

On the Structure of Certain Excursions of a Markov Process

B. Maisonneuve*

Université de Grenoble II, I.M.S.S. 47X-38040 Grenoble Cédex

Summary. Let X be a Markov process and M a homogeneous random set. For $t \geq 0$, we set $G_t = \text{Sup} \{s \leq t: s \in M\}$. The stochastic dependence between the past and the future of G_T is investigated for certain stopping times T . This gives some insight to recent results of Gettoor concerning the excursion straddling t and the first excursion exceeding a in length.

I. Introduction

Let $(\Omega, \mathcal{F}, \mathcal{F}_t, X_t, \theta_t, P^x)$ be a strong Markov process and let M be a homogeneous closed well measurable set (for instance $M = \{t: X_t \in B\}$, where B is a Borel subset of the state space E). Set

$$G_t = \text{Sup} \{s \leq t: s \in M\}, \quad A_t = t - G_t, \quad t \geq 0.$$

We will establish that if T is a stopping time of the family (\mathcal{F}_{G_t}) , the past and the future relative to G_T are conditionally independent given (A_T, X_{G_T}) . More precisely, there exists a family $(P^{a,x})$ of probabilities on Ω such that $P^{0,x} = P^x$ and

$$E^*[f \circ \theta_{G_T} | A_T, \mathcal{F}_{G_T}] = P^{A_T, X_{G_T}}(f) \quad \text{a.s. on } \{T < \infty\}$$

for all (\mathcal{F}_{G_t}) -stopping times T and all positive measurable functions f on Ω ($P^{a,x}(f)$ denotes $\int f dP^{a,x}$).

In particular if we set

$$T^a = \text{Inf} \{t: A_t > a\}, \quad G^a = G_{T^a}, \quad a > 0,$$

one has $A_{T^a} = a$ on $\{G^a < \infty\}$ and the preceding formula specializes as follows

$$E^*[f \circ \theta_{G^a} | \mathcal{F}_{G^a}] = P^{a, X_{G^a}}(f) \quad \text{a.s. on } \{G^a < +\infty\}.$$

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Therefore the past and the future of G^a are conditionally independent given X_{G^a} .

These formulas generalize and make more understandable the following result recently established by Gettoor [1]: the distribution of the excursion straddling t , conditionally on $A_t=a, X_{G_t}=x$, is the same as the distribution of the first excursion exceeding a in length, conditionally on $X_{G^a}=x$.

We shall also give analogous results for conditioning on $\mathcal{F}_{G_T^-}$, under somewhat stronger assumptions. The reader will find a lot of nice applications and explicit computations for the excursions from a point in Gettoor's paper [1].

II. Notations and Preliminary Results

We shall use the notations and assumptions of [3]. $(\Omega, \mathcal{F}, \mathcal{F}_t, X_t, \theta_t, P^x)$ is the canonical right continuous realization of a Markov semi-group (P_t) satisfying the "hypothèses droites" of Meyer. The state space E is Lusin; in E we distinguish a point δ . We assume that δ is absorbing and that the lifetime $\zeta = \inf\{t: X_t = \delta\}$ is infinite P^x -a.s. for each $x \neq \delta$.

M is a closed random subset of \mathbb{R}_+ , well measurable (or optional) and homogeneous in $]0, \infty[$:

$$(M - t) \cap]0, \infty[= M \circ \theta_t \cap]0, \infty[, \quad t \geq 0.$$

We associate with M the following notations:

$$\begin{aligned} R &= \inf\{t > 0: t \in M\}, & R_t &= R \circ \theta_t, & D_t &= t + R_t, & t &\geq 0, \\ G_t &= \sup\{s \leq t: s \in M\}, & 0 &\leq t \leq \infty, \\ A_t &= t - G_t, & t &\geq 0, & A_\infty &= \infty, \\ G &= \{t > 0: R_{t-} = 0, R_t > 0\}, \\ F &= \{x \in E: P^x\{R = 0\} = 1\}. \end{aligned}$$

The set $G(\omega)$ is the set of the left endpoints in $]0, \infty[$ of the intervals that are contiguous to $M(\omega)$.

R is assumed to be \mathcal{F}^* -measurable - \mathcal{F}^* denotes the universal completion of $\mathcal{F}^0 = \sigma\{X_s, s \geq 0\}$. Then one has the following result (see [3]).

