

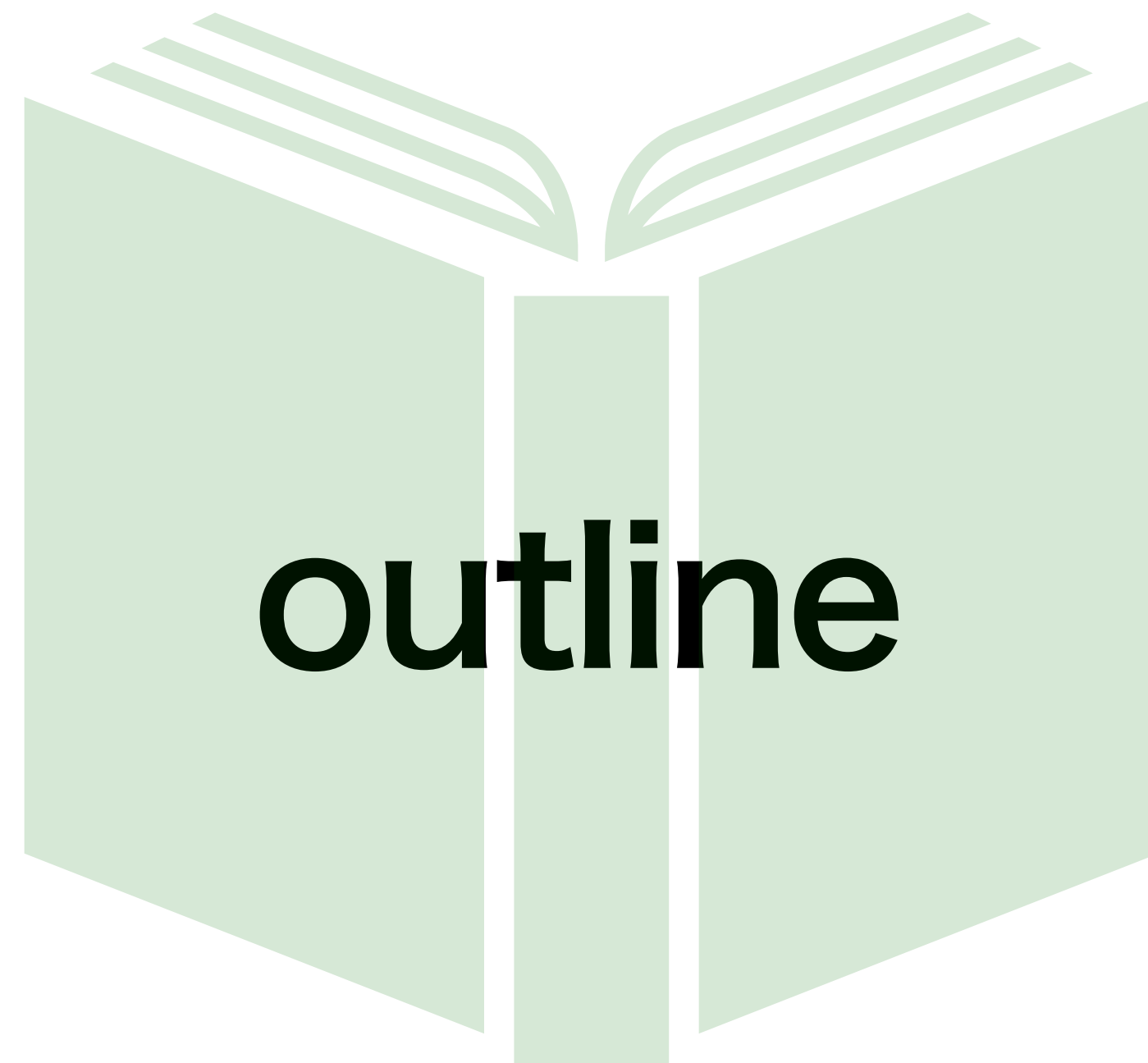
# **Classify Japan climatic regions using cluster analysis**

Research Project A

Computer Science and Engineering, 3rd grade

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Wednesday, 29, September 2020



1. Background
2. Process
3. Make suitable data for clustering
4. Link latitude and longitude
5. Clustering
6. Discussion
7. Plot
8. Future Work

# 1. Background

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Many researchers classified climate regions in Japan.

A researcher [1] classified  
by using a **year** the **daily-average precipitation** data.

[1] 草薙浩, (2016). 平年日降水量時系列のクラスター分析による日本の9気候地域区分の提案. 天気, 63(1), 5-12.

# 1. Background and Proposal

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- (1) In Japan, the climate changes significantly from **month** to **month**.
- (2) The data of the **daily-average temperature** also seems to be a good factor for using clustering.



Analyze the data of the **daily-average precipitation and temperature** with clustering every **month**.

## 2. Process

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**Make suitable data for clustering**



**Link latitude and longitude of all AMeDAS for plot**



**Do clustering and plot the result**

# 3. Make suitable data for clustering

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Get data (.csv) from JMA [2] (Japan Meteorological Agency)

- (1) Original data have **much unnecessary information**.
- (2) The size of the data available at one time was **limited**.  
(It needs to extract the data for clustering)
- Make some C program and shell script  
to remove unnecessary one and to extract the data of  
**the daily-average precipitation, temperature, date & place name.**

[2] JMA, 過去の気象データ・ダウンロード.

