

Gibbs Distribution and the Repairman Problem

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Abstract: In this paper, we obtain weak convergence results for a family of Gibbs measures depending on the parameter $\theta > 0$ in the following form $dP_\theta(x) = Z_\theta \exp(-H_\theta(x)/\theta)dQ(x)$, where we show that the limit distribution is concentrated in the set of the global minima of the limit Gibbs potential. We also give an explicit calculus for the limit distribution. Here, we use the above as an alternative to Lyapunov's function or to direct methods for stationary probability convergence and apply it to the repairman problem. Finally, we illustrate this method with a numerical example.

Keywords: Gibbs measure; Gibbs potential; Laplace's method; weak convergence; repairman problem

MSC: 60E99; 60F99; 60G99; 60J99

1. Introduction

In this paper, we study the weak convergence of a family of Gibbs probability measures via Laplace's method ([1–4]). This method is interpreted as a weak convergence of probability measures, which was used by Hwang [5]. Yu. Kaniovski and G. Pflug [6] investigated Laplace's method to study the limit stationary distribution of the birth and death process.

In this paper, we suppose that the Gibbs potential H_θ depends on a parameter θ , and Q is a probability measure on the Euclidean space \mathbb{R}^r , such that it dominates P_θ for any θ . To show the tightness of the family

$$\frac{dP_\theta}{dQ}(x) = Z_\theta \exp(-H_\theta(x)/\theta), \quad \theta > 0, \quad (1)$$

we give some additional conditions for the probability measure Q and for the limit Gibbs potential H of H_θ when $\theta \rightarrow 0$. Under these conditions, the limit distribution P of P_θ , as $\theta \rightarrow 0$, is concentrated on the set of the global minima of the limit Gibbs potential, so we give an explicit calculus for the limit probability P . We use these results to prove that the stationary probability of the repairman problem converges to some probability measure as its state space goes to infinity.

The repairman problem was introduced very early in queuing theory, and in the 1960s, important results concerning stochastic approximations, in particular diffusion approximation, were given by Iglehart [7], and later on, more detailed results on averaging and diffusion approximation were given by Korolyuk [8], see also [4,9,10]. The repairman problem is as follows. Consider n identical devices working independently and simultaneously with lifetimes exponentially distributed with parameter $\lambda > 0$. It is also supposed that we have the possibility of repairing $r \leq n$ failed devices at one time, where the service times are independent and exponentially distributed with mean value $1/\mu$. Suppose that we have a stock of m separate devices of the same type assumed as not broken while waiting for replacements. As soon as a device breaks down, we replace it with another identical



Citation: Chetouani, H.; Limnios, N. Gibbs Distribution and the Repairman Problem. *Mathematics* **2023**, *11*, 4120. <https://doi.org/10.3390/math11194120>

Academic Editors: Alexander Bochkov and Gurami Tsitsiashvili

Received: 4 September 2023
Revised: 26 September 2023
Accepted: 28 September 2023
Published: 29 September 2023



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one [7,8,11]. This is a special case of the birth and death process, but with interesting features, see, e.g., [4,8,12,13].

The aim of this paper is to generalize the Hwang theorem [5] and apply it to obtain the weak convergence of the stationary distribution of the repairman problem. The method used here is that proposed by Kaniovski and Pflug in [6], where the stationary distribution is written in Gibbs form, and then we apply the weak convergence of the last stationary distribution.

Hence, we propose an alternative to Lyapunov's function ([14]) or to direct methods ([11]) for stationary probability convergence in a series scheme (i.e., a functional setting). For weak convergence, see, e.g., [1–3,15].

Section 2 presents the problem setting and notation. Section 3 presents the tightness of the family of probability measures $\{P_\theta, \theta > 0\}$, the limit probability concentration on the set of global minima of the limit Gibbs potential, and the main results and an explicit calculus of the limit distribution. Section 4 presents an application, where we study the stationary distribution of the repairman problem with limited service and spare devices [8], which has a Gibbs representation, using the results of Sections 3. It is proven that the limit stationary distribution is concentrated and uniformly distributed on the set of global minima of the limit Gibbs potential. Finally, in Section 5, we present some conclusions and perspectives.