Theorem 1. *There exists an "exit system" (B, \hat{P}) : B is a well measurable random measure, \hat{P} is a kernel from (E, \mathcal{E}^*) to (Ω, \mathcal{F}^*) such that*

$$E^* \left[\sum_{s \in G} Z_s f \circ \theta_s \right] = E^* \left[\int_0^\infty Z_s \hat{P}^{X_s}(f) dB_s \right] \tag{1}$$

for all positive well-measurable Z and \mathcal{F}^* -measurable f .

In view of [3], one may and will choose the system (B, \hat{P}) with the following properties: $dB_t = dK_t + \sum_{\substack{s \in G \\ X_s \notin F}} \varepsilon_s(dt)$, where K is a continuous additive functional carried by F , with 1-potential ≤ 1 ; $\hat{P}^x = P^x$ if $x \notin F$; $\hat{P}^x\{R = 0\} = 0$ and $\hat{P}^x(1 - e^{-R}) \leq 1$ for all x (which implies that $\hat{P}^x\{R > a\} < \infty$ for all x and all $a > 0$); and finally $(X_t)_{t > 0}$ is strong Markov with semi-group (P_t) relative to the measure \hat{P}^x , for all x .

III. Conditioning on (A_T, \mathcal{F}_{G_T})

For each positive random variable S , \mathcal{F}_S denotes the σ -field of the sets $A \in \mathcal{F}$ such that $I_A = Z_S$ on $\{S < \infty\}$ for some well measurable process Z . For example $S, X_S(X_\infty = \delta)$ are \mathcal{F}_S -measurable. Here is a technical result.

Proposition 1. 1) The family $(\check{\mathcal{F}}_t) = (\mathcal{F}_{G_t})$ is increasing.

2) For each (\mathcal{F}_t) -stopping time T , one has $\mathcal{F}_{G_T} \subset \check{\mathcal{F}}_T \subset \mathcal{F}_T$.

Proof. 1) Let (Z_t) be a right continuous adapted process. The process (Z_{G_t}) is adapted, since $G_t \leq t$, and right continuous, since (G_t) is increasing and right continuous. Therefore (Z_{G_t}) is well measurable for each well measurable (Z_t) . Let $u \geq 0$, set $Z_t^u = Z_{G_t \wedge u}$; the process $(Z_t^u)_{t \geq 0}$ is still well measurable and the equality $Z_{G_u} = Z_{G_u}^u$, $u \leq v$, shows that $\mathcal{F}_{G_u} \subset \mathcal{F}_{G_v}$ if $u \leq v$.

2) Let T be a stopping time of the increasing family $(\check{\mathcal{F}}_t)$. We have already observed that, if (Z_t) is (\mathcal{F}_t) -well measurable, (Z_{G_t}) is (\mathcal{F}_t) -well measurable; therefore $\mathcal{F}_{G_t} \subset \mathcal{F}_t$ for all t , and (Z_{G_t}) is $(\check{\mathcal{F}}_t)$ -well measurable, which implies that Z_{G_T} is $\check{\mathcal{F}}_T$ -measurable (set $Z_\infty = 0$) and that $\mathcal{F}_{G_T} \subset \check{\mathcal{F}}_T$.

Remark. The variable A_T is $\check{\mathcal{F}}_T$ -measurable, since (A_t) is right continuous and $(\check{\mathcal{F}}_t)$ -adapted, but not necessarily \mathcal{F}_{G_T} -measurable.

Theorem 2. For all positive \mathcal{F}^* -measurable functions f on Ω and all $(\check{\mathcal{F}}_t)$ -stopping times T one has

$$\hat{P}^{A_T, X_{G_T}}(1) > 0 \quad \text{a.s. on } \{G_T < T < \infty\}, \quad (2)$$

$$E^* [f \circ \theta_{G_T} | \check{\mathcal{F}}_T] = P^{A_T, X_{G_T}}(f) \quad \text{a.s. on } \{T < \infty\}, \quad (3)$$

where we set

$$\begin{aligned} \hat{P}^{a, x}(f) &= \hat{P}^x(f 1_{(R > a)}) \quad \text{if } a > 0, \\ P^{a, x}(f) &= \frac{\hat{P}^{a, x}(f)}{\hat{P}^{a, x}(1)} \quad \text{if } a > 0 \quad \left(\frac{0}{0} = 0 \right), \\ P^{0, x}(f) &= P^x(f). \end{aligned}$$

