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## **Extreme Value Charts and Analysis of Means (ANOM) Based on the Log Logistic Distribution**

B. Srinivasa Rao

*R.V.R & J.C. College of Engineering, Guntur, Andhra Pradesh, India*

J. Pratapa Reddy

*St. Ann's College for Women, Guntur, Andhra Pradesh, India*

G. Sarath Babu

*Chebrolu Hanumaiah Institute of Pharmaceutical Sciences, Guntur, Andhra Pradesh, India*

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## Extreme Value Charts and Analysis of Means (ANOM) Based on the Log Logistic Distribution

B. Srinivasa Rao

R. V. R. & J. C. College of  
 Engineering,  
 Guntur, Andhra Pradesh, India

J. Pratapa Reddy

St. Ann's College for Women,  
 Guntur, Andhra Pradesh, India

G. Sarath Babu

Chebrolu Hanumaiah Institute of  
 Pharmaceutical Sciences,  
 Guntur, Andhra Pradesh, India

A probability model of a quality characteristic is assumed to follow a log logistic distribution. This article proposes variable control charts, termed extreme value charts, based on the extreme values of each subgroup. The control chart constants depend on the probability model of the extreme order statistics and the size of each subgroup. The analysis of means (ANOM) technique for a skewed population is applied with respect to log logistic distribution. Results are illustrated using examples based on real data.

Key words: ANOM, LLD, in control, equi-tailed, Q-Q plot.

### Introduction

The probability density function (PDF) of a log logistic distribution (LLD) with shape parameter  $b$  and scale parameter  $\sigma$  is given by

$$f(x, b, \sigma) = \frac{b \left( \frac{x}{\sigma} \right)^{b-1}}{\sigma \left( 1 + \left( \frac{x}{\sigma} \right)^b \right)^2}, \quad x > 0, \sigma > 0, b > 1. \quad (1.0.1)$$

and its cumulative distribution function (CDF) is

$$F(x, b, \sigma) = \frac{\left( \frac{x}{\sigma} \right)^b}{1 + \left( \frac{x}{\sigma} \right)^b}, \quad x > 0, \sigma > 0, b > 1. \quad (1.0.2)$$

When  $\sigma = 1$  and  $b > 1$  these equations are termed standard PDF and CDF. In order to construct a control chart using extreme observations of a subgroup drawn from a production process with the quality variate following a LLD, the percentiles of extreme order statistics from LLD samples are needed. Specifically, the test statistic on the extreme value control chart is the original sample vector  $X = (x_1, x_2, \dots, x_n)$  from ongoing production. In this chart all individual sample observations are plotted into the control chart without calculating any statistics. A corrective action is taken after one, or either, of the extreme values – namely  $x_{(1)}$  (sample minimum) and  $x_{(n)}$  (sample maximum) – of the sample respectively fall above or below specified lines (limits); this is why the chart is called an extreme value controlled chart.

The Shewhart (1986) controlled chart is a common quality control statistical tool: When a Shewhart chart indicate the presence of an assignable cause, a process adjustment can be made if the remedy is known; otherwise the

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B. Srinivasa Rao is an Associate Professor of Statistics in the Department of Mathematics and Humanities. Email him at: boyapatisirinu@yahoo.com. J. Pratapa Reddy is an Associate Professor of Statistics in the Department of Computer Applications. Email him at: jakkampratapa@yahoo.co.in. G. Sarath Babu is an Assistant Professor of Statistics. Email him at: gorantla.sarath@gmail.com.

suspected presence of assignable cause is regarded as an indication of heterogeneity of the subgroup statistic for which the control chart was developed. For example, if the statistic is the sample mean, this leads to heterogeneity of the process mean and indicates departures from the target mean. Such an analysis is generally carried out by dividing a collection of a given number of subgroup means into categories, such that means within a category are homogenous and those between categories are heterogeneous. This procedure, developed by Ott (1967) is called analysis of means (ANOM).

When using the ANOM technique the concept of a control chart for means is viewed differently, grouping of plotted means that fall within or outside control limits. For the homogeneity of the means it is necessary that all means fall within the control limits. If  $(1 - \alpha)$  is the confidence coefficient, then the probability that all subgroup means will fall within the control limits is  $(1 - \alpha)$ . Assuming independence of the subgroup, the probability statement becomes  $n^{\text{th}}$  power of the probability that a subgroup mean will fall within the limits. This means that, in the sampling distribution of  $\bar{x}$ , the confidence interval for  $\bar{x}$  to lie between two specified limits should be equal to  $(1 - \alpha)^{1/n}$ . This same principle is adapted through log logistic distribution in this study. This article explores ANOM using control limits of extreme value statistics considering only control chart aspects. (See Rao (2005) for a detailed description of ANOM; other related works include: Ramig, 1983; Bakir, 1994; Bernard & Wludyka, 2001; Wludyka, et al., 2001; Montgomery, 2001; Nelson & Dudewicz, 2002; Rao & Prankumar, 2002; Farnum, 2004; Guirguis & Tobias, 2004; Srinivasa Rao & Kantam, 2012.)

