Human-Information Systems Laboratory

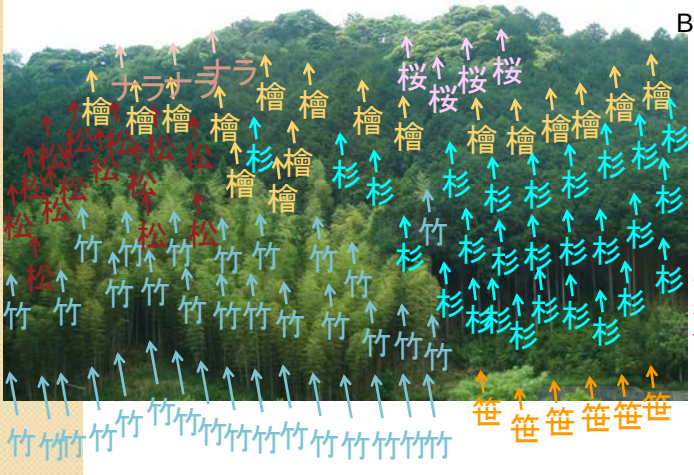
Cluster Analysis and Data Mining

Cluster Analysis is a fundamental technique for summarizing large scale data sets by partitioning similar objects into clusters. In this lecture, the basic concept of clustering is introduced with its applications.

Concept of Clustering

- How can we recognize a forest?

  Case 1： Machine-like approach (Enumeration)



Detailed information
But Too Complicated !

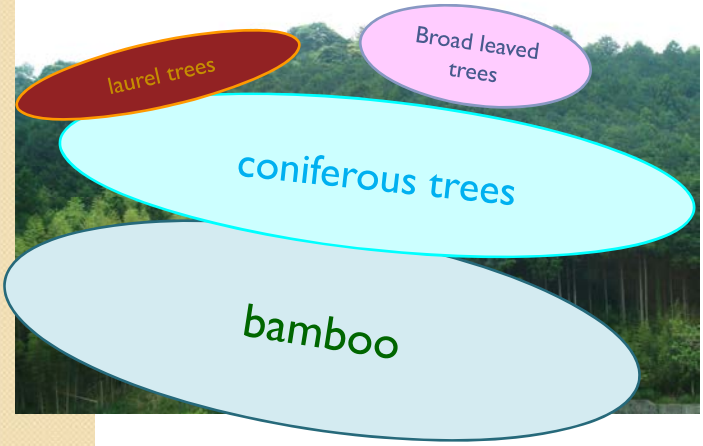
Useless !

Some people cannot see the forest for the trees.

Concept of Clustering

- How can we recognize a forest?

  Case 2：Human-like approach（Summarization）



Rough sketch

↓

Rough summary

↓

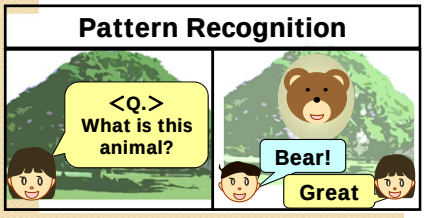
Useful !

Clustering = Grouping Activity

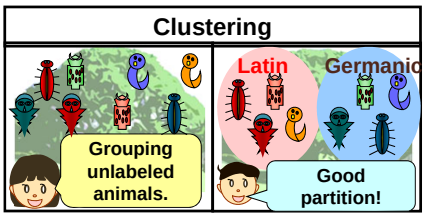
•Clustering（Cluster analysis）

- ➢Unsupervised classification
- ➢Grouping of unlabeled objects

•Difference from Pattern Recognition



Classifying a new object into a known category.



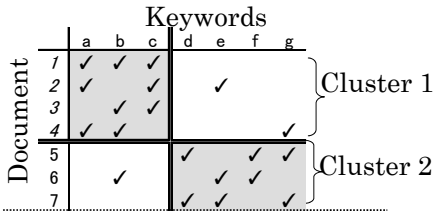
Finding an unknown category structure.

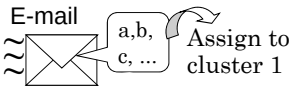
Application of Clustering

- Automatic Classification in Google News

Automatic search of similar news documents

- Automatic E-mail Classification by document clustering



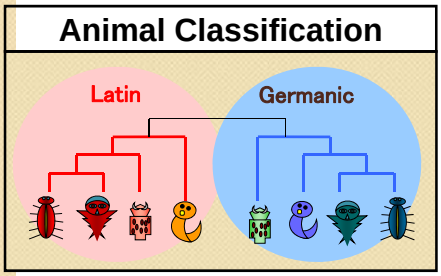


Clustering is a hot topic!

