

# STATISTICAL ANALYSIS ON EQUALITY OF IIP FOR FOOD, TOBACCO AND BEVERAGES PRODUCTION IN INDIA CONSIDERING BASE 2011-12 BY USING DISTRIBUTION FREE ADAPTIVE TEST

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#### Abstract

The index of industrial production (IIP) is as index that declares the performance of various industrial sectors of Indian economy. This is calculated and published by Central Statistical Organisation (CSO) every month. This is a composite indicator of the general level of industrial activity in the economy.

Now this study attempts to make a statistical analysis on equality of IIP for food, tobacco and beverages production in India. Secondary data is collected for the study. The conclusion and interpretation are made by the use of adaptive distribution free test.

*Key words:* Beta Generated family, Generalized Marshall-Olkin family, Power weighted moments, AIC, A and W.

# 1. INTRODUCTION

For the month of March 2023, the quick estimates of IIP with base 2011-12 stands at 150.5. According to the report food, tobacco and beverages in India was at 10.68% in 2019. This value is based on the World Bank Collection of development indicators and it was compiled from officially recognised sources. This chapter is based on secondary source and this India – food, tobacco and beverages actual values, historical data, forecasts and projections were sourced from the World Bank on June of 2023.



# 2. OBJECTIVE

The objective of the present study is to test whether the IIP for food, Tobacco and beverages production in India considering base 2011-12 is equal or not. Accordingly, this study is considering the null hypothesis that the IIP of all these three sectors in India is equal against the alternative that they are not equal.

# 3. METHODOLOGY

In this present study we are going to test our null hypothesis by using Adaptive test. This is a distribution free test. In this Adaptive test after collection of data we calculate the measure of skewness and tailweight of the distribution of the data set that we have collected. Then depending upon value of these two selection statistics we decide the test statistics which will help us to give the interpretation on the data set.

# **3.1 Selection Statistics**

Here we will discuss the use of selection statistics  $S = (Q_1, Q_2)$ , where  $Q_1$  and  $Q_2$  are Hogg's measure of skewness and tail weight defined by –

$$Q_1 = \frac{\overline{U}_{5\%} - \overline{M}_{50\%}}{\overline{M}_{50\%} - \overline{L}_{5\%}} \text{ and } Q_2 = \frac{\overline{U}_{5\%} - \overline{L}_{5\%}}{\overline{U}_{50\%} - \overline{L}_{50\%}}$$

Where,  $\overline{U}_{5\%}$ ,  $\overline{M}_{50\%}$  and  $\overline{L}_{5\%}$  are the averages of the upper 5%, middle 50% and lower 5% of the order statistics in the combined sample.  $\overline{U}_{50\%}$  and  $\overline{L}_{50\%}$  are the averages of the upper 50% and lower 50% of the order statistics in the combined sample.

Table 3.1: Theoretical Values of Q1 and Q2 for Some Selected Distributions

Distributions	Q1	Q <sub>2</sub>
Uniform	1	1.9
Normal	1	2.585
Logistic	1	3.204
Double exponential	1	3.302
Exponential	4.569	2.864



Now let us define four categories of S-

$$D_{1} = \{S/0 \le Q_{1} \le 2, 1 \le Q_{2} \le 2\}$$
$$D_{2} = \{S/0 \le Q_{1} \le 2; 2 \le Q_{2} \le 3\}$$
$$D_{3} = \{S/Q_{1} \ge 0; Q_{2} \ge 3\}$$

 $D_4 \!=\! \{S\!/\,Q_1\!\!>\!\!2;\,1\!\!\leq\!\!Q_2\!\leq\!\!3\}$ 

This means that the distribution is short tailed or medium tailed if S falls in the category  $D_1$  or  $D_2$  respectively and is long tailed if S falls in the category  $D_3$  and right skewed if it falls in the category  $D_4$ .

