

The Data Mining Methods Conceptual map

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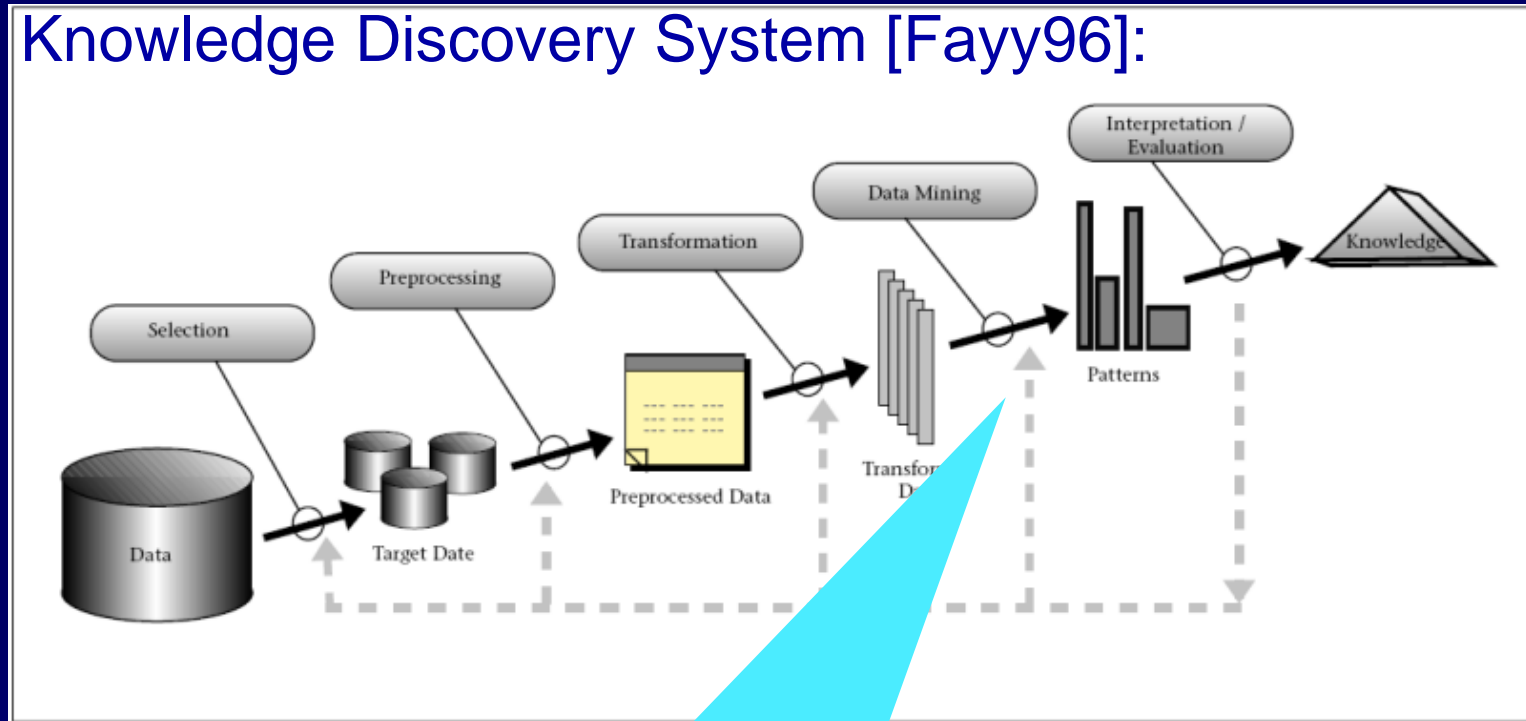
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Data Mining and Knowledge Discovery

- Knowledge Discovery System [Fayy96]:



DM method choice

- Critical
 - Suitable tools increase
 - Methodological expertise required

Choosing DM-technique [iEMSs10][EMSO18]

- Computational DM systems: Catalogues with many options
- Abundant Literature
- Choices grouped under different criteria:
 - Technical proximity [Gibert et al, 2008]
 - Research area, ...
- Expert decision criteria (EDC):
 - Goal of problem to be solved
 - Structure of available data set
 - Method's properties
 - Future use of the resulting model
- Conceptual map of DM methods based on EDC [Gibert et al, 2015]
 - Modelling decision process itself
 - Good-practice guidelines for non-expert users
 - Building intelligent recommenders