Remarks. a) If $a > 0$, $\hat{P}^{a, x}(1) = \hat{P}^x\{R > a\} < \infty$, as already observed.

b) $(a, x) \rightarrow \hat{P}^{a, x}(f)$ is measurable on $]0, \infty[\times E$ due to the right continuity in a , and $(a, x) \rightarrow P^{a, x}(f)$ is measurable on $[0, \infty[\times E$.

c) X_{G_T} is \mathcal{F}_{G_T} -measurable. Therefore equality (3) implies the following:

$$E^* [f \circ \theta_{G_T} | A_T, \mathcal{F}_{G_T}] = P^{A_T, X_{G_T}}(f) \quad \text{a.s. on } \{T < \infty\}. \quad (3')$$

d) Let us set, for a positive h on E

$$\begin{aligned} \hat{Q}_a(x, h) &= \hat{P}^x(h \circ X_a I_{(R > a)}), \quad a > 0, \\ q_a(x, h) &= \frac{\hat{Q}_a(x, h)}{\hat{Q}_a(x, 1)} \quad \text{if } a > 0 \quad \left(\frac{0}{0} = 0 \right), \\ &= h(x) \quad \text{if } a = 0, \end{aligned}$$

like in [3]. We have $\hat{Q}_a(x, h) = \hat{P}^{a, x}(h \circ X_a)$, $q_a(x, h) = P^{a, x}(h \circ X_a)$ and formula (3) implies

$$E^*[h \circ X_T | \check{\mathcal{F}}_T] = q_{A_T}(X_{G_T}, h) \quad \text{a.s. on } \{T < \infty\}.$$

Hence Theorem 2 is a direct generalization of Theorem (7.10) of [3].

Proof. Condition (2) was already established in [3], in the form $\hat{Q}_{A_T}(X_{G_T}, 1) > 0$ a.s. on $\{G_T < T < \infty\}$. On the set $\{G_T = T < \infty\}$, equality (3) follows from $\check{\mathcal{F}}_T \subset \mathcal{F}_T$ and from the Markov property at time T . One can reduce the proof of (3) on the set $\{G_T < T < \infty\}$ to the case $T = t$ constant by the argument of [3]. So we have to establish that, for each positive well measurable Z ,

$$E^*[Z_{G_t} f \circ \theta_{G_t} I_{\{G_t < t\}}] = E^*[Z_{G_t} P^{A_t, X_{G_t}}(f) I_{\{G_t < t\}}]. \quad (4)$$

Let us firstly give the following extension of formula (1): if we set $G^0 = G$ if $R = 0$, $G \cup \{0\}$ if $R > 0$ and $dB_t^0 = I_{\{R > 0\}} \varepsilon_0(dt) + dB_t$, one has

$$E^*\left[\sum_{s \in G^0} Z_s F(s, X_s, \theta_s)\right] = E^*\left[\int_{[0, \infty[} Z_s(\omega) dB_s^0(\omega) \int F(s, X_s(\omega), \omega') \hat{P}^{X_s(\omega)}(d\omega')\right] \quad (5)$$

for each positive universally measurable function F on $\mathbb{R}_+ \times E \times \Omega$.

Now the basic observation is that $s = G_t < t$ iff $s \in G^0$ and $0 < t - s < R \circ \theta_s$. Therefore the left side of (4) equals

$$E^*\left[\int_{[0, t[} Z_s \hat{P}^{t-s, X_s}(f) dB_s^0\right]$$

(take $F(s, x, \omega) = f(\omega) I_{\{0 < t-s < R\}}(\omega)$ in (5)) and the right side of (4) equals

$$E^*\left[\int_{[0, t[} Z_s P^{t-s, X_s}(f) \hat{P}^{t-s, X_s}(1) dB_s^0\right]$$

(take $F(s, x, \omega) = P^{t-s, x}(f) I_{\{0 < t-s < R\}}(\omega)$ in (5)). This yields (4).