Retrieved 04, 2020, from <https://www.data.jma.go.jp/gmd/risk/obsdl/index.php>

# 3. Make suitable data for clustering

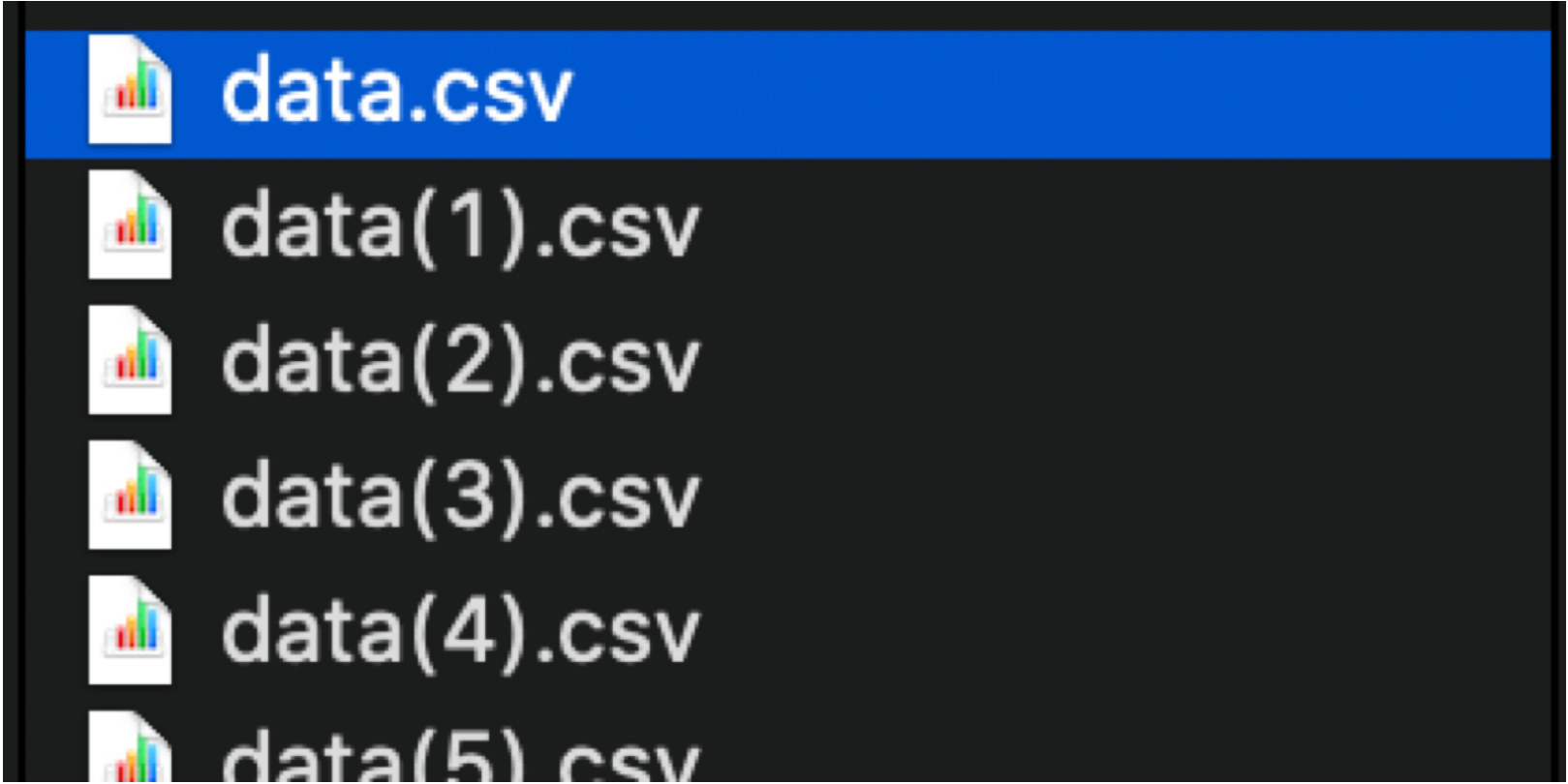


Fig.1 Downloaded CSV file

ダウンロードした時刻：2020/07/20 14:20:31												
	愛知	愛知	愛知	愛知	愛知	愛知	愛知	愛知	愛知	愛知	愛知	愛知
	一宮	一宮	一宮	一宮	一宮	一宮	一宮	一宮	一宮	一宮	豊山	豊山
年月日	降水量の合計(mm)	降水量の合計(mm)	降水量の合計(mm)	降水量の合計(mm)	降水量の合計(mm)	平均気温(℃)	平均気温(℃)	平均気温(℃)	平均気温(℃)	平均気温(℃)	降水量の合計(mm)	降水量
				平年値(mm)	平年値(mm)				平年値(℃)	平年値(℃)		
		品質情報	均質番号		品質情報		品質情報	均質番号		品質情報		品質情報
2019/1/1	0.0	8	1	1.7	8		0	1		0		
2019/1/2	0.0	8	1	1.8	8		0	1		0		