#### Extreme Value Charts

The given sample observations are assumed to follow log logistic mode. The controlled lines are determined by the theory of extreme order statistics based on a half logistic model. The controlled lines are determined in such a way that an arbitrarily chosen  $x_i$  of  $X = (x_1, x_2, \dots, x_n)$  lies within the limits  $(1 - \alpha)^{1/n}$  with the probability  $(1 - \alpha)^{1/n}$ . This can be formulated as a

probability inequality as:  $P(x_i \leq L) = \alpha/2$  and  $P(x_i \geq U) = \alpha/2$ . The theory of order statistics states that the cumulative distribution function of the least and highest order statistics in a sample of size  $n$  from any continuous population are  $[F(x)]^n$  and  $1-[1-F(x)]^n$ , respectively, where  $F(x)$  is a cumulative distribution function (CDF) of the population. If  $1-\alpha$  is desired at 0.9973, then  $\alpha$  would equal 0.0027. Taking  $F(x)$  as the CDF of a standard log logistic model results in solutions of the equations  $1-[1-F(x)]^n = 0.00135$  and  $[F(x)]^n = 0.99865$  which, in turn, can be used to develop the controlled limits of an extreme value chart. The solutions for the two equations for  $n = 2$  (1) 10 with  $b = 2, 3, 4$  and 5 are shown in table 2.1 and denoted as  $Z_* = Z_{(1)0.00135}$  and  $Z^{**} = Z_{(n)0.99865}$ .

The values shown in table 2.1 indicate the following probability statements:

$$P\left(\frac{Z_{(1)0.00135} < Z_i < Z_{(n)0.99865}}{\forall i=1,2,\dots,n}\right) = 0.9973 \quad (2.0.3)$$

and

$$P\left(\frac{\sigma Z_{(1)0.00135} < x_i < \sigma Z_{(n)0.99865}}{\forall i=1,2,\dots,n}\right) = 0.9973 \quad (2.0.4)$$

Taking  $\frac{\bar{x}}{1.5708}, \frac{\bar{x}}{1.0472}, \frac{\bar{x}}{0.7854}$  and  $\frac{\bar{x}}{0.6283}$  as unbiased estimates of  $\sigma$  when  $b = 2, b = 3, b = 4$  and  $b = 5$ , respectively, the equation becomes

$$P\left(\frac{D_3^* \bar{x} < x_i < D_4^* Z_{(n)0.99865}}{\forall i=1,2,\dots,n}\right) = 0.9973 \quad (2.0.5)$$

where, for  $b = 2$ :

$$D_3^* = \frac{Z(1)(0.00135)}{1.5708}$$

and

$$D_4^* = \frac{Z(n)(0.99865)}{1.5708} \quad \text{and} \quad D_4^* = \frac{Z(n)(0.99865)}{0.7854}.$$

For b = 3:

$$D_3^* = \frac{Z(1)(0.00135)}{1.0472}$$

and

$$D_4^* = \frac{Z(n)(0.99865)}{1.0472}.$$

For b = 4:

$$D_3^* = \frac{Z(1)(0.00135)}{0.7854}$$

For b = 5:

$$D_3^* = \frac{Z(1)(0.00135)}{0.6283}$$

and

$$D_4^* = \frac{Z(n)(0.99865)}{0.6283}.$$

Thus,  $D_3^*$  and  $D_4^*$  constitute the control chart constants for the extreme value charts (see Table 2.2 for n = 2(1)10).

Table 2.1: Control Chart Limits of Extreme Value Charts

n	b=2		b=3		b=4		b=5	
	Z*	Z**	Z*	Z**	Z*	Z**	Z*	Z**
2	0.0259	38.4705	0.0877	11.3959	0.1612	6.2024	0.2322	4.3057
3	0.0212	47.1192	0.0766	13.0456	0.1456	6.8643	0.2141	4.6695
4	0.0183	54.4101	0.0696	14.3588	0.1355	7.3763	0.2021	4.9462
5	0.0164	60.8334	0.0646	15.4677	0.1282	7.7995	0.1933	5.1719
6	0.0150	66.6404	0.0608	16.4371	0.1225	8.1633	0.1864	5.3640
7	0.0138	71.9804	0.0577	17.3038	0.1178	8.4841	0.1807	5.5320
8	0.0129	76.9508	0.0552	18.0915	0.1139	8.7721	0.1760	5.6817
9	0.0122	81.6190	0.0531	18.8160	0.1106	9.0343	0.1719	5.8172
10	0.0116	86.0343	0.0513	19.4886	0.1078	9.2754	0.1683	5.9411

