

# Development and Validation of Predictive & Prognostic Biomarkers with High Dimensional Data

Noah Simon & Richard Simon

# Development and Validation of Predictive & Prognostic Biomarkers that Inform Treatment Decisions with High Dimensional Data

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# Kinds of Biomarkers

- Early detection
- Diagnostic
- Prognostic
- Predictive
- Endpoint
  - Pharmacodynamic, intermediate, surrogate

# Prognostic Biomarker

- Measured before treatment to indicate long-term outcome of patient without treatment or receiving standard treatment
- Can be used to identify which patients have such good prognosis on “minimal” treatment that they don't require more intensive regimens

# Predictive Biomarker

- Measured before treatment to indicate who is likely or unlikely to benefit from a particular treatment

# Kinds of biomarkers

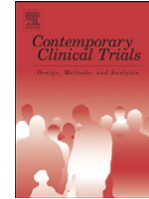
- Measurement of single analyte
- Scalar function of measurements of multiple analytes
  - $p > n$  or  $p < n$



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## Contemporary Clinical Trials

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# Overfitting in prediction models – Is it a problem only in high dimensions?



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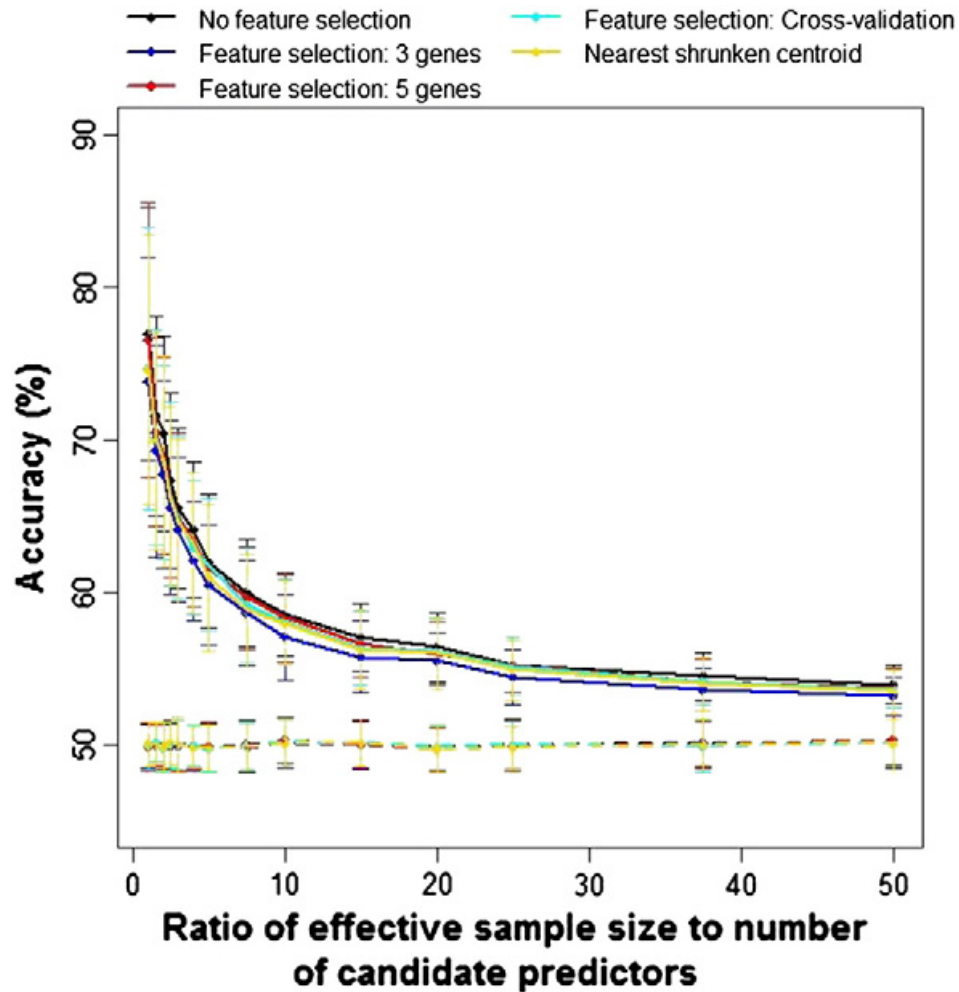
Clinical trials

