

شبكة المعلومات الجامعية

بسم الله الرحمل الرحم





شبكة المعلومات الجامعية التوثيق الالكتروني والميكروفيلم



شبكة المعلومات الجامعية

جامعة عين شمس

التوثيق الالكتروني والميكروفيلم

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Institute of Statistical Studies and Research
Department of Operations Research

A Comparative Study for Approximation Techniques to Nonlinear Optimization Problems

By

Gamal El-Dean Abdel Hakim Mohamed El-Emam

A Thesis Submitted to the Institute of Statistical Studies & Research

Cairo University

In Partial Fulfillment of the Requirements for the Degree of

Master of Science

In

Operations Research

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ABSTRACT

Constrained nonlinear programming problems often arise in many engineering applications and much of today's engineering analysis consists of running complex computer codes. Despite a steady increase in computing power, the complexity of engineering analyses seems to advance at the same rate. The use of statistical techniques to build approximations of expensive computer analysis codes pervades much of today's engineering design. These statistical approximations, or meta-models, are used to replace the actual expensive computer analyses, facilitating multidisciplinary and concept exploration. In this thesis, we review some of these techniques, especially Response Surface Methodology (RSM) and Kriging. Both methods are applied to thirty widely used classes' of single objective optimization problems. We compare the results of both Response Surface Methodology and Kriging model with the Generally solutions of the original nonlinear optimization solutions. speaking, Kriging method is able to return almost the same solution as the original model optimum for the majority of problems. Response Surface Methodology comes next to Kriging is finding the optimum to the original As a remedy, a new approach is proposed to split the model into model. several smaller sub-models. The problem is transformed into a number of sub-problems with an impact on increased model CPU processing time. A set of test bed problems is verified using both approximation methods. When Splitting is employed prior to approximation, both techniques are able to capture system optima consistently. Monotonicity and continuity are still important and valid concerns in any approximation.