

```

# comparacion mediante pruebas en pareja para medir significancia estadistica

# arboles = 100, v1d PROM= 0.986607
rf = [0.99678686, 0.99964298, 1 ,      1 ,      1 ,      1 , 1 ,      1 ,      1

# arboles= 500 , v3  PROM= 0.98232
rf2 = [0.99393074, 0.99964298, 1. ,      1. ,      1. ,      1., 1. ,      1. ,

# arboles = 25,   v2 prom= 0.978500318763707
rf3 = [0.99143163, 0.99964298, 1. ,      0.99928571, 1.,      1. , 1. ,      1. ,

# de la prueba pareada entres los tres rf, ninguno es estadisticamete mas sigificativo, enton

from scipy import stats

stats.ttest_rel(rf2, rf)

Ttest_relResult(statistic=-1.0625793669394634, pvalue=0.31565170839933626)

#stats.ttest_rel(rf2, rf2)

#stats.ttest_rel(rf3, rf2)

stats.ttest_rel(rf3, rf)

Ttest_relResult(statistic=-1.078184934383557, pvalue=0.30899927516686543)

# svm RADIAL 0 RBF accPROM 0.904903771
svm = [0.87325955, 0.82113531, 0.88964286, 0.90821429, 0.92892857, 0.94285714, 0.94428571, 0
# svm 2 lineal accPROM 0.85854763604835
svm2 = [0.86719029, 0.79900036, 0.84392857, 0.865 ,      0.8425 ,      0.85142857, 0.88107143, 0
# svm 3 poly alfa 8,  accPROM 0.7308073392155863
svm3 = [0.70581935, 0.68868261, 0.73892857, 0.73464286, 0.71392857, 0.72392857, 0.73642857, 0

from scipy import stats

#stats.ttest_rel(df['bp_before'], df['bp_after'])
stats.ttest_rel(svm2, svm) # svm radial  es mas significativo

Ttest_relResult(statistic=-5.121446640281181, pvalue=0.0006267555728178404)

```

```

from scipy import stats

#stats.ttest_rel(df['bp_before'], df['bp_after'])
stats.ttest_rel(svm3, svm) # svm radial es mas significativo

Ttest_relResult(statistic=-12.545525125959326, pvalue=5.26789191645834e-07)

from scipy import stats

#stats.ttest_rel(df['bp_before'], df['bp_after'])
stats.ttest_rel(rf, svm)
#stats.ttest_rel(gh2, gh)

Ttest_relResult(statistic=5.2117566811001295, pvalue=0.0005552702643440211)

#print(stats.__version__)

import scipy

print(scipy.__version__)

#print('scipy'.__version__)

1.4.1

stats.ttest_rel(svm, rf)

Ttest_relResult(statistic=-5.2117566811001295, pvalue=0.0005552702643440211)

# El valor p devuelto (0.0005552) es menor que el umbral 'a' (0.05). Por lo tanto, se rechaza
# que el rendimiento de 'rf' es significativamente diferente y mejorado en comparación con el

# https://www.coursehero.com/file/p4hol16a/Aqu%C3%AD-prepresenta-la-proporci%C3%B3n-poblacion

ann = [0.92771084, 0.99776786, 0.92767857, 0.98883929, 0.99776786, 0.98660714, 0.925, 0

# 98.3753008
ann77 = [0.93261937, 0.99464286, 0.99910714 ,0.99776786, 0.99910714 ,0.99107143, 0.99955357,

#ann78 = [1, 1, 1 ,1, 1 ,1, 1, 1, 1, 1]

# ann78 = [1, 1, 1 ,1, 1 ,1, 1, 1, 1, 1,1,1,1,1,1,1,1]

# prom= 89.61651

