We Only Got Started With Dakota

Bob Cochran

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What Did We Cover?

- Using Dakota
 - Dakota command line usage
 - Interfacing your simulation program
 - Text based input and output files
- Sensitivity Analysis
 - Centered Parameter Study
 - Sampling Study
 - Latin hypercue sampling (LHS)
 - Variance Based Decomposition
 - Direct approach
 - Surrogate model
 - Surrogate Models
 - Polynomial Chaos Expansion

What Did We Cover? (continued)

- Uncertainty Quantification
 - Describing uncertainty using probability distribution functions
- Uncertainty Propagation
 - Sampling
 - Variance Based Decomposition

Many Methods in One Tool



 Sensitivity Analysis Designs: MC/LHS, DACE, sparse grid, one-at-a-time Analysis: correlations, scatter, Morris effects, Sobol indices 	 Uncertainty Quantification MC/LHS/Adaptive Sampling Reliability Stochastic expansions Epistemic methods
 Optimization Gradient-based local Derivative-free local Global/heuristics Surrogate-based 	 Calibration Tailored gradient-based Use any optimizer Bayesian inference

Interface Dakota to your simulation once, then apply various algorithms depending on your goal...

How related tools compare to Dakota DARO Sandia National

Software	Methods	Simulation Interface	Hybrid Analyses	R&D	Parallel Computing
NASA UQTools	UQ	???	no	no	???
OpenTURNS	UQ	???	no	no	no
LLNL PSUADE	SA	yes	no	no	???
MIT MUQ	UQ, Opt	custom	no	yes	no
SNL UQTk	UQ, Cal	yes	no	yes	no
OpenMDAO	Opt	yes	no	yes	yes
COIN-OR	Opt	no	no	yes	some
NLOpt	Opt	yes	no	no	no
Nessus	UQ	yes	no	???	no
GoldSim	UQ	???	no	???	no
PEST	Cal	yes	no	???	yes

aboratories

Other SA Approaches Typically Only Require Changing the Method Block



Dakota Reference Manual guides in specifying keywords

method,
sampling
sample_type lhs
seed = 52983
samples = 100

LHS Sampling

method,
sampling
sample_type lhs
seed = 52983
samples = 500
<pre>variance_based_decomp</pre>

Variance-based Decomposition using LHS Sampling method, dace oas main_effects seed = 52983 samples = 500

Main Effects Analysis using Orthogonal Arrays

method,
 psuade_moat
 partitions = 3
 seed = 52983
 samples = 100

Morris One-at-a-Time

Dakota SA Methods Summary



Category	Dakota method names	univariate trends	correlations	modified mean, s.d.	main effects Sobol inds.	importance factors / local sensis	
Parameter	centered, vector, list	Ρ					
studies	grid		D		Р		
Sampling	sampling, dace lhs, dace random, fsu_quasi_mc, fsu_cvt with variance_based_decomp	Ρ	D		D		e F
DACE (DOE- like)	dace {oas, oa_lhs, box_behnken, central_composite}		D		D		
MOAT	psuade_moat			D			
PCE, SC	polynomial_chaos, stoch_collocation				D	D	
Mean value	local_reliability					D	

also multipurpose!

D: Dakota-generated

P: Post-processing required (3rd party tools)

Guide to Optimization Methods See Usage Guidelines in User's Manual



Category	Dakota method names	Continuous Variables	Categorical/ Discrete Variables	Bound Constraints	General Constraints
	optpp_cg	x			
Gradient-Based	dot_bfgs, dot_frcg, conmin_frcg	x		x	
Local (smooth)	<pre>npsol_sqp, nlpql_sqp, dot_mmfd, dot_slp, dot_sqp, conmin_mfd, optpp_newton, optpp_q_newton, optpp_fd_newton</pre>	x		x	x
Gradient-Based Global (smooth)	hybrid, multi_start	x		x	x
Derivative-Free Local (nonsmooth)	optpp_pds	x		x	
	coliny_cobyla, coliny_pattern_search, coliny_solis_wets, surrogate_based_local	x		x	x
	asynch_pattern_search, mesh_adaptive_search	x	x	x	x
	<pre>ncsu_direct, genie_direct, genie_opt_darts</pre>	x		x	
Derivative-Free Global (nonsmooth)	coliny_direct, efficient_global, surrogate_based_global	x		x	x
	coliny_ea, soga, moga (multiobjective)	x	x	x	x

