

The University of Reading

CSMDM16 - Data Analytics and Mining

KDD Development Environments: KNIME

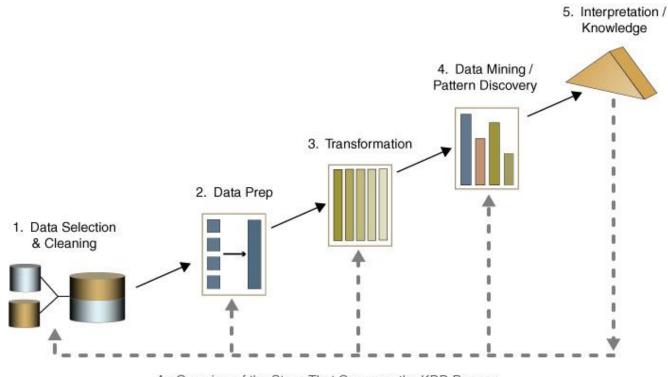
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KDD Development Environments

- Increasing demand for integrated environments to facilitate the KDD process
- Data mining workflow systems that integrates analytical data mining methods for prediction, discovery, classification, etc., with data management and information visualization.



KDD Development Environments – Open Source



Weka 3, Data Mining Software in Java



Orange, a component-based data mining software (C++)



 MLC++ is a library of C++ classes for supervised machine learning

D2K - Data to Knowledge™



D2K, Data to Knowledge (Java)



KNIME, Konstanz Information Miner (Java)

KDD Development Environments - Commercial



RapidMiner (formerly YALE, Yet Another Learning Environment) (Java) – free trial version available







Pentaho – free trial/light version available
 Also free community edition: http://community.pentaho.com/

 (Note: Pentaho Data Mining is based on Weka)



IBM SPSS (Statistical Package for Social Science)



SAS



STATISTICA (Dell acquired StatSoft in March 2014)

2014 KDnuggets Poll

http://www.kdnuggets.com/polls/2014/analytics-data-mining-data-science-software-used.html

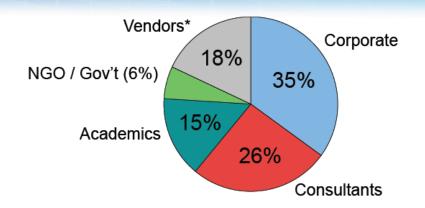
- ☐ The 15th annual KDnuggets Software Poll with over 3,000 voters
- □ Poll: what Analytics, Data Mining, Data Science software/tools you used in the past 12 months for a real project
- > The top 10 tools by share of users were:
 - RapidMiner, 44.2% share (39.2% in 2013)
 - R, 38.5% (37.4% in 2013)
 - Excel, 25.8% (28.0% in 2013)
 - SQL, 25.3% (na in 2013)
 - Python, 19.5% (13.3% in 2013)
 - Weka, 17.0% (14.3% in 2013)
 - KNIME, 15.0% (5.9% in 2013)
 - Hadoop, 12.7% (9.3% in 2013)
 - SAS base, 10.9% (10.7% in 2013)
 - Microsoft SQL Server, 10.5% (7.0% in 2013)

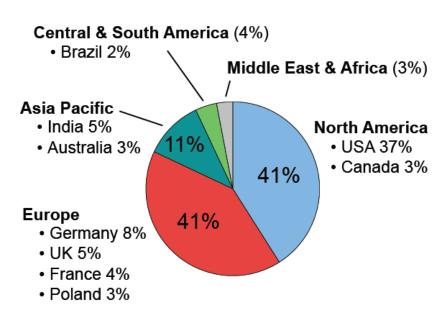
2013 Data Miner Survey: Overview

Vendors are included in this analysis.

#6

- 6th survey since 2007
- 68 questions
- 10,000+ invitations emailed, plus promoted by newsgroups, vendors, and bloggers
- Respondents: 1,259 data miners from 75 countries
- Data collected in first half of 2013





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^{*}Data from software vendors is excluded from analyses in this presentation unless otherwise noted.

Some Key Findings:

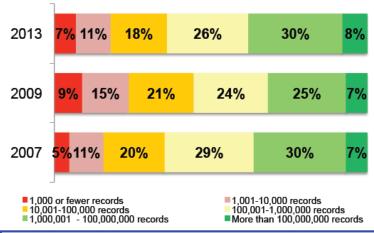
- **BIG DATA**: Many in the field are talking about the phenomena of Big Data. There are clearly some areas in which the volume and sources of data have grown. However it is unclear how much Big Data has impacted the typical data miner. While data miners believe that the size of their datasets have increased over the past year, data from previous surveys indicate that the size of datasets have been fairly consistent over time.
- THE ASCENDANCE OF R: The proportion of data miners using R is rapidly growing, and since 2010, R has been the most-used data mining tool. While R is frequently used along with other tools, an increasing number of data miners also select R as their primary tool.
- **ENGAGEMENT & JOB SATISFACTION**: The Data Miners in our survey are highly engaged with the analytic community: consuming and producing content, entering competitions and searching for education and growth within their jobs. All of these activities lead to high job satisfaction, which has been increasing over time.
- ANALYTIC SOFTWARE: Data miners are a diverse group who are looking for different things from their data mining tools. Ease-of-use and cost are two distinguishing dimensions. Software packages vary in their strengths and features. STATISTICA, KNIME, SAS JMP and IBM SPSS Modeler all receive high satisfaction ratings.

Big Data: Hype or Reality?

There is a lot of talk in the business and technical press about Big Data. Clearly some businesses and scientific areas are working with very large data sets. However, it is unclear how much Big Data has impacted the typical data miner.

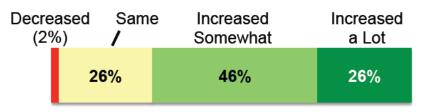
In 2013, the general perception among data miners is that data volumes have increased (72% say it has). However, the datasets they report using are of similar size to what was reported in 2007. Additionally, only 13% report that their company has an active big data program.

Typical Data Set Size



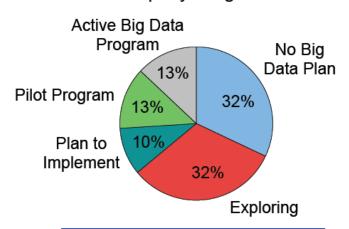
Question: What size data sets did you typically data mine in the past year?

2013: Perception of Data Size Increase



Question: Has the volume/size of data that you use in your analyses increased in the last two years?

