

Generalized Confidence Interval for the Largest Normal Mean

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Abstract. In this article we consider inferences on the largest mean of $k(\geq 2)$ normal populations with unequal unknown variances. The coverage probability and expected length of the two confidence intervals for the largest mean are compared using Monte Carlo simulation. We found that the coverage probability of the generalized confidence interval and optimal confidence interval are close to nominal level but the expected length of generalized confidence interval is smaller than the optimal confidence interval. Finally an example is provided.

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1. Introduction

Let $\pi_1, \pi_2, \dots, \pi_k$ be $k(\geq 2)$ normal populations with means μ_i and variances σ_i^2 , and n_i observations taken from each population, $i = 1, 2, \dots, k$.

The ordered values of μ_i 's, denoted by $\mu_{[1]} \leq \mu_{[2]} \leq \dots \leq \mu_{[k]}$, provide some important and useful information about the populations. For

example, when the variances are equal, the ordering of μ_i 's is equivalent to ordering of the populations. In reliability and lifetesting, ranked parameters are applied to study a system.

The problem of ranking of means of several independent normal populations, specially the largest and smallest mean, has been studied by many authors; Dudewicz (1972) showed that the largest sample mean, as a natural estimator of the largest mean, is biased. However, Chen (1976) noted that the largest sample mean is both strongly consistent and asymptotically unbiased. For the largest mean, when the common population variance is known, Saxena and Tong (1969) and Dudewicz (1972) analyzed confidence intervals that were not optimal, but Dudewicz and Tong (1971) proposed an optimal confidence interval, and then Tong (1973) provided percentage points for this kind of interval. When the variances are equal but unknown, Saxena (1976) gave a large sample approximation, Chen and Chen (1999) obtained a nearly optimal, and Chen and Chen (2004) found an optimal confidence interval. When the variances are unknown and possibly unequal, Chen and Dudewicz (1976), and Chen (1977a,b) proposed a class of intersecting intervals, and then Chen and Wen (2006) found an optimal confidence interval.

For the ranked mean, $\mu_{[i]}$ there are two articles, by Dudewicz (1972), who obtained a class of two-sided intervals when variances are known,

and by Alam and Saxena (1974), who gave a confidence interval such that the distribution of each population is stochastically increasing in mean.

In this paper, we first review two methods that are applicable for interval estimation ranked means, $\mu_{[i]}$ in Section 2. First method is based on the concepts of generalized confidence interval and generalized p-value, and is given by Chang and Huang (2000). The concepts of generalized p-value and generalized confidence interval that provide exact tests and exact confidence intervals, proposed by Tsui and Weerahandi (1989) and Weerahandi (1993). Second method is optimal confidence interval that is given by Chen and Wen (2006). In Section 3, by a numerical example, we compare these methods with each other. Simulation studies are presented in Section 4, to compare the coverage probabilities and the expected lengths of these methods.

2. Inferences for the Mean of the Best Population

Suppose $X_{i1}, X_{i2}, \dots, X_{in_i}$, $i = 1, \dots, k$ are k random samples with size n_i from normal populations with means μ_i and unequal variances σ_i^2 . The largest normal mean, $\mu^* = \mu_{[k]} = \max(\mu_1, \dots, \mu_k)$ provides some important and useful information about populations. The population that has the largest mean, μ^* is known as the best population.

The problem of interest is to find confidence interval for the largest mean. In this Section we first study the confidence interval of μ^* based on a generalized pivotal variable that is given by Chang and Huang (2000). Then briefly we review the method of Chen and Chen (2006) that is an optimal confidence interval for the largest mean.

2.1 Generalized Confidence Interval for the Largest Mean

Let, for the i th population, $\bar{X}_i = \sum_{j=1}^{n_i} X_{ij}/n_i$ and $S_i^2 = \sum_{j=1}^{n_i} (X_{ij} - \bar{X}_i)^2 / (n_i - 1)$ be the sample mean and sample variance, respectively, and \bar{x}_i and s_i^2 denote the observed values of \bar{X}_i and S_i^2 , $i = 1, \dots, k$.

The generalized pivotal variable for μ_i is given by

$$T_i = \bar{x}_i - \frac{t_{n_i-1} s_i}{\sqrt{n_i}}, \quad i = 1, \dots, k, \quad (1)$$

where $t_{n_i-1} = \sqrt{n_i} \frac{\bar{X}_i - \mu_i}{S_i}$ follows a student's t-distribution with $n_i - 1$ degrees of freedom.

The generalized pivotal variable for the largest mean is the largest generalized pivotal variables, T_i 's of μ_i 's, i.e.

$$T^* = \max(T_1, T_2, \dots, T_k). \quad (2)$$

The exact $(1 - \alpha)$ confidence interval is given by

$$[T^*(\alpha/2), T^*(1 - \alpha/2)], \quad (3)$$

where $T^*(\gamma)$ stands for the γ th quantile of T^* .

