Improved sequential linear programming formulation for structural weight minimization

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Abstract

This paper presents an optimization algorithm for weight minimization of structures. The algorithm—denoted as LSTRLP (line search trust region linear programming)—combines sequential linear programming (SLP) and Trust region methods (TRM). LSTRLP solves a linearized sub-problem in each design cycle and accepts or rejects intermediate designs based on a line-search strategy which detects if the eventual improvement in cost is actually the largest possible. It is to be noticed that the present work is the closure to several studies carried out by the present authors in order to improve the overall efficiency and robustness of the sequential linear programming method.

The LSTRLP algorithm is implemented by an optimization code written in Fortran 90. The optimization code is tested in eight cases of weight minimization of bar truss and frame structures. The test cases include examples of large-scale and configuration optimization. The results obtained here are compared to those presented in literature. The optimizations are run also with sequential quadratic programming (SQP) routines implemented in commercial software. The results indicate that LSTRLP is competitive with recently published algorithms and commercial software. © 2004 Elsevier B.V. All rights reserved.

Keywords: Structural optimization; SLP; Trust region; Line-search

1. Introduction

Sequential linear programming (SLP) is very popular in practical engineering since the linear solvers utilized to solve the linearized sub-problems are easily available to designers. In optimum design of structures, SLP is more attractive than other optimization methods because it requires structural analysis only for computing gradients of the cost function and constraints. Trust region methods (TRM) are universally acknowledged robust and versatile optimization algorithms because of their excellent global
Nomenclature

\( N \) number of optimization variables
\( j \) counter of optimization variables
\( NC \) number of inequality constraint functions
\( m \) counter of inequality constraints
\( NC_{\text{act}} \) number of active constraint functions
\( m_a \) counter of active constraints
\( NCV \) number of violated constraint functions
\( v \) counter of violated constraints
\( X(\mathbf{x}_1, \mathbf{x}_2, \ldots, \mathbf{x}_N) \) design vector containing the optimization variables
\( W(X) \) cost function of the optimization problem
\( \mathbf{g}_m(X) \) nonlinear inequality constraint functions
\( x^l \) lower bound of the \( j \)th optimization variable
\( x^u \) upper bound of the \( j \)th optimization variable
\( i \) iteration counter of the current design cycle
\( X_i(\mathbf{x}_1, \mathbf{x}_2, \ldots, \mathbf{x}_N) \) design vector found in the \( i \)th optimization cycle by solving the current linearized sub-problem
\( X_{\text{OPT}}(x_{\text{opt},1}, x_{\text{opt},2}, \ldots, x_{\text{opt},N}) \) optimum design vector updated each time the current iteration results in design improvements
\( P_i^0 \) linearization point used in the \( i \)th optimization cycle: point of the design space about which the nonlinear problem is linearized
\( X_i^0(x_{0,1}^0, x_{0,2}^0, \ldots, x_{0,N}^0) \) design vector defining the linearization point \( P_i^0 \) used in the \( i \)th design cycle
\( \mathbf{e}_{\text{TR}}(x_1 - x_{0,1}^i, x_2 - x_{0,2}^i, \ldots, x_N - x_{0,N}^i) \) search direction vector originating from the linearization point \( P_i^0 \) of the current design cycle
\( X \) generic point of the design space: it is defined by the \( X = X_i^0 + \mathbf{e}_{\text{TR}} \) vector where the \( N \) design variable values are stored
\( \rho(X) \) trust region parameter computed at the \( X \) point of the design space
\( W_{\text{APP}}(X) \) approximate cost function computed at the \( X \) point of the design space
\( \epsilon_{\text{LIN}}(X) \) linearization error computed at the \( X \) point of the design space in the \( i \)th optimization cycle
\( \epsilon_{\text{LIN}} \) allowable linearization error computed in the \( i \)th design cycle
\( \epsilon_{\text{INV}} \) linearization error computed at the intermediate solution of \( i \)th design cycle
\( \epsilon_{\text{INV}} \) linearization error computed for the intermediate solution obtained in the \((i-1)\)th design cycle \((X_{i-1})\) if the nonlinear functions are linearized about the current intermediate design point \((X_i)\)
\( S_k \) descent directions defined by perturbing the \( k \)th design variable
\( k \) counter of perturbed variables and search directions \( S_k \)
\( S_{\text{des},r} \) limit directions orthogonal to the gradient of cost function in the linearization point \( P_i^0 \)
\( r \) counter of the limit directions orthogonal to the gradient of cost function in \( P_i^0 \)
\( p \) counter of the feasible sub-segments \((\text{i.e., parts in which the solution step size } ||X_{\text{SOL}} - X_i|| \text{ is divided})\) actually found when the check on local non-convexity of the constraint domain is performed
\( \nabla \) gradient operator
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