

AUTOMATIC STUDY IN STOCHASTIC CONTROL

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Abstract :

The purpose of this paper is to give an example of automatic generation of a complete study in stochastic control done by an expert system designed at INRIA by the authors. This study includes the:

- *automatic choice of an algorithm,*
- *automatic checking of the mathematical well posedness of the problem,*
- *automatic generation of a numerical routine,*
- *automatic test of this routine on a numerical example,*
- *automatic generation of graphics,*
- *automatic writing of a report describing all the obtained results .*

1) INTRODUCTION

We give here the present state of an expert system about optimization and identification of dynamic systems, started four years ago. In this system we want to code the state of art in the domain, using the more advanced techniques in computer science: -inference engine, -symbolic manipulation, -numerical techniques. This system must be able to derive by itself classical theoretical results, to choose a method, to apply algorithms, to write programs, to make a numerical study, to design some graphics and to write a complete report from a specification in a specialized command language.

The present code has about 15000 lines of a mixture of LISP, MACSYMA, and PROLOG and needs a lot of improvements but has actually all the possibilities described above.

After a brief description of its mathematical capabilities in optimal control we give an example of a session to give an idea of the level of mathematical background needed by the user. Then we present the report generated by the expert system. This is an example where the dynamic programming approach has been chosen. It is the most developed part of the system.

The computer science aspect of the system has been described in Chancellier-Gomez-Quadrat-Sulem [43], the mathematical aspects briefly reported here are described more precisely in Theosys [21] .

A similar work for nonlinear filtering is developed by the Blankenship team at the Maryland University.

2) STOCHASTIC CONTROL

We define the stochastic control of diffusion processes following Kushner[25] Fleming-Rishel[15], Bensoussan[4], Bensoussan-Lions[6], by :

i) its dynamics :

$$(1) \quad dX_t = b(X_t, U_t) dt + h dW_t$$

with :

- t the time,
- X_t the state of the system,
- U_t the control,
- W_t a brownian motion modelling of the stochastic perturbations,
- h the local standard deviation of the noise,
- b the drift of the system;

ii) the criterium to be minimized is one of the three following types:

$$(2) \quad \text{Inf}_U E \{ \int_0^T c(X_t, U_t) dt \}$$

with:

- T the horizon which can be a stopping time,
- c an instantaneous cost ;

$$(3) \quad \text{Inf}_U E \{ \int_0^\infty e^{-lt} c(X_t, U_t) dt \}$$

with l the discount rate,

$$(4) \quad \text{Inf}_U \text{Lim}_{T \rightarrow \infty} \{ 1/T \int_0^T c(X_t, U_t) dt \} .$$

3) ALGORITHMS

The system uses four kinds of methods to solve more or less completely the problem :

- dynamic programming approach,
- decoupling method,
- stochastic gradient technique,
- perturbation methods.

Each approach has its most efficient domain of application.

3.1) DYNAMIC PROGRAMMING

This approach leads to solve one of the following kinds of partial differential equation (PDE).

- The finite horizon case (2) :

$$(5) \quad D_t V + \text{Inf}_U \{ b(x, u) D_x V + c(x, u) \} + \text{tr}(a D_{xx} V) = 0$$

where V denotes the dynamic programming function defined by:

$$(6) \quad V(t,x) = \text{Inf } E \left\{ \int_t^T c(X_s, U_s) ds \mid X_t = x \right\} ;$$

- The infinite horizon case with a discounted cost (10) :

$$(7) \quad -lV + \text{Inf}_u \{ b(x,u) \cdot D_x V + c(x,u) \} + \text{tr}(a \cdot D_{xx} V) = 0 ,$$

with :

$$(8) \quad V(x) = \text{Inf } E \left\{ \int_0^\infty e^{-ls} c(X_s, U_s) ds \mid X_0 = x \right\} ;$$

- The average cost by unit of time case (11) ::

$$(9) \quad -m + \text{Inf}_u \{ b(x,u) \cdot D_x W + c(x,u) \} + \text{tr}(a \cdot D_{xx} W) = 0 ,$$

with m defined by :

$$(10) \quad m = \text{Inf } \lim_{T \rightarrow \infty} 1/T \int_0^T c(X_s, U_s) ds .$$

We have used the notation:

$$(11) \quad a = 1/2h \cdot h^* .$$

This approach is the best one when it can be used, that is when the dimension of the system is less than 3 or 4. Indeed we have to solve a discretized version of the PDE numerically Kushner[25], Quadrat[34],[35], Goursat-Quadrat[22].

It is the only approach which gives a complete answer under general hypotheses needed in practise. More over we can use the aggregation method, for example Turgeon[42], Torrion[41], to reduce the dimension of the system.

This is the approach chosen by the system in the session example in the report generated. We can see the explanation about the discretization of the dynamic programming equation and the proof of the existence and uniqueness of a solution of the Bellman Equation. In the future the system will use the better result of P.L.Lions described in these proceedings.

3.2) DECOUPLING METHOD

If the system has a special structure : -uncoupled dynamics, -computable criterium without numerical evaluation of multiple integrals on a large dimension space, then it is possible to optimize in the class of local feedbacks.

If the system has not this special structure, by changing the feedback it is sometimes possible to come down to this structure. It is the classical problem of decoupling in the automatic control litterature (Claude[8],

Geromel-Lévine-Willis[16], in the non linear case).

