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## DIFFUSION APPROXIMATION FOR A CONTROLLED SERVICE SYSTEM

VĚRA LÁNSKÁ

The aim of the paper is to suggest a procedure how to control a service system under the possibility of the system's ruin, when the additions of the capital are random variables with a given distribution function. Using diffusion approximation for the capital the original problem is converted into the problem of controlling continuous Markov processes. A procedure how to compute the optimal control policy is presented.

### 1. INTRODUCTION

A service system  $M/M/1$  without possibility of queueing is considered. Its input is composed of  $n$  mutually independent Poisson processes with arrival rates  $a_i q > 0$ ,  $i = 1, \dots, n$ . The service time of the  $i$ -th type of customer is a random variable having exponential distribution with a service rate  $b_i q > 0$ ,  $i = 1, \dots, n$ . (The parameter  $q$  is sufficiently large). The behaviour of the system is described by a random process  $\{i_t, t \in [0, T]\}$  with a finite set of states  $\{0, 1, \dots, n\}$ .  $i_t = 0$  means that the system is vacant at time  $t$ ,  $i_t = j, j = 1, \dots, n$ , means that the system serves a customer of  $j$ -th type. Further, we assume that the functioning of the system depends on a quantity varying in time according to its performance. The quantity is called capital, it is denoted by  $\{V_t, t \in [0, T]\}$ , and it has a positive initial value  $V_0 > 0$ . During the service of a customer of type  $i$ ,  $i = 1, \dots, n$ , the capital increases by a random variable  $X$  per unit time with a given distribution function  $F_i(x)$ ,  $i = 1, \dots, n$ . The yield  $X'$  of the next served customer is independent with the distribution function  $F_k(x)$ , where  $k$  denotes his type, etc. The distribution function has the following properties:

$$(1) \quad \int_{-\infty}^{\infty} x \, dF_i(x) = c_i > 0, \quad \int_{-\infty}^{\infty} x^2 \, dF_i(x) = d_i q > 0, \\ \int_{-\infty}^{\infty} x^4 \, dF_i(x) = O(q^2), \quad i = 1, \dots, n.$$

Let  $x_t$  denote the rate of increase of the capital at time  $t$ . Thus,  $x_t$  is piecewise constant. Following the state change a new value of  $x_t$  is selected.

If  $i_t = 0$ , the capital has a constant decrease  $c_0 < 0$  per unit time. If the capital reaches zero, the ruin occurs, and the system ceases to work. The ruin moment is a random variable  $\tau = \inf \{t, V_t \leq 0\}$ . To measure the utility of the system's performance, we introduce the optimality criterion

$$E_y \left\{ \int_0^\tau e^{-\lambda t} dV_t - N e^{-\lambda \tau} \right\},$$

where  $E_y$  is the mathematical expectation under the condition that the initial capital equals  $y$ , the discount factor  $\lambda$  is a chosen positive number.  $N > 0$  denotes the penalty for the ruin. With regard to the danger of the ruin the strategy has to depend on the actual capital. It is given by a vector function  $u(y) = (u_1(y), \dots, u_n(y))$ , where  $0 \leq u_i(y) \leq 1$  has the following meaning: if the system is vacant and if the capital equals  $y$  and the  $i$ -th customer arrives, then  $u_i(y)$  denotes the probability of his accepting. The strategies with a bounded derivative are admissible and their totality is denoted by  $\mathcal{U}$ .

The system with  $d_i = c_i^2$ ,  $i = 1, \dots, n$  was investigated in [2]. No diffusion approximation was used; the system of Bellman's equations was derived directly for the expected discounted criterion.

## 2. LIMIT DISTRIBUTION OF THE CAPITAL

We shall prove a limit theorem for  $\{V_t, t \in [0, T]\}$  under the assumption that the parameter  $q$  tends to infinity and  $u(y) \in \mathcal{U}$  is a stationary control policy.

