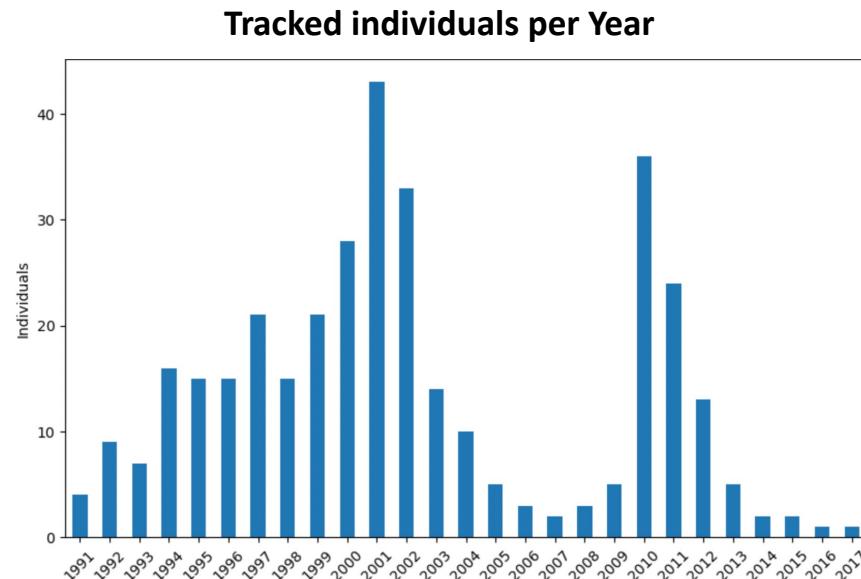


27 years of data

204 different birds

181 017 GPS locations

Including variables like: timestamp, longitude, latitude, altitude, identification number etc.



# Datasets

## Temperature Data

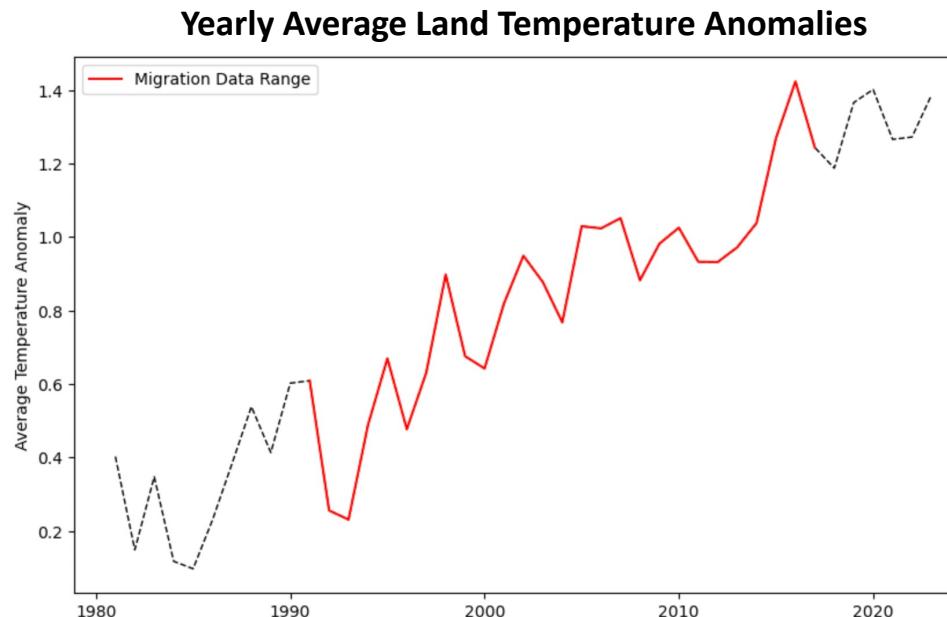
Estimated Global Land-Surface TAVG (1750-2023)      BERKELEY EARTH™

274 years of monthly data

Based on 50 691 time series with 21 339 392 data points

Anomalies compared to 1951-1980 average

Including variables like: year, month, monthly anomaly, annual anomaly etc.



# Research question

**Climate change leads to differential shifts in the timing of annual cycle stages in a migratory bird**

Barbara M. Tomotani ✉, Henk van der Jeugd, Phillip Gienapp, Iván de la Hera, Jos Pilzecker, Corry Teichmann, Marcel E. Visser

First published: 06 December 2017 | <https://doi.org/10.1111/gcb.14006> | Citations: 58

1 December 2002

**Predicting Life-Cycle Adaptation of Migratory Birds to Global Climate Change**

*Timothy Coppock, Christiaan Both*

*"Global warming can lead to increased or decreased intervals between annual cycle stages in migratory birds. Therefore, climate change will also alter the time constraints across the annual cycle."*

*"In different species annual breeding and migration cycles can also be controlled primarily by endogenous rhythms and photoperiodic cues which do not relate to temperature."*

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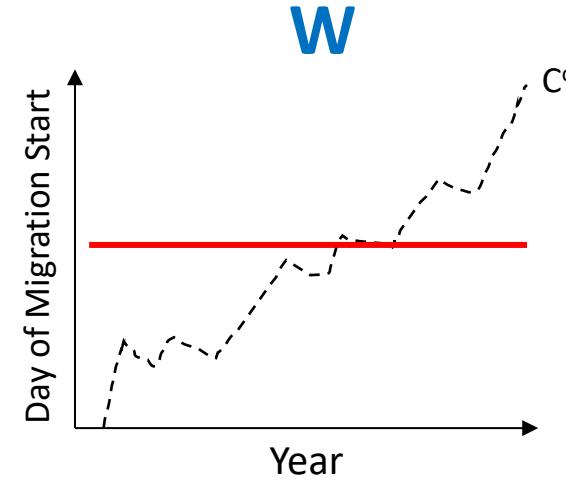
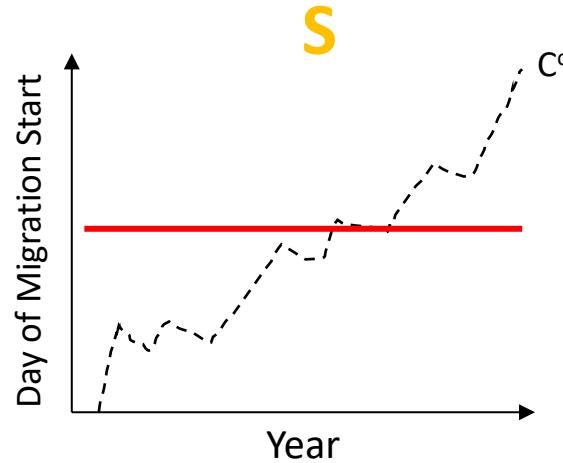
*"In different species annual breeding and migration cycles can also be controlled primarily by endogenous rhythms and photoperiodic cues which do not relate to temperature."*

### Research question:

Has the global average temperature increase caused a shift in the migration starting dates for the white storks?

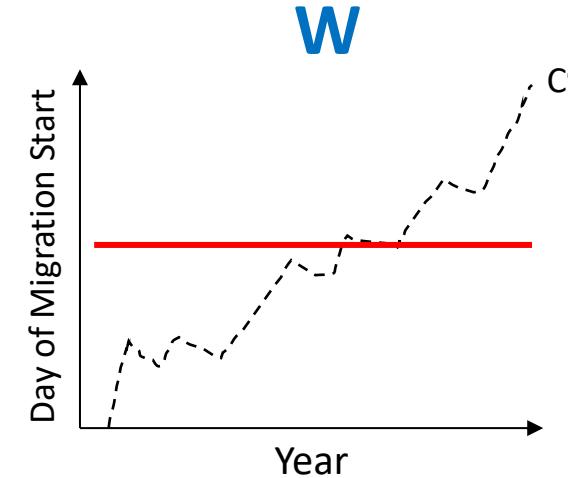
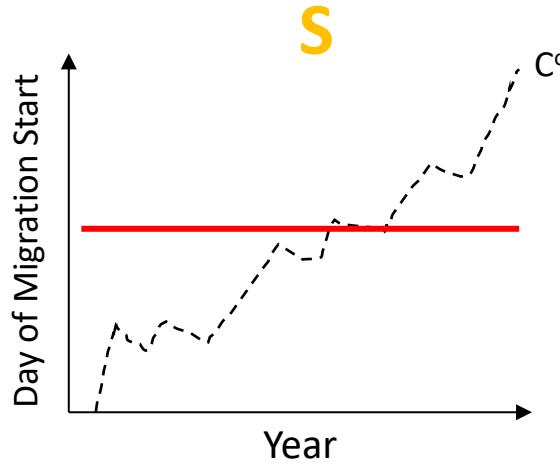
# Research question

**H0:** Average land temperature increase has had no effect on the migration start over the years

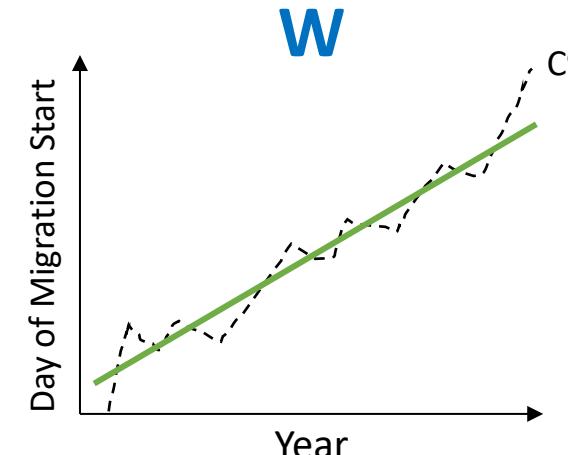
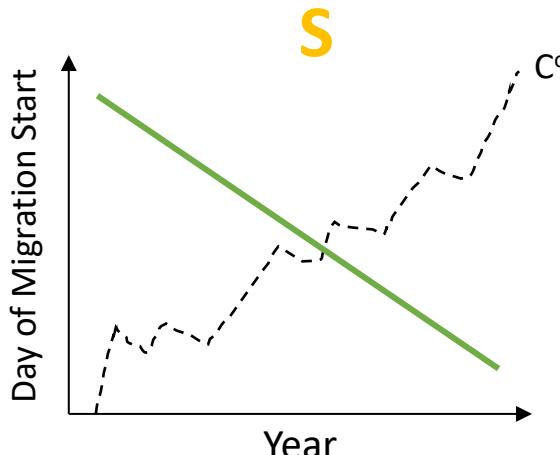


# Research question

**H0:** Average land temperature increase has had no effect on the migration start over the years



**H1:** Average land temperature increase has had an effect on the migration start over the years  
(- effect on **summer migration** start; + effect on **winter migration** start)



