

Modelling strategies to combine multiple serum tumor biomarkers for early prediction of immunotherapy non-response in non-small cell lung cancer

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Introduction:

- Current practice still relies on PD-L1 testing to identify patients likely to respond to immune checkpoint inhibitor (ICI) therapy.
- Serum tumor markers (STMs) are known to reflect tumor activity and might therefore be useful in response prediction.

Aim:

To compare several methods in their ability to accurately predict non-response in NSCLC patients receiving ICI therapy by combining multiple sequentially measured STMs.

Methods:

- 412 NSCLC patients assigned to a training (75%) and validation (25%) cohort.
- Bi-weekly measurements of CYFRA, CEA, CA125, NSE, and SCC.
- 9 prediction methods: Logistic regression (LR), quadratic discriminant analysis (QDA), LASSO, random forest (RF), bagging, boosting, neural network (NN), support vector machines (SVM), recurrent neural network (RNN-GRU).
- 95% specificity in model training to assure a low false positive rate.
- 1000 bootstrap samples to assess diagnostic accuracy.



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Results:

	Training set	Validation set
Patients (n (%))	307 (74.5%)	105 (25.5%)
Mean age (years (SD))	63.7 (91.6%)	62.7 (10.1%)
Male sex (n (%))	159 (51.8%)	65 (61.9%)
Nivolumab (n (%))	272 (88.6%)	100 (95.2%)
Pembrolizumab (n (%))	35 (11.4%)	5 (4.8%)
Number of patients with PD at 6 months (n (%))	210 (68.4%)	71 (67.7%)
Mean survival after treatment start (days (SD))	232 (198)	255 (225)
Patients with biomarker measurements		
CYFRA (n (%))	306 (99.7%)	103 (98.1%)
CEA (n (%))	299 (97.4%)	101 (96.2%)
CA-125 (n (%))	305 (99.3%)	102 (97.1%)
NSE (n (%))	305 (99.3%)	102 (97.1%)
SCC (n (%))	258 (84.4%)	80 (76.2%)

Table 1: Description of the patient cohort used in this study. Progressive disease: PD, standard deviation: SD.

Best performance: Sensitivity & Specificity

• Training data:

Sensitivity: 79.5% / Specificity: 95.5%
Method: Boosting - CYFRA, CEA, CA125, NSE.

• Validation data:

Sensitivity: 68.8% / Specificity: 45.5%.
Method: QDA - CYFRA, CEA, CA125, NSE, SCC.

• Bootstrap average:

Sensitivity: 75.8% / Specificity: 92.7%
Method: Boosting - CYFRA, CEA, CA125, NSE.

Best performance: ROC curve

• Training data

AUC: 0.960
Method: Boosting – CYFRA, CEA, CA125, NSE.

• Validation data

AUC: 0.895
Method: Bagging – CYFRA, CEA, CA125, NSE, SCC.

