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Numerical simulations versus theoretical predictions for a non-Gaussian noise induced escape problem in application to full counting statistics

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A theoretical approach for characterizing the influence of asymmetry of noise distribution on the escape rate of a multistable system is presented. This was carried out via the estimation of an action, which is defined as an exponential factor in the escape rate, and discussed in the context of full counting statistics paradigm. The approach takes into account all cumulants of the noise distribution and demonstrates an excellent agreement with the results of numerical simulations. An approximation of the third-order cumulant was shown to have limitations on the range of dynamic stochastic system parameters. The applicability of the theoretical approaches developed so far is discussed for an adequate characterization of the escape rate measured in experiments.

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I. INTRODUCTION

Shot noise\textsuperscript{[1]} characterizes transport properties of mesoscopic conductors. Therefore, studying properties of the shot noise is essential for understanding the behavior of the mesoscopic carriers. The properties can be described by the full counting statistics approach\textsuperscript{[2]} which considers the third- and higher-order cumulants, also known as irreducible correlators. At the same time, first and second cumulants specify properties of equilibrium symmetrical Johnson-Nyquist noise which is different from the shot noise. Typically\textsuperscript{[1]}, the statistics of the shot noise are characterized by a nonsymmetrical distribution, e.g., binomial or Poisson. A scheme for a qualitative characterization of the distribution asymmetry via measurements of the escape rate of an auxiliary multistable system driven by fluctuations has been recently suggested\textsuperscript{[3,4]}. The main idea was to study the escape rate of a noise detector for characterization of acting fluctuations which are the output of a mesoscopic system. This scheme was extensively discussed in a number of theoretical papers\textsuperscript{[5–7]} and implemented experimentally with the Josephson junction as a noise detector driven by Poisson noise\textsuperscript{[8,9]}. A combination of theoretical, numerical, and experimental investigations was presented\textsuperscript{[10]} showing some correspondences and disagreements between the theory, numerics, and experiment. However, despite the progress made in the theoretical description of the short noise and, in turn, properties of the noise-detector scheme, certain questions still remain. For example, accuracy of the suggested theoretical approaches was not validated by numerical simulations. There are also certain disagreements between several published theoretical approaches\textsuperscript{[7]}, thus resulting in controversy\textsuperscript{[11]}. Notably, the reports indicate poor correspondence between the theoretical and experimental results\textsuperscript{[7,10]}, and several outstanding issues have been identified\textsuperscript{[7]}. In particular, the literature cited two important aspects: (i) validity of the use of a third-order-cumulant approximation in theoretical approaches and (ii) omission of the prefactor in the expression of the escape rate\textsuperscript{[7]}. There is, however, reported research\textsuperscript{[7]} which places the scheme into a strong nonlinear regime for maximizing distribution asymmetry.

In the majority of published papers, a similar theoretical model of the noise-detector scheme has been used that corresponds to a nonlinear oscillator driven by a mixture of white Gaussian and Poisson noises. The presence of the Poissonian process leads to asymmetry of the noise distribution. The characterization of the degree of this asymmetry has become the focus of such theoretical considerations. To make the model analytically tractable, an approximation of the third-order cumulant is often applied together with a high-barrier assumption\textsuperscript{[5–7]}. If the latter corresponds to a typical experimental set\textsuperscript{[8,9]}, the former places limitations on the theoretical predictions, and the degree of the limitations is still not well understood. Several theoretical approaches\textsuperscript{[4,11,12]} assume a high-friction limit resulting in overdamped model dynamics, whereas weak damping (underdamped) dynamics is experimentally observed\textsuperscript{[8,9]}.

A revision of the existing approaches explained earlier is necessary in order to shed light on the non-Gaussian noise induced escape problem. The aim of this research is therefore to present a generic approach beyond the third-cumulant limit, and to compare the theoretical predictions with the results of numerical simulations. The revision in this work follows the approaches adopted in previous studies\textsuperscript{[6,7]} demonstrating the presence of constraints on the noise characteristics when theoretical predictions are compared with numerical and/or experimental results. We also show that the absence of a prefactor in the estimation of the escape rate may result in a large error in theoretical predictions.

