

E. Z. FERENSTEIN and A. SIEROCIŃSKI (Warszawa)

OPTIMAL STOPPING OF A RISK PROCESS

Abstract. Optimal stopping time problems for a risk process $U_t = u + ct - \sum_{n=0}^{N(t)} X_n$ where the number $N(t)$ of losses up to time t is a general renewal process and the sequence of X_i 's represents successive losses are studied. $N(t)$ and X_i 's are independent. Our goal is to maximize the expected return before the ruin time. The main results are closely related to those obtained by Boshuizen and Gouweleew [2].

1. Introduction. Let $\{N(t), t \geq 0\}$ be a renewal process representing the stream of losses of an insurance company, so $N(t)$ is the number of losses up to the time t . If T_i denotes the time of occurrence of the i th loss, then random variables (r.v.'s) $S_i = T_i - T_{i-1}$ are independent identically distributed (i.i.d.) with a cumulative distribution function (c.d.f.) F , $T_0 = 0$. Let X_1, X_2, \dots be a sequence of i.i.d. r.v.'s with c.d.f. H , representing the successive losses. As a capital assets model for the insurance company we take the risk process

$$(1) \quad U_t = u + ct - \sum_{n=0}^{N(t)} X_n,$$

where $u > 0$ represents the initial capital and $c > 0$ is a constant rate of income from the insurance premium, $X_0 = 0$. The return at time t will be defined by the process $\{Z(t), t \geq 0\}$ where

$$(2) \quad Z(t) = \begin{cases} g_1(U_t) \mathbf{I}\{U_s > 0, s \leq t\} & \text{if } t \leq t_0, \\ 0 & \text{if } t > t_0, \end{cases}$$

where g_1 is a utility function. For simplicity define $g(u, t) = g_1(u) \mathbf{I}\{t \geq 0\}$.

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Then

$$(3) \quad Z(t) = g(U_t, t_0 - t) \prod_{j=1}^{N(t)} \mathbf{I}\{U_{T_j} > 0\}.$$

Let

$$(4) \quad \mathcal{F}(t) = \sigma(U_s, s \leq t) = \sigma(X_1, T_1, \dots, X_{N(t)}, T_{N(t)})$$

be the σ -field generated by all events up to time t , $t \geq 0$, and \mathcal{T} be the set of stopping times with respect to the family $\{\mathcal{F}(t), t \geq 0\}$. Moreover, for $n = 0, 1, 2, \dots, n < K$, denote by $\mathcal{T}_{n,K}$ the subset of \mathcal{T} such that

$$(5) \quad \tau \in \mathcal{T}_{n,K} \quad \text{if and only if} \quad T_n \leq \tau \leq T_K \quad \text{a.s.}$$

Set $\mathcal{F}_n \triangleq \mathcal{F}(T_n)$. We will be interested in finding optimal stopping times τ^* , $\tau_{n,K}^*$, τ_K^* such that

$$(6) \quad EZ(\tau^*) = \sup\{EZ(\tau) : \tau \in \mathcal{T}\},$$

$$(7) \quad EZ(\tau_K^*) = \sup\{EZ(\tau) : \tau \in \mathcal{T}_{0,K}\},$$

$$(8) \quad E\{Z(\tau_{n,K}^*) \mid \mathcal{F}_n\} = \text{ess sup}\{E(Z(\tau) \mid \mathcal{F}_n) : \tau \in \mathcal{T}_{n,K}\}.$$

The crucial role in the subsequent considerations is played by the following representation theorem for stopping times (see for example Davis [4]):

LEMMA 1. *If $\tau \in \mathcal{T}_{n,K}$, then there exists a positive \mathcal{F}_n -measurable r.v. R_n such that*

$$(9) \quad \tau \wedge T_{n+1} = (T_n + R_n) \wedge T_{n+1} \quad \text{a.s.}$$

2. Finite horizon case. In this section we will find the form of optimal stopping rules in the finite horizon case, i.e. optimal in the class $\mathcal{T}_{0,K}$, where K is finite and fixed. First, in Theorem 1, we will derive dynamic programming equations satisfied by

$$(10) \quad \Gamma_{n,K} = \text{ess sup}\{E(Z(\tau) \mid \mathcal{F}_n) : \tau \in \mathcal{T}_{n,K}\}, \quad n = K, K-1, \dots, 1.$$