Then the problem can be reduced to a PDE control problem. Indeed one controls the marginal laws of the density of probability of the state. By hypotheses on the structure, these laws are independent and satisfy Fokker-Planck equations .

More precisely given :

- an uncoupled dynamic :

$$(12) \quad dX_{it} = b_i(X_{it}, U_{it})dt + h_i dW_{it}$$

(where i denotes the number of the subsystem)

- a coupling criterium computable by convolutions or gaussian approximation from the law of the subsystem:

$$(13) \quad E \int_0^T f(Z_t - \sum_i c_i(X_{it}, U_{it})) dt ,$$

- Z_t a function of time ,

we minimize (13) in the class of local feedbacks.

The problem comes down to the control of the marginal laws p_i of each subsystem given by :

$$(14) \quad - D_t p_i - D_x \{ b(x_i, u_i) p_i \} + D_{xx} (a_{ii} p_i) = 0$$

$$(15) \quad p(0, x) \text{ known ,}$$

with the criterium :

$$(16) \quad \text{Min} \int_0^T f(Z_t - \sum_i c_i(x_i, u_i(x_i))) \prod_i p_i(t, x_i) dx_i$$

where $p_i(t, x_i)$ denotes the density for X_{it} to be at x_i at time t .

This problem can be solved numerically. A theoretic study of the algorithm for a discrete version of the problem can be found in Quadrat-Viot[37]. An application to hydropower management is given in Delebecque-Quadrat[11]. A beautiful study of the loss of optimality is given in Torrion[41].

With this kind of approach we can solve Quadrat_Viot[37] some steady state control problem of Jackson type queuing network BCMP[2].

The possibilities of the system for automatic generation of numerical routines based on this method have been described in Gomez-Quadrat-Sulem[20].

3.3) STOCHASTIC GRADIENT

Sometimes we have an idea of a good class of feedback in which we would like to optimize. Then it is possible to use stochastic gradient techniques.

Denoting by $U(x, a)$ this parametrized class, where a is a parameter, a becomes an open loop control to be optimized in the following control problem:

$$(17) \quad dX_t = b(X_t, U(X_t, A_t)) dt + h dW_t ,$$

$$(18) \quad \text{Inf}_A \ E \{ \int_0^T c(X_t, U(X_t, A_t)) dt \} .$$

Then we have to compute the sequence indexed by n :

$$(19) \quad A_{n+1} = A_n - r_n D_A J(w_n, A_n)$$

$$(20) \quad J(w_n, A_n) = \int_0^T c(X_{tn}, U(X_{tn}, A_{tn})) dt$$

$$(21) \quad \sum_n r_n = \infty, \quad r_n > 0, \quad r_n \rightarrow 0$$

where w_n denotes independent simulation of trajectories of the noise.

A_n converges to the optimum if we have made some convexity assumptions Polyak[31], [32], Kushner-Clark[26], Dodu-Goursat-Hertz-Quadrat-Viot[12]. The computational cost increases linearly with the dimension of the system so it is particularly useful for large system. For small size systems dynamic programming gives a more complete answer.

3.4) REGULAR PERTURBATION

When the size of the noise is small, we can compute affine feedback good up to order 4 Cruz[10], Fleming[14], Bensoussan[5], for enough regular systems.

We are in finite horizon problem case (13) with :

-the dynamics :

$$(22) \quad dX_t = b(X_t, U_t) dt + h dW_t , \quad h \ll 1,$$

-the affine following feedback :

$$(23) \quad u(t, x) = u_0(t) + K(t)(x - x_0(t))$$

where (u_0, x_0) is the solution of the following deterministic control problem ($h=0$):

$$(24) \quad dx_0(t) = b(x_0(t), u_0(t)) dt ,$$

$$(25) \text{ Inf } \int_0^T c(x_0(t), u_0(t)) dt ,$$

and $K(t)$ is the solution of the tangent linear quadratic problem :

$$(26) K = -H_{uu} (H_{ux} + b_u P) ,$$

$$(27) P' + A^*P + PA - PSP + Q = 0 ,$$

$$(28) A = b_x - b_u H_{uu}^{-1} H_{ux} ,$$

$$(29) S = b_u H_{uu}^{-1} b_u^* ,$$

$$(30) Q = H_{xx} - H_{ux}^* H_{uu}^{-1} H_{ux} ,$$

(31) $H=b.p+c$ where p denotes the dual variables of the states, gives, when (26).....(31) is well posed, a feedback which has a loss of optimality of order h^4 .

This method is the best one for stabilization problems.

The system knows all these methods and is able to generate a study based on all of these point of view. The more advanced part of the system is the dynamic programming part. It is for example the only method for which the system is able to make the theoretical work.

It is able to chose the method then it checks the well-posedness of the HJB equation if the dynamic programming approach has been chosen, it generates numerical subroutines and test them on a numerical example. Finally it generates the report.

The system is written in LISP,MACSYMA,PROLOG. The generated numerical program is in FORTRAN. Prolog is used to organize the program, Macsyma is used for the algebraic manipulations and for building the numerical routines, the 3D plotting, the scientific editing. Lisp is used for a good integration of Prolog and Macsyma and for general purpose programming. Let us recall that the Prolog used, called Oblogis, is written in Lisp like Macsyma. The system is currently under development on the Symbolics Lisp Machine and a former version exists on Multics.

4) A SESSION EXAMPLE OF THE PRESENT SYSTEM

We give an example of the interaction of the system with the user and the generated report .