Theoretical considerations\textsuperscript{[5–7,10,12]}, as well as this work, are motivated by the use of the escape rate for characterization of shot noise in mesoscopic conductors. However, the general theoretical setting is applicable for a wider range of problems including vibrations in civil structures\textsuperscript{[13]}, switching in microelectromechanical system (MEMS) and nanoelectromechanical system (NEMS)\textsuperscript{[14,15]}, neuronal dynamics\textsuperscript{[16]}, and ion channel permittivity\textsuperscript{[17,18]}. It is important to mention that noise induced escape and the corresponding mean first-passage time problem in the presence of non-Gaussian noise (also known as white shot noise) have not been discussed as comprehensively as white or colored...
Gaussian noise [19]. Nonetheless, the literature is extensive and nearly all publications have dealt with one-dimensional potential systems [20–24] for the overdamped case. In some cases, additional limitations on noise characteristics [20,21,23] allowed the problem to be solved analytically and made the task analytically intractable for other cases [21].

The approach adapted in this work is based on the practical realizations of a noise detector used for full counting statistics [8,9]. This approach provides an experimental basis for studying the escape problem in the presence of non-Gaussian noise. The dynamic behavior of the detector is underdamped and the acting noise is white (uncorrelated) with a finite second cumulant; there are no explicit limitations on the shape of the noise distribution. This experimental setting is quite broad, involving wide-ranging research applications as mentioned above, and thus extends beyond the detection of noise statistics.

The dynamic system, noise properties, as well as the theoretical approach for the estimation of the escape rate for non-Gaussian noise are presented in Sec. II. We compare theoretical and numerical results in Sec. III and discuss applicability of the third-cumulant limit in Sec. IV. The main conclusions are summarized in Sec. V.

II. DYNAMIC SYSTEM AND THEORETICAL APPROACH

A simplified experimental scheme, presented in Fig. 1(a), shows a noise source and a threshold detector, both biased by current and based on the Josephson junction [8–10]. For this system, the voltage drop in the noise source is larger than thermal fluctuations. Tunneling events are a dominant source of carriers, implying that the noise source is in a pure shot-noise regime. An additional current bias [middle of Fig. 1(a)] is used to remove a constant component ∫ of the shot noise ζ(t) indicating that zero-mean shot noise η(t) is acting on the detector. In contrast, the detector is in the thermal regime as the voltage drop of the detector is smaller than thermal fluctuations. It is noted that the experiments [8–10] are carried out in a low-temperature environment with a minimal temperature of around 20 mK. For such low-temperature values, the quantum effects in the detector should be strong as predicted by the theory (for example, Ref. [25]). However, previous experimental results [8–10,26,27] state that the dynamics of the detector can also be described within the classical limit. Another potentially important factor refers to the presence of the feedback effect of the detector to the noise source [28,29]. This factor can be neglected in our consideration due to the feedback being considered small in the reported experiments [7–10].

The Josephson junction as a noise source produces shot noise with Poissonian statistics [1,30–32]. Poisson noise is known as a rare event process with an asymmetric probability distribution [33]. The presence of the asymmetry means that depending on the sign of the currents I [left and middle current sources in Fig. 1(a)], the right and left tails of the probability distribution are of different widths [Fig. 1(b)]. This asymmetry is reflected in the difference (asymmetry) of escape rates in the detector that has a multistable potential [Fig. 1(c)]. Thus, the difference between the escape rates characterizes the asymmetry of the noise distribution and, consequently, the degree of non-Gaussianity. In order to extract quantitative information on non-Gaussianity from the experimental measurements, a theoretical description (in the form of a mathematical expression in the simplest case) is required that conjoins the escape rate with the asymmetry parameter(s). So far, the third cumulant, as mentioned above, has been considered as the main asymmetry parameter, although all cumulants of higher than the second order contribute [34,35]. Note that the values of the first and second cumulants as well as detector parameters can be defined by using well-established measurement techniques [10,27], and therefore these values
are considered as “known”. The validity of the theoretical prediction is crucial for the entire experimental approach. The dynamic system under consideration, which describes the experimental setup [7,9,10], is as follows:

\[ \dot{x} + \alpha \dot{x} + \frac{dU}{dx} = \sqrt{\alpha D} \xi(t) \pm \eta(t) , \]  

(1)

\[ \eta(t) = \xi(t) - I . \]  

(2)