Then, in Theorem 2, we will find optimal stopping times $\tau_{n,K}^*$ and τ_K^* and corresponding optimal conditional mean rewards and optimal mean rewards, respectively. Define

$$(11) \quad \mu_n = \prod_{j=1}^n \mathbf{I}(U_{T_j} > 0), \quad \mu_0 = 1.$$

Note that

$$(12) \quad \Gamma_{K,K} = Z(T_K) = g(U_{T_K}, t_0 - T_K)\mu_K.$$

THEOREM 1. (i) For $n = K - 1, K - 2, \dots, 0$,

$$\Gamma_{n,K} = \text{ess sup}\{\mu_n \bar{F}(R_n)g(U_{T_n} + cR_n, t_0 - T_n - R_n) + E(\mathbf{I}\{R_n \geq S_{n+1}\}\Gamma_{n+1,K} | \mathcal{F}_n) : R_n \geq 0, R_n \text{ is } \mathcal{F}_n\text{-measurable}\} \text{ a.s.,}$$

where $\bar{F} = 1 - F$ denotes the survival function.

(ii) For $n = K, K - 1, \dots, 0$,

$$(13) \quad \Gamma_{n,K} = \mu_n \gamma_{K-n}(U_{T_n}, T_n) \quad \text{a.s.,}$$

where the sequence of functions $\{\gamma_j(u, t), u \in \mathbb{R}, t \geq 0\}$ is defined recursively as follows:

$$(14) \quad \gamma_0(u, t) = g(u, t_0 - t),$$

$$(15) \quad \gamma_j(u, t) = \sup_{r \geq 0} \left[\bar{F}(r)g(u + cr, t_0 - t - r) + \int_0^r dF(s) \int_0^{u+cs} \gamma_{j-1}(u + cs - x, t + s) dH(x) \right], \quad j = 1, 2, \dots$$

Proof. (i) Let $\tau \in \mathcal{T}_{n,K}$, $0 \leq n < K < \infty$. From Lemma 1 we get

$$A_n \triangleq \{\tau < T_{n+1}\} = \{T_n + R_n < T_{n+1}\} = \{R_n < S_{n+1}\}$$

and

$$\bar{A}_n = \{\tau \geq T_{n+1}\} = \{R_n \geq S_{n+1}\}.$$

Then, using the properties of the conditional expectation, we can obtain the conditional expectation of the return at τ :

$$E(Z(\tau) | \mathcal{F}_n) = E(Z(\tau)\mathbf{I}_{A_n} | \mathcal{F}_n) + E(Z(\tau)\mathbf{I}_{\bar{A}_n} | \mathcal{F}_n) \triangleq \alpha_n + \beta_n,$$

where

$$\begin{aligned} \alpha_n &= E(\mathbf{I}\{R_n < S_{n+1}\}g(U_\tau, t_0 - \tau)\mu_n | \mathcal{F}_n) \\ &= \mu_n E(\mathbf{I}\{R_n < S_{n+1}\}g(U_{T_n} + cR_n, t_0 - T_n - R_n) | \mathcal{F}_n) \\ &= \mu_n \bar{F}(R_n)g(U_{T_n} + cR_n, t_0 - T_n - R_n). \end{aligned}$$

Note that β_n can be expressed as follows:

$$\beta_n = E[\mathbf{I}\{S_{n+1} \leq R_n\}E(Z(\tau') | \mathcal{F}_{n+1}) | \mathcal{F}_n],$$

where $\tau' = \tau \vee T_{n+1} \in \mathcal{T}_{n+1,K}$. Hence,

$$\begin{aligned} E(Z(\tau) | \mathcal{F}_n) &= \mu_n \bar{F}(R_n)g(U_{T_n} + cR_n, t_0 - T_n - R_n) \\ &\quad + E[\mathbf{I}\{S_{n+1} \leq R_n\}E(Z(\tau') | \mathcal{F}_{n+1}) | \mathcal{F}_n]. \end{aligned}$$

Now, following the standard reasoning of optimal stopping theory, we get the dynamic programming equation for $\Gamma_{n,K}$, $n = K, K - 1, \dots, 0$, given in (i), with $\Gamma_{K,K} = \mu_K g(U_{T_K}, t_0 - T_K)$.

