Application of Chance-Constrained Programming to Solution of the So-called "Savings and Loan Association" Type of Problem
Application of Chance-Constrained Programming to Solution of the So-called "Savings and Loan Association" Type of Problem

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This is the third in a series of papers that deal with the problem of finding the optimal decision rules for n-period chance-constrained programming models. The first two papers in this series are entitled "Optimal Decision Rules for the E Model of Chance-Constrained Programming" and "Optimal Decision Rules for the Triangular E Model of Chance-Constrained Programming."

This paper shows how the mathematical results developed in the above-mentioned papers can be successfully applied to analysis of problems in planning under uncertainty. Such a problem has been formulated in the literature as the "Savings and Loan Problem" although the theory has relevance in many other situations, some directly concerned with military problems. The same model could be used as an aid to decision making in any situation where decisions must be made periodically over an n-period horizon. This includes such problems as planning research and development projects. The particular instance chosen to illustrate the theory was selected because earlier work has been done on this problem, and it is desirable to be able to compare the results contained here with those previously obtained by other authors.

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Application of Chance-Constrained Programming to Solution of the So-called "Savings and Loan Association" Type of Problem
ABSTRACT

This paper contains an application of chance-constrained programming to a problem in financial planning. In particular, the problem is one of planning for liquidity in a savings and loan association first discussed by Charnes and Thore. The optimal rules for this problem are found and compared with the optimal linear rules given by Charnes and Thore. The discontinuous nature of the optimal rules is discussed from economic and control theory viewpoints.
Introduction

This paper gives an application of the method of chance-constrained pro-
gramming to a problem in financial planning. Such problems are particularly
well suited to analysis by chance-constrained programming because they deal
with questions of planning in the face of an uncertain future and are such that
the chance elements enter into both the objective function and the constraints.
These are precisely the kinds of problems that chance-constrained program-
ing was designed to handle.

In most of the work done thus far in chance-constrained programming,
problems have been solved by converting them into equivalent deterministic
and, in general, nonlinear problems. This transformation has been accomplished
chiefly by restricting the class of admissible decision rules to the class of
linear decision rules.¹ ³

Even purely deterministic models have proved useful in the analysis of
problems of financial budgeting. Specifically, the work of Charnes, Cooper,
and Miller⁴ on the problem of the costing of funds in a simple warehouse model
is mentioned. There in addition to the usual set of constraints in a warehouse
model, is added a simple deterministic liquidity constraint requiring that the
planned accumulation of cash be greater than or equal to the difference between
the minimum cash balance considered acceptable and the initial cash holdings.

The specific problem presented here is a two-period problem of planning
for liquidity in a savings and loan association. The model used was first dis-
cussed by Charnes and Thore.⁵ Their assumption was that the admissible class
of decision rules was the class of linear rules. The deterministic equivalent
was found and the resulting nonlinear programming problem was solved to get
the optimal decision rules in terms of the parameters of the model.

The results presented in this paper differ from these in one extremely
significant way. It is no longer assumed that the class of decision rules under
consideration is linear. In particular, the decision rules will be arbitrary func-
tions of the random observations and decision rules of previous periods, sub-
ject only to certain regularity conditions that permit the use of the isoperimetric
theory of the calculus of variations. Thus the admissible class of decision rules
in this formulation is much larger than the class of only linear rules.

In Charnes and Kirby⁶ necessary conditions were developed for decision
rules, restricted only to the class described above, to be optimal for the type
of problem to be considered here. Most of the mathematical results in this
paper will be based on the results contained in this paper.

It is shown that under certain circumstances the optimal decision rule in
this class of feasible rules is, in fact, the optimal linear rule. The result is
of particular importance because it shows that good reason exists, other than
the fact that it is mathematically more manageable, for limiting oneself to find-
ing the optimal linear decision rule.
It will also be shown that in all other cases (i.e., where the rule is not linear) the optimal rule is discontinuous. This result may seem to be highly surprising at first but by economic arguments it will be shown that it does, in fact, appear to be reasonable.

**Description of the Model**

To provide a background for subsequent mathematical treatment of the problem a brief description of the model with some remarks about its institutional setting will be given (Ref 5, pp 5-22).

The problem is planning for liquidity in a savings and loan association. There are several reasons why the association needs to provide voluntarily for liquidity. These include the fact that the association must be ready to pay its savers the money that belongs to them on request. In other words the association must be prepared to handle withdrawals from its total savings capital.

At the same time a continuous flux exists in the mortgage-loan portfolio of an association owing to people requesting and receiving new loans, while a stream of repayments from outstanding mortgage loans brings funds back to the association. Hence in any period the association must have cash available to meet the excess of new loans over the stream of repayments.

These liquidity needs are usually met by the inflow of new savings and of mortgage-loan repayments. There are, however, other sources such as the stock liquidity, which is available by selling the US government securities contained in the association’s stock portfolio. Cash can also be obtained by borrowing from both the Federal Home Loan Bank and commercial banks.

There are also legal restrictions on the amount of liquid assets that the association must hold. At the present time the required minimum ratio of cash plus US securities to savings capital is 7 percent. Thus, if an association wants to hold liquid assets, to be available in case of liquidity needs, it must hold liquid funds above this 7 percent requirement.

The problem can then be expressed as follows: Given the different needs for liquidity of a savings and loan association and given the different sources of liquidity available, how should the association choose between these alternative sources of funds in order to provide for the given needs of liquidity?

In the very simple model discussed in this paper the essential feature is that the association holds only two types of assets: cash and loans. The association's liquid assets, cash, yield no earnings at all, and the illiquid assets, loans, yield earnings through the interest paid on them although they are perfectly illiquid. Thus the association is faced with a clear-cut choice between liquidity and profitability.

Furthermore it is assumed that the making of new loans is a decision variable that is completely under the control of the association; i.e., there is perfect competition in the loan market so that the association can make any number of loans that it wishes on current standard terms.

The stochastic part of the model arises from the random variations of the savings capital of the association. The increase in withdrawable savings capital during period \( t \) will be treated as a random variable. This random variable can assume both positive and negative values. If it is positive, then more money has been deposited than withdrawn during that period, whereas if the random variable is negative the converse is true.
As we stated earlier the association needs free cash above the legal liquidity requirement to enable it to make new loans and to provide for savers who want to withdraw part of their investment. Although it is true that the association could borrow to obtain the cash it needs, there are certain chance-constrained limits on the frequency of such borrowing. Thus the problem is one of planning for liquidity in the face of the uncertain variations in savings capital, subject to the restriction that the association does not want to borrow too often.