Equation (1) is in a dimensionless form and all the parameters are normalized. An equivalent non-normalized equation with corresponding values of parameters can be found in published papers [7,10]. Normalized coordinate \( x \) corresponds to the current of the Josephson junction with biased potential \( U(x) = \sin(x) - sx \) and damping coefficient \( \alpha \). The bias \( s \) is selected in such a way that a tilted multistable (washboard) potential is formed in the system, and a noise induced escape from one of the stable states via the lowest potential barrier [Fig. 1(c)] is analyzed. In the case of a tilted form, the system (1), after escape, evolves along the potential (so-called running mode) and this motion can be easily detected in experiments [9,10]. Noise \( \xi(t) \) corresponds to the thermal noise in the detector. It is Gaussian white noise with unit variance and zero-mean value: \( \langle \xi(t) \rangle = 0, \langle \xi(t) \xi(0) \rangle = \delta(t); \) \( D \) defines noise intensity. Term \( \eta(t) \) in (1) describes a shot noise of the source and consists of two components (2): \( I \) is bias current applied to the source and \( \xi(t) \) corresponds to the Poisson white noise which can be represented as a sum of independent pulses [36]

\[ \zeta(t) = \sum_{i=1}^{N} z_i g(t - t_i). \]  

(3)

In expression (3), \( z_i \) are independent random amplitudes of pulses, function \( g(t - t_i) \) describes the pulse shape, \( t_i \) are independent random times of pulse appearance. Time intervals \( t_i = t_{i+1} - t_i \) between two subsequent pulses have exponential distribution

\[ p(t_i) = \Lambda \exp(-\Lambda t_i), \]  

(4)

where \( \Lambda \) is a parameter of the Poisson noise and defines the frequency of the events. The number of pulses \( n \) within the time interval \( \Delta t \) follows a Poisson distribution

\[ p(n) = \frac{(\Delta t \Lambda)^n \exp(-\Delta t \Lambda)}{n!} . \]  

(5)

Following previous approaches [5–7,10], we consider \( \delta \) impulses of the same amplitude \( \lambda \):

\[ \zeta(t) = \sum_{i=1}^{N} \lambda \delta(t - t_i). \]  

(6)

The Poisson noise (6) is characterized by an infinite number of nonzero cumulants defined as

\[ \chi_s(0, t_1, \ldots, t_N) = \Lambda \lambda^s \delta(t_1) \ldots \delta(t_N) , \]  

(7)

where \( \chi_s \) represents an \( s \)-order cumulant. Since the first cumulant is nonzero, the noise produces a bias \( \lambda \Lambda \) which is removed by term \( I \) in (2) [the middle current source in Fig. 1(a)], therefore, \( I = \lambda \Lambda \). Thus, we consider a zero-mean non-Gaussian noise \( \eta(t) \) acting together with a white noise \( \xi(t) \) on the system (1). All cumulants of \( \eta(t) \) are equal to cumulants \( \chi_s \) except the first cumulant, which is equal to zero.

The task consists of the estimation of the difference between two mean first-passage times \( (T_+ \) and \( T_-) \) corresponding to the opposite signs in front of \( \eta(t) \) in (1). Both \( T_+ \) and \( T_- \) are experimentally measured quantities which lead to an asymmetry factor [7,10]

\[ \Gamma_T = \frac{T_+}{T_-} - 1. \]  

(8)

The value of \( \Gamma_T \) characterizes the asymmetry of the noise distribution and the lower index \( T \) is used to stress that the factor is derived from measurements of times \( T_{\pm} \). Due to escape having an activation character, the times \( T_{\pm} \) can be expressed in the following form [19]:

\[ T_{\pm} = Z_{\pm} \exp \left( \frac{S_{\pm}}{\theta} \right) , \]  

(9)

where \( Z_{\pm} \) and \( S_{\pm} \) are the prefactor and action, respectively, and \( \theta \) is an effective intensity of both Gaussian and non-Gaussian noise in (1). Further, it is implicitly assumed [5–7,10,12] that prefactor \( Z_{\pm} \) can be omitted leading to the asymmetry factor in the form

\[ \Gamma_S = \exp \left( \frac{S_+ - S_-}{\theta} \right) - 1. \]  

(10)

The subscript \( S \) in (10) indicates that we need to measure actions \( S_{\pm} \) rather than times \( T_{\pm} \). Omission of the prefactors \( Z_{\pm} \) is equivalent to the assumption that \( Z_+ = Z_- \). The validity of this assumption was not verified and we discuss this matter in the following. Thus, the actions \( S_{\pm} \) are the subject of theoretical approaches, whereas the times \( T_{\pm} \) are measured experimentally. This indicates that the actions rather than \( T_{\pm} \) must be extracted from experiments or numerical simulations for the proper use of the theory. Note that this aspect has not been addressed in published papers.