(ii) We will prove (ii) using the backward induction method for $n = K - 1, \dots, 1$. First note that (ii) is satisfied for $n = K$ since $\Gamma_{K,K} = \mu_K \gamma_0(U_{T_K}, T_K)$.

Let $n = K - 1$. Then from (i) and the definition of the risk process (1) we get

$$\begin{aligned} \Gamma_{K-1,K} &= \text{ess sup} \{ \mu_{K-1} \bar{F}(R_{K-1}) g(U_{T_{K-1}} + cR_{K-1}, t_0 - T_{K-1} - R_{K-1}) \\ &\quad + E(\mu_K \mathbf{I}\{R_{K-1} \geq S_K\}) \gamma_0(U_{T_{K-1}} + cS_K - X_K, T_{K-1} + S_K) \mid \mathcal{F}_{K-1} \} : \\ &\quad R_{K-1} \text{ is } \mathcal{F}_{K-1}\text{-measurable, } R_{K-1} \geq 0 \}. \end{aligned}$$

Now, to get $\Gamma_{K-1,K} = \mu_{K-1} \gamma_1(U_{T_{K-1}}, T_{K-1})$ it is sufficient to note that $\mu_K = \mu_{K-1} \mathbf{I}\{U_{T_{K-1}} + cS_K - X_K > 0\}$. Moreover, the random variables S_K , X_K and the σ -field \mathcal{F}_{K-1} are independent; S_K and X_K have c.d.f. F and H , respectively.

Let $1 \leq n \leq K - 1$ and suppose that $\Gamma_{n,K} = \mu_n \gamma_{K-n}(U_{T_n}, T_n)$. From (i) we have

$$\begin{aligned} \Gamma_{n-1,K} &= \text{ess sup} \{ \mu_{n-1} \bar{F}(R_{n-1}) g(U_{T_{n-1}} + cR_{n-1}, t_0 - T_{n-1} - R_{n-1}) \\ &\quad + \mu_n E(\mathbf{I}\{R_{n-1} \geq S_n\}) \gamma_{K-n}(U_{T_n}, T_n) \mid \mathcal{F}_{n-1} \} : \\ &\quad R_{n-1} \geq 0, R_{n-1} \text{ is } \mathcal{F}_{n-1}\text{-measurable} \}. \end{aligned}$$

The second term under ess sup can be rewritten in the following way:

$$\begin{aligned} \mu_{n-1} E[\mathbf{I}\{R_{n-1} \geq S_n\} \mathbf{I}\{U_{T_{n-1}} + cS_n - X_n > 0\} \\ \times \gamma_{K-n}(U_{T_{n-1}} + cS_n - X_n, T_{n-1} + S_n) \mid \mathcal{F}_{n-1}]. \end{aligned}$$

Then we have

$$\begin{aligned} \Gamma_{n-1,K} &= \mu_{n-1} \text{ess sup} \left\{ \bar{F}(R_{n-1}) g(U_{T_{n-1}} + cR_{n-1}, t_0 - T_{n-1} - R_{n-1}) \right. \\ &\quad \left. + \int_0^{R_{n-1}} dF(s) \int_0^{U_{T_{n-1}} + cs} \gamma_{K-n}(U_{T_{n-1}} + cs - x, T_n + s) dH(x) : \right. \\ &\quad \left. R_{n-1} \geq 0, R_{n-1} \text{ is } \mathcal{F}_{n-1}\text{-measurable} \right\} \\ &= \mu_{n-1} \cdot \gamma_{K-(n-1)}(U_{T_{n-1}}, T_{n-1}). \quad \blacksquare \end{aligned}$$

To find the form of optimal stopping times τ_K^* we need to analyze properties of the sequence of functions $\{\gamma_n, n \geq 0\}$ defined in the second part of Theorem 1.

