




# Japan-Korea Micro Geo Data International Symposium (December 1<sup>st</sup>, 2015)



Ritsu Sakuramachi, Yuki Akeyama  
(Shibasaki Lab., The University of Tokyo)

# outline

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- ▶ What is the inter-firm transaction big data?
- ▶ Introduction of studies using the inter-firm transaction big data ① (Akeyama)
- ▶ Introduction of studies using the inter-firm transaction big data and GPS big data ② (Sakuramachi)

# What is the inter-firm transaction big data?

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**This data contains information of**

- **1,600,000 firms (1996-2013)**
- **4,300,000 transactions (2008-2013)**

**Data Producer: Teikoku Databank, LTD (TDB)**

**(Largest credit research company of Japan)**

Japanese largest transaction data developed by questionnaire surveys and interviews for firms by 1700 investigators with firm credit check

# Inter-firm transaction big data

## Inter-firm transaction data (2008-2013)

Database of approx. 4.3 millions  
Inter-firm transactions with  
following attributes

- Client and contractor firm ID
- Transaction item
- Transaction date
- Estimated transaction value<sup>1)</sup>

## Firm data (1996-2013)

Database of approx. 1.6 million firms  
involved in transactions with following  
attributes

- Enterprise ID
- Sales amount
- Address
- Capital
- Business category
- etc...

Item : Milk  
Date : 201401  
Value : 1 billion Yen



ID : 1 Sales amounts : 5 billion Yen  
Address : Nemuro city, Hokkaido  
Business : Milk product etc...

ID : 2 Sales amounts : 20 billion Yen  
Address : Fukuoka city, Fukuoka  
Business : Wholesale trade etc...

1) Tamura, K., Miura, W., Takayasu, M., Takayasu, H., Kitajima, S., and Goto, H., 2012, "Estimation of Flux Between Interacting Nodes on Huge Inter-firm Networks", International Journal of Modern Physics: Conference Series, 16,93–104.

# Inter-firm transaction big data

## Inter-firm transaction data (2008-2013)

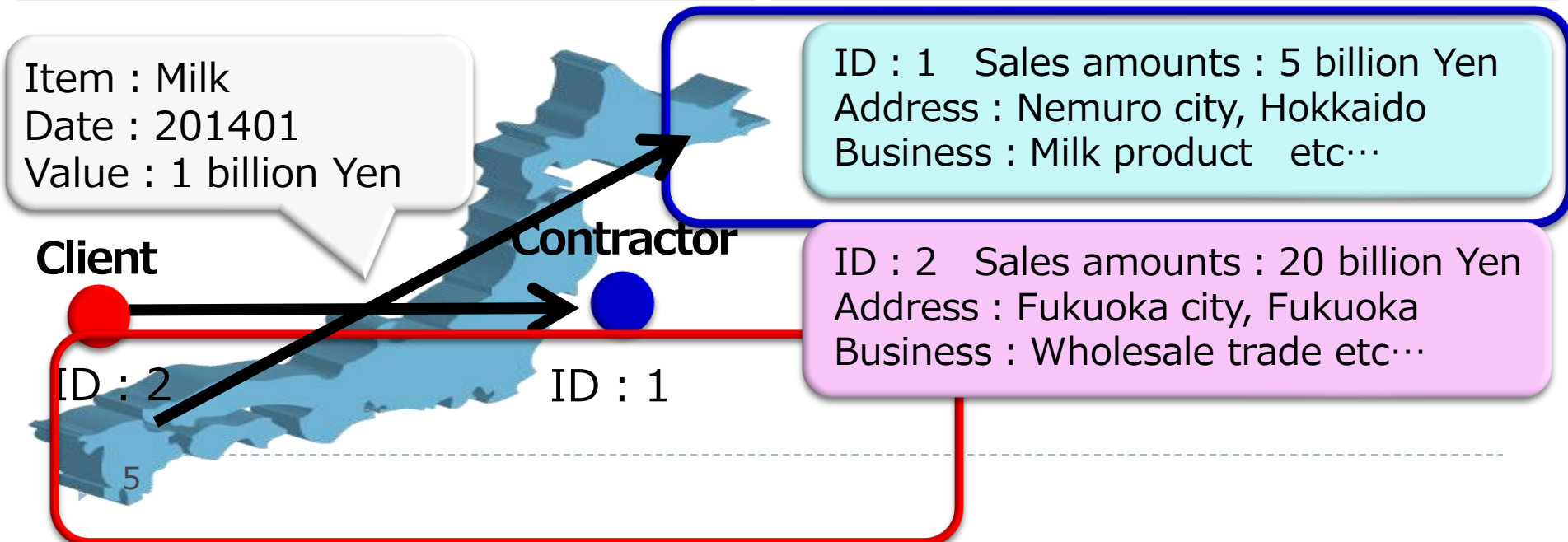
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## Firm data (1996-2013)

Database of approx. 1.6 million firms  
involved in transactions with following  
attributes

- Enterprise ID • Sales amount
- Address • Capital • Business category  
etc...



# Visualization of network between regions using large-scale business transaction data

Yuki Akeyama – Civil Engineering, U.T.

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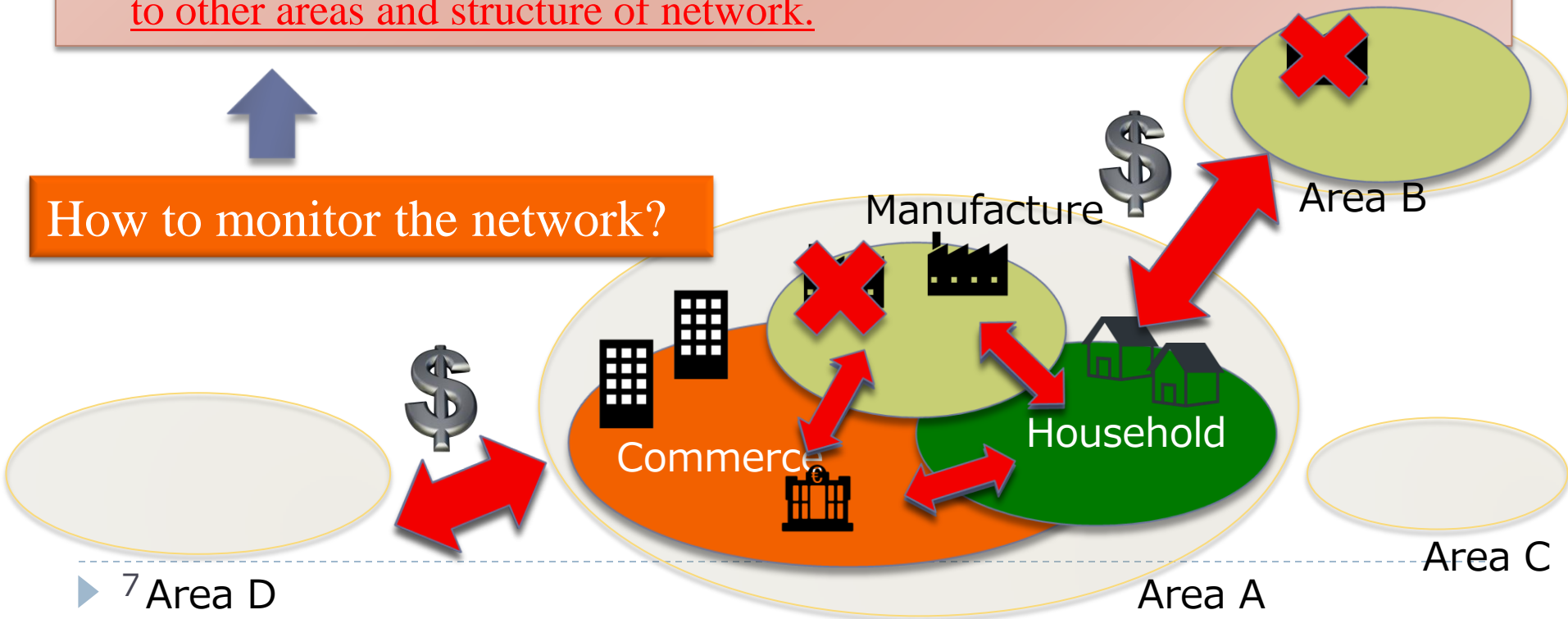
Yuki Akiyama – CSIS, U.T.

Ryosuke Shibasaki – CSIS, U.T.

# Background

- Increase or decrease of firm sales  $\Rightarrow$  impact to trading partners  
 $\Rightarrow$  impact to consumption of retails in the area where firms are locate
- Example : Great East Japan Earthquake(2011) happened...  
 $\Rightarrow$  Damage to car component suppliers  
 $\Rightarrow$  decline of sales of car manufactures or other component suppliers
- it's necessary to understand how each local area have economical relationship to other areas and structure of network.

How to monitor the network?



# Economic analysis using Big Data

Previous study : using governmental statistics or I/O table

Problem: the data is aggregated in administrative area.

⇒ It's difficult to analyze in area of smaller scale.

