

$$\begin{aligned}
& u_x^+(c,t) + \alpha u^+(c,t) = -f(t) \\
& \qquad\qquad\qquad \qquad\qquad\qquad \qquad\qquad\qquad \} \text{ te } [0,T]; \\
& u^+(s(t),t) = 0 \\
& s(0) = b \\
(1.3) \quad & \{ \dot{s}(t) = -\delta_1 u_x^-(s(t),t) + \delta_2 u_x^+(s(t),t), \text{ te } [0,T].
\end{aligned}$$

Here $a < b < c$, $T > 0$, α_1 , α_2 , α , δ_1 and δ_2 are given positive constants.

The problem of finding a triple $\{u^-(\dots), u^+(\dots), s(\dots)\}$ which satisfies (1.1) - (1.3) is known in mathematical physics as the two-phase Stefan problem.

Let flux $f(t)$ be determined from the control equation

$$(1.4) \quad B \dot{f}(t) + f(t) = p(t), \text{ t} > 0, \text{ f}(0) = 0,$$

where $p(t)$ is the decision of controller C to which the system reacts with a certain time lag $\theta \geq 0$ (we suppose that $s(t) = s_0(t)$, $t \in [-\theta, 0]$, is observed previously).

Moreover, there is a region desirable for $s(\dots)$:

$$s_1(t) \leq s(t) \leq s_2(t),$$

$s_1(\dots)$ and $s_2(\dots)$ being given such that

$$(A1) \quad a < s_1(t) < s_2(t) < c,$$

$$s_2(t) - s_1(t) \geq \mu > 0, \text{ t} \geq 0.$$

Finally, let S^0 (S^+) be the set of passive decisions (the set of active decisions) of the controller C :

$$(A2) \quad 0 \in S^0 \subseteq [0,1], \text{ 1} \in S^+ \subseteq [0,1],$$

$$p^0 \in S^0, p^+ \in S^+ \rightarrow p^0 < p^+.$$

So, observing the evolution of $s(t)$, the controller C takes at time moment t

a passive decision $p^0 \in S^0$ if $s(t-\theta) = s_1(t-\theta)$
and the decision was active just before;

an active decision $p^+ \in S^+$ if $s(t-\theta) = s_2(t-\theta)$
and the decision was passive just before.

In the case $S^0 \equiv \{0\}$, $S^+ \equiv \{1\}$ and without any time lag

this hysteresis behaviour has been considered by Hoffmann and Sprekels (1984). In the present paper we shall first develop the results of these authors to the case where S^0 and S^+ have only restriction (A2). Our special attention will be paid to estimating explicitly the time interval of realization of a decision change regime and the number of decision changes by a taken decision change regime and the parameters of the system. Then, in §3, we shall use the freedom in choosing decision change regimes to minimize the maximal deviation of inter-face $s(\cdot)$ from some given ideal one, or to minimize the total cost for going out from the desirable zone.

§2. REAL-TIME DECISION CHANGE REGIME

It was proved in Hoffmann and Sprekels (1984) that for every control $p(t)$, $0 \leq p(t) \leq 1$, $t \in [0, T]$, the problem (1.1)-(1.4) has a unique solution $\{u^-(p(\cdot); \cdot, \cdot), u^+(p(\cdot); \cdot, \cdot), s(p(\cdot); \cdot)\}$ and moreover, $s(p(\cdot); \cdot)$ oscillates at most N^* times between $s_1(t)$ and $s_2(t)$ in $[0, T]$, N^* being finite and independent of $p(\cdot)$. In what follows it will be shown that N^* can be determined a priori.

From now on we shall write "regime" instead of "decision change regime", for short.

Definition 2.1. A sequence $p := \{p_i, i = 1, 2, \dots, N^*\}$ is called a pre-regime provided that

$$p_0 \in \begin{cases} S^0 & \text{if } s(-\theta) \leq s_1(-\theta), \\ S^+ & \text{if } s(-\theta) \geq s_2(-\theta), \\ S^0 \cup S^+ & \text{if } s_1(-\theta) < s(-\theta) < s_2(\theta), \end{cases}$$

and

$$p_{k+1} \in \begin{cases} S^0 & \text{if } p_k \in S^+, \\ S^+ & \text{if } p_k \in S^0, \end{cases}$$

$$k = 0, 1, \dots, N^* - 1.$$

Let p be a given pre-regime.

Definition 2.2. A sequence $r := \{p_k, t_k, k = 0, 1, \dots, N \leq N^*\}$ is said to be a regime (or a realization of pre-regime p) if the following relations hold:



$$\begin{aligned}
t_0 &:= 0, \\
t_{k+1} & \begin{cases} \inf M_{k+1} & \text{if } M_{k+1} \neq \varphi, \\ +\infty & \text{otherwise,} \end{cases} \\
M_{k+1} &:= \{t \in (t_k, T) : s(p; t-\theta) = \\
& \quad \begin{cases} s_1(t-\theta) & \text{if } p_k \in S^+ \\ s_2(t-\theta) & \text{if } p_k \in S^0 \end{cases} \\
s(p; t) &:= s(p(\cdot); t), \\
p(t) = p(r; t) &:= p_k, \quad t_k \leq t < t_{k+1}, \\
k &= 0, 1, 2, \dots, N.
\end{aligned}$$

Consider a regime $r = \{r_i, t_i, i = 0, 1, \dots, N\}$. Our goal is to establish some a priori estimates.