# EXTREME VALUE CHARTS AND ANOM BASED ON LOG LOGISTIC DISTRIBUTION

Analysis of Means (ANOM): Log Logistic Distribution

When the data variate follows log logistic distribution, suppose  $\bar{x}_1, \bar{x}_2, \dots, \bar{x}_k$  are arithmetic means of  $k$  subgroups of size  $n$  drawn from a log logistic model. The subgroups means are used to develop control charts to assess whether the population from which these subgroups are drawn is operating with admissible quality variations. Depending on the basic population model, control chart constants may be used. In general, the process may be said to be in control if all subgroup means are within the control limits; otherwise the process is said to lack control. If  $\alpha$  is the level of significance of this decision, the following probability statements apply:

$$P(LCL < x_i, \forall i=1,2, \dots, k < UCL) = 1 - \alpha \quad (3.0.6)$$

using the notion of independent subgroups,

(3.0.6) becomes

$$P(LCL < x_i < UCL) = (1 - \alpha)^{1/k} \quad (3.0.7)$$

With equi-tailed probability for each subgroup mean, two constants, for example  $L^*$  and  $U^*$ , may be found such that

$$P(x_i < L^*) = P(\bar{x}_i > U^*) = \frac{1 - (1 - \alpha)^{1/k}}{2} \quad (3.0.8)$$

For skewed populations, such as the LLD, it is necessary to calculate  $L^*$ ,  $U^*$  separately from the sampling distribution of  $\bar{x}_i$ . Accordingly, these depend on the subgroup size  $n$  and number of subgroups  $k$ . The percentiles of the sampling distribution of  $\bar{x}$  in samples from a log logistic distribution for  $b = 2, b = 3, b = 4$  and  $b = 5$  with  $\sigma = 1$  were calculated using Monte-Carlo simulations (see Tables 3.1, 3.2, 3.3 and 3.4).

Table 2.2: Control Chart Limits of Extreme Value Charts

n	b = 2		b = 3		b = 4		b = 5	
	$D_3^*$	$D_4^*$	$D_3^*$	$D_4^*$	$D_3^*$	$D_4^*$	$D_3^*$	$D_4^*$
2	0.0165	24.4910	0.0837	10.8823	0.2052	7.8971	0.3695	6.8529
3	0.0135	29.9969	0.0731	12.4576	0.1854	8.7398	0.3423	7.4319
4	0.0116	34.6384	0.0664	13.7116	0.1726	9.3917	0.3216	7.8723
5	0.0104	38.7276	0.0617	14.7705	0.1632	9.9306	0.3076	8.2315
6	0.0095	42.4246	0.0580	15.6962	0.1559	10.3938	0.2966	8.5373
7	0.0088	45.8240	0.0551	16.5238	0.1500	10.8022	0.2876	8.8047
8	0.0082	48.9882	0.0527	17.2760	0.1450	11.1690	0.2801	9.0429
9	0.0078	51.9601	0.0507	17.9679	0.1408	11.5028	0.2735	9.2586
10	0.0074	54.7710	0.0489	18.6102	0.1372	11.8098	0.2678	9.4558

Table 3.1: Percentiles of Sample Mean in LLD with  $b = 2$ 

n	0.99865	0.99	0.975	0.95	0.05	0.025	0.01	0.00135
2	12.9560	4.8648	3.2368	2.363	0.2646	0.2144	0.1624	0.1002
3	11.3630	4.4994	3.0711	2.3002	0.3390	0.2887	0.2368	0.1772
4	10.5971	3.8004	2.7757	2.0875	0.3848	0.3313	0.2767	0.2002
5	9.8124	3.5276	2.5244	1.9696	0.4228	0.3709	0.3180	0.2237
6	9.6424	3.47156	2.4932	1.9646	0.4518	0.4008	0.3446	0.2658
7	7.0472	3.2050	2.3660	1.9001	0.4692	0.4176	0.3662	0.2896
8	6.5408	2.9775	2.2500	1.8469	0.4965	0.4435	0.3914	0.3202
9	5.8094	2.8794	2.2181	1.8242	0.5144	0.4681	0.4201	0.3434
10	6.1672	2.7692	2.2176	1.8063	0.5284	0.4827	0.4331	0.3635

Table 3.2: Percentiles of Sample Mean in LLD with  $b = 3$ 

n	0.99865	0.99	0.975	0.95	0.05	0.025	0.01	0.00135
2	4.8927	3.1605	2.6322	2.2986	0.7921	0.7115	0.6210	0.4927
3	4.3935	2.7927	2.4511	2.1816	0.8913	0.8211	0.7476	0.6406
4	3.3256	2.6008	2.2806	2.0491	0.9423	0.8750	0.8007	0.6739
5	3.1616	2.4128	2.1578	1.9784	0.9857	0.9256	0.8598	0.7290
6	3.2322	2.3737	2.1150	1.9448	1.0203	0.9629	0.8933	0.7799
7	2.9030	2.2476	2.0506	1.9071	1.0350	0.9802	0.9162	0.8115
8	2.7292	2.1760	2.0073	1.8728	1.0608	1.0041	0.9447	0.8522
9	2.5742	2.1361	1.9732	1.8443	1.0779	1.0272	0.9693	0.8833
10	2.5429	2.1235	1.9520	1.8268	1.0961	1.0490	0.9930	0.9017