```

```

annv2=[0.90138331 ,0.903125 , 0.89553571, 0.89464286, 0.89642857, 0.89955357, 0.89375 , 0.

# prom = 95.9779873
# annv3=[0.99393074, 0.99964298, 1. , 1. , 1. , 1., 1. , 1. ,

annv3=[0.93083445, 0.98303571, 0.92321429, 0.92723214 ,0.9875 , 0.99375, 0.99553571, 0.938

#stats.ttest_rel(annv2, ann)

#stats.ttest_rel(ann, ann77)

stats.ttest_rel(annv2, annv3)
# resulta mas significativo annv3

Ttest_relResult(statistic=-6.538442668121652, pvalue=0.0001065875379783153)

stats.ttest_rel(annv3, ann77)
# la mas significativa es la ann77

Ttest_relResult(statistic=-2.4026432030499154, pvalue=0.03972535093203618)

stats.ttest_rel(svm, ann)

Ttest_relResult(statistic=-3.8274191226309147, pvalue=0.004043678125891304)

# El valor p devuelto (0.0040436) es menor que el umbral 'a' (0.05). Por lo tanto, se rechaza
# que el rendimiento de 'ann' es significativamente diferente y mejorado en comparación con e

stats.ttest_rel(rf, ann)

Ttest_relResult(statistic=1.8770201400069235, pvalue=0.09324865532455924)

stats.ttest_rel(ann,rf)

Ttest_relResult(statistic=-1.8770201400069235, pvalue=0.09324865532455924)

# El valor p devuelto (0.093248655) es mayor que el umbral 'a' (0.05). Por lo tanto, se acept
# que el rendimiento de 'rf' es NO ES significativamente diferente y mejorado en comparación

# a H0 es las medias de las dos muestras son iguales

# Si el valor p es menor que un cierto nivel de significancia  $\alpha$  fijado por el usuario a prior
#estadisticamente significativo y H0 se debe rechazar.

```

```
#Que un resultado sea estadísticamente significativo quiere
#decir que es improbable que suceda por azar.
```

```
#Un valor p menor a 0,05 indica una fuerte evidencia en
#contra de H0 porque habría una probabilidad menor al
# 5 % que H0 sea correcta y nos lleva a aceptar H1.
```

```
#Un valor p superior a 0,05 indica una fuerte evidencia a
#favor de H0 y al rechazo de H1.
```

```
stats.ttest_rel(ann, rf)
```

```
Ttest_relResult(statistic=-1.8770201400069235, pvalue=0.09324865532455924)
```

```
# Resultado: el estadístico de la prueba t de muestras emparejadas es -1.8770201400069235 y e
# El valor p es mayor que el nivel de significancia = 0.05.
# La prueba t de muestras pareadas no rechaza la hipótesis nula: la media de "ann" y la media
```

```
import numpy as np
np.mean(rf)
```

```
0.987571555
```

```
np.std(rf)
```

```
0.036107795683153865
```

```
np.mean(svm)
```

```
0.9049037710000001
```

```
np.std(svm)
```

```
0.03622486618036772
```

```
np.mean(ann)
```

```
0.9651371559999999
```

```
np.std(ann)
```

```
0.03303353508427252
```

```
stats.ttest_rel(ann, svm)
```

```
stats.ttest_rel(svm, ann77)
```

```
Ttest_relResult(statistic=3.8274191226309147, pvalue=0.004043678125891304)
```

```
stats.ttest_rel(svm, ann77)
```

```
Ttest_relResult(statistic=-5.714908029509426, pvalue=0.00028880820708700487)
```

```
stats.ttest_rel(rf, ann77)
```

```
Ttest_relResult(statistic=0.23564252830958335, pvalue=0.8189847522277838)
```

```
# no hay significancia pero el rendimiento es mejor en Ann
```

```
stats.ttest_rel(ann77, rf)
```

```
Ttest_relResult(statistic=-0.23564252830958335, pvalue=0.8189847522277838)
```

```
# stats.ttest_rel(ann78, rf)
```

```
np.mean(ann77)
```

```
0.9837530079999999
```

```
np.std(ann77)
```

```
0.027051125415198097
```

```
# https://machinelearningmastery.com/how-to-code-the-students-t-test-from-scratch-in-python/
```

```
# t-test for dependent samples
```

```
from math import sqrt
```

```
from numpy.random import seed
```

```
from numpy.random import randn
```

```
from numpy import mean
```

```
from scipy.stats import t
```

```
# function for calculating the t-test for two dependent samples
```

```
def dependent_ttest(data1, data2, alpha):
```

```
    # calculate means
```

```
    mean1, mean2 = mean(data1), mean(data2)
```

```
    # number of paired samples
```

```
    n = len(data1)
```

```
    # sum squared difference between observations
```

```
    d1 = sum([(data1[i]-data2[i])**2 for i in range(n)])
```

```
    # sum difference between observations
```

```
    d2 = sum([data1[i]-data2[i] for i in range(n)])
```

```
    # standard deviation of the difference between means
```

```
    sd = sqrt((d1 - (d2**2 / n)) / (n - 1))
```

```
    # standard error of the difference between the means
```

```

    sed = sd / sqrt(n)
    # calculate the t statistic
    t_stat = (mean1 - mean2) / sed
    # degrees of freedom
    df = n - 1
    # calculate the critical value
    cv = t.ppf(1.0 - alpha, df)
    # calculate the p-value
    p = (1.0 - t.cdf(abs(t_stat), df)) * 2.0
    # return everything
    return t_stat, df, cv, p

# seed the random number generator
seed(1)
# generate two independent samples (pretend they are dependent)
data1 = 5 * randn(100) + 50
data2 = 5 * randn(100) + 51
# calculate the t test
alpha = 0.05
#t_stat, df, cv, p = dependent_ttest(data1, data2, alpha)
t_stat, df, cv, p = dependent_ttest(ann77, rf, alpha)
print('t=%.3f, df=%d, cv=%.3f, p=%.3f' % (t_stat, df, cv, p))
# interpret via critical value
if abs(t_stat) <= cv:
    print('Accept null hypothesis that the means are equal.')
else:
    print('Reject the null hypothesis that the means are equal.')
# interpret via p-value
if p > alpha:
    print('Accept null hypothesis that the means are equal.')
else:
    print('Reject the null hypothesis that the means are equal.')

    t=-0.236, df=9, cv=1.833, p=0.819
    Accept null hypothesis that the means are equal.
    Accept null hypothesis that the means are equal.