Fig.2 Contents of the CSV file



# 3. Make suitable data for clustering

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フォルダ	フォルダ	書類	
abashiri_kitami_monbetsu	チミケップ山	遠軽_1_1_2_28.txt	遠軽 5/1,1.6,8.0 5/2,1.6,8.2 5/3,1.7,8.4 5/4,1.7,8.5 5/5,1.7,8.7 5/6,1.8,8.8 5/7,1.8,9.0
aichi	宇登呂	遠軽_3_1_4_30.txt	
akita	遠軽	遠軽_5_1_6_30.txt	
aomori	丸瀬布	遠軽_7_1_8_31.txt	
chiba	境野	遠軽_9_1_10_31.txt	
chimo	興部	遠軽_11_1_12_31.txt	

Fig.3 After making suitable data for clustering



# 4. Link latitude and longitude

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Get geographic data from JMA. [3]

- From the **place name**, make a list of the geographic data.

```
北見, キタミ, KITAMI, 43, 46.6, 143, 50.5, 0104  
境野, サカイノ, SAKAINO, 43, 42.4, 143, 38.6, 0184  
常呂, トコロ, TOKORO, 44, 06.9, 144, 02.2, 0003  
斜里, シャリ, SHARI, 43, 53.1, 144, 42.0, 0015  
津別, ツベツ, TSUBETSU, 43, 42.1, 144, 02.0, 0100  
湧別, ユウベツ, YUBETSU, 44, 12.8, 143, 37.1, 0005  
滝上, タキノウエ, TAKINOUE, 44, 10.6, 143, 03.7, 0165  
白滝, シラタキ, SHIRATAKI, 43, 51.9, 143, 09.2, 0475  
紋別, モンベツ, MOMBETSU, 44, 20.7, 143, 21.3, 0016  
網走, アバシリ, ABASHIRI, 44, 01.0, 144, 16.7, 0038  
美幌, ビホロ, BIHORO, 43, 46.2, 144, 10.3, 0060  
興部, オコッペ, OKOPPE, 44, 28.2, 143, 06.5, 0008
```

Fig.4 List of the geographic data

[3] JMA, 観測概要と観測所一覧.

Retrieved 04, 2020, from [https://www.data.jma.go.jp/obd/stats/data/mdrr/man/kansoku\\_gaiyou.html](https://www.data.jma.go.jp/obd/stats/data/mdrr/man/kansoku_gaiyou.html)

# 5. Clustering

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Clustering by using scipy.

- **Normalization** before clustering
- Clustering by **ward** and **euclidean**.

From the dendrogram and the graph of each cluster,  
decide the number of clusters.

# 5. Clustering

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## Dendrogram

- The figure that **integrated similar data** until all data are integrated.