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Table 3.3: Percentiles of Sample Mean in LLD with  $b = 4$

n	0.99865	0.99	0.975	0.95	0.05	0.025	0.01	0.00135
2	6.2330	3.4078	2.6332	2.1740	0.5154	0.4459	0.3739	0.2719
3	5.8539	2.9913	2.4142	2.0571	0.6056	0.5447	0.4763	0.3970
4	4.0852	2.6589	1.2396	1.8961	0.6549	0.5949	0.5293	0.4205
5	4.0895	2.4914	2.0797	1.8209	0.6973	0.6402	0.5764	0.4649
6	3.9497	2.4311	2.0150	1.7868	0.7278	0.6728	0.6120	0.5129
7	3.3344	2.2616	1.9481	1.7396	0.7435	0.6919	0.6317	0.5374
8	3.1361	2.1661	1.8742	1.7040	0.7702	0.7172	0.6603	0.5799
9	2.8196	2.0899	1.8578	1.6766	0.7869	0.7402	0.6875	0.6204
10	2.9277	2.1209	1.8353	1.6523	0.8039	0.7567	0.7036	0.6275

Table 3.4: Percentiles of Sample Mean in LLD with  $b = 5$

n	0.99865	0.99	0.975	0.95	0.05	0.025	0.01	0.00135
2	4.5322	3.2326	2.8121	2.5223	1.0811	0.9880	0.8909	0.7344
3	4.1151	2.9163	2.6339	2.4091	1.1882	1.1093	1.0315	0.9027
4	3.3010	2.7435	2.4671	2.2876	1.2398	1.1642	1.0868	0.9514
5	3.0996	2.5609	2.3640	2.2206	1.2812	1.2195	1.1485	1.0053
6	2.9154	2.4838	2.3201	2.1834	1.3056	1.2545	1.1927	1.0291
7	2.8736	2.4256	2.2716	2.1508	1.3344	1.2759	1.2103	1.0960
8	2.7778	2.3577	2.2310	2.1206	1.3587	1.3036	1.2420	1.1449
9	2.6438	2.3260	2.1990	2.0882	1.3776	1.3254	1.2603	1.1749
10	2.6299	2.3118	2.1803	2.0749	1.3973	1.3463	1.2899	1.1963

The percentiles shown in Tables 3.1 – 3.4 are used in equation (3.0.8) for specified  $n$  and  $k$  to determine  $L^*$  and  $U^*$  for  $\alpha = 0.05$  (see Tables 3.5, 3.6, 3.7 and 3.8). A control chart for averages showing in control conclusions indicates that all subgroups means, though varying among themselves, are homogenous in

some cells. This is the null hypothesis in an analysis of variance technique, hence, the constants shown in tables 3.5 - 3.8 can be used as an alternative to analysis of variance techniques. For a normal population Ott's (1967) tables can be used, and for a LLD the tables shown herein can be used.

Table 3.5: LLD Constants for Analysis of Means for  $b = 2$ ,  $(1-\alpha) = 0.95$ 

$k$	$n$									
	2		3		4		5		6	
1	0.3303	5.1477	0.4539	4.7591	0.5212	4.3511	0.5768	4.0660	0.6266	3.9113
2	0.2639	7.0248	0.9864	6.6701	0.4599	5.8876	0.5161	5.4263	0.5798	4.9528
3	0.2365	8.9046	0.3569	7.7521	0.4292	6.9287	0.4819	6.2444	0.5517	5.7732
4	0.2192	10.4238	0.3355	8.9742	0.4080	7.7136	0.4591	7.0343	0.5307	6.3341
5	0.2068	11.5178	0.3169	9.5190	0.3742	8.4092	0.4451	7.4928	0.5140	7.3627
6	0.2015	12.4982	0.3054	10.1839	0.3697	9.1787	0.4303	8.0430	0.4153	7.5537
7	0.1972	12.9499	0.3007	10.8149	0.3592	9.9402	0.4199	8.7234	0.4779	7.8955
8	0.1890	13.4191	0.2950	11.0937	0.3544	10.4101	0.4131	8.8267	0.4708	8.3140
9	0.1866	13.9979	0.2931	12.1805	0.3532	10.8970	0.4056	9.1813	0.4665	8.8593
10	0.1716	14.6854	0.2626	12.4166	0.3374	11.0593	0.3878	9.6395	0.4592	9.0119
20	0.1331	17.9113	0.2247	16.9479	0.3130	14.8274	0.3423	12.3236	0.4244	11.9848
30	0.1028	23.7632	0.2097	20.7346	0.3016	18.0296	0.3300	14.8560	0.3885	13.4620
40	0.0946	33.6935	0.1780	24.0424	0.2751	19.7351	0.3118	16.4754	0.3687	16.1631
50	0.0929	34.7754	0.1686	25.1053	0.2745	22.5189	0.3088	18.1307	0.3637	16.4374