Buning (1996) proposed the following adaptive test A:

$$A = \begin{cases} G & if \ S \in D_1 \\ KW & if \ S \in D_2 \\ LT & if \ S \in D_3 \\ HFR & if \ S \in D_4 \end{cases}$$

Where G, KW, LT and HFR the Gastwrith, Kruskal-Wallis, Long-tailed and Hogg-Fisher-Randle statistics respectively describe in section 3.3.

# **3.3 Test Procedures**

Let us consider,  $X_{i1}, X_{i2}, ..., X_{in_i}$ , i=1, 2,....,c be independent random variables with absolutely continuous distribution function  $F(x-\theta_i)$ .

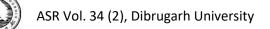
Here our null hypothesis is  $H_0: \theta_1 = \theta_2 = \cdots = \theta_c$ 

and the alternative hypothesis is  $H_1: \theta_r \neq \theta_s$  for at least one pair (r,s),  $r \neq s$ .

# 3.3.1 F-Test

For normally distributed random samples with equal variance, for testing equality of means, the likelihood ratio F test is the best one. The test statistics defined as

$$\mathbf{F} = \frac{(N-c)\sum_{i=1}^{C} n_i (\overline{X_i} - \overline{X})^2}{(C-1)\sum_{i=1}^{C} \sum_{j=1}^{n_i} (X_{ij} - \overline{X_i})^2}$$



where N=
$$\sum_{i=1}^{c} n_i, \bar{X}_i = \frac{1}{n_i} \sum_{j=1}^{n_i} X_{ij}, \bar{X} = \frac{1}{N} \sum_{i=1}^{c} n_i \bar{X}_i$$

Under H<sub>0</sub> the test statistics follows F distribution with c-1 and N-c degrees of freedom.

#### 3.3.2 Kruskal-Wallis (KW) Test

Let  $R_{ij}$  be the rank of the observation  $X_{ij}$  in the pooled sample. The Kruskal-Wallis test for two-sided alternative which based on the statistic

$$\begin{split} \mathbf{KW} &= \frac{12}{N(N+1)} \sum_{i=1}^{k} \frac{1}{n_i} \left[ R_i - \frac{n_i(N+1)}{2} \right]^2 \\ \mathbf{KW} &= \frac{12}{N(N+1)} \sum_{i=1}^{c} \frac{1}{n_i} \left[ R_i - \frac{n_i(N+1)}{2} \right]^2 \\ &= \frac{12}{N(N+1)} \sum_{i=1}^{c} \frac{R_i^2}{n_i} - 3(N+1) \\ &\text{where } R_i = \sum_{j=1}^{n_i} R_{ij} \quad \text{and } \mathbf{N} = \sum_{i=1}^{c} n_i \; . \end{split}$$

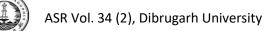
for sample size  $n_i$  and sample number k large,  $H_o$  is rejected if KW> $\chi^2_{\alpha,(k-1)}$ . When k is small and  $n_i$  are small then exact distribution table for KW can be used.

#### 3.3.3 Adaptive Test

Now let us define the suggested Adaptive test which involves, a linear rank statistics. Let us consider a the combined ordered sample  $X_{(1),}X_{(2),...,}X_{(N)}$  of  $X_{11,}...X_{1n_{1,...}}X_{c1,}...X_{cn_{c}}$  and indicator variables  $V_{ik}$  given by

$$V_{ik} = \begin{cases} 1 & if X_{(k)} & belong to the ith sample, i = 1, ..., c, k = 1, ..., N \\ 0 & otherwise \end{cases}$$
  
where N= $\sum_{i=1}^{c} n_i$ ,

Let a(k), k = 1, 2, ..., N be real valued score with mean  $\bar{a} = \frac{1}{N} \sum_{k=1}^{N} a(k)$ . now we define for each sample a statistics  $A_i$  in the following way:



$$A_{i} = \frac{1}{n_{i}} \sum_{k=1}^{N} a(k) V_{ik, i} \leq i \leq c$$

Then the linear rank statistics  $L_N$  is given by

$$L_{\rm N} = \frac{(N-1)\sum_{i=1}^{c} n_i (A_i - \bar{a})^2}{\sum_{k=1}^{N} (a(k) - \bar{a})^2}$$

Under  $H_0$ ,  $L_N$  is distribution free. It follows asymptotically chi- square distribution with c-1 degrees of freedom.

