

```
In [ ]: import numpy as np
import pandas as pd
import statsmodels.api as sm
from statsmodels.formula.api import ols
from statsmodels.stats.anova import anova_lm

# Data
C = [16.30, 11.90, 19.30, 23.60, 15.10, 16.80, 8.50, 18.70, 15.20, 9.80, 23.60, 9.10, 18.30, 13.20, 12.00, 15.30, 23.50, 9.50, 13.60, 26.00]
F1 = [22.60, 13.00, 30.10, 14.10, 21.00, 14.20, 19.80, 31.80, 27.00, 17.20, 16.60, 22.50, 14.10, 14.70, 25.70, 19.20, 23.20, 28.10, 19.00, 12.00]
F2 = [14.60, 16.80, 16.90, 18.10, 20.20, 15.30, 17.60, 28.10, 19.10, 19.90, 18.70, 21.30, 16.00, 25.30, 13.30, 24.40, 11.40, 11.10, 18.40, 19.90]
F3 = [26.10, 27.80, 19.70, 16.90, 23.30, 26.00, 23.90, 20.50, 21.00, 27.80, 23.10, 22.80, 24.90, 22.00, 23.40, 21.60, 24.00, 30.60, 29.10, 31.00]

data = {
    'WeightLoss': C + F1 + F2 + F3,
    'Group': ['C']*len(C) + ['F1']*len(F1) + ['F2']*len(F2) + ['F3']*len(F3)
}

df = pd.DataFrame(data)
model = ols('WeightLoss ~ Group', data=df).fit()

anova_table = anova_lm(model)
print(anova_table)
```

	df	sum_sq	mean_sq	F	PR(>F)
Group	3.0	746.352375	248.784125	10.363037	0.000009
Residual	76.0	1824.522500	24.006875	NaN	NaN

```
In [ ]: import pandas as pd
import numpy as np
from statsmodels.stats.multicomp import pairwise_tukeyhsd
import matplotlib.pyplot as plt

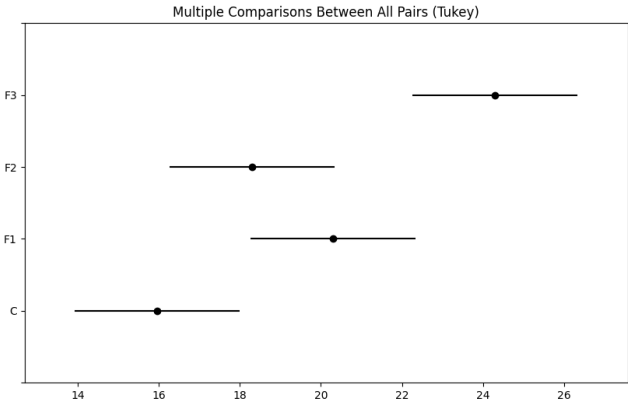
# Data
data = {
    'C': [16.30, 11.90, 19.30, 23.60, 15.10, 16.80, 8.50, 18.70, 15.20, 9.80, 23.60, 9.10, 18.30, 13.20, 12.00, 15.30, 23.50, 9.50, 13.60, 26.00],
    'F1': [22.60, 13.00, 30.10, 14.10, 21.00, 14.20, 19.80, 31.80, 27.00, 17.20, 16.60, 22.50, 14.10, 14.70, 25.70, 19.20, 23.20, 28.10, 19.00, 12.00],
    'F2': [14.60, 16.80, 16.90, 18.10, 20.20, 15.30, 17.60, 28.10, 19.10, 19.90, 18.70, 21.30, 16.00, 25.30, 13.30, 24.40, 11.40, 11.10, 18.40, 19.90],
    'F3': [26.10, 27.80, 19.70, 16.90, 23.30, 26.00, 23.90, 20.50, 21.00, 27.80, 23.10, 22.80, 24.90, 22.00, 23.40, 21.60, 24.00, 30.60, 29.10, 31.00]
}

df = pd.DataFrame(data)
stacked_data = df.stack().reset_index()
stacked_data.columns = ['Index', 'Group', 'Weight Loss']

tukey = pairwise_tukeyhsd(stacked_data['Weight Loss'], stacked_data['Group'])
print(tukey.summary())
tukey.plot_simultaneous()
plt.show()
```

