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DIFFUSION APPROXIMATION FOR A CONTROLLED SERVICE SYSTEM

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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, ..., 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, ..., n. (The parameter *q* is sufficiently large). The behaviour of the system is described by a random process $\{i_i, t \in [0, T]\}$ with a finite set of states $\{0, 1, ..., n\}$. $i_t = 0$ means that the system is vacant at time $t, i_t = j, j = 1, ..., n$, 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, ..., n, the capital increases by a random variable X per unit time with a given distribution function $F_i(x)$, i = 1, ..., 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)
$$\int_{-\infty}^{\infty} x \, dF_i(x) = c_i > 0 , \quad \int_{-\infty}^{\infty} x^2 \, dF_i(x) = d_i q > 0 ,$$
$$\int_{-\infty}^{\infty} x^4 \, dF_i(x) = O(q^2) , \quad i = 1, ..., n .$$

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

$$\mathsf{E}_{\mathsf{y}}\left\{\int_{0}^{\mathsf{r}} \mathrm{e}^{-\lambda t} \, \mathrm{d} V_{t} - N \, \mathrm{e}^{-\lambda t}\right\},\,$$

where E_y is the mathematical expectation under the condition that the initial capital equals y, the discount factor λ 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 $u(y) = (u_1(y), \ldots, u_n(y))$, where $0 \le u_i(y) \le 1$ 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, ..., 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 $\{V_i, t \in [0, T]\}$ 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, ..., n, as a solution of the following system of equations

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

$$c_{0} + \sum_{k=1}^{\infty} a_{k} q u_{k}(y) \int_{-\infty}^{-\infty} w(k, x, y) dF_{k}(x) - \theta(y) = 0,$$

(3)
$$w(i, x, y)^2 - b_i q \, w_2(i, x, y) - \sigma_q^2(y) = 0, \quad i = 1, ..., n,$$
$$\sum_{k=1}^n a_k q \, u_k(y) \int_{-\infty}^{\infty} \left[w(k, x, y)^2 + w_2(k, x, y) \right] dF_k(x) - \sigma_q^2(y) = 0$$

(We set w(0, x, y) and $w_2(0, x, y)$ zero.) Letting q to infinity we obtain

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

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

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

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 \mathscr{C}_t be the σ -algebra on C_T generated by the sets

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

The random function $\{y_t, t \in [0, T]\}$ is defined on (C_T, \mathscr{C}_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 \mathscr{P}_T^q induced on (C_T, \mathscr{C}_T) by $\{V_t, t \in [0, T]\}$.

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

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

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

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

Lemma 1. Let

$$M_{t} = V_{t} - V_{0} - \int_{0}^{t} \theta(V_{s}) \, \mathrm{d}s + \int_{0}^{t} (w(s) - w(s -)) \, \mathrm{d}N_{s},$$

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

Proof. Let *△* be arbitrarily small.

$$E[M_{t+A} - M_t | (i_t, x_t, V_t) = (i \neq 0, x, y)] =$$

= $(1 - b_i q \Delta) (x \Delta - \theta(y) \Delta) - b_i q \Delta w(i, x, y) + o(\Delta) = o(\Delta).$

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.

The above mentioned martingale has the following property.

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

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

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

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

$$\mathsf{E}(M_{t_2} - M_{t_1})^4 = \mathsf{E}(\sum_{k=0}^{n-1} Y_k)^4 = \mathsf{E}(\sum_{k=0}^{n-1} Y_k^4 + 4\sum_{m=0}^{n-1} (\sum_{k < m} Y_k) Y_m^3 + 6\sum_{m=0}^{n-1} (\sum_{k < m} Y_k)^2 Y_m^2).$$

Letting $n \to \infty$ we get

$$\begin{split} \mathsf{E}(M_{t_2} - M_{t_1})^4 &= \mathsf{E}\left(\int_{t_1}^{t_2} (w(s) - w(s-))^4 \, \mathrm{d}N_s + \right. \\ &+ 4 \int_{t_1}^{t_2} (M_{s-} - M_{t_1}) \, (w(s) - w(s-))^3 \, \mathrm{d}N_s + \\ &+ 6 \int_{t_1}^{t_2} (M_{s-} - M_{t_1})^2 \, (w(s) - w(s-))^2 \, \mathrm{d}N_s \Big). \end{split}$$

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

$$\overline{w}_k(i, y) = \int_{-\infty}^{\infty} w(i, x, y)^k \, \mathrm{d}F_i(x) \, ,$$

we have

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

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

Lemma 3. For $t \in [0, T]$ $\int_{0}^{t} (w(s) - w(s-)) dN_{s} = w(t) - w(0) - \int_{0}^{t} x_{s} w'(s) ds,$ here $w'(s) = (\delta/\delta w) w(s, x, V)$

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.