# Data Wrangling Roadmap

## Initial data cleaning

Unfiltered data

181 017 instances

Remove marked outliers

Individual = ID + year

One location per day

$\geq 3000$  km of total movement

Exclude ind. w 1000+ km between two days

$\geq 90$  days of data

## Migration labelling

Rolling sum distance of last 5 days

Latitude change from previous day

Applying the labelling rules

Migration statistics

## Secondary data cleaning and temperature data merging

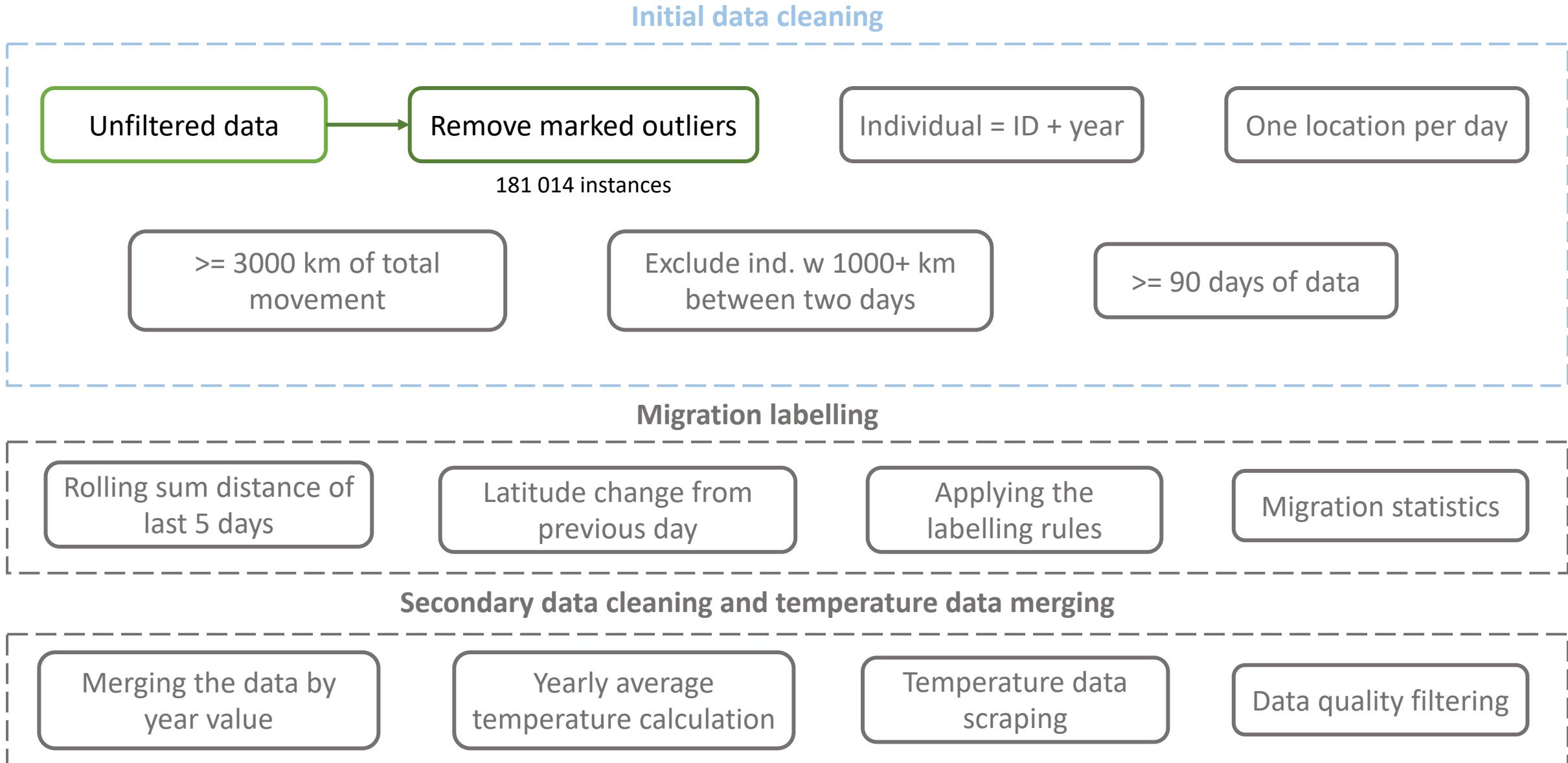
