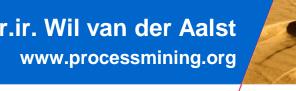
Process Mining: Data Science in Action

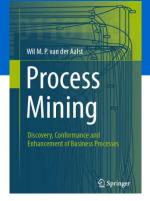
#### **Cluster Analysis**



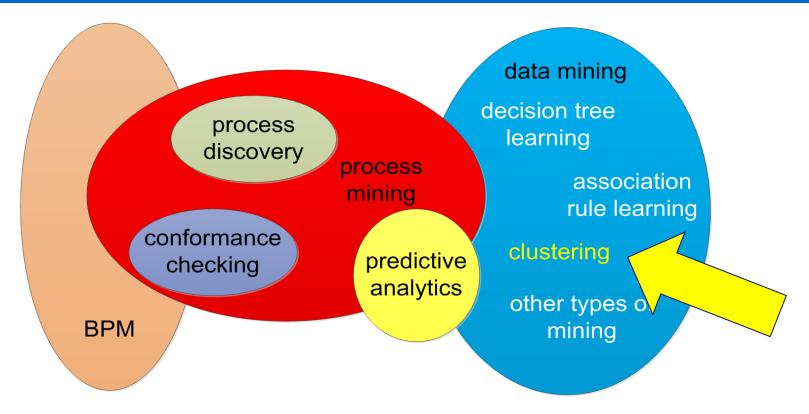




Where innovation starts

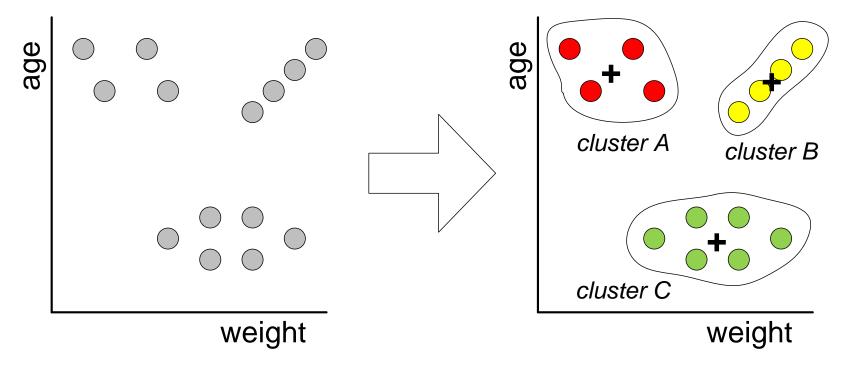


# Clustering (unsupervised learning: no response variable)



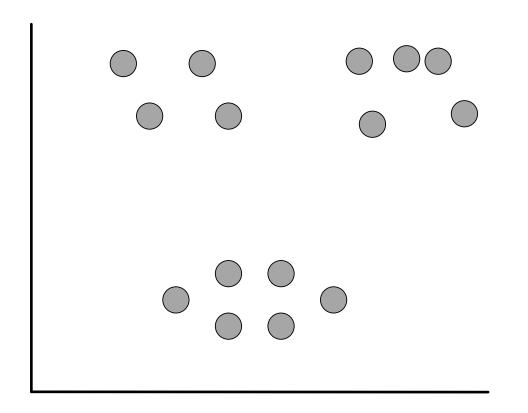


### Clustering





### k-means clustering

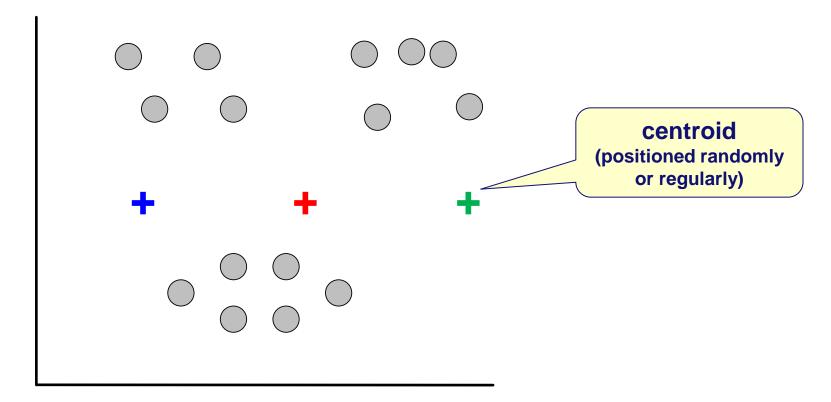


#### assume k=3

Just an illustration.

