



Roberval Monteiro Bezerra de Lima Teresa Paula Costa Azinheira Oliveira Editores Técnicos







Empresa Brasileira de Pesquisa Agropecuária Embrapa Amazônia Ocidental Ministério da Agricultura, Pecuária e Abastecimento

Conferência Internacional da Amazônia em Estatística Experimental de Risco



Roberval Monteiro Bezerra de Lima Teresa Paula Costa Azinheira Oliveira Editores Técnicos

EmbrapaBrasília, DF
2015



Experimental Design and Data Analysis: Methods for the Response Optimization and Applications Using R

¹Teresa A. Oliveira; ²Conceição Leal; ³Roberval Lima ¹Universidade Aberta, DCeT and CEAUL, Portugal, teresa.oliveira@uab.pt; ²Universidade Aberta, MEMeC, Portugal, conceicao.leal2010@gmail.com; ³Embrapa Amazônia Ocidental, Brasil, roberval.lima@cpaa.embrapa.br

Abstract

Experimental Design is used by an increasing number of researchers to explore relationships between factors of interest. The factor levels are changed systematically according to the design, in order to evaluate their effects on the process. The aim is to reach a maximum amount of information with minimum cost in a minimal amount of time. This work presents a general review on Experimental Design techniques and a discussion of methods for the response optimization, followed by some applications to real data.

In Response Surface Methodology (RSM), a significant number of studies relates to a single response system or to responses analyzed one-by-one. However, for most of applications it is necessary and useful considering all responses simultaneously, since it is common to find optimization problems with several conflicting responses. The methodologies to multiple responses problems optimization, associated with the RSM, boil down to an informal overlap of contour maps of different responses to select the optimal solution. In most recent research, associated with the increasing computer capabilities, mathematical programming techniques were developed, including heuristic search methods, which allow locating the combinations of variables that lead to the best compromise between the different specifications of multivariate responses. The desirability function is a useful approach to simultaneous optimization of multiple responses.



Considering the experimental designs developed by Lima (2010) in forestry production and by Mondim (2014) in ceramic industry, we will use R to illustrate the role of RSM and of the methodology of genetic algorithms on the optimization of a problem. RSM will be used to optimizing one-response-at-a-time. The desirability function and the genetic algorithm will be used to optimizing one-response-at-a-time and all response simultaneously. Results will be compared and discussed.

Keywords: experimental design, response surface methodology, genetic algorithms, desirability, multiple response.

References

ALVAREZ, M. J.; ILZARBE, L.; VILES, E.; TANCO, M. The use of genetic algorithms in response surface methodology. **Quality Technology and Quantitative Management**, Taiwan, v. 6, n. 3, p. 295-307, 2009.

BILES, W. A response surface method for experimental optimization of multi response processes. **Industrial and Engineering Chemistry, Process Design and Deployment**, Washington, DC, v. 14, p. 152-158, 1975.

BOX, G. E. P.; DRAPER, N. R. Empirical model-building and response surfaces. New York: J. Wiley & Sons, 1987. 688 p. (Wiley Series in Probability and Mathematical Statistics, 157).

BOX, G. E.; DRAPER, N. R. Response surfaces, mixtures, and ridge analyses. New York: John Wiley & Sons, 2007. v. 649.

BOX, G. E.; WILSON, K. B. On the experimental attainment of optimum conditions. **Journal of the Royal Statistical Society**. **Series B: Methodological**, London, v. 13, n. 1, p. 1-45, 1951.



BUSACCA, P. G.; MARSEGUERRA, M.; ZIO, E. Multi-objective optimization by genetic algorithms: application to safety systems. **Reliability Engineering System Safety**, Barking, v. 72, n. 1, p. 59-74, 2001.

CARLYLE, W.; MONTGOMERY, D. C.; RUNGER, G. C. Optimization problems and methods in quality control and improvement. **Journal of Quality Technology**, Milwaukee, v. 32, n. 1, p. 1-17, jan. 2000.

COPELAND, K. A.; NELSON, P. R. Dual response optimization via direct function minimization. **Journal of Quality Technology**, Milwaukee, v. 28, n. 3, p. 331-336, jul. 1996.