To describe the model more rigorously the following notation is introduced:

Let

\[ M \] = total cash held by the association
\[ L \] = mortgage loans
\[ S \] = the legal liquidity minimum
\[ S_t \] = total withdrawable savings capital at time \( t \)
\[ i \] = dividend rate
\[ \Delta S_t = S_t - S_{t-1} (1+i) \] = the increase in total withdrawable savings capital during period \( t \) above dividends credited to accounts (i.e., the net increase in savings in period \( t \))
\[ \ell_t \] = the new mortgage loans obtained by the association in period \( t \)
\[ M_t \] = cash holdings at the end of period \( t \)
\[ a_t \] = annuities received during period \( t \)

The assumption is that the frequency by which prepayments of loans and complete loan payoff occur in the total loan portfolio does not change over time. Assuming further that the association expects to charge a constant rate of interest on its loans over time, the total amount of interest the loan \( \ell_t \) will bring into the association over the life of the loan is \( R \ell_t \), where \( R \) is the interest rate on loans.

Hence the entire interest earned by loans made during periods \( t = 1, \ldots, T \) is

\[ \sum_{t=1}^{T} R \ell_t \]

It is assumed that the objective of the association is to maximize the expected interest accrued by \( \ell_t \), \( t = 1, \ldots, T \), i.e., it is desired to maximize

\[ E(\sum_{t=1}^{T} R \ell_t) \]

It is supposed that a known constant \( \lambda \) exists such that the association is required by law to keep cash balances amounting to at least 100\( \lambda \) percent of the total savings capital \( S \). Moreover, borrowing is not allowed when the association holds cash above the legal minimum, i.e., when \( M_t \geq \lambda S_t \). It is also assumed that the association cannot and does not want to borrow too often and therefore needs to check the frequency of its borrowing. This leads to the liquidity constraint

\[ \Pr(M_t \geq \lambda S_t) \leq \alpha_t \quad t = 1, \ldots, T \]

(1)

where \( \alpha_t \) is some preassigned probability.

The admissible class of decision rules—distribution of the random variables \( \Delta S_t \) will be discussed in a moment—is such that it allows \( \ell_t \) to be negative. This shall be interpreted as term borrowing on the same length of time and with the same interest charge as given by the average term loan. It is assumed
that such long-term borrowing cannot occur too frequently, therefore the constraint
\[ P(t_i > 0) \leq \beta_i, \quad i = 1, \ldots, T, \tag{2} \]

where \( \beta_i \) is a preassigned probability, is imposed on \( t_i \).

In each period \( i, \ t = 1, \ldots, T \), the budget constraint (balance-sheet identity) is
\[ M_i + f_i = M_{i-1} + a_i + \Delta S_i \tag{3} \]

This says that the sum of the cash holdings at the end of period \( t \) and the new loans made during period \( t \) must equal the sum of the initial cash held at the start of period \( t \), the annuities received during period \( t \), and the net inflow of new savings during period \( t \) above dividends credited to accounts. In brief, the left side of Eq 3 gives the use of funds in period \( t \) and the right side gives the total loanable funds that become available during period \( t \).

Thus the problem is to maximize
\[ E \left( \sum_{t=1}^{T} R_t \right) \tag{4} \]
subject to Eqs 1, 2, and 3.

Using the data given in Charnes and Thore\(^5\) the following expressions were obtained for the annuities \( a_1 \) and \( a_2 \), in the two-period problem \( T = 2 \)
\[ a_1 = 13,104 \]
\[ a_2 = 11,762 + 0.173 \ t_1 \]
in millions of dollars.

Inserting these expressions into the budget constraint Eq 3 the following expressions for \( M_1 \) and \( M_2 \) are derived:
\[ M_1 = -t_1 + \Delta S_1 + 13,104 + M_0 \]
\[ M_2 = -0.827t_1 - t_2 + \Delta S_1 + \Delta S_2 + 24,866 + M_0. \]

By substituting these expressions into the liquidity constraints Eq 1 the problem can be mathematically expressed as maximize
\[ E(R_1, R_2) \]
subject to
\[ P(- t_1 - \Delta S_1 + 13,104 + M_0 - \Delta S_1) \leq \beta_1, \]
\[ P(-0.827t_1 - t_2 - \Delta S_1 - 24,866 + M_0 - \Delta S_2) \leq \beta_2, \]
\[ P(t_1 > 0) = \beta_1, \]
\[ P(t_2 > 0) = \beta_2. \tag{5} \]

It is known from the definition of \( \Delta S_i \) that \( S_i = \Delta S_i + S_{i-1}(1+i) \). Therefore, \( S_1 = \Delta S_1 + (1+i)S_0 \) and \( S_2 = \Delta S_2 + (1+i)S_1 = \Delta S_2 + (1+i)\Delta S_1 + (1+i)^2 S_0 \), thus these expressions can be substituted into Eq 5. Moreover, since \( R > 0 \) and appears in the objective function as the coefficient of both \( t_1 \) and \( t_2 \), it can be dropped from the problem. Hence Eq 5 becomes maximize
\[ E(f_1, f_2) \]
subject to

\[ P[I - f_1 + (1 - \lambda) S_1 - M_0 - 13.104] \leq \beta_1, \]
\[ P[I - 0.827 f_1 - f_2 + (1 - \lambda) S_1 + (1 - \lambda) S_2 - M_0 - 24.866] \leq \beta_2, \]
\[ P(f_1 \geq 0) \leq \beta_1, \]
\[ P(f_2 \geq 0) \leq \beta_2. \]  \hspace{1cm} (6)

Finally defining

\[ \begin{align*}
K_1 &= (1 - \lambda) S_1 - M_0 - 13.104, \\
K_2 &= (1 - \lambda) S_2 - M_0 - 24.866,
\end{align*} \]  \hspace{1cm} (7)

and so Eq 6 can be written as

maximize \[ E(f_1, f_2) \]
subject to

\[ P[I - f_1 + (1 - \lambda) S_1 - K_1] \leq \beta_1, \]
\[ P[I - 0.82 f_1 - f_2 + (1 - \lambda) S_1 + (1 - \lambda) S_2 - K_2] \leq \beta_2, \]
\[ P(f_1 \geq 0) \leq \beta_1, \]
\[ P(f_2 \geq 0) \leq \beta_2. \]  \hspace{1cm} (8)

In the development of the model thus far nothing has been said about what distributions will be assumed for the random variables \( \Delta S_1 \) and \( \Delta S_2 \). In Charnes and Thore\(^5\) it was assumed that \( \Delta S_1, t = 1, 2, \) were independent random variables each being normally distributed with mean \( \bar{\varepsilon} \) and standard deviation \( \delta \).

