# ANOVA Testing: Visualization with Python

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I. DEFINITIONS: SST, SSB, SSW

TABLE I

Group $(M=4)$	Observations (There are $N=6$ observations for each group)						Mean	Grand mean
(A)	$a_1 = 7$	$a_2 = 8$	$a_3 = 15$	$a_4 = 11$	$a_5 = 9$	$a_6 = 10$	$\overline{a} = 10$	$\widehat{\mu} = \frac{\overline{a} + \overline{b} + \overline{c} + \overline{d}}{M} \approx 15.9583$
(B)	$b_1 = 12$	$b_2 = 17$	$b_3 = 13$	$b_4 = 18$	$b_5 = 19$	$b_6 = 15$	$\bar{b} = 15.667$	
(C)	$c_1 = 14$	$c_2 = 18$	$c_3 = 19$	$c_4 = 17$	$c_5 = 16$	$c_6 = 18$	$\overline{c} = 17$	
(D)	$d_1 = 19$	$d_2 = 25$	$d_3 = 22$	$d_4 = 23$	$d_5 = 18$	$d_6 = 20$	$\overline{d} = 21.167$	

Denote SST as the *total sum of squares*. Denote SSB as the *between-group sum of squares*. Denote SSW as the *within-group sum of squares*. We have

$$SST = SSB + SSW. (1)$$

Using the numbers provided in the table, we can calculate the SSB as follows:

$$SSB = N \sum_{m=1}^{M} (\widehat{\mu} - \mu_m)^2$$

$$= 6 \times \left[ (15.9583 - 10)^2 + (15.9583 - 15.667)^2 + (15.9583 - 17)^2 + (15.9583 - 21.167)^2 \right]$$

$$\approx 382.8$$
(2)

where  $\mu_m \in \{\overline{a}, \overline{b}, \overline{c}, \overline{d}\}$ , denotes the group mean. Meanwhile, the SSW can be calculated as

$$SSW = \sum_{n=1}^{N} (a_n - \overline{a})^2 + \sum_{n=1}^{N} (b_n - \overline{b})^2 + \sum_{n=1}^{N} (c_n - \overline{c})^2 + \sum_{n=1}^{N} (d_n - \overline{d})^2$$

$$= 40 + 39.33 + 16 + 34.83$$

$$\approx 130.2$$
(3)

As a result, the SST is equal to SST = 382.8 + 130.2 = 513.

### II. ONE-WAY ANOVA TEST

Let us define the two following hypotheses:

**Null hypothesis**  $(H_0): \mu_1 = \mu_2 = ... = \mu_M$ 

**Alternative hypothesis**  $(H_1)$ : At least one group mean is different from the others.

In order to accept/reject  $(H_0)$ , we will compare  $F_0$  to  $F_{\alpha,M-1,M(N-1)}$ . What is  $F_0$  and what is  $F_{\alpha,M-1,M(N-1)}$ ? In the ANOVA,  $F_0$  is the F-test statistic, which can be derived from the SSB and SSW as follows:

$$F_0 = \frac{\text{SSB}/(M-1)}{\text{SSW}/(M(N-1))} = \frac{382.8/(4-1)}{130.2/(4(6-1))} \approx 20$$
 (4)

On the other hand,  $F_{\alpha,M-1,M(N-1)}$  is deduced from the F distribution that has a confidence level of  $\alpha$  and degrees of freedom (M-1) and M(N-1). Note that the p-value is

$$p=1-\alpha=\Pr\{F_0\leq F_{\alpha,M-1,M(N-1)}\}=$$
 probability of accepting  $H_0,$ 

which can be understood as the probability of rejecting  $(H_0)$ . In short, we use the F distribution to find the value  $F_{\alpha,M-1,M(N-1)}$  and then compare it to  $F_0$ .

If  $F_0 > F_{\alpha,M-1,M(N-1)}$ , we can reject **null hypothesis**  $(H_0)$  and accept the **alternative hypothesis**  $(H_1)$ . In the case of  $F_0 \le F_{\alpha,M-1,M(N-1)}$ , we can accept  $(H_0)$  and reject  $(H_1)$ .

#### III. DATA VISUALIZATION WITH PYTHON

### A. DataFrames and Table Styles

In Python, we use Pandas to store the data in a dataframe. The dataframe can be displayed in the form of a table. But, a question arises: what table style should we use to display the dataframe? There are many table styles in Python, thus it depends on our specific purposes. Since we will apply the one-way ANOVA on the data, we want to display the dataframe in the following table style:

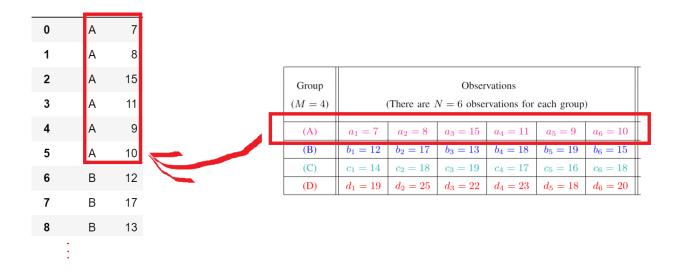


Fig. 1. (On the left side): The table style in which the dataframe is displayed looks like this.