Evidenced by
real experiences



Towards
Integral KDD systems
construction

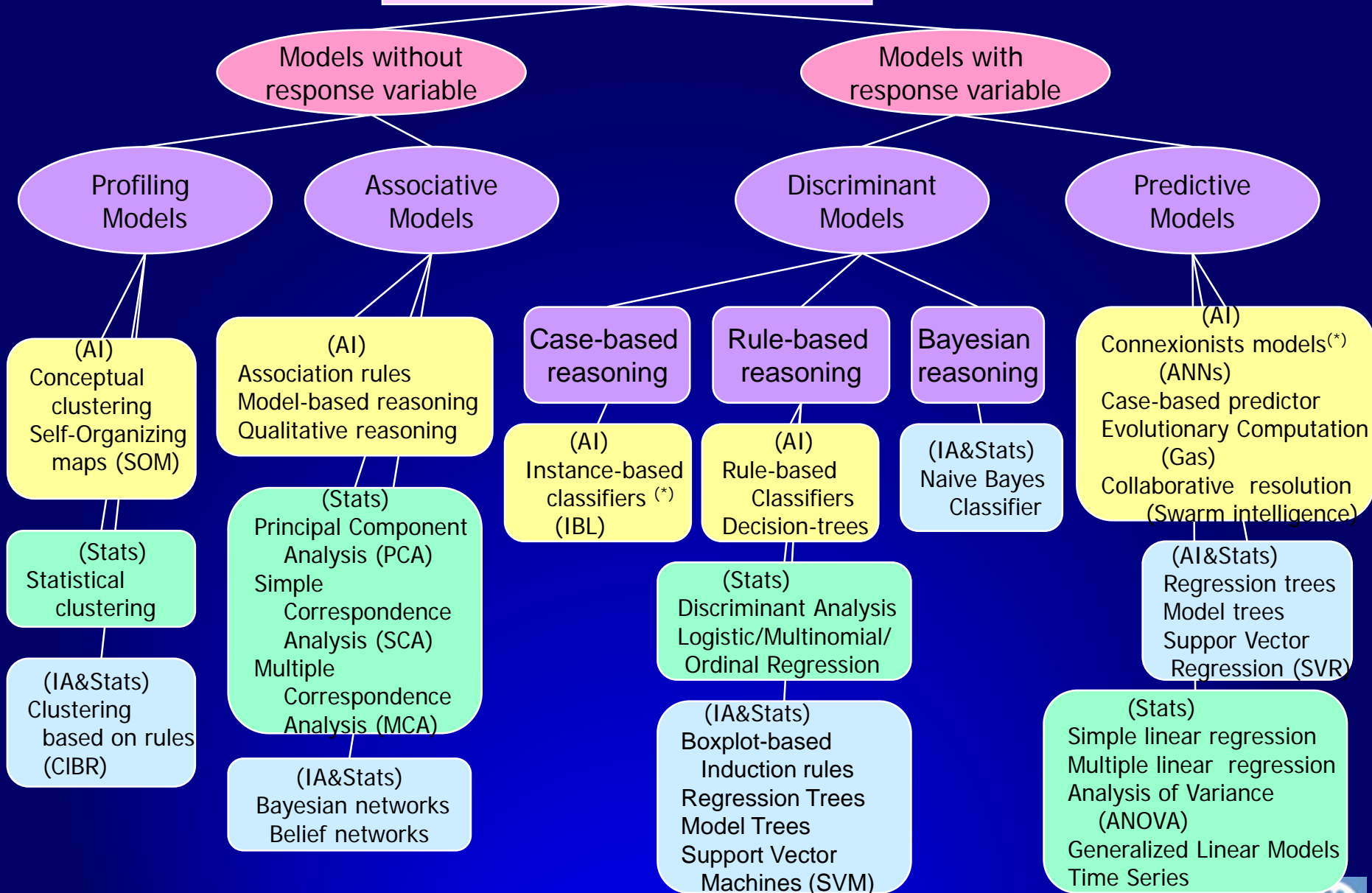
Gibert K, J. Spate, M. Sánchez-Marrè, I. Athanasiadis, J. Comas (2008): Data Mining for Environmental Systems. In Environmental Modeling, Software and Decision Support. State of the art and New Perspectives. IDEA Series v3 (Jackeman, A. J., Voinov, A., Rizzoli, A., and Chen, S. eds), pp 205-228. Elsevier NL.

Gibert K, M. Sánchez-Marrè (2015) Improving ontological knowledge with reinforcement in recommending the data mining method for real problems. In Proceedings of Conferencia de la Asociación Española para la Inteligencia Artificial (CAEPIA) 2015:769-778, Albacete, nov 2015

Knowledge Models

Is there a response variable?