Let $B = B[(-\infty, \infty) \times [0, \infty)]$ be the space of all bounded continuous functions, with the norm $\|\delta\| = \sup_{u,t} |\delta(u, t)|$, and

$$(16) \quad B^0 = \{\delta : \delta(u, t) = \delta_1(u, t) \mathbf{I}\{t \leq t_0\} \text{ and } \delta_1 \in B\}.$$

For any $\delta \in B^0$ and any $u \in \mathbb{R}$, $t, r \geq 0$ define

$$(17) \quad \phi_\delta(r, u, t) \triangleq \bar{F}(r)g(u + cr, t_0 - t - r) + \int_0^r dF(s) \left[\int_0^{u+cs} \delta(u + cs - x, t + s) dH(x) \right].$$

Note that the properties of the c.d.f. F imply that $\phi_\delta(r, u, t)$ has an at most countable number of points of discontinuity with respect to r and is continuous with respect to (u, t) provided that $g_1(\cdot)$ is continuous and $t \neq t_0 - r$. In what follows we will use the following

ASSUMPTION 1. *The function $g_1(\cdot)$ is bounded and continuous.*

For all $\delta \in B^0$ define

$$(18) \quad (\Phi\delta)(u, t) = \sup_{r \geq 0} \{\phi_\delta(r, u, t)\}.$$

LEMMA 2. *For any $\delta \in B^0$ we have*

$$(\Phi\delta)(u, t) = \max_{0 \leq r \leq t_0 - t} \{\phi_\delta(r, u, t)\} \in B^0$$

and there exists a function r_δ such that $(\Phi\delta)(u, t) = \phi_\delta(r_\delta(u, t), u, t)$.

PROOF. Observe that for all $\delta \in B^0$ and for any $r > t_0 - t$ we have

$$(19) \quad \phi_\delta(r, u, t) = \int_0^{t_0 - t} dF(s) \left[\int_0^{u+cs} \delta(u + cs - x, t + s) dH(x) \right].$$

Hence, Assumption 1 and the fact that F has an at most finite number of discontinuity points in the compact interval $[0, t_0]$ imply the form of Φ . ■

Observe that for $i = 1, 2, \dots, u \in \mathbb{R}, t \geq 0, \gamma_i(u, t)$ can be rewritten as follows:

$$(20) \quad \gamma_i(u, t) = \begin{cases} (\Phi\gamma_{i-1})(u, t) & \text{if } u \geq 0 \text{ and } t \leq t_0, \\ 0 & \text{otherwise,} \end{cases}$$

and from Lemma 2 there exist functions $r_i \triangleq r_{\gamma_{i-1}}$ such that

$$(21) \quad \gamma_i(u, t) = \begin{cases} \phi_{\gamma_{i-1}}(r_i(u, t), u, t) & \text{if } u \geq 0 \text{ and } t \leq t_0, \\ 0 & \text{otherwise.} \end{cases}$$

To determine the form of optimal stopping times $\tau_{n,K}^*$ we need to define the following r.v.'s:

$$(22) \quad R_i^* \triangleq r_{K-i}(U_{T_i}, T_i)$$

and

$$(23) \quad \sigma_{n,K} = K \wedge \inf\{i \geq n : R_i^* < S_{i+1}\}.$$

THEOREM 2. *Let*

$$(24) \quad \tau_{n,K}^* = T_{\sigma_{n,K}} + R_{\sigma_{n,K}}^* \quad \text{and} \quad \tau_K^* = \tau_{0,K}^*.$$

Then, for any $0 \leq n \leq K$, we have

$$(25) \quad \Gamma_{n,K} = E(Z(\tau_{n,K}^*) \mid \mathcal{F}_n) \quad \text{a.s.} \quad \text{and} \quad \Gamma_{0,K} = E(Z(\tau_K^*)) = \gamma_K(u, 0).$$

Proof. This is a straightforward consequence of the formulas (22)–(24) and Theorem 1.

3. Infinite horizon case. In this section we will show that there exists an optimal stopping rule τ^* in the infinite horizon case, maximizing over \mathcal{T} the mean return (2), i.e. (6) is fulfilled. Moreover, the optimal stopping time τ^* can be defined as a limit of the finite horizon optimal stopping times.

ASSUMPTION 2. $F(t_0) < 1$.

LEMMA 3. The operator $\Phi : B^0 \rightarrow B^0$ defined by (18) is a contraction.