2013: Your Company's Big Data Plan

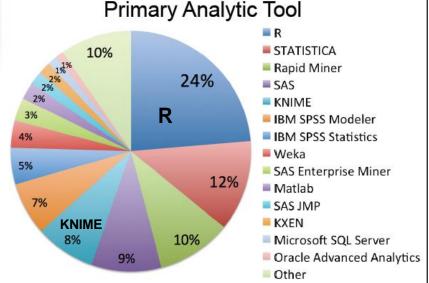


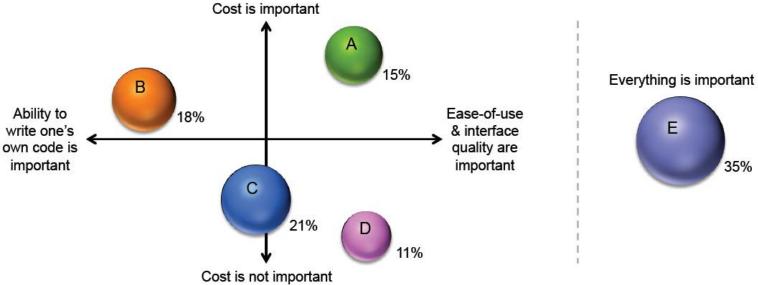
Question: What is your company / organization doing with regards to Big Data?

Tool Selection

Data miners are a diverse group who are looking for different things from their data mining tools. They report using multiple tools to meet their analytic needs, and even the most popular tool is identified as their primary tool by just 24% of data miners. Over the years, R and Rapid Miner have shown substantial increases.

Cluster analysis* reveals that, in their tool-selection preferences, data miners fall into 5 groups. The primary dimensions that distinguish them are price sensitivity and code-writing / interface / ease-of-use preferences.





^{*}Cluster analysis was conducted on data miners' ratings of the importance of 22 tool selection factors.

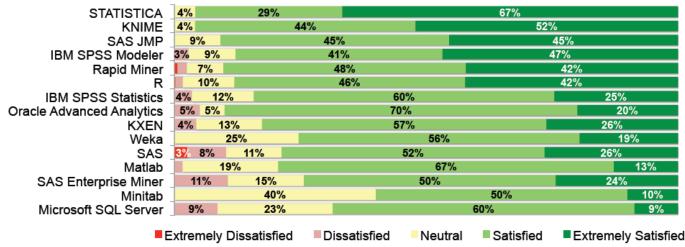
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Tool Satisfaction

Most data miners are happy with their analytic software. STATISTICA and KNIME have particularly high satisfaction ratings (they also had the highest ratings in the 2011 survey). SAS JMP, IBM SPSS Modeler, Rapid Miner and R also have high ratings. While people are more satisfied with their primary tools, the patterns of primary and secondary tool satisfaction are generally similar. However, people choosing IBM SPSS Statistics as their secondary tool give it high ratings, while people using SAS Enterprise Miner and IBM SPSS Modeler as their secondary tools give these tools lower ratings.

Most people also report that they will continue using their primary tools – the highest continuation rate is among people choosing KNIME as their primary tool: 85% report that they are "extremely likely" to continue using it as their primary tool for the next 3 years. R and STATISTICA users also report especially high continuation plans. Across all tools, when people say they are likely to switch primary tools, many are choosing R (see page 16).





Satisfaction question: Please rate your overall satisfaction with [insert name of previously identified software package].

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Tool Satisfaction: Details

Overall, data miners express the most satisfaction with the quality and accuracy of their tools' model performance and with the variety of algorithms their tools make available to them. Data miners are least satisfied with their tools' help functions, their graphical visualization of models, and their ability to handle large data sets. STATISTICA received strong ratings across many dimensions.

	Overall	IBM SPSS Statistics	IBM SPSS Modeler	KNIME	R	Rapid Miner	SAS	SAS Enterprise Miner	STATISTICA	Weka
Quality and accuracy of model performance	4.28	3.96	4.15	4.30	4.39	4.25	4.20	4.48	4.62	4.16
Variety of available algorithms	4.27	3.66	4.05	4.36	4.74	4.55	3.91	4.23	4.59	4.46
Data manipulation capabilities	4.19	3.91	4.36	4.54	4.24	4.07	4.50	3.74	4.52	3.48
Dependability/Stability of software	4.19	4.02	3.96	4.27	4.24	4.07	4.28	4.16	4.51	4.03
Ability to automate repetitive tasks	4.18	3.79	3.76	4.42	4.35	4.18	4.26	4.10	4.44	3.76
Quality of output / Ease of interpretation	4.11	3.87	3.89	4.17	4.10	4.18	3.84	4.10	4.59	3.82
Ease of use	4.11	4.10	4.67	4.58	3.59	4.39	3.77	4.27	4.58	4.03
Good metrics of model quality	4.08	3.72	3.89	3.91	4.19	4.17	4.01	4.17	4.50	4.06
Data quality assessment & data preparation capabilities	4.05	3.72	4.27	4.37	4.02	4.00	4.26	3.77	4.41	3.47
Ability to easily incorporate data at different levels of granularity (e.g. transaction data and customer data)	4.03	3.87	4.25	4.21	3.94	4.04	4.14	4.10	4.30	3.59
Cost of software	4.03	3.02	2.89	4.85	4.93	4.86	2.33	2.70	3.91	4.89
Ability to modify algorithm options to fine-tune analyses	4.01	3.26	3.63	3.80	4.35	4.10	3.91	3.94	4.28	4.18
Good variable discovery, profiling and selection	4.00	3.64	4.16	4.07	4.03	4.06	3.78	4.23	4.42	3.77
Quality of user interface	3.97	4.02	4.47	4.54	3.49	4.37	3.66	4.10	4.53	3.54
Ease of model deployment (scoring to other data sets)	3.97	3.42	4.01	4.21	3.87	4.19	3.92	4.00	4.43	3.75
Speed	3.95	3.62	4.01	4.00	3.69	3.95	3.93	3.97	4.54	3.70
Enables mining within one's database	3.92	3.59	4.18	4.08	3.92	3.83	3.78	3.93	4.26	3.69
Ability to handle very large data sets	3.84	3.65	4.19	3.90	3.27	3.59	4.35	4.30	4.56	3.18
Strong graphical visualization of models	3.83	2.94	3.60	3.90	4.14	4.01	3.09	3.77	4.58	3.38
Useful help menu, demos and tutorials	3.82	3.87	3.82	4.05	3.86	3.54	3.67	3.90	4.23	3.50

Mean satisfaction rating on 1-5 scale

Higher Satisfaction

Lower Satisfaction

Question: Rate how satisfied you are with the performance of your primary data mining package (identified earlier) on each of these factors.