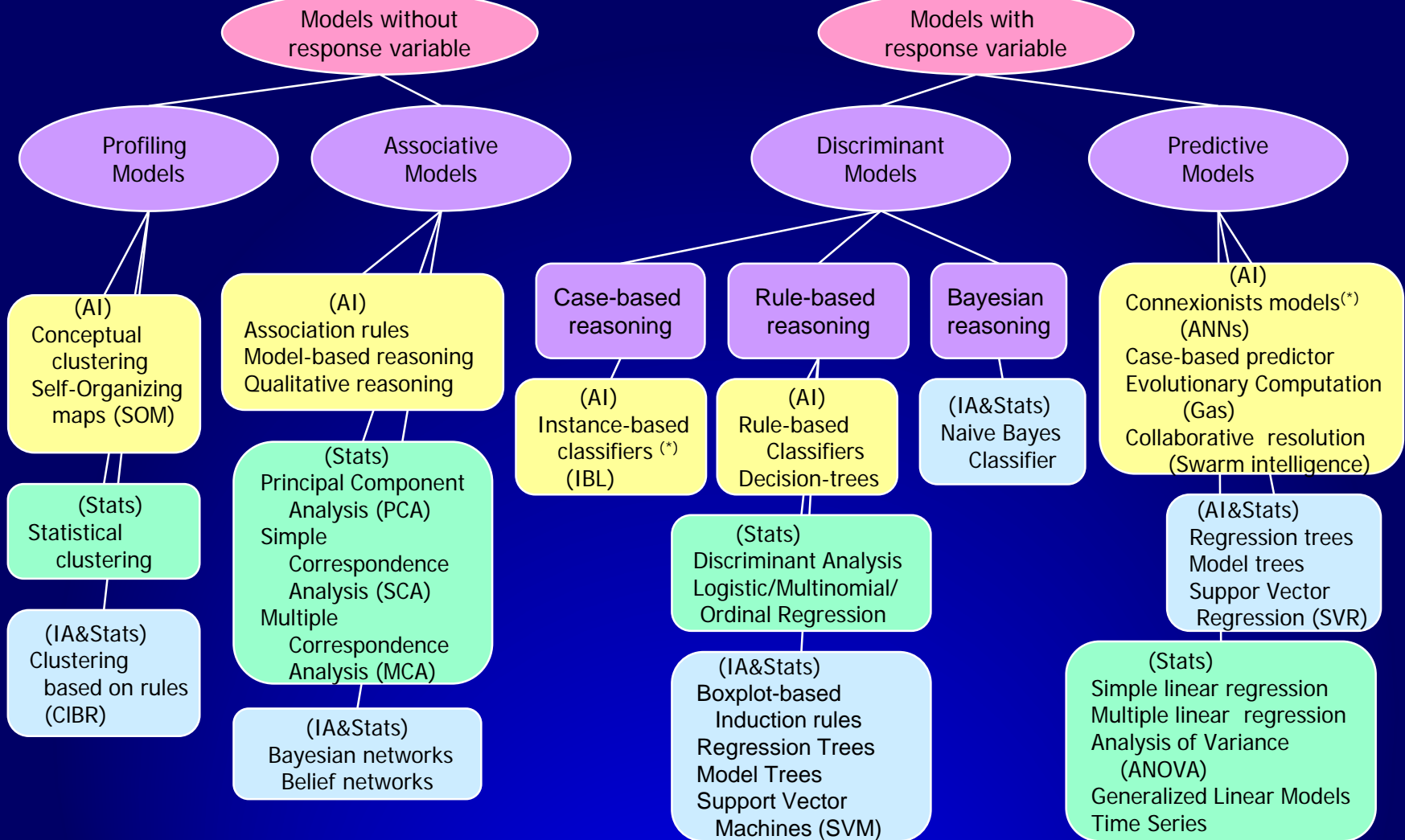
[Gibert et al 2010, iEMSS]
[Gibert et al 2010b, CAEPIA]



Knowledge Models

Is there a response variable?

[Gibert et alt 2010, iEMSS]
[Gibert et alt 2010b, CAEPIA]



Gibert, K, M. Sánchez-Marrè, V. Codina (2010) Choosing the right data mining technique: classification of methods and intelligent