(robot)

NOUS ALLONS ESSAYER DE RESOUDRE VOTRE PROBLEME DE CONTROLE STOCHASTIQUE
ENONCEZ LE PROBLEME

EN CAS DE DIFFICULTES TAPEZ help

PRECISEZ TOUT D'ABORD LE NUMERO OU LE NOM DE VOTRE PROBLEME EN TAPANT PAR EXEMPLE pr
obleme 0

PRECISEZ EGALEMENT LA LANGUE DANS LAQUELLE LE RAPPORT DEVRA ETRE REDIGE EN TAPANT ve
rsion francaise OU version anglaise

====> VERSION ANGLAISE

Loading LM1:>sulem>new>clauses-text.lisp into package MACSYMA

Loading LM1:>sulem>new>textes-anglais.lisp into package MACSYMA

====> HELP

VOUS AVEZ BESOIN D ' AIDE POUR

- ENTRER LA DYNAMIQUE DU PROBLEME (1)

- DONNER L ' HORIZON ET LES CONDITIONS AUX LIMITES (2)

- QUESTIONNER OU MODIFIER LA BASE DE CONNAISSANCE DU ROBOT (3)

- GENERER LES PROGRAMMES FORTRAN , OBTENIR LES RESULTATS THEORIQUES ET NUMERIQUES ,

GENERER LE RAPPORT ET LES GRAPHES (4)

TAPER LE NUMERO CORRESPONDANT

1

LE ROBOT COMPREND LES PHRASES DU TYPE SUIVANT :

ON PEUT UTILISER N ' IMPORTE LAQUELLE DES EXPRESSIONS SEPARÉES PAR LE SYMBOLE /

PROBLEME DE COMMANDE/EVALUATION_COUT_MOYEN/CALCUL_DENSITE_PROBABILITE

X EST UN ETAT/COMMANDE/PARAMETRE

1 EST LE COUT-INSTANTANE

LA MOYENNE-DYNAMIQUE/VARIANCE-DYNAMIQUE EN X1 EST 1

DANS LE CAS D ' UN CALCUL DE DENSITE DE PROBABILITE , PRECISEZ :

INVARIANTE/MARGINALES EST LA LOI

LA SIGNIFICATION/UNITE/DOMAIN-DE-VARIATION DE X EST ...

====> PROBLEME 0

PROBLEME-GLOBAL 0

====> X1 IS A STATE VARIABLE

JE NE COMPRENDS PAS

====> X1 IS A STATE

JE NE COMPRENDS PAS

====> BIEN-POSE

WHAT IS THE TYPE (type)

- COMMAND (commande) ,

- EVALUATION OF AN AVERAGE COST (evaluation_cout_moyen),

- PROBABILITY DENSITY COMPUTATION (calcul_densite_probabilite) OF THE PB 0 ?
COMMANDE;

WHAT IS THE horizon

-FINITE (fini),

-INFINITE (infini),

-ERGODIC (ergodique) FOR THE PB 0 ?

INFINI;

WHAT IS THE TIME NAME (temps) FOR THE PB 0 ?

T;

WHAT IS THE LIST OF THE STATE NAMES (etat) FOR THE PB 0 ?
[X1,X2];

WHAT IS THE LIST OF THE PARAMETER NAMES (parametre) FOR THE PB 0 ?
[];

WHAT IS THE LIST OF THE PRICE NAMES (prix) DU PB 0 ?
[P1,P2];

WHAT IS THE LIST OF THE PRICE DERIVATIVE NAMES (derivee-prix) FOR THE PB 0 ?
[Q1,Q2];

WHAT IS THE LIST OF THE COMMAND NAMES (commande) OF THE PB 0 ?
[U1,U2];

WHAT IS THE NAME OF THE OPTIMAL COST (cout-optimal) FOR THE PB 0 ?
V;

WHAT IS THE AVERAGE OF THE DYNAMIC (moyenne-dynamique), FOR THE PB 0 ,OF X1 ?
 $3 \cdot X1 + U1$;

WHAT IS THE VARIANCE OF THE DYNAMIC (variance-dynamique), FOR THE PB 0 ,OF X1 ?
 $1 + X1^2$;

WHAT IS THE AVERAGE OF THE DYNAMIC (moyenne-dynamique), FOR THE PB 0 ,OF X2 ?
 $3 \cdot X2 + U2$;

WHAT IS THE VARIANCE OF THE DYNAMIC (variance-dynamique), FOR THE PB 0 ,OF X2 ?
 $1 + X2^2$;

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF X1 ?
[0,1];

WHAT IS THE BOUNDARY CONDITION (condition-frontiere) FOR THE PB 0 AT X1 = 0 ?
[REFLECHI,0];

WHAT IS THE BOUNDARY CONDITION (condition-frontiere) FOR THE PB 0 AT X1 = 1 ?
[REFLECHI,0];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF X2 ?
[0,1];

WHAT IS THE BOUNDARY CONDITION (condition-frontiere) FOR THE PB 0 AT X2 = 0 ?
[ARRET,0];

WHAT IS THE BOUNDARY CONDITION (condition-frontiere) FOR THE PB 0 AT X2 = 1 ?
[ARRET,0];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF U1 ?
[-1,1];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF U2 ?
[-1,1];