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KNIME

KNIME Konstanz Information Miner



Copyright, 2003 - 2006
Konstanz University, Germany
Chair for Bioinformatics and Information Mining
Prof. Dr. Michael B. Berthold



- Developed at the Department of Computer and Information Science, University of Konstanz, Germany
- Under continuous evolution and extension
 - 1st release in April 2006
 - ver.3 released in Dec. 2015
 - current version: 3.2.1

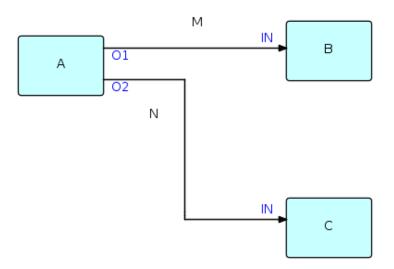


First release and first publication in 2006:

M. Berthold, N. Cebron, F. Dill, G. Di Fatta, T. Gabriel, F. Georg, T. Meinl, P. Ohl, C. Sieb, B. Wiswedel, "KNIME: the Konstanz Information Miner", Proc. of Workshop on Multi-Agent Systems and Simulation (MAS&S), 4th Annual Industrial Simulation Conference (ISC), Palermo, Italy, June 5-7, 2006, pp.58-61.

Flow-based Programming

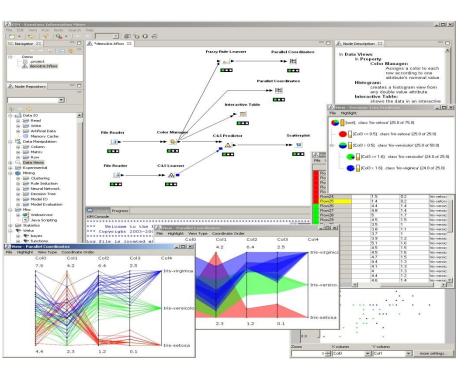
Flow-based Programming (FBP) is a programming paradigm that defines applications as networks of "black box" processes, which exchange data across predefined connections by message passing, where the connections are specified externally to the processes. These black box processes can be reconnected endlessly to form different applications without having to be changed internally. FBP is thus naturally component-oriented.



J. Paul Morrison, Flow-Based Programming, 2nd Edition: A New Approach to Application Development, CreateSpace, 2010

KNIME

Knime: Interactive Data Exploration



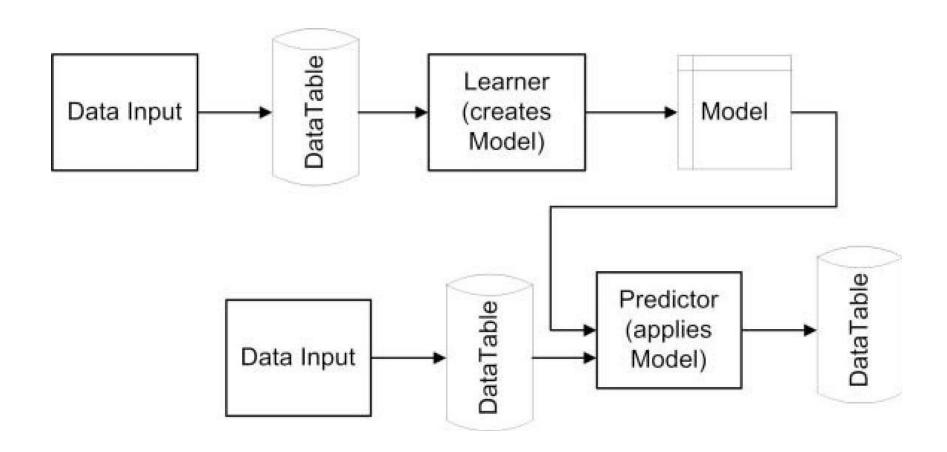
Features:

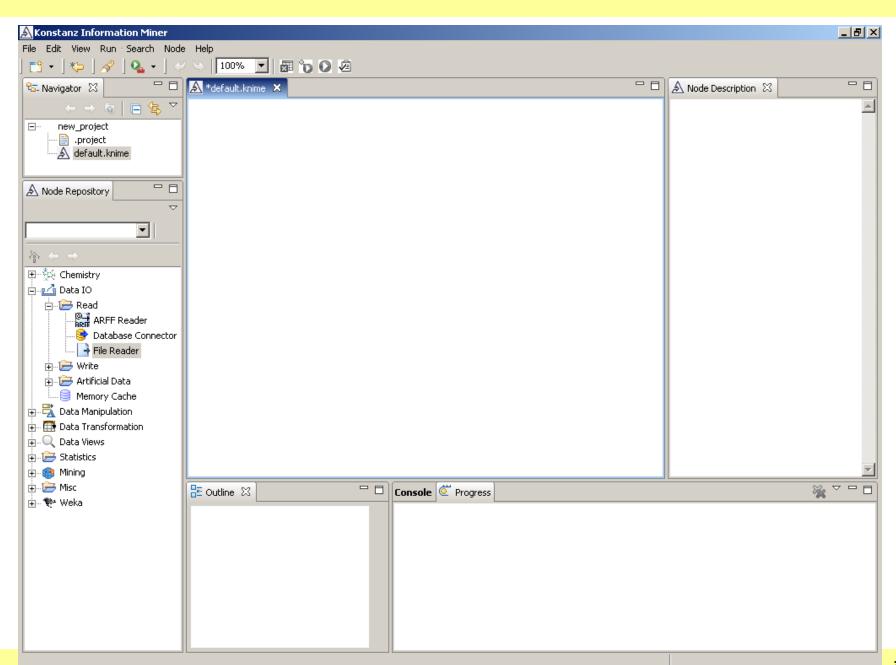
- Modular Data Pipeline Environment
- Large collection of Data Mining techniques
- Data and Model Visualizations
- Interactive Views on Data and Models
- Java Code Base as Open Source Project
- Seamless Integration: R Library, Weka, etc.
- Based on the Eclipse Plug-in technology