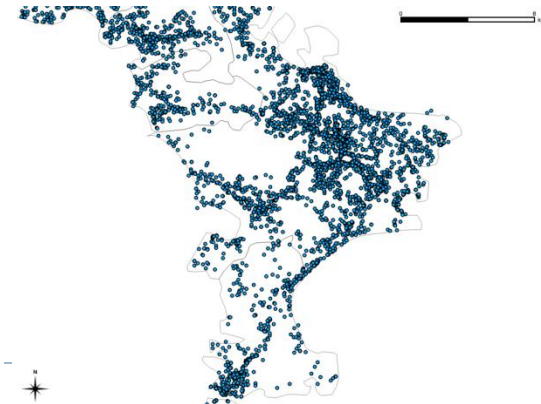
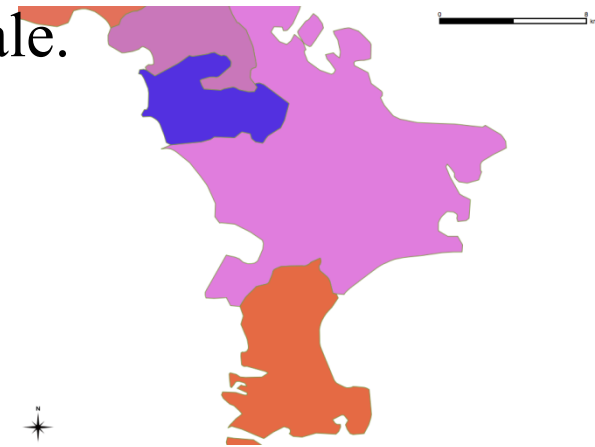


This study: using large-scale transaction data

Non-aggregated data

⇒ we can analyze economy in “various scale”.

(city block ▪ damaged area ▪ commutable area...)





# Our objective is...

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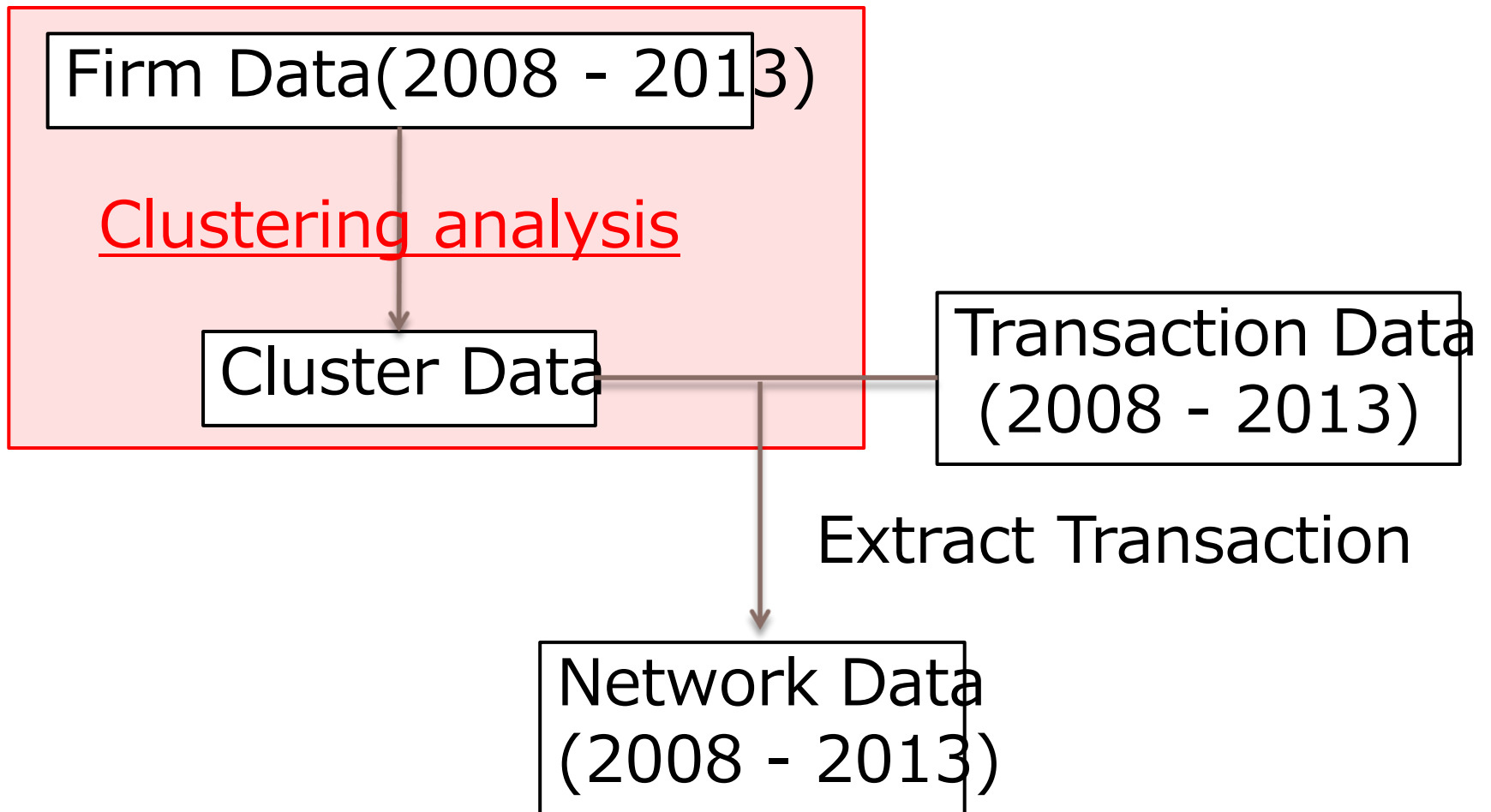
1. Defining local area with clustering method.  
(non-administrative division)
2. Visualizing economic network between local area  
in Japan using transaction data.



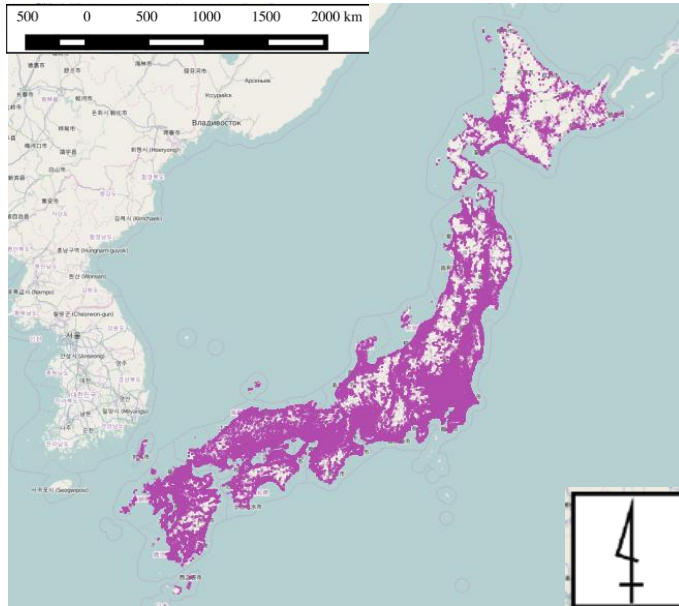
Understanding important counterparts  
for each local area.

# Processing flow

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# Result-Clustering

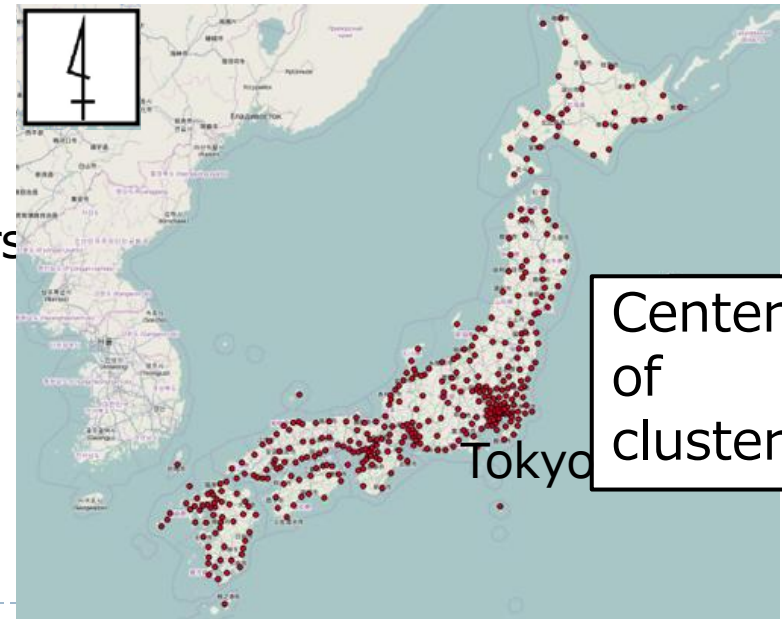
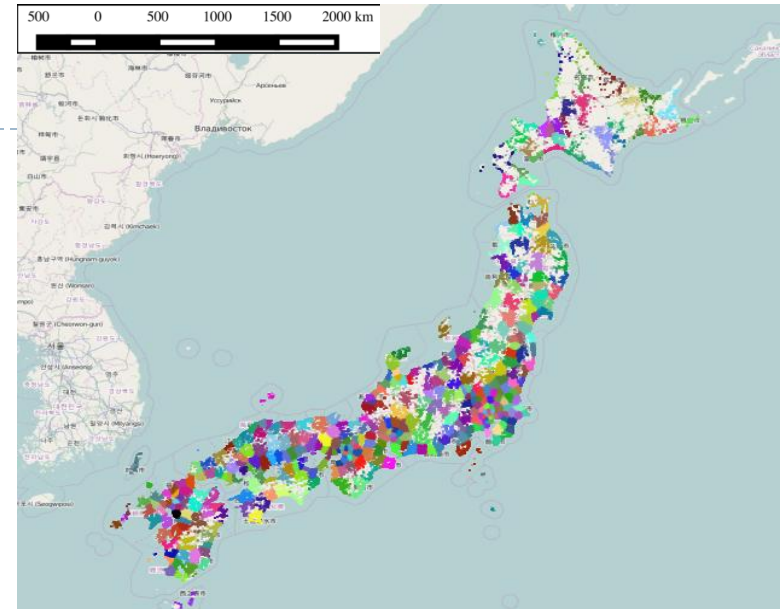


