Ovarian tumor subtype prediction using Gene Expression Data

Valentin Bernu 17/12

Plan

- 1. Data Exploration
- 2. Feature selection 1: Kruskal-Wallis test
- 3. Model benchmark
- 4. Feature selection 2: Sequential Feature Selection
- 5. Prediction & Evaluation

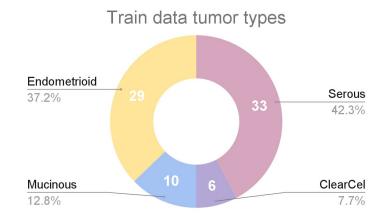
1. Data exploration

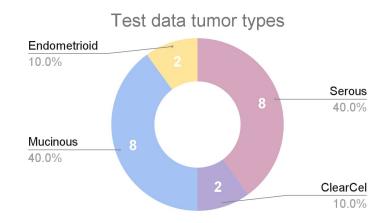
GSE6008 dataset

22286 columns

78 rows for training

20 rows for testing

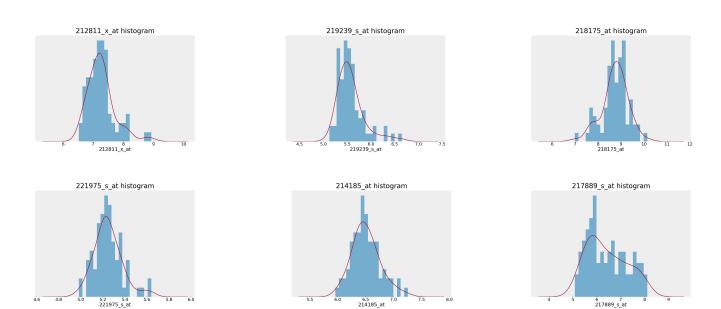




1. Data exploration

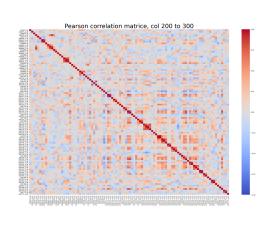
GSE6008 dataset

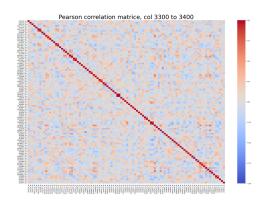
22283 gene expression columns

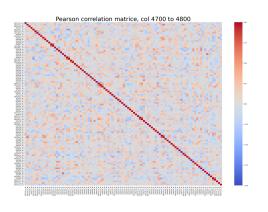


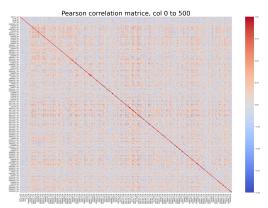
1. Data exploration

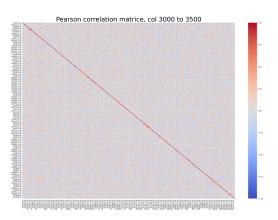
Correlation matrices

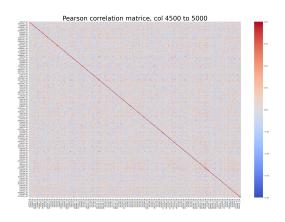








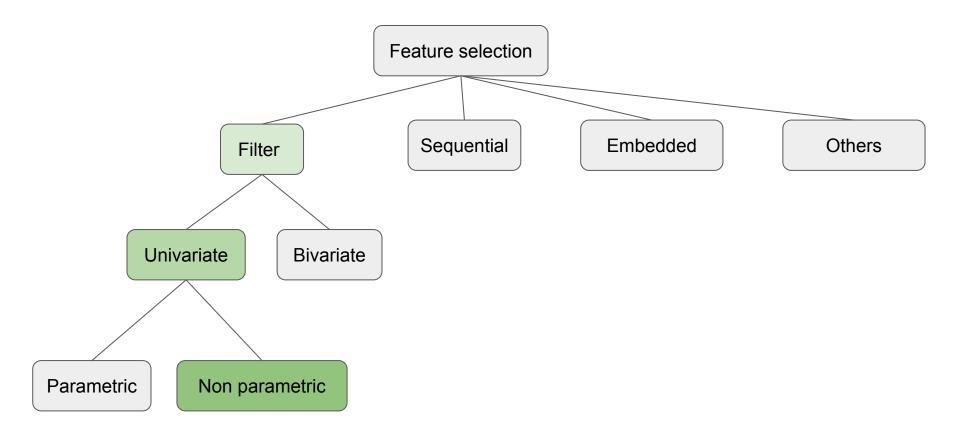




GSE6008 dataset

- Tumor types differently distributed in train and test data.
- Gene expression data present different distribution shapes, with different means and standard deviations.
- Normal distribution are frequent.
- Not strong correlation pattern detected in correlation matrices.

2. Feature Selection 1: Kruskal-Wallis test

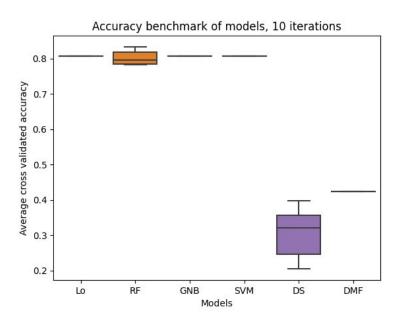


3. Model benchmark

- Logistic Regression (Lo)
- Random Forest (RF)
- Gaussian Naive Bayes (GNB)
- Support Vector Machines (SVM)
- Dummy Stratified (DS) -generates predictions by respecting the training set's class distribution.
- Dummy Most Frequent (DMF) -always predicts the most frequent label in the training set.

Method: 10 iterations with 3-fold cross validation

3. Model benchmark



4. Feature Selection 2: Sequential Feature Selection (SFS)

What is SFS?

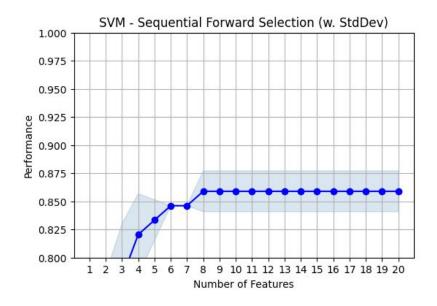
"At each stage, this estimator chooses the best feature to add or remove based on the cross-validation score of an estimator", from scikit-learn documentation.

Pros

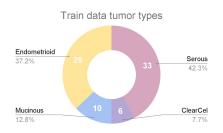
- Good performance
- Robust to redundancy

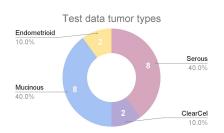
Cons

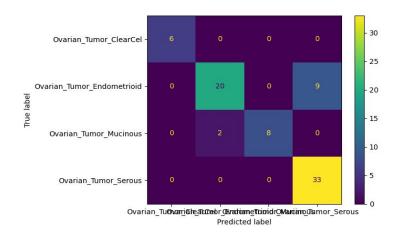
- Requires high computational power
- Can be trapped in local minima

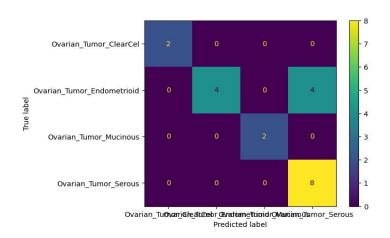


5. Performance evaluation









Future work directions

- Confidence score on predictions
- Study of selected features
- Class imbalance handling
- Regularization (embedded feature selection)

Thank you!