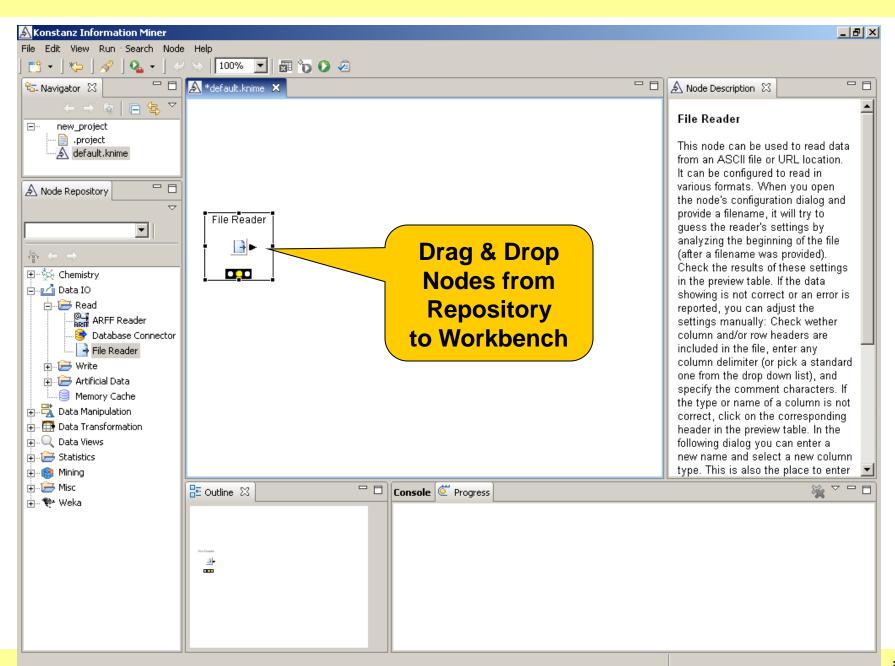
Easy extendibility

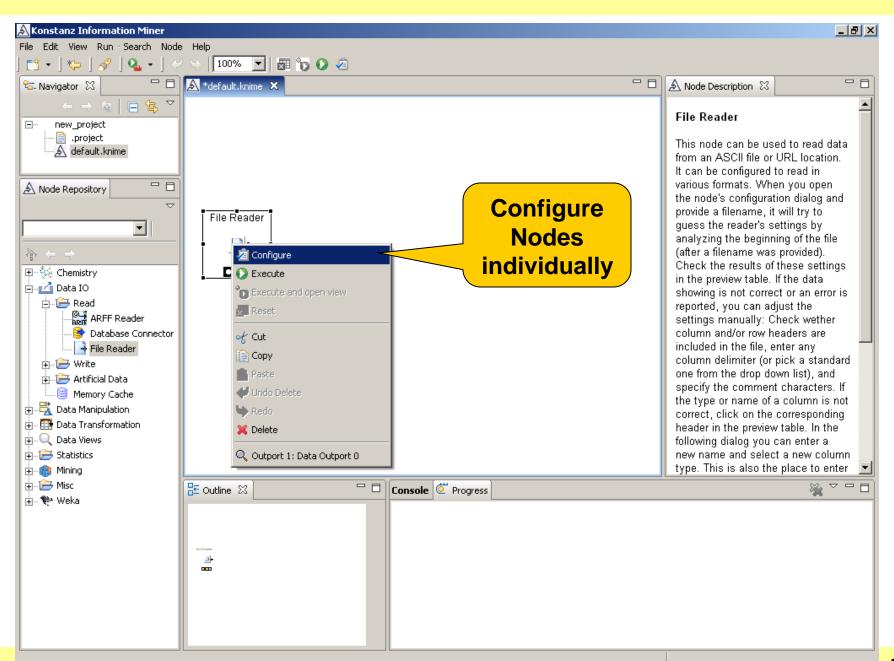
New nodes via open API and integrated wizard