2. Problem Statement and Existence of the Limit Probability

Let Q be a fixed probability measure on $(\mathbb{R}^r, \mathcal{B}_r)$, where $r \in \mathbb{N}^* = \mathbb{N} \setminus \{0\}$, and \mathcal{B}_r is the Borel σ -algebra of the Euclidean space \mathbb{R}^r , with the scalar product denoted by $\langle \cdot, \cdot \rangle$ and the induced Euclidean norm denoted by $\|\cdot\|$. Let $(H_\theta(\cdot))_{\theta > 0}$ be a family of real-valued functions defined on \mathbb{R}^r , and $H(\cdot)$ be a real-valued continuous function defined on \mathbb{R}^r with a finite global minimum denoted by

$$H^* := \min_{y \in \mathbb{R}^r} H(y).$$

Define the set of points where the global minimum of $H(\cdot)$ is reached as

$$N = \{x \mid H(x) = H^*\},$$

and denote the neighborhood of the set N by

$$N^\delta := \{x \mid H(x) - H^* < \delta\}, \quad \text{for all } \delta > 0.$$

In what follows, we suppose that for any $\delta > 0$, we have

$$Q\{x \mid H(x) - H^* < \delta\} > 0, \quad (2)$$

and H is continuous on some neighborhood of its global minimum.

The space of integrable and bounded functions is equipped with the uniform norm

$$\|H_\theta - H\|_\infty := \sup_x |H_\theta(x) - H(x)|.$$

Laplace's method is used here to define a probability measure P on the set of the global minima of the limit Gibbs potential as weak convergence of the family $(P_\theta, \theta > 0)$, defined as

$$\frac{dP_\theta}{dQ}(x) = Z_\theta \exp\left(-\frac{H_\theta(x)}{\theta}\right), \quad \theta > 0,$$

where

$$Z_\theta = \left[\int \exp\left(-\frac{H_\theta(x)}{\theta}\right) dQ(x) \right]^{-1},$$

as $\theta \rightarrow 0$. If P_θ converges weakly to P , then P is the searched probability measure. We investigate this method with some additional conditions to show that the limit probability P exists and is explicitly defined on the set of global minima of the limit Gibbs potential.

Subsequently, when the integration domain is not indicated, it is supposed to be the whole space \mathbb{R}^r .

Let us suppose here that H_θ converges uniformly to H , i.e.,

$$\lim_{\theta \rightarrow 0} \|H_\theta - H\|_\infty = 0. \quad (3)$$

We have the following result, similar to Proposition 1 in [5], but here in a more general setting where the Gibbs potential depends on a parameter.

Proposition 1. *If P_θ is tight, then H has a global finite minimum.*

Proof. We provide the proof by contradiction. We suppose that the minimum of H exists and it is equal to 0 ($H^* = 0$), and that P_θ is not tight, i.e., there exists $\varepsilon_0 > 0$, and $\{P_{\theta_k}, k = 0, 1, 2, \dots\}$, $\theta_k \rightarrow 0$ as $k \rightarrow \infty$, such that for any compact set K in \mathbb{R}^r ,

$$P_{\theta_k}(K^c) > \varepsilon_0, \quad \text{for all } k \geq 0. \quad (4)$$

Since H is continuous on a neighborhood of its global minimum, then there exists $\delta_0 > 0$ such that for all $\delta \in (0, \delta_0)$, $K = [H \leq \delta] := \{x \mid H(x) \leq \delta\}$ is a compact set. Hence,

$$\begin{aligned} P_{\theta_k}(K^c) &= P_{\theta_k}[H > \delta] \\ &= Z_{\theta_k} \int_{[H > \delta]} \exp\left(-\frac{H_{\theta_k}(x)}{\theta_k}\right) dQ(x) \\ &< Z_{\theta_k} \exp\left(-\frac{\delta}{\theta_k}\right) \int_{[H > \delta]} \exp\left(-\frac{H_{\theta_k}(x) - H(x)}{\theta_k}\right) dQ(x). \end{aligned}$$

The uniform convergence (3) implies that for $\delta > 0$, there exists k_0 such that for all $k \geq k_0$

$$-\frac{\delta}{2} < H_{\theta_k}(\cdot) - H(\cdot) < \frac{\delta}{2}.$$

Then, we obtain the following inequality

$$\begin{aligned} P_{\theta_k}[H > \delta] &< Z_{\theta_k} \exp\left(-\frac{\delta}{\theta_k}\right) \exp\left(\frac{\delta}{2\theta_k}\right) \\ &< \exp\left(-\frac{\delta}{2\theta_k}\right) \left[\int_{N^{\frac{\delta}{3}}} \exp(-H_{\theta_k}(x)/\theta_k) dQ(x) \right]^{-1} \\ &< \exp\left(-\frac{\delta}{2\theta_k}\right) \left[\int_{N^{\frac{\delta}{3}}} \exp\left(-\frac{H_{\theta_k}(x) - H(x)}{\theta_k}\right) \exp\left(-\frac{H(x)}{\theta_k}\right) dQ(x) \right]^{-1} \\ &< \exp\left(-\frac{\delta}{2\theta_k}\right) \left[\exp\left(-\frac{\delta}{3\theta_k}\right) \int_{N^{\frac{\delta}{3}}} \exp\left(-\frac{H_{\theta_k}(x) - H(x)}{\theta_k}\right) dQ(x) \right]^{-1}. \end{aligned}$$

