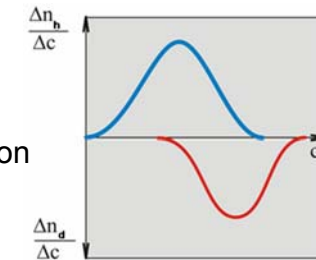
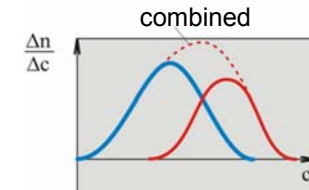
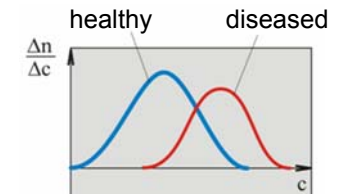


KAD 2013.11.26

Overlapping distributions

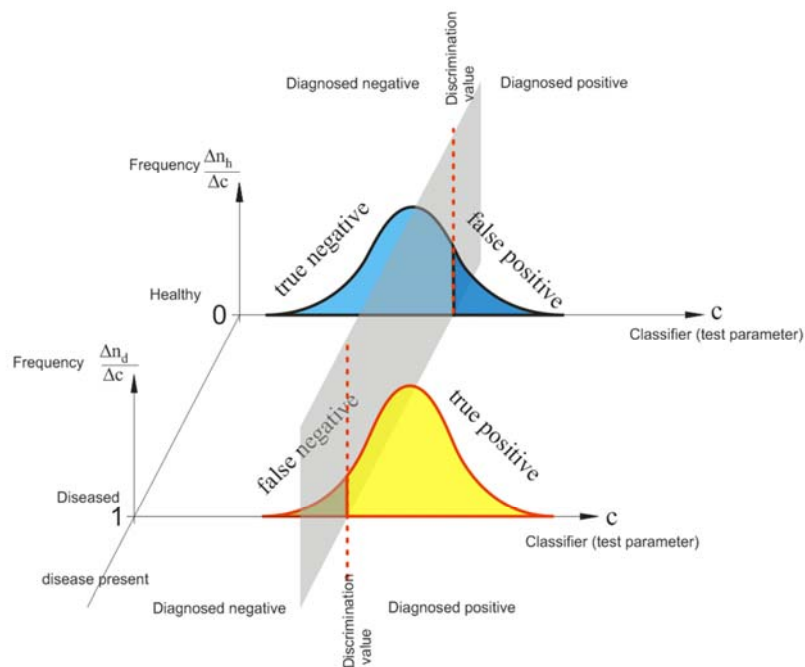
assumption:

a classifier value
(e.g. serum concentration)
changes (e.g. increases)
in diseased subpopulation



novel
representation

2

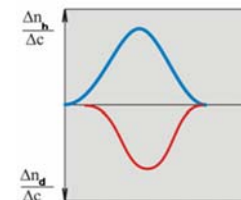


3

Based on overlap magnitude:

