Support Vector Machines for Large Scale Text Mining in R

Ingo Feinerer\textsuperscript{1}  Alexandros Karatzoglou\textsuperscript{2}

\textsuperscript{1}Vienna University of Technology, Austria
\textsuperscript{2}Telefonica Research, Spain

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Motivation

- Machine learning and data mining require classification
- Large amounts of data
- Use R for data intensive operations
- Text mining is especially resource hungry
- Highly sparse matrices
- Need of scalable implementations
Large Scale Linear Support Vector Machines

Modified Finite Newton $l_2$-SVM

Given

- $m$ binary labeled examples $\{x_i, y_i\}$ with $y_i \in \{-1, +1\}$, and
- the SVM optimization problem

$$w^* = \arg\min_{w \in \mathbb{R}^d} \frac{1}{2} \sum_{i=1}^{m} c_i l_2(y_i w^T x_i) + \frac{\lambda}{2} \|w\|^2$$

the modified finite Newton $l_2$-SVM method gives an efficient primal solution.
R Extension Package svmlin

Features

Implements $l_2$-SVM algorithm.

- Extends original C++ version of svmlin by Sindhwani and Keerthi (2007).

Adds support for

- multi-class classification (one-against-one and one-against-all voting schemes),
- cross-validation, and
- a broad range of sparse matrix formats (SparseM, Matrix, slam).
R Extension Package svmlin

Interface

```r
model <- svmlin(matrix,
                 labels,
                 lambda = 0.1,
                 cross = 3)
```

- Regularization parameter of $\lambda = 0.1$
- 3-fold cross-validation
- model can be used with the predict() function
Text mining framework in R

- Functionality for managing text documents
- Abstracts the process of document manipulation
- Eases the usage of heterogeneous text formats (XML, ...)
- Meta data management
- Preprocessing via transformations and filters

Exports

- (Sparse) term-document matrices
- Interfaces to string kernels

Available via CRAN
Data

Reuters-21578
- News articles by Reuters news agency from 1987
- 21578 short to medium length documents in XML format
- Wide range of topics (M&A, finance, politics, ...)

SpamAssassin
- Public mail corpus
- Authentic e-mail communication with classification into normal and unsolicited mail of various difficulty levels
- 4150 ham and 1896 spam documents

20 Newsgroups
- 19997 e-mail messages taken from 20 different newsgroups
- Wide field of topics, e.g., atheism, computer graphics, or motorcycles
Preprocessing

Creation of term-document matrices

- 42 seconds for Reuters-21578
- 31 seconds for SpamAssassin
- 75 seconds for 20 Newsgroups

Term-document matrix size

- Reuters-21578: 65973 terms, 21578 documents, 24 MB
- SpamAssassin: 151029 terms, 6046 documents, 24 MB
- 20 Newsgroups: 175685 terms, 19997 documents, 46 MB
Protocol

Compare \( \text{SVM} \) implementations

- Runtime of \( \text{svm} \) (package \text{e1071}) vs. \text{svmlin}
- For \( \text{svm} \) we use a linear kernel and set the cost parameter to \( \frac{1}{\lambda} \)
- Initially sample \( \frac{1}{10} \) from data set for training
- Increase training data in \( \frac{1}{10} \) steps
- Compare classification performance using 10-fold cross-validation
Results

SpamAssassin
Results

20 Newsgroups

Portion of data used for training

Training time in seconds

e1071
svmlin
Conclusion

- svmlin extension package
- Takes advantage of sparse data
- Computations are done in primal space (no kernel necessary)
- Comparison with state-of-the-art svm
- Linear scaling, faster training times