Again, assumption (3) implies that, for $\delta > 0$, there exists k_1 such that, for all $k \geq k_1$,

$$-\frac{\delta}{8} < H_{\theta_k}(\cdot) - H(\cdot) < \frac{\delta}{8},$$

Finally, we obtain

$$P_{\theta_k}[H > \delta] < \frac{\exp\left(-\frac{\delta}{24\theta_k}\right)}{Q\left(N^{\frac{\delta}{3}}\right)},$$

By assumption (2), we have

$$Q\left(N^{\frac{\delta}{3}}\right) > 0.$$

Therefore, $P_{\theta_k}(K^c) = P_{\theta_k}[H > \delta]$ converges to 0 as $k \rightarrow +\infty$. Hence, a contradiction arises with hypothesis (4). \square

Remark 1. It is worth noting here that we cannot conclude from the above first inequality that $Z_{\theta_k} \exp\left(-\frac{\delta}{2\theta_k}\right)$ goes to zero, since Z_{θ_k} can go to infinity, as θ_k goes to zero.

In the following, we suppose that the minimum H^* of the limit Gibbs potential H exists and is finite.

Proposition 2. For all $\varepsilon > 0$, the following convergences hold

$$P_{\theta}[|H_{\theta} - H^*| > \varepsilon] \rightarrow 0, \text{ and } P_{\theta}[H - H^* > \varepsilon] \rightarrow 0, \text{ as } \theta \rightarrow 0 \quad (5)$$

exponentially fast in $1/\theta$.

To prove Proposition 2, we need the following lemma:

Lemma 1. For all $\varepsilon > 0$, there exists θ_0 such that for all $\theta < \theta_0$

$$\left\{x \mid H(x) - H^* < \frac{\varepsilon}{2}\right\} \subset \{x \mid |H_{\theta}(x) - H^*| < \varepsilon\}.$$

Proof of Lemma 1. Let $x \in \{x \mid H(x) - H^* < \frac{\varepsilon}{2}\}$, then

$$|H_{\theta}(x) - H^*| \leq |H_{\theta}(x) - H(x)| + H(x) - H^*,$$

By the uniform convergence (3), there exists θ_0 such that for all $\theta \leq \theta_0$, we have

$$\|H_{\theta} - H\|_{\infty} < \frac{\varepsilon}{2},$$

Hence,

$$|H_{\theta}(x) - H^*| < \varepsilon,$$

i.e.,

$$\left\{x \mid H(x) - H^* < \frac{\varepsilon}{2}\right\} \subset \{x \mid |H_{\theta}(x) - H^*| < \varepsilon\}.$$

\square

Proof of Proposition 2. Owing to Lemma 1, there exists θ_0 such that for all $\theta < \theta_0$,

$$\begin{aligned} P_{\theta}[|H_{\theta} - H^*| > \varepsilon] &\leq P_{\theta}[H - H^* > \varepsilon/2] \\ &\leq Z_{\theta} \int_{[H-H^* > \varepsilon/2]} \exp\left(-\frac{H_{\theta}(x)}{\theta}\right) dQ(x) \\ &\leq Z_{\theta} \exp\left(-\frac{H^*}{\theta}\right) \int_{[H-H^* > \varepsilon/2]} \exp\left(-\frac{H_{\theta}(x) - H(x)}{\theta}\right) \\ &\quad \times \exp\left(-\frac{H(x) - H^*}{\theta}\right) dQ(x), \end{aligned}$$

By (3), there exists θ_0 such that for all $\theta < \theta_0$, we have

$$|H_{\theta}(\cdot) - H(\cdot)| < \frac{\varepsilon}{4},$$

Hence,

$$\begin{aligned}
 P_\theta[H - H^* > \varepsilon/2] &\leq Z_\theta \exp\left(-\frac{H^*}{\theta}\right) \exp\left(\frac{-\varepsilon}{2\theta}\right) \exp\left(\frac{\varepsilon}{4\theta}\right) \\
 &\leq \exp\left(\frac{-\varepsilon}{4\theta}\right) \left[\int_{[|H_\theta - H^*| < \frac{\varepsilon}{8}]} \exp\left(-\frac{H_\theta(x) - H^*}{\theta}\right) dQ(x) \right]^{-1} \\
 &\leq \exp\left(\frac{-\varepsilon}{4\theta}\right) \exp\left(\frac{\varepsilon}{8\theta}\right) \\
 &\leq \exp\left(\frac{-\varepsilon}{8\theta}\right).
 \end{aligned}$$

The above inequality concludes the proof. \square

Remark 2. Proposition 2 means that the limit probability P is concentrated on the set of global minima of the limit Gibbs potential $H(\cdot)$.