$k$	$n$							
	7		8		9		10	
1	0.6592	3.6837	0.6945	3.6131	0.7253	3.5156	0.7536	3.4029
2	0.5999	4.6640	0.6423	4.5937	0.6657	4.4740	0.6873	4.2973
3	0.5712	5.4824	0.6103	5.2898	0.6424	5.1286	0.6513	4.8740
4	0.5561	5.9729	0.5949	5.7324	0.6134	5.7535	0.6309	5.0412
5	0.5378	6.7590	0.5838	6.2880	0.6025	6.2461	0.6199	5.8142
6	0.5280	7.1752	0.5682	6.9044	0.5905	6.6190	0.6140	6.0589
7	0.5243	7.7579	0.5596	7.2190	0.5833	6.9793	0.5962	6.3483
8	0.5224	7.9220	0.5499	7.6895	0.5821	7.2779	0.5916	6.6536
9	0.5213	8.6406	0.5493	7.8990	0.5800	7.6447	0.5863	7.0156
10	0.5022	8.6938	0.5437	8.0013	0.5672	7.7415	0.5752	7.0463
20	0.4771	11.6854	0.5106	10.6656	0.5192	9.8981	0.5427	9.9847
30	0.4510	13.1287	0.4994	12.2854	0.5059	11.3192	0.5237	10.3004
40	0.4249	14.5202	0.4815	14.1756	0.4751	13.1701	0.4990	12.7299
50	0.4242	16.1104	0.4777	14.5343	0.4677	13.2737	0.4972	13.4114

## EXTREME VALUE CHARTS AND ANOM BASED ON LOG LOGISTIC DISTRIBUTION

 Table 3.6: LLD Constants for Analysis of Means for  $b = 3$ ,  $(1-\alpha) = 0.95$ 

k	<i>n</i>									
	2		3		4		5		6	
1	0.4695	2.7584	0.5683	2.5524	0.6248	2.3464	0.6666	2.2172	0.6988	2.1238
2	0.4059	3.3751	0.5101	3.0785	0.5720	2.7479	0.6177	2.5241	0.6660	2.3903
3	0.3733	3.8157	0.4875	3.3458	0.5469	2.9638	0.5929	2.8291	0.6448	2.5548
4	0.3593	4.1241	0.4584	3.6189	0.5203	3.0344	0.5695	2.9484	0.6255	2.6804
5	0.3427	4.5531	0.4451	3.8694	0.5063	3.4598	0.5526	0.0452	0.6095	2.8544
6	0.3388	4.7973	0.4414	3.9541	0.4956	3.5578	0.5437	3.1117	0.5999	2.9205
7	0.3368	4.9647	0.4295	4.0572	0.4881	3.6225	0.5349	3.2163	0.5868	3.0714
8	0.3145	5.2882	0.4167	4.1558	0.4787	3.6740	0.5208	3.3446	0.5772	3.1013
9	0.3136	5.3570	0.4015	4.2656	0.4700	3.7688	0.5152	3.3985	0.5730	3.2142
10	0.3099	5.4002	0.3986	4.3335	0.4645	3.8646	0.5135	3.4171	0.5712	3.2412
20	0.2624	6.1439	0.3477	5.6020	0.4415	4.6598	0.4769	4.1357	0.5375	3.6726
30	0.2455	8.6207	0.3273	6.1971	0.4184	5.0747	0.4486	4.4602	0.5233	4.0896
40	0.2144	8.7280	0.2993	6.3545	0.4108	5.4735	0.4358	4.5999	0.5046	4.3879
50	0.2044	9.0887	0.2835	6.6497	0.3955	5.7045	0.4300	4.6674	0.5240	4.4110

k	<i>n</i>							
	7		8		9		10	
1	0.7261	2.0437	0.7474	1.9820	0.7730	1.9419	0.7902	1.9098
2	0.6783	2.2991	0.7126	2.2268	0.7322	2.1810	0.7442	2.1333
3	0.6559	2.4746	0.6947	2.3405	0.7071	2.3472	0.7205	2.2488
4	0.6404	2.5873	0.6761	2.4314	0.6957	2.4292	0.7071	2.3594
5	0.6281	2.6892	0.6663	2.8787	0.6826	2.5798	0.6962	2.4315
6	0.6225	2.8124	0.6627	2.6851	0.6744	2.6859	0.6848	2.5297
7	0.6145	2.8789	0.6419	2.7863	0.6701	2.7588	0.6824	2.5859
8	0.6110	3.0330	0.6362	2.8695	0.6638	2.7963	0.6731	2.6091
9	0.6087	3.1342	0.6344	2.9247	0.6527	2.8245	0.6703	2.6647
10	0.6048	3.1402	0.6317	2.9568	0.6522	2.8585	0.6652	2.7014
20	0.5782	3.6224	0.6163	3.2793	0.6165	3.1551	0.6323	2.9422
30	0.5664	3.9450	0.6036	3.6797	0.6092	3.6316	0.6161	3.3713
40	0.5332	4.0987	0.5951	3.8300	0.5934	3.5372	0.6112	3.3857
50	0.5277	4.2365	0.5902	3.9501	0.5824	3.6456	0.6035	3.4009