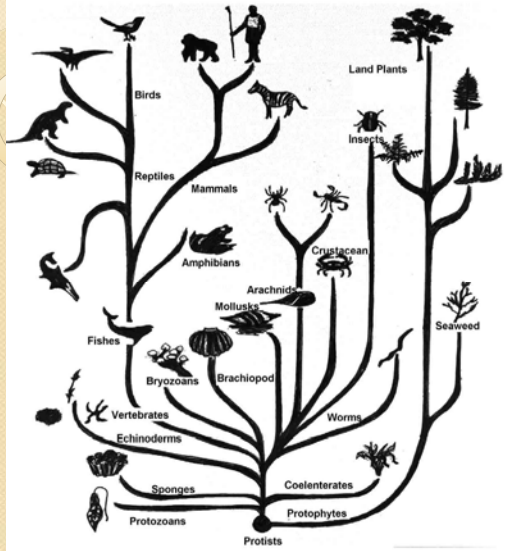
Variants of Clustering

•Hierarchical Method

- Gradually increasing (decreasing) the number of clusters.
- Summary by Tree-like structure



- Various clusters with various cutting levels
- Computationally expensive for large data sets



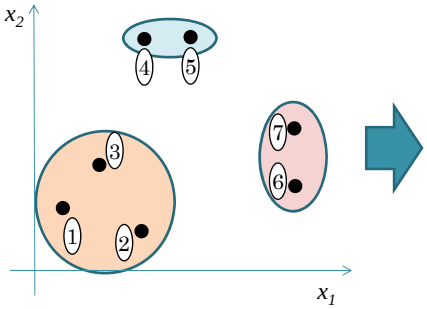
A sample of "Tree of Life Gallery"

Hierarchical clustering cannot applied to larger data sets.

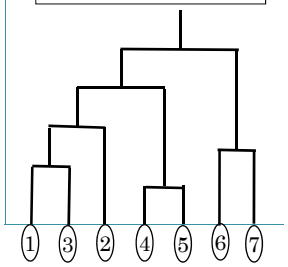
Batch process by non-hierarchical method is preferred.

Aggregation Process

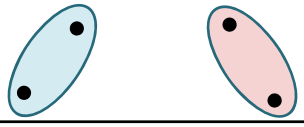
- All disjoint objects are merged into clusters one by one such that most familiar clusters are merged in each step.



Dendrogram



- What are the `most familiar clusters'?



Various distances between multi-object clusters.

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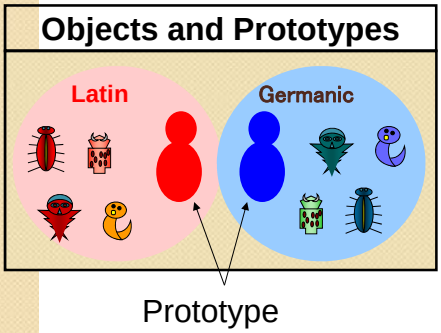
Variants of Clustering

•Non-hierarchical Method

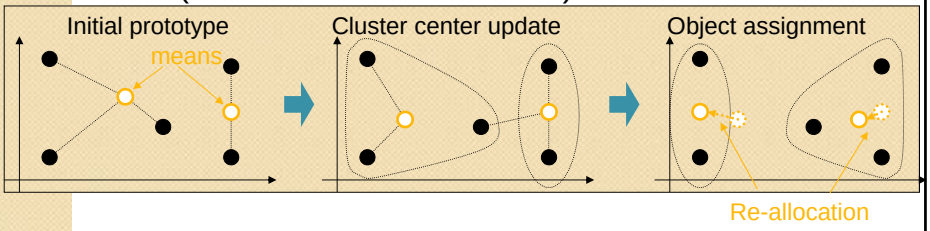
➢ Extracting a pre-defined (fixed) number of clusters, each of which represented by a cluster prototype.



