

Valuing Guaranteed Minimum Death Benefits in Variable Annuities with Knock-Out Options

Xiao Wang

Ph.D. Candidate
Department of Statistics and Actuarial Science
The University of Iowa

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- 1 The motivation – basic problem
- 2 Lapses and surrenders incorporation
- 3 Main results – value up-and-out option
- 4 Main results – a generalization of our formula

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we are to value a **K -strike put option** that is exercised at time T_x .



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Exponential case is sufficient

Our problem is reduced to finding

$$E[e^{-\delta T} b(S(T))],$$

where T is an **exponential** random variable **independent** of the stock price process $\{S(t)\}$.



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Therefore, instead of the payoff $[K - S(T)]^+$, we may want to consider the following payoff,

$$1(\max_{0 \leq t \leq T} S(t) < U) [K - S(T)]^+,$$

where $1(\cdot)$ is an indicator function and U is a barrier.

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This is the payoff of an **up-and-out put option**.



Our valuation problem with lapses and surrenders

We aim to determine the following expected present value for an **up-and-out option**,

$$E[e^{-\delta T} 1_{(\max_{0 \leq t \leq T} S(t) < U)} b(S(T))],$$

where $b(\cdot)$ is a death benefit function, and T is an **independent exponential** exercise date with mean $1/\lambda$.



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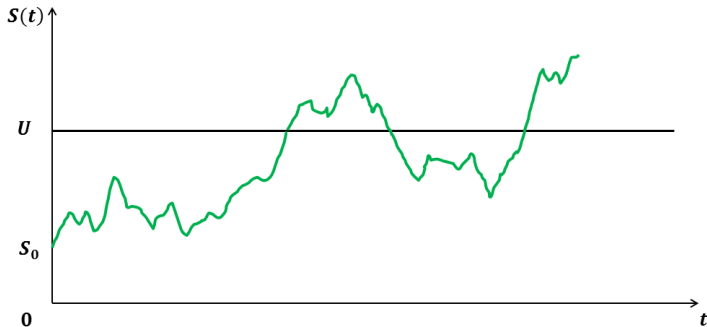
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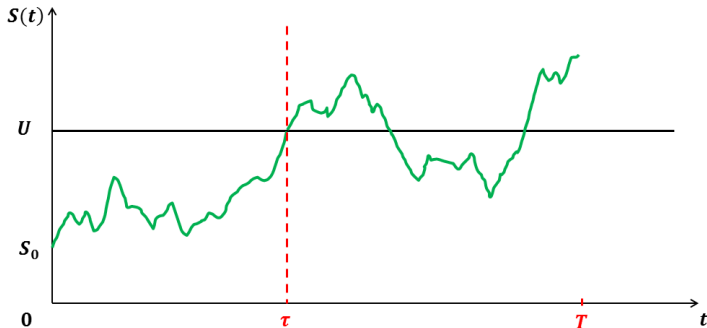
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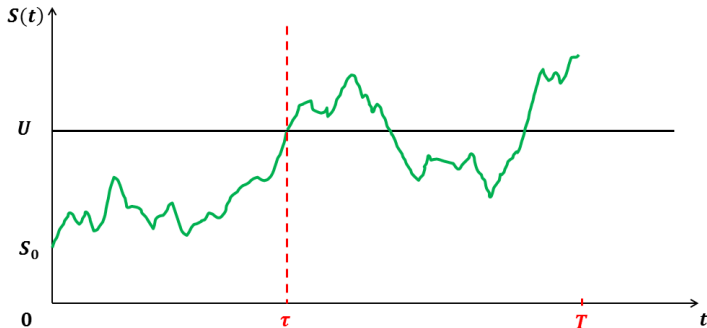
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Note that $\max_{0 \leq t \leq T} S(t) < U \Leftrightarrow \tau > T$, or $\max_{0 \leq t \leq T} S(t) \geq U \Leftrightarrow \tau \leq T$.



