

8(1): 1-8, 2020; Article no.AJPAS.58491 ISSN: 2582-0230

Optimal Prediction Variance Capabilities of Inscribed Central Composite Designs

Julius C. Nwanya^{1*} and Kelechukwu C. N. Dozie¹

¹Department of Statistics, Imo State University, Owerri, Nigeria.

Authors' contributions

This work was carried out in collaboration between both authors. Author JCN designed the study, performed the statistical analysis, wrote the protocol and wrote the first draft of the manuscript. Author KCND managed the analyses and literature searches of the study. Both authors read and approved the final manuscript.

Article Information

DOI: 10.9734/AJPAS/2020/v8i130194 <u>Editor(s):</u> (1) Dr. Manuel Alberto M. Ferreira, Lisbon University, Portugal. (1) Gladys Gakenia Njoroge, Chuka University, Kenya. (2) Manpreet Singh Bhatti, Guru Nanak Dev University, India. (3) Krishnadhan Sarkar, Raniganj Girls' College, India. Complete Peer review History: <u>http://www.sdiarticle4.com/review-history/58491</u>

Original Research Article

Received: 28 April 2020 Accepted: 03 July 2020 Published: 14 July 2020

Abstract

This study looks at the effects of replication on prediction variance performances of inscribe central composite design especially those without replication on the factorial and axial portion (ICCD1), inscribe central composite design with replicated axial portion (ICCD2) and inscribe central composite design whose factorial portion is replicated (ICCD3). The G-optimal, I-optimal and FDS plots were used to examine these designs. Inscribe central composite design without replicated factorial and axial portion (ICCD1) has a better maximum scaled prediction variance (SPV) at factors k = 2 to 4 while inscribe central composite design with replicated factorial portion (ICCD3) has a better maximum and average SPV at 5 and 6 factor levels. The fraction of design space (FDS) plots show that the inscribe central composite design is superior to ICCD3 and inscribe central composite design with replicated factorial portion (ICCD2) from 0.0 to 0.5 of the design space while inscribe central composite design with replicated factorial portion (ICCD2) and ICCD3 is superior to ICCD1 and ICCD2 from 0.6 to 1.0 of the design space for factors k = 2 to 4.

Keywords: Inscribe central composite design; scaled prediction variance; fraction of design space.

^{*}Corresponding author: E-mail: nwjulius@yahoo.com;

1 Introduction

The inscribed central composite design (ICCD) is a design that enables one to study the full ranges of the experiment variables while excluding non allowable operating conditions at one or more of the extremes of the design region. It is used when the region of operability coincides with the region of interest. The factorial points of ICCD according to [1] are brought into the interior of the design space and set at a distance from the centre point that preserves the proportional distance of the factorial points to the axial points. The prediction variance stability and capability of CCD performance when replication is at the axial and factorial portion has been of interest to many scholars of second order design. Replication being one of the basic principles of design of experiment help researchers to provide increased precision and to obtain an estimate of the experimental error. Replication has been variously applied by many researchers in the study of central composite design. In [2], the author(s) considered the replication of the factorial and axial portion of rotatable central composite and orthogonal central composite designs for factors $k = 2, 3, \dots, 8$. His results show that replicating the axial portion of both designs have a better potential for improved precision of prediction than replicating the factorial portion. Some second order design prediction capability was studied by [3]. They used fraction of design space (FDS) plots. Their plots revealed that the CCD with three center points was superior to other designs. The prediction variance properties by [4] assessed some second-order designs for cuboidal regions using the G- and I-optimality criteria. They recommended the G-efficient design over any other standard design presented in their work and recommend Face centered cube (FCC) and Hoke designs for small number of design variables. In [5] in the spherical region, compared prediction variance performance of central composite design, minimum resolution v design and small composite design. Their conclusion is that central composite design has a better prediction variance over minimum resolution v designs and small composite design. The partially replicated factorial and axial portions of the rotatable central composite and orthogonal central composite designs was evaluated by [6] using the Doptimality criterion. In their conclusion, they pointed out that replicating the factorial portion enhances the D-optimality performance of both central composite designs better than the replicating the axial portion. The fraction of design space plots and variance dispersion graphs were used by [7], to study the prediction capabilities of partially replicated orthogonal central composite design in a spherical region. They pointed out that replicating the axial portions of the orthogonal central composite design by far reduces the prediction variance, thus improving the G-efficiency in the spherical region. Also in [8], the author(s) considered variations of partially replicated central composite designs in the hypercube using the G- and Ioptimality criteria as well as the fraction of design space plots. They concluded that the replicated-star options always give smaller G- and I-optimal values than the corresponding replicated-cube options with equal number of replications. In [9] the author(s) used quantile plots to compare the prediction variance of partially replicated face center cube and rotatable central composite design. They concluded that the performance of the factorial with replicated axial points in both design is preferred to the replicated factorial portion with one axial portion. There are situation where the replicated factorial portion with one axial portion performs better than non-replicated factorial portion with replicated axial portion but as one moves towards the design perimeter, the later performs better. In [10], the author(s) applied inscribe central composite design for modeling and optimization of marble quality. Belt speed, rotational speed and pressure of the polishing head were used as factors with two responses. The used analysis of variance (ANOVA) to assess the fit of the models generated with the experimental data. Their study showed that the inscribed central composite design could efficiently be applied for the modelling of polishing machine for surface quality of marble strips. In [11], the author(s) studied prediction variance of a central composite design with missing observation, they derived an expression for the inflation amount of the prediction variance. This derived expression showed that rotatable central composite design inflation amount of the prediction variance depends only on the Euclidean norms and the inner product of the two vectors of factor values at which the observation is missing and the response is predicted. In [12,13,14,15], the author(s) have highlighted the benefits of scaled prediction variance in model assessment of second order designs. The Gand I-optimality criteria are prediction variance oriented which are single number criteria. Single number criteria, may not reflect the prediction variance characteristics of a design. In [3], fraction of design space plot is an alternative to single-number criteria. The FDS plot shows the characteristics of the prediction

