OPTIMIZATION OF COMPLEX SYSTEMS IN THE PRESENCE OF UNCERTAINTY

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Optimization of Complex Systems in the Presence of Uncertainty and Approximations

Engineering decisions are invariably made under substantial uncertainty about current and future system cost and response, including cost and response associated with low-probability but high-consequence events. Such events motivate approaches that typically have centered on constraining or minimizing probability of failure, in contrast to the risk-neutral approach of constraining or minimizing expected values. The research under this proposal has, instead, developed concepts of risk-averse decision making between these extremes with the aim of achieving an advanced methodology better able to deal with risks and reliability in engineering design.

Measures of risk that go beyond statistical quantiles to so-called superquantiles (CVaR) and their mixtures have been the main focus. The results have explored their superior properties and enhanced computability along with surprising implications that standard least-squares regression in statistical approximations might better be supplanted by generalizations like quantile and even superquantile regression. Superquantile regression, which provides a cautious and powerful tool, is completely new. It is entirely a product of this grant research.
Summary of Effort

The research was a collaborative effort with Johannes Royset of the Naval Postgraduate School, who had separate funding from AFOSR. The PI received one month of salary in each of the three years for his research, which closely followed the lines of the proposal.

Accomplishments and New Findings

Engineering decisions are invariably made under substantial uncertainty about current and future system cost and response, including cost and response associated with low-probability, high-consequence events. A risk-neutral decision maker would rely on expected values when comparing designs, while a risk-averse decision maker might adopt nonlinear utility functions or failure probability criteria. Paper [2] shows that these models for making decisions fall within a framework of risk measures that includes many other possibilities. General recommendations for selecting risk measures lead to decision models for risk-averse decision making that comprehensively represent risks in engineering systems, avoid paradoxes, and accrue substantial benefits in subsequent risk, reliability, and cost optimization. The paper provides an overview of the framework of decision making based on risk measures.

Superquantile risk, also known as conditional value-at-risk (CVaR), is widely used as a coherent measure of risk due to its improved properties over those of quantile risk (value-at-risk). The latest paper [1] considers second-order superquantile/CVaR measures of risk, which represent further “smoothing” by averaging the classical quantities. It also steps further and examines the more general “mixed” superquantile/CVaR measures of risk with fundamental importance in dual utility theory. It establishes representations of these mixed and second-order superquantile risk measures in terms of risk profiles, risk envelopes, and risk identifiers. The expressions facilitate the development of dual methods for mixed and
second-order superquantile risk minimization as well as superquantile regression, a second-order version of quantile regression.

Superquantiles (also called conditional values-at-risk) are useful tools in risk modeling and optimization, with expanding roles beyond these areas. Paper [5] provides a broad overview of superquantiles and their versatile applications. We see that superquantiles are as fundamental to the description of a random variable as the cumulative distribution function (cdf), they can recover the corresponding quantile function through differentiation, they are dual in some sense to superexpectations, which are convex functions uniquely defining the cdf, and they also characterize convergence in distribution. A superdistribution function defined by superquantiles leads to higher-order superquantiles as well as new measures of risk and error, with important applications in risk modeling and generalized regression.

Random variables can be described by their cumulative distribution functions, a class of nondecreasing functions on the real line. Those functions can in turn be identified, after the possible vertical gaps in their graphs are filled in, with maximal monotone relations, and this is the theme of paper [3], which provides a new foundation for superquantile analysis. Maximal monotone relations are known to be the subdifferentials of convex functions. Analysis of these connections yields fresh insights. The generalized inversion operation between distribution functions and quantile functions corresponds to graphical inversion of monotone relations. In subdifferential terms, it corresponds to passing to conjugate convex functions under the Legendre-Fenchel transform. Among other things, this shows that convergence in distribution for sequences of random variables is equivalent to graphical convergence of the monotone relations and epigraphical convergence of the associated convex functions. Measures of risk that employ quantiles (VaR) and superquantiles (CVaR), either individually or in mixtures, are illuminated in this way. Formulas for their calculation are seen from a perspective that reveals how they were discovered. The approach leads further to developments in which the superquantiles for a given distribution are interpreted as the quantiles for an overlying “superdistribution.” In this way a generalization of Koenker-Basset error is derived which provides a platform for superquantile regression as a higher-order extension
of quantile regression.

Paper [4] presents a generalized regression technique centered on a superquantile (also called conditional value-at-risk) that is consistent with that coherent measure of risk and yields more conservatively fitted curves than classical least-squares and quantile regression. In contrast to other generalized regression techniques that approximate conditional superquantiles by various combinations of conditional quantiles, we directly and in perfect analog to classical regression obtain superquantile regression functions as optimal solutions of certain error minimization problems. We show the existence and possible uniqueness of regression functions, discuss the stability of regression functions under perturbations and approximation of the underlying data, and propose an extension of the coefficient of determination R-squared for assessing the goodness of fit. The paper presents two numerical methods for solving the error minimization problems and illustrates the methodology in several numerical examples in the areas of uncertainty quantification, reliability engineering, and financial risk management.

**Personnel Supported**

Personnel supported under the grant: Prof. R.T. Rockafellar, University of Washington, Seattle. Students involved but not supported: S. Miranda (Naval Postgraduate School).

**Publications**

The following publications were written under the project:

see http://www.math.washington.edu/~rtr/mypage.html for access to the papers.


**Interactions/Transitions**

The project resulted in the following presentations by R.T. Rockafellar:


New discoveries, inventions, or patent disclosures

No patents, but several discoveries and advances as described above.