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```
In []: !pip install statsmodels
!pip install scikit-learn
!pip install scikit-posthocs
!pip install openpyxl
!pip install scikit-posthocs
!pip install bioinfokit
```

## Librerias

```
import pandas as pd
import statsmodels.api as sm
from statsmodels.formula.api import ols
from statsmodels.stats.multicomp import pairwise_tukeyhsd
import scikit_posthocs as sp
from bioinfokit.analys import stat
```

# 1. TOROS6

```
In [12]: # Leer el archivo Excel
    df = pd.read_excel("toros6.xlsx")
    df['Toros'] = df['Toros'].astype(str)

# ANOVA
    modelo = ols('Peso ~ C(Toros)', data=df).fit()
    print(sm.stats.anova_lm(modelo, typ=2))

# Duncan
    #print(sp.posthoc_duncan(df, val_col='Peso', group_col='Toros'))

# Tukey
    tukey = pairwise_tukeyhsd(df['Peso'], df['Toros'])
    print(tukey)
```

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```
df
                                  PR(>F)
            sum sq
C(Toros)
         22.883333
                   5.0 1.842953 0.119954
Residual 134.100000 54.0
                           NaN
Multiple Comparison of Means - Tukey HSD, FWER=0.05
_____
group1 group2 meandiff p-adj lower upper reject
   -0.4 0.9927 -2.4822 1.6822 False
               0.8 0.8645 -1.2822 2.8822 False
    1
    1
          4
               -0.2 0.9997 -2.2822 1.8822 False
          5
    1
               1.3 0.4468 -0.7822 3.3822 False
    1
               0.8 0.8645 -1.2822 2.8822 False
          6
    2
               1.2 0.536 -0.8822 3.2822 False
          3
    2
          4
                0.2 0.9997 -1.8822 2.2822 False
    2
          5
               1.7 0.1703 -0.3822 3.7822 False
    2
          6
               1.2 0.536 -0.8822 3.2822 False
    3
          4
               -1.0 0.7155 -3.0822 1.0822 False
    3
          5
                0.5 0.98 -1.5822 2.5822 False
    3
               0.0
                     1.0 -2.0822 2.0822 False
    4
          5
               1.5 0.2886 -0.5822 3.5822 False
                1.0 0.7155 -1.0822 3.0822 False
    4
    5
          6
               -0.5 0.98 -2.5822 1.5822 False
```

### 2. ALFALFA

```
In [18]:
         df = pd.read_excel("alfalfa.xlsx")
         df['Parcela'] = df['Parcela'].astype(str)
         # ANOVA
         modelo = ols('Dosis ~ C(Parcela)', data=df).fit()
         print(sm.stats.anova_lm(modelo, typ=2))
         # Duncan
         #print(sp.posthoc duncan(df, val col='Dosis', group col='Parcela'))
         # Tukey
         tukey = pairwise tukeyhsd(df['Dosis'], df['Parcela'])
         print(tukey)
                     sum sq
                              df
                                    F
                                                    PR(>F)
```

```
C(Parcela)
                4.0 265.978723 4.947594e-17
         2000.16
Residual
           37.60 20.0
                          NaN
Multiple Comparison of Means - Tukey HSD, FWER=0.05
_____
group1 group2 meandiff p-adj lower upper reject
_____
              4.2 0.0008 1.6051 6.7949
    1
         2
    1
         3
              15.0
                   0.0 12.4051 17.5949
                                      True
         4
    1
              14.6
                    0.0 12.0051 17.1949
                                      True
    1
             25.4 0.0 22.8051 27.9949
                                      True
    2
             10.8 0.0 8.2051 13.3949
         3
                                      True
    2
         4
              10.4
                    0.0 7.8051 12.9949
                                      True
    2
         5
             21.2 0.0 18.6051 23.7949
                                     True
    3
         4
             -0.4 0.99 -2.9949 2.1949 False
    3
         5
              10.4
                    0.0 7.8051 12.9949
                                      True
    4
         5
              10.8
                    0.0 8.2051 13.3949
                                      True
```

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# **2. TOROS5**

```
In [19]: df = pd.read excel("toros5.xlsx")
       df['Toros'] = df['Toros'].astype(str)
       modelo = ols('Peso ~ C(Toros)', data=df).fit()
       print(sm.stats.anova_lm(modelo, typ=2))
       # Duncan
       #print(sp.posthoc_duncan(df, val_col='Peso', group_col='Toros'))
       tukey = pairwise_tukeyhsd(df['Peso'], df['Toros'])
       print(tukey)
               sum_sq df
                            F
                                    PR(>F)
      C(Toros)
               777.6
                     4.0 2.778113 0.038046
      Residual 3148.9 45.0 NaN
       Multiple Comparison of Means - Tukey HSD, FWER=0.05
          _____
      group1 group2 meandiff p-adj lower
                                       upper reject
       -----
                     -7.4 0.2931 -18.0299 3.2299 False
                     -7.5 0.2803 -18.1299 3.1299 False
           1
                4 -12.2 0.0172 -22.8299 -1.5701 True
           1
               5 -5.4 0.6035 -16.0299 5.2299 False
           2
               3
                    -0.1
                           1.0 -10.7299 10.5299 False
                    -4.8 0.7027 -15.4299 5.8299 False
           2
                4
          2
               5
                     2.0 0.9832 -8.6299 12.6299 False
           3
                    -4.7 0.7187 -15.3299 5.9299 False
               4
           3
               5
                     2.1 0.9799 -8.5299 12.7299 False
                     6.8 0.3764 -3.8299 17.4299 False
```

## 4. CEBADA

```
In [20]: df = pd.read_excel("cebada.xlsx")
    df['Cebada'] = df['Cebada'].astype(str)

# ANOVA
modelo = ols('Cosecha ~ C(Cebada)', data=df).fit()
print(sm.stats.anova_lm(modelo, typ=2))

# Duncan
#print(sp.posthoc_duncan(df, val_col='Cosecha', group_col='Cebada'))

# Tukey
tukey = pairwise_tukeyhsd(df['Cosecha'], df['Cebada'])
print(tukey)
```

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sum\_sq df F PR(>F) C(Cebada) 24775.0 3.0 12.234568 0.000206 Residual 10800.0 16.0 NaN NaN Multiple Comparison of Means - Tukey HSD, FWER=0.05 \_\_\_\_\_ group1 group2 meandiff p-adj upper reject lower \_\_\_\_\_ 4.0 0.9947 -43.0114 51.0114 False -84.0 0.0005 -131.0114 -36.9886 True 1 3 1 4 -22.0 0.553 -69.0114 25.0114 False 2 3 -88.0 0.0003 -135.0114 -40.9886 True 2 4 -26.0 0.4155 -73.0114 21.0114 False 3 4 62.0 0.0081 14.9886 109.0114 True