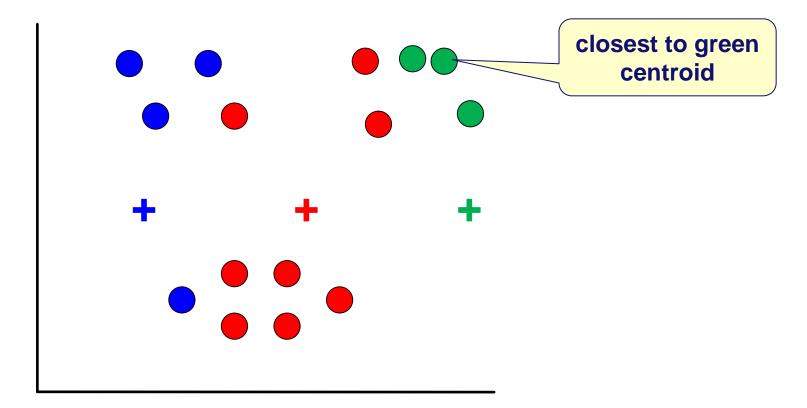
May be misleading:
often many
dimensions! TU/e

#### Generate k=3 centroids (e.g., random)



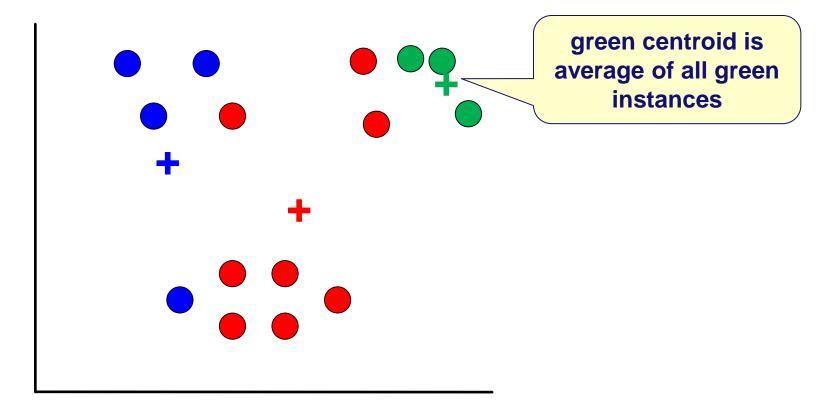


#### Assign instances to closest centroid



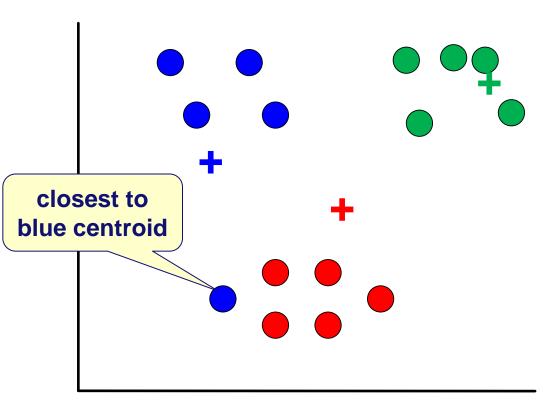


# Recompute centroids based on assigned instances



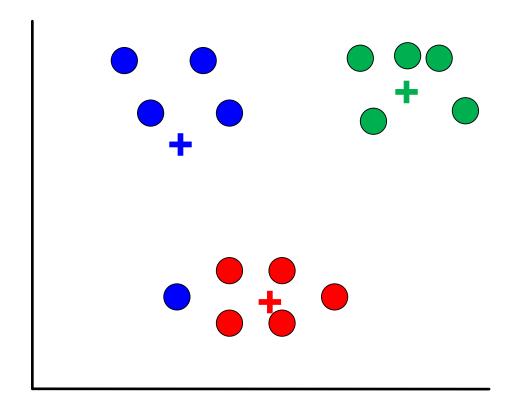


#### Assign instances to closest centroid



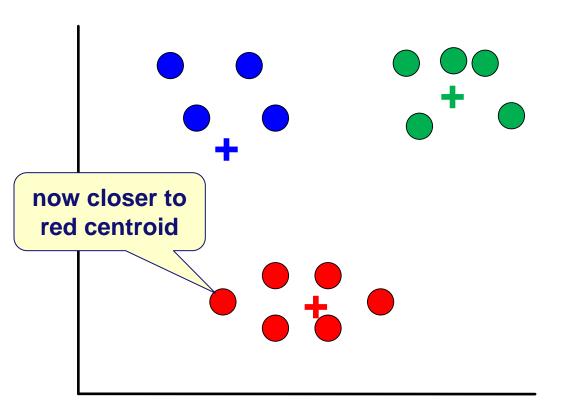


# Recompute centroids based on assigned instances



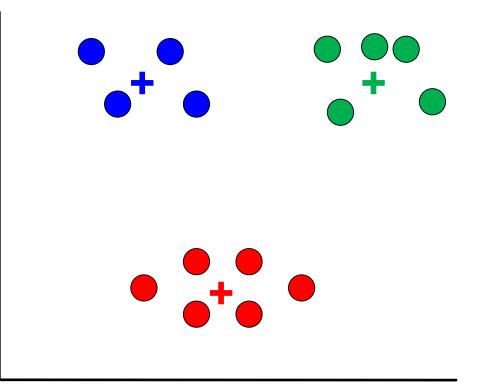