Although all previously suggested theoretical approaches [5–7,10,12] are very similar, there is no unified framework to follow. Therefore, we suggest our version which is based on the approaches described in publications [6,7]. This will enable us to compare two approaches: the first approach with only the third cumulant taken into account and the second approach with all cumulants (7) considered. In this way, the importance of higher-order cumulants will be investigated.

The starting point of the theoretical development is the Fokker-Plank equation (FPE) corresponding to the Langevin equation (1) [7] (see papers [36,37] for details of the derivation of the term describing non-Gaussian noise):

\[ \frac{\partial P}{\partial t} = -\frac{\partial}{\partial x}(yP) - \frac{\partial}{\partial y} \left[ \left( -\alpha y - \frac{dU(x)}{dx} \right) P \right] + \frac{1}{2}\alpha D \frac{\partial^2}{\partial y^2} P + \Lambda \left[ \exp \left( \mp \lambda \frac{\partial}{\partial y} \right) - 1 \right] P . \]  

(11)

In Eq. (11), a new variable \( y = \dot{x} \) is introduced and \( P \equiv P(x,y,t) \) is probability density. The exponent in the last term in (11) describes Poisson noise and has the following
Maclaurin series representation [38]:

$$\exp\left(\mp \lambda \frac{\partial}{\partial y}\right) = \sum_{j=0}^{\infty} \frac{(-\lambda y)^j}{j!}.$$  

(12)

Let us consider the solution of Eq. (11) in exponential form [7], that is,

$$P \propto \exp\left(\frac{S}{\theta}\right).$$  

(13)

with action $S$ and the effective noise intensity $\theta$. Note that an exponential form has been used in all recent theoretical approaches [5–7,10,12]. However, particular forms of $\theta$ are varied from one approach to another. Importantly, the proportionality symbol is used in (13) because equality would require an additional prefactor $Z$.

Effective noise intensity $\theta$ is an asymptotic parameter of the problem. Assuming that higher cumulants of a non-Gaussian noise are smaller than the second cumulant, $\theta$ can be chosen [7] in the form $\theta = \alpha D + \lambda^2 \lambda$, i.e., $\theta$ is proportional to the second moment of the sum of the Gaussian and non-Gaussian noises. Note that the second moment is a quantity measured experimentally. The selection of $\theta$ and the form of (13) require an additional verification, which is performed later.

Following the Wentzel-Kramers-Brillouin (WKB) approximation [7,39], we substitute (13) into the FPE (11) and keep the leading-order (1/$\theta$) terms only. The latter formally means the use of the zero-noise limit as $\theta \to 0$. The final result can be written as the following Hamiltonian system of equations:

$$\dot{x} = y,$n

$$\dot{y} = -\alpha y - \frac{\alpha D}{\theta} p_y \pm \lambda \Lambda \left[\exp\left(\mp \lambda p_y \theta / \theta\right) - 1\right],$$

(14)

$$\dot{p}_x = \frac{d^2 U}{dx^2},$$

$$\dot{p}_y = -p_x + a p_y.$$

In (14), $p_x \equiv \frac{\partial U}{\partial x}$ and $p_y \equiv \frac{\partial U}{\partial y}$ are conjugated moments. In contrast to the Gaussian case [39], the asymptotic parameter $\theta$ is not eliminated; all the parameters characterizing noise also are present in the final system of Eqs. (14). The approach, however, can be only applied formally for $S \gg \theta$. In the following, we return to this point.

System (14) has to be completed by two boundary conditions corresponding to a transition from the minimum of the potential $(x = x_{\text{min}}, \ y = 0)$ to its maximum $(x = x_{\text{max}}, \ y = 0)$ [inset in Fig. 1(c)]. These boundary conditions [39] specify a heteroclinic (connecting two saddle states) trajectory of system (14) and these are the following:

$$\text{for } t_i \to -\infty : x = x_{\text{min}}, \ y = 0, \ p_x = 0, \ p_y = 0,$$

$$\text{for } t_f \to \infty : x = x_{\text{max}}, \ y = 0, \ p_x = 0, \ p_y = 0,$$

