



The Genetic and Evolutionary Computation Conference

Algorithm and Experiment Design with HeuristicLab

An Open Source Optimization Environment for Research and Education

Stefan Wagner & Gabriel Kronberger

Heuristic and Evolutionary Algorithms Laboratory (HEAL)
School of Informatics, Communications and Media
Upper Austria University of Applied Sciences
Hagenberg, Austria

support@heuristiclab.com
<http://dev.heuristiclab.com>

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1



Instructor Biographies

❖ Stefan Wagner

- MSc in computer science (2004)
Johannes Kepler University Linz, Austria
- PhD in technical sciences (2009)
Johannes Kepler University Linz, Austria
- Associate professor (2005 – 2009)
Upper Austria University of Applied Sciences
- Full professor for complex software systems (since 2009)
Upper Austria University of Applied Sciences
- Co-founder of the HEAL research group
- Project manager and chief architect of HeuristicLab
- <http://heal.heuristiclab.com/team/wagner>



❖ Gabriel Kronberger

- MSc in computer science (2005)
Johannes Kepler University Linz, Austria
- PhD in technical sciences (2010)
Johannes Kepler University Linz, Austria
- Research assistant (since 2005)
Upper Austria University of Applied Sciences
- Member of the HEAL research group
- Architect of HeuristicLab
- <http://heal.heuristiclab.com/team/kronberger>



2



Agenda

- ❖ Objectives of the Tutorial
- ❖ Introduction
- ❖ Where to get HeuristicLab?
- ❖ Plugin Infrastructure
- ❖ Graphical User Interface
- ❖ Available Algorithms & Problems
- ❖ Demonstration
- ❖ Some Additional Features
- ❖ Planned Features
- ❖ Team
- ❖ Suggested Readings
- ❖ Bibliography
- ❖ Questions & Answers



3



Objectives of the Tutorial

- ❖ Introduce general motivation and design principles of HeuristicLab
- ❖ Show where to get HeuristicLab
- ❖ Explain basic GUI usability concepts
- ❖ Demonstrate basic features
- ❖ Demonstrate editing and analysis of optimization experiments
- ❖ Demonstrate custom algorithms and graphical algorithm designer
- ❖ Demonstrate data-based modeling features
- ❖ Outline some additional features

4



Introduction



HeuristicLab
A Paradigm-Independent and Extensible
Environment for Heuristic Optimization

❖ Motivation and Goals

- graphical user interface
- paradigm independence
- multiple algorithms and problems
- large scale experiments and analyses
- parallelization
- extensibility, flexibility and reusability
- visual and interactive algorithm development
- multiple layers of abstraction

❖ Facts

- development of HeuristicLab started in 2002
- based on Microsoft .NET and C#
- used in research and education
- second place at the Microsoft Innovation Award 2009
- open source (GNU General Public License)
- version 3.3.0 released on May 18th, 2010
- latest version 3.3.5 released in June 2011



5



Where to get HeuristicLab?

❖ Download binaries

- deployed as ZIP archives
- latest stable version 3.3.5
 - released in June 2011
- daily trunk build
- <http://dev.heuristiclab.com/download>

❖ Check out sources

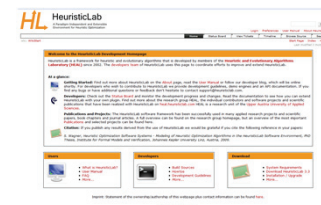
- SVN repository
- HeuristicLab 3.3.5 tag
 - <http://dev.heuristiclab.com/svn/hl/core/tags/3.3.5>
- current development trunk
 - <http://dev.heuristiclab.com/svn/hl/core/trunk>

❖ License

- GNU General Public License (Version 3)

❖ System requirements

- Microsoft .NET Framework 4.0 Full Version
- enough RAM and CPU power ;-)



6



Plugin Infrastructure

❖ HeuristicLab consists of many assemblies

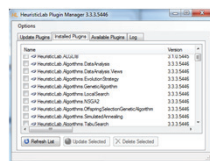
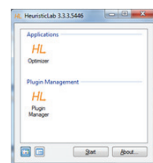
- for example 74 plugins in HeuristicLab 3.3.3
- plugins can be loaded or unloaded at runtime
- plugins can be updated via internet
- application plugins provide GUI frontends

❖ Extensibility

- developing and deploying new plugins is easy
- dependencies are explicitly defined, automatically checked and resolved
- automatic discovery of interface implementations (service locator pattern)

❖ Plugin Manager

- GUI to check, install, update or delete plugins



7



Graphical User Interface

❖ HeuristicLab GUI is made up of views

- views are visual representations of content objects
- views are composed in the same way as their content
- views and content objects are loosely coupled
- multiple different views may exist for the same content

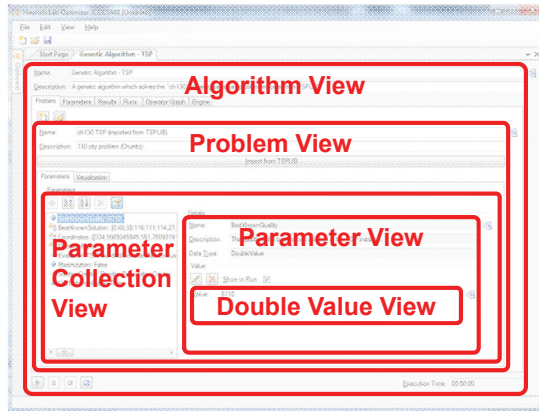
❖ Drag & Drop

- views support drag & drop operations
- content objects can be copied (default) or moved (shift key)
- enabled for collection items and content objects

8



Graphical User Interface



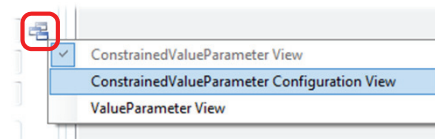
9



Graphical User Interface

❖ ViewHost

- control which hosts views
- right-click on windows icon to switch views
- double-click on windows icon to open another view
- drag & drop windows icon to copy contents



10



Available Algorithms & Problems

❖ Algorithms

- Genetic Algorithm
- Island Genetic Algorithm
- Offspring Selection Genetic Algorithm
- Island Offspring Selection Genetic Algorithm
- SASEGASA
- Evolution Strategy
- NSGA-II
- Particle Swarm Optimization
- Local Search
- Simulated Annealing
- Tabu Search
- Variable Neighborhood Search
- Linear Regression
- Linear Discriminant Analysis
- Support Vector Machine
- k-Means
- User-defined Algorithm

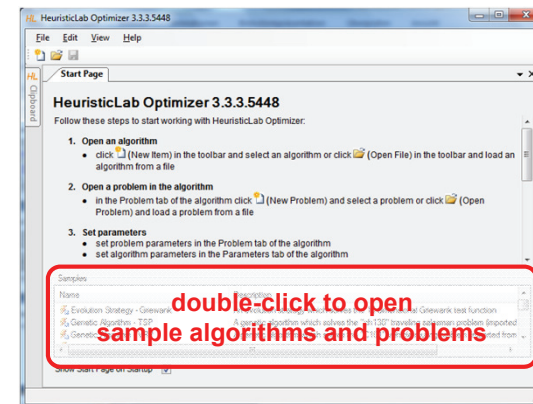
❖ Problems

- Single-Objective Test Function
- Traveling Salesman Problem
- Quadratic Assignment Problem
- Vehicle Routing Problem
- Scheduling
- Knapsack
- OneMax
- Data Analysis
- Regression
- Symbolic Regression
- Classification
- Symbolic Classification
- Clustering
- Artificial Ant
- External Evaluation Problem
- User-defined Problem