Let us define  $\theta(u(y))$ ,  $\sigma^2(u(y))$  (further the abbreviated denotation  $\theta(y)$ ,  $\sigma^2(y)$  will be used) together with  $w(i, x, y)$  and  $w_2(i, x, y)$ ,  $i = 1, \dots, n$ , as a solution of the following system of equations

$$(2) \quad x - b_i q w(i, x, y) - \theta(y) = 0, \quad i = 1, \dots, n,$$

$$c_0 + \sum_{k=1}^n a_k q u_k(y) \int_{-\infty}^{\infty} w(k, x, y) dF_k(x) - \theta(y) = 0,$$

$$(3) \quad w(i, x, y)^2 - b_i q w_2(i, x, y) - \sigma_q^2(y) = 0, \quad i = 1, \dots, n,$$

$$\sum_{k=1}^n a_k q u_k(y) \int_{-\infty}^{\infty} [w(k, x, y)^2 + w_2(k, x, y)] dF_k(x) - \sigma_q^2(y) = 0.$$

(We set  $w(0, x, y)$  and  $w_2(0, x, y)$  zero.)

Letting  $q$  to infinity we obtain

$$(4) \quad \theta(y) = \frac{c_0 + \sum_{k=1}^n \frac{a_k c_k}{b_k} u_k(y)}{1 + \sum_{k=1}^n \frac{a_k}{b_k} u_k(y)}, \quad \sigma^2(y) = \frac{2 \sum_{k=1}^n \frac{a_k d_k}{b_k^2} u_k(y)}{1 + \sum_{k=1}^n \frac{a_k}{b_k} u_k(y)}.$$

We are going to show that the evolution of  $V_t$  will be sufficiently closely described by the stochastic differential equation

$$(5) \quad dv_t = \theta(v_t) dt + \sigma(v_t) dW_t, \quad v_0 = V_0, \quad t \in [0, T],$$

where  $\{W_t, t \in [0, T]\}$  is a standardized Wiener process.

Let  $C_T$  be the space of all continuous functions on  $[0, T]$  with the uniform metric. Further, for  $t \in [0, T]$ , let  $\mathcal{G}_t$  be the  $\sigma$ -algebra on  $C_T$  generated by the sets

$$\{f \in C_T; f(s) \leq x\}, \quad s \in [0, t], \quad x \in (-\infty, \infty).$$

The random function  $\{y_t, t \in [0, T]\}$  is defined on  $(C_T, \mathcal{G}_T)$  by the relation  $y_t(f) = f(t)$ ,  $t \in [0, T]$ ,  $f \in C_T$ . The probability distribution of  $\{V_t, t \in [0, T]\}$  is the probability measure  $\mathcal{P}_T^q$  induced on  $(C_T, \mathcal{G}_T)$  by  $\{V_t, t \in [0, T]\}$ .

**Theorem.** Let the stationary control  $u(y)$  have a bounded derivative on  $(-\infty, \infty)$ . Then  $\mathcal{P}_T^q$  converges, as  $q \rightarrow \infty$ , weakly to the probability distribution  $\mathcal{P}_T$  of a random process  $\{v_t, t \in [0, T]\}$  such that

$$(6) \quad \begin{aligned} dv_t &= \theta(v_t) dt + \sigma(v_t) dW_t, \quad t \in [0, T], \\ \mathcal{P}_T(v_0 = V_0) &= 1, \end{aligned}$$

where  $\{W_t, t \in [0, T]\}$  is a standardized Wiener process.

According to the result of [5]  $\mathcal{P}_T$  is unique. The proof of the theorem will be decomposed into a sequence of lemmas. Let  $\mathcal{F}_t$  be the  $\sigma$ -algebra of random events defined by the history of the service system up to time  $t$ .

**Lemma 1.** Let

$$M_t = V_t - V_0 - \int_0^t \theta(V_s) ds + \int_0^t (w(s) - w(s-)) dN_s,$$

where  $w(s) = w(i_s, x_s, V_s)$  and  $N_t = \sum_i \chi\{v_i \leq t\}$ ,  $v_i$  are the moments of state changes. Then  $\{M_t, t \in [0, T]\}$  is a martingale with respect to  $\{\mathcal{F}_t, t \in [0, T]\}$ .

