Design of Early Validation Trials of Biomarkers

Daniel Normolle, Ph.D.

Great Lakes-New England Clinical Epidemiological Center Early Detection Research Network

Department of Radiation Oncology Comprehensive Cancer Center Biostatistics Unit University of Michigan

1 Example

We have laboratory results of an interesting marker:

Twenty normals and twenty cancers of unknown provenance

Unblinded laboratory analysis

Two-standard deviation difference between samples

What is the next step?

1 Example (continued)

Weak idea for the design:

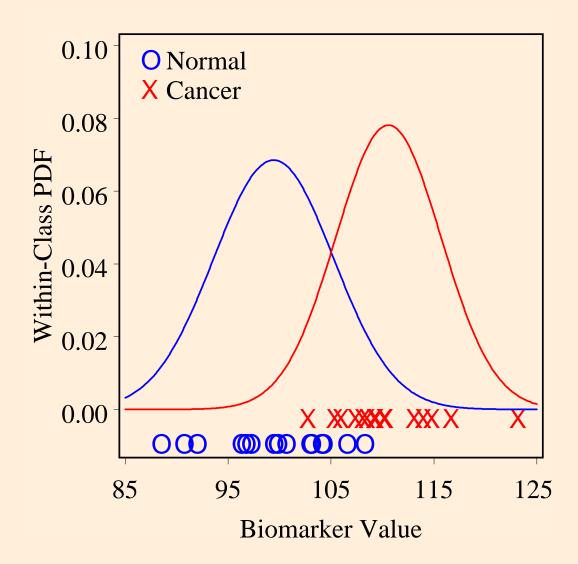
Twenty well-characterized cancers and normals

Two-sample t-test

80% power for two-standard-deviation difference

1 Example (continued)

Results of the trial:



1 Example (continued)

Why is the design weak?

Is a test that identifies cancers useful?

Two standard deviations may not be sufficient

Testing sample = training sample

Test sample is unblinded

How does the design relate to the decision?

2 Design Criteria

What are the goals of the marker?

Screen of an untested population

No other information about subjects

Potentially tens of millions of tests (e.g., PSA)

Screen of a clinical population

Demographic, risk, other marker data

Fewer tests, high-risk subjects

Combine with other markers to make a panel

2 Design Criteria (continued)

What are the goals of the study?

Feasibility/Technical

Further tweaking of assay

Characterize features of the assay

Identify threats to assay validity (e.g., sample processing)

Developmental/Pivotal

Estimate sensitivity and specificity

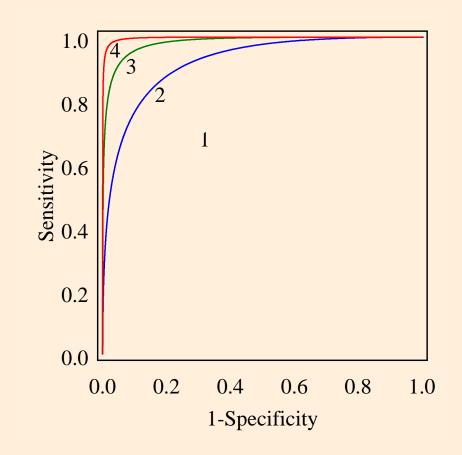
Verify the validity of the marker

Determine quality of reported signal (pathway activation in tissue)

3 Estimate Sensitivty and Specificity

What is sufficient sensitivity and specificity?