Merging the data by year value

Yearly average temperature calculation

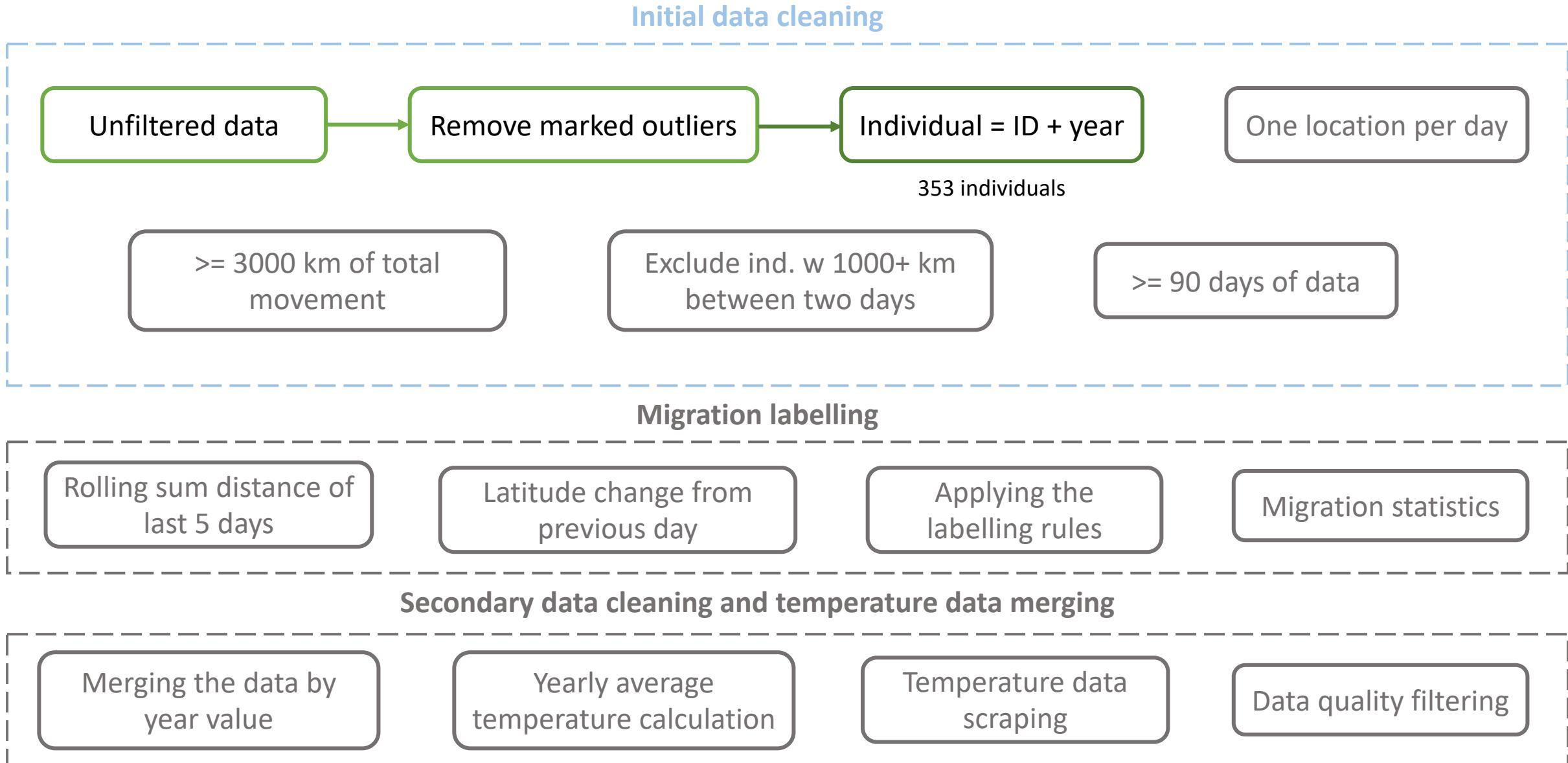
Temperature data scraping

Data quality filtering

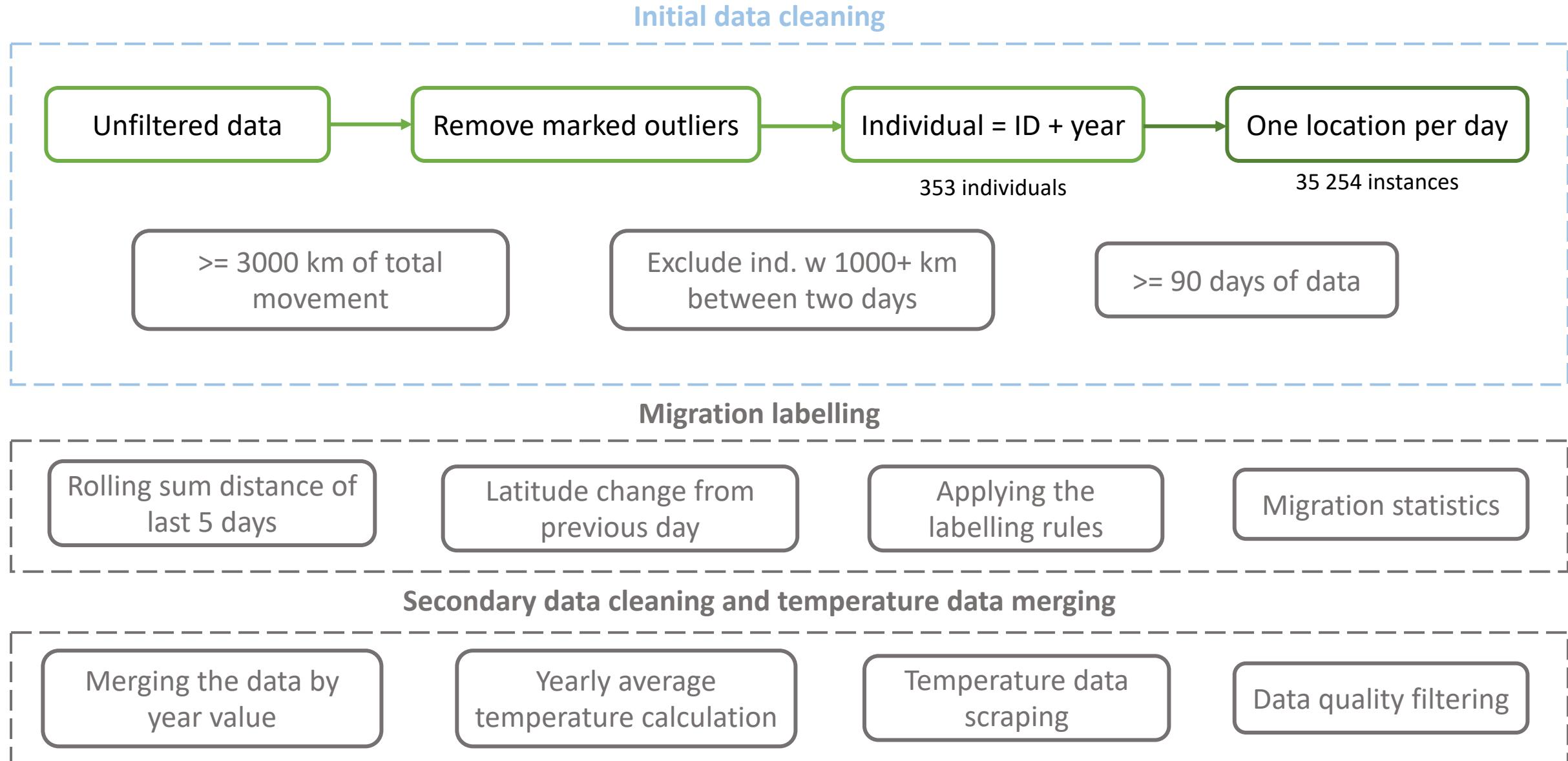
# Data Wrangling Roadmap



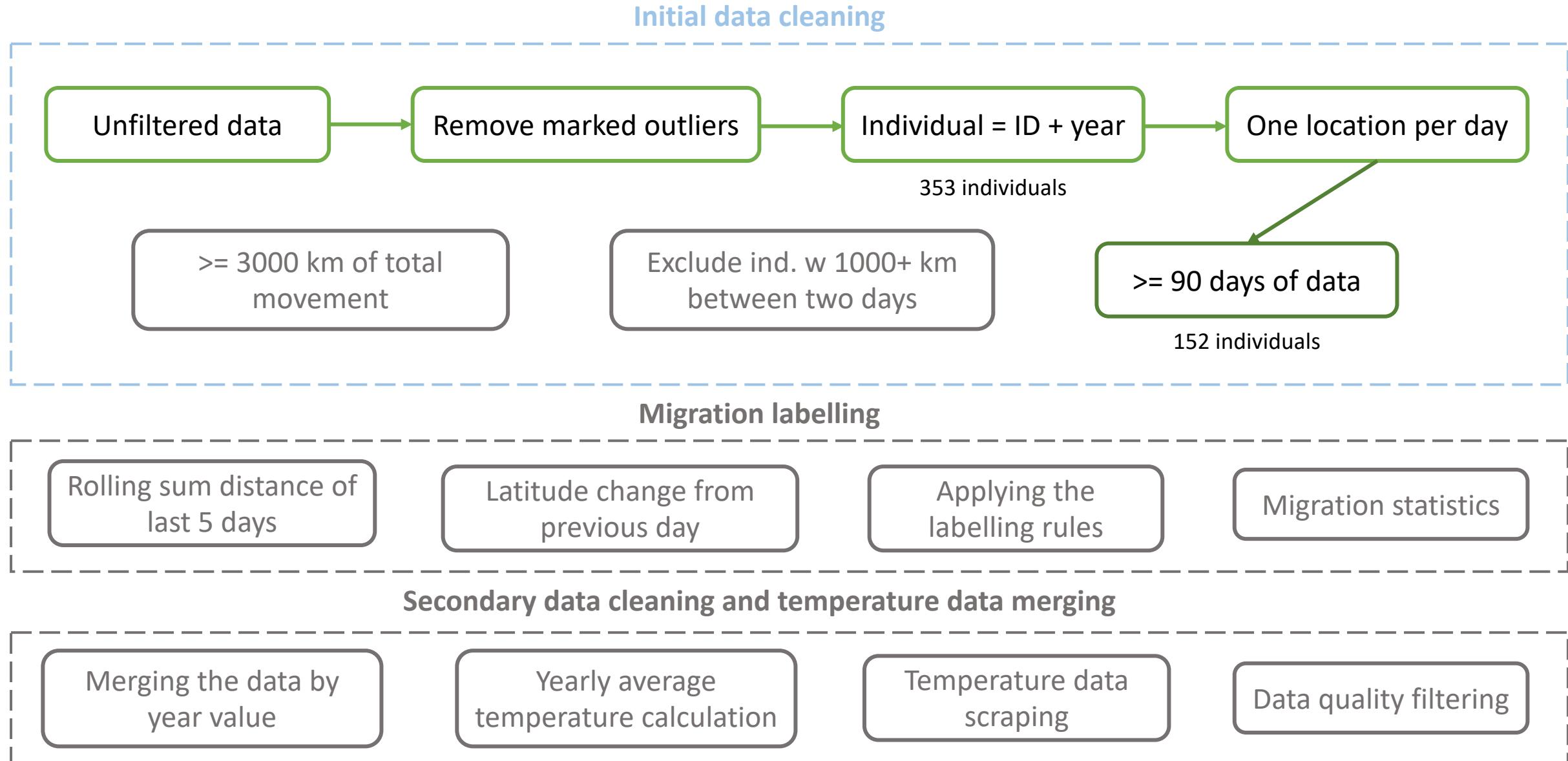
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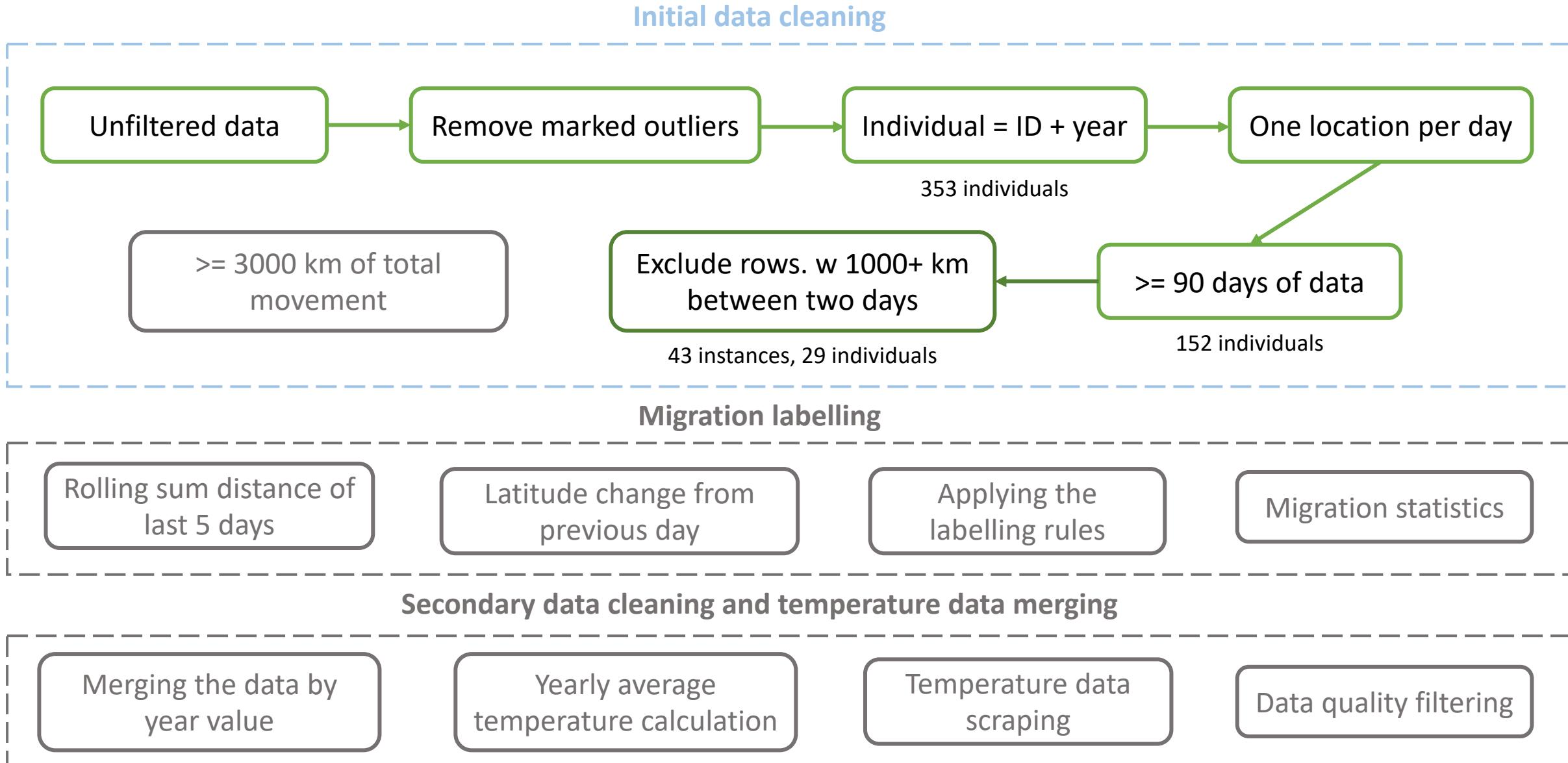
