



Japan-Korea Micro Geo Data International Symposium (December 1st, 2015)



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

outline

- ▶ 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?



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 value¹⁾

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.

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Item : Milk
Date : 201401
Value : 1 billion Yen

Client

ID : 2

5

Contractor

ID : 1

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...

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.

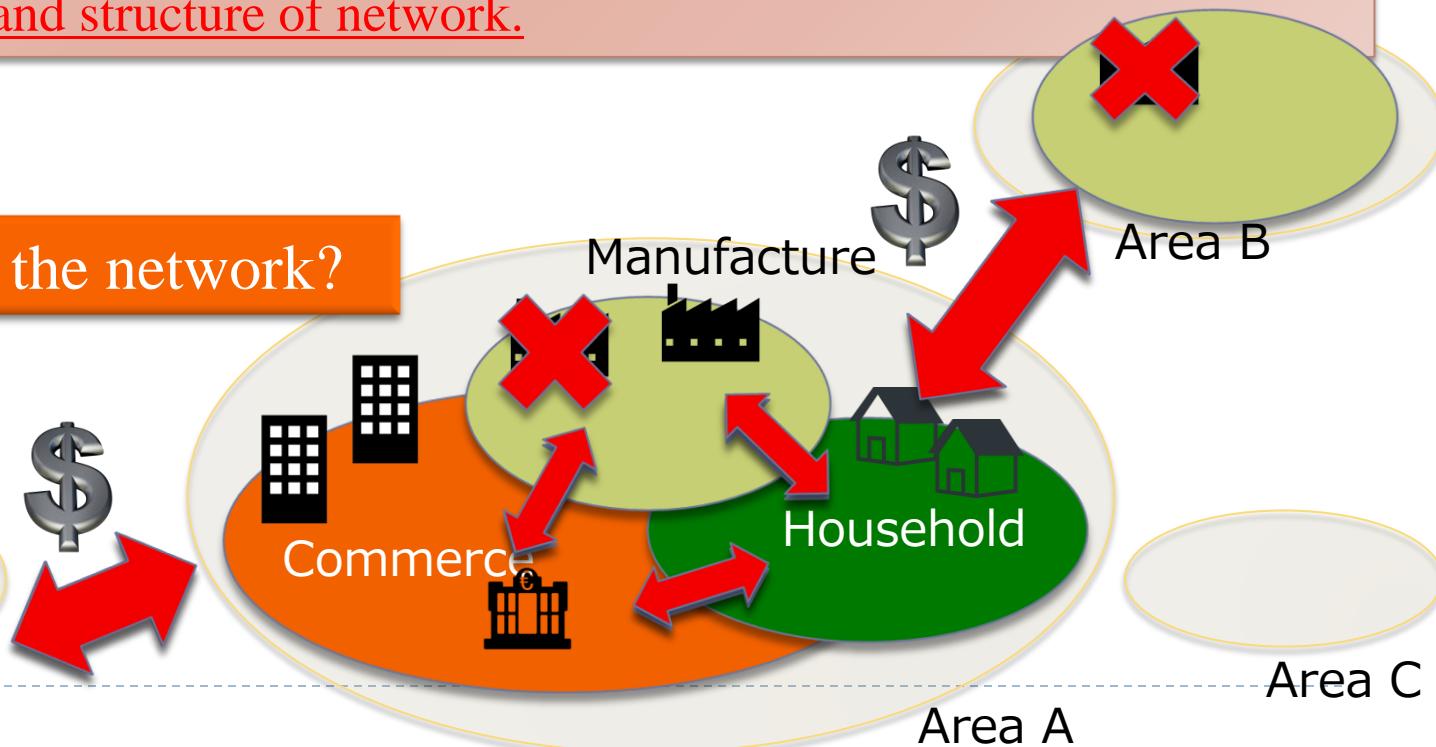
Ryosuke Shibasaki – CSIS, U.T.

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.



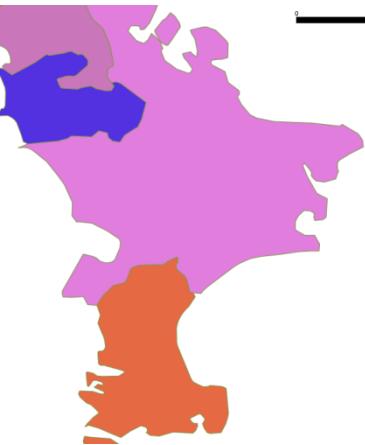
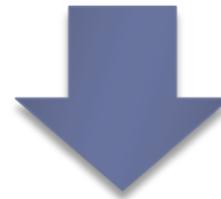
How to monitor the network?



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.

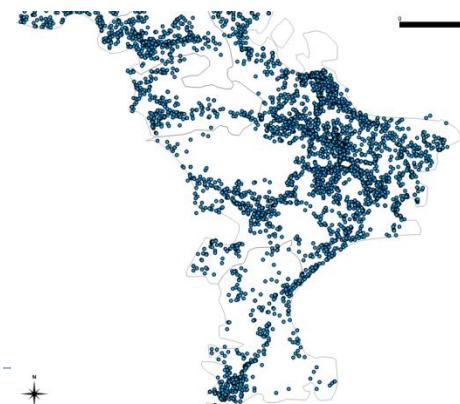


This study: using large-scale transaction data

Non-aggregated data

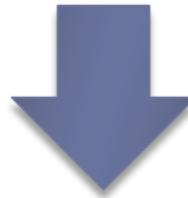
⇒ we can analyze economy in “various scale”.

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



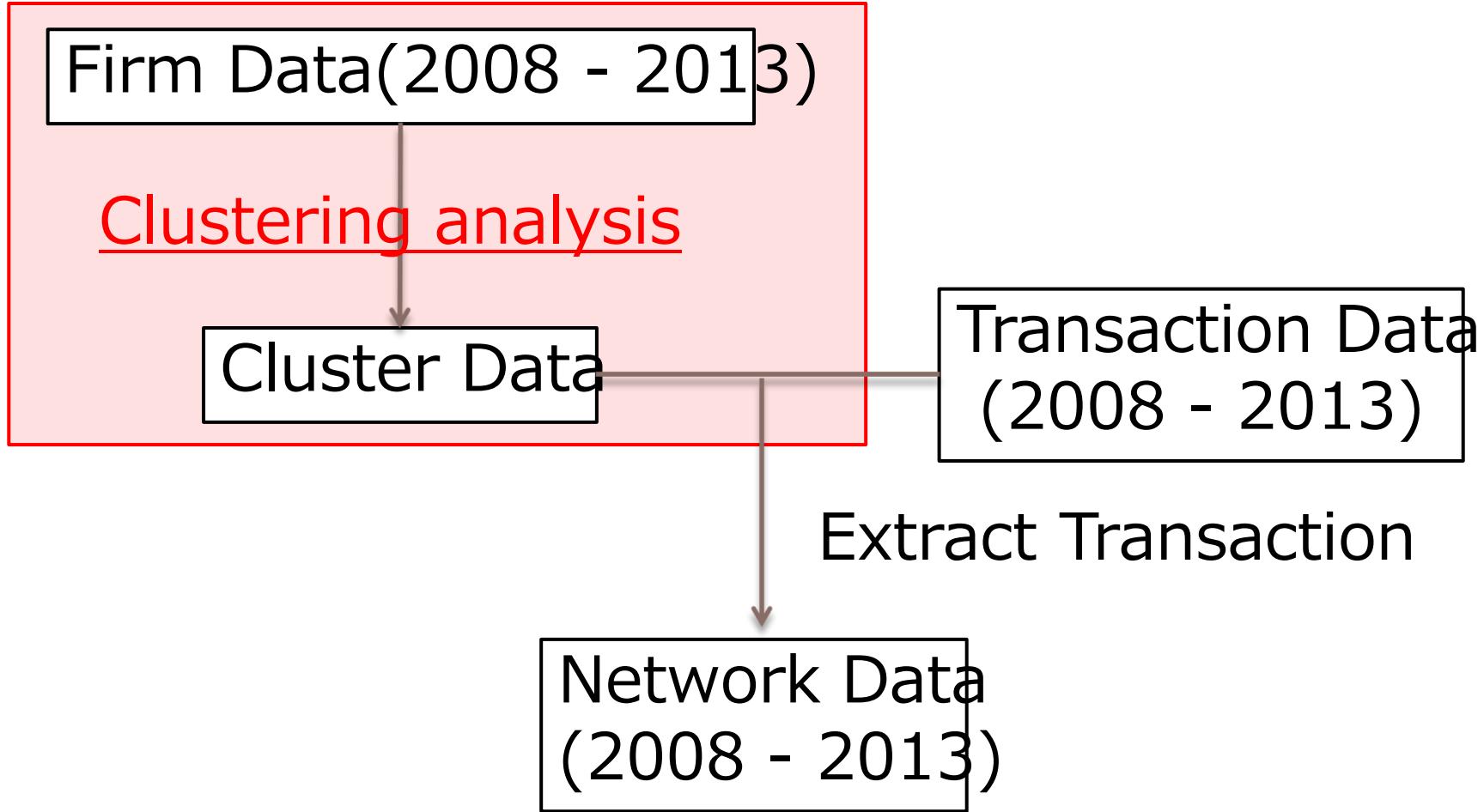
Our objective is...

