Effects of the tempered aging and its Fokker-Planck equation

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In the renewal processes, if the waiting time probability density function is a tempered power-law distribution, then the process displays a transition dynamics; and the transition time depends on the parameter $\lambda$ of the exponential cutoff. In this paper, we discuss the aging effects of the renewal process with the tempered power-law waiting time distribution. By using the aging renewal theory, the $p$-th moment of the number of renewal events $n_a(t, t)$ in the interval $(t_a, t_a + t)$ is obtained for both the weakly and strongly aged systems; and the corresponding surviving probabilities are also investigated. We then further analyze the tempered aging continuous time random walk and its Einstein relation, and the mean square displacement is attained. Moreover, the tempered aging diffusion equation is derived.

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I. INTRODUCTION

In 1975, Scher and Motroll [1] used the continuous time random walk (CTRW) to study non-Gaussian anomalous diffusion. Nowadays, the CTRW model becomes popular in describing anomalous diffusion and a lot of chemical, physical, and biological processes [2–4], such as, the transport of electric charge in a complex system, diffusion in a low dimensional chaotic system, and the anomalous diffusion when cooling the mental solid and the twinkling of single quantum dot. In 1996, Montuschi and Bouchaud introduce a CTRW framework for describing the aging phenomena in glasses [5]. This generalized CTRW is called aging continuous time random walk (ACTRW) in [6]. The complex dynamical systems displaying aging behaviour are quite extensive, including the fluorescence of single nanocrystals [7], aging effect in a single-particle trajectory averages [8].

Research on statistics, based on power-law distributions with a heavy tail, yields many of significant results. Often the power-law distribution doesn’t extend indefinitely, due to the finite life span of particles, the boundedness of physical space. For this reason, in 1994, Mantegna and Stanley omit the large steps to study the truncated Lévy flights [9,10]. While the tempered power-law distribution [11] uses a different approach, exponentially tempering the probability of large jumps. Exponential tempering offers technical advantages since the tempered process is still an infinitely divisible Lévy process which makes it convenient to identify the governing equation and compute the transition densities at any scale [12].

By tempering, the distribution changes from heavy tail to semi-heavy, and the existence of conventional moments is ensured, which is useful in some practical applications. Recently tempered power-law distributions [13] have been observed for many geophysical processes at various scales [10,12,14,16], including interplanetary solar-wind velocity and magnetic field fluctuations measured in the alluvial aquifers [17].

In this paper, we discuss the aging effects of the renewal processes with exponentially tempered power-law waiting time probability density function (PDF)

$$\varphi(t) = L_\alpha(t)e^{\lambda t-\lambda t} \sim \frac{1}{-\Gamma(-\alpha)}t^{-(1+\alpha)}e^{-\lambda t},$$  \hspace{1cm} (1)

where $0 < \alpha < 1$, $L_\alpha(t)$ is the one side Lévy distribution [18,19], and $\lambda > 0$ is generally a small parameter. The semi-heavy tails and scale-free waiting time properties of $\varphi(t)$ play a particularly prominent role in diffusion phenomena.

FIG. 1: Random variables (r.v.) generated by Eq. (1) with $\alpha = 0.6$. And the parameter $\lambda$ is chosen, respectively, as (a) $\lambda = 10$, (b) $\lambda = 10^{-1}$, (c) $\lambda = 10^{-3}$, (d) $\lambda = 10^{-5}$ and $k = 1, 2, \ldots, 100$.

From Eq. (1), it can be noted that if $0.1 \ll t \ll 1/\lambda$, $\varphi(t) \sim t^{-(1+\alpha)}$, while if $t \gg 1/\lambda$, $\varphi(t) \sim \exp(-\lambda t)$. For the random variables generated by Eq. (1), Fig. 1 shows that the maximum and range of fluctuations vary dramatically with the change of $\lambda$. The introduced tempering forces the renewal process to converge from non-Gaussian to Gaussian. But the convergence is very slow, requiring a long time to find the trend. So, with the time passed by, both the non-Gaussian and Gaussian processes can be described.
This paper is organized as follows. In Sec. II, we use the aging renewal theory to obtain the p-th moment of the number of renewals \( n_a(t_a, t) \) within the time interval \((t_a, t + t_a)\) of the renewal process starting from time zero. The survival probability is, respectively, discussed in weakly and strongly aging system. We then turn to discuss the tempered ACTRW in Sec. III, and the mean square displacement is obtained for the cases \( t_a \ll t \) and \( t \ll t_a \). The numerical simulations confirm the analytical expressions of the mean square displacement. And the propagator function is also numerically obtained. In Sec. IV, we discuss the Einstein relation of the tempered ACTRW. The diffusion equation for the tempered ACTRW is derived in Sec. V, describing the time evolution of the PDF of the position. Finally, we conclude the paper with some remarks.

II. TEMPERED AGING RENEWAL THEORY

First, we briefly outline the main ingredients in the CTRW and ACTRW. The standard CTRW assumes that the jumping transitions begin at time \( t = 0 \), and observation of dynamics starts at \( t_a \). The ACTRW modifies the statistic of time interval for first jump, namely, the waiting time PDF to the first jump is \( \omega(t_a, t) \). It describes a CTRW process having the aging time interval \((0, t_a)\), while \( t_a \) corresponds to the initial observation time \( t = 0 \). Aging means that the number of renewals in the time interval \((t_a, t_a + t)\) depends on the aging time \( t_a \), even when the former is long. Thus generally ACTRW and CTRW exhibit different behaviors.

More concretely, ACTRW describes the following process: a walker is trapped on the origin for time \( t_1 \), then jumps to \( x_1 \); the walker is further trapped on \( x_1 \) for time \( t_2 \), and then jumps to a new position; this process is then renewed. Thus, ACTRW process is characterized by a set of waiting times \( \{t_1, t_2, \cdots, t_n, \cdots\} \) and displacements \( \{x_1, x_2, \cdots, x_n, \cdots\} \). Here \( \omega(t_a, t) \) is the PDF of the first waiting time \( t_1 \). In ACTRW process, the random walk starts from the time \( t_a \), therefore \( \omega(t_a, t_a) \) may depend on the aging time of the process \( t_a \). The waiting times \( t_n \) with \( n > 1 \) are independent and identically distributed (i.i.d.) with a common probability density \( \varphi(t) \). And the jump lengths \( \{x_1, x_2, \cdots, x_n, \cdots\} \) are i.i.d. random variables, described by the probability density \( f(x) \).

