

Identification of Material Parameters

with LS-OPT®

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1

Overview

- About LS-OPT
- Introduction
 - Parameter identification
- Optimization algorithm
 - Sequential Response Surface Method
- Matching of scalar values and curve matching metrics
- Setup of a parameter identification problem in LS-OPT
 - Live demonstration
- Parameterization of LS-DYNA input curves
 - Live demonstration
- Remarks

About LS-OPT

■LS-OPT is a standalone optimization software → can be linked to any simulation code

Interface to LS-DYNA and MSC-Nastran

User-defined Interface

Current production version is LS-OPT 5.1

■ LS-OPT Support web page → <u>www.lsoptsupport.com</u>

Download of Executables

Tutorials

HowTos / FAQs

Documents

te to LS-OPT Support Site... -.... 💠 LS-OPT Support LS-OPT, the graphical optimization tool that interfaces perfectly with LS-DYI lows the user to structure the design process, explore the design space and compute optimal designs according to cified constraints and objectives. The program is also hig ted to the solution of system identification p and stochastic analysis The graphical tool LS-OPTui interfaces with LS-DYNA and provides an environment to specify on onitor and control parallel simulations and post-process optimization data, as well as viewing multiple designs S-PREPOST Applications: Design Optimization, Design of Experiments (Sensitivity A in LS-OPT 4.2 Ontimizatio Aug 18, 2011 Full Vehicle MDC Example Jun 03, 201 Constraints, mixed continuous/discrete variables, multiple load cases, etc. ase of LS-OPTO Multi-Objective optimization (Pareto Frontier) liability based design optimization May 13, 2011 Official Release of LS-TaSC (Topology and vstem/Parameter Identification available Material parameter eva Apr 19, 2011 Calibration of test results eta Release of LS-TaSC Feb 02, 201 Design Exploration Meta Models: Interrelation design variables vs. system response · Study of design changes Sensitivity Studies

Parameter identification: Objectives

- Parameter Identification problems are non-linear inverse problems solved using optimization
- A computed curve (from LS-DYNA[®]), dependent on parameters, is matched to an experimental curve
- Optimization provides a calibration of the unknown parameters
- Principle technologies involved:
 - Optimization algorithm
 - Curve Matching metric





Optimization Algorithm -Sequential Response Surface Method (SRSM)

Optimization Algorithm

Response Surface Methodology - Optimization Process



Optimization Algorithm Find an Optimum on the Response Surface (one iteration) Starting value on response surface Optimization of sub-problem (response surface) using e.g. LFOPC algorithm Optimum (predicted by response surface) Optimum (computed by simulation using design variables)

Optimization Algorithm

Sequential Response Surface Methodology (SRSM)





Objective Functions matching of scalar values and curve matching metrics

Matching of scalar values

Standard Composite Functions

Targeted Formulation

$$F = \sum_{j=1}^{m} W_j \left[\frac{f_j(\mathbf{x}) - G_j}{s_j} \right]^2$$

 $f_j(\mathbf{x})$: simulation response as function of variable vector \mathbf{x}

- G_i : target value
 - ; weighting factor
- S_i : normalization factor

Matching of scalar values

Targeted Formulation









Partial Curve Mapping

Partial Curve Mapping algorithm

- Normalize the curves to the test (experimental) curve
 - Avoids problems with different magnitudes for abscissa and ordinate
 - Unit independent
- Map the short curve onto the long curve so that the lengths are equal (mild filtering of curves by user is recommended)
- The distance is defined by the area between the short curve and the mapping
- Optimize the offset to find the smallest distance between the curves



Parameter Identification with Test Curves

Interface for Curve Mapping

| | Curve Matching | Composite | | X | |
|--------------------|-----------------------------------|-----------|------------|----------------|----|
| Name: | | | | | |
| Residual2 | | | | | |
| Algorithm | | Importe | ed experim | ental cu | r١ |
| | or (difference in curve Y values) | in 2-col | umn forma | t | |
| Curve Mapping (s) | spe of area between curves) | | | | |
| Target curve: | | | | | |
| test2 | | | ✓ add net | w file history | |
| Computed curve: | | | | | |
| s2_vs_e2 | K | | | ~ | |
| Regression Points | | | | | |
| From target curv | | | | | |
| O Fixed number (ed | quidistant, interpolated) | | | | |
| | | | | | |
| Г | | | | | |
| | Computed history/c | rossplot | | | |
| | | | Cancel | ок | |
| | | | | | |



Setup of a parameter identification problem in LS-OPT -Example

Example

Example: Material properties of a foam





Live Demonstration



Parameterization of LS-DYNA input

curves



Parameterization of LS-DYNA input curves

- User-defined preprocessor (Python, Perl, ...) that generates LS-DYNA include file containing the curve
- Parameterized analytical function
 - spline, polynomial, exponential function, ...





Live Demonstration

Remarks

- Make sure to evaluate exactly the same entities from simulation and test (filtering, ...)
- The result can never be better than the (material-) model
- Use <u>appropriate</u> analytical function for parameterization of LS-DYNA input curves
- Ranges for parameters?
 → increase if optimal value is bound and result not good enough (if parameter is sensitive!)
- Additional objective functions like max value, time of failure, ... might improve the results
- Multiple load cases: objectives might be in conflict