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

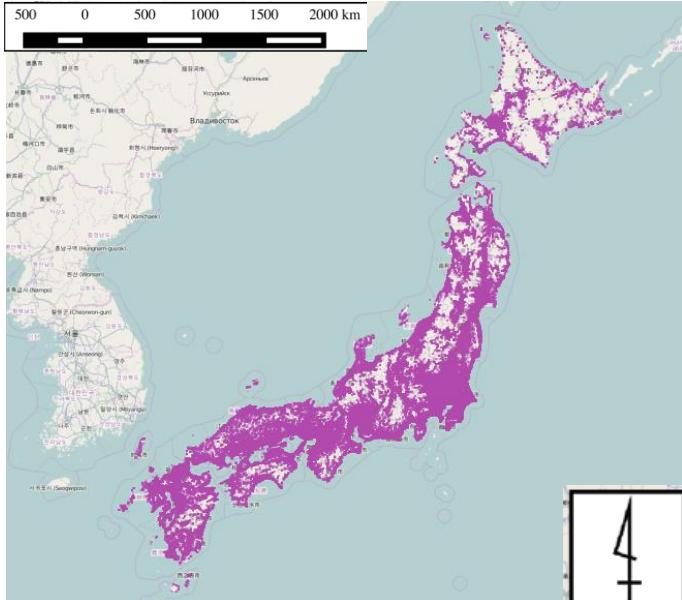


Understanding important counterparts
for each local area.

Processing flow

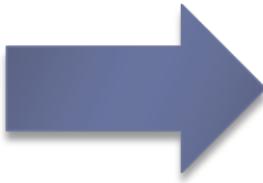


Result-Clustering

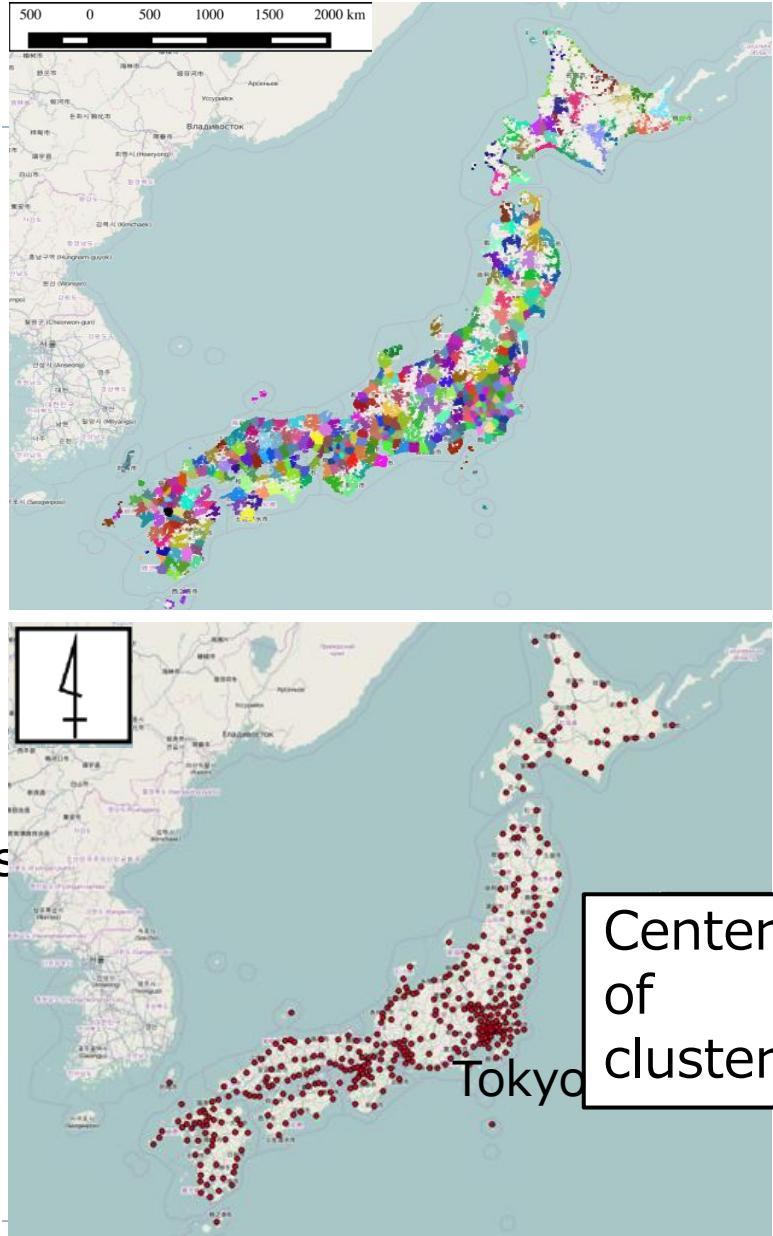


Mapping of 1.6 million firms (400 clusters)
(point data)

