

**Evidence-Based Medicine
Diagnosis**

Component 2 / Unit 5

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Using EBM to assess questions about diagnosis

- Diagnostic process involves logical reasoning and pattern recognition
- Consists of two essential steps
 - Enumerate diagnostic possibilities and estimate their relative likelihood, generating *differential diagnosis*
 - Incorporate new information from *diagnostic tests* to change probabilities, rule out some possibilities, and choose most likely diagnosis
- Two variations on diagnosis also to be discussed
 - Screening
 - Clinical prediction rules

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Diagnostic (un)certainty can be expressed as probabilities

- Probability is expressed from 0.0 to 1.0
 - Probability of heads on a coin flip = 0.5
- Alternative expression is odds
 - Odds = Probability of event occurring / Probability of event not occurring
 - Odds of heads on a coin flip = $0.5/0.5 = 1$
- Rolling a die
 - Probability of any number = $1/6$
 - Odds of any number = $1/5$

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Assessing the diagnostic value of a test

| | Disease present | Disease absent | Total |
|---------------|--------------------------|-----------------------------|--------------------------------|
| Test positive | True positive (TP) | False positive (FP) | All with positive test (TP+FP) |
| Test negative | False negative (FN) | True negative (TN) | All with negative test (FN+TN) |
| Total | All with disease (TP+FN) | All without disease (FP+TN) | All (TP+FN+FP+TN) |

Sensitivity and specificity

- Sensitivity or true positive rate (TPR) is proportion of patients with disease who have positive test
 - $TPR = TP / (TP + FN)$
 - "Positivity in disease" (PID)
 - Better at ruling out disease - SnNout
- Specificity or true negative rate (TNR) is proportion of patients without disease who have negative test
 - $TNR = TN / (TN + FP)$
 - "Negativity in health" (NIH)
 - Better at ruling in disease - SpPin

Other statistics to calculate disease probability

- Prevalence of disease is the proportion of people with a disease
 - $Prevalence = Total\ with\ disease / Total$
 - $Prevalence = (TP + FN) / (TP + FN + FP + TN)$
 - Can be a good starting point for pre-test probability of disease
- Likelihood ratio positive measures how many times more likely test is positive in disease
 - $LR+ = Sensitivity / (1 - Specificity) = TPR / FPR$
- Likelihood ratio negative measures how many times more likely test is negative in health
 - $LR- = (1 - Sensitivity) / Specificity = FNR / TNR$

Calculating probability of disease with Bayes' Theorem

- Need pre-test probability
 - Can be prevalence, known risk, or estimate
- Convert to pre-test odds
 - Pre-test odds = (pre-test prob)/(1-pre-test prob)
- Calculate post-test odds with LR
 - Post-test odds = pre-test odds * LR+ (or LR-)
- Convert to post-test probability
 - Post-test prob = post-test odds/(1+post-test odds)

Other statistics – predictive value

- Predictive value positive is proportion of people with positive test who have disease
 - $PV+ = TP/(TP+FP)$
 - Note: Sensitivity is $TP/(TP+FN)$
- Predictive value negative is proportion of people with negative test who do not have disease
 - $PV- = TN/(TN+FN)$

Limitations of diagnostic tests

- There are many real and potential sources of bias (Whiting, 2004)
- Diagnostic tests improve probability estimations of diseases but still require diagnostic judgment
- Results can be influenced by pre-test probability estimation that may be incorrect
- Bayes' Theorem assume conditional independence of all data, but clinicians often have multiple pieces of evidence and tests, which can
 - Make calculations very complex
 - Violate assumptions of independence

Example of Bayes' Theorem application

- Detection of colonic polyps >6 mm in size in patients at high risk for colon cancer (Rockey, 2005)
- Air-contrast barium enema (ACBE)
 - Sens = 41%, spec = 82%, LR+ = 2.28, LR- = 0.72
- Computer colonic tomography
 - Sens = 55%, spec = 89%, LR+ = 5.00, LR- = 0.51
- Colonoscopy
 - Sens = 99%, spec = 99.6%, LR+ = 248, LR- = 0.01
- Number of polyps detected in this study of 614 patients was 155 patients with 234 lesions, 152 of which were adenomas or cancer

Comparing ACBE and colonoscopy

- Estimate pre-test probability = 0.25
- Pre-test odds = $0.25 / (1 - 0.25) = 0.33$
- With positive tests
 - ACBE
 - Post-test odds = $0.33 * 2.28 = 0.75$
 - Post-test probability = $0.75 / (1 + 0.75) = 0.43$ (43%)
 - Colonoscopy
 - Post-test odds = $0.33 * 248 = 81.8$
 - Post-test probability = $81.8 / (1 + 81.8) = 0.99$ (99%)

Comparing ACBE and colonoscopy

- With negative tests
 - ACBE
 - Post-test odds = $0.33 * 0.72 = 0.24$
 - Post-test probability = $0.24 / (1 + 0.24) = 0.19$ (19%)
 - Colonoscopy
 - Post-test odds = $0.33 * 0.01 = 0.0033$
 - Post-test probability = $0.0033 / (1 + 0.0033) = 0.0033$ (0.3%)

Screening tests for disease

- “Identification of unrecognized disease”
- Aim to keep disease (or complications) from occurring (1° prevention) or stop progression (2° prevention)
- Requirements for a screening test
 - Low cost
 - Intervention effective
 - High sensitivity – do not want to miss any cases; usually follow up with test of high specificity

Americans love screening tests despite lack of evidence

- Despite their limitations, screening tests for cancer are very popular with Americans (Schwartz, 2004)
- But cost of FP tests is substantial; in one study of screening for prostate, lung, colorectal, and ovarian cancer (Lafata, 2004)
 - 43% of sample had at least one FP test
 - Increased medical spending in following year by over \$1000
- Despite lack of evidence for benefit of Pap smear in women with hysterectomy, procedure is still widely done (Sirovich, 2004)
- Despite lack of evidence for benefit of annual physical exam, two-thirds of physicians still believe it is necessary (Prochazka, 2005)

Clinical prediction rules

- Use of results of multiple “tests” to predict diagnosis
- Best evidence establishes rule in one population and validates in another independent one
- Examples of clinical prediction rules
 - Predicting deep venous thrombosis (DVT) (Wells, 2000; Wells, 2006)
 - High sensitivity, moderate specificity
 - Better for ruling out than ruling disease
 - Coronary risk prediction – newer risk markers do not add more to known basic risk factors (Folsom, 2006)