Because the following mathematical results are applicable for a fairly large class of random variables consideration at this time will not be limited to only normal random variables. Instead it is assumed that \( \Delta S_1, \Delta S_2 \) are independent, continuous random variables with frequency functions \( f_t(\Delta S_1), t = 1, 2, \) and distribution functions \( F_t(\Delta S_1), t = 1, 2, \) respectively. It is assumed also that either

\[ \frac{df_2}{d \Delta S_2} \left[ F_2^{-1}(1) \right] = 0, \text{ or } F_2^{-1}(1) = \infty, \]

and either

\[ F_2^{-1}(0) = -\infty \]

or

\[ \int_0^1 f_2 \left[ F_2^{-1}(D) \right] dD = \int_0^1 f_2 \left[ F_2^{-1}(0) \right] \text{ for all } D \in (0, 1) \]

or

\[ [1 - \lambda(1 - \lambda)] \Delta S_1 - 0.827 f_1 - K_2 + (1 - \lambda) F_2^{-1}(0) = 0 \]

implies

\[ \Delta S_1 \leq F_2^{-1}(0) \text{ for all } f_1 \geq 0. \]
It is emphasized that conditions (a) and (b) do not restrict the distribution of \( \Delta S_1 \). Hence \( \Delta S_1 \) can be any continuous random variable. Since the 0 and 1 fractile points of a normal random variable are \(-\infty\) and \(+\infty\) respectively, it can be seen that \( \Delta S_1 \) and \( \Delta S_2 \), being normally distributed, are admissible random variables.

**Mathematical Treatment**

The problem now is solving Eq 8 for the optimal decision rules \( t_1' \) and \( t_2' \). As is customary in multi-period problems in chance-constrained programming it is desired that the \( j \)th period decision rule be an explicit function of the random variables whose values will have been observed at the time the \( j \)th period decision rule is put into effect, but it is not a function of the random variables of the \( j \)th or future period.\(^2\)\(^,\)\(^6\)\(^,\)\(^7\) Thus it is required that \( t_2 \) be such that \( t_2 = t_2(\Delta S_1) \), i.e., \( t_2 \) is a function of \( \Delta S_1 \) but not \( \Delta S_2 \). This agrees with the above requirement. Since the second-period decision \( t_2 \) must be made before the random variable \( \Delta S_2 \) of the second period is observed, the knowledge of the observed value of \( \Delta S_1 \) can be used in making the decision \( t_2 \). Similarly it is required that \( t_1 \) be a zero-order rule (Ref 2), since \( t_1 \) is not to be an explicit function of either \( \Delta S_1 \) or \( \Delta S_2 \).

The probability and the expectation will be computed in Eq 8 using the joint distribution of \( \Delta S_1, \Delta S_2 \). Hence

\[
P(\Delta S_1, \Delta S_2) = \int_{\delta_1}^{\delta_2} \int_{\delta_1}^{\delta_2} f_1(\Delta S_1) f_2(\Delta S_2) \, d\Delta S_1 \, d\Delta S_2,
\]

where \( A \) is the set of points for which

\[
-0.827 f_1(\delta_1) f_2(\delta_2) + (1 - \lambda)(1 - \delta_1) \Delta S_1 + (1 - \lambda) \Delta S_2 \leq K_2.
\]

This integration over \( A \) can be written in the form

\[
\int_a^b \int_{\delta_1}^{\delta_2} f_1(\Delta S_1) f_2(\Delta S_2) \, d\Delta S_1 \, d\Delta S_2,
\]

where

\[
a = \frac{K_2 - 0.827 f_1(\delta_1) f_2(\delta_2) - (1 - \lambda)(1 - \delta_1) \Delta S_1 - (1 - \lambda) \Delta S_2 \leq K_2}{(1 - \lambda)}.
\]

Thus the integration over \( A \) is equal to

\[
1 - \int_a^b \int_{\delta_1}^{\delta_2} \left\{ \frac{K_2 - 0.827 f_1(\delta_1) f_2(\delta_2) - (1 - \lambda)(1 - \delta_1) \Delta S_1 - (1 - \lambda) \Delta S_2 \leq K_2}{(1 - \lambda)} \right\} f_1(\Delta S_1) \, d\Delta S_1,
\]

where \( g, h \) are the smallest and largest values of \( \Delta S_1 \) for which \( f_1(\Delta S_1) > 0 \).\(^*\)

Thus

\[
P(-0.827 f_1(\delta_1) f_2(\delta_2) + (1 - \lambda)(1 - \delta_1) \Delta S_1 + (1 - \lambda) \Delta S_2 \leq K_2) \geq 1 - \delta_2,
\]

if and only if

\[
\int_a^b \int_{\delta_1}^{\delta_2} \left\{ \frac{K_2 - 0.827 f_1(\delta_1) f_2(\delta_2) - (1 - \lambda)(1 - \delta_1) \Delta S_1 - (1 - \lambda) \Delta S_2 \leq K_2}{1 - \lambda} \right\} f_1(\Delta S_1) \, d\Delta S_1 \geq 1 - \delta_2,
\]

\( \ast g, h \) may be \(-\infty\) and \(+\infty\), respectively.
Thus write Eq 8 as

\[
\begin{array}{c}
\text{maximize} \\
\text{subject to}
\end{array}
\]

\[
\mathcal{F}_1 \left[ \frac{f_1 + K_1}{1 - \lambda} \right] \leq 1 - \alpha_1,
\]

\[
\mathcal{F}_2 \left\{ \frac{0.827 f_1 + K_2 + \frac{f_2(b_1) - [1 - \Lambda (1 + \iota)] b_1 + K_1}{1 - \lambda}}{f_1(b_1) d(b_1) \leq 1 - \alpha_2.}
\]

(9)

where \( b_1 = \Delta S_1 \) to conform to the notation used in Charnes and Kirby.\(^6\)

In order to solve Eq 9 assume that \( t_1 \) is fixed and then proceed to find the optimal \( t_2 \), namely, \( t_2^* \), in terms of \( t_1 \) and \( b_1 \). Having found \( t_2^* \) put it into the objective function of Eq 9 to find \( E(t_1) \) in terms of \( t_1 \) and then solve the resulting problem for \( t_1^* \), the optimal value of \( t_1 \).