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



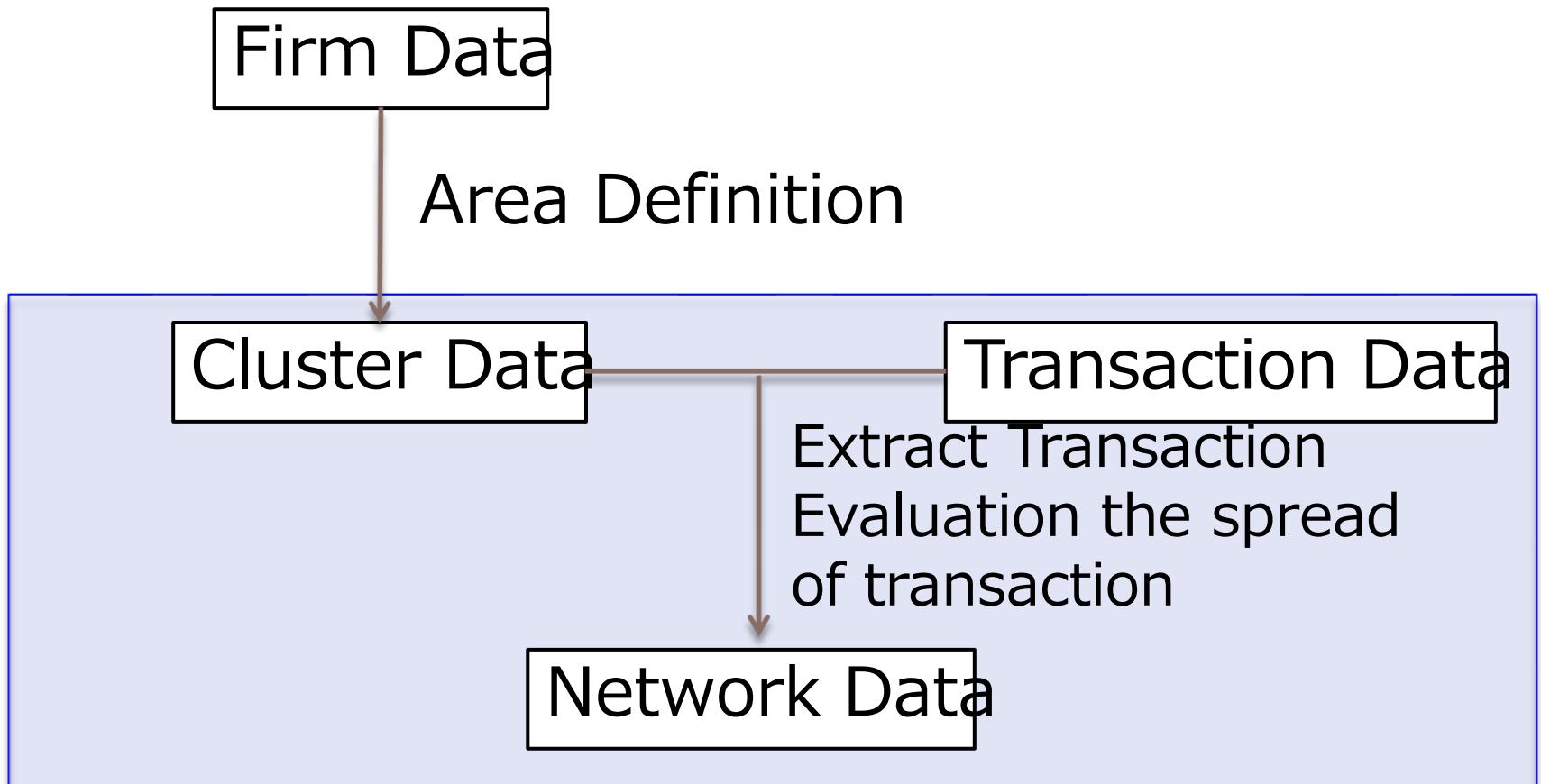
K-means++
clustering



Centers
of
cluster

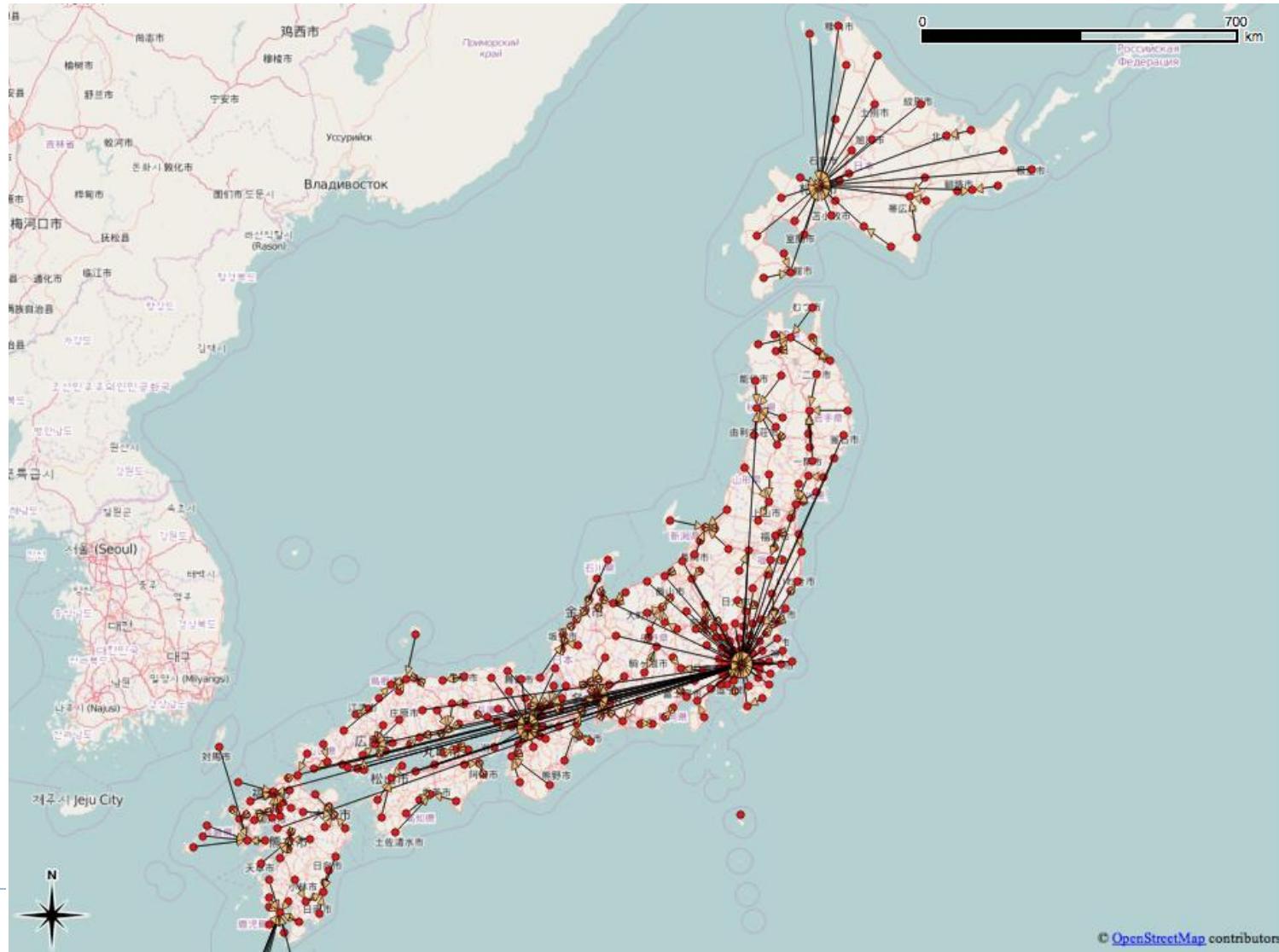
Tokyo

Processing Flow



Result & Discussion

Arrows: 1st transaction counterpart of each cluster



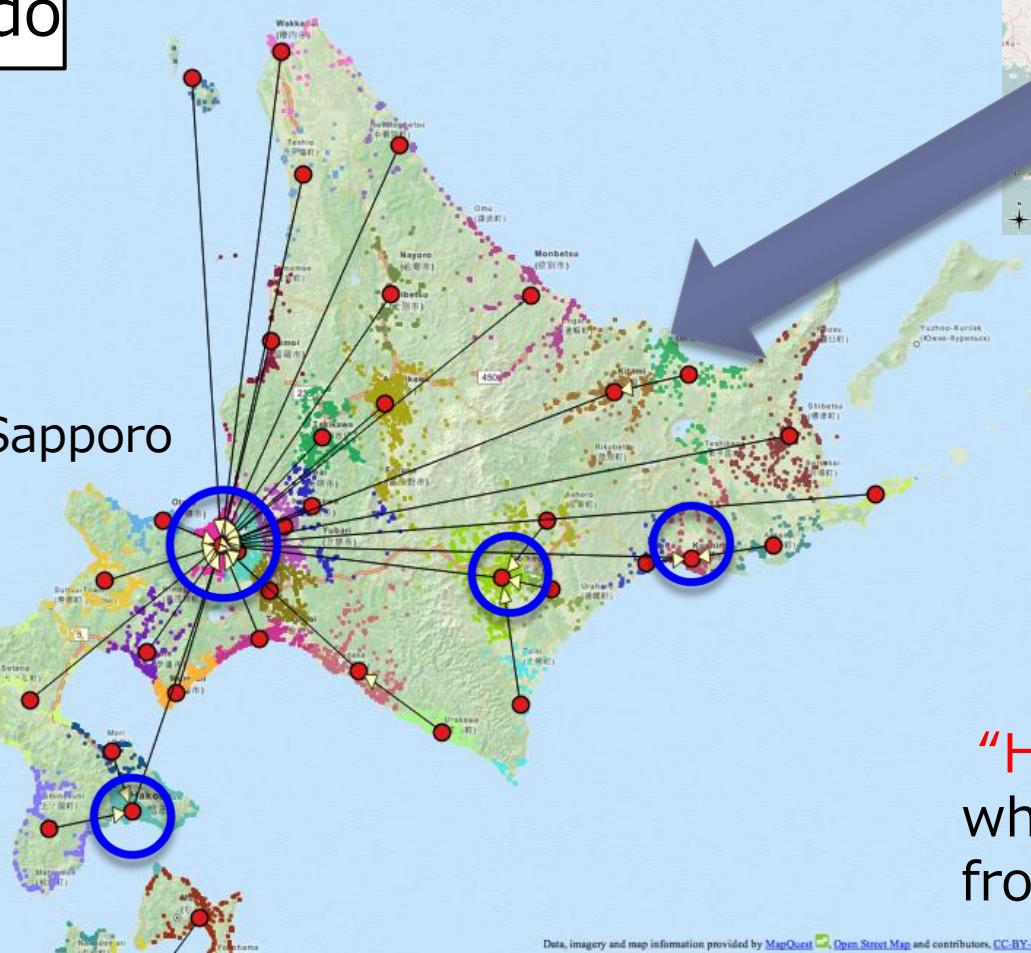
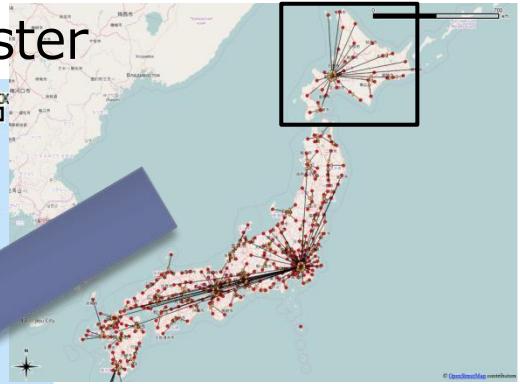
Result & Discussion