Table 3.7: LLD Constants for Analysis of Means for  $b = 4$ ,  $(1-\alpha) = 0.95$ 

k	<i>n</i>									
	2		3		4		5		6	
1	0.5538	2.0708	0.6462	1.9460	0.6916	1.8023	0.7255	1.7171	0.7522	1.6570
2	0.4998	2.3840	0.5967	2.1913	0.6472	1.9817	0.6821	1.8793	0.7205	1.8782
3	0.4692	2.5861	0.5711	2.3067	0.6235	2.0966	0.6641	1.9984	0.7016	1.8626
4	0.4501	2.7621	0.5495	2.4355	0.6032	2.2188	0.6423	2.0633	0.6858	1.9325
5	0.4424	2.9285	0.5382	2.5761	0.5927	2.3001	0.6290	2.1164	0.6741	1.9779
6	0.4363	2.9917	0.5261	2.6173	0.5881	2.3841	0.6182	2.1564	0.6696	2.0239
7	0.4280	3.0573	0.5194	2.6437	0.5769	2.4234	0.6101	2.1760	0.6782	2.0083
8	0.4122	3.1024	0.5090	2.6733	0.5636	2.4560	0.6063	2.2113	0.6521	2.0791
9	0.4077	3.1713	0.5000	2.7301	0.5562	2.4843	0.5925	2.2169	0.6491	2.1297
10	0.4024	3.2275	0.4962	2.7719	0.5518	2.5069	0.5880	2.2393	0.6472	2.1553
20	0.3486	3.5220	0.4395	3.2248	0.5232	2.7649	0.5633	2.4762	0.6228	2.3591
30	0.3057	4.0758	0.4203	0.5425	0.5083	3.0566	0.5348	2.5192	0.5973	2.3924
40	0.2952	4.3930	0.3891	3.6392	0.4958	3.1190	0.5131	2.6812	0.5904	2.5104
50	0.2793	4.4353	0.3874	3.7004	0.4843	3.1396	0.5121	2.7258	0.5870	2.5539

k	<i>n</i>							
	7		8		9		10	
1	0.7698	1.6077	0.7891	1.5732	0.8069	1.5495	0.8201	1.5341
2	0.7335	1.7229	0.7605	1.6883	0.7728	1.6479	0.7835	1.6446
3	0.7123	1.8107	0.7444	1.7368	0.7550	1.7385	0.7666	1.6908
4	0.7018	1.8610	0.7310	1.5922	0.7477	1.7852	0.7558	0.7328
5	0.6192	1.9104	0.7250	1.8373	0.7378	1.8349	0.7467	1.7653
6	0.6841	1.5945	0.7181	1.8821	0.7316	1.8768	0.7398	1.7819
7	0.7104	1.9209	0.7298	1.9051	0.7338	1.8111	0.7552	1.5561
8	0.6744	2.0342	0.6980	1.9459	0.7190	1.9222	0.7306	1.8189
9	0.6711	2.0588	0.6964	1.9618	0.7152	1.9513	0.7281	1.8303
10	0.6693	2.0766	0.6943	1.9997	0.7093	1.9538	0.7224	1.8422
20	0.6504	2.2074	0.6809	2.1062	0.6810	2.0959	0.6985	1.9428
30	0.6345	2.3967	0.6672	2.2398	0.6731	2.1668	0.6848	2.0243
40	0.6109	2.4582	0.6621	2.3577	0.6608	2.1886	0.6739	2.0624
50	0.6092	2.5502	0.6585	2.3803	0.6519	2.2144	0.6729	2.0980

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 Table 3.8: LLD Constants for Analysis of Means for  $b = 5$ ,  $(1-\alpha) = 0.95$ 