Multiple Comparison of Means – Tukey HSD, FWER=0.05

group1	group2	meandiff	p-adj	lower	upper	reject
C	F1	4.33	0.0326	0.26	8.4	True
C	F2	2.34	0.4365	-1.73	6.41	False
C	F3	8.325	0.0	4.255	12.395	True
F1	F2	-1.99	0.5756	-6.06	2.08	False
F1	F3	3.995	0.0563	-0.075	8.065	False
F2	F3	5.985	0.0013	1.915	10.055	True



```
In [ ]: import numpy as np
from scipy.stats import dunnett

c = np.array([16.30, 11.90, 19.30, 23.60, 15.10, 16.80, 8.50, 18.70, 15.20, 9.80, 23.60, 9.10, 18.30, 13.20, 12.00, 15.30, 23.50, 9.50, 13.60, 26.00])
f1 = np.array([22.60, 13.00, 30.10, 14.10, 21.00, 14.20, 19.80, 31.80, 27.00, 17.20, 16.60, 22.50, 14.10, 14.70, 25.70, 19.20, 23.20, 28.10, 19.00, 12.00])
f2 = np.array([14.60, 16.80, 16.90, 18.10, 20.20, 15.30, 17.60, 28.10, 19.10, 19.90, 18.70, 21.30, 16.00, 25.30, 13.30, 24.40, 11.40, 11.10, 18.40, 19.90])
f3 = np.array([26.10, 27.80, 19.70, 16.90, 23.30, 26.00, 23.90, 20.50, 21.00, 27.80, 23.10, 22.80, 24.90, 22.00, 23.40, 21.60, 24.00, 30.60, 29.10, 31.00])

res = dunnett(f1, f2, f3, control=c)
res.pvalue
```

Out[]: array([1.80154936e-02, 3.07340568e-01, 2.45973916e-06])

```
In [ ]: import numpy as np
import scipy.stats as stats

group_means = np.array([15.965, 20.295, 18.305, 24.290])
j = 20
mse = 24.006875
alpha = 0.05
k = 5

alpha_adj = alpha / k

t_stats = []
t_stats.append((group_means[0] - group_means[1])/(np.sqrt(mse / 20 + mse / 2)))
t_stats.append((group_means[0] - group_means[2])/(np.sqrt(mse / 20 + mse / 2)))
t_stats.append((group_means[0] - group_means[3])/(np.sqrt(mse / 20 + mse / 2)))
t_stats.append((group_means[0] - (group_means[1] + group_means[2] + group_means[3]) / 4) / (np.sqrt(mse / 20 + mse / 2)))
t_stats.append((group_means[1] - group_means[2])/(np.sqrt(mse / 20 + mse / 2)))

df = 76
t_critical = stats.t.ppf(1 - alpha_adj / 2, df)

results = []

for t_stat in t_stats:
    results.append(abs(t_stat) > t_critical)

for result in results:
    print(result)

True
False
True
True
False
```

```
In [ ]: means = np.array([15.965, 20.295, 18.305, 24.290])
J = 20
I = len(means)
df1 = I - 1
df2 = I * (J - 1)
alpha = 0.05
mse = 24.006875
F_obs = 10.363

grand_mean = np.mean(means)

SSTr = J * np.sum((means - grand_mean) ** 2)
lambda_ = SSTr / mse

f_crit = f.ppf(1 - alpha, df1, df2)
power = 1 - ncf.cdf(f_crit, df1, df2, lambda_)

print(f"Noncentrality parameter (lambda): {lambda_:.4f}")
print(f"Critical F value: {f_crit:.4f}")
print(f"Estimated power: {power:.4f}")

Noncentrality parameter (lambda): 31.0891
Critical F value: 2.7249
Estimated power: 0.9981
```

```
In [ ]: import pandas as pd
import statsmodels.api as sm
import statsmodels.formula.api as smf

data = [100.7, 102.4, 104.3, 109.2, 101.7, 99.4, 103.5, 104.0, 107.7, 105.9,
        101.2, 102.1, 103.8, 105.5, 106.2, 107.1, 108.0, 109.1, 110.0, 111.0]
block = ['B1', 'B1', 'B2', 'B2', 'B3', 'B3'] * 4
dose = ['Control'] * 6 + ['Dose 1'] * 6 + ['Dose 2'] * 6 + ['Dose 3'] * 6

df = pd.DataFrame({
    'Response': data,
    'Block': block,
    'Dose': dose
})

df['Block'] = df['Block'].astype('category')
df['Dose'] = df['Dose'].astype('category')

model = smf.ols('Response ~ C(Dose) + C(Block)', data=df).fit()
anova_table = sm.stats.anova_lm(model, typ=1)

print(anova_table)
```

```
In [ ]: SSTr = 0
for drug in ['Control', 'Dose 1', 'Dose 2', 'Dose 3']:
    SSTr += 6 * ((df[df['Drug'] == drug].sum()[0] / 6) - 102.42916667)**2
SSTr
```

```
In [ ]: SSB = 0
for block in ['B1', 'B2', 'B3']:
    SSB += 8 * ((df[df['Block'] == block].sum()[0] / 8) - 102.429166667)**2
SSB
```