Corollary 1. *If $a > 0$, let $T^a = \inf\{t: A_t > a\}$, $G^a = G_{T^a}$. Then*

$$E^*[f \circ \theta_{G^a} | \check{\mathcal{F}}_{G^a}] = P^{a, X_{G^a}}(f) \quad \text{a.s. on } \{G^a < \infty\} \quad (6)$$

for all positive $\check{\mathcal{F}}^*$ -measurable f .

Proof. T^a is an $(\check{\mathcal{F}}_t)$ -stopping time since

$$\{T^a \leq t\} = \{A_t = a\} \cup \left(\bigcup_{\substack{r < t \\ r \text{ rational}}} \{A_r > a\}\right)$$

and (A_t) is $(\check{\mathcal{F}}_t)$ -adapted. Notice also that $A_{T^a} = a$ on $\{G^a < \infty\} = \{T^a < \infty\}$. Therefore applying (3) to $T = T^a$ yields (6).

Remark. (6) can also be established directly by observing that $s = G^a < \infty$ iff $s \in G^0$, $R \circ \theta_s > a$ and $s \leq T^a$, and by using equality (5) with $Z_t = I_{\{t \leq T^a\}}$, like in the proof of (4).

From Theorem 2 one obtains the distribution of the excursion straddling T , conditionally on A_T, X_{G_T} (this was proved by Gettoor for $T = t$).

Corollary 2. *Let T be an $(\overline{\mathcal{F}}_T)$ -stopping time. If $0 < a < r$, $x \in E$, $0 < t_1 < \dots < t_n < r$, one has (symbolically)*

$$\begin{aligned} & P^* \{ \theta_{G_T}^{-1} \{ X_{t_1} \in dy_1, \dots, X_{t_n} \in dy_n, R > r \} \mid A_T = a, X_{G_T} = x \} \\ &= \widehat{Q}_a(x, 1)^{-1} \widehat{Q}_{t_1}(x, dy_1) Q_{t_2-t_1}(y_1, dy_2) \dots Q_{t_n-t_{n-1}}(y_{n-1}, dy_n) P^{y_n} \{ R > r - t_n \} \\ &= P^* \{ \theta_{G^a}^{-1} \{ X_{t_1} \in dy_1, \dots, X_{t_n} \in dy_n, R > r \} \mid X_{G^a} = x \} \end{aligned}$$

where

$$\widehat{Q}_t(x, dy) = \widehat{P}^x \{ X_t \in dy, R > t \}, \quad Q_t(x, dy) = P^x \{ X_t \in dy, R > t \}.$$

Proof. By the Markov property of the measure \widehat{P}^x , one has

$$\begin{aligned} & P^{a, x} \{ X_{t_1} \in dy_1, \dots, X_{t_n} \in dy_n, R > r \} \\ &= \widehat{P}^{a, x}(1)^{-1} \widehat{P}^x \{ X_{t_1} \in dy_1, R > t_1 \} P^{y_1} \{ X_{t_2-t_1} \in dy_2, R > t_2 - t_1 \} \\ &\quad \dots P^{y_{n-1}} \{ X_{t_n-t_{n-1}} \in dy_n, R > t_n - t_{n-1} \} P^{y_n} \{ R > r - t_n \}. \end{aligned}$$

Hence Corollary 2 follows from Theorem 2 and Corollary 1.

IV. Conditioning on $(A_T, \overline{\mathcal{F}}_{G_T-})$

For each positive random variable S , $\overline{\mathcal{F}}_{S-}$ denotes the σ -field of the sets $A \in \mathcal{F}$ such that $I_A = Z_S$ on $\{S < \infty\}$ for some predictable process Z . For example S is $\overline{\mathcal{F}}_{S-}$ -measurable. From now on we shall assume for simplicity that (X_t) is a *Hunt process*. Then X_{S-} is $\overline{\mathcal{F}}_{S-}$ -measurable (we set $X_{\infty-} = \delta$).

In the sequel we shall also assume that M has (a.s.) no isolated point, which implies the equivalence $G_t \leq u \Leftrightarrow t \leq D_u$ for all $t, u \geq 0$. It follows that the process (Z_{G_t}) is predictable for all predictable Z . In fact if $Z_t = I_{\{t \leq U\}}$, where U is a stopping time, $Z_{G_t} = I_{\{t \leq D_U\}}$ is a predictable process, since D_U is a stopping time. This observation leads to the following result, whose proof is entirely analogous to the proof of Proposition 1.