#### Assign instances to closest centroid





# Recompute centroids based on assigned instances

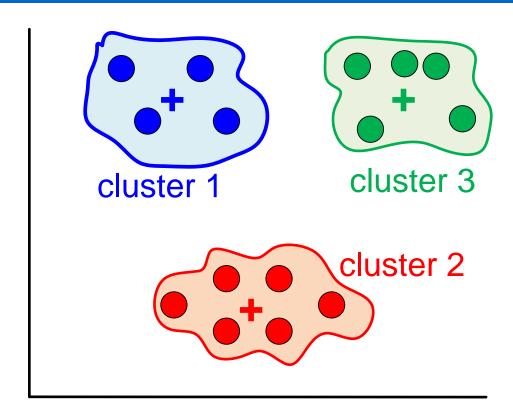


Fixpoint has been reached: Nothing changes anymore.

Due to non-deterministic nature (random initial centroids), experiment may be repeated multiple times. Select "best clustering" at end.



#### Clusters returned

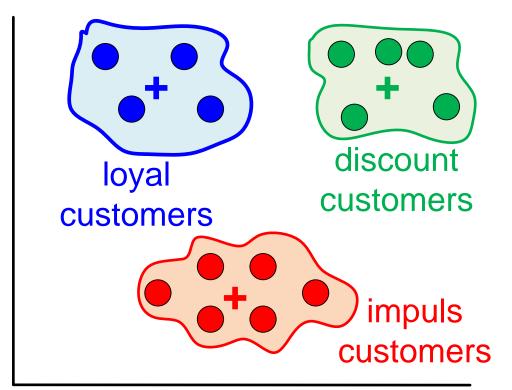


#### Main idea:

The instances in a cluster are more similar to each other than to those in other clusters.