Thus consider the problem

\[
\begin{array}{c}
\text{maximize} \\
\text{subject to}
\end{array}
\]

\[
\mathcal{F}_2 \left\{ \frac{0.827 f_1 + f_2 - [1 - \Lambda (1 + \iota)] b_1 + K_2}{1 - \lambda} \right\} f_1(b_1) d(b_1) \leq 1 - \alpha_2.
\]

\[
P(t_1 \geq 0) \geq \beta_1,
\]

\[
P(t_2 \geq 0) \geq \beta_2.
\]

(10)

where it is assumed that \( t_1 \) is a known constant.

In order to use the results contained in Charnes and Kirby\(^6\) one further assumption is required.

There exists a partition of the set of points for which \( f_1(b_1) > 0 \) such that for each interval \([y', z']\) in the partition the following properties hold:

(a) \( \int_{y'}^{z'} f_1(b_1) d(b_1) > 0 \),

(b) \( f_2 \left\{ \frac{0.827 f_1 + f_2 - [1 - \Lambda (1 + \iota)] b_1 + K_2}{1 - \lambda} \right\} \) are continuous in \([y', z']\),

and

(c) \( f_2 \) is of constant sign in \([y', z']\).

This partition shall be referred to as the "optimal partition" of \([g, h]\).

From theorem 3 in Charnes and Kirby\(^6\), it is known that in any interval \([y', z']\), \( f_2(b_1) \) is defined by one of the following four equations:

\[
f_2(b_1) = 0,
\]

\[
f_2(b_1) - [1 - \Lambda (1 + \iota)] b_1 = 0.827 f_1 - K_2 - (1 - \Lambda) f_2^{-1}(D^-),
\]

(11)

and

(12)
\[ f_2(b_1) = (1 - \lambda) b_1 - 0.827 f_1 - K_2 + (1 - \lambda) \bar{F}_2^{-1}(0), \] (13)

\[ f_2(b_1) = (1 - \lambda) b_1 - 0.827 f_1 - K_2 + (1 - \lambda) \bar{F}_2^{-1}(1), \] (14)

where \( D' \) satisfies \( 0 < D' < 1 \) and is a constant.

From theorem 7\( ^8 \) it is known that a further necessary condition that \( t'_2(b_1) \) be defined by Eq 14 over any interval \([y', z']\) in the optimal partition is that \( \bar{F}_2^{-1}(1) < + \infty \) and satisfy

\[ \frac{df_2}{db_2} \left[ \bar{F}_2^{-1}(1) \right] \geq 0. \]

The admissible class of random variables violates this inequality when \( \bar{F}_2^{-1}(1) < + \infty \); hence it is concluded that \( t'_2(b_1) \) is not given by Eq 14 for any interval in the optimal partition.

From theorem 8\( ^6 \) a necessary condition that \( t'_2(b_1) \) be given by Eq 13 over some interval \([y', z']\) in the optimal partition is that \( \bar{F}_2^{-1}(0) > + \infty \), and either

\[ f_2(\bar{F}_2^{-1}(0)) < f_2(\bar{F}_2^{-1}(D'^-)), \]

or

\[ (1 - \lambda) b_1 - 0.827 f_1 - K_2 + (1 - \lambda) \bar{F}_2^{-1}(0) = 0 \]

for some point \( \bar{b}_1 \) in \([y', z']\).

Again, it can be seen that the admissible class of random variables fails to satisfy these conditions. Therefore \( t'_2 \) is not given by Eq 13 anywhere.

Hence it has been shown, using the results in Charnes and Kirby\( ^6 \) and the assumptions concerning the distributions of \( b_1 \) and \( b_2 \), that a necessary condition that \( t'_2(b_1) \) be optimal for Eq 10 is that there exists a partition of \([g, h]\), which is denoted by \([y', z']\), with properties (a), (b), and (c) previously described and such that in each \([y', z']\) either

\[ f_2(b_1) = 0, \]

or

\[ f_2(b_1) = (1 - \lambda) b_1 - 0.827 f_1 - K_2 + (1 - \lambda) \bar{F}_2^{-1}(D'^-). \]

where \( D' \) satisfies \( 0 < D' < 1 \) and \( D' \) is a constant.

There are two possible cases, one in which the constraint \( P(t_2 \geq 0) \geq \beta_2 \) is binding and the other where it is not binding. In the latter case theorem 6 Charnes and Kirby\( ^6 \) can be used to conclude that \( t'_2(b_1) \) is not identically 0 over any interval in our optimal partition. Thus

Lemma 1: If \( P(t_2 \geq 0) \geq \beta_2 \) is not binding in Eq 10, then for \( g \leq b_1 \leq h \),

\[ f_2(b_1) = (1 - \lambda) b_1 - 0.827 f_1 - K_2 + (1 - \lambda) \bar{F}_2^{-1}(D'^-). \]

An examination of the first constraint of Eq 10 shows that it must be satisfied as an equality at the optimum. Otherwise, \( t'_2(b_1) \) could be increased for some value of \( b_1 \), as the integrand of the constraint is a monotonic increasing function of \( t_2 \), thus increasing the value of the objective function and therefore...
contradicting the optimality of \( t_2'(b_1) \). This equality in the \( 1 - \alpha_2 \) constraint will give

\[
D^* = 1 - \alpha_2
\]

when Lemma 1 holds.

The case in which the constraint \( P(t_2' > 0) > \beta_2 \) is binding is next considered.

Let \( I = \{ j : t_2' > 0 \text{ with equality on at most a set of measure 0 in } (y_j', z_j') \} \).

Let \( J = \{ j : t_2' < 0 \text{ with equality on at most a set of measure 0 in } (y_j', z_j') \} \).

Let \( K = \{ j : t_2' = 0 \text{ in } (y_j', z_j') \} \).

Suppose that \( D' \) is known. Then the problem of determining \( y_j', z_j' \), for \( j \in I \) and \( j \in J \), can be expressed as maximize

\[
\max \left[ \sum_{j \in I} \left( \left( 1 - \lambda \right) \int_{y_j}^{z_j} f_1(b_1) \, db_1 \right) - K_2 - 0.827 f_2 \right] + \sum_{j \in J} \left[ \left( 1 - \lambda \right) \int_{y_j}^{z_j} f_1(b_1) \, db_1 \right]
\]

subject to

\[
D^* \sum_{j \in I} \int_{y_j}^{z_j} f_1(b_1) \, db_1 - \sum_{j \in J} \int_{y_j}^{z_j} f_1(b_1) \, db_1 = 1 - \beta_2,
\]

\[
y_j \geq z_j \text{ for all } j \in I \text{ and } j \in J,
\]

\[
z_{j-1} \geq y_j \text{ for all } j \in I \text{ and } j \in J,
\]

\[
g \geq y_j \text{ for all } j \in I \text{ and } j \in J,
\]

\[
z_j \geq h \text{ for all } j \in I \text{ and } j \in J,
\]

and

\[
D^* \geq \frac{0.827 f_1 + K_2 - 0.827 \lambda (1 + \alpha) b_1}{1 - \lambda} \text{ for all } b_1 \text{ in } [y_j, z_j] \text{ when } j \in I,
\]

\[
D^* \geq \frac{0.827 f_1 + K_2 - 0.827 \lambda (1 + \alpha) b_1}{1 - \lambda} \text{ for all } b_1 \text{ in } [y_j, z_j] \text{ when } j \in J.
\]