k	<i>n</i>									
	2		3		4		5		6	
1	0.6183	1.7596	0.7003	1.6595	0.7380	1.5590	0.7661	1.5036	0.7871	1.4560
2	0.5683	1.9464	0.6561	1.8041	0.6999	1.6716	0.7287	1.6030	0.7602	1.5422
3	0.5418	2.0700	0.6306	1.8956	0.6762	1.7528	0.7123	1.6647	0.7428	1.5898
4	0.5224	2.1907	0.6141	1.9736	0.6657	1.8008	0.6928	1.7119	0.7313	1.6292
5	0.5175	2.2863	0.6031	2.0374	0.6534	1.8694	0.6816	1.7496	0.7201	1.6586
6	0.5076	2.3282	0.5957	2.0855	0.6480	1.9026	0.6732	1.7694	0.7145	1.6720
7	0.5044	2.3661	0.5802	2.1047	0.6395	1.9335	0.6691	1.7829	0.7093	1.6910
8	0.4913	2.4061	0.5747	2.1234	0.6248	1.9751	0.6587	1.7873	0.7057	1.7134
9	0.4858	2.4285	0.5708	2.1327	0.6195	2.0120	0.6550	1.8075	0.7002	1.7293
10	0.4778	2.4406	0.5634	2.1710	0.6175	2.0183	0.6414	1.8256	0.6942	1.7310
20	0.4292	2.6741	0.5022	2.3900	0.5846	2.1100	0.6218	1.9631	0.6724	1.8581
30	0.3870	2.9332	0.4963	2.6166	0.5771	2.2720	0.5977	2.0171	0.6607	1.9060
40	0.3741	2.9979	0.4683	2.6941	0.5573	2.3353	0.5760	2.0587	0.6469	1.9297
50	0.3563	3.0159	0.4598	2.7000	0.5521	2.3364	0.5729	2.0806	0.6426	1.9329

k	<i>n</i>							
	7		8		9		10	
1	0.8032	1.4224	0.8184	1.3976	0.8342	1.3850	0.8433	1.3707
2	0.7721	1.4942	0.7943	1.4671	0.8050	1.4502	0.8148	1.4294
3	0.7546	1.5472	0.7805	1.5109	0.7900	1.4973	0.8000	1.4701
4	0.7436	1.5846	0.7699	1.5274	0.7838	1.5264	0.7906	1.4873
5	0.7366	1.6084	0.7647	1.5548	0.7760	1.5550	0.7837	1.5172
6	0.7297	1.6367	0.7592	1.5695	0.7694	1.5742	0.7753	1.5216
7	0.7256	1.6484	0.7497	1.5892	0.7669	1.5986	0.7737	1.5360
8	0.7213	1.6674	0.7437	1.6120	0.7580	1.6170	0.7684	1.5413
9	0.7154	1.6722	0.7402	1.6340	0.7507	1.6787	0.7655	1.5481
10	0.7131	1.6751	0.7396	1.6457	0.7500	1.6457	0.7616	1.5503
20	0.6973	1.7975	0.7269	1.7162	0.7264	1.6930	0.7450	1.6237
30	0.6866	1.8671	0.7179	1.8126	0.7159	1.7138	0.7351	1.6402
40	0.6674	1.9378	0.7092	1.8794	0.7115	1.7430	0.7326	1.6644
50	0.6673	1.9386	0.7085	1.8884	0.7012	1.7599	0.7163	1.6892

**Example 1 (Wadsworth, 1986)**

The following 25 observations are from a metal product manufacturing site. Variations in iron content were suspected in raw material supplied by 5 different suppliers. Five ingots were randomly selected from each of the suppliers. The following table contains the iron determinations for each ingot by weight from each of the 5 suppliers.

	Supplier				
	1	2	3	4	5
Ingot Iron Content (g)	3.46	3.59	3.51	3.38	3.29
	3.48	3.46	3.64	3.4	3.46
	3.56	3.42	3.46	3.37	3.37
	3.39	3.49	3.52	3.46	3.32
	3.4	3.5	3.49	3.39	3.38

**Example 2**

In a study of 3 brands of batteries, it was suspected that the life (in weeks) of the three brands was different. Five of each brand of battery were tested with the following results:

Battery Life (weeks)		
Brand 1	Brand 2	Brand 3
100	76	108
96	80	100
92	75	96
96	84	98
92	82	100

**Example 3**

Four catalysts that may affect the concentration of a component in a three component liquid mixture were investigated.

The following concentrations were obtained:

Catalyst			
1	2	3	4
58.2	56.3	50.1	52.9
57.2	54.5	54.2	49.9
58.4	57.0	55.4	50.0
55.8	55.3	54.9	51.7

The goodness of fit of data, as revealed by a Q-Q plot (correlation coefficient), for the 3 examples are summarized Table 3.9, which shows that the log logistic distribution is a better model than the normal because it exhibits a significant linear relation between sample and population quantiles.

**Table 3.9: Goodness of Fit Data from Q-Q Plot**

Example	b	LLD	Normal
1	2	0.9306	0.2067
	3	0.9673	
	4	0.9801	
	5	0.9854	
2	2	0.8484	0.4149
	3	0.8986	
	4	0.9206	
	5	0.9324	
3	2	0.8424	0.2067
	3	0.8981	
	4	0.9223	
	5	0.9351	

Treating the observations in the data as single samples, the decision limits for the normal and the LLD populations were calculated and are shown in Tables 3.10 and 3.11 respectively.