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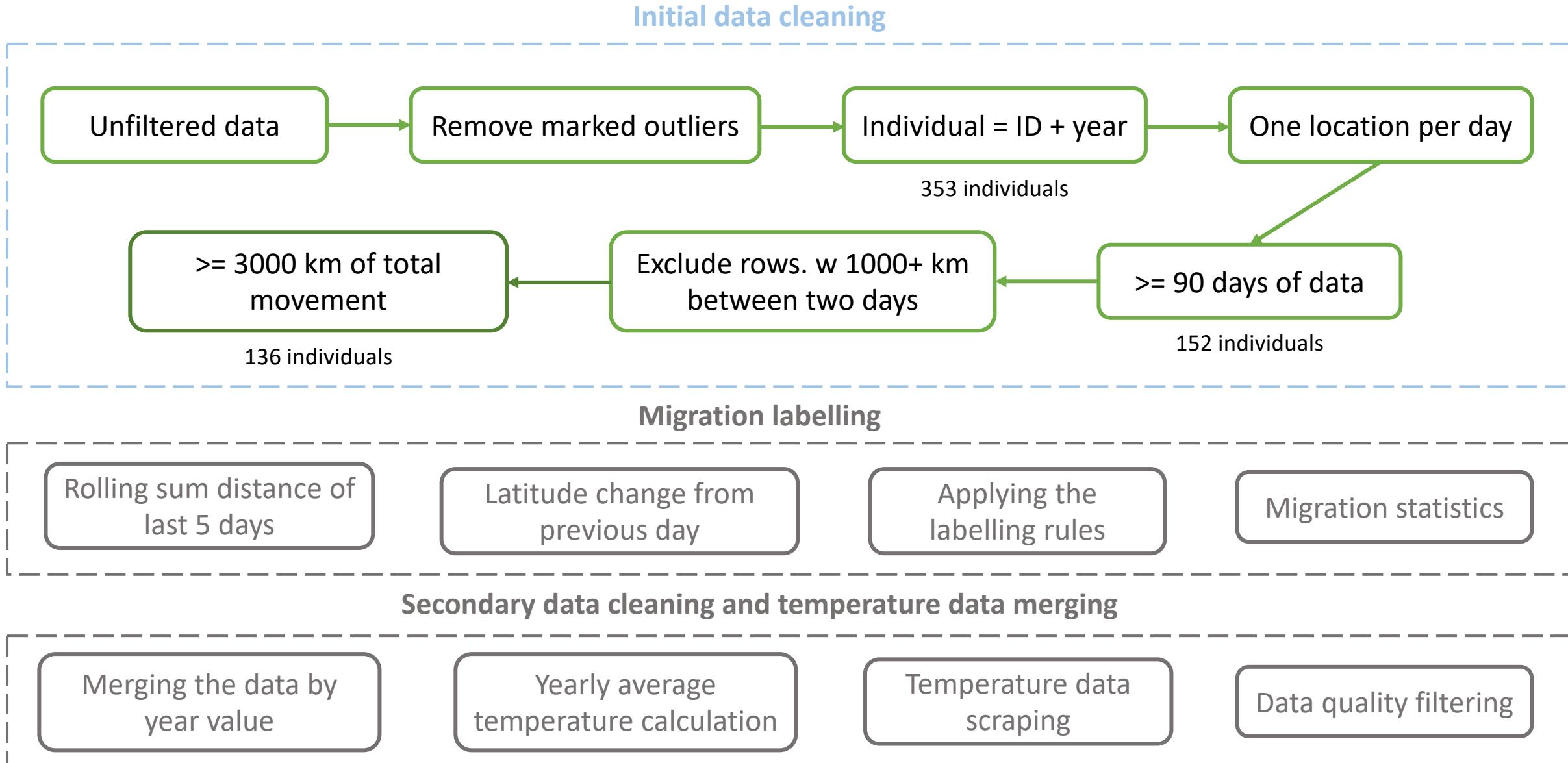
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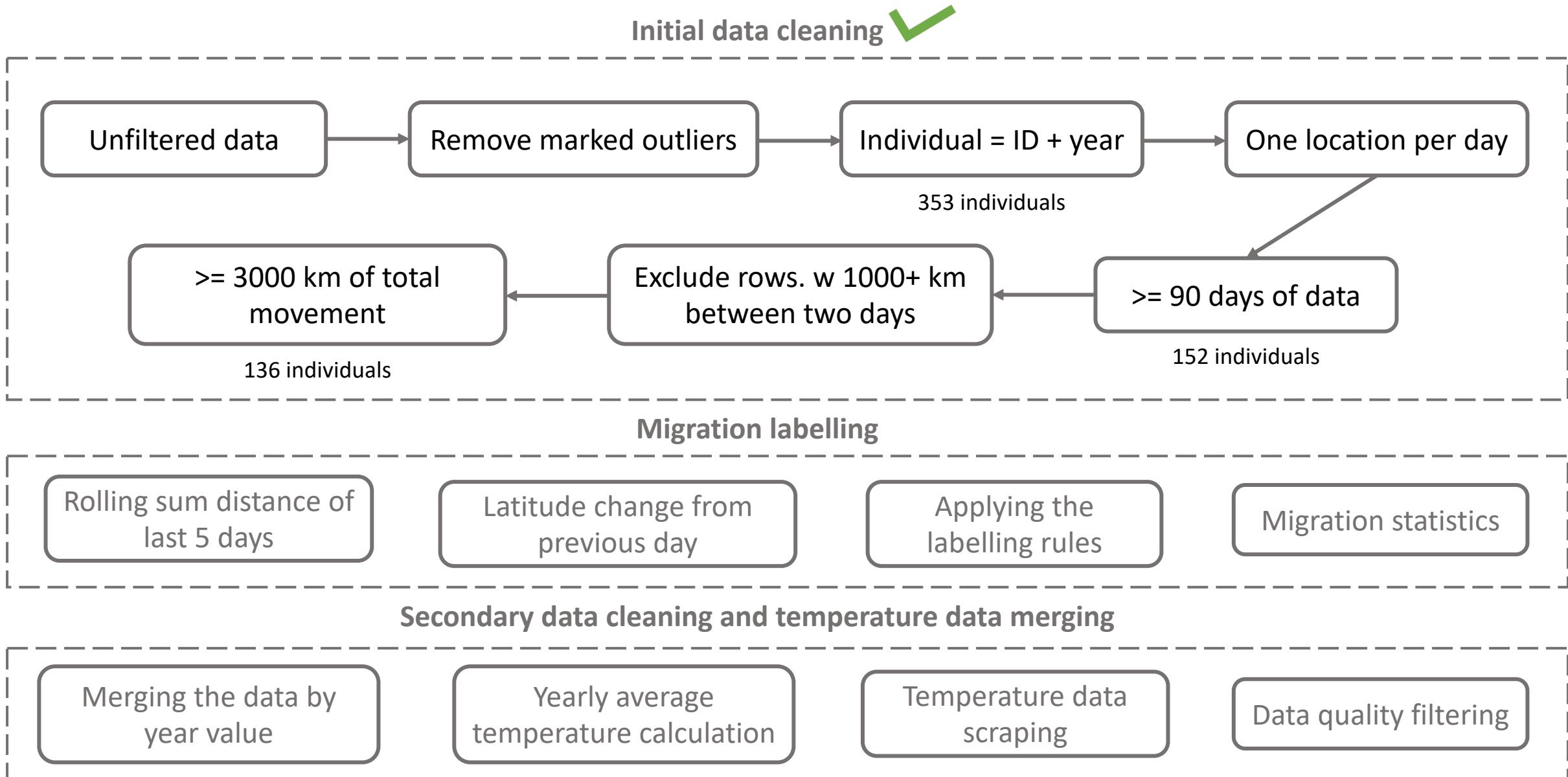
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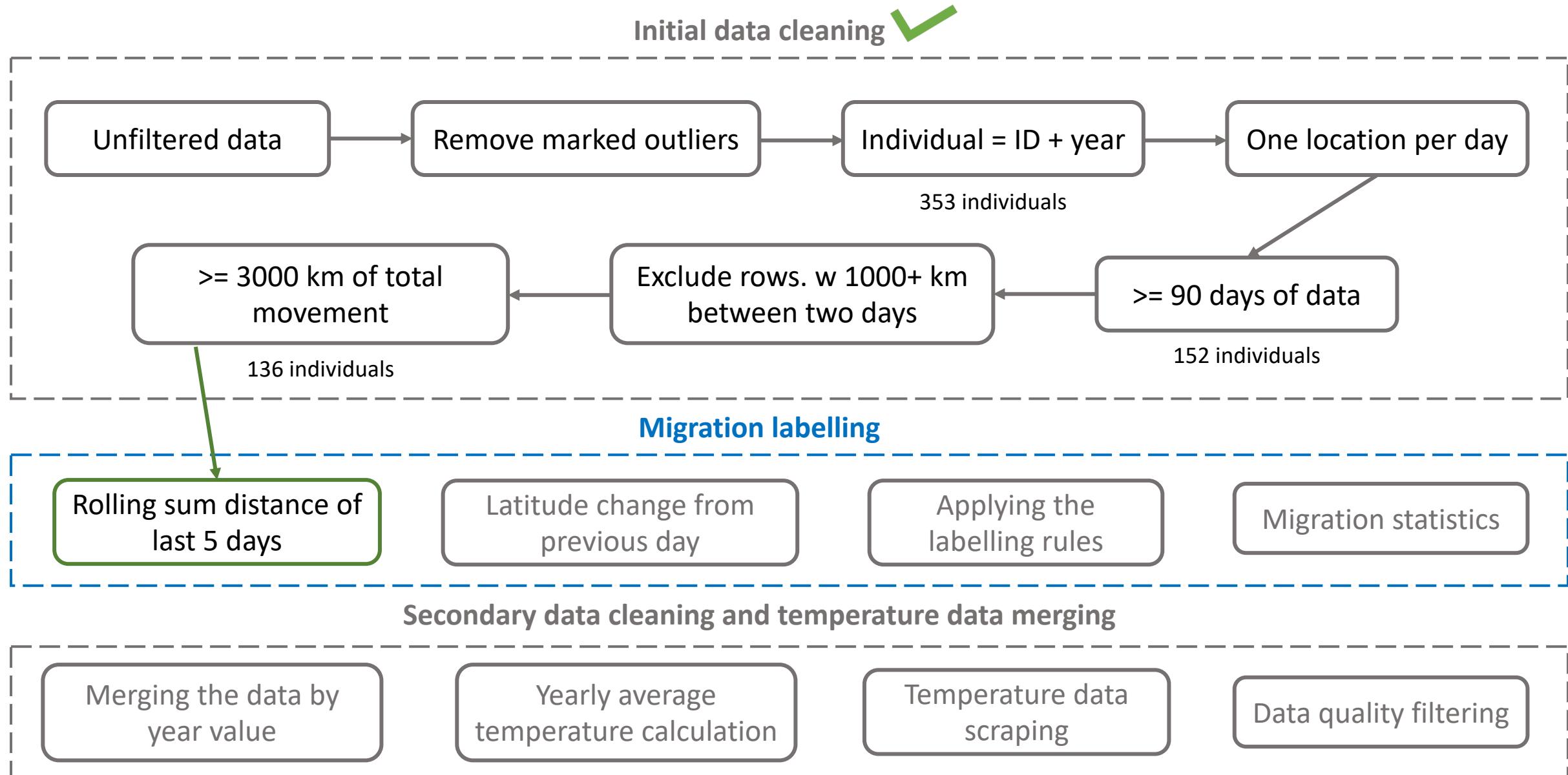
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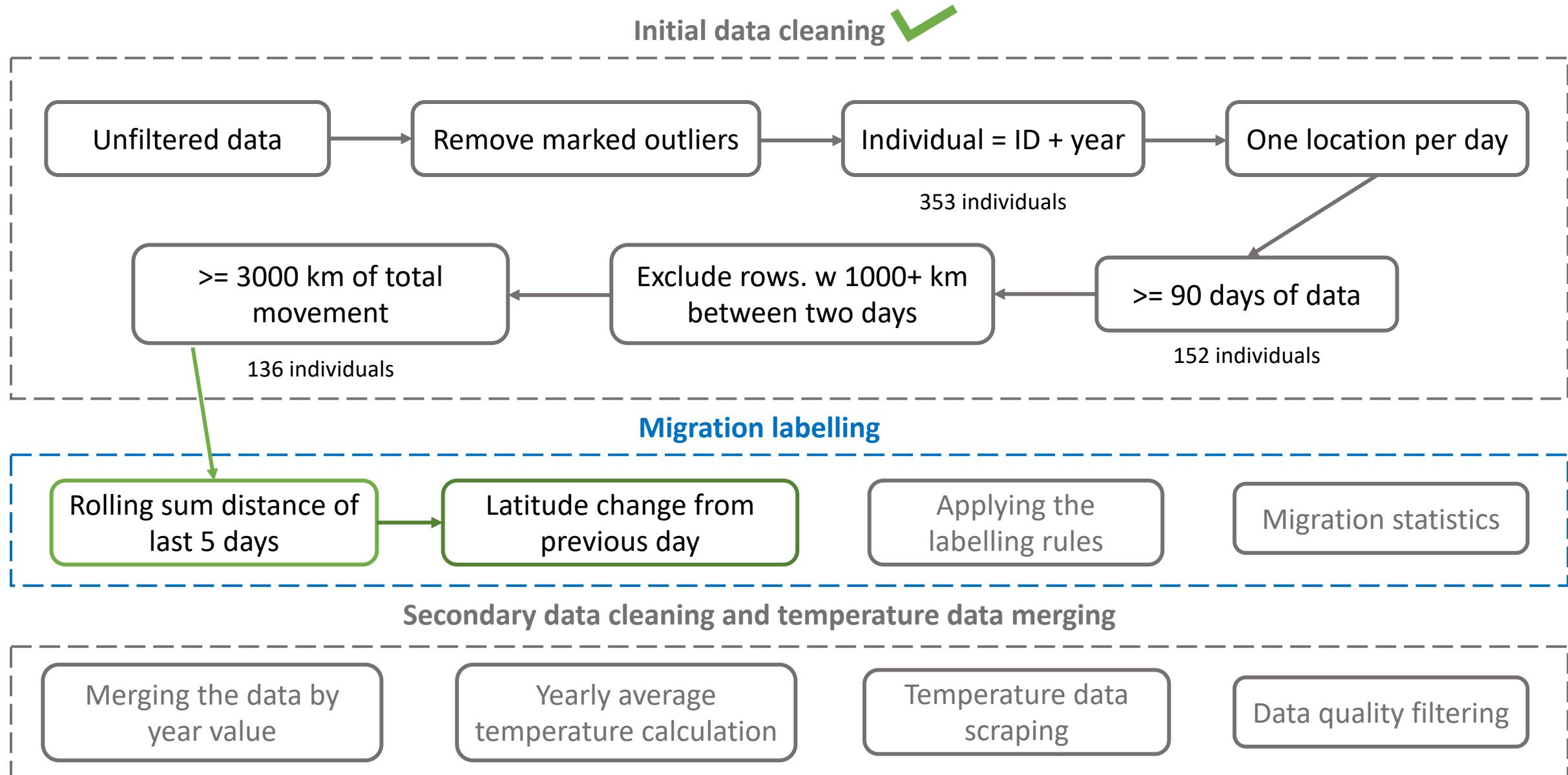
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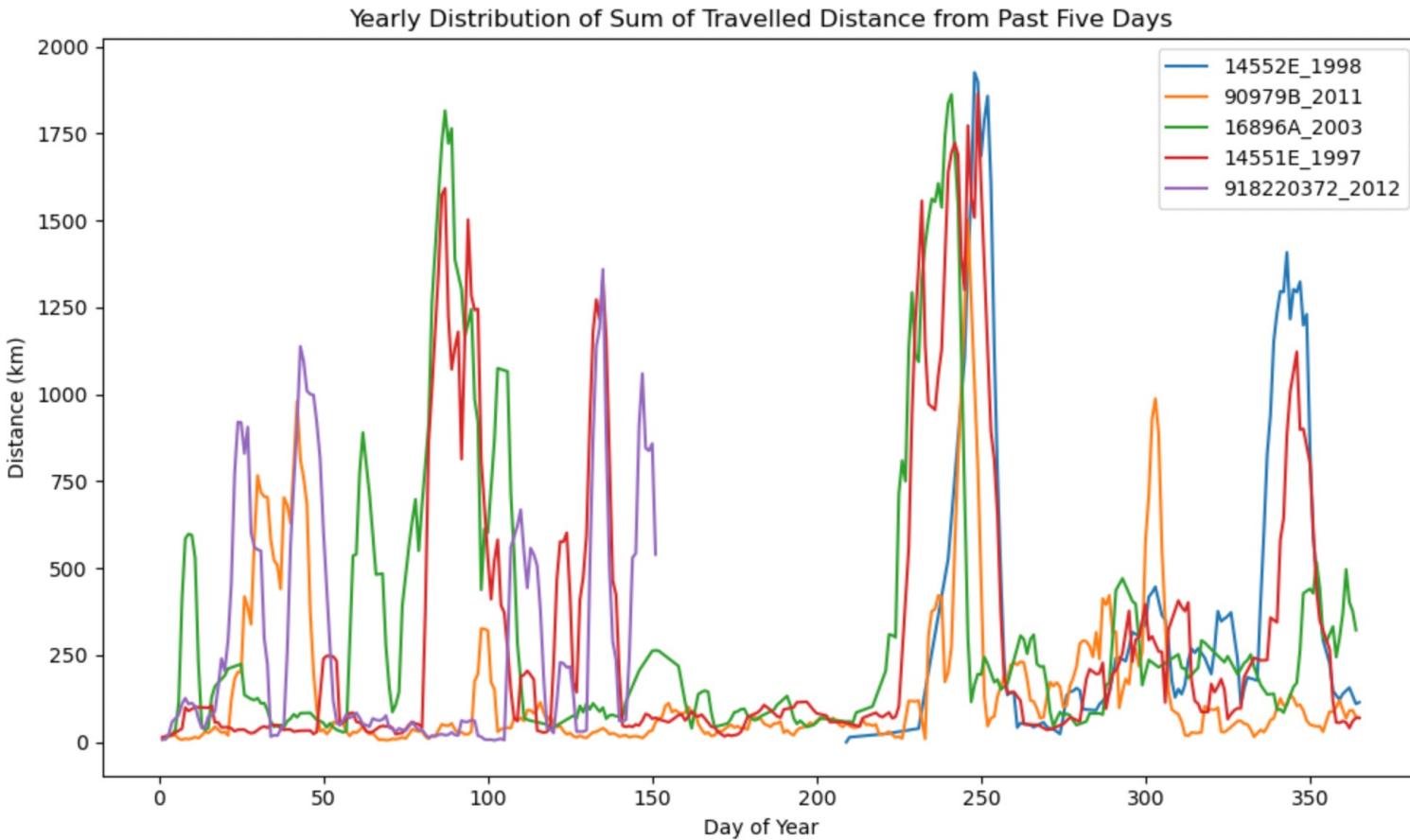
# Data Wrangling Roadmap