3. Characterization of the Limit Probability

The uniform convergence in (3) implies that for any $x, y \in \mathbb{R}^r$

$$H_\theta(x) - H_\theta(y) = H(x) - H(y) + \Delta_\theta(x, y),$$

with $\lim_{\theta \rightarrow 0} \Delta_\theta(x, y) = 0$.

In this section, we suppose that

(A1) For all $x, y \in N$, $\lim_{\delta \rightarrow 0} \lim_{\|x-y\| < \delta} \Delta_\theta(x, y) = 0$.

(A2) For all $x, y \in N$, $\frac{\Delta_\theta(x, y)}{\theta}$ is bounded as $\theta \rightarrow 0$.

3.1. Case When $Q(N) > 0$

Theorem 1. If assumptions (A1) and (A2) are verified, then the limit probability P of P_θ as $\theta \rightarrow 0$ is concentrated on the set N and its density with respect to Q is

$$f(x) = \begin{cases} \left[\int_N \exp(-\Delta(x, y)) Q(dy) \right]^{-1}, & \text{if } x \in N, \\ 0, & \text{otherwise,} \end{cases}$$

where $\Delta(x, y) := \lim_{\theta \rightarrow 0} \frac{\Delta_\theta(x, y)}{\theta}$.

Proof. To prove Theorem 1, we must prove that the density

$$f_\theta(x) = Z_\theta \exp\left(-\frac{H_\theta(x)}{\theta}\right),$$

converged to $f(x)$.

For $x \in N$, we have

$$\begin{aligned}
 f_\theta(x) &= \frac{\exp\left(-\frac{H_\theta(x)}{\theta}\right)}{\int \exp\left(-\frac{H_\theta(y)}{\theta}\right) Q(dy)} \\
 &= \frac{1}{\int \exp\left(-\frac{H_\theta(y) - H_\theta(x)}{\theta}\right) Q(dy)}. \tag{6}
 \end{aligned}$$

The denominator in (6) can be written as follows:

$$\begin{aligned} \int \exp\left(-\frac{H_\theta(y) - H_\theta(x)}{\theta}\right) Q(dy) &= \int \exp\left(-\frac{H(y) - H(x) + \Delta_\theta(y, x)}{\theta}\right) Q(dy) \\ &= \int_N \exp\left(-\frac{\Delta_\theta(y, x)}{\theta}\right) Q(dy) \\ &\quad + \int_{N^c} \exp\left(-\frac{H(y) - H(x)}{\theta}\right) \times \\ &\quad \exp\left(-\frac{\Delta_\theta(y, x)}{\theta}\right) Q(dy). \end{aligned}$$

Since $\frac{\Delta_\theta(x, y)}{\theta} \rightarrow \Delta(x, y)$ as $\theta \rightarrow 0$, then, by the dominated convergence theorem, we obtain

$$\int_N \exp\left(-\frac{\Delta_\theta(y, x)}{\theta}\right) Q(dy) \rightarrow \int_N \exp(-\Delta(y, x)) Q(dy),$$

and

$$\int_{N^c} \exp\left(-\frac{H(y) - H(x)}{\theta}\right) \exp\left(-\frac{\Delta_\theta(y, x)}{\theta}\right) Q(dy) \rightarrow 0,$$

as $\theta \rightarrow 0$.

We can prove in the same way that for $x \notin N$, $f(x) = 0$. \square

Corollary 1. If $\Delta(\cdot, \cdot) = 0$, then the probability P is the uniform distribution on the set N , with density $1/Q(N)$.

Example 1. Let $Q(dx) = \frac{dx}{4} \mathbf{1}_{\{|x| \leq 2\}}$ be a probability measure on \mathbb{R} and

$$H_\theta(x) = \begin{cases} x^2 - \theta x, & 1 < |x| < 2, \\ 1 - \theta x, & |x| \leq 1, \\ 0, & \text{otherwise.} \end{cases}$$

We observe that H_θ converges uniformly to the function

$$H(x) = \begin{cases} x^2, & 1 < |x| < 2, \\ 1, & |x| \leq 1, \\ 0, & \text{otherwise.} \end{cases}$$

The global minimum for $H(\cdot)$ is 1, and the set of its global minimum is

$$N = \{x \in \mathbb{R} \mid |x| \leq 1\} = [-1; 1].$$

The Gibbs potential H is continuous on some neighborhood of the set of its global minima N and for all $x, y \in N$

$$\Delta(x, y) = -(x - y),$$

Then,

$$f(x) = \left(\int_{-\infty}^{\infty} e^{x-y} \mathbf{1}_{\{|y| \leq 1\}} \frac{dy}{4} \right)^{-1} \mathbf{1}_{\{|x| \leq 1\}} = \frac{4 e^{-x}}{e - e^{-1}} \mathbf{1}_{\{|x| \leq 1\}}.$$

3.2. Case When $Q(N) = 0$

In the sequel, we denote by $O(\theta)$, $\Delta(\theta, \delta)$ all quantities tending to 0 as $\theta \rightarrow 0$ or $\delta \rightarrow 0$, even if they are not equal, and by $o(\theta)$, all quantities such that $o(\theta)/\theta \rightarrow 0$ as $\theta \rightarrow 0$.