- Minimizing the distances between objects and prototypes
- Suitable for large data sets
- Various extensions with different prototypes

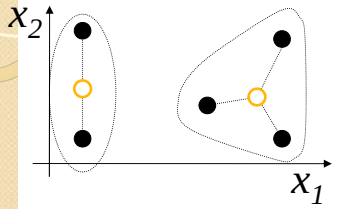
Basic method

- k-means clustering algorithm [MacQueen 1967]
  - Iterative algorithm composed of two phases
  - Prototype estimation phase (Cluster center update)
  - Nearest prototype assignment （Cluster membership update）
  - Prototype = k mean vector (k is the number of clusters)



Mathematics of k-means

•From the view point of minimizing an objective function...



n objects

$\boldsymbol{x}_1, \boldsymbol{x}_2, \cdots, \boldsymbol{x}_n$

K prototypes

$\boldsymbol{b}_1, \boldsymbol{b}_2, \cdots, \boldsymbol{b}_K$

Within-cluster errors

$$\text{min.} \quad L_{KM} = \sum_{k=1}^{K} \sum_{i \in G_k} \| \boldsymbol{x}_i - \boldsymbol{b}_k \|^2$$

← Errors in cluster G_k

Rewritten with crisp memberships

$$\text{min.} \quad L_{KM} = \sum_{k=1}^{K} \sum_{i=1}^{n} u_{ki} \| \boldsymbol{x}_i - \boldsymbol{b}_k \|^2$$

← u_{ki} has 1 only in a cluster
⇒Winner takes all !

$$\text{s.t.} \quad u_{ki} \in \{0,1\}, \quad \sum_{k=1}^{K} u_{ki} = 1$$

Conclusion

- Clustering is a computational realization of important human-like activity of `grouping'.
- We have two main schemes of `hierarchical approach' and `non-hierarchical approach'.
- Hierarchical approach can derive dendrogram, which is useful for deriving various numbers of clusters, but useless for large scale data.
- Non-hierarchical approach is useful for handling large scale data, which iterates object assignment and prototype updating.
- Clustering is available in many real applications.

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知能情報特論I 26th July, 2017
Advanced Intelligent Information Systems I

Collaborative Recommendation System

Human Information Systems Laboratory
Katsuhiro Honda

Agenda

- Review on Data Mining
- Concept of Collaborative Filtering
- Algorithm in Amazon recommendation

Background

- **Information Pollution** ⇒ **Information explosion**
  In advanced information society, we have too many information to handle them.
- **Data Mining**
  Extraction of useful information with computational supports.
- **Collaborative Filtering** (Information Filtering)
  Automatic selection of information for users
  Computational realization of "Word-of-Mouth" through user collaboration

Human friendly information societies!

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What is data mining?

•Data
= electrically stored information



Purchase　medical　education

(Info-plosion in various fields)

•Mining・・・gold, coal, etc.

•Data Mining
Treasure hunting from data mount..

Computational support

Knowledge discovery for future decision support

3

What is useful knowledge?

- Example: a market

Strategic marketing with weather information

•Conventional statistics
⇒Statistical significance (trivial)
「Few customers in rainy days」 Trivial…

•Smart strategy
⇒Strategic knowledge
Seasonal difference of Weather influences
Complex features

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3 key elements of Data Mining

- Observation･･･We must carefully observe data.

- Sampling･･･We should extract meaningful group.
  Clustering! (My early lecture)
  Extension from conventional statistics

- Correlation ･･･We can find pattern in the group.
  (correlation (association) rule analysis)
  Utilization of conventional statistics !

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For Human-friendly information society･･･

- IT method for supporting finding preferable information from info-mountain

Not everyone searching for Wally!

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An example...

How do you find YOUR interesting information from data masses?

I have nothing to do today.
Let's enjoy a "Funny" video content.
How can we find a "Funny" one?

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Limit of conventional information retrieval

- **Ranking**: Popularity-based Recommendation

  Majority tendency = Ignoring personal feelings

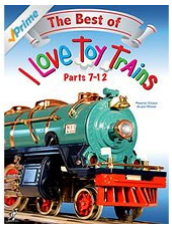
  Unable to recommend based on personal preferences

- **Keyword search**: Content-based recommendation

  • Comparison between keyword and content label

  Label may not fit user's feeling.

  Funny ≠ Feeling for Content



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How will you find a funny content in real world?