# Data Wrangling Roadmap



# Migration Labelling



How to “find” the **summer** and **winter** migration windows?

# Migration Labelling

```
for ind in data.id_year.unique():
```

```
    for loc in data[data.id_year == ind]:
```

Migration already started? — YES → Distance sum of the past 5 days  $\geq 100\text{km}$ ? — YES → Do nothing



Distance sum of the past 5 days  $\geq 500\text{km}$ ?

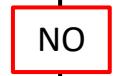


Do nothing

All latitude changes in the past 5 days **positive**?



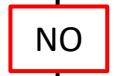
Start **summer** migration



All latitude changes in the past 5 days **negative**?

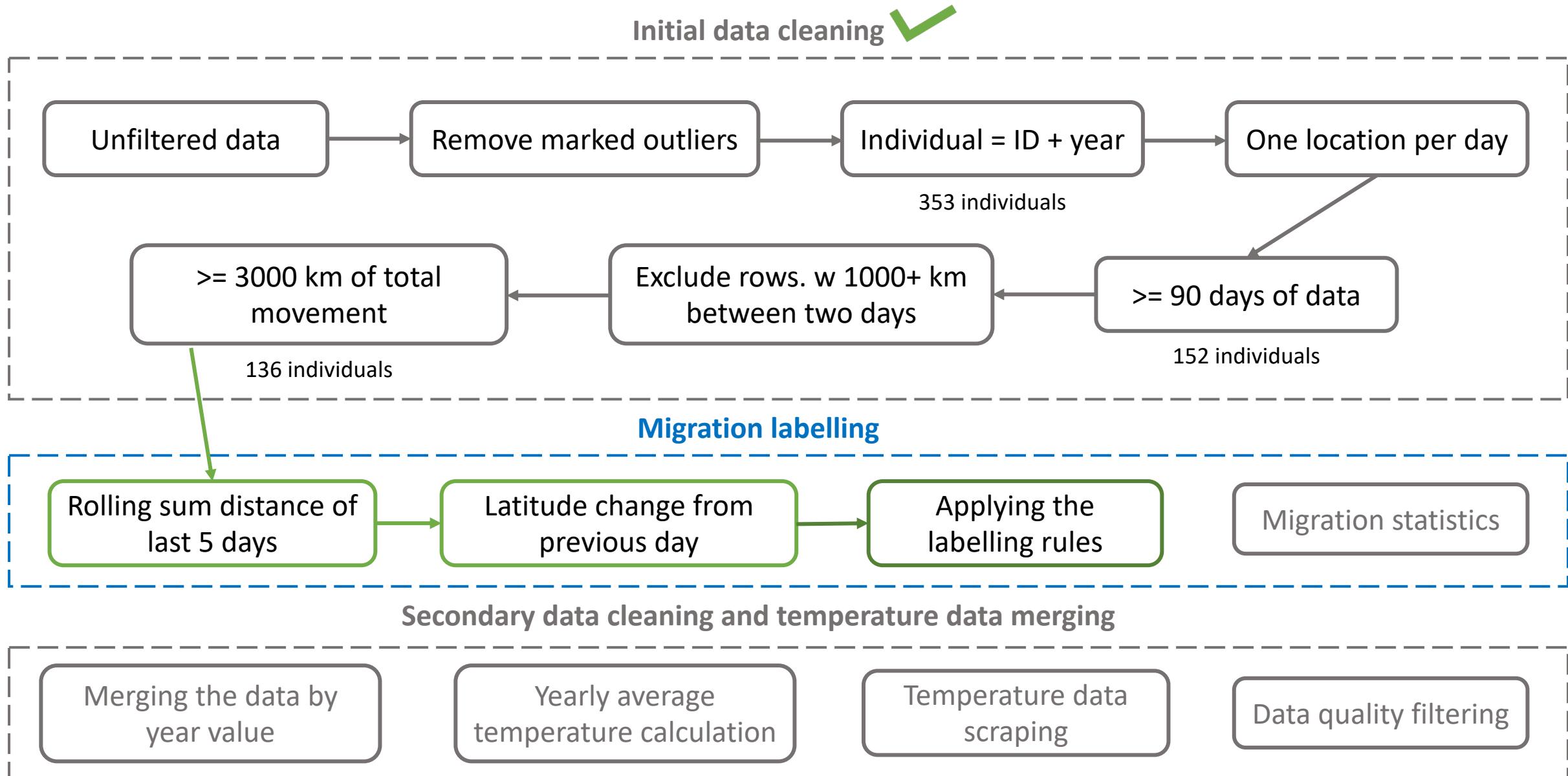


Start **winter** migration



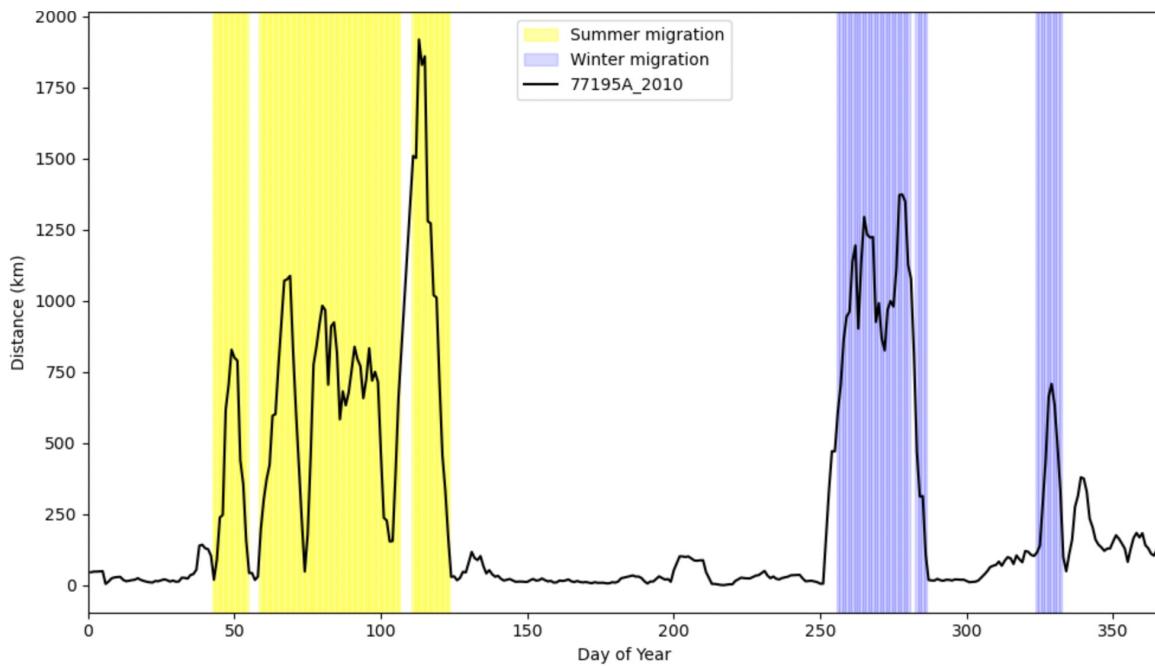
Do nothing

# Data Wrangling Roadmap

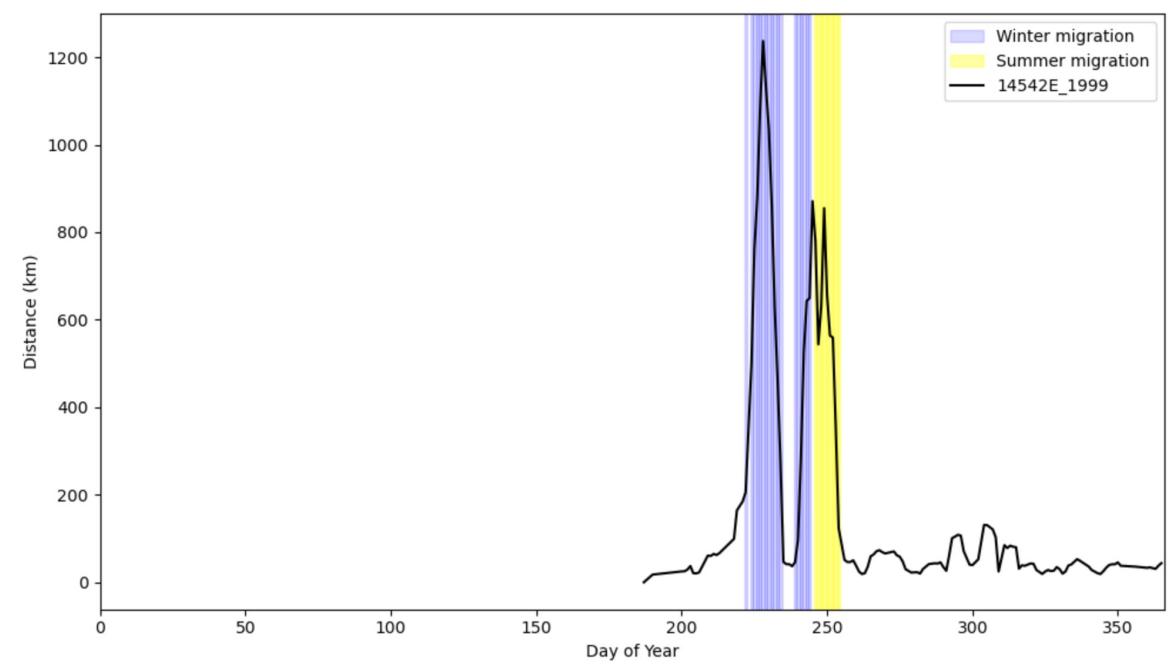


# Migration Labelling

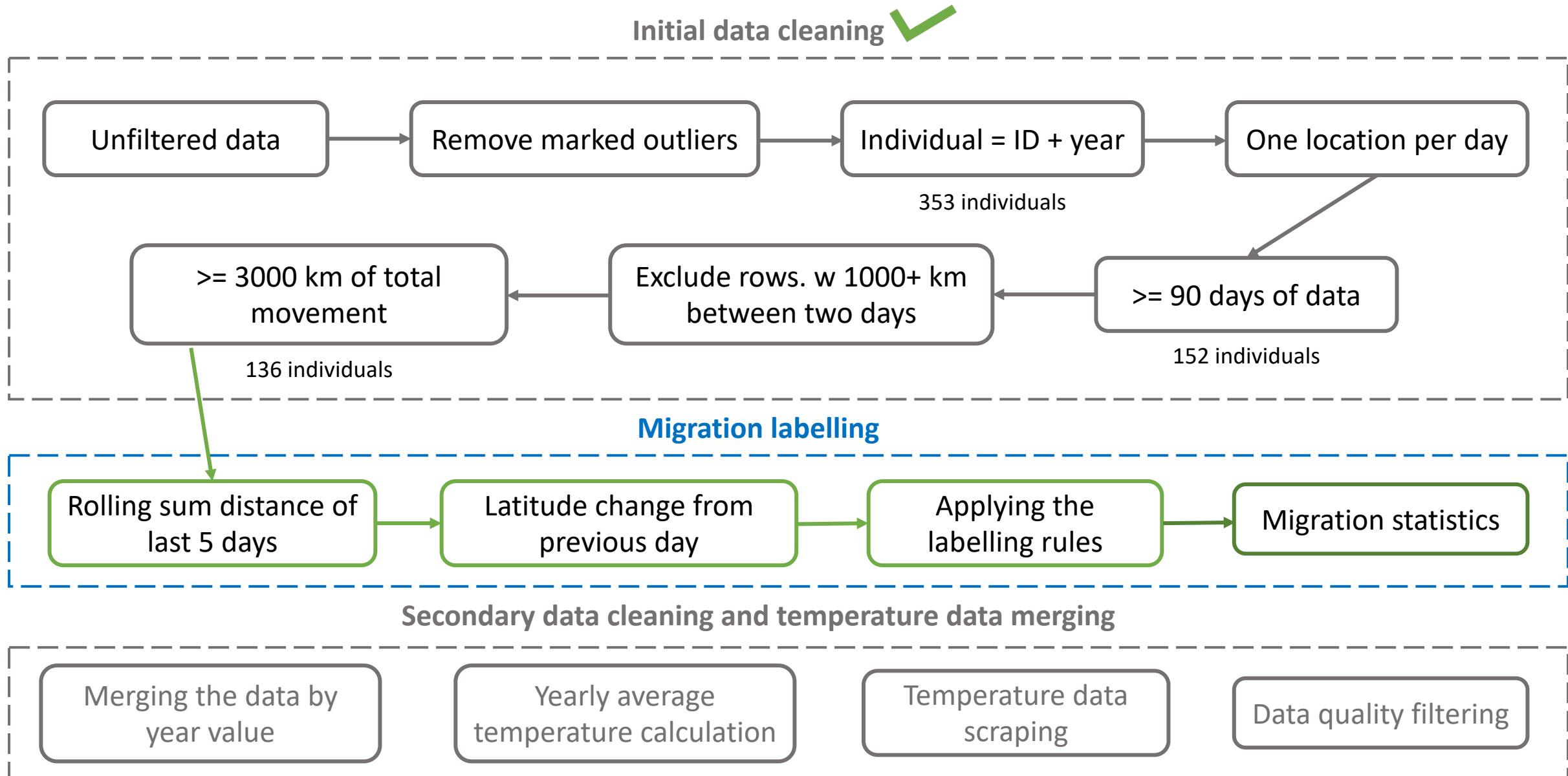
**Successful labelling**



**Failed labelling**



# Data Wrangling Roadmap



# Migration Statistics

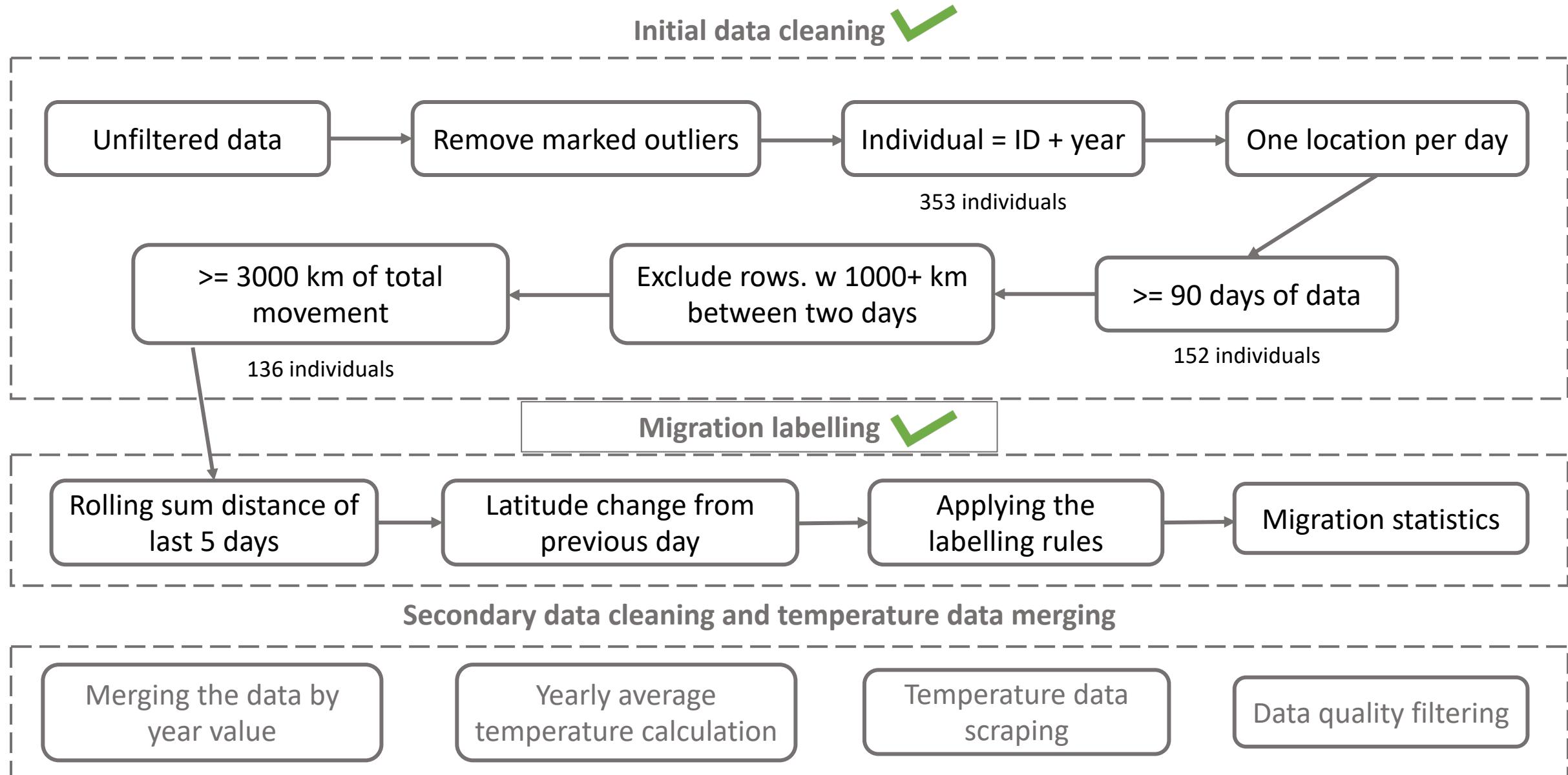
This step involves characterising different aspects of the migration windows of each individual

Further time series quality filtering and data analysis will be based on variables calculated here