4

K-means++  
clustering

Mapping of 1.6 million firms (400 clusters  
(point data))

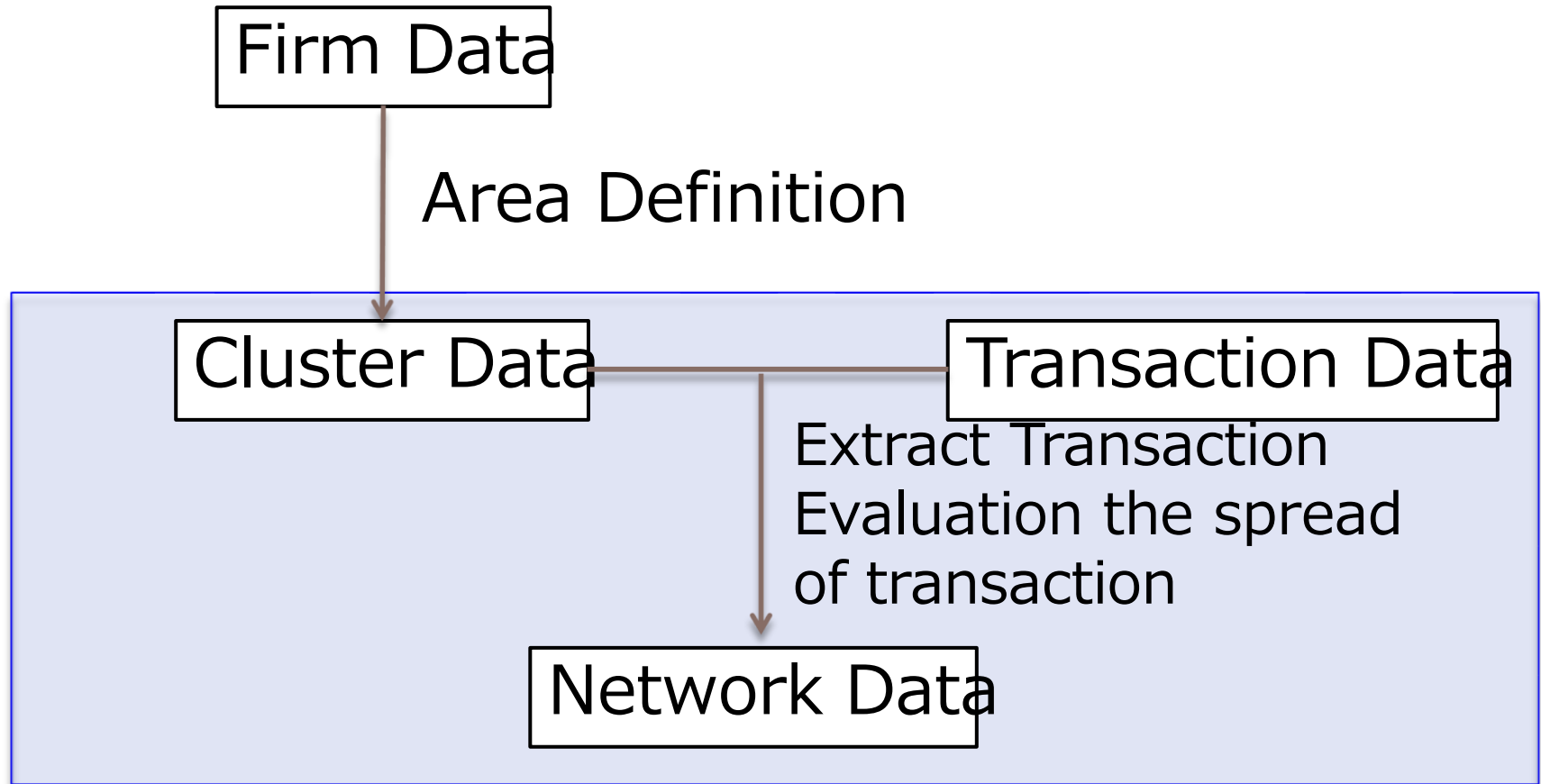
Each color means each cluster.  
Many clusters are in urban area.