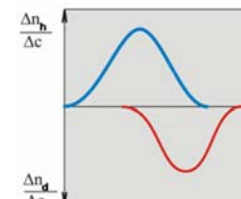
full
overlap

useless method



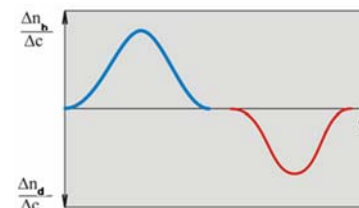
partial
overlap

real-life situation



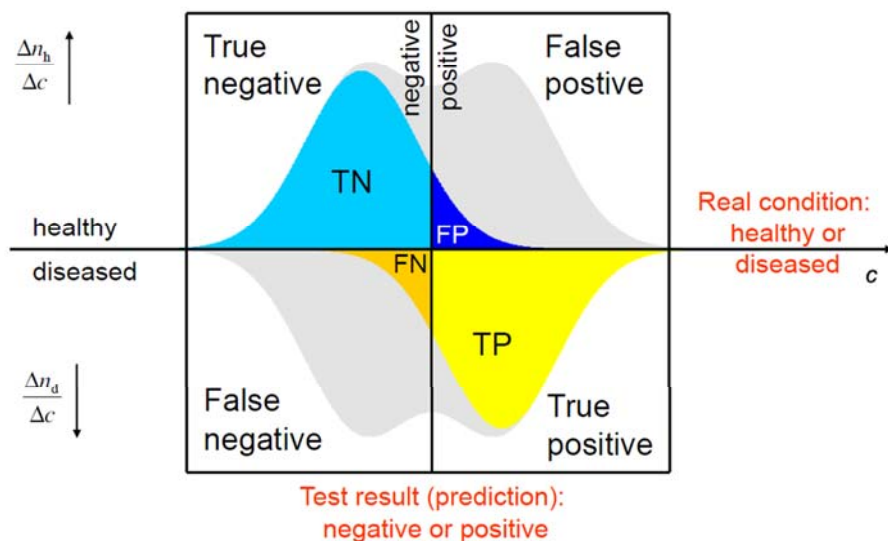
complete
separation

perfect method



4

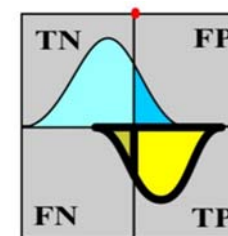
Contingency table: Confusion matrix (binary classification)



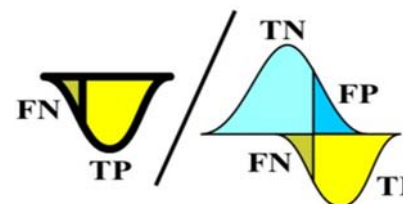
5

Prevalence

$$(w = \frac{de - sp}{se - sp})$$



Prevalence: measure of how common the disease is; frequency of diseased in examined population, probability prior to test (a-priori-probability),

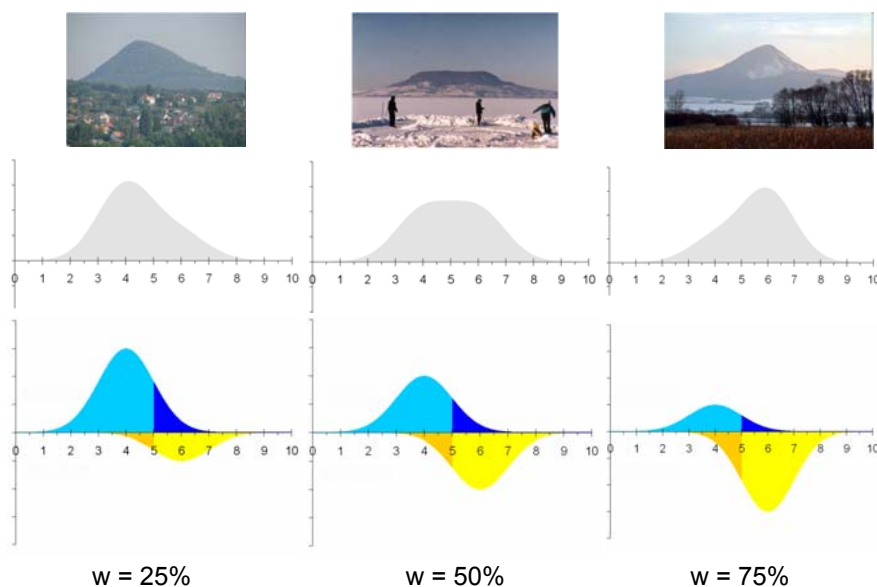


$$w = \frac{TP + FN}{TP + TN + FN + FP}$$

$$w = \frac{\text{diseased}}{\text{total}}$$

6

Shape of combined distributions



7

Parameters of diagnostic „goodness”

The goodness of a test can be described in terms of the following diagnostic parameters

Sensitivity

Specificity

PPV, relevance

NPV, segregation

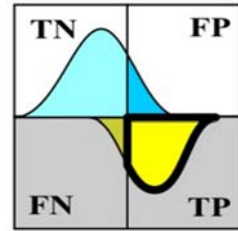
Every method must be compared with a reference-method (gold standard)

