Types of errors

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Basic terms

In general, **Positive** = identified and **negative** = rejected. Therefore:

True positive = correctly identified

False positive = incorrectly identified

True negative = correctly rejected

False negative = incorrectly rejected

Medical testing example:

True positive = Sick people correctly diagnosed as sick

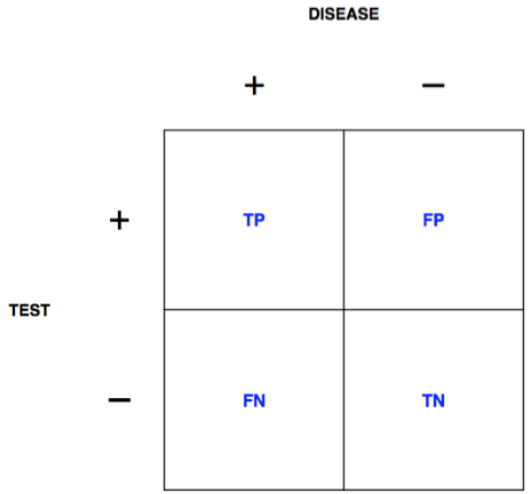
False positive= Healthy people incorrectly identified as sick

True negative = Healthy people correctly identified as healthy

False negative = Sick people incorrectly identified as healthy.

http://en.wikipedia.org/wiki/Sensitivity_and_specificity

Key quantities

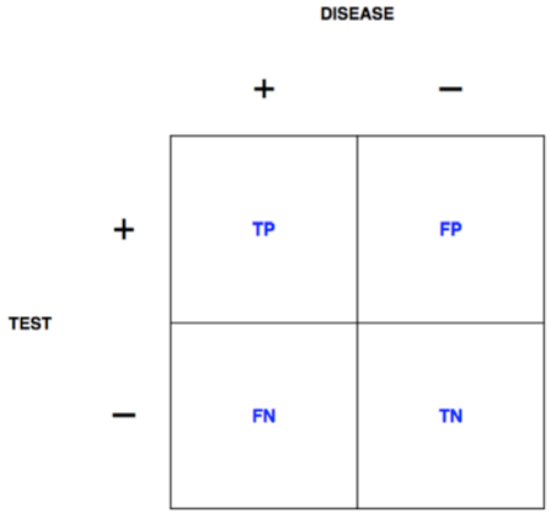


Sensitivity	→	Pr (positive test | disease)
Specificity	→	Pr (negative test | no disease)
Positive Predictive Value	→	Pr (disease | positive test)
Negative Predictive Value	→	Pr (no disease | negative test)
Accuracy	→	Pr (correct outcome)

http://en.wikipedia.org/wiki/Sensitivity_and_specificity

http://www.biostat.jhsph.edu/~iruczins/teaching/140.615/

Key quantities as fractions



Sensitivity	→	TP / (TP+FN)
Specificity	→	TN / (FP+TN)
Positive Predictive Value	→	TP / (TP+FP)
Negative Predictive Value	→	TN / (FN+TN)
Accuracy	→	(TP+TN) / (TP+FP+FN+TN)

http://www.biostat.jhsph.edu/~iruczins/teaching/140.615/

Screening tests

Assume that some disease has a 0.1% prevalence in the population. Assume we have a test kit for that disease that works with 99% sensitivity and 99% specificity. What is the probability of a person having the disease given the test result