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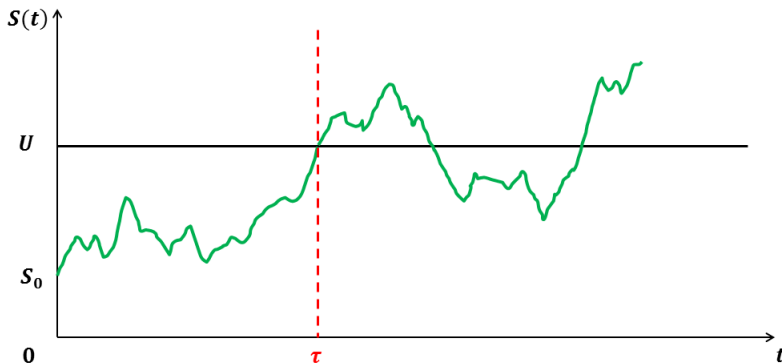
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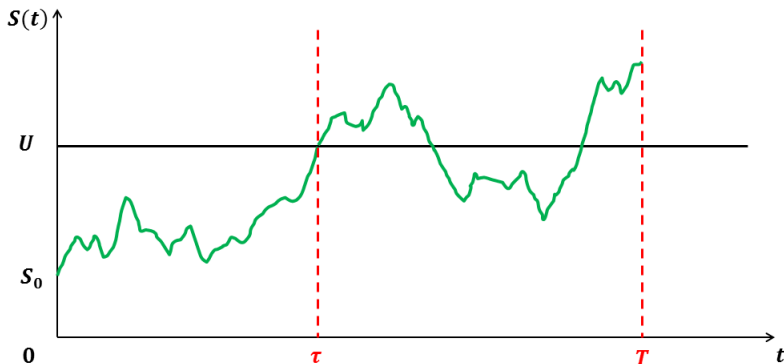
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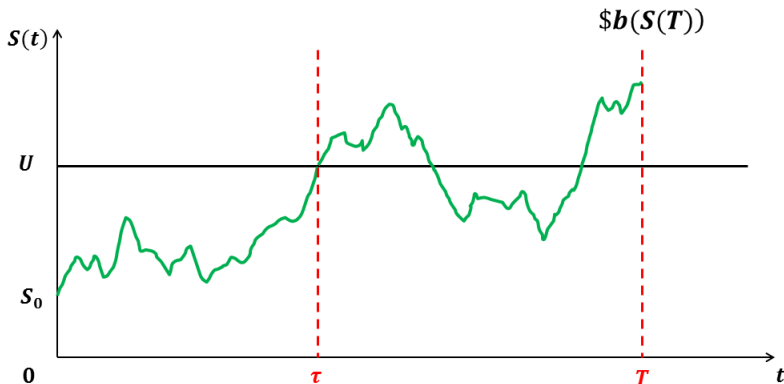
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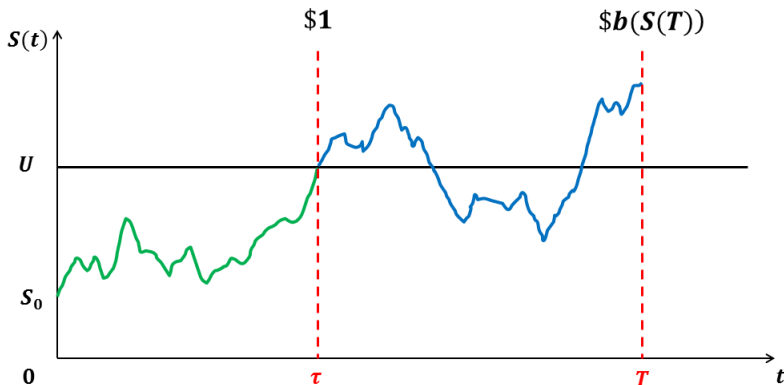
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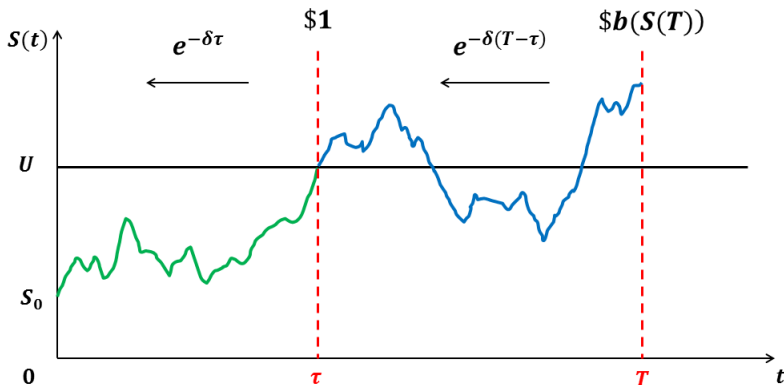
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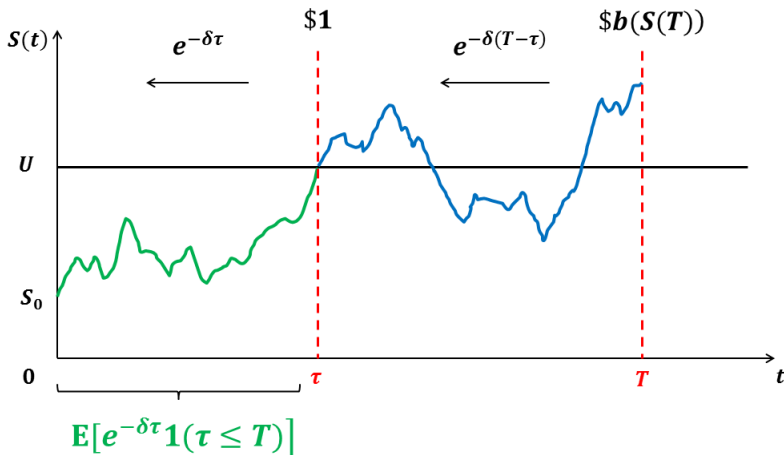