When \( t_a = 0 \), we have \( \omega(t_a, t_1) = \varphi(t_1) \), which is just the well known Montroll-Weiss nonequilibrium process. In order to investigate ACTRW, we should first discuss the aging renewal process. In what follows, we suppose that \( P_{n_a}(t_a, t) \) is the probability of the renewal process \( n_a(t_a, t) \), where \( n_a(t_a, t) = n(t_a + t) - n(t) \) and \( n(t) \) denotes the number of renewals by time \( t \), i.e., \( n_a(t_a, t) \) is the number of renewals in time interval \((t_a, t_a + t)\) for a process starts at the time \( 0 \). Our main work is to discuss the properties of the renewal process \([20, 21]\) in the time interval \((t_a, t_a + t)\).

According to the renewal theory developed by Gordèche and Luck [20],

\[
\omega(s, u) = \frac{1}{1 - \varphi(s)} \frac{\varphi(s) - \varphi(u)}{u - s},
\]

where \( \omega(s, u) \) is the double Laplace transform of the PDF of the first waiting time \( \omega(t_a, t_1) \), and \( \varphi(u) \) is the Laplace transform of \( \varphi(t) \). This paper focuses on taking \( \varphi(t) \) as Eq. (1), and its Laplace transform \((t \rightarrow u)\) has the asymptotic form

\[
\varphi(u) = e^{-(u + \lambda)^\alpha} \sim 1 + \lambda^\alpha - (u + \lambda)^\alpha, \quad 0 < \alpha < 1;
\]

and if \( \alpha = 1 \), \( \varphi(u) \sim 1 - u \).

The double Laplace transform of the PDF of \( n_a(t_a, t) \) reads \([6, 23]\), \( t_a \rightarrow s, t \rightarrow u \),

\[
P_{n_a}(s, u) = \begin{cases} 
\frac{1 - s\omega(s, u)}{su}, & n_a = 0, \\
\omega(s, u)\varphi^{n_a-1}(u)\frac{1-\varphi(u)}{u}, & n_a \geq 1, 
\end{cases}
\]

For the particular case \( \alpha = 1 \), from \([23]\), \( \omega(s, u) \sim 1/s \); and from Eq. (4) we can get \( P_{n_a}(t_a, t) \sim \delta(n_a - t) \), for this case \( P_{n_a}(t_a, t) \) is independent of \( t_a \). Since \( \omega(s, u) \) plays a key role in our discussion, we now derive the analytical formulation of \( \omega(t_a, t) \) \([23]\),

\[
\omega(s, t) = \frac{1}{1 - \varphi(s)} \exp(st) \int_t^{\infty} \exp(sy)\varphi(y)dy \\
\sim \frac{(s + \lambda)^\alpha}{\lambda^\alpha - (s + \lambda)^\alpha} \exp(st) \Gamma(-\alpha, (s + \lambda)t),
\]

where \( \Gamma(\alpha, x) = \int_x^{\infty} \exp(-t) t^{\alpha-1}dt \) is an incomplete Gamma function. Using the Laplace transform of incomplete Gamma function \([24]\), we have,

\[
\omega(t_a, t) = \frac{\exp(-\lambda t)}{\Gamma(-\alpha)} g(t_a) * (\exp(-\lambda t_a)(t_a + t)^{-\alpha-1}),
\]

where \( g(t_a) = t_a^{-\alpha} \exp(-\lambda t_a) E_{\alpha, \alpha}(\lambda t_a^n) \). From the second line of Eq. (4), if \( s \ll \lambda, i.e., t_a \gg 1/\lambda \), Eq. (4) can be given by

\[
\omega(t_a, t) \sim \lambda \exp(-\lambda t \sin(\pi \alpha)) \frac{1}{\alpha \pi t^n} \int_0^{t_a} \exp(-\lambda t) \frac{1}{\pi + t} t^n dt.
\]

If \( s \gg \lambda, i.e., t_a \ll 1/\lambda \), then there exists \( \frac{(s + \lambda)^\alpha}{\lambda^\alpha} \sim 1 \). Eq. (5) can be further simplified as

\[
\omega(t_a, t) \sim \frac{\sin(\pi \alpha)}{\pi} \frac{1}{t + t_a} \left( \frac{t_a}{t} \right)^\alpha,
\]

under the further assumption \( t \ll 1/\lambda, i.e., \lambda t \ll 1 \), there exists

\[
\omega(t_a, t) \sim \frac{\sin(\pi \alpha)}{\pi} \frac{1}{t + t_a} \left( \frac{t_a}{t} \right)^\alpha,
\]

being the same as the one given in \([6, 23]\) for the power law waiting time, i.e., \( \lambda = 0 \). When \( \lambda \) is sufficiently large,
Consider the survival probability $P_0(t_a, t)$, which gives the probability of making no jumps during the interval $t_a$ up to $t_a + t$,

$$P_0(s, u) \sim \frac{1}{su} \frac{(s + \lambda)^\alpha - (u + \lambda)^\alpha}{u(u - s)[\lambda^\alpha - (s + \lambda)^\alpha]}.$$  \hfill (11)

It is instructive to consider two different limits. If $s \ll u$, i.e., $t_a \gg t$, there exists

$$P_0(s, u) \sim \frac{1}{su} \frac{(u + \lambda)^\alpha - \lambda^\alpha}{u^2[(s + \lambda)^\alpha - \lambda^\alpha]}.$$  \hfill (12)