recommenders. Proc. of the IEMSS'10, 5th biennial meeting (III DMTES Workshop), S23.03.1-S23.03.9. 2010

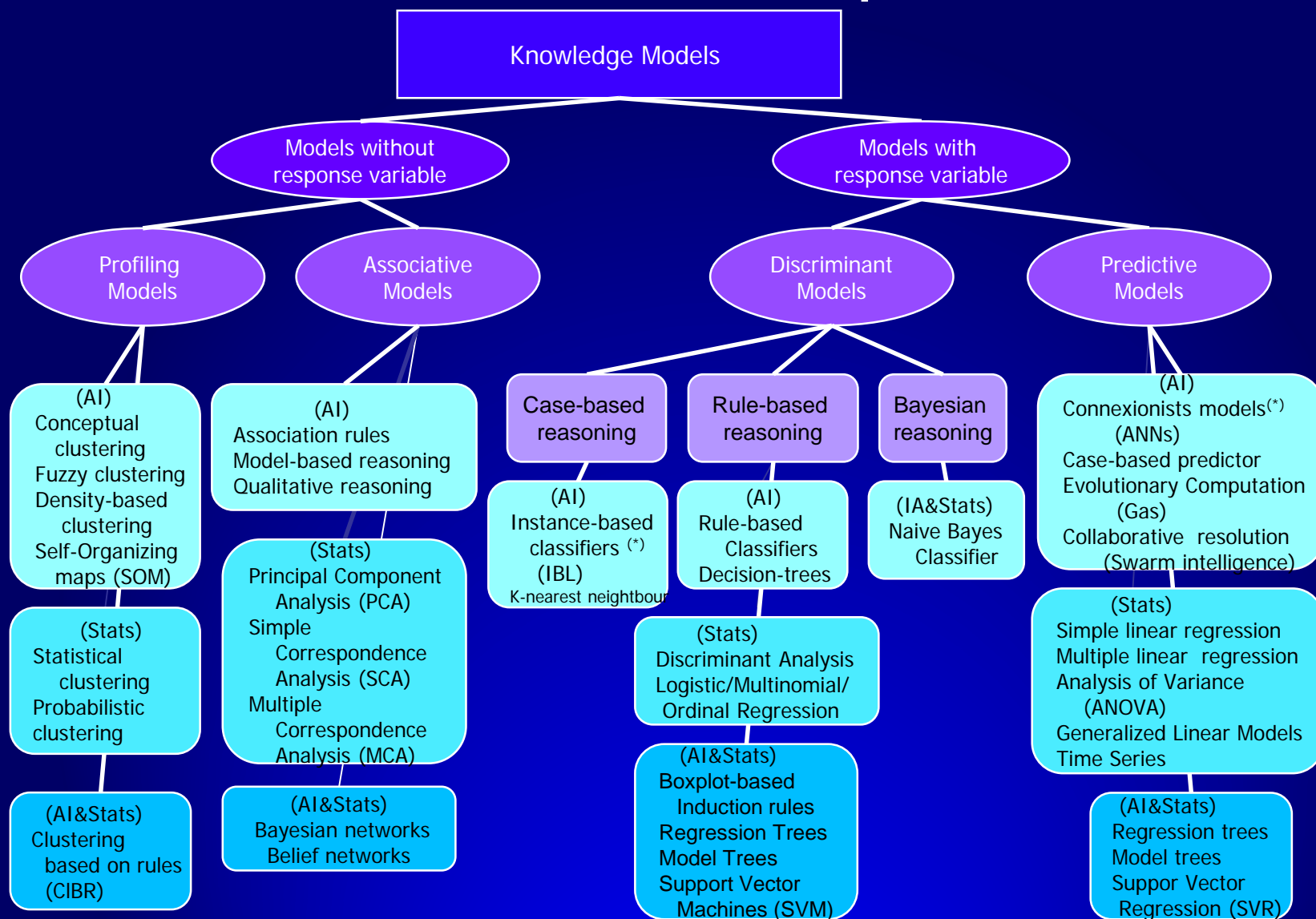
Gibert K, M. Sánchez-Marrè (2010b) Elección de la técnica de minería de datos: Mapa conceptual de técnicas. Actas del V simposio de teoría y aplicaciones de minería de datos: TAMIDA 2010. pp: 37—44. Ibergaceta.2010