DEL CASTILLO, E. Multiresponse process optimization via constrained confidence regions. **Journal of Quality Technology**, Milwaukee, v. 28, n. 1, jan. 1996.

DEL CASTILLO, E.; MONTGOMERY, D. A nonlinear programming solution to the dual response problem. **Journal of Quality Technology**, Milwaukee, v. 25, n. 3, jul. 1993.

DERRINGER, G.; SUICH, R. Simultaneous optimization of several response variables. **Journal of Quality Technology**, Milwaukee, v. 12, n. 4, p. 214-219, oct. 1980.

GOLDBERG, D. E. Genetic algorithm in search, optimization and machine learning. Reading, MA: Addison-Wesley, 1989. 432 p.

HOLLAND, J. **Adaptation in natural and artificial systems**. Ann Arbor, MI: The University of Michigan Press, 1975.

HOOKE, R.; JEEVES, T. A. Direct search solution of numerical and statistical problems. **Journal of the ACM**, New York, v. 8, n. 2, p. 212-229, 1961.



- KHURI, A. I.; CONLON, M. Simultaneous optimization of multiple responses represented by polynomial regression functions. **Technometrics**, Richmond, v. 23, n. 4, p. 363-375, 1981.
- KIM, K. J.; BYUN, J. H.; MIN, D.; JEONG, I. J. **Multiresponse surface optimization**: concept, methods, and future direction (Tutorial). Seoul: Korea Society for Quality Management, 2001.
- KIM, D.; RHEE, S.; PARK, H. Modelling and optimization of a GMA welding process by genetic algorithm and response surface methodology. **International Journal of Production Research**, London, v. 40, n. 7, p. 1699-1711, 2002.
- LASDON, L. S.; FOX, R. L.; RATNER, M. W. Nonlinear optimization using the generalized reduced gradient method. **RAIRO Operations Research Recherche Opérationnelle**, Paris, v. 8, n. 3, p. 73-103, 1974.
- LENTH, R. V. Response-surface methods in R, using rsm. **Journal of Statistical Software**, Los Angeles, v. 32, n. 7, p. 1-17, 2009.
- LIMA, R. M. B. de. **Técnicas de experimentação estatística**: planos factoriais e aplicações. 2010. 78 f. Dissertação (Mestrado em Estatística, Matemática e Computação) Universidade Aberta, Lisboa.
- MONDIM, A. F. da S. **Metodologias de superfície de resposta**: uma investigação no âmbito da indústria cerâmica e do vidro. 2014. 109 f. Dissertação (Mestrado em Estatística, Matemática e Computação) Universidade Aberta, Lisboa.
- MYERS, R. H.; CARTER, W. H. Response surface techniques for dual response systems. **Technometrics**, Richmond, v. 15, p. 301-317, 1973.



MYERS, R. H.; MONTGOMERY, D. C.; ANDERSON-COOK, C. M. **Response surface methodology**: process and product optimization using designed experiments. New York: John Wiley & Sons, 2009. v. 705.

NELDER, J. A.; MEAD, R. A simplex method for function minimization. **The Computer Journal**, Oxford, v. 7, n. 4, p. 308-313, 1965.

PASANDIDEH, S. H. R.; NIAKI, S. T. A. Multi-response simulation optimization using genetic algorithm within desirability function framework. **Applied Mathematics and Computation**, New York, v. 175, n. 1, p. 366-382, 2006.

PLANTE, R. D. Process capability: a criterion for optimizing multiple response product and process design. **IIE Transactions**, Philadelphia, v. 33, n. 6, p. 497-509, 2001.

R DEVELOPMENT CORE TEAM. R: a language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing, 2012.

SCRUCCA, L. GA: a package for genetic algorithms in R. **Journal of Statistical Software**, Los Angeles, v. 53, p. 1-37, 2012.

ZADBOOD, A.; NOGHONDARIAN, K. Considering preference parameters in multi response surface optimization approaches. **International Journal of Modeling and Optimization**, Jurong West, v. 1, n. 2, p. 158-162, jun. 2011.