Gold standard: method known to always work; often autopsy

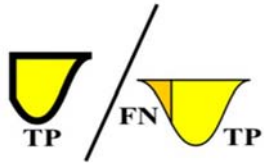


8

Diagnostic sensitivity (se)



Probability that the test finds the diseased positive; positive within diseased

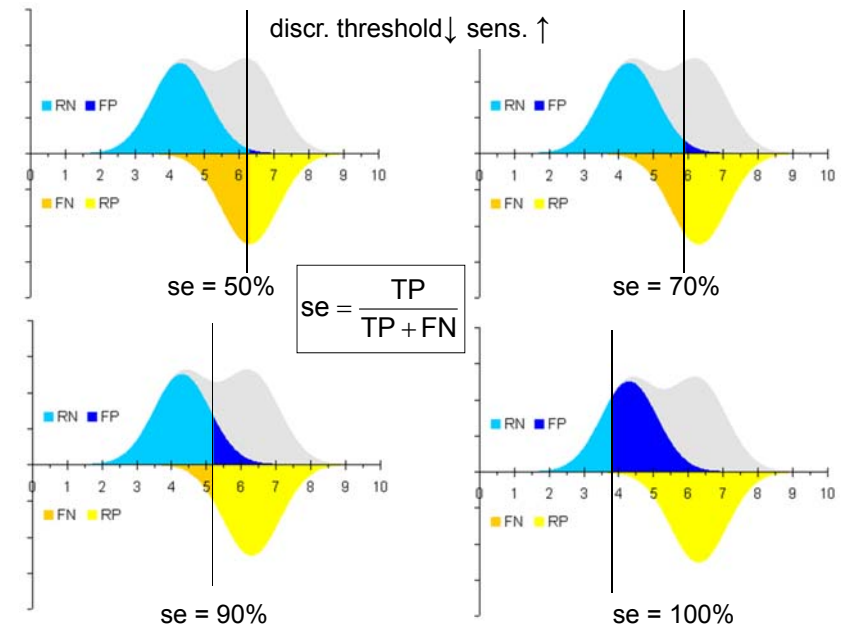


$$se = \frac{\text{true positive}}{\text{diseased}} = \frac{TP}{TP + FN}$$

Large-sensitivity tests are required:

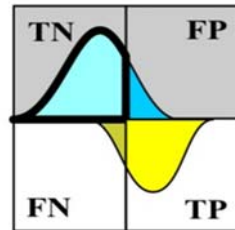
In early diagnosis (screening) so that few patients remain unrecognized.
If the risk of disease is greater than the risk of treatment.

9

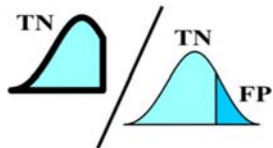


10

Diagnostic specificity (sp)