11

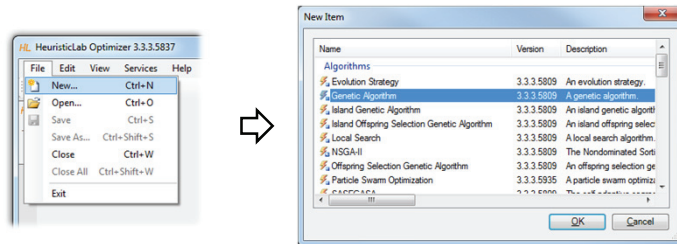


HeuristicLab Optimizer



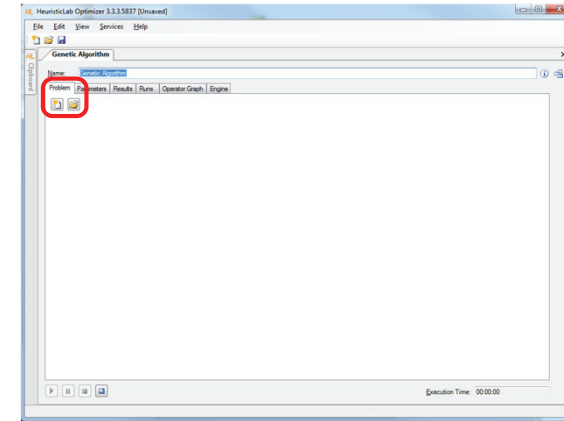
12

GECCO Create Algorithm



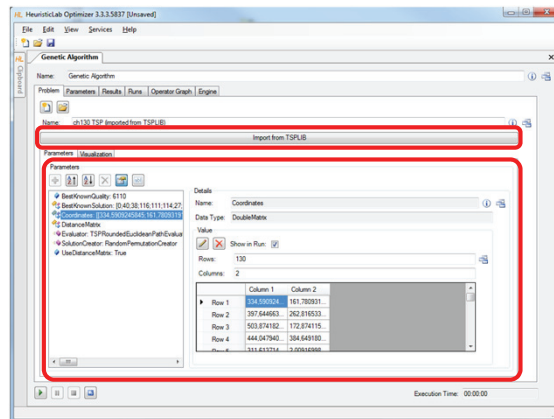
13

GECCO Create or Load Problem



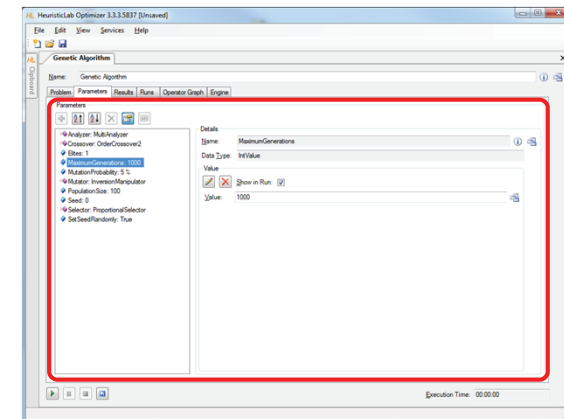
14

GECCO Import or Parameterize Problem Data



15

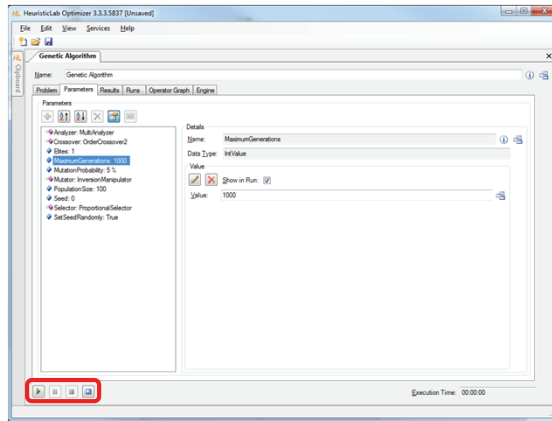
GECCO Parameterize Algorithm



16



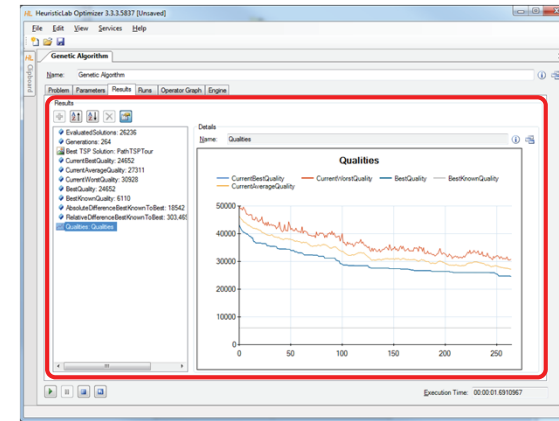
Start, Pause, Resume, Stop and Reset



17



Inspect Results

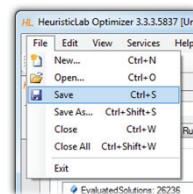


18



Save and Load

- ❖ Save to and load from disk
 - HeuristicLab items (i.e., algorithms, problems, experiments, ...) can be saved to and loaded from a file
 - algorithms can be paused, saved, loaded and resumed
 - data format is custom compressed XML
 - saving and loading files might take several minutes
 - saving and loading large experiments requires some memory

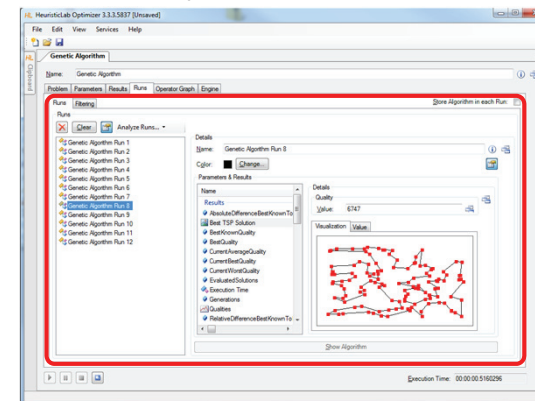


19



Compare Runs

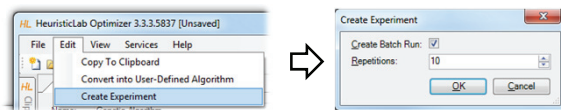
- ❖ A run is created each time when the algorithm is stopped
 - runs contain all results and parameter settings
 - previous results are not forgotten and can be compared



20

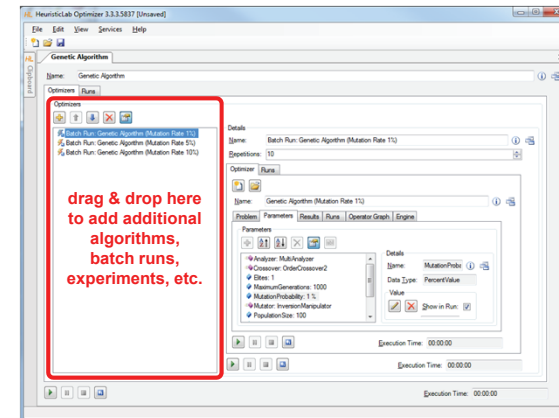
Create Batch Runs and Experiments

- ❖ Batch runs
 - execute the same optimizer (e.g. algorithm, batch run, experiment) several times
- ❖ Experiments
 - execute different optimizers
 - suitable for large scale algorithm comparison and analysis
- ❖ Experiments and batch runs can be nested
- ❖ Generated runs can be compared afterwards



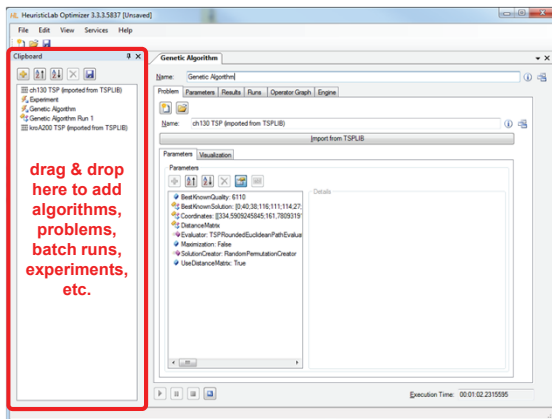
21

Create Batch Runs and Experiments



22

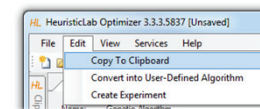
Clipboard



23

Clipboard

- ❖ Store items
 - click on the buttons to add or remove items
 - drag & drop items on the clipboard
 - use the menu to add a copy of a shown item to the clipboard

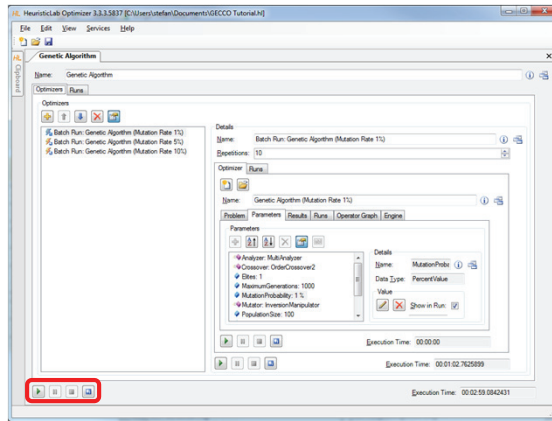


- ❖ Show items
 - double-click on an item in the clipboard to show its view
- ❖ Save and restore clipboard content
 - click on the save button to write the clipboard content to disk
 - clipboard is automatically restored when HeuristicLab is started the next time

24



Start, Pause, Resume, Stop, Reset



25



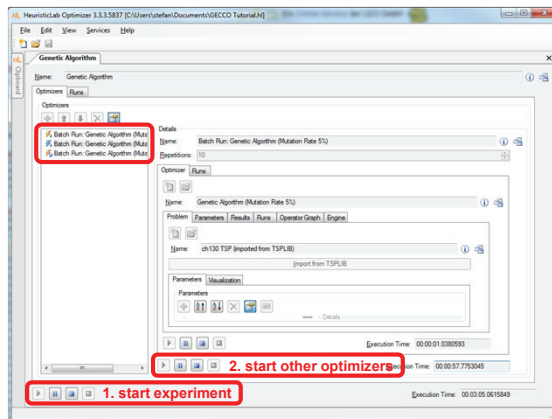
Multi-core CPUs and Parallelization

- ❖ Parallel execution of optimizers in experiments
 - optimizers in an experiment are executed sequentially from top to bottom per default
 - experiments support parallel execution of their optimizers
 - select a not yet executed optimizer and start it manually to utilize another core
 - execution of one of the next optimizers is started automatically after an optimizer is finished
- ❖ Parallel execution of algorithms
 - HeuristicLab provides special operators for parallelization
 - engines decide how to execute parallel operations
 - sequential engine executes everything sequentially
 - parallel engine executes parallel operations on multiple cores
 - Hive engine (under development) executes parallel operations on multiple computers
 - all implemented algorithms support parallel solution evaluation

26



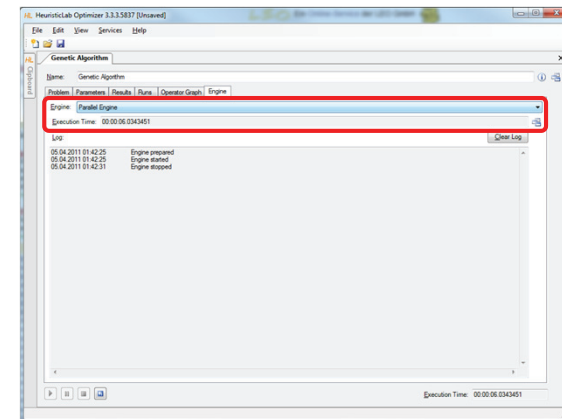
Parallel Execution of Experiments



27



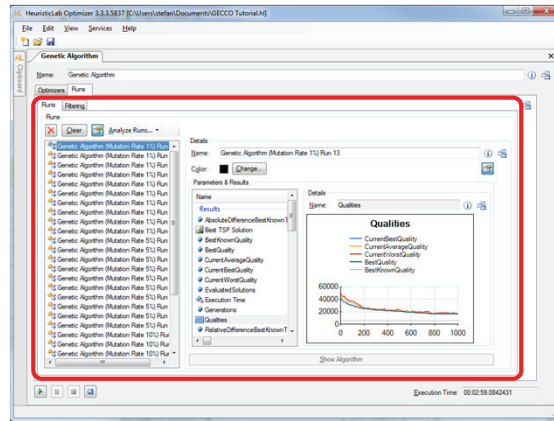
Parallel Execution of Algorithms



28



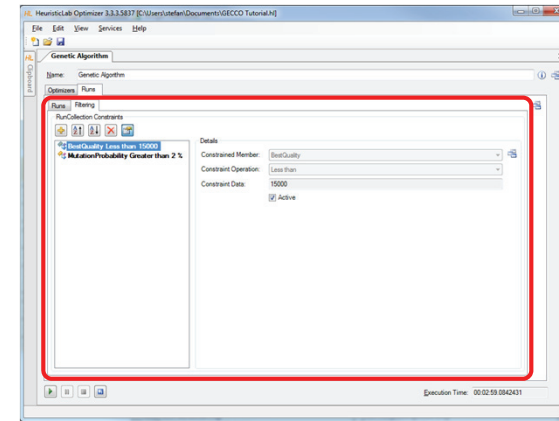
Compare Runs



29



Filter Runs

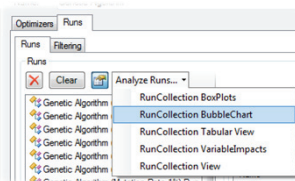


30



Analyze Runs

- ❖ HeuristicLab provides interactive views to analyze and compare all runs of a run collection
 - textual analysis
 - RunCollection Tabular View
 - graphical analysis
 - RunCollection BubbleChart
 - RunCollection BoxPlots
- ❖ Filtering is automatically applied to all open run collection views