Denote by $f(r; \cdot)$ the unique solution of (1.4) with $p(\cdot) = p(r; \cdot)$. One can see that $f(r; \cdot)$ is monotone in every interval (t_k, t_{k+1}) and the following inequalities hold:

$$(2.1) \quad \min \{f_k(r), f_{k+1}(r)\} \leq f(r; t) \leq \max \{f_k(r), f_{k+1}(r)\} = \begin{cases} f_k(r) & \text{if } f_k(r) \geq p_k, \\ f_{k+1}(r) & \text{if } f_k(r) < p_k, \quad t_k \leq t < t_{k+1}, \end{cases}$$

where

$$f_k(r) := f(r; k) = \exp(-t_k/B) \sum_{i=0}^{k-1} [\exp(t_{i+1}/B) - \exp(t_i/B)];$$

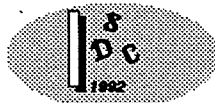
$$(2.2) \quad 0 \leq F_k(r) \leq f(r; t) \leq F^k(r) \leq 1,$$

$$\text{where } F_k(r) := \min \{f_i(r), i = 0, 1, \dots, k+1\} \leq f(r; t) \leq \max \{f_i(r), i = 0, 1, \dots, k+1\} =: F^k(r), \quad t \in [0, t_{k+1}];$$

$$(2.3) \quad 0 \geq u_+(r; x, t) \geq -\tau_k \geq -\tau, \quad (x, t) \in \text{Cl} \Omega_k^+(s(r; \cdot))$$

where

$$\begin{aligned}
\tau_k &:= \max \{\alpha_3 F^k(r); |\Phi(x)|, b \leq x \leq c\}, \\
\tau &:= \max \{\alpha_3; |\Phi(x)|, b \leq x \leq c\}, \\
\alpha_3 &:= \min \{1, 1/\alpha\},
\end{aligned}$$



$$\Omega_k^+(s(r; \cdot)) := \{(x, t) \in R^2: s(r; t) < x \\ < c, 0 < t < t_{k+1}\},$$

$k = 0, 1, \dots, N;$

$$(2.4) \quad |\dot{s}(r; t)| \leq \max \{ \delta_1 \phi_1', i = 1, 2; \delta_2 F^k(r); \delta_2 \alpha \tau_k \} \\ =: E_k(r), t \in [0, t_{k+1}], k = 0, 1, \dots, N;$$

$$(2.5) \quad |\dot{s}(r; t)| \leq \max \{ \delta_1 \phi_1', i = 1, 2; \delta_2; \delta_2 \alpha \tau \} =: E, \\ p(\cdot), t \in [0, T],$$

where

$$\phi_1' := \max \{ |\phi'(x)|, a \leq x \leq b \}, \\ \phi_2' := \max \{ |\phi'(x)|, b \leq x \leq c \}.$$

It is worth noting that relations (2.4)-(2.5) enable us to estimate $\dot{s}(t)$ a priori and explicitly by a given regime and by the parameters of the system. To continue we need the function $\phi(\cdot)$ to have the following property:

(A3) $a-b < \Gamma < c-b$ where

$$\Gamma := (\delta_1/\alpha_1) \int_a^b \phi(x) dx + (\delta_2/\alpha_2) \int_b^c \phi(x) dx.$$

We are going now to formulate the main results of the section.

Result 1 (concerning the time interval of realization of decision change pre-regimes). Every decision change pre-regime p is realized uniquely by a decision change regime

$r := \{r_i, t_i, i = 0, 1, \dots, N\}, t_N < T(p) \leq t_{N+1}$, where $N = N(p)$

is the minimal from non-negative integers k satisfying the following two conditions (i) - (ii):

$$(i) \quad (2.6) \quad \sigma_{1, k-1}(r) := \sum_{j=0}^{k-1} \omega_j(r) \max\{f_j(r), f_{j+1}(r)\} \\ < (1/\delta_2)(\Gamma - a + b),$$

$$(2.7) \quad \sigma_{2,k-1}(r) := \sum_{j=0}^{k-1} \omega_j(r) [\alpha r_j - \min\{f_j(r), f_{j+1}(r)\}]$$

$$< (1-\delta_2)(c-b-\Gamma),$$

$$\omega_j(r) := t_{j+1} - t_j;$$

(ii) at least one of the following inequalities holds:

$$(2.8) \quad \sigma_{1,k}(r) \geq (1/\delta_2)(\Gamma-a+b),$$

$$(2.9) \quad \sigma_{2,k}(r) \geq (1-\delta_2)(c-b-\Gamma),$$

(by definition $\sigma_{1,-1} = \sigma_{2,-1} = 0$, so (i)-(ii) hold automatically for $k = 0$).