#### Many use cases

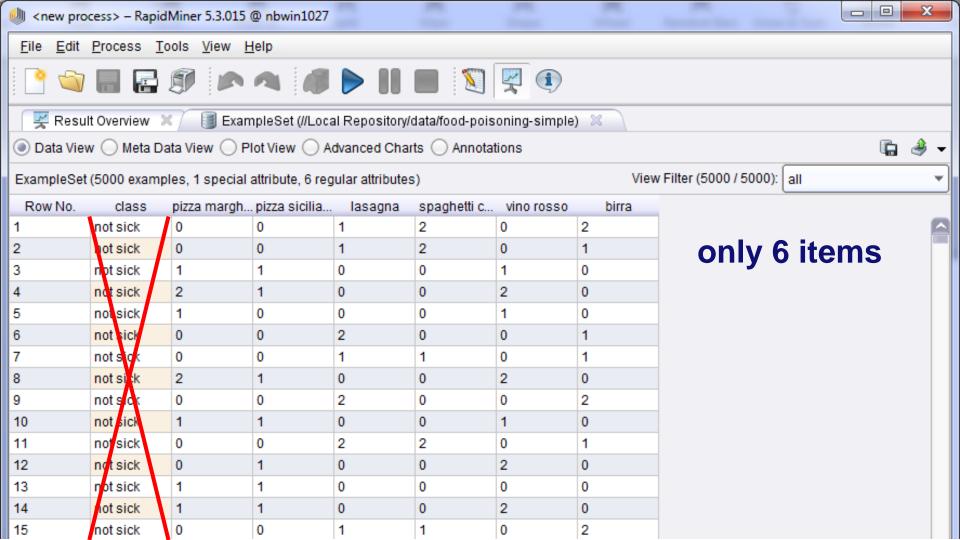


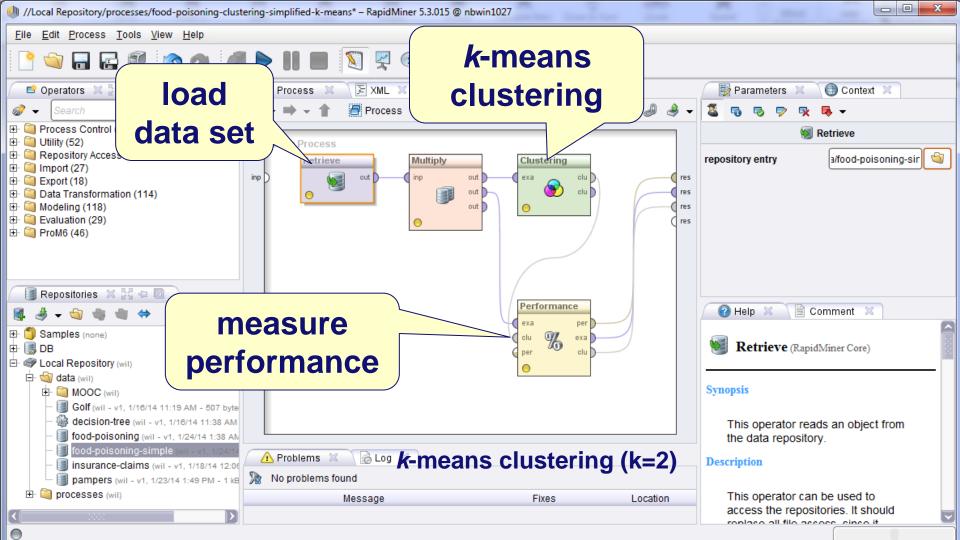
For example: finding homogenous groups of customers, patients, sessions, students, etc.

After clustering: apply additional mining techniques on the partitioned input data.









#### Results *k*-means clustering (k=2)

#### Two clusters:

- Cluster 0: 2551 instances

- Cluster 1: 2449 instances

#### Centroids

Text View Fold	der View 🔘 Grap	ph View ( Ce
Attribute	cluster_0	cluster_1
pizza margherita	0.008	1.026
pizza siciliana	0.011	1.028
lasagna	0.987	0.011
spaghetti carbonara	0.984	0.008
vino rosso	0.011	1.046
birra	0.995	0.007