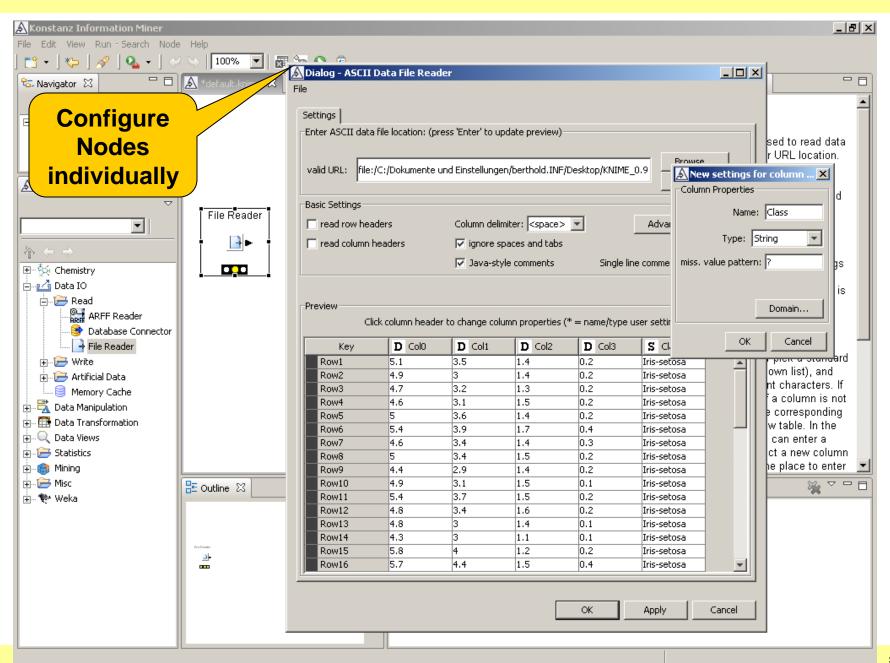
Data Pipeline



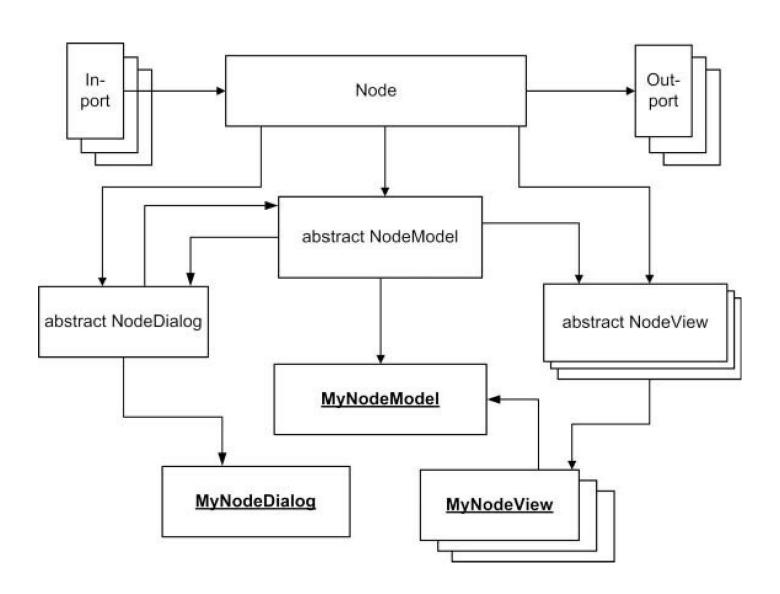


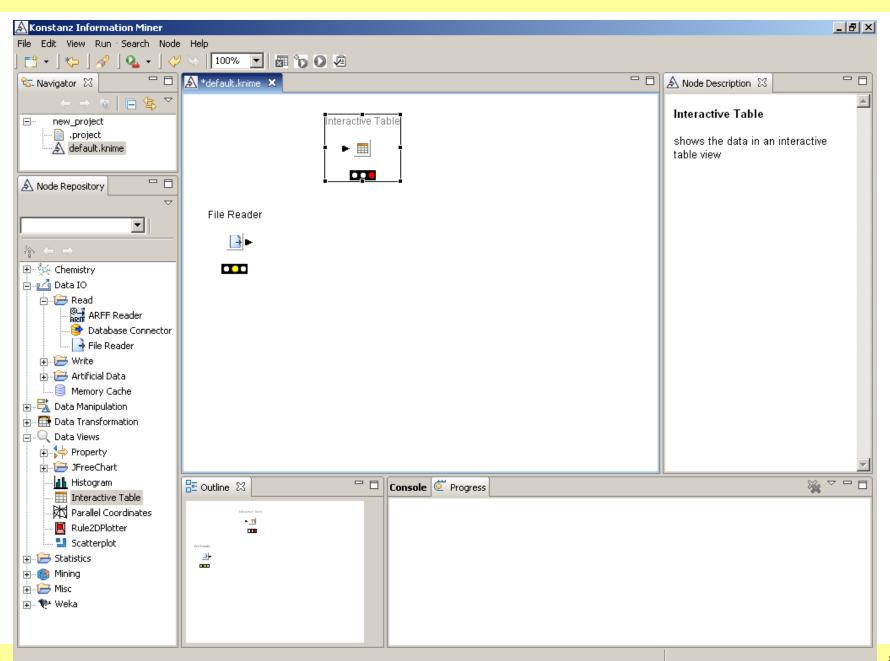


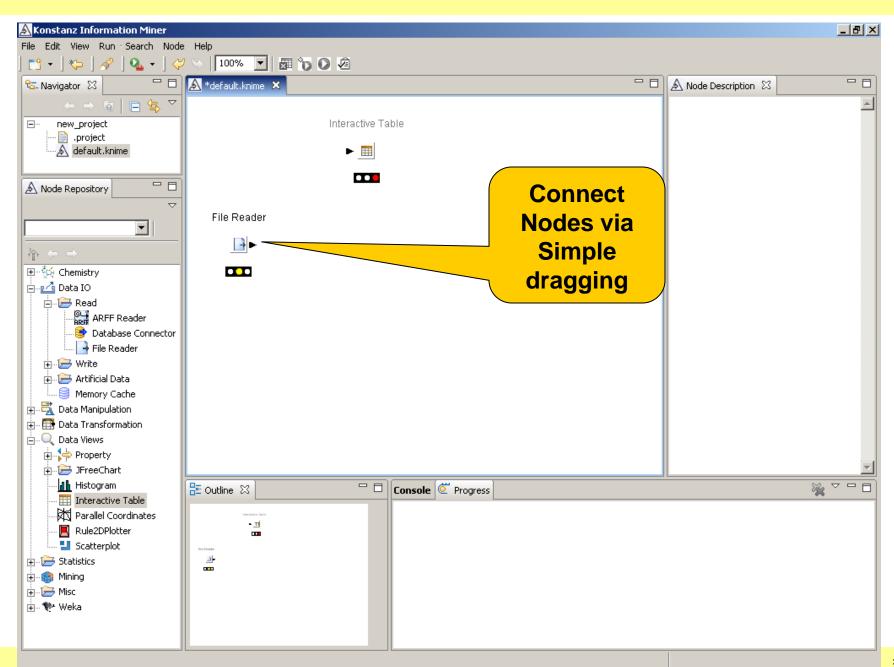


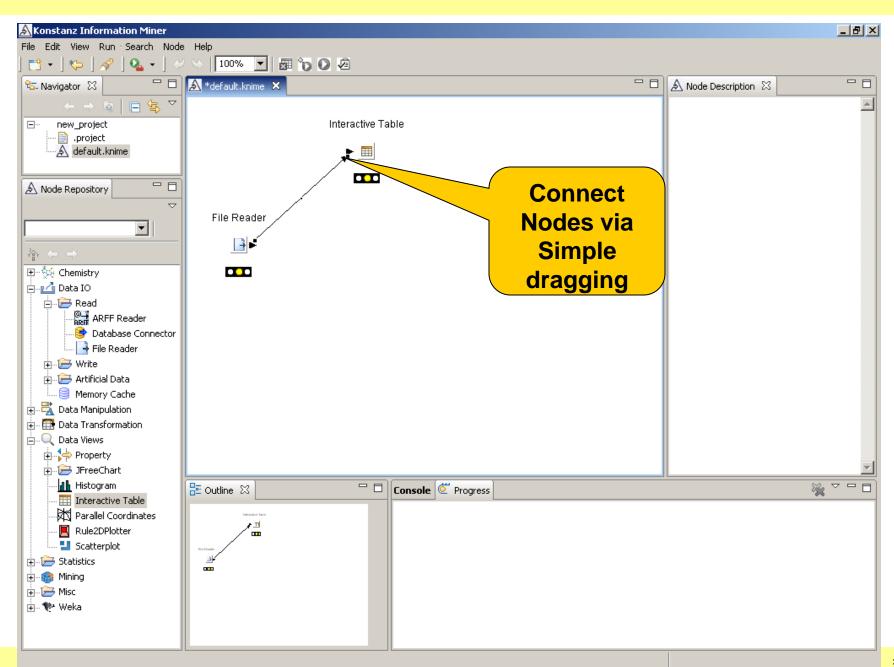


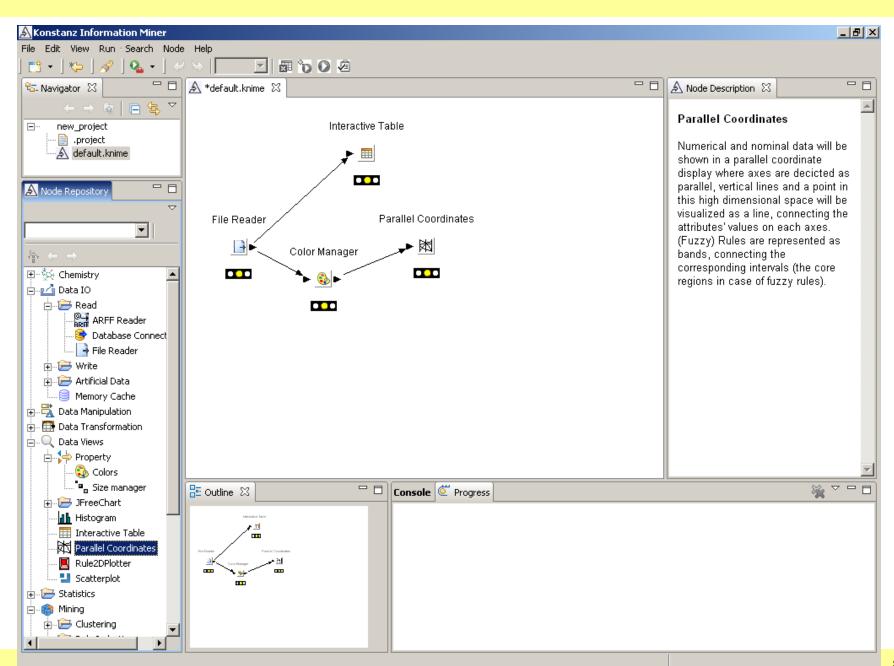
Node Model

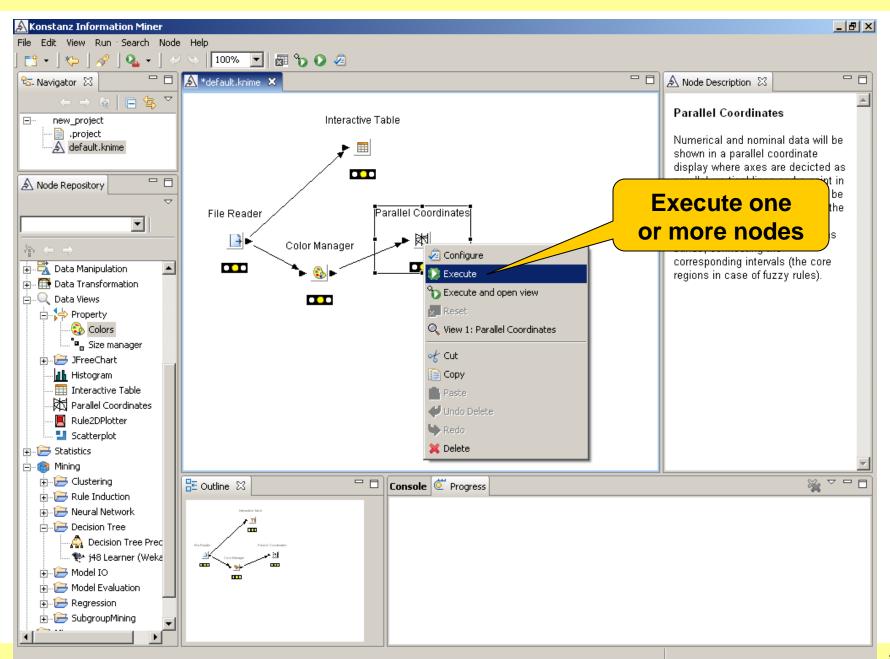


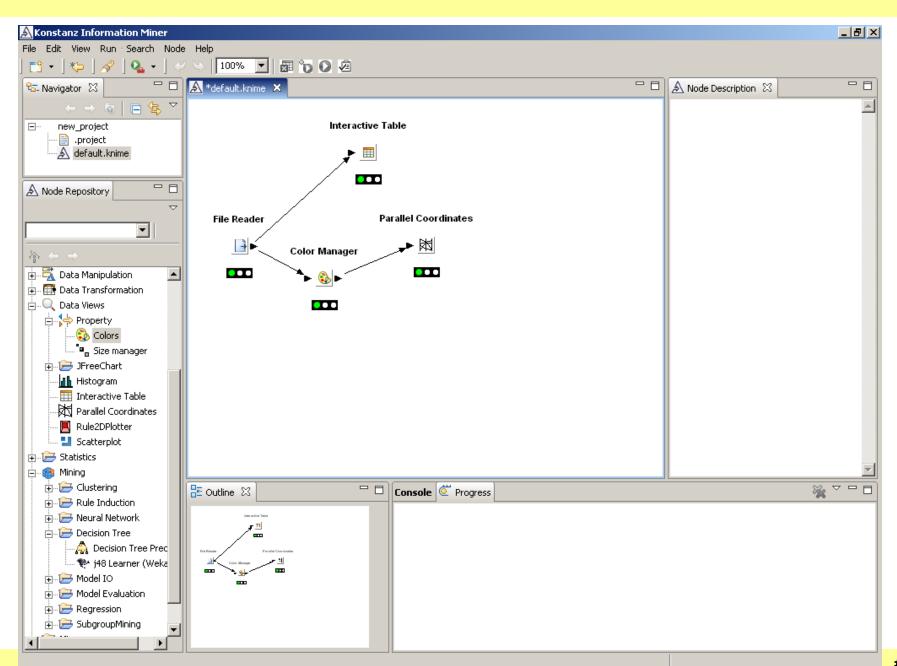


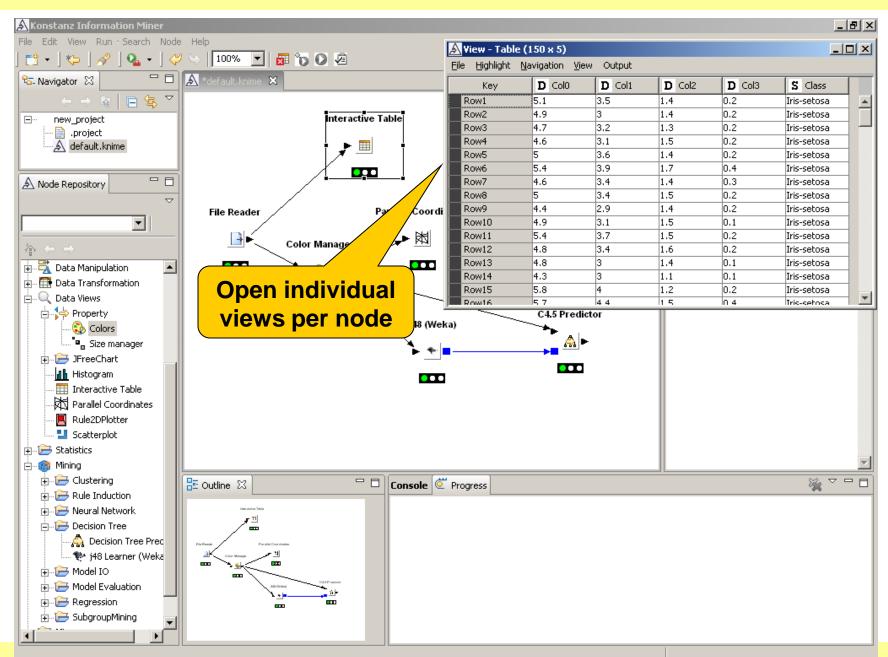


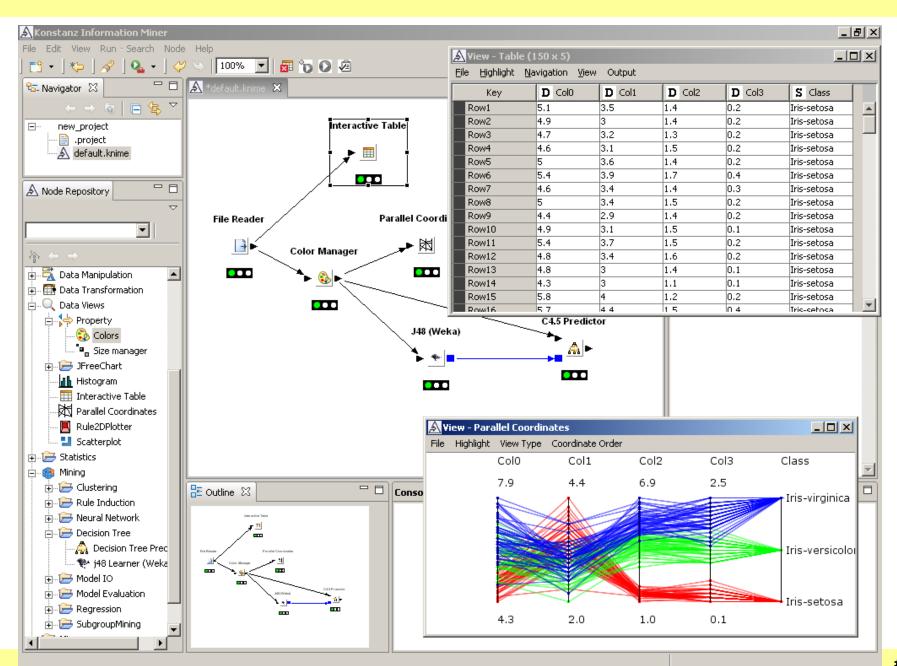


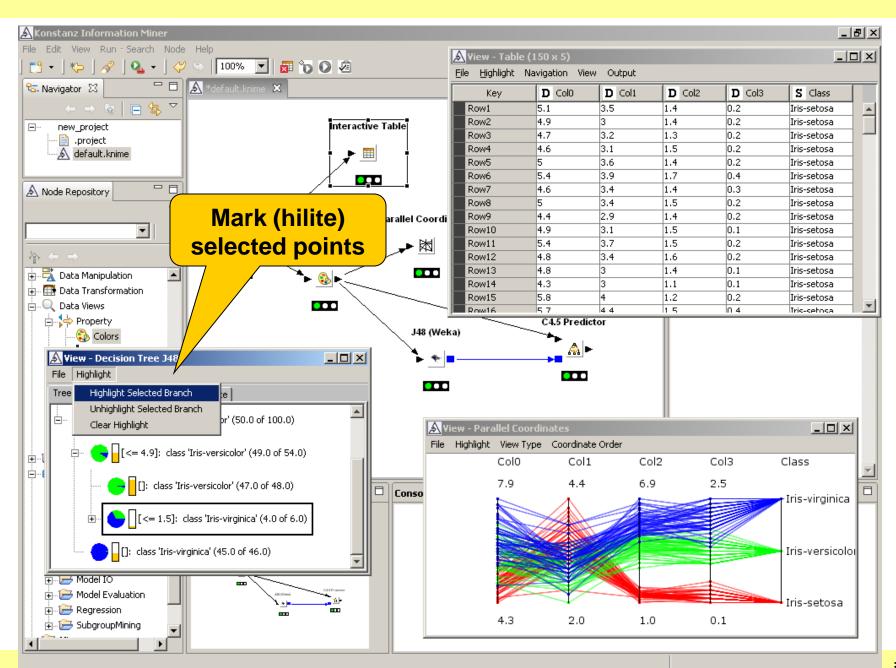


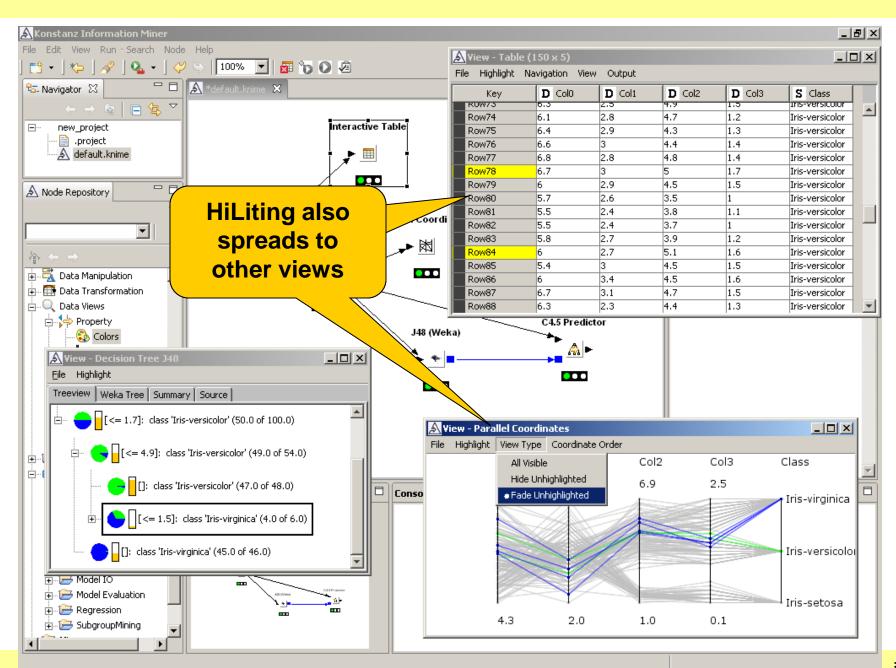


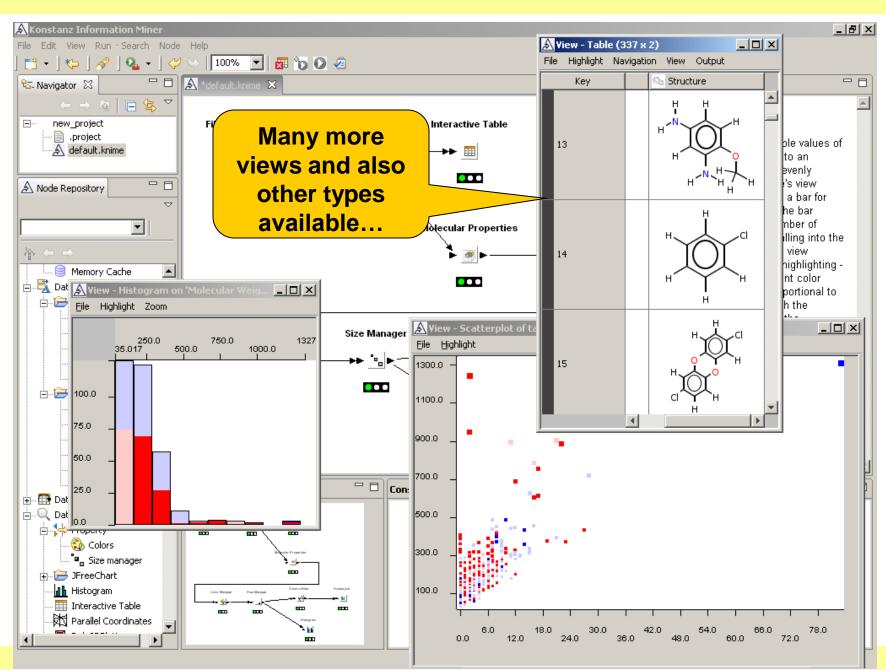








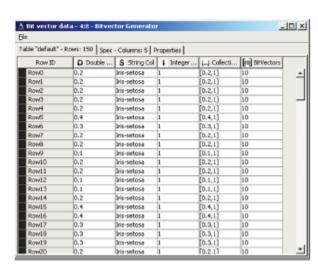




Data Table

- Contains meta information (spec)
 - data types
 - domains
 - # of rows/cols
- Large tables are buffered on disc
- Blob cell support for large data cells e.g. images





Data Types

- Common data types
 - Double Value



- Int Value



String Value



- Collections



- · Sets
- Lists
- Bit vectors