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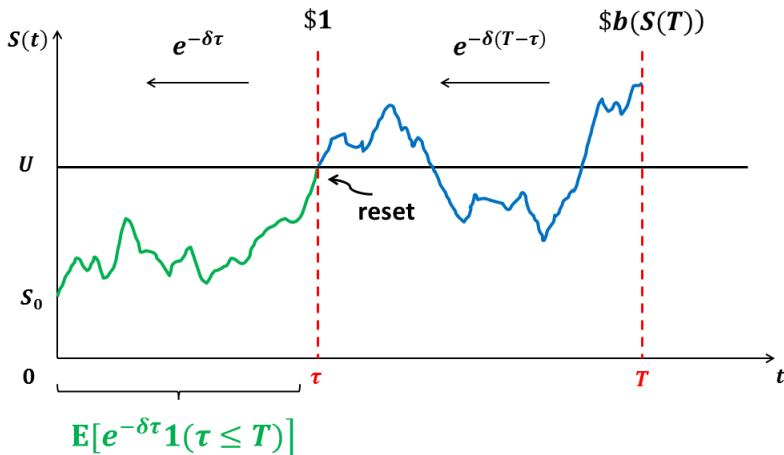
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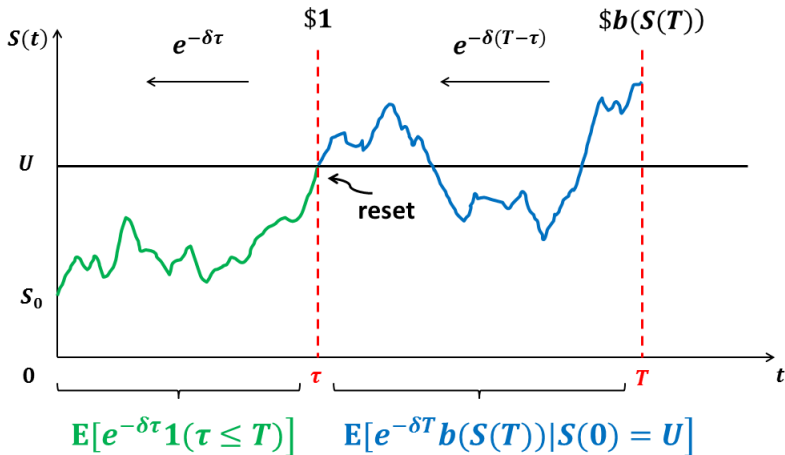
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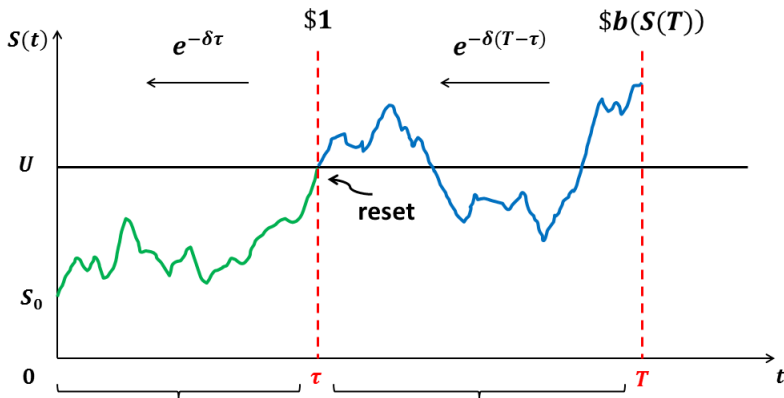
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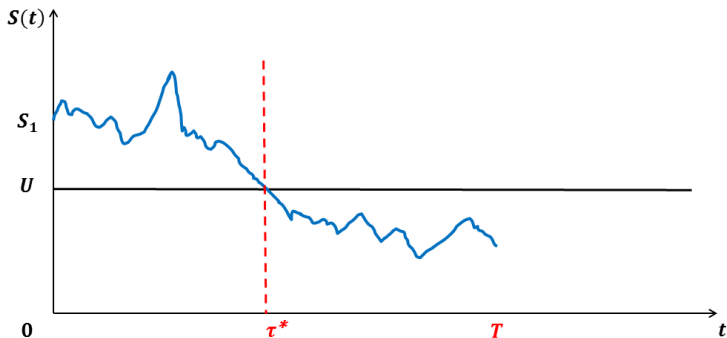
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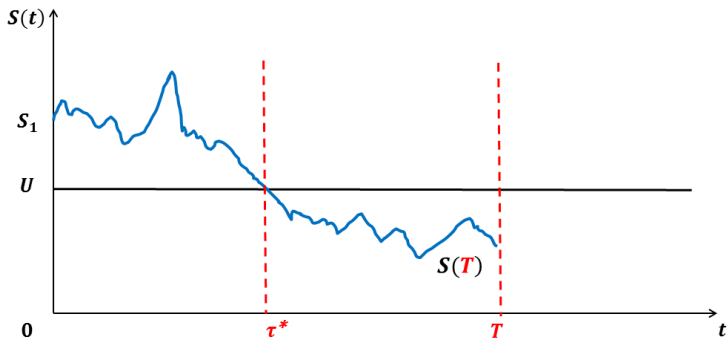
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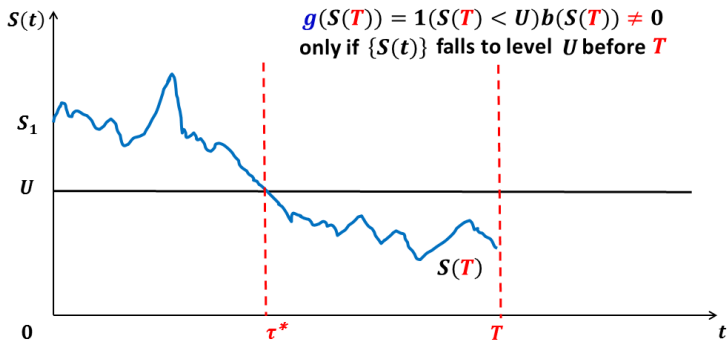