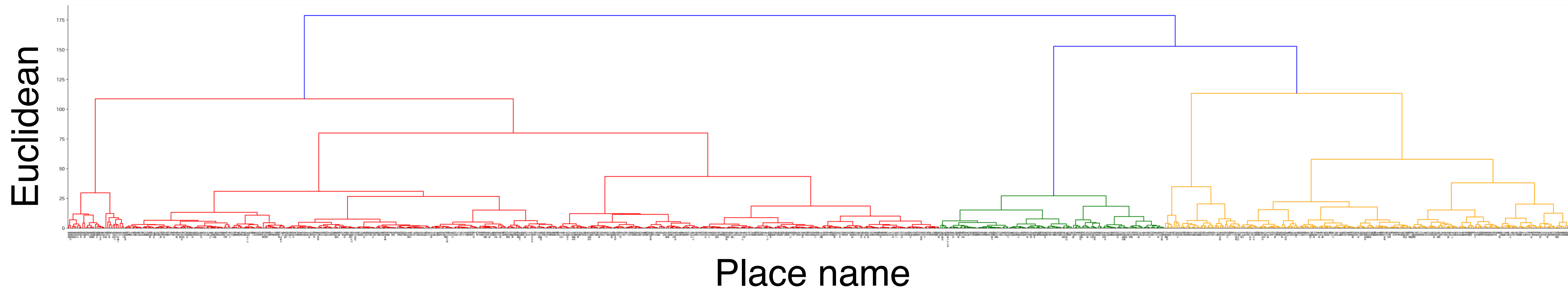


Fig.5 Dendrogram in January

# 5. Clustering

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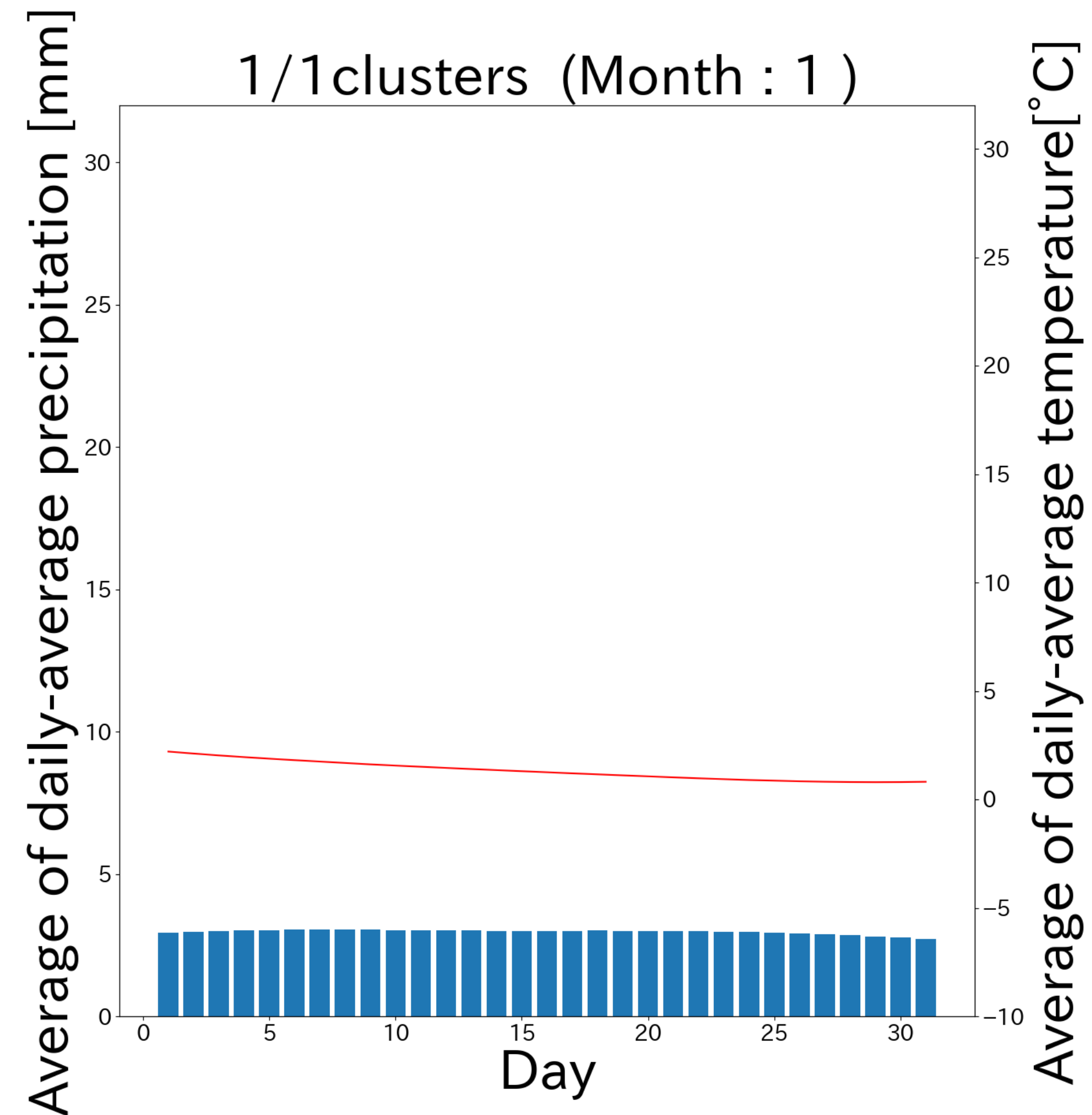


Fig.6 Average score of each cluster in January (1 cluster)