Patient selection

### ABSTRACT

The growing recognition that human diseases are molecularly heterogeneous has stimulated interest in the development of prognostic and predictive classifiers for patient selection and stratification. In the process of classifier development, it has been repeatedly emphasized that in situations where the number of candidate predictor variables is much larger than the number of observations, the apparent (training set, resubstitution) accuracy of the classifiers can be highly optimistically biased and hence, classification accuracy should be reported based on evaluation of the classifier on a separate test set or using complete cross-validation. Such evaluation methods have however not been the norm in the case of low-dimensional,  $p < n$  data that arise, for example, in clinical trials when a classifier is developed on a combination of clinico-pathological variables and a small number of genetic biomarkers selected from an understanding of the biology of the disease. We undertook simulation studies to investigate the existence and extent of the problem of overfitting with low-dimensional data. The results indicate that overfitting can be a serious problem even for low-dimensional data, especially if the relationship of outcome to the set of predictor variables is not strong. We hence encourage the adoption of either a separate test set or complete cross-validation to evaluate classifier accuracy, even when the number of candidate predictor variables is substantially smaller than the number of cases.

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**Fig. 1.** Illustration of overfit in prediction models under the null. The dashed line represents the accuracy of classification as evaluated in the test set and the solid lines represent the accuracy of classification in the training set. The error bars represent  $\pm 1$  standard deviation.



# Prognostic Biomarkers

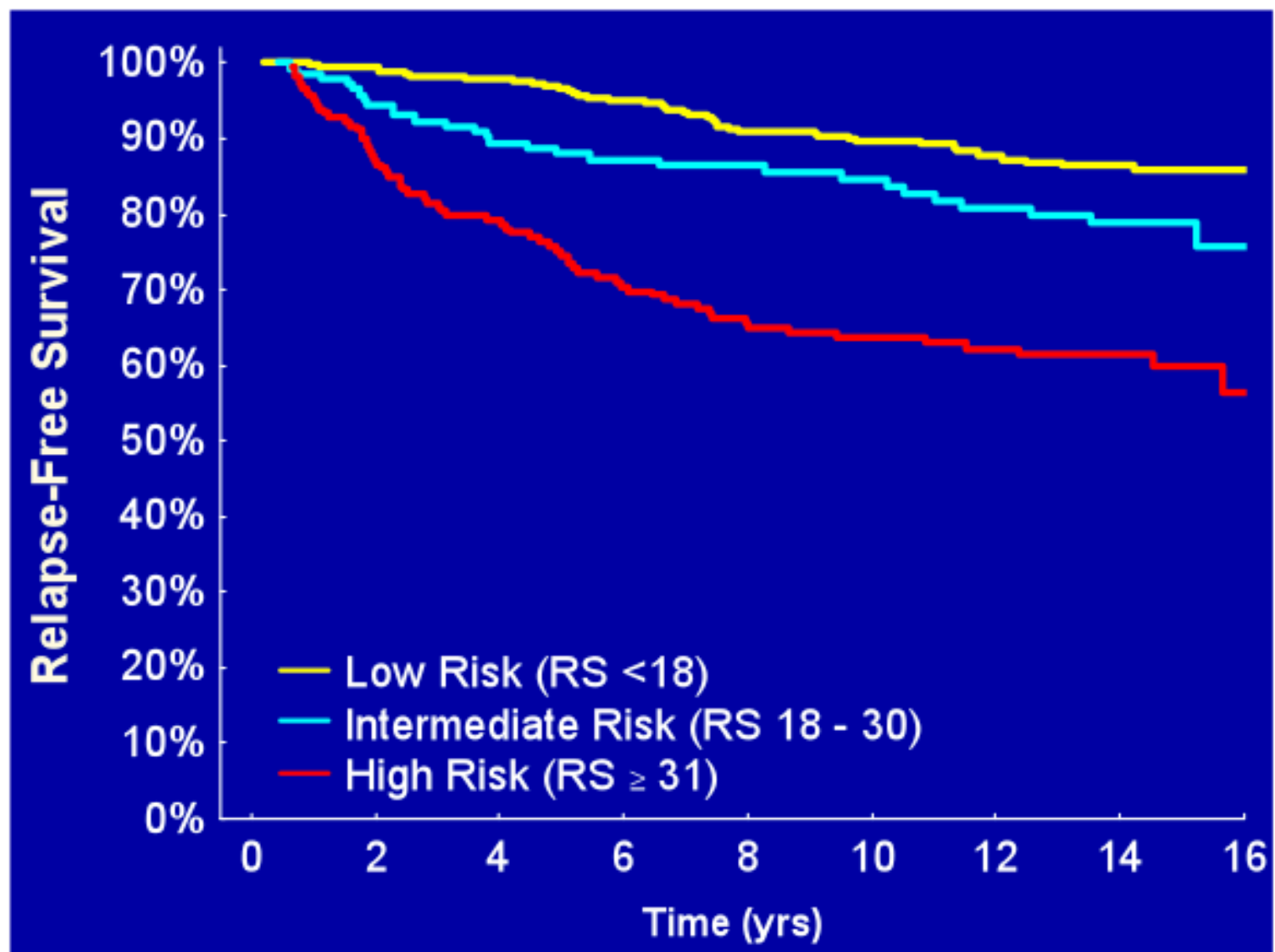
- Many prognostic studies do not develop models or biomarkers that inform treatment decisions
- They use heterogeneous convenience samples of cases
- The selection of cases and analysis are not driven by an intended use
- Over-emphasis on statistical significance and hazard ratios
- Biased estimates of prediction accuracy

