

Real Options for Project Schedules (ROPS)

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Abstract

Real Options for Project Schedules (ROPS) has three recursive sampling/optimization shells. An outer Adaptive Simulated Annealing (ASA) optimization shell optimizes parameters of strategic Plans containing multiple Projects containing ordered Tasks. A middle shell samples probability distributions of durations of Tasks. An inner shell samples probability distributions of costs of Tasks. PATHTREE is used to develop options on schedules. Algorithms used for Trading in Risk Dimensions (TRD) are applied to develop a relative risk analysis among projects.

KEYWORDS: options; simulated annealing; risk management; copula; nonlinear; statistical

[†] L. Ingber, "Real Options for Project Schedules (ROPS)", Report 2007:ROPS, Lester Ingber Research, Ashland, OR, 2007. URL http://www.ingber.com/markets07_rops.pdf. Early drafts can be downloaded from <http://arxiv.org/abs/0704.0090> and <http://ssrn.com/abstract=977839>.

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1. Introduction

This paper is a brief description of a methodology of developing options (in the sense of financial options, e.g., with all Greeks defined below) to schedule complex projects, e.g., such as the massive US Army project Future Combat Systems (FCS).

The major focus is to develop Real Options for non-financial projects, as discussed in other earlier papers [2,4,18]. Data and some guidance on its use has been reported in a previous study of FCS [1,5]. The need for tools for fairly scheduling and pricing such a complex project has been emphasized in Recommendations for Executive Action in a report by the U.S. General Accounting Office (GAO) on FCS [20], and they also emphasize the need for management of FCS business plans [19,21].

The concept of Real Options is the subject of many papers. Some papers give good advice on how to apply these tools [3,22,23]. The approach given in this paper includes important optimization and options algorithms not present in previous papers, and a process of aggregating tasks to permit top-level control of projects, which enables development of more real Real Options.

2. Goals

A given Plan results in $S(t)$, money allocated by the client/government is defined in terms of Projects $S_i(t)$,

$$S(t) = \sum_i S_i(t) \quad (1)$$

where $a_i(t)$ may be some scheduled constraints. PATHTREE processes a probability tree developed over the life of the plan T , divided into N nodes at times $\{t_n\}$, each with mean epoch length dt [16]. Options, including all Greeks defined below, familiar to financial markets, are calculated for quite arbitrary nonlinear means and variances of multiplicative noise [6,10]. Options are calculated as expected values along stochastic paths of their underlying markets, and the Greeks are functional derivatives of these options. This enables fair pricing of expected values of a market over extended periods and including changes to underlying variables that can occur along these stochastic paths. The ability to process nonlinear functions defining probability distributions is essential for real-world applications.

Each Task has a range of durations, with nonzero A_i , with a disbursement of funds used, defining $S_i(t_n)$. Any Task dependent on a Task completion is slaved to its precursor(s).

We develop the Plan conditional probability density (CPD) in terms of differenced costs, dS ,

$$P(S \pm dS; t_n + dt | S; t_n)$$

P is modeled/cast/fit into the functional form

$$P(S \pm dS; t_n + dt | S; t_n) = (2\pi g^2 dt)^{-\frac{1}{2}} \exp(-Ldt) \quad (2)$$

$$L = \frac{(dS - fdt)^2}{(2g^2 dt^2)}$$

where f and g are nonlinear function of cost S and time t . The g^2 variance function absorbs the multiple Task cost and schedule statistical spreads, to determine $P(dS, t)$, giving rise to the stochastic nature of dollars spent on the Plan.

A given Project i with Task k has a mean duration i_{ik} , with a mean cost S_{ik} . The spread in dS has two components arising from: (1) a stochastic duration around the mean duration, and (2) a stochastic spread of mean dollars around a deterministic disbursement at a given time. Different finite-width asymmetric distributions are used for durations and costs. For example, the distribution created for Adaptive Simulated Annealing (ASA) [8], originally called Very Fast Simulated Re-annealing [7], is a finite-ranged distribution with shape determined by a parameter “temperature” q .

$$ASA(x; q) = \frac{1}{2(|x| + q) \ln(1 + \frac{1}{q})} \quad (3)$$

For each state (whether duration or cost): (a) A random binary choice can be made to be higher or lower than the mean, using any ratio of probabilities selected by the client. (b) Then, an ASA distribution is used on the chosen side. Each side has a different q , each falling off from the mean. This is illustrated and further described in Fig. 1.