- **Word-of-Mouth Communication**
  - You often prefer to ask your friends for promising recommendation!
  - You often recommend favorite items to your friends!

Why?

You expect that your friend have similar preference to you!

Word-of-Mouth communication = Recommendation from Friends!

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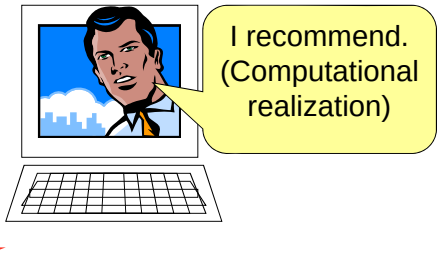
Collaborative Filtering = Preference-based information filtering

- **Considering how users feel for items**

⇒Comparison of users feelings such as ratings on items

Personalized recommendation is achieved by finding `Friends'!



Recommendation by neighborhood user search

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Collaborative filtering: Virtual Word-of-Mouth

Advantages of collaborative filtering

(i) **Available without content analysis**
We cannot check all labels of all contents.

(ii) **Recommend based on item quality or tastes**
We need not present appropriate keywords.

(iii) **Provide serendipitous recommendation**
We can find (unknown) valuable content.

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Famous application of collaborative filtering

Information filtering and E-commerce

Customers who bought this item also bought

Page 1 of 4

Hero's Life Death and Transfiguration
Richard S ···
CD
¥ 858

Great Mass in C
Benjamin ···
★★★★★ 12
CD
¥ 1,229 prime

Wagner: Orchestral Music
Richard W ···
★★★★★ 1
CD
¥ 2,099 prime

What other items do customers buy after viewing this item?

ドヴォルザーク:交響曲第8番&第9番「新世界より」 CD
カラヤン(ヘルベルト・フォン)
★★★★½ 14
¥ 1,571 prime

Efficient marketing based on user preferences

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Personalized Recommendation in Amazon

Your Store | Your Browsing History | Recommended For You | Improve Your Recommendations | Your Profile | Help

Your Store › Recommended for you
(If you're not 本多克宏, please sign in.)

Recommendations

- Amazon Video
- Apps for Android
- Baby & Maternity
- Car Products
- Digital Music
- DVD
- Electronics, Cameras & AV
- Fashion
- Food, Beverage & Alcohol
- Foreign Books
- Home & Kitchen
- Japanese Books

These recommendations are based on items you own and more.

view: **All** | New Releases | Coming Soon

1. 

Party Queen(AL+DVD2枚組) [CD+DVD]
~ 浜崎あゆみ (March 21, 2012)
Average Customer Review: ★★★½☆ (154)
In Stock

List Price: ~~¥ 7,560~~
Price: ¥ 1,728
Amazon Points: 5pt
34 used & new from ¥ 16

☐ I Own It ☐ Not interested ☒ ☆☆☆☆☆ Rate this item

Recommended because you purchased **NEXT LEVEL【初回限定生産】(2CD+DVD)(ジャケットA)** and more (Fix this)

Personalized recommendation based on my personal purchase history

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Collaborative filtering

✓personalized recommendation by mutual collaboration of users

✓Missing value estimation in evaluation matrix

- **Evaluation matrix: ratings for items by users**

	Golf	Soccer	Ski	Tennis
Andy	5		4	5
Bob	2	5		1
Clark		4	1	2
Dick	5	1	5	

Missing element

Items with high prediction values are recommended.

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Neighborhood-based algorithm

- **Famous application to net news: GroupLens**

1. Neighborhood search
2. Averaging in neighborhood)

	Golf	Soccer	Ski	Tennis
Andy	5		4	5
Bob	2	5		1
Clark		4	1	2
Dick	5	1	5	

neighborhood

⇒ Similar to Dick
⇒ Not similar
⇒ Not similar

Andy likes tennis ⇒ Dick would also like tennis

Sampling and Correlation analysis ⇒ Data mining!