# 5. Clustering

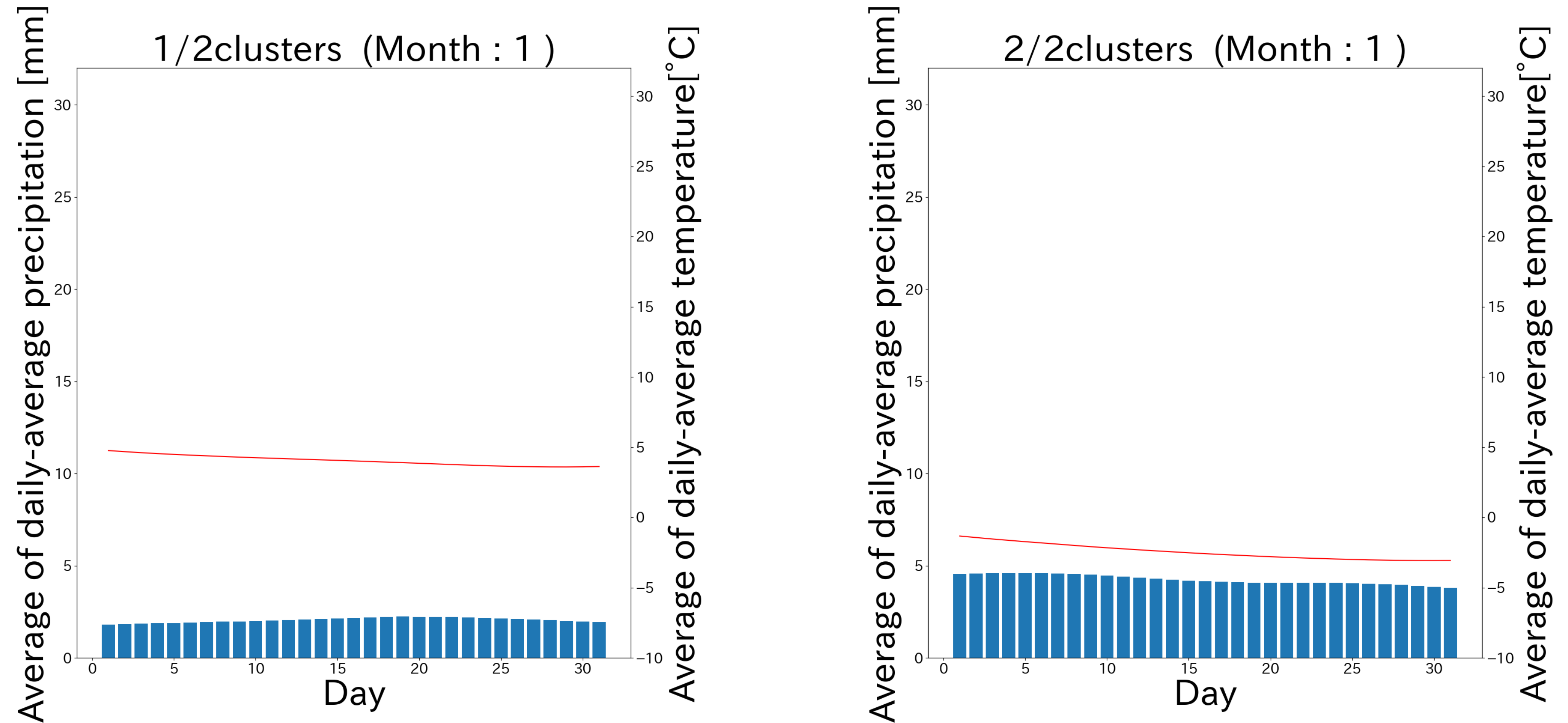


Fig.7 Average score of each cluster in January (2 clusters)



# 5. Clustering

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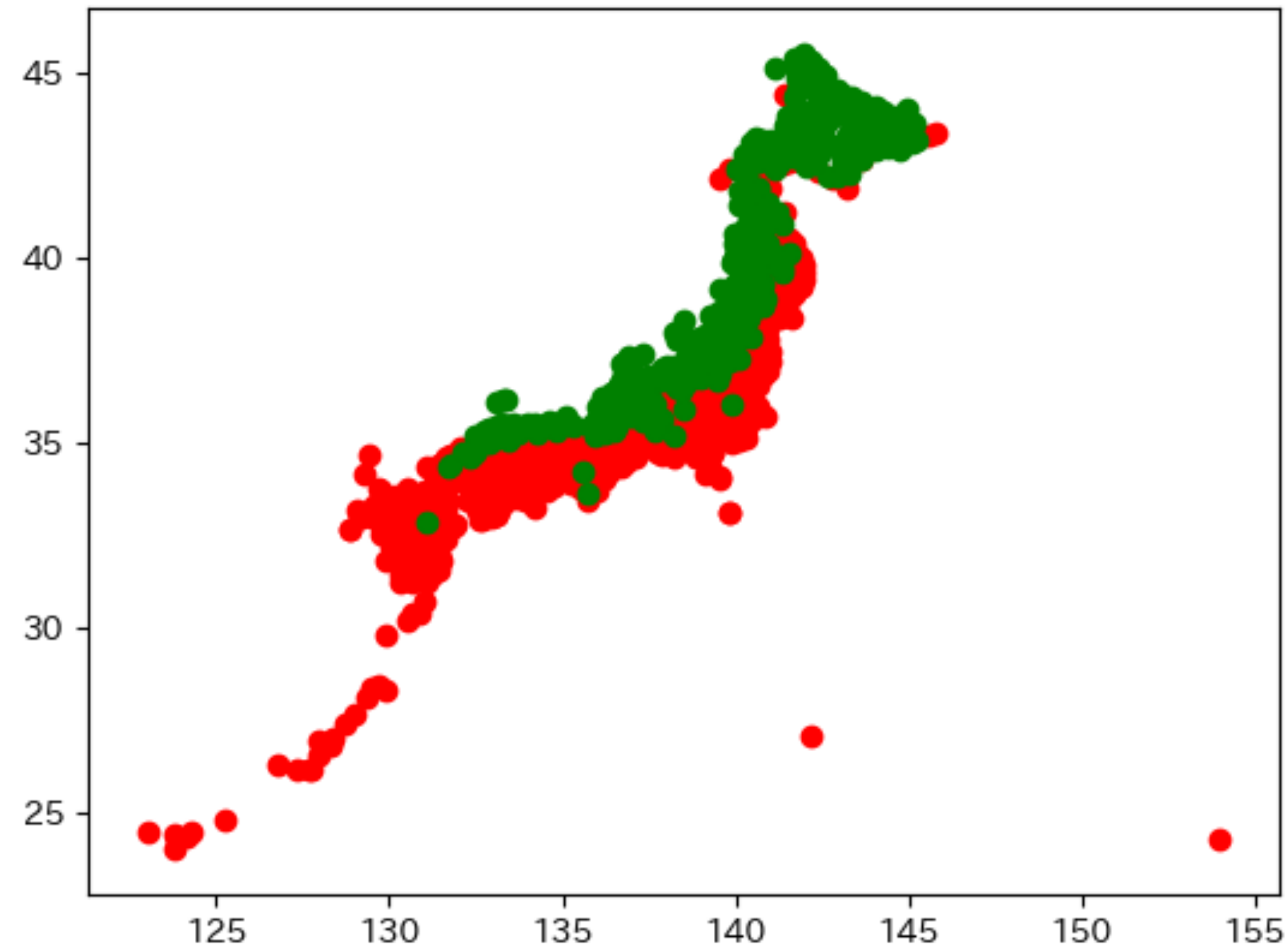


Fig.8 Scatter of each cluster in January (2 clusters)