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Let us denote

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

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

Lemma 4. The family of \mathscr{R}_T^q is tight.

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

(8)
$$\lim_{\delta \to 0} \overline{\lim_{q \to \infty}} \mathscr{R}^{q}_{T} (\sup_{|t-s| < \delta} |y_{t} - y_{s}| > \varepsilon) = 0, \quad \varepsilon > 0.$$

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

$$\begin{aligned} \mathscr{R}_{T}^{t}\left(\sup_{|t-s|<\delta}|y_{s}-y_{t}|>\varepsilon\right) &= P\left(\sup_{|t-s|<\delta}|Y_{s}-Y_{t}|>\varepsilon\right) \leq \\ &\leq P\left(\sup_{|t-s|<\delta}|M_{t}-M_{s}|>\frac{\varepsilon}{3}\right) + P\left(\sup_{|t-s|<\delta}|w(t)-w(s)|>\frac{\varepsilon}{3}\right) + \\ &+ P\left(\sup_{|t-s|<\delta}|M_{t}-M_{j\delta}|>\frac{\varepsilon}{3}\right) \leq \sum_{j=0}^{|T/\delta|} P\left(\sup_{|\delta\leq s\leq (j+1)\delta}|M_{s}-M_{j\delta}|>\frac{\varepsilon}{9}\right) + \\ &+ P\left(\sup_{0\leq s\leq T}2|w(s)|>\frac{\varepsilon}{3}\right) + P\left(\int_{0}^{T}|x_{s}w'(s)|\,\mathrm{d}s>\frac{\varepsilon}{3}\right) \leq \\ &\leq \sum_{j=0}^{|T/\delta|} \left(\frac{9}{\varepsilon}\right)^{4} \mathsf{E}\left(M_{(j+1)\delta}-M_{j\delta}\right)^{4} + \left(\frac{\delta}{\varepsilon}\right)^{4} \mathsf{E}\sup_{0\leq s\leq T}|w(s)|^{4} + \frac{3}{\varepsilon} \mathsf{E}\int_{0}^{T}|x_{s}w'(s)|\,\mathrm{d}s . \end{aligned}$$

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

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

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

The weak limit of any convergent subsequence $\mathscr{R}_{I}^{q}, q_{j} \to \infty$, is denoted by \mathscr{R}_{T} . Its existence is guaranteed by Lemma 4.

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

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

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

From the proof of Lemma 4 results

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

Using (9)

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

as $q \to \infty$.

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

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

(10)
$$\mathscr{E}_T\{(y_t - y_s)^2 \mid \mathscr{C}_s\} = \mathscr{E}_T\{\int_s^t \sigma^2(v_u) \, \mathrm{d}u \mid \mathscr{C}_s\}$$

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

(11)
$$v_t = V_0 + y_t + \int_0^t \theta(v_s) \, \mathrm{d}s \, , \ t \in [0, T] \, .$$

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

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

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

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

(12) $\mathsf{E}(M_t - M_s)^2 f(s_1, ..., Y_{s_k}) - \mathsf{E}(Y_t - Y_s)^2 f(Y_{s_1}, ..., Y_{s_k}) \to 0$ as $q \to \infty$. Let t < s, then

(13)
$$\mathsf{E}\{(M_t - M_s)^2 \mid \mathscr{F}_s\} = \mathsf{E}\left\{\int_s^t \sigma_q^2(V_u) \,\mathrm{d}u + w_2(s) - w_2(t) + \int_s^t x_u \, w_2'(u) \,\mathrm{d}u \mid \mathscr{F}_s\right\}.$$

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

$$\overline{M}_{t} = \int_{0}^{t} (w(s) - w(s-))^{2} dN_{s} - \int_{0}^{t} \sigma_{q}^{2}(V_{s}) ds + \int_{0}^{t} (w_{2}(s) - w_{2}(s-)) dN_{s}$$

The same method as in Lemma 1 is used with respect to (3). Lemma 3 also holds for w_2 . From (13)

(14)
$$\mathsf{E}(M_t - M_s)^2 f(Y_{s_1}, ..., Y_{s_k}) - \mathsf{E} \int_0^t \sigma_q^2(V_u) \, \mathrm{d} u \, f(Y_{s_1}, ..., Y_{s_k}) \to 0$$

as $q \to \infty$.