In Eq 16 the objective is obtained by using the expression for \( t_2'(b_1) \) given in Eq 12 since it is only necessary to perform the integration over those intervals for which \( t_2'(b_1) \) is not identically 0, i.e., for \( j \in I \) and \( j \in J \). The first constraint of Eq 16 corresponds to the first constraint of Eq 10, except that it is written in such a way that it involves only the intervals \([y_j, z_j]\) for which \( j \in I \) and \( j \in J \). The second constraint says the \( P(t_2' < 0) = 1 - \beta_2 \), which is true if and only if \( P(t_2' > 0) = \beta_2 \); therefore it corresponds to the second constraint of Eq 10. The remaining conditions assure that a collection of nonoverlapping intervals will be secured such that \( t_2' > 0 \) for \( j \in I \) and \( t_2' < 0 \) for \( j \in J \). The fact that \( t_2' = 0 \) on at most a set of measure 0 in any of the intervals for \( j \in I \) and \( j \in J \) is guaranteed by the fact that \( f_1 \frac{[0.827 t_1 + \beta_2 - (1 - \lambda (1 + \alpha) b_1)]}{1 - \lambda} \) is a
strictly monotonic decreasing function of \( b \) over the sets where \( \mathcal{f}_1(b) > 0 \), as \( [1 - \lambda (1 + i)] > 0 \).

Using this strict monotonicity the last two constraints of Eq 16 can be replaced by

\[
D^* = \mathcal{f}_2 \left\{ \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] y_j}{1 - \lambda} \right\} \text{ for all } j, f.
\]

and

\[
D^* = \mathcal{f}_2 \left\{ \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] z_j}{1 - \lambda} \right\} \text{ for all } j, f.
\]

Equation 16 will now be solved using the Lagrange multiplier technique to derive necessary conditions for \( y_j, z_j \) to be optimal for \( j \in I, j \in J \). To do this begin by forming the Lagrangian function

\[
L = [(1 - \lambda) \mathcal{F}_1^- (D^*) - K_2 - 0.827 \mathcal{f}_1] \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1(b) \text{ d}(b_i)
\]

\[
+ [(1 - \lambda) \mathcal{F}_1^- (D^*) - K_2 - 0.827 \mathcal{f}_1] \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1(b) \text{ d}(b_i)
\]

\[
+ \Omega \left\{ \mathcal{f}_1 \left[ (y_j - z_j) - \mathcal{f}_1 (y_j) \right] - \mathcal{f}_2 \left( \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] b_j}{1 - \lambda} \right) \int_{b_1}^{b_2} \mathcal{f}_1 \text{ d}(b) - (1 - \theta_2) - Q \right\}
\]

\[
+ \lambda \left\{ \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 (z_j) - \mathcal{f}_1 (y_j) - (1 - \beta_2) \right\}
\]

\[
+ \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 (y_j - z_j - 1 - R_j^2)
\]

\[
+ \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 (y_j - \theta - Q_j^2) + \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 (h - z_j - p_j^2)
\]

\[
+ \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 \left( \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] y_j}{1 - \lambda} \right) \int_{b_1}^{b_2} \mathcal{f}_1 \text{ d}(b) - (1 - \beta_2) - Q \right\}
\]

\[
+ \sum_{I \cap J} \int_{b_{i-1}}^{b_i} \mathcal{f}_1 \left( \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] z_j}{1 - \lambda} \right) \int_{b_1}^{b_2} \mathcal{f}_1 \text{ d}(b).
\]

where \( \Delta, \Omega, \eta, \xi, \delta, \phi, \lambda, \theta \), are Lagrange multipliers and

\[
Q = \int_{b_1}^{b_2} \left( \frac{0.827 \mathcal{f}_1 + K_2 - [1 - \lambda (1 + \theta)] b_j}{1 - \lambda} \right) \mathcal{f}_1 \text{ d}(b).
\]

The following equations and inequalities provide some necessary conditions that \( y_j', z_j' \) maximize \( L \) and hence solve Eq 16.

Let \( j \in I \)

Then \( \Delta L / \Delta z_j = 0 \) implies

\[
z_j = \frac{\eta_j \xi_j + \phi_j}{[1 - \lambda (1 + \theta)] \mathcal{f}_1 (y_j)} - \left[ \frac{(1 - \lambda) \mathcal{F}_1^- (D^*) - K_2 - 0.827 \mathcal{f}_1}{1 - \lambda (1 + \theta)} \right]
\]

\[
- \left( \frac{\eta_j \xi_j + \phi_j}{[1 - \lambda (1 + \theta)] \mathcal{f}_1 (z_j)} - \left[ \frac{(1 - \lambda) \mathcal{F}_1^- (D^*) - K_2 - 0.827 \mathcal{f}_1}{1 - \lambda (1 + \theta)} \right] \right)
\]

(17)
and $\delta L/\partial y_j = 0$ implies

$$y_j = \frac{\eta_j \cdot \xi_j + \delta_j}{1 - \lambda (1 + \vartheta)} \cdot \frac{1}{f_1 (y_j)} \cdot \frac{1}{1 - \lambda (1 + \vartheta)} \cdot \left[ (1 - \lambda) F_2^{-1} (D_y) - K_2 - 0.827 f_1 \right]$$

$$- \left( \Omega D^v - \Omega F_2 \right) \frac{0.827 f_1 - [1 - \lambda (1 + \vartheta)] y_j + K_2}{1 - \lambda}$$

$$\lambda f_2 (y_j) \left[ 1 - \lambda \right]$$

(18)