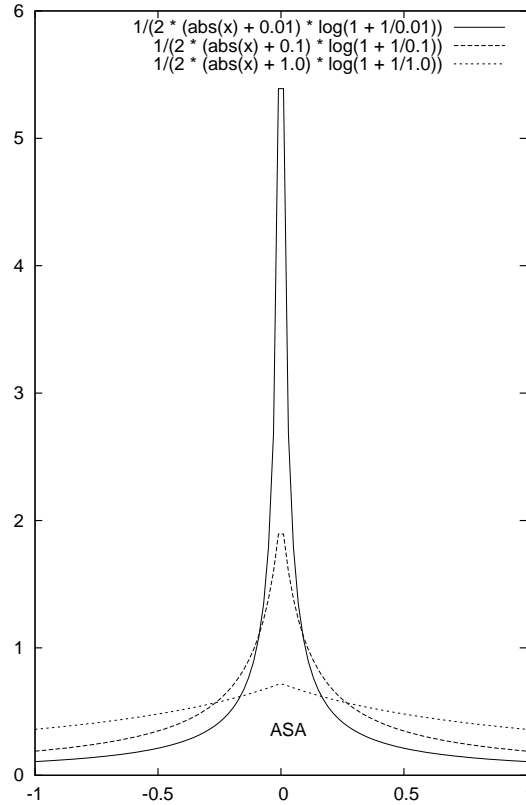


Fig. 1. The ASA distribution can be used to develop finite-range asymmetric distributions from which a value can be chosen for a given state of duration or cost. (a) A random binary distribution is selected for a lower-than or higher-than mean, using any ratio of probabilities selected by the client. Each side of the mean has its own temperature q . Here ASA distributions are given for $q = \{0.01, 0.1, 1.0\}$. The range can be scaled to any finite interval and the mean placed within this range. (b) A uniform random distribution selects a value from $[-1,1]$, and a normalized ASA value is read off for the given state.

At the end of the tree at a time T (T also can be a parameter), there is a total cost at each node $S(T)$, called a final “strike” in financial language. (A final strike might also appear at any node before T due to cancellation of the Project using a particular kind of schedule alternative.) Working backwards, options are calculated at time t_0 . Greeks defined below assess sensitivity to various variables, e.g., like those discussed in previous papers [18], but here we deliver precise numbers based on as much real-world information as is available.

3. Data

The following data are used to develop Plan CPD. Each Task i has

- (a) a Projected allocated cost, C_i
- (b) a Projected time schedule, T_i
- (c) a CPD with a statistical width of funds spent, SW_{S_i}
- (d) a distribution with a statistical width of duration, SW_{T_i}
- (e) a range of durations, R_{T_i}
- (f) a range of costs, R_{S_i}

Expert guesses need to be provided for (c)-(f) for the prototype study.

A given Plan must be constructed among all Tasks, specified the ordering of Tasks, e.g., obeying any sequential constraints among Tasks.

4. Three Recursive Shells

4.1. Outer Shell

There may be several parameters in the Project, e.g., as coefficients of variables in means and variances of different CPD. These are optimized in an outer shell using ASA [8]. This end product, including MULTI_MIN states returned by ASA, gives the client flexibility to apply during a full Project [18]. We may wish to minimize $Cost/T$, or $(CostOverrun - CostInitial)/T$, etc.

4.2. Middle Shell

To obtain the Plan CPD, an middle shell of Monte Carlo (MC) states are generated from recursive calculations. A Weibull or some other asymmetric finite distribution might be used for Task durations. For a given state in the outer middle, a MC state has durations and mean cost disbursements defined for each Task.

4.3. Inner Shell

At each time, for each Task, the differenced cost $(S_{ik}(t + dt) - S_{ik}(t))$ is subjected to a inner shell stochastic variation, e.g., some asymmetric finite distribution. The net costs $dS_{ik}(t)$ for each Project i and Task k are added to define $dS(t)$ for the Plan. The inner shell cost CPD is re-applied many times to get a set of $\{dS\}$ at each time.

5. Real Options

5.1. Plan Options

After the Outer MC sampling is completed, there are histograms generated of the Plan's $dS(t)$ and $dS(t)/S(t - dt)$ at each time t . The histograms are normalized at each time to give $P(dS, t)$. At each time t , the data representing P is "curve-fit" to the form of Eq. (2), where f and g are functions needed to get good fits, e.g., fitting coefficients of parameters $\{x\}$

$$f = x_{f0} + x_{f1}S + x_{f2}S^2 + \dots$$

$$g = x_{g0} + x_{g1}S + x_{g2}S^2 + \dots \quad (4)$$

At each time t , the functions f and g are fit to the function $\ln(P(dS, t))$, which includes the prefactor containing g and the function L which may be viewed as a Padé approximate of these polynomials.

Complex constraints as functions of $S_{ik}(t)$ can be easily incorporated in this approach, e.g., due to regular reviews by funding agencies or executives. These P 's are input into PATHTREE to calculate options for a given strategy or Plan.