variance in the entire design space and the characteristics of scaled prediction variance (SPV) in a multidimensional region of a single two-dimensional graph with a single curve.

In this work, the G- and I-optimality criteria as well as the fraction of design space plots were used to compare partially replicated variation of the inscribe central composite design which are the inscribe central composite design with replicated factorial portion plus one axial portion (ICCD3), the inscribe central composite design with replicated axial portion (ICCD2) and the inscribe central composite without replicated factorial and axial portion (ICCD1).

2 Model Development

	x_0	x_1	<i>x</i> ₂	x_1^2	x_{2}^{2}	$x_1 x_2$	
	1	$\frac{-1}{\alpha}$	$\frac{-1}{\alpha}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	
X =	1	$\frac{-1}{\alpha}$	$\frac{1}{\alpha}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	$\frac{-1}{\alpha^2}$	$n_c f$
	1	$\frac{1}{\alpha}$	$\frac{-1}{\alpha}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	$\frac{-1}{\alpha^2}$	
	1	$\frac{1}{\alpha}$	$\frac{1}{\alpha}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	$\frac{1}{\alpha^2}$	
	1	-1	0	1	0	0	
	1	1	0	1	0	0	21
	1	0	-1	0	1	0	2kr
	1	0	1	0	1	0	
	1	0	0	0	0	0	n_0
	1	0	0	0	0	0	

Table 1. Design matrix of inscribe central composite design for K = 2

The relationship between the response variable y and the design variables x_1, x_2, \dots, x_k , in an *N*-run experiment is fitted into a second-order model as

$$y = \beta_0 + \sum_{i=1}^k \beta_i x_i + \sum_{i=1}^k \beta_{ii} x_{ii}^2 + \sum_{j=i+1}^k \sum_{i=1}^{k-1} \beta_{ij} x_i x_j + \varepsilon_{ij}$$
(1)

Which in a matrix form is given as

$$Y = X\beta + \varepsilon \tag{2}$$

Where Y is an $N \ge 1$ vector of responses, X is an $N \ge p$ expanded design matrix. β is a vector of unknown coefficients to be estimated by least square methods from experimental data and ε is the random error term which is distributed as NID $(0, \sigma^2)$. The rows of X represent the experimental run, and the column corresponds to the experimental settings of k design variables. In the expanded design matrix X, the number of columns p, is the number of parameters in the model given as

$$p = \frac{(k+1)(k+2)}{2}$$
(3)