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Table 3.10: Normal Distribution

Examples	[LDL, UDL] (Ott, 1967)	Number of Counts			
		In	P = in/k	Out	Out/k
1: n=5, k=5, $\alpha=0.05$	[3.517, 3.879]	3	0.6	2	0.4
2: n=5, k=3, $\alpha=0.05$	[87.82, 95.52]	2	0.7	1	0.3
3: n=4, k=4, $\alpha=0.05$	[26.14, 82.84]	2	0.5	2	0.5

Table 3.11: Log Logistic Distribution

Examples	[LDL, UDL]	Number of Counts			
		In	P=in/k	Out	Out/k
<b>b = 2</b>					
1 n=5, k=5, $\alpha=0.05$	[1.5345, 25.8322]	5	1	0	0
2 n=5, k=3, $\alpha=0.05$	[44.1805, 572.4044]	3	1	0	0
3 n=4, k=4, $\alpha=0.05$	[22.2330, 420.2953]	4	1	0	0
<b>b = 2</b>					
1 n=5, k=5, $\alpha=0.05$	[1.9053, 10.4989]	5	1	0	0
2 n=5, k=3, $\alpha=0.05$	[54.3510, 259.3351]	3	1	0	0
3 n=4, k=4, $\alpha=0.05$	[28.3547, 180.0506]	4	1	0	0
<b>b = 3</b>					
1 n=5, k=5, $\alpha=0.05$	[2.1685, 7.2966]	5	1	0	0
2 n=5, k=3, $\alpha=0.05$	[60.8822, 183.1867]	3	1	0	0
3 n=4, k=4, $\alpha=0.05$	[32.8685, 120.9012]	4	1	0	0
<b>b = 5</b>					
1 n=5, k=5, $\alpha=0.05$	[2.3499, 6.0319]	5	1	0	0
2 n=5, k=3, $\alpha=0.05$	[65.2987, 152.5993]	3	1	0	0
3 n=4, k=4, $\alpha=0.05$	[36.2766, 98.1243]	4	1	0	0

### Conclusion

Ott's (1967) ANOM tables are designed for normal distributions, the number of homogenous means for each data set was 3, 2, 2, respectively, and those not homogeneous are 2, 1 and 2, respectively. When the ANOM tables of the proposed model (LLD) are used for the same data sets, the number of homogenous means are 5, 3 and 4, respectively, and they do not exhibit deviation from homogeneity for values of  $b = 2$ ,  $b = 3, b = 4$  and  $b = 5$ . Use of the normal model resulted in homogeneity for some means and deviation for other means, thus indicating possible rejection of those means. The rejection decision is valid if a normal distribution is a good fit for the data. However, by comparison, results show that the LLD is a better model than the normal. Results are supported by the Q-Q plot correlation coefficient for each data set with the normal and with the LLD. It is therefore assumed that more error is likely to be associated with the decision process when data are from a normal distribution, thus, making all the means homogenous using LLD (see Table 3.11) is recommended over using the normal-ANOM procedure.

### References

- Bakir, S. T. (1994). Means using the ranks for randomized complete block designs. *Communications in Statistics-Simulation and Computation*, 23, 547-568.
- Bernard, A. J., & Wludyka, P. S. (2001). Robust I-sample analysis of means type randomization tests for variance. *Journal of Statistical Computation and Simulation*, 69, 57-88.
- Farnum, N. R. (2004). Analysis of means table using mathematical processors. *Quality Engineering*, 16, 399-405.
- Guirguis, G. H., & Tobias, R. D. (2004). On the computation of the distribution for the analysis of means. *Communication in Statistics-Simulation and Computation*, 16, 861-887.
- Montgomery, D. C. (2000). *Design and Analysis of Experiments*, 5<sup>th</sup> Ed. New York, NY: John Wiley and Sons Inc.
- Nelson, P. R., & Dudewicz, E. J. (2002). Exact analysis of means with unequal variances. *Technometrics*, 44, 152-160.
- Ott, E. R. (1967). Analysis of means: A graphical procedure. *Industrial Quality Control*, 24, 101-109.
- Ramig, P. F. (1983). Applications of analysis of means. *Journal of Quality Technology*, 15, 19-25.
- Rao, C. V. (2005). Analysis of means: A review. *Journal of Quality Technology*, 37, 308-315.
- Rao, C. V., & Prankumar, M. (2002). ANOM-type graphical methods for testing the equality of several correlation coefficients. *Gujarat Statistical Review*, 29, 47-56.
- Shewart, W. A. (1986). *Statistical method from the view point of quality control*, 14<sup>th</sup> Ed. New York, NY: Mineola, Dover Publications.
- Srinivasa Rao, B., & Kantam, R. R. L. (2012). Extreme value charts and analysis of means based on half logistic distribution. *International Journal of Quality, Reliability and Management*, 29(5), 501-511.
- Wadsworth, H. M., Stephens, K. S., & Godfrey, A. F. (2001). *Modern methods of quality control and improvement*. New York, NY: John Wiley and Sons, Inc.
- Wludyka, P. S., Nelson, P. R., & Silva, P. R. (2001). Power curves for analysis of means for variances. *Journal of Quality Technology*, 3, 60-75.