WHAT IS THE VALUE OF THE INSTANTANEOUS COST (cout-instantane) FOR THE PB 0 ?
 $U1^2 + U2^2 + (X1 - 1/2)^2 + X2^2$;

WHAT IS THE actualisation RATE OF THE PB 0 ?
3;

====> METHODE
METHODE : PROGRAMMATION-DYNAMIQUE

====> RESOUDRE ?PPD

DO YOU WANT THE SUBPROGRAM OUTPUT (sortie) YES (oui),NO (non) FOR THE PB 0 ?
OUI;

WHAT IS THE WANTED precision FOR THE PB 0 ?
0.1;

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF P1 ?
[-1,1];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF P2 ?
[-1,1];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF Q1 ?
[-1,1];

WHAT IS THE DOMAIN OF VARIATION (domaine-de-variation) , FOR THE PB 0 ,OF Q2 ?
[-1,1];

====> HELP

VOUS AVEZ BESOIN D ' AIDE POUR

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 - GENERER LES PROGRAMMES FORTRAN , OBTENIR LES RESULTATS THEORIQUES ET NUMERIQUES ,
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- TAPER LE NUMERO CORRESPONDANT

4

ON GENERE LE SOUS-PROGRAMME/PROGRAMME PRINCIPAL EN TAPANT PAR EX :

SOUS-PROGRAMME/PROGRAMME-PRINCIPAL ?PPD

POUR GENERER LE SOUS-PROGRAMME ,LE PROGRAMME PRINCIPAL ET EXECUTER ON TAPE:

RESOUDRE ?PPD

POUR AVOIR LE GRAPHE DE V EN FONCTION DE (X1 X2) POUR LA METHODE PROGRAMMATION DYNAMIQUE , ON TAPE :

GRAPHE V [X1,X2] ?PPD

POUR AVOIR LES FIGURES DE V ET U AVEC LES COURBES DE NIVEAU :

FIGURES V U ?PPD

POUR AVOIR LES RESULTATS THEORIQUES (EXISTENCE D'UNE SOLUTION) :

DEMONSTRATION

POUR GENERER LE RAPPORT :

RAPPORT

====> DEMONSTRATION

In the hamiltonian H(V) :

$$H(V) = \min \left(U \frac{dV}{dX} + U \frac{dV}{dX} + U + U \right)$$

$\begin{matrix} 2 & & 1 & & 2 & & 1 \\ & 2 & & 1 & & 2 & & 1 \end{matrix}$

- the coefficients [U₁ , U₂] are bounded by 1

- the coefficient U₂ + U₁ is bounded by 2.

The hamiltonian H(V) is lipschitzian with coefficient 1.4142135.

The variational formulation of the problem in the space H^1 is :

$$C(V, W) = - \int_0^1 (H(V) W) + A(V, W) = \int_0^1 ((X^2 + (X - 0.5)^2) W)$$

where $A(V, W)$ is the linear part :

$$A(V, W) = \int_0^1 ((X^2 + 1) \frac{dV}{dX} \frac{dW}{dX}) + \int_0^1 ((X^2 + 1) \frac{dV}{dX} \frac{dW}{dX}) - \int_0^1 (X^2 \frac{dV}{dX} W)$$

$$- \int_0^1 (X \frac{dV}{dX} W) + 3 \int_0^1 (V W)$$

The linear part is coercive in the space H^1 :

$$A(V, V) \geq \frac{(9235521.0 \text{ TETA}^2 + 2.9180478e7 \text{ TETA} + 1.0) \text{ NORM}(V, L)^2}{1.191288e7 \text{ TETA} + 3453520.0}$$

$$- 2.5510203e-4 (3039.0 \text{ TETA} - 3039.0) \text{ NORM}(V, H)$$

with $0.0 < \text{TETA} < 1.0$.

The problem is bounded in the space H^1 because the coefficients of the hamiltonian are bounded.

The problem is hemicontinuous in the space H^1 because the hamiltonian is lipschitzian.

The problem is monotonous in the space H^1 with $0.11699876 < \text{TETA} < 0.693925$

The problem is inf-compact in the space H^1 with $0.355051 < \text{TETA} < 0.717224$

The problem has a unique solution in the space H^1 .

====> GRAPHE V [X1,X2] ?PPD
 ====> GRAPHE U1 [X1,X2] ?PPD
 ====> FIGURES V U ?PPD
 ====> RAPPORT

Written: LM1:>quadrat>fortran>rapport.fortran.61

====> HELP

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TAPER LE NUMERO CORRESPONDANT

3

ON QUESTIONNE LE ROBOT PAR DES PHRASE DU TYPE :

QUELLE EST LA COMMANDE
QUELLES SONT LES CONDITION-FRONTIERE
POUR SUPPRIMER UNE CLAUSE ON TAPE :
SUPPRIMER LE COUT-FINAL
POUR LISTER LES DONNEES DU PROBLEME ON TAPE :
DONNEES
POUR VERIFIER QUE LE PROBLEME EST BIEN POSE ON TAPE :
BIEN-POSE
POUR SORTIR DU SYSTEME ON TAPE
STOP