Receiver Operating Characteristic curve: CYFRA & CEA & CA125 & NSE

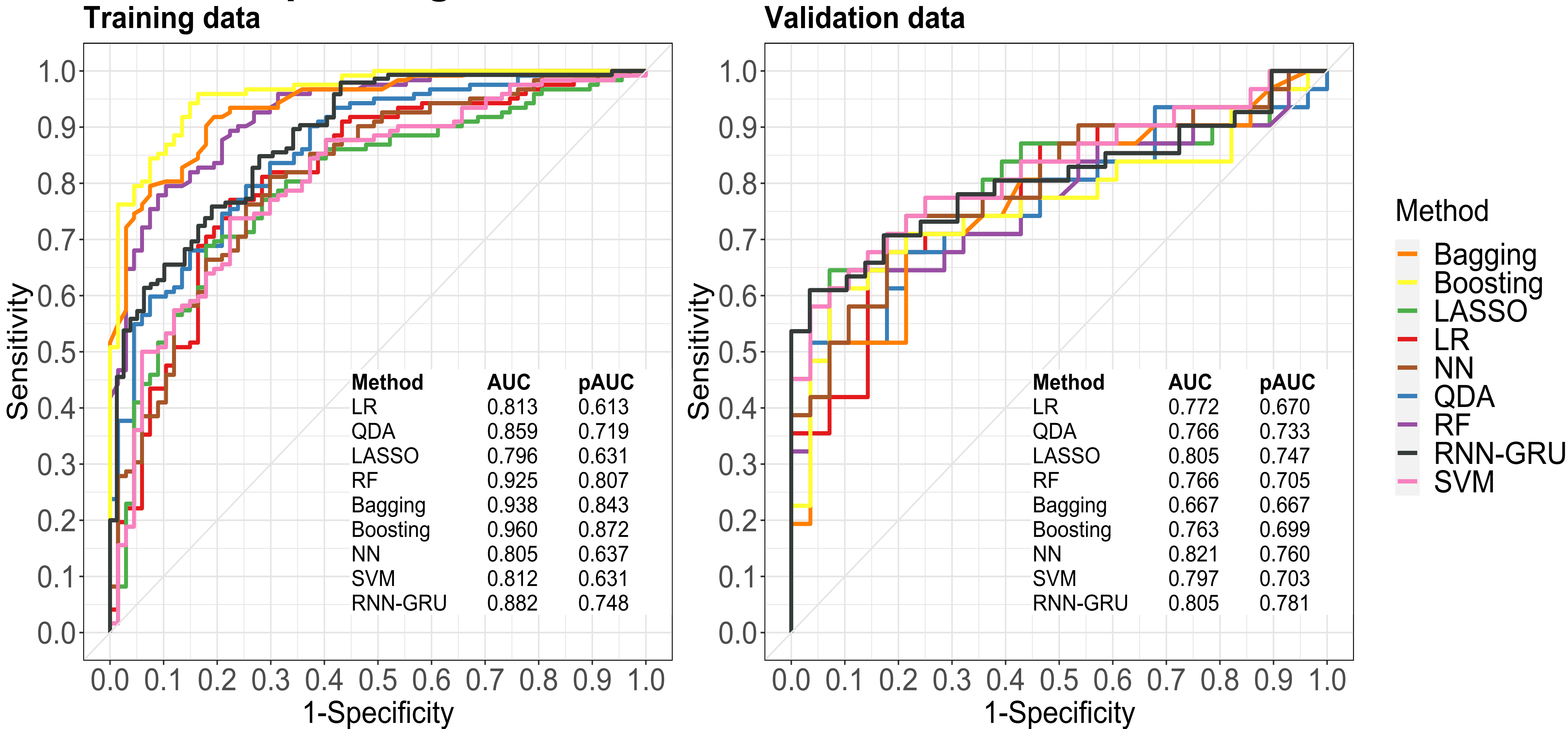


Figure 2: ROC curves for model training and validation on the STM combination of CYFRA, CEA, CA125, and NSE. The pAUC is calculated for the specificity range of 0.9 to 1.

Discussion:

- Increasing the number of STMs in the model leads to marginal gains and might result in a decrease in the specificity of the model.
- Sensitivity results are based on a 95% specificity during model training, thus narrowing the ROC curve to a single point. While the pAUC results indicate that multiple methods provide a good predictive performance.

Conclusion:

- Multiple sequentially measured STMs can be combined in a prediction model to predict ICI non-response in NSCLC patients.
- Boosting provided the best performance across all STM combinations included in this study.
- The Boosting model based on CYFRA, and CEA measurements should be subject of further evaluation.

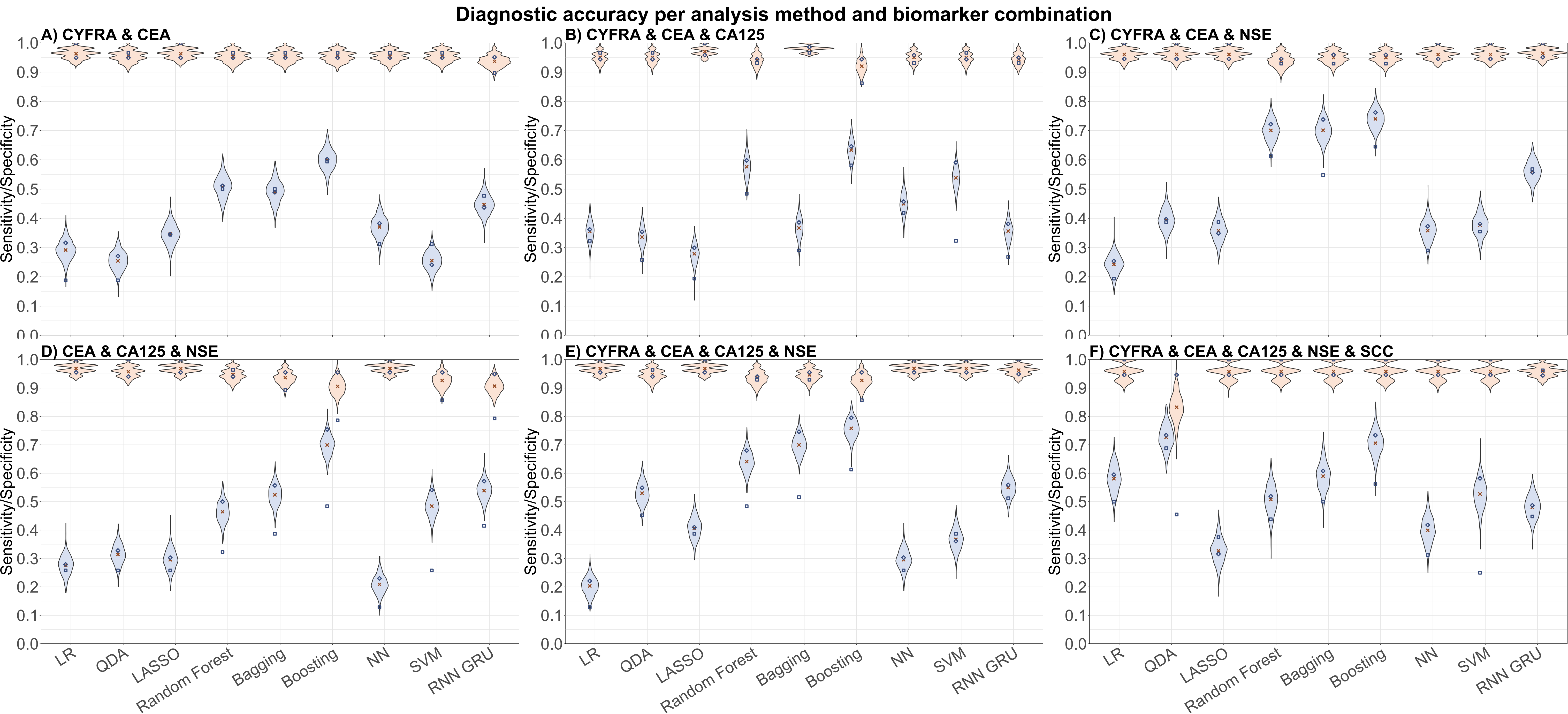


Figure 1: Sensitivity and specificity results found for model training, validation, and the bootstrap analysis.