5.2. Risk Management of Project Options

If some measure of risk among Projects is desired, then during the MC calculations developed for the top-level Plan, sets of differenced costs for each Project, $dS_i(t)$ and $dS_i(t)/S_i(t - dt)$, stored from each of the Project's Tasks. Then, histograms and Project CPDs are developed, similar to the development of the Plan CPD. A copula analysis, coded in TRD for risk management of financial markets, are applied to develop a relative risk analysis among these projects [11,15]. In such an analysis, the Project marginal CPDs are all transformed to Gaussian spaces, where it makes sense to calculate covariances and correlations. An audit trail back the original Project spaces permits analysis of risk dependent on the tails of the Project CPDs. Some details are given in papers dealing with generic portfolios [12-14].

It is *not* appropriate to simply use a standard Gaussian second moment of a non-Gaussian distribution (most common in real-world data) to assess “risk”. It is *not* appropriate to simply use standard Gaussian second moments of non-Gaussian distributions (most common in real-world data) to assess correlations of variables.

5.2.1. Post-Processing of Multiple Optima/“Mutations”

Within a chosen resolution of variables and parameters to be sampled, often the huge numbers of possible states to be importance-sampled presents multiple optima and sometimes multiple optimal states. It should be understood that any sampling algorithm processing a huge number of states can find many multiple optima. As used in TRD, relevant to this ROPS approach, ASA’s MULTI_MIN OPTIONS are used to save multiple optima during sampling. Some algorithms might label these states as “mutations” of optimal states. It is important to be able to include them in final decisions, e.g., to apply additional metrics of performance specific to applications. Experience shows that all criteria are not best considered by lumping them all into one cost function, but rather good judgment on risk versus benefit should be applied to multiple stages of pre-processing and post-processing when performing such sampling.

5.3. Greeks

Traders on financial options use “Greeks” to assess value of markets. Consider a position Π on an option. The change in value associated with changes in time dt , underlying dS , risk-free interest-rate ρ , and volatility $d\sigma$ is given by a sum over the “Greeks,” as derived from a Taylor expansion of the differenced position [10],

$$d\Pi = \frac{\partial\Pi}{\partial S} dS + \frac{1}{2} \frac{\partial^2\Pi}{\partial S^2} dS^2 + \frac{\partial\Pi}{\partial\sigma} d\sigma + \frac{\partial\Pi}{\partial t} dt + \frac{\partial\Pi}{\partial r} dr + \frac{\partial^2\Pi}{\partial S \partial\sigma} dS d\sigma + \frac{1}{2} \frac{\partial^2\Pi}{\partial\sigma^2} d\sigma^2 + \frac{1}{6} \frac{\partial^3\Pi}{\partial S^2 \partial\sigma} dS^2 d\sigma + \dots$$

$$d\Pi = \Delta dS + \frac{1}{2} \Gamma dS^2 + K d\sigma + \Theta dt + \rho dr + \Delta' dS d\sigma + \frac{1}{2} K' d\sigma^2 + \frac{1}{6} \Gamma' dS^2 d\sigma + \dots \quad (5)$$

where Δ = Delta, Γ = Gamma, K = Kappa (sometimes called the Vega), Θ = Theta, ρ = Rho. The Greeks and the covariance matrix are functions of a strike X at the value the underlying of the option is priced to be when it is exercised [6,16], The Greeks are calculated using probability-distribution models selected by traders. They can draw from Black Scholes (most common), Ornstein-Uhlenbeck or more general uni-variable or multi-variable models (which are more appropriate to real-world data) [9,16,17].

The Greeks are useful numbers to gauge the sensitivity of the price of the option with respect to several of its underlying variables.

6. Generic Applications

ROPS can be applied to any complex scheduling of tasks similar to the FCS project. The need for government agencies to plan and monitor such large projects is becoming increasingly difficult and necessary [24]. Many large businesses have similar projects and similar requirements to manage their complex projects.

Similarly, large investment funds should at least internally develop estimates of the Real Options “fair value” of their revenue streams, based on realistic underlying probability distributions of their returns. They then can use these estimates as complementary measures of risk, and decide on what products they might invest in as insurance to protect their assets.

