



## SVM OPTIMIZER

The SVM *Optimizer* is a graphical user interface for efficiently developing SVM two-class classifiers in Matlab. In particular it implements powerful and straightforward parameter optimization approaches developed by J. von Frese and taught in his SVM courses. The SVM *Optimizer* guides the user through the whole classifier optimization and testing process.

The application of support vector machines has often been hampered by the fact that existing software is neither user-friendly nor allows for a systematic, rigorous parameter tuning. When used properly, SVMs are actually a very simple and straightforward tool and even a novice user can develop optimal models.

- Method	Data	
Linear C-SVM	Training data:	
C Linear nu-SVM		
© RBF-SVM	Select Y Y	SVM Optimizer
- Parameters	n-fold Cross-Validation	Test Set-
C 10.^[-6:0.5:6]	Number of splits (n) 10	None / Cross-validation
	Number of repetitions 10	Split training data
Nu 0.05:0.05:0.5		Fraction for test set (0 - 100 %): 20
[]	Grouping variable (e.g. replicates):	
σ	(optional) Select	Select Xtest Xtest
Default	☑ Balanced splits	Select Ytest (optional) Ytest
Pre-processing	Optimize Train Cross-valida	tion Test Export Results
Save Classifier	Load Classifier Help Quit	

Existing commercial software usually applies a grid search over a pre-defined parameter space and suggests the model with the lowest cross-validation error. This has severe disadvantages:

- The optimal parameters might actually lie outside the pre-defined search grid. As the lowest achievable error is unknown and no diagnostics for over- or underfitting is provided, the found solution might actually be far from optimal.
- As no data specific knowledge is used, the search grid is usually laid out on a logarithmic scale and thus might offer no or only limited coverage of the optimal region and therefore result in suboptimal or instable models.
- As cross-validation has been used for parameter optimization, the respective error estimate is optimistically biased and a further test set would be required for a valid error assessment.





Features

- Offers the linear C- and nu-SVM as well as the non-linear C RBF-SVM as essential toolset for tackling all numerical two-class classification problems.
- Uses the efficient LIBSVM optimization module.
- Fast training even for large data sets
- Provides several pre-processing options: Optional auto-scaling for variables measured on different scales, advanced custom preprocessing through a seamless interface to the commercial PLS-Toolbox (www.eigenvector.com, not included). For applying the non-linear RBF-SVM to high-dimensional data, a prior dimensionality reduction by PCA or PLS can be performed.
- Provides unique diagnostics for optimizing the linear SVM parameters C or nu, establishing for each trial solution, if over- or underfitting occurred.
- Provides efficient diagnostics from the data at hand for choosing the RBF parameter  $\boldsymbol{\sigma}$  reliably.
- Repeated n-fold cross-validation for error assessment
- Allows the rigorous definition of cross-validation partitions (e.g. leaving all replicates of a certain sample or a specific subgroup out).
- When using an independent test set, often the rather wide confidence intervals for such an estimate are neglected. The SVM *Optimizer* also provides this crucial information. The corresponding Matlab function can also be used independently.
- Ability to inspect the model
- Ability to save and apply existing models

Note: This is version 1.0 of the software. - For efficient computation, the SVM optimizations are performed within a compiled C library ("MEX-file"). This library is provided for Windows 32 and 64 bit systems.

Introductory Price:

Single user commercial license Other licenses on request 300 € (+ 19 % VAT)

The price includes the next update which will include the option for double cross-validation, thus providing further validation for the chosen parameters.

Please refer to our SVM Starter Page for more information on SVMs!