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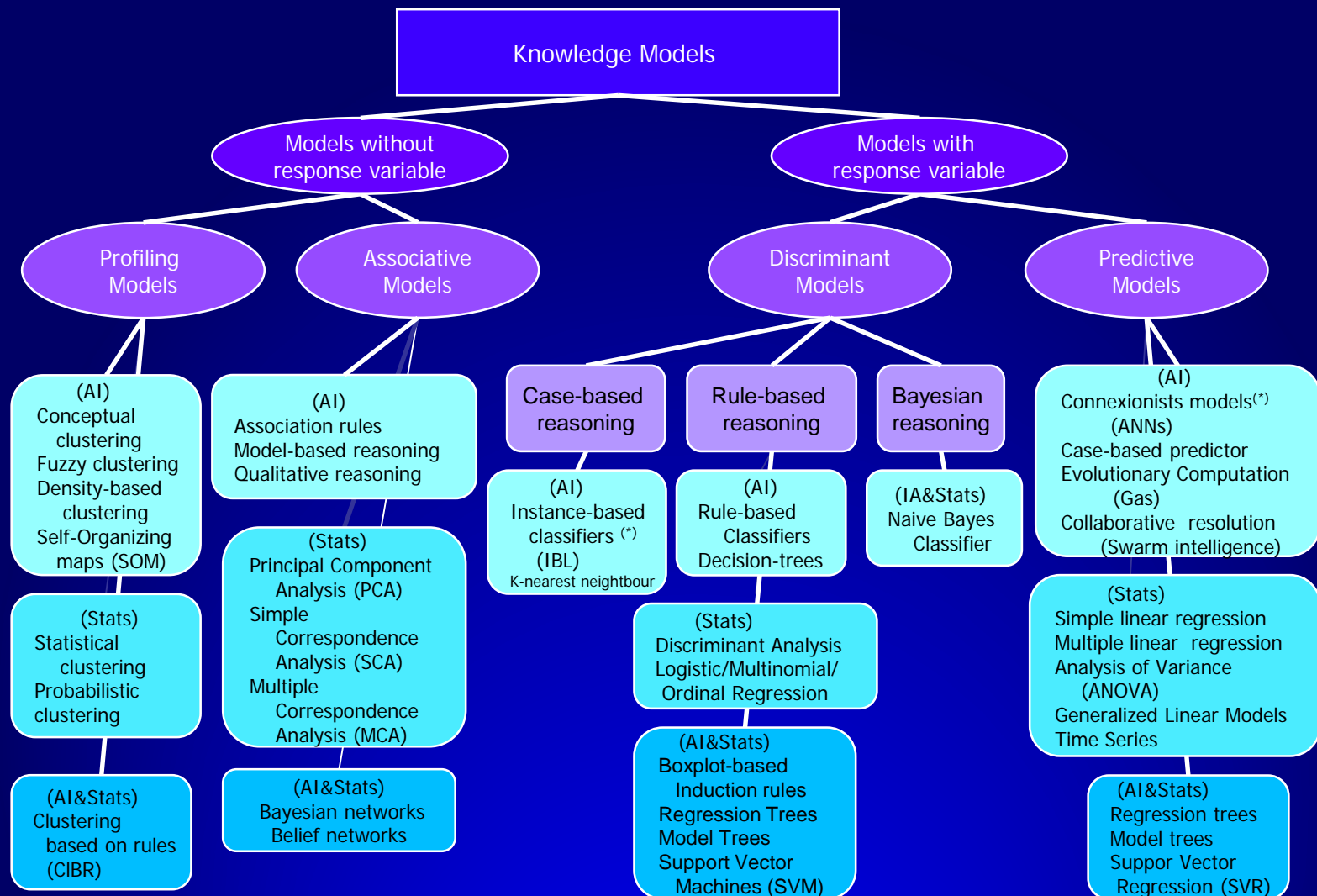
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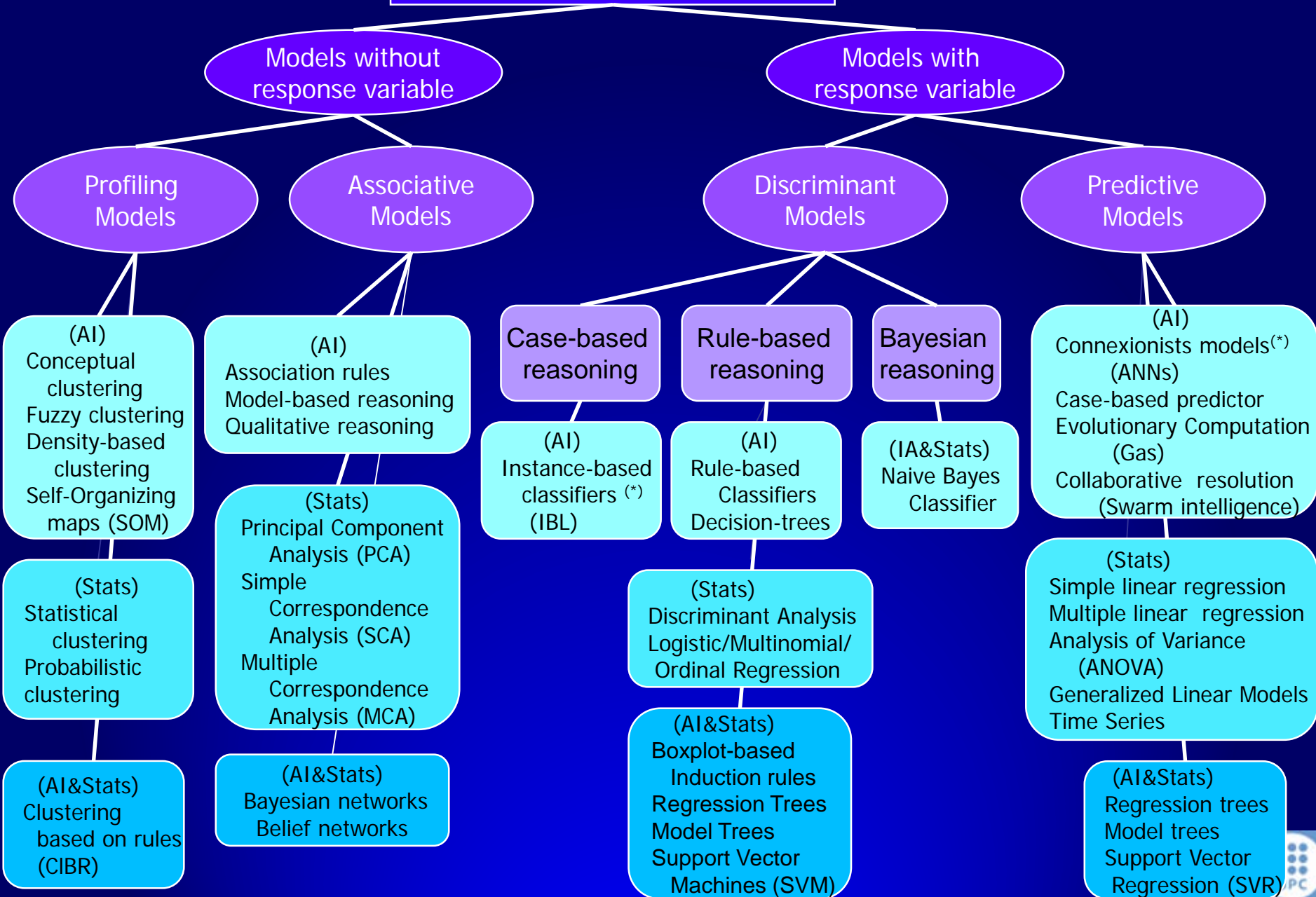
DMMCM map *[Gibert et al 2018]*