Arrows: 1st transaction counterpart of each cluster

Hokkaido

Sapporo

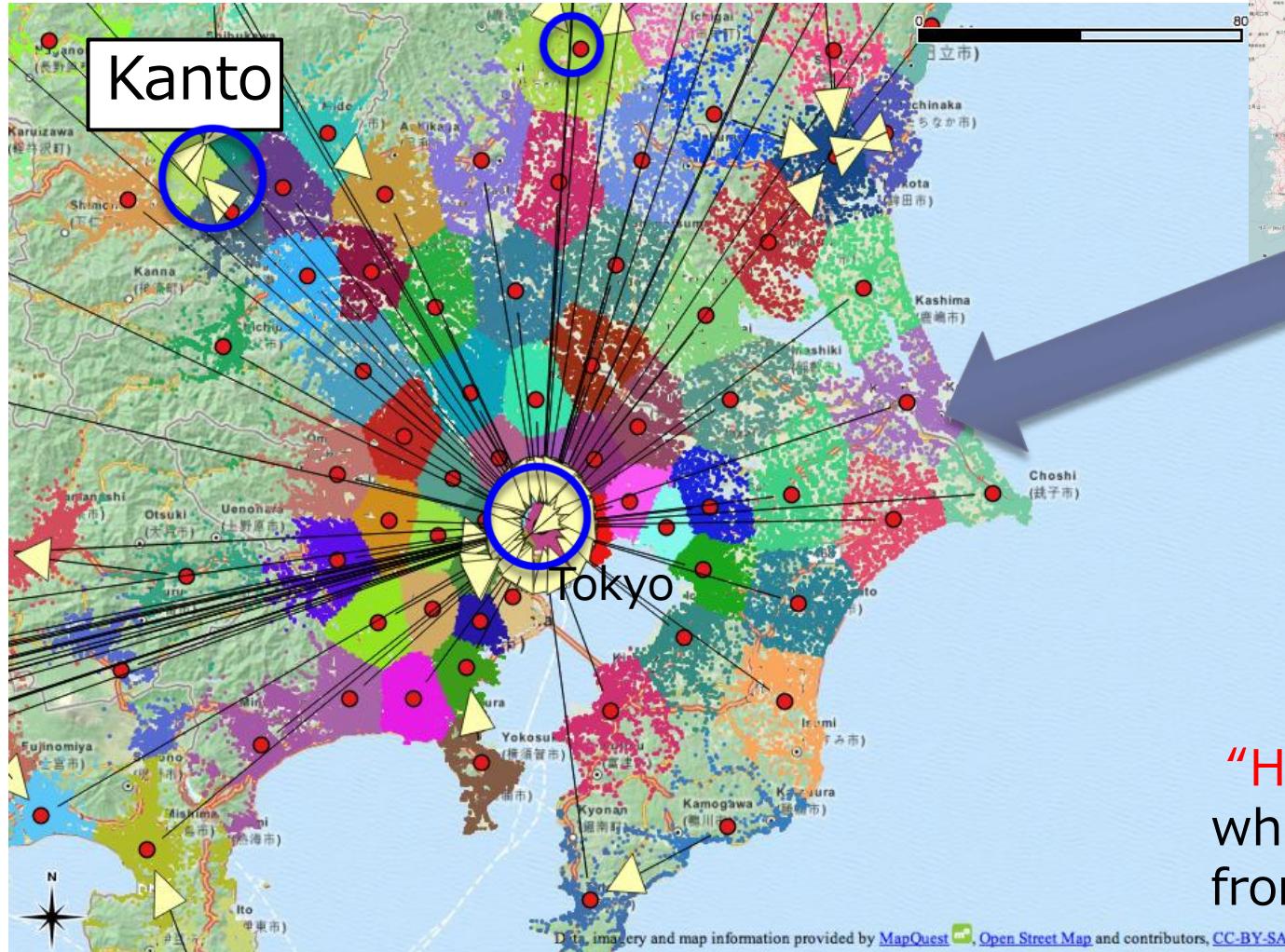
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“Hub cluster”
which collect arrows
from other clusters