Probability that the test finds a healthy negative; negative among healthy

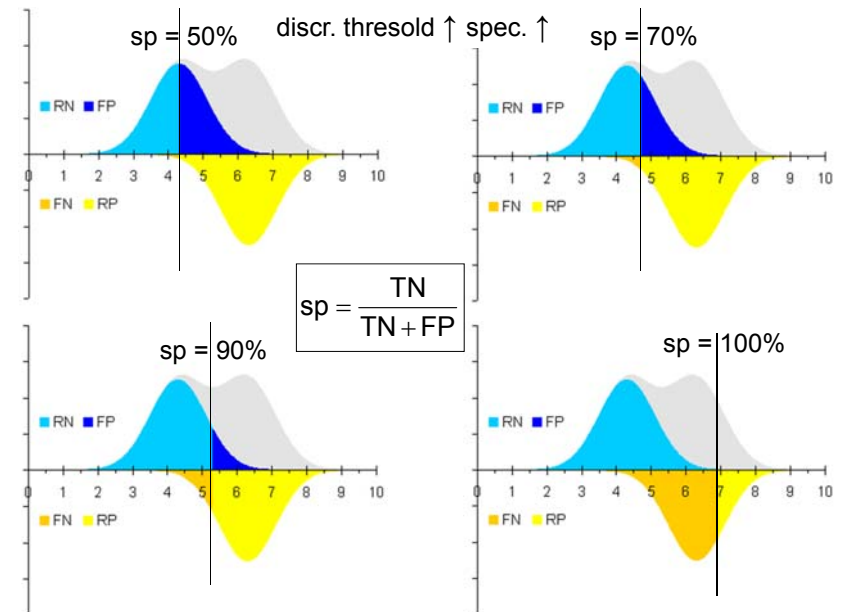


$$sp = \frac{\text{true negative}}{\text{healthy}} = \frac{TN}{TN + FP}$$

High-specificity tests are important:

When the false positive values have severe consequences (e.g. surgery).
When the risk of treatment is greater than the risk of disease.

11

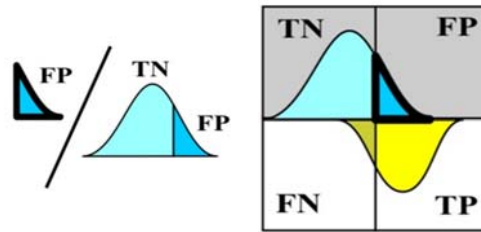


12

Diagnostic False Positive Rate

$$1 - sp = \frac{FP}{TN + FP}$$

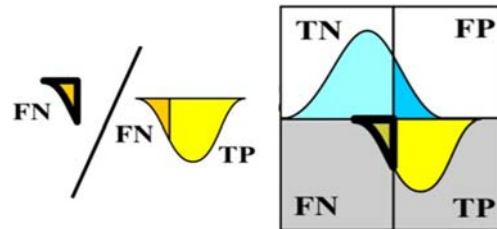
(Type-I error)



Diagnostic False Negative Rate

$$1 - se = \frac{FN}{FN + TP}$$

(Type-II error)



13

Predictive values (vertical rates)

a-posteriori-probabilities; they depend on prevalence

Diagnostic relevance

PPV,
positive predictive value

$$PPV = \frac{se \cdot w}{se \cdot w + (1 - sp) \cdot (1 - w)}$$

probability of disease if test is positive,
diseased among positive



$$PPV = \frac{\text{true positive}}{\text{positive}} = \frac{TP}{TP + FP}$$

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Diagnostic segregation

NPV,
negative predictive value

$$NPV = \frac{sp \cdot (1 - w)}{sp \cdot (1 - w) + (1 - se) \cdot w}$$

probability of healthiness if test is negative;

healthy among positives

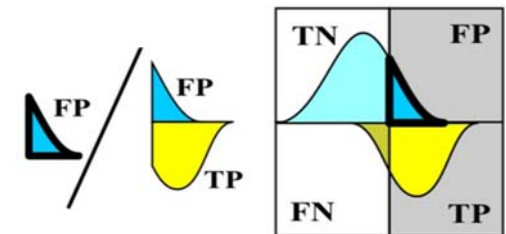


$$PVN = \frac{\text{true negative}}{\text{negative}} = \frac{TN}{TN + FN}$$

15

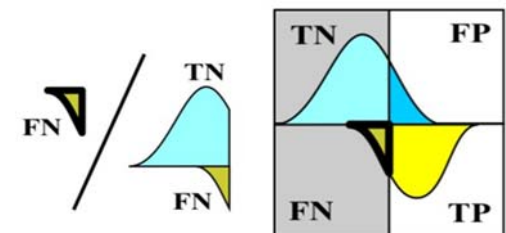
False alarm rate

$$1 - PPV = \frac{FP}{FP + TP}$$



False reassurance rate

$$1 - NPV = \frac{FN}{FN + TN}$$



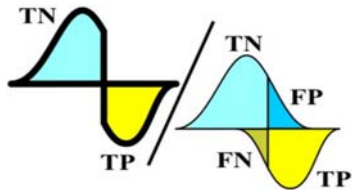
16

Diagnostic efficacy/efficiency

accuracy

correct classification rate

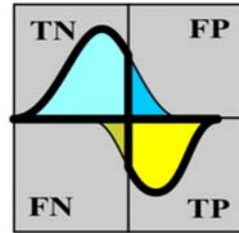
$$de = se \cdot w + sp \cdot (1 - w)$$



$$de = \frac{\text{true positive} + \text{true negative}}{\text{healthy} + \text{diseased}}$$