Proposition 2. 1) *The family $(\overline{\mathcal{F}}_t) = (\overline{\mathcal{F}}_{G_t-})$ is increasing.*

2) *For each $(\overline{\mathcal{F}}_T)$ -stopping time T , one has $\overline{\mathcal{F}}_{G_T-} \subset \overline{\mathcal{F}}_T \subset \mathcal{F}_T$.*

In this paragraph we shall further assume that there exists a “predictable exit system” $(\overline{B}, \overline{P})$: \overline{B} is a predictable random measure and \overline{P} is a kernel from (E, \mathcal{E}^*) to (Ω, \mathcal{F}^*) such that for all positive predictable Z and \mathcal{F}^* -measurable f

$$E^* \left[\sum_{s \in G} Z_s f \circ \theta_s \right] = E^* \left[\int_0^\infty Z_s \overline{P}^{X_s}(f) d\overline{B}_s \right].$$

Furthermore $\overline{P}^x(R=0) = 0$ and $\overline{P}^x(1 - e^{-R}) \leq 1$ for all x , and the process $(X_t)_{t > 0}$ is strong Markov with respect to (P_t) for each measure \overline{P}^x . The existence of such a system is not always satisfied as Gettoor and Sharpe showed in [2]. The reader will find sufficient conditions for this existence in the appendix.

Under all previous assumptions, one has the following results (the proofs are analogous to the proofs of paragraph 3 and left to the reader).

Theorem 3. Let T be an $(\overline{\mathcal{F}}_t)$ -stopping time. For all positive \mathcal{F}^* -measurable f , $\bar{P}^{A_T, X_{G_T-}}(1) > 0$ a.s. on $\{0 < G_T < T < \infty\}$ and

$$E^*[f \circ \theta_{G_T} | \overline{\mathcal{F}}_T] = \tilde{P}^{A_T, X_{G_T-}}(f) \quad \text{a.s. on } \{0 < G_T < T < \infty\} \quad (7)$$

where we set

$$\begin{aligned} \bar{P}^{a, x}(f) &= \bar{P}^x(f I_{(R > a)}) \quad \text{if } a > 0, \\ \tilde{P}^{a, x}(f) &= \frac{\bar{P}^{a, x}(f)}{\bar{P}^{a, x}(1)} \end{aligned}$$

Remarks. A_T is $\overline{\mathcal{F}}_T$ -measurable since (A_t) is $(\overline{\mathcal{F}}_t)$ -adapted and right continuous. Proposition 2, equality (7) and the fact that X_{G_T-} is \mathcal{F}_{G_T-} -measurable imply

$$E^*[f \circ \theta_{G_T} | A_T, \mathcal{F}_{G_T-}] = \tilde{P}^{A_T, X_{G_T-}}(f) \quad \text{a.s. on } \{0 < G_T < T < \infty\}. \quad (7')$$

Corollary 3. With the notations of Corollary 1, one has

$$E^*[f \circ \theta_{G^a} | \mathcal{F}_{G^a-}] = \tilde{P}^{a, X_{G^a-}}(f) \quad \text{a.s. on } \{0 < G^a < \infty\}. \quad (8)$$

Remarks. 1) Corollary 3 follows from Theorem 3 by observing that T^a is an $(\overline{\mathcal{F}}_t)$ -stopping time, and that $A_{T^a} = a$ on $\{G^a < \infty\}$. It can also be established directly along the lines of the remark following Corollary 1 (notice that, $Z_t = I_{\{t \leq T^a\}}$ is predictable). This method has the advantage that it does not require that M has no isolated point.

2) In the case where $M = \{t: X_t = x_0\}$, $X_{G^a-} = x_0$ a.s. on $\{0 < G^a < \infty\}$, since M has no isolated point; therefore the strict past of G^a and the future of G^a are independent, as a consequence of Corollary 3.

Finally Corollary 2 can be rephrased for an $(\overline{\mathcal{F}}_t)$ -stopping time by replacing X_{G_T} , X_{G^a} , \hat{Q}_t by X_{G_T-} , X_{G^a-} , \hat{Q}_t , where $\hat{Q}_t(x, dy) = \bar{P}^x\{X_t \in dy, R > t\}$.