## B. Python Code

In the following, I will present two ways to enter the data into a list and then convert the list into a dataframe (using pandas). Depending on whether the dataframe is displayed in a suitable table style or not, I will employ the pandas.melt method to re-arrange the dataframe into the suitable table style (as seen in the figure above).

1) Way 1: If I create the following dataframe, then its table style will not be the same as the required table style.

Fig. 2. This dataframe is **not** in the table style that is required.

To obtain a suitable dataframe that has the same table style as Fig. 1, I have to use pandas.melt to create the following dataframe.

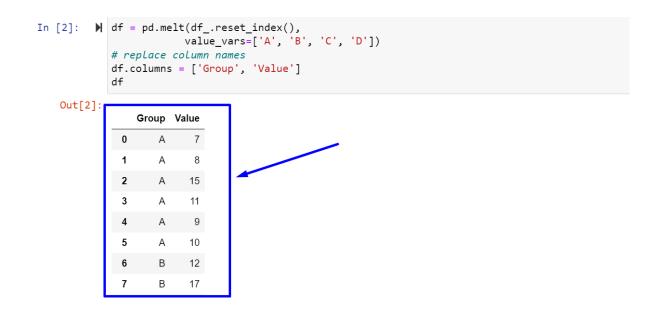


Fig. 3. Now, the new dataframe is in the suitable table style.

2) Way 2: If I create the following dataframe, then its table style will be directly the same as the required table style.

```
import pandas as pd
In [1]:
                    data = [{'Group': 'A', 'Value': 7},
                                 {'Group': 'A', 'Value': 8},
{'Group': 'A', 'Value': 15},
                                 {'Group': 'A', 'Value': 11},
{'Group': 'A', 'Value': 9},
{'Group': 'A', 'Value': 10},
                                 {'Group': 'B', 'Value': 12},
{'Group': 'B', 'Value': 17},
{'Group': 'B', 'Value': 13},
{'Group': 'B', 'Value': 18},
                                 {'Group': 'B', 'Value': 19},
                                 {'Group': 'B', 'Value': 15},
                                 {'Group': 'C', 'Value': 14},
{'Group': 'C', 'Value': 18},
                                 {'Group': 'C', 'Value': 18},
                                 {'Group': 'C', 'Value': 17},
{'Group': 'C', 'Value': 16},
                                 {'Group': 'C', 'Value': 18},
                                 {'Group': 'D', 'Value': 19},
{'Group': 'D', 'Value': 25},
{'Group': 'D', 'Value': 22},
                                 {'Group': 'D', 'Value': 23},
                                 {'Group': 'D', 'Value': 18},
                                 {'Group': 'D', 'Value': 20}]
                   df = pd.DataFrame(data)
                   df
```

Fig. 4. This is a direct way to create the required dataframe.

As a result, it is seen that the dataframe has the required table style.

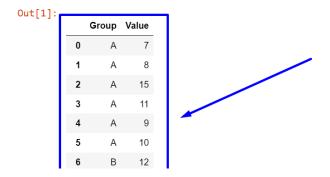


Fig. 5. The required table style that is exactly the same as the one in Figure 1.

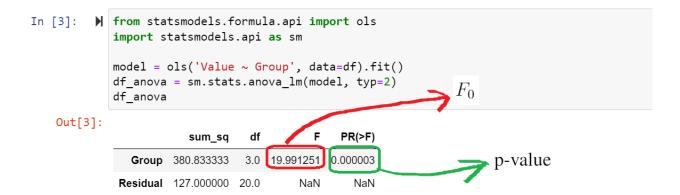
3) Data Visualization: Now, we can portray the data distribution using seaborn.boxplot as can be seen below.

```
# Use boxplot to see the data distribution
In [2]:
             import seaborn as sns
             ax = sns.boxplot(x='Group', y='Value', data=df, color='#9cd000')
             ax = sns.swarmplot(x="Group", y="Value", data=df, color='#7d0000')
                25.0
                22.5
                20.0
                17.5
             Value
                15.0
                12.5
                10.0
                 7.5
                                                 Ċ
                                         Group
```

Fig. 6. The data distribution.

### IV. ONE-WAY ANOVA TEST WITH PYTHON

To calculate  $F_0$  and p-value from the data, I call the statsmodels library and use the following syntax:



Before making a decision, I continue to calculate  $F_{\alpha,M-1,M(N-1)}$ . Assume that  $\alpha=0.01$ , then I obtain  $F_{\alpha,M-1,M(N-1)}\approx 5$ .

```
In [4]: M import scipy.stats M = 4 N = 6 scipy.stats.f.ppf(q=1-0.01, dfn=M-1, dfd=M*(N-1)) Out[4]: 4.938193382310539
```

It is clear that  $F_0 \approx 20 > F_{\alpha,M-1,M(N-1)} \approx 5$ , I reject the null hypothesis  $(H_0)$ . Looking at the p-value, it is readily seen that  $p \approx 3 \times 10^{-6}$  is too small, which means that the probability of accepting  $(H_0)$  is too small. Conclusion is reached!  $(H_0)$  is rejected.