===> DONNEES

ACTUALISATION 0 3

COMMANDE 0 U1

COMMANDE 0 U2

CONDITION-FRONTIERE 0 X1 0 REFLECHI 0

CONDITION-FRONTIERE 0 X1 1 REFLECHI 0

CONDITION-FRONTIERE 0 X2 0 ARRET 0

CONDITION-FRONTIERE 0 X2 1 ARRET 0

2 1 2 2 2
COUT-INSTANTANE 0 X2 + (X1 - -) + U2 + U1
2

COUT-OPTIMAL 0 V

DERIVEE-PRIX 0 Q1

DERIVEE-PRIX 0 Q2

DOMAINE-DE-VARIATION 0 X1 0 1

DOMAINE-DE-VARIATION 0 X2 0 1

DOMAINE-DE-VARIATION 0 U1 - 1 1

DOMAINE-DE-VARIATION 0 U2 - 1 1

DOMAINE-DE-VARIATION 0 P1 - 1 1

DOMAINE-DE-VARIATION 0 P2 - 1 1

DOMAINE-DE-VARIATION 0 Q1 - 1 1

DOMAINE-DE-VARIATION 0 Q2 - 1 1

ETAT 0 X1

ETAT 0 X2

HORIZON 0 INFINI

MOYENNE-DYNAMIQUE 0 X1 3 X1 + U1

MOYENNE-DYNAMIQUE 0 X2 3 X2 + U2

PRECISION 0 0.1

PRIX 0 P1

PRIX 0 P2

PROBLEME 0

SORTIE 0 OUI

TEMPS 0 T

TYPE 0 COMMANDE

2
VARIANCE-DYNAMIQUE 0 X1 X1 + 1
2

VARIANCE-DYNAMIQUE 0 X2 X2 + 1
METHODE-POSSIBLE 0 PROGRAMMATION-DYNAMIQUE

DISCRETISATION 0 ESPACE [0, 0]

OPTIMISATION 0 GRADIENT-PROJECTION

NB-PT-DISCRETISATION-ESPACE 0 [11, 11]

NB-PT-DISCRETISATION-TEMPS 0 11

NB-MAXI-ITERATION-OPTIMISATION 0 24

NB-MAXI-ITERATION-IMPLICITE 0 1232

PAS-GRADIENT 0 0.25

PAS-IMPLICITE 0 0.00124533

COEF-CONTRACTION 0 0.0037359898

PRECISION-RES-IMPLICITE 0 0.0015068494

PLOT 0 VPERSP

PLOT 0 VCONTOUR
PLOT 0 U1PERSP
PLCT 0 U1CONTOUR

==> STOP
AU REVOIR

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ABSTRACT

We consider a static system the 2-dimensional state of which is modelled by a controlled diffusion process defined on $[0,1] \times [0,1]$. The purpose is to minimize the expected discounted cost that includes a integral cost and stopping and reflecting costs. The optimal cost satisfies a Bellman equation derived from the optimal principle of Dynamic Programming. The Dynamic Programming equation is discretized and then solved numerically.

1 - Notations

State variables: $[X1, X2]$

Control variables : $[U1, U2]$

Time: T

Optimal cost: V

State dimension : N

ith state variable : XI

Derivative Operator with respect to the time variable: D_0

Derivative Operator with respect to XI : D_I

2 - Evolution equation of the system:

We consider the control diffusion process defined by the dynamic equation :

$$(0-2) \quad dX_2 = (3 X_2 + U_2) dT + \text{SQRT}(2 X_2^2 + 2) \cdot dW_2$$

$$(1-1) \quad dX_1 = (3 X_1 + U_1) dT + \text{SQRT}(2 X_1^2 + 2) \cdot dW_1 - dZ_1 + dZ_1$$

where

$X1$ belongs to $[0, 1]$

$X2$ belongs to $[0, 1]$

$U1$ belongs to $[- 1, 1]$

$U2$ belongs to $[- 1, 1]$

W_t is a Wiener process, i.e. a continuous gaussian process

with independent increments.

Z_t is an increasing process, strictly increasing when $X_t = J$

is on the boundary $X_t = J$.

X_t is reflected on the boundary $X_t = 0$.

X_t is reflected on the boundary $X_t = 1$.

This process is well defined [7]. It is the limit when the time step h goes to 0 of a markovian discrete process X_n

which satisfies:

$$(i) \quad E(X_{n+1} - X_n | F_n) = h \begin{bmatrix} 3X_n + U_1 \\ 3X_n + U_2 \end{bmatrix} + O(h)$$

$$(ii) \quad E((X_{n+1} - X_n)^2 | F_n) = h \begin{bmatrix} X_n^2 + 1 & 0 \\ 0 & X_n^2 + 1 \end{bmatrix} + O(h)$$

(iii) A uniform integrability condition of the increment $X_{n+1} - X_n$

where F_n denotes the sigma-algebra generated by X_0, X_1, \dots, X_n .

3 - Value function

The stopping time T_F is defined by :

$$T_F = \min(T_1, T_2)$$

with

$$T_1 = \inf \{T \mid X_t \leq 0\}$$

$$T_2 = \inf \{T \mid X_t \geq 1\}$$

The problem is to minimise the expectation of the function :

$$(2) \quad J(S) = E \int_0^{T_F} (X_t^2 + (X_t - \frac{1}{2})^2 + U_2 + U_1) dt$$

in the feedbacks class, i.e. the applications :

$$S : [X_1, X_2] \rightarrow [U_1, U_2] .$$

4 - Optimality conditions :

