

# Adaptive Multilevel Splitting Method for Rare Event Estimation

Josephine Frankfort\*

Department of Mathematics Engineering, University of Hamburg, Hamburg, Germany

## Abstract

Rare event estimation is crucial in many fields, such as finance, engineering, and environmental science. These events, although infrequent, can have significant consequences, making their accurate prediction and understanding vital. Traditional methods often fall short due to the immense computational power required or lack of accuracy. The Adaptive Multilevel Splitting (AMS) method offers a robust alternative, providing a practical approach to estimating the probability of rare events in complex systems. The core of the AMS method is its splitting mechanism, where simulations that reach a certain intermediate threshold are duplicated. This process enhances the sampling of rare events, increasing the likelihood of observing them without requiring an excessive number of initial simulations. Define the rare event of interest and the corresponding threshold. Initialize a large number of independent simulations. Run the simulations until they reach the predefined threshold or fail. The successful paths are then analyzed to determine the next threshold. Simulations that reach the threshold are split, creating multiple copies that are slightly perturbed. This step increases the sample size for subsequent levels.

**Keywords:** Splitting • Dimensional • Method

## Introduction

Rare events are typically defined as occurrences with a very low probability but potentially catastrophic consequences. Examples include financial market crashes, structural failures in engineering, and extreme weather events. Estimating the likelihood of these events is challenging due to their infrequency, making direct simulation approaches computationally infeasible [1]. The AMS method is designed to address these challenges by efficiently estimating the probability of rare events through a series of intermediate steps [2].

## Literature Review

The AMS method is a sophisticated simulation technique that improves the efficiency of rare event estimation. It operates by dividing the problem into multiple levels or stages, progressively refining the estimate at each step. This approach contrasts with traditional Monte Carlo simulations, which require a prohibitively large number of samples to achieve accurate results for rare events. The AMS method adapts to the problem at hand, dynamically adjusting its parameters to improve accuracy and efficiency. This adaptability is crucial for dealing with complex, high-dimensional systems where fixed parameters may not yield optimal results. By breaking down the estimation process into multiple levels, the AMS method reduces the variance of the estimator. Each level focuses on a more probable event than the previous, gradually narrowing down to the rare event of interest [3].

Repeat the level advancement and splitting process until the rare event threshold is reached. Each iteration refines the estimate, improving accuracy. Once the simulations have reached the final threshold, calculate the probability of the rare event based on the proportion of successful paths. By focusing computational effort on the most relevant parts of the sample space, the AMS method significantly reduces the number of required simulations. The multilevel approach and adaptive nature of the method result in more

accurate estimates of rare event probabilities, even in complex systems. The AMS method can be applied to a wide range of problems across different fields, from financial risk assessment to structural reliability analysis [4].

In finance, rare events such as market crashes can have devastating effects. Traditional risk assessment methods often underestimate the probability of such events due to their low frequency. The AMS method has been applied to model the probability of extreme losses in financial portfolios, providing more reliable estimates that help in better risk management and regulatory compliance. Engineering structures, such as bridges and buildings, must be designed to withstand rare but potentially catastrophic events like earthquakes. The AMS method has been used to estimate the probability of structural failure under extreme loads, leading to safer and more cost-effective design practices.

## Discussion

Rare environmental events, such as extreme weather conditions, pose significant challenges for prediction and mitigation. The AMS method has been employed to estimate the probability of extreme weather events, aiding in the development of more robust disaster preparedness plans and climate models. The method's complexity requires a deep understanding of the underlying system and careful tuning of parameters, which can be time-consuming and resource-intensive. Although more efficient than traditional methods, the AMS method still requires significant computational resources, especially for very high-dimensional problems. The accuracy of the AMS method depends heavily on the initial setup, including the choice of thresholds and the number of initial simulations. Poor choices can lead to suboptimal results [5,6].

## Conclusion

The Adaptive Multilevel Splitting method represents a significant advancement in the field of rare event estimation. The AMS method is an active area of research, with ongoing efforts to address its limitations and expand its applicability. Developing automated techniques for parameter tuning and threshold selection to reduce the method's complexity and make it more user-friendly. Leveraging advances in parallel computing to further reduce the computational demands of the AMS method, making it accessible for larger and more complex problems. Its adaptive, multilevel approach offers a practical and efficient solution for estimating the probability of rare events in complex systems. Despite its challenges, the AMS method's flexibility and accuracy make it a valuable tool across various domains, from finance and engineering to environmental science. Continued research and development are likely to further enhance its capabilities, making it an indispensable asset in the toolkit of risk analysts and decision-makers.

\*Address for Correspondence: Josephine Frankfort, Department of Mathematics Engineering, University of Hamburg, Hamburg, Germany; E-mail: osepinerankfort@gmail.com

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## Conflict of Interest

None.

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