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Comparison of users

- **Correlation analysis in GoupLens**

	Golf	Soccer	Ski	Tennis	rugby	cricket	cycling	Judo	sumo	karate
Andy	5	3	1	4	4	1	3	5	2	2
Dick	4	2	1	3	2	1	4	4	3	2

- Statistical measure : Pearson Correlation Coefficient

$$\frac{\sum_{i=1}^{n}(x_i-\bar{x})(y_i-\bar{y})}{\sqrt{\sum_{i=1}^{n}(x_i-\bar{x})^2}\sqrt{\sum_{i=1}^{n}(y_i-\bar{y})^2}}$$

$\bar{x}$: **Average of** x

$\bar{y}$: **Average of** y

Statistical similarity between x and y
= How correctly y will be predicted from x

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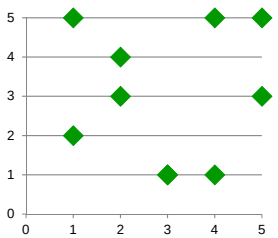
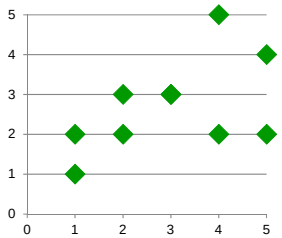
Correlation… revisited

- Comparison of correlation coefficients

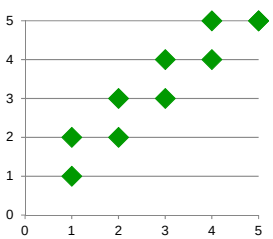
	Golf	Soccer	Ski	Tennis	rugby	cricket	cycling	Judo	sumo	karate
Andy	5	3	1	4	4	1	3	5	2	2
Dick	4	2	1	3	2	1	4	4	3	2



Correlation:0.04



Correlation:0.51



Correlation:0.94

Unpredictable(0.0) < Weakly related(0.5) < Equivalent(1.0)

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Representative model for Collaborative Filtering

•User Neighborhood-based Algorithm

A user neighborhood having similar preferences are first searched. The applicability of new items are calculated as the weighted average of their ratings.

（Ex.）GroupLens

$$p_{ij} = \bar{x}_i + \frac{\sum_{u=1}^{n} (x_{uj} - \bar{x}_u) \times \omega_{iu}}{\sum_{u=1}^{n} \omega_{iu}}$$

Correlation coefficients among users

GroupLens uses the deviation from average for ignoring users' biases.

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Several Approaches to Collaborative Filtering

- **Memory-based algorithm**
  GroupLens or MovieLenz
  ⇒**All evaluations are stored in `Memory'**
  Calculation for each user arrival
  Realization of human word-of-mouth
- **Amazon.com Recommendation System**
  GroupLens-based recommendation with `item lists'
  Hybrid of Memory-based and Model-based systems
- **Model-based algorithm**

Purchase history is summarized in `Model' for quick action.
⇒ Without evaluation data ⇒ Low memory requirement
⇒ Low Calculation costs ⇒ Low network traffics
We developed a clustering-based approach

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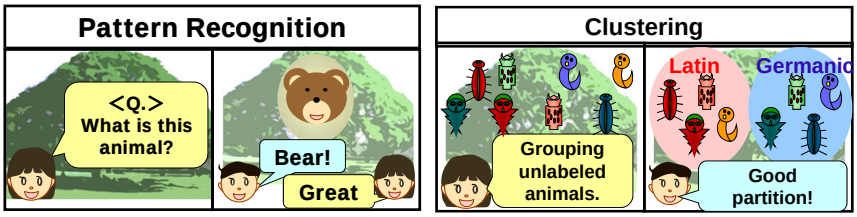
Clustering = Grouping Activity

- Clustering（Cluster analysis）
  - ➢Unsupervised classification
  - ➢Grouping of unlabeled objects
- Difference from Pattern Recognition



Classifying a new object into a known category.

Finding an unknown category structure.

Recommendation by Co-clustering

- Recommendation based on user-item groups

	item1	item2	item3	item4	item5	item6
user a	O	O				O
user b	O	O			O	
user c		O	O			
user d		O	O	O		
user e					O	O
user f		O		O		

Pair-wise clusters of user-item

	item1	item5	item6	item3	item4	item
user a	O		O			O
user b	O	O				O
user e		O	O			
user c				O		O
user d				O	O	O
user f					O	O

Information summary by co-clusters

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Conclusions

- In this lecture, I introduced the basic concept of collaborative filtering, which can be a computational realization of human Word-of-Mouth communication.
- In practical use such as Amazon online shop, some modifications were implemented for reducing computational costs.
- It may be possible to make it more effective with model-based algorithms. It will be nice if you are interested in such technologies.

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