In what follows, we suppose that $N = \{x^1, x^2, \dots, x^n, \dots\}$ and there exists a continuous positive real function $\psi_i(\cdot)$ in the neighborhood of 0, with $\psi_i(0) = 0$, such that in the neighborhood of x^i ,

$$H(x) = H^* + \psi_i(x - x^i).$$

The measure Q is also supposed to be absolutely continuous with respect to the Lebesgue measure, with continuous density f . We also define

$$P_\delta^{i,\theta} = P_\theta(V_i^\delta),$$

where $V_i^\delta = \{x \in \mathbb{R}^r \mid \|x - x^i\| < \delta\}$, and $V^\delta = \{x \in \mathbb{R}^r \mid \|x\| < \delta\}$.

Theorem 2. Let assumptions (A1) and (A2) hold, and for all $i \in \mathbb{N}^*$, there exists $\alpha_i > 0$ such that

$$\lim_{\delta \rightarrow 0} \sup_{\|x\| < \delta} \left| \frac{\psi_i(x)}{\|x\|^2} - \alpha_i \right| = 0,$$

Then, in the case of $f(x^i) \neq 0$, we obtain

$$\lim_{\theta \rightarrow 0} \lim_{\delta \rightarrow 0} \theta \log \left(\frac{P_\delta^{i,\theta} \alpha_i^{1/2}}{Z_\theta \exp(-H^*/\theta) f(x^i) \pi^{r/2} \theta^{1/2}} \right) = 0,$$

and, in the case of $f(x^i) = 0$, we obtain

$$P\{x^i\} = 0.$$

Proof. Let $V_i^\delta, i \in \mathbb{N}^*$ be open neighborhoods of $x^i, i \in \mathbb{N}^*$, with a radius less than $\delta > 0$. We have

$$\begin{aligned} P_\delta^{i,\theta} &= P_\theta(V_i^\delta) \\ &= Z_\theta \int_{V_i^\delta} \exp\left(-\frac{H_\theta(x)}{\theta}\right) f(x) dx \\ &= Z_\theta A_i(\theta, \delta), \end{aligned}$$

where

$$\begin{aligned} A_i(\theta, \delta) &= \int_{V_i^\delta} \exp\left(-\frac{H_\theta(x)}{\theta}\right) f(x) dx \\ &= \exp\left(-\frac{H^*}{\theta}\right) \int_{V_i^\delta} \exp\left(-\frac{H_\theta(x) - H^*}{\theta}\right) f(x) dx \\ &= \exp\left(-\frac{H^*}{\theta}\right) \int_{V_i^\delta} \exp\left(-\frac{H_\theta(x) - H(x)}{\theta}\right) \times \\ &\quad \exp\left(-\frac{H(x) - H^*}{\theta}\right) f(x) dx \\ &= \exp\left(-\frac{H^*}{\theta}\right) \exp\left(-\frac{H_\theta(x^i) - H(x^i)}{\theta}\right) \times \\ &\quad \int_{V_i^\delta} \exp\left(-\frac{\Delta_\theta(x, x^i)}{\theta}\right) \exp\left(-\frac{H(x) - H^*}{\theta}\right) f(x) dx. \end{aligned}$$

According to assumption (A1), the following estimation holds

$$1 - O(\delta) \leq \exp\left(-\frac{\Delta_\theta(x, x^i)}{\theta}\right) \leq 1 + O(\delta).$$

It follows that

$$(1 - O(\delta))B_i(\theta, \delta) \leq A_i(\theta, \delta) \leq (1 + O(\delta))B_i(\theta, \delta),$$

where

$$B_i(\theta, \delta) = \exp\left(\frac{-H^*}{\theta}\right) \exp\left(-\frac{H_\theta(x^i) - H(x^i)}{\theta}\right) \int_{V_i^\delta} \exp\left(-\frac{H(x) - H^*}{\theta}\right) f(x) dx.$$

Define also

$$\begin{aligned} D_i(\theta, \delta) &= \int_{V_i^\delta} \exp\left(-\frac{H(x) - H^*}{\theta}\right) f(x) dx \\ &= \int_{V_i^\delta} \exp\left(-\frac{\psi_i(x - x^i)}{\theta}\right) f(x) dx. \end{aligned}$$