2.2 Optimal Confidence Interval

Chen and Wen (2006) proposed a two-stage procedure for obtaining an optimal confidence interval for the largest or smallest mean of k independent normal population, where the population variances are unknown and possibly unequal. This procedure requires additional samples and it may not be practicable in real problems. Therefore they employed a one-stage sampling procedure that is proposed by Chen and Lam (1989) for interval estimation. This one-stage sampling procedure for interval estimation is stated as follows:

Let X_{ij} ($j = 1, \dots, n_i$) be an independent random sample from normal population π_i , ($i = 1, \dots, k$) with unknown and unequal variance σ_i^2 , $i = 1, \dots, k$.

Choose n'_i ($n'_i \geq 3$) observations, $X'_{i1}, \dots, X'_{in'_i}$ from each random samples, X_{i1}, \dots, X_{in_i} .

Employ the first n_0 observations, ($2 \leq n_0 < n'_i$) and calculate the sample mean and sample variance of these observations, respectively by

$$\tilde{X}_i = \frac{1}{n_0} \sum_{j=1}^{n_0} X'_{ij} \quad , \quad \tilde{S}_i^2 = \frac{1}{n_0 - 1} \sum_{j=1}^{n_0} (X'_{ij} - \tilde{X}_i)^2.$$

It is noted that Chen and Lam (1989) suggested taking n_0 to be $\min \{n'_i - 1, i = 1, \dots, k\}$.

Choose the weights for the i th sample observations to be

$$U_i = \frac{1}{n'_i} + \frac{1}{n'_i} \sqrt{\frac{n'_i - n_0}{n_0} \left(\frac{n'_i h^*}{\tilde{S}_i^2} - 1 \right)}, \quad V_i = \frac{1}{n'_i} - \frac{1}{n'_i} \sqrt{\frac{n_0}{n'_i - n_0} \left(\frac{n'_i h^*}{\tilde{S}_i^2} - 1 \right)},$$

where h^* is the maximum of $\left\{ \frac{\tilde{S}_1^2}{n_1}, \dots, \frac{\tilde{S}_k^2}{n_k} \right\}$.

Let final weighted sample mean using all observations be defined by

$$\tilde{Y}_i = U_i \sum_{j=1}^{n_0} X'_{ij} + V_i \sum_{j=n_0+1}^{n'_i} X'_{ij}.$$

and \tilde{Y}^* represents the largest one of $\tilde{Y}_1, \dots, \tilde{Y}_k$. Then the $100\%(1 - \alpha)$ confidence interval for largest mean, μ^* is

$$(\tilde{Y}^* - d_1 \sqrt{h^*}, \tilde{Y}^* + d_2 \sqrt{h^*}), \quad (4)$$

where d_1 and d_2 are left entry and right entry percentage points, respectively, that are given by Chen and Wen (2006).

3. Numerical Example

The data, taken from Bishop and Dudewicz (1978), is an experiment for studying the bacterial killing. The experiment involved testing four types of solvents for their effects on the ability of a fungicide methyl-2-benzimidazole-carbamate to destroy the fungus *Penicillium expansum*. The fungicide was diluted in exactly the same manner in the four different types of solvents and sprayed on the fungus, and the percentage of fungus destroyed was measured. Let μ_i denote the mean percentage

of fungus destroyed by solvent i . Our interest is in finding an interval estimation for the largest mean percentage, $\mu^* = \max(\mu_1, \mu_2, \mu_3, \mu_4)$.

The summary statistics of all data for different solvents and other needed values for one-stage sampling procedure by $n_0 = 15$, are given in Table 1 (For details see Chen and Wen (2006)).

The 95% proposed generalized confidence interval in [3] and one-stage confidence interval in [4] for the largest mean, μ^* are (97.038, 97.953) and (96.290, 97.868), respectively.

Table 1: Summary statistics of bacterial killing ability example

Statistics	n_i	\bar{x}_i	s_i^2	n'_i	\tilde{x}_i	\tilde{s}_i^2
Solvent 1	19	97.18	2.091	18	96.8420	2.1099
Solvent 2	28	95.41	4.112	25	94.6860	3.1708
Solvent 3	52	95.42	4.764	49	94.3833	5.8842
Solvent 4	16	97.37	0.753	16	97.3327	0.7797

4. Simulation Study

For comparing the coverage probabilities and expected length of the confidence intervals for the largest mean, simulations were studied for the case of $k = 4$ only. The observations are generated with size n_i from four normal populations with means μ_i and variances σ_i^2 , $i = 1, 2, 3, 4$. Also we considered that $\mu^* = \mu_1 = 10$, that is, the largest mean is 10 and the first population is the best population.

The generalized confidence interval (GC) in (3) that is considered by Chang and Huang (2000), compared by the optimal confidence interval (OC) of Chen and Chen (2006) in (4). The results are given in Tables 2 and 3. We found that the coverage probabilities of generalized confidence interval and optimal interval are close to significant level and in some cases the coverage probabilities of optimal interval is greater than significant level. Also the expected length of generalized confidence interval is smaller than expected length optimal confidence interval.