# Oncotype DX

- Identify a subset of stage I, ER+ breast cancer patients who have such good outcome with only anti-estrogen therapy that they do not need chemotherapy

# Key Features of Oncotype DX Development

- Select cases with stage I, ER+ patients who received anti-estrogen therapy alone
- Analysis driven by objective of trying to identify a subset with such good outcome that they don't need other therapy
  - Not by FDR or what genes are significant or which model has the greatest separation of survival curves
- Separation of data used for model development from data used for validation



# Key Features of Oncotype DX Development

- Avoided problems of multiplicity by not doing any model development or tweaking on the data used for model validation
- Strong analytical validation

# Kinds of Validation

- Analytical validation
  - Does it accurately measure the analyte and is it reproducible
- Clinical validation
  - Does it correlate with some clinical feature like outcome or stage
- Medical utility
  - Is it actionable in a way that benefits the patient
  - Requires clarity on intended use

# Single Arm Study with Binary Response

- Pathologic complete remission following pre-operative chemotherapy for patients with locally advanced breast cancer
- PCr indicates treatment effect on tumor even without a control group
- Whether that study informs treatment decisions depends on context of what other effective regimens are available

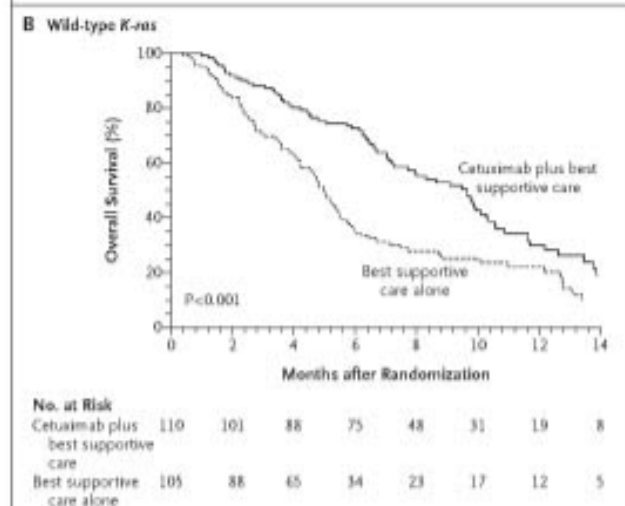
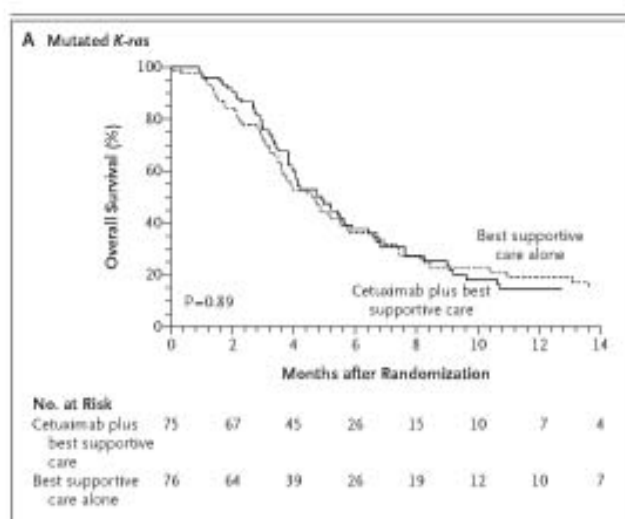
- So a prognostic biomarker can in some cases be established as useful for informing treatment decisions based on a single arm study with a survival (time-to-event) endpoint
  - Prospective clinical trial
  - “Prospective-retrospective” trial