**Proof.** Let  $\Delta$  be arbitrarily small.

$$\begin{aligned} \mathbb{E}[M_{t+\Delta} - M_t | (i_t, x_t, V_t) = (i \neq 0, x, y)] &= \\ &= (1 - b_i q \Delta)(x \Delta - \theta(y) \Delta) - b_i q \Delta w(i, x, y) + o(\Delta) = o(\Delta). \end{aligned}$$

The last equality holds in virtue of (2). The same is valid for  $i = 0$ . Thus  $\{M_t, t \in [0, T]\}$  is the martingale.  $\square$

The above mentioned martingale has the following property.

**Lemma 2.** For  $0 \leq t_1 \leq t_2 \leq T$

$$E(M_{t_2} - M_{t_1})^4 \leq B(t_2 - t_1)^2 + A(t_2 - t_1),$$

where  $A \rightarrow 0$  as  $q \rightarrow \infty$  and  $B$  is a constant with respect to  $q$ .

**Proof.** Let  $\Delta = (t_2 - t_1)n^{-1}$ ,  $Y_k = M_{(k+1)\Delta} - M_{k\Delta}$ .

Then

$$\begin{aligned} E(M_{t_2} - M_{t_1})^4 &= E\left(\sum_{k=0}^{n-1} Y_k\right)^4 = E\left(\sum_{k=0}^{n-1} Y_k^4 + 4 \sum_{m=0}^{n-1} \left(\sum_{k<m} Y_k\right) Y_m^3 + \right. \\ &\quad \left. + 6 \sum_{m=0}^{n-1} \left(\sum_{k<m} Y_k\right)^2 Y_m^2\right). \end{aligned}$$

Letting  $n \rightarrow \infty$  we get

$$\begin{aligned} E(M_{t_2} - M_{t_1})^4 &= E\left(\int_{t_1}^{t_2} (w(s) - w(s-))^4 dN_s + \right. \\ &\quad \left. + 4 \int_{t_1}^{t_2} (M_{s-} - M_{t_1}) (w(s) - w(s-))^3 dN_s + \right. \\ &\quad \left. + 6 \int_{t_1}^{t_2} (M_{s-} - M_{t_1})^2 (w(s) - w(s-))^2 dN_s\right). \end{aligned}$$

Let us denote by  ${}^iN_t$  the counting process of transitions into state  $i$  and by  ${}^iQ_t$  the corresponding transition rate. If we define

$$\bar{w}_k(i, y) = \int_{-\infty}^{\infty} w(i, x, y)^k dF_i(x),$$

we have

$$\begin{aligned} E(M_{t_2} - M_{t_1})^4 &= E\left[\int_{t_1}^{t_2} w(s-)^4 {}^0Q_s ds - 4 \int_{t_1}^{t_2} (M_{s-} - M_{t_1}) \right. \\ &\quad \left. w(s-)^3 {}^0Q_s ds + 6 \int_{t_1}^{t_2} (M_{s-} - M_{t_1})^2 w(s-)^2 {}^0Q_s ds + \right. \\ &\quad \left. + \sum_{i=1}^n \left( \int_{t_1}^{t_2} \bar{w}_4(i, V_s) {}^iQ_s ds + 4 \int_{t_1}^{t_2} (M_{s-} - M_{t_1}) \bar{w}_3(i, V_s) {}^iQ_s ds + \right. \right. \\ &\quad \left. \left. + 6 \int_{t_1}^{t_2} (M_{s-} - M_{t_1})^2 \bar{w}_2(i, V_s) {}^iQ_s ds \right) \right]. \end{aligned}$$

Using Hölder inequality and the fact that  $E(M_{t_2} - M_{t_1})^4 = f(t_2)$  is a non-decreasing function in  $t_2$ , we obtain

$$f(t_2) \leq A_1(t_2 - t_1) + A_2(t_2 - t_1)f(t_2)^{1/4} + B_1(t_2 - t_1)f(t_2)^{1/2},$$

where  $A_1, A_2 \rightarrow 0$  as  $q \rightarrow \infty$  and  $B_1$  is independent of  $q$ . The statement of the lemma follows from the above inequality.  $\square$

**Lemma 3.** For  $t \in [0, T]$

$$\int_0^t (w(s) - w(s-)) dN_s = w(t) - w(0) - \int_0^t x_s w'(s) ds,$$

where  $w'(s) = (\partial/\partial y) w(i_s, x_s, V_s)$ .