(15)

where $t_i$ and $t_f$ are the initial and final time moments. If the solution of the boundary problem exists [39], it can be used to calculate the action difference $S_{\text{max}} - S_{\text{min}}$ corresponding to the minimal energy required for the system to migrate from the bottom $x_{\text{min}}$ to the top potential $x_{\text{max}}$. We assume that $S_{\text{min}} = 0$ and therefore $S = S_{\text{max}}$. Denoting coordinates of a heteroclinic trajectory as $$(\hat{x}, \hat{y}, \hat{p}_x, \hat{p}_y),$$ the action is determined by the expression

$$S_{\pm} = \int_{t_i}^{t_f} dt \left\{ -\frac{\alpha D}{2\theta} \hat{p}_y + \theta \Lambda \left[\exp\left(\mp \lambda \hat{p}_y \theta / \theta\right) - 1 \pm \frac{\lambda}{\theta} \hat{p}_y\right] + \frac{\lambda^2}{\theta^2} \hat{p}_y^2 \right\},$$

(16)

where $t_i$ and $t_f$ are initial and final time moments, respectively; signs of $S_{\pm}$ correspond to the signs of noise $\eta(t)$ (2). Action $S_{\pm}$ corresponds to the potential $U(x)$ of the system if it is affected by Gaussian noise only, whereas action is different in the presence of non-Gaussian noise. The WKB approximation and action $S$ are being extensively applied for the analysis of fluctuations in nonequilibrium systems [40–42], whereas action $S$ specified a quasipotential and has the same meaning as potential in an equilibrium case. The latter means that the mean first-passage times $T_{\pm}$ can be presented in the exponential form

$$T_{\pm} \propto \exp\left(\frac{S_{\pm}}{\theta}\right),$$

(17)

where $S_{\pm}$ is defined by (16) with boundary conditions (15).

The theoretical approach presented above takes into account all cumulants of non-Gaussian noise. The third-cumulant approximation can be obtained from Eqs. (14) by expanding the exponential function into a series (12) and truncating all terms above $p_y^3$. The resulting Hamiltonian system is

$$\dot{x} = y,$n

$$\dot{y} = -\alpha y - \frac{\alpha D}{\theta} p_y + \lambda^2 \Lambda \left(-p_y \pm \frac{\lambda}{2\theta} p_y^2\right),$$

(18)

$$\dot{p}_x = p_x \frac{d^2 U}{dx^2},$$

$$\dot{p}_y = -p_x + a p_y,$$

and the corresponding action is defined as

$$S_{3,\pm} = \int_{t_i}^{t_f} dt \left\{ -\frac{\alpha D}{2\theta} \hat{p}_y - \frac{\lambda^2}{2\theta} \hat{p}_y^2 \pm \frac{\lambda^3}{3\theta^2} \hat{p}_y^3\right\},$$

(19)

where index “3” in (19) is used to indicate the third-order-cumulant approximation. Thus, systems (14) and (18) with boundary conditions (15) and corresponding actions (16) and (19) describe effects of the presence of non-Gaussian noise. We solve the boundary value problem using custom software [43], following the approach described in paper [44].

III. NUMERICAL SIMULATIONS VERSUS THEORY

Numerical simulations of the Langevin equation (1) are extremely computationally demanding because the escape time $T_{\pm}$ should be lower than a characteristic relaxation time of the system (1) by a factor $10^6$ for mimicking the experiments [10]. Therefore, for accelerating the simulations, we replace the periodic potential $U(x) = \sin(x) - sx$ in (1) by the third-order polynomial $U(x) = ax - bx^3/3$ with parameters $a$ and $b$ to approximate one well of the periodic potential [see Fig. 1(c)]. Note that the relative difference between the theoretically calculated actions $S_{\pm}$ for these polynomial and periodic potentials is less than 0.01%.
Numerical simulations were performed using a Heun difference scheme, details of which can be found in publication [45]. The Poisson noise term was constant in each integration step and was calculated [10] as \( \lambda \, p n(\Delta t \lambda) \), where \( \lambda \) and \( \Delta t \) are the amplitude and frequency of the Poisson noise, respectively; \( \Delta t \) is the integration step size and the value of \( p n \) is produced by a pseudorandom numbers generator with a Poisson distribution (5). The applied scheme was verified against known theoretical results [35] for linear systems.