Result & Discussion

Arrows: 1st transaction counterpart of each cluster

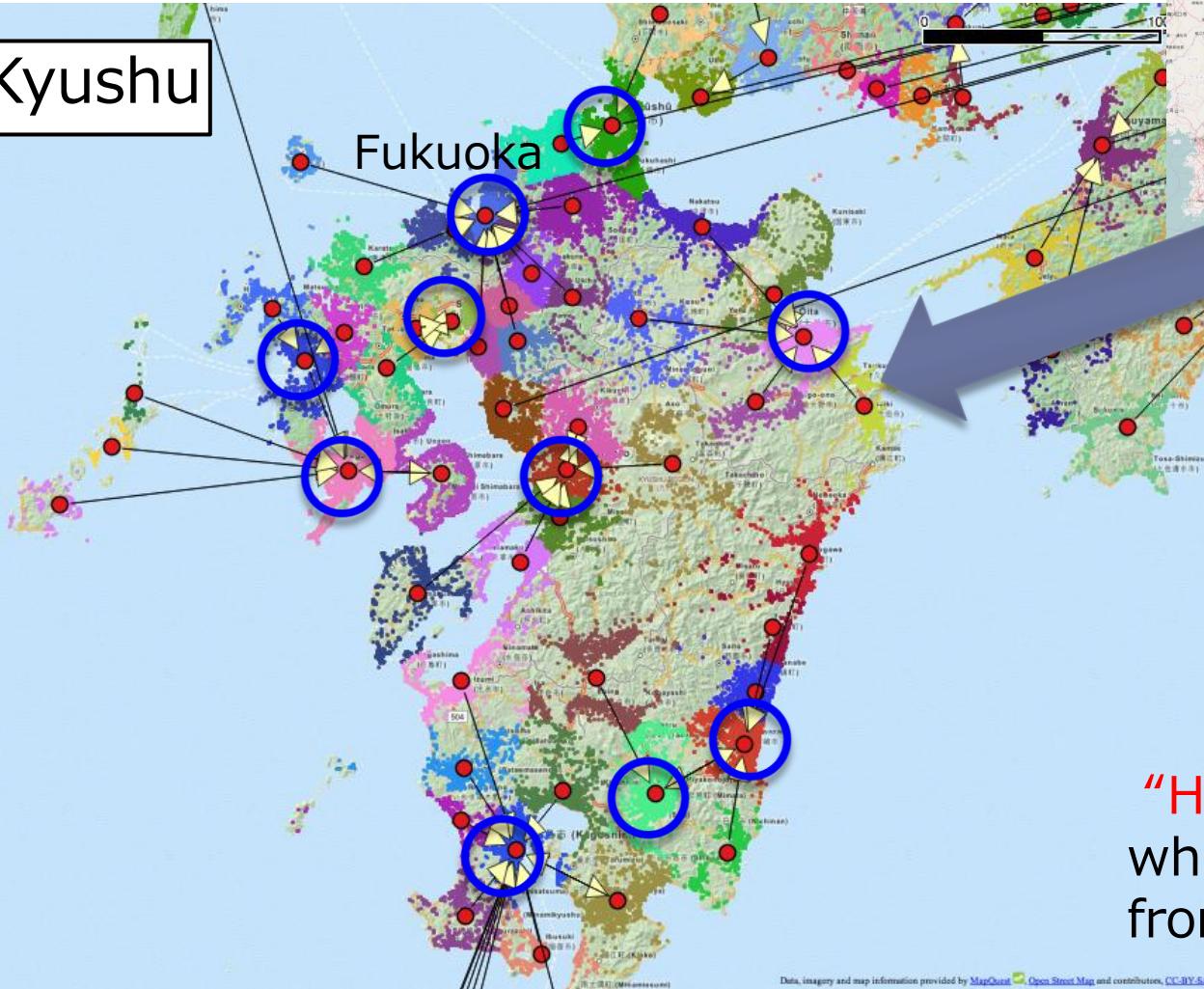


“Hub cluster”
which collect arrows
from other clusters

Result & Discussion

Arrows: 1st transaction counterpart of each cluster

Kyushu



“Hub cluster”
which collect arrows
from other clusters

Conclusion

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

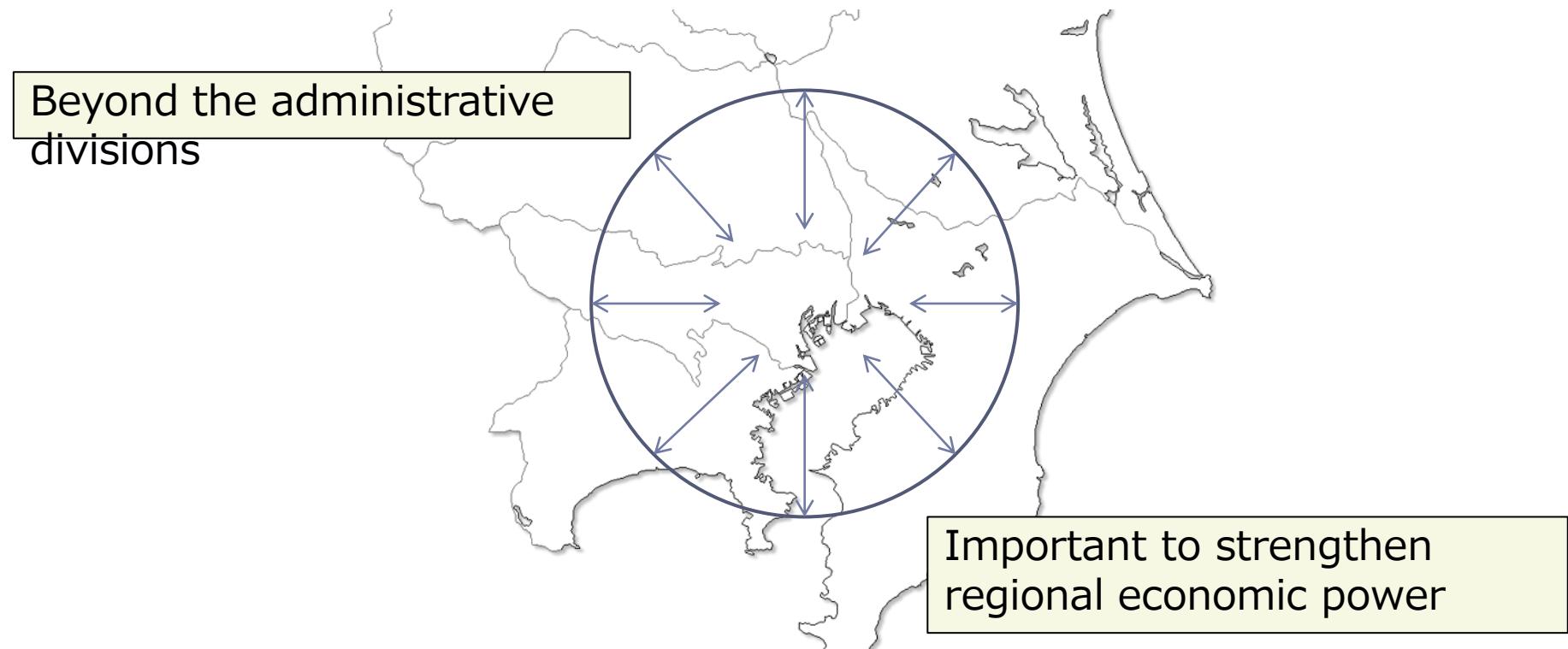


A Network Theory-based Delineation of Metropolitan areas with Mass People Flow Data

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What is a metropolitan area?

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



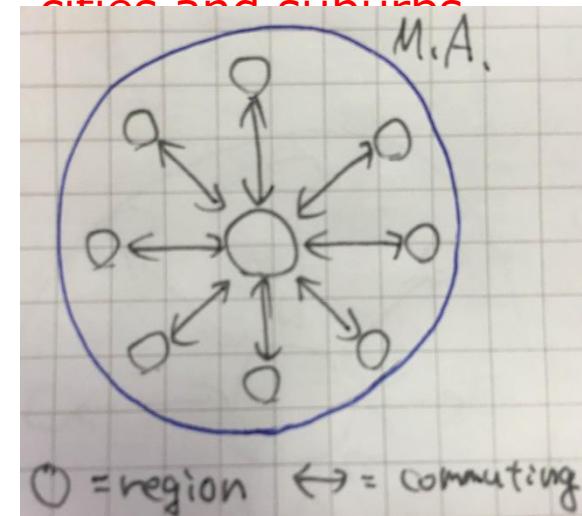
Previous research

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)	<ol style="list-style-type: none">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。 <ol style="list-style-type: none">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。
森川 (1990)	<ol style="list-style-type: none">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。
Kawashima et al. (1993)	<ol style="list-style-type: none">1. 常住人口が10万人以上。2. 昼夜間人口比率が1.0以上。 <ol style="list-style-type: none">1. 中心都市への通勤比率が5%以上、または、5百人以上。
総務庁統計局 (1999)	<ol style="list-style-type: none">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。
日本産業消費研究所 (2000)	<ol style="list-style-type: none">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。
金本・徳岡 (2002)	<ol style="list-style-type: none">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。 <ol style="list-style-type: none">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**



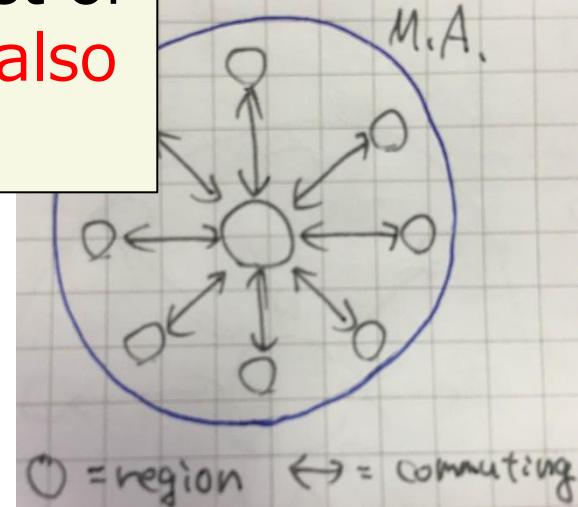
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Many previous researches define metropolitan areas based on commuting between central and suburbs