Moreover, the time interval $[0, T(p)]$ of realization of decision change pre-regime p is defined as follows:

$$(iii) \quad T(p) = T_1(p)$$

$$\text{where } T_1(p) := \inf\{t \in [t_k, t_{k+1}]: \sigma_{1,k-1}(r) + (t-t_k)\max\{f_k(r), f_{k+1}(r)\} = (1/\delta_2)(\Gamma-a+b),$$

if only (2.8) takes place;

$$T(p) = T_2(p)$$

$$\text{where } T_2(p) := \inf\{t \in [t_k, t_{k+1}]: \sigma_{2,k-1}(r) + (t-t_k)[\alpha r_k - \min\{f_k(r), f_{k+1}(r)\}] = (1-\delta_2)(c-b-\Gamma),$$

if only (2.9) takes place;

$$T(p) = \min\{T_1(p), T_2(p)\},$$

if both (2.8) and (2.9) take place.

Thanks to Result 1 every decision change pre-regime p can be identified with its unique realization $r = \{r_i, t_i, i = 0, 1, \dots, N = N(p)\}$, as we shall do henceforth.

Set for $0 \leq t_* < t^* \leq T$

$$\mu(t_*, t^*) := \min\{s_2(t) - s_1(t), t_* - \theta \leq t \leq t^* - \theta\},$$

$$\mu(t^*) := \mu(0, t^*), \quad \mu_k(p) := \mu(t_{k+1});$$

$$D_i(t_*, t^*) := \max\{|s_i(t), t_* - \theta \leq t \leq t^* - \theta\}, i=1, 2,$$

$$D(t_*, t^*) := \min\{D_i(t_*, t^*), i=1, 2\},$$

$$D^i(t^*) := D^i(0, t^*), \quad i=1, 2,$$

$$D(t^*) := \min\{D^i(t^*), i=1, 2\},$$



$$D_k(r) := D(t_{k+1}), \quad D := D(T)$$

and let $N(t_*, t^*)$ denote the number of decision changes in the interval $[t_*, t^*]$.

Result 2 (concerning the number of decision changes and their frequency). The following estimates hold:

$$(2.10) \quad N(t_*, t^*) \leq 1 + \frac{t^* - t}{\mu(t_*, t^*)} [E + D(t_*, t^*)],$$

$$0 \leq t_* < t^* \leq T;$$

$$(2.11) \quad \omega_k(r) \geq \frac{\mu_k(r)}{E_k(r) + D_k(r)}, \quad k = 0, 1, \dots, N(r).$$

§3. OPTIMAL DECISION CHANGE REGIME

It follows from the results obtained in §2 that the control to the system being considered is completely determined by the decision change pre-regime p and it remains to choose the latter. In the sequel we shall try to optimize this operation, i.e. to use the freedom in choosing p to minimize the maximal deviation of $s(\cdot)$ from a given ideal interface $\sigma(\cdot)$:

$$(3.1) \quad J_1(p) := \max \{ |s(p; t) - \sigma(t)|, \quad 0 \leq t \leq T(p) \}$$

or the total cost which must be paid for leaving the desirable zone:

$$(3.2) \quad J_2(p) := \int_0^{T(p)} J_{p,s}(t) dt$$

where

$$J_{p,s}(t) := \begin{cases} \pi_1(s_1(t) - s(t)) & \text{if } s(t) < s_1(t), \\ 0 & \text{if } s_1(t) \leq s(t) \leq s_2(t), \\ \pi_2(s(t) - s_2(t)) & \text{if } s(t) > s_2(t), \end{cases}$$

π_1 and π_2 being given positive coefficients.



Denote by PR the set of all possible decision change pre-regimes corresponding to a taken pair (S^0, S^+) . Thus we are led to the following optimization problem:

(OP) Minimize $J_1(p)$ (or $J_2(p)$) subject to $p \in PR$.

Result 3. Every optimizing sequence of decision change pre-regimes $\{p^{(n)}, n=1,2,\dots\}$, i.e.

$$J_1(p^{(n)}) \rightarrow m := \min \{J_1(p), p \in PR\}$$

as $n \rightarrow \infty$, contains a subsequence $\{p^{(k)}, k=1,2,\dots\}$ which converges to an optimal decision change pre-regime

p_0 :

$$p^{(k)} \rightarrow p_0 \text{ as } k \rightarrow \infty, J_1(p_0) = m.$$

In practice this result allows us to orient our decision to an optimal one and to find an approximate solution.

Let us omit the mathematical proof of Results 1-3, because it needs much more than one page.

CONCLUSION REMARK. The present paper summarizes a qualitative study of the problem. In practice, especially if we confine ourselves to some finite sets S_0 and S_1 , the technique applied facilitates the analysis of simulation and gaming process to obtain, on the one hand, the complete information about possible decision change pre-regimes, and on the other hand, an optimal one.

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