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Evaluation of diagnostic tests

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ABSTRACT

Measures of test efficacy are its repeatability or reliability and its validity. Validity is computed by comparing it against an older 'gold standard' test, which is supposed to unequivocally give the diagnosis. Sensitivity and specificity are inherent properties of a test. Predictive values give the probability of disease or no disease when the test result is known and vary with the prevalence of disease in the setting it is used. Receiver operator characteristic curves are used to determine the best cut off and also to choose between tests with numerical values. Likelihood ratios give the odds of disease at a particular test result.

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One of the most important tasks, a clinician has to perform when faced with a patient is that of reaching a diagnosis. Most astute clinicians use a thorough knowledge of literature along with a judicious use of diagnostic tests, good judgment, and a ready approach to organize the information. While newer diagnostic tests are continually coming into use, not much was said till recently about assessment of the test itself. The science of Clinical Epidemiology addresses the diagnostic process and the interpretation and evaluation of diagnostic data, both clinical and laboratory.

1. Measures of test efficacy

Most diagnostic testing concerns measurements. A good test is easy, inexpensive, safe and easily available. One of the attributes of a good test is its *reliability* or *repeatability*. Repeat measurements are likely to vary even in the same subjects. These differences increase the “noise” around any measurement and should be minimized. Validity of a test is an

expression of the degree to which it is supposed to measure, and how well it discriminates between diseased and non-diseased.

1.1. Probability of a diagnosis

Whenever we consider a diagnosis, we talk about probabilities. On the basis of the history and clinical examination, we already have an idea of the probability of a diagnosis (prior probability). This prior or pretest probability depends on clinical judgment and prevalence in the clinical setting in which the patient is seen. Above a certain level of probability, we should treat the patient and below a certain level we would not. In between these two levels, we would take the help of diagnostic tests. These tests are expected to move us along the scale of probability into either the “treat” or “not treat” zones (post-test probability) (Fig. 1).

How much the test affects probability of the diagnosis will depend in how good or discriminatory the test is.

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Fig. 1 – Probability of a diagnosis.

1.2. The gold standard

To study the validity of a test, we need to compare it with an older, more established test – the 'gold standard'. The gold standard test may be more difficult, risky, expensive or not easily available but it should unequivocally tell us whether the disease is present or absent. The gold standard result is equated with presence or absence of the disease in question. Sometimes an ideal gold standard is not available and a combination of tests, follow-up data or response to treatment, etc. are taken as the gold standard. For example, to study the accuracy of a new test for enteric fever, a combination of the Widal test and blood culture (either one or both) may be taken as the gold standard. The new test being evaluated and the gold standard test are done uniformly in all subjects in a blinded fashion. This would give us 4 groups of subjects as shown in the 2×2 table (Fig. 2).

1.3. Sensitivity and specificity

Two of the attributes of a test are its sensitivity and specificity. Sensitivity is the ability of a test to pick out those patients, who really have the disease. It is synonymous with positivity in disease (PiD) rates or true positivity. Specificity is the ability of a test to confirm that the disease is absent when it is truly absent and is synonymous with negative in health (NiH) rate or true negativity. Sensitivity and specificity can be derived by comparing the test with the "real answer" about whether the disease is present or absent in a simple 2×2 table given.

$$\text{Sensitivity} = a/a + c; \quad \text{Specificity} = d/b + d$$

Sensitivity and specificity are attributes of a test. Some tests are more sensitive, while others more specific. Very often there is a trade off between sensitivity and specificity. Sensitive tests are more useful for ruling out a diagnosis, if they are negative. They are used early in the diagnostic process such as in screening. A sensitive test is also used for dangerous but treatable conditions when it is important not to miss a single patient. A specific test is more useful for ruling in a diagnosis, if

it is positive and therefore more useful for final diagnosis. They are used when a false diagnosis or label would cause alarm. What degree of sensitivity/specificity one chooses also depends on the clinical situation and the trade offs involved. For example, one would like a highly sensitive test, when one is considering a diagnosis of bacterial meningitis as the condition is fatal if left untreated, and one would rather err on the side of treating them. Before conveying a diagnosis of say, malignancy one would want to be very sure and therefore use a very specific test.

2. Predictive values

Sensitivity and specificity are good for describing a test but in a clinical situation, the clinician has the test result and wants to know whether the disease is present or not. For this purpose, he needs to know its predictive value.

Positive predictive value (PPV) is the proportion of those with a positive test, who actually have the disease, i.e. $\text{PPV} = \text{true positive}/\text{total positives}$. From the 2×2 table above, $\text{PPV} = a/a + b$.

Negative predictive value (NPV) is the proportion of those, who test negative and are actually non-diseased, i.e. $\text{NPV} = \text{true negatives}/\text{total negatives} = d/c + d$.

Predictive values are thus calculated horizontally.

It is important to understand, however, that predictive value of a positive or negative test is not constant but changes with the prevalence of the disease in the situation, in which the test is used. Predictive values can be increased or decreased by choosing the type of patients to be tested. For example, PPV of amniotic fluid alpha fetoprotein (AFP) for neural tube defect can be increased by doing it in mothers with high serum AFP. This in effect means increasing the prevalence or pretest probability of the condition in the tested population. PPV will also decrease greatly, if the test is used for screening in the community. PPV also equals the post-test probability mentioned above.