Progress of data

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…

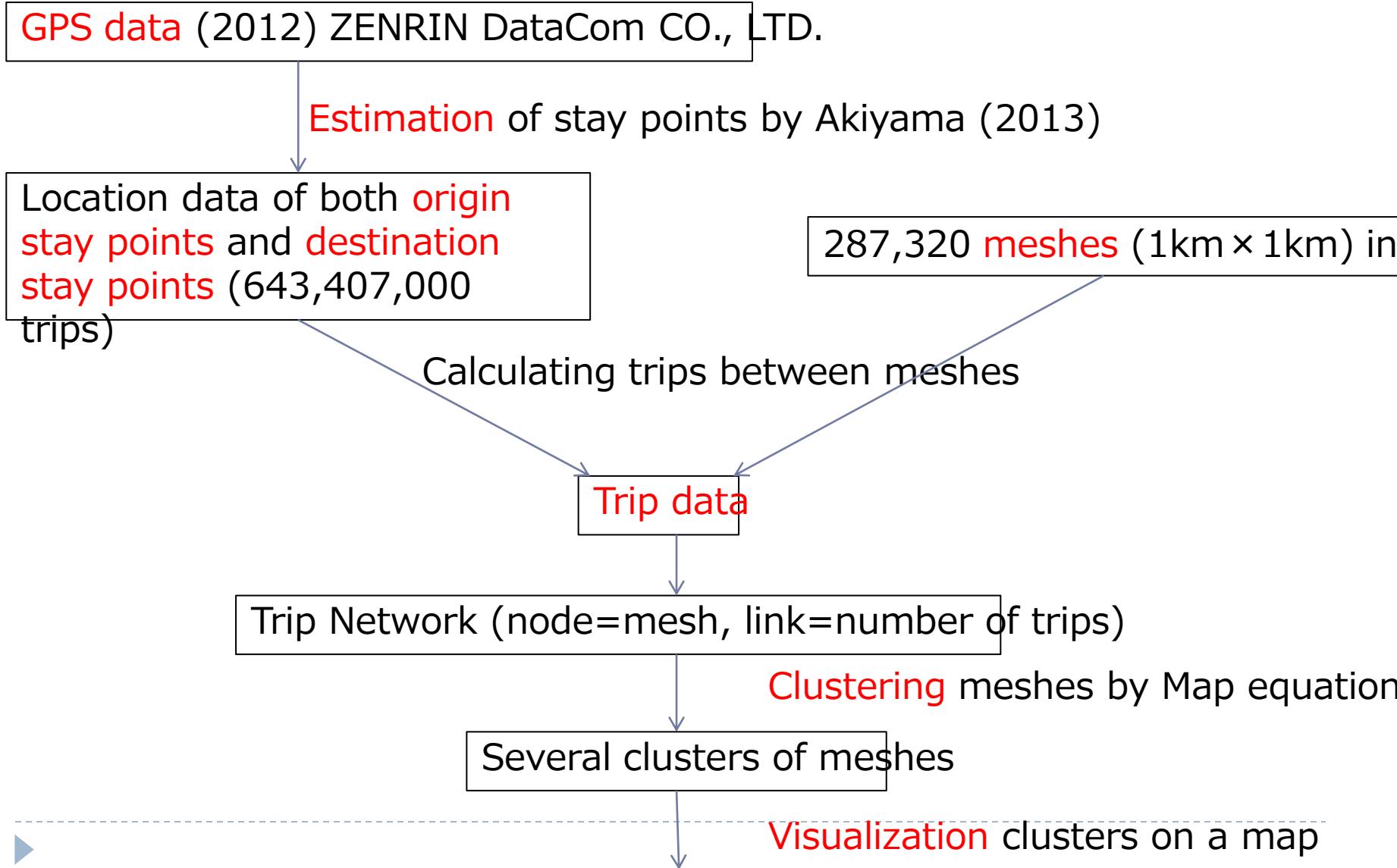
Mesh
ID:
53394611

32 trips

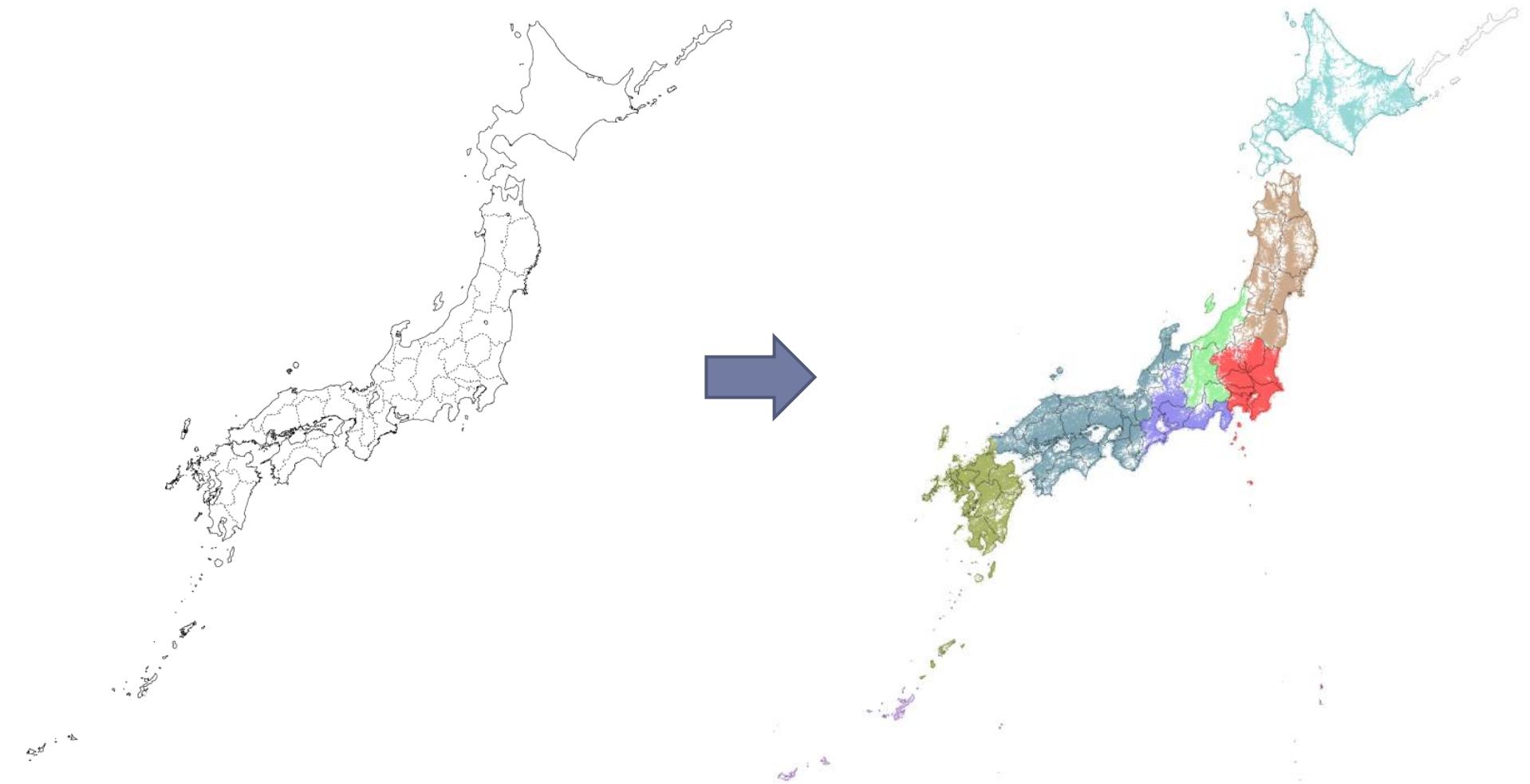
Mesh
ID:
53394449



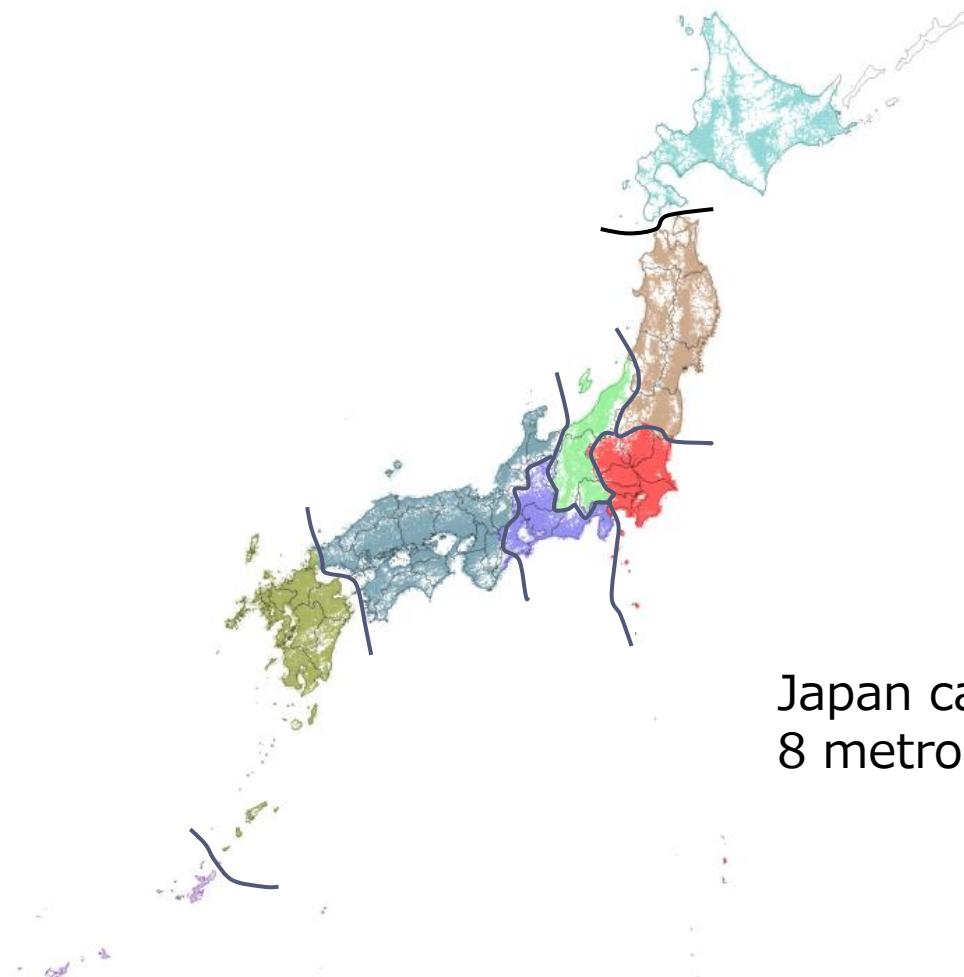
Study flow



Result of Clustering

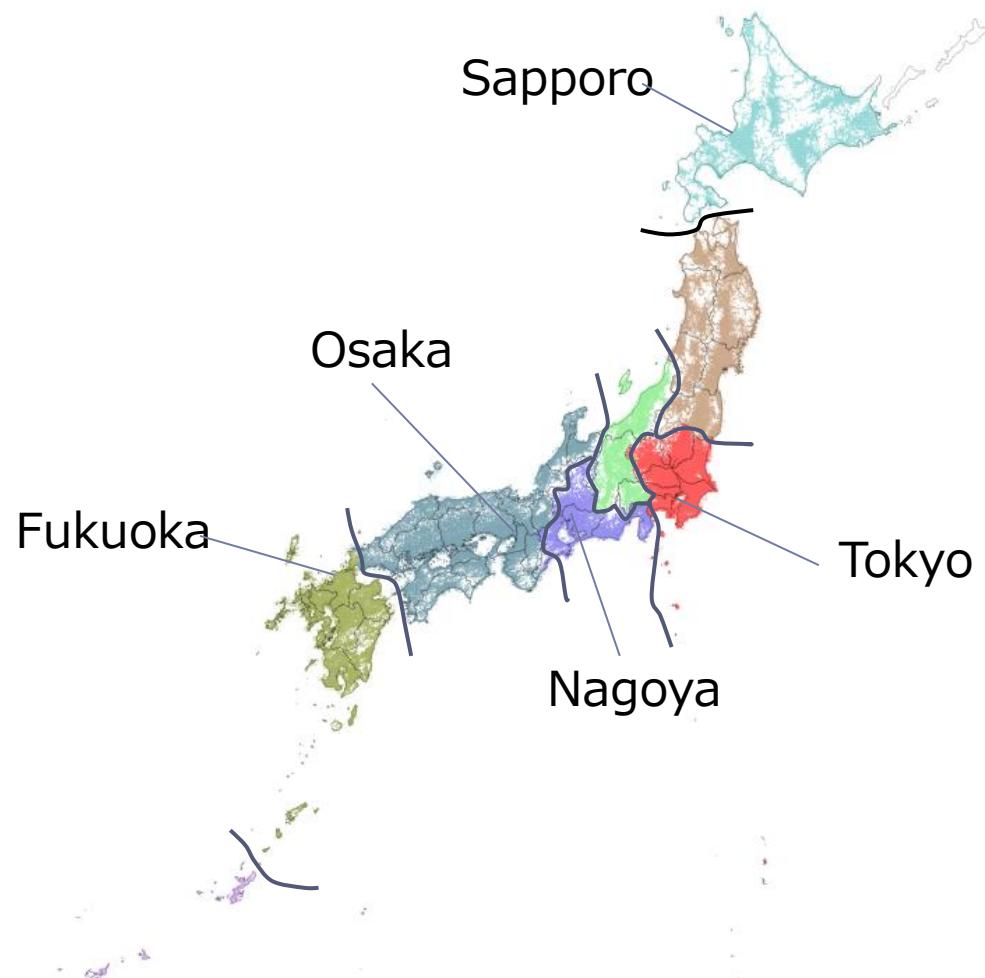


Result of Clustering