First, we checked the scaling (9) alongside the selection of the effective noise intensity as \( \theta = \alpha D + \lambda^2 \Lambda \). As it was mentioned above, our theoretical approach contains a self-contradiction: the asymptotic character of the WKB approximation aims to remove the explicit value of \( \theta \), but \( \theta \) appears explicitly in the final expression. The same contradiction exists in previous theoretical developments [5–7,10,12] too because \( \theta \) includes both Gaussian \( D \) and non-Gaussian \( \lambda^2 \Lambda \) parts and this puts constraints on the way the parameter \( \theta \) can be varied in numerical simulations to evaluate scaling (9). As such, the following factors have to be kept unchanged:

\[
\frac{\alpha D}{\theta} = C_1, \quad \lambda \Lambda = C_2, \quad \frac{\lambda}{\theta} = C_3, \quad \theta \Lambda = C_4,
\]

(20)

where \( C_1, C_2, C_3, \) and \( C_4 \) are constants, whereas the value of \( \theta \) is varied. Importantly, these constraints keep the bias \( I = \lambda \Lambda \) constant. The bias \( I \) characterizes the asymmetry of the Poisson noise and it is the parameter of consideration [7]. The first condition in (20) means that relative contributions of Gaussian and non-Gaussian noises are constant. Note that this is equivalent to keeping the ratio \( \frac{\lambda^2}{\theta} \) constant; this ratio characterizes the relative contribution of Poisson noise. In order to calculate constants using (20), the values of bias \( I = \lambda \Lambda \), ratio \( \alpha D/\theta \) (or ratio \( \frac{\lambda^2}{\theta} \)), and \( \lambda \) should be specified. Knowing the constants, we can vary \( \theta \) in order to change the values of \( D, \lambda, \) and \( \Lambda \) as follows:

\[
\alpha D = C_1 \theta, \quad \lambda \Lambda = C_2, \quad \frac{\lambda}{\theta} = C_3, \quad \theta \Lambda = C_4.
\]

(21)

It is evident that experimental implementation of the procedure described above is a nontrivial task; this is, however, the only means to verify theoretical actions (16) and (19). In the absence of Poisson noise, these actions are equal to the potential barrier \( \Delta U \):

\[
\Delta U = U(x_{\text{max}}) - U(x_{\text{min}}) = \frac{4}{3} \sqrt{\frac{a^3}{b}}
\]

(22)

with \( x_{\text{max}} = \sqrt{a/b} \) and \( x_{\text{min}} = -\sqrt{a/b} \) corresponding to the maximum and minimum of \( U(x) = ax - bx^3/3 \), respectively. For the selected values \( a = 0.009, 3101, b = 0.4965 \), the potential barrier is \( \Delta U = 0.0017 \).

Let us consider an underdamped regime (weak damping) of system (1) by fixing the damping coefficient to \( \alpha = 0.5 \); also, set ratio \( \frac{\lambda^2}{\theta} = 0.3 \) and bias \( I = 0.015 \). These values correspond to experimental conditions typically reported [8–10]. Following the approach described above, we varied \( \theta \) and calculated the mean escape times \( T_\pm \) for the two different signs of the noise \( \eta(t) \). More than 20 000 escape events were used for the estimation of the values of \( T_\pm \). Escape times \( T_\pm \), as functions of \( 1/\theta \), are shown in Fig. 2. Exponential scaling (9) is clearly observed and supports the validity of the selected form (13) and the use of \( \theta = \alpha D + \lambda^2 \Lambda \) as an effective noise intensity. Significantly, the exponential scaling is observed in a wide range of \( 1/\theta \) up to the value of \( \theta \) close to the magnitude of the potential barrier \( \Delta U \), and therefore action \( S \) [see upper axis in Fig. 2(b)]. This result means that despite the asymptotic character \( \theta \ll S \) of the WKB approximation used, it is also applicable for \( \theta \lesssim S \). This relaxes the condition constraining the use of a high barrier in experiments [10]. The range of experimental parameters for which the theoretical description is valid can therefore be significantly extended.
The least-squares fitting (solid lines in Fig. 2) of the numerical results by a linear function
\[
\ln(T_\pm) = S_\pm \frac{1}{\theta} + \ln(Z_\pm)
\] allows us to extract both the values of actions \( S_\pm \) and prefactors \( Z_\pm \ln(9) \) for conducting a comparison between the theory and numerical simulations.