31



RunCollection Tabular View

Run	BestKnownQuality	BestKnownSolution	BestQuality	Coordinates	Crossover	CurrentAverageQuality
Genetic Algorithm (Mutation Rate 1%) Run 13	1110	[0.40.38.116.111.114...	16405	[334.5500245...	OrderCrossover	16543.13
Genetic Algorithm (Mutation Rate 1%) Run 14	1110	[0.40.38.116.111.114...	14783	[334.5500245...	OrderCrossover	16205.02
Genetic Algorithm (Mutation Rate 1%) Run 15	1110	[0.40.38.116.111.114...	14252	[334.5500245...	OrderCrossover	14202.89
Genetic Algorithm (Mutation Rate 1%) Run 16	1110	[0.40.38.116.111.114...	13243	[334.5500245...	OrderCrossover	13245.95
Genetic Algorithm (Mutation Rate 1%) Run 17	1110	[0.40.38.116.111.114...	13703	[334.5500245...	OrderCrossover	13749.88
Genetic Algorithm (Mutation Rate 1%) Run 18	1110	[0.40.38.116.111.114...	13984	[334.5500245...	OrderCrossover	13991.09
Genetic Algorithm (Mutation Rate 1%) Run 19	1110	[0.40.38.116.111.114...	15421	[334.5500245...	OrderCrossover	15431.74
Genetic Algorithm (Mutation Rate 1%) Run 20	1110	[0.40.38.116.111.114...	14409	[334.5500245...	OrderCrossover	15147
Genetic Algorithm (Mutation Rate 1%) Run 21	1110	[0.40.38.116.111.114...	13771	[334.5500245...	OrderCrossover	13954.56
Genetic Algorithm (Mutation Rate 1%) Run 22	1110	[0.40.38.116.111.114...	14029	[334.5500245...	OrderCrossover	14512.3
Genetic Algorithm (Mutation Rate 1%) Run 23	1110	[0.40.38.116.111.114...	13095	[334.5500245...	OrderCrossover	13642.7
Genetic Algorithm (Mutation Rate 1%) Run 24	1110	[0.40.38.116.111.114...	12403	[334.5500245...	OrderCrossover	12810.09
Genetic Algorithm (Mutation Rate 1%) Run 25	1110	[0.40.38.116.111.114...	14091	[334.5500245...	OrderCrossover	14653.98
Genetic Algorithm (Mutation Rate 1%) Run 26	1110	[0.40.38.116.111.114...	12555	[334.5500245...	OrderCrossover	13247.99
Genetic Algorithm (Mutation Rate 1%) Run 27	1110	[0.40.38.116.111.114...	12782	[334.5500245...	OrderCrossover	13264.35
Genetic Algorithm (Mutation Rate 1%) Run 28	1110	[0.40.38.116.111.114...	12711	[334.5500245...	OrderCrossover	13151.19
Genetic Algorithm (Mutation Rate 1%) Run 29	1110	[0.40.38.116.111.114...	12226	[334.5500245...	OrderCrossover	12625.78
Genetic Algorithm (Mutation Rate 1%) Run 30	1110	[0.40.38.116.111.114...	13345	[334.5500245...	OrderCrossover	13777.85
Genetic Algorithm (Mutation Rate 1%) Run 31	1110	[0.40.38.116.111.114...	12887	[334.5500245...	OrderCrossover	13284.81
Genetic Algorithm (Mutation Rate 1%) Run 32	1110	[0.40.38.116.111.114...	12741	[334.5500245...	OrderCrossover	13115.18
Genetic Algorithm (Mutation Rate 1%) Run 33	1110	[0.40.38.116.111.114...	15821	[334.5500245...	OrderCrossover	16084.04
Genetic Algorithm (Mutation Rate 1%) Run 34	1110	[0.40.38.116.111.114...	16384	[334.5500245...	OrderCrossover	16059.36

32



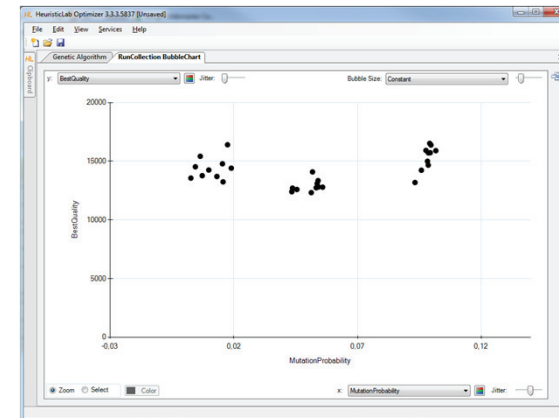
RunCollection Tabular View

- ❖ Sort columns
 - click on column header to sort column
 - Ctrl-click on column header to sort multiple columns
- ❖ Show or hide columns
 - right-click on table to open dialog to show or hide columns
- ❖ Compute statistical values
 - select multiple numerical values to see count, sum, minimum, maximum, average and standard deviation
- ❖ Select, copy and paste into other applications

33



RunCollection BubbleChart



34



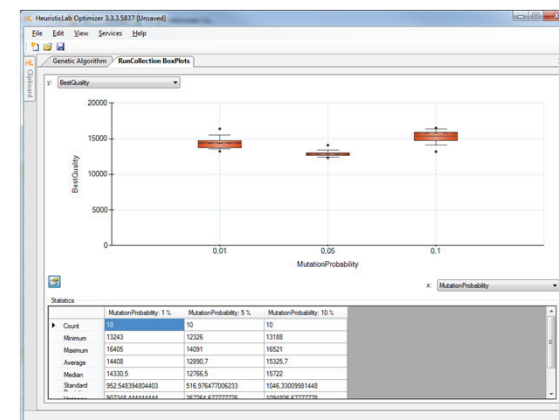
RunCollection BubbleChart

- ❖ Choose values to plot
 - choose which values to show on the x-axis, the y-axis and as bubble size
 - possible values are all parameter settings and results
- ❖ Add jitter
 - add jitter to separate overlapping bubbles
- ❖ Zoom in and out
 - click on Zoom and click and drag in the chart area to zoom in
 - double click on the chart area background or on the circle buttons beside the scroll bars to zoom out
- ❖ Color bubbles
 - click on Select, choose a color and click and drag in the chart area to select and color bubbles
 - apply coloring automatically by clicking on the axis coloring buttons
- ❖ Show runs
 - double click on a bubble to open its run
- ❖ Export image
 - right-click to open context menu to copy or save image
 - save image as pixel (BMP, JPG, PNG, GIF, TIF) or vector graphics (EMF)
- ❖ Show box plots
 - right-click to open context menu to show box plots view

35



RunCollection BoxPlots



36



RunCollection BoxPlots

- ❖ Choose values to plot
 - choose which values to show on the x-axis and y-axis
 - possible values are all parameter settings and results
- ❖ Zoom in and out
 - click on Zoom and click and drag in the chart area to zoom in
 - double click on the chart area background or on the circle buttons beside the scroll bars to zoom out
- ❖ Show or hide statistical values
 - click on the lower left button to show or hide statistical values
- ❖ Export image
 - right-click to open context menu to copy or save image
 - save image as pixel (BMP, JPG, PNG, GIF, TIF) or vector graphics (EMF)

37



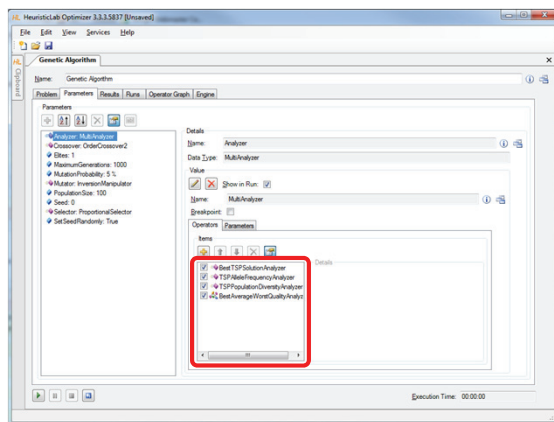
Analyzers

- ❖ Special operators for analysis purposes
 - are executed after each iteration
 - serve as general purpose extension points of algorithms
 - can be selected and parameterized in the algorithm
 - perform algorithm-specific and/or problem-specific tasks
 - some analyzers are quite costly regarding runtime and memory
 - implementing and adding custom analyzers is easy
- ❖ Examples
 - TSPAlleleFrequencyAnalyzer
 - TSPPopulationDiversityAnalyzer
 - SuccessfulOffspringAnalyzer
 - SymbolicDataAnalysisVariableFrequencyAnalyzer
 - SymbolicRegressionSingleObjectiveTrainingBestSolutionAnalyzer
 - ...