Some of the scores to obtain more powerful test for types of distribution according to Buning(1991,1994) are as follows:

# (i) Gastwrith Test G (Short Tailed):

$$a_{G}(k) = \begin{cases} k - \frac{(N+1)}{4} ifk \le \frac{(N+1)}{4} \\ 0 \quad if \frac{(N+1)}{4} \le k \le \frac{3(N+1)}{4} \\ k - \frac{3(N+1)}{4} ifk \ge \frac{3(N+1)}{4} \end{cases}$$

# (ii) Kruskal Wallis Test KW (Medium Tails):

If 
$$a_{KW}(k) = k$$
, test transform to above KW test

#### (iii) LT Test (Long Tailed):

$$a_{LT} = \begin{cases} -\left(\left[\frac{N}{4}\right] + 1 & ifk < \left[\frac{N}{4}\right] + 1 \\ k - \frac{(N+1)}{2} if\left[\frac{N}{4}\right] + 1 \le k \le \left[\frac{3(N+1)}{4}\right] \\ \left[\frac{N}{4}\right] + 1 & ifk > \left[\frac{3(N+1)}{4}\right] \end{cases}$$



# (iv) Hogg Fisher Randles Test HFR (Right Skewed):

$$a_{HFR} = \begin{cases} k - \frac{(N+1)}{2} ifk \le \frac{(N+1)}{2} \\ 0 & ifk > \frac{(N+1)}{2} \end{cases}$$

For left-skewed distributions we change the terms k-(N+1)/2 and 0 in the definition of the scores above.

# **Collected Data:**

#### **Index of Industrial Production**

	Food	Tobacco	Beverages
Jan-21	111.4	72.4	54.8
Feb-21	111.3	74	82
Mar-21	130.1	96.5	85.2
Apr-21	110.3	95.6	85.3
May-21	117.3	86.9	86.3
Jun-21	118.8	95	86.8
Jul-21	115.3	81.7	89.9
Aug-21	116.2	79.4	90.2
Sep-21	127.8	70	91.5
Oct-21	145.1	108.2	91.5
Nov-21	118.1	75.8	91.8
Dec-21	151.8	101.8	92
Jan-22	147	93.3	92.9
Feb-22	139.9	79.9	94.4
Mar-22	136.7	79.2	94.7
Apr-22	115.6	94.6	95.3
May-22	149	103.7	95.4
Jun-22	155.5	90.6	98.1
Jul-22	113.2	72.2	102
Aug-22	157	92.5	104.3
Sep-22	145.8	77.9	104.7

# Base 2011-12



Oct-22	138.3	114	110.7
Nov-22	150.5	107.3	116.1
Dec-22	132.7	85.4	118.2
Jan -23	117	113	119.5
FEB -23	141.7	96.4	119.9
Mar-23	122.7	87.9	126.2

# 4. CALCULATION AND CONCLUSION:

<b>Q</b> <sub>1</sub>	0.838892
Q <sub>2</sub>	1.738072

According to the values of Q1 and Q2 the collected data falls in the category of D1. That means the data follows the characteristics of a short tailed distribution and as per our theory Gastwirth test (or short tailed test) will be more preferable for the data to proceed further.

The calculated value of test statistics is 208.41 and the tabulated value of the F test statistics with (2,24) df is 3.40 at 5% level of significance. Hence, we may reject our null hypothesis H<sub>0</sub> at 5% level of significance and conclude that the IIP of the Manufactured products namely Food, Tobacco and Beverages are not equal.

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