*Proof.* Let  $(v_a, v_{a+1})$  be the interval between two transitions and  $v_a < t < v_{a+1}$ . Then  $x_s = x$ ,  $i_s = i$ ,  $V_s = y + x(s - v_a)$  for  $s \in (v_a, t)$  and

$$w(t) - w(v_a) = w(i, x, y + x(t - v_a)) - w(i, x, y) = \int_{v_a}^t x w'(i, x, V_s) ds.$$

Composing all such intervals we obtain the assertion of the lemma.  $\square$

Let us denote

$$(7) \quad Y_t = V_t - V_0 - \int_0^t \theta(V_s) ds, \quad t \in [0, T],$$

and let  $\mathcal{R}_T^q$  be the probability distribution of  $\{Y_t, t \in [0, T]\}$ .

**Lemma 4.** The family of  $\mathcal{R}_T^q$  is tight.

*Proof.* According to [1] it is sufficient to prove that

$$(8) \quad \lim_{\delta \rightarrow 0} \overline{\lim}_{q \rightarrow \infty} \mathcal{R}_T^q \left( \sup_{|t-s| < \delta} |y_t - y_s| > \varepsilon \right) = 0, \quad \varepsilon > 0.$$

By Lemma 3

$$Y_t = M_t - w(t) + w(0) + \int_0^t x_s w'(s) ds.$$

Take  $\varepsilon > 0$ ,  $\delta > 0$  and  $q > 0$ . Then

$$\begin{aligned} \mathcal{R}_T^q \left( \sup_{|t-s| < \delta} |y_s - y_t| > \varepsilon \right) &= P \left( \sup_{|t-s| < \delta} |Y_s - Y_t| > \varepsilon \right) \leq \\ &\leq P \left( \sup_{|t-s| < \delta} |M_t - M_s| > \frac{\varepsilon}{3} \right) + P \left( \sup_{|t-s| < \delta} |w(t) - w(s)| > \frac{\varepsilon}{3} \right) + \\ &+ P \left( \sup_{|t-s| < \delta} \left| \int_s^t x_u w'(u) du \right| > \frac{\varepsilon}{3} \right) \leq \sum_{j=0}^{[T/\delta]} P \left( \sup_{j\delta \leq s \leq (j+1)\delta} |M_s - M_{j\delta}| > \frac{\varepsilon}{9} \right) + \\ &+ P \left( \sup_{0 \leq s \leq T} 2|w(s)| > \frac{\varepsilon}{3} \right) + P \left( \int_0^T |x_s w'(s)| ds > \frac{\varepsilon}{3} \right) \leq \\ &\leq \sum_{j=0}^{[T/\delta]} \left( \frac{9}{\varepsilon} \right)^4 E(M_{(j+1)\delta} - M_{j\delta})^4 + \left( \frac{6}{\varepsilon} \right)^4 E \sup_{0 \leq s \leq T} |w(s)|^4 + \frac{3}{\varepsilon} E \int_0^T |x_s w'(s)| ds. \end{aligned}$$

In the last step submartingale inequality was used. From Lemma 2

$$\mathcal{R}_T^q \left( \sup_{|t-s| < \delta} |y_s - y_t| > \varepsilon \right) \leq \left( \frac{T}{\delta} + 1 \right) \left( \frac{9}{\varepsilon} \right)^4 (B\delta^2 + A\delta) + \left( \frac{6}{\varepsilon} \right)^4 C + \frac{3}{\varepsilon} D,$$

where each of  $A, C, D$  tends to zero as  $q \rightarrow \infty$ . Thus (8) is immediately obtained.  $\square$

The weak limit of any convergent subsequence  $\mathcal{R}_j^q, q_j \rightarrow \infty$ , is denoted by  $\mathcal{R}_T$ . Its existence is guaranteed by Lemma 4.

**Lemma 5.**  $\{y_t, t \in [0, T]\}$  is on  $(C_T, \mathcal{C}_T, \mathcal{R}_T)$  a quadratically integrable martingale with respect to  $\{\mathcal{C}_t, t \in [0, T]\}$ .