```

```

# t-test for dependent samples
from math import sqrt
from numpy.random import seed
from numpy.random import randn
from numpy import mean
from scipy.stats import t

# function for calculating the t-test for two dependent samples
def dependent_ttest(data1, data2, alpha):
    # calculate means
    mean1, mean2 = mean(data1), mean(data2)
    # number of paired samples
    n = len(data1)
    # sum squared difference between observations
    d1 = sum([(data1[i]-data2[i])**2 for i in range(n)])

```

```

d1 = sum([(data1[i]-data2[i])**2 for i in range(n)])
# sum difference between observations
d2 = sum([data1[i]-data2[i] for i in range(n)])
# standard deviation of the difference between means
sd = sqrt((d1 - (d2**2 / n)) / (n - 1))
# standard error of the difference between the means
sed = sd / sqrt(n)
# calculate the t statistic
t_stat = (mean1 - mean2) / sed
# degrees of freedom
df = n - 1
# calculate the critical value
cv = t.ppf(1.0 - alpha, df)
# calculate the p-value
p = (1.0 - t.cdf(abs(t_stat), df)) * 2.0
# return everything
return t_stat, df, cv, p

# seed the random number generator
seed(1)
# generate two independent samples (pretend they are dependent)
data1 = 5 * randn(100) + 50
data2 = 5 * randn(100) + 51
# calculate the t test
alpha = 0.05
alpha = 0.01
#t_stat, df, cv, p = dependent_ttest(data1, data2, alpha)
#t_stat, df, cv, p = dependent_ttest(ann, rf, alpha)
t_stat, df, cv, p = dependent_ttest(rf, rf2, alpha)
print('t=%.3f, df=%d, cv=%.3f, p=%.3f' % (t_stat, df, cv, p))
# interpret via critical value
if abs(t_stat) <= cv:
    print('Accept null hypothesis that the means are equal.')
else:
    print('Reject the null hypothesis that the means are equal.')
# interpret via p-value
if p > alpha:
    print('Accept null hypothesis that the means are equal.')
else:
    print('Reject the null hypothesis that the means are equal.')

    t=1.063, df=9, cv=2.821, p=0.316
    Accept null hypothesis that the means are equal.
    Accept null hypothesis that the means are equal.

#pr

#c1=[97.07]

c1= [0.9978678 , 0.99964311,1.      ,      1.      ,      1.      ,      1.      , 1.      ,      1.      ,

# c2=[97.92]

```

```
c2= [1. ,      1. ,      1. ,      0.99857347, 1. ,      1., 1. ,      1. ,  
stats.ttest_rel(c1, c2)  
Ttest_relResult(statistic=0.9915687549955755, pvalue=0.34731640777152717)
```