Now let us vary \( \alpha D \) from zero to \( \theta \), that is, between the two extreme cases of pure Gaussian noise and pure Poisson noise. Mimicking experimental conditions, we fix the amplitude \( \lambda = 0.0084 \) and vary frequency \( \Lambda \). This results in varying the ratio \( \frac{\lambda^2 \Delta}{\theta} \) between 0 to 1. Note that the bias \( I \) also changes. A comparison of numerical (markers) and theoretical (solid lines) normalized actions \( S_\pm/\Delta U \) is presented in Fig. 3. The results of the numerical simulations and the theoretical predictions are in close agreement thus proving the applicability of the theoretical approach presented here for the analysis of the non-Gaussian features of the noise. It is seen (Fig. 3) that the difference between \( S_- \) and \( S_+ \) increases with the increase of the relative contribution of Poisson noise which, in turn, corresponds to increased asymmetry of noise distribution. The actual difference between the two values of \( S_- \) and \( S_+ \) provides a qualitative descriptive of the asymmetry.

To reiterate, experimentally a non-Gaussian feature of noise is characterized by the asymmetry factor \( \Gamma_T \) given by (8), whereas the theoretical approach estimates the factor \( \Gamma_S \) via (10). Numerical simulations allow us to estimate both factors \( \Gamma_T \) and \( \Gamma_S \) by calculating the dependencies \( T_\pm(\theta) \) and using the fitting expression (23) to extract the actions \( S_\pm \). We denote numerically obtained factors (8) and (10) by an upper index \( n \), that is, \( \Gamma^T_n \) and \( \Gamma^S_n \), respectively, and compare these factors with the theoretical \( \Gamma_S \) for different values of the ratio \( \lambda^2 \Delta/\theta \). As can be seen in Fig. 4, the theoretical factor \( \Gamma_S \) is close to \( \Gamma^S_0 \). This reflects once again the validity of the use of scaling (9) as well as the applicability of the theory developed. The theoretical factor \( \Gamma_S \) is also close to \( \Gamma^S_0 \) for small values of the ratio \( \lambda^2 \Delta/\theta \), when the asymmetry of the noise distribution is small. However, with this ratio approaching 1, there is a growing difference between \( \Gamma_T \) and \( \Gamma_S \). The presence of such a difference means that neglecting prefactor \( Z \) in (10) can lead to an error when approximation (10) is used instead of factor (8) arising from experimental measurements. The maximum of the ratio \( \lambda^2 \Delta/\theta \) was reported to be around 0.6 from experiments [8–10] and is within the range of negligible difference between \( \Gamma_T \) and \( \Gamma_S \). As a result, the use of \( \Gamma_S \) in a theoretical consideration instead of the experimentally measured \( \Gamma_T \) can not be a cause for a poor correspondence between theory and experiments [7,10]. However, the ratio \( \lambda^2 \Delta/\theta \) can be larger in experiments [26], and in this case the theory will produce an error in the estimation of the asymmetry factor \( \Gamma_T \).

IV. COMPARISON OF THE THIRD-ORDER-CUMULANT AND ALL-CUMULANTS APPROACHES

Previous theoretical developments [5–7,10,12] aimed to derive an analytical expression with the third-order-cumulant approximation. In this section, the importance of keeping all cumulants is considered via a comparison of two actions \( S_{\pm} \) and \( S^3_{\pm} \) calculated according to (16) and (19), respectively, with the numerically obtained action via scaling (9). For maximizing the effects of non-Gaussianity of noise, the pure Poisson noise has been investigated, i.e., when \( D = 0 \). The bias \( I = \lambda \Lambda \) is selected as a varying parameter characterizing the asymmetry of the noise distribution. Non-Gaussian effects are maximized in the \( I \to 0 \) limit; another limit as \( I \to \infty \) corresponds to the Gaussian case. The dependencies of the theoretical and numerical actions as functions of the inverse
current $1/I$ are shown in Fig. 5. Several remarkable features can be seen.

First, the third-cumulant approach provides the solution for the limited range of the inverse current $1/I < 32$ only. Outside this range, the solution of the boundary value problem (15) does not exist for negative $\eta(t)$, whereas this is not the case for the positive sign. For large values of bias $I$ ($I > 0.2$ or for the inverse $1/I < 5$), the difference between the all-cumulants and third-cumulant approaches [Fig. 5(b)] is small, however the difference increases with the decrease of the bias $I$ (greater $1/I$). Note that in the experiments [10], the current is relatively large and, consequently, the third-cumulant approach provides high-accuracy predictions.