**Proof.** Let  $0 \leq s_1 < s_2 < \dots < s_k < s < t \leq T$  and let  $f(x_1, \dots, x_k)$  be a bounded continuous function on  $\mathbb{R}^k$ . From the martingale property of  $\{M_t, t \in [0, T]\}$  follows

$$E(M_t - M_s)f(M_{s_1}, \dots, M_{s_k}) = 0.$$

From the proof of Lemma 4 results

$$(9) \quad E \sup_{0 \leq t \leq T} |M_t - Y_t|^4 \rightarrow 0 \quad \text{as } q \rightarrow \infty.$$

Using (9)

$$E(Y_s - Y_t)f(Y_{s_1}, \dots, Y_{s_k}) - E(M_t - M_s)f(M_{s_1}, \dots, M_{s_k}) \rightarrow 0$$

as  $q \rightarrow \infty$ .

This gives the martingale property of  $\{y_t, t \in [0, T]\}$ . The integrability of its square follows from (9) and Lemma 2.  $\square$

**Lemma 6.** On  $(C_T, \mathcal{C}_T, \mathcal{R}_T)$

$$(10) \quad \mathcal{E}_T\{(y_t - y_s)^2 \mid \mathcal{C}_s\} = \mathcal{E}_T\left\{\int_s^t \sigma^2(v_u) du \mid \mathcal{C}_s\right\}$$

holds for  $0 \leq s < t \leq T$ , where  $\{v_t, t \in [0, T]\}$  is the solution of

$$(11) \quad v_t = V_0 + y_t + \int_0^t \theta(v_s) ds, \quad t \in [0, T].$$

( $\mathcal{E}_T$  denotes the mathematical expectation with respect to  $\mathcal{R}_T$ ).

**Proof.** Note that  $\theta(y)$  is Lipschitz continuous, and hence (11) has the unique solution. As in preceding proof, to establish (10) it suffices to show that

$$\int (y_t - y_s)^2 f(y_{s_1}, \dots, y_{s_k}) d\mathcal{R}_T = \int \left( \int_s^t \sigma^2(v_u) du f(y_{s_1}, \dots, y_{s_k}) d\mathcal{R}_T \right),$$

when  $s_1, \dots, s_k$  and  $f(x_1, \dots, x_k)$  are the same as in Lemma 5. From (9)

$$(12) \quad E(M_t - M_s)^2 f(s_1, \dots, s_{s_k}) - E(Y_t - Y_s)^2 f(Y_{s_1}, \dots, Y_{s_k}) \rightarrow 0 \quad \text{as } q \rightarrow \infty.$$

Let  $t < s$ , then

$$(13) \quad E\{(M_t - M_s)^2 \mid \mathcal{F}_s\} = E\left\{\int_s^t \sigma_q^2(V_u) du + w_2(s) - w_2(t) + \int_s^t x_u w_2'(u) du \mid \mathcal{F}_s\right\}.$$

The relation (13) is proved by proving martingale property for

$$\bar{M}_t = \int_0^t (w(s) - w(s-))^2 dN_s - \int_0^t \sigma_q^2(V_s) ds + \int_0^t (w_2(s) - w_2(s-)) dN_s.$$

The same method as in Lemma 1 is used with respect to (3). Lemma 3 also holds for  $w_2$ .

From (13)

$$(14) \quad E(M_t - M_s)^2 f(Y_{s_1}, \dots, Y_{s_k}) - E \int_0^t \sigma_q^2(V_u) du f(Y_{s_1}, \dots, Y_{s_k}) \rightarrow 0$$

as  $q \rightarrow \infty$ .

(12) and (14) together give the assertion of this lemma. □

**Corollary.**

$$y_t = y_0 + \int_0^t \sigma(v_s) dW_s, \quad t \in [0, T],$$

where  $\{W_t, t \in [0, T]\}$  is a Wiener process on  $(C_T, \mathcal{G}_T, \mathcal{R}_T)$ .

Proof.  $\{W_t = \int_0^t \sigma(v_s)^{-1} dy_s, t \in [0, T]\}$  is a martingale, which satisfies

$$\mathcal{E}_T\{(W_t - W_s)^2 | \mathcal{G}_s\} = t - s \quad \text{for } 0 \leq s < t \leq T.$$

This relation is a characteristic property of a Wiener process. □

The proof of the Theorem follows with regard to (7) and (11) from the fact that (6) holds on  $(C_T, \mathcal{G}_T, \mathcal{R}_T)$ .