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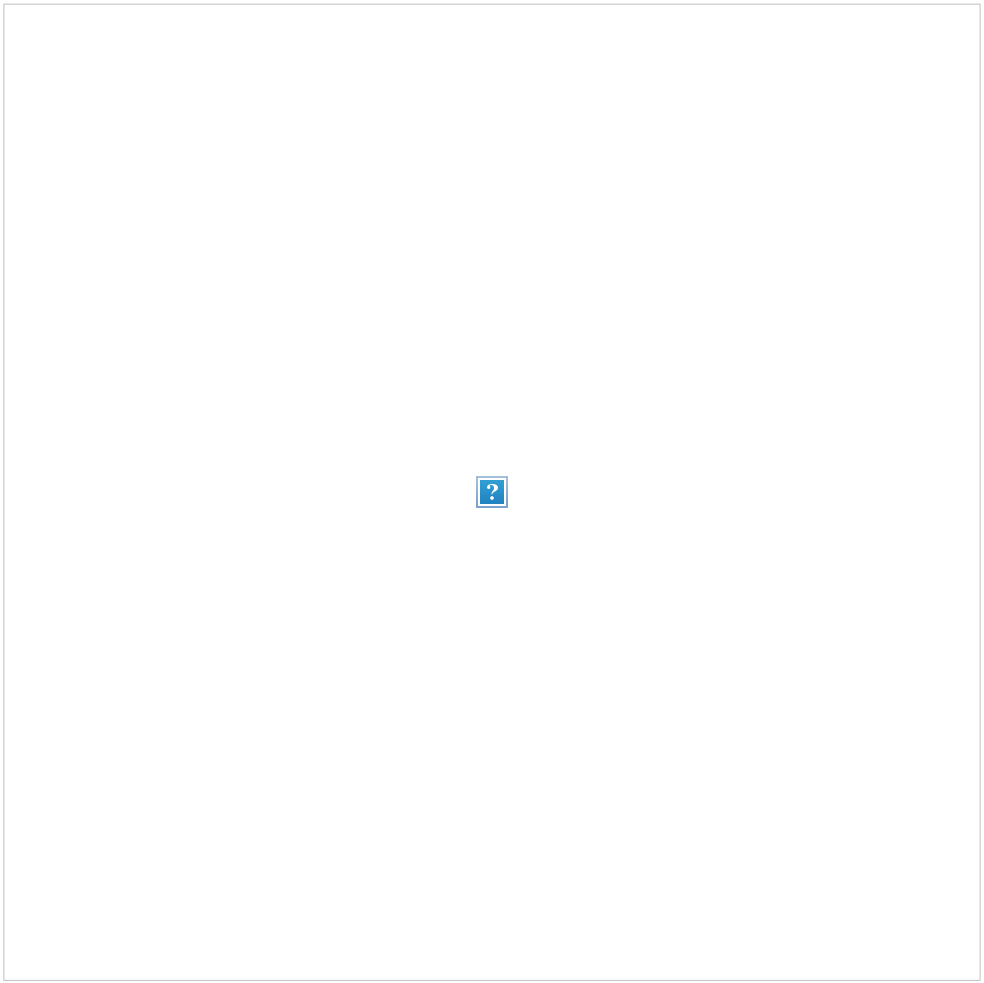
General population



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General population as fractions

TEST \ DISEASE	+	−
+	99	999
−	1	98901

Sensitivity → 99 / (99+1) = 99%

Specificity → 98901 / (999+98901) = 99%

Positive Predictive Value → 99 / (99+999) ~ 9%

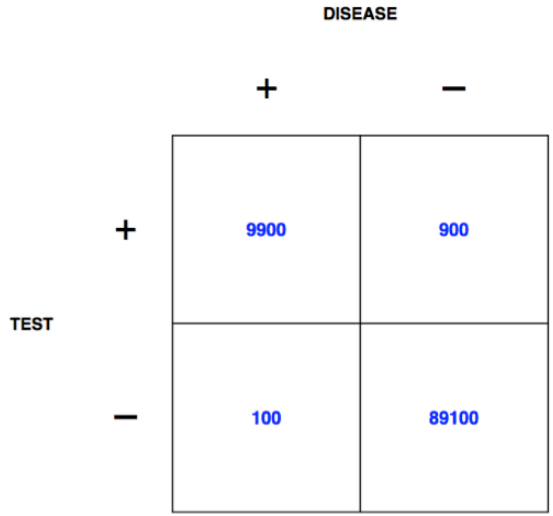
http://www.biostat.jhsph.edu/~iruczins/teaching/140.615/