Let $j \in J$

Then $\delta L/\partial z_j = 0$ implies

$$z_j = \frac{\eta_j \cdot \xi_j + \delta_j}{1 - \lambda (1 + \vartheta)} \cdot \frac{1}{f_1 (z_j)} \cdot \frac{1}{1 - \lambda (1 + \vartheta)} \cdot \left[ (1 - \lambda) F_2^{-1} (D_y) - K_2 - 0.827 f_1 \right]$$

$$- \left( \Omega D^v - \Omega F_2 \right) \frac{0.827 f_1 - [1 - \lambda (1 + \vartheta)] z_j + K_2}{1 - \lambda}$$

$$\lambda f_2 (z_j) \left[ 1 - \lambda \right]$$

(19)

and $\delta L/\partial y_j = 0$ implies

$$y_j = \frac{\eta_j \cdot \xi_j + \delta_j}{1 - \lambda (1 + \vartheta)} \cdot \frac{1}{f_1 (y_j)} \cdot \frac{1}{1 - \lambda (1 + \vartheta)} \cdot \left[ (1 - \lambda) F_2^{-1} (D_y) - K_2 - 0.827 f_1 \right]$$

$$- \left( \Omega D^v - \Omega F_2 \right) \frac{0.827 f_1 - [1 - \lambda (1 + \vartheta)] y_j + K_2}{1 - \lambda}$$

$$\lambda f_2 (y_j) \left[ 1 - \lambda \right]$$

(20)

$$\frac{\partial L}{\partial T_j} = 0 \cdot \eta_j T_j = 0, \quad j \in I \text{ and } j \neq j.$$  \quad (21)

$$\frac{\partial L}{\partial R_j} = 0 \cdot \xi_j R_j = 0, \quad j \in I \text{ and } j \neq j.$$  \quad (22)

$$\frac{\partial^2 L}{\partial R_j^2} \leq 0 \cdot \xi_j \leq 0, \quad j \in I \text{ and } j \neq j.$$  \quad (23)

$$\frac{\partial L}{\partial S_j} = 0 \cdot \lambda_j S_j = 0, \quad j \in I.$$  \quad (24)

$$\frac{\partial^2 L}{\partial S_j^2} \leq 0 \cdot \lambda_j \leq 0, \quad j \in I.$$  \quad (25)

13
From the expression Eq 12 for \( t'_2 \) in any interval \([y_j, z_j]\) for which \( j \in I \), it is seen that \( t'_2 \) is a straight line of positive slope as \( [1 - \lambda (1 + i)] > 0 \). Therefore if at some point, say \( b_1, t'_2 (b_1) > 0 \) then \( t'_2 (b_1) > 0 \) for all \( b_1 > b_1 \). This conclusion follows directly from the geometry of the situation.

Let \( m \) be the smallest value of \( j \) for which \( t'_2 \geq 0 \) almost everywhere in \([y_j, z_j]\). Then \( t'_2 \geq 0 \) in all intervals \([y_j, z_j]\) for which \( j = m \).

Similarly \( t'_2 \leq 0 \) in all intervals \([y_j, z_j]\) for which \( j < m \).

Next, it is observed from Eq 21 that \( \eta_j \neq 0 \rightarrow T_j = 0 \rightarrow y_j = z_j \) from the equation \( \partial L / \partial \eta_j = 0 \). But \( y_j = z_j \) implies that Eq 16 could just as well be solved by dropping the value of \( j \) from the problem. So the assumption now is that \( \eta_j = 0 \) for all \( j \).

Moreover from Eq 22 \( \xi_{j+1} \neq 0 \rightarrow R_{j+1} = 0 \rightarrow y_{j+1} = z_j \) from the equation \( \partial L / \partial \xi_{j+1} = 0 \). But when \( j = m \) it is known that \( t'_2 \neq 0 \rightarrow t'_2 > 0 \), so that \( y_{j+1} = z_j \) means that in the intervals \([y_j, z_j]\) and \([y_{j+1}, z_{j+1}]\), \( t'_2 \) is defined by \( t'_2 = [1 - \lambda (1 + i)] b_1 - 0.827 t_1 - K_2 + (1 - \lambda) F_2 [F_1(D^*)] \). If this is the case, then the intervals \([y_j, z_j], [y_{j+1}, z_{j+1}]\) can be written as one continuous interval \([y_j, z_{j+1}]\). Thus the indexes could be renumbered dropping one interval from the problem. It is assumed that \( \xi_{j+1} = 0, j \geq m \).

Similarly it is assumed that \( \xi_{j+1} = 0 \) for \( j < m - 2 \). Therefore the only \( \xi_j \) that can be nonzero is \( \xi_m \).

Using an analogous argument the only \( \phi_i \) that can be nonzero is \( \delta_k \), and the only \( \phi_t \) that can be nonzero is \( \phi_t \), where \( t \) is such that \( z_t = z_{k}, j \in I \), and \( k \) is such that \( y_k = y_k \), \( j \in I \), \( j \in I \).

Now if \( j \in I \) then \( \xi_{j+1} = 0 \), and if \( \phi = 0 \) also, then Eq 17 implies that there is only one \( z_j \) for \( j \in I \). This follows from the fact that the solution of the equation

\[
\frac{[1 - \lambda] F_1^{-1} (D^*) - K_2 - 0.827 t_1}{1 - \lambda (1 + i)}
\]

is unique, as \( F_2 \) is a monotonic function of \( z_j \) and it is known that Eq 17 must hold for all \( j \in I \).

If \( \phi_j \neq 0 \) then \( j = t \) from the work above. Hence using \( \partial L / \partial P_t = 0 \) it is found that \( P_t = 0 \rightarrow z_t = h \) from the equation \( \partial L / \partial \phi_t = 0 \), so again \( z_t \) is uniquely determined. Hence, there exists at most one interval \([y_j, z_j]\) for \( j \in I \).

Similarly by using Eq 20 and our definition of \( \delta_k \), it is shown that there exists at most one interval \([y_j, z_j]\) for \( j \in I \).

Using the fact that \( t'_2 \) is a monotonic increasing function of \( b_1 \) in \([y'_j, z'_j]\), \( j \in I \) and \( j \in I \), and that \( l, l \) contain at most one index \( j \), it can be seen that the constraints restricting \( t'_2 > 0, j \in I \), and \( t'_2 < 0, j \in I \), can, in fact, be replaced by \( t'_2 > 0 \) when \( j = m \), \( t'_2 < 0 \) when \( j = m - 1 \), and \( t'_2 = 0 \) for \( j \neq m - 1, m \). Note
also that it is necessary to have \( \delta_m = 0 \), as \( \delta_m \neq 0 \rightarrow y_m = g \rightarrow \) is empty, which is impossible since \( \beta_2 \neq 0 \).

Since Eq 16 is being solved for \( y' \), \( z' \) for \( j \in I \) and \( j \in J \), it is now only necessary to consider the cases in which \( j = m \) and \( j = m - 1 \).

Let \( j = m \) and suppose \( z_m < h \) so that \( P_m \neq 0 \). Then \( \phi_m = 0 \) from the equation \( \partial L / \partial P_m = 0 \), and the following possible cases exist.