Out[]: 161.68083333333323

```
In [ ]: ((df[df['Block'] == 'B1'].sum()[0]))
```

Out[]: 807.8

```
In [ ]: for i in range(4):
    print(i)
df.loc[0].Response
0
1
2
3
```

Out[]: 100.7

```
In [ ]: import pandas as pd
import numpy as np
from sklearn.preprocessing import OneHotEncoder

drug_levels = ['Control'] * 6 + ['Dose 1'] * 6 + ['Dose 2'] * 6 + ['Dose 3']
block_levels = ['B1', 'B1', 'B2', 'B2', 'B3', 'B3'] * 4

df = pd.DataFrame({'Drug': drug_levels, 'Block': block_levels})

X = pd.get_dummies(df, columns=['Drug', 'Block'], drop_first=False)
X.insert(0, 'Intercept', 1)

X = X[['Intercept', 'Drug_Control', 'Drug_Dose 1', 'Drug_Dose 2', 'Drug_Dose 3',
      'Block_B1', 'Block_B2', 'Block_B3']]
X
```

Out[]:

	Intercept	Drug_Control	Drug_Dose 1	Drug_Dose 2	Drug_Dose 3	Block_B1	Block_B2	Block_B3
0	1	1	0	0	0	1	0	0
1	1	1	0	0	0	1	0	0
2	1	1	0	0	0	0	0	1
3	1	1	0	0	0	0	0	1
4	1	1	0	0	0	0	0	0
5	1	1	0	0	0	0	0	0
6	1	0	1	0	0	1	0	0
7	1	0	1	0	0	1	0	0
8	1	0	1	0	0	0	0	1
9	1	0	1	0	0	0	0	1
10	1	0	1	0	0	0	0	0
11	1	0	1	0	0	0	0	0
12	1	0	0	1	0	0	1	0
13	1	0	0	1	0	1	0	0
14	1	0	0	1	0	0	0	1
15	1	0	0	1	0	0	0	1
16	1	0	0	1	0	0	0	0
17	1	0	0	1	0	0	0	0
18	1	0	0	0	1	1	0	0
19	1	0	0	0	1	1	0	0
20	1	0	0	0	1	0	0	1
21	1	0	0	0	1	0	0	1
22	1	0	0	0	1	0	0	0
23	1	0	0	0	1	0	0	0

```

In [ ]: import pandas as pd
import numpy as np

response = [
    100.7, 102.4, 104.3, 109.2, 101.7, 99.4, # Control
    103.5, 104.0, 107.7, 105.9, 104.5, 102.3, # Dose 1
    97.8, 94.6, 105.2, 102.9, 98.0, 99.9, # Dose 2
    102.6, 102.2, 106.8, 106.6, 100.1, 96.0 # Dose 3
]

drug = ['Control'] * 6 + ['Dose 1'] * 6 + ['Dose 2'] * 6 + ['Dose 3'] * 6
block = (['B1', 'B1', 'B2', 'B2', 'B3', 'B3']) * 4

df = pd.DataFrame({
    'Response': response,
    'Drug': drug,
    'Block': block
})

X = pd.get_dummies(df[['Drug', 'Block']], drop_first=False)
X.insert(0, 'Intercept', 1)

X_np = X.values
y_np = df['Response'].values

# least squares solution using pseudoinverse
beta_hat = np.linalg.pinv(X_np) @ y_np

coeff_names = X.columns
coefficients = pd.Series(beta_hat, index=coeff_names)
print(coefficients)

Intercept      64.692105
Drug_Control    16.693860
Drug_Dose 1     18.393860
Drug_Dose 2     13.477193
Drug_Dose 3     16.127193
Block_B1        20.109868
Block_B2        25.209868
Block_B3        19.372368
dtype: float64

In [ ]: import numpy as np
import pandas as pd
from numpy.linalg import inv

data = [100.7, 102.4, 104.3, 109.2, 101.7, 99.4, # Control
        103.5, 104.0, 107.7, 105.9, 104.5, 102.3, # Dose 1