Centers  
of  
cluster

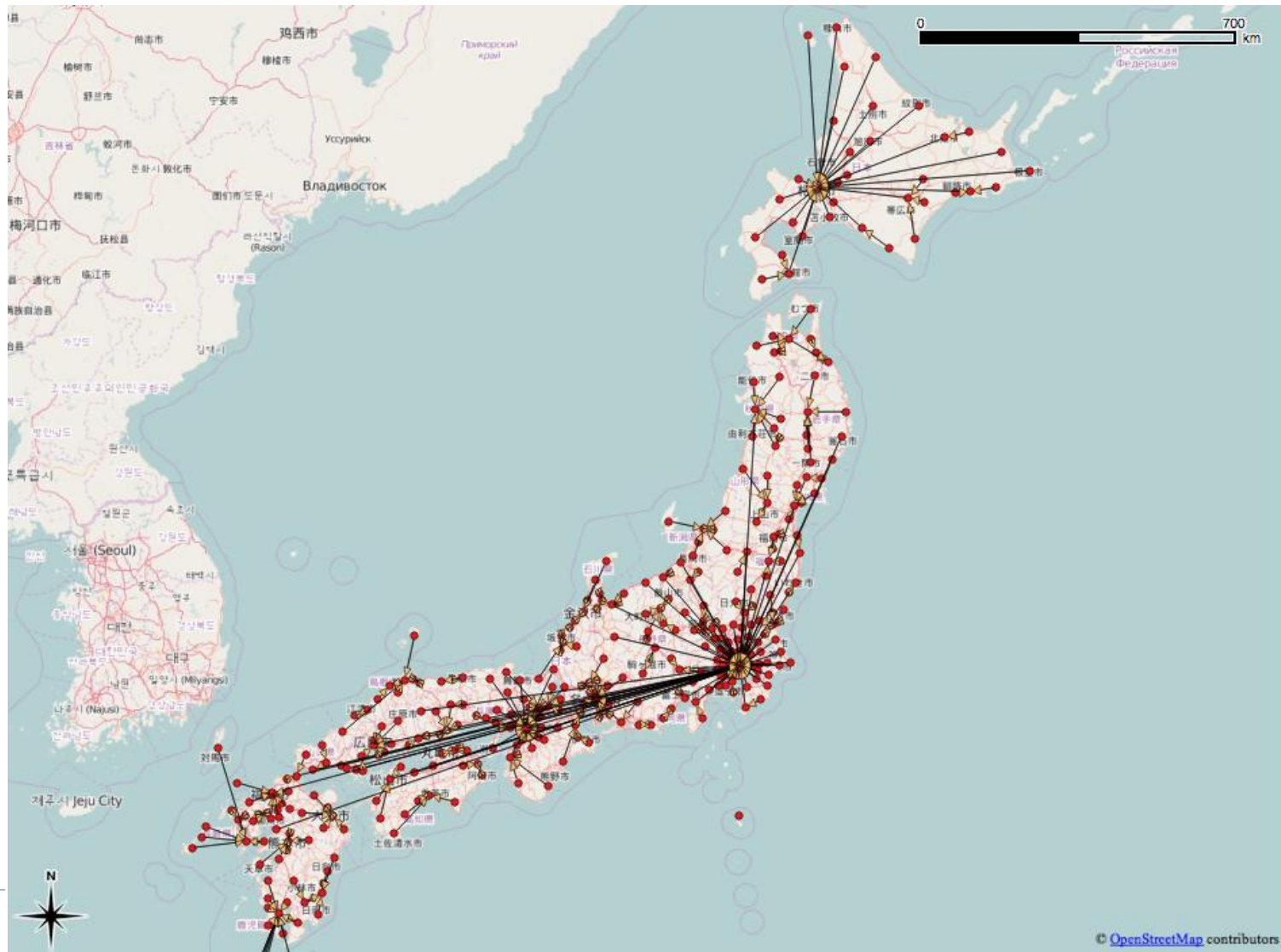
# Processing Flow

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# Result & Discussion

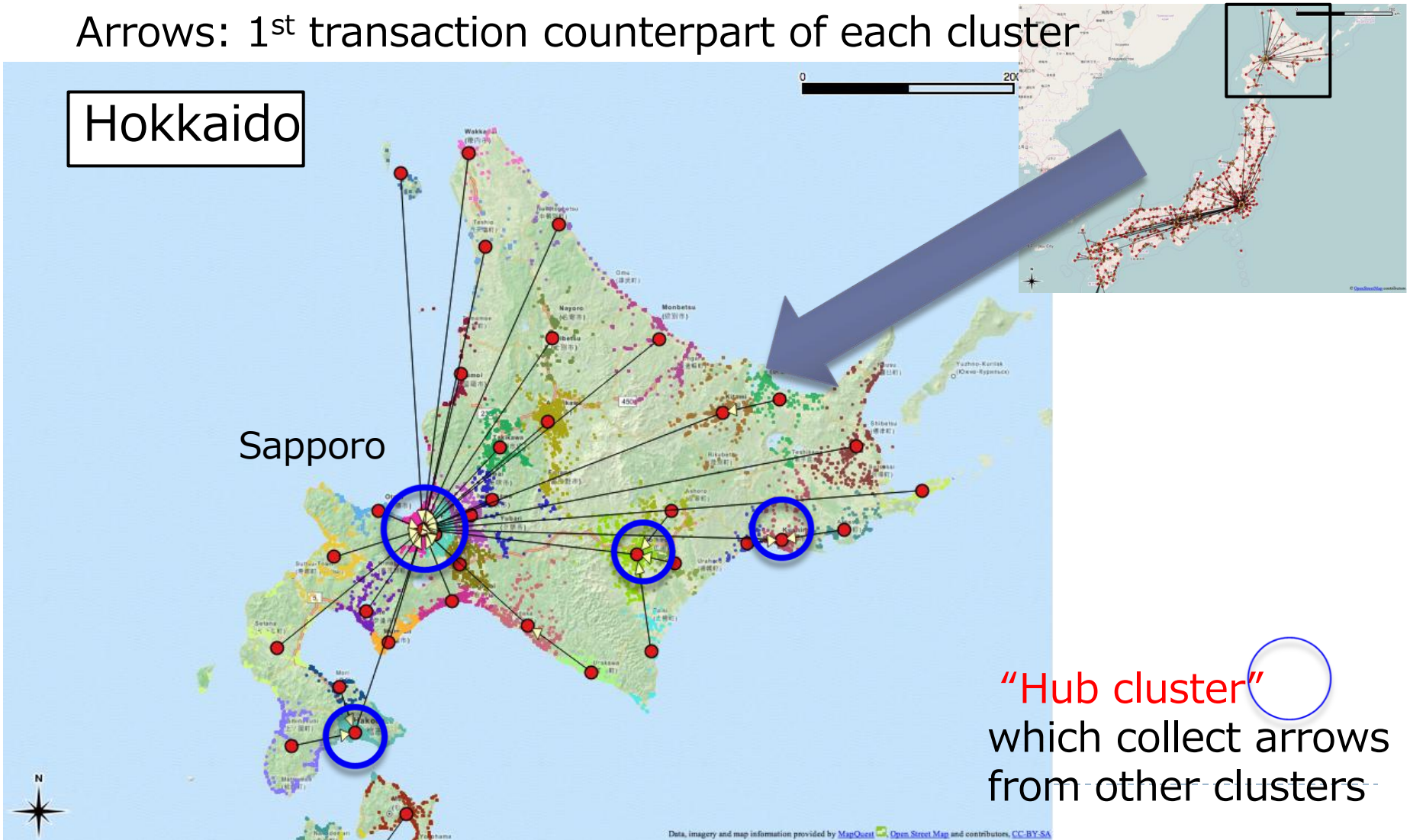
Arrows: 1<sup>st</sup> transaction counterpart of each cluster



Hokkaido

# Sapporo

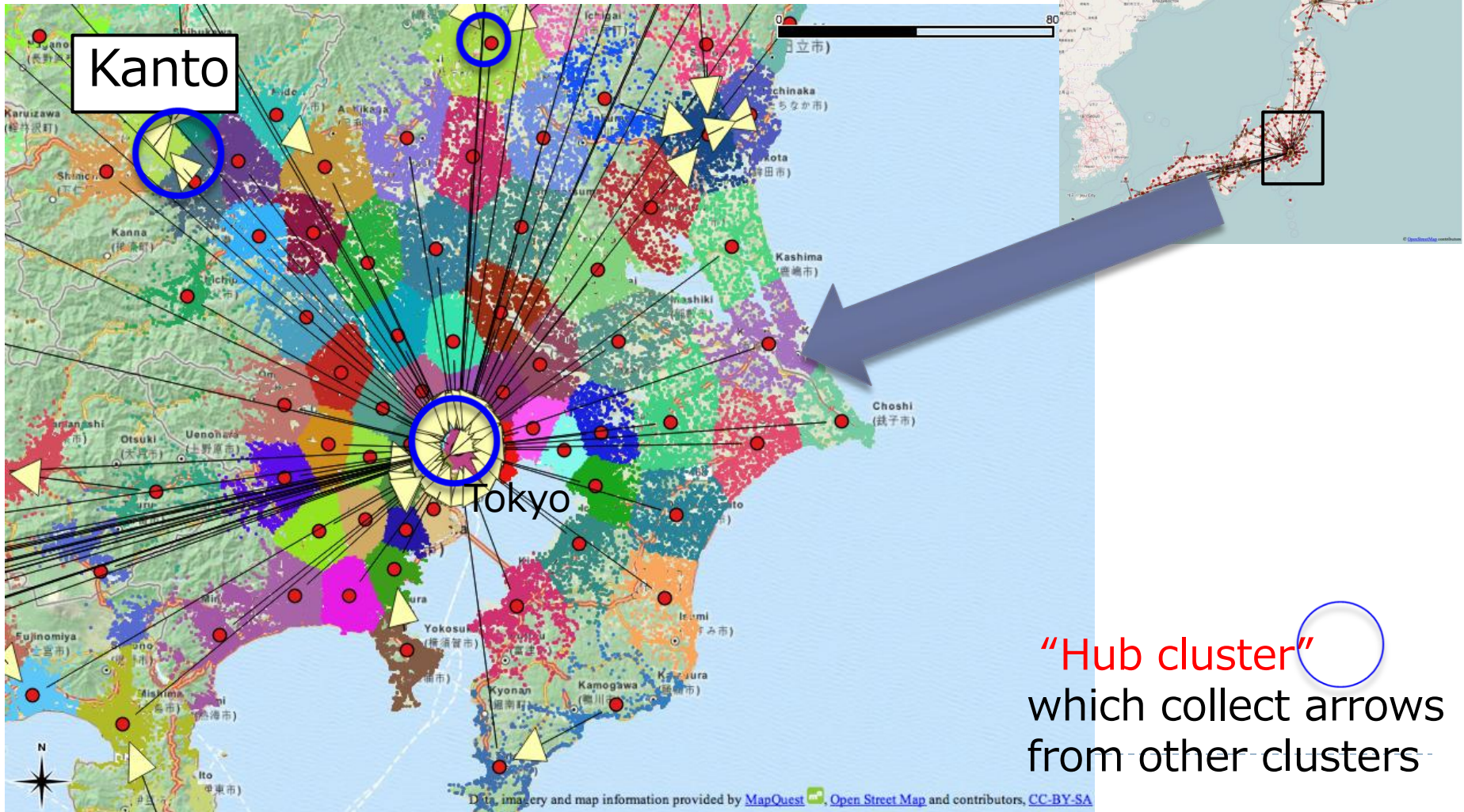
“Hub cluster”  
which collect arrow  
from other cluster