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Table 2: Simulation of %95 coverage probabilities with unequal variances and $\mu^* = \mu_1 = 10$

μ_2, μ_3, μ_4	1, 1, 1		1, 2, 5		2, 5, 7	
	<i>OC</i>	<i>GC</i>	<i>OC</i>	<i>GC</i>	<i>OC</i>	<i>GC</i>
$\sigma^2 = (1, 1, 2, 2)$						
$n = (10, 10, 10, 10)$	0.951	0.958	0.947	0.951	0.954	0.954
$n = (10, 10, 20, 20)$	0.944	0.944	0.956	0.947	0.953	0.939
$n = (20, 10, 10, 30)$	0.937	0.969	0.959	0.948	0.956	0.944
$n = (10, 30, 30, 30)$	0.961	0.943	0.942	0.951	0.947	0.955
$n = (30, 30, 30, 30)$	0.951	0.946	0.950	0.958	0.956	0.965
$\sigma^2 = (4, 1, 2, 3)$						
$n = (10, 10, 10, 10)$	0.937	0.957	0.944	0.953	0.944	0.951
$n = (10, 10, 20, 20)$	0.941	0.943	0.947	0.959	0.955	0.942
$n = (20, 10, 10, 30)$	0.947	0.954	0.955	0.950	0.959	0.950
$n = (10, 30, 30, 30)$	0.956	0.938	0.954	0.954	0.952	0.946
$n = (30, 30, 30, 30)$	0.944	0.941	0.941	0.961	0.955	0.947
$\sigma^2 = (1, 10, 2, 1)$						
$n = (10, 10, 10, 10)$	0.947	0.942	0.961	0.965	0.961	0.953
$n = (10, 10, 20, 20)$	0.941	0.954	0.956	0.949	0.960	0.950
$n = (20, 10, 10, 30)$	0.951	0.942	0.942	0.952	0.944	0.947
$n = (10, 30, 30, 30)$	0.950	0.959	0.941	0.947	0.942	0.945
$n = (30, 30, 30, 30)$	0.946	0.939	0.948	0.958	0.962	0.949
$\sigma^2 = (5, 2, 1, 10)$						
$n = (10, 10, 10, 10)$	0.948	0.938	0.963	0.949	0.957	0.962
$n = (10, 10, 20, 20)$	0.947	0.945	0.942	0.949	0.960	0.957
$n = (20, 10, 10, 30)$	0.953	0.946	0.954	0.954	0.955	0.954
$n = (10, 30, 30, 30)$	0.956	0.945	0.943	0.953	0.951	0.941
$n = (30, 30, 30, 30)$	0.945	0.948	0.951	0.955	0.962	0.956

Table 3: Simulation of expected length of %95 confidence interval with unequal variances and $\mu^* = \mu_1 = 10$

μ_2, μ_3, μ_4	1, 1, 1		1, 2, 5		2, 5, 7	
	<i>OC</i>	<i>GC</i>	<i>OC</i>	<i>GC</i>	<i>OC</i>	<i>GC</i>
$\sigma^2 = (1, 1, 2, 2)$						
$n = (10, 10, 10, 10)$	5.202	1.387	5.162	1.401	5.317	1.401
$n = (10, 10, 20, 20)$	4.210	1.373	4.147	1.372	4.129	1.393
$n = (20, 10, 10, 30)$	4.462	0.923	4.304	0.927	3.716	0.927
$n = (10, 30, 30, 30)$	3.441	1.397	3.431	1.395	3.324	1.391
$n = (30, 30, 30, 30)$	2.902	0.738	2.878	0.744	2.781	0.743
$\sigma^2 = (4, 1, 2, 3)$						
$n = (10, 10, 10, 10)$	6.733	2.781	6.709	2.787	6.840	2.770
$n = (10, 10, 20, 20)$	5.839	2.805	5.934	2.825	5.929	2.773
$n = (20, 10, 10, 30)$	5.324	1.854	5.903	1.842	5.298	1.854
$n = (10, 30, 30, 30)$	5.158	2.767	5.220	2.804	5.152	2.804
$n = (30, 30, 30, 30)$	3.741	1.782	3.771	1.484	3.767	1.482
$\sigma^2 = (1, 10, 2, 1)$						
$n = (10, 10, 10, 10)$	8.936	1.386	8.903	1.397	8.445	1.391
$n = (10, 10, 20, 20)$	8.469	1.384	8.081	1.381	8.026	1.389
$n = (20, 10, 10, 30)$	8.142	0.915	8.151	0.916	8.163	0.931
$n = (10, 30, 30, 30)$	5.606	1.392	5.430	1.404	5.558	1.386
$n = (30, 30, 30, 30)$	4.788	0.739	5.004	0.741	4.908	0.743
$\sigma^2 = (5, 2, 1, 10)$						
$n = (10, 10, 10, 10)$	10.97	4.354	11.07	4.318	10.83	4.138
$n = (10, 10, 20, 20)$	9.477	4.429	9.305	4.416	9.266	4.174
$n = (20, 10, 10, 30)$	8.035	2.938	7.901	2.955	7.764	2.861
$n = (10, 30, 30, 30)$	8.225	4.449	8.530	4.476	8.389	4.147
$n = (30, 30, 30, 30)$	5.967	2.346	6.229	2.360	5.921	2.329

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