# 5. Clustering

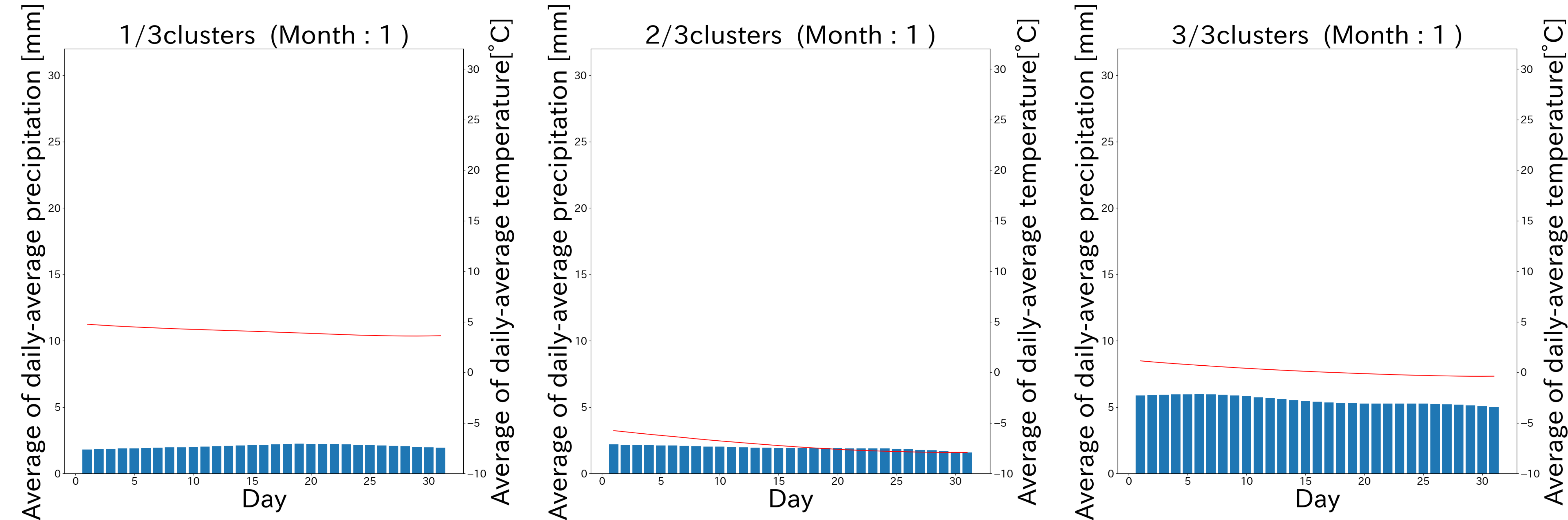


Fig.9 Average score of each cluster in January (3 clusters) 15

# 5. Clustering

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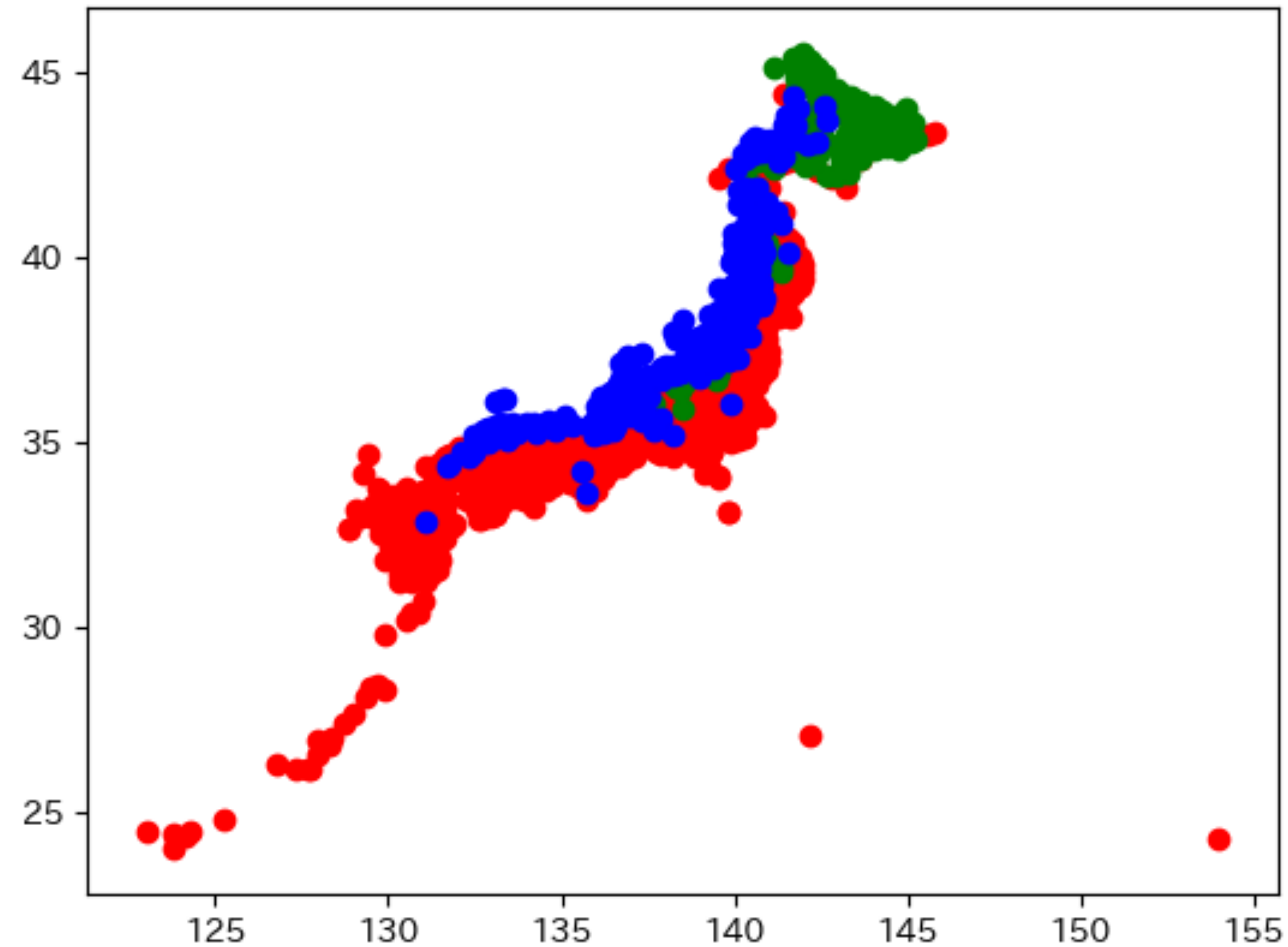