        97.8, 94.6, 105.2, 102.9, 98.0, 99.9, # Dose 2
        102.6, 102.2, 106.8, 106.6, 100.1, 96.0] # Dose 3
block = ['B1', 'B1', 'B2', 'B2', 'B3', 'B3'] * 4
drug = ['Control'] * 6 + ['Dose 1'] * 6 + ['Dose 2'] * 6 + ['Dose 3'] * 6

df = pd.DataFrame({
    'Response': data,
    'Block': block,
    'Drug': drug
})

# Columns: Intercept, Drug_Control, Drug_Dose1, Drug_Dose2, Block_B1, Block_
X = np.array([[1, 1, 0, 0, 1, 0], # Control, Block B1
              [1, 1, 0, 0, 1, 0], # Control, Block B1
              [1, 1, 0, 0, 0, 1], # Control, Block B2
              [1, 1, 0, 0, 0, 1], # Control, Block B2
              [1, 1, 0, 0, -1, -1], # Control, Block B3
              [1, 1, 0, 0, -1, -1], # Control, Block B3
              [1, 0, 1, 0, 1, 0], # Dose 1, Block B1
              [1, 0, 1, 0, 1, 0], # Dose 1, Block B1
              [1, 0, 1, 0, 0, 1], # Dose 1, Block B1
              [1, 0, 1, 0, 0, 1], # Dose 1, Block B1
              [1, 0, 1, 0, -1, -1], # Dose 1, Block B1
              [1, 0, 1, 0, -1, -1], # Dose 1, Block B1
              [1, 0, 0, 1, 1, 0], # Dose 2, Block B1
              [1, 0, 0, 1, 1, 0], # Dose 2, Block B1
              [1, 0, 0, 1, 0, 1], # Dose 2, Block B1
              [1, 0, 0, 1, 0, 1], # Dose 2, Block B1
              [1, 0, 0, 1, -1, -1], # Dose 2, Block B1
              [1, 0, 0, 1, -1, -1], # Dose 2, Block B1
              [1, -1, -1, -1, 1, 0], # Dose 3, Block B1
              [1, -1, -1, -1, 1, 0], # Dose 3, Block B1
              [1, -1, -1, -1, 0, 1], # Dose 3, Block B1
              [1, -1, -1, -1, 0, 1], # Dose 3, Block B1
              [1, -1, -1, -1, -1, -1], # Dose 3, Block B1
              [1, -1, -1, -1, -1, -1], # Dose 3, Block B1
])

X_df = pd.DataFrame(X, columns=["Intercept", "Drug_Control", "Drug_Dose1", "
Y = np.array(df['Response'])
beta_hat = np.linalg.inv(X.T @ X) @ X.T @ Y

beta_hat_df = pd.Series(beta_hat, index=["mu", "alpha1", "alpha2", "alpha3",
print("Full-Rank Least Squares Estimate (beta_hat):")
print(beta_hat_df)
print("alpha 4 calculated = - (alpha1 + alpha2 + alpha3) = " + str(-1 * ( be
print("beta 3 calculated = - (b1 + b2) = " + str(-1 * ( beta_hat_df['beta1']

Full-Rank Least Squares Estimate (beta_hat):
mu      102.429167
alpha1    0.520833
alpha2    2.220833
alpha3   -2.695833
beta1    -1.454167
beta2     3.645833
dtype: float64
alpha 4 calculated = - (alpha1 + alpha2 + alpha3) = -0.045833333333333317405
beta 3 calculated = - (b1 + b2) = -2.1916666666666665