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

Corollary.

$$y_t = y_0 + \int_0^t \sigma(v_s) \, \mathrm{d} W_s \,, \quad t \in [0, T] \,,$$

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

Proof.
$$\{W_t = \int_0^t \sigma(v_s)^{-1} dy_s, t \in [0, T]\}$$
 is a martingale, which satisfies
 $\mathscr{E}_T\{(W_t - W_s)^2 | \mathscr{E}_s\} = t - s \text{ for } 0 \le s < t \le T.$

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

3. OPTIMALITY

Let us make a slight change in denotation.

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

For the limiting diffusion the optimality criterion has the form

$$\mathsf{E}_{y}\left\{\int_{0}^{\tau}\mathsf{e}^{-\lambda t}\,\mathsf{d}v_{t}-N\,\mathsf{e}^{-\lambda \tau}\right\}, \quad \lambda>0\,,$$

or

(15)
$$v(y) = \mathsf{E}_{y} \left\{ \int_{0}^{\tau} \mathrm{e}^{-\lambda t} \, \theta(v_{t}) \, \mathrm{d}t - N \, \mathrm{e}^{-\lambda t} \right\},$$

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

$$\theta(u)\frac{\mathrm{d}}{\mathrm{d}y}+\varkappa(u)\frac{\mathrm{d}^2}{\mathrm{d}y^2}$$

in such a way that (15) is maximal.

265

We define

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

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

(17)
$$\max_{u \in U} \left\{ \varkappa(u) \, \hat{v}''(y) + \theta(u) \left(\hat{v}'(y) + 1 \right) - \lambda \, \hat{v}(y) \right\} = 0$$

(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

$$\theta(\hat{u}(y)) \frac{\mathrm{d}}{\mathrm{d}y} + \varkappa(\hat{u}(y)) \frac{\mathrm{d}^2}{\mathrm{d}y^2}$$

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 $y \ge 0$

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

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 chossing K two cases can occur:

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

(19)
$$(\varkappa(\hat{u}(\infty)) - \varkappa(u)) v''(0) + (\theta(\hat{u}(\infty)) - \theta(u))(v'(0) + 1) \ge 0$$
for all $u \in U$.

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

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

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

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

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

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)
$$\min_{u\in U\setminus \{\hat{u}(\infty)\}} \{ (\varkappa(^{0}u) - \varkappa(u)) v''(y_{N}) + (\theta(^{0}u) - \theta(u)) (v'(y_{N}) + 1) \} \ge 0,$$

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

$$\min_{u \in U - \{u(\infty)\}} \{ (\varkappa(^{0}u) - \varkappa(u)) v''(y_{a}) + (\theta(^{0}u) - \theta(u)) (v'(y_{a}) + 1) \} = 0$$

The minimizing strategy is denoted by ${}^{1}u$ 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

$$\hat{u}(y) = {}^{j}u(y + y_{N})$$
 for $y_{j} \le y + y_{N} < y_{j-1}$.

Its optimality follows from the construction.

4. EXAMPLE

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

$$a_{1} = 1 \qquad a_{2} = 0, 1 \qquad c_{0} = -1$$

$$b_{1} = 1 \qquad b_{2} = 0, 5 \qquad \lambda = -1$$

$$c_{1} = 1 \qquad c_{2} = 50$$

$$d_{1} = 1 \qquad d_{2} = 100$$

1. According to the foregoing section the optimal strategy for $y = +\infty$ equals $\hat{u}(\infty) = (\hat{u}_1(\infty), \hat{u}_2(\infty)) = (0, 1)$ and $\theta(\hat{u}(\infty)) = 7, 5, \varkappa(\hat{u}(\infty)) = 33, 33.$

2. The solution of (18) is

$$v(y) = K e^{-0.32y} + 7.5.$$

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

$$8,64v''(y) + 4,55(v'(y) + 1) - v(y) = 0$$

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 $v(y_N) = -N$.

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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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