

Predictive values

Predictive value positive (PV+): $P(\text{disease} \mid \text{test}+)$

Predictive value negative (PV-): $P(\text{no disease} \mid \text{test}-)$

Example. A: mammogram positive, B: developing breast cancer in next 2 years

Suppose that 7% of the general population of women will have a positive mammogram.

What is the probability of developing breast cancer over the next 2 years among women in the general population?

$$P(\text{breast cancer} \mid \text{mammogram}+) = .1$$

$$P(\text{breast cancer} \mid \text{mammogram}-) = .0002$$

$$P(B) = P(\text{breast cancer})$$

$$= P(\text{breast cancer} \mid \text{mammogram}+)P(\text{mammogram}+) + P(\text{breast cancer} \mid \text{mammogram}-)P(\text{mammogram}-) = .1(.07) + .0002(.93) = 0.00719$$

$$PV+ = P(\text{breast cancer} \mid \text{mammogram}+) = .1$$

$$PV- = P(\text{no breast cancer} \mid \text{mammogram}-) = 1 - P(\text{breast cancer} \mid \text{mammogram}-) = 1 - .0002 = .9998$$

Sensitivity and specificity

Sensitivity of a symptom is the probability that the symptom is present given that the person has a disease = $P(\text{symptom} \mid \text{disease})$

Specificity of a symptom is the probability that the symptom is not present given that the person does not have a disease = $P(\text{no symptom} \mid \text{no disease})$

A false negative is defined as a person who tests out as negative but who is actually positive.

A false positive is defined as a person who tests out as positive but who is actually negative.

Bayes' Rule

Let A = symptom and B = disease. Then

$$PV+ = P(B \mid A) = \frac{P(A \mid B)P(B)}{P(A \mid B)P(B) + P(A \mid B^c)P(B^c)}$$

This can be written as

$$PV+ = \frac{sensitivity \times x}{sensitivity \times x + (1 - specificity) \times (1 - x)}$$

where $x = P(B)$ = probability of disease in the reference population.

Example: (Cancer) Suppose the disease is lung cancer and the symptom is cigarette smoking. If we assume 90% of people with lung cancer and 30 % of people without lung cancer are smokers, What is the sensitivity and specificity? Symptom: smoking, Disease: lung cancer

Sensitivity = $P(\text{symptom} \mid \text{disease}) = .9$

Specificity = $P(\text{no symptom} \mid \text{no disease}) = 1 - P(\text{symptom} \mid \text{no disease}) = .7$

Example: (Hypertension) Suppose 84% of hypertensive and 23% of normotensives are classified as hypertensive by an automated blood-pressure machine. What are the predictive value positive and predictive value negative of the machine, assuming 20% of the adult population is hypertensive? The sensitivity = $P(\text{symptom} \mid \text{disease}) = .84$ and specificity = $P(\text{no symptom} \mid \text{no disease}) = 1 - .23 = .77$. From Bayes rule $PV+ = (sensitivity \times x) / (sensitivity \times x + (1 - specificity) \times (1 - x))$

$PV- = (specificity \times (1 - x)) / (specificity \times (1 - x) + (1 - sensitivity) \times x)$

$PV+ = (.84)(.2) / [(.84)(.2) + (.23)(.8)] = .168 / .352 = .48$

$PV- = (.77)(.8) / [(.77)(.8) + (.16)(.2)] = .616 / .648 = .95$

An example in radiology

CT rating by radiologist

True Disease status	Definitely normal (1)	Probably normal (2)	Questionable (3)	Probably abnormal (4)	Definitely abnormal (5)	Total
Normal	33	6	6	11	2	58
Abnormal	3	2	2	11	33	51
Total	36	8	8	22	35	109

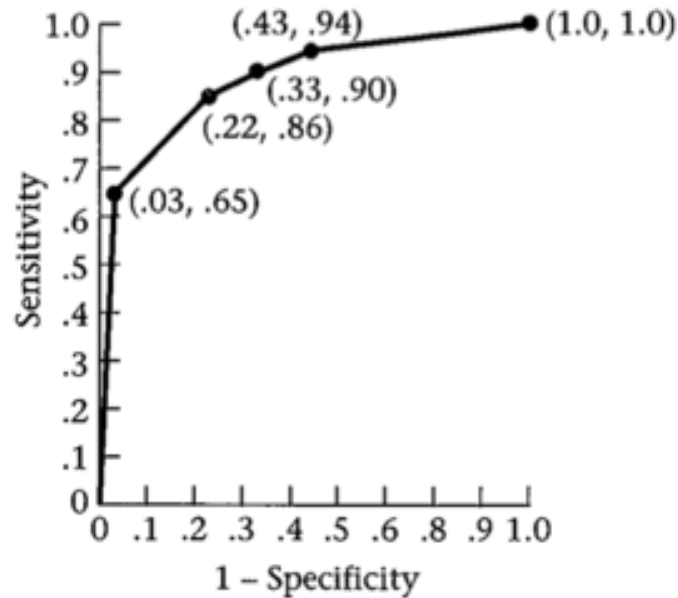
Table 1: Sensitivity vs. Specificity for different test criteria

Test positive criteria	Sensitivity	Specificity
1+	1	0
2+	0.94	0.57
3+	0.90	0.67
4+	0.86	0.78
5+	0.65	0.97
6+	0	1.0

Receiving operating characteristic (ROC) curve

ROC curve is a plot of the sensitivity versus (1-specificity) of a screening test, where the different points on the curve correspond to different cutoff points used to designate test positive.

Figure 3.7 ROC curve for the data in Table 3.4*



*Each point represents (1 - specificity, sensitivity) for different test-positive criteria.

Practical

<http://www.amstat.org/publications/jse/v13n2/datasets.kahn.html>