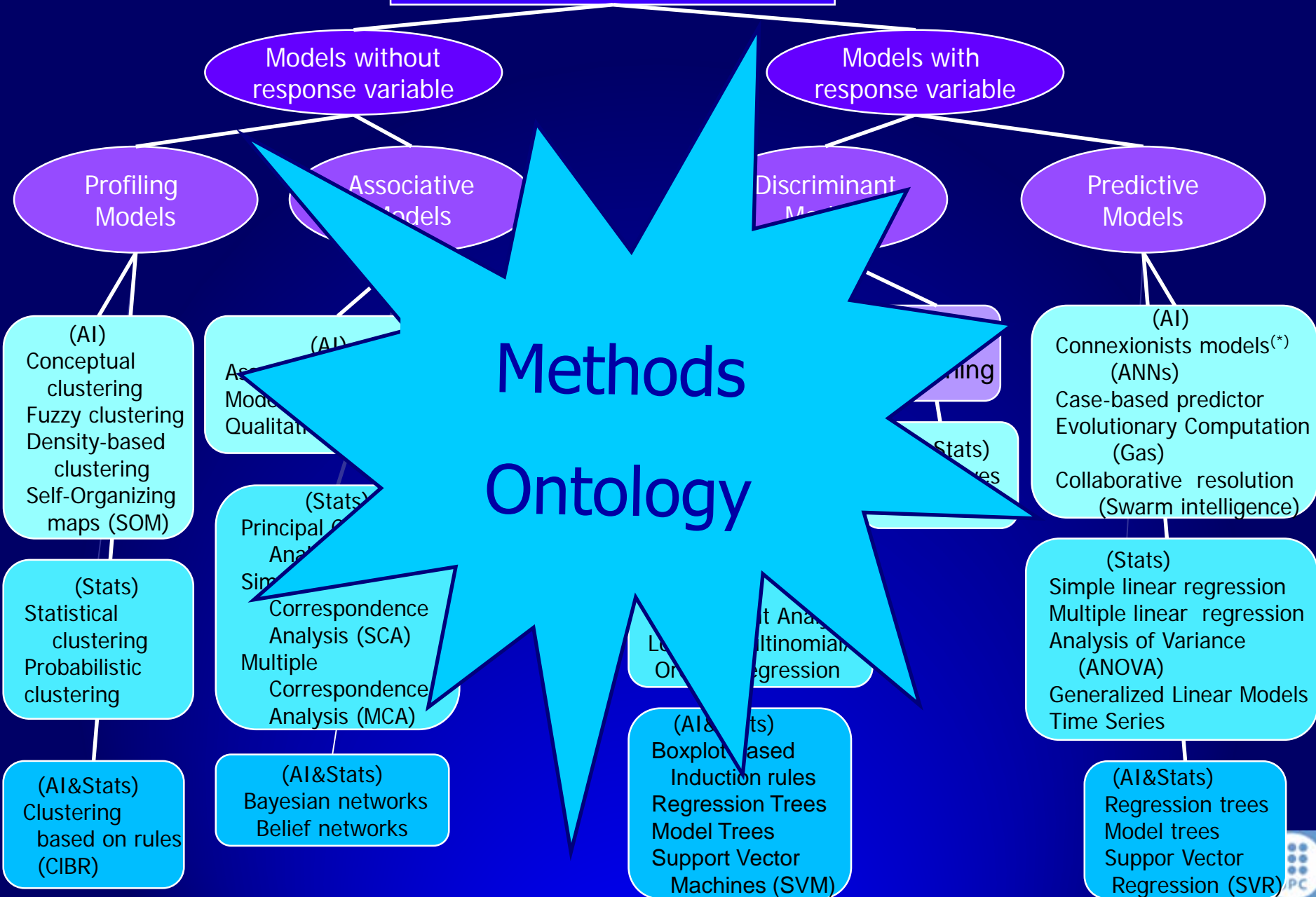


The Data Mining Methods Conceptual Map

DMMCM map *[Gibert et al 2018]*







Role of variables in Data Set (linked with goals)

Profiling Models

(AI)
Conceptual clustering
Fuzzy clustering
Density-based clustering
Self-Organizing maps (SOM)

(Stats)
Statistical clustering
Probabilistic clustering

(AI&Stats)
Clustering based on rules (CIBR)

Associative Models

(AI)
Association rules
Model-based reasoning
Qualitative reasoning

(Stats)
Principal Component Analysis (PCA)
Simple Correspondence Analysis (SCA)
Multiple Correspondence Analysis (MCA)

(AI&Stats)
Bayesian networks
Belief networks

Discriminant Models

Case-based reasoning

(AI)
Instance-based classifiers (*)
(IBL)

Rule-based reasoning

(AI)
Rule-based Classifiers
Decision-trees

Bayesian reasoning

(IA&Stats)
Naive Bayes Classifier

(Stats)
Discriminant Analysis
Logistic/Multinomial/
Ordinal Regression

(AI&Stats)
Boxplot-based
Induction rules
Regression Trees
Model Trees
Support Vector
Machines (SVM)

Predictive Models

(AI)
Connexionists models(*)
(ANNs)
Case-based predictor
Evolutionary Computation
(Gas)
Collaborative resolution
(Swarm intelligence)

(Stats)
Simple linear regression
Multiple linear regression
Analysis of Variance
(ANOVA)
Generalized Linear Models
Time Series

(AI&Stats)
Regression trees
Model trees
Support Vector
Regression (SVR)



DMMCMMap

Knowledge Models

[Gibert et al iEMSs'2018]

Models without
response variable

Models with
response variable

Profiling
Models

Associative
Models

Discriminant
Models

Predictive
Models

Cognition

Re-cognition

Conceptual
clusters
Fuzzy

Case-based
reasoning

Case-based

Statistics models(*)
(s)

(AI)
Instance
classification
(IBL)

Resolution
intelligence)

Statistical
Probabilistic
clustering

(AI&Stats)
Clustering
based on rules
(CIBR)

Bayesian
Analysis (MCA)

(AI&Stats)
Bayesian networks
Belief networks

Ordinal Regression

(AI&Stats)
Boxplot-based
Induction rules
Regression Trees
Model Trees
Support Vector
Machines (SVM)

Linear regression
variance

Generalized Linear Models
Time Series

(AI&Stats)
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DMMCMMap

Knowledge Models

[Gibert et al iEMSs'2018]

Models without
response variable

Models with
response variable

Profiling
Models

Main problem goal

Associative
Models

Discriminative
Models

Nature of response variable

Regressive
Models

(AI)
Conceptual
clustering
Fuzzy clustering
Density-based
clustering
Self-Organizing
maps (SOM)

(Stats)
Statistical
clustering
Probabilistic
clustering

(AI&Stats)
Clustering
based on rules
(CIBR)

(AI)
Association rules
Model-based reasoning
Qualitative reasoning

(Stats)
Principal Component
Analysis (PCA)
Simple
Correspondence
Analysis (SCA)
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Bayesian networks
Belief networks