Fig.10 Scatter of each cluster in January (3 clusters)

# 5. Clustering

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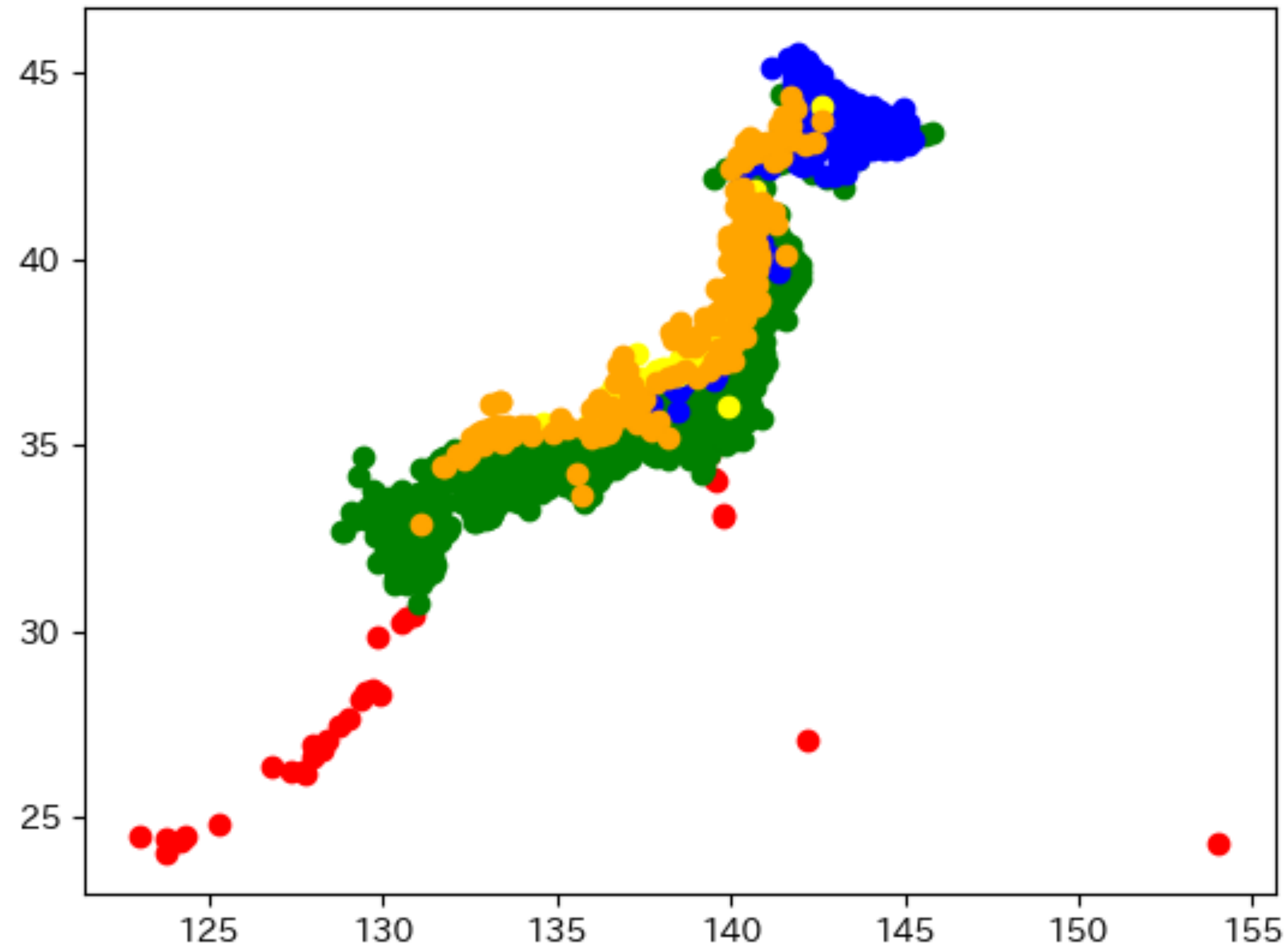


Fig.11 Scatter of each cluster in January (5 clusters)

# 5. Clustering

Name	Representative location	Number	precipitation	Temperature
Island	Okinawa	31	High	Very high
Pacific	Tokyo	460	Low	High
Eastern Hokkaido	Kushiro	126	Low	Very Low
Sea of Japan (Heavy Snow)	Kanazawa	42	Very High	Low
Sea of Japan	Akita	185	High	Low