```

```

In [ ]: print("alphas in reference to Mean:")
for drug in ['Control', 'Dose 1', 'Dose 2', 'Dose 3']:
    print((df[df['Drug'] == drug].sum()[0] / 6) - 102.429166667)

alphas in reference to Mean:
0.5208333329999988
2.2208333329999874
-2.6958333336666698
-0.04583333366666409

In [ ]: print("betas in reference to mean:")
for block in ['B1', 'B2', 'B3']:
    print((df[df['Block'] == block].sum()[0] / 8) - 102.429166667)

betas in reference to mean:
-1.4541666670000097
3.645833332999999
-2.191666669999927

In [ ]: coefficients # Rank Deficient
beta_hat_df # Full rank

# C1
print("C1")
c1_rd = np.array([0, 1, -1, 0, 0, 0, 0, 0]) # Contrast between Dose 1 and D
c1_fr = np.array([0, 1, -1, 0, 0, 0, 0])
print("RD", np.dot(c1_rd, coefficients)) # First contrast
print("FR", np.dot(c1_fr, beta_hat_df)) # First contrast

# C2
print("C2")
c2_rd = np.array([0, 1, -1/3, -1/3, -1/3, 0, 0, 0]) # Contrast between Dose
c2_fr = np.array([0, 4/3, 0, 0, 0, 0, 0])
print("RD", np.dot(c2_rd, coefficients)) # First contrast
print("FR", np.dot(c2_fr, beta_hat_df)) # First contrast

# C3
c3_rd = np.array([0, 1, 1, 0, 0, -1, -1, 0]) # Contrast between Dose 1 and
c3_fr = np.array([0, 1, 1, 0, 0, -1, -1])
print("RD", np.dot(c3_rd, coefficients)) # First contrast
print("FR", np.dot(c3_fr, beta_hat_df)) # First contrast

C1
RD -1.6999999999999957
FR -1.7000000000000066
C2
RD 0.6944444444444535
FR 0.6944444444444283
RD -10.232017543859627
FR 0.5499999999999843

In [ ]: from scipy.stats import kruskal

group1 = [8, 10, 12, 10, 13, 12, 12, 15, 13, 9]
group2 = [8, 14, 16, 14, 10, 11, 10, 9, 9, 12]
group3 = [11, 12, 11, 23, 19, 11, 17, 17, 16, 16]
group4 = [13, 17, 20, 15, 11, 17, 16, 5, 11, 20]

statistic, p_value = kruskal(group1, group2, group3, group4)

print("Kruskal-Wallis Statistic:", statistic)
print("P-value:", p_value)

Kruskal-Wallis Statistic: 9.27796080658906
P-value: 0.025814669132235304

```

```

In [ ]: import numpy as np
import pandas as pd
from scipy.stats import rankdata, chi2

Group1 = [8, 10, 12, 10, 13, 12, 12, 15, 13, 9]
Group2 = [8, 14, 16, 14, 10, 11, 10, 9, 9, 12]
Group3 = [11, 12, 11, 23, 19, 11, 17, 17, 16, 16]
Group4 = [13, 17, 20, 15, 11, 17, 16, 5, 11, 20]

data = Group1 + Group2 + Group3 + Group4
groups = (['Group1'] * len(Group1) +
          ['Group2'] * len(Group2) +
          ['Group3'] * len(Group3) +
          ['Group4'] * len(Group4))

N = len(data)
k = 4

ranks = rankdata(data)

df = pd.DataFrame({'Data': data, 'Group': groups, 'Rank': ranks})
print(df)

group_stats = df.groupby('Group').agg(n=('Rank', 'size'),
                                      R=('Rank', 'sum'))

print("\nGroup Stats:")
print(group_stats)

H = (12 / (N * (N + 1))) * np.sum(group_stats['R']**2 / group_stats['n']) -
print("\nKruskal-Wallis H statistic (uncorrected):", H)

tie_counts = np.array([np.sum(np.array(data)==val) for val in np.unique(data)])
tie_term = np.sum(tie_counts**3 - tie_counts)
T = 1 - tie_term / (N**3 - N)
print("Tie correction factor T:", T)

H_corrected = H / T
print("Kruskal-Wallis H statistic (corrected):", H_corrected)

df_chi2 = k - 1 # 3

p_value = 1 - chi2.cdf(H_corrected, df_chi2)
print("\np-value:", p_value)