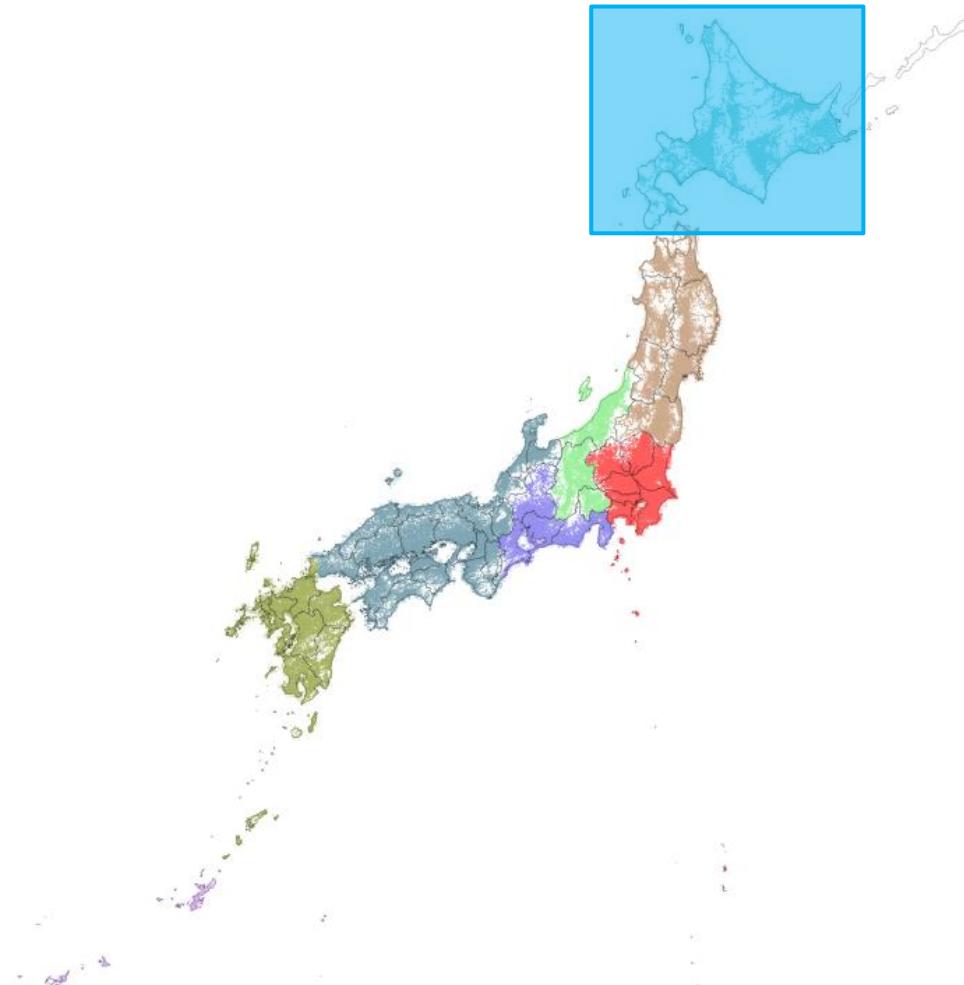


Japan can be divided into
8 metropolitan areas

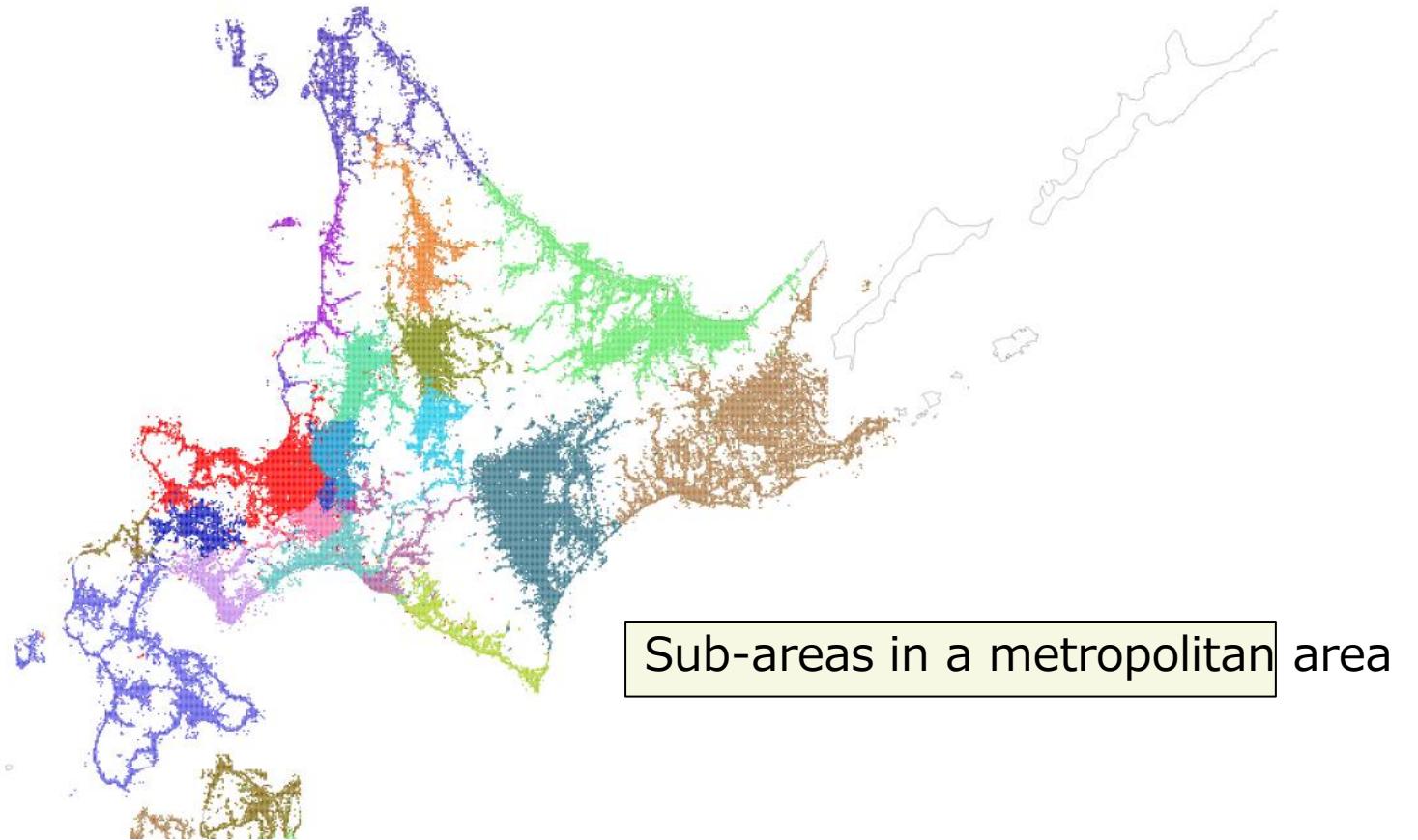
Result of Clustering



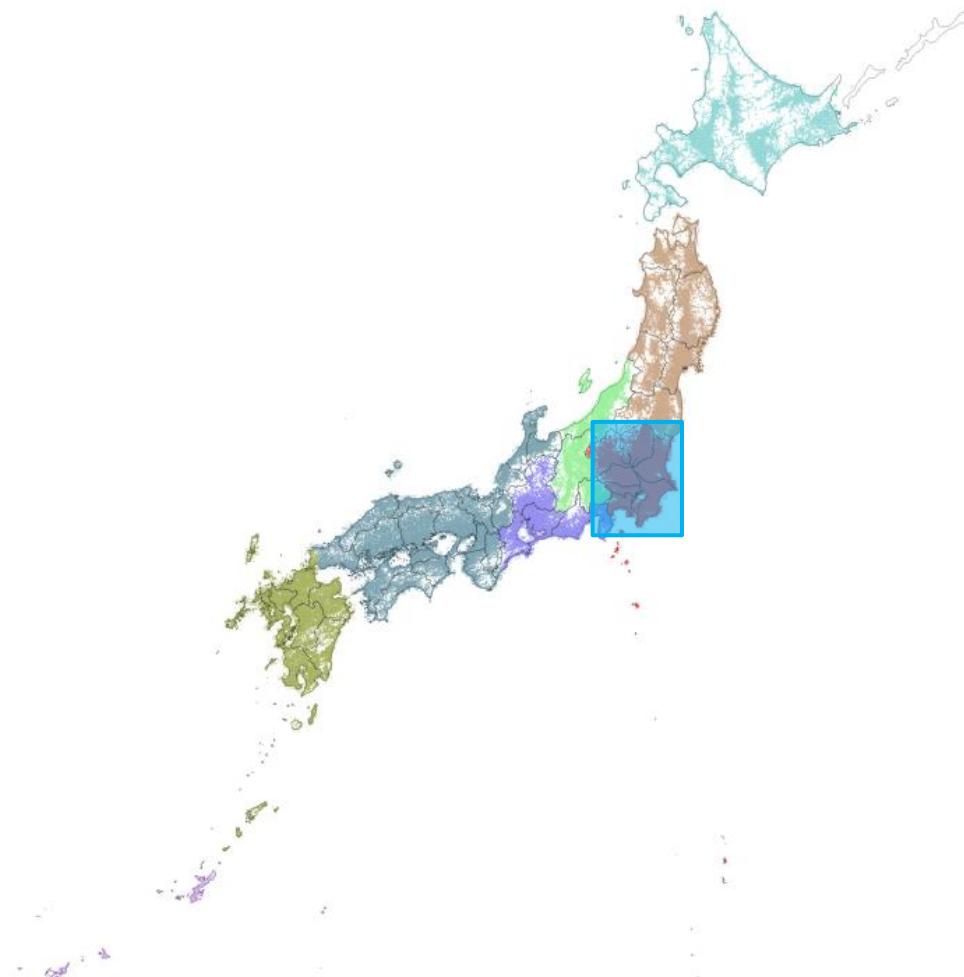
Result of Clustering



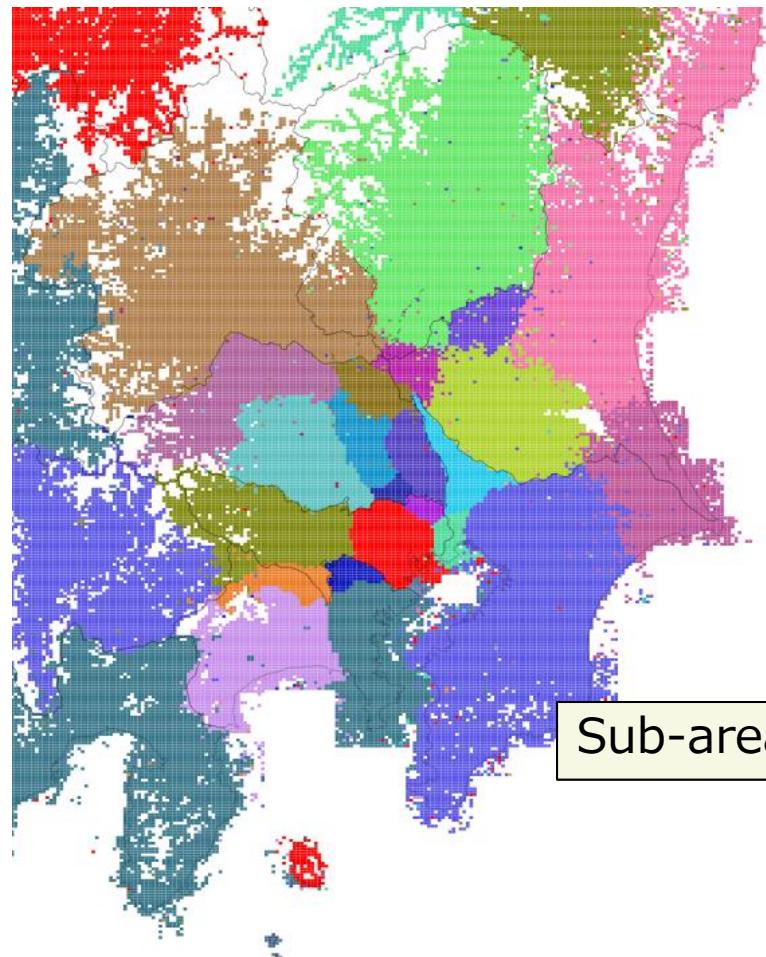
Result of Clustering



Result of Clustering

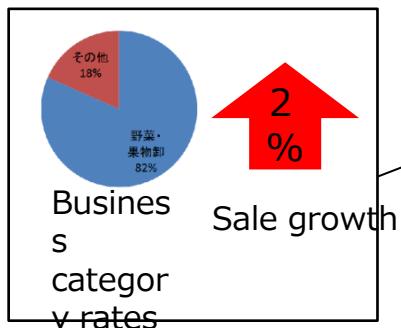