For multi-objective problems: use weighted sum with any method, pareto_set, or moga.

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Category	Dakota method names	Continuous Variables	Categorical/ Discrete Variables	Bound Constraints	General Constraints
	optpp_cg	x			
Gradient-Based	dot_bfgs, dot_frcg, conmin_frcg	x		x	
Local (smooth)	<pre>npsol_sqp, nlpql_sqp, dot_mmfd, dot_slp, dot_sqp, conmin_mfd, optpp_newton, optpp_q_newton, optpp_fd_newton</pre>	x		x	x
Gradient-Based Global (smooth)	hybrid, multi_start	x		x	x
Derivative-Free Local (nonsmooth)	optpp_pds	x		x	
	coliny_cobyla, coliny_pattern_search, coliny_solis_wets, surrogate_based_local	x		x	x
	asynch_pattern_search, mesh_adaptive_search	x	x	x	x
	<pre>ncsu_direct, genie_direct, genie_opt_darts</pre>	x		x	
Derivative-Free Global (nonsmooth)	coliny_direct, efficient_global, surrogate_based_global	x		x	x
	coliny_ea, soga, moga (multiobjective)	x	x	x	x

For multi-objective problems: use weighted sum with any method, pareto_set, or moga.

Guide to Calibration Methods See Usage Guidelines in User's Manual



Category	Specialized Calibration Methods	General Optimization Methods	Continuous Variables	Categorical/ Discrete Variables	Bound Constraints	General Constraints
		optpp_cg	x			
Gradient-	nl2sol	<pre>dot_bfgs, dot_frcg, conmin_frcg</pre>	x		x	
Based Local (smooth)	nlssol_sqp, optpp_g_newton	<pre>npsol_sqp, nlpql_sqp, dot_mmfd, dot_slp, dot_sqp, conmin_mfd, optpp_newton, optpp_q_newton, optpp_fd_newton</pre>	x		x	x
Gradient- Based Global (smooth)	hybrid*, multi_start*	hybrid, multi_start	x		x	x
		optpp_pds	x		x	
Derivative- Free Local (nonsmooth)	surrogate_ based_local*	coliny_cobyla, coliny_pattern_search, coliny_solis_wets	x		x	x
		asynch_pattern_search, mesh_adaptive_search	x	x	x	x
Derivative- Free Global		<pre>ncsu_direct, genie_direct, genie_opt_darts</pre>	x		x	
		<pre>coliny_direct, efficient_global, surrogate_based_global</pre>	x		x	x
(nonsmooth)		coliny_ea, soga	x	x	x	x

*: in conjunction with a specialized gradient-based method (nl2sol, nlssol, optpp_g_newton)

Dakota UQ Methods Summary



character	method class	problem character	variants
aleatory	probabilistic sampling	nonsmooth, multimodal, modest cost, # variables	Monte Carlo, LHS, importance
	local reliability	smooth, unimodal, more variables, failure modes	mean value and MPP, FORM/SORM,
	global reliability	nonsmooth, multimodal, low dimensional	EGRA
	stochastic expansions	nonsmooth, multimodal, low dimension	polynomial chaos, stochastic collocation
epistemic	interval estimation	simple intervals	global/local optim, sampling
	evidence theory	belief structures	global/local evidence
both	nested UQ	mixed aleatory / epistemic	nested

Also see Usage Guidelines in User's Manual