V. Appendix

In this section, we shall give sufficient conditions for the existence of a “predictable exit system”. We assume for simplicity that (X_t) is a Hunt process, and that there exists a universally measurable subset S of F^c such that if $s \in G$, $X_{s-} \in F^c$, $X_s \in F^c$ then

$$X_{s-} \neq X_s \Leftrightarrow X_{s-} \in S.$$

Then one has the following.

Theorem 4. Under the previous assumptions, there exists a predictable exit system (\bar{B}, \bar{P}) such that $\bar{P}^x(R=0)=0$, $\bar{P}^x(1-e^{-R}) \leq 1$ and $(X_t)_{t>0}$ is strong Markov with semi-group (P_t) relative to \bar{P}^x , for all x .

Proof. The set G is the union of the following sets:

$$\begin{aligned} {}^1G &= \{s \in G: X_s \in F\}, \\ {}^2G &= \{s \in G: X_{s-} \in F, X_s \in F^c\}, \end{aligned}$$

$$\begin{aligned}
 {}^3G &= \{s \in G: X_{s-} \in S, X_s \in F^c\}, \\
 {}^4G &= \{s \in G: X_{s-} \in S^c \cap F^c, X_s \in F^c\}.
 \end{aligned}$$

It is known that ${}^1G \cap [T] = \emptyset$ a.s. for each stopping time T , whereas $G \setminus {}^1G$ is a countable union of graphs of stopping times.

For each positive, bounded, \mathcal{F}^* -measurable function f , define

$${}^i u_f = E^* \left[\sum_{s \in {}^i G} e^{-s} (1 - e^{-R_s}) f \circ \theta_s \right], \quad i = 1, 2, 3.$$

For each predictable stopping time T , one has ${}^i G \cap [T] = \emptyset$ a.s., $i = 1, 2, 3$, since $X_{s-} \neq X_s$ for $s \in {}^2G \cup {}^3G$. Hence the functions ${}^i u_f$ are regular 1-potentials. By the argument of [3], Theorem (4.1), there exist continuous additive functionals ${}^i K$ and kernels ${}^i N$ such that

$${}^i u_f = E^* \left[\int_0^\infty e^{-s} {}^i N^{X_s}(f) d {}^i K_s \right], \quad i = 1, 2, 3$$

for all f . 1K and 2K are carried by F , whereas 3K is carried by S . Let ${}^1d, {}^2d$ be densities of ${}^1K, {}^2K$ with respect to ${}^1K + {}^2K$ and set

$$\begin{aligned}
 \bar{P}^x(f) &= \sum_{i=1,2} {}^i d(x) {}^i N^x(f/1 - e^{-R}) && \text{if } x \in F, \\
 &= {}^3 N^x(f/1 - e^{-R}) && \text{if } x \in S, \\
 &= P^x(f) && \text{if } x \in S^c \cap F^c, \\
 d\bar{B}_t &= \sum_{i=1,2,3} d {}^i K_t + \sum_{s \in {}^4 G} \varepsilon_s(dt).
 \end{aligned}$$

The random measure \bar{B} is predictable, since $X_{s-} = X_s$ for $s \in {}^4G$. By the same arguments as in [3], one shows that, except for x in a \bar{B} -null set N_0 , one has $\bar{P}^x\{R=0\}=0$, $\bar{P}^x(1 - e^{-R}) \leq 1$ and the strong Markov property of \bar{P}^x with respect to (P_t) . If we change the definition of \bar{P} by setting $\bar{P}^x = 0$, if $x \in N_0$, (\bar{B}, \bar{P}) is a predictable exit system with all the desired properties.

Remark. By construction $E^* \left[\int_0^\infty e^{-s} \bar{P}^{X_s}(1 - e^{-R}) d\bar{B}_s \right] \leq 1$. Therefore the process $\left(\int_0^t \bar{P}^{X_s}(1 - e^{-R}) d\bar{B}_s \right)$ is a predictable additive functional with 1-potential ≤ 1 .

References

1. Gettoor, R.K.: Excursions of a Markov process. [To appear in Ann. Probability]
2. Gettoor, R.K., Sharpe, M.J.: Last exit decompositions and distributions. Indiana Univ. Math. J., **23**, 377-404 (1973)
3. Maisonneuve, B.: Exit systems. Ann. Probability, **3**, 309-411 (1975)