The Bellman function V is defined by:

$$(3) V(Y_1, Y_2) = \min_S (E [J(S) \mid [X_1 = Y_1, X_2 = Y_2]])$$

V satisfies the Dynamic Programming equation [2] [1]:

$$(4) \min_{U_1, U_2} \left((X_2^2 + 1) \frac{d^2 V}{dX_2^2} + (3X_2 + U_2) \frac{dV}{dX_2} + (X_1^2 + 1) \frac{d^2 V}{dX_1^2} + (3X_1 + U_1) \frac{dV}{dX_1} + X_2^2 + (X_1 - \frac{1}{2})^2 + U_2 + U_1 \right) - 3V(X_1, X_2) = 0$$

$$V(X_1, 0) = 0$$

$$V(X_1, 1) = 0$$

$$\frac{d}{dX_1} V(0, X_2) = 0$$

$$\frac{d}{dX_1} V(1, X_2) = 0$$

5 - Theoretical Analysis

In the hamiltonian H(V) :

$$H(V) = \min_U \left(U^2 \frac{dV}{dX_1} + U \frac{dV}{dX_2} + U^2 + U \right)$$

- the coefficients $[U_1, U_2]$ are bounded by 1

- the coefficient $U^2 + U$ is bounded by 2.

The hamiltonian H(V) is lipschitzian with coefficient 1.4142135.

The variational formulation of the problem in the space H is :

$$C(V, W) = - \int (H(V) W) + A(V, W) = \int \left((X_1^2 + (X_2 - 0.5)^2) W \right)$$

where $A(V, W)$ is the linear part :

$$A(V, W) = \frac{1}{2} \left[(X + 1) \frac{dV}{dX} \frac{dW}{dX} \right] + \frac{1}{2} \left[(X + 1) \frac{dV}{dX} \frac{dW}{dX} \right] - \frac{1}{2} \left[(X + 1) \frac{dV}{dX} \frac{dW}{dX} \right] + 3 \frac{1}{2} (V W)$$

The linear part is coercive in the space H^1 :

$$A(V, V) \geq \frac{(9235521.0 \text{ TETA}^2 + 2.9180478e7 \text{ TETA} + 1.0) \text{ NORM}(V, L)^2}{1.191288e7 \text{ TETA} + 3453520.0}$$

$$- 2.5510203e-4 (3039.0 \text{ TETA} - 3039.0) \text{ NORM}(V, H)$$

with $0.0 < \text{TETA} < 1.0$.

The problem is bounded in the space H^1 because the coefficients of the hamiltonian are bounded.

The problem is hemicontinuous in the space H^1 because the hamiltonian is lipschitzian.

The problem is monotonous in the space H^1 with $0.11699876 < \text{TETA} < 0.693925$

The problem is inf-compact in the space H^1 with $0.355051 < \text{TETA} < 0.717224$

The problem has a unique solution in the space H^1 .

6 - Dynamic programming method

Our purpose is to solve the Bellman equation (4) after discretization [5] [6] [3] [4]. This is possible because the state dimension is small.

6-1 Discretization:

We denote:

- h_i : discretization step for the i -th space variable X_i

We define the following operators:

$$S_i : V(X_1, \dots, X_i, \dots, X_N) \rightarrow V(X_1, \dots, X_i + h_i, \dots, X_N) \quad i=1, \dots, N$$

$$\delta = \frac{S - 1}{I \quad HI}$$

$$\partial = \frac{\delta}{I} + \frac{\delta}{2 \frac{S}{I}}$$

$$\gamma = \frac{\delta}{I \quad S}$$

We thus approximate:

$$\frac{dV}{dXI} \text{ by } \gamma (V)$$

$$\frac{dV}{dX1} \text{ by } \partial (V)$$

$$\frac{dV}{dX2} \text{ by } \partial (V)$$

The discretized Bellman equation is:

$$(5) \text{ MIN}_{U1, U2} \left((3 X2 + U2) \frac{\partial (V)}{2} + (X2^2 + 1) \frac{\gamma (V)}{2} + (3 X1 + U1) \frac{\partial (V)}{1} \right. \\ \left. + (X1^2 + 1) \frac{\gamma (V)}{1} + X2^2 + (X1 - \frac{1}{2})^2 + U2^2 + U1^2 \right) - 3 V(X1, X2) = 0$$

- reordering with respect to S we get:

$$(6) \text{ MIN}_{U1, U2} \left(- \left(\frac{8 X2^2}{H2} + \frac{8 X1^2}{H1} + \frac{8}{H2} + \frac{8}{H1} \right) V \right)$$

$$\begin{aligned}
& + (S_2 \cdot V) \left(- \frac{4 X_2^2}{H_2} + \frac{6 X_2}{H_2} + \frac{2 U_2}{H_2} - \frac{4}{H_2} \right) \\
& + (S_2 \cdot V) \left(- \frac{4 X_2^2}{H_2} - \frac{6 X_2}{H_2} - \frac{2 U_2}{H_2} - \frac{4}{H_2} \right) - 4 X_2^2 \\
& + (S_1 \cdot V) \left(- \frac{4 X_1^2}{H_1} + \frac{6 X_1}{H_1} + \frac{2 U_1}{H_1} - \frac{4}{H_1} \right) \\
& + (S_1 \cdot V) \left(- \frac{4 X_1^2}{H_1} - \frac{6 X_1}{H_1} - \frac{2 U_1}{H_1} - \frac{4}{H_1} \right) - 4 X_1^2 + 4 X_1 - 4 U_2 - 4 U_1 - 1) / 4) \\
& - 3 V(X_1, X_2) = 0
\end{aligned}$$