(a) \( \xi_m = 0 \) and \( \lambda_m = 0 \). Then Eq 17 and Eq 18 imply \( y_m = z_m \); i.e., \( I \) is empty.

(b) \( \xi_m = 0 \) and \( \lambda_m \neq 0 \). Then \( S_m = 0 \) from Eq 24; hence, from \( \partial L / \partial \lambda_m = 0 \), there is

\[
\lambda_m = \frac{\left[ - \left( 1 - \lambda \right) \tilde{F}_2^{-1} \left( D'' \right) \cdot 0.827 f_1 \cdot k_2 \right]}{\left( 1 - \lambda \right) \tilde{F}_2 \left( y_m \right)}.
\]

Putting this into Eq 18 with \( j = m \), using the fact that \( \delta_m = 0 \), the result is

\[
\left. \begin{array}{c}
\lambda_m \tilde{F}_2^{-1} \left( 0.827 f_1 \cdot k_2 \cdot 1 \cdot 1 - \lambda \cdot 1 \cdot 1 \right) \\
\lambda_m \tilde{F}_2^{-1} \left( \frac{1 - \lambda, 1 \cdot 1 \cdot 1}{1 - \lambda} \right) \\
\lambda_m \tilde{F}_2^{-1} \left( 0.827 f_1 \cdot k_2 \cdot 1 \cdot 1 - \lambda \cdot 1 \cdot 1 \right)
\end{array} \right\} = 0,
\]

or

\[
\frac{\lambda_m \tilde{F}_2^{-1} \left( 0.827 f_1 \cdot k_2 \cdot 1 \cdot 1 - \lambda \cdot 1 \cdot 1 \right)}{\left( 1 - \lambda \right) \tilde{F}_2 \left( y_m \right)} = 0.
\]

Hence, the result is

\[
\frac{\lambda_m \tilde{F}_2^{-1} \left( 0.827 f_1 \cdot k_2 \cdot 1 \cdot 1 - \lambda \cdot 1 \cdot 1 \right)}{\left( 1 - \lambda \right) \tilde{F}_2 \left( y_m \right)} = 0.
\]

But \( \tilde{F}_2 \left[ \tilde{F}_2^{-1} \left( D'' \right) \right] \not\equiv 0 \) as \( 0 < D'' < 1 \), therefore Eq 29 implies \( \lambda_m = 0 \), which is a contradiction. Therefore \( \phi_m = 0 \), \( \xi_m \neq 0 \) and \( \lambda_m \neq 0 \) is impossible.

(c) \( \xi_m \neq 0 \) and \( \lambda_m = 0 \). Then \( R_m = 0 \rightarrow y_m = z_m-1 \). But \( t_2'(y_m) \geq 0 \) and \( t_2(z_{m-1}) \leq 0 \) from the definition of \( m \), hence \( t_2'(y_m) = 0 \). This implies that Eq 28 must again hold, therefore, by inserting Eq 28 into Eq 18 the following equation must be true:

\[
\frac{\xi_m}{\left[ 1 - \lambda \cdot 1 \cdot 1 \right] \tilde{F}_2 \left( y_m \right)} = 0, \quad \text{or} \quad \xi_m = 0,
\]

which is a contradiction. Therefore

\[
\phi_m = 0, \quad \xi_m \neq 0 \quad \text{and} \quad \lambda_m = 0
\]
is impossible.
(d) $\xi_m \neq 0$ and $\lambda_m \neq 0$. Again Eq 28 holds; hence

\[
\frac{\lambda_m \int_{y_m}^{\xi_m} (D^*)}{(1-\lambda) \int_{y_m}^{\xi_m}} = \frac{\xi_m}{[1-\lambda (1+i)] \int_{y_m}^{\xi_m}} = 0.
\]

Therefore $\lambda_m = 0$ and $\xi_m = 0$ as by Eqs 23 and 25 $\lambda_m \neq 0$, $\xi_m \neq 0$, and it is known that $(1-\lambda), [1-\lambda (1+i)], \int_{y_m}^{\xi_m}$, $\int_{y_m}^{\xi_m}$, and $\int_{2}^{\xi_m} (D^*)$ are all positive. Thus a contradiction exists.

Therefore

\[
\begin{cases}
\delta_m = 0, & 1 \text{ is empty,} \\
\delta_m = 0, & z_m = h.
\end{cases}
\] (30)

Using an analogous development by assuming that $\delta_m = 0$ similar contradictions are derived and it is concluded that

\[
\begin{cases}
\delta_m = 0, & 1 \text{ is empty,} \\
\delta_m = 0, & z_m = a.
\end{cases}
\] (31)

It has already been shown that at most only one interval $[y_m, z_m]$ exists in $1$ or $2$. So using the constraint that $\sum_i \epsilon_i [\int_{d_i}^{y_i} (z_i) - \int_{d_i}(y_i)] = 1 - \beta_2$, it is seen that $\beta^*$ cannot be empty; hence $y_{m-1} = y^*, z_{m-1} = \int_{m-1}^{m+1} (1-\beta_2)$.

Therefore

\[
\gamma_i^* (b_i) = \left[1 - \lambda (1+i) \right] \left[ b_i - 0.827 \int_{d_i}^{y_i} - K_2 \right] (1-\lambda) \int_{y_i}^{\xi_i} (D^*)
\]

for all $b_i : \int_{y_i}^{\xi_i} = \int_{m-1}^{m+1} (1-\beta_2)$. (32)

Moreover using Eqs 30 and 32 Eq 9 can be written as the following non-linear problem in the three variables $D^2$, $y_m$, and $\xi_m$:

maximize

\[
\gamma_i^* (b_i) = \left[1 - \lambda (1+i) \right] \left[ b_i - 0.827 \int_{d_i}^{y_i} - K_2 \right] (1-\lambda) \int_{y_i}^{\xi_i} (D^*)
\]

subject to

\[
\begin{align*}
\int_{d_i}^{y_i} & = 0, \\
D & = \int_{d_i}^{y_i} \left( 0.827 \int_{d_i}^{y_i} - K_2 \right) \left[ \lambda (1+i) \right] \left[ b_i - 0.827 \int_{d_i}^{y_i} - K_2 \right] (1-\lambda) \int_{y_i}^{\xi_i} (D^*)
\end{align*}
\]

for all $b_i : \int_{y_i}^{\xi_i} = \int_{m-1}^{m+1} (1-\beta_2)$. (33)

By solving Eq 33 and then putting $\xi_m$, $D^*$, and $y_m^*$ into Eq 12 and using Eq 32 the optimal $\xi_m^*(b_i)$ is derived. Hence Eq 9 has been solved.
Conclusions

It has been shown in the preceding section that when the constraint \( P(t_2 > 0) \geq \beta_2 \) is not binding the result is

\[
\gamma_1 = 1 - \lambda (1 + i) \quad \text{for} \quad \alpha = b_1 - h.
\]

In other words, \( t_2'(b_1) \) is a linear function of \( b_1 \) and as such it coincides with the optimal linear rule found by Charnes and Thore" for this same case. It is emphasized again that in finding the optimal linear rule Charnes and Thore" required \( b_1 \) and \( b_2 \) to be normal random variables, although the results hold for a much more general class of random variables.