$$de = \frac{TP + TN}{TN + FP + FN + TP}$$

often: discrimination threshold is chosen so that de is maximized



Effect of prevalence

Case 1: $w = 50\%$

		NPV = 90%		
Case 1: w = 50%		test		
		negative	positive	
SP = 90%	Gold standard	healthy	90	10
		diseased	10	90

Sensitivity (TPR) = 90%

(ACC, de = 90%) Precision, PPV = 90%

Case 2: $w = 10\%$

Case 2: $w = 10\%$

NPV = 99%

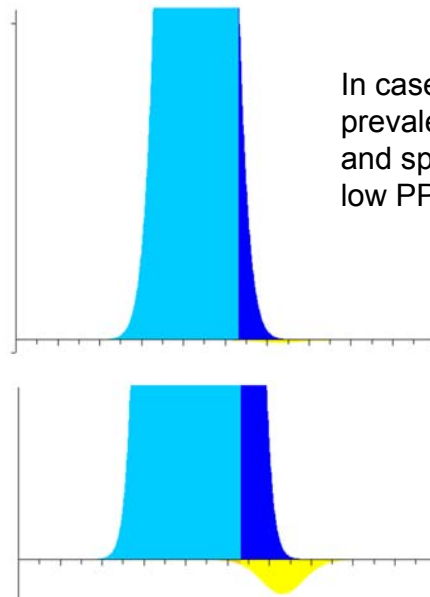
		test		
		negative	positive	
SP = 90%	Gold standard	healthy	810	90
		diseased	10	90

Sensitivity (TPR) = 90%

(ACC, de = 90%) PPV = 50%

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Effect of prevalence



In case of very small prevalence a highly sensitive and specific test could be of low PPV.