Now, setting $y = x - x^i, i \in \mathbb{N}^*$, we obtain

$$\begin{aligned} D_i(\theta, \delta) &= \int_{V^\delta} \exp\left(-\frac{\psi_i(y)}{\theta}\right) f(y + x^i) dy \\ &= \int_{V^\delta} \exp\left(-\frac{\alpha_i \|y\|^2 + o(\|y\|^2)}{\theta}\right) f(y + x^i) dy \end{aligned}$$

and

$$(1 - o(\delta^2)/\theta)F_i(\theta, \delta) \leq D_i(\theta, \delta) \leq (1 + o(\delta^2)/\theta)F_i(\theta, \delta),$$

where

$$F_i(\theta, \delta) = \int_{V^\delta} \exp\left(-\frac{\alpha_i \|y\|^2}{\theta}\right) f(y + x^i) dy.$$

We have

$$\begin{aligned}
 F_i(\theta, \delta) &= \int_{V^\delta} \exp(-\|y\|^2) f\left(\sqrt{\frac{\theta}{\alpha_i}} y + x^i\right) \sqrt{\frac{\theta}{\alpha_i}} dy \\
 &= \int_{V^{\frac{\delta}{\sqrt{\theta}}}} \exp(-\|y\|^2) \left(f(x^i) + \Delta(\theta, y)\right) \sqrt{\frac{\theta}{\alpha_i}} dy \\
 &= \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \int_{V^{\frac{\delta}{\sqrt{\theta}}}} \exp(-\|y\|^2) dy + \sqrt{\frac{\theta}{\alpha_i}} \int_{V^{\frac{\delta}{\sqrt{\theta}}}} \exp(-\|y\|^2) \Delta(\theta, y) dy \\
 &= \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \int_{V^{\frac{\delta}{\sqrt{\theta}}}} \exp(-\|y\|^2) dy + \Delta(\theta, \delta) \\
 &= \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \int \exp(-\|y\|^2) dy (1 - O(\theta)) + \Delta(\theta, \delta) \\
 &= \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \left(\int_{-\infty}^{+\infty} \exp(-u^2) du\right)^r (1 - O(\theta)) + \Delta(\theta, \delta) \\
 &= \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \pi^{r/2} (1 - O(\theta)) + \Delta(\theta, \delta).
 \end{aligned}$$

If $f(x^i) = 0$, then

$$\lim_{\delta \rightarrow 0} \lim_{\theta \rightarrow 0} P_\delta^{i, \theta} = 0,$$

and if $f(x^i) \neq 0$, then

$$\lim_{\delta \rightarrow 0} \lim_{\theta \rightarrow 0} \theta \log \left(\frac{P_\delta^{i, \theta} \alpha_i^{1/2}}{Z_\theta \exp(-H^*/\theta) f(x^i) \pi^{r/2} \theta^{1/2}} \right) = 0.$$

□

Corollary 2. Under the assumption of Theorem 2, if for all $x^i \in N$, we have

$$H_\theta(x^i) - H^* = a_i \theta + o(\theta), \quad (7)$$

where $a_i \in \mathbb{R}$, and if there exists $x^k \in N$ such that $f(x^k) \neq 0$, then

$$P\{x^i\} = \frac{f(x^i) \alpha_i^{-1/2} \exp(-a_i)}{\sum_{k=1}^{+\infty} f(x^k) \alpha_k^{-1/2} \exp(-a_k)}.$$

Proof. According to the proof of Theorem 2, we obtain

$$P_\theta(V_i^\delta) \asymp \pi^{r/2} Z_\theta \sqrt{\frac{\theta}{\alpha_i}} f(x^i) \exp\left(\frac{-H^*}{\theta}\right) \exp\left(-\frac{H_\theta(x^i) - H(x^i)}{\theta}\right), \quad (8)$$

and by (7),

$$\exp\left(-\frac{H_\theta(x^i) - H(x^i)}{\theta}\right) \asymp \exp(-a_i).$$

Then, we obtain that

$$\pi^{r/2} Z_\theta \sqrt{\theta} \exp\left(\frac{-H^*}{\theta}\right) \sum_{k \geq 1} f(x^k) \alpha_k^{-1/2} \exp(-a_k) \asymp 1,$$

Hence,

$$Z_{\theta}^{-1} \asymp \pi^{r/2} \sqrt{\theta} \exp\left(\frac{-H^*}{\theta}\right) \sum_{k \geq 1} f(x^k) \alpha_k^{-1/2} \exp(-a_k). \quad (9)$$

Finally, (8) and (9) give

$$P\{x^i\} = \frac{f(x^i) \alpha_i^{-1/2} \exp(-a_i)}{\sum_{k=1}^{+\infty} f(x^k) \alpha_k^{-1/2} \exp(-a_k)}.$$