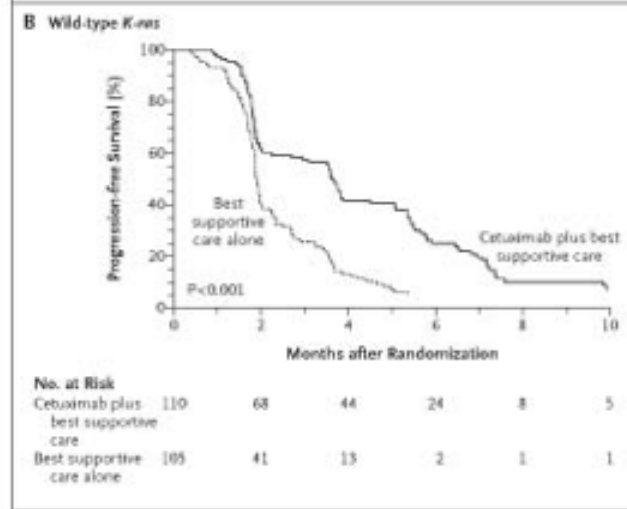
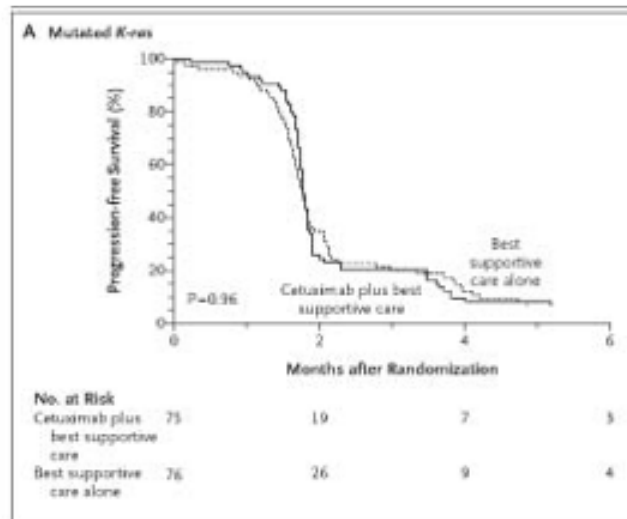
- Development of a predictive biomarker model that informs selection of a new treatment or control often requires data from an RCT

# Development and Validation of Predictive Biomarker Usually Requires RCT





**Figure 1. Kaplan-Meier Curves for Overall Survival According to Treatment.** Panel A shows results for patients with mutated K-ras tumors, and Panel B for patients with wild-type K-ras tumors. Cetuximab as compared with best supportive care alone was associated with improved overall survival among patients with wild-type K-ras tumors but not among those with mutated K-ras tumors. The difference in treatment effect according to mutation status was significant (test for interaction,  $P=0.01$ ).



**Figure 2. Kaplan-Meier Curves for Progression-free Survival According to Treatment.**

Panel A shows results for patients with mutated K-ras tumors, and Panel B for patients with wild-type K-ras tumors. Cetuximab as compared with best supportive care alone was associated with improved progression-free survival among patients with wild-type K-ras tumors but not among those with mutated K-ras tumors. The difference in treatment effect according to mutation status was significant (test for interaction,  $P<0.001$ ).