Case-based
reasoning

(AI)
Instance-based
classifiers (*)
(IBL)

Rule-based
reasoning

(AI)
Rule-based
Classifiers
Decision-trees

(Stats)
Discriminant Analysis
Logistic/Multinomial/
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DMMCMap

Knowledge Models

[Gibert et al iEMSs'2018]

Models without
response variable

Models with
response variable

Profiling
Models

Associative
Models

Discriminant
Models

Predictive
Models

Correlation
(AI)

Association rules
(AI)
Model-based
Qualitative

Case-based
reasoning

Rule-based
reasoning

Bayesian
reasoning

Connectionist models(*)
(AI)
Neural networks
Statistical

Relations
among
objects

Relations
among
variables

Explain a
Qualitative
variable

Explain a
Quantitative
variable

Principles
(AI)

Principal
Analysis
Simple
Correlation
Analysis
Multiple
Correlation
Analysis

(AI)
Instance-based
classification
(IBL)

(Stats)

(AI)
Decision trees
Classification
trees

(AI&Stats)
Clustering
based on rules
(CIBR)

(AI&Stats)
Bayesian networks
Belief networks

Bottom-up
Inductive
Regression
trees
Model Trees
Support Vector
Machines (SVM)

General
Time series
models

(AI&Stats)
Regression trees
Model trees
Support Vector
Regression (SVR)



DMMCMMap

Knowledge Models

[Gibert et al iEMSs'2018]

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response variable

Models with
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Type of output

Combination of
Research area

Data structure

Technical assumptions

Type of model

Type of output

Expected use of model

DMMCMMap

Knowledge Models

[Gibert et al *iEMSs'2018*]

Models without
response variable

Models with
response variable

Profiling
Models

Associative
Models

Discriminant
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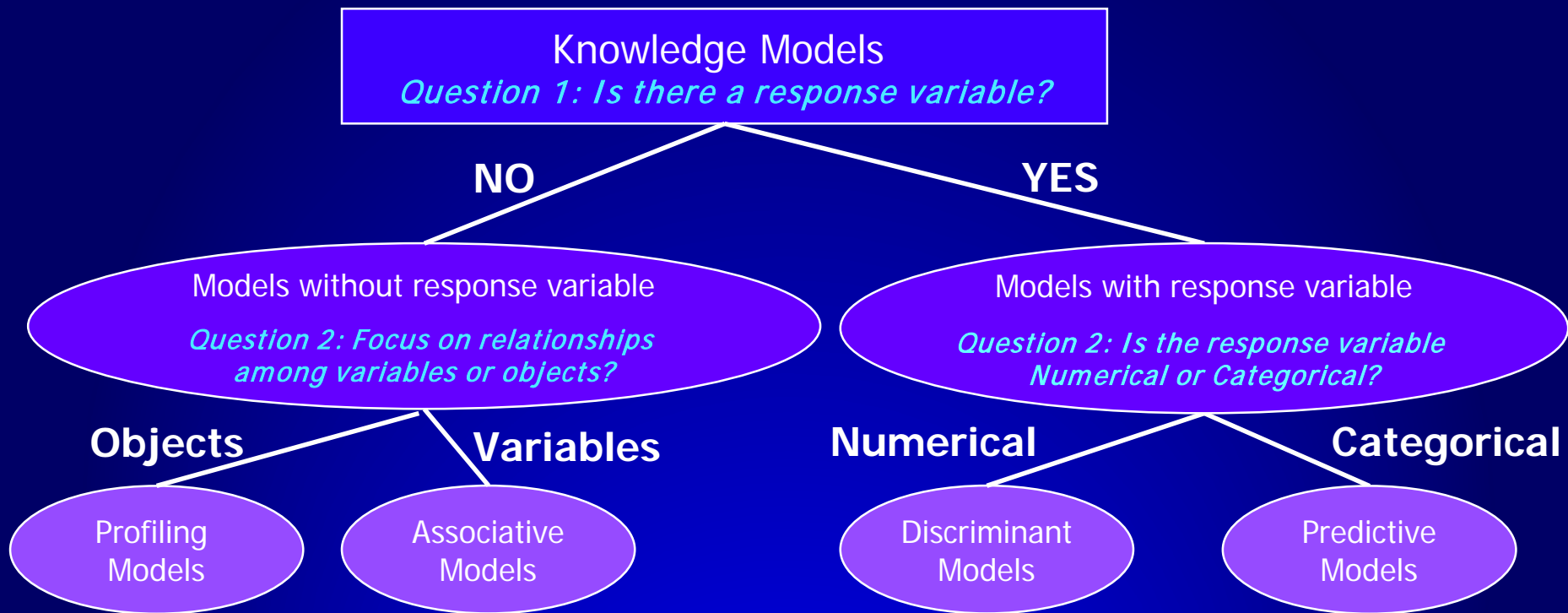
The associated decision process *[iEMSs18]*

Two steps decision process:

1. Identify the **structure** of the target **problem**:
Determine the main branch of the DMMCM
questions associated to nodes in the DMMCM
2. Identify appropriate **technique** within the DMMCM **branch**:
 - i. Find DMMT of selected branch and identify a particular box in DMMCM map

The associated decision process [iEMSs18]

Three questions for Step 1



Select one brach of the DMMCM map

The associated decision process [iEMSs18]

Formal framework for methods

- i) **Technical requirements:** related to dataset structure.
- Type of explanatory/response variable?
Numerical, ordered qualitative, non-ordered qualitative, all

Missmatch:
incorrect use

- ii) **Non restrictive technical properties:** related to data structure.