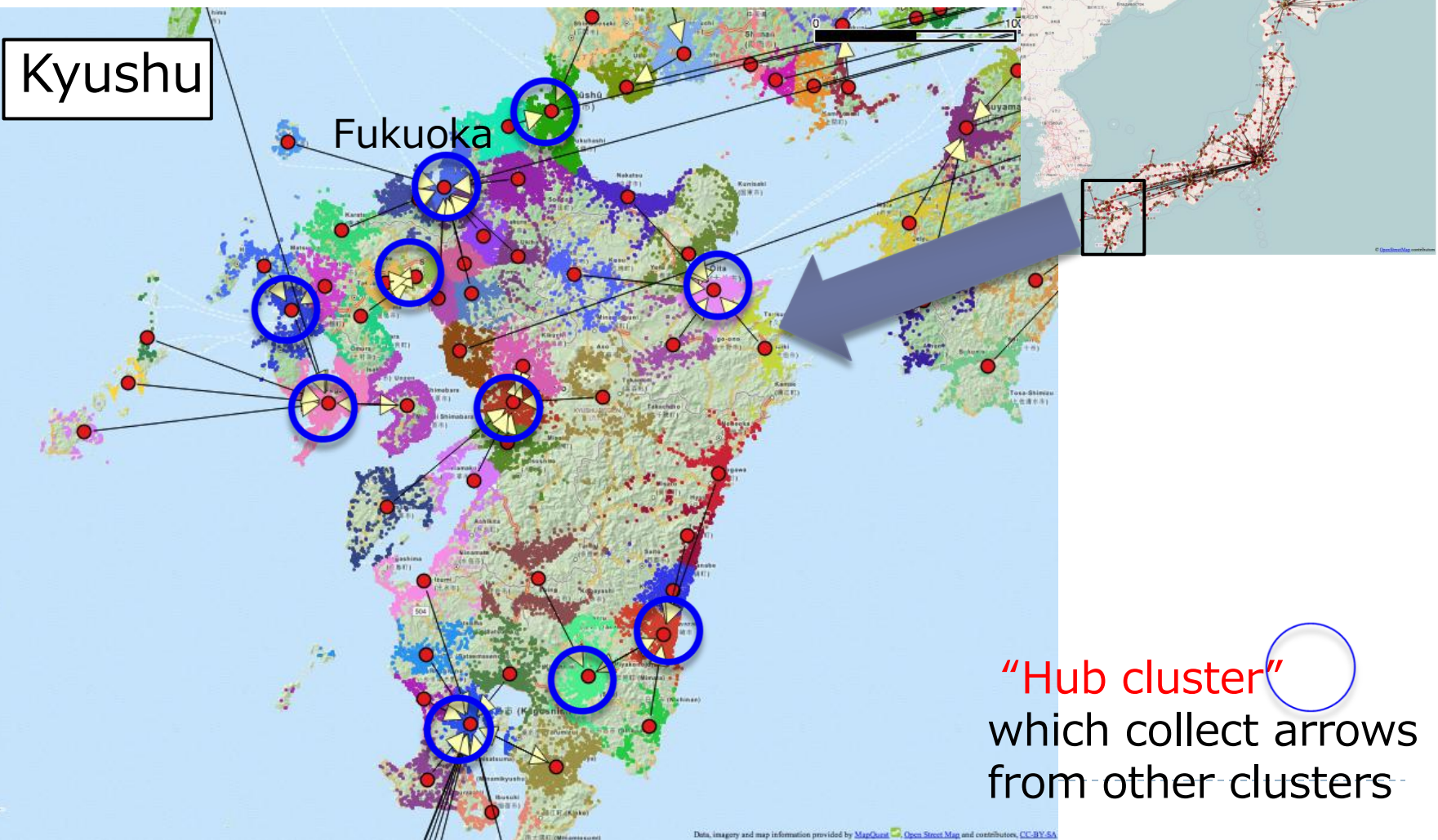
# Result & Discussion

Arrows: 1<sup>st</sup> transaction counterpart of each cluster



# Result & Discussion

Arrows: 1<sup>st</sup> transaction counterpart of each cluster





# Conclusion

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1. New definition of Local area  
→ clustering based on geospatial data of firms
2. Visualizing economic network between local area  
→ Dependent cluster  $\Leftrightarrow$  Hub cluster



**Support Strategic Economic Policy**  
- selection of most important  
economic counterparts

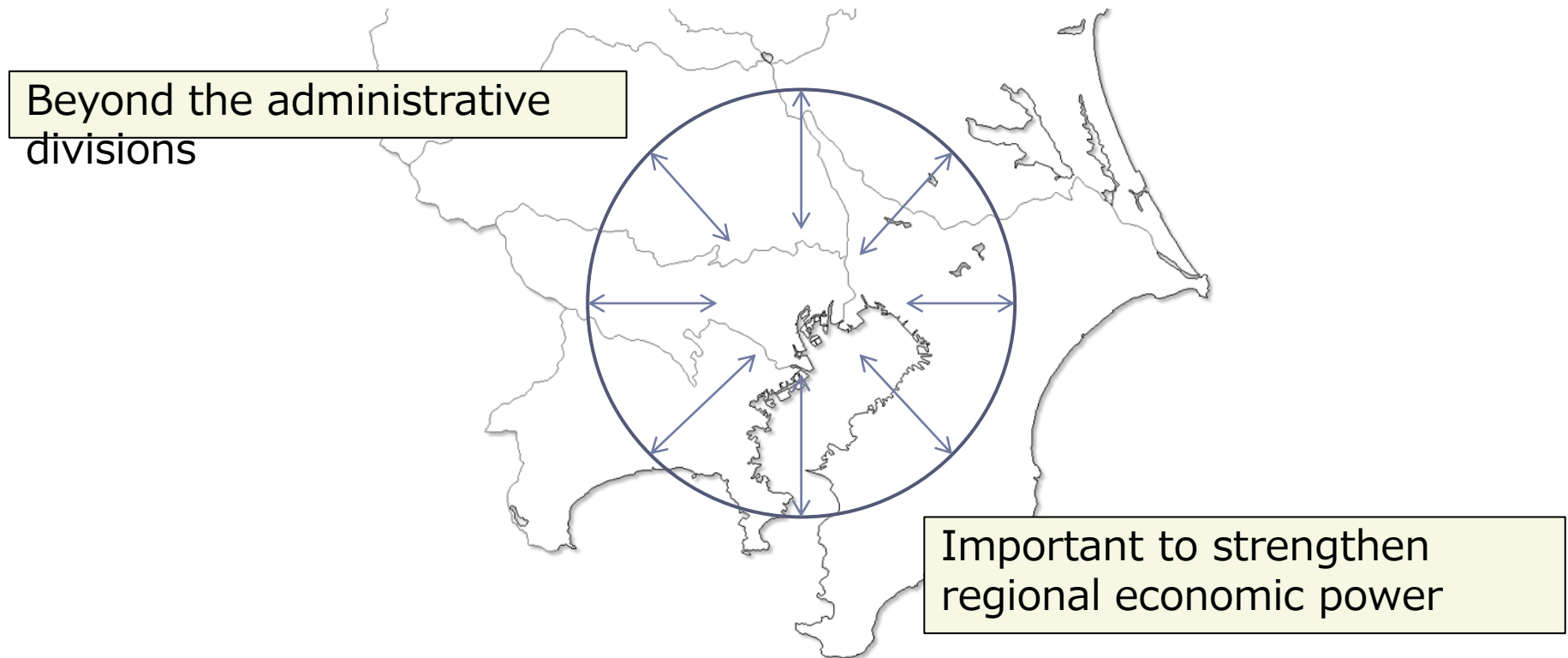


# A Network Theory-based Delineation of Metropolitan areas with Mass People Flow Data

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rsakura@csis.u-tokyo.ac.jp

# What is a metropolitan area?

- ▶ **Metropolitan areas** are regions defined with focusing on **spatial expanse of daily economic activities**.



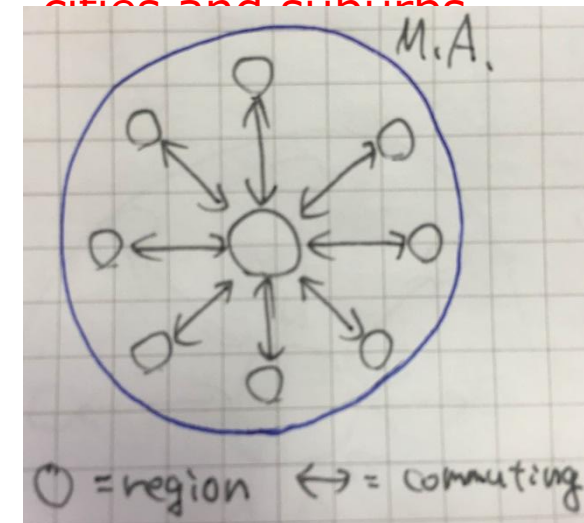
# Previous research

表 1 中心都市と郊外都市の設定基準

	上段：中心都市の設定基準
	下段：郊外都市の設定基準
山田・徳岡 (1983)	1. 常住人口が5万人以上。 2. 鉱業を除く非1次産業就業人口比率が75%以上。 3. 昼夜間人口比率が1.0以上。 4. 他の特定の中心都市への流出就業者比率が15%未満。 5. 総流出就業者比率が30%未満。
森川 (1990)	1. 中心都市への流出就業者比率が10%以上。 2. 鉱業を除く非1次産業就業人口比率が75%以上。
Kawashima et al. (1993)	1. 卸売・小売業、サービス業従業者数が3千人以上。 1. 中心都市への通勤者比率が5%以上。
総務庁統計局 (1999)	1. 常住人口が10万人以上。 2. 昼夜間人口比率が1.0以上。 1. 中心都市への通勤比率が5%以上、または、5百人以上。
日本産業消費研究所 (2000)	1. 東京特別区部および政令指定都市（大都市圏）。大都市圏に属さない人口50万人以上の市（都市圏）。 1. 中心都市への流出出勤・通学者の常住人口に占める割合が1.5%以上。
金本・徳岡 (2002)	1. 周辺市町村からの通勤・通学者比率が10%以上。 1. 中心都市への通勤・通学者比率が10%以上。 1. DID人口が1万人以上。 2. 郊外市町村の条件を満たすが、従業常住人口比率が1以上で、DID人口が中心都市の3分の1以上か、10万人以上。 1. 中心都市への通勤率が10%以上。 2. 郊外市町村への通勤率が10%以上。

(出所) 山田・徳岡 (1983), 森川 (1990), Kawashima et al. (1993), 総務庁統計局 (1999), 日本産業消費研究所 (2000), 金本・徳岡 (2002) より作成。

Many previous researches define metropolitan areas based on **commuting rates** between **central cities** and **suburbs**.