6-2 Probabilistic Interpretation of the discretized equation :

The discretization of the Bellman equation

$$(7) \text{MIN}_{U_1, U_2} (A V + C(U_1, U_2)) - \lambda V = 0$$

can be interpreted as a control problem of Markov chain with discount factor k and instantaneous cost kC . The associated cost function is

$$(8) \quad k \sum_{N=0}^{\infty} C(X_N, U_N) (\lambda k + 1)^{-N-1}$$

and the Markov matrix M :

$$M = k A + I$$

where I is the Identity matrix,

and k : inverse of the maximum of the diagonal of A

is given by:

INITIAL_PT	FINAL_PT	TRANSITION_PROBABILITY
[X1, X2]	[X1, X2]	0
[X1, X2]	[X1 + H1, X2]	$\frac{2 X1^2}{H1} + \frac{3 X1^2}{H1} + \frac{U1}{H1} + \frac{2}{H1}$
[X1, X2]	[X1, X2 + H2]	$2 \left(\frac{2 X2^2}{H2} + \frac{2 X1^2}{H1} + \frac{2}{H2} + \frac{2}{H1} \right)$
[X1, X2]	[X1, X2 + H2]	$\frac{2 X2^2}{H2} + \frac{3 X2^2}{H2} + \frac{U2}{H2} + \frac{2}{H2}$
[X1, X2]	[X1 - H1, X2]	$\frac{2 X1^2}{H1} - \frac{3 X1^2}{H1} - \frac{U1}{H1} + \frac{2}{H1}$
[X1, X2]	[X1, X2 - H2]	$2 \left(\frac{2 X2^2}{H2} + \frac{2 X1^2}{H1} + \frac{2}{H2} + \frac{2}{H1} \right)$
[X1, X2]	[X1, X2 - H2]	$\frac{2 X2^2}{H2} - \frac{3 X2^2}{H2} - \frac{U2}{H2} + \frac{2}{H2}$

Indeed if the following conditions are satisfied:

$$(9-1) \quad H1 \leq \sup_{X1, X2, U1, U2} \frac{2(X1^2 + 1)}{\text{ABS}(3X1 + U1)}$$

$$(9-2) \quad H2 \leq \sup_{X1, X2, U1, U2} \frac{2(X2^2 + 1)}{\text{ABS}(3X2 + U2)}$$

the matrix coefficients are positive and the sum of the coefficients on a same line is equal to 1. The matrix M is thus a transition matrix of a Markov chain. Moreover the optimal cost obeys:

$$(10) \quad (\lambda k + 1) V = \min_{U1, U2} (M(U1, U2) V + k C(U1, U2))$$

Thus we can use the contraction iteration:

$$(11) \quad V_{N+1} = \frac{\min_{U1, U2} (M(U1, U2) \cdot V_N + k C(U1, U2))}{\lambda k + 1}$$

6-3 Optimisation method:

We denote by H the Hamiltonian defined by:

$$(12) \quad H = (X2^2 + 1) \frac{dV}{dX2} + (3X2 + U2) \frac{dV}{dX2} + (X1^2 + 1) \frac{dV}{dX1} + (3X1 + U1) \frac{dV}{dX1} + X2^2 + (X1 - \frac{1}{2})^2 + U2^2 + U1^2$$

H is minimised by a projected gradient method:

$$(13) \quad U_{N+1} = \text{PROJ}_{[-1, 1]} \left(U_N - \frac{d}{dU} (H(U)) \right)$$

that is:

$$\begin{aligned}
 & \left[\begin{array}{l} U_1 \\ N + 1 \end{array} \right] = \text{PROJ}_{[-1, 1]} \left(\begin{array}{l} U_1 - R \left(\frac{dV}{dX_1} + 2 U_1 \right) \\ N \end{array} \right) \\
 (14) & \left[\begin{array}{l} U_2 \\ N + 1 \end{array} \right] = \text{PROJ}_{[-1, 1]} \left(\begin{array}{l} U_2 - R \left(\frac{dV}{dX_2} + 2 U_2 \right) \\ N \end{array} \right)
 \end{aligned}$$

This algorithm converges when the step R satisfies:

$$(15) \quad 0 < R < \frac{2m}{M}$$

with:

$$(16) \quad m |V| \leq \frac{D}{U} H(V) \cdot V \leq M |V|$$

6-4 Numerical results

Annex 1 Main Program

PROGRAM M_PPD

REAL V(11,11),U(2,11,11)

C

do 1002 K1=1,11

C

do 1001 K2=1,11

V(K1,K2)=0.0

C

do 1000 J=1,2

U(J,K1,K2)=0.0

1000

CONTINUE

C

fin de do

C

1001 CONTINUE

C

fin de do

C

1002 CONTINUE

C

fin de do

C

CALL PPD(11,11,V,0.0015068494,1232,0.00124533,U,0.010000001,24,0.2

1 5)

STOP

END

Annex 2 Subroutine solving the dynamic programming equation

SUBROUTINE PPD(N1,N2,V,EPSIMP,IMPMAX,RO,U,EPS,NMAX,ROG)

DIMENSION V(N1,N2),U(2,N1,N2)

Resolution de l equation de Kolmogorov dans le cas ou:

Les parametres sont

L etats-temps est: X1 X2

La dynamique du systeme est decrite par l operateur

plus($Q_2 X_2^2 + X_2^2 + 3 P_2 X_2 + Q_1 X_1^2 + X_1^2 + 3 P_1 X_1 - X_1 + Q_2$

$+ Q_1 + 0.25$, Minu($U_2^2 + P_2 U_2 + U_1^2 + P_1 U_1$))

ou v(..) et w designe le cout

ou pi designe sa derivee premiere par rapport a xi

ou qi designe sa derivee seconde par rapport a xi

Le probleme est statique

Les conditions aux limites sont:

X2 = 0 V = 0

X2 = 1 V = 0

X1 = 0 -p1 = 0

X1 = 1 p1 = 0

Les nombres de points de discretisation sont: N1 N2

X2 = 1 correspond a I2 = N2

X2 = 0 correspond a I2 = 1

X1 = 1 correspond a I1 = N1 - 1

X1 = 0 correspond a I1 = 2

Le taux d actualisation vaut: 3

impmax designe le nbre maxi d iterations du systeme implicite

epsimp designe l erreur de convergence du systeme implicite

ro designe le pas de la resolution du systeme implicite

par une methode iterative

P2 est discretise par difference divise symetrique

P1 est discretise par difference divise symetrique

Minimisation par la methode de gradient avec projection

de l'Hamiltonien:

$U_2^2 + P_2 U_2 + U_1^2 + P_1 U_1$

contraintes sur le controle:

- 1 =< U2 =< 1

- 1 =< U1 =< 1

nmax designe le nombre maxi d iteration de la methode de

gradient avec projection

eps designe l erreur de convergence de la methode de

gradient avec projection

rog designe le pas, qui est constant, dans la methode de gradi#

ent

H2 = 0.999999/(N2-1)

H1 = 0.999999/(N1-3)

U2 = U(2,1,1)

U1 = U(1,1,1)

HIH2 = H2**2

HIH1 = H1**2

H22 = 2*H2

H21 = 2*H1

NM2 = N2-1

NM1 = N1-1

do 1019 I2=1,N2,1

do 1019 I1=1,N1,1

V(I1,I2) = 0.0


```

1019 CONTINUE
    IMITER = 1
1013 CONTINUE
    ERIMP = 0
    do 1011 I1=1,N1,1
        X1 = H1*(I1-2)
        V(I1,N2) = 0
        V(I1,1) = 0
1011 CONTINUE
    do 1009 I2=2,NM2,1
        X2 = H2*(I2-1)
        V(N1,I2) = V(N1-2,I2)
        V(1,I2) = V(3,I2)
1010 CONTINUE
    do 1009 I1=2,NM1,1
        X1 = H1*(I1-2)
        Q2 = (V(I1,I2+1)-2*V(I1,I2)+V(I1,I2-1))/HIH2
        Q1 = (V(I1+1,I2)-2*V(I1,I2)+V(I1-1,I2))/HIH1
        P2 = (V(I1,I2+1)-V(I1,I2-1))/H22
        P1 = (V(I1+1,I2)-V(I1-1,I2))/H21
        W = V(I1,I2)
        NITER = 0
        W0 = -1.0e20
1001 CONTINUE
    NITER = NITER+1
    if (NITER-NMAX) 1002,1002,1003
1003 CONTINUE
    WRITE(8,1801) I1,I2
1801  FORMAT(' descente n a pas converge', 2 i3)
    GOTO 1004
1002 CONTINUE
    UN1 = (1-2*ROG)*U1-P1*ROG
    UN2 = (1-2*ROG)*U2-P2*ROG
    U1 = UN1
    U2 = UN2
    U1 = AMAX1(U1,-1)
    U1 = AMIN1(U1,1)
    U2 = AMAX1(U2,-1)
    U2 = AMIN1(U2,1)
    WW = U2**2+P2*U2+U1**2+P1*U1
    ER = ABS(WW-W0)
    if (ER-EPS) 1004,1004,1005
1005 CONTINUE
    W0 = WW
    GOTO 1001
1004 CONTINUE
    U(1,I1,I2) = U1
    U(2,I1,I2) = U2
    W0 = WW
    W1 = Q2*X2**2+X2**2+3*P2*X2+Q1*X1**2+X1**2+3*P1*X1-X1+Q2+Q1+0.25
    W0 = W1+W0
    W0 = W0-3*V(I1,I2)
    VNEW = RO*W0+V(I1,I2)
    V(I1,I2) = VNEW
    ERIMP = ABS(W0)+ERIMP
1009 CONTINUE
    IMITER = IMITER+1
    if (IMITER-IMPMAX) 1016,1015,1015
1016 CONTINUE
    if (EPSIMP-ERIMP) 1013,1012,1012

```

```
1015 CONTINUE
      WRITE(8,1807)
1807  FORMAT(' schema implicite n a pas converge')
1012 CONTINUE
      do 1017 I1=1,N1,1
      do 1017 I2=1,N2,1
      WRITE(8,1800) I1,I2,V(I1,I2)
1800  FORMAT(' FTNV[', (i3,','), i3,']:', e14.7,'$')
      WRITE(8,1901) I1,I2,U(1,I1,I2)
1901  FORMAT(' FTNU1[', (i3,','), i3,']:', e14.7,'$')
      WRITE(8,1902) I1,I2,U(2,I1,I2)
1902  FORMAT(' FTNU2[', (i3,','), i3,']:', e14.7,'$')
1017 CONTINUE
      RETURN
      END
```

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