38



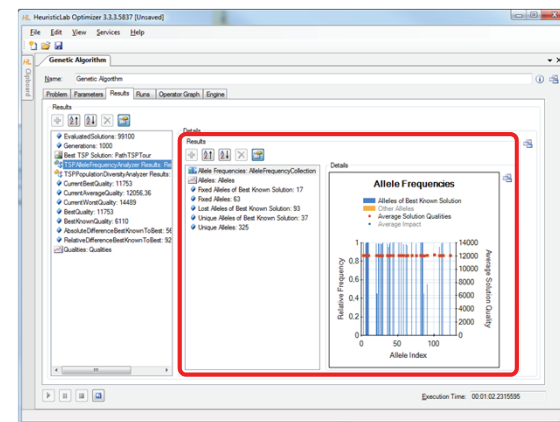
Analyzers



39



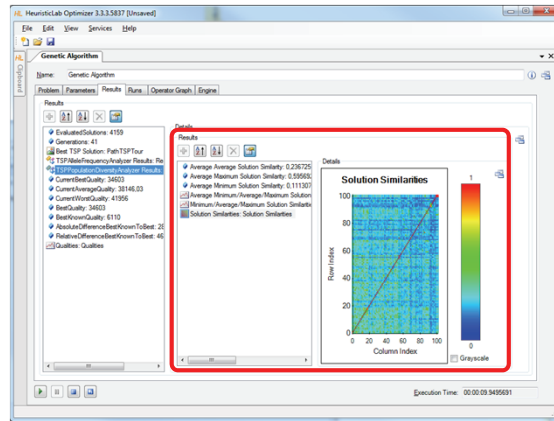
TSPAlleleFrequencyAnalyzer



40



TSPPopulationDiversityAnalyzer

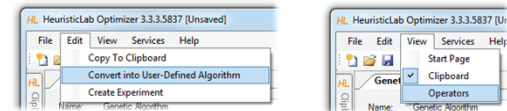


41



Building User-Defined Algorithms

- ❖ Operator graphs
 - algorithms are represented as operator graphs
 - operator graphs of user-defined algorithms can be changed
 - algorithms can be defined in the graphical algorithm designer
 - use the menu to convert a standard algorithm into a user-defined algorithm

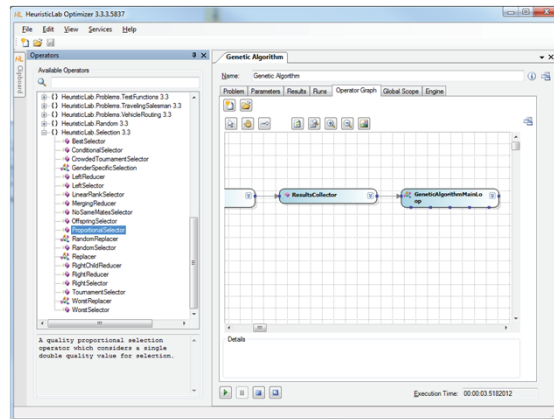


- ❖ Operators sidebar
 - drag & drop operators into an operator graph
- ❖ Programmable operators
 - add programmable operators in order to implement custom logic in an algorithm
 - no additional development environment needed
- ❖ Debug algorithms
 - use the debug engine to obtain detailed information during algorithm execution

42



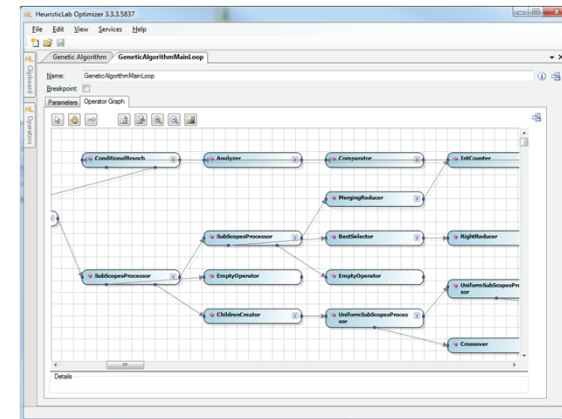
Building User-Defined Algorithms



43



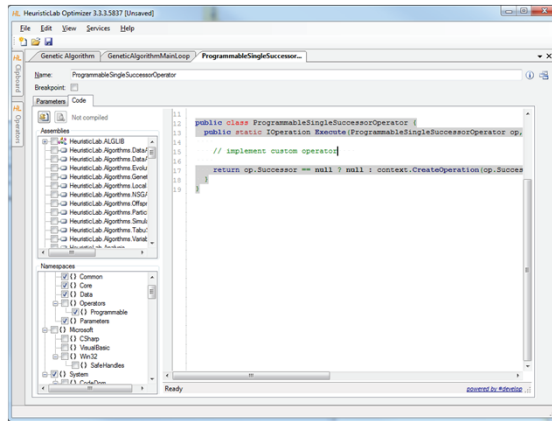
Building User-Defined Algorithms



44



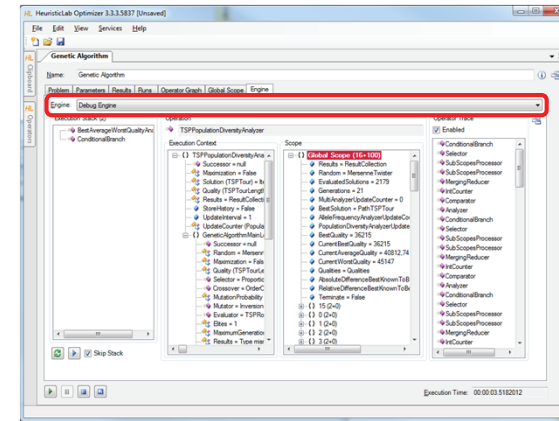
Programmable Operators



45



Debugging Algorithms



46



Introduction to Data-based Modeling

- ❖ Dataset: Matrix $(x_{ij})_{i=1..N, j=1..K}$
 - N observations of K input variables
 - x_{ij} = i-th observation of j-th variable
 - Additionally: Vector of labels $(y_1, \dots, y_N)^T$
- ❖ Goal: learn association of input variable values to labels
- ❖ Common tasks
 - Regression (real-valued labels)
 - Classification (discrete labels)
 - Clustering (no labels, group similar observations)

47



Data-based Modeling Algorithms in HeuristicLab

- ❖ Symbolic regression and classification based on genetic programming
- ❖ External Libraries:
 - Support Vector Machines for Regression and Classification
 - Linear Regression
 - Linear Discriminate Analysis
 - K-Means clustering

48



In the next 60 minutes: case studies

❖ Regression

- Artificial benchmark problem dataset *Poly-10*
- Algorithms:
 - Linear regression
 - Symbolic regression using Genetic Programming

❖ Classification

- Real world medical *Mammographic Mass* dataset from the UCI Machine Learning Repository
- Algorithms:
 - Symbolic classification

49



Case study: Regression

❖ Poly-10 benchmark problem dataset

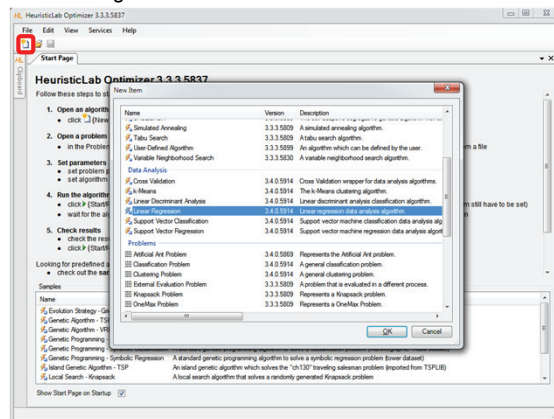
- 10 input variables $x_1 \dots x_{10}$
- $y = x_1 x_2 + x_3 x_4 + x_5 x_6 + x_1 x_7 x_9 + x_3 x_6 x_{10}$
- Non-linear modeling approach necessary
- Frequently used in GP literature
- Download:
 - <http://dev.heuristiclab.com/AdditionalMaterial#GECCO2011>