For $\lambda \ll u$, i.e., $t_0 \ll t \ll 1/\lambda$, then $\lambda^\alpha t \sim (1 + \lambda/u)^\alpha \sim u^\alpha$. Performing double inverse Laplace transform on the above equation results in

$$P_0(t_a, t) \sim 1 - t_a^{\alpha - 1} \exp(-\lambda t_a) \mathcal{E}_{\alpha,\alpha} \left(\lambda^\alpha t_a^{1-\alpha}\right) \frac{-\lambda^\alpha t}{\Gamma(2-\alpha)}.$$  \hfill (13)

For $t_a \ll 1/\lambda$, Eq. (13) can be simplified as

$$P_0(t_a, t) \sim 1 - \frac{t_a^{1-\alpha}}{\Gamma(2-\alpha)}.$$  \hfill (14)

It can be noted that $t_a^{1-\alpha}/\Gamma(2-\alpha)$ is larger than $\lambda^\alpha t$ in the parenthesis of Eq. (13), since $\lambda^\alpha t = (\lambda t)^{1-\alpha} \ll t_a^{1-\alpha}$. For $t_a \gg 1/\lambda$, from Eq. (13) and Eq. (15), we obtain

$$P_0(t_a, t) \sim 1 - \frac{t_a^{1-\alpha}}{\Gamma(2-\alpha)}.$$  \hfill (15)

being confirmed by FIG. 3, i.e., the lines tend to be close for big $t_a$.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{fig3.png}
\caption{Probability of particles making jumps during the time interval $[t_a, t_a + t]$ for $t_a \gg 1/\lambda$. The parameters are taken as $\alpha = 0.6, t_a = 10^4$, and $t = 1000$; and the symboled lines are obtained by averaging 5000 trajectories with different $\lambda$. The solid lines from down to up corresponding to the increased $\lambda$ are the theoretical results of Eq. (6). When $\lambda$ is large, the probability reaches 1 quickly.}
\end{figure}

Eq. (11) plays a dominant role. In the following, we analyze the asymptotic form of $\omega(t_a, t)$ with $\alpha \in (0, 1)$. From Eq. (2) and Eq. (3), there exists

$$\omega(s, u) \sim \frac{(s + \lambda)^\alpha - (u + \lambda)^\alpha}{(u - s)[\lambda^\alpha - (s + \lambda)^\alpha]}.$$  \hfill (10)

For Eq. (11), taking the limit $s \to \infty$, we have $\omega(s, u) \sim s^{-1}$; then $\omega(0, t) \sim \delta(t)$. Using the relation $P_0(t_a, t) = 1 - \int_0^t \omega(t_a, t) dt$ leads to $P_0(0, t) = 0$, i.e., all the particles move at time $t = 0$, no aging phenomenon.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{fig4.png}
\caption{Time evolution of $P_0(t_a, t)$ with different $t_a$. The parameters are taken as $\alpha = 0.6, \lambda = 10^{-4}$, and $t = 1000$. The lines are obtained by averaging 5000 trajectories. The (red) dashed line, $1 - 0.05 t^{0.4}$, is the fitting result for small $t$ and big $t_a$, which agrees with Eq. (15).}
\end{figure}

And for $t \ll t_a \ll 1/\lambda$, we have $P_0(t_a, t) \sim 1$, i.e., for small $\lambda, \varphi(t) \sim t^{-1-\alpha}$, the waiting time is generally long;
in a small observation time \( t \), we cannot find movement of the particles. Eq. (11) can also be rewritten as
\[
P_0(s, u) \sim \frac{1}{su} + \frac{1}{u(u - s)} + \frac{(u + \lambda)^\alpha - \lambda^\alpha}{u(s - u)(s + \lambda)^\alpha - \lambda^\alpha}.
\] (16)

For the case \( u \ll s \), i.e., \( t_a \ll t \), Eq. (16) yields
\[
P_0(s, u) \sim \frac{(u + \lambda)^\alpha - \lambda^\alpha}{us[(s + \lambda)^\alpha - \lambda^\alpha]}.
\] (17)

under the further assumption \( u \ll \lambda \), we have
\[
P_0(t_a, t) \sim 1 \ast g(t_a) \frac{t^{-\alpha}}{\Gamma(1 - \alpha)},
\] (18)

where \( g(t_a) \) is defined in Eq. (17), i.e., when \( t_a \ll t \ll 1/\lambda \),
\[
P_0(t_a, t) \sim \frac{\sin(\pi \alpha)}{\pi \alpha} \left( \frac{t}{t_a} \right)^{-\alpha},
\] (19)

being the same as the result given in [20] for the pure power law case (\( \lambda = 0 \)).

![Figure 5: Time evolution of \( P_0(t_a, t) \) with different \( \lambda \). The parameters are taken as \( \alpha = 0.6 \), \( t_a = 500 \), and \( t = 50000 \). The lines are obtained by averaging 5000 trajectories. The dashed line with arrow is for the indicator of slope \(-0.6\), confirming Eq. (19).](image)

When the aging time is sufficiently long relative to the observation time and \( \lambda \) is small, the probability of making no jumps during the time interval \( (t_a, t_a + t) \) approaches to one, i.e., the system is completely trapped. On the contrary, if \( t_a \) is short, while the observation time is long enough, then the particles are unacted on the aging time. So, at least one jump will be made, namely, the possibility of making no jumps is zero. Indeed, from Eq. (19), it can be easily obtained that \( P_0(t_a, t) \sim 0 \) when \( t/t_a \rightarrow \infty \); and \( P_0(t_a, t) \sim 0 \) can also be directly obtained from Eq. (11) under the assumption \( t \gg t_a \).