- Additional data types
 - Terms and Documents
 - Image
 - Network
 - Chemical types
 - Molecules i.e. CDK, Smiles, SDF, ...
 - Distance Matrix
 - Custom data types

KNIME Features

- Node types
 - I/O
 - Data manipulation
 - Learners
 - Predictors
 - Views
- Highlighting
- Metanodes
- Quickforms
- Loops and flow variables
- Error handling: "try-catch" nodes
- Extensions (KNIME plugins)

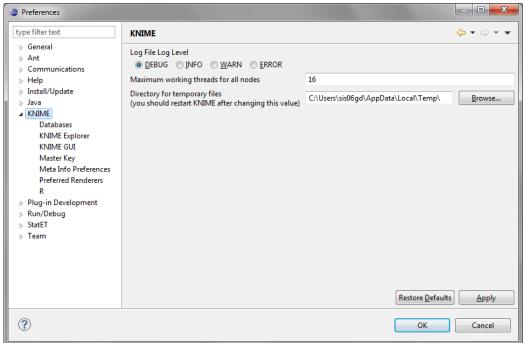
Perspective and Preferences

 In the SDK version (Eclipse), you need to select the KNIME perspective. Communications
CVS Repository Exploring
Debug
Java (default)
Java Browsing
Java Type Hierarchy
KIMIE
Plug-in Development
Resource
StatET
Team Synchronizing
OK Cancel

Open Perspective

- - X

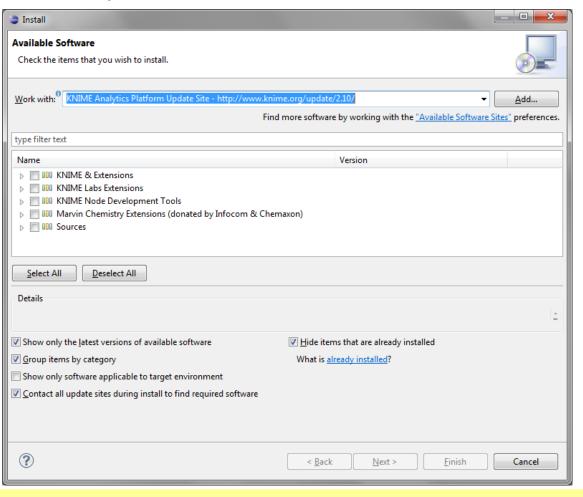
In Menu Windows, select Preferences



KNIME Updates and Extensions

KNIME can be updated and extended by means of the Eclipse "Install New Software" mechanism.

select the KNIME update site



KNIME Extensions