50



Linear Regression

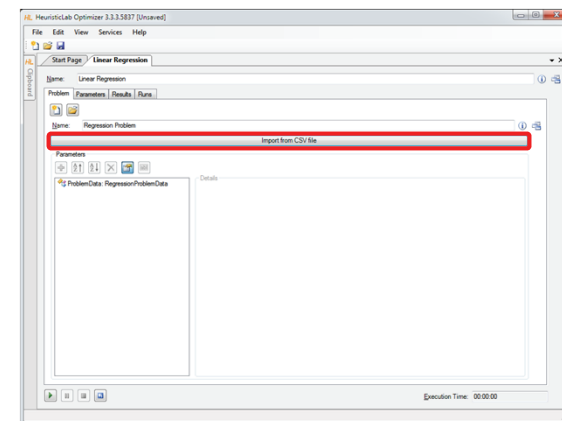
❖ Create new algorithm



51



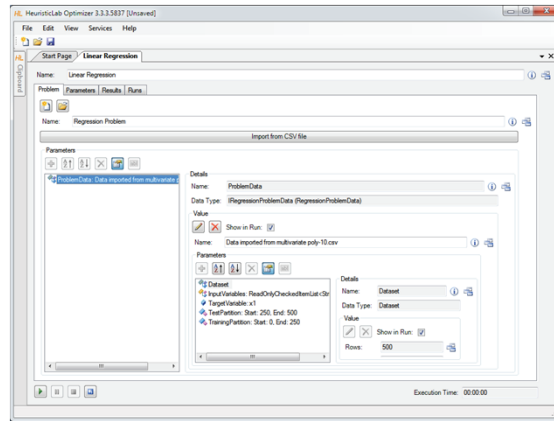
Import Data from CSV-File



52



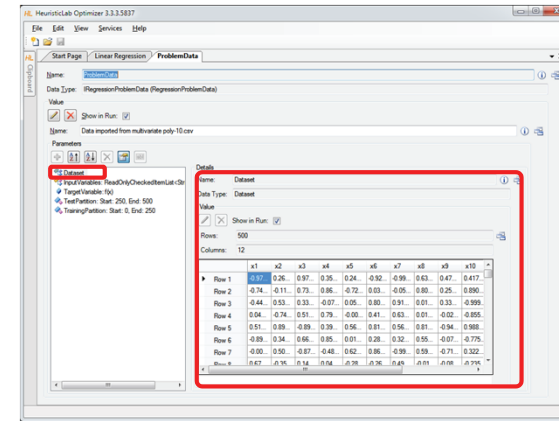
Inspect And Configure Dataset



53



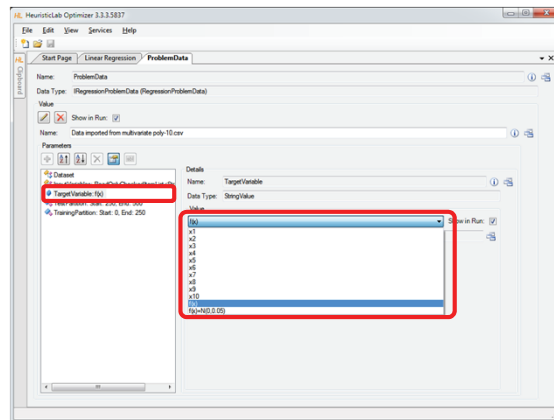
Inspect Imported Data



54



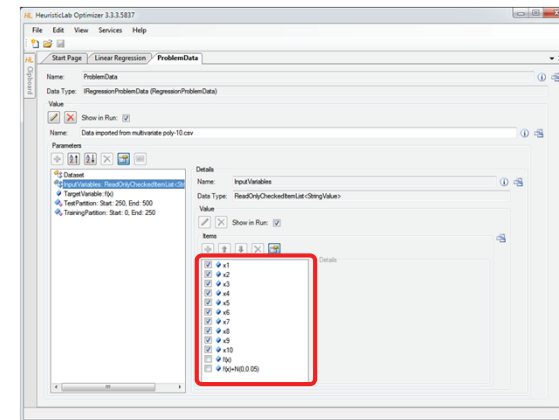
Set Target Variable



55



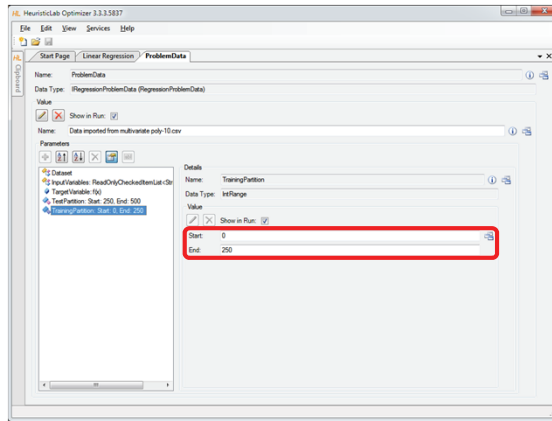
Select Input Variables



56



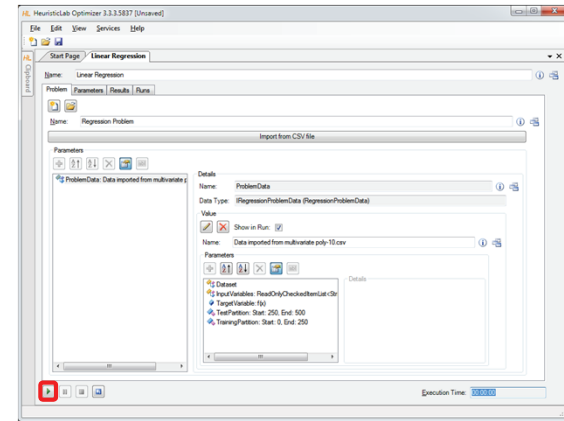
Configure Training and Test Partitions



57



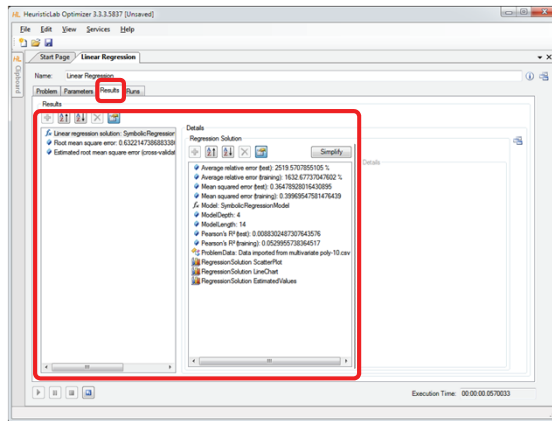
Run Linear Regression



58



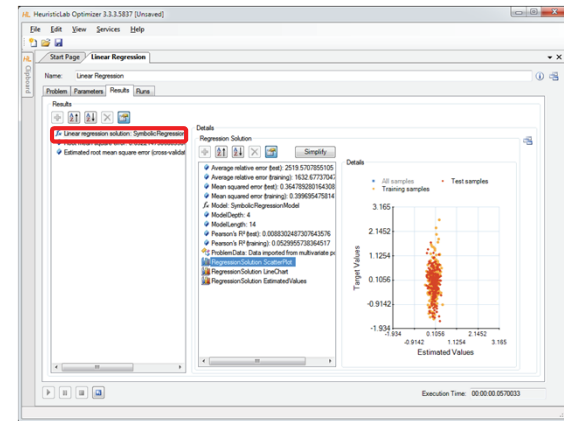
Inspect Results



59



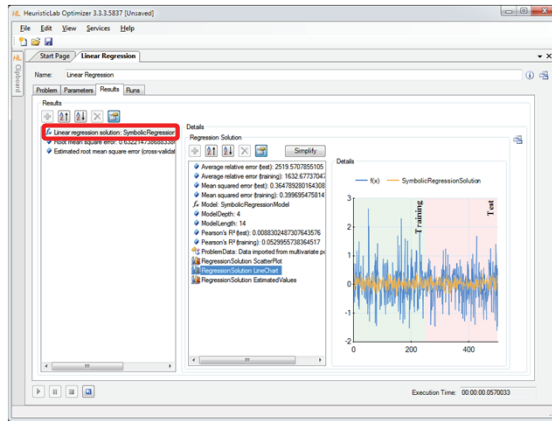
Inspect Scatterplot of Predicted and Target Values



60



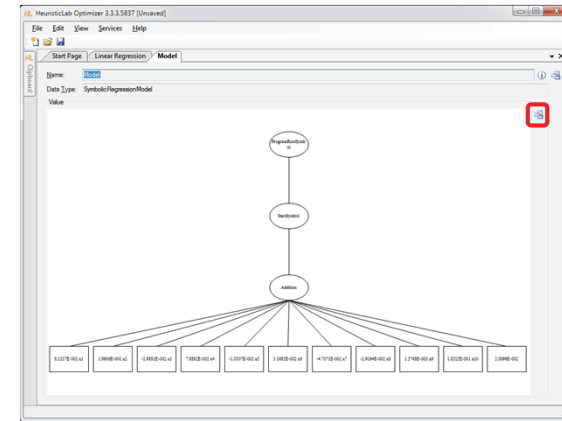
Inspect Linechart



61



Inspect Graphical Representation of Model

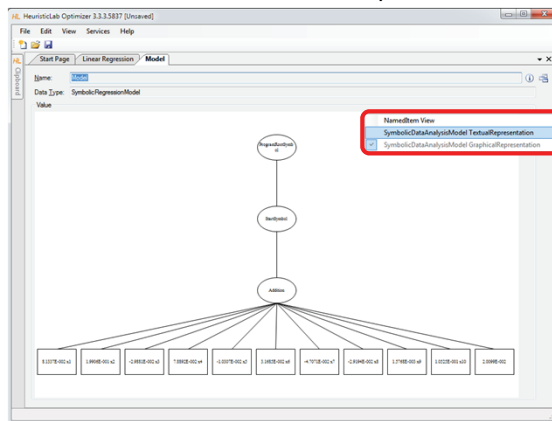


62



Textual Representations Are Also Available

- ❖ Use ViewHost to switch to textual representation view.



63

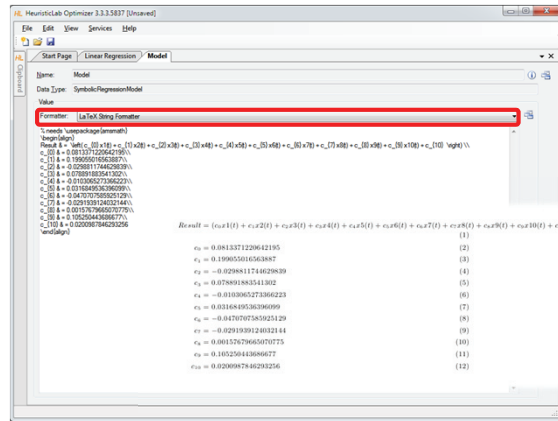


Default Textual Representation for Model Export

64



Textual Representation for Export to LaTeX



65

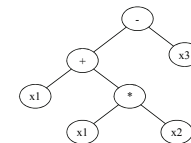


Nonlinear Modeling: Symbolic Regression

- ❖ Linear regression produced an inaccurate model.
- ❖ Next: produce a nonlinear symbolic regression model using genetic programming

Genetic programming

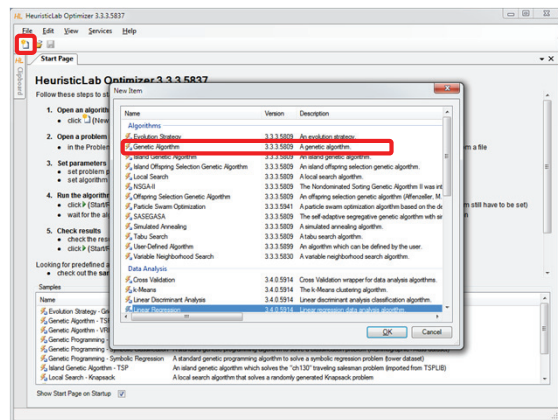
- Evolve variable-length models
- Model representation: symbolic expression tree
- Structure and model parameters are evolved side-by-side
- White-box models