From Eq. (3), we can write the double Laplace transform of the PDF of \( n_a(t_a, t) \) as
\[
P_{n_a}(s, u) = \frac{\delta(n_a)}{u} \left[ \frac{1}{s - \omega(s, u)} + \omega(s, u) \varphi_{\alpha - 1}(u) \frac{1 - \varphi(u)}{u} \right].
\] (20)

Inserting Eq. (23) into the above equation yields
\[
P_{n_a}(s, u) \sim \omega(s, u) \exp[-(u + \lambda)^\alpha - \lambda^\alpha] \frac{(u + \lambda)^\alpha - \lambda^\alpha}{u} + \frac{\delta(n_a)}{u} \left[ \frac{1}{s - \omega(s, u)} \right].
\] (21)

From now on, we start to calculate the \( p \)-th moment of the aging renewal process \( n_a(t_a, t) \) [22], which reads
\[
\langle n_{n_a}^p(s, u) \rangle = \frac{\Gamma(p + 1)\omega(s, u)}{u[(u + \lambda)^\alpha - \lambda^\alpha]^p}.
\] (22)

For the cases that \( u \gg s \) and \( \lambda \ll u \) or \( u \ll \lambda \), there exist
\[
\langle n_{n_a}^p(s, u) \rangle \sim \left\{ \begin{array}{ll}
\frac{\Gamma(p + 1)}{((s + \lambda)^\alpha - \lambda^\alpha)^{p + \alpha(p - 1)}} & \text{for } s \ll u, \\
\frac{\Gamma(p + 1)}{(\alpha\lambda^{\alpha - 1})^p \lambda^{1+p}} & \text{for } u \ll s.
\end{array} \right.
\] (23)

By double inverse Laplace transform we have
\[
\langle n_{n_a}^p(t_a, t) \rangle \sim \left\{ \begin{array}{ll}
\frac{\Gamma(p + 1)}{1(2+\alpha-p-\alpha)} g(t_a) t^{\alpha p - \alpha + 1} & \text{for } t \ll t_a, \\
(\frac{t}{\tau})^p & \text{for } t_a \ll t.
\end{array} \right.
\] (24)

Taking \( p = 1 \) in [22] leads to
\[
\langle n_a(s, u) \rangle = \frac{(s + \lambda)^\alpha - (u + \lambda)^\alpha}{u(u - s)[(u + \lambda)^\alpha - (s + \lambda)^\alpha]}
\] (25)

which can be rewritten as
\[
\langle n_a(s, u) \rangle = \frac{1}{u(s - u)} \left( \frac{(u + \lambda)^\alpha - \lambda^\alpha}{(s + \lambda)^\alpha - \lambda^\alpha} - \frac{1}{(s + \lambda)^\alpha - \lambda^\alpha} \right).
\] (26)

We will confirm that if both \( t_a \) and \( t \) are large scales, \( \langle n_a(t_a, t) \rangle \sim t/\langle \tau \rangle \), which is an important result for normal diffusion. For small \( u \) and \( s \), using the Taylor expansion \( (u + \lambda)^\alpha \sim \lambda^\alpha + \alpha\lambda^{\alpha - 1}u \) and \( (s + \lambda)^\alpha \sim \lambda^\alpha + \alpha\lambda^{\alpha - 1}s \), from Eq. (25), we have
\[
\langle n_a(s, u) \rangle \sim \frac{1}{\alpha\lambda^{\alpha - 1}u^2s}.
\] (27)

Performing double Laplace transform on the above equation yields
\[
\langle n_a(t_a, t) \rangle \sim \frac{t}{\alpha\lambda^{\alpha - 1}} = \frac{t}{\langle \tau \rangle},
\] (28)

where \( \langle \tau \rangle = \alpha\lambda^{\alpha - 1} \) (see Appendix [C]).
For the slightly aging system, $t \gg t_a$, i.e., $u \ll s$; performing the double inverse Laplace transform on both sides of Eq. (29) yields

$$
\langle n_a(t_a, t) \rangle \sim (t_a^{\alpha-1} e^{-\lambda t_a} E_{\alpha, \alpha} (\lambda^\alpha t_a^\alpha)) * 1
- (t_a^{\alpha-1} e^{-\lambda t_a} E_{\alpha, \alpha} (\lambda^\alpha t_a^\alpha)) * 1
\sim (t_a^{\alpha-1} e^{-\lambda t_a} E_{\alpha, \alpha} (\lambda^\alpha t_a^\alpha)) * 1.
$$

(29)

For the special case, $\lambda = 0$, it can be noted that $\langle n_a(t_a, t) \rangle \sim t^\alpha$. For $t \gg 1/\lambda$, using the asymptotic expansion of Mittag-Leffler function $E_{\alpha, \alpha}$, from Eq. (29), we again obtain $\langle n_a(t_a, t) \rangle \sim t^\alpha$. In the long time scale, the process converges to the Gaussian process, and then the first moment of the number of renewal events grows linearly with the observation time $t$. For $t \ll 1/\lambda$, from Eq. (29) we have

$$
\langle n_a(t_a, t) \rangle \sim \frac{1}{\Gamma(1 + \alpha)} t^\alpha.
$$

(30)

It can be seen that when $t \gg t_a$ the first moment of $n_a$ is not relevant to the aging time $t_a$. From Eq. (29) and Eq. (30), we can see that $\lambda$ plays an important role in our discussion as expected.

![Figure 6](image6.png)

**FIG. 6:** Time evolution of the ensemble average of the renewal times $\langle n_a(t_a, t) \rangle$ with the waiting time PDF Eq. (1) for slight aging. The parameters are taken as $\alpha = 0.6$, $t_a = 3$, and $t = 1000$. The real lines are for the analytical result Eq. (29) and the other lines are obtained by averaging 5000 trajectories.

While for the strongly aging system, $t_a \gg t$, i.e., $s \ll u$; there exists

$$
\langle n_a(s, u) \rangle \sim \frac{1}{u^2 [(s + \lambda)^\alpha - \lambda^\alpha]},
$$

(31)

which yields

$$
\langle n_a(t_a, t) \rangle \sim t t_a^{\alpha-1} e^{-\lambda t_a} E_{\alpha, \alpha} (\lambda^\alpha t_a^\alpha).
$$

(32)

Following the way used above, for $t_a \gg 1/\lambda$, the term $t_a^{\alpha-1} E_{\alpha, \alpha} (\lambda^\alpha t_a^\alpha) e^{-\lambda t_a}$ tends to $\lambda^{1-\alpha}/\alpha$. Then we have

$$
\langle n_a(t_a, t) \rangle \sim \frac{t}{\tau}.
$$

(33)

For $t_a \ll 1/\lambda$, there exists

$$
\langle n_a(t_a, t) \rangle \sim \frac{1}{\Gamma(\alpha)} t t_a^{\alpha-1}.
$$

(34)

The above results for the first moment of $n_a(t_a, t)$ can be summarized as: 1. if $t_a$ or $t$ is greater than $1/\lambda$, then $\langle n_a(t_a, t) \rangle \sim t/\tau$; 2. for $t \gg t_a$ and $t \ll 1/\lambda$ (i.e. $t^{-\alpha-1} \exp(-\lambda t) \sim t^{-\alpha-1}$), $\langle n_a(t_a, t) \rangle \sim t^\alpha/\alpha$; 3. for $t \ll t_a$ and $t \ll 1/\lambda$, $\langle n_a(t_a, t) \rangle$ behaves as $t t_a^{\alpha-1}$.