- Experimental Extensions: http://tech.knime.org/knime-labs
 - Modular Data Generators
 - Network Mining
 - Perl Scripting
 - Text Processing
 - etc.
- Community Contributions: http://tech.knime.org/community
 - Chemoinformatics
 - High Content Screening
 - Image Processing
 - Next Generation Sequencing
 - R/Groovy/Matlab/Python Scripting
 - STARK
 - etc.
- KNIME is designed to be extended!
 - You just need to use the SDK version.

KNIME Extensions (Plugins)

Some available plugins include:

- Chemistry types and features
- Distance Matrix
- Ensemble Learning
- HTML/PDF Writer
- Item Set Mining
- R Statistics Integration
- Report Designer
- Webservice Client
- Weka Data Mining Integration
- XLS Support
- XML Processing



Conclusions on KNIME

- Modularity and extendibility
 - General and extendible data structure (DataTable and DataCell)
 - Nodes encapsulate computational processing tasks (algorithms)
- A workflow management system
 - directed edges connects nodes to create data pipelines
 - a workflow is, in general, a directed acyclic graph
 - multi-threading
 - Meta-nodes (nested workflows)
- New releases
 - Enhanced GUI and performance
 - Include more and more modules and features

KNIME Useful Resources

KNIME User → desktop version: "KNIME Analytics Platform"

- http://www.knime.org/knime
- <u>http://www.knime.org/downloads/datasets</u>
- http://www.knime.org/introduction/examples

Workflow examples:

- parallel coordinates on Iris data
- PCA on Iris data
- decision tree on Iris data

KNIME Developer → SDK version (Eclipse):

- <u>http://tech.knime.org/developer-guide</u>
- http://tech.knime.org/developer/example
- API: for example see the DataTable interface in
 http://tech.knime.org/docs/api/org/knime/core/data/package-summary.html
- Simple exercise: create a new KNIME node to compute column stats

Rosaria's blog with lots of examples:

<u>http://www.dataminingreporting.com/</u>