Cluster 0: 2551 instances

birra

Cluster 1: 2449 instances

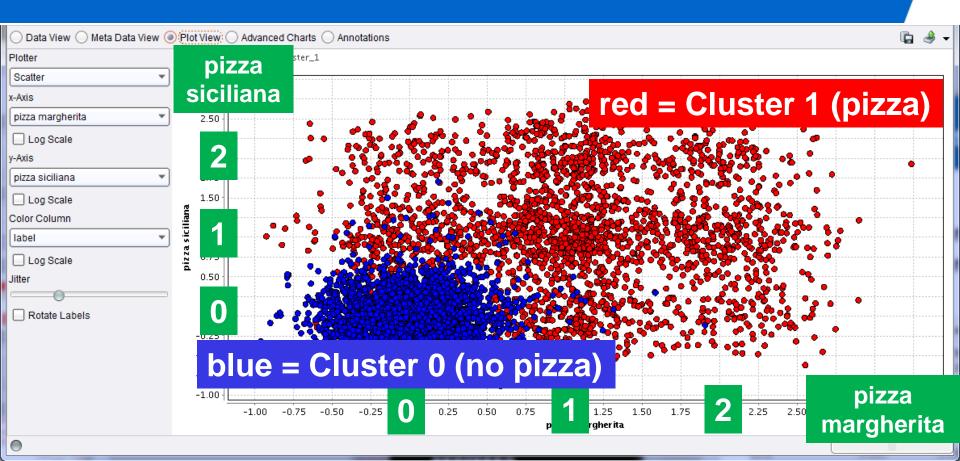
vino rosso



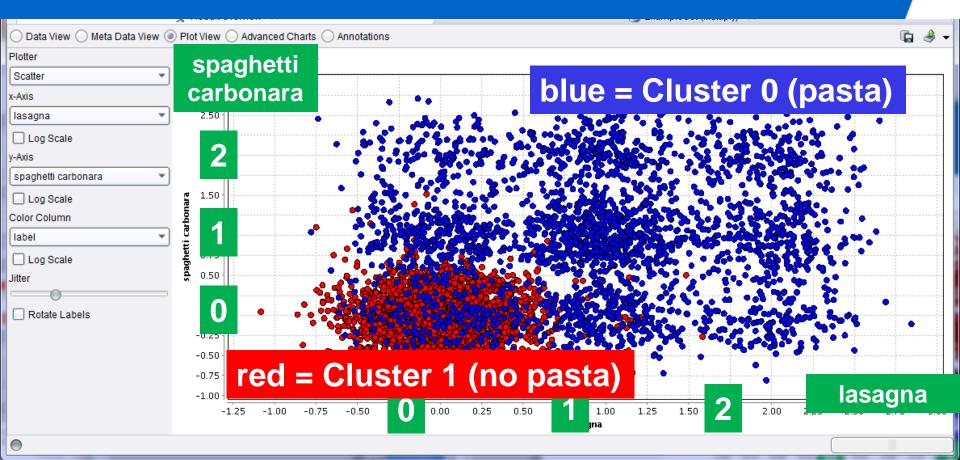
lasagna spaghetti carbonara

pizza siciliana pizza margherita

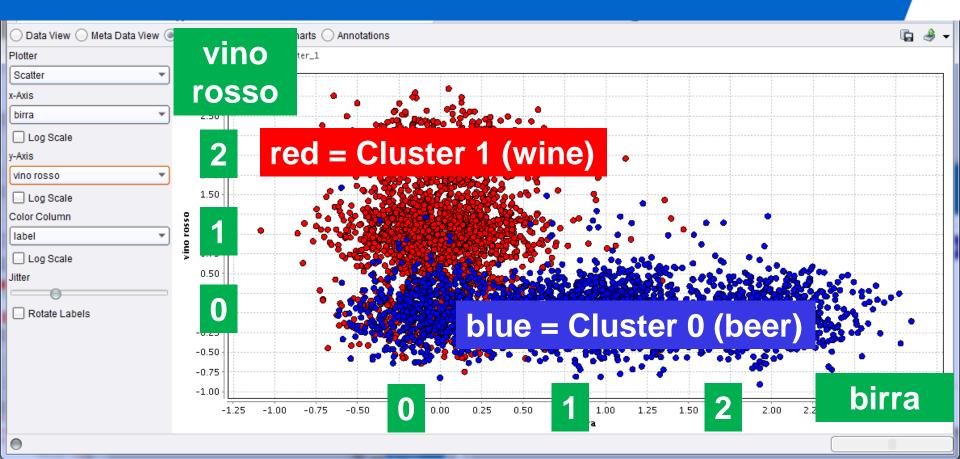
### Scatter plot (with jitter)