![Figure 7](image7.png)

**FIG. 7:** The relation between the $\langle n_a(t_a, t) \rangle$ and $\lambda$ for $t_a \gg t$ according to the trajectories and the theory. The number of particles is $3000$, $t_a = 1000$, $t = 5$, $\alpha = 0.6$, and $\lambda = 10^{-2}$, $10^{-3}$, $10^{-4}$, and $10^{-5}$. The real lines are for the analytical result Eq. (33) and the other symbol lines are obtained by averaging 5000 trajectories.

### III. TEMPERED ACTRW

#### A. Mean squared displacement

After understanding the statistics of the number of renewals, we go further to discuss the tempered ACTRW with the waiting time distribution Eq. (1). The process of ACTRW has been described in Sec. I. This paper focuses on the symmetric random walk, i.e., the distribution of jump lengths $f(x) = f(-x)$; and $M_2 = \int_{-\infty}^{\infty} x^2 f(x) dx$ is finite. For such a random walk, we denote $P(x, t_a, t)$ as the PDF of particles’ position in the decoupled tempered ACTRW with aging time $t_a$. Then

$$
P(x, t_a, t) = \sum_{n_a=0}^{\infty} P_{n_a}(t_a, t) f_{n_a}(x),
$$

(35)
FIG. 8: Time evolution of the ensemble average of the renewal times \( \langle n_a(t_a, t) \rangle \) for strong aging \( (t_a \gg t) \). The parameters \( \alpha = 0.6, t_a = 1000, \) and \( \lambda = 10^{-3} \) for (a); \( \lambda = 10^{-2} \) for (b); \( \lambda = 10^{-3} \) for (c); \( \lambda = 10^{-4} \) for (d). The lines are obtained by averaging \( 10^4 \) trajectories. It can be seen that \( \langle n_a(t_a, t) \rangle \) grows linearly with time \( t \) for fixed \( t_a \), for all values of \( \lambda \), which confirms the analytical result Eq. (32).

where \( P_{n_a}(t_a, t) \) means the probability of jumping \( n_a \) steps in the time interval \( (t_a, t_a + t) \), and \( f_{n_a}(x) \) the probability of jumping to the position \( x \) after \( n_a \) steps. In the Fourier-Laplace domain,

\[
P(k, s, u) = \sum_{n_a=0}^{\infty} P_{n_a}(s, u) f^{n_a}(k).
\] (36)

Inserting Eq. (11) into Eq. (36) leads to

\[
P(k, s, u) = \frac{1 - s \omega(s, u) + \omega(s, u)(1 - \varphi(u))}{su} \frac{f(k)}{u - 1 - \varphi(u) f(k)}.
\] (37)

By differentiating Eq. (37) two times with respect to \( k \) and setting \( k = 0 \), we derive the second order moment of the random walks, i.e.,

\[
\langle r^2(s, u) \rangle = \frac{\omega(s, u) M_2}{u[1 - \varphi(u)]}.
\] (38)

For the mean square displacement, we present the results of the slightly aging and strongly aging system, i.e.,

\[
\langle r^2(s, u) \rangle \sim \begin{cases} 
M_2 u^\alpha [(u + \lambda)^{\alpha-1} - u^{\alpha}], & \text{for } u \ll s, \\
M_2 u^\alpha [(s + \lambda)^{\alpha-1} - s^{\alpha}], & \text{for } u \gg s.
\end{cases}
\] (39)

Performing double inverse Laplace transform on \( \langle r^2(s, u) \rangle \) yields

\[
\langle r^2(t_a, t) \rangle \sim \begin{cases} 
M_2 g(t) \ast 1, & \text{for } t \gg t_a, \\
M_2 t g(t_a), & \text{for } t_a \gg t,
\end{cases}
\] (40)

where \( g(z) = z^{\alpha-1} \exp(-\lambda z) E_{\alpha,\alpha}(\lambda^\alpha z^\alpha) \).

From FIG. (10), we can see the large fluctuations of \( \langle r^2(t_a, t) \rangle \) even if the number of trajectories is 10000. This is because that most of particles are trapped in the initial position for \( t_a \gg t \ll 1/\lambda \), which is consistent with Eq. (13). This is related to population splitting [25].

It can be noted that when \( t \gg t_a \), the mean squared displacement has no aging effect; while \( t \ll t_a (\lambda t_a \ll 1) \), it is deeply affected by the aging time \( t_a \). The surprising result is that \( \langle r^2(t_a, t) \rangle \) for strongly aging \( \langle n_a(t_a, t) \rangle \), when the second order moment of the jump length is finite; the same things happen for the pure power law to normal diffusion.

FIG. 9: The relation between the \( \langle r^2(t_a, t) \rangle \) and the observation time \( t \) for \( t_a \ll t \) getting from the trajectories of the particles (dashed line) and Eq. (40) (real line). The parameters \( \alpha = 0.6, \lambda = 10^{-3}, t_a = 1, \) and the number of trajectories is 5000. With the increase of \( \lambda \), the characteristic of \( \langle r^2(t_a, t) \rangle \) changes from Power law to normal diffusion.