66



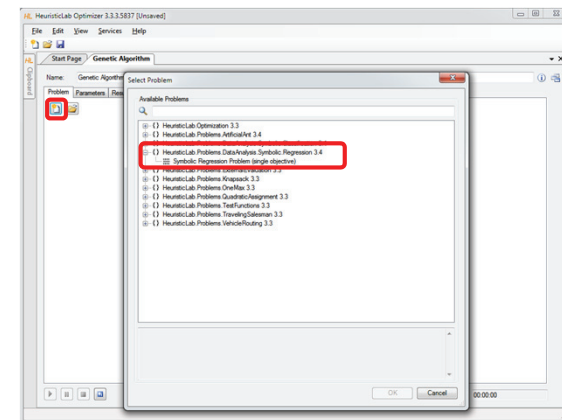
Create New Genetic Algorithm



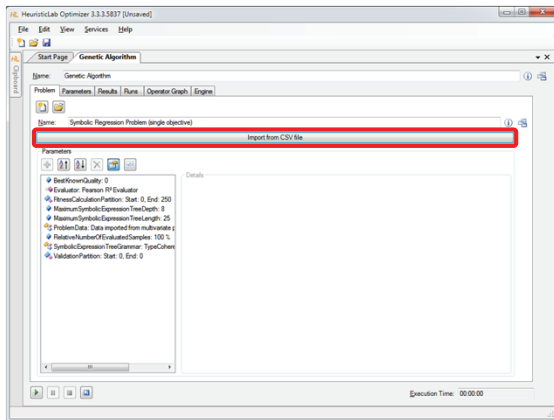
67



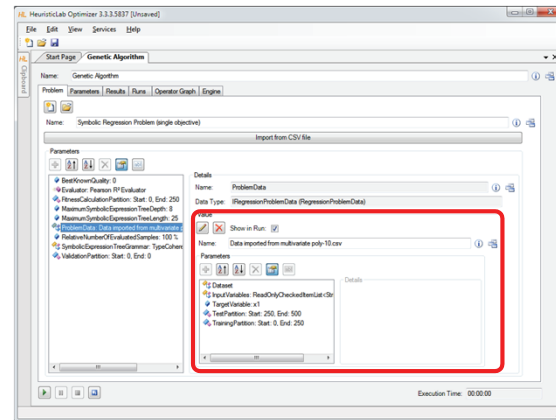
Create New Symbolic Regression Problem



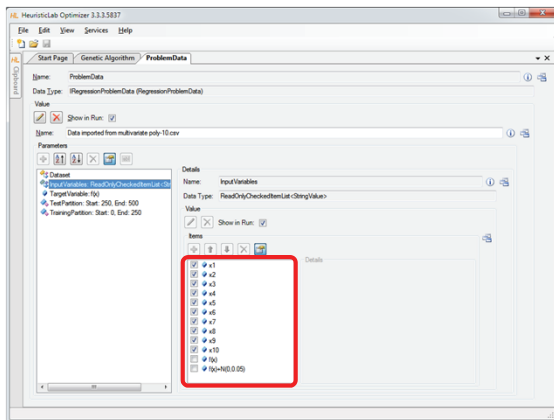
68



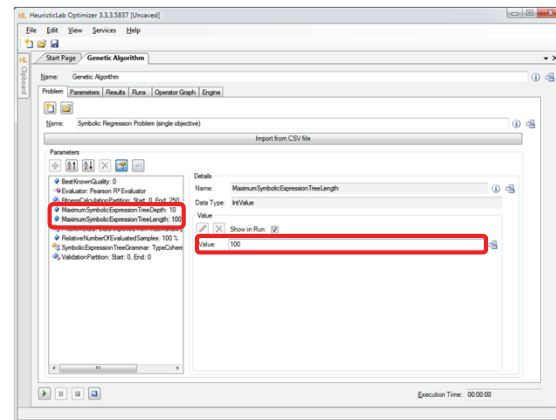
69



70



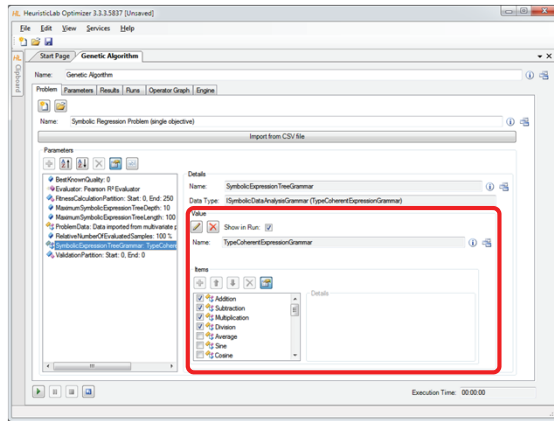
71



72



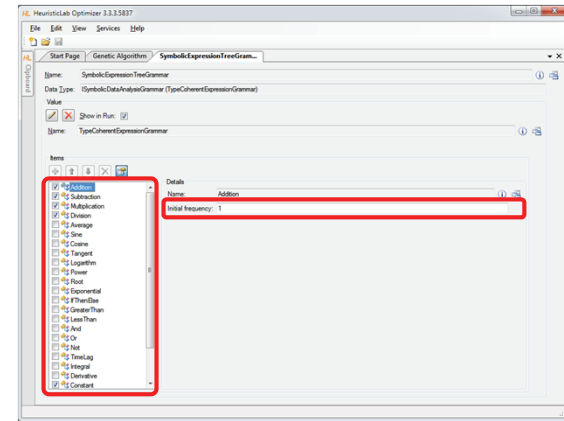
Configure Function Set (Grammar)



73



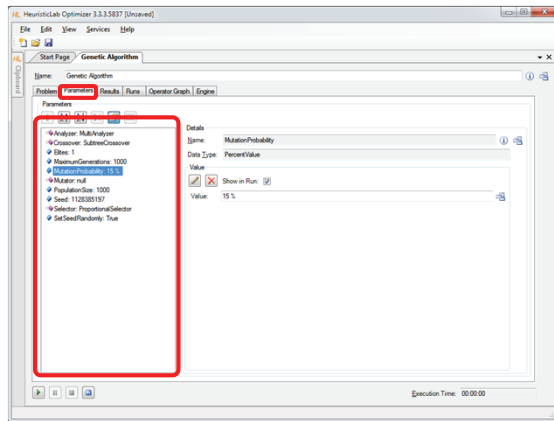
Configure Function Set (Grammar)



74



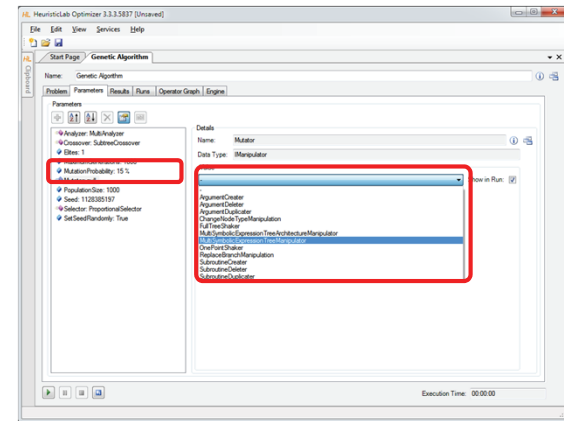
Configure Algorithm Parameters



75



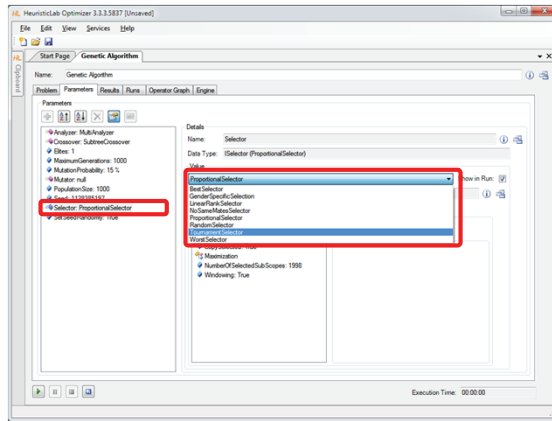
Configure Mutation Operator



76



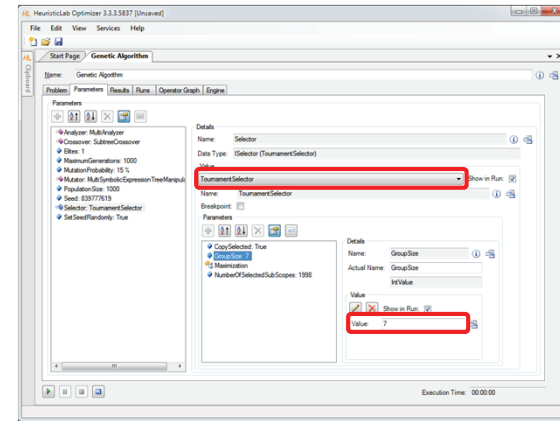
Configure Selection Operator



77



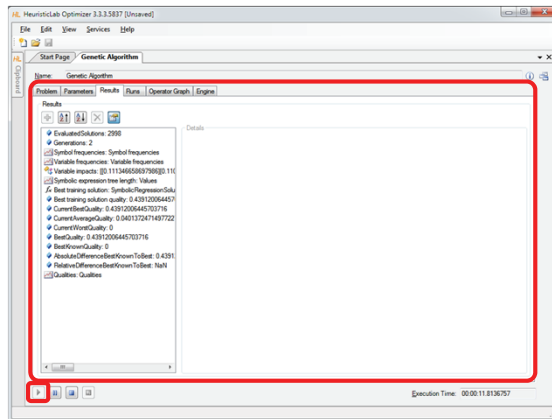
Configure Tournament Group Size



78



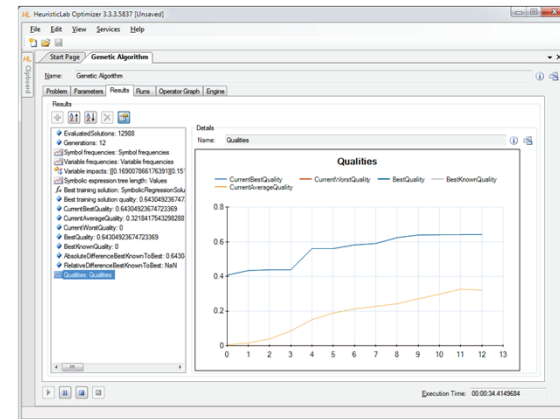
Start Algorithm and Inspect Results



79



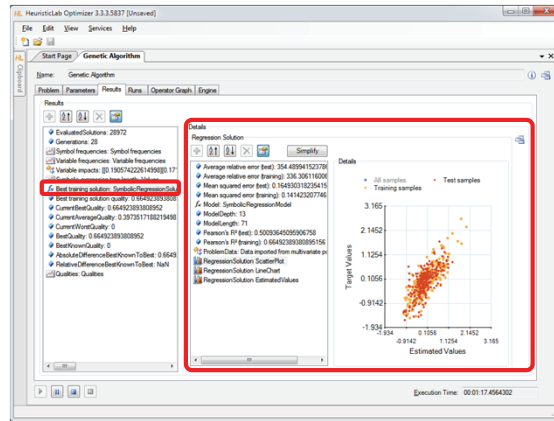
Inspect Quality Chart



80



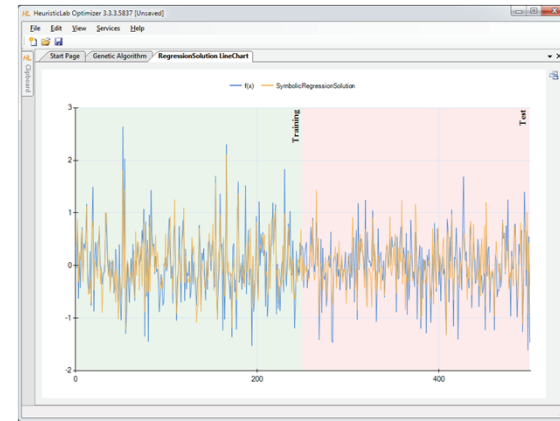
Inspect Best Model on Training Partition