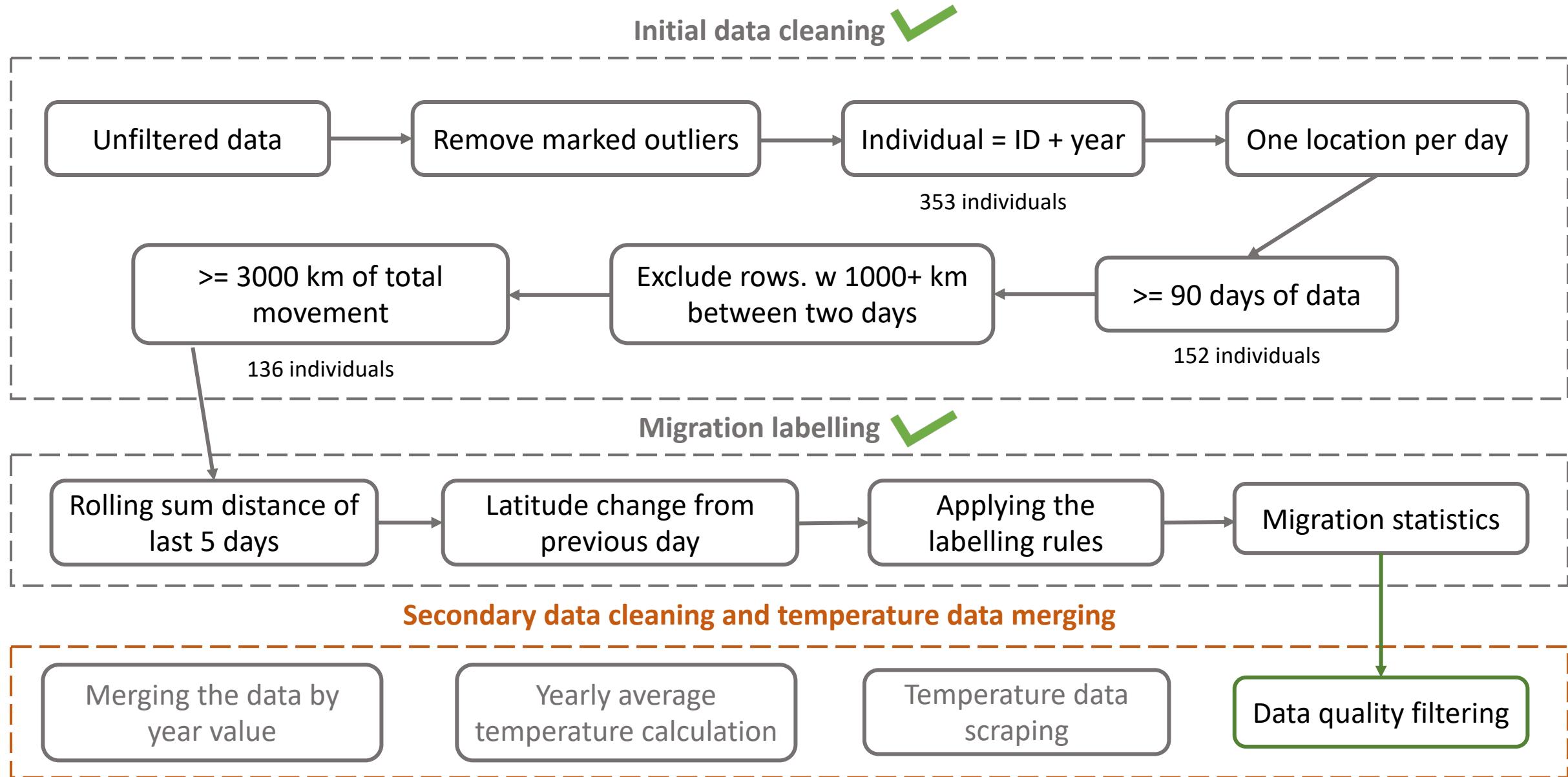
## Migration variables:

- Total migration period in days (**S** & **W** separate)
- First day of migration (**S** & **W** separate)
- Last day of migration (**S** & **W** separate)
- Number of migration windows (**S** & **W** separate)
- Length of the shortest window (**S** & **W** separate)
- **S** window before **W** window (0/1)

# Data Wrangling Roadmap



# Data Wrangling Roadmap



# Data Quality Filtering

How to filter out the timeseries where migration labelling “worked”?

**High Quality (HQ) timeseries:**

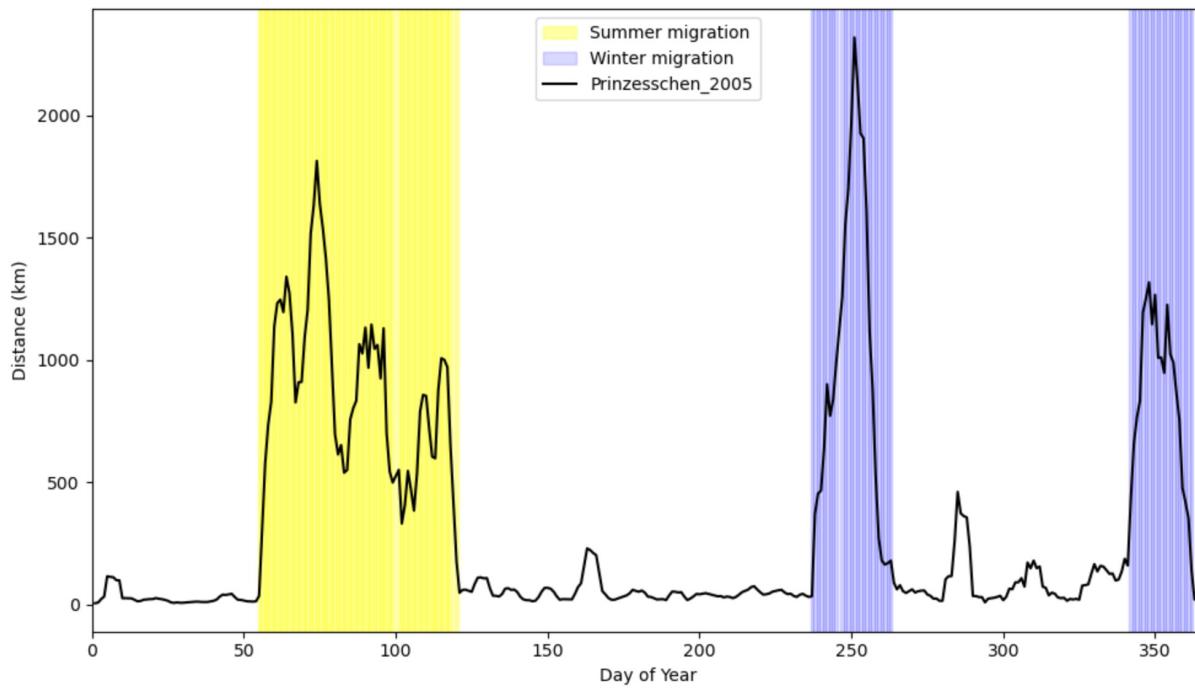
```
'S_before_W' == 1 &  
'shortest_S_window' >= 14 &  
'shortest_W_window' >= 14 & → 24 individuals  
'first_S' <= (366/2) &  
'first_W' >= (366/2)
```

**Medium Quality (MQ) timeseries:**

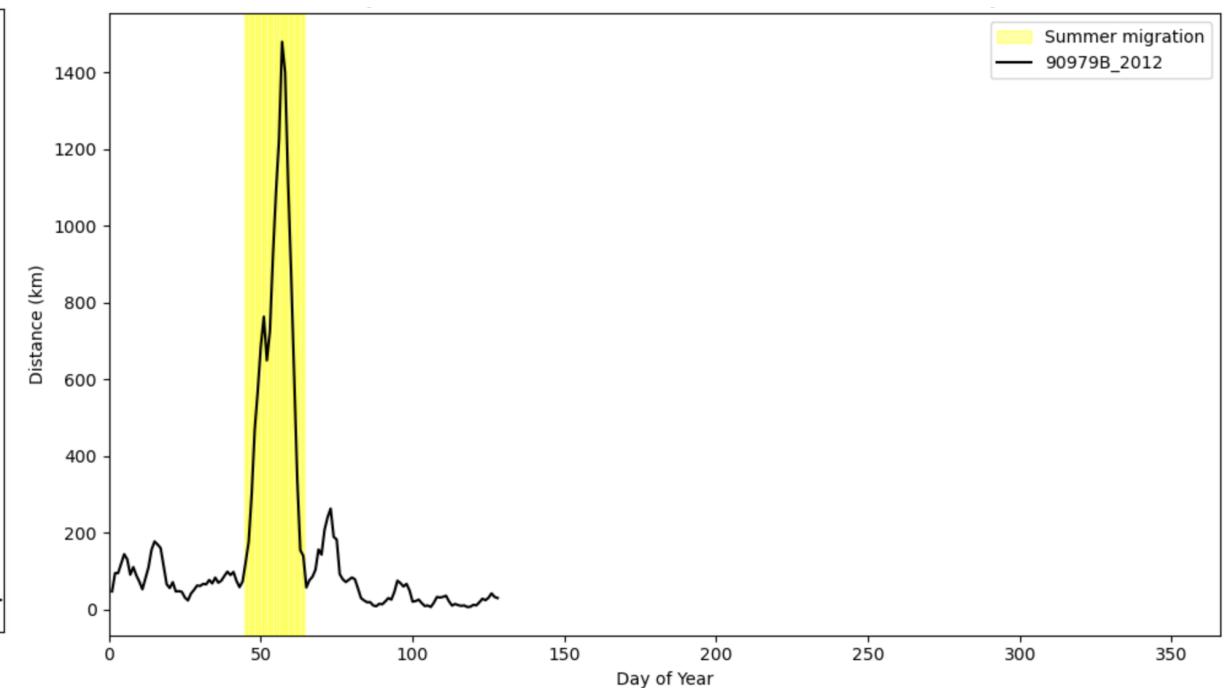
```
'S_before_W' == 1 &  
'shortest_S_window' >= 14 | 'shortest_S_window' == -1 &  
'shortest_W_window' >= 14 | 'shortest_W_window' == -1 & → 35 individuals  
'first_S' <= (366/2) | first_S == -1 &  
'first_W' >= (366/2) | first_W == -1
```