### 3. OPTIMALITY

Let us make a slight change in denotation.

$$\theta(u) = \theta(u(y)), \quad \kappa(u) = \frac{1}{2}\sigma^2(u(y)) \quad \text{for } u(y) = u \in U = [0, 1]^n$$

For the limiting diffusion the optimality criterion has the form

$$E_y \left\{ \int_0^{\tau} e^{-\lambda t} dv_t - N e^{-\lambda \tau} \right\}, \quad \lambda > 0,$$

or

$$(15) \quad v(y) = E_y \left\{ \int_0^{\tau} e^{-\lambda t} \theta(v_t) dt - N e^{-\lambda \tau} \right\},$$

where  $\{v_t, t \in [0, T]\}$  satisfies (5) with  $v_0 = y$  and  $\tau = \inf\{t > 0, v_t \leq 0\}$ .

The problem of maximization of  $v(y)$  is the problem of controlling the one-dimensional Markov process with differential generator

$$\theta(u) \frac{d}{dy} + \kappa(u) \frac{d^2}{dy^2}$$

in such a way that (15) is maximal.



We define

$$(16) \quad \hat{v}(y) = \sup_{u \in U} v(y).$$

$\hat{v}(y)$  fulfils the Bellman equation

$$(17) \quad \max_{u \in U} \{ \kappa(u) \hat{v}''(y) + \theta(u) (\hat{v}'(y) + 1) - \lambda \hat{v}(y) \} = 0$$

(see for example [3]) with initial condition  $\hat{v}(0) = -N$  and  $\hat{v}(\infty) < \infty$ . The primes denote the derivatives with respect to  $y$ . The optimal process has the generator

$$\theta(\hat{u}(y)) \frac{d}{dy} + \kappa(\hat{u}(y)) \frac{d^2}{dy^2},$$

where  $\hat{u}(y)$  is the maximizer of the curly bracket in (17).  $\hat{u}(y)$  is not necessarily an element of  $\mathcal{U}$ .

Now, we shall construct the optimal strategy  $\hat{u}(y)$ . The whole construction is divided into four parts.

1. Let us put  $y = +\infty$ . Then

$$\hat{v}(\infty) = \lambda^{-1} \max_{u \in U} \theta(u) = \lambda^{-1} \theta(\hat{u}(\infty)).$$

2. Further, we solve for  $y \geq 0$

$$(18) \quad \kappa(\hat{u}(\infty)) v''(y) + \theta(\hat{u}(\infty)) (v'(y) + 1) - \lambda v(y) = 0.$$

The only bounded solution has the form

$$v(y) = K e^{py} + \lambda^{-1} \theta(\hat{u}(\infty)),$$

where  $p$  is the only negative root of the quadratic equation corresponding to (18) and  $K$  is an unknown constant. We shall assume  $K < 0$ . In such case  $v$  is the increasing function. When choosing  $K$  two cases can occur:

(i) There exists  $K$  so that  $v(0) = -N$  and simultaneously

$$(19) \quad (\kappa(\hat{u}(\infty)) - \kappa(u)) v''(0) + (\theta(\hat{u}(\infty)) - \theta(u)) (v'(0) + 1) \geq 0$$

for all  $u \in U$ .

Then the construction is finished and the optimal strategy  $\hat{u}(y) = \hat{u}(\infty)$  for all  $y \geq 0$ .