▶ Hayashi (2014) Journal of economics and sociology, Kagoshima University, 83: 125-137

# Previous research

表 1 中心都市と郊外都市の設定基準

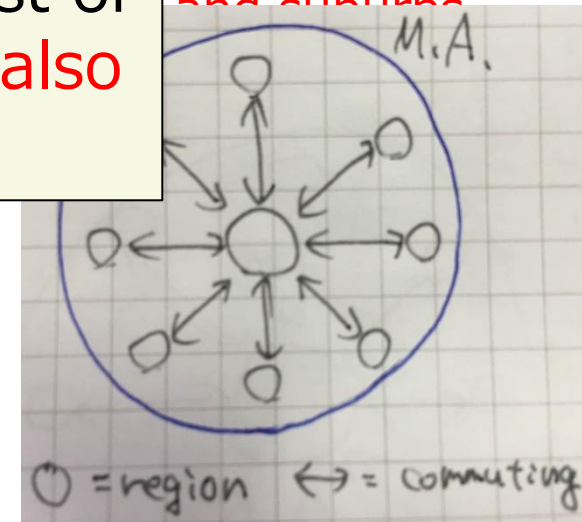
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Many previous researches define metropolitan areas

based on commuting between central and suburbs

However, our daily economic activities consist of not only commuting but also other activities (buying behavior, tourism, etc.)

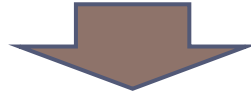


▶ Hayashi (2014) Journal of economics and sociology, Kagoshima University, 83: 125-137

# Progress of data

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National census (per 5 years)



GPS data (24 hours in 365 days!)

comprehensive and continuous mass people flow data  
with high spatiotemporal resolution

# Study flow

GPS data (2012) ZENRIN DataCom CO., LTD.

Estimation of stay points by Akiyama (2013)

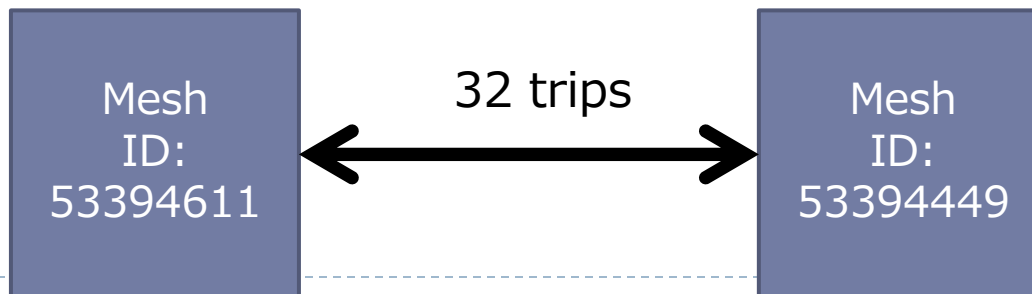
Location data of both origin stay points and destination stay points (643,407,000 trips)

287,320 meshes (1km × 1km) in

Calculating trips between meshes

Trip data

Trip data including information like...



# Study flow

GPS data (2012) ZENRIN DataCom CO., LTD.

Estimation of stay points by Akiyama (2013)

Location data of both origin stay points and destination stay points (643,407,000 trips)

287,320 meshes (1km × 1km) in

Calculating trips between meshes

Trip data

Trip Network (node=mesh, link=number of trips)