At risk subpopulation

		DISEASE +	DISEASE −
TEST	+	9900	900
	−	100	89100

http://www.biostat.jhsph.edu/~iruczins/teaching/140.615/

At risk subpopulation as fraction



Sensitivity → 9900 / (9900+100) = 99%

Specificity → 89100 / (900+89100) = 99%

Positive Predictive Value → 9900 / (9900+900) ≈ 92%

Negative Predictive Value → 89100 / (100+89100) ≈ 99.9%

http://www.biostat.jhsph.edu/~iruczins/teaching/140.615/

Key public health issue

Vast Study Casts Doubts on Value of Mammograms

By GINA KOLATA FEB. 11, 2014

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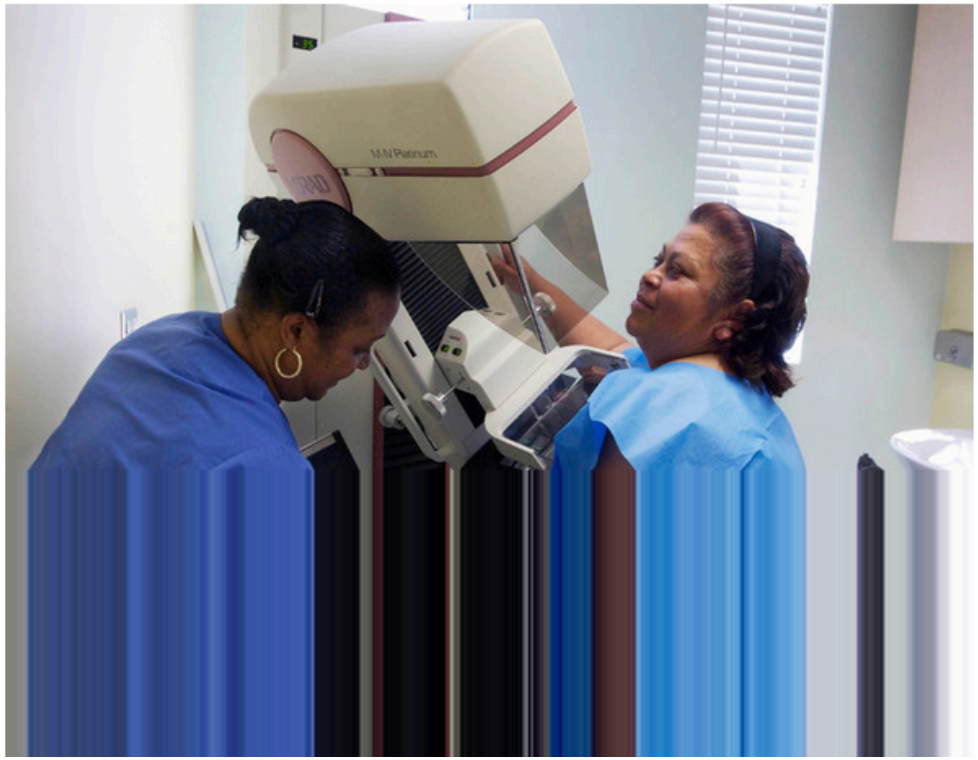
One of the largest and most meticulous studies of mammography ever done, involving 90,000 women and lasting a quarter-century, has added powerful new doubts about the value of the screening test for women of any age.

It found that the death rates from breast cancer and from all causes were the same in women who got mammograms and those who did not. And the screening had harms: One in five cancers found with



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Key public health issue

The New York Times

Business Day

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Looser Guidelines Issued on Prostate Screening

By ANDREW POLLACK
Published: May 3, 2013

In a major shift, the American Urological Association has pulled back its strong support of prostate cancer screening, saying that the testing should be considered primarily by men aged 55 to 69.

The association had staunchly defended the benefits of screening men with the prostate test, even after a government advisory committee, the United States Preventive Services Task Force, said in 2011 that healthy men should not be screened because far more men would be harmed by unnecessary prostate cancer treatments than would be saved from death.

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For continuous data

Mean squared error (MSE):

$$\frac{1}{n}\sum_{i=1}^{n}(Prediction_i - Truth_i)^2$$

Root mean squared error (RMSE):

$$\sqrt{\frac{1}{n}\sum_{i=1}^{n}(Prediction_i - Truth_i)^2}$$

Common error measures

1. Mean squared error (or root mean squared error)
    - Continuous data, sensitive to outliers
2. Median absolute deviation
    - Continuous data, often more robust
3. Sensitivity (recall)
    - If you want few missed positives
4. Specificity
    - If you want few negatives called positives
5. Accuracy
    - Weights false positives/negatives equally
6. Concordance
    - One example is kappa