References

- [1] G.G. Brown, R.T. Grose, and R.A. Koyak, "Estimating total program cost of a long-term, high-technology, high-risk project with task durations and costs that may increase over time," *Military Operations Research* **11**, 41-62 (2006). [URL http://www.nps.navy.mil/orfacpag/resumePages/papers/Brownpa/Estimating_total_program_cost.pdf]
- [2] T.E. Copeland and P.T. Keenan, "Making real options real," *McKinsey Quarterly* 128-141 (1998). [URL http://faculty.fuqua.duke.edu/~charvey/Teaching/BA456_2006/McK98_3.pdf]
- [3] A. Damodaran, "The Promise and Peril of Real Options," Report Stern School Business, New York U, New York, 2007. [URL <http://www.stern.nyu.edu/~adamodar/pdf/papers/realopt.pdf>]
- [4] G. Glaros, "Real options for defense," *Transformation Trends* **June**, 1-11 (2003). [URL http://www.oft.osd.mil/library/library_files/trends_205_transformation_trends_9_june%202003_issue.pdf]
- [5] R. Grose, "Cost-constrained project scheduling with task durations and costs that may increase over time: Demonstrated with the U.S. Army future combat systems," Thesis, Naval Postgraduate School, Monterey, CA, 2004. [URL <http://www.stormingmedia.us/75/7594/A759424.html>]
- [6] J.C. Hull, *Options, Futures, and Other Derivatives, 4th Edition* (Prentice Hall, Upper Saddle River, NJ, 2000).
- [7] L. Ingber, "Very fast simulated re-annealing," *Mathl. Comput. Modelling* **12**, 967-973 (1989). [URL http://www.ingber.com/asa89_vfsr.pdf]
- [8] L. Ingber, "Adaptive Simulated Annealing (ASA)," Global optimization C-code, Caltech Alumni Association, Pasadena, CA, 1993. [URL <http://www.ingber.com/#ASA-CODE>]
- [9] L. Ingber, "High-resolution path-integral development of financial options," *Physica A* **283**, 529-558 (2000). [URL http://www.ingber.com/markets00_highres.pdf]
- [10] L. Ingber, "Statistical mechanics of portfolios of options," Report 2002:SMPO, Lester Ingber Research, Chicago, IL, 2002. [URL http://www.ingber.com/markets02_portfolio.pdf]
- [11] L. Ingber, "Trading in Risk Dimensions (TRD)," Report 2005:TRD, Lester Ingber Research, Ashland, OR, 2005. [URL http://www.ingber.com/markets05_trd.pdf]
- [12] L. Ingber, "Ideas by statistical mechanics (ISM)," Report 2006:ISM, Lester Ingber Research, Ashland, OR, 2006. [URL http://www.ingber.com/smni06_ism.pdf]
- [13] L. Ingber, "Ideas by Statistical Mechanics (ISM)," *J. Integrated Systems Design and Process Science* **11**, 31-54 (2007). [Special Issue: Biologically Inspired Computing.]
- [14] L. Ingber, "AI and Ideas by Statistical Mechanics (ISM)," in *Encyclopedia of Artificial Intelligence*, ed. by J.R. Rabuñal, J. Dorado, and A.P. Pazos (Information Science Reference, New York, 2008), p. 58-64. [ISBN 978-1-59904-849-9]
- [15] L. Ingber, "Trading in Risk Dimensions (TRD)," in *Handbook of Trading*, ed. by G.N. Gregoriou (Wiley, New York, 2010), p. (to be published).
- [16] L. Ingber, C. Chen, R.P. Mondescu, D. Muzzall, and M. Renedo, "Probability tree algorithm for general diffusion processes," *Phys. Rev. E* **64**, 056702-056707 (2001). [URL http://www.ingber.com/path01_pathtree.pdf]
- [17] L. Ingber and J.K. Wilson, "Volatility of volatility of financial markets," *Mathl. Computer Modelling* **29**, 39-57 (1999). [URL http://www.ingber.com/markets99_vol.pdf]
- [18] K.J. Leslie and M.P. Michaels, "The real power of real options," *McKinsey Quarterly* 4-22 (1997). [URL http://faculty.fuqua.duke.edu/~charvey/Teaching/BA456_2006/McK97_3.pdf]
- [19] General Accounting Office, "Future Combat System Risks Underscore the Importance of Oversight," Report GAO-07-672T, GAO, Washington DC, 2007. [URL <http://www.gao.gov/cgi-bin/getrpt?GAO-07-672T>]

- [20] General Accounting Office, “Key Decisions to Be Made on Future Combat System,” Report GAO-07-376, GAO, Washington DC, 2007. [URL <http://www.gao.gov/cgi-bin/getrpt?GAO-07-376>]
- [21] General Accounting Office, “Key Considerations for Planning Future Army Combat Systems,” Report GAO-09-410T, GAO, Washington DC, 2009. [URL www.gao.gov/highlights/d09410thigh.pdf]
- [22] U. Pape and S. Schmidt-Tank, “Valuing Joint Ventures Using Real Options,” Report ESCP-EAP Working Paper No. 7, ESCP-EAP European School Management, Paris, 2005. [URL <http://ssrn.com/abstract=695821>]
- [23] G. Sick and A. Gamba, “Some Important Issues Involving Real Options,” Haskayne School of Business, U Calgary, Calgary, 2005. [URL <http://ssrn.com/abstract=645581>]
- [24] B. Wysocki, Jr, “Is U.S. Government ‘Outsourcing Its Brain’?,” *Wall Street Journal* **March 30**, 1 (2007).