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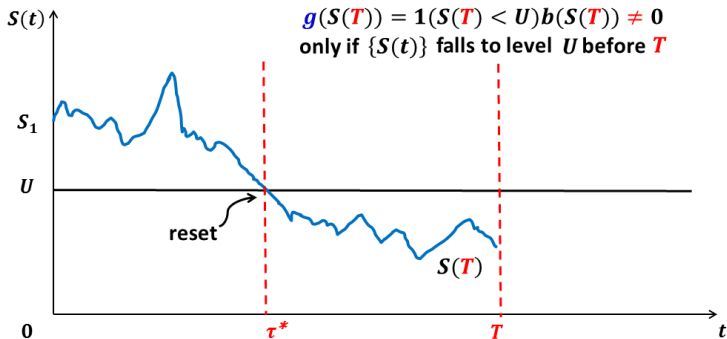
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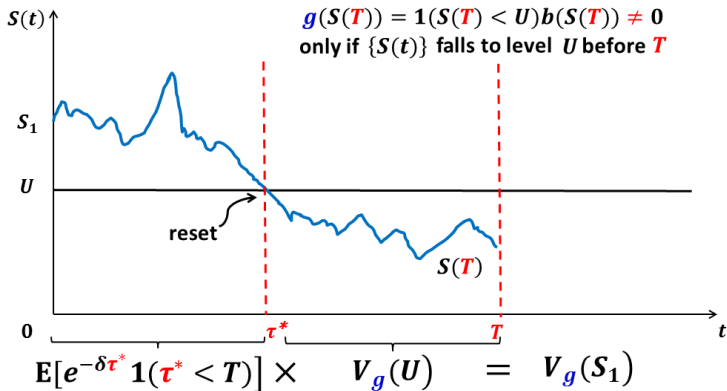
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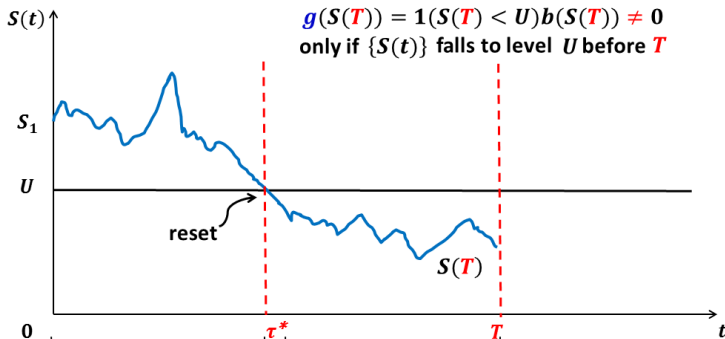
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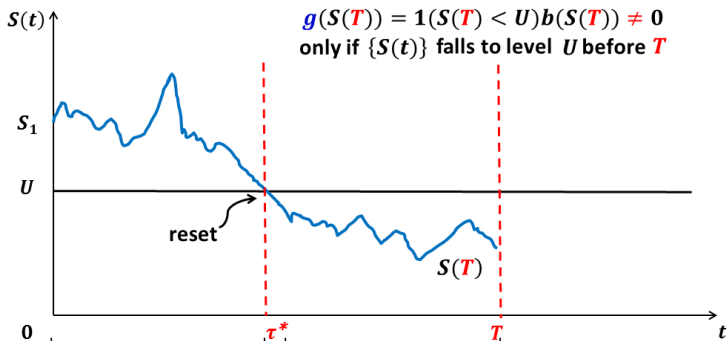
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- Single and double barrier options with **exponentially curved boundaries**. Example, $U(t) = Ue^{\theta t}, t \geq 0$.