Now, suppose that the function $H \in \mathcal{C}^2(\mathbb{R}^r)$, with Hessian matrix

$$A(x) = \left(\frac{\partial^2 H}{\partial x_i \partial x_j}(x) \right)_{1 \leq i, j \leq r},$$

is invertible for all $x \in N$, then we have the following result. \square

Corollary 3. If $H \in \mathcal{C}^2(\mathbb{R}^r)$ and

$$\sup_{x \in N} |H_{\theta}(x) - H^*| = o(\theta),$$

and there exists $k \in \mathbb{N}$ such that $f(x^k) \neq 0$, then

$$P\{x^i\} = \frac{f(x^i) |\det(A(x^i))|^{-1/2}}{\sum_{k=1}^{+\infty} f(x^k) |\det(A(x^k))|^{-1/2}}.$$

Proof. To prove Corollary 3, it is sufficient to put in Corollary 2,

$$a_i = 0, \text{ and } \alpha_i = |\det(A(x^i))|,$$

for all x^i belonging to N . \square

Remark 3. Equation (9) implies that

$$\sum_{k \geq 1} f(x^k) \alpha_k^{-1/2} < +\infty.$$

4. The Repairman Problem

Let us apply the previous results to obtain the limit distribution of the stationary one in the repairman problem [7,8] in the averaging scheme. This is an alternative method to Lyapunov's function method in the diffusion approximation scheme to prove convergence of stationary probabilities [11].

This system can be described by a Markov birth and death process $x_n(t)$, $t > 0$, representing the number of failed components at time t , with state space [7,8,11]

$$E^n = \{0, 1, \dots, m+n\},$$

and jump intensities given by

$$\lambda_i = \begin{cases} n\lambda, & 0 \leq i \leq m \\ (m+n-i)\lambda, & m \leq i \leq m+n, \end{cases} \quad \mu_i = \begin{cases} i\mu, & 0 \leq i \leq r \\ r\mu, & r \leq i \leq m+n. \end{cases}$$

In what follows, suppose that $m =: m_0 n$ and $r = r_0 n$, where m_0 and r_0 are constants.

Now, consider a Markov process $v_n(t)$ with state space $E_n = \{v_i = i/n \mid 0 \leq i \leq n + m\}$ and intensity functions given by

$$\lambda(v_i) = \begin{cases} \lambda, & 0 \leq v_i \leq m_0 \\ (m_0 + 1 - v_i)\lambda, & m_0 \leq v_i \leq m_0 + 1 \end{cases} \quad \mu(v_i) = \begin{cases} v_i\mu, & 0 \leq v_i \leq r_0 \\ r_0\mu, & r_0 \leq v_i \leq m_0 + 1 \end{cases}$$

where $v_i = i/n$.

Then, we have

$$x_n(t) = nv_n(nt).$$

Consider now the normalized process defined by

$$\xi_n(t) := v_n(nt) = x_n(t)/n,$$

with the velocity of jumps defined by

$$C(v) := \lambda(v) - \mu(v).$$

This function enables us to classify the repairman problem depending on the position of the equilibrium point p defined by

$$C(p) = 0. \quad (10)$$

Under the condition $r_0 = \lambda/\mu < m_0$, the interval $V^* = [r_0, m_0]$ is an equilibrium set of a repairable system. Our main objective is to describe the stationary distribution of this repairable system on the equilibrium set V^* .

4.1. Gibbs Potential for the Stationary Distribution

The stationary distribution ρ^n of the process $x_n(t), t \geq 0$, may be written as follows:

$$\rho^n(k) = \rho_0^n \prod_{i=1}^k [\lambda(v_{i-1})/\mu(v_i)], \quad 1 \leq k \leq m + n, \quad (11)$$

where

$$\rho_0^n = \left[1 + \sum_{k=1}^N \prod_{i=1}^k [\lambda(v_{i-1})/\mu(v_i)] \right]^{-1}. \quad (12)$$

By using the Gibbs potential, the stationary distribution (11) is represented as follows (see [6]):

$$\rho^n(v_k) = \rho_0^n \exp[-NH_n(v_k)], \quad 1 \leq k \leq r,$$

with $N = m + n$ and where H_n is the Gibbs potential determined by the following relation

$$H_n(v_k) = -\frac{a}{n} \sum_{i=1}^k \ln[F(v_i)],$$

where $a = n/N = (m_0 + 1)^{-1}$ and $F(v) := \lambda(v)/\mu(v)$ is the kernel of the Gibbs potential H_n .