Proof. Let $\delta_1, \delta_2 \in B^0$. By Lemma 2 there exist $\varrho_i \triangleq r_{\delta_i}(u, t)$, $i = 1, 2$, such that $(\Phi\delta_i)(u, t) = \phi_{\delta_i}(\varrho_i, u, t)$, $i = 1, 2$. Since $\phi_{\delta_2}(\varrho_2, u, t) \geq \phi_{\delta_2}(\varrho_1, u, t)$ we obtain the inequalities

$$\begin{aligned} (\Phi\delta_1)(u, t) - (\Phi\delta_2)(u, t) &\leq \int_0^{\varrho_1} dF(s) \int_0^{u+cs} [\delta_1 - \delta_2](u + cs - x, t + s) dH(x) \\ &\leq \|\delta_1 - \delta_2\| \int_0^{\varrho_1} dF(s) \int_0^{u+cs} dH(x) \leq \varrho \|\delta_1 - \delta_2\|, \end{aligned}$$

where

$$(26) \quad \varrho = \sup_{u>0} \int_0^{t_0} dF(s) \int_0^{u+cs} dH(x) \leq F(t_0) < 1.$$

Similarly, we get $(\Phi\delta_2)(u, t) - (\Phi\delta_1)(u, t) \leq \varrho \|\delta_1 - \delta_2\|$. Hence, $\|\Phi\delta_2 - \Phi\delta_1\| \leq \varrho \|\delta_1 - \delta_2\|$. ■

Since $\gamma_0(u, t) = g(u, t_0 - t)$ it follows that $\gamma_i \in B^0$ for all i . Hence, from the Fixed Point Theorem we get the following lemma.

LEMMA 4. There exists $\gamma \in B^0$ such that

$$(27) \quad \gamma = \Phi\gamma \quad \text{and} \quad \lim_{K \rightarrow \infty} \|\gamma_K - \gamma\| = 0.$$

Remark 1. Note that all optimal stopping times are less than t_0 a.s., which is a consequence of the definition (2) of the return.

THEOREM 3. Assume that the utility function g_1 is differentiable and nondecreasing, and F has the density function f . Then

- (i) for $n = 0, 1, \dots$, the limit $\hat{\tau}_n \triangleq \lim_{K \rightarrow \infty} \tau_{n,K}^*$ exists and $\hat{\tau}_n$ is an optimal stopping rule in $\mathcal{T} \cap \{\tau \geq T_n\}$,
- (ii) $E[Z(\hat{\tau}_n) \mid \mathcal{F}_n] = \mu_n \gamma(U_{T_n}, T_n)$ a.s.

Proof. (i) Let $n \geq 0$. Note that $\tau_{n,K}^* \leq \tau_{n,K+1}^*$ a.s. Hence, the stopping rule $\hat{\tau}_n = \lim_{K \rightarrow \infty} \tau_{n,K}^* \geq T_n$ exists.

To prove optimality of $\hat{\tau}_n$ we will apply similar arguments to those used by Boshuizen and Goeweleeuw [2] in the proof of the existence of optimal stopping times for semi-Markov processes. Let $\xi_t = (t, U_t, Y_t, V_t)$, $Y_t = t - T_{N(t)}$, $V_t = \mu_{N(t)}$, $t \geq 0$. Then $\xi = \{\xi_t : t \geq 0\}$ is a Markov process with the state space $\mathbb{R}_+^1 \times \mathbb{R}^1 \times \mathbb{R}_+^1 \times \{0, 1\}$. Note that the return $Z(t)$ is a function, say \tilde{g} , of ξ_t . Let A be a strong generator of ξ . Then we get

$$(28) \quad (A\tilde{g})(t, u, y, v) = \left\{ cg'_1(u) - \frac{f(y)}{\bar{F}(y)} \left[g_1(u) - \int_0^u g_1(u-x) dH(x) \right] \right\} v,$$

where $t < t_0$, $y \geq 0$ and $v \in \{0, 1\}$.

Now, note that $\tilde{g}(\xi_t) - \tilde{g}(\xi_0) - \int_0^t (A\tilde{g})(\xi_s) ds$, $t \geq 0$, is a martingale with respect to $\sigma(\xi_s, s \leq t)$, which is the same as $\mathcal{F}(t)$ (see [3], p. 31). Applying the optional sampling theorem ([3], p. 22) we get

$$(29) \quad E[\tilde{g}(\xi_{\tau_{n,K}^*}) | \xi_{T_n}] - \tilde{g}(\xi_{T_n}) = E \left[\int_{T_n}^{\tau_{n,K}^*} (A\tilde{g})(\xi_s) ds \mid \mathcal{F}_n \right] \quad \text{a.s.}$$