3. Tests with numerical values

The above discussion was centered on tests, which gave a positive or a negative answer, for example, biopsy findings of Reid Sterberg cells in Hodgkin's disease or Western Blot assay in HIV. Many laboratory tests give the answer in numerical values on a continuous scale. It is often up to the clinician to decide the cut-off he chooses, above or below which the measurement is considered abnormal. Once a cut-off point is chosen, the continuous data are converted into dichotomous positive or negative, and the sensitivity and specificity at that cut-off can be calculated as above. One can also calculate sensitivity and specificity at a range of cut-off values. Usually

	Gold Standard +ve	Gold Standard –ve
Test +ve	a	b
Test –ve	c	d

Fig. 2 – 2×2 table. a – patients with disease and a positive test; b – patients without disease but a positive test; c – patients with disease but a negative test; d patients without disease and a negative test.

Fig. 3 – ROC curve.

there is a trade off between sensitivity and the specificity as the cut point changes, i.e. sensitivity decreases as specificity increases and vice versa.

4. Receiver operator characteristic (ROC) curves

These are the curves used to decide the best cut off for a test and also to compare 2 tests. To make an ROC curve, one calculates specificity and sensitivity over different cut points. Then one plots the sensitivity on the Y-axis and 1 – specificity on the X-axis to get several points on the graph. The point nearest the left upper corner is the best cut-off point (Fig. 3). When comparing two tests, the test with larger area under the curve is the better (more discriminatory) test and the test with a curve nearer to a diagonal line is a less useful one. However, as already stressed, the clinical significance of a particular diagnosis would often dictate the cut-off or test chosen.

5. Likelihood ratios

When one chooses a cut-off for a test that gives numerical values, there is a danger of losing important information. This is because one often does not consider how abnormal (high or low) the test reading was. A very abnormally high or low value would naturally be expected to be more significant than a borderline one. Likelihood ratios (LR) are a way of getting over this difficulty. A positive likelihood ratio (LR+) expresses in a single value the odds that a particular test result is likely to come from a patient with the disease. LRs can be calculated at different cut-offs and also over a range of readings.

$$\begin{aligned} \text{LR + ve} &= \frac{\text{Proportion of subjects with disease having that result}}{\text{Proportion of subjects without disease having that test result}} \\ &= a/a + c \div b/b + d \\ &= \text{Sensitivity}/1 - \text{Specificity} \end{aligned}$$

For a test with a continuous value, LR+ve can be calculated at any value of the test result by dividing the probability of that test result in patients with disease by the probability of the same test result in patients without disease.

LRs, since they are calculated vertically, do not change with prevalence of disease in the population being tested.

Also, Pretest odds $\times$ LR = Post-test odds.

Odds can be calculated by the formula: Odds = Probability/1 – Probability.

6. Clinical prediction rules

The principles concerning diagnostic tests can be applied to clinical data or a constellation of clinical findings as well. For example, one could talk about sensitivity and specificity of fever + petechiae + palpable spleen in acute leukemia. Based on this, one may go a step further and develop a 'clinical prediction rule' for a particular diagnosis. In fact, clinical prediction rules are far more cost-effective and possibly less erroneous than sophisticated laboratory tests since the latter often involve several steps leaving room for error at each step.

7. Sample size in diagnostic test evaluation studies

As these are also a form of prevalence (cross-sectional) studies, sample size is calculated as below:

$$N = \frac{z_{\alpha}^2 P(1-P)}{W^2}$$

where P is the desired proportion (sensitivity or specificity), z_{α} is the standard normal deviate for a 2 sided α and W is the desired total width of the confidence interval.

Steps in conducting a study on diagnostic test evaluation:

1. The investigator focuses on a new test which needs evaluation.
2. He looks up relevant literature on the test and possible gold standard test to be used and decides on the gold standard, which must be internationally acceptable.
3. He then decides how and where he will enroll subjects (study population). Generally subjects should be those, in whom the test is likely to be used. For example, if it is a test for typhoid fever, subjects may be patients with fever reporting at a clinic. Inclusion and exclusion criteria need to be decided.
4. Sample size calculations can be based on desired sensitivity and specificity (whichever gives higher sample size) using the formula for prevalence studies. He also needs to consider, whether enough subjects with and without the disease in question in the projected time frame and whether it will be feasible for all to have the new test and the gold standard test done. Data Collection Forms, informed consent forms, etc. should be designed.
5. He then starts data collection and performs the new test as well as gold standard test in all enrolled subjects in a uniform and blinded fashion.
6. Results are analyzed from the 2×2 table. Sensitivity, specificity and predictive values are calculated.
7. If the test has a numerical value, ROC curve can be constructed for choosing the best cut off and giving area under the curve.
8. Finally, likelihood ratios can be calculated for different values of the test result.

Conflicts of interest

The author has none to declare.

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