However at a point x in the design space, the prediction variance is given as

$$\operatorname{var}[\hat{y}(x)] = \sigma^2 f^T(x) (X^T X)^{-1} f(x)$$
(4)

where $(X^T X)^{-1}$ is the variance-covariance matrix of the design matrix X, f(x) is the vector of coordinates of point in the region of interest expanded to model form. That is, $f^T(x) = [1, x_1, \dots, x_k, x_1^2, \dots, x_k^2, x_1 x_2, \dots, x_{k-1} x_k]$ of Equation (4) is scaled by multiplying by N, the total number of runs, and dividing by σ^2 , the process variance. The resulting expression

$$\frac{N\operatorname{var}[\hat{y}(x)]}{\sigma^{2}} = Nf^{T}(x)(X^{T}X)^{-1}f(x)$$
(5)

is the scaled prediction variance (SPV). This SPV gives a good measure of the precision of the estimated response at any point in the design space.

The G-optimality criterion minimizes the maximum scaled prediction variance (SPV). That is

$$\min\{\max N \operatorname{var} \hat{y}(x)\} = \min\{N \max f^{T}(x) (X^{T} X)^{-1} f(x)\}$$
(6)

The I-optimality criterion minimizes the average SPV. That is

$$\min\{averageN \operatorname{var} \hat{y}(x)\} = \frac{1}{A} \int_{R} \operatorname{var}[y(x)] dx$$
(7)

where A is the volume of the design space R.

3 Partial Replication of the Design

Let the factorial portion $f = 2^k$ replicated n_c times, the axial portion 2k replicated r times and the center point replicated n_0 times, then the inscribed central composite design use a total of $N = 2^k n_c + 2kr + n_0$ number of observations or runs for model parameter estimation.

In this study, three versions of the inscribed central composite design (ICCD) are generated by replicating the factorial and axial portions. The first version is where the factorial and axial portion are not replicated. It is denoted as ICCD1. The second version where factorial portion is not replicated and the axial portion is replicated twice and it is denoted as ICCD2. Finally, the third version is where the factorial portion is replicated twice and the axial portion not replicated. It is denoted is as ICCD3. The number of center points used in all the versions of the ICCD in this study is three.

4 Comparison of the Design

The results in Table 2 show that the inscribed central composite design without replicated factorial and axial portion (ICCD1) has a better maximum scaled prediction variance (SPV) at factors k = 2 to 4. Also the inscribe central composite design with replicated factorial portion (ICCD3) has a better maximum SPV at 5 and 6 factor levels when half replicate of the factorial portion were used while ICCD2 has a better maximum SPV when full replicate of the factorial were used.





Fig. 4. FDS Plot of designs for k = 5



Fig. 5. FDS Plot of designs for k = 6

The average SPV of ICCD1, ICCD2 and ICCD3 are relatively the same at factors k = 2, 3 and 4. The ICCD3 has a better average SPV at factors k = 5 and 6 when half replicate of the factorial portion were used while at full replicate, all the versions of the ICCD1 are relatively the same.

Design	N	Max SPV	Ave. SPV	k
ICCD1	11	6.875	5.995	
ICCD 2	15	8.925	6.000	2
ICCD 3	15	8.925	6.000	
ICCD1	17	11.39	9.996	
ICCD 2	23	14.90	10.005	3
ICCD 3	25	14.45	10.000	
ICCD1	27	15.741	15.066	
ICCD 2	35	19.88	15.015	4
ICCD 3	43	24.381	15.007	
ICCD1	45	25.290	21.015	
ICCD 2	55	24.200	21.010	5 (full)
ICCD 3	77	43.274	21.021	
ICCD1	29	25.520	16.037	
ICCD2	39	33.579	20.202	5 (half replicate)
ICCD 3	45	20.295	14.040	
ICCD1	79	43.687	27.966	
ICCD 2	91	29.757	28.028	6 (full)
ICCD 3	143	79.508	28.028	
ICCD1	47	29.657	20.21	
ICCD 2	59	36.58	23.836	6 (half replicate)
ICCD3	79	25.438	18.486	

	Table 2.	Summary	statistics	for	versions	of	IC	CD
--	----------	---------	------------	-----	----------	----	----	----

The fraction of design space (FDS) plots (Figs. 1 to 3) show that the inscribe central composite design (ICCD1) is superior to ICCD3 and ICCD2 from 0.0 to 0.5 of the design space while inscribe central composite design with replicated factorial portion (ICCD3) is superior to ICCD1 and ICCD2 from 0.6 to 1.0 of the design space for factors k = 2 to 4. The ICCD2 is superior to ICCD1 and ICCD3 at factor k = 6 (See Fig. 5). The result of the FDS plot in this study is in line with [3] in the literature.