81



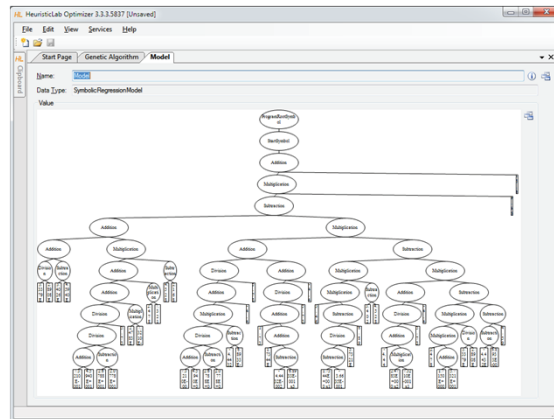
Inspect Linechart of Best Model on Trainingset



82



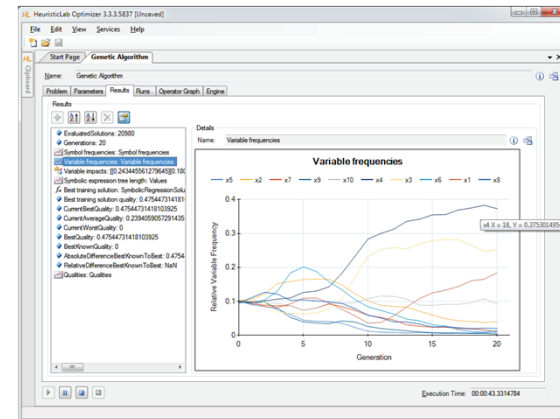
Inspect Structure of Best Model on Trainingset



83



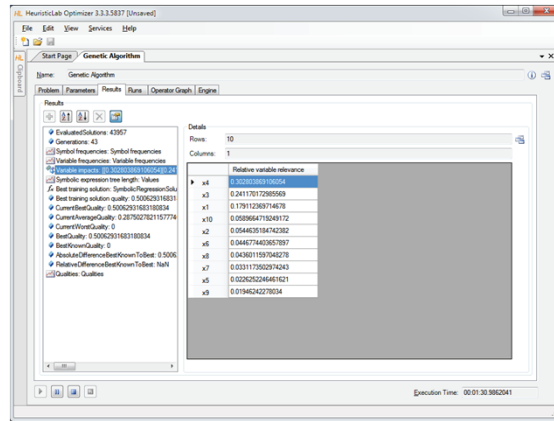
Inspect Variable Frequency Chart



84



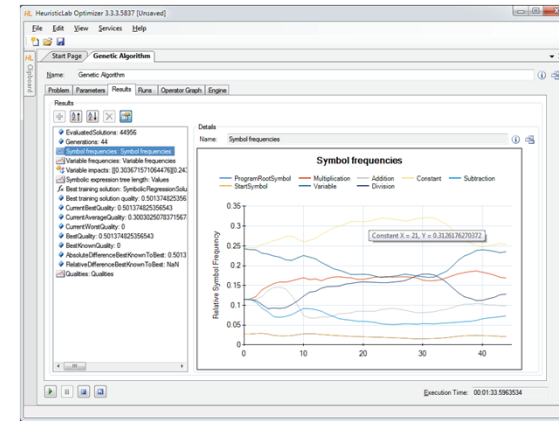
Inspect Variable Impacts



85



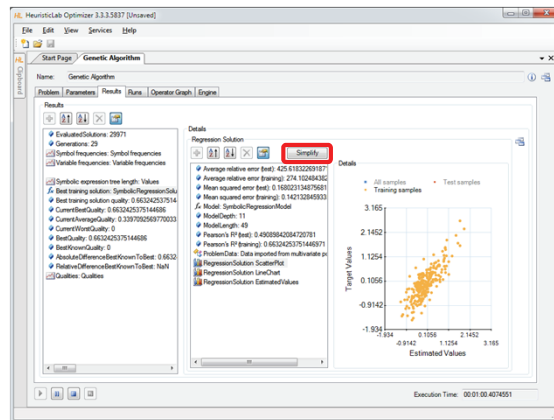
Inspect Symbol Frequencies



86



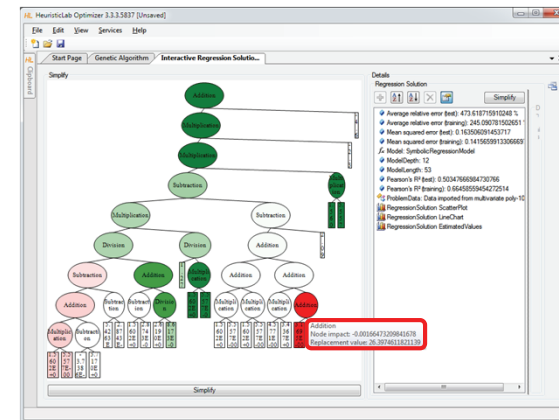
Detailed Model Analysis and Simplification



87

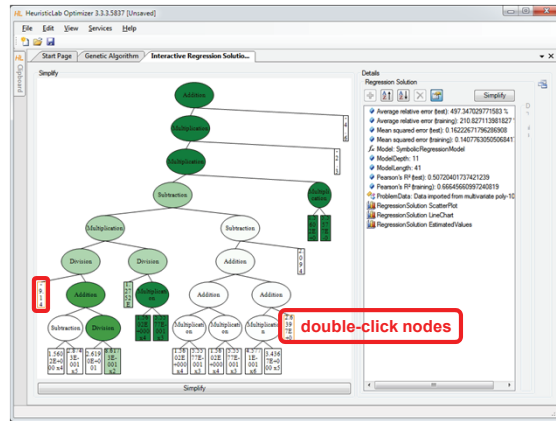


Symbolic Simplification and Node Impacts



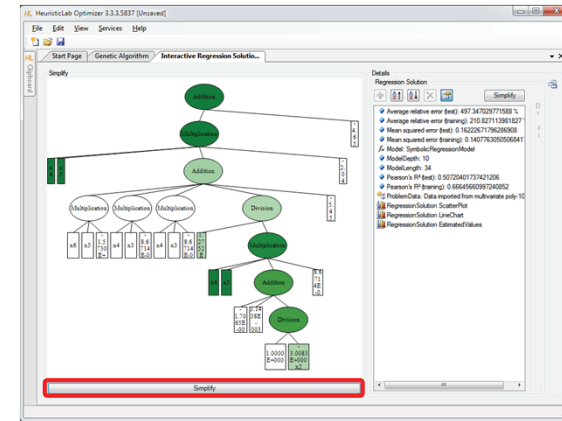
88

GECCO Manual Simplification



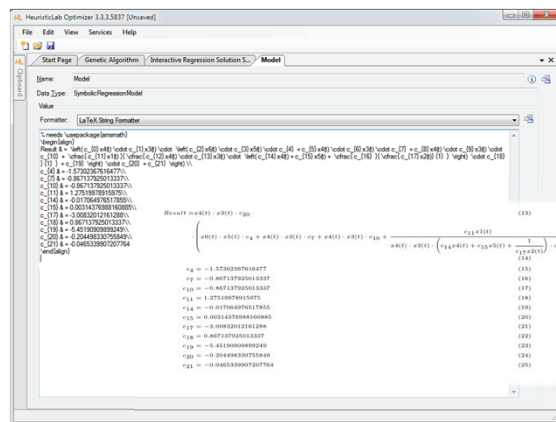
89

GECCO Automatic Symbolic Simplification



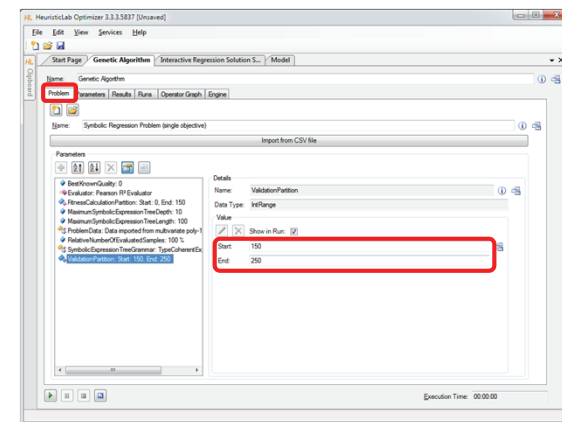
90

GECCO LaTeX Export



91

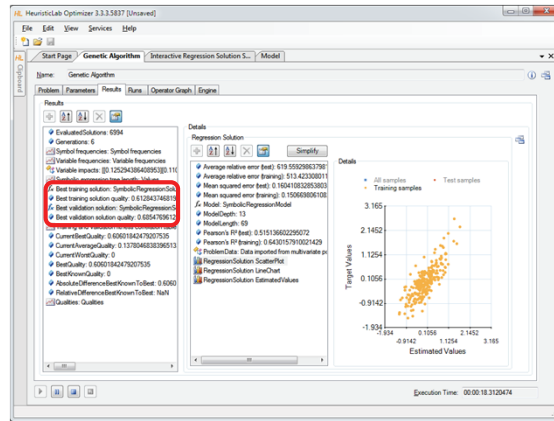
GECCO Configuration of Validation Partition