In order to give a brief interpretation of these linear rules write

\[
\begin{align*}
\gamma_1' & = \gamma_1 \\
\gamma_2' & = \gamma_2 (b_1 - E_1),
\end{align*}
\]

where \( \gamma_1, \gamma_2, \gamma_2' \) are constants and \( E_1 \) is the mean of the random variable \( b_1 \) (i.e., \( \Delta b_1 \)).

Since \( \gamma_2 = 1 - \lambda (1 + i) > 0 \), the optimal decision rule \( t_2' \) states that if there has been an inflow of savings during period 1 that exceeds the mean, free cash holdings during period 2 will be lowered by an amount \( \gamma_2 (b_1 - E_1) \), as opposed to the case in which \( t_2' \) is chosen to be a zero-order rule (i.e., \( t_2' = \gamma_2 \)).

It is also clear from the solution that the holdings of free cash will be equal to their minimum values as long as these minimum values do not interfere with the limitations of long-term borrowing given by the \( \{ t_i > 0 \} \cdot \beta_i \), \( i = 1, 2 \), constraints. When these constraints become binding the amount of free cash held will be greater than this minimum value. Thus the association's profits drop, owing to the loss in interest that would be obtained if less money were held and, consequently, more money loaned.

It is worth noting at this point that if Eq 5 were expanded to be a full \( n \)-dimensional problem and if it were assumed that the constraints \( P(t_i > 0) \cdot \beta_i \), \( j = 1, \ldots, n \), are not binding then by theorem 6 in Charnes and Kirby" it can be seen that the optimal rule for \( t_i \) will, in fact, be the optimal linear rule. This result certainly gives great justification for using linear decision rules in cases where the computation of the optimal decision rule from a much larger class of decision rules is extremely difficult to do.

Consider the case where the constraint \( P(t_2 > 0) \cdot \beta_2 \) is binding. The first observation is that in this case \( t_2' \) is discontinuous. The fact that it must have at least one discontinuity follows directly from Eq 32, since it is known that \( y_m' > \frac{1}{1 - \lambda (1 + i)} \) when \( P(t_2 > 0) \cdot \beta_2 \) is binding. In general, however, \( t_2' \) will also be discontinuous at \( b_1 = F_1^{-1} (1 - \beta_2) \).

\[
\begin{align*}
t_2'(b_1) \text{ will be continuous at } b_1 = F_1^{-1} (1 - \beta_2) \text{ if and only if } t_2' - F_1^{-1} (1 - \beta_2) \left. \right|_{b_1} = 0, \text{ i.e., if and only if }
\end{align*}
\]

It is known then that to get \( t_2' \) it is necessary to increase the measure of the
set of points for which \( t_2^0 \cdot 0 \) where \( t_2^0 \) is the optimal rule when \( \beta_2 = \beta_2^0 \), i.e., \( t_2^0 \) must be increased for some values of \( b_1 \) to get \( t_2^* \). The \( 1 - \alpha_2 \) constraint being satisfied by \( t_2^0 \) as an equality means that \( t_2^0 \) must also decrease for some values of \( b_1 \), otherwise this constraint would be violated. The question then arises, for what values of \( b_1 \) should \( t_2^0 \) be increased and for what values should \( t_2^0 \) be decreased in order to get \( t_2^* \). It seems reasonable that \( t_2^* \) should be as close to \( t_2^0 \) as possible. Therefore \( t_2^0 \) should be increased for those values of \( b_1 \) for which \( t_2^2 < 0 \) but is almost 0. Similarly \( t_2^0 \) should be decreased where it is positive but almost 0. This accounts for the location of the region \( t_2^2 = 0 \).

For all remaining values of \( b_1 \), \( t_2^0 \) is changed as little as possible by making \( t_2^* \) linear with the same slope as \( t_2^0 \) and changing the intercept term just enough to satisfy the \( 1 - \alpha_2 \) constraint.

If it is assumed that \( b_1 \) and \( b_2 \) are \( N(E_i, \delta_i) \), \( i = 1, 2 \), random variables, it can be shown that the optimal linear rule when the \( \beta_2 \) constraint is binding is

\[
(1 - A) \cdot (1 - i) \left( \Phi^{-1} (1 - \beta_2) \delta_1 \mid 1 - A (1 - i) \right)
\]

(34)

where \( \Phi \) is the distribution function of a \( N(0, 1) \) random variable.

Now suppose that \( \beta_2 > \frac{1}{2} \), then \( \Phi^{-1} (1 - \beta_2) < 0 \), so that the slope of \( t_2^* \) in Eq 34 is greater than the slope of \( t_2^0 \) in Eq 32. This means that the optimal linear rule would recommend borrowing more when \( b_1 \) is very negative and lending more when \( b_1 \) is very positive than would the optimal rule given by Eq 32.

Conversely if \( \beta < \frac{1}{2} \), the slope of Eq 34 is less than the slope of Eq 32 when \( t_2^* \neq 0 \). In this case using the linear rule recommends borrowing less and lending less for small and large values of \( b_1 \) respectively, than would Eq 32.

In conclusion, there is one further reason for not being too surprised at the discontinuity of \( t_2^* \) under certain circumstances. Equation 10 can be regarded as a problem for finding the "optimal control" \( t_2 (b_1) \). From control theory it is known that in many cases the optimal control is discontinuous. It takes on one value for certain time intervals and then switches abruptly to another value. This is precisely what \( t_2^* \) does. It is defined by Eq 32 for all values of \( b_1 \) in \( \delta_i, \delta_i \), \( 1 - \beta_2 \), it then switches to \( t_2^* = 0 \) from \( \delta_i \) \( 1 - \beta_2 \) to \( \gamma_i \), and then switches again to the line defined by Eq 32. Thus \( t_2^* \) behaves similarly to the optimal controls in many control-theory problems.
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ADDITIONAL REFERENCES