

Comparing algorithm - tests.

- * t-test - The accuracy of the models is compared.
- * It is used to compare different classifiers such as SVM, DT, RF, etc.
- * First ^{take} mean.

↳ * But mean will not always give the accurate results.

* Because mean is always ^{highly} affected by the datapoints. Whenever new data point is added the change in mean occurs.

* So we will go to t-test.

Why t-test?

* The choice of applying test depends on the sample size.

* Z-test is used when there is a large sample. t-test is used for small sample.

* In machine learning, ~~work~~ we use real time data, where we need to make assumption normality. t-test is more robust in this case.

Hypothesis testing :-

1. - It is a statistical technique which is used to make decisions based on sample data.

Steps :-

1. Define null hypothesis and alternate hypothesis.

H_0 :- There is no difference in the mean accuracy.

H_1 :- There is difference in the accuracy of the model.

* Choose significance level -

It is the probability of rejecting the null hypothesis.

* Usually 0.05 is taken as the level of significance - 95 % confidence H_0 is true
5 % confidence H_0 is false.

* Collect the data.

* Calculate t-statistic

⇒ If we are comparing same classifiers then it is known as related t-test.

⇒ If we are comparing different classifiers then it is known as paired independent t-test.

* p-value is determined.

* p-value is the probability of obtaining a test statistic as extreme or more extreme than observed — assuming H_0 is true

↳ area under t-distribution

formula of test test:

curve that is more extreme than observed value of t

$$t = \frac{(\bar{x} - \mu)}{\frac{s}{\sqrt{n}}}$$

⇒ \bar{x} - sample mean

μ - population mean

s - standard deviation

n - sample size.

$$t = (\bar{x}_1 - \bar{x}_2) / (s_p * \sqrt{1/n_1 + 1/n_2})$$

s_p = pooled standard deviation

$$\bar{A}_1 = \bar{A}_2$$

$$s_p = \sqrt{(s_1^2 + s_2^2) / 2}$$

s_1 & s_2 are standard deviations of two samples.

n_1 & n_2 are sample size