prevalence = 0.1 %

sensitivity = 98 %

specificity = 98 %

PPV = 4 %

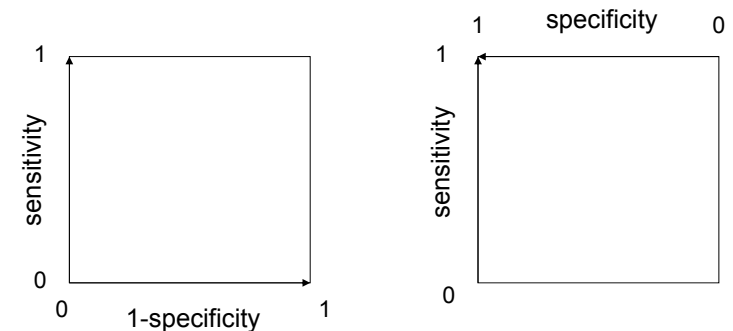
19

Comparison of diagnostic tests: the ROC space

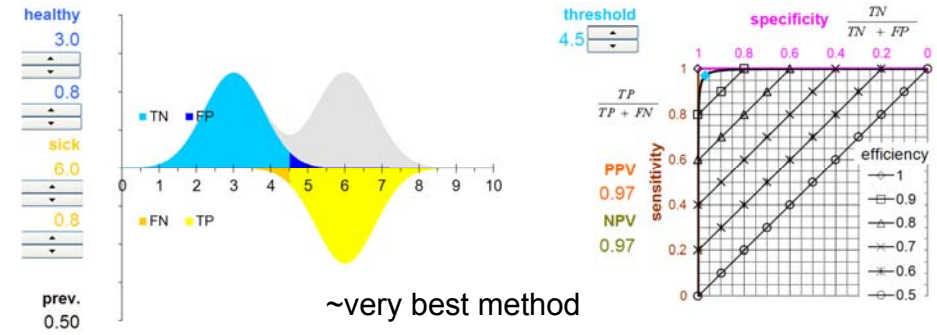
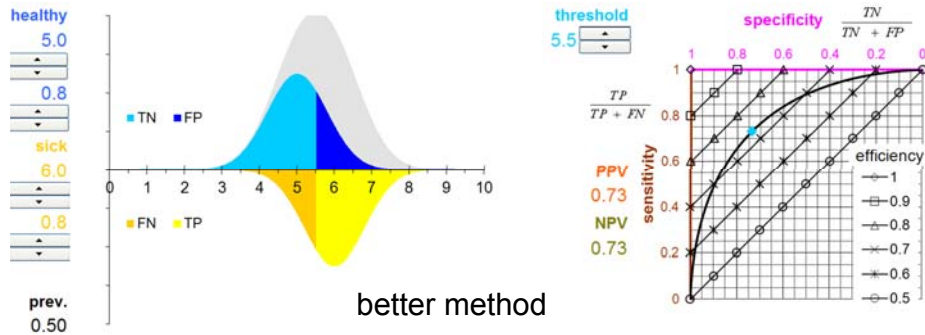
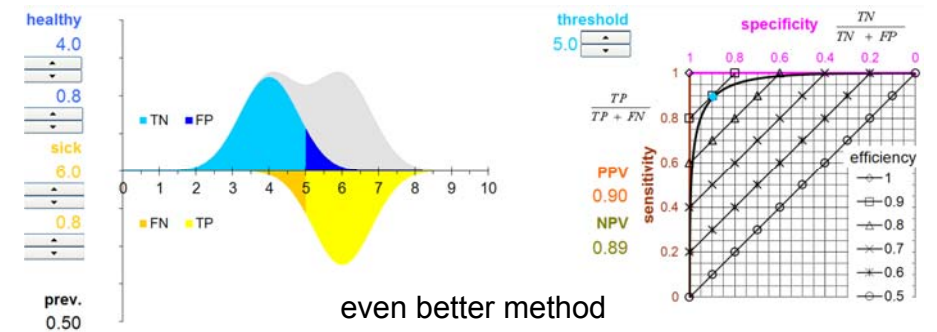
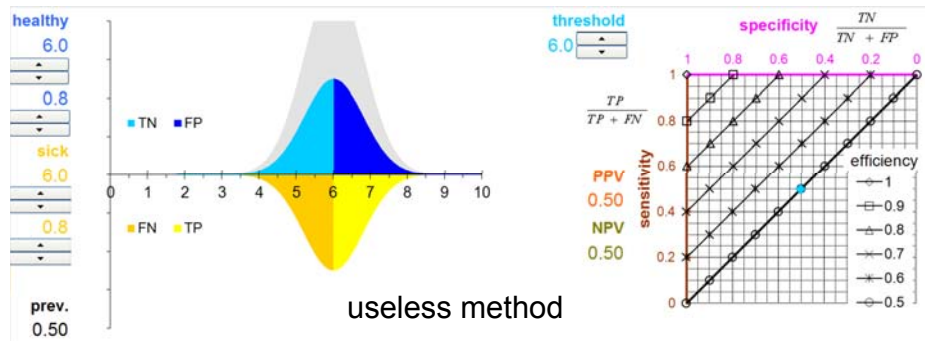
ROC: receiver-operator (operating) characteristic

~ 1950: first ROC Analysis (receiver: Radar)

~ 1970: first medical applications

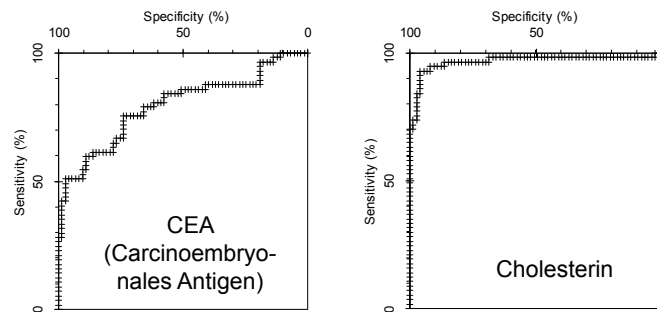


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E.g.: Tumor markers in the ascites

increased CEA and/or cholesterol concentrations in ascites are diagnostic markers for carcinomatosis



Which method is better? What discrimination threshold should be used?