Clustering meshes by Map equation

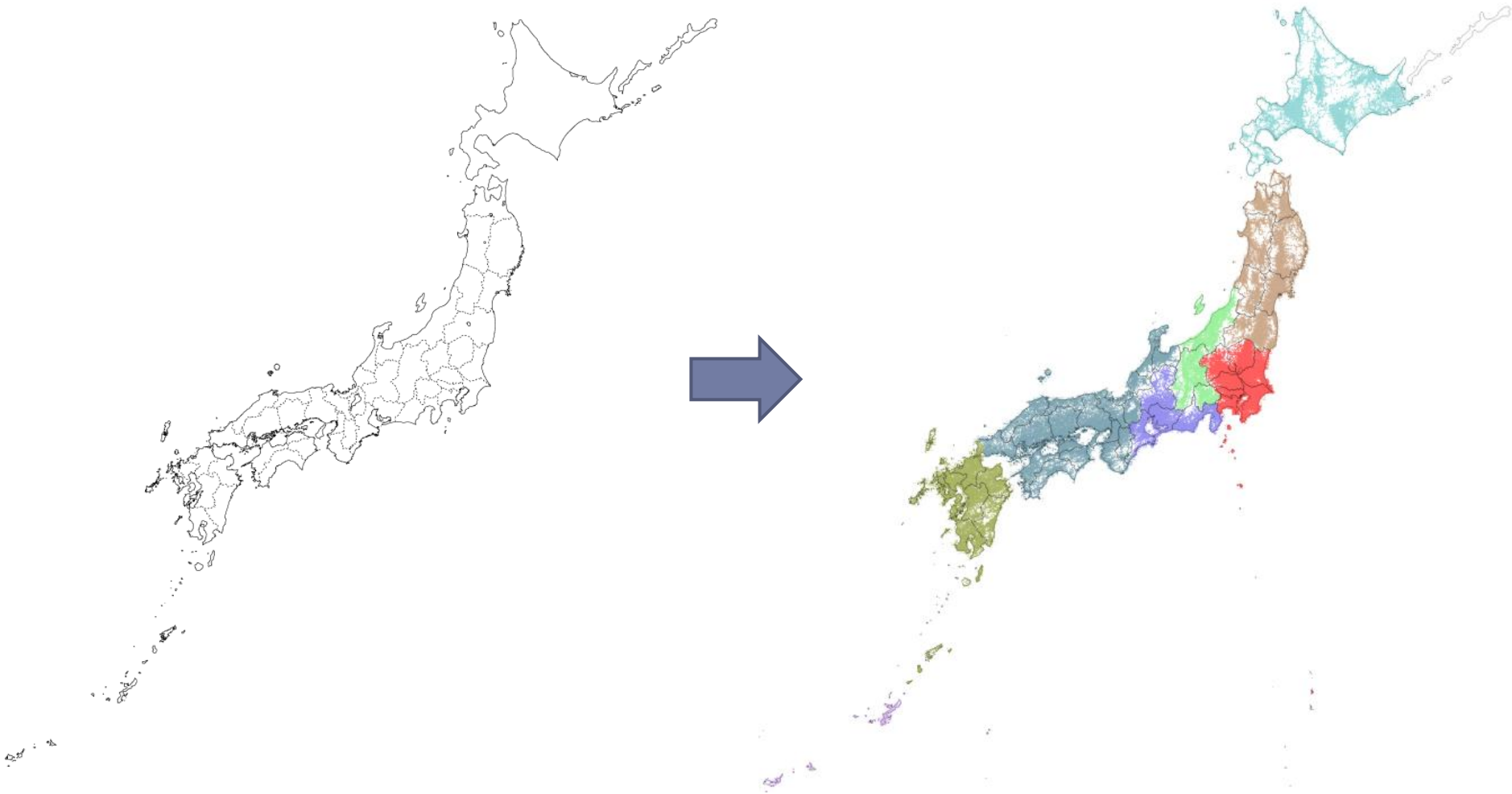
Several clusters of meshes

Visualization clusters on a map



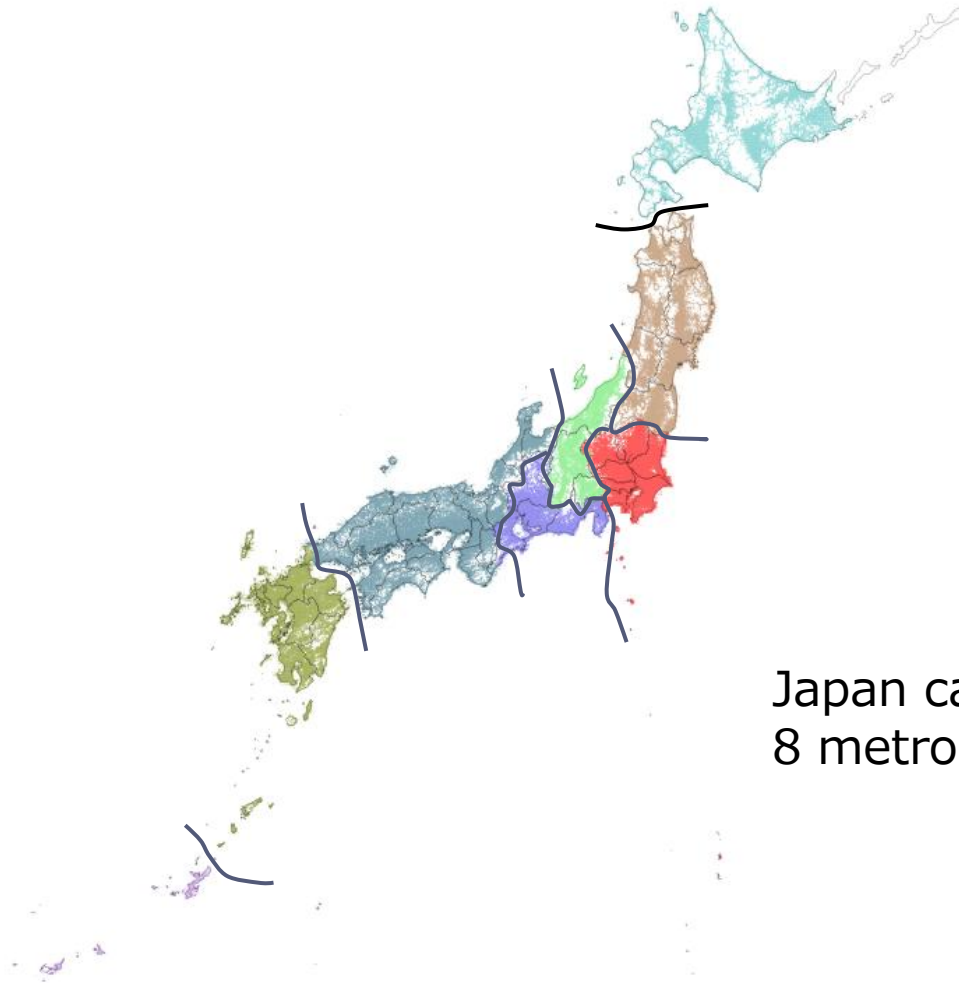
# Result of Clustering

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# Result of Clustering

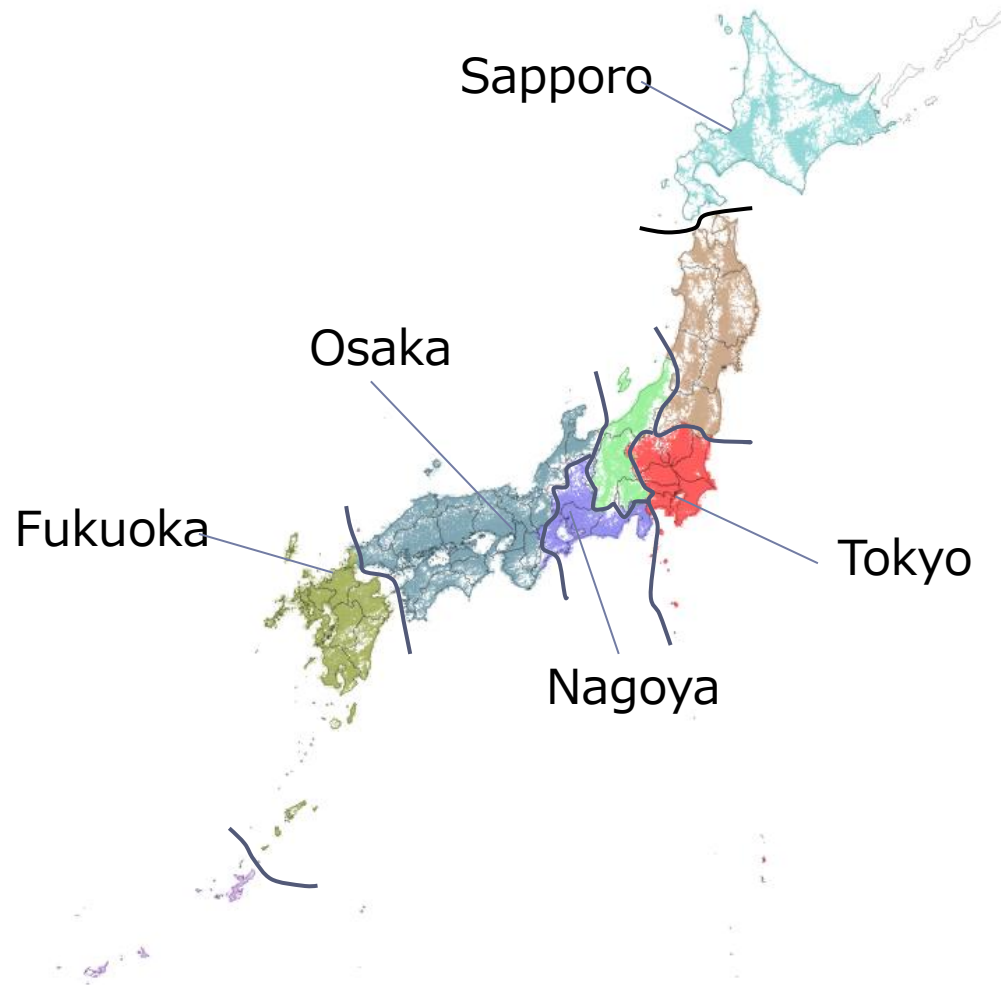
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Japan can be divided into  
8 metropolitan areas