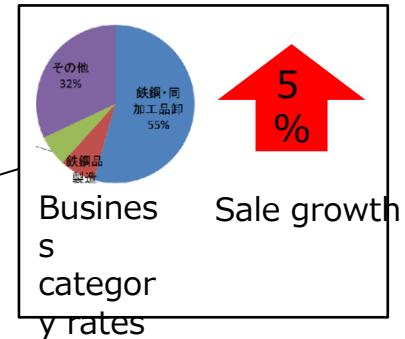
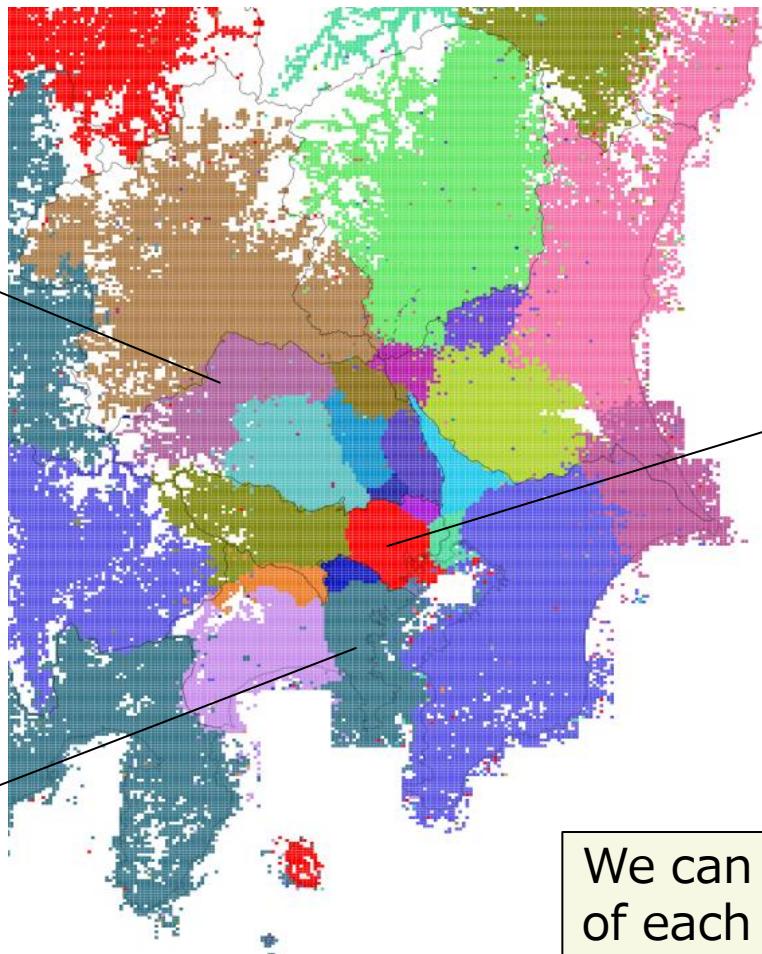
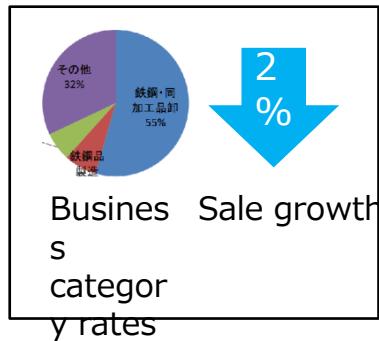


Result of Clustering



Sub-areas in a metropolitan area

Combination of metropolitan area and inter-firm big data



We can know the economic situation of each regions
→ helpful for policy making