Gulyás M, Kaposi AD, Elek G, Szollár LG, Hjerpe A, Value of carcinoembryonic antigen (CEA) and cholesterol assays of ascitic fluid in cases of inconclusive cytology, J Clinical Pathology 2001 (54) 831-835

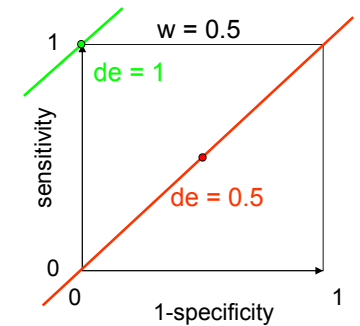
$$de = se \cdot w + sp \cdot (1 - w)$$

$$\frac{de}{1 - w} = \frac{w}{1 - w} se + (sp - 1) + 1$$

$$(1 - sp) + \frac{de}{1 - w} - 1 = \frac{w}{1 - w} se$$

$$se = \frac{1 - w}{w} (1 - sp) + \frac{1}{w} de + \frac{w - 1}{w}$$

slope intercept



if $w = 0.5$: $se = 1 \cdot (1 - sp) + 2 \cdot de - 1$

The points have the same diagnostic efficiency belong to a line with a slope of 1.

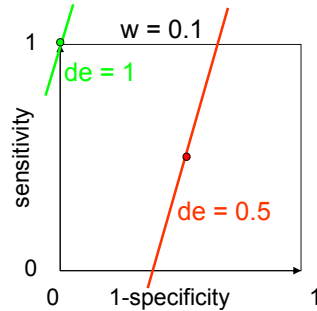
If $de = 0.5$, the intercept is 0.

$$se = \underbrace{\frac{1-w}{w}}_{\text{slope}} (1-sp) + \underbrace{\frac{1}{w} de + \frac{w-1}{w}}_{\text{intercept}}$$

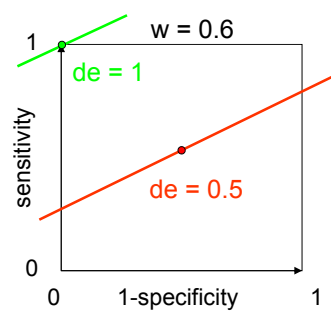
if $w < 0.5$: the slope of lines at identical diagnostic efficiencies is greater than 1.

if $w > 0.5$: the slope of lines at identical diagnostic efficiencies is smaller than 1.