Since

$$(30) \quad (A\tilde{g})(\xi_s) = \left\{ cg'_1(U_s) + \frac{f(s - T_{N(s)})}{\bar{F}(s - T_{N(s)})} \times \left[\int_0^{U_s} g_1(U_s - x) dH(x) - g_1(U_s) \right] \right\} \mu_{N(s)},$$

the right hand side of (29) can be expressed as the difference $E(I_{n,K}^1 | \mathcal{F}_n) - E(I_{n,K}^2 | \mathcal{F}_n)$, where

$$I_{n,K}^2 = \int_{T_n}^{\tau_{n,K}^*} \frac{f(s - T_{N(s)})}{\bar{F}(s - T_{N(s)})} g_1(U_s) \mu_{N(s)} ds.$$

Now, $I_{n,K}^1$, $I_{n,K}^2$ are positive r.v.'s and $I_{n,K}^2$ is bounded by $g_1(u + ct_0) \times E(L)/\bar{F}(t_0)$, where $L = \inf\{n \in \mathbb{N} : T_n < t_0, T_{n+1} \geq t_0\}$. Note that

$$E(L) = \sum_{n=1}^{\infty} F^{*(n)}(t_0) \leq \sum_{n=1}^{\infty} [F(t_0)]^n < \infty.$$

Hence, from the convergence of $\tau_{n,K}^*$ to $\hat{\tau}_n$ as $K \rightarrow \infty$ and the Monotone Convergence Theorem we see that the right hand side of (29) converges to

$$E \left[\int_{T_n}^{\hat{\tau}_n} (A\tilde{g})(\xi_s) ds \mid \mathcal{F}_n \right].$$

Again, applying Dynkin's formula, since $\hat{\tau}_n < \infty$ a.s. we get

$$(31) \quad E \left[\int_{T_n}^{\hat{\tau}_n} (A\tilde{g})(\xi_s) ds \mid \mathcal{F}_n \right] = E[\tilde{g}(\xi_{\hat{\tau}_n}) \mid \mathcal{F}_n] - \tilde{g}(\xi_{T_n}) \quad \text{a.s.}$$

Hence, we have

$$(32) \quad E[\tilde{g}(\xi_{\tau_{n,K}^*}) \mid \mathcal{F}_n] \xrightarrow{K \rightarrow \infty} E[\tilde{g}(\xi_{\hat{\tau}_n}) \mid \mathcal{F}_n] \quad \text{a.s.}$$

Now, we will prove that $\hat{\tau}_n$ is optimal in the class $\mathcal{T} \cap \{\tau : \tau \geq T_n\}$. Let τ be any stopping rule from $\mathcal{T} \cap \{\tau : \tau \geq T_n\}$. Then, as $\tau_{n,K}^*$ is optimal in $\mathcal{T}_{n,K}$, we have for any K ,

$$(33) \quad E[\tilde{g}(\xi_{\tau_{n,K}^*}) \mid \mathcal{F}_n] \geq E[\tilde{g}(\xi_{\tau \wedge T_n}) \mid \mathcal{F}_n] \quad \text{a.s.}$$

Hence, a reasoning similar to that which led to (32) gives

$$(34) \quad E[\tilde{g}(\xi_{\hat{\tau}_n}) \mid \mathcal{F}_n] \geq E[\tilde{g}(\xi_{\tau}) \mid \mathcal{F}_n] \quad \text{a.s.},$$

which completes the proof of (i).

(ii) $E[\tilde{g}(\xi_{\tau_{n,K}^*}) \mid \mathcal{F}_n] = \mu_n \gamma_{K-n}(U_{T_n}, T_n)$ from Theorem 1(ii). Now, Lemma 4 and (34) give

$$(35) \quad E[\tilde{g}(\xi_{\tau_{n,K}^*}) \mid \mathcal{F}_n] \xrightarrow{K \rightarrow \infty} E[\tilde{g}(\xi_{\hat{\tau}_n}) \mid \mathcal{F}_n] = \mu_n \gamma(U_{T_n}, T_n) \quad \text{a.s.} \quad \blacksquare$$

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Elżbieta Z. Ferenstein and Andrzej Sierociński
 Institute of Mathematics
 Warsaw University of Technology
 Pl. Politechniki 1
 00-661 Warszawa, Poland

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