(ii) Case (i) does not hold. Then we choose  $K$  so that

$$\min_{u \in U} [(\kappa(\hat{u}(\infty)) - \kappa(u)) v''(0) + (\theta(\hat{u}(\infty)) - \theta(u)) (v'(0) + 1)] = 0.$$

3. Let (ii) occur. The strategy by which the minimum is achieved is denoted by  ${}^0u$  and we solve the following equation for  $y \leq 0$

$$\kappa({}^0u) v''(y) + \theta({}^0u) (v'(y) + 1) - \lambda v(y) = 0$$

with terminal conditions

$$v(0) = K + \lambda^{-1} \theta(\hat{u}(\infty)), \quad v'(0) = Kp.$$

Either, for  $y_N < 0$  such that  $v(y_N) = -N$  the following inequality holds

$$(20) \quad \min_{u \in U \setminus \{\hat{u}(\infty)\}} \{(\alpha^0 u - \alpha(u)) v^0(y_N) + (\theta^0 u - \theta(u)) (v(y_N) + 1)\} \geq 0,$$

and the construction is completed. Or, there exists  $0 > y_a > y_N$ , such that

$$\min_{u \in U \setminus \{\hat{u}(\infty)\}} \{(\alpha^0 u - \alpha(u)) v^0(y_a) + (\theta^0 u - \theta(u)) (v(y_a) + 1)\} = 0$$

The minimizing strategy is denoted by  ${}^1u$  and the whole procedure is repeated for  $y \leq y_a$  so many times till we obtain such  $y_N$  that  $v(y_N) = -N$  and corresponding inequality (20) holds.

4. The last step of the construction is the shifting of the end point  $y_N$  into zero. The resulting strategy is thus

$$\hat{u}(y) = {}^j u(y + y_N) \quad \text{for} \quad y_j \leq y + y_N < y_{j-1}.$$

Its optimality follows from the construction.

#### 4. EXAMPLE

In this section a numerical example is given. Only two types of customers are considered. The following parameters are chosen:

$$\begin{aligned} a_1 &= 1 & a_2 &= 0,1 & c_0 &= -1 \\ b_1 &= 1 & b_2 &= 0,5 & \lambda &= 1 \\ c_1 &= 1 & c_2 &= 50 \\ d_1 &= 1 & d_2 &= 100 \end{aligned}$$

1. According to the foregoing section the optimal strategy for  $y = +\infty$  equals

$$\hat{u}(\infty) = (\hat{u}_1(\infty), \hat{u}_2(\infty)) = (0, 1) \quad \text{and} \\ \theta(\hat{u}(\infty)) = 7, 5, \quad \alpha(\hat{u}(\infty)) = 33, 33.$$

2. The solution of (18) is

$$v(y) = K e^{-0,32y} + 7,5.$$

3. When choosing  $K$ , case (ii) occurs. We get  ${}^0u = (1, 1)$ ,  $\theta({}^0u) = 4,55$ ,  $\alpha({}^0u) = 18,64$  and the parameter  $K = -5,57$ . For  $y \leq 0$  the following equation is solved

$$18,64v''(y) + 4,55(v'(y) + 1) - v(y) = 0$$

with terminal conditions  $v(0) = 1,93$  and  $v'(0) = 9,28$ . The solution has the form

$$v(y) = -18,54 e^{-0,38y} + 15,92 e^{0,14y} + 4,55.$$

For fixed  $N$  we find  $y_N < 0$  such that  $v(y_N) = -N$ .

The following table gives several mutually corresponding values.

$N$	1	5	10	100
$y_N$	-0.32	-0.75	-1.29	-10.98

4. The optimal strategy  $\hat{u}(y)$  equals

$$\hat{u}(y) = \begin{cases} (1, 1) & \text{for } 0 \leq y < -y_N \\ (0, 1) & \text{for } -y_N \leq y. \end{cases}$$

*Remark.* If  $d_2 \leq 15,00$  the optimal strategy would equal (0,1) for all  $y \geq 0$ .

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#### REFERENCES

- [1] P. Billingsley: *Convergence of Probability Measures*. Wiley, New York 1968.
- [2] V. Lánská: On optimizing a one-server system with several types of customers. *Math. Operationsforsch. Statist., Ser. Optim.* 11 (1980), 2, 333–339.
- [3] P. Mandl: *Analytical Treatment of One-dimensional Markov Processes*. Academia, Prague – Springer - Verlag, Berlin 1968.
- [4] P. Mandl: On aggregating controlled Markov chains. In: *Contributions to Statistics* (J. Jurečková, ed.), Academia, Prague 1979, 136–156.
- [5] D. W. Stroock, S. R. Varadhan: Diffusion processes with continuous coefficients. *Comm. Pure Appl. Math.* XXII (1969), 345–400, 479–530.

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