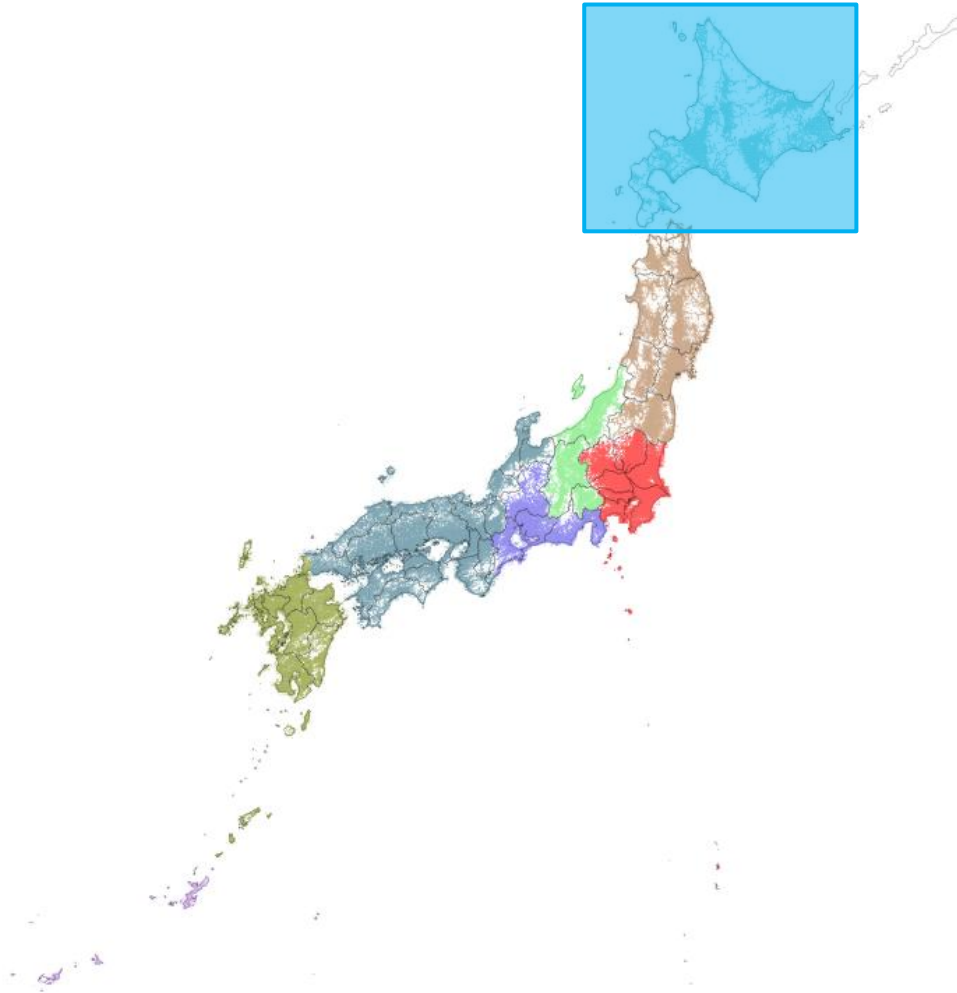
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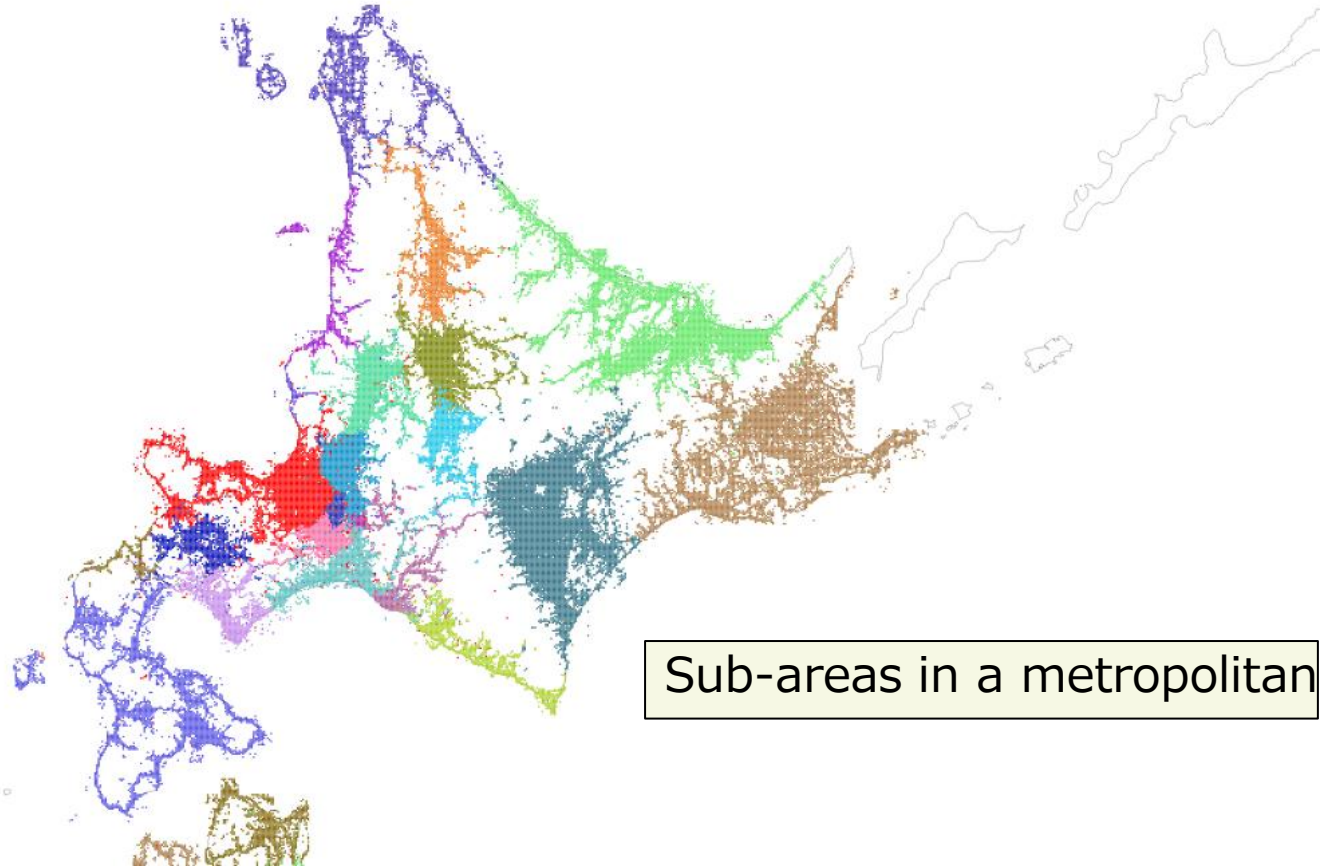
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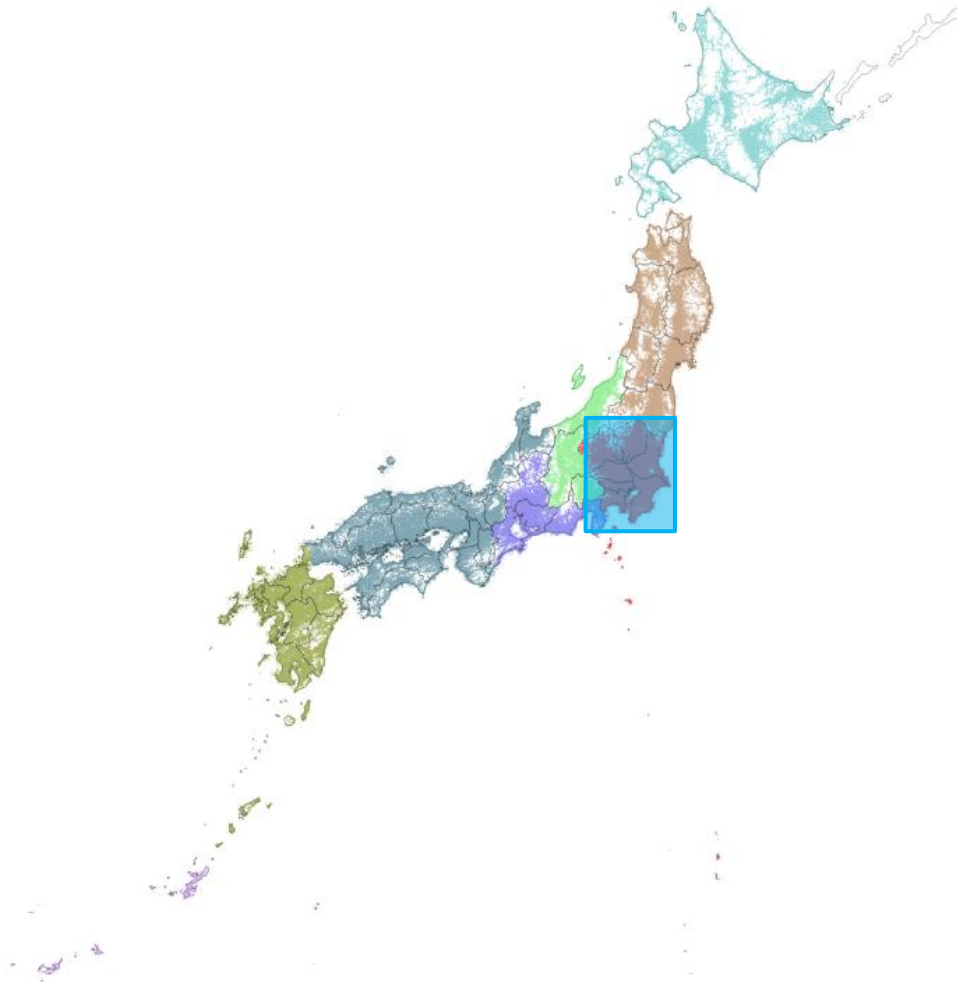
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Sub-areas in a metropolitan area

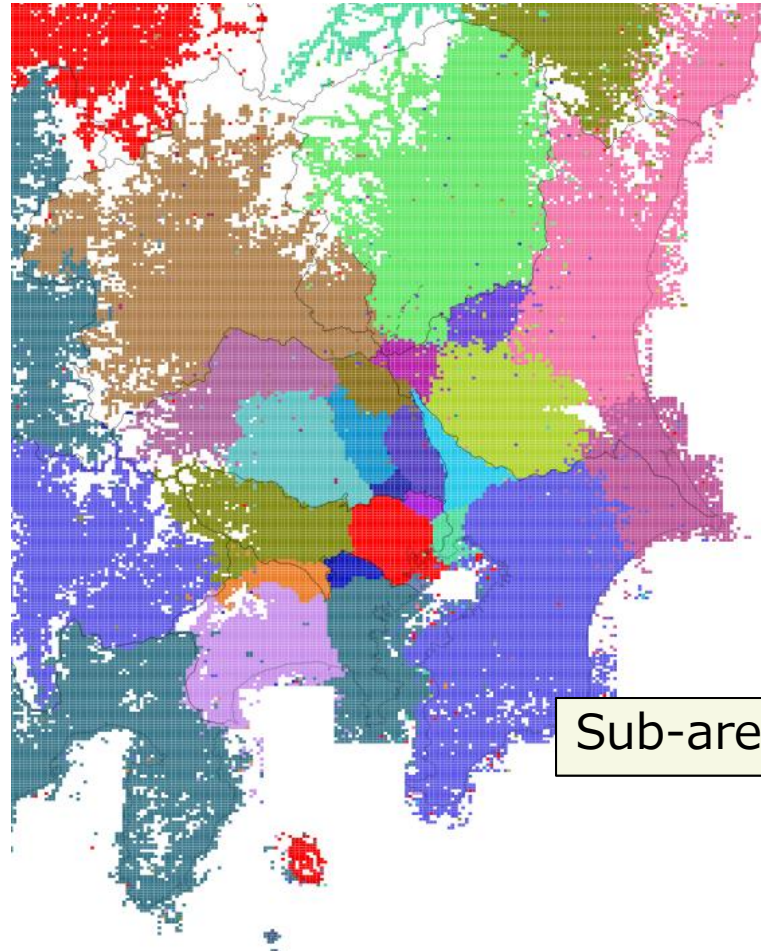
# Result of Clustering

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# Result of Clustering

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Sub-areas in a metropolitan area



# Combination of metropolitan area and inter-firm big data