FIG. 10: The relation between \( \langle r^2(t_a, t) \rangle \) and the observation time \( t \) for \( t_a \gg t \). It can be noted that \( \langle r^2(t_a, t) \rangle \) increases linearly with \( t \) for fixed \( t_a \) and the fluctuations are large because of population splitting. The parameter \( t_a = 1000, \) \( t = 5, \lambda = 10^{-4} \) for (a); \( \lambda = 10^{-2} \) for (b); \( \lambda = 10^{-3} \) for (c); \( \lambda = 10^{-4} \) for (d), \( \alpha = 0.6 \) and the number of the trajectories is 10000.
the characteristics of\( \lambda\).

Contrary to FIG. (12), FIG. (13) displays the behaviors of the\( \alpha\) stable distribution for all kinds of\( \lambda\).

From the numerical results and the theory we can see that the ‘\( \alpha\) stable distribution’ characteristics can be found for small\( \lambda\). Both the distributions for\( \lambda = 10^{-3}\) and\( \lambda = 10^{-4}\) have the sharp peak and the tail of Eq. (43) decays slowly. While for large\( \lambda\), the top of the distribution for Eq. (42) is smooth, being different from the case of small\( \lambda\). Therefore, depending on the choice of\( \lambda\), one can control the behaviors of the propagator.

### IV. STRONG RELATION BETWEEN THE FLUCTUATION AND RESPONSE

In this section, we discuss the aging from a new point of view. Based on the CTRW model, consider such a process: the particles begin to move at time\( t = 0\) and undergo unbiased diffusion in the time interval\( (0, t_a)\); then an external field is switched on the system starting from\( t_a\). If the averaged response of the particles depends on\( t_a\), the process is said to exhibit aging. Generally speaking, giving some disturbance to a system, some characteristics (parameters of thermodynamics) of the system...
will change, being called response \cite{26}. Under the small disturbance of external field, if the change of the parameter of thermodynamics is proportional to the force of external field, then it is called linear response. It seems important to use drift diffusion to consider aging. Using the method given in \cite{27, 28}, we discuss the tempered aging Einstein relation.

Let us consider a simple example of random walk on a one-dimensional lattice; the length of the lattice is $c$, and the particles can only move to its neighboring sites. Waiting times of different steps of the random walk are considered independent and have the same distribution $\psi(t)$. Jumps to the right (left) are performed with the probability $1/2 + h/2$ (1/2 $- h/2$). The total time is $t = t_a + t_b$: $[0, t_a]$ is called aging interval with $h = 0$; and $(t_a, t_a + t_b)$ is called response interval with $0 < h < 1$. Let $x = x_a + x_b$, where $x_a = \sum_{i=1}^{n_a} x_i^a$ is the displacement performed in the aging time interval and $x_b = \sum_{i=1}^{n_b} x_i^b$ is the displacement performed in the response time interval, and $x_i^a, x_i^b$ are the step lengths and $n_a, n_b$ are the number of events happened in the two time intervals, respectively.

We consider the correlation function $\langle x_a^2 \rangle$ which shows the impact between $(x_a)^2$ in the aging interval and $x_b$ in the response interval. And define a parameter $F_R$ to show the relation between fluctuation and response \cite{29}.

$$ F_R = \frac{\langle (x_a)^2 \rangle^2 \langle x_b \rangle}{\langle (x_a)^2 \rangle \langle x_b \rangle} - 1. \tag{44} $$

If $F_R = 0$, it shows that $x_a^2$ and $x_b$ are independent with each other. Using the relation $\langle x_a^2 \rangle = c^2(n_a)$ and $\langle x_b \rangle = hc(n_b)$, then $F_R$ can be shown in another way,

$$ F_R = \frac{\langle n_a n_b \rangle}{\langle n_a \rangle \langle n_b \rangle} - 1. \tag{45} $$

We further introduce $X_{t_a,t_b}(n_a, n_b)$, the probability to occur $n_a$ events in the aging interval and $n_b$ events in the response interval. Following the result given in \cite{29},

$$ X_{t_a,t_b}(n_a, n_b) = \begin{cases} I(t_{n_a} < t_a < t_{n_a+1}) \\
I(t_{n_a+n_b} < t_a + t_b < t_{n_a+n_b+1}) \end{cases}, \tag{46} $$

where $I(t_{n_a} < t_a < t_{n_a+1}) = 1$ if the event inside the parenthesis occurs, and 0 if not. Using double Laplace transform, if $n_b = 0,$

$$ X_{a,u}(n_a, n_b) = \frac{\varphi^{n_a}(s)}{u} \left[ \frac{1 - \varphi(s)}{s} - \frac{\varphi(u) - \varphi(s)}{s - u} \right]; \tag{47} $$

and if $n_b \geq 1$,

$$ X_{a,u}(n_a, n_b) = \frac{\varphi^{n_a(s)}(s)\varphi^{n_b-1}(u)}{u(s - u)}[1 - \varphi(u)][\varphi(u) - \varphi(s)]. $$

Summing $n_a, n_b$ from 0 to $\infty$ leads to

$$ \langle n_a n_b \rangle_{s,u} = \frac{[\varphi(u) - \varphi(s)]\varphi(s)}{u(s - u)(1 - \varphi(s))^2(1 - \varphi(u))}. $$

Using the Laplace transform of $\varphi(t)$ \cite{4}, we have, when $t_a \ll t_b$,

$$ \langle n_a n_b \rangle_{t_a,t_b} \sim t_b [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)] * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]; \tag{48} $$

if $t_a \ll t_b$, there exists

$$ \langle n_a n_b \rangle_{t_a,t_b} \sim (1 * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]) \cdot (1 * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]) - 1 * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)] * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]. \tag{49} $$

Following the results given by \cite{21},

$$ \langle n_a \rangle_{s,u} = \frac{\varphi(s)}{us(1 - \varphi(s))}. $$

there exists

$$ \langle n_a \rangle_{t_a,t_b} = (1 * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]). \tag{50} $$

And $\langle n_b \rangle_{s,u}$ is the same as Eq. \cite{29}. For $t_a \gg t_b$, there exists

$$ F_R(t_a,t_b) \sim \frac{[a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)] * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)]}{(1 * [a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)])t_a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a)}, \tag{51} $$

From Eq. \cite{34}, we know that $F_R(t_a, t_b)$ does depend on $t_b$.