Then, the limit Gibbs potential is represented as

$$H(v) = -a \int_0^v \ln[F(u)] du, \quad 0 \leq v \leq 1 + m_0. \quad (13)$$

Therefore, in what follows and in the case when $r_0 = \lambda/\mu$, we will use the Gibbs potential for a stationary distribution with the kernel represented on the interval $(0, 1 + m_0]$ by

$$F(v) = \begin{cases} r_0/v, & 0 < v \leq r_0 \\ 1, & r_0 \leq v \leq m_0 \\ 1 + m_0 - v, & m_0 \leq v \leq 1 + m_0. \end{cases} \quad (14)$$

Alternatively, in explicit form on the left interval $[0, r_0]$

$$H(v) = av[\ln(v/r_0) - 1],$$

and on the right interval $[m_0, 1 + m_0]$

$$H(v) = a(m_0 + 1 - v) \ln(m_0 + 1 - v) - a(m_0 + r_0 - v), \quad m_0 \leq v \leq m_0 + 1.$$

Owing to (10), on the interval, $V^* = [r_0, m_0]$

$$H(v) = H(r_0) = -ar_0 = -r_0/(1 + m_0),$$

It is easy to verify that the set V^* is an equilibrium set for the limit Gibbs potential (13)

$$H^* = \min_{0 \leq v \leq 1 + m_0} H(v) = H(r_0) = -ar_0.$$

4.2. Limit Stationary Distribution

The stationary distribution of a repairable system induces a stationary distributed random variable x_n^* , with

$$\mathbb{P}\{x_n^* = k\} = \rho_k^n, \quad 0 \leq k \leq N.$$

In this example, Q is the probability measure and $Q(dx) = p^{-1} \mathbf{1}_{[r_0, m_0]}(x) dx$ on $[r_0, m_0]$, where $p = m_0 - r_0$.

We can show that

$$|\Delta_n(x, y)| \leq \frac{C}{n^2},$$

which means that

$$\lim_{n \rightarrow +\infty} n\Delta_n(x, y) = 0.$$

Then, the random variable x_n^* converges weakly to some random variable uniformly distributed on the equilibrium set V^* with density $f(x) = (m_0 - r_0)^{-1} \mathbf{1}_{[r_0, m_0]}(x)$.

Finally, we give the following numerical example (see Figure 1), which clearly shows that the limit stationary distribution is uniformly distributed on the equilibrium set V^* .

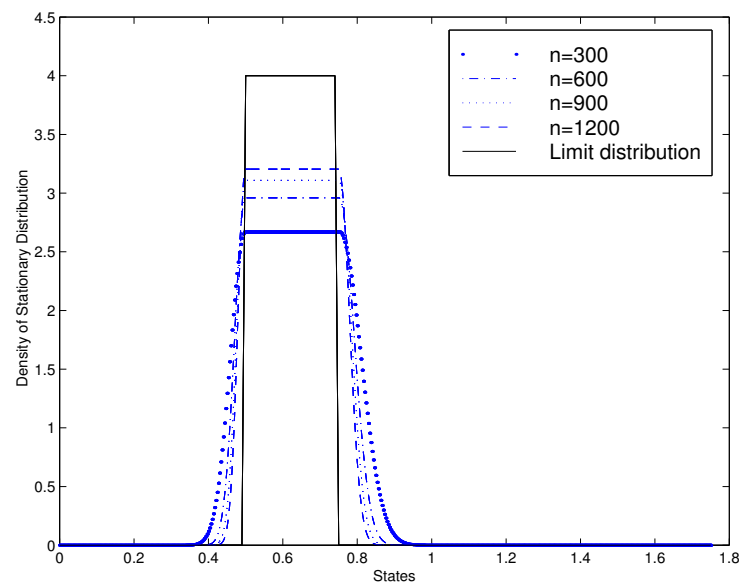


Figure 1. Convergence of the stationary distribution density for the repairman problem (with $\lambda = 0.5$, $\mu = 1$, $r_0 = \lambda/\mu = 0.5$ and $m_0 = 0.75$).

5. Concluding Remarks

We have presented the use of Gibbs distributions as a way to prove the convergence of the stationary distribution of the repairman problem in the averaging series scheme. The key contributions of this work are the weak convergence Gibbs distribution when the potential depends on a parameter θ and the transformation of the repairman stationary probabilities to Gibbs distributions. It will be of interest, on the one hand, to try to establish the same kind of results in a diffusion approximation scheme. Beyond the repairman problem, it would be of interest to apply this method to other similar problems. This problem must be rescaled in time when the finite state space becomes infinite. On the other hand, other work of interest can be performed in the direction of large deviations to provide a rate of convergence to the stationary distribution, see, e.g., [16].

Author Contributions: Methodology, H.C. and N.L. All authors have read and agreed to the published version of the manuscript.

Funding: This research received no external funding.

Data Availability Statement: Not applicable.

Acknowledgments: We are indebted to the three anonymous referees for their useful comments that greatly improved the presentation of this paper.

Conflicts of Interest: The authors declare no conflict of interest.

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