```

	Data	Group	Rank
0	8	Group1	2.5
1	10	Group1	8.5
2	12	Group1	19.0
3	10	Group1	8.5
4	13	Group1	23.0
5	12	Group1	19.0
6	12	Group1	19.0
7	15	Group1	27.5
8	13	Group1	23.0
9	9	Group1	5.0
10	8	Group2	2.5
11	14	Group2	25.5
12	16	Group2	30.5
13	14	Group2	25.5
14	10	Group2	8.5
15	11	Group2	13.5
16	10	Group2	8.5
17	9	Group2	5.0
18	9	Group2	5.0
19	12	Group2	19.0
20	11	Group3	13.5
21	12	Group3	19.0
22	11	Group3	13.5
23	23	Group3	40.0
24	19	Group3	37.0
25	11	Group3	13.5
26	17	Group3	34.5
27	17	Group3	34.5
28	16	Group3	30.5
29	16	Group3	30.5
30	13	Group4	23.0
31	17	Group4	34.5
32	20	Group4	38.5
33	15	Group4	27.5
34	11	Group4	13.5
35	17	Group4	34.5
36	16	Group4	30.5
37	5	Group4	1.0
38	11	Group4	13.5
39	20	Group4	38.5

```

Group Stats:
      n      R
Group
Group1 10 155.0
Group2 10 143.5
Group3 10 266.5
Group4 10 255.0

Kruskal-Wallis H statistic (uncorrected): 9.193536585365877
Tie correction factor T: 0.9909005628517824
Kruskal-Wallis H statistic (corrected): 9.27796080658906

p-value: 0.02581466913223529

```

```
In [ ]: # Total ranks
avg_ranks = []
for group in ['Group1', 'Group2', 'Group3', 'Group4']:
    avg_ranks.append(df[df['Group'] == group].sum()['Rank'] / 10)

total_ranks_mean = sum(avg_ranks) / 4
for group in ['Group1', 'Group2', 'Group3', 'Group4']:
    avg_ranks.append(df[df['Group'] == group].sum()['Rank'])

H = 0
num = 0
denom = 0
for group in ['Group1', 'Group2', 'Group3', 'Group4']:
    num += (10 * (df[df['Group'] == group].sum()['Rank']/10 - total_ranks_mean))
for rank in df['Rank']:
    denom += (rank - total_ranks_mean)**2
H = (40 - 1) * num / denom
H
```

Out[]: 9.277960806589036

```
In [ ]: for group in ['Group1', 'Group2', 'Group3', 'Group4']:
    print(10 * (df[df['Group'] == group].sum()['Rank']/10 - total_ranks_mean))

250.0
378.225
378.22499999999985
250.0
```

```
In [ ]: import numpy as np

Group1 = np.array([8, 10, 12, 10, 13, 12, 12, 15, 13, 9])
Group2 = np.array([8, 14, 16, 14, 10, 11, 10, 9, 9, 12])
Group3 = np.array([11, 12, 11, 23, 19, 11, 17, 17, 16, 16])
Group4 = np.array([13, 17, 20, 15, 11, 17, 16, 5, 11, 20])

def calc_stat(groups):
    """Calculates difference in means (test statistic) between the group means
    means = [np.mean(group) for group in groups]
    return np.max(means) - np.min(means)

observed_stat = calc_stat([Group1, Group2, Group3, Group4])

def permutation_test(groups, n_permutations=10000):
    """Perform permutation test"""
    combined_data = np.concatenate(groups)
    stat_list = []
    for _ in range(n_permutations):
        np.random.shuffle(combined_data)
        new_groups = np.split(combined_data, 4)
        stat_list.append(calc_stat(new_groups))
    return stat_list

perm_stats = permutation_test([Group1, Group2, Group3, Group4])

p_value = np.mean(np.array(perm_stats) >= observed_stat)

print('observed_stat = ', observed_stat)
print('perm_stats example = ', perm_stats[1:4])
print('p_value = ', p_value)

observed_stat = 4.0
perm_stats example = [3.4000000000000004, 3.5999999999999996, 1.0]
p_value = 0.0912
```

```
In [ ]: print(np.mean(Group1))
print(np.mean(Group2))
print(np.mean(Group3))
print(np.mean(Group4))

11.4
11.3
15.3
14.5
```