- Required data size
- Variable independence required
- Normality required
- Outliers non acceptable
- Recommended data size

Missmatch: loose
of performance

- iii) **Non restrictive preference properties:** user preferences/goals

- Is running speed a priority?
- Is interpretability of results a priority?
- Is machine readability required?

Missmatch: loose
of preference

DMMT (Data Mining Method Template)

Formal framework for methods description

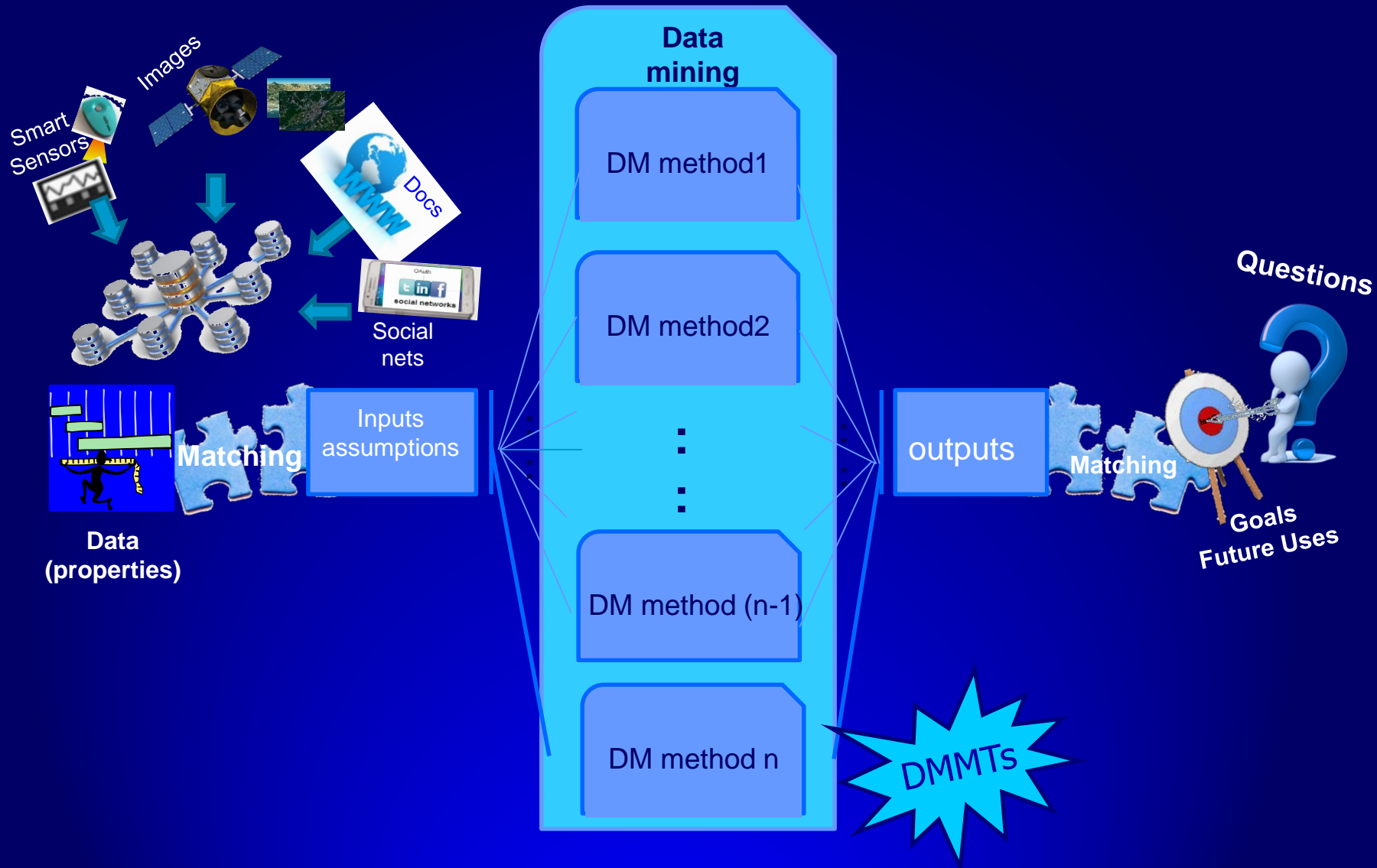
- Main goals of a family of methods
- Brief discussion of the main principles of the family
- Type of input required
- Technical assumptions to be assessed on data
- Requirements
- Type of output expected from the method
- Applications and references

DMMT (Data Mining Method Template)

- Profiling methods
- Associative
- Discriminant
 - CBR
- Predictive
 - ANN
 - Evolutionary Computation

The associated decision process *[iEMSs18]*

Browse on corresponding DMMT in Step 2



Conclusions and future work

- DMMCM: (non-exhaustive) ontology of Data Mining methods
- A 2-steps decision process is proposed to choose a DM method for a real problem
 - Step 1 determines a family of problems with 3 simple questions
 - Step 2 determines a concrete suitable method in the family
- Properties of methods are critical to choose
- A formal framework to describe methods is proposed
 - DMMT
 - Restrictions system
 - Allows dynamic growth of DM methods according to SoTA
- Data structure and model uses are relevant criteria
- Currently an intelligent methodological recommender is built using DMMT and restrictions framework