## Scatter plot (with jitter)

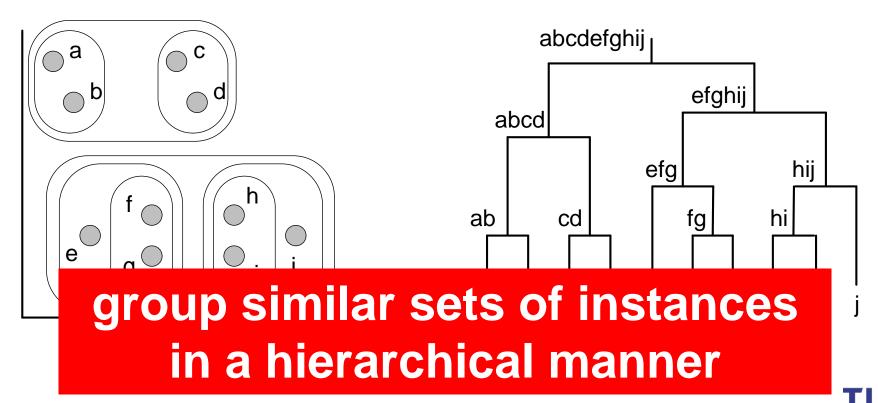


## Scatter plot (with jitter)

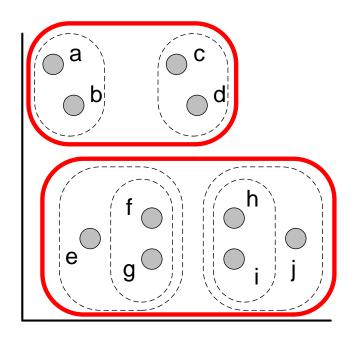


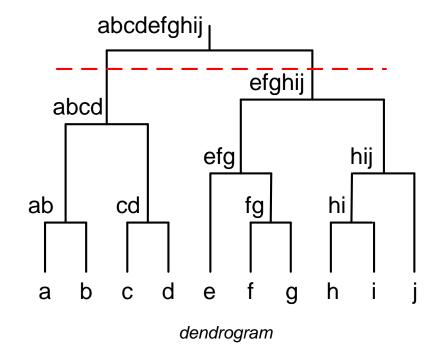


### Agglomerative hierarchical clustering



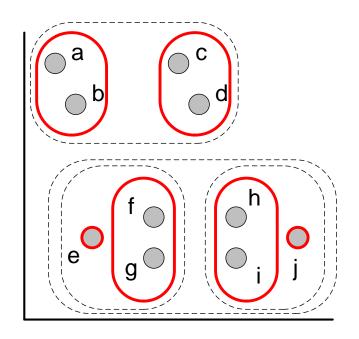
#### Two larger clusters

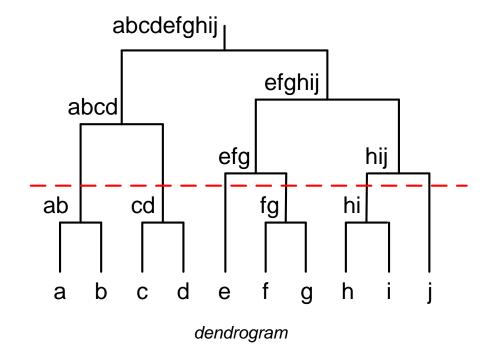






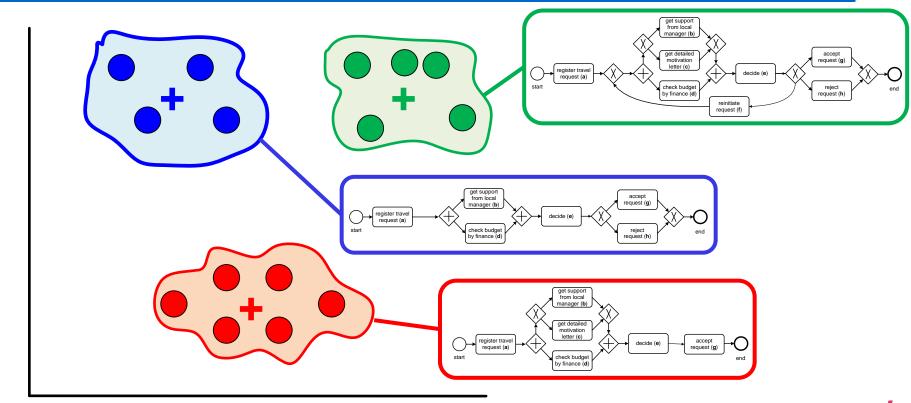
#### Six smaller clusters





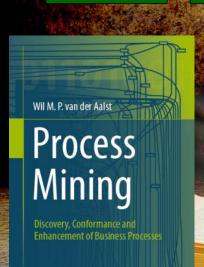


#### Clustering can be used to split event logs

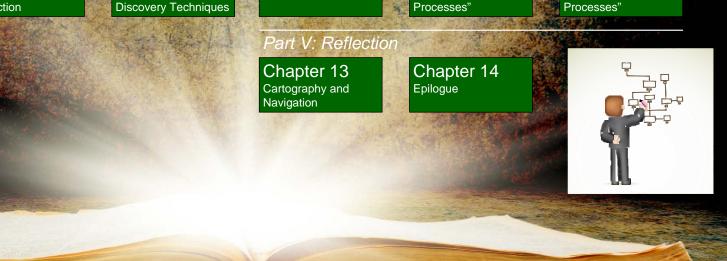




#### Part I: Preliminaries Part III: Beyond Process Discovery Chapter 2 Chapter 3 Chapter 7 Chapter 8 Chapter 1 Process Modeling and Data Mining Introduction Conformance Mining Additional **Operational Support** Analysis Checking Perspectives Part II: From Event Logs to Process Models Part IV: Putting Process Mining to Work Chapter 11 Chapter 4 Chapter 5 Chapter 6 Chapter 10 Process Discovery: An Getting the Data Advanced Process **Tool Support** Analyzing "Lasagna Analyzing "Spaghetti Introduction



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Chapter 9

Chapter 12