5 Conclusion

This work has shown how replication of any part of the inscribed central composite design affect prediction variance performance. Inscribed central composite design without replicated factorial and axial portion (ICCD1) has a better maximum scaled prediction variance (SPV) at low factor levels (k = 2 - 4), while inscribe central composite design with replicated factorial portion (ICCD3) has better maximum and average SPV at 5 and 6 factor levels. Also the fraction of design space (FDS) plots show that the inscribe central composite design without replicated factorial and axial portion (ICCD1) is superior to ICCD 3 and ICCD 2 from 0.0 to 0.5 of the design space while inscribe central composite design with replicated factorial portion (ICCD3) is superior to ICCD 1 and ICCD2 from 0.6 to 1.0 of the design space for factors k = 2 to 4.

Competing Interests

Authors have declared that no competing interests exist.

References

- Verseput R. Selecting the right central composite design for response surface methodology applications; 2001. Available:www.qualitydigest.com/
- [2] Dykstra Jr. O. Partial duplication of response surface designs. Technometrics. 1960;2(2):185–195.
- [3] Zahran A, Anderson-Cook CN. Fraction of design space to assess the prediction capability of response surface designs. Journal of Quality Technology. 2003;35:377-386.
- [4] Park Y, Richardson DE, Montgomery DC, Ozol-Godfrey A, Borrer CM, Anderson-Cook CM. Prediction variance properties of second-order designs for cuboidal region. Journal of Quality Technology. 2005;37(4):253–266.
- [5] Chigbu PE, Ukaegbu EC, Nwanya JC. On comparing the prediction variances of some central composite designs in spherical region: A review. Statistica. Anno, LXIX. 2009;4:285-298.
- [6] Chigbu PE, Ohaegbulem UO. On the preference of replicating factorial runs to axial runs in restricted second-order designs. Journal of Applied Sciences. 2011;11(22):3732–3737.
- [7] Ukaegbu EC, Chigbu PE. Graphical evaluation of the prediction capabilities of partially replicated orthogonal central composite design. Quality and Reliability Engineering International. 2014;31:701– 717.
- [8] Ukaegbu EC, Chigbu PE. Comparison of the prediction capabilities of partially replicated central composite designs in Cuboidal Region. Communications in Statistics-Theory and Methods. 2015;44:406-427. Available:http://dx.doi.org/10.1080/03610926.2012.745561

- [9] Umelo-Ibemere NC, Amuji HO. Quantile plots of the prediction variance for partially replicated central composite design. International Journal of Statistics and Probability. 2015;4(2):55–60.
- [10] Sümeyra Cevheroglu Çira, Ahmet Dag, Askeri Karakus. Application of response surface methodology and central composite inscribed design for modeling and optimization of marble surface quality. Advances in Materials Science and Engineering; 2016. Available:https://doi.org/10.1155/2016/2349476
- Kohei Fujiwara, Shun Matsuura. Prediction variance of a central composite design with missing observation; 2019.
 Available:https://doi.org/10.1080/03610926.2019.1625925
- [12] Giovannitti-Jensen A, Myers RH. Graphical assessment of the prediction capability of response surface designs. Technometrics. 1989;31(2):159-171.
- Borkowski JJ. Spherical prediction variance properties of central composite and Box-Behnken designs. Technometrics. 1995;37:399-410.
- [14] Montgomery DC. Design and analysis of experiments, 6th Edition. John Wiley and Sons, Inc. N.J.; 2005.
- [15] Anderson-Cook CM, Borror CM, Montgomery DC. Response surface design evaluation and comparison. Journal of Statistical Planning and Inference. 2009;139:629-641.

© 2020 Nwanya and Dozie; This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Peer-review history:

The peer review history for this paper can be accessed here (Please copy paste the total link in your browser address bar) http://www.sdiarticle4.com/review-history/58491