Second, for the all-cumulants approach, there is a very good correspondence between theoretical and numerical results for a wide range of $1/I$. For $1/I > 90$ [Fig. 5(a)] the exponential scaling (17) is not observed in numerical dependencies $T(\theta)$ for the negative sign of $\eta(t)$.

Third, the theory predicts a bell shape of the dependence $S_-(1/I)$ [red dashed line in Fig. 5(a)] with a clear maximum. This feature was not confirmed by numerical simulations for the selected parameters, but it was observed for a different set of parameters. Further discussion of this feature is out of the scope of this paper.

Finally, numerical and theoretical asymmetrical factors $\Gamma$ calculated as functions of the inverse bias $1/I$ (Fig. 6) have been compared. There is an excellent correspondence between the numerical factor $\Gamma^\nu_3$ and the theoretical factor $\Gamma_3$ which was obtained by the all-cumulants approach. The third-order-cumulant approximation $\Gamma^\nu_3$ has a limited range of $1/I$ where the theoretical prediction is close to numerical results. The difference between the factors $\Gamma_S$ and $\Gamma_T$ is observed in a wide range of the inverse bias $1/I$ and, moreover, the difference reaches a value of one order of magnitude. The latter demonstrates a significant contribution of the prefactors $Z_\pm$ in the estimation of the asymmetry factor $\Gamma_T$ for the case of strong asymmetry of noise distribution.

V. CONCLUSIONS AND DISCUSSION

We presented a theoretical background for calculating the action (an exponential factor of the mean escape time) for an underdamped oscillator driven by a mixture of white Gaussian and Poisson noises. Note that this approach can be extended to systems of any dimension and to any non-Gaussian noise with finite cumulants. The validity of the theoretical approach suggested here has been confirmed by numerical simulations. We showed that this approach is able to provide a qualitative prediction for actions $S_\pm$ for a wide range of parameters.

Theoretical considerations as presented here and published elsewhere [4–7,10] include the asymptotic parameter $\theta$ explicitly in the final expressions. This places constraints (20) on the range of the parameters in experiments or numerical simulations within which they can be varied in order to be consistent with the theoretical approach. This implies that the dependence of the mean first-passage time $T$ on noise intensity is not exponential (17) for the case where intensity of only one noisy component in a mixture of two components (Gaussian and non-Gaussian) is varied. Note that the exponential scaling of escape rate as a function of intensity of the Poissonian component [Gaussian component (temperature) was constant] was used for a comparison between the theory and experiments [9,10]. According to our results (Fig. 3), varying the intensity of just one component changes the ratio between the components and it changes action $S$, which is represented by an exponential factor in $T$. Since this change of $S$ is relatively
small, the deviation of $T$ from exponential scaling is weak but still present in experiments [9,10]. It is noticeable that there is a difference between experimental and theoretical results [for example, compare dashed and solid lines in Fig. 2(b) in Le Masne et al. [9]]; this difference can be explained by the action being dependent on the intensity ratios between Gaussian and Poissonian components of noise.

It was stated in the Introduction that there are discrepancies between the published theoretical predictions and experiments. As a possible explanation [7] of the discrepancies, the use of a third-order-cumulant approximation and omission of a prefactor in the escape rate have been mentioned. Our comparative study shows that these are not relevant for the range of parameters in experiments. The discrepancy can be explained by the additional approximations made in Eq. (19) necessary for deriving an analytical expression for the action $S$. In contrast, we solved the equation numerically, and showed excellent correspondence between the theoretical approach and numerical simulations, thus confirming the validity of the general theoretical framework.

We showed that the third-order-cumulant approach is applicable for a limited region of parameters, within which the non-Gaussian effects are relatively weak. The all-cumulants approach does not have such limitations and demonstrates an excellent correspondence with the results of numerical simulations. We further demonstrated that both the current and all previous theoretical approaches [5–7,10,12] are not capable of providing a quantitative description of noise with a strong asymmetry since these do not take the prefactor $Z_{\pm}$ into account in the expression for the mean escape rate (9) and subsequently in (8) for the asymmetry factor $\Gamma_T$. These approaches can only provide a qualitative prediction. Furthermore, since the theory is able to accurately predict the actions $S_{\pm}$, experiments need to be designed so as to extract the actions rather than the mean escape time. Such experiments would then lead to new challenges as these will require simultaneous tuning of several parameters in order to satisfy four conditions (20).

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