References

-  Gerber, H.U., Shiu, E.S.W., Yang, H., 2012. Valuing equity-linked death benefits and other contingent options: a discounted density approach. *Insurance: Mathematics and Economics* 51, 73-92.
-  Gerber, H.U., Shiu, E.S.W., Yang, H., 2013. Valuing equity-linked death benefits in jump diffusion models. *Insurance: Mathematics and Economics*, 53, 615-623.
-  Dufresne, D., 2007. Fitting combinations of exponentials to probability distributions. *Applied Stochastic Models in Business and Industry* 23, 23-48.

Thank you!



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Because of **Optional Sampling Theorem**, if we replace fixed t by τ , the condition for martingale still holds for θ^+ ,

$$E[e^{-\delta\tau}1(\tau \leq T)[S(\tau)]^{\theta^+}] = [S(0)]^{\theta^+},$$

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Similarly, define τ^* as the **first** time when $\{S(t)\}$ **falls** to level U with an initial price $S_1 > U$.

Find $E[e^{-\delta\tau}1(\tau \leq T)]$ cont'd

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$$\Leftrightarrow E[e^{-\delta\tau}1(\tau \leq T)] = \left[\frac{S_0}{U}\right]^{\theta^+}.$$

Similarly, define τ^* as the **first** time when $\{S(t)\}$ **falls** to level U with an initial price $S_1 > U$. Then

$$E[e^{-\delta\tau^*}1(\tau^* \leq T)] = \left[\frac{S_1}{U}\right]^{\theta^-}.$$