92



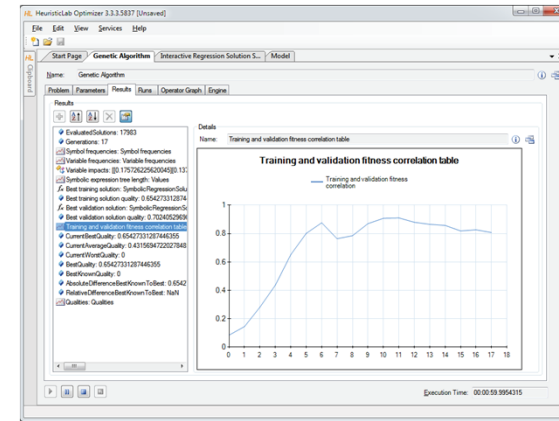
Inspect Best Model on Validation Partition



93



Inspect Linechart of Correlation of Training and Validation Fitness



94



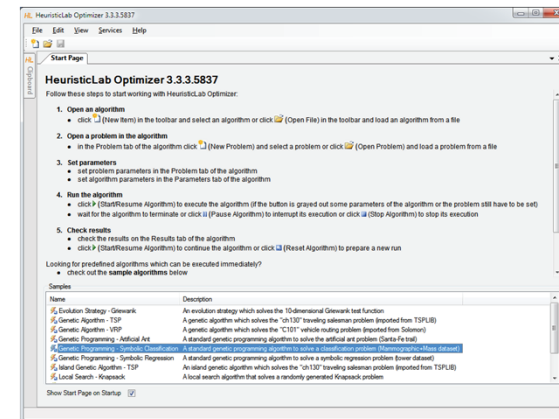
Case study: Classification

- ❖ Real world medical dataset (*Mammographic Mass*) from UCI Machine Learning Repository (Frank & Asuncion)
 - data from non-invasive mammography screening
 - variables:
 - patient age
 - visual features of inspected mass lesions: shape, margin, density
 - target variable: severity (malignant, benign)
- Download:
 - <http://dev.heuristiclab.com/AdditionalMaterial#GECCO2011>

95



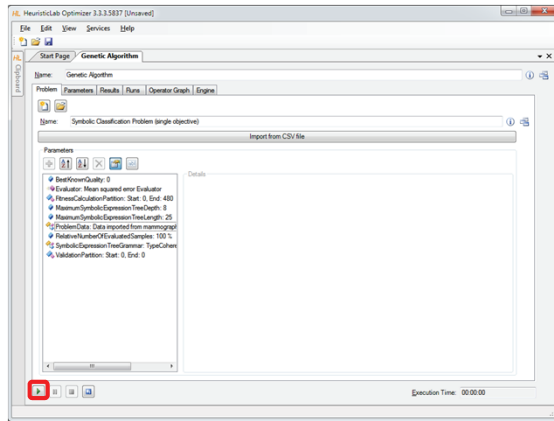
Open Sample



96



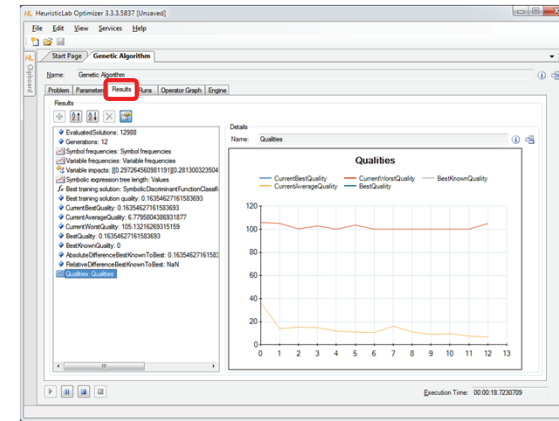
Configure and Run Algorithm



97



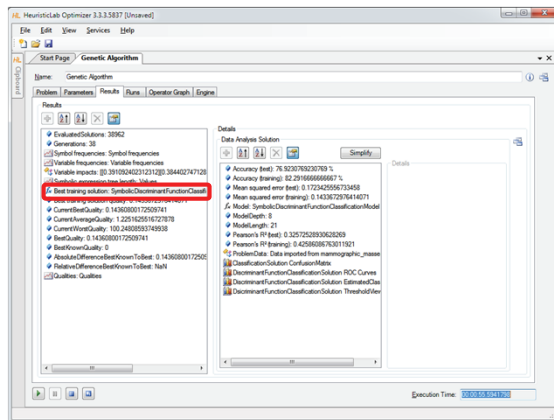
Inspect Quality Linechart



98



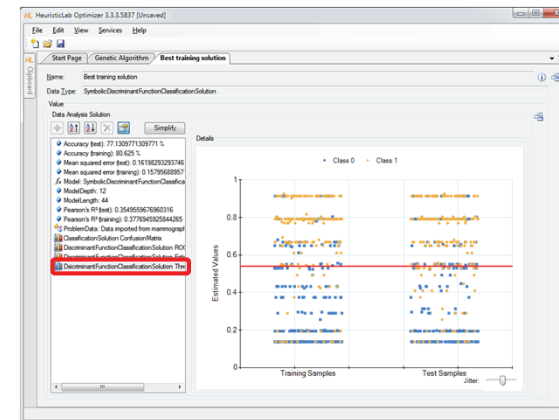
Inspect Best Training Solution



99



Inspect Model Output and Thresholds



100

Inspect Confusion Matrix



Inspect ROC curve



Analyse and Simplify Best Training Solution

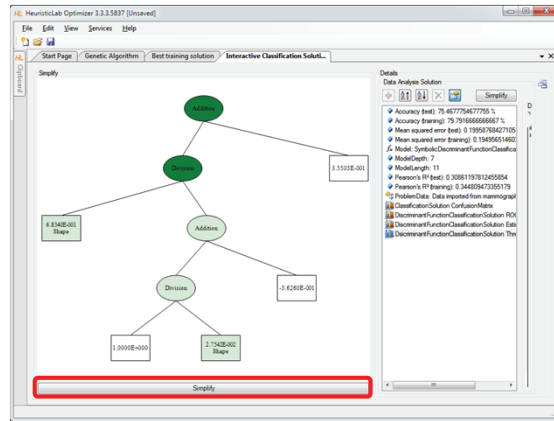


Analyse and Simplify Model





Symbolically Simplified Model



105



Some Additional Features

- ❖ HeuristicLab Hive
 - parallel and distributed execution of algorithms and experiments on many computers in a network
- ❖ Optimization Knowledge Base (OKB)
 - database to store algorithms, problems, parameters and results
 - open to the public
 - open for other frameworks
 - analyze and store characteristics of problem instances and problem classes
- ❖ External solution evaluation and simulation-based optimization
 - interface to couple HeuristicLab with other applications (MatLab, AnyLogic, ...)
 - supports different protocols (command line parameters, TCP, ...)
- ❖ Parameter grid tests and meta-optimization
 - automatically create experiments to test large ranges of parameters
 - apply heuristic optimization algorithms to find optimal parameter settings for heuristic optimization algorithms



106



Planned Features

- ❖ Algorithms & Problems
 - steady-state genetic algorithm
 - unified tabu search for vehicle routing
 - scatter search
 - ...
- ❖ Cloud Computing
 - port HeuristicLab Hive to Windows Azure
- ❖ Linux
 - port HeuristicLab to run on Mono and Linux machines
- ❖ Have a look at the HeuristicLab roadmap
 - <http://dev.heuristiclab.com/trac/hl/core/roadmap>
- ❖ Any other ideas, requests or recommendations?
 - please write an e-mail to support@heuristiclab.com

107



HeuristicLab Team



Heuristic and Evolutionary Algorithms Laboratory (HEAL)
School of Informatics, Communications and Media
Upper Austria University of Applied Sciences

Softwarepark 11
A-4232 Hagenberg
AUSTRIA

<http://heal.heuristiclab.com>

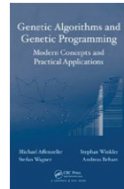
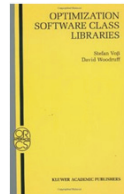


108



Suggested Readings

- ❖ S. Voß, D. Woodruff (Edts.)
Optimization Software Class Libraries
Kluwer Academic Publishers, 2002
- ❖ M. Affenzeller, S. Winkler, S. Wagner, A. Beham
**Genetic Algorithms and Genetic Programming
Modern Concepts and Practical Applications**
CRC Press, 2009



109



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HeuristicLab: A generic and extensible optimization environment
Adaptive and Natural Computing Algorithms, pp. 538-541
Springer, 2005
- ❖ S. Wagner, S. Winkler, R. Braune, G. Kronberger, A. Beham, M. Affenzeller
Benefits of plugin-based heuristic optimization software systems
Computer Aided Systems Theory - EUROCAST 2007, Lecture Notes in Computer Science, vol. 4739, pp. 747-754
Springer, 2007
- ❖ S. Wagner, G. Kronberger, A. Beham, S. Winkler, M. Affenzeller
Modeling of heuristic optimization algorithms
Proceedings of the 20th European Modeling and Simulation Symposium, pp. 106-111
DIPTeM University of Genova, 2008
- ❖ S. Wagner, G. Kronberger, A. Beham, S. Winkler, M. Affenzeller
Model driven rapid prototyping of heuristic optimization algorithms
Computer Aided Systems Theory - EUROCAST 2009, Lecture Notes in Computer Science, vol. 5717, pp. 729-736
Springer, 2009
- ❖ S. Wagner
Heuristic optimization software systems - Modeling of heuristic optimization algorithms in the HeuristicLab software environment
Ph.D. thesis, Johannes Kepler University Linz, Austria, 2009.
- ❖ S. Wagner, A. Beham, G. Kronberger, M. Kommenda, E. Pitzer, M. Kofler, S. Vonofen, S. Winkler, V. Dorfer, M. Affenzeller
HeuristicLab 3.3: A unified approach to metaheuristic optimization
Actas del séptimo congreso español sobre Metaheurísticas, Algoritmos Evolutivos y Bioinspirados (MAEB'2010), 2010
- ❖ Detailed list of all publications of the HEAL research group: <http://research.fh-ooe.at/de/orguniti/detail/356#showpublications>

110



Questions & Answers

<http://dev.heuristiclab.com>

support@heuristiclab.com

111