e.g. $w = 0.1$, slope: 9

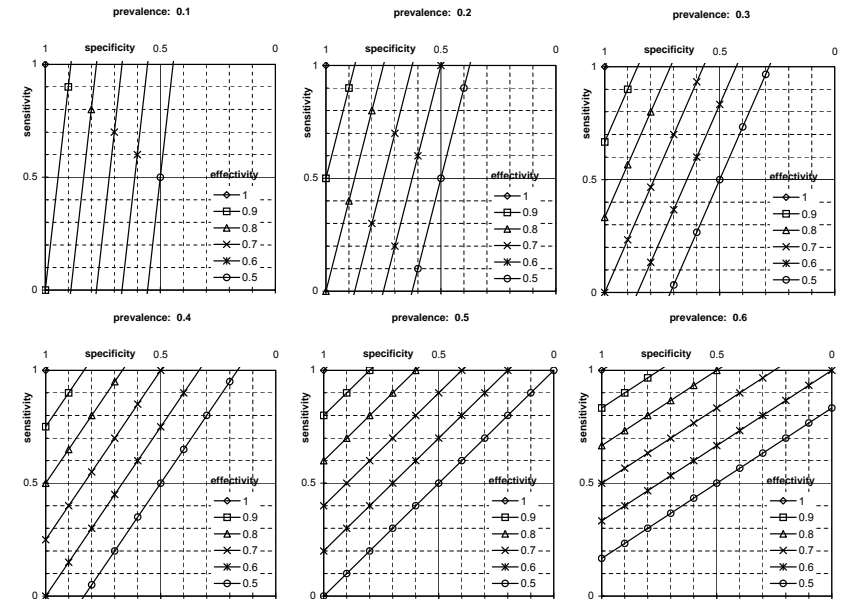


e.g. $w = 0.6$, slope: 0.66



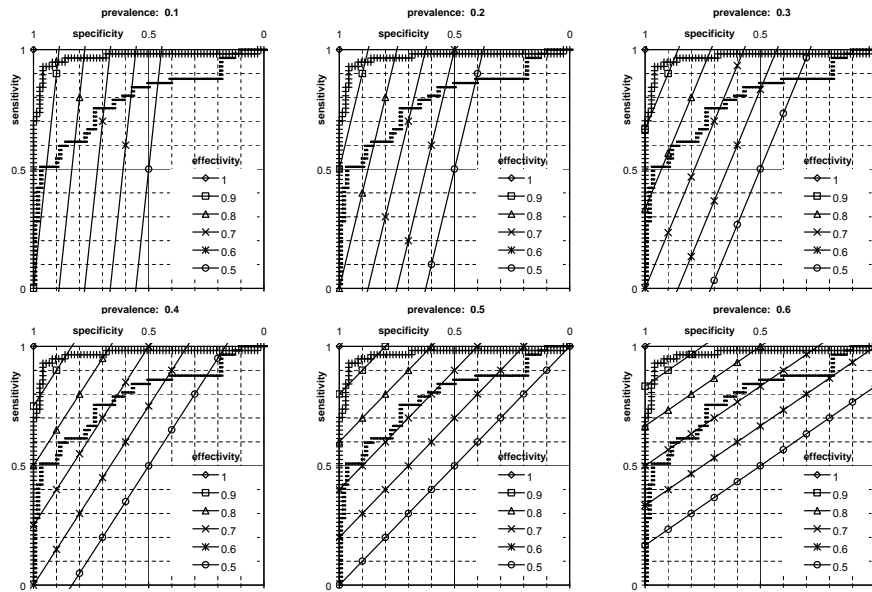
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Isoeffective curves on the ROC



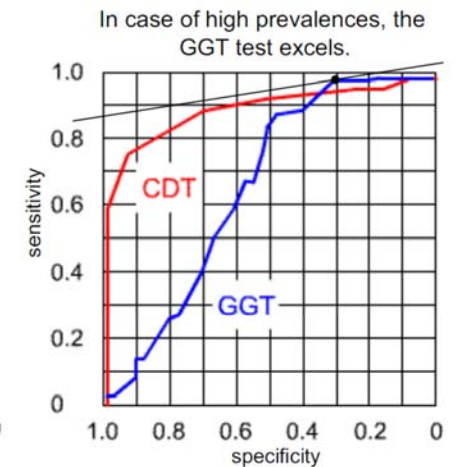
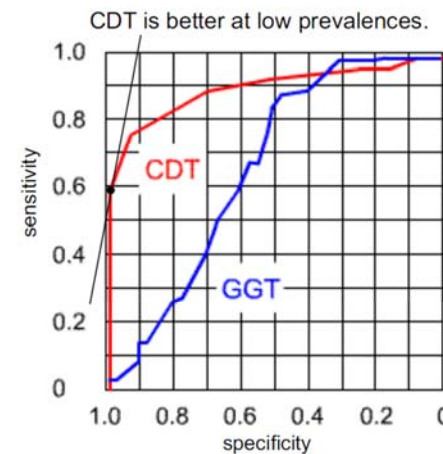
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Ascites (+ Cholesterin, – CEA)



Twist in the ROC space...

Alcoholism diagnostics with CDT (carbohydrate deficient transferrin) and γ -GT (gamma-Glutamyltransferase)



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