Table.1 Features of each cluster in January (5 clusters)

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# 6. Discussion

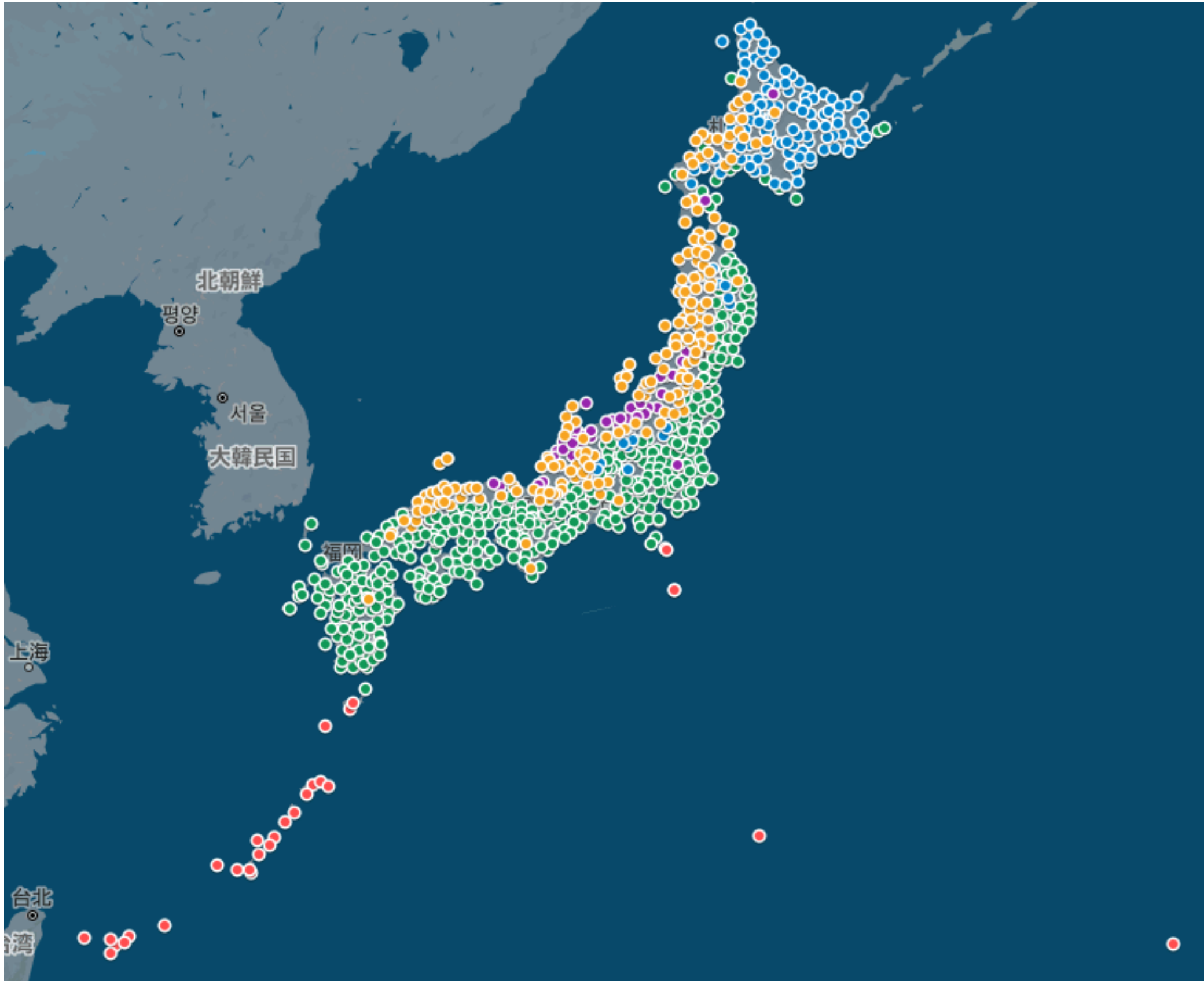
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Good climate analysis on January.

- There is **a lot of snow** on the **Sea of Japan** side and **little precipitation** on the **Pacific** side in January.
- **Warmer** in **the south**, **cooler** in **the north**.

# 7. Plot

Plot the data by using Google Map



Name	Color
Island	Red
Pacific	Green
Eastern Hokkaido	Blue
Sea of Japan (Heavy Snow)	Purple
Sea of Japan	Orange

Fig.12 Result of classification in January (5 clusters)



# 7. Future Work

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Is there a better way to decide cluster numbers ?

- I decided on a hunch.

Proposal

- The number of AMeDAS that included the cluster
- The information of each cluster (euclidean)

# Reference

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- [1] 草薙浩, (2016). 平年日降水量時系列のクラスター分析による日本の9気候地域区分の提案. 天気, 63(1), 5-12.
- [2] JMA, *過去の気象データ・ダウンロード*. Retrieved 04, 2020, from <https://www.data.jma.go.jp/gmd/risk/obsdl/index.php>
- [3] JMA, *観測概要と観測所一覧*. Retrieved 04, 2020, from [https://www.data.jma.go.jp/obd/stats/data/mdrr/man/kansoku\\_gaiyou.html](https://www.data.jma.go.jp/obd/stats/data/mdrr/man/kansoku_gaiyou.html)