Example: ROC curves for normally distributed markers Number indicates separation in standard deviations



Minimum cost classification rule:

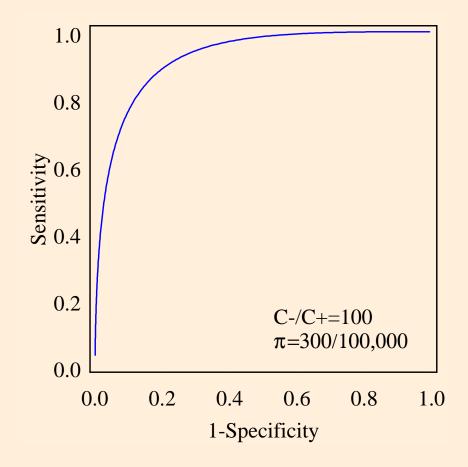
$$\text{If} \quad \frac{\hat{\mathbf{P}}(\boldsymbol{x}|+)}{\hat{\mathbf{P}}(\boldsymbol{x}|-)} > \frac{(1-\pi)}{\pi} \cdot \frac{C_{+}}{C_{-}} \quad \text{then} \quad \text{Classify } \boldsymbol{x} \in +$$

Otherwise

Classify $x \in -$

The Minimum cost point can be identified on the ROC curve:

$$ROC' = \frac{(1-\pi)}{\pi} \cdot \frac{C_+}{C_-}$$



Classification criteria (RHS of the minimum cost rule):

| | | | π | | |
|---------------|-----|-------------|-------------|------------|------------|
| | | 300/100,000 | 100/100,000 | 30/100,000 | 10/100,000 |
| C_{-}/C_{+} | 300 | 1.11 | 3.3 | 11.1 | 33.3 |
| | 100 | 3.3 | 9.99 | 33.3 | 99.99 |
| | 30 | 11.1 | 33.3 | 111.1 | 333.3 |
| | 10 | 33.2 | 99.9 | 333.2 | 999.9 |

Markers intended for population screens will require very high specificities

If the marker is intended to be used as a population screen, the minimum cost rule is appropriate, and will tend to recommend very high specificity

In clinical decision-making, risk reduction may be a more relevant criterion, so the bar may be set lower

In either case, a pivotal trial will recommend a marker for further development if it meets minimum criteria for sensitivity and specificity:

Determine minimum criteria for sensitivity and specificity

Design the trial as parallel single-sided tests of the null hypothesis of inadequate sensitivity and specificity

Power the tests appropriately

Using the training sample as the testing sample tends to overestimate the sensitivity and specificity

Split the testing and the training sample

Blind the testing sample

Cross-validation

Simulation Example: Generate 1000 data sets with 15 controls and 15 cases, one discriminating variable and five noise variables

| | | 1 | | 1+5 Noise | |
|--------|-----------------|-------|-------|-----------|-------|
| | | Sens. | Spec. | Sens. | Spec. |
| Method | Resubstitution | .696 | .694 | .767 | .772 |
| | Holdout | .683 | .694 | .630 | .638 |
| | Crossvalidation | .689 | .686 | .625 | .629 |

While there are systematic criteria for the size of the testing sample, no such standards exist for the training sample

4 Verify the validity of the marker

What are the biases in testing a potential screen or clincal marker?

Cases that differ from controls in a way that affects the test

Case and control samples that have been treated differently

Wrong type of cases (pre-cancers versus cancers)

Wrong type of disease

Wrong marker

5 Determine quality of reported signal

Tumor tissues will not be assayed in real markers

Ideally, markers signalling different pathways will be combined

If there are multiple potential pathways, and the sensitivity of a marker is 30%, is it because:

The pathway is only activated in 30% of cases, or

The activation signal in the tumor isn't detected in the tested product?

Estimate the sensitivity and specificity relative to activated tumor, rather than patients

Pivotal trial will accept a sample marker for use in a panel if it is sufficently senstitive and specific for the marker in target tissue

6 Summary

A pivotal study should produce results that directly inform the next decision to be made

Minimal criteria for sensitivity and specificity are the most likely for determining sample size

A marker intended to be used as a population screen must be very sensitive and extremely specific

Sensitivity and specificity should not be directly estimated from the training sample

A blinded testing sample is more convincing

As the dimensionality of the marker increases, the risk of over-fitting increases

A sample marker that indicates a single tissue pathway should be validated against that pathway