



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

		DISEASE	
		+	-
TEST	+	TP	FP
	-	FN	TN

$$TP + TN = \text{Accuracy}$$

Sensitivity

→ $\Pr(\text{positive test} \mid \text{disease})$

Specificity

→ $\Pr(\text{negative test} \mid \text{no disease})$

Positive Predictive Value

→ $\Pr(\text{disease} \mid \text{positive test})$

Negative Predictive Value

→ $\Pr(\text{no disease} \mid \text{negative test})$

Accuracy

→ $\Pr(\text{correct outcome})$

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

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

Key quantities as fractions

		DISEASE	
		+	-
TEST	+	TP	FP
	-	FN	TN

Sensitivity

$$\rightarrow TP / (TP+FN)$$

Specificity

$$\rightarrow TN / (FP+TN)$$

Positive Predictive Value

$$\rightarrow TP / (TP+FP)$$

Negative Predictive Value

$$\rightarrow TN / (FN+TN)$$

Accuracy

$$\rightarrow (TP+TN) / (TP+FP+FN+TN)$$

<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 is positive**, if we randomly select a subject from

- ▶ the general population? `also Prevalence = 0.1%`
- ▶ a high risk sub-population with 10% disease prevalence?

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

General population

		DISEASE	
		+	-
TEST	+	99	999
	-	1	98901

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

General population as fractions

		DISEASE	
		+	-
TEST	+	99	999
	-	1	98901

Sensitivity

$$\rightarrow 99 / (99+1) = 99\%$$

Specificity

$$\rightarrow 98901 / (999+98901) = 99\%$$

Positive Predictive Value

$$\rightarrow 99 / (99+999) \approx 9\%$$

Negative Predictive Value

$$\rightarrow 98901 / (1+98901) > 99.9\%$$

Accuracy

$$\rightarrow (99+98901) / 100000 = 99\%$$

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

At risk subpopulation

		DISEASE	
		+	-
TEST	+	9900	900
	-	100	89100

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

At risk subpopulation as fraction

		DISEASE	
		+	-
TEST	+	9900	900
	-	100	89100

Sensitivity

$$\rightarrow 9900 / (9900 + 100) = 99\%$$

Specificity

$$\rightarrow 89100 / (900 + 89100) = 99\%$$

Positive Predictive Value

$$\rightarrow 9900 / (9900 + 900) \approx 92\%$$

Negative Predictive Value

$$\rightarrow 89100 / (100 + 89100) \approx 99.9\%$$

Accuracy

$$\rightarrow (9900 + 89100) / 100000 = 99\%$$

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

Key public health issue -> false positives

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 mammography and treated was not a threat to the woman's health and did not need treatment such as chemotherapy, surgery or radiation.

The study, published Tuesday in The British Medical Journal, is one of the few rigorous



Nearly 75 percent of American women 40 and over say they had a mammogram in the past year. Damian Dovarganes/Associated Press

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

Key public health issue

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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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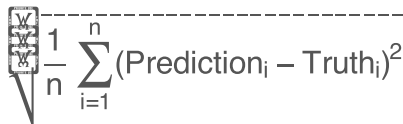
For continuous data

Mean squared error (MSE):

$$\frac{1}{n} \sum_{i=1}^n (\text{Prediction}_i - \text{Truth}_i)^2$$

Root mean squared error (RMSE):

to get the same unit:



The diagram illustrates the derivation of the RMSE formula from the MSE formula. It shows the MSE formula $\frac{1}{n} \sum_{i=1}^n (\text{Prediction}_i - \text{Truth}_i)^2$ with a large square root symbol $\sqrt{}$ positioned to its left. A dashed horizontal line extends from the top of the square root symbol across the entire formula, indicating that the square root is applied to the entire expression.

$$\sqrt{\frac{1}{n} \sum_{i=1}^n (\text{Prediction}_i - \text{Truth}_i)^2}$$

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