In the following, we further consider the Einstein relation \cite{27, 31} for the tempered aging process. Denoting $\langle x(t_a, t_b) \rangle$ as the first order moment of the displacement under the influence of a force $F$, from Eq. \cite{10} and

\begin{align*}
\langle x(t_a, t_b) \rangle_F &= hc\langle n_b \rangle, \quad \text{we get that for } t_a \gg t_b, \\
\langle x(t_a, t_b) \rangle_F &\sim hc t_b a^{-1}E_{a,a}(\lambda^a t_a) \exp(-\lambda t_a). \tag{52}
\end{align*}

Denoting $\langle x^2(t_a, t_b) \rangle_0$ as the mean squared displacement of the random walk without external force, from Eq. \cite{10}
we obtain that for $t_a \gg t_b$,

$$\langle r^2(t_a, t_b) \rangle_0 \sim M_2 t_b t_a^{-1} E_{\alpha, \alpha}(\lambda t_a^\alpha) \exp(-\lambda t_a)$$  \hspace{1cm} (53)

with $M_2 = 1/2e^2 + 1/2(-e)^2 = e^2$. Under the assumption $h = \frac{t}{2K_\alpha T} \ll 1$, we obtain the following relation,

$$\langle x(t_a, t_b) \rangle_F \sim \frac{F}{2K_\alpha T} \langle r^2(t_a, t_b) \rangle_0.$$  \hspace{1cm} (54)

V. FOKKER-PLANCK EQUATION FOR THE TEMPERED ACTRW

We now derive the Fokker-Planck equation of the tempered ACTRW, which can be used to solve the tempered aging diffusion problems with different types of boundary and initial condition. Omitting the motionless part of Eq. 47 and taking $f(x)$ as Gaussian (i.e., $f(k) \sim 1 - \frac{1}{2}k^2$), we have

$$P(k, s, u) = \frac{\omega(s, u)}{u} \left(1 + \lambda^\alpha - (u + \lambda)^\alpha + \frac{2}{\lambda} \frac{k^2}{1 + \lambda^\alpha - (u + \lambda)^\alpha}\right).$$  \hspace{1cm} (55)

Define the Riemann-Liouville fractional derivative from 0 to $t$ as,

$$\partial_t^\alpha D_t^n y(t) = \frac{1}{\Gamma(n-\alpha)} \int_0^t (t-\tau)^{n-\alpha-1} y(\tau) d\tau,$$  \hspace{1cm} (56)

where $n = [p] + 1$ and $[\ ]$ denotes the integer part of $p$. For $0 < p < 1$, taking the Laplace transform of the Riemann-Liouville fractional derivative results in

$$\mathcal{L}[\partial_t^\alpha D_t^n y(t)] = s^p \mathcal{L}[y(t)] = s^p y(s).$$  \hspace{1cm} (57)

From the property of the inverse Fourier transform, we have

$$\mathcal{F}^{-1}[k^2 y(k)] = \frac{-\partial^2}{\partial x^2} y(x).$$  \hspace{1cm} (58)

From Eq. 57 and Eq. 58, performing once inverse Fourier transform and double inverse Laplace transforms leads to

$$- e^{-\lambda t} \partial_t^\alpha \left(e^{\lambda t} P(x, t_a, t)\right) + \lambda^\alpha P(x, t_a, t)$$

$$= - \frac{1}{2} (1 + \lambda^\alpha) \frac{\partial^2}{\partial x^2} P(x, t_a, t) + e^{-\lambda t} \partial_t^\alpha \left(e^{\lambda t} \frac{\partial^2}{\partial x^2} P(x, t_a, t)\right)$$

$$+ \lambda^\alpha [\omega(t_a, t) * 1] \delta(x) - [1 * e^{-\lambda t} \partial_t^\alpha \left(e^{\lambda t} \omega(t_a, t)\right)] \delta(x),$$  \hspace{1cm} (59)

where the notation $*$ represents the convolution of the functions w.r.t. $t$. Eq. 59 is the Fokker-Planck equation of the Green function $P(x, t_a, t)$ in the case that the waiting time distribution is the tempered power-law $1/\lambda^\alpha$.

For the non tempered case, namely, $\lambda = 0$ and $\varphi(t) \sim t^{-1-\alpha}$, taking $\lambda = 0$ in Eq. 55 results in

$$P(k, s, u) \sim \frac{\omega(s, u)}{u} \frac{u^{-1}}{u^\alpha + \frac{k^2}{2} (1 - u^\alpha)}.$$  \hspace{1cm} (60)

Performing inverse Fourier transform and double inverse Laplace transform on Eq. 57 yields the corresponding aging diffusion equation

$$\partial_t^\alpha \partial_t^n P(x, t_a, t) = \frac{1}{2} \frac{\partial^2}{\partial x^2} P(x, t_a, t) - \partial_t^\alpha \frac{\partial^2}{\partial x^2} P(x, t_a, t)$$

$$- \omega(t_a, t) \frac{t^{-\alpha}}{\Gamma(1 - \alpha)} \delta(x).$$  \hspace{1cm} (61)

As expected, taking $\lambda = 0$ in Eq. 59 also arrives at Eq. 61. There are also other forms of Eq. 59, e.g., adding the motionless part of 47 to the equation; for the longer time scale $t \gg 1$, then $u^\alpha k^2$ can be reasonably omitted $32$, i.e., the term $\partial_t^\alpha \frac{\partial^2}{\partial x^2} P(x, t_a, t)$ can be omitted in Eq. 61.