# Data Quality Filtering

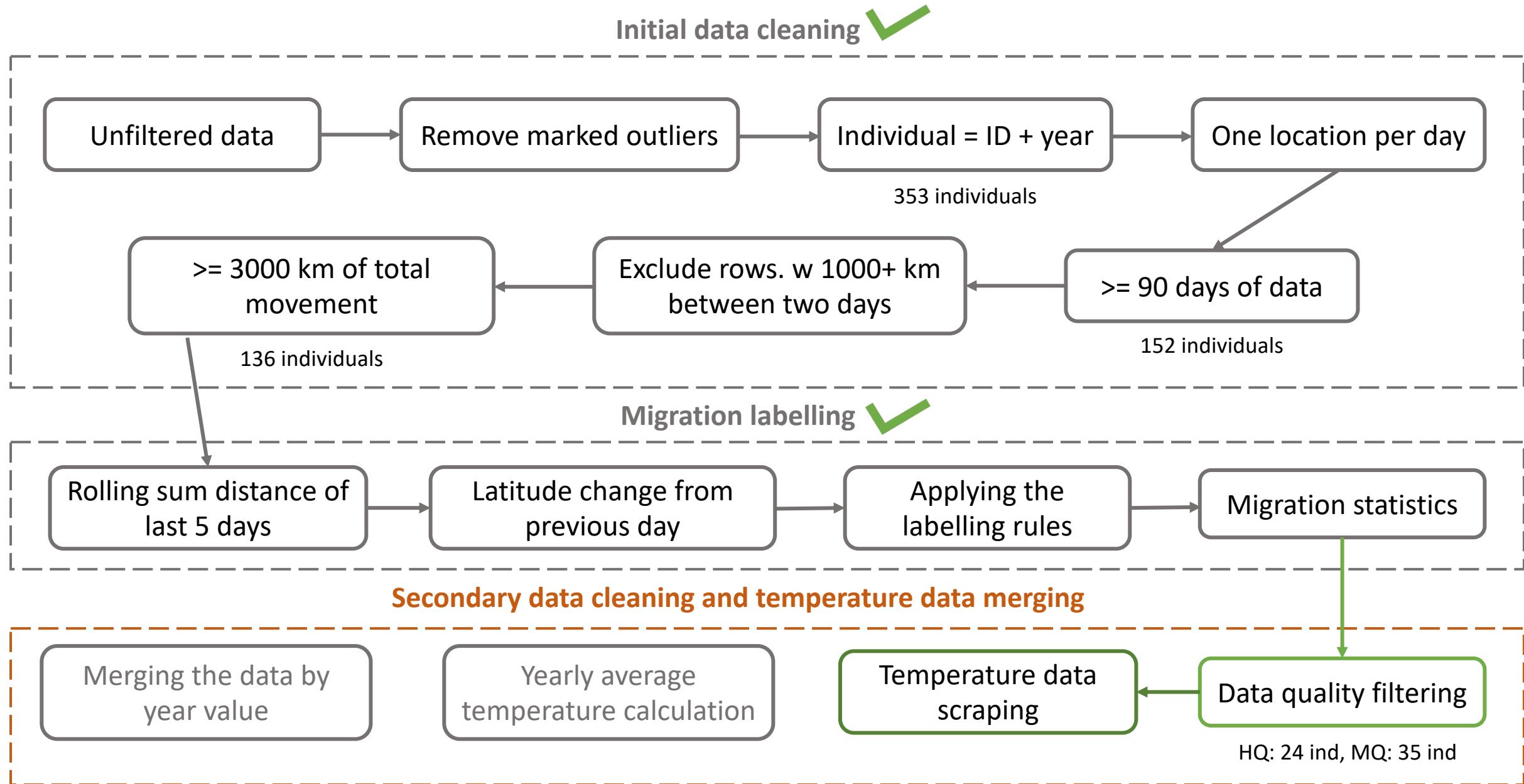
**High Quality** timeseries



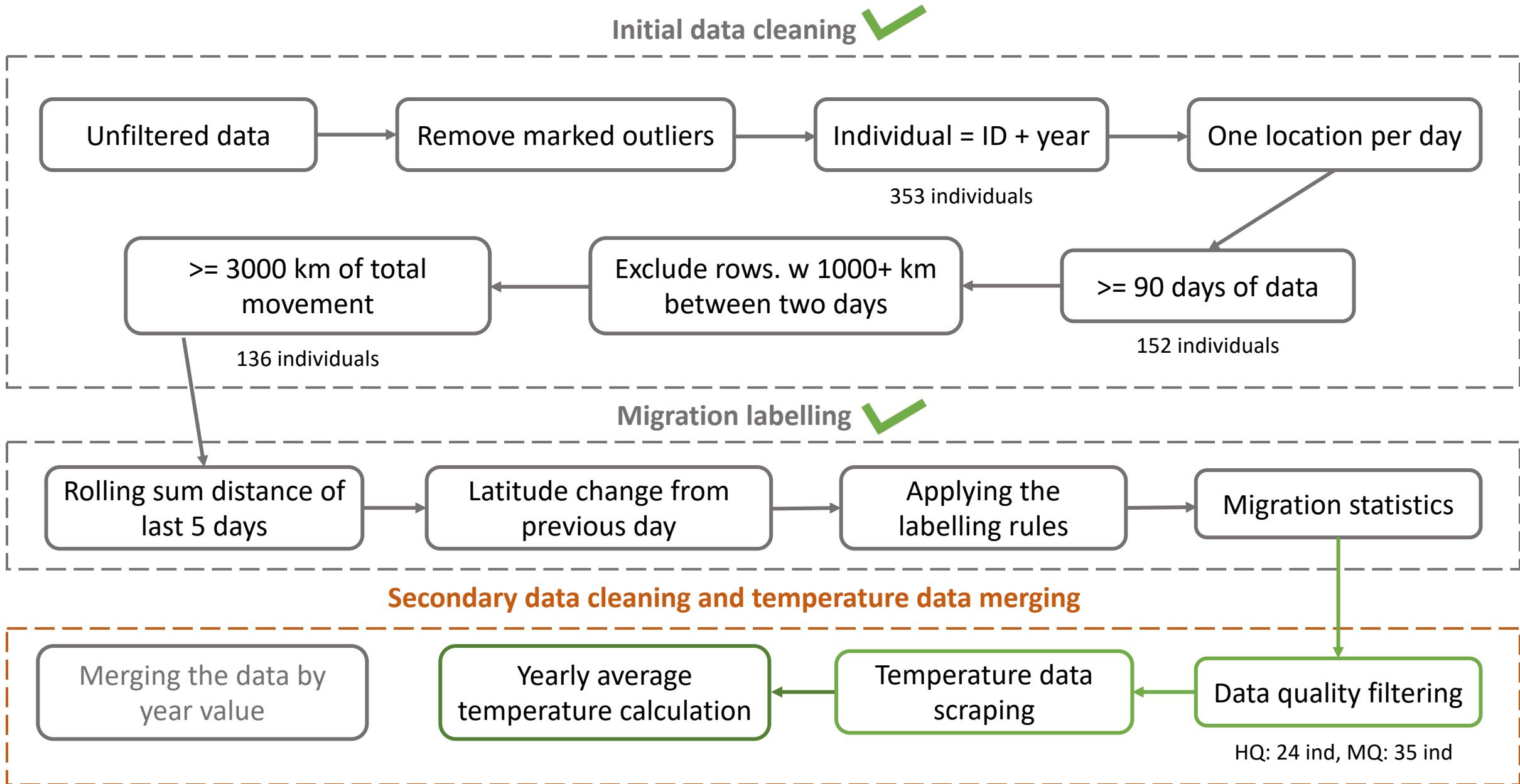
**Medium Quality** timeseries



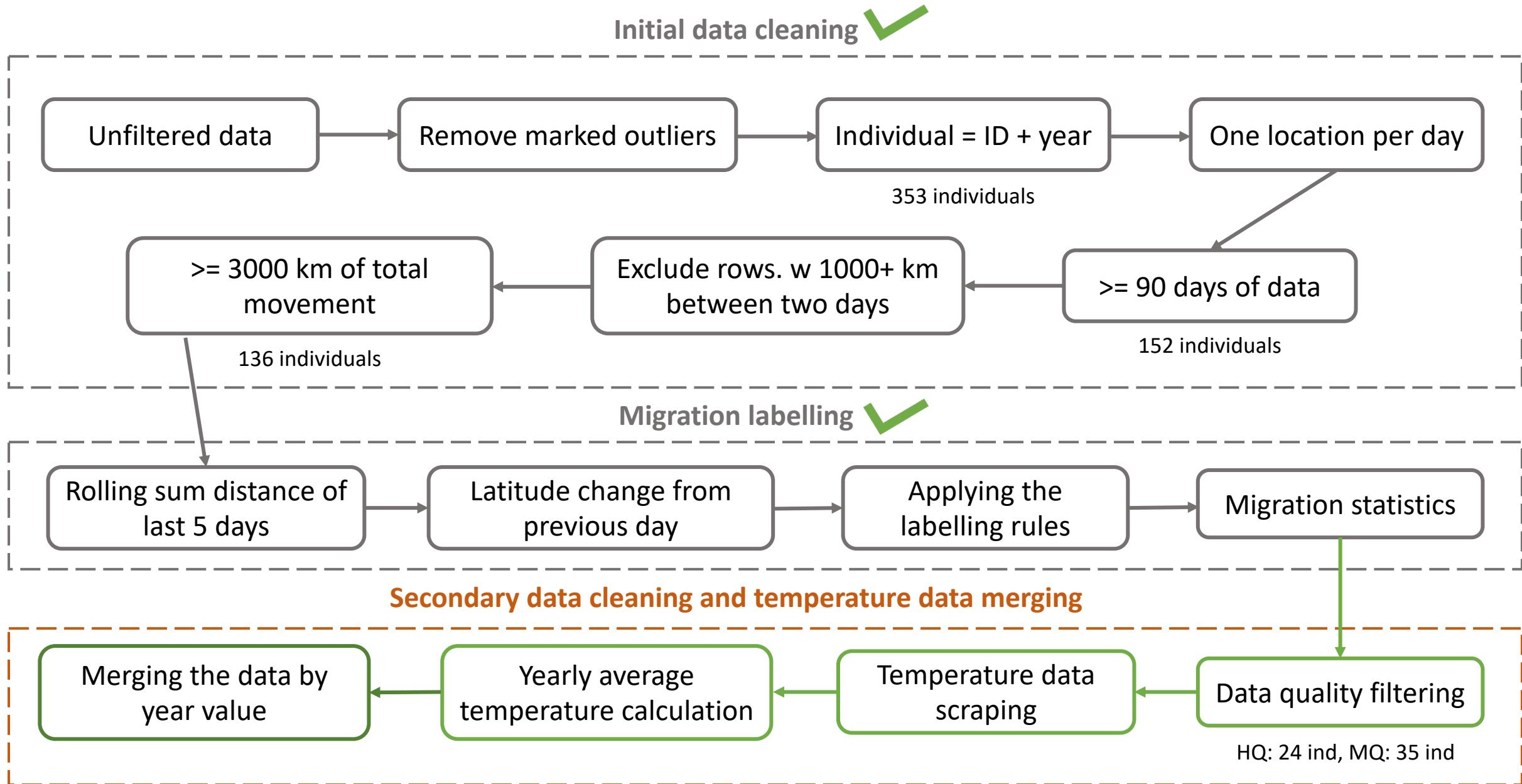
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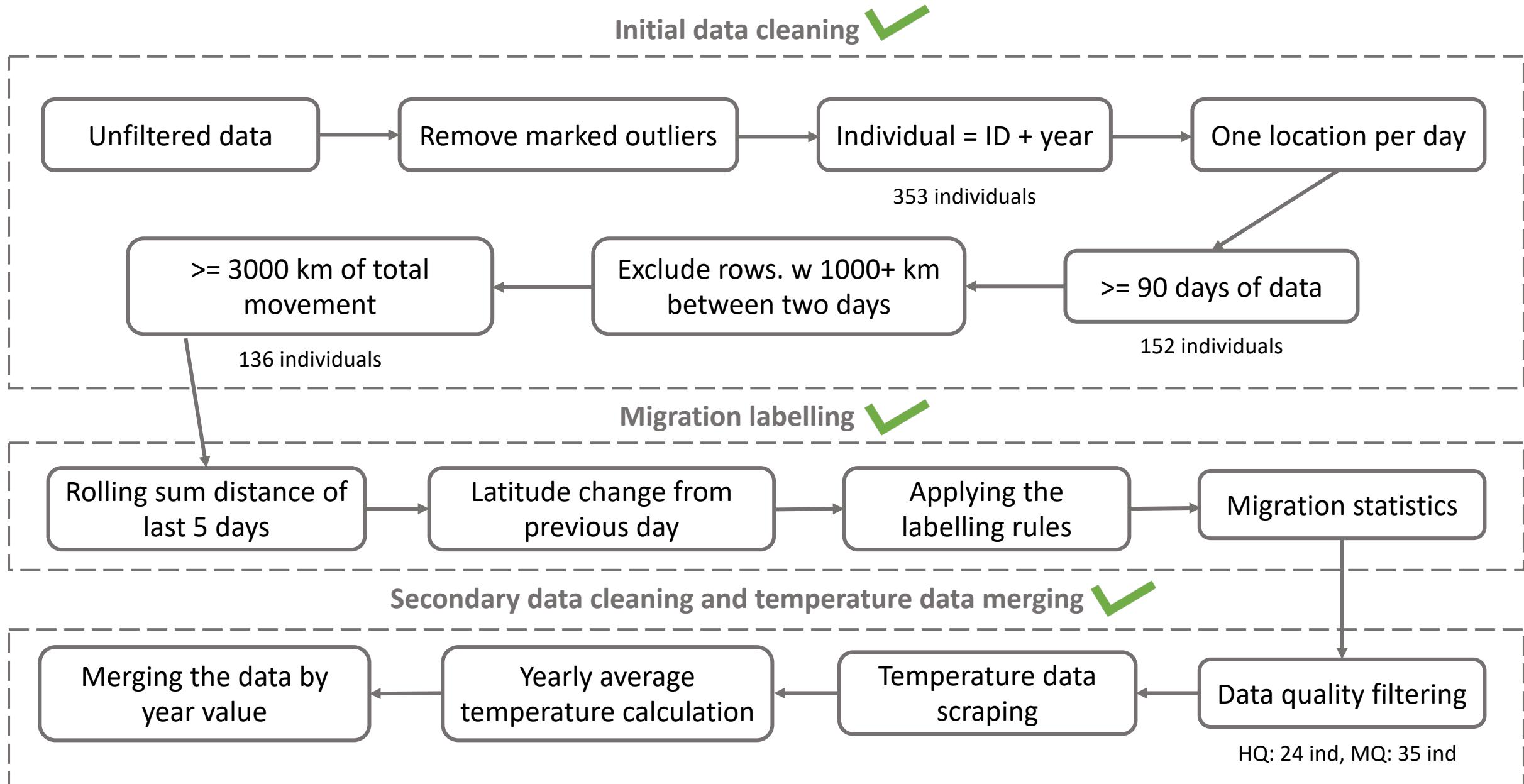
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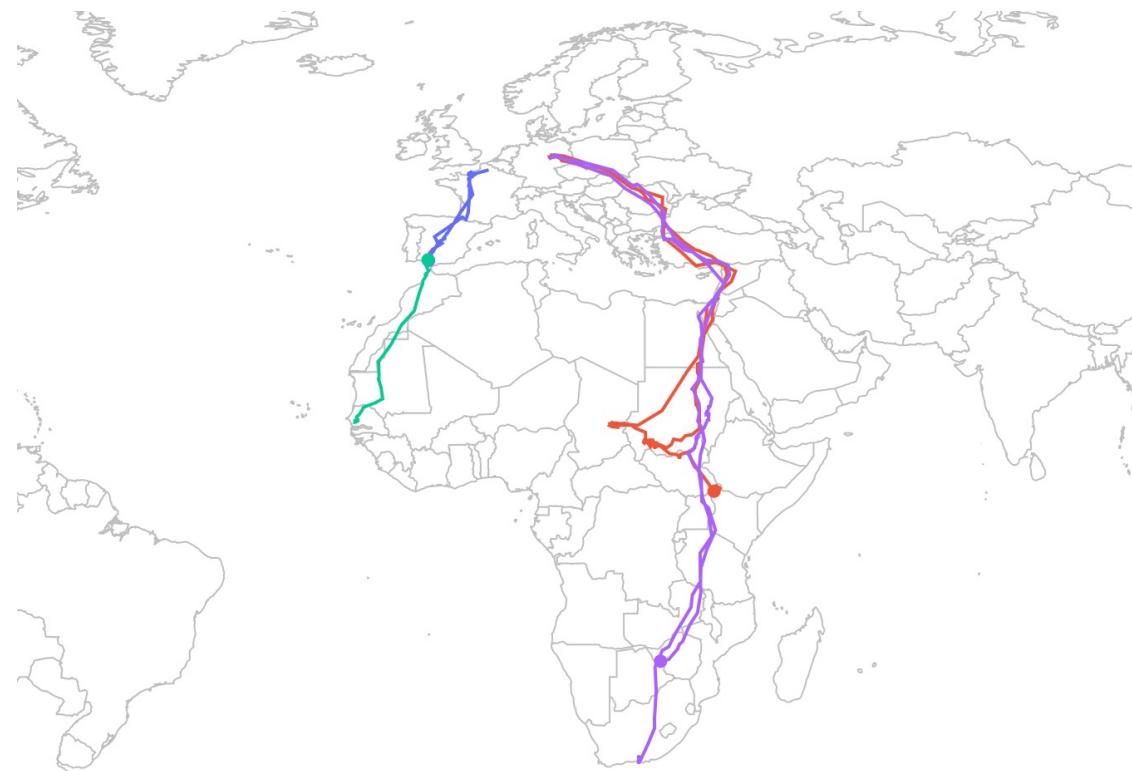
# Data Wrangling Roadmap



# Migration Path Clustering

White storks use **multiple** migration paths

Choice of the path can have a considerable effect on the  
migration behaviour parameters



How to cluster line objects meaningfully?

# Migration Path Clustering

Calculate **overview features** of the Geotrail objects (individual lines)



total distance, maximum/minimum latitude & longitude



Standardize the features & apply **K-means** clustering  
(use elbow plot to determine the k-value)

**HQ** data



$K = 4$

**MQ** data



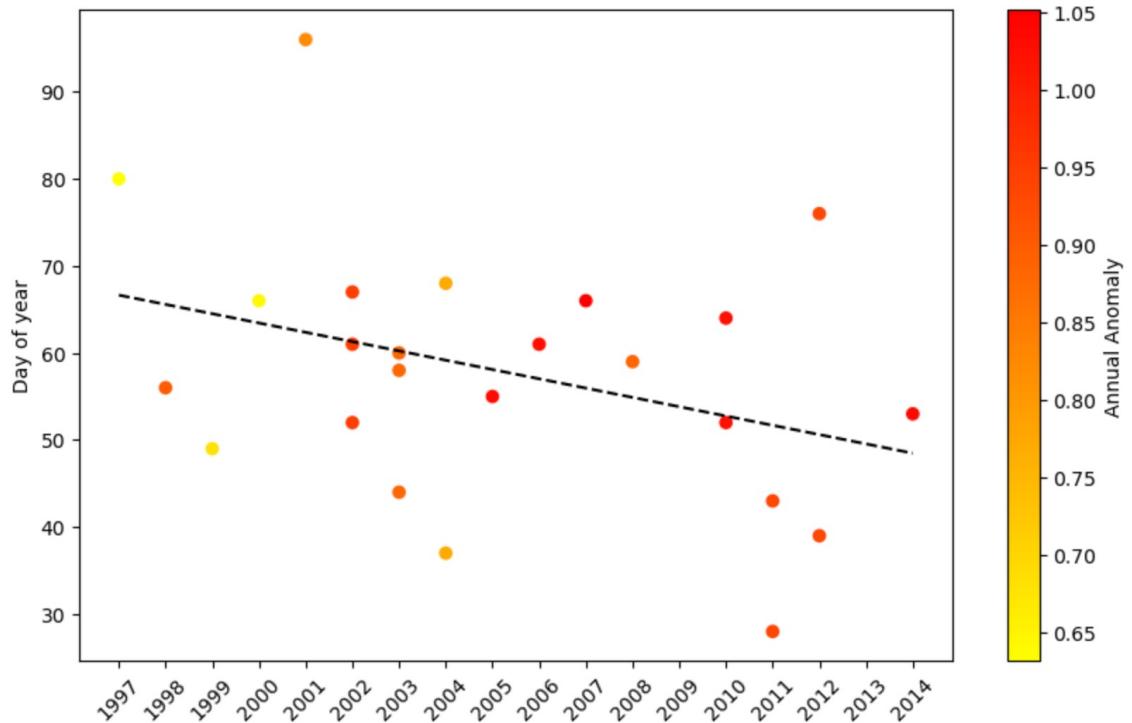
$K = 2$

# Correlation Analysis

Start of migration x Year  
HQ data

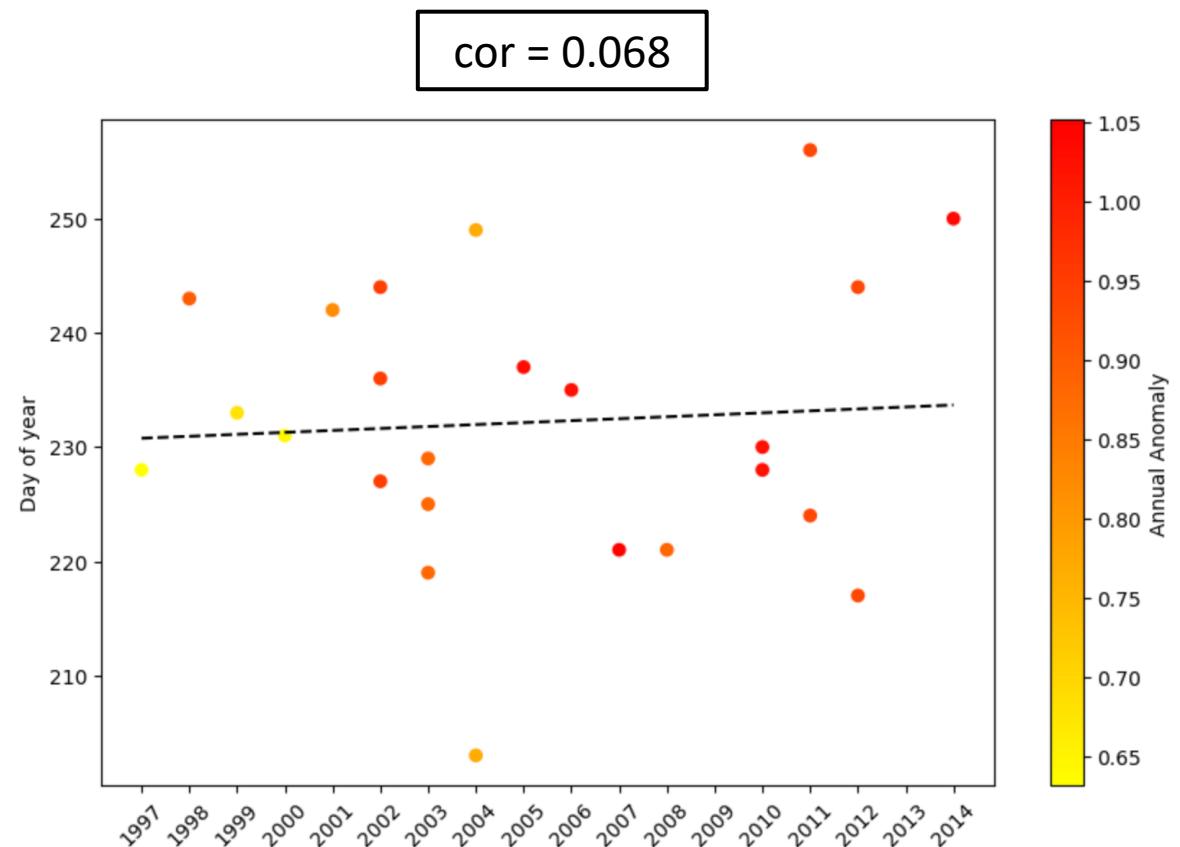
Summer migration

cor = -0.355



Winter migration

cor = 0.068

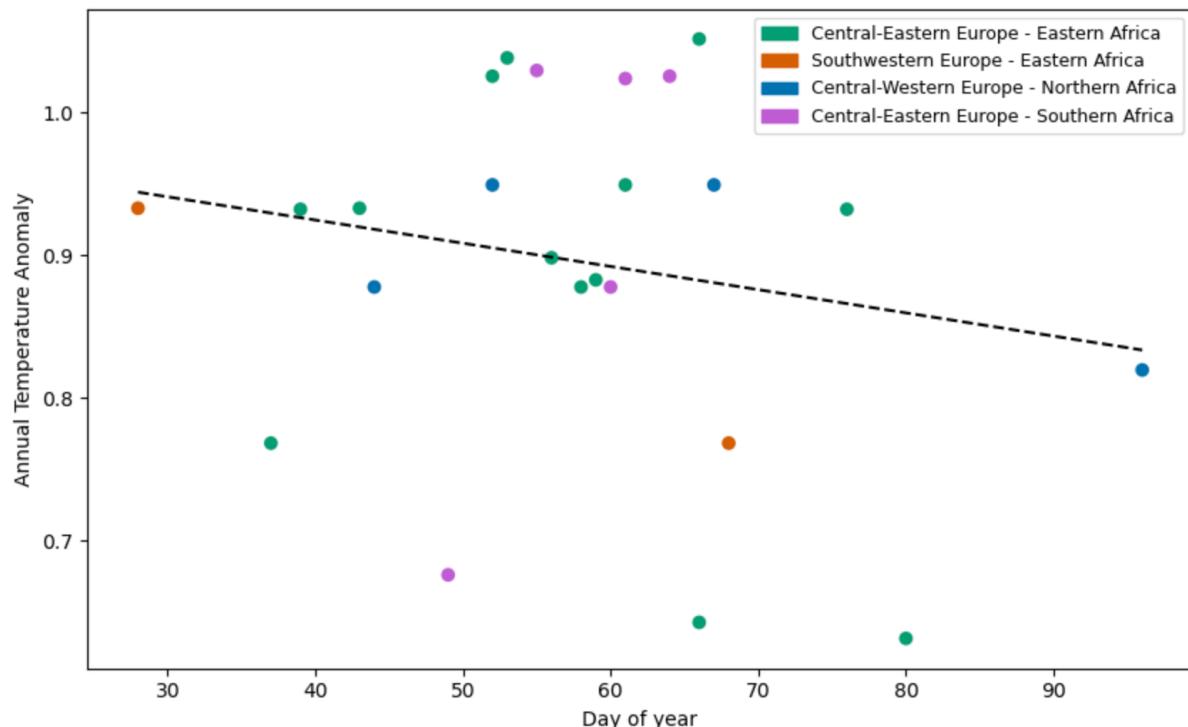


# Correlation Analysis

Start of migration x Temperature anomaly  
HQ data

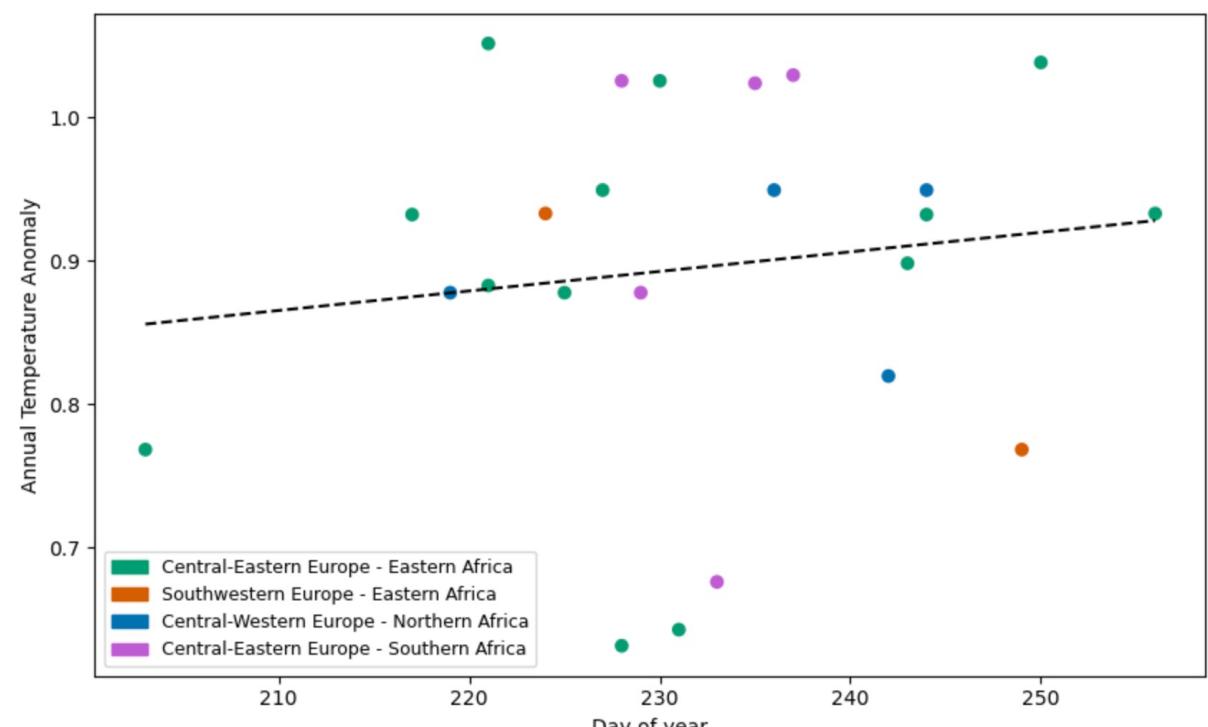
Summer migration

cor = -0.192



Winter migration

cor = 0.135



# Statistical Analysis?

Is there a statistically significant relationship between **migration starting times** and **temperature anomalies**?

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Only **24** (or 35) **individuals**

**16** different **test groups**

**1-3 individuals** per group

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# Conclusion

Migration behaviour change tendencies can be observed over the years on the tracked white storks

Individuals tend to start their **summer** migration **earlier** and begin the **winter** migration **later**

But it must be kept in mind that (weak) correlation **≠ causation!**

Unfortunately the experiment set up does not allow for a reliable statistical analysis...

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## Possible improvements:

- Make the **labelling algorithm** more robust
- Hand-label some individuals to create a training set for a possible **ML labelling solution**
- Make the data quality filtering conditions more realistic
- Find more optimal methods for migration path clustering
- Use **location-based** temperature data
- Gather more quality GPS data(!)
- Discuss the approaches with someone who knows about white stork **phenology**



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