VI. CONCLUSIONS

Because of the boundedness of physical space and the finiteness of the lifetime of particles, sometimes it is a more physical choice to use tempered power-law jump length or waiting time distribution instead of the pure power-law distribution. This paper discusses the renewal process and ACTRW with the tempered power-law waiting time distribution $\varphi(t) \sim e^{-\lambda t^{-1-\alpha}}$. Since the tempered power-law distribution lies between the pure power-law and exponential distributions, as expected, the transition dynamics is found with the time evolution and the turning point depends on $\lambda$. By using the aging renewal theory, the $p$-th moments of the renewal times $n_a(t_a, t)$ are analytically obtained and numerically con-
firmed by simulating the particles’ trajectories. In particular, the first order moment of \( n_a(t_a, t) \) is more detailedly discussed. Similarly, the mean squared displacement of the tempered ACTRW is analytically got and numerically verified. Based on the \( p \)-th moment of \( n_a(t_a, t) \) and mean squared displacement of the tempered ACTRW, the aging effects are deeply analyzed. Finally, the tempered aging Einstein relation is attained and the corresponding Fokker-Planck equation is derived from the tempered ACTRW.

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Appendix A: Generation of random variables (FIG. 1)

When generating the random variables with the PDF Eq. (1) to plot FIG. 1, the Monte Carlo statistical methods [31] is used. We first rewrite Eq. (1) to plot FIG. 1, the Monte Carlo statistical method was verified. Based on the tempered ACTRW is analytically got and numerically verified. The mean squared displacement of the tempered ACTRW.

**Algorithm:**

1. Generate a r.v. \( x_1 \) with PDF \( f_1(x) = \alpha t^\alpha_0 z^{-\alpha} \) with \( t_0 \) being a small number. Denote the maximum of \( H(x) \) as \( M \).
2. Accept \( x_1 \), if \( M x \leq H(x_1) \); otherwise, reject.
3. Return to Step 1.

Appendix B: Mittag-Leffler function

The two-parameter function of the Mittag-Leffler type plays a very important role in the fractional calculus, being introduced by G.M. Mittag-Leffler and studied by A. Wiman [32]. The two-parameter Mittag-Leffler function is defined by the series expansion

\[
E_{\alpha,\beta}(z) = \sum_{k=0}^{\infty} \frac{z^k}{\Gamma(\alpha k + \beta)},
\]

where \( \alpha > 0 \) and \( \beta > 0 \). Its one-parameter form (\( \beta = 1 \)) is given as

\[
E_{\alpha}(z) = \sum_{k=0}^{\infty} \frac{z^k}{\Gamma(\alpha k + 1)}.
\]

The asymptotic expansions of Mittag-Leffler are important for obtaining the various useful estimates of the long time or short time fractional dynamics. For small \( z \), there exists

\[
E_{\alpha,\beta}(z) \sim \frac{1}{\Gamma(\beta)} + \frac{z}{\Gamma(\alpha + \beta)}. \tag{B3}
\]

in the special case, \( E_{\alpha}(z) \sim 1 + \frac{z}{\Gamma(\alpha + 1)} \). Another important and useful formula is the asymptotic expansion of large scale for the Mittag-Leffler function. For \( 0 < \alpha < 2 \), \( \beta \) is an arbitrary complex number and \( \mu \) is an arbitrary real number such that \( \pi \mu/2 < \mu < \min\{\pi, \pi \alpha\} \), then for an arbitrary integer \( p \geq 1 \), the following expansion holds,

\[
E_{\alpha,\beta}(z) \sim \frac{1}{\alpha} z^{(1-\beta)/\alpha} \exp(z^{1/\alpha}) \tag{B4}
\]

with \( z \to \infty \) and \( \arg(z) \leq \mu \). For \( z \to +\infty \), from Eq. (B4), we have

\[
E_{\alpha,\beta}(z) \sim \frac{1}{\alpha} z^{(1-\beta)/\alpha} \exp(z^{1/\alpha}). \tag{B5}
\]

And if \( z \to -\infty \),

\[
E_{\alpha,\beta}(z) \sim \frac{1}{z^{\beta/\alpha}} z^{-2\beta/\alpha} \exp(-2z^{1/\alpha}); \tag{B6}
\]

when \( \alpha = \beta \), we have \( |\Gamma(0)| \to \infty \). Therefore \( E_{\alpha,\alpha}(z) \sim \frac{1}{z^{\beta/\alpha}} z^{-2\beta/\alpha} \).

In the analysis of this paper, we use the following function several times.

\[
g(z) = z^{\alpha-1} \exp(-\lambda z) E_{\alpha,\alpha}(\lambda z^\alpha). \tag{B7}
\]

When \( z \to +\infty \), from (B5), there exists

\[
g(z) \sim \frac{\lambda^{1-\alpha} z^{\alpha}}{\alpha}; \tag{B8}
\]

see Fig. 13.

While for \( z \to 0 \), from (B3), we have \( g(z) \sim 1/\Gamma(\alpha) z^{\alpha-1} \).

Appendix C: Laplace transform of \( \varphi(t) \)

Here we present the Laplace transform of Eq. (1). From the definition of the survival probability on a site, i.e., the probability that the waiting time on a site exceeds \( t \),

\[
\Psi(t) = \int_t^\infty \varphi(\tau) d\tau = 1 - \int_0^t \varphi(\tau) d\tau. \tag{C1}
\]

Using the definition of incomplete Gamma function, for Eq. (1) we have

\[
\Psi(t) \sim \lambda \alpha \int_0^\infty \exp(-z) z^{-\alpha-1} dz \tag{C2}
\]

\[
= \lambda \alpha \Gamma(-\alpha, \lambda t).
\]
According to the Laplace transform of the incomplete Gamma function $\mathcal{L}[\Gamma(-\alpha, \lambda t)] = \Gamma(-\alpha)[1 - (\frac{\lambda}{\lambda + t})^\alpha]/u$ with $\Re(-\alpha) > -1$ and $\Psi(t)$, we have

$$\psi(u) \sim 1 + \lambda^\alpha - (u + \lambda)^\alpha. \quad (C3)$$

From Eq. (C3), we have two useful asymptotics. For the long time scale, $t \gg 1/\lambda$, (i.e., $u \ll \lambda$) by the Taylor expansion, we have

$$\psi(u) \sim 1 + u^\alpha \left[ \left( \frac{\alpha}{u} \right)^\alpha - \left( 1 + \frac{\lambda}{u